

Package ‘cograph’

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Title Analysis and Visualization of Complex Networks

Version 2.1.1

Description Provides tools for the analysis, visualization, and manipulation of dynamical, social (Saqr et al. (2024) <doi:10.1007/978-3-031-54464-4_10>) and complex networks (Saqr et al. (2025) <doi:10.1145/3706468.3706513>). The package supports multiple network formats and offers flexible tools for heterogeneous, multi-layer, and hierarchical network analysis with simple syntax and extensive toolset.

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URL <https://sonsoles.me/cograph/>,
<https://github.com/sonsoleslp/cograph>

BugReports <https://github.com/sonsoleslp/cograph/issues>

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abbrev_label	<i>Abbreviate Labels</i>
--------------	--------------------------

Description

Abbreviates labels to a maximum length, adding ellipsis if truncated.

Usage

```
abbrev_label(label, abbrev = NULL, n_labels = NULL)
```

```
label_abbrev(label, abbrev = NULL, n_labels = NULL)
```

Arguments

label	Character vector of labels to abbreviate.
abbrev	Abbreviation control: <ul style="list-style-type: none">• NULL: No abbreviation (return labels unchanged)• Integer: Maximum character length (truncate + ellipsis)• "auto": Adaptive abbreviation based on label count
n_labels	Number of labels (used for "auto" mode). If NULL, uses length(label).

Value

Character vector of (possibly abbreviated) labels.

Examples

```
labels <- c("VeryLongStateName", "Short", "AnotherLongName")

# No abbreviation
abbrev_label(labels, NULL)

# Fixed max length
abbrev_label(labels, 5) # "Very...", "Short", "Anot..."

# Auto-adaptive
abbrev_label(labels, "auto")
```

aggregate_layers	<i>Aggregate Layers</i>
------------------	-------------------------

Description

Combines multiple network layers into a single network.

Usage

```
aggregate_layers(  
  layers,  
  method = c("sum", "mean", "max", "min", "union", "intersection"),  
  weights = NULL  
)  
  
lagg(  
  layers,  
  method = c("sum", "mean", "max", "min", "union", "intersection"),  
  weights = NULL  
)
```

Arguments

layers	List of adjacency matrices
method	Aggregation: "sum", "mean", "max", "min", "union", "intersection"
weights	Optional layer weights (for weighted sum)

Value

Aggregated adjacency matrix

Examples

```
# layers <- list(L1 = mat1, L2 = mat2, L3 = mat3)  
# aggregate_layers(layers, "sum")           # Total  
# aggregate_layers(layers, "mean")         # Average  
# aggregate_layers(layers, "union")        # Any edge  
# aggregate_layers(layers, "intersection") # All edges
```

aggregate_weights	<i>Aggregate Edge Weights</i>
-------------------	-------------------------------

Description

Aggregates a vector of edge weights using various methods. Compatible with igraph's edge.attr.comb parameter.

Usage

```
aggregate_weights(w, method = "sum", n_possible = NULL)
```

```
wagg(w, method = "sum", n_possible = NULL)
```

Arguments

w	Numeric vector of edge weights
method	Aggregation method: "sum", "mean", "median", "max", "min", "prod", "density", "geomean"
n_possible	Number of possible edges (for density calculation)

Value

Single aggregated value

Examples

```
w <- c(0.5, 0.8, 0.3, 0.9)
aggregate_weights(w, "sum") # 2.5
aggregate_weights(w, "mean") # 0.625
aggregate_weights(w, "max") # 0.9
```

assortativity	<i>Degree Assortativity Coefficient</i>
---------------	---

Description

Computes the degree assortativity coefficient, measuring the tendency of nodes to connect to other nodes with similar degree. Positive values indicate assortative mixing (high-degree nodes connect to high-degree nodes), negative values indicate disassortative mixing.

Usage

```
assortativity(x, directed = NULL, type = NULL, digits = NULL, ...)
```

Arguments

<code>x</code>	Network input: matrix, igraph, network, cograph_network, or tna object.
<code>directed</code>	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.
<code>type</code>	Character string specifying which degree correlation to compute. One of "out-in" (default for directed), "in-in", "out-out", "in-out", or "degree" (for undirected). Ignored for undirected networks.
<code>digits</code>	Integer or NULL. Round result to this many decimal places. Default NULL (no rounding).
<code>...</code>	Additional arguments passed to to_igraph .

Details

The degree assortativity coefficient is defined as the Pearson correlation coefficient between the degrees of nodes at either end of each edge (Newman 2002):

$$r = \frac{\sum_{jk} jk(e_{jk} - q_j q_k)}{\sigma_q^2}$$

where e_{jk} is the fraction of edges connecting degree- j to degree- k vertices, q_k is the excess degree distribution, and σ_q^2 its variance.

For directed networks, different degree combinations (in/out) at source and target ends can be specified via the `type` parameter.

Value

An object of class "cograph_assortativity" with components:

coefficient Numeric scalar: the assortativity coefficient in $[-1, 1]$.

type Character: the degree type used.

directed Logical: whether the network was treated as directed.

n_nodes Integer: number of nodes.

n_edges Integer: number of edges.

network The original input network.

References

Newman, M.E.J. (2002). Assortative mixing in networks. *Physical Review Letters*, 89(20), 208701. doi:10.1103/PhysRevLett.89.208701

See Also

[assortativity_attribute](#), [centrality](#), [network_summary](#)

Examples

```
# Assortative network (high-degree connect to high-degree)
adj <- matrix(c(
  0, 1, 1, 1, 0,
  1, 0, 1, 1, 0,
  1, 1, 0, 0, 1,
  1, 1, 0, 0, 1,
  0, 0, 1, 1, 0
), 5, 5)
rownames(adj) <- colnames(adj) <- LETTERS[1:5]
cograph::assortativity(adj)
```

assortativity_attribute

Attribute Assortativity (Homophily)

Description

Computes assortativity with respect to a node attribute, measuring the tendency of nodes to connect to others with similar attribute values. For categorical attributes, this computes the modularity-based nominal assortativity. For numeric attributes, this computes the Pearson correlation between attribute values at edge endpoints.

Usage

```
assortativity_attribute(x, values, directed = NULL, digits = NULL, ...)
```

```
homophily(x, values, directed = NULL, digits = NULL, ...)
```

Arguments

<code>x</code>	Network input: matrix, igraph, network, cograph_network, or tna object.
<code>values</code>	Named vector of attribute values (names must match node names) or an unnamed vector in node order.
<code>directed</code>	Logical or NULL. If NULL (default), auto-detect.
<code>digits</code>	Integer or NULL. Round result. Default NULL.
<code>...</code>	Additional arguments passed to to_igraph .

Details

For categorical (nominal) attributes, the coefficient is:

$$r = \frac{\text{tr}(\mathbf{e}) - \|\mathbf{e}^2\|}{1 - \|\mathbf{e}^2\|}$$

where \mathbf{e} is the mixing matrix with e_{ij} = fraction of edges connecting type i to type j .

For numeric (scalar) attributes, the coefficient is the Pearson correlation between attribute values at edge endpoints.

Value

An object of class "cograph_assortativity" with components:

coefficient Numeric scalar: assortativity coefficient.

type Character: "nominal" or "scalar".

directed Logical.

n_nodes Integer.

n_edges Integer.

attribute_values The attribute values used.

network Original input.

References

Newman, M.E.J. (2003). Mixing patterns in networks. *Physical Review E*, 67(2), 026126. doi:10.1103/PhysRevE.67.026126

See Also

[assortativity](#), [detect_communities](#)

Examples

```
adj <- matrix(c(0,1,1,0, 1,0,0,0, 1,0,0,1, 0,0,1,0), 4, 4)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
groups <- c(A = "x", B = "x", C = "y", D = "y")
cograph::assortativity_attribute(adj, groups)
```

as_cograph

Convert to Cograph Network

Description

Creates a lightweight cograph_network object from various network inputs. The resulting object is a named list with all data accessible via \$.

Usage

```
as_cograph(x, directed = NULL, simplify = FALSE, ...)
```

```
to_cograph(x, directed = NULL, ...)
```

Arguments

x	Network input. Can be: <ul style="list-style-type: none"> • A square numeric matrix (adjacency/weight matrix) • A data frame with edge list (from, to, optional weight columns) • An igraph object • A statnet network object • A qgraph object • A tna object • An existing cograph_network object (returned as-is)
directed	Logical. Force directed interpretation. NULL for auto-detect.
simplify	Logical or character. If FALSE (default), every transition from tna sequence data is a separate edge. If TRUE or a string ("sum", "mean", "max", "min"), duplicate edges are aggregated.
...	Additional arguments (currently unused).

Details

The cograph_network format is designed to be:

- Lean: Only essential data stored, computed values derived on demand
- Modern: Uses named list elements instead of attributes for clean \$ access
- Compatible: Works seamlessly with `splot()` and other cograph functions

Use getter functions for programmatic access: [get_nodes](#), [get_edges](#), [get_labels](#), [n_nodes](#), [n_edges](#)

Use setter functions to modify: [set_nodes](#), [set_edges](#), [set_layout](#)

Value

A cograph_network object: a named list with components:

`nodes` Data frame with id, label, (x, y if layout applied)

`edges` Data frame with from, to, weight columns

`directed` Logical indicating if network is directed

`weights` Full nxn weight matrix (for `to_matrix` round-trip)

`data` Original estimation data (sequence matrix, edge list, etc.), or NULL

`meta` Consolidated metadata list with sub-fields: `source` (input type string), `layout` (layout info list or NULL), `tna` (TNA metadata or NULL)

`node_groups` Optional node groupings data frame

A cograph_network object. See [as_cograph](#).

See Also

[get_nodes](#) to extract the nodes data frame, [get_edges](#) to extract edges as a data frame, [n_nodes](#) and [n_edges](#) for counts, [is_directed](#) to check directedness, [splot](#) for plotting

Examples

```

# From adjacency matrix
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)

# Direct $ access to core data
net$nodes      # nodes data frame
net$edges      # edges data frame
net$directed   # TRUE/FALSE

# Getter functions (recommended for programmatic use)
get_nodes(net) # nodes data frame
get_edges(net) # edges data frame (from, to, weight)
get_labels(net) # character vector of labels
n_nodes(net)   # 3
n_edges(net)   # 3
cograph::is_directed(net) # FALSE (symmetric matrix)

# Setter functions
net <- set_nodes(net, data.frame(id = 1:3, label = c("A", "B", "C")))
net <- set_edges(net, data.frame(from = c(1,2), to = c(2,3), weight = c(0.5, 0.8)))
net <- set_layout(net, data.frame(x = c(0, 1, 0.5), y = c(0, 0, 1)))

# Plot it
splot(net)

# From igraph (if installed)
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_ring(10)
  net <- as_cograph(g)
  splot(net)
}
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- to_cograph(mat)

```

as_mcml

Convert to mcml

Description

Convert various objects to the mcml class – a clean, tna-independent representation of a multilayer cluster network.

Usage

```

as_mcml(x, ...)

## S3 method for class 'cluster_summary'
as_mcml(x, ...)

```

```
## S3 method for class 'group_tna'
as_mcml(x, clusters = NULL, method = "sum", type = "tna", directed = TRUE, ...)

## S3 method for class 'mcml'
as_mcml(x, ...)

## Default S3 method:
as_mcml(x, ...)
```

Arguments

x	Object to convert.
...	Additional arguments passed to methods.
clusters	Integer or character vector of row-to-group assignments. Required when the group_tna has the same labels across all groups (row-level clustering from tna::group_model(cluster_data(...))).
method	Aggregation method for macro weights (default "sum").
type	Transition type (default "tna").
directed	Logical; whether the network is directed (default TRUE).

Value

An mcml object with components macro, clusters, cluster_members, and meta.

An mcml object.

An mcml object. When clusters is provided, macro\$data contains the cluster assignments and macro\$weights is NULL (the macro is the sequence of clusters, not a summary).

The input mcml object unchanged.

See Also

[build_mcml](#), [as_tna](#)

Examples

```
# From cluster_summary
mat <- matrix(c(0.5, 0.2, 0.3,
               0.1, 0.6, 0.3,
               0.4, 0.1, 0.5), 3, 3, byrow = TRUE,
             dimnames = list(c("A", "B", "C"), c("A", "B", "C")))
clusters <- list(G1 = c("A", "B"), G2 = c("C"))
cs <- cluster_summary(mat, clusters, type = "tna")
m <- as_mcml(cs)
m$macro$weights

## Not run:
# From group_tna with cluster assignments (requires tna + Nestimate)
seqs <- data.frame(T1 = c("A", "B", "A"), T2 = c("B", "A", "B"))
```

```

mod <- tna::tna(seqs)
cl <- Nestimate::cluster_data(mod, k = 2)
gt <- tna::group_model(cl)
m <- as_mcml(gt, clusters = cl$assignments)

## End(Not run)

```

as_tna

Convert cluster_summary to tna Objects

Description

Converts a `cluster_summary` object to proper `tna` objects that can be used with all functions from the `tna` package. Creates a macro (cluster-level) `tna` model and per-cluster `tna` models (internal transitions within each cluster), returned as a flat `group_tna` object.

Usage

```

as_tna(x)

## S3 method for class 'cluster_summary'
as_tna(x)

## S3 method for class 'mcml'
as_tna(x)

## Default S3 method:
as_tna(x)

```

Arguments

`x` A `cluster_summary` object created by `cluster_summary`. The `cluster_summary` should typically be created with `type = "tna"` to ensure row-normalized transition probabilities. If created with `type = "raw"`, the raw counts will be passed to `tna::tna()` which will normalize them.

Details

This is the final step in the MCML workflow, enabling full integration with the `tna` package for centrality analysis, bootstrap validation, permutation tests, and visualization.

Requirements:

The `tna` package must be installed. If not available, the function throws an error with installation instructions.

Workflow:

```

# Full MCML workflow
net <- cograph(edges, nodes = nodes)
net$nodes$clusters <- group_assignments
cs <- cluster_summary(net, type = "tna")
tna_models <- as_tna(cs)

# Now use tna package functions
plot(tna_models$macro)
tna::centralities(tna_models$macro)
tna::bootstrap(tna_models$macro, iter = 1000)

# Analyze per-cluster patterns
plot(tna_models$ClusterA)
tna::centralities(tna_models$ClusterA)

```

Excluded Clusters:

A per-cluster tna cannot be created when:

- The cluster has only 1 node (no internal transitions possible)
- Some nodes in the cluster have no outgoing edges (row sums to 0)

These clusters are silently excluded. The macro (cluster-level) model still includes all clusters.

Value

A `group_tna` object (S3 class) – a flat named list of tna objects. The first element is named "macro" and represents the cluster-level transitions. Subsequent elements are named by cluster name and represent internal transitions within each cluster.

macro A tna object representing cluster-level transitions. Contains `$weights` (k x k transition matrix), `$inits` (initial distribution), and `$labels` (cluster names). Use this for analyzing how learners/entities move between high-level groups or phases.

<cluster_name> Per-cluster tna objects, one per cluster. Each tna object represents internal transitions within that cluster. Contains `$weights` ($n_i \times n_i$ matrix), `$inits` (initial distribution), and `$labels` (node labels). Clusters with single nodes or zero-row nodes are excluded (tna requires positive row sums).

A `group_tna` object (flat list of tna objects: macro + per-cluster).

A `group_tna` object (flat list of tna objects: macro + per-cluster).

A tna object constructed from the input.

See Also

[cluster_summary](#) to create the input object, [plot_mcml](#) for visualization without conversion, `tna::tna` for the underlying tna constructor

Examples

```

# -----
# Basic usage

```

```

# -----
mat <- matrix(runif(36), 6, 6)
diag(mat) <- 0
rownames(mat) <- colnames(mat) <- LETTERS[1:6]

clusters <- list(
  G1 = c("A", "B"),
  G2 = c("C", "D"),
  G3 = c("E", "F")
)

cs <- cluster_summary(mat, clusters, type = "tna")
tna_models <- as_tna(cs)

# Print summary
tna_models

# -----
# Access components
# -----
# Macro (cluster-level) tna
tna_models$macro
tna_models$macro$weights # 3x3 transition matrix
tna_models$macro$inits   # Initial distribution
tna_models$macro$labels  # c("G1", "G2", "G3")

# Per-cluster tnas
names(tna_models)       # "macro", "G1", "G2", "G3"
tna_models$G1           # tna for cluster G1
tna_models$G1$weights   # 2x2 matrix (A, B)

# -----
# Plot via cograph's own renderer – avoids tna's plot dispatch,
# which pulls in `cluster` for layout and breaks
# `_R_CHECK_DEPENDS_ONLY_=TRUE` CRAN checks.
# -----
splot(tna_models$macro)
splot(tna_models$G1)

# -----
# Centrality analysis via tna (optional dependency)
# -----
## Not run:
if (requireNamespace("tna", quietly = TRUE)) {
  tna::centralities(tna_models$macro)
  tna::centralities(tna_models$G1)
  tna::centralities(tna_models$G2)
}

## End(Not run)

## Not run:
# Bootstrap requires a tna object built from raw sequence data (has $data)

```

```

# as_tna() returns weight-matrix-based tnas which don't satisfy that requirement
if (requireNamespace("tna", quietly = TRUE)) {
  boot <- tna::bootstrap(tna_models$macro, iter = 1000)
  summary(boot)
}

## End(Not run)

# -----
# Check which within-cluster models were created
# -----
cs <- cluster_summary(mat, clusters, type = "tna")
tna_models <- as_tna(cs)

# All cluster names
names(cs$cluster_members)

# Clusters with valid per-cluster models
setdiff(names(tna_models), "macro")

# Clusters excluded (single node or zero rows)
setdiff(names(cs$cluster_members), names(tna_models))

```

 build_mcml

Build MCML from Raw Transition Data

Description

Builds a Multi-Cluster Multi-Level (MCML) model from raw transition data (edge lists or sequences) by recoding node labels to cluster labels and counting actual transitions. Unlike [cluster_summary](#) which aggregates a pre-computed weight matrix, this function works from the original transition data to produce the TRUE Markov chain over cluster states.

Usage

```

build_mcml(
  x,
  clusters = NULL,
  method = c("sum", "mean", "median", "max", "min", "density", "geomean"),
  type = c("tna", "frequency", "cooccurrence", "semi_markov", "raw"),
  directed = TRUE,
  compute_within = TRUE
)

```

Arguments

x Input data. Accepts multiple formats:
data.frame with from/to columns Edge list. Columns named from/source/src/v1/node1/i and to/target/tgt/v2/node2/j are auto-detected. Optional weight column (weight/w/value/strength).

	<p>data.frame without from/to columns Sequence data. Each row is a sequence, columns are time steps. Consecutive pairs (t, t+1) become transitions.</p> <p>tna object If x\$data is non-NULL, uses sequence path on the raw data. Otherwise falls back to cluster_summary.</p> <p>cograph_network If x\$data is non-NULL, detects edge list vs sequence data. Otherwise falls back to cluster_summary.</p> <p>cluster_summary Returns as-is.</p> <p>square numeric matrix Falls back to cluster_summary.</p> <p>non-square or character matrix Treated as sequence data.</p>
clusters	<p>Cluster/group assignments. Accepts:</p> <p>named list Direct mapping. List names = cluster names, values = character vectors of node labels. Example: <code>list(A = c("N1", "N2"), B = c("N3", "N4"))</code></p> <p>data.frame A data frame where the first column contains node names and the second column contains group/cluster names. Example: <code>data.frame(node = c("N1", "N2", "N3"), group = c("A", "A", "B"))</code></p> <p>membership vector Character or numeric vector. Node names are extracted from the data. Example: <code>c("A", "A", "B", "B")</code></p> <p>column name string For edge list data.frames, the name of a column containing cluster labels. The mapping is built from unique (node, group) pairs in both from and to columns.</p> <p>NULL Auto-detect from <code>cograph_network\$nodes</code> or <code>\$node_groups</code> (same logic as cluster_summary).</p>
method	Aggregation method for combining edge weights: "sum", "mean", "median", "max", "min", "density", "geomean". Default "sum".
type	Post-processing: "tna" (row-normalize), "cooccurrence" (symmetrize), "semi_markov", or "raw". Default "tna".
directed	Logical. Treat as directed network? Default TRUE.
compute_within	Logical. Compute within-cluster matrices? Default TRUE.

Value

A `cluster_summary` object with `meta$source = "transitions"`, fully compatible with [plot_mcml](#), [as_tna](#), and [splot](#).

See Also

[cluster_summary](#) for matrix-based aggregation, [as_tna](#) to convert to tna objects, [plot_mcml](#) for visualization

Examples

```
# Edge list with clusters
edges <- data.frame(
  from = c("A", "A", "B", "C", "C", "D"),
  to   = c("B", "C", "A", "D", "D", "A"),
  weight = c(1, 2, 1, 3, 1, 2)
```

```

)
clusters <- list(G1 = c("A", "B"), G2 = c("C", "D"))
cs <- build_mcml(edges, clusters)
cs$macro$weights

# Sequence data with clusters
seqs <- data.frame(
  T1 = c("A", "C", "B"),
  T2 = c("B", "D", "A"),
  T3 = c("C", "C", "D"),
  T4 = c("D", "A", "C")
)
cs <- build_mcml(seqs, clusters, type = "raw")
cs$macro$weights

```

centrality

Calculate Network Centrality Measures

Description

Computes centrality measures for nodes in a network and returns a tidy data frame. Accepts matrices, igraph objects, cograph_network, or tna objects.

Usage

```

centrality(
  x,
  measures = "all",
  mode = "all",
  normalized = FALSE,
  weighted = TRUE,
  directed = NULL,
  loops = TRUE,
  simplify = "sum",
  digits = NULL,
  sort_by = NULL,
  cutoff = -1,
  invert_weights = NULL,
  alpha = 1,
  damping = 0.85,
  personalized = NULL,
  transitivity_type = "local",
  isolates = "nan",
  lambda = 1,
  k = 3,
  states = NULL,
  decay_parameter = 0.5,
  dmnc_epsilon = 1.7,

```

```

membership = NULL,
katz_alpha = 0.1,
hubbell_weight = 0.5,
...
)

```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object)
measures	Which measures to calculate. Default "all" calculates all available measures (87 total). Can be a character vector of measure names. Core (igraph-backed): "degree", "strength", "betweenness", "closeness", "eigenvector", "pagerank", "authority", "hub", "eccentricity", "coreness", "constraint", "transitivity", "harmonic", "alpha", "power", "subgraph". Native : "diffusion", "leverage", "kreach", "laplacian", "load", "current_flow_closeness", "current_flow_betweenness", "voterank", "percolation". Distance-based : "radiality", "lin", "decay", "residual_closeness", "dangalchev", "generalized_closeness", "harary", "average_distance", "barycenter", "wiener", "closeness_vitality". Spectral/walk : "communicability", "communicability_betweenness", "random_walk". Path-based : "stress", "flow_betweenness". Local/neighborhood : "lobby", "entropy", "semilocal", "clusterrank", "bottleneck", "centroid", "mnc", "dmnc", "lac", "topological_coefficient", "bridging", "local_bridging", "effective_size", "diversity", "cross_clique", "markov". Influence : "integration", "expected", "gilschmidt". Directed-only : "salsa", "leaderank", "trophic_level", "pairwisedis", "prestige_domain", "prestige_domain_proximity". Community-aware (require membership): "participation", "within_module_z", "gateway", "brokerage_coordinator", "brokerage_itinerant", "brokerage_representative", "brokerage_gatekeeper", "brokerage_liaison" (the last 5 also require a directed graph; see centrality_brokerage_coordinator). Zoo (batch 2) : "gravity", "collective_influence", "local_hindex", "hindex_strength", "onion", "second_order", "infection", "nonbacktracking", "spanning_tree". Classical (batch 3, reference-validated) : "katz" (Katz 1953), "hubbell" (Hubbell 1965), "information" (Stephenson-Zelen 1989), "reaching_local" (Mones et al. 2012). See centrality_katz , centrality_hubbell , centrality_information , centrality_pairwisedis , centrality_reaching_local .
mode	For directed networks: "all", "in", or "out". Affects degree, strength, closeness, eccentricity, coreness, and harmonic centrality.
normalized	Logical. Normalize values to 0-1 range by dividing by max. For closeness, this is passed directly to igraph (proper normalization).
weighted	Logical. Use edge weights if available. Default TRUE.
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.
loops	Logical. If TRUE (default), keep self-loops. Set to FALSE to remove them before calculation.
simplify	How to combine multiple edges between the same node pair. Options: "sum" (default), "mean", "max", "min", or FALSE/"none" to keep multiple edges.
digits	Integer or NULL. Round all numeric columns to this many decimal places. Default NULL (no rounding).

sort_by	Character or NULL. Column name to sort results by (descending order). Default NULL (original node order).
cutoff	Maximum path length to consider for betweenness and closeness. Default -1 (no limit). Set to a positive value for faster computation on large networks at the cost of accuracy.
invert_weights	Logical or NULL. For path-based measures (betweenness, closeness, harmonic, eccentricity, kreach), should weights be inverted so that higher weights mean shorter paths? Default NULL which auto-detects: TRUE for tna objects (transition probabilities), FALSE otherwise (matching igrph/sna). Set explicitly to TRUE for strength/frequency weights (qgraph style) or FALSE for distance/cost weights.
alpha	Numeric. Exponent for weight transformation when invert_weights = TRUE. Distance is computed as $1 / \text{weight}^{\alpha}$. Default 1. Higher values increase the influence of weight differences on path lengths.
damping	PageRank damping factor. Default 0.85. Must be between 0 and 1.
personalized	Named numeric vector for personalized PageRank. Default NULL (standard PageRank). Values should sum to 1.
transitivity_type	Type of transitivity to calculate: "local" (default), "global", "undirected", "localundirected", "barrat" (weighted), or "weighted".
isolates	How to handle isolate nodes in transitivity calculation: "nan" (default) returns NaN, "zero" returns 0.
lambda	Diffusion scaling factor for diffusion centrality. Default 1.
k	Path length parameter for geodesic k-path centrality. Default 3.
states	Named numeric vector of percolation states (0-1) for percolation centrality. Each value represents how "activated" or "infected" a node is. Default NULL (all nodes get state 1, equivalent to betweenness).
decay_parameter	Numeric. Decay parameter for decay and generalized closeness centrality. Default 0.5. Must be between 0 and 1.
dmnc_epsilon	Numeric. Epsilon exponent for DMNC (Density of Maximum Neighborhood Component). Default 1.7 as recommended by Lin et al. (2008). centiserve uses 1.67 (four-community assumption). Must be between 1 and 2.
membership	Integer vector of community assignments (one per node) for community-aware measures: participation, within_module_z, gateway. Default NULL. Required when requesting these measures.
katz_alpha	Attenuation factor for Katz centrality. Must satisfy $\alpha < 1/\rho(A)$. Default 0.1 (matches centiserve and NetworkX conventions). Only used when "katz" is in measures.
hubbell_weight	Weight factor w for Hubbell centrality. Must satisfy $w \cdot \rho(W) \leq 1$ for solvability. Default 0.5. Only used when "hubbell" is in measures.
...	Additional arguments (currently unused)

Details

The following centrality measures are available:

degree Count of edges (supports mode: in/out/all)

strength Weighted degree (supports mode: in/out/all)

betweenness Shortest path centrality

closeness Inverse distance centrality (supports mode: in/out/all)

eigenvector Influence-based centrality

pagerank Random walk centrality (supports damping and personalization)

authority HITS authority score

hub HITS hub score

eccentricity Maximum distance to other nodes (supports mode)

coreness K-core membership (supports mode: in/out/all)

constraint Burt's constraint (structural holes)

transitivity Local clustering coefficient (supports multiple types)

harmonic Harmonic centrality - handles disconnected graphs better than closeness (supports mode: in/out/all)

diffusion Diffusion degree centrality - sum of scaled degrees of node and its neighbors (supports mode: in/out/all, lambda scaling)

leverage Leverage centrality - measures influence over neighbors based on relative degree differences (supports mode: in/out/all)

kreach Geodesic k-path centrality - count of nodes reachable within distance k (supports mode: in/out/all, k parameter)

alpha Alpha/Katz centrality - influence via paths, penalized by distance. Similar to eigenvector but includes exogenous contribution

power Bonacich power centrality - measures influence based on connections to other influential nodes

subgraph Subgraph centrality - participation in closed loops/walks, weighting shorter loops more heavily

laplacian Laplacian centrality using Qi et al. (2012) local formula. Matches NetworkX and `centiserve::laplacian()`

load Load centrality - fraction of all shortest paths through node, similar to betweenness but weights paths by $1/\text{count}$

current_flow_closeness Information centrality - closeness based on electrical current flow (requires connected graph)

current_flow_betweenness Random walk betweenness - betweenness based on current flow rather than shortest paths (requires connected graph)

voterank VoteRank - identifies influential spreaders via iterative voting mechanism. Returns normalized rank (1 = most influential)

percolation Percolation centrality - importance for spreading processes. Uses node states (0-1) to weight paths. When all states equal, equivalent to betweenness. Useful for epidemic/information spreading analysis.

radiality Radiality centrality (centiserve). Sum of $(\text{diam} + 1 - d)$ normalized by $n-1$.

lin Lin's centrality. Reachable nodes squared divided by sum of distances.

decay Decay centrality. Sum of δ^d for parameter δ .

residual_closeness Residual closeness. Sum of $1/2^d$.

dangalchev Dangalchev closeness (alias for residual closeness).

generalized_closeness Generalized closeness. Sum of α^d .

harary Harary centrality. Sum of $1/d^2$ for all reachable pairs.

average_distance Average distance (centiserve). Sum of distances / $(n+1)$.

barycenter Barycenter centrality. $1 / \text{sum of distances}$.

wiener Wiener index. Total sum of shortest path distances from node.

closeness_vitality Closeness vitality. Drop in Wiener index when node removed.

communicability Total communicability. Row sums of matrix exponential.

communicability_betweenness Communicability betweenness. Fraction of communicability through each node.

random_walk Random walk centrality. Inverse sum of random walk distances (requires connected graph).

stress Stress centrality. Number of shortest paths through node.

flow_betweenness Flow betweenness. Max-flow based betweenness.

lobby Lobby index (h-index of neighborhood).

entropy Graph entropy centrality. Entropy change on node removal.

semilocal Semi-local centrality. Triple-nested neighborhood sum.

clusterrank ClusterRank. Clustering coefficient times neighbor degree sum.

bottleneck Bottleneck centrality. Count of shortest path trees where node is critical.

centroid Centroid value. Minimum $f(v,i)$ across all nodes.

mnc Maximum Neighborhood Component size.

dmnc Density of Maximum Neighborhood Component.

topological_coefficient Topological coefficient. Shared neighbor ratio.

bridging Bridging centrality. Betweenness times bridging coefficient.

local_bridging Local bridging. $(1/\text{degree})$ times bridging coefficient.

effective_size Burt's effective size. Degree minus redundancy.

diversity Diversity centrality. Shannon entropy of edge weight distribution.

cross_clique Cross-clique connectivity. Count of cliques containing node.

markov Markov centrality. Inverse mean first passage time (requires connected graph).

integration Integration centrality. Distance-based influence.

expected Expected centrality. Sum of neighbor degrees.

gilschmidt Gil-Schmidt power index. Sum of $1/d$ normalized by $n-1$.

salsa SALSA authority scores (directed graphs only).

leaderrank LeaderRank. PageRank with ground node (directed graphs only).

participation Participation coefficient. Diversity of inter-community connections (requires membership).

within_module_z Within-module degree z-score. Intra-community connectivity (requires membership).

gateway Gateway coefficient. Inter-community brokerage weighted by centrality (requires membership).

Value

A data frame with columns:

- node: Node labels/names
- One column per measure, with mode suffix for directional measures (e.g., degree_in, closeness_all)

Examples

```
# Basic usage with matrix
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality(adj)

# Specific measures
centrality(adj, measures = c("degree", "betweenness"))

# Directed network with normalization
centrality(adj, mode = "in", normalized = TRUE)

# Sort by pagerank
centrality(adj, sort_by = "pagerank", digits = 3)

# PageRank with custom damping
centrality(adj, measures = "pagerank", damping = 0.9)

# Harmonic centrality (better for disconnected graphs)
centrality(adj, measures = "harmonic")

# Global transitivity
centrality(adj, measures = "transitivity", transitivity_type = "global")
```

centrality_alpha	<i>Alpha (Katz) Centrality</i>
------------------	--------------------------------

Description

Influence via all paths penalized by distance. Similar to eigenvector centrality but includes an exogenous contribution, making it well-defined even for directed acyclic graphs.

Usage

```
centrality_alpha(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of alpha centrality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_eigenvector](#) for a related measure.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_alpha(adj)
```

centrality_authority *HITS Authority and Hub Scores*

Description

Kleinberg's HITS algorithm. `centrality_authority` scores nodes pointed to by good hubs. `centrality_hub` scores nodes that point to good authorities.

Usage

```
centrality_authority(x, ...)
```

```
centrality_hub(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`...` Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of authority or hub scores.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 0, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_authority(adj)
centrality_hub(adj)
```

centrality_average_distance
Average Distance Centrality

Description

Sum of shortest path distances divided by $(n + 1)$. Lower values indicate more central nodes.

Usage

```
centrality_average_distance(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of average distance values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_average_distance(adj)
```

centrality_barycenter *Barycenter Centrality*

Description

Inverse of the total distance to all reachable nodes.

Usage

```
centrality_barycenter(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of barycenter centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_barycenter(adj)
```

centrality_betweenness

Betweenness Centrality

Description

Fraction of shortest paths passing through each node. Nodes with high betweenness act as bridges connecting different parts of the network.

Usage

```
centrality_betweenness(x, ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed, cutoff, invert_weights).

Value

Named numeric vector of betweenness values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_load](#) for a related measure.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_betweenness(adj)
```

centrality_bottleneck *Bottleneck Centrality*

Description

Number of shortest path trees where the node appears in more than $n/4$ paths.

Usage

```
centrality_bottleneck(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named integer vector of bottleneck centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0), 4, 4)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_bottleneck(adj)
```

centrality_bridging *Bridging Centrality*

Description

Product of betweenness and bridging coefficient. Identifies nodes that bridge communities.

Usage

```
centrality_bridging(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`...` Additional arguments passed to [centrality](#).

Value

Named numeric vector of bridging centrality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_local_bridging](#) for the local variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_bridging(adj)
```

centrality_brokerage_coordinator
Gould-Fernandez Brokerage — Coordinator Role

Description

Coordinator brokerage (w_I): count of open directed 2-paths $A \rightarrow V \rightarrow A$ passing through node V , where all three nodes belong to V 's group. The broker mediates contact between two in-group members.

Usage

```
centrality_brokerage_coordinator(x, membership = NULL, ...)
```

Arguments

<code>x</code>	Directed network input (matrix, igraph, cograph_network, tna object).
<code>membership</code>	Integer or character vector of group assignments, length equal to the number of nodes. Required.
<code>...</code>	Additional arguments passed to centrality .

Details

Bit-exact match against `sna::brokerage$raw.nli[, "w_I"]`. Counts OPEN 2-paths only — those where no direct edge from *a* to *c* exists. Directed-only; returns NA with a warning on undirected input.

Value

Named integer vector of coordinator role counts.

References

Gould, R. V., & Fernandez, R. M. (1989). Structures of mediation: A formal approach to brokerage in transaction networks. *Sociological Methodology*, 19, 89-126.

See Also

[centrality](#), [centrality_brokerage_itinerant](#), [centrality_brokerage_representative](#), [centrality_brokerage_g](#), [centrality_brokerage_liaison](#).

Examples

```
adj <- matrix(c(0,1,1,0, 0,0,1,1, 0,0,0,1, 1,0,0,0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_brokerage_coordinator(adj, membership = c(1, 1, 2, 2))
```

centrality_brokerage_gatekeeper

Gould-Fernandez Brokerage — Gatekeeper Role

Description

Gatekeeper brokerage (`b_OI`): count of open directed 2-paths $A \rightarrow V \rightarrow B$ where *V* and *B* are in the same group and *A* is in a different group. The broker acts as a gate letting in-group members receive contact from outside.

Usage

```
centrality_brokerage_gatekeeper(x, membership = NULL, ...)
```

Arguments

x	Directed network input (matrix, igraph, cograph_network, tna object).
membership	Integer or character vector of group assignments, length equal to the number of nodes. Required.
...	Additional arguments passed to centrality .

Details

Bit-exact match against `sna::brokerage$raw.nli[, "b_OI"]`. Directed-only.

Value

Named integer vector of gatekeeper role counts.

References

Gould & Fernandez (1989).

See Also

[centrality_brokerage_coordinator](#).

Examples

```
adj <- matrix(c(0,1,1,0, 0,0,1,1, 0,0,0,1, 1,0,0,0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_brokerage_gatekeeper(adj, membership = c(1, 1, 2, 2))
```

centrality_brokerage_itinerant

Gould-Fernandez Brokerage — Itinerant (Consultant) Role

Description

Itinerant brokerage (`w_O`): count of open directed 2-paths $A \rightarrow V \rightarrow A$ where the two endpoints are in the same group but the broker V is in a different group. The broker mediates within another group as an outsider.

Usage

```
centrality_brokerage_itinerant(x, membership = NULL, ...)
```

Arguments

x	Directed network input (matrix, igraph, cograph_network, tna object).
membership	Integer or character vector of group assignments, length equal to the number of nodes. Required.
...	Additional arguments passed to centrality .

Details

Bit-exact match against `sna::brokerage$raw.nli[, "w_0"]`. Directed-only.

Value

Named integer vector of itinerant role counts.

References

Gould & Fernandez (1989).

See Also

[centrality_brokerage_coordinator](#).

Examples

```
adj <- matrix(c(0,1,1,0, 0,0,1,1, 0,0,0,1, 1,0,0,0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_brokerage_itinerant(adj, membership = c(1, 1, 2, 2))
```

centrality_brokerage_liaison

Gould-Fernandez Brokerage — Liaison Role

Description

Liaison brokerage (`b_0`): count of open directed 2-paths $A \rightarrow V \rightarrow B$ where all three nodes belong to different groups. The broker mediates between two groups to neither of which they belong.

Usage

```
centrality_brokerage_liaison(x, membership = NULL, ...)
```

Arguments

<code>x</code>	Directed network input (matrix, igraph, cograph_network, tna object).
<code>membership</code>	Integer or character vector of group assignments, length equal to the number of nodes. Required.
<code>...</code>	Additional arguments passed to centrality .

Details

Bit-exact match against `sna::brokerage$raw.nli[, "b_0"]`. Directed-only.

Value

Named integer vector of liaison role counts.

References

Gould & Fernandez (1989).

See Also

[centrality_brokerage_coordinator](#).

Examples

```
adj <- matrix(c(0,1,1,0, 0,0,1,1, 0,0,0,1, 1,0,0,0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_brokerage_liaison(adj, membership = c(1, 1, 2, 2))
```

centrality_brokerage_representative

Gould-Fernandez Brokerage — Representative Role

Description

Representative brokerage (b_IO): count of open directed 2-paths $A \rightarrow V \rightarrow B$ where A and V are in the same group and B is in a different group. The broker represents their group outward.

Usage

```
centrality_brokerage_representative(x, membership = NULL, ...)
```

Arguments

x	Directed network input (matrix, igraph, cograph_network, tna object).
membership	Integer or character vector of group assignments, length equal to the number of nodes. Required.
...	Additional arguments passed to centrality .

Details

Bit-exact match against `sna::brokerage$raw.nli[, "b_IO"]`. Directed-only.

Value

Named integer vector of representative role counts.

References

Gould & Fernandez (1989).

See Also

[centrality_brokerage_coordinator](#).

Examples

```
adj <- matrix(c(0,1,1,0, 0,0,1,1, 0,0,0,1, 1,0,0,0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_brokerage_representative(adj, membership = c(1, 1, 2, 2))
```

centrality_centroid *Centroid Value*

Description

Minimum difference between own and competitor's closer-node count. Measures how much a node is at the center of the graph.

Usage

```
centrality_centroid(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of centroid values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0), 4, 4)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_centroid(adj)
```

centrality_closeness *Closeness Centrality*

Description

Inverse of the average shortest path distance from a node to all others. For directed networks, `centrality_incloseness` and `centrality_outcloseness` measure incoming and outgoing closeness.

Usage

```
centrality_closeness(x, mode = "all", ...)  
centrality_incloseness(x, ...)  
centrality_outcloseness(x, ...)
```

Arguments

<code>x</code>	Network input (matrix, igraph, network, cograph_network, tna object).
<code>mode</code>	For directed networks: "all" (default), "in", or "out".
<code>...</code>	Additional arguments passed to <code>centrality</code> (e.g., normalized, weighted, directed).

Value

Named numeric vector of closeness values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_harmonic](#) for a variant that handles disconnected graphs.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)  
rownames(adj) <- colnames(adj) <- c("A", "B", "C")  
centrality_closeness(adj)
```

centrality_closeness_vitality
Closeness Vitality

Description

Drop in the Wiener index when a node is removed. Higher values indicate more critical nodes for overall connectivity.

Usage

```
centrality_closeness_vitality(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of closeness vitality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_closeness_vitality(adj)
```

centrality_clusterrank
ClusterRank Centrality

Description

Product of clustering coefficient and sum of (neighbor degree + 1).

Usage

```
centrality_clusterrank(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of ClusterRank values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_clusterrank(adj)
```

centrality_communicability

Communicability Centrality

Description

Total communicability: row sums of the matrix exponential of the adjacency matrix. Measures a node's ability to broadcast information through all paths.

Usage

```
centrality_communicability(x, ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
...	Additional arguments passed to centrality .

Value

Named numeric vector of communicability values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_subgraph](#) for the diagonal-only variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_communicability(adj)
```

centrality_communicability_betweenness

Communicability Betweenness Centrality

Description

Fraction of total communicability that passes through each node.

Usage

```
centrality_communicability_betweenness(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
... Additional arguments passed to [centrality](#).

Value

Named numeric vector of communicability betweenness values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_communicability_betweenness(adj)
```

centrality_constraint *Burt's Constraint*

Description

Network constraint measuring the extent to which a node's connections are redundant. Low constraint indicates access to structural holes (brokerage opportunities).

Usage

```
centrality_constraint(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`...` Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of constraint values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_constraint(adj)
```

centrality_coreness *K-Core Decomposition (Coreness)*

Description

Assigns each node to its maximum k-core. A k-core is a maximal subgraph where every node has at least k connections within the subgraph.

Usage

```
centrality_coreness(x, mode = "all", ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
mode For directed networks: "all" (default), "in", or "out".
... Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of coreness values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_coreness(adj)
```

centrality_cross_clique

Cross-Clique Connectivity

Description

Count of all cliques (not just maximal) containing each node. Measures embeddedness in dense substructures.

Usage

```
centrality_cross_clique(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
... Additional arguments passed to [centrality](#).

Value

Named integer vector of cross-clique counts.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_cross_clique(adj)
```

centrality_current_flow_betweenness

Current Flow Betweenness Centrality

Description

Betweenness based on electrical current flow rather than shortest paths. Uses the Laplacian pseudoinverse. Requires a connected graph.

Usage

```
centrality_current_flow_betweenness(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
... Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of current flow betweenness values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_betweenness](#) for the shortest-path variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_current_flow_betweenness(adj)
```

`centrality_current_flow_closeness`*Current Flow Closeness Centrality*

Description

Information centrality based on electrical current flow through the network. Uses the pseudoinverse of the Laplacian matrix. Requires a connected graph.

Usage

```
centrality_current_flow_closeness(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`...` Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of current flow closeness values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_closeness](#) for the shortest-path variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_current_flow_closeness(adj)
```

`centrality_dangalchev` *Dangalchev Closeness Centrality*

Description

Alias for residual closeness centrality: sum of $1/2^d$.

Usage

```
centrality_dangalchev(x, mode = "all", ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
 mode For directed networks: "all" (default), "in", or "out".
 ... Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of Dangalchev closeness values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_residual_closeness](#) (equivalent).

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_dangalchev(adj)
```

centrality_decay *Decay Centrality*

Description

Sum of δ^d over all nodes, where d is the shortest path distance. Nodes near many others get higher scores. The `decay_parameter` controls the distance penalty.

Usage

```
centrality_decay(x, mode = "all", decay_parameter = 0.5, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
 mode For directed networks: "all" (default), "in", or "out".
 decay_parameter Numeric between 0 and 1. Default 0.5.
 ... Additional arguments passed to [centrality](#).

Value

Named numeric vector of decay centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_decay(adj, decay_parameter = 0.5)
```

centrality_degree	<i>Degree Centrality</i>
-------------------	--------------------------

Description

Number of edges connected to each node. For directed networks, `centrality_indegree` counts incoming edges and `centrality_outdegree` counts outgoing edges.

Usage

```
centrality_degree(x, mode = "all", ...)
```

```
centrality_indegree(x, ...)
```

```
centrality_outdegree(x, ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of degree values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_strength](#) for the weighted version.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_degree(adj)
```

centrality_diffusion *Diffusion Centrality*

Description

Sum of scaled degrees of a node and its neighbors, measuring the node's potential for spreading information through the network.

Usage

```
centrality_diffusion(x, mode = "all", lambda = 1, ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
lambda	Scaling factor for neighbor contributions. Default 1.
...	Additional arguments passed to centrality (e.g., weighted, directed).

Value

Named numeric vector of diffusion centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_diffusion(adj)
```

centrality_diversity *Diversity Centrality*

Description

Shannon entropy of the edge weight distribution per node. Measures how evenly a node distributes its connections.

Usage

```
centrality_diversity(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
 ... Additional arguments passed to [centrality](#).

Value

Named numeric vector of diversity centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
mat <- matrix(c(0, .5, .3, .5, 0, .8, .3, .8, 0), 3, 3)
rownames(mat) <- colnames(mat) <- c("A", "B", "C")
centrality_diversity(mat)
```

centrality_dmnc	<i>Density of Maximum Neighborhood Component (DMNC)</i>
-----------------	---

Description

Edge count divided by max component size^{1.5} in the neighborhood subgraph.

Usage

```
centrality_dmnc(x, mode = "all", dmnc_epsilon = 1.7, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
 mode For directed networks: "all" (default), "in", or "out".
 dmnc_epsilon Numeric. Epsilon exponent for DMNC. Default 1.7 as recommended by Lin et al. (2008). centiserve uses 1.67 (four-community assumption). Must be between 1 and 2.
 ... Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of DMNC values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_mnc](#) for the size-only variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_dmnc(adj)
```

centrality_eccentricity

Eccentricity

Description

Maximum shortest path distance from a node to any other node. For directed networks, `centrality_ineccentricity` and `centrality_outeccentricity` use incoming and outgoing paths.

Usage

```
centrality_eccentricity(x, mode = "all", ...)
```

```
centrality_ineccentricity(x, ...)
```

```
centrality_outeccentricity(x, ...)
```

Arguments

<code>x</code>	Network input (matrix, igraph, network, cograph_network, tna object).
<code>mode</code>	For directed networks: "all" (default), "in", or "out".
<code>...</code>	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of eccentricity values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_eccentricity(adj)
```

`centrality_effective_size`*Effective Size (Burt's)*

Description

Network effective size: degree minus redundancy. Measures non-redundant contacts in ego network.

Usage

```
centrality_effective_size(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`...` Additional arguments passed to [centrality](#).

Value

Named numeric vector of effective size values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_constraint](#) for a related structural holes measure.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_effective_size(adj)
```

`centrality_eigenvector`*Eigenvector Centrality*

Description

Influence-based centrality where a node's score depends on the scores of its neighbors. Nodes connected to other high-scoring nodes get higher scores.

Usage

```
centrality_eigenvector(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
... Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of eigenvector centrality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_pagerank](#) for a random walk variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_eigenvector(adj)
```

centrality_entropy *Entropy Centrality*

Description

Graph-theoretic entropy based on shortest path distribution in the residual graph after removing the node.

Usage

```
centrality_entropy(x, mode = "all", ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
mode For directed networks: "all" (default), "in", or "out".
... Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of entropy centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_entropy(adj)
```

centrality_expected	<i>Expected Centrality</i>
---------------------	----------------------------

Description

Sum of neighbor degrees. Simple but effective influence proxy.

Usage

```
centrality_expected(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of expected centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_expected(adj)
```

centrality_flow_betweenness

Flow Betweenness Centrality

Description

Max-flow based betweenness centrality.

Usage

```
centrality_flow_betweenness(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
... Additional arguments passed to [centrality](#).

Value

Named numeric vector of flow betweenness values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_betweenness](#) for shortest-path variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_flow_betweenness(adj)
```

centrality_gateway

Gateway Coefficient

Description

Inter-community brokerage weighted by centrality. Combines participation with degree information. Requires community membership.

Usage

```
centrality_gateway(x, membership = NULL, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
membership	Integer vector of community assignments (one per node).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality .

Value

Named numeric vector of gateway coefficient values (0-1).

See Also

[centrality](#) for computing multiple measures at once, [centrality_participation](#) for the simpler participation coefficient.

Examples

```
adj <- matrix(c(0,1,1,0,0, 1,0,1,0,0, 1,1,0,1,0, 0,0,1,0,1, 0,0,0,1,0), 5, 5)
rownames(adj) <- colnames(adj) <- LETTERS[1:5]
centrality_gateway(adj, membership = c(1, 1, 1, 2, 2))
```

centrality_generalized_closeness
Generalized Closeness Centrality

Description

Sum of α^d over all nodes. Generalization of decay centrality matching tidygraph's implementation.

Usage

```
centrality_generalized_closeness(x, mode = "all", decay_parameter = 0.5, ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
decay_parameter	Numeric between 0 and 1 (the alpha parameter). Default 0.5.
...	Additional arguments passed to centrality .

Value

Named numeric vector of generalized closeness values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_decay](#) (equivalent formulation).

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_generalized_closeness(adj)
```

centrality_giltschmidt *Gil-Schmidt Power Index*

Description

Sum of $1/d(v,w)$ normalized by $(n-1)$. Variant of closeness using harmonic mean of distances.

Usage

```
centrality_giltschmidt(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of Gil-Schmidt power index values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_harmonic](#) for a related measure.

Examples

```
adj <- matrix(c(0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0), 4, 4)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_giltschmidt(adj)
```

centrality_harary *Harary Centrality*

Description

Sum of $1/d^2$ over all reachable node pairs. Robust to disconnected graphs.

Usage

```
centrality_harary(x, mode = "all", ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
mode For directed networks: "all" (default), "in", or "out".
... Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of Harary centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)  
rownames(adj) <- colnames(adj) <- c("A", "B", "C")  
centrality_harary(adj)
```

centrality_harmonic *Harmonic Centrality*

Description

Sum of inverse shortest path distances to all other nodes. Unlike closeness, harmonic centrality handles disconnected graphs naturally (unreachable nodes contribute 0 instead of making the measure undefined).

Usage

```
centrality_harmonic(x, mode = "all", ...)
```

```
centrality_inharmonic(x, ...)
```

```
centrality_outharmonic(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`mode` For directed networks: "all" (default), "in", or "out".
`...` Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of harmonic centrality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_closeness](#) for the traditional variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_harmonic(adj)
```

centrality_hubbell *Hubbell Centrality*

Description

Hubbell (1965) input-output centrality: $C = (I - wW)^{-1}\mathbf{1}$, where W is the (weighted) adjacency matrix and w is a weight factor that must satisfy $w \cdot \rho(W) < 1$ for the system to be solvable.

Usage

```
centrality_hubbell(x, hubbell_weight = 0.5, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`hubbell_weight` Attenuation factor w . Default 0.5. If $w \cdot \rho(W) \geq 1$, the function returns NA with a warning.
`...` Additional arguments passed to [centrality](#).

Details

Bit-exact match against `centiserve::hubbell` when edge weights are passed explicitly (cograph mirrors `centiserve`'s full-inverse LAPACK call path).

Value

Named numeric vector of Hubbell centrality values (or NA if the system is not solvable).

Note on centiserve equivalence

`centiserve::hubbell(g, weights = NULL)` silently resets all edge weights to 1, ignoring the graph's weight attribute. To reproduce `cograph`'s values with `centiserve` on a weighted graph, pass `weights = igraph::E(g)$weight` explicitly.

References

Hubbell, C. H. (1965). An input-output approach to clique identification. *Sociometry*, 28(4), 377-399.

See Also

[centrality](#), [centrality_katz](#).

Examples

```
# Small weighted path graph; spectral radius permits weightfactor = 0.5
adj <- matrix(0, 4, 4)
adj[1,2] <- adj[2,1] <- adj[2,3] <- adj[3,2] <- adj[3,4] <- adj[4,3] <- 0.3
rownames(adj) <- colnames(adj) <- LETTERS[1:4]
centrality_hubbell(adj, hubbell_weight = 0.5)
```

centrality_information

Information Centrality (Stephenson-Zelen)

Description

Information centrality (Stephenson & Zelen 1989) measures a node's importance in terms of the "information" contained in all paths (not only shortest) passing through it. Defined via the inverse of a Laplacian-like matrix, yielding per-node $IC_i = 1/(C_{ii} + (\text{tr}(C) - 2R_i)/n)$ where $C = A^{-1}$ and R_i is the row sum of C .

Usage

```
centrality_information(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`...` Additional arguments passed to [centrality](#).

Details

Bit-exact match against `sna::infocent` on connected undirected graphs (`cograph` mirrors `sna`'s exact construction and call sequence).

Value

Named numeric vector of information centrality values.

References

Stephenson, K., & Zelen, M. (1989). Rethinking centrality: Methods and examples. *Social Networks*, 11(1), 1-37.

See Also

[centrality](#), [centrality_current_flow_closeness](#).

Examples

```
adj <- matrix(c(0,1,1,0, 1,0,1,1, 1,1,0,1, 0,1,1,0), 4, 4)
rownames(adj) <- colnames(adj) <- LETTERS[1:4]
centrality_information(adj)
```

centrality_integration

Integration Centrality

Description

Distance-based influence: sum of $1 - (d-1)/\max(d)$ over all nodes.

Usage

```
centrality_integration(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of integration centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_integration(adj)
```

centrality_katz	<i>Katz Centrality</i>
-----------------	------------------------

Description

Katz (1953) status index: $C = (I - \alpha A^T)^{-1} \mathbf{1}$. Each node's score sums attenuated walks of every length back to it, with attenuation α applied per step. Rankings are identical to Bonacich's alpha centrality with a uniform exogenous vector.

Usage

```
centrality_katz(x, katz_alpha = 0.1, ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
katz_alpha	Attenuation factor. Must satisfy $\alpha < 1/\rho(A)$ where $\rho(A)$ is the spectral radius. Default 0.1 matches centiserve and NetworkX conventions.
...	Additional arguments passed to centrality .

Details

Equivalence is verified bit-exact against `centiserve::katzcent` (cograph mirrors centiserve's exact LAPACK call sequence) and at machine epsilon against `igraph::alpha_centrality(exo = 1)` and `networkx.katz_centrality_numpy`.

Value

Named numeric vector of Katz centrality values.

References

Katz, L. (1953). A new status index derived from sociometric analysis. *Psychometrika*, 18(1), 39-43.

See Also

[centrality](#), [centrality_eigenvector](#), [centrality_pagerank](#).

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_katz(adj)
```

centrality_kreach *Geodesic K-Path Centrality*

Description

Count of nodes reachable within shortest path distance k . Measures how many nodes a given node can reach quickly.

Usage

```
centrality_kreach(x, mode = "all", k = 3, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`mode` For directed networks: "all" (default), "in", or "out".
`k` Maximum path length. Default 3.
`...` Additional arguments passed to [centrality](#) (e.g., weighted, directed, invert_weights).

Value

Named numeric vector of k -reach centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0), 4, 4)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_kreach(adj, k = 2)
```

centrality_lac *Local Average Connectivity (LAC)*

Description

Average degree of neighbors within the neighborhood subgraph. Measures how interconnected a node's neighbors are. Proposed by Li et al. (2011) for identifying essential proteins in PPI networks.

Usage

```
centrality_lac(x, mode = "all", ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
mode For directed networks: "all" (default), "in", or "out".
... Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of LAC values.

References

Li, M., Wang, J., Chen, X., Wang, H., & Pan, Y. (2011). A local average connectivity-based method for identifying essential proteins from the network level. *Computational Biology and Chemistry*, 35(3), 143-150.

See Also

[centrality](#) for computing multiple measures at once, [centrality_dmnc](#) for another neighborhood density measure.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_lac(adj)
```

centrality_laplacian *Laplacian Centrality*

Description

Energy drop from the graph Laplacian when a node is removed (Qi et al. 2012). Measures a node's importance to the overall network energy.

Usage

```
centrality_laplacian(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
... Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of Laplacian centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_laplacian(adj)
```

centrality_leaderrank *LeaderRank Centrality*

Description

PageRank variant with a ground node connected to all nodes. Requires a directed graph.

Usage

```
centrality_leaderrank(x, ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object). Must be directed.
...	Additional arguments passed to centrality .

Value

Named numeric vector of LeaderRank values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_pagerank](#) for standard PageRank.

Examples

```
## Not run:
adj <- matrix(c(0, 1, 0, 0, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_leaderrank(adj)

## End(Not run)
```

centrality_leverage *Leverage Centrality*

Description

Measures a node's influence over its neighbors based on relative degree differences. Positive values indicate the node has more connections than its average neighbor.

Usage

```
centrality_leverage(x, mode = "all", ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`mode` For directed networks: "all" (default), "in", or "out".
`...` Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of leverage centrality values (range -1 to 1).

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0), 4, 4)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_leverage(adj)
```

centrality_lin *Lin Centrality*

Description

Reachable nodes squared divided by sum of distances. Well-defined for disconnected graphs.

Usage

```
centrality_lin(x, mode = "all", ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
 mode For directed networks: "all" (default), "in", or "out".
 ... Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of Lin centrality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_closeness](#) for a related measure.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_lin(adj)
```

centrality_load	<i>Load Centrality</i>
-----------------	------------------------

Description

Fraction of all shortest paths passing through a node, similar to betweenness but weighting paths by $1/\text{count}$ (Goh et al. 2001).

Usage

```
centrality_load(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
 ... Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of load centrality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_betweenness](#) for the standard variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_load(adj)
```

centrality_lobby	<i>Lobby Index (H-Index of Neighborhood)</i>
------------------	--

Description

Largest k such that the node's closed neighborhood contains at least k nodes with degree $\geq k$.
Network analogue of the h-index.

Usage

```
centrality_lobby(x, mode = "all", ...)
```

Arguments

<code>x</code>	Network input (matrix, igraph, network, cograph_network, tna object).
<code>mode</code>	For directed networks: "all" (default), "in", or "out".
<code>...</code>	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named integer vector of lobby index values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_lobby(adj)
```

centrality_local_bridging
Local Bridging Centrality

Description

(1/degree) times bridging coefficient. Local measure of inter-community connectivity.

Usage

```
centrality_local_bridging(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
... Additional arguments passed to [centrality](#).

Value

Named numeric vector of local bridging values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_bridging](#) for the betweenness-weighted variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_local_bridging(adj)
```

centrality_markov *Markov Centrality*

Description

Inverse of column means of the mean first passage time matrix. Requires a connected graph.

Usage

```
centrality_markov(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
... Additional arguments passed to [centrality](#).

Value

Named numeric vector of Markov centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_markov(adj)
```

centrality_mnc

Maximum Neighborhood Component (MNC)

Description

Size of the largest connected component in the node's neighborhood subgraph.

Usage

```
centrality_mnc(x, mode = "all", ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
mode For directed networks: "all" (default), "in", or "out".
... Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named integer vector of MNC values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_dmnc](#) for the density variant.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_mnc(adj)
```

centrality_pagerank *PageRank Centrality*

Description

Random walk centrality measuring node importance. Simulates a random walker that follows edges with probability damping and jumps to a random node with probability $1 - \text{damping}$.

Usage

```
centrality_pagerank(x, damping = 0.85, personalized = NULL, ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
damping	Damping factor (probability of following an edge). Default 0.85.
personalized	Named numeric vector for personalized PageRank. Values should sum to 1. Default NULL (uniform).
...	Additional arguments passed to centrality (e.g., weighted, directed).

Value

Named numeric vector of PageRank values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_eigenvector](#) for a related measure.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_pagerank(adj)
centrality_pagerank(adj, damping = 0.9)
```

centrality_pairwisedis

Pairwise Disconnectivity (Potapov et al. 2008)

Description

For a directed network, `pairwisedis(v)` is the fraction of ordered reachable pairs (s, t) that become unreachable when node v is removed:

$$PD(v) = (|P(G)| - |P(G - v)|) / |P(G)|$$

where $|P(G)|$ is the number of ordered pairs (s, t) , $s \neq t$ with a directed path from s to t .

Usage

```
centrality_pairwisedis(x, ...)
```

Arguments

`x` Directed network input (matrix, igraph, cograph_network, tna object).
`...` Additional arguments passed to [centrality](#).

Details

Bit-exact match against `centiserve::pairwisedis` on directed graphs. Requires the input to be directed; returns NA with a warning on undirected inputs.

Value

Named numeric vector of pairwise disconnectivity values in $[0, 1]$.

References

Potapov, A. P., Voss, N., Sasse, N., & Wingender, E. (2008). Topology of mammalian transcription networks. *Genome Informatics*, 18, 193-204.

See Also

[centrality](#), [robustness](#).

Examples

```
adj <- matrix(c(0,1,0, 0,0,1, 1,0,0), 3, 3, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_pairwisedis(adj)
```

centrality_participation

Participation Coefficient

Description

Measures diversity of inter-community connections. Nodes connecting to many communities have high participation. Requires community membership.

Usage

```
centrality_participation(x, membership = NULL, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
membership	Integer vector of community assignments (one per node).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality .

Value

Named numeric vector of participation coefficient values (0-1).

See Also

[centrality](#) for computing multiple measures at once, [centrality_within_module_z](#) for within-community connectivity.

Examples

```
adj <- matrix(c(0,1,1,0,0, 1,0,1,0,0, 1,1,0,1,0, 0,0,1,0,1, 0,0,0,1,0), 5, 5)
rownames(adj) <- colnames(adj) <- LETTERS[1:5]
centrality_participation(adj, membership = c(1, 1, 1, 2, 2))
```

centrality_percolation

Percolation Centrality

Description

Importance for spreading processes using node states. Each node has a state (0-1) representing how activated it is. When all states are equal, equivalent to betweenness.

Usage

```
centrality_percolation(x, states = NULL, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`states` Named numeric vector of node states (0-1). Default NULL (all nodes get state 1).
`...` Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of percolation centrality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_betweenness](#) which this generalizes.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_percolation(adj)
centrality_percolation(adj, states = c(A = 0.8, B = 0.2, C = 0.5))
```

centrality_power

Bonacich Power Centrality

Description

Measures influence based on connections to other influential nodes. The power parameter controls whether connections to well-connected nodes increase or decrease centrality.

Usage

```
centrality_power(x, mode = "all", ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`mode` For directed networks: "all" (default), "in", or "out".
`...` Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of power centrality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_eigenvector](#) for a related measure.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_power(adj)
```

centrality_prestige_domain

Domain Prestige

Description

Directed-graph prestige measure: for each node v , the number of other nodes that can reach v via a directed path.

$$\text{domain}(v) = |\{u \neq v : u \rightarrow^* v\}|$$

Usage

```
centrality_prestige_domain(x, ...)
```

Arguments

x Directed network input (matrix, igraph, cograph_network, tna object).
 \dots Additional arguments passed to [centrality](#).

Details

Bit-exact match against `sna::prestige(cmode = "domain")`. Directed-only; returns NA with a warning on undirected input.

Value

Named numeric vector of domain prestige values in $\{0, 1, \dots, N - 1\}$.

References

Wasserman, S., & Faust, K. (1994). *Social Network Analysis: Methods and Applications*. Cambridge University Press.

See Also

[centrality](#), [centrality_reaching_local](#) for the dual "out-reachability" measure, [centrality_pairwisedis](#) for a related reachability-based directed measure.

Examples

```
# Directed 3-cycle: every node reaches every other node
adj <- matrix(c(0,1,0, 0,0,1, 1,0,0), 3, 3, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_prestige_domain(adj)
```

centrality_prestige_domain_proximity
Domain Proximity Prestige

Description

Distance-weighted variant of domain prestige. For each directed node v :

$$PD(v) = R_v^2 / (D_v \cdot (n - 1))$$

where R_v is the number of other nodes that reach v , and D_v is the sum of geodesic distances from those reachers to v . A node that is reachable quickly from many others scores high; unreachable nodes score 0.

Usage

```
centrality_prestige_domain_proximity(x, ...)
```

Arguments

x Directed network input (matrix, igraph, cograph_network, tna object).
 \dots Additional arguments passed to [centrality](#).

Details

Bit-exact match against `sna::prestige(cmode = "domain.proximity")` on strongly connected directed graphs. Directed-only; returns NA with a warning on undirected input.

Value

Named numeric vector of domain proximity prestige values in $[0, 1]$.

Divergence from sna on disconnected graphs

`sna`'s formula computes $(\text{counts} > 0) * \text{gdist}$ element-wise and then sums to get the denominator. For any pair where $\text{gdist} = \text{Inf}$ (unreachable), R evaluates $\text{FALSE} * \text{Inf} = \text{NaN}$, so the entire denominator becomes NaN and `sna` zeros every node via `p[is.nan(p)] <- 0`. `cograph` masks with `is.finite()` before summing, producing mathematically correct values on any directed graph, including those with disconnected components.

References

Wasserman, S., & Faust, K. (1994). *Social Network Analysis: Methods and Applications*. Cambridge University Press.

See Also

[centrality](#), [centrality_prestige_domain](#) for the unweighted count, [centrality_reaching_local](#) for the dual out-reachability measure.

Examples

```
# Directed 3-cycle: each node is reached by both others at distance 1 and 2
adj <- matrix(c(0,1,0, 0,0,1, 1,0,0), 3, 3, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_prestige_domain_proximity(adj)
```

centrality_radiality *Radiality Centrality*

Description

Centrality based on sum of (diameter + 1 - distance) normalized by n-1. Nodes closer to others (on average) have higher radiality.

Usage

```
centrality_radiality(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of radiality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_closeness](#) for a related measure.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_radiality(adj)
```

`centrality_random_walk`*Random Walk Centrality*

Description

Inverse sum of random walk distances. Requires a connected graph.

Usage

```
centrality_random_walk(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`...` Additional arguments passed to [centrality](#).

Value

Named numeric vector of random walk centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_random_walk(adj)
```

`centrality_reaching_local`*Local Reaching Centrality (Mones, Vicsek & Vicsek 2012)*

Description

Local reaching centrality measures how much of the network is reachable from a node.

Usage

```
centrality_reaching_local(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality .

Details

- Directed unweighted: $LRC(v) = |\{u : u \neq v, v \rightarrow u\}| / (N - 1)$.
- Undirected unweighted: average of $1/d(v, u)$ over all $u \neq v$, divided by $N - 1$. Numerically equal to `igraph::harmonic_centrality(normalized = TRUE)`.
- Weighted: NetworkX convention, where edge weights are interpreted as strengths and path length is $\sum_e (\text{total_weight} / w_e)$. Per-path score is the mean of original edge weights along the shortest path.

Bit-exact match against `networkx.local_reaching_centrality` across all three branches. Bit-exact match against `igraph::harmonic_centrality(normalized = TRUE)` for the undirected unweighted branch. See [reaching_global](#) for the graph-level hierarchy measure derived from per-node LRC.

Value

Named numeric vector of local reaching centrality values.

References

Mones, E., Vicsek, L., & Vicsek, T. (2012). Hierarchy measure for complex networks. *PLoS ONE*, 7(3), e33799.

See Also

[centrality](#), [centrality_harmonic](#), [reaching_global](#).

Examples

```
# Directed path A -> B -> C
adj <- matrix(c(0,1,0, 0,0,1, 0,0,0), 3, 3, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_reaching_local(adj, mode = "out")
```

centrality_residual_closeness
Residual Closeness Centrality

Description

Sum of $1/2^d$ for all nodes, including self. Robust to disconnected graphs.

Usage

```
centrality_residual_closeness(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of residual closeness values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_dangalchev](#) (alias).

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_residual_closeness(adj)
```

centrality_salsa *SALSA Authority Centrality*

Description

Stochastic Approach for Link-Structure Analysis. Returns authority scores. Requires a directed graph.

Usage

```
centrality_salsa(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object). Must be directed.

`...` Additional arguments passed to [centrality](#).

Value

Named numeric vector of SALSA authority scores.

See Also

[centrality](#) for computing multiple measures at once, [centrality_authority](#) for HITS authority.

Examples

```
## Not run:
adj <- matrix(c(0, 1, 0, 0, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_salsa(adj)

## End(Not run)
```

centrality_semilocal *Semi-Local Centrality*

Description

Triple-nested neighborhood computation measuring 4-hop local influence.

Usage

```
centrality_semilocal(x, mode = "all", ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).

`mode` For directed networks: "all" (default), "in", or "out".

`...` Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of semi-local centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_semilocal(adj)
```

centrality_strength *Strength Centrality (Weighted Degree)*

Description

Sum of edge weights connected to each node. For directed networks, `centrality_instrength` sums incoming weights and `centrality_outstrength` sums outgoing weights.

Usage

```
centrality_strength(x, mode = "all", ...)
```

```
centrality_instrength(x, ...)
```

```
centrality_outstrength(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`mode` For directed networks: "all" (default), "in", or "out".
`...` Additional arguments passed to [centrality](#) (e.g., normalized, weighted, directed).

Value

Named numeric vector of strength values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_degree](#) for the unweighted version.

Examples

```
mat <- matrix(c(0, .5, .3, .5, 0, .8, .3, .8, 0), 3, 3)
rownames(mat) <- colnames(mat) <- c("A", "B", "C")
centrality_strength(mat)
```

centrality_stress *Stress Centrality*

Description

Number of shortest paths passing through each node. Unlike `betweenness`, does not normalize by the total number of shortest paths.

Usage

```
centrality_stress(x, ...)
```

Arguments

`x` Network input (matrix, igraph, network, cograph_network, tna object).
`...` Additional arguments passed to `centrality`.

Value

Named numeric vector of stress centrality values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_betweenness](#) for the normalized variant.

Examples

```
adj <- matrix(c(0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0), 4, 4)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
centrality_stress(adj)
```

centrality_subgraph *Subgraph Centrality*

Description

Participation in closed loops (walks), weighting shorter loops more heavily. Based on the diagonal of the matrix exponential of the adjacency matrix.

Usage

```
centrality_subgraph(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
 ... Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of subgraph centrality values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_subgraph(adj)
```

centrality_topological_coefficient
Topological Coefficient

Description

Fraction of shared second-order neighbors, measuring topological overlap between a node and its neighbors.

Usage

```
centrality_topological_coefficient(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
 ... Additional arguments passed to [centrality](#).

Value

Named numeric vector of topological coefficient values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_topological_coefficient(adj)
```

`centrality_transitivity`*Local Transitivity (Clustering Coefficient)*

Description

Proportion of triangles around each node relative to the number of possible triangles. Measures how tightly clustered a node's neighborhood is.

Usage

```
centrality_transitivity(x, transitivity_type = "local", isolates = "nan", ...)
```

Arguments

<code>x</code>	Network input (matrix, igraph, network, cograph_network, tna object).
<code>transitivity_type</code>	Type of transitivity: "local" (default), "global", "undirected", "localundirected", "barrat" (weighted), or "weighted".
<code>isolates</code>	How to handle isolate nodes: "nan" (default) or "zero".
<code>...</code>	Additional arguments passed to centrality (e.g., weighted, directed).

Value

Named numeric vector of transitivity values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_transitivity(adj)
```

centrality_voterank *VoteRank Centrality*

Description

Identifies influential spreaders via an iterative voting mechanism. Returns normalized rank (1 = most influential). Based on Zhang et al. (2016).

Usage

```
centrality_voterank(x, ...)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).
... Additional arguments passed to [centrality](#) (e.g., weighted, directed).

Value

Named numeric vector of VoteRank values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)  
rownames(adj) <- colnames(adj) <- c("A", "B", "C")  
centrality_voterank(adj)
```

centrality_wiener *Wiener Index Centrality*

Description

Total sum of shortest path distances from a node to all others. Higher values indicate less central (more peripheral) nodes.

Usage

```
centrality_wiener(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality (e.g., normalized, weighted, directed).

Value

Named numeric vector of Wiener index values.

See Also

[centrality](#) for computing multiple measures at once.

Examples

```
adj <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
centrality_wiener(adj)
```

centrality_within_module_z

Within-Module Degree Z-Score

Description

Z-score of intra-community connectivity. High values indicate hubs within their own community. Requires community membership.

Usage

```
centrality_within_module_z(x, membership = NULL, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
membership	Integer vector of community assignments (one per node).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality .

Value

Named numeric vector of within-module z-score values.

See Also

[centrality](#) for computing multiple measures at once, [centrality_participation](#) for between-community diversity.

Examples

```
adj <- matrix(c(0,1,1,0,0, 1,0,1,0,0, 1,1,0,1,0, 0,0,1,0,1, 0,0,0,1,0), 5, 5)
rownames(adj) <- colnames(adj) <- LETTERS[1:5]
centrality_within_module_z(adj, membership = c(1, 1, 1, 2, 2))
```

centralization	<i>Centralization index</i>
----------------	-----------------------------

Description

Computes Freeman's centralization for degree, betweenness, closeness, or eigenvector centrality.

Usage

```
centralization(
  x,
  measure = c("degree", "betweenness", "closeness", "eigenvector"),
  directed = NULL,
  mode = "all",
  ...
)
```

Arguments

x	Network input
measure	One of "degree", "betweenness", "closeness", "eigenvector"
directed	Logical or NULL
mode	"all", "in", or "out"
...	Additional arguments passed to to_igraph()

Value

Numeric scalar in $[0, 1]$

Examples

```
star <- matrix(0, 5, 5)
star[1, 2:5] <- 1; star[2:5, 1] <- 1
cograph::centralization(star, "degree")
```

cluster_quality *Cluster Quality Metrics*

Description

Computes per-cluster and global quality metrics for network partitioning. Supports both binary and weighted networks.

Usage

```
cluster_quality(x, clusters, weighted = TRUE, directed = TRUE)
```

```
cqual(x, clusters, weighted = TRUE, directed = TRUE)
```

Arguments

x	Adjacency matrix
clusters	Cluster specification (list or membership vector)
weighted	Logical; if TRUE, use edge weights; if FALSE, binarize
directed	Logical; if TRUE, treat as directed network

Value

A cluster_quality object with:

per_cluster	Data frame with per-cluster metrics
global	List of global metrics (modularity, coverage)

See [cluster_quality](#).

Examples

```
mat <- matrix(runif(100), 10, 10)
diag(mat) <- 0
clusters <- c(1,1,1,2,2,2,3,3,3,3)

q <- cluster_quality(mat, clusters)
q$per_cluster    # Per-cluster metrics
q$global        # Modularity, coverage
mat <- matrix(runif(100), 10, 10)
diag(mat) <- 0
cqual(mat, c(1,1,1,2,2,2,3,3,3,3))
```

cluster_significance Test Significance of Community Structure

Description

Compares observed modularity against a null model distribution to assess whether the detected community structure is statistically significant.

Usage

```
cluster_significance(  
  x,  
  communities,  
  n_random = 100,  
  method = c("configuration", "gnm"),  
  seed = NULL  
)  
  
csig(  
  x,  
  communities,  
  n_random = 100,  
  method = c("configuration", "gnm"),  
  seed = NULL  
)
```

Arguments

x	Network input: adjacency matrix, igraph object, or cograph_network.
communities	A communities object (from communities or igraph) or a membership vector (integer vector where communities[i] is the community of node i).
n_random	Number of random networks to generate for the null distribution. Default 100.
method	Null model type: "configuration" Preserves degree sequence (default). More stringent test. "gnm" Erdos-Renyi model with same number of edges. Tests against random baseline.
seed	Random seed for reproducibility. Default NULL.

Details

The test works by:

1. Computing the modularity of the provided community structure
2. Generating n_random random networks using the specified null model
3. For each random network, detecting communities with Louvain and computing modularity

4. Comparing the observed modularity to this null distribution

A significant result (low p-value) indicates that the community structure is stronger than expected by chance for networks with similar properties.

Value

A `cograph_cluster_significance` object with:

observed_modularity Modularity of the input communities

null_mean Mean modularity of random networks

null_sd Standard deviation of null modularity

z_score Standardized score: $(\text{observed} - \text{null_mean}) / \text{null_sd}$

p_value One-sided p-value (probability of observing equal or higher modularity by chance)

null_values Vector of modularity values from null distribution

method Null model method used

n_random Number of random networks generated

See [cluster_significance](#).

References

Reichardt, J., & Bornholdt, S. (2006). Statistical mechanics of community detection. *Physical Review E*, 74, 016110.

See Also

[communities](#), [cluster_quality](#)

Examples

```
## Not run:
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")
  comm <- community_louvain(g)

  # Test significance
  sig <- cluster_significance(g, comm, n_random = 100, seed = 123)
  print(sig)

  # Configuration model (stricter test)
  sig2 <- cluster_significance(g, comm, method = "configuration")
}

## End(Not run)
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")
  comm <- community_louvain(g)
  csig(g, comm, n_random = 20, seed = 1)
}
```

cluster_summary	<i>Cluster Summary Statistics</i>
-----------------	-----------------------------------

Description

Aggregates node-level network weights to cluster-level summaries. Computes both macro (cluster-to-cluster) transitions and per-cluster transitions (how nodes connect inside each cluster).

Usage

```
cluster_summary(
  x,
  clusters = NULL,
  method = c("sum", "mean", "median", "max", "min", "density", "geomean"),
  type = c("tna", "cooccurrence", "semi_markov", "raw"),
  directed = TRUE,
  compute_within = TRUE
)

csum(
  x,
  clusters = NULL,
  method = c("sum", "mean", "median", "max", "min", "density", "geomean"),
  type = c("tna", "cooccurrence", "semi_markov", "raw"),
  directed = TRUE,
  compute_within = TRUE
)
```

Arguments

<code>x</code>	<p>Network input. Accepts multiple formats:</p> <ul style="list-style-type: none"> matrix Numeric adjacency/weight matrix. Row and column names are used as node labels. Values represent edge weights (e.g., transition counts, co-occurrence frequencies, or probabilities). cograph_network A cograph network object. The function extracts the weight matrix from <code>x\$weights</code> or converts via <code>to_matrix()</code>. Clusters can be auto-detected from node attributes. tna A tna object from the tna package. Extracts <code>x\$weights</code>. cluster_summary If already a cluster_summary, returns unchanged.
<code>clusters</code>	<p>Cluster/group assignments for nodes. Accepts multiple formats:</p> <ul style="list-style-type: none"> NULL (default) Auto-detect from <code>cograph_network</code>. Looks for columns named 'clusters', 'cluster', 'groups', or 'group' in <code>x\$nodes</code>. Throws an error if no cluster column is found. This option only works when <code>x</code> is a <code>cograph_network</code>.

	<p>vector Cluster membership for each node, in the same order as the matrix rows/columns. Can be numeric (1, 2, 3) or character ("A", "B"). Cluster names will be derived from unique values. Example: <code>c(1, 1, 2, 2, 3, 3)</code> assigns first two nodes to cluster 1.</p> <p>data.frame A data frame where the first column contains node names and the second column contains group/cluster names. Example: <code>data.frame(node = c("A", "B", "C"), group = c("G1", "G1", "G2"))</code></p> <p>named list Explicit mapping of cluster names to node labels. List names become cluster names, values are character vectors of node labels that must match matrix row/column names. Example: <code>list(Alpha = c("A", "B"), Beta = c("C", "D"))</code></p>
method	<p>Aggregation method for combining edge weights within/between clusters. Controls how multiple node-to-node edges are summarized:</p> <p>"sum" (default) Sum of all edge weights. Best for count data (e.g., transition frequencies). Preserves total flow.</p> <p>"mean" Average edge weight. Best when cluster sizes differ and you want to control for size. Note: when input is already a transition matrix (rows sum to 1), "mean" avoids size bias. Example: cluster with 5 nodes won't have 5x the weight of cluster with 1 node.</p> <p>"median" Median edge weight. Robust to outliers.</p> <p>"max" Maximum edge weight. Captures strongest connection.</p> <p>"min" Minimum edge weight. Captures weakest connection.</p> <p>"density" Sum divided by number of possible edges. Normalizes by cluster size combinations.</p> <p>"geomean" Geometric mean of positive weights. Useful for multiplicative processes.</p>
type	<p>Post-processing applied to aggregated weights. Determines the interpretation of the resulting matrices:</p> <p>"tna" (default) Row-normalize so each row sums to 1. Creates transition probabilities suitable for Markov chain analysis. Interpretation: "Given I'm in cluster A, what's the probability of transitioning to cluster B?" Required for use with tna package functions. Diagonal is zero; per-cluster data is in <code>\$clusters</code>.</p> <p>"raw" No normalization. Returns aggregated counts/weights as-is. Use for frequency analysis or when you need raw counts. Compatible with igraph's <code>contract + simplify</code> output.</p> <p>"cooccurrence" Symmetrize the matrix: $(A + t(A)) / 2$. For undirected co-occurrence analysis.</p> <p>"semi_markov" Row-normalize with duration weighting. For semi-Markov process analysis.</p>
directed	<p>Logical. If TRUE (default), treat network as directed. A->B and B->A are separate edges. If FALSE, edges are undirected and the matrix is symmetrized before processing.</p>
compute_within	<p>Logical. If TRUE (default), compute per-cluster transition matrices for each cluster. Each cluster gets its own $n_i \times n_i$ matrix showing internal node-to-node</p>

transitions. Set to FALSE to skip this computation for better performance when only the macro (cluster-level) summary is needed.

Details

This is the core function for Multi-Cluster Multi-Level (MCML) analysis. Use `as_tna` to convert results to tna objects for further analysis with the tna package.

Workflow:

Typical MCML analysis workflow:

```
# 1. Create network
net <- cograph(edges, nodes = nodes)
net$nodes$clusters <- group_assignments

# 2. Compute cluster summary
cs <- cluster_summary(net, type = "tna")

# 3. Convert to tna models
tna_models <- as_tna(cs)

# 4. Analyze/visualize
plot(tna_models$macro)
tna::centralities(tna_models$macro)
```

Between-Cluster Matrix Structure:

The `macro$weights` matrix has clusters as both rows and columns:

- Off-diagonal (row *i*, col *j*): Aggregated weight from cluster *i* to cluster *j*
- Diagonal (row *i*, col *i*): Per-cluster total (sum of internal edges in cluster *i*)

When `type = "tna"`, rows sum to 1 and diagonal values represent "retention rate" - the probability of staying inside the same cluster.

Choosing method and type:

Input data	Recommended	Reason
Edge counts	<code>method="sum", type="tna"</code>	Preserves total flow, normalizes to probabilities
Transition matrix	<code>method="mean", type="tna"</code>	Avoids cluster size bias
Frequencies	<code>method="sum", type="raw"</code>	Keep raw counts for analysis
Correlation matrix	<code>method="mean", type="raw"</code>	Average correlations

Value

A `cluster_summary` object (S3 class) containing:

macro A tna object representing the macro (cluster-level) network:

weights $k \times k$ matrix of cluster-to-cluster weights, where k is the number of clusters. Row i , column j contains the aggregated weight from cluster i to cluster j . Diagonal contains aggregated intra-cluster weight (retention / self-loops). Processing depends on type.

inits Numeric vector of length k . Initial state distribution across clusters, computed from column sums of the original matrix. Represents the proportion of incoming edges to each cluster.

clusters Named list with one element per cluster. Each element is a tna object containing:

weights $n_i \times n_i$ matrix for nodes inside that cluster. Shows internal transitions between nodes in the same cluster.

inits Initial distribution for the cluster.

NULL if `compute_within = FALSE`.

cluster_members Named list mapping cluster names to their member node labels. Example:

```
list(A = c("n1", "n2"), B = c("n3", "n4", "n5"))
```

meta List of metadata:

type The type argument used ("tna", "raw", etc.)

method The method argument used ("sum", "mean", etc.)

directed Logical, whether network was treated as directed

n_nodes Total number of nodes in original network

n_clusters Number of clusters

cluster_sizes Named vector of cluster sizes

See [cluster_summary](#).

See Also

[as_tna](#) to convert results to tna objects, [plot_mcml](#) for two-layer visualization, [plot_mtna](#) for flat cluster visualization

Examples

```
# -----
# Basic usage with matrix and cluster vector
# -----
mat <- matrix(runif(100), 10, 10)
diag(mat) <- 0
rownames(mat) <- colnames(mat) <- LETTERS[1:10]

clusters <- c(1, 1, 1, 2, 2, 2, 3, 3, 3, 3)
cs <- cluster_summary(mat, clusters)

# Access results
cs$weights      # 3x3 cluster transition matrix
cs$inits        # Initial distribution
cs$clusters$1$weights # Per-cluster 1 transitions
cs$meta        # Metadata

# -----
# Named list clusters (more readable)
# -----
clusters <- list(
  Alpha = c("A", "B", "C"),
```

```

    Beta = c("D", "E", "F"),
    Gamma = c("G", "H", "I", "J")
  )
  cs <- cluster_summary(mat, clusters, type = "tna")
  cs$macro$weights      # Rows/cols named Alpha, Beta, Gamma
  cs$clusters$Alpha     # Per-cluster Alpha network

# -----
# Auto-detect clusters from cograph_network
# -----
net <- as_cograph(mat)
net$nodes$clusters <- c(1, 1, 1, 2, 2, 2, 3, 3, 3, 3)
cs <- cluster_summary(net) # No clusters argument needed

# -----
# Different aggregation methods
# -----
cs_sum <- cluster_summary(mat, clusters, method = "sum") # Total flow
cs_mean <- cluster_summary(mat, clusters, method = "mean") # Average
cs_max <- cluster_summary(mat, clusters, method = "max") # Strongest

# -----
# Raw counts vs TNA probabilities
# -----
cs_raw <- cluster_summary(mat, clusters, type = "raw")
cs_tna <- cluster_summary(mat, clusters, type = "tna")

rowSums(cs_raw$macro$weights) # Various sums
rowSums(cs_tna$macro$weights) # All equal to 1

# -----
# Skip within-cluster computation for speed
# -----
cs_fast <- cluster_summary(mat, clusters, compute_within = FALSE)
cs_fast$clusters # NULL

# -----
# Convert to tna objects for tna package
# -----
cs <- cluster_summary(mat, clusters, type = "tna")
tna_models <- as_tna(cs)
# tna_models$macro      # tna object
# tna_models$Alpha     # tna object (cluster network)
mat <- matrix(c(0.5, 0.2, 0.3, 0.1, 0.6, 0.3, 0.4, 0.1, 0.5), 3, 3,
             byrow = TRUE,
             dimnames = list(c("A", "B", "C"), c("A", "B", "C")))
csum(mat, list(G1 = c("A", "B"), G2 = c("C")))

```

Description

The main entry point for `cograph`. Accepts adjacency matrices, edge lists, `igraph`, `statnet` network, `qgraph`, or `tna` objects and creates a visualization-ready network object.

Usage

```
cograph(
  input,
  layout = NULL,
  directed = NULL,
  nodes = NULL,
  seed = 42,
  simplify = FALSE,
  ...
)
```

Arguments

<code>input</code>	Network input. Can be: <ul style="list-style-type: none"> • A square numeric matrix (adjacency/weight matrix) • A data frame with edge list (from, to, optional weight columns) • An <code>igraph</code> object • A <code>statnet</code> network object • A <code>qgraph</code> object • A <code>tna</code> object
<code>layout</code>	Layout algorithm: "circle", "spring", "groups", "grid", "random", "star", "bipartite", or "custom". Default NULL (no layout computed). Set to a layout name to compute immediately, or use <code>sn_layout()</code> later.
<code>directed</code>	Logical. Force directed interpretation. NULL for auto-detect.
<code>nodes</code>	Node metadata. Can be NULL or a data frame with node attributes. If data frame has a <code>label</code> or <code>labels</code> column, those are used for display.
<code>seed</code>	Random seed for deterministic layouts. Default 42. Set NULL for random.
<code>simplify</code>	Logical or character. If FALSE (default), every transition from <code>tna</code> sequence data is a separate edge. If TRUE or a string ("sum", "mean", "max", "min"), duplicate edges are aggregated.
<code>...</code>	Additional arguments passed to the layout function.

Value

A `cograph_network` object that can be further customized and rendered.

See Also

[splot](#) for base R graphics rendering, [soplot](#) for grid graphics rendering, [sn_nodes](#) for node customization, [sn_edges](#) for edge customization, [sn_layout](#) for changing layouts, [sn_theme](#) for visual themes, [sn_palette](#) for color palettes, [from_qgraph](#) and [from_tna](#) for converting external objects

Examples

```
# From adjacency matrix (no layout computed yet - fast!)
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- cograph(adj)

# Layout computed automatically when plotting
splot(net) # Uses spring layout by default

# From edge list
edges <- data.frame(from = c(1, 1, 2), to = c(2, 3, 3))
cograph(edges)

# Compute layout immediately if needed
cograph(adj, layout = "circle") |> splot()

# With customization (pipe-friendly workflow)
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
cograph(adj) |>
  sn_nodes(fill = "steelblue") |>
  sn_edges(color = "gray50") |>
  splot(layout = "circle")

# Weighted network with automatic styling
w_adj <- matrix(c(0, 0.5, -0.3, 0.5, 0, 0.4, -0.3, 0.4, 0), nrow = 3)
cograph(w_adj) |>
  sn_edges(color = "weight", width = "weight") |>
  splot()

# With igraph (if installed)
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_ring(10)
  cograph(g) |> splot()
}
```

CographLayout

CographLayout R6 Class

Description

Class for managing layout algorithms and computing node positions.

Value

A CographLayout R6 object.

Methods

Public methods:

- [CographLayout\\$new\(\)](#)

- [CographLayout\\$compute\(\)](#)
- [CographLayout\\$normalize_coords\(\)](#)
- [CographLayout\\$get_type\(\)](#)
- [CographLayout\\$get_params\(\)](#)
- [CographLayout\\$print\(\)](#)
- [CographLayout\\$clone\(\)](#)

Method `new()`: Create a new CographLayout object.

Usage:

```
CographLayout$new(type = "circle", ...)
```

Arguments:

`type` Layout type (e.g., "circle", "spring", "groups").

... Additional parameters for the layout algorithm.

Returns: A new CographLayout object.

Method `compute()`: Compute layout coordinates for a network.

Usage:

```
CographLayout$compute(network, ...)
```

Arguments:

`network` A CographNetwork object.

... Additional parameters passed to the layout function.

Returns: Data frame with x, y coordinates.

Method `normalize_coords()`: Normalize coordinates to 0-1 range with padding.

Usage:

```
CographLayout$normalize_coords(coords, padding = 0.1)
```

Arguments:

`coords` Matrix or data frame with x, y columns.

`padding` Numeric. Padding around edges (default 0.1).

Returns: Normalized coordinates.

Method `get_type()`: Get layout type.

Usage:

```
CographLayout$get_type()
```

Returns: Character string.

Method `get_params()`: Get layout parameters.

Usage:

```
CographLayout$get_params()
```

Returns: List of parameters.

Method `print()`: Print layout summary.

Usage:

```
CographLayout$print()
```

Method `clone()`: The objects of this class are cloneable with this method.

Usage:

```
CographLayout$clone(deep = FALSE)
```

Arguments:

`deep` Whether to make a deep clone.

Examples

```
# Create a circular layout
layout <- CographLayout$new("circle")

# Apply to network
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- CographNetwork$new(adj)
coords <- layout$compute(net)
```

CographNetwork

CographNetwork R6 Class

Description

Core class representing a network for visualization. Stores nodes, edges, layout coordinates, and aesthetic mappings.

Value

A CographNetwork R6 object.

Active bindings

`n_nodes` Number of nodes in the network.

`n_edges` Number of edges in the network.

`is_directed` Whether the network is directed.

`has_weights` Whether edges have weights.

`node_labels` Vector of node labels (priority: labels > label).

Methods**Public methods:**

- `CographNetwork$new()`
- `CographNetwork$clone_network()`
- `CographNetwork$set_nodes()`
- `CographNetwork$set_edges()`
- `CographNetwork$set_directed()`
- `CographNetwork$set_weights()`
- `CographNetwork$set_layout_coords()`
- `CographNetwork$set_node_aes()`
- `CographNetwork$set_edge_aes()`
- `CographNetwork$set_theme()`
- `CographNetwork$get_nodes()`
- `CographNetwork$get_edges()`
- `CographNetwork$get_layout()`
- `CographNetwork$get_node_aes()`
- `CographNetwork$get_edge_aes()`
- `CographNetwork$get_theme()`
- `CographNetwork$set_layout_info()`
- `CographNetwork$get_layout_info()`
- `CographNetwork$set_plot_params()`
- `CographNetwork$get_plot_params()`
- `CographNetwork$print()`
- `CographNetwork$clone()`

Method `new()`: Create a new `CographNetwork` object.

Usage:

```
CographNetwork$new(
  input = NULL,
  directed = NULL,
  nodes = NULL,
  simplify = FALSE
)
```

Arguments:

`input` Network input (matrix, edge list, or igraph object).

`directed` Logical. Force directed interpretation. `NULL` for auto-detect.

`nodes` Node metadata. Can be `NULL` or a data frame with node attributes. If data frame has a `label` or `labels` column, those are used for display.

`simplify` Logical or character. If `FALSE` (default), every transition from tna sequence data is a separate edge. If `TRUE` or a string ("`sum`", "`mean`", "`max`", "`min`"), duplicate edges are aggregated.

Returns: A new `CographNetwork` object.

Method `clone_network()`: Clone the network with optional modifications.

Usage:

```
CographNetwork$clone_network()
```

Returns: A new CographNetwork object.

Method `set_nodes()`: Set nodes data frame.

Usage:

```
CographNetwork$set_nodes(nodes)
```

Arguments:

`nodes` Data frame with node information.

Method `set_edges()`: Set edges data frame.

Usage:

```
CographNetwork$set_edges(edges)
```

Arguments:

`edges` Data frame with edge information.

Method `set_directed()`: Set directed flag.

Usage:

```
CographNetwork$set_directed(directed)
```

Arguments:

`directed` Logical.

Method `set_weights()`: Set edge weights.

Usage:

```
CographNetwork$set_weights(weights)
```

Arguments:

`weights` Numeric vector of weights.

Method `set_layout_coords()`: Set layout coordinates.

Usage:

```
CographNetwork$set_layout_coords(coords)
```

Arguments:

`coords` Matrix or data frame with x, y columns.

Method `set_node_aes()`: Set node aesthetics.

Usage:

```
CographNetwork$set_node_aes(aes)
```

Arguments:

`aes` List of aesthetic parameters.

Method `set_edge_aes()`: Set edge aesthetics.

Usage:

`CographNetwork$set_edge_aes(aes)`

Arguments:

`aes` List of aesthetic parameters.

Method `set_theme()`: Set theme.

Usage:

`CographNetwork$set_theme(theme)`

Arguments:

`theme` CographTheme object or theme name.

Method `get_nodes()`: Get nodes data frame.

Usage:

`CographNetwork$get_nodes()`

Returns: Data frame with node information.

Method `get_edges()`: Get edges data frame.

Usage:

`CographNetwork$get_edges()`

Returns: Data frame with edge information.

Method `get_layout()`: Get layout coordinates.

Usage:

`CographNetwork$get_layout()`

Returns: Data frame with x, y coordinates.

Method `get_node_aes()`: Get node aesthetics.

Usage:

`CographNetwork$get_node_aes()`

Returns: List of node aesthetic parameters.

Method `get_edge_aes()`: Get edge aesthetics.

Usage:

`CographNetwork$get_edge_aes()`

Returns: List of edge aesthetic parameters.

Method `get_theme()`: Get theme.

Usage:

`CographNetwork$get_theme()`

Returns: CographTheme object.

Method `set_layout_info()`: Set layout info.

Usage:

```
CographNetwork$set_layout_info(info)
```

Arguments:

info List with layout information (name, seed, etc.).

Method `get_layout_info()`: Get layout info.

Usage:

```
CographNetwork$get_layout_info()
```

Returns: List with layout information.

Method `set_plot_params()`: Set plot parameters.

Usage:

```
CographNetwork$set_plot_params(params)
```

Arguments:

params List of all plot parameters used.

Method `get_plot_params()`: Get plot parameters.

Usage:

```
CographNetwork$get_plot_params()
```

Returns: List of plot parameters.

Method `print()`: Print network summary.

Usage:

```
CographNetwork$print()
```

Method `clone()`: The objects of this class are cloneable with this method.

Usage:

```
CographNetwork$clone(deep = FALSE)
```

Arguments:

deep Whether to make a deep clone.

Examples

```
# Create network from adjacency matrix
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- CographNetwork$new(adj)

# Access properties
net$n_nodes
net$n_edges
net$is_directed
```

CographTheme

CographTheme R6 Class

Description

Class for managing visual themes for network plots.

Value

A CographTheme R6 object.

Active bindings

name Theme name.

Methods

Public methods:

- [CographTheme\\$new\(\)](#)
- [CographTheme\\$get\(\)](#)
- [CographTheme\\$set\(\)](#)
- [CographTheme\\$get_all\(\)](#)
- [CographTheme\\$merge\(\)](#)
- [CographTheme\\$clone_theme\(\)](#)
- [CographTheme\\$print\(\)](#)
- [CographTheme\\$clone\(\)](#)

Method `new()`: Create a new CographTheme object.

Usage:

```
CographTheme$new(  
  name = "custom",  
  background = "white",  
  node_fill = "#4A90D9",  
  node_border = "#2C5AA0",  
  node_border_width = 1,  
  edge_color = "gray50",  
  edge_positive_color = "#2E7D32",  
  edge_negative_color = "#C62828",  
  edge_width = 1,  
  label_color = "black",  
  label_size = 10,  
  title_color = "black",  
  title_size = 14,  
  legend_background = "white"  
)
```

Arguments:

name Theme name (optional).
background Background color.
node_fill Default node fill color.
node_border Default node border color.
node_border_width Default node border width.
edge_color Default edge color.
edge_positive_color Color for positive edge weights.
edge_negative_color Color for negative edge weights.
edge_width Default edge width.
label_color Default label color.
label_size Default label size.
title_color Title color.
title_size Title size.
legend_background Legend background color.

Returns: A new CographTheme object.

Method `get()`: Get a theme parameter.

Usage:

```
CographTheme$get(name)
```

Arguments:

name Parameter name.

Returns: Parameter value.

Method `set()`: Set a theme parameter.

Usage:

```
CographTheme$set(name, value)
```

Arguments:

name Parameter name.

value Parameter value.

Method `get_all()`: Get all theme parameters.

Usage:

```
CographTheme$get_all()
```

Returns: List of parameters.

Method `merge()`: Merge with another theme.

Usage:

```
CographTheme$merge(other)
```

Arguments:

other Another CographTheme or list of parameters.

Returns: A new merged CographTheme.

Method `clone_theme()`: Clone the theme.

Usage:

```
CographTheme$clone_theme()
```

Returns: A new CographTheme.

Method `print()`: Print theme summary.

Usage:

```
CographTheme$print()
```

Method `clone()`: The objects of this class are cloneable with this method.

Usage:

```
CographTheme$clone(deep = FALSE)
```

Arguments:

`deep` Whether to make a deep clone.

Examples

```
# Create a custom theme
theme <- CographTheme$new(
  background = "white",
  node_fill = "steelblue",
  edge_color = "gray60"
)
```

color_communities *Color Nodes by Community*

Description

Generate colors for nodes based on community membership. Designed for direct use with `splot()` `node.color` parameter.

Usage

```
color_communities(x, method = "louvain", palette = NULL, ...)
```

Arguments

<code>x</code>	Network input: matrix, igraph, network, cograph_network, or tna object.
<code>method</code>	Community detection algorithm. See detect_communities for available methods. Default "louvain".
<code>palette</code>	Color palette to use. Can be: <ul style="list-style-type: none"> • NULL (default): Uses a colorblind-friendly palette • A character vector of colors • A function that takes n and returns n colors • A palette name: "rainbow", "colorblind", "pastel", "viridis"
<code>...</code>	Additional arguments passed to detect_communities .

Value

A named character vector of colors (one per node), suitable for use with `splot()` `node.color` parameter.

See Also

[detect_communities](#), [splot](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")

# Basic usage with splot
splot(adj, node_fill = color_communities(adj))

# Custom palette
splot(adj, node_fill = color_communities(adj, palette = c("red", "blue")))
```

communities

Community Detection

Description

Detects communities/clusters in networks using various algorithms. Provides a unified interface to `igraph`'s community detection functions with full parameter exposure.

Usage

```
communities(
  x,
  method = c("louvain", "leiden", "fast_greedy", "walktrap", "infomap",
            "label_propagation", "edge_betweenness", "leading_eigenvector", "spinglass",
            "optimal", "fluid"),
  community = NULL,
  weights = NULL,
  resolution = 1,
  directed = NULL,
  seed = NULL,
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
method	Community detection algorithm. One of: <ul style="list-style-type: none"> • "louvain" - Louvain modularity optimization (default, fast) • "leiden" - Leiden algorithm (improved Louvain) • "fast_greedy" - Fast greedy modularity optimization • "walktrap" - Random walk-based detection • "infomap" - Information theoretic approach • "label_propagation" - Label propagation (very fast) • "edge_betweenness" - Girvan-Newman algorithm • "leading_eigenvector" - Leading eigenvector method • "spinglass" - Spinglass simulation • "optimal" - Exact modularity optimization (slow) • "fluid" - Fluid communities algorithm
community	Optional integer or character vector. If supplied, the returned data frame is filtered to rows whose community column matches one of the given values. Default NULL (keep all communities).
weights	Edge weights. If NULL, uses edge weights from the network if available, otherwise unweighted. Set to NA for explicitly unweighted.
resolution	Resolution parameter for modularity-based methods (louvain, leiden). Higher values yield more communities. Default 1.
directed	Logical; whether to treat the network as directed. Default NULL (auto-detect).
seed	Random seed for reproducibility. Only applies to stochastic algorithms (louvain, leiden, infomap, label_propagation, spinglass).
...	Additional parameters passed to the specific algorithm. See individual functions for details.

Details**Algorithm Selection Guide:**

Algorithm	Best For	Time Complexity
louvain	Large networks, general use	$O(n \log n)$
leiden	Large networks, better quality than louvain	$O(n \log n)$
fast_greedy	Medium networks	$O(n^2 \log n)$
walktrap	Networks with clear community structure	$O(n^2 \log n)$
infomap	Directed networks, flow-based	$O(E)$
label_propagation	Very large networks, speed critical	$O(E)$
edge_betweenness	Small networks, hierarchical	$O(E^2 n)$
leading_eigenvector	Networks with dominant structure	$O(n^2)$
spinglass	Small networks, allows negative weights	$O(n^3)$
optimal	Tiny networks only (<50 nodes)	NP-hard
fluid	When k is known	$O(E k)$

Value

A tidy `cograph_communities` data frame with columns:

node Node label (character)

community Community assignment (integer)

Metadata stored as attributes: "algorithm", "modularity", "network" (original input), "igraph_result".

See Also

[community_louvain](#), [community_leiden](#), [community_fast_greedy](#), [community_walktrap](#), [community_infomap](#), [community_label_propagation](#), [community_edge_betweenness](#), [community_leading_eigenvector](#), [community_spinglass](#), [community_optimal](#), [community_fluid](#)

Examples

```
# Create a network with community structure
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")

  # Default (Louvain)
  comm <- cograph::communities(g)
  print(comm)

  # Walktrap
  comm2 <- cograph::communities(g, method = "walktrap")
  print(comm2)
}
```

community_consensus *Consensus Community Detection*

Description

Runs a stochastic community detection algorithm multiple times and finds consensus communities via co-occurrence matrix thresholding. This approach produces more robust and stable community assignments than single runs.

Usage

```
community_consensus(
  x,
  method = c("louvain", "leiden", "infomap", "label_propagation", "spinglass"),
  n_runs = 100,
  threshold = 0.5,
  seed = NULL,
  ...
)
```

```

com_consensus(
  x,
  method = c("louvain", "leiden", "infomap", "label_propagation", "spinglass"),
  n_runs = 100,
  threshold = 0.5,
  seed = NULL,
  ...
)

```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
method	Community detection algorithm to use. Default "louvain". Must be a stochastic method (louvain, leiden, infomap, label_propagation, spinglass).
n_runs	Number of times to run the algorithm. Default 100.
threshold	Co-occurrence threshold for consensus. Default 0.5. Nodes that appear together in \geq threshold proportion of runs are placed in the same community.
seed	Optional seed for reproducibility. If provided, seeds for individual runs are derived from this seed.
...	Additional arguments passed to the community detection method.

Details

The algorithm works as follows:

1. Run the specified algorithm `n_runs` times (without seeds to allow variation)
2. Build a co-occurrence matrix counting how often each pair of nodes appears in the same community
3. Normalize to proportions (0-1)
4. Threshold to create a consensus graph (edge if co-occurrence \geq threshold)
5. Run walktrap on the consensus graph to get final communities

Value

A `cograph_communities` object with consensus membership.

References

Lancichinetti, A., & Fortunato, S. (2012). Consensus clustering in complex networks. *Scientific Reports*, 2, 336.

See Also

[communities](#), [community_louvain](#)

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")

  # Consensus from 50 Louvain runs
  cc <- community_consensus(g, method = "louvain", n_runs = 50)
  print(cc)

  # Stricter threshold for more robust communities
  cc2 <- community_consensus(g, threshold = 0.7, n_runs = 100)
}
```

community_edge_betweenness

Edge Betweenness Community Detection

Description

Girvan-Newman algorithm. Iteratively removes edges with highest betweenness centrality to reveal community structure.

Usage

```
community_edge_betweenness(
  x,
  weights = NULL,
  directed = TRUE,
  edge.betweenness = TRUE,
  merges = TRUE,
  bridges = TRUE,
  modularity = TRUE,
  membership = TRUE,
  ...
)

com_eb(
  x,
  weights = NULL,
  directed = TRUE,
  edge.betweenness = TRUE,
  merges = TRUE,
  bridges = TRUE,
  modularity = TRUE,
  membership = TRUE,
  ...
)
```

Arguments

x	Network input
weights	Edge weights. NULL uses network weights, NA for unweighted.
directed	Logical; treat graph as directed? Default TRUE.
edge.betweenness	Logical; return edge betweenness values? Default TRUE.
merges	Logical; return merge matrix? Default TRUE.
bridges	Logical; return bridge edges? Default TRUE.
modularity	Logical; return modularity scores? Default TRUE.
membership	Logical; return membership vector? Default TRUE.
...	Additional arguments passed to to_igraph

Value

A cograph_communities object

A cograph_communities object. See [detect_communities](#).

References

Girvan, M., & Newman, M.E.J. (2002). Community structure in social and biological networks. *PNAS*, 99(12), 7821-7826.

Examples

```
g <- igraph::make_graph("Zachary")
comm <- community_edge_betweenness(g)
membership(comm)

net <- as_cograph(matrix(runif(25), 5, 5))
com_eb(net)
```

community_fast_greedy *Fast Greedy Community Detection*

Description

Hierarchical agglomeration using greedy modularity optimization. Produces a dendrogram of community merges.

Usage

```
community_fast_greedy(  
  x,  
  weights = NULL,  
  merges = TRUE,  
  modularity = TRUE,  
  membership = TRUE,  
  ...  
)  
  
com_fg(  
  x,  
  weights = NULL,  
  merges = TRUE,  
  modularity = TRUE,  
  membership = TRUE,  
  ...  
)
```

Arguments

x	Network input
weights	Edge weights. NULL uses network weights, NA for unweighted.
merges	Logical; return merge matrix? Default TRUE.
modularity	Logical; return modularity scores? Default TRUE.
membership	Logical; return membership vector? Default TRUE.
...	Additional arguments passed to to_igraph

Value

A `cograph_communities` object with optional dendrogram
A `cograph_communities` object. See [detect_communities](#).

References

Clauset, A., Newman, M.E.J., & Moore, C. (2004). Finding community structure in very large networks. *Physical Review E*, 70, 066111.

Examples

```
g <- igraph::make_graph("Zachary")  
comm <- community_fast_greedy(g)  
membership(comm)
```

community_fluid *Fluid Communities Detection*

Description

Simulates fluid dynamics where communities compete for nodes. Requires specifying the number of communities.

Usage

```
community_fluid(x, no.of.communities, ...)
```

```
com_fl(x, no.of.communities, ...)
```

Arguments

x Network input
no.of.communities Number of communities to detect. Required.
... Additional arguments passed to [to_igraph](#)

Value

A `cograph_communities` object
A `cograph_communities` object. See [detect_communities](#).

References

Pares, F., Gasulla, D.G., Vilalta, A., Moreno, J., Ayguade, E., Labarta, J., Cortes, U., & Suzumura, T. (2018). Fluid communities: A competitive, scalable and diverse community detection algorithm. *Studies in Computational Intelligence*, 689, 229-240.

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {  
  g <- igraph::make_graph("Zachary")  
  
  # Detect exactly 2 communities  
  comm <- community_fluid(g, no.of.communities = 2)  
}  
  
m <- matrix(runif(25), 5, 5); diag(m) <- 0  
net <- as_cograph(m)  
com_fl(net, no.of.communities = 2)
```

community_infomap *Infomap Community Detection*

Description

Information-theoretic community detection based on random walk dynamics. Minimizes the map equation (description length of random walks).

Usage

```
community_infomap(
  x,
  weights = NULL,
  v.weights = NULL,
  nb.trials = 10,
  modularity = TRUE,
  seed = NULL,
  ...
)
```

```
com_im(
  x,
  weights = NULL,
  v.weights = NULL,
  nb.trials = 10,
  modularity = TRUE,
  seed = NULL,
  ...
)
```

Arguments

x	Network input
weights	Edge weights for transitions. NULL uses network weights.
v.weights	Vertex weights (teleportation weights).
nb.trials	Number of optimization trials. Default 10.
modularity	Logical; calculate modularity? Default TRUE.
seed	Random seed for reproducibility. Default NULL.
...	Additional arguments passed to to_igraph

Value

A `cograph_communities` object

A `cograph_communities` object. See [detect_communities](#).

References

Rosvall, M., & Bergstrom, C.T. (2008). Maps of random walks on complex networks reveal community structure. *PNAS*, 105(4), 1118-1123.

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")
  comm <- community_infomap(g, nb.trials = 20)
}
```

community_label_propagation

Label Propagation Community Detection

Description

Fast semi-synchronous label propagation algorithm. Each node adopts the most frequent label among its neighbors.

Usage

```
community_label_propagation(
  x,
  weights = NULL,
  mode = c("out", "in", "all"),
  initial = NULL,
  fixed = NULL,
  seed = NULL,
  ...
)

com_lp(
  x,
  weights = NULL,
  mode = c("out", "in", "all"),
  initial = NULL,
  fixed = NULL,
  seed = NULL,
  ...
)
```

Arguments

x	Network input
weights	Edge weights. NULL uses network weights, NA for unweighted.
mode	For directed graphs: "out" (default), "in", or "all".

initial	Initial labels (integer vector or NULL for unique labels).
fixed	Logical vector indicating which labels are fixed.
seed	Random seed for reproducibility. Default NULL.
...	Additional arguments passed to to_igraph

Value

A cograph_communities object

A cograph_communities object. See [detect_communities](#).

References

Raghavan, U.N., Albert, R., & Kumara, S. (2007). Near linear time algorithm to detect community structures in large-scale networks. *Physical Review E*, 76, 036106.

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")

  # Basic label propagation
  comm <- community_label_propagation(g)

  # With some nodes fixed to specific communities
  initial <- rep(NA, igraph::vcount(g))
  initial[1] <- 1 # Node 1 in community 1
  initial[34] <- 2 # Node 34 in community 2
  fixed <- !is.na(initial)
  initial[is.na(initial)] <- seq_len(sum(is.na(initial)))
  comm2 <- community_label_propagation(g, initial = initial, fixed = fixed)
}

net <- as_cograph(matrix(runif(25), 5, 5))
com_lp(net)
```

community_leading_eigenvector

Leading Eigenvector Community Detection

Description

Detects communities using the leading eigenvector of the modularity matrix. Hierarchical divisive algorithm.

Usage

```

community_leading_eigenvector(
  x,
  weights = NULL,
  steps = -1,
  start = NULL,
  options = igraph::arpack_defaults(),
  callback = NULL,
  extra = NULL,
  env = parent.frame(),
  ...
)

com_le(
  x,
  weights = NULL,
  steps = -1,
  start = NULL,
  options = igraph::arpack_defaults(),
  callback = NULL,
  extra = NULL,
  env = parent.frame(),
  ...
)

```

Arguments

x	Network input
weights	Edge weights. NULL uses network weights, NA for unweighted.
steps	Maximum number of splits. Default -1 (until modularity decreases).
start	Starting community structure (membership vector).
options	ARPACK options list. Default uses igraph::arpack_defaults().
callback	Optional callback function called after each split.
extra	Extra argument passed to callback.
env	Environment for callback evaluation.
...	Additional arguments passed to to_igraph

Value

A cograph_communities object

A cograph_communities object. See [detect_communities](#).

References

Newman, M.E.J. (2006). Finding community structure using the eigenvectors of matrices. *Physical Review E*, 74, 036104.

Examples

```
g <- igraph::make_graph("Zachary")
comm <- community_leading_eigenvector(g)
membership(comm)

net <- as_cograph(matrix(runif(25), 5, 5))
com_le(net)
```

community_leiden *Leiden Community Detection*

Description

Leiden algorithm - an improved version of Louvain that guarantees well-connected communities. Supports CPM and modularity objectives.

Usage

```
community_leiden(
  x,
  weights = NULL,
  resolution = 1,
  objective_function = c("CPM", "modularity"),
  beta = 0.01,
  initial_membership = NULL,
  n_iterations = 2,
  vertex_weights = NULL,
  seed = NULL,
  ...
)

com_ld(
  x,
  weights = NULL,
  resolution = 1,
  objective_function = c("CPM", "modularity"),
  beta = 0.01,
  initial_membership = NULL,
  n_iterations = 2,
  vertex_weights = NULL,
  seed = NULL,
  ...
)
```

Arguments

x	Network input
weights	Edge weights. NULL uses network weights, NA for unweighted.
resolution	Resolution parameter. Default 1.
objective_function	Optimization objective: "CPM" (Constant Potts Model) or "modularity". Default "CPM".
beta	Parameter for randomness in refinement step. Default 0.01.
initial_membership	Initial community assignments (optional).
n_iterations	Number of iterations. Default 2. Use -1 for convergence.
vertex_weights	Vertex weights for CPM objective.
seed	Random seed for reproducibility. Default NULL.
...	Additional arguments passed to to_igraph

Value

A `cograph_communities` object

A `cograph_communities` object. See [detect_communities](#).

References

Traag, V.A., Waltman, L., & van Eck, N.J. (2019). From Louvain to Leiden: guaranteeing well-connected communities. *Scientific Reports*, 9, 5233.

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")

  # Standard Leiden
  comm <- community_leiden(g)

  # Higher resolution for more communities
  comm2 <- community_leiden(g, resolution = 1.5)

  # Modularity objective
  comm3 <- community_leiden(g, objective_function = "modularity")
}
```

community_louvain	<i>Louvain Community Detection</i>
-------------------	------------------------------------

Description

Multi-level modularity optimization using the Louvain algorithm. Fast and widely used for large networks.

Usage

```
community_louvain(x, weights = NULL, resolution = 1, seed = NULL, ...)
```

```
com_lv(x, weights = NULL, resolution = 1, seed = NULL, ...)
```

Arguments

x	Network input
weights	Edge weights. NULL uses network weights, NA for unweighted.
resolution	Resolution parameter. Higher values = more communities. Default 1 (standard modularity).
seed	Random seed for reproducibility. Default NULL.
...	Additional arguments passed to to_igraph

Value

A `cograph_communities` object

A `cograph_communities` object. See [detect_communities](#).

References

Blondel, V.D., Guillaume, J.L., Lambiotte, R., & Lefebvre, E. (2008). Fast unfolding of communities in large networks. *Journal of Statistical Mechanics*, P10008.

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {  
  g <- igraph::make_graph("Zachary")  
  comm <- community_louvain(g)  
  membership(comm)  
  
  # Reproducible result with seed  
  comm1 <- community_louvain(g, seed = 42)  
  comm2 <- community_louvain(g, seed = 42)  
  identical(membership(comm1), membership(comm2))  
}
```

community_optimal *Optimal Community Detection*

Description

Finds the optimal community structure by maximizing modularity exactly. Very slow - only use for small networks (<50 nodes).

Usage

```
community_optimal(x, weights = NULL, ...)
```

```
com_op(x, weights = NULL, ...)
```

Arguments

x	Network input
weights	Edge weights. NULL uses network weights, NA for unweighted.
...	Additional arguments passed to to_igraph

Value

A `cograph_communities` object

A `cograph_communities` object. See [detect_communities](#).

Note

This is an NP-hard problem. Use only for tiny networks.

References

Brandes, U., Delling, D., Gaertler, M., Gorke, R., Hofer, M., Nikoloski, Z., & Wagner, D. (2008). On modularity clustering. *IEEE Transactions on Knowledge and Data Engineering*, 20(2), 172-188.

Examples

```
g <- igraph::make_ring(10)
comm <- community_optimal(g)
membership(comm)

net <- as_cograph(matrix(runif(25), 5, 5))
com_op(net)
```

community_sizes	<i>Get Community Sizes</i>
-----------------	----------------------------

Description

Get Community Sizes

Usage

```
community_sizes(x)
```

Arguments

x A cograph_communities object

Value

Named integer vector of community sizes

Examples

```
g <- igraph::make_graph("Zachary")
comm <- community_louvain(g)
community_sizes(comm)
```

community_spinglass	<i>Spinglass Community Detection</i>
---------------------	--------------------------------------

Description

Statistical mechanics approach using simulated annealing. Can handle negative edge weights.

Usage

```
community_spinglass(
  x,
  weights = NULL,
  vertex = NULL,
  spins = 25,
  parupdate = FALSE,
  start.temp = 1,
  stop.temp = 0.01,
  cool.fact = 0.99,
  update.rule = c("config", "random", "simple"),
  gamma = 1,
```

```

implementation = c("orig", "neg"),
gamma.minus = 1,
seed = NULL,
...
)

com_sg(
  x,
  weights = NULL,
  vertex = NULL,
  spins = 25,
  parupdate = FALSE,
  start.temp = 1,
  stop.temp = 0.01,
  cool.fact = 0.99,
  update.rule = c("config", "random", "simple"),
  gamma = 1,
  implementation = c("orig", "neg"),
  gamma.minus = 1,
  seed = NULL,
  ...
)

```

Arguments

x	Network input
weights	Edge weights. NULL uses network weights, NA for unweighted.
vertex	Vertex to find community for (single community mode). NULL for full partitioning.
spins	Number of spins (maximum communities). Default 25.
parupdate	Parallel update mode. Default FALSE.
start.temp	Starting temperature. Default 1.
stop.temp	Stopping temperature. Default 0.01.
cool.fact	Cooling factor. Default 0.99.
update.rule	Update rule: "config" (default), "random", or "simple".
gamma	Gamma parameter for modularity. Default 1.
implementation	"orig" (default) or "neg" (for negative weights).
gamma.minus	Gamma for negative weights in "neg" implementation.
seed	Random seed for reproducibility. Default NULL.
...	Additional arguments passed to to_igraph

Value

A `cograph_communities` object

A `cograph_communities` object. See [detect_communities](#).

References

Reichardt, J., & Bornholdt, S. (2006). Statistical mechanics of community detection. *Physical Review E*, 74, 016110.

Examples

```
g <- igraph::make_graph("Zachary")
comm <- community_spinglass(g)
membership(comm)

net <- as_cograph(matrix(runif(25), 5, 5))
com_sg(net)
```

community_walktrap *Walktrap Community Detection*

Description

Detects communities via random walks. Nodes within the same community tend to have short random walk distances.

Usage

```
community_walktrap(
  x,
  weights = NULL,
  steps = 4,
  merges = TRUE,
  modularity = TRUE,
  membership = TRUE,
  ...
)

com_wt(
  x,
  weights = NULL,
  steps = 4,
  merges = TRUE,
  modularity = TRUE,
  membership = TRUE,
  ...
)
```

Arguments

x	Network input
weights	Edge weights. NULL uses network weights, NA for unweighted.
steps	Number of random walk steps. Default 4.
merges	Logical; return merge matrix? Default TRUE.
modularity	Logical; return modularity scores? Default TRUE.
membership	Logical; return membership vector? Default TRUE.
...	Additional arguments passed to to_igraph

Value

A cograph_communities object

A cograph_communities object. See [detect_communities](#).

References

Pons, P., & Latapy, M. (2006). Computing communities in large networks using random walks. *Journal of Graph Algorithms and Applications*, 10(2), 191-218.

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")

  # Default 4 steps
  comm <- community_walktrap(g)

  # More steps for larger communities
  comm2 <- community_walktrap(g, steps = 8)
}
```

compare_communities *Compare Community Structures*

Description

Compares two community structures using various similarity measures.

Usage

```
compare_communities(
  comm1,
  comm2,
  method = c("vi", "nmi", "split.join", "rand", "adjusted.rand")
)
```

Arguments

comm1	First community structure (communities object or membership vector)
comm2	Second community structure (communities object or membership vector)
method	Comparison method: "vi" (variation of information), "nmi" (normalized mutual information), "split.join", "rand" (Rand index), "adjusted.rand"

Value

Numeric similarity/distance value

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")
  c1 <- community_louvain(g)
  c2 <- community_leiden(g)
  compare_communities(c1, c2, "nmi")
}
```

 core_periphery

Detect Core-Periphery Structure

Description

Identifies core-periphery structure in a network using either continuous (Borgatti-Everett) or discrete methods. Core nodes are densely interconnected, while periphery nodes connect primarily to the core.

Usage

```
core_periphery(
  x,
  method = c("continuous", "discrete"),
  directed = NULL,
  iter = 100,
  digits = NULL,
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
method	Character string; either "continuous" (default, Borgatti-Everett model) or "discrete" (binary core/periphery assignment).
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.

iter	Integer; maximum number of iterations for the continuous algorithm. Default 100.
digits	Integer or NULL. Round numeric outputs to this many decimal places. Default NULL (no rounding).
...	Additional arguments passed to to_igraph

Details

Continuous method (Borgatti-Everett): Finds a coreness vector c (values 0-1) that maximizes the correlation between the adjacency matrix and the ideal rank-1 pattern matrix (the outer product of the coreness vector with itself). The algorithm initializes from eigenvector centrality and iteratively refines via power iteration until convergence.

Discrete method: Produces a binary core (1) / periphery (0) assignment. Starts from the continuous solution, thresholds at the median, then greedily swaps node assignments to maximize fitness. Discrete fitness is defined by high density within the core and low density within the periphery.

Value

A list with class "cograph_core_periphery" containing:

scores Named numeric vector of continuous coreness scores (0 = periphery, 1 = core)

assignment Named character vector of "core" or "periphery" labels

fitness Correlation between adjacency matrix and ideal core-periphery pattern

core_density Edge density within the core subgraph

periphery_density Edge density within the periphery subgraph

References

Borgatti, S.P. & Everett, M.G. (2000). Models of core/periphery structures. *Social Networks*, 21(4), 375-395. doi:[10.1016/S03788733\(99\)000192](https://doi.org/10.1016/S03788733(99)000192)

See Also

[centrality](#), [network_summary](#)

Examples

```
# Core-periphery in a simple network
adj <- matrix(c(
  0, 1, 1, 1, 0,
  1, 0, 1, 1, 0,
  1, 1, 0, 1, 1,
  1, 1, 1, 0, 1,
  0, 0, 1, 1, 0
), 5, 5)
rownames(adj) <- colnames(adj) <- LETTERS[1:5]
cp <- cograph::core_periphery(adj)
cp
```

```
# Discrete assignment
cp_disc <- cograph::core_periphery(adj, method = "discrete")
cp_disc$assignment
```

degree_distribution *Degree Distribution Visualization*

Description

Creates a histogram or cumulative distribution plot of node degrees. By default, bins are integer-aligned (one bar per degree value) so each bar maps to an exact degree.

Usage

```
degree_distribution(
  x,
  mode = "all",
  directed = NULL,
  loops = TRUE,
  simplify = "sum",
  cumulative = FALSE,
  breaks = NULL,
  bins = NULL,
  bin_width = NULL,
  normalize = FALSE,
  log = "",
  main = "Degree Distribution",
  xlab = "Degree",
  ylab = NULL,
  col = "steelblue",
  border = "white",
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object.
mode	For directed networks: "all", "in", or "out". Default "all".
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.
loops	Logical. If TRUE (default), keep self-loops. Set FALSE to remove them.
simplify	How to combine multiple edges between the same node pair. Options: "sum" (default), "mean", "max", "min", or FALSE/"none" to keep multiple edges.
cumulative	Logical. If TRUE, show CCDF (complementary cumulative distribution: $P(\text{degree} \geq k)$) instead of frequency. Default FALSE.

<code>breaks</code>	Bin specification passed to <code>hist</code> . Can be a numeric vector of breakpoints, a single number giving the number of bins, or a character string naming an algorithm (e.g. "Sturges", "FD", "scott"). Overrides <code>bins</code> and <code>bin_width</code> . Default NULL (auto-detect).
<code>bins</code>	Integer. Approximate number of bins. Overrides <code>bin_width</code> . Default NULL.
<code>bin_width</code>	Numeric. Width of each bin. Default NULL (auto: 1 when the degree range is ≤ 50 , otherwise Freedman-Diaconis).
<code>normalize</code>	Logical. If TRUE, the y-axis shows proportions (bars sum to 1) instead of counts. Default FALSE.
<code>log</code>	Character. Axis log-scaling: "" (none, default), "x", "y", or "xy". For cumulative plots, "xy" produces log-log CCDF (standard for power-law inspection).
<code>main</code>	Character. Plot title. Default "Degree Distribution".
<code>xlab</code>	Character. X-axis label. Default "Degree".
<code>ylab</code>	Character. Y-axis label. Default auto-chosen based on <code>normalize</code> and <code>cumulative</code> .
<code>col</code>	Character. Bar/line fill color. Default "steelblue".
<code>border</code>	Character. Bar border color. Default "white".
<code>...</code>	Additional graphical arguments passed to <code>barplot</code> (histogram) or <code>plot</code> (cumulative).

Value

Invisibly returns a list with components:

degree Named integer vector of per-node degrees.

table Table of degree frequencies.

breaks Breakpoints used for the histogram (non-cumulative only).

counts Bin counts (non-cumulative only).

proportions Bin proportions (non-cumulative only).

Examples

```
# Basic usage - integer-aligned bins by default
adj <- matrix(c(0, 1, 1, 0,
               1, 0, 1, 1,
               1, 1, 0, 1,
               0, 1, 1, 0), 4, 4, byrow = TRUE)
cograph::degree_distribution(adj)

# Cumulative (CCDF)
cograph::degree_distribution(adj, cumulative = TRUE)

# Control bins
## Not run:
cograph::degree_distribution(large_net, bins = 15)
cograph::degree_distribution(large_net, bin_width = 5)
cograph::degree_distribution(large_net, breaks = c(0, 5, 10, 20, 50, 100))
```

```
## End(Not run)

# For directed networks
directed_adj <- matrix(c(0, 1, 0, 0,
                        0, 0, 1, 0,
                        1, 0, 0, 1,
                        0, 1, 0, 0), 4, 4, byrow = TRUE)
cograph::degree_distribution(directed_adj, mode = "in",
                             main = "In-Degree Distribution")

# With igraph
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::sample_gnp(100, 0.1)
  cograph::degree_distribution(g, col = "coral")
}
```

detect_communities *Detect Communities in a Network*

Description

Detects communities (clusters) in a network using various community detection algorithms. Returns a data frame with node-community assignments.

Usage

```
detect_communities(x, method = "louvain", directed = NULL, weights = TRUE)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object.
method	Community detection algorithm to use. One of: <ul style="list-style-type: none"> "louvain": Louvain method (default, fast and accurate) "walktrap": Walktrap algorithm based on random walks "fast_greedy": Fast greedy modularity optimization "label_prop": Label propagation algorithm "infomap": Infomap algorithm based on information flow "leiden": Leiden algorithm (improved Louvain)
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.
weights	Logical. Use edge weights for community detection. Default TRUE.

Value

A data frame with columns:

- node: Node labels/names
- community: Integer community membership

Examples

```
# Basic usage
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
detect_communities(adj)

# Different algorithm
detect_communities(adj, method = "walktrap")
```

disparity_filter *Disparity Filter*

Description

Extracts the statistically significant backbone of a weighted network using the disparity filter method (Serrano, Boguna, & Vespignani, 2009).

Usage

```
disparity_filter(x, level = 0.05, ...)

## Default S3 method:
disparity_filter(x, level = 0.05, ...)

## S3 method for class 'matrix'
disparity_filter(x, level = 0.05, ...)

## S3 method for class 'tna'
disparity_filter(x, level = 0.05, ...)

## S3 method for class 'cograph_network'
disparity_filter(x, level = 0.05, ...)

## S3 method for class 'igraph'
disparity_filter(x, level = 0.05, ...)
```

Arguments

x	A weight matrix, tna object, or cograph_network.
level	Significance level (default 0.05). Lower values result in a sparser backbone (fewer edges retained).
...	Additional arguments (currently unused).

Details

The disparity filter identifies edges that carry a disproportionate fraction of a node's total weight, based on a null model where weights are distributed uniformly at random.

For each node i with degree k_i , and each edge (i, j) with normalized weight $p_{ij} = w_{ij}/s_i$ (where s_i is the node's strength), the p-value is:

$$p = (1 - p_{ij})^{(k_i - 1)}$$

Edges are significant if $p < level$ for either endpoint.

Value

For matrices: a binary matrix (0/1) indicating significant edges. For `tna`, `cograph_network`, and `igraph` objects: a `tna_disparity` object containing the significance matrix, original weights, filtered weights, and summary statistics.

References

Serrano, M. A., Boguna, M., & Vespignani, A. (2009). Extracting the multiscale backbone of complex weighted networks. *Proceedings of the National Academy of Sciences*, 106(16), 6483-6488.

See Also

[bootstrap](#) for bootstrap-based significance testing

Examples

```
# Create a weighted network
mat <- matrix(c(
  0.0, 0.5, 0.1, 0.0,
  0.3, 0.0, 0.4, 0.1,
  0.1, 0.2, 0.0, 0.5,
  0.0, 0.1, 0.3, 0.0
), nrow = 4, byrow = TRUE)
rownames(mat) <- colnames(mat) <- c("A", "B", "C", "D")

# Extract backbone at 5% significance level
backbone <- disparity_filter(mat, level = 0.05)
backbone

# More stringent filter (1% level)
backbone_strict <- disparity_filter(mat, level = 0.01)
```

dispersion *Dispersion (Backstrom-Kleinberg 2014)*

Description

Per-pair measure of tie strength from the Facebook relationship-inference paper. For each pair (u, v) where v is a neighbor of u :

Usage

```
dispersion(x, u = NULL, v = NULL, normalized = TRUE, alpha = 1, b = 0, c = 0)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
u	Optional source node (1-based index or node name). If NULL (default), compute for all sources.
v	Optional target node. If NULL, compute for all neighbors of u.
normalized	Logical. If TRUE (default), return the normalized form; otherwise the raw count.
alpha	Numeric normalization exponent. Default 1.
b	Numeric bias added to dispersion before exponentiation. Default 0.
c	Numeric bias added to embeddedness in the denominator. Default 0.

Details

1. Let $S_T = N(u) \cap N(v)$ be their mutual friends (embeddedness).
2. Count pairs $(s, t) \subset S_T$ such that:
 - s and t are not directly connected, AND
 - s and t share no common neighbor inside $N(u)$ other than u and v .
3. The raw dispersion is this count. When `normalized = TRUE`, the result is $(\text{dispersion} + b)^\alpha / (\text{embeddedness} + c)$ (normalization is skipped when $\text{embeddedness} + c == 0$).

Matches `networkx.dispersion` bit-exact for all three call modes (single pair, single source, full matrix).

Value

- Scalar if both u and v are specified.
- Named numeric vector if exactly one of u, v is given (names are the other endpoints).
- A data frame with columns `from`, `to`, `dispersion` when neither u nor v is given (one row per ordered edge).

References

Backstrom, L., & Kleinberg, J. (2014). Romantic partnerships and the dispersion of social ties: A network analysis of relationship status on Facebook. In *Proceedings of CSCW* (pp. 831-841). ACM. <https://arxiv.org/pdf/1310.6753v1.pdf>

Examples

```
g <- igraph::make_graph("Zachary")
# Node 0 (R index 1) to node 33 (R index 34)
dispersion(g, u = 1, v = 34)
# All pairs from node 1
head(dispersion(g, u = 1))
```

edge_centrality *Calculate Edge Centrality Measures*

Description

Computes centrality measures for edges in a network and returns a tidy data frame. Unlike node centrality, these measures describe edge importance.

Usage

```
edge_centrality(
  x,
  measures = "all",
  weighted = TRUE,
  directed = NULL,
  cutoff = -1,
  invert_weights = NULL,
  alpha = 1,
  digits = NULL,
  sort_by = NULL,
  ...
)

edge_betweenness(x, ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object)
measures	Which measures to calculate. Default "all" calculates all available edge measures. Options: "betweenness", "weight", "overlap", "simmelian", "reciprocity".
weighted	Logical. Use edge weights if available. Default TRUE.
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.

cutoff	Maximum path length for betweenness. Default -1 (no limit).
invert_weights	Logical or NULL. Invert weights for path-based measures? Default NULL (auto-detect: TRUE for tna objects, FALSE otherwise).
alpha	Numeric. Exponent for weight inversion. Default 1.
digits	Integer or NULL. Round numeric columns. Default NULL.
sort_by	Character or NULL. Column to sort by (descending). Default NULL.
...	Additional arguments passed to to_igraph

Details

Edge measures available:

betweenness Number of shortest paths passing through the edge.

weight Original edge weight.

overlap Jaccard neighborhood overlap of edge endpoints.

simmelian Number of triangles the edge participates in.

reciprocity Whether the reverse edge exists (directed only). Adds columns: reciprocated, reverse_weight, weight_ratio.

Value

A data frame with columns from, to, and one column per requested measure.

Named numeric vector of edge betweenness values (named by "from->to").

Examples

```
# Create test network
mat <- matrix(c(0,1,1,0, 1,0,1,1, 1,1,0,0, 0,1,0,0), 4, 4)
rownames(mat) <- colnames(mat) <- c("A", "B", "C", "D")

# All edge measures
edge_centrality(mat)

# Just betweenness
edge_centrality(mat, measures = "betweenness")

# Sort by betweenness to find bridge edges
edge_centrality(mat, sort_by = "betweenness")
mat <- matrix(c(0,1,1,0, 1,0,1,1, 1,1,0,0, 0,1,0,0), 4, 4)
rownames(mat) <- colnames(mat) <- c("A", "B", "C", "D")
edge_betweenness(mat)
```

edge_reciprocity	<i>Edge Reciprocity</i>
------------------	-------------------------

Description

Convenience wrapper around [edge centrality](#) that returns only reciprocity information for directed networks.

Usage

```
edge_reciprocity(x, top = NULL, directed = NULL, digits = NULL, ...)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object.
top	Integer or NULL. Return only the top N edges. Default NULL.
directed	Logical or NULL. Default NULL (auto-detect).
digits	Integer or NULL. Round numeric columns. Default NULL.
...	Additional arguments passed to edge centrality .

Value

A data frame with columns: from, to, weight, reciprocated, reverse_weight, weight_ratio.

See Also

[edge centrality](#)

Examples

```
adj <- matrix(c(0, 0.8, 0, 0.3, 0, 0.5, 0.7, 0, 0), 3, 3, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
cograph::edge_reciprocity(adj, directed = TRUE)
```

estrada_index	<i>Estrada Index</i>
---------------	----------------------

Description

A graph-level spectral invariant derived from subgraph centrality:

$$EE(G) = \sum_{i=1}^n e^{\lambda_i}$$

where λ_i are the eigenvalues of the adjacency matrix. The Estrada index equals the total number of closed walks in the graph, weighted by walk length: $EE(G) = \sum_k M_k/k!$ where M_k is the number of closed walks of length k . It is the sum of subgraph centralities across all nodes.

Usage

```
estrada_index(x)
```

Arguments

x Network input (matrix, igraph, network, cograph_network, tna object).

Details

Matches `networkx.estrada_index` at machine epsilon (max relative difference $\sim 5e-15$ across random test graphs).

Value

A single numeric value — the Estrada index of the graph.

References

Estrada, E. (2000). Characterization of 3D molecular structure. *Chemical Physics Letters*, 319(5-6), 713-718.

See Also

[centrality_subgraph](#) for the per-node equivalent (sum of `subgraph_centrality(x)` equals `estrada_index(x)`).

Examples

```
# Karate club
g <- igraph::make_graph("Zachary")
estrada_index(g)
```

filter_edges

Filter Edges by Metadata

Description

Filter edges using dplyr-style expressions on any edge column. Returns a `cograph_network` object by default (universal format), or optionally the same format as input when `keep_format = TRUE`.

Usage

```
filter_edges(
  x,
  ...,
  .keep_isolates = FALSE,
  keep_format = FALSE,
  directed = NULL
)
```

```
subset_edges(
  x,
  ...,
  .keep_isolates = FALSE,
  keep_format = FALSE,
  directed = NULL
)
```

Arguments

<code>x</code>	Network input: <code>cograph_network</code> , <code>matrix</code> , <code>igraph</code> , <code>network</code> , or <code>tna</code> object.
<code>...</code>	Filter expressions using any edge column (e.g., <code>weight > 0.5</code> , <code>weight > mean(weight)</code> , <code>abs(weight) > 0.3</code>).
<code>.keep_isolates</code>	Logical. Keep nodes with no remaining edges? Default <code>FALSE</code> .
<code>keep_format</code>	Logical. If <code>TRUE</code> , return the same format as input (<code>matrix</code> returns <code>matrix</code> , <code>igraph</code> returns <code>igraph</code> , etc.). Default <code>FALSE</code> returns <code>cograph_network</code> (universal format).
<code>directed</code>	Logical or <code>NULL</code> . If <code>NULL</code> (default), auto-detect from matrix symmetry. Set <code>TRUE</code> to force directed, <code>FALSE</code> to force undirected. Only used for non- <code>cograph_network</code> inputs.

Value

A `cograph_network` object with filtered edges. If `keep_format = TRUE`, returns the same type as input (`matrix`, `igraph`, `network`, etc.).

See [filter_edges](#).

See Also

[filter_nodes](#), [splot](#), [subset_edges](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")

# Keep only strong edges (returns cograph_network)
filter_edges(adj, weight > 0.5)

# Keep format: matrix in, matrix out
filter_edges(adj, weight > 0.5, keep_format = TRUE)

# Keep edges above mean weight
splot(filter_edges(adj, weight >= mean(weight)))
```

```

# With cograph_network (pipe-friendly)
net <- as_cograph(adj)
net |>
  filter_edges(weight > 0.3) |>
  filter_nodes(degree >= 2) |>
  splot()

# Keep isolated nodes
filter_edges(net, weight > 0.7, .keep_isolates = TRUE)

# With igraph (keep_format = TRUE returns igraph)
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_ring(5)
  filter_edges(g, weight > 0, keep_format = TRUE) # Returns igraph
}

```

filter_nodes

Filter Nodes by Metadata or Centrality

Description

Filter nodes using dplyr-style expressions on any node column or centrality measure. Returns a `cograph_network` object by default (universal format), or optionally the same format as input when `keep_format = TRUE`.

Usage

```

filter_nodes(
  x,
  ...,
  .keep_edges = c("internal", "none"),
  keep_format = FALSE,
  directed = NULL
)

subset_nodes(
  x,
  ...,
  .keep_edges = c("internal", "none"),
  keep_format = FALSE,
  directed = NULL
)

```

Arguments

`x` Network input: `cograph_network`, `matrix`, `igraph`, `network`, or `tna` object.

`...` Filter expressions using any node column or centrality measure. Available variables include:

	Node columns All columns in the nodes dataframe: id, label, name, x, y, inits, color, plus any custom
	Centrality measures degree, indegree, outdegree, strength, instrength, outstrength, betweenness, closeness, eigenvector, pagerank, hub, authority
	Examples: <code>degree >= 3, label %in% c("A", "B"), pagerank > 0.1 & degree >= 2.</code>
<code>.keep_edges</code>	How to handle edges. One of: "internal" (default) Keep only edges between remaining nodes "none" Remove all edges
<code>keep_format</code>	Logical. If TRUE, return the same format as input (matrix returns matrix, igraph returns igraph, etc.). Default FALSE returns <code>cograph_network</code> (universal format).
<code>directed</code>	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected. Only used for non- <code>cograph_network</code> inputs.

Value

A `cograph_network` object with filtered nodes. If `keep_format = TRUE`, returns the same type as input (matrix, igraph, network, etc.).

See Also

[filter_edges](#), [splot](#), [subset_nodes](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")

# Keep only high-degree nodes (returns cograph_network)
filter_nodes(adj, degree >= 3)

# Keep format: matrix in, matrix out
filter_nodes(adj, degree >= 3, keep_format = TRUE)

# Filter by node label
splot(filter_nodes(adj, label %in% c("A", "C")))

# Combine centrality and metadata filters
splot(filter_nodes(adj, degree >= 2 & label != "D"))

# With cograph_network (pipe-friendly)
net <- as_cograph(adj)
net |>
```

```

filter_edges(weight > 0.3) |>
filter_nodes(degree >= 2) |>
splot()

# With igraph (keep_format = TRUE returns igraph)
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_ring(5)
  filter_nodes(g, degree >= 2, keep_format = TRUE) # Returns igraph
}

```

fit_degree_distribution

Fit Statistical Distributions to Degree Sequence

Description

Fits one or more statistical distributions to the degree sequence of a network via maximum likelihood estimation and evaluates goodness-of-fit using Kolmogorov-Smirnov tests. Returns a comparison table sorted by AIC.

Usage

```

fit_degree_distribution(
  x,
  distributions = NULL,
  mode = "all",
  directed = NULL,
  xmin = NULL,
  ...
)

```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object.
distributions	Character vector of distributions to fit. Options: "power_law", "exponential", "poisson", "geometric". Default NULL fits all four.
mode	For directed networks: "all", "in", or "out". Determines which degree to extract. Default "all".
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.
xmin	Minimum degree to include in fitting. For power-law, NULL triggers automatic estimation (Clauset et al. 2009 via igraph). For other distributions, NULL defaults to 1.
...	Additional arguments (currently unused).

Details

Power-law (Pareto Type I): $P(k) \sim k^{-\alpha}$. When `igraph` is available, uses `igraph::fit_power_law()` implementing the Clauset et al. (2009) method. Otherwise, computes the simple MLE: $\alpha = 1 + n / \sum \log(k/k_{min})$.

Exponential: $P(k) \sim e^{-\lambda k}$. MLE: $\lambda = 1/\bar{k}$.

Poisson: $P(k) \sim \lambda^k e^{-\lambda} / k!$. MLE: $\lambda = \bar{k}$. Note: the KS test uses a continuous approximation for a discrete distribution; p-values are approximate.

Geometric: $P(k) \sim (1-p)^k p$. MLE: $p = 1/(1 + \bar{k})$.

Value

An object of class "cograph_degree_fit" containing:

fits Named list, one entry per distribution, each with: `distribution`, `parameters` (named list of fitted params), `loglik`, `aic`, `bic`, `ks_stat`, `ks_p`.

comparison Data frame sorted by AIC with columns: `distribution`, `aic`, `bic`, `ks_stat`, `ks_p`.

best Name of the best-fitting distribution (lowest AIC).

degree The degree vector used for fitting.

References

Clauset, A., Shalizi, C. R., & Newman, M. E. J. (2009). Power-law distributions in empirical data. *SIAM Review*, 51(4), 661–703.

See Also

[degree_distribution](#), [centrality](#)

Examples

```
adj <- matrix(c(0, 1, 1, 0, 0,
               1, 0, 1, 1, 0,
               1, 1, 0, 1, 1,
               0, 1, 1, 0, 1,
               0, 0, 1, 1, 0), 5, 5, byrow = TRUE)
rownames(adj) <- colnames(adj) <- LETTERS[1:5]
fit <- cograph::fit_degree_distribution(adj,
  distributions = c("exponential", "poisson"))
print(fit)
```

from_qgraph	<i>Convert a qgraph object to cograph parameters</i>
-------------	--

Description

Extracts the network, layout, and all relevant arguments from a qgraph object and passes them to a cograph plotting engine. Reads resolved values from graphAttributes rather than raw Arguments.

Usage

```
from_qgraph(
  qgraph_object,
  engine = c("splot", "soplot"),
  plot = TRUE,
  weight_digits = 2,
  show_zero_edges = FALSE,
  ...
)
```

Arguments

qgraph_object	Return value of qgraph::qgraph()
engine	Which cograph renderer to use: "splot" or "soplot". Default: "splot".
plot	Logical. If TRUE (default), immediately plot using the chosen engine.
weight_digits	Number of decimal places to round edge weights to. Default 2. Edges that round to zero are removed unless show_zero_edges = TRUE.
show_zero_edges	Logical. If TRUE, keep edges even if their weight rounds to zero. Default: FALSE.
...	Override any extracted parameter. Use qgraph-style names (e.g., minimum) or cograph names (e.g., threshold).

Details

Parameter Mapping:

The following qgraph parameters are automatically extracted and mapped to cograph equivalents:

Node properties:

- labels/names -> labels
- color -> node_fill
- width -> node_size (scaled by 1.3x)
- shape -> node_shape (mapped to cograph equivalents)
- border.color -> node_border_color
- border.width -> node_border_width
- label.cex -> label_size

- label.color -> label_color

Edge properties:

- labels -> edge_labels
- label.cex -> edge_label_size (scaled by 0.5x)
- lty -> edge_style (numeric to name conversion)
- curve -> curvature
- asize -> arrow_size (scaled by 0.3x)

Graph properties:

- minimum -> threshold
- maximum -> maximum
- groups -> groups
- directed -> directed
- posCol/negCol -> edge_positive_color/edge_negative_color

Pie/Donut:

- pie values -> donut_fill with donut_inner_ratio=0.8
- pieColor -> donut_color

Important Notes:

- **edge_color and edge_width are NOT extracted** because qgraph bakes its cut-based fading into these vectors, producing near-invisible edges. cograph applies its own weight-based styling instead.
- The cut parameter is also not passed because it causes faint edges with hanging labels.
- Layout coordinates from qgraph are preserved with rescale=FALSE.
- If you override layout, rescale is automatically re-enabled.

Value

Invisibly, a named list of cograph parameters that can be passed to `splot()` or `soplot()`.

See Also

[cograph](#) for creating networks from scratch, [splot](#) and [soplot](#) for plotting engines, [from_tna](#) for tna object conversion

Examples

```
# Convert and plot a qgraph object
adj <- matrix(c(0, .5, .3, .5, 0, .4, .3, .4, 0), 3, 3)
q <- qgraph::qgraph(adj)
from_qgraph(q) # Plots with splot

# Use soplot engine instead
from_qgraph(q, engine = "soplot")

# Override extracted parameters
from_qgraph(q, node_fill = "steelblue", layout = "circle")
```

```
# Extract parameters without plotting
params <- from_qgraph(q, plot = FALSE)
names(params) # See what was extracted

# Works with themed qgraph objects
q_themed <- qgraph::qgraph(adj, theme = "colorblind", posCol = "blue")
from_qgraph(q_themed)
```

from_tna

Convert a tna object to cograph parameters

Description

Extracts the transition matrix, labels, and initial state probabilities from a tna object and plots with cograph. Initial probabilities are mapped to donut fills.

Usage

```
from_tna(
  tna_object,
  engine = c("splot", "soplot"),
  plot = TRUE,
  weight_digits = NULL,
  show_zero_edges = FALSE,
  ...
)
```

Arguments

tna_object	A tna object from tna::tna()
engine	Which cograph renderer to use: "splot" or "soplot". Default: "splot".
plot	Logical. If TRUE (default), immediately plot using the chosen engine.
weight_digits	Number of decimal places to round edge weights to. Default 2. Edges that round to zero are removed unless show_zero_edges = TRUE.
show_zero_edges	Logical. If TRUE, keep edges even if their weight rounds to zero. Default: FALSE.
...	Additional parameters passed to the plotting engine (e.g., layout, node_fill, donut_color).

Details

Conversion Process:

The tna object's transition matrix becomes edge weights, labels become node labels, and initial state probabilities (inits) are mapped to donut_fill values to visualize starting state distributions.

TNA networks are always treated as directed because transition matrices represent directional state changes.

The default donut_inner_ratio of 0.8 creates thin rings that effectively visualize probability values without obscuring node labels.

Parameter Mapping:

The following tna properties are automatically extracted:

- **weights:** Transition matrix -> edge weights
- **labels:** State labels -> node labels
- **inits:** Initial probabilities -> donut_fill (0-1 scale)

TNA Visual Defaults:

The following visual defaults are applied for TNA plots (all can be overridden via ...):

- layout = "oval": Oval/elliptical node arrangement
- node_fill: Colors from TNA palette (Accent/Set3 based on state count)
- node_size = 7: Larger nodes for readability
- arrow_size = 0.61: Prominent directional arrows
- edge_color = "#003355": Dark blue edges
- edge_labels = TRUE: Show transition weights on edges
- edge_label_size = 0.6: Readable edge labels
- edge_label_position = 0.7: Labels positioned toward target
- edge_start_style = "dotted": Dotted line at edge source
- edge_start_length = 0.2: 20% of edge is dotted

Value

Invisibly, a named list of cograph parameters that can be passed to `splot()` or `soplot()`.

See Also

[cograph](#) for creating networks from scratch, [splot](#) and [soplot](#) for plotting engines, [from_qgraph](#) for qgraph object conversion

Examples

```
# Convert and plot a tna object
model <- tna::tna(tna::group_regulation)
from_tna(model) # Plots with donut rings showing initial probabilities

# Use soplot engine instead
from_tna(model, engine = "soplot")
```

```
# Customize the visualization
from_tna(model, layout = "circle", donut_color = c("steelblue", "gray90"))

# Extract parameters without plotting
params <- from_tna(model, plot = FALSE)
# Modify and plot manually
params$node_fill <- "coral"
do.call(splot, params)
```

get_data

Get Original Data from Cograph Network

Description

Extracts the original estimation data stored in a `cograph_network` object. This is the raw input data (e.g., sequence matrix from `tna`, edge list data frame) preserved for reference.

Usage

```
get_data(x)
```

Arguments

`x` A `cograph_network` object.

Value

The original data object, or `NULL` if not stored.

See Also

[as_cograph](#), [get_meta](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
get_data(net) # NULL (matrices don't store raw data)
```

get_edges	<i>Get Edges from Cograph Network</i>
-----------	---------------------------------------

Description

Extracts the edges data frame from a `cograph_network` object.

Usage

```
get_edges(x)
```

Arguments

`x` A `cograph_network` object.

Value

A data frame with columns: `from`, `to`, `weight`.

See Also

[as_cograph](#), [n_edges](#), [get_nodes](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
get_edges(net)
```

get_edge_list	<i>Extract Raw Edge List from TNA Model</i>
---------------	---

Description

Extract individual-level transition counts as an edge list from a `tna` object.

Usage

```
get_edge_list(x, by_individual = TRUE, drop_zeros = TRUE)
```

Arguments

`x` A `tna` object created by `tna::tna()`

`by_individual` Logical. If `TRUE` (default), returns edge list with individual IDs. If `FALSE`, aggregates across all individuals.

`drop_zeros` Logical. If `TRUE` (default), excludes edges with zero count.

Value

A data frame with columns:

id Individual identifier (only if `by_individual = TRUE`)

from Source state label

to Target state label

count Number of transitions

See Also

[extract_motifs\(\)](#) for motif analysis using edge lists

Other motifs: [extract_motifs\(\)](#), [extract_triads\(\)](#), [motif_census\(\)](#), [motifs\(\)](#), [plot.cograph_motif_analysis\(\)](#), [plot.cograph_motifs\(\)](#), [subgraphs\(\)](#), [triad_census\(\)](#)

Examples

```
Mod <- tna::tna(tna::group_regulation)

# Get edge list by individual
edges <- get_edge_list(Mod)
head(edges)

# Aggregate across individuals
agg_edges <- get_edge_list(Mod, by_individual = FALSE)
```

get_groups

Get Node Groups from Cograph Network

Description

Extracts the node groupings from a `cograph_network` object.

Usage

```
get_groups(x)
```

Arguments

`x` A `cograph_network` object.

Value

A data frame with node groupings, or NULL if not set. The data frame has columns:

- `node`: Node labels
- One of `layer`, `cluster`, or `group`: Group assignment

See Also

[set_groups](#), [splot](#)

Examples

```
mat <- matrix(runif(25), 5, 5)
rownames(mat) <- colnames(mat) <- LETTERS[1:5]
net <- as_cograph(mat)
net <- set_groups(net, list(G1 = c("A", "B"), G2 = c("C", "D", "E")))
get_groups(net)
```

get_labels

Get Labels from Cograph Network

Description

Extracts the node labels vector from a `cograph_network` object.

Usage

```
get_labels(x)
```

Arguments

`x` A `cograph_network` object.

Value

A character vector of node labels.

See Also

[as_cograph](#), [get_nodes](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
get_labels(net)
```

get_layout	<i>Get a Registered Layout</i>
------------	--------------------------------

Description

Get a Registered Layout

Usage

```
get_layout(name)
```

Arguments

name	Character. Name of the layout.
------	--------------------------------

Value

The layout function, or NULL if not found.

Examples

```
get_layout("circle")
```

get_meta	<i>Get Metadata from Cograph Network</i>
----------	--

Description

Extracts the consolidated metadata list from a `cograph_network` object. The metadata contains source type, layout info, and TNA metadata.

Usage

```
get_meta(x)
```

Arguments

x	A <code>cograph_network</code> object.
---	--

Value

A list with components:

source Character string indicating input type

layout List with layout name and seed, or NULL

tna List with TNA metadata (type, group_name, group_index), or NULL

See Also

[as_cograph](#), [get_source](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
get_meta(net)
```

get_nodes

Get Nodes from Cograph Network

Description

Extracts the nodes data frame from a `cograph_network` object.

Usage

```
get_nodes(x)
```

Arguments

`x` A `cograph_network` object.

Value

A data frame with columns: `id`, `label`, `name`, `x`, `y` (and possibly others).

See Also

[as_cograph](#), [n_nodes](#), [get_edges](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
get_nodes(net)
```

get_shape	<i>Get a Registered Shape</i>
-----------	-------------------------------

Description

Get a Registered Shape

Usage

```
get_shape(name)
```

Arguments

name	Character. Name of the shape.
------	-------------------------------

Value

The shape drawing function, or NULL if not found.

Examples

```
get_shape("circle")
```

get_source	<i>Get Source Type from Cograph Network</i>
------------	---

Description

Extracts the source type string from a `cograph_network` object's metadata.

Usage

```
get_source(x)
```

Arguments

x	A <code>cograph_network</code> object.
---	--

Value

A character string indicating the input type (e.g., "matrix", "tna", "igraph", "edgelist"), or "unknown" if not set.

See Also

[as_cograph](#), [get_meta](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
get_source(net) # "matrix"
```

get_theme	<i>Get a Registered Theme</i>
-----------	-------------------------------

Description

Get a Registered Theme

Usage

```
get_theme(name)
```

Arguments

name Character. Name of the theme.

Value

The theme object, or NULL if not found.

Examples

```
get_theme("classic")
```

ggplot_robustness	<i>Compare Network Robustness (ggplot2)</i>
-------------------	---

Description

Creates a ggplot2 faceted visualization comparing robustness across multiple networks. Produces publication-quality figures similar to those in Nature Scientific Reports.

Usage

```
ggplot_robustness(
  ...,
  networks = NULL,
  measures = c("betweenness", "degree", "random"),
  strategy = "sequential",
  colors = NULL,
  title = NULL,
  n_iter = 1000,
```

```

    seed = NULL,
    type = "vertex",
    ncol = NULL,
    free_y = FALSE
  )

```

Arguments

...	Named arguments: network names as names, network objects as values.
networks	Named list of networks (alternative to ...).
measures	Attack strategies to compare. Default c("betweenness", "degree", "random").
strategy	Character string; "sequential" (default) recalculates centrality after each removal, "static" uses initial centrality ranking throughout.
colors	Named vector of colors for measures.
title	Overall title. Default NULL.
n_iter	Iterations for random. Default 1000.
seed	Random seed. Default NULL.
type	Removal type. Default "vertex".
ncol	Columns in facet. Default NULL (auto).
free_y	If TRUE, allow different y-axis scales per facet. Default FALSE.

Value

A ggplot2 object.

Examples

```

if (requireNamespace("igraph", quietly = TRUE) &&
    requireNamespace("ggplot2", quietly = TRUE)) {

  g1 <- igraph::sample_pa(40, m = 2, directed = FALSE)
  g2 <- igraph::sample_gnp(40, 0.15)

  ggplot_robustness(
    "Teaching network" = g1,
    "Collaborative network" = g2,
    n_iter = 20
  )
}

```

group centrality *Group Centrality (Everett-Borgatti 1999)*

Description

Group centrality measures the importance of a *set* of nodes $C \subseteq V$ rather than a single node. Three variants are supported:

Usage

```
group centrality(
  x,
  nodes,
  measure = c("betweenness", "closeness", "degree"),
  mode = c("all", "out", "in"),
  normalized = TRUE
)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
nodes	Integer vector of node indices (1-based) or character vector of node names identifying the group C .
measure	One of "betweenness", "closeness", "degree".
mode	For directed graphs with measure = "degree": "all" (both directions), "out" (outgoing), or "in" (incoming). Ignored for undirected graphs and other measures.
normalized	Logical, for "betweenness" only. If TRUE (default), divide by $(V - C)(V - C - 1)$.

Details

betweenness $GBC(C) = \sum_{s,t \in V \setminus C, s \neq t} \sigma(s, t | C) / \sigma(s, t)$, where $\sigma(s, t)$ is the number of shortest s - t paths and $\sigma(s, t | C)$ is the number of those paths passing through at least one node in C . Normalized by $1 / ((|V| - |C|)(|V| - |C| - 1))$.

closeness $GCC(C) = (|V| - |C|) / \sum_{v \in V \setminus C} d(v, C)$, where $d(v, C) = \min_{c \in C} d(v, c)$ is the shortest distance from v to any group member. Unreachable nodes contribute 0 to the denominator sum (matching NetworkX convention). For directed graphs, cograph uses $d(v, c)$ in the original direction, equivalent to NetworkX's "reverse then multi-source".

degree $GDC(C) = |N(C) \setminus C| / (|V| - |C|)$, the fraction of non-group nodes adjacent to at least one group member. mode = "in" / "out" pick the corresponding directed neighborhood.

Value

A single numeric scalar — the group centrality of the set nodes.

Divergence from NetworkX on betweenness

`networkx.group_betweenness_centrality` uses the Puzis-Yahalom-Elovici iterative algorithm, which produces results that diverge from the textbook Everett-Borgatti / Puzis 2008 "at least one node in C" definition on some graph topologies (verified via an independent Python brute-force). `cograph` implements the textbook formula directly; `group_closeness` and `group_degree` match NetworkX exactly.

References

Everett, M. G., & Borgatti, S. P. (1999). The centrality of groups and classes. *Journal of Mathematical Sociology*, 23(3), 181-201.

Puzis, R., Yahalom, R., & Elovici, Y. (2008). Augmentative data collection for betweenness centrality. In *Advances in Social Networks Analysis and Mining* (pp. 196-200). IEEE.

See Also

[centrality](#) for per-node measures.

Examples

```
g <- igraph::make_graph("Zachary")
group_centrality(g, nodes = c(1, 2, 3), measure = "betweenness")
group_centrality(g, nodes = c(1, 2, 3), measure = "closeness")
group_centrality(g, nodes = c(1, 2, 3), measure = "degree")
```

hai_datasets

Human-AI Interaction Coding Sequences

Description

Coded sequences of human-AI programming interactions from 34 projects across 429 sessions. Actions are coded at two granularity levels (broad categories vs fine-grained codes) and split by actor (Human, AI, or both combined). Each row is one session (project + session_id); columns T1, T2, ... hold the sequential actions. NA indicates the session ended before that time step.

Usage

`coding`

`coding_detailed`

`ai_coding`

`ai_detailed`

`human_ai`

`human_ai_detailed`

Format

coding 429 x 164 data.frame. Human actions by category (9 states: Command, Correct, Frustrate, Inquire, Interrupt, Refine, Request, Specify, Verify).

coding_detailed 429 x 164 data.frame. Human actions by fine-grained code (15 states: Accept, Arguing, Ask, Command, Context, Correction, Direct, Frustration, Interrupt, Refinement, Reject, Request, Specification, Thinking, Verification).

ai_coding 428 x 138 data.frame. AI actions by category (8 states: Ask, Delegate, Execute, Explain, Investigate, Plan, Repair, Report).

ai_detailed 428 x 138 data.frame. AI actions by fine-grained code (18 states: Acknowledge, Apologize, Ask, Comply, Delegate, Diagnose, Escape, Execute, Explain, Hedge, Investigate, Plan, Refuse, Report, Retry, Scaffold, Suggest, Warn).

human_ai 429 x 302 data.frame. Both actors combined, by category (17 states).

human_ai_detailed 429 x 302 data.frame. Both actors combined, by fine-grained code (32 states).

An object of class `data.frame` with 429 rows and 164 columns.

An object of class `data.frame` with 429 rows and 164 columns.

An object of class `data.frame` with 428 rows and 138 columns.

An object of class `data.frame` with 428 rows and 138 columns.

An object of class `data.frame` with 429 rows and 287 columns.

An object of class `data.frame` with 429 rows and 287 columns.

Value

A `data.frame` where each row is one session. The first columns identify the session; the remaining columns (T1, T2, ...) hold the sequential action codes, with NA indicating the session ended before that time step. Six variants are provided: `coding` (human actions by category, 9 states), `coding_detailed` (human actions by fine-grained code, 15 states), `ai_coding` (AI actions by category, 8 states), `ai_detailed` (AI actions by fine-grained code, 18 states), `human_ai` (both actors by category, 17 states), and `human_ai_detailed` (both actors by fine-grained code, 32 states).

Source

Human-AI programming interaction study, 34 projects, 429 sessions.

Examples

```
data(coding)
head(coding[, 1:6])
dim(coding)
```

is_bipartite

Check if a Matrix Could Be Bipartite

Description

Tests whether a matrix could represent a bipartite incidence matrix. A non-square matrix is considered bipartite by default. For square matrices, checks whether the corresponding graph has bipartite structure (i.e., nodes can be partitioned into two groups with edges only between groups).

Usage

```
is_bipartite(x)
```

Arguments

x A numeric matrix.

Details

For non-square matrices, returns TRUE since they naturally represent two-mode data (rows and columns are distinct node types).

For square matrices, the function checks whether the corresponding undirected graph is bipartite by attempting a two-coloring via `igraph::is_bipartite()` when `igraph` is available. Without `igraph`, it uses a BFS-based two-coloring algorithm.

Value

Logical. TRUE if the matrix could represent a bipartite network, FALSE otherwise.

Examples

```
# Non-square matrix is bipartite
inc <- matrix(c(1, 0, 1, 1, 1, 0), 2, 3)
cograph::is_bipartite(inc)

# Square bipartite-compatible adjacency
adj <- matrix(c(0, 0, 1, 1,
               0, 0, 1, 0,
               1, 1, 0, 0,
               1, 0, 0, 0), 4, 4, byrow = TRUE)
cograph::is_bipartite(adj)

# Non-bipartite (triangle)
tri <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
cograph::is_bipartite(tri)
```

is_directed	<i>Check if Network is Directed</i>
-------------	-------------------------------------

Description

Checks whether a `cograph_network` is directed.

Usage

```
is_directed(x)
```

Arguments

`x` A `cograph_network` object.

Value

Logical: TRUE if directed, FALSE if undirected.

See Also

[as_cograph](#)

Examples

```
# Symmetric matrix -> undirected
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
cograph::is_directed(net) # FALSE

# Asymmetric matrix -> directed
mat2 <- matrix(c(0, 1, 0, 0, 0, 1, 0, 0, 0), nrow = 3)
net2 <- as_cograph(mat2)
cograph::is_directed(net2) # TRUE
```

is_tna_network	<i>Check if Network is TNA-based</i>
----------------	--------------------------------------

Description

Checks whether a `cograph_network` was created from a `tna` or `group_tna` object.

Usage

```
is_tna_network(x)
```

Arguments

x A `cograph_network` object.

Value

Logical: TRUE if the network was created from a TNA object, FALSE otherwise.

See Also

[as_cograph](#)

Examples

```
# Non-TNA network
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
is_tna_network(net) # FALSE
```

```
model <- tna::tna(tna::group_regulation)
net_tna <- as_cograph(model)
is_tna_network(net_tna) # TRUE
```

k_shortest_paths *Find K Shortest Loopless Paths (Yen's Algorithm)*

Description

Computes up to k shortest loopless paths between two nodes using Yen's algorithm. Each path is a sequence of distinct nodes from source to target.

Usage

```
k_shortest_paths(x, from, to, k = 3, weights = NULL, directed = NULL, ...)
```

Arguments

x Network input: `matrix`, `igraph`, `network`, `cograph_network`, or `tna` object

from Character or numeric node identifier for the source node.

to Character or numeric node identifier for the target node.

k Integer; number of shortest paths to find. Default 3.

weights Edge weight handling: `NULL` (default) auto-detects from edge attributes, `NA` forces unweighted distances, or a numeric vector of custom weights.

directed Logical or `NULL`. If `NULL` (default), auto-detect from matrix symmetry. Set `TRUE` to force directed, `FALSE` to force undirected.

... Additional arguments passed to [to_igraph](#)

Details

Yen's algorithm finds the k shortest loopless (simple) paths in a graph. It works by:

1. Finding the shortest path via Dijkstra's algorithm
2. For each subsequent path, systematically exploring deviations from previously found paths by temporarily removing edges, finding spur paths, and selecting the shortest candidate

The algorithm may return fewer than k paths if fewer distinct loopless paths exist between the two nodes.

Value

A list with class "cograph_k_paths" containing:

paths List of up to k character vectors, each containing node names in path order

distances Numeric vector of path lengths (sum of edge weights or hop count)

from Source node name

to Target node name

k Number of paths requested

References

Yen, J.Y. (1971). Finding the K shortest loopless paths in a network. *Management Science*, 17(11), 712-716. doi:10.1287/mnsc.17.11.712

See Also

[shortest_paths](#)

Examples

```
# Find 3 shortest paths in a small network
adj <- matrix(c(
  0, 1, 1, 0, 0,
  0, 0, 1, 1, 0,
  0, 0, 0, 1, 1,
  0, 0, 0, 0, 1,
  0, 0, 0, 0, 0
), 5, 5, byrow = TRUE)
rownames(adj) <- colnames(adj) <- LETTERS[1:5]
kp <- cograph::k_shortest_paths(adj, from = "A", to = "E", k = 3)
kp
```

 layer_degree_correlation

Degree Correlation Between Layers

Description

Measures hub consistency across layers via degree correlation.

Usage

```
layer_degree_correlation(layers, mode = c("total", "in", "out"))
ldegcor(layers, mode = c("total", "in", "out"))
```

Arguments

layers	List of adjacency matrices
mode	Degree type: "total", "in", "out"

Value

Correlation matrix between layer degree sequences

Examples

```
mat1 <- matrix(c(0, 1, 0, 1, 0, 1, 0, 1, 0), 3, 3)
mat2 <- matrix(c(0, 0, 1, 1, 0, 0, 0, 1, 0), 3, 3)
layers <- list(L1 = mat1, L2 = mat2)
layer_degree_correlation(layers, mode = "total")
```

 layer_similarity

Layer Similarity

Description

Computes similarity between two network layers.

Usage

```
layer_similarity(
  A1,
  A2,
  method = c("jaccard", "overlap", "hamming", "cosine", "pearson")
)

lsim(A1, A2, method = c("jaccard", "overlap", "hamming", "cosine", "pearson"))
```

Arguments

A1	First adjacency matrix
A2	Second adjacency matrix
method	Similarity method: "jaccard", "overlap", "hamming", "cosine", "pearson"

Value

Numeric similarity value

Examples

```
A1 <- matrix(c(0,1,1,0, 1,0,0,1, 1,0,0,1, 0,1,1,0), 4, 4)
A2 <- matrix(c(0,1,0,0, 1,0,1,0, 0,1,0,1, 0,0,1,0), 4, 4)

layer_similarity(A1, A2, "jaccard") # Edge overlap
layer_similarity(A1, A2, "cosine") # Weight similarity
```

```
layer_similarity_matrix
      Pairwise Layer Similarities
```

Description

Computes similarity matrix for all pairs of layers.

Usage

```
layer_similarity_matrix(
  layers,
  method = c("jaccard", "overlap", "cosine", "pearson")
)

lsim_matrix(layers, method = c("jaccard", "overlap", "cosine", "pearson"))
```

Arguments

layers	List of adjacency matrices (one per layer)
method	Similarity method

Value

Symmetric matrix of pairwise similarities

Examples

```
# layers <- list(T1 = mat1, T2 = mat2, T3 = mat3)
# layer_similarity_matrix(layers, "cosine")
```

layout_circle	<i>Circular Layout</i>
---------------	------------------------

Description

Arrange nodes evenly spaced around a circle.

Usage

```
layout_circle(network, order = NULL, start_angle = pi/2, clockwise = TRUE, ...)
```

Arguments

network	A CographNetwork object.
order	Optional vector specifying node order (indices or labels).
start_angle	Starting angle in radians (default: pi/2 for top).
clockwise	Logical. Arrange nodes clockwise? Default TRUE.
...	Additional arguments (ignored).

Value

Data frame with x, y coordinates.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- CographNetwork$new(adj)
coords <- layout_circle(net)
```

layout_groups	<i>Group-based Layout</i>
---------------	---------------------------

Description

Arrange nodes based on group membership. Groups are positioned in a circular arrangement around the center, with nodes within each group also arranged in a circle.

Usage

```
layout_groups(
  network,
  groups,
  group_positions = NULL,
  inner_radius = 0.15,
  outer_radius = 0.35
)
```

Arguments

network	A CographNetwork object.
groups	Vector specifying group membership for each node. Can be numeric, character, or factor.
group_positions	Optional list or data frame with x, y coordinates for each group center.
inner_radius	Radius of nodes within each group (default: 0.15).
outer_radius	Radius for positioning group centers (default: 0.35).

Value

Data frame with x, y coordinates.

Examples

```
# Create a network with groups
adj <- matrix(0, 9, 9)
adj[1, 2:3] <- 1; adj[2:3, 1] <- 1 # Group 1
adj[4, 5:6] <- 1; adj[5:6, 4] <- 1 # Group 2
adj[7, 8:9] <- 1; adj[8:9, 7] <- 1 # Group 3
net <- CographNetwork$new(adj)
groups <- c(1, 1, 1, 2, 2, 2, 3, 3, 3)
coords <- layout_groups(net, groups)
```

layout_oval

Oval Layout

Description

Arrange nodes evenly spaced around an ellipse. This creates an oval-shaped network layout that is wider than it is tall (or vice versa depending on ratio).

Usage

```
layout_oval(
  network,
  ratio = 1.5,
  order = NULL,
  start_angle = pi/2,
  clockwise = TRUE,
  rotation = 0,
  ...
)
```

Arguments

network	A CographNetwork object.
ratio	Aspect ratio (width/height). Values > 1 create horizontal ovals, values < 1 create vertical ovals. Default 1.5.
order	Optional vector specifying node order (indices or labels).
start_angle	Starting angle in radians (default: pi/2 for top).
clockwise	Logical. Arrange nodes clockwise? Default TRUE.
rotation	Rotation angle in radians to tilt the entire oval. Default 0.
...	Additional arguments (ignored).

Value

Data frame with x, y coordinates.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- CographNetwork$new(adj)
coords <- layout_oval(net, ratio = 1.5)
```

layout_spring

Fruchterman-Reingold Spring Layout

Description

Compute node positions using the Fruchterman-Reingold force-directed algorithm. Nodes connected by edges are attracted to each other while all nodes repel each other.

Usage

```
layout_spring(
  network,
  iterations = 200,
  cooling = 0.95,
  repulsion = 1.5,
  attraction = 1,
  seed = NULL,
  initial = NULL,
  max_displacement = NULL,
  anchor_strength = 0,
  area = 1.5,
  gravity = 0,
  init = c("random", "circular"),
  cooling_mode = c("exponential", "vcf", "linear"),
  ...
)
```

Arguments

network	A CographNetwork object.
iterations	Number of iterations (default: 200).
cooling	Rate of temperature decrease for exponential cooling (default: 0.95).
repulsion	Repulsion constant (default: 1.5).
attraction	Attraction constant (default: 1).
seed	Random seed for reproducibility.
initial	Optional initial coordinates (matrix or data frame). For animations, pass the previous frame's layout to ensure smooth transitions.
max_displacement	Maximum distance a node can move from its initial position (default: NULL = no limit). Useful for animations to prevent large jumps between frames. Values like 0.05-0.1 work well.
anchor_strength	Strength of force pulling nodes toward initial positions (default: 0). Higher values (e.g., 0.5-2) keep nodes closer to their starting positions. Only applies when initial is provided.
area	Area parameter controlling node spread (default: 1.5). Higher values spread nodes further apart.
gravity	Gravity force pulling nodes toward center (default: 0). Higher values (e.g., 0.5-2) prevent nodes from drifting apart.
init	Initialization method: "random" (default) or "circular".
cooling_mode	Cooling schedule: "exponential" (default, uses cooling parameter), "vcf" (Variable Cooling Factor - adapts based on movement), or "linear" (linear decrease over iterations).
...	Additional arguments (ignored).

Value

Data frame with x, y coordinates.

Examples

```
adj <- matrix(c(0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 0), nrow = 4)
net <- CographNetwork$new(adj)
coords <- layout_spring(net, seed = 42)

# For animations: use previous layout as initial with constraints
coords2 <- layout_spring(net, initial = coords, max_displacement = 0.05)

# With gravity to keep nodes centered
coords3 <- layout_spring(net, gravity = 0.5, area = 2, seed = 42)

# With circular initialization and VCF cooling
coords4 <- layout_spring(net, init = "circular", cooling_mode = "vcf", seed = 42)
```

list_layouts	<i>List Available Layouts</i>
--------------	-------------------------------

Description

List Available Layouts

Usage

```
list_layouts()
```

Value

Character vector of registered layout names.

Examples

```
list_layouts()
```

list_palettes	<i>List Available Color Palettes</i>
---------------	--------------------------------------

Description

Returns the names of all registered color palettes.

Usage

```
list_palettes()
```

Value

Character vector of palette names.

Examples

```
list_palettes()
```

<code>list_shapes</code>	<i>List Available Shapes</i>
--------------------------	------------------------------

Description

List Available Shapes

Usage

```
list_shapes()
```

Value

Character vector of registered shape names.

Examples

```
list_shapes()
```

<code>list_svg_shapes</code>	<i>List Registered SVG Shapes</i>
------------------------------	-----------------------------------

Description

Get names of all registered custom SVG shapes.

Usage

```
list_svg_shapes()
```

Value

Character vector of registered shape names.

Examples

```
list_svg_shapes()
```

<code>list_themes</code>	<i>List Available Themes</i>
--------------------------	------------------------------

Description

List Available Themes

Usage

```
list_themes()
```

Value

Character vector of registered theme names.

Examples

```
list_themes()
```

<code>membership</code>	<i>Get Community Membership</i>
-------------------------	---------------------------------

Description

Extracts a named membership vector from a communities result. Works with both `cograph_communities` data frames and `igraph_communities` objects.

Usage

```
membership(x)
```

Arguments

`x` A `cograph_communities` or `igraph_communities` object.

Value

Named integer vector of community assignments.

Examples

```
g <- igraph::make_graph("Zachary")
comm <- community_louvain(g)
membership(comm)
```

Description

Two modes of motif analysis for networks:

- **Census** (`named_nodes = FALSE`, default): Counts MAN type frequencies with significance testing. Nodes are exchangeable.
- **Instances** (`named_nodes = TRUE`, or use `subgraphs()`): Lists specific node triples forming each pattern. Nodes are NOT exchangeable.

Usage

```
motifs(  
  x,  
  named_nodes = FALSE,  
  actor = NULL,  
  window = NULL,  
  window_type = c("rolling", "tumbling"),  
  pattern = c("triangle", "network", "closed", "all"),  
  include = NULL,  
  exclude = NULL,  
  significance = TRUE,  
  n_perm = 1000L,  
  min_count = if (named_nodes) 5L else NULL,  
  edge_method = c("any", "expected", "percent"),  
  edge_threshold = 1.5,  
  min_transitions = 5,  
  top = NULL,  
  seed = NULL  
)  
  
## S3 method for class 'cograph_motif_result'  
print(x, ...)  
  
## S3 method for class 'cograph_motif_result'  
plot(  
  x,  
  type = c("triads", "types", "significance", "patterns"),  
  n = 15,  
  ncol = 5,  
  colors = c("#2166AC", "#B2182B"),  
  ...  
)
```

Arguments

x	Input data: a tna object, cograph_network, matrix, igraph, or data.frame (edge list).
named_nodes	Logical. If FALSE (default), performs census (type-level counts). If TRUE, extracts specific node triples (instance-level). subgraphs() is a convenience wrapper that sets this to TRUE.
actor	Character. Column name in the edge list metadata to group by. If NULL (default), auto-detects standard column names (session_id, session, actor, user, participant). If no grouping column found, performs aggregate analysis.
window	Numeric. Window size for windowed analysis. Splits each actor's transitions into windows of this size. NULL (default) means no windowing.
window_type	Character. Window type: "rolling" (default) or "tumbling". Only used when window is set.
pattern	Which MAN triad types to include in the analysis: "triangle" (default) Only the 7 closed triangle types: 030C, 030T, 120C, 120D, 120U, 210, 300. Excludes trivial open patterns (empty triads, single edges, chains, stars, mutual pairs). "network" All types except trivially open ones. Excludes 003 (empty), 012 (single edge), 021C (chain). "closed" Like "network" but also excludes 120C (mixed regulated). Excludes 003, 012, 021C, 120C. "all" All 16 MAN types, including empty and trivial patterns.
include	Character vector of MAN types to include exclusively. Overrides pattern.
exclude	Character vector of MAN types to exclude. Applied after pattern filter.
significance	Logical. Run permutation significance test? Default TRUE.
n_perm	Number of permutations for significance. Default 1000.
min_count	Minimum observed count to include a triad (instance mode only). Default 5 for instances, NULL for census.
edge_method	Method for determining edge presence: "any" (default), "expected", or "percent".
edge_threshold	Threshold for "expected" or "percent" methods. Default 1.5.
min_transitions	Minimum total transitions for a unit to be included. Default 5.
top	Return only the top N results. NULL returns all.
seed	Random seed for reproducibility.
...	Additional arguments passed to internal plot helpers.
type	Plot type: "triads" Network diagrams of specific node triples (instance mode) or falls back to patterns (census mode). Arranged in a grid. "types" Bar chart of MAN type frequencies. "significance" Z-score plot showing over- and under-represented types relative to a null model. Requires significance = TRUE in the motifs() call.

	"patterns" Abstract MAN pattern diagrams showing the edge structure of each triad type.
n	Maximum number of items to plot. Default 15.
ncol	Number of columns in the triad/pattern grid. Default 5.
colors	Two-element color vector: first color for over-represented or positive values, second for under-represented or negative values. Default <code>c("#2166AC", "#B2182B")</code> (blue/red).

Details

Detects input type and analysis level automatically. For inputs with individual/group data (tna objects, cograph networks from edge lists with metadata), performs per-group analysis. For aggregate inputs (matrices, igraph), analyzes the single network.

Value

A `cograph_motif_result` object with:

results Data frame of results. Census: type, count, (z, p, sig). Instances: triad, type, observed, (z, p, sig).

type_summary Named counts by MAN type

level Analysis level: "individual" or "aggregate"

named_nodes Whether nodes are identified (TRUE) or exchangeable (FALSE)

n_units Number of units analyzed

params List of parameters used

Invisibly returns the input `x`.

See Also

`subgraphs()`, `motif_census()`, `extract_motifs()`

Other motifs: `extract_motifs()`, `extract_triads()`, `get_edge_list()`, `motif_census()`, `plot.cograph_motif_analy`, `plot.cograph_motifs()`, `subgraphs()`, `triad_census()`

Examples

```
# Census from a matrix (no significance test – fastest path)
mat <- matrix(c(0,3,2,0, 0,0,5,1, 0,0,0,4, 2,0,0,0), 4, 4, byrow = TRUE)
rownames(mat) <- colnames(mat) <- c("Plan","Execute","Monitor","Adapt")
motifs(mat, significance = FALSE)
```

```
# With a minimal significance test (set n_perm >= 500 in practice)
motifs(mat, n_perm = 10L, seed = 1)
## Not run:
if (requireNamespace("tna", quietly = TRUE)) {
  # tna object input – keep n_perm small for example speed
  Mod <- tna::tna(tna::group_regulation)
  motifs(Mod, n_perm = 10L, seed = 1)
```

```

  subgraphs(Mod, n_perm = 10L, seed = 1)
}

## End(Not run)

```

neighborhood_overlap *Neighborhood Overlap (Jaccard) for Each Edge*

Description

Convenience wrapper around [edge centrality](#) that returns only the overlap measure sorted by overlap descending.

Usage

```
neighborhood_overlap(x, top = NULL, directed = NULL, digits = NULL, ...)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object.
top	Integer or NULL. Return only the top N edges. Default NULL.
directed	Logical or NULL. Default NULL (auto-detect).
digits	Integer or NULL. Round numeric columns. Default NULL.
...	Additional arguments passed to edge centrality .

Value

A data frame sorted by overlap (descending) with columns: from, to, weight (if weighted), overlap, shared_neighbors.

See Also

[edge centrality](#), [simmelian strength](#)

Examples

```

adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
cograph::neighborhood_overlap(adj)

```

network_bridges	<i>Bridge Edges</i>
-----------------	---------------------

Description

Finds edges whose removal would disconnect the network. These are critical edges for network connectivity.

Usage

```
network_bridges(x, count_only = FALSE, ...)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
count_only	Logical. If TRUE, return only the count. Default FALSE.
...	Additional arguments passed to to_igraph

Value

If count_only = FALSE, data frame with from/to columns. If count_only = TRUE, integer count.

Examples

```
# Two triangles connected by single edge
adj <- matrix(0, 6, 6)
adj[1,2] <- adj[2,1] <- adj[1,3] <- adj[3,1] <- adj[2,3] <- adj[3,2] <- 1
adj[4,5] <- adj[5,4] <- adj[4,6] <- adj[6,4] <- adj[5,6] <- adj[6,5] <- 1
adj[3,4] <- adj[4,3] <- 1 # Bridge
network_bridges(adj) # Edge 3-4
network_bridges(adj, count_only = TRUE) # 1
```

network_clique_size	<i>Largest Clique Size</i>
---------------------	----------------------------

Description

Finds the size of the largest clique (complete subgraph) in the network. Also known as the clique number or omega of the graph.

Usage

```
network_clique_size(x, ...)
```

Arguments

x Network input: matrix, igraph, network, cograph_network, or tna object
 ... Additional arguments passed to [to_igraph](#)

Value

Integer: size of the largest clique

Examples

```
# Triangle embedded in larger graph
adj <- matrix(c(0,1,1,1, 1,0,1,0, 1,1,0,0, 1,0,0,0), 4, 4)
network_clique_size(adj) # 3
```

network_cut_vertices *Cut Vertices (Articulation Points)*

Description

Finds nodes whose removal would disconnect the network. These are critical nodes for network connectivity.

Usage

```
network_cut_vertices(x, count_only = FALSE, ...)
```

Arguments

x Network input: matrix, igraph, network, cograph_network, or tna object
 count_only Logical. If TRUE, return only the count. Default FALSE.
 ... Additional arguments passed to [to_igraph](#)

Value

If count_only = FALSE, vector of node indices (or names if graph is named). If count_only = TRUE, integer count.

Examples

```
# Bridge node connecting two components
adj <- matrix(c(0,1,1,0,0, 1,0,1,0,0, 1,1,0,1,0, 0,0,1,0,1, 0,0,0,1,0), 5, 5)
network_cut_vertices(adj) # Node 3 is cut vertex
network_cut_vertices(adj, count_only = TRUE) # 1
```

network_girth	<i>Network Girth (Shortest Cycle Length)</i>
---------------	--

Description

Computes the girth of a network - the length of the shortest cycle. Returns Inf for acyclic graphs (trees, DAGs).

Usage

```
network_girth(x, ...)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
...	Additional arguments passed to to_igraph

Value

Integer: length of shortest cycle, or Inf if no cycles exist

Examples

```
# Triangle has girth 3
triangle <- matrix(c(0,1,1, 1,0,1, 1,1,0), 3, 3)
network_girth(triangle) # 3

# Tree has no cycles (Inf)
tree <- matrix(c(0,1,0, 1,0,1, 0,1,0), 3, 3)
network_girth(tree) # Inf
```

network_global_efficiency	<i>Global Efficiency</i>
---------------------------	--------------------------

Description

Computes the global efficiency of a network - the average of the inverse shortest path lengths between all pairs of nodes. Higher values indicate better global communication efficiency. Handles disconnected graphs gracefully (infinite distances contribute 0).

Usage

```
network_global_efficiency(
  x,
  directed = NULL,
  weights = NULL,
  invert_weights = NULL,
  alpha = 1,
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
directed	Logical. Consider edge direction? Default TRUE for directed graphs.
weights	Edge weights (NULL for unweighted). Set to NA to ignore existing weights.
invert_weights	Logical or NULL. Invert weights so higher weights = shorter paths? Default NULL which auto-detects: TRUE for tna objects, FALSE otherwise (matching igraph/sna). Set TRUE for strength/frequency weights (qgraph style).
alpha	Numeric. Exponent for weight inversion: distance = 1/weight^alpha. Default 1.
...	Additional arguments passed to to_igraph

Value

Numeric in [0, 1]: global efficiency

Examples

```
# Complete graph has efficiency 1
k4 <- matrix(1, 4, 4); diag(k4) <- 0
network_global_efficiency(k4) # 1

# Star has lower efficiency
star <- matrix(c(0,1,1,1, 1,0,0,0, 1,0,0,0, 1,0,0,0), 4, 4)
network_global_efficiency(star) # ~0.83
```

network_local_efficiency

Local Efficiency

Description

Computes the average local efficiency across all nodes. Local efficiency of a node is the global efficiency of its neighborhood subgraph (excluding the node itself). Measures fault tolerance and local integration.

Usage

```
network_local_efficiency(
  x,
  weights = NULL,
  invert_weights = NULL,
  alpha = 1,
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
weights	Edge weights (NULL for unweighted). Set to NA to ignore existing weights.
invert_weights	Logical or NULL. Invert weights so higher weights = shorter paths? Default NULL which auto-detects: TRUE for tna objects, FALSE otherwise (matching igraph/sna). Set TRUE for strength/frequency weights (qgraph style).
alpha	Numeric. Exponent for weight inversion. Default 1.
...	Additional arguments passed to to_igraph

Value

Numeric in [0, 1]: average local efficiency

Examples

```
# Complete graph: removing any node leaves complete subgraph, so local efficiency = 1
k5 <- matrix(1, 5, 5); diag(k5) <- 0
network_local_efficiency(k5) # 1

# Star: neighbors not connected to each other
star <- matrix(c(0,1,1,1,1, 1,0,0,0,0, 1,0,0,0,0, 1,0,0,0,0, 1,0,0,0,0), 5, 5)
network_local_efficiency(star) # 0
```

network_radius	<i>Network Radius</i>
----------------	-----------------------

Description

Computes the radius of a network - the minimum eccentricity across all nodes. The eccentricity of a node is the maximum shortest path distance to any other node. The radius is the smallest such maximum distance.

Usage

```
network_radius(x, directed = NULL, ...)
```

Arguments

x Network input: matrix, igraph, network, cograph_network, or tna object
 directed Logical. Consider edge direction? Default TRUE for directed graphs.
 ... Additional arguments passed to [to_igraph](#)

Value

Numeric: the network radius

Examples

```
# Star graph: center has eccentricity 1, leaves have 2, so radius = 1
star <- matrix(c(0,1,1,1, 1,0,0,0, 1,0,0,0, 1,0,0,0), 4, 4)
network_radius(star) # 1
```

network_rich_club *Rich Club Coefficient*

Description

Computes the rich club coefficient for a given degree threshold k . Measures the tendency of high-degree nodes to connect to each other. A normalized version compares to random graphs.

Usage

```
network_rich_club(x, k = NULL, normalized = FALSE, n_random = 10, ...)
```

Arguments

x Network input: matrix, igraph, network, cograph_network, or tna object
 k Degree threshold. Only nodes with degree $> k$ are included. If NULL, uses median degree.
 normalized Logical. Normalize by random graph expectation? Default FALSE.
 n_random Number of random graphs for normalization. Default 10.
 ... Additional arguments passed to [to_igraph](#)

Value

Numeric: rich club coefficient (> 1 indicates rich club effect when normalized)

Examples

```
# Scale-free networks often show rich-club effect
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::barabasi.game(50, m = 2)
  network_rich_club(g, k = 5)
}
```

network_small_world *Small-World Coefficient (Sigma)*

Description

Computes the small-world coefficient σ , defined as: $\sigma = (C / C_{\text{rand}}) / (L / L_{\text{rand}})$ where C is clustering coefficient, L is mean path length, and $_{\text{rand}}$ are values from equivalent random graphs.

Usage

```
network_small_world(x, n_random = 10, ...)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
n_random	Number of random graphs for comparison. Default 10.
...	Additional arguments passed to to_igraph

Details

Values > 1 indicate small-world properties. Typically small-world networks have $\sigma \gg 1$.

Value

Numeric: small-world coefficient σ

Examples

```
# Watts-Strogatz small-world graph
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::sample_smallworld(1, 20, 3, 0.1)
  network_small_world(g) # Should be > 1
}
```

network_summary *Network-Level Summary Statistics*

Description

Computes comprehensive network-level statistics for a network. Returns a data frame with one row containing various metrics including density, centralization scores, transitivity, and more.

Usage

```
network_summary(
  x,
  directed = NULL,
  weighted = TRUE,
  mode = "all",
  loops = TRUE,
  simplify = "sum",
  detailed = FALSE,
  extended = FALSE,
  digits = 3,
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.
weighted	Logical. Use edge weights for strength/centrality calculations. Default TRUE.
mode	For directed networks: "all", "in", or "out". Affects degree-based calculations. Default "all".
loops	Logical. If TRUE (default), keep self-loops. Set FALSE to remove them.
simplify	How to combine multiple edges between the same node pair. Options: "sum" (default), "mean", "max", "min", or FALSE/"none" to keep multiple edges.
detailed	Logical. If TRUE, include mean/sd centrality statistics. Default FALSE returns 18 basic metrics; TRUE returns 29 metrics.
extended	Logical. If TRUE, include additional structural metrics (girth, radius, clique size, cut vertices, bridges, efficiency). Default FALSE.
digits	Integer. Round numeric results to this many decimal places. Default 3.
...	Additional arguments (currently unused)

Value

A data frame with one row containing network-level statistics:

Basic measures (always computed):

node_count	Number of nodes in the network
edge_count	Number of edges in the network
density	Edge density (proportion of possible edges)
component_count	Number of connected components
diameter	Longest shortest path in the network
mean_distance	Average shortest path length
min_cut	Minimum cut value (edge connectivity)

centralization_degree Degree centralization (0-1)
centralization_in_degree In-degree centralization (directed only)
centralization_out_degree Out-degree centralization (directed only)
centralization_betweenness Betweenness centralization (0-1)
centralization_closeness Closeness centralization (0-1)
centralization_eigen Eigenvector centralization (0-1)
transitivity Global clustering coefficient
reciprocity Proportion of mutual edges (directed only)
assortativity_degree Degree assortativity coefficient
hub_score Maximum hub score (HITS algorithm)
authority_score Maximum authority score (HITS algorithm)

Extended measures (when extended = TRUE):

girth Length of shortest cycle (Inf if acyclic)
radius Minimum eccentricity (shortest max-distance from any node)
vertex_connectivity Minimum nodes to remove to disconnect graph
largest_clique_size Size of the largest complete subgraph
cut_vertex_count Number of articulation points (cut vertices)
bridge_count Number of bridge edges
global_efficiency Average inverse shortest path length
local_efficiency Average local efficiency across nodes

Detailed measures (when detailed = TRUE):

mean_degree, sd_degree, median_degree Degree distribution statistics
mean_strength, sd_strength Weighted degree statistics
mean_betweenness Average betweenness centrality
mean_closeness Average closeness centrality
mean_eigenvector Average eigenvector centrality
mean_pagerank Average PageRank
mean_constraint Average Burt's constraint
mean_local_transitivity Average local clustering coefficient

Examples

```

# Basic usage with adjacency matrix
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
network_summary(adj)

# With detailed statistics
network_summary(adj, detailed = TRUE)

```

```

# With extended structural metrics
network_summary(adj, extended = TRUE)

# All metrics
network_summary(adj, detailed = TRUE, extended = TRUE)

# From igraph object
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::sample_gnp(20, 0.3)
  network_summary(g)
}

```

```

network_vertex_connectivity
      Network Vertex Connectivity

```

Description

Computes the vertex connectivity of a network - the minimum number of vertices that must be removed to disconnect the graph (or make it trivial). Higher values indicate more robust network structure.

Usage

```
network_vertex_connectivity(x, ...)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
...	Additional arguments passed to to_igraph

Value

Integer: minimum vertex cut size

Examples

```

# Complete graph K4 has vertex connectivity 3
k4 <- matrix(1, 4, 4); diag(k4) <- 0
network_vertex_connectivity(k4) # 3

# Path graph has vertex connectivity 1
path <- matrix(c(0,1,0,0, 1,0,1,0, 0,1,0,1, 0,0,1,0), 4, 4)
network_vertex_connectivity(path) # 1

```

nodes	<i>Get Nodes from Cograph Network (Deprecated)</i>
-------	--

Description

Extracts the nodes data frame from a `cograph_network` object. **Deprecated:** Use `get_nodes` instead.

Usage

```
nodes(x)
```

Arguments

`x` A `cograph_network` object.

Value

A data frame with columns: `id`, `label`, `name`, `x`, `y` (and possibly others).

See Also

[get_nodes](#), [as_cograph](#), [n_nodes](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
nodes(net) # Deprecated, use get_nodes(net) instead
```

<code>n_communities</code>	<i>Get Number of Communities</i>
----------------------------	----------------------------------

Description

Get Number of Communities

Usage

```
n_communities(x)
```

Arguments

`x` A `cograph_communities` object

Value

Integer count of communities

Examples

```
g <- igraph::make_graph("Zachary")
comm <- community_louvain(g)
n_communities(comm)
```

n_edges	<i>Get Number of Edges</i>
---------	----------------------------

Description

Returns the number of edges in a `cograph_network`.

Usage

```
n_edges(x)
```

Arguments

x A `cograph_network` object.

Value

Integer: number of edges.

See Also

[as_cograph](#), [n_nodes](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
n_edges(net) # 3
```

n_nodes	<i>Get Number of Nodes</i>
---------	----------------------------

Description

Returns the number of nodes in a `cograph_network`.

Usage

```
n_nodes(x)
```

Arguments

`x` A `cograph_network` object.

Value

Integer: number of nodes.

See Also

[as_cograph](#), [n_edges](#), [get_nodes](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
n_nodes(net) # 3
```

overlay_communities	<i>Overlay Community Blobs on a Network Plot</i>
---------------------	--

Description

Render a network with [splot](#) and overlay smooth blob shapes highlighting node communities.

Usage

```
overlay_communities(  
  x,  
  communities,  
  blob_colors = NULL,  
  blob_alpha = 0.25,  
  blob_linewidth = 0.7,  
  blob_line_alpha = 0.8,  
  ...  
)
```

Arguments

x	A network object passed to <code>splot</code> : tna, matrix, igraph, or cograph_network.
communities	Community assignments in any format: a method name (e.g., "walktrap", "louvain"), a numeric membership vector (e.g., c(1, 1, 2, 2, 3)), a named list of character vectors, a cograph_communities object, or a tna_communities object.
blob_colors	Character vector of fill colors for blobs. Recycled if shorter than the number of communities.
blob_alpha	Numeric. Fill transparency (0-1).
blob_linewidth	Numeric. Border line width.
blob_line_alpha	Numeric. Border line transparency (0-1).
...	Additional arguments passed to <code>splot</code> .

Value

The `splot` result (invisibly).

Examples

```
## Not run:
mat <- matrix(runif(25), 5, 5,
              dimnames = list(LETTERS[1:5], LETTERS[1:5]))
diag(mat) <- 0
overlay_communities(mat, list(g1 = c("A", "B"), g2 = c("C", "D", "E")))

## End(Not run)

if (requireNamespace("tna", quietly = TRUE)) {
  model <- tna::tna(tna::group_regulation)

  # With a named list
  overlay_communities(model, list(
    Regulatory = c("plan", "monitor", "adapt"),
    Social      = c("cohesion", "emotion", "consensus"),
    Task        = c("discuss", "synthesis", "coregulate")
  ))

  # With a cograph_communities object (infomap supports directed graphs)
  comm <- cograph::communities(model$weights, method = "infomap")
  overlay_communities(model, comm)
}
```

palettes

Color Palettes

Description

Built-in color palettes for network visualization.

Value

A character vector of hex color codes.

Examples

```
palette_blues(5)
palette_reds(5)
```

palette_blues

Blues Palette

Description

Generate a blue sequential palette.

Usage

```
palette_blues(n, alpha = 1)
```

Arguments

n	Number of colors to generate.
alpha	Transparency (0-1).

Value

Character vector of colors.

Examples

```
palette_blues(5)
```

palette_colorblind *Colorblind-friendly Palette*

Description

Generate a colorblind-friendly palette using Wong's colors.

Usage

```
palette_colorblind(n, alpha = 1)
```

Arguments

n	Number of colors to generate.
alpha	Transparency (0-1).

Value

Character vector of colors.

Examples

```
palette_colorblind(5)
```

palette_diverging *Diverging Palette*

Description

Generate a diverging color palette (blue-white-red).

Usage

```
palette_diverging(n, alpha = 1, midpoint = "white")
```

Arguments

n	Number of colors to generate.
alpha	Transparency (0-1).
midpoint	Color for midpoint.

Value

Character vector of colors.

Examples

```
palette_diverging(5)
```

palette_pastel	<i>Pastel Palette</i>
----------------	-----------------------

Description

Generate a soft pastel color palette.

Usage

```
palette_pastel(n, alpha = 1)
```

Arguments

n	Number of colors to generate.
alpha	Transparency (0-1).

Value

Character vector of colors.

Examples

```
palette_pastel(5)
```

palette_rainbow	<i>Rainbow Palette</i>
-----------------	------------------------

Description

Generate a rainbow color palette.

Usage

```
palette_rainbow(n, alpha = 1)
```

Arguments

n	Number of colors to generate.
alpha	Transparency (0-1).

Value

Character vector of colors.

Examples

```
palette_rainbow(5)
```

palette_reds	<i>Reds Palette</i>
--------------	---------------------

Description

Generate a red sequential palette.

Usage

```
palette_reds(n, alpha = 1)
```

Arguments

n	Number of colors to generate.
alpha	Transparency (0-1).

Value

Character vector of colors.

Examples

```
palette_reds(5)
```

palette_viridis	<i>Viridis Palette</i>
-----------------	------------------------

Description

Generate colors from the viridis palette.

Usage

```
palette_viridis(n, alpha = 1, option = "viridis")
```

Arguments

n	Number of colors to generate.
alpha	Transparency (0-1).
option	Viridis option: "viridis", "magma", "plasma", "inferno", "cividis".

Value

Character vector of colors.

Examples

```
palette_viridis(5)
```

```
plot.cograph_cluster_significance
      Plot Cluster Significance
```

Description

Creates a histogram of the null distribution with the observed value marked.

Usage

```
## S3 method for class 'cograph_cluster_significance'
plot(x, ...)
```

Arguments

```
x          A cograph_cluster_significance object
...        Additional arguments passed to hist
```

Value

Invisibly returns x

Examples

```
## Not run:
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::make_graph("Zachary")
  comm <- community_louvain(g)
  sig <- cluster_significance(g, comm, n_random = 20, seed = 42)
  plot(sig)
}

## End(Not run)
```

```
plot.cograph_communities
      Plot Community Structure
```

Description

Visualizes network with community coloring using splot.

Usage

```
## S3 method for class 'cograph_communities'
plot(x, network = NULL, ...)
```

Arguments

x A cograph_communities object
network The original network (required if not stored)
... Additional arguments passed to splot

Value

Invisibly returns the plot

Examples

```
g <- igraph::make_graph("Zachary")
comm <- community_louvain(g)
mat <- igraph::as_adjacency_matrix(g, sparse = FALSE)
plot(comm, network = mat)
```

plot.cograph_core_periphery

Plot Core-Periphery Structure

Description

Visualizes the network with core nodes highlighted (larger, red) and periphery nodes de-emphasized (smaller, blue).

Usage

```
## S3 method for class 'cograph_core_periphery'
plot(
  x,
  core_color = "#E41A1C",
  periphery_color = "#377EB8",
  core_size = 12,
  periphery_size = 6,
  ...
)
```

Arguments

x A cograph_core_periphery object from [core_periphery](#).
core_color Color for core nodes. Default "#E41A1C".
periphery_color Color for periphery nodes. Default "#377EB8".
core_size Numeric size for core nodes. Default 12.
periphery_size Numeric size for periphery nodes. Default 6.
... Additional arguments passed to [splot](#).

Value

Invisible x.

plot.cograph_degree_fit

Plot method for cograph_degree_fit

Description

Overlays fitted distribution curves on a histogram of observed degrees.

Usage

```
## S3 method for class 'cograph_degree_fit'
plot(
  x,
  which = NULL,
  log = "",
  cols = NULL,
  lwd = 2,
  main = "Degree Distribution Fit",
  ...
)
```

Arguments

x	A cograph_degree_fit object from fit_degree_distribution .
which	Character vector of distribution names to display. Default NULL shows all fitted distributions.
log	Character string for log-scale axes: "" (default), "x", "y", or "xy".
cols	Named or unnamed character vector of colors for distribution curves. Default uses a built-in palette.
lwd	Line width for fitted curves. Default 2.
main	Plot title. Default "Degree Distribution Fit".
...	Additional arguments passed to hist .

Value

Invisible NULL.

Examples

```
## Not run:
adj <- matrix(c(0, 1, 1, 0, 0,
               1, 0, 1, 1, 0,
               1, 1, 0, 1, 1,
               0, 1, 1, 0, 1,
               0, 0, 1, 1, 0), 5, 5, byrow = TRUE)
fit <- cograph::fit_degree_distribution(adj)
plot(fit)
plot(fit, which = c("exponential", "poisson"), log = "y")

## End(Not run)
```

plot.cograph_motifs *Plot Network Motifs*

Description

Visualize motif frequencies and their statistical significance.

Usage

```
## S3 method for class 'cograph_motifs'
plot(
  x,
  type = c("bar", "heatmap", "network"),
  show_nonsig = FALSE,
  top_n = NULL,
  colors = c("#2166AC", "#F7F7F7", "#B2182B"),
  ...
)
```

Arguments

x	A cograph_motifs object from <code>motif_census()</code>
type	Plot type: "bar" (default) Bar chart of motif frequencies, colored by significance direction (over/under-represented). "heatmap" Heatmap of z-scores across motif types. "network" Network diagrams of the top motifs by lz-scorel.
show_nonsig	Show non-significant motifs? Default FALSE.
top_n	Show only top N motifs by lz-scorel. Default NULL (all).
colors	Three-element color vector for under-represented, neutral, and over-represented motifs. Default <code>c("#2166AC", "#999999", "#B2182B")</code> (blue/gray/red).
...	Additional arguments passed to plotting functions

Value

A ggplot2 object (invisibly)

See Also

[motif_census\(\)](#) for the analysis that produces this object

Other motifs: [extract_motifs\(\)](#), [extract_triads\(\)](#), [get_edge_list\(\)](#), [motif_census\(\)](#), [motifs\(\)](#), [plot.cograph_motif_analysis\(\)](#), [subgraphs\(\)](#), [triad_census\(\)](#)

Examples

```
mat <- matrix(sample(0:1, 100, replace = TRUE, prob = c(0.7, 0.3)), 10, 10)
diag(mat) <- 0
m <- motif_census(mat, directed = TRUE, n_random = 50)
plot(m)
plot(m, type = "network")
```

plot.cograph_motif_analysis

Plot Motif Analysis Results

Description

Create visualizations for motif analysis results including network diagrams of triads, bar plots of type distributions, and significance plots.

Usage

```
## S3 method for class 'cograph_motif_analysis'
plot(
  x,
  type = c("triads", "types", "significance", "patterns"),
  n = 20,
  colors = c("#2166AC", "#B2182B"),
  res = 72,
  node_size = 5,
  label_size = 7,
  title_size = 7,
  stats_size = 5,
  ncol = 5,
  legend = TRUE,
  color = "#800020",
  spacing = 1,
  ...
)
```

Arguments

x	A cograph_motif_analysis object from extract_motifs()
type	Plot type: "triads" (default) Network diagrams of specific named triads, arranged in a grid. Each cell shows the three nodes and their edges. "types" Bar chart of MAN type frequencies. "significance" Z-score plot showing over- and under-represented types. Requires significance = TRUE in extract_motifs() . "patterns" Abstract MAN pattern diagrams showing edge structure of each triad type without specific node labels.
n	Number of triads/patterns to show. Default 20.
colors	Two-element color vector for the types/significance plots: first color for over-represented, second for under-represented. Default <code>c("#2166AC", "#B2182B")</code> (blue/red).
res	Resolution for scaling (kept for backwards compatibility). Default 72.
node_size	Size of nodes in triad diagrams (1-10 scale). Default 5.
label_size	Font size for node labels (3-letter abbreviations). Default 7.
title_size	Font size for motif type title (e.g., "120C"). Default 7.
stats_size	Font size for statistics text (n, z, p). Default 5.
ncol	Number of columns in the plot grid. Default 5.
legend	Show abbreviation legend at bottom? Default TRUE.
color	Color for nodes, edges, and labels in triad diagrams. Default "#800020" (maroon).
spacing	Spacing multiplier between grid cells (0.5-2). Default 1.
...	Additional arguments (unused).

Value

Invisibly returns NULL for triad plots, or a ggplot2 object for types/significance/patterns plots.

See Also

[extract_motifs\(\)](#) for the analysis that produces this object, [motif_census\(\)](#) for statistical motif analysis

Other motifs: [extract_motifs\(\)](#), [extract_triads\(\)](#), [get_edge_list\(\)](#), [motif_census\(\)](#), [motifs\(\)](#), [plot.cograph_motifs\(\)](#), [subgraphs\(\)](#), [triad_census\(\)](#)

Examples

```
mat <- matrix(c(0,3,2,0, 0,0,5,1, 0,0,0,4, 2,0,0,0), 4, 4, byrow = TRUE)
rownames(mat) <- colnames(mat) <- c("Plan", "Execute", "Monitor", "Adapt")
m <- extract_motifs(mat, significance = FALSE)
plot(m)
## Not run:
```

```
plot(m, type = "types")  
## End(Not run)
```

plot.cograph_network *Plot cograph_network Object*

Description

Plot cograph_network Object

Usage

```
## S3 method for class 'cograph_network'  
plot(x, ...)
```

Arguments

x A cograph_network object.
... Additional arguments passed to sn_render.

Value

The input object x, invisibly.

Examples

```
## Not run:  
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)  
net <- cograph(adj)  
plot(net)  
  
## End(Not run)
```

plot.cograph_rich_club

Plot Rich Club Results

Description

Two plot types: "curve" (default) shows the rich club coefficient across thresholds with null model bands. "network" highlights rich club members on the network at a given threshold.

Usage

```
## S3 method for class 'cograph_rich_club'
plot(x, type = c("curve", "network"), k = NULL, col = "#E41A1C", ...)
```

Arguments

x	A cograph_rich_club data frame.
type	Character. "curve" (default) or "network".
k	Numeric. For type = "network", the degree/strength threshold to visualize. If NULL, uses the threshold with the highest phi_norm (or phi if not normalized).
col	Line/node color for rich club. Default "#E41A1C".
...	Additional arguments passed to plot (curve) or splot (network).

Value

Invisible x.

plot.cograph_vulnerability
Plot Node Vulnerability

Description

Plot Node Vulnerability

Usage

```
## S3 method for class 'cograph_vulnerability'
plot(x, top = NULL, col = "steelblue", ...)
```

Arguments

x	A cograph_vulnerability object.
top	Integer or NULL. Show only top N nodes. Default NULL (all).
col	Bar color. Default "steelblue".
...	Additional arguments passed to barplot .

Value

Invisible x.

plot.tna_bootstrap *Plot Bootstrap Results*

Description

Visualizes bootstrap analysis results with styling to distinguish significant from non-significant edges. Works with tna_bootstrap objects from the tna package.

Usage

```
## S3 method for class 'tna_bootstrap'
plot(x, ...)

splot.tna_bootstrap(
  x,
  display = c("styled", "significant", "full", "ci"),
  edge_style_sig = 1,
  edge_style_nonsig = 2,
  color_nonsig = "#888888",
  show_ci = FALSE,
  show_stars = TRUE,
  width_by = NULL,
  inherit_style = TRUE,
  ...
)
```

Arguments

x	A tna_bootstrap object (from tna::bootstrap).
...	Additional arguments passed to splot().
display	Display mode: <ul style="list-style-type: none"> • "styled" (default): All edges with styling to distinguish significant/non-significant • "significant": Only significant edges • "full": All edges without significance styling • "ci": Show CI bands on edges
edge_style_sig	Line style for significant edges (1=solid). Default 1.
edge_style_nonsig	Line style for non-significant edges (2=dashed). Default 2.
color_nonsig	Color for non-significant edges. Default "#888888" (grey).
show_ci	Logical: overlay CI bands on edges? Default FALSE.
show_stars	Logical: show significance stars (*, **, ***) on edges? Default TRUE.
width_by	Optional: "cr_lower" to scale edge width by lower consistency range bound.
inherit_style	Logical: inherit colors/layout from original TNA model? Default TRUE.

Details

The function expects a `tna_bootstrap` object containing:

- `weights` or `weights_orig`: Original weight matrix
- `weights_sig`: Significant weights only (optional)
- `p_values`: P-value matrix
- `ci_lower`, `ci_upper`: Confidence interval bounds
- `level`: Significance level (default 0.05)
- `model`: Original TNA model for styling inheritance

Edge styling in "styled" mode:

- Significant edges: solid dark blue, bold labels with stars, rendered on top
- Non-significant edges: dashed pink, plain labels, rendered behind

Value

Invisibly returns the plot.

Examples

```
## Not run:
# Create a mock tna_bootstrap object with synthetic data
set.seed(42)
w <- matrix(c(0, 0.3, 0.1, 0.2, 0, 0.4, 0.3, 0.1, 0), 3, 3)
rownames(w) <- colnames(w) <- c("A", "B", "C")
p <- matrix(c(1, 0.01, 0.5, 0.03, 1, 0.001, 0.2, 0.8, 1), 3, 3)
boot <- list(
  weights = w,
  p_values = p,
  ci_lower = w - 0.05,
  ci_upper = w + 0.05,
  level = 0.05,
  model = list(weights = w, labels = c("A", "B", "C"))
)
class(boot) <- c("tna_bootstrap", "list")
splot(boot)
splot(boot, display = "significant")

## End(Not run)
```

plot.tna_disparity *Plot Disparity Filter Result*

Description

Plot Disparity Filter Result

Usage

```
## S3 method for class 'tna_disparity'  
plot(x, type = c("backbone", "comparison"), ...)
```

Arguments

x	A tna_disparity object.
type	Plot type: "backbone" (default) or "comparison".
...	Additional arguments passed to splot.

Value

Invisibly returns NULL. Called for the side effect of producing a plot.

Examples

```
## Not run:  
mat <- matrix(c(  
  0.0, 0.5, 0.1, 0.0,  
  0.3, 0.0, 0.4, 0.1,  
  0.1, 0.2, 0.0, 0.5,  
  0.0, 0.1, 0.3, 0.0  
) , nrow = 4, byrow = TRUE)  
rownames(mat) <- colnames(mat) <- c("A", "B", "C", "D")  
disp <- disparity_filter(cograph(mat), level = 0.05)  
plot(disp)  
plot(disp, type = "comparison")  
  
## End(Not run)
```

plot_alluvial	<i>Plot Alluvial Diagram</i>
---------------	------------------------------

Description

Creates an alluvial (Sankey) diagram showing aggregated flows between states. This is an alias for `plot_transitions()` with aggregated flows (default).

Usage

```
plot_alluvial(  
  x,  
  from_title = "From",  
  to_title = "To",  
  title = NULL,  
  from_colors = NULL,  
  to_colors = NULL,  
  flow_fill = "#888888",  
  flow_alpha = 0.4,  
  flow_color_by = NULL,  
  flow_border = NA,  
  flow_border_width = 0.5,  
  node_width = 0.08,  
  node_border = NA,  
  node_spacing = 0.02,  
  label_size = 3.5,  
  label_position = c("beside", "inside", "above", "below", "outside"),  
  label_halo = TRUE,  
  label_color = "black",  
  label_fontface = "plain",  
  label_nudge = 0.02,  
  title_size = 5,  
  title_color = "black",  
  title_fontface = "bold",  
  curve_strength = 0.6,  
  show_values = FALSE,  
  value_position = c("center", "origin", "destination", "outside_origin",  
    "outside_destination"),  
  value_size = 3,  
  value_color = "black",  
  value_halo = NULL,  
  value_fontface = "bold",  
  value_nudge = 0.03,  
  value_min = 0,  
  show_totals = FALSE,  
  total_size = 4,  
  total_color = "white",
```

```

total_fontface = "bold",
conserve_flow = TRUE,
min_flow = 0,
threshold = 0,
value_digits = 2,
column_gap = 1
)

```

Arguments

x	<p>Input data in one of several formats:</p> <ul style="list-style-type: none"> • A transition matrix (rows = from, cols = to, values = counts) • Two vectors: pass before as x and after as second argument (contingency table computed automatically, like chi-square) • A 2-column data frame (raw observations; table computed automatically) • A data frame with columns: from, to, count • A list of matrices for multi-step transitions
from_title	Title for the left column. Default "From". For multi-step, use a vector of titles (e.g., c("T1", "T2", "T3", "T4")).
to_title	Title for the right column. Default "To". Ignored for multi-step.
title	Optional plot title. Applied via <code>ggplot2::labs(title = title)</code> .
from_colors	Colors for left-side nodes. Default uses palette.
to_colors	Colors for right-side nodes. Default uses palette.
flow_fill	Fill color for flows. Default "#888888" (grey). Ignored if <code>flow_color_by</code> is set.
flow_alpha	Alpha transparency for flows. Default 0.4.
flow_color_by	Color flows by "source", "destination", or NULL (use <code>flow_fill</code>). Default NULL.
flow_border	Border color for flows. Default NA (no border).
flow_border_width	Line width for flow borders. Default 0.5.
node_width	Width of node rectangles (0-1 scale). Default 0.08.
node_border	Border color for nodes. Default NA (no border).
node_spacing	Vertical spacing between nodes (0-1 scale). Default 0.02.
label_size	Size of node labels. Default 3.5.
label_position	Position of node labels: "beside" (default), "inside", "above", "below", "outside". Applied to first and last columns. See <code>mid_label_position</code> for middle columns.
label_halo	Logical: add white halo around labels for readability? Default TRUE.
label_color	Color of state name labels. Default "black".
label_fontface	Font face of state name labels ("plain", "bold", "italic", "bold.italic"). Default "plain".
label_nudge	Distance between node edge and label (in plot units). Default 0.02. Increase for more spacing.

<code>title_size</code>	Size of column titles. Default 5.
<code>title_color</code>	Color of column title text. Default "black".
<code>title_fontface</code>	Font face of column titles. Default "bold".
<code>curve_strength</code>	Controls bezier curve shape (0-1). Default 0.6.
<code>show_values</code>	Logical: show transition counts on flows? Default FALSE.
<code>value_position</code>	Position of flow values: "center", "origin", "destination", "outside_origin", "outside_destination". Default "center".
<code>value_size</code>	Size of value labels on flows. Default 3.
<code>value_color</code>	Color of value labels. Default "black".
<code>value_halo</code>	Logical: add halo around flow value labels? Default NULL (inherits from <code>label_halo</code>).
<code>value_fontface</code>	Font face of flow value labels. Default "bold".
<code>value_nudge</code>	Distance of value labels from node edge when using "origin" or "destination" positions. Default 0.03.
<code>value_min</code>	Minimum count to show a flow value label. Default 0 (show all). Use to hide small flows (e.g., <code>value_min = 100</code>).
<code>show_totals</code>	Logical: show total counts on nodes? Default FALSE.
<code>total_size</code>	Size of total labels. Default 4.
<code>total_color</code>	Color of total labels. Default "white".
<code>total_fontface</code>	Font face of total labels. Default "bold".
<code>conserve_flow</code>	Logical: should left and right totals match? Default TRUE. When FALSE, each side scales independently (allows for "lost" or "gained" items).
<code>min_flow</code>	Minimum flow value to display. Default 0 (show all).
<code>threshold</code>	Minimum edge weight to display. Flows below this value are removed. Combined with <code>min_flow</code> : effective minimum is <code>max(threshold, min_flow)</code> . Default 0.
<code>value_digits</code>	Number of decimal places for flow value labels and node totals. Default 2.
<code>column_gap</code>	Horizontal spread of columns (0-1). Default 1 uses full width. Use smaller values (e.g., 0.6) to bring columns closer together.

Value

A `ggplot2` object.

See Also

[plot_transitions](#), [plot_trajectories](#)

Examples

```
## Not run:
# From a transition matrix
mat <- matrix(c(50, 10, 5, 15, 40, 10), 2, 3)
rownames(mat) <- c("A", "B")
colnames(mat) <- c("X", "Y", "Z")
plot_alluvial(mat)

## End(Not run)
```

```
plot_centrality_distribution
      Plot Centrality Distribution
```

Description

Histogram or density plot of any centrality measure. Accepts the output of [centrality](#) directly.

Usage

```
plot_centrality_distribution(
  x,
  measure = "degree_all",
  type = c("histogram", "density"),
  normalize = FALSE,
  bins = NULL,
  log = "",
  col = "steelblue",
  border = "white",
  main = NULL,
  xlab = NULL,
  ...
)
```

Arguments

x	A data frame from centrality , or a network input (matrix, igraph, cograph_network, tna).
measure	Character. Which centrality measure to plot. Default "degree_all". Must match a column name in the centrality output.
type	Character. "histogram" (default) or "density".
normalize	Logical. Show proportions instead of counts. Default FALSE.
bins	Integer or NULL. Number of bins. Default NULL (auto).
log	Character. Log scaling: "", "x", "y", or "xy". Default "".

col	Fill color. Default "steelblue".
border	Border color. Default "white".
main	Plot title. Default auto-generated from measure name.
xlab	X-axis label. Default auto-generated.
...	Additional arguments passed to barplot or plot .

Value

Invisibly returns the centrality values plotted.

Examples

```
adj <- matrix(c(0,1,1,0, 1,0,1,1, 1,1,0,1, 0,1,1,0), 4, 4)
rownames(adj) <- colnames(adj) <- LETTERS[1:4]
cograph::plot_centrality_distribution(adj, measure = "degree_all")
```

plot_chord	<i>Chord Diagram</i>
------------	----------------------

Description

Draw a chord diagram where nodes are arcs on the outer ring and edges are curved ribbons (chords) connecting them. Arc size is proportional to total flow through each node and chord width is proportional to edge weight.

Usage

```
plot_chord(
  x,
  directed = NULL,
  segment_colors = NULL,
  segment_border_color = "white",
  segment_border_width = 1,
  segment_pad = 0.02,
  segment_width = 0.08,
  chord_color_by = "source",
  chord_alpha = 0.5,
  chord_border = NA,
  self_loop = TRUE,
  labels = NULL,
  label_size = 1,
  label_color = "black",
  label_offset = 0.05,
  label_threshold = 0,
  threshold = 0,
  ticks = FALSE,
```

```

    tick_interval = NULL,
    tick_labels = TRUE,
    tick_size = 0.6,
    tick_color = "grey30",
    start_angle = pi/2,
    clockwise = TRUE,
    title = NULL,
    title_size = 1.2,
    background = NULL,
    ...
)

```

Arguments

x	A weight matrix, <code>cograph_network</code> , <code>tna</code> , or <code>igraph</code> object.
directed	Logical. If <code>NULL</code> (default), auto-detected from matrix symmetry.
segment_colors	Colors for the outer ring segments. <code>NULL</code> uses a built-in vibrant palette.
segment_border_color	Border color for segments.
segment_border_width	Border width for segments.
segment_pad	Gap between segments in radians.
segment_width	Radial thickness of the outer ring as a fraction of the radius.
chord_color_by	How to color chords: "source" (default), "target", or a color vector of length matching the number of non-zero edges.
chord_alpha	Alpha transparency for chords.
chord_border	Border color for chords. <code>NA</code> for no border.
self_loop	Logical. Show self-loop chords?
labels	Node labels. <code>NULL</code> uses row names, <code>FALSE</code> suppresses labels.
label_size	Text size multiplier for labels.
label_color	Color for labels.
label_offset	Radial offset of labels beyond the outer ring.
label_threshold	Hide labels for nodes whose flow fraction is below this value.
threshold	Minimum absolute weight to show a chord.
ticks	Logical. Draw tick marks along the outer ring to indicate magnitude?
tick_interval	Spacing between ticks in the same units as the weight matrix. <code>NULL</code> (default) auto-selects a nice interval.
tick_labels	Logical. Show numeric labels at major ticks?
tick_size	Text size multiplier for tick labels.
tick_color	Color for tick marks and labels.
start_angle	Starting angle in radians (default <code>pi/2</code> , top).

clockwise	Logical. Lay out segments clockwise?
title	Optional plot title.
title_size	Text size multiplier for the title.
background	Background color for the plot. NULL (default) uses the current device background.
...	Additional arguments (currently ignored).

Details

The diagram is drawn entirely with base R graphics using `polygon()` for segments and chords, and `bezier_points()` for the curved ribbons.

For directed networks, each segment is split into an outgoing half and an incoming half so that chords attach to the correct side. For undirected networks each edge is drawn once and the full segment arc is shared.

Value

Invisibly returns a list with components `segments` (data frame of segment angles and flows) and `chords` (data frame of chord endpoints and weights).

Examples

```
# Weighted directed matrix
mat <- matrix(c(
  0, 25, 5, 15,
  10, 0, 20, 8,
  3, 18, 0, 30,
  20, 5, 10, 0
), 4, 4, byrow = TRUE,
dimnames = list(c("A", "B", "C", "D"), c("A", "B", "C", "D")))

plot_chord(mat)
plot_chord(mat, chord_alpha = 0.6, ticks = TRUE)

if (requireNamespace("tna", quietly = TRUE)) {
  # TNA transition network
  model <- tna::tna(tna::group_regulation)
  plot_chord(model, ticks = TRUE, segment_width = 0.10)
}
```

plot_compare

Plot Network Difference

Description

Plots the difference between two networks ($x - y$) using `splot`. Positive differences ($x > y$) are shown in green, negative ($x < y$) in red. Optionally displays node-level differences (e.g., initial probabilities) as donut charts.

Usage

```

plot_compare(
  x,
  y = NULL,
  i = NULL,
  j = NULL,
  pos_color = "#009900",
  neg_color = "#C62828",
  labels = NULL,
  title = NULL,
  inits_x = NULL,
  inits_y = NULL,
  show_inits = NULL,
  donut_inner_ratio = 0.8,
  force = FALSE,
  ...
)

```

Arguments

x	First network: matrix, CographNetwork, tna, igraph object, OR a group_tna object. For group_tna with 2 groups, compares them directly. For more groups, plots all pairwise comparisons (or specify i, j).
y	Second network: same type as x. Ignored if x is group_tna.
i	Index/name of first group when x is group_tna. NULL for all pairs.
j	Index/name of second group when x is group_tna. NULL for all pairs.
pos_color	Color for positive differences ($x > y$). Default "#009900" (green).
neg_color	Color for negative differences ($x < y$). Default "#C62828" (red).
labels	Node labels. NULL uses rownames or defaults.
title	Plot title. NULL for auto-generated title.
inits_x	Node values for x (e.g., initial probabilities). NULL to auto-extract from tna.
inits_y	Node values for y. NULL to auto-extract from tna.
show_inits	Logical: show node differences as donuts? Default TRUE if inits available.
donut_inner_ratio	Inner radius ratio for donut (0-1). Default 0.8.
force	Logical: force plotting when more than 4 groups (many comparisons). Default FALSE.
...	Additional arguments passed to splot().

Details

The function computes element-wise subtraction of the weight matrices. Edge colors indicate direction of difference:

- Green edges: x has higher weight than y

- Red edges: y has higher weight than x

When initial probabilities (inits) are provided or extracted from tna objects, nodes display donut charts showing the absolute difference, colored by direction:

- Green donut: x has higher initial probability
- Red donut: y has higher initial probability

For lists of networks (e.g., group_tna), specify which elements to compare using i and j parameters.

Value

Invisibly returns a list with difference matrix and inits difference.

Examples

```
## Not run:
# Compare two adjacency matrices
set.seed(42)
m1 <- matrix(runif(25), 5, 5)
m2 <- matrix(runif(25), 5, 5)
rownames(m1) <- colnames(m1) <- LETTERS[1:5]
rownames(m2) <- colnames(m2) <- LETTERS[1:5]
plot_compare(m1, m2)

# With node-level differences
inits1 <- c(0.3, 0.2, 0.2, 0.15, 0.15)
inits2 <- c(0.1, 0.4, 0.2, 0.2, 0.1)
plot_compare(m1, m2, inits_x = inits1, inits_y = inits2)

## End(Not run)
```

plot_comparison_heatmap

Plot Comparison Heatmap

Description

Creates a heatmap visualization comparing two networks.

Usage

```
plot_comparison_heatmap(
  x,
  y = NULL,
  type = c("difference", "x", "y"),
  name_x = "x",
  name_y = "y",
```

```

    low_color = "blue",
    mid_color = "white",
    high_color = "red",
    limits = NULL,
    show_values = FALSE,
    value_size = 3,
    digits = 2,
    title = NULL,
    xlab = "Target",
    ylab = "Source"
  )

```

Arguments

x	First network: matrix, CographNetwork, tna, or igraph object.
y	Second network: same type as x. NULL to plot just x.
type	What to display: "difference" (x - y), "x", or "y".
name_x	Label for first network in title. Default "x".
name_y	Label for second network in title. Default "y".
low_color	Color for low/negative values. Default "blue".
mid_color	Color for zero/middle values. Default "white".
high_color	Color for high/positive values. Default "red".
limits	Color scale limits. NULL for auto. Use c(-1, 1) for normalized.
show_values	Logical: display values in cells? Default FALSE.
value_size	Text size for cell values. Default 3.
digits	Decimal places for cell values. Default 2.
title	Plot title. NULL for auto-generated.
xlab	X-axis label. Default "Target".
ylab	Y-axis label. Default "Source".

Value

A ggplot2 object.

Examples

```

## Not run:
set.seed(42)
m1 <- matrix(runif(25), 5, 5)
m2 <- matrix(runif(25), 5, 5)
rownames(m1) <- colnames(m1) <- LETTERS[1:5]
rownames(m2) <- colnames(m2) <- LETTERS[1:5]

# Difference heatmap
plot_comparison_heatmap(m1, m2)

```

```
# Show just one network
plot_comparison_heatmap(m1, type = "x")

## End(Not run)
```

plot_degree_correlation

Plot Degree-Degree Correlation

Description

Scatter plot of each node's degree against the average degree of its neighbors. Reveals assortative (positive slope) or disassortative (negative slope) mixing patterns.

Usage

```
plot_degree_correlation(
  x,
  mode = "all",
  directed = NULL,
  col = "steelblue",
  main = "Degree-Degree Correlation",
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna.
mode	Character. For directed networks: "all", "in", or "out". Default "all".
directed	Logical or NULL. Default NULL (auto-detect).
col	Point color. Default "steelblue".
main	Title. Default "Degree-Degree Correlation".
...	Additional arguments passed to plot .

Value

Invisibly returns a data frame with columns node, degree, avg_neighbor_degree.

See Also

[centrality](#), [degree_distribution](#), [network_summary](#)

Examples

```
## Not run:
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::sample_pa(100, m = 3, directed = FALSE)
  cograph::plot_degree_correlation(g)
}

## End(Not run)
```

plot_edge_diff_forest *Forest Plot for Bootstrap Edge Differences*

Description

Visualises pairwise edge weight differences from a `boot_glasso` object. Each row (linear) or spoke (circular) is one edge pair; the CI bar spans the bootstrap CI of the difference; a dashed line/ring marks zero. Red = first edge larger; blue = second edge larger.

Usage

```
plot_edge_diff_forest(x, ...)

## S3 method for class 'boot_glasso'
plot_edge_diff_forest(
  x,
  alpha = NULL,
  layout = c("linear", "circular", "chord", "tile"),
  show_nonsig = FALSE,
  nonzero_only = FALSE,
  sort_by = c("estimate", "significance", "name"),
  n_top = NULL,
  pos_color = "#C0392B",
  neg_color = "#2C6E8A",
  nonsig_color = "#AAAAAA",
  ring_color = "#C8C8C8",
  label_size = 2.3,
  label_color = NULL,
  point_size = if (match.arg(layout) == "circular") 2 else 3,
  r_inner = 0.38,
  r_outer = 0.72,
  title = NULL,
  subtitle = NULL,
  ...
)
```

Arguments

<code>x</code>	A <code>boot_glasso</code> object with <code>\$boot_edges</code> and <code>\$edge_diff_p</code> .
<code>...</code>	Currently unused.
<code>alpha</code>	Significance threshold. Default: inherits from object.
<code>layout</code>	"linear" (default), "circular", or "chord". The chord layout places all edge names on a unit circle and connects significant pairs with bezier arcs; arc width and colour encode the mean bootstrap difference.
<code>show_nonsig</code>	Include non-significant pairs? Default FALSE.
<code>nonzero_only</code>	If TRUE, restrict to edges that are non-zero in the original network (identified via <code>\$original_pcor</code>). Useful for EBICglasso results where many edges are regularised to exactly zero. Default FALSE.
<code>sort_by</code>	"estimate" (default), "significance", or "name" (linear only).
<code>n_top</code>	Restrict to top N pairs by absolute difference.
<code>pos_color</code>	Colour when <code>edge1 > edge2</code> . Default crimson.
<code>neg_color</code>	Colour when <code>edge1 < edge2</code> . Default teal.
<code>nonsig_color</code>	Colour for non-significant pairs.
<code>ring_color</code>	Ring colour (circular/chord). Default light grey.
<code>label_size</code>	Text size. Default 2.3.
<code>label_color</code>	Fixed label colour (NULL = inherit).
<code>point_size</code>	Size of estimate square (linear/circular).
<code>r_inner</code>	Inner ring radius (circular). Default 0.38.
<code>r_outer</code>	Outer ring radius (circular). Default 0.72.
<code>title</code>	Plot title.
<code>subtitle</code>	Plot subtitle.

Value

A `ggplot` object.

Examples

```
set.seed(1)
data1 <- as.data.frame(matrix(rnorm(60), 20, 3, dimnames = list(NULL, c("A", "B", "C"))))
bg <- Nestimate::boot_glasso(data1, iter = 50, centrality = c("strength", "expected_influence"))
plot_edge_diff_forest(bg)
```

plot_edge_weights *Plot Edge Weight Distribution*

Description

Histogram of edge weights in a network.

Usage

```
plot_edge_weights(  
  x,  
  normalize = FALSE,  
  bins = NULL,  
  log = "",  
  directed = NULL,  
  col = "steelblue",  
  border = "white",  
  main = "Edge Weight Distribution",  
  xlab = "Weight",  
  ...  
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna.
normalize	Logical. Show proportions. Default FALSE.
bins	Integer or NULL. Number of bins. Default NULL (auto).
log	Character. Log scaling. Default "".
directed	Logical or NULL. Default NULL (auto-detect).
col	Fill color. Default "steelblue".
border	Border color. Default "white".
main	Title. Default "Edge Weight Distribution".
xlab	X-axis label. Default "Weight".
...	Additional arguments passed to barplot .

Value

Invisibly returns the weight vector.

Examples

```
adj <- matrix(c(0, 2, 3, 2, 0, 1, 3, 1, 0), 3, 3)  
rownames(adj) <- colnames(adj) <- c("A", "B", "C")  
cograph::plot_edge_weights(adj)
```

`plot_heatmap`*Plot Network as Heatmap*

Description

Visualizes a network adjacency/weight matrix as a heatmap. Supports single networks, multi-cluster networks (block diagonal), and multi-layer networks (group_tna).

Usage

```
plot_heatmap(  
  x,  
  cluster_list = NULL,  
  cluster_spacing = 0,  
  show_legend = TRUE,  
  legend_position = "right",  
  legend_title = "Weight",  
  colors = "viridis",  
  limits = NULL,  
  midpoint = NULL,  
  na_color = "grey90",  
  show_values = FALSE,  
  value_size = 2.5,  
  value_color = "black",  
  value_fontface = "plain",  
  value_fontfamily = "sans",  
  value_halo = NULL,  
  value_digits = 2,  
  show_diagonal = TRUE,  
  diagonal_color = NULL,  
  cluster_labels = TRUE,  
  cluster_borders = TRUE,  
  border_color = "black",  
  border_width = 0.5,  
  row_labels = NULL,  
  col_labels = NULL,  
  show_axis_labels = TRUE,  
  axis_text_size = 8,  
  axis_text_angle = 45,  
  title = NULL,  
  subtitle = NULL,  
  xlab = NULL,  
  ylab = NULL,  
  threshold = 0,  
  aspect_ratio = 1,  
  ...  
)
```

Arguments

<code>x</code>	Network input: matrix, CographNetwork, tna, igraph, group_tna, or any object cograph accepts.
<code>cluster_list</code>	Optional list of character vectors defining node clusters. Creates a block-structured heatmap with clusters along diagonal.
<code>cluster_spacing</code>	Gap size between clusters (in cell units). Default 0.
<code>show_legend</code>	Logical: display color legend? Default TRUE.
<code>legend_position</code>	Position: "right" (default), "left", "top", "bottom", "none".
<code>legend_title</code>	Title for legend. Default "Weight".
<code>colors</code>	Color palette: vector of colors for gradient, or a palette name ("viridis", "heat", "blues", "reds", "diverging"). Default "viridis".
<code>limits</code>	Numeric vector c(min, max) for color scale. NULL for auto.
<code>midpoint</code>	Midpoint for diverging scales. NULL for auto (0 if data spans neg/pos).
<code>na_color</code>	Color for NA values. Default "grey90".
<code>show_values</code>	Logical: display values in cells? Default FALSE.
<code>value_size</code>	Text size for cell values. Default 2.5.
<code>value_color</code>	Color for cell value text. Default "black".
<code>value_fontface</code>	Font face for values: "plain", "bold", "italic", "bold.italic". Default "plain".
<code>value_fontfamily</code>	Font family for values: "sans", "serif", "mono". Default "sans".
<code>value_halo</code>	Halo color behind value labels for readability on dark cells. Set to a color (e.g., "white") to enable, or NULL (default) to disable.
<code>value_digits</code>	Decimal places for values. Default 2.
<code>show_diagonal</code>	Logical: show diagonal values? Default TRUE.
<code>diagonal_color</code>	Optional color for diagonal cells. NULL uses scale.
<code>cluster_labels</code>	Logical: show cluster/layer labels? Default TRUE.
<code>cluster_borders</code>	Logical: draw borders around clusters? Default TRUE.
<code>border_color</code>	Color for cluster borders. Default "black".
<code>border_width</code>	Width of cluster borders. Default 0.5.
<code>row_labels</code>	Row labels. NULL for auto (rownames or indices).
<code>col_labels</code>	Column labels. NULL for auto (colnames or indices).
<code>show_axis_labels</code>	Logical: show axis tick labels? Default TRUE.
<code>axis_text_size</code>	Size of axis labels. Default 8.
<code>axis_text_angle</code>	Angle for x-axis labels. Default 45.
<code>title</code>	Plot title. Default NULL.

subtitle	Plot subtitle. Default NULL.
xlab	X-axis label. Default NULL.
ylab	Y-axis label. Default NULL.
threshold	Minimum absolute value to display. Values with $\text{abs}(\text{value}) < \text{threshold}$ are set to zero. Default 0.
aspect_ratio	Aspect ratio. Default 1 (square cells).
...	Additional arguments (currently unused).

Details

For multi-cluster networks, provide `cluster_list` as a named list where each element is a vector of node names belonging to that cluster. The heatmap will be reordered to show clusters as blocks along the diagonal.

For `group_tna` objects (multiple separate networks), each network becomes a diagonal block. Off-diagonal blocks are empty (no inter-layer edges) unless the networks share nodes.

Value

A `ggplot2` object.

Examples

```
## Not run:
# Single network
m <- matrix(runif(25), 5, 5)
rownames(m) <- colnames(m) <- LETTERS[1:5]
plot_heatmap(m)

# With clusters
clusters <- list(Group1 = c("A", "B"), Group2 = c("C", "D", "E"))
plot_heatmap(m, cluster_list = clusters)

# Custom colors and legend
plot_heatmap(m, colors = "heat", limits = c(0, 1), show_values = TRUE)

## End(Not run)

## Not run:
# Multi-layer (group_tna) - requires tna package and sequence data
if (requireNamespace("tna", quietly = TRUE)) {
  mod <- tna::tna(tna::group_regulation)
  plot_heatmap(mod)
}

## End(Not run)
```

plot_htna

Plot Heterogeneous TNA Network (Multi-Group Layout)

Description

Plots a TNA model with nodes arranged in multiple groups using geometric layouts:

- 2 groups: Bipartite (two vertical columns or horizontal rows)
- 3+ groups: Polygon (nodes along edges of a regular polygon)

Supports triangle (3), rectangle (4), pentagon (5), hexagon (6), and beyond.

Usage

```
plot_htna(  
  x,  
  node_list = NULL,  
  community = NULL,  
  layout = "auto",  
  use_list_order = TRUE,  
  jitter = FALSE,  
  jitter_amount = 0.8,  
  jitter_side = "first",  
  orientation = "vertical",  
  group1_pos = -2,  
  group2_pos = 2,  
  group_spacing = NULL,  
  node_spacing = NULL,  
  columns = 1,  
  column_spacing = NULL,  
  layout_margin = 0.15,  
  curvature = 0.4,  
  group1_color = "#4FC3F7",  
  group2_color = "#fbb550",  
  group1_shape = "circle",  
  group2_shape = "square",  
  group_colors = NULL,  
  group_shapes = NULL,  
  angle_spacing = 0.15,  
  edge_colors = NULL,  
  intra_curvature = NULL,  
  legend = TRUE,  
  legend_position = "bottomright",  
  extend_lines = FALSE,  
  scale = 1,  
  nodes = NULL,  
  label_abbrev = NULL,
```

```

    ...
)

htna(
  x,
  node_list = NULL,
  community = NULL,
  layout = "auto",
  use_list_order = TRUE,
  jitter = FALSE,
  jitter_amount = 0.8,
  jitter_side = "first",
  orientation = "vertical",
  group1_pos = -2,
  group2_pos = 2,
  group_spacing = NULL,
  node_spacing = NULL,
  columns = 1,
  column_spacing = NULL,
  layout_margin = 0.15,
  curvature = 0.4,
  group1_color = "#4FC3F7",
  group2_color = "#fbb550",
  group1_shape = "circle",
  group2_shape = "square",
  group_colors = NULL,
  group_shapes = NULL,
  angle_spacing = 0.15,
  edge_colors = NULL,
  intra_curvature = NULL,
  legend = TRUE,
  legend_position = "bottomright",
  extend_lines = FALSE,
  scale = 1,
  nodes = NULL,
  label_abbrev = NULL,
  ...
)

```

Arguments

<code>x</code>	A tna object, weight matrix, or <code>cograph_network</code> .
<code>node_list</code>	Node groups can be specified as: <ul style="list-style-type: none"> • A list of character vectors (node names per group) • A string column name from nodes data (e.g., "groups") • NULL to auto-detect from columns named: groups, cluster, community, etc. • NULL with community specified for algorithmic detection

community	Community detection method to use for auto-grouping. If specified, overrides node_list. See detect_communities for available methods: "louvain", "walk-trap", "fast_greedy", "label_prop", "infomap", "leiden".
layout	Layout type: "auto" (default), "bipartite", "polygon", or "circular". When "auto", uses bipartite for 2 groups and polygon for 3+ groups. "circular" places groups along arcs of a circle. Legacy values "triangle" and "rectangle" are supported as aliases for "polygon".
use_list_order	Logical. Use node_list order (TRUE) or weight-based order (FALSE). Only applies to bipartite layout.
jitter	Controls horizontal spread of nodes. Options: <ul style="list-style-type: none"> • FALSE (default) or 0: No jitter (nodes aligned in columns) • TRUE: Auto-compute jitter based on edge connectivity • Numeric (0-1): Amount of jitter (0.3 = spread nodes 30\) • Named list: Manual per-node offsets by label (e.g., list(Wrong = -0.2)) • Numeric vector of length n: Direct x-offsets for each node Only applies to bipartite layout.
jitter_amount	Base jitter amount when jitter=TRUE. Default 0.5. Higher values spread nodes more toward the center. Only applies to bipartite layout.
jitter_side	Which side(s) to apply jitter: "first", "second", "both", or "none". Default "first" (only first group nodes are jittered toward center). Only applies to bipartite layout.
orientation	Layout orientation for bipartite: "vertical" (two columns, default), "horizontal" (two rows), "facing" (both groups on same horizontal line, group1 left, group2 right, tip-to-tip), or "circular" (two facing semicircles with a gap between them). Ignored for triangle/rectangle layouts.
group1_pos	Position for first group in bipartite layout. Default -2. Overridden by group_spacing if specified.
group2_pos	Position for second group in bipartite layout. Default 2. Overridden by group_spacing if specified.
group_spacing	Numeric. Distance between the two groups in bipartite layout. Overrides group1_pos/group2_pos. For example, group_spacing = 6 places groups at x = -3 and x = 3. Default NULL (uses group1_pos/group2_pos).
node_spacing	Numeric. Vertical (or horizontal) gap between nodes within a group. Default NULL (auto-computed from the largest group size). Increase for more space between nodes (e.g., 0.5 or 0.8).
columns	Integer or vector of length 2. Number of sub-columns per group. A single value applies to both groups. A vector of 2 sets columns per group independently (e.g., c(2, 1) puts the first group in 2 columns). Nodes are distributed evenly across sub-columns. Default 1.
column_spacing	Numeric. Horizontal distance between sub-columns within a group. Default NULL (auto: node_spacing * 2).
layout_margin	Margin around the layout (0-1). Default 0.15. Increase if labels or self-loops are clipped at the edges.

curvature	Edge curvature amount. Default 0.4 for visible curves.
group1_color	Color for first group nodes. Default "#4FC3F7".
group2_color	Color for second group nodes. Default "#fb5555".
group1_shape	Shape for first group nodes. Default "circle".
group2_shape	Shape for second group nodes. Default "square".
group_colors	Vector of colors for each group. Overrides group1_color/group2_color. Required for 3+ groups if not using defaults.
group_shapes	Vector of shapes for each group. Overrides group1_shape/group2_shape. Required for 3+ groups if not using defaults.
angle_spacing	Controls empty space at corners (0-1). Default 0.15. Higher values create larger empty angles at vertices. Only applies to triangle/rectangle layouts.
edge_colors	Vector of colors for edges by source group. If NULL (default), uses darker versions of group_colors. Set to FALSE to use default edge color.
intra_curvature	Numeric. Curvature amount for intra-group edges (edges between nodes in the same group). When set, intra-group edges are drawn separately with curves that arc away from the opposing group. Default NULL (intra-group edges drawn normally by splot). Typical values: 0.3 to 1.0.
legend	Logical. Whether to show a legend. Default TRUE for polygon layouts.
legend_position	Position for legend: "topright", "topleft", "bottomright", "bottomleft", "right", "left", "top", "bottom". Default "bottomright".
extend_lines	Logical or numeric. Draw extension lines from nodes. Only applies to bipartite layout. <ul style="list-style-type: none"> • FALSE (default): No extension lines • TRUE: Draw lines extending toward the other group (default length 0.1) • Numeric: Length of extension lines
scale	Scaling factor for spacing parameters. Use scale > 1 for high-resolution output (e.g., scale = 4 for 300 dpi). This multiplies group positions and polygon/circular radius to maintain proper proportions at higher resolutions. Default 1.
nodes	Node metadata. Can be: <ul style="list-style-type: none"> • NULL (default): Use existing nodes data from cograph_network • Data frame: Must have label column for matching; if labels column exists, uses it for display text Display priority: labels column > label column (identifiers).
label_abbrev	Label abbreviation: NULL (none), integer (max chars), or "auto" (adaptive based on node count). Applied before passing to tplot.
...	Additional parameters passed to tplot().

Value

Invisibly returns the result from tplot().

Examples

```
# Create a 6-node network
mat <- matrix(runif(36, 0, 0.3), 6, 6)
diag(mat) <- 0
colnames(mat) <- rownames(mat) <- c("A", "B", "C", "D", "E", "F")

# Bipartite layout (2 groups)
groups <- list(Group1 = c("A", "B", "C"), Group2 = c("D", "E", "F"))
plot_htna(mat, groups)

# Polygon layout (3 groups)
groups3 <- list(X = c("A", "B"), Y = c("C", "D"), Z = c("E", "F"))
plot_htna(mat, groups3)
## Not run:
mat <- matrix(runif(36, 0, 0.3), 6, 6)
diag(mat) <- 0
colnames(mat) <- rownames(mat) <- c("A", "B", "C", "D", "E", "F")
groups <- list(Group1 = c("A", "B", "C"), Group2 = c("D", "E", "F"))
htna(mat, groups)

## End(Not run)
```

plot_mcml

Plot Multi-Cluster Multi-Layer Network

Description

Produces a two-layer hierarchical visualization of a clustered network. The **bottom layer** shows every node arranged inside elliptical cluster shells with full within-cluster and between-cluster edges drawn at the individual-node level. The **top layer** collapses each cluster into a single summary pie-chart node whose colored slice represents the proportion of within-cluster flow, with edges carrying the aggregated between-cluster weights. Dashed inter-layer lines connect each detail node to its corresponding summary node, making the hierarchical mapping explicit.

Usage

```
plot_mcml(
  x,
  cluster_list = NULL,
  mode = c("weights", "tna"),
  layer_spacing = NULL,
  spacing = 3,
  shape_size = 1.2,
  summary_size = 4,
  skew_angle = 60,
  aggregation = c("sum", "mean", "max"),
  minimum = 0,
  colors = NULL,
```

```
legend = TRUE,
show_labels = TRUE,
nodes = NULL,
label_size = NULL,
label_abbrev = NULL,
node_size = 1.8,
node_shape = "circle",
cluster_shape = "circle",
title = NULL,
subtitle = NULL,
title_size = 1.2,
subtitle_size = 0.9,
legend_position = "right",
legend_size = 0.7,
legend_pt_size = 1.2,
summary_labels = TRUE,
summary_label_size = 0.8,
summary_label_position = 3,
summary_label_color = "gray20",
summary_arrows = TRUE,
summary_arrow_size = 0.1,
between_arrows = FALSE,
edge_width_range = c(0.3, 1.3),
between_edge_width_range = c(0.5, 2),
summary_edge_width_range = c(0.5, 2),
edge_alpha = 0.35,
between_edge_alpha = 0.6,
summary_edge_alpha = 0.7,
inter_layer_alpha = 0.5,
edge_labels = FALSE,
edge_label_size = 0.5,
edge_label_color = "gray40",
edge_label_digits = 2,
summary_edge_labels = FALSE,
summary_edge_label_size = 0.6,
top_layer_scale = c(0.8, 0.25),
inter_layer_gap = 0.6,
node_radius_scale = 0.55,
shell_alpha = 0.15,
shell_border_width = 2,
node_border_color = "gray30",
summary_border_color = "gray20",
summary_border_width = 2,
label_color = "gray20",
label_position = 3,
...
)
```

Arguments

x	A weight matrix, tna object, cograph_network, or cluster_summary object. When a cluster_summary is provided (e.g., from <code>cluster_summary</code>), all aggregation has already been performed and the cluster_list, aggregation, and nodes parameters are ignored. See the Input Formats section for details.
cluster_list	How to assign nodes to clusters. Accepts: <ul style="list-style-type: none"> • A named list of character vectors — each element contains the node names belonging to that cluster, and the list names become the cluster labels (e.g., <code>list(GroupA = c("A", "B"), GroupB = c("C", "D"))</code>). • A string giving a column name in the node metadata (from a cograph_network) to use as the grouping variable. • NULL — attempt auto-detection from common column names (cluster, group, etc.) in node metadata. Ignored when x is a cluster_summary.
mode	What values to display on edges: <p>"weights" (default) Shows raw aggregated edge values. Useful when absolute magnitudes (e.g., total co-occurrences) matter.</p> <p>"tna" Row-normalizes the summary matrix so each row sums to 1, producing transition probabilities. Automatically enables edge_labels and summary_edge_labels unless you explicitly set them to FALSE.</p>
layer_spacing	Vertical distance between the bottom and top layers. NULL (default) auto-calculates a gap that prevents overlap based on cluster positions and shell sizes. Increase for more vertical separation; decrease to make the plot more compact.
spacing	Distance from the center to each cluster's position in the bottom layer. Larger values spread clusters farther apart. Default 3.
shape_size	Radius of each cluster's elliptical shell in the bottom layer. Increase when nodes overlap or shells feel cramped. Default 1.2.
summary_size	Size of the pie-chart summary nodes in the top layer. Controls the visual radius of each pie chart. Default 4.
skew_angle	Perspective tilt angle in degrees (0–90). At 0 the bottom layer is viewed from directly above (fully circular); at 90 it collapses to a flat line. Values around 45–70 give a natural table-top perspective. Default 60.
aggregation	Method for collapsing individual edge weights into between-cluster and within-cluster summaries: <p>"sum" (default) Total flow — appropriate when you care about the volume of all transitions between clusters.</p> <p>"mean" Average flow per node pair — useful when clusters differ in size and you want a size-normalized comparison.</p> <p>"max" Strongest single edge — highlights the dominant connection between each pair of clusters.</p> Ignored when x is a cluster_summary.
minimum	Edge weight threshold. Edges with absolute weight below this value are not drawn. Set to a small positive value (e.g., 0.01) to remove visual noise from near-zero edges. Default 0 (show all).

colors	Character vector of colors for the clusters. The first color is applied to the first cluster, and so on. Must have length equal to the number of clusters, or it will be recycled. When NULL (default), colors are auto-generated from a colorblind-safe palette.
legend	Logical. Whether to draw a legend mapping cluster names to colors. Default TRUE.
show_labels	Logical. Show node labels in the bottom layer. Default TRUE. Set to FALSE for dense networks where labels create clutter.
nodes	Node metadata data frame for custom display labels. Must contain a label column whose values match the row/column names of the weight matrix. If a labels column also exists, those values are used as display text (e.g., full names instead of codes). Display priority: labels column > label column. Ignored when x is a cluster_summary or cograph_network (which carries its own node metadata).
label_size	Text size (cex) for bottom-layer node labels. NULL (default) auto-scales to 0.6. Increase for readability in publication figures; decrease for dense networks.
label_abbrev	Controls label abbreviation to reduce overlap: <ul style="list-style-type: none"> • NULL — no abbreviation (show full labels). • An integer — truncate labels to this many characters. • "auto" — adaptively abbreviates based on the total number of nodes: more nodes triggers shorter abbreviations.
node_size	Size of individual detail nodes in the bottom layer. This controls the pie-chart radius for each node. Default 1.8.
node_shape	Shape for detail nodes in the bottom layer. Supported values: "circle", "square", "diamond", "triangle". Can be a single value applied to all nodes or a character vector of length equal to the number of nodes (one shape per node). Default "circle".
cluster_shape	Shape for summary nodes in the top layer. Same supported values as node_shape. Can be a single value or a vector of length equal to the number of clusters. Default "circle".
title	Main plot title displayed above the figure. Default NULL (no title).
subtitle	Subtitle displayed below the title. Default NULL (no subtitle).
title_size	Text size (cex.main) for the title. Default 1.2.
subtitle_size	Text size (cex.sub) for the subtitle. Default 0.9.
legend_position	Where to place the legend: "right", "left", "top", "bottom", or "none" to suppress it entirely. Default "right".
legend_size	Text size (cex) for legend labels. Default 0.7.
legend_pt_size	Point size (pt.cex) for legend symbols. Default 1.2.
summary_labels	Logical. Show cluster name labels next to the summary pie-chart nodes in the top layer. Default TRUE.
summary_label_size	Text size for summary labels. Default 0.8.

summary_label_position	Position of summary labels relative to nodes: 1 = below, 2 = left, 3 = above, 4 = right. Default 3 (above).
summary_label_color	Color for summary labels. Default "gray20".
summary_arrows	Logical. Draw arrowheads on summary-layer directed edges. Set to FALSE for undirected networks. Default TRUE.
summary_arrow_size	Size of arrowheads on summary edges. Default 0.10.
between_arrows	Logical. Draw arrowheads on between-cluster edges in the bottom layer. Default FALSE.
edge_width_range	Numeric vector $c(\min, \max)$ controlling the line-width range for within-cluster edges in the bottom layer. The weakest edge gets \min and the strongest gets \max . Default $c(0.3, 1.3)$.
between_edge_width_range	Numeric vector $c(\min, \max)$ for between-cluster edges in the bottom layer (shell-to-shell lines). Default $c(0.5, 2.0)$.
summary_edge_width_range	Numeric vector $c(\min, \max)$ for summary edges in the top layer. Default $c(0.5, 2.0)$.
edge_alpha	Transparency (0–1) for within-cluster edges. Lower values make these edges more subtle, keeping focus on between-cluster structure. Default 0.35.
between_edge_alpha	Transparency (0–1) for between-cluster edges in the bottom layer. Default 0.6.
summary_edge_alpha	Transparency (0–1) for summary-layer edges. Default 0.7.
inter_layer_alpha	Transparency (0–1) for the dashed inter-layer lines connecting detail nodes to their summary node. Lower values make these scaffolding lines less visually dominant. Default 0.5.
edge_labels	Logical. Show numeric weight labels on within-cluster edges. Default FALSE (automatically set to TRUE when mode = "tna").
edge_label_size	Text size for within-cluster edge labels. Default 0.5.
edge_label_color	Color for within-cluster edge labels. Default "gray40".
edge_label_digits	Number of decimal places for edge weight labels on both layers. Default 2.
summary_edge_labels	Logical. Show numeric weight labels on summary-layer edges. Default FALSE (automatically set to TRUE when mode = "tna").
summary_edge_label_size	Text size for summary edge labels. Default 0.6.

top_layer_scale	Numeric vector $c(x_scale, y_scale)$ controlling the horizontal and vertical radii of the oval on which summary nodes are placed, as multiples of spacing. Widen with $c(1.0, 0.25)$ or flatten with $c(0.8, 0.15)$ to adjust the top-layer shape. Default $c(0.8, 0.25)$.
inter_layer_gap	Vertical gap between the top of the bottom layer and the bottom of the top layer, as a multiple of spacing. Increase to separate the layers more. Default 0.6.
node_radius_scale	Radius of the circle on which nodes are arranged inside each cluster shell, as a fraction of shape_size. Increase to push nodes outward toward the shell border; decrease to pack them tighter. Default 0.55.
shell_alpha	Fill transparency (0–1) for cluster shells. Higher values make shells more opaque, giving stronger visual grouping but potentially obscuring edges. Default 0.15.
shell_border_width	Line width for cluster shell borders. Default 2.
node_border_color	Border color for detail nodes in the bottom layer. Default "gray30".
summary_border_color	Border color for summary pie-chart nodes. Default "gray20".
summary_border_width	Border line width for summary nodes. Default 2.
label_color	Text color for detail node labels. Default "gray20".
label_position	Position of detail node labels: 1 = below, 2 = left, 3 = above, 4 = right. Default 3.
...	Additional arguments (currently unused).

Details

Use `plot_mcml` when you need a simultaneous micro/macro view of cluster structure — the bottom layer reveals internal cluster dynamics while the top layer provides a bird's-eye summary. For a flat multi-cluster plot without the summary layer, see [plot_mtna](#). For stacked multilevel/multiplex layers, see [plot_mlna](#).

Two workflows:

1. **Direct:** pass a weight matrix (or `tna / cograph_network` object) together with `cluster_list`. The function calls `cluster_summary` internally to compute aggregated weights.
2. **Pre-computed:** call `cluster_summary` yourself, inspect or modify the result, then pass the `cluster_summary` object as `x`. This avoids redundant computation when you plot the same clustering repeatedly with different visual settings.

Mode:

- "weights" (default) — displays raw aggregated edge values. Use this when the absolute magnitude of transitions matters.
- "tna" — row-normalizes the summary matrix to transition probabilities (rows sum to 1) and automatically enables edge labels on both layers (unless you explicitly set `edge_labels` or `summary_edge_labels` to `FALSE`).

Layout logic: Bottom-layer clusters are arranged on a circle of radius spacing, flattened by the perspective skew_angle. Nodes inside each cluster sit on a smaller circle of radius shape_size * node_radius_scale. The top-layer summary nodes are placed on an oval above the bottom layer whose proportions are controlled by top_layer_scale.

Value

Invisibly returns the cluster_summary object used for plotting. This object can be passed back to plot_mcml() to avoid recomputation, inspected with print(), or fed to as_tna for further analysis.

Input Formats

x accepts four types:

matrix A square numeric weight matrix with row/column names matching the node identifiers in cluster_list.

tna A TNA model object. The \$weights matrix is extracted automatically.

cograph_network A cograph network object. Weights are extracted via to_matrix() and node metadata (display labels) is read from the \$nodes data frame.

cluster_summary A pre-computed summary from cluster_summary. When this type is passed, the cluster_list, aggregation, and nodes parameters are ignored because the summary already contains everything needed.

Edge Types

The plot contains four distinct edge categories, each with its own set of visual parameters:

Within-cluster (bottom) Edges connecting nodes inside the same cluster shell. Controlled by edge_width_range, edge_alpha, edge_labels, edge_label_size, edge_label_color, and edge_label_digits.

Between-cluster (bottom) Edges from one cluster shell to another, drawn between shell borders. Controlled by between_edge_width_range and between_edge_alpha.

Summary (top) Edges between summary pie-chart nodes in the top layer. Controlled by summary_edge_width_range, summary_edge_alpha, summary_edge_labels, summary_edge_label_size, summary_arrows, and summary_arrow_size.

Inter-layer (dashed) Dashed lines connecting each detail node to its cluster's summary node. Controlled by inter_layer_alpha.

Customization Quick Reference

Visual element	Key parameters
Cluster spacing / perspective	spacing, skew_angle
Cluster shell appearance	shape_size, shell_alpha, shell_border_width, colors
Detail nodes	node_size, node_shape, node_border_color
Detail labels	show_labels, label_size, label_abbrev, label_color, label_position
Summary nodes	summary_size, cluster_shape, summary_border_color, summary_border_width

Summary labels	summary_labels, summary_label_size, summary_label_color, summary_label_position
Within-cluster edges	edge_width_range, edge_alpha, edge_labels
Between-cluster edges	between_edge_width_range, between_edge_alpha
Summary edges	summary_edge_width_range, summary_edge_alpha, summary_edge_labels, summary_arrow
Inter-layer lines	inter_layer_alpha
Top-layer layout	top_layer_scale, inter_layer_gap
Title / legend	title, subtitle, legend, legend_position

See Also

[cluster_summary](#) for pre-computing aggregated cluster data, [plot_mtna](#) for flat multi-cluster visualization (no summary layer), [plot_mlna](#) for stacked multilevel/multiplex layer visualization, [aggregate_weights](#) for the low-level weight aggregation used internally, [detect_communities](#) for algorithmic cluster detection

Examples

```
# --- Setup: create a test matrix ---
mat <- matrix(runif(36), 6, 6)
diag(mat) <- 0
colnames(mat) <- rownames(mat) <- LETTERS[1:6]

clusters <- list(
  Cluster1 = c("A", "B"),
  Cluster2 = c("C", "D"),
  Cluster3 = c("E", "F")
)

# 1. Basic usage – pass matrix + clusters directly
plot_mcml(mat, clusters)

# 2. Pre-compute with cluster_summary for reuse
cs <- cluster_summary(mat, clusters)
plot_mcml(cs)

## Not run:
# 3. TNA mode – transition probabilities with edge labels
plot_mcml(mat, clusters, mode = "tna")

# 4. Custom shapes – different shape per cluster
plot_mcml(mat, clusters,
  node_shape = "diamond",
  cluster_shape = c("circle", "square", "triangle")
)

# 5. Styling – custom colors, transparency, edge widths
plot_mcml(mat, clusters,
  colors = c("#1b9e77", "#d95f02", "#7570b3"),
```

```
    edge_alpha = 0.5,
    between_edge_alpha = 0.8,
    shell_alpha = 0.25,
    edge_width_range = c(0.5, 2.0)
  )

# 6. Edge labels on both layers
plot_mcml(mat, clusters,
  edge_labels = TRUE,
  summary_edge_labels = TRUE,
  edge_label_digits = 1
)

# 7. Layout tuning – adjust spacing, perspective, and layer gap
plot_mcml(mat, clusters,
  spacing = 4,
  skew_angle = 45,
  top_layer_scale = c(1.0, 0.3),
  inter_layer_gap = 0.8
)

# 8. With mean aggregation for size-normalized comparison
plot_mcml(mat, clusters,
  aggregation = "mean",
  title = "Mean-aggregated cluster network"
)

# 9. Label abbreviation for dense networks
big_mat <- matrix(runif(400), 20, 20)
diag(big_mat) <- 0
colnames(big_mat) <- rownames(big_mat) <- paste0("Node_", 1:20)
big_clusters <- list(
  Alpha = paste0("Node_", 1:7),
  Beta = paste0("Node_", 8:14),
  Gamma = paste0("Node_", 15:20)
)
plot_mcml(big_mat, big_clusters, label_abbrev = "auto")

# 10. Minimal clean plot – no legend, no labels, no arrows
plot_mcml(mat, clusters,
  legend = FALSE,
  show_labels = FALSE,
  summary_labels = FALSE,
  summary_arrows = FALSE
)

## End(Not run)
```

Description

Plot a network combining symmetric (undirected) and asymmetric (directed) matrices with appropriate edge styling.

Creates a network visualization combining edges from a symmetric matrix (rendered as straight undirected edges) and an asymmetric matrix (rendered as curved directed edges).

Usage

```
plot_mixed_network(
  sym_matrix,
  asym_matrix,
  layout = "oval",
  sym_color = "ivory4",
  asym_color = COGRAPH_SCALE$tna_edge_color,
  curvature = 0.3,
  edge_width = NULL,
  node_size = 7,
  title = NULL,
  threshold = 0,
  edge_labels = TRUE,
  arrow_size = 0.61,
  edge_label_size = 0.6,
  edge_label_position = 0.7,
  initial = NULL,
  ...
)
```

Arguments

sym_matrix	A symmetric matrix representing undirected relationships. These edges will be drawn straight without arrows.
asym_matrix	An asymmetric matrix representing directed relationships. These edges will be drawn curved with arrows. Reciprocal edges curve in opposite directions.
layout	Layout algorithm or coordinate matrix. Default "oval".
sym_color	Color for symmetric/undirected edges. Default "#457B9D" (steel blue).
asym_color	Color for asymmetric/directed edges. Can be a single color or a vector of two colors for positive/negative directions. Default "#003355" (dark blue, matching TNA style).
curvature	Curvature magnitude for directed edges. Default 0.3.
edge_width	Edge width(s). If NULL (default), scales automatically by edge weight like TNA plots. Pass a numeric value to override.
node_size	Node size. Default 7.
title	Plot title. Default NULL.
threshold	Minimum absolute edge weight to display. Values with $\text{abs}(\text{value}) < \text{threshold}$ are set to zero (edge removed). Default 0. Zero-weight edges are always removed regardless of this setting.

edge_labels	Show edge weight labels. Default TRUE.
arrow_size	Arrow head size for directed edges. Default 0.61 (TNA style).
edge_label_size	Size of edge labels. Default 0.6.
edge_label_position	Position of edge labels along edge (0-1). Default 0.7.
initial	Optional named numeric vector of initial state probabilities (length = number of nodes). When provided, nodes are drawn as donuts with the fill proportion equal to the initial probability. Default NULL.
...	Additional arguments passed to <code>splot()</code> .

Value

Invisibly returns the combined `cograph_network` object.

Examples

```
# Create symmetric matrix (undirected)
sym <- matrix(0, 4, 4, dimnames = list(LETTERS[1:4], LETTERS[1:4]))
sym[1,2] <- sym[2,1] <- 0.5
sym[3,4] <- sym[4,3] <- 0.6

# Create asymmetric matrix (directed)
asym <- matrix(0, 4, 4, dimnames = list(LETTERS[1:4], LETTERS[1:4]))
asym[1,3] <- 0.7
asym[3,1] <- 0.3
asym[2,4] <- 0.8
asym[4,2] <- 0.4

# Plot combined network
plot_mixed_network(sym, asym, title = "Mixed Network")
```

Description

Visualizes multilevel/multiplex networks where multiple layers are stacked in a 3D perspective view. Each layer contains nodes connected by solid edges (within-layer), while dashed lines connect nodes between adjacent layers (inter-layer edges). Each layer is enclosed in a parallelogram shell giving a pseudo-3D appearance.

Usage

```
plot_mlna(  
  model,  
  layer_list = NULL,  
  community = NULL,  
  layout = "horizontal",  
  layer_spacing = 4,  
  layer_width = 8,  
  layer_depth = 4,  
  skew_angle = 25,  
  node_spacing = 0.7,  
  colors = NULL,  
  shapes = NULL,  
  edge_colors = NULL,  
  within_edges = TRUE,  
  between_edges = TRUE,  
  between_style = 2,  
  show_border = TRUE,  
  legend = TRUE,  
  legend_position = "topright",  
  curvature = 0.15,  
  node_size = 3,  
  minimum = 0,  
  scale = 1,  
  show_labels = TRUE,  
  nodes = NULL,  
  label_abbrev = NULL,  
  ...  
)
```

```
mlna(  
  model,  
  layer_list = NULL,  
  community = NULL,  
  layout = "horizontal",  
  layer_spacing = 4,  
  layer_width = 8,  
  layer_depth = 4,  
  skew_angle = 25,  
  node_spacing = 0.7,  
  colors = NULL,  
  shapes = NULL,  
  edge_colors = NULL,  
  within_edges = TRUE,  
  between_edges = TRUE,  
  between_style = 2,  
  show_border = TRUE,  
  legend = TRUE,
```

```

    legend_position = "topright",
    curvature = 0.15,
    node_size = 3,
    minimum = 0,
    scale = 1,
    show_labels = TRUE,
    nodes = NULL,
    label_abbrev = NULL,
    ...
)

```

Arguments

model	A tna object, weight matrix, or cograph_network.
layer_list	Layers can be specified as: <ul style="list-style-type: none"> • A list of character vectors (node names per layer) • A string column name from nodes data (e.g., "layer") • NULL to auto-detect from columns named: layer, layers, groups, etc. • NULL with community specified for algorithmic detection
community	Community detection method to use for auto-layering. If specified, overrides layer_list. See detect_communities for available methods: "louvain", "walk-trap", "fast_greedy", "label_prop", "infomap", "leiden".
layout	Node layout within layers: "horizontal" (default) spreads nodes horizontally, "circle" arranges nodes in an ellipse, "spring" uses force-directed placement based on within-layer connections.
layer_spacing	Vertical distance between layer centers. Default 2.5.
layer_width	Horizontal width of each layer shell. Default 5.
layer_depth	Depth of each layer (for 3D effect). Default 2.5.
skew_angle	Angle of perspective skew in degrees. Default 25.
node_spacing	Node placement ratio within layer (0-1). Default 0.7. Higher values spread nodes closer to the layer edges.
colors	Vector of colors for each layer. Default auto-generated.
shapes	Vector of shapes for each layer. Default cycles through "circle", "square", "diamond", "triangle".
edge_colors	Vector of edge colors by source layer. If NULL (default), uses darker versions of layer colors.
within_edges	Logical. Show edges within layers (solid lines). Default TRUE.
between_edges	Logical. Show edges between adjacent layers (dashed lines). Default TRUE.
between_style	Line style for between-layer edges. Default 2 (dashed). Use 1 for solid, 3 for dotted.
show_border	Logical. Draw parallelogram shells around layers. Default TRUE.
legend	Logical. Whether to show legend. Default TRUE.

legend_position	Position for legend. Default "topright".
curvature	Edge curvature for within-layer edges. Default 0.15.
node_size	Size of nodes. Default 2.5.
minimum	Minimum edge weight threshold. Edges below this are hidden. Default 0.
scale	Scaling factor for spacing parameters. Use scale > 1 for high-resolution output (e.g., scale = 4 for 300 dpi). This multiplies layer_spacing, layer_width, and layer_depth to maintain proper proportions at higher resolutions. Default 1.
show_labels	Logical. Show node labels. Default TRUE.
nodes	Node metadata. Can be: <ul style="list-style-type: none"> • NULL (default): Use existing nodes data from cograph_network • Data frame: Must have label column for matching; if labels column exists, uses it for display text Display priority: labels column > label column (identifiers).
label_abbrev	Label abbreviation: NULL (none), integer (max chars), or "auto" (adaptive based on node count).
...	Additional parameters (currently unused).

Value

Invisibly returns NULL.

See [plot_mlna](#).

Examples

```
## Not run:
# Create multilevel network
set.seed(42)
nodes <- paste0("N", 1:15)
m <- matrix(runif(225, 0, 0.3), 15, 15)
diag(m) <- 0
colnames(m) <- rownames(m) <- nodes

# Define 3 layers
layers <- list(
  Macro = paste0("N", 1:5),
  Meso = paste0("N", 6:10),
  Micro = paste0("N", 11:15)
)

# Basic usage
plot_mlna(m, layers)

# Customized
plot_mlna(m, layers,
  layer_spacing = 2.5,
  layer_width = 5,
```

```

        between_style = 2, # dashed
        minimum = 0.1)

# Circle layout within layers
plot_mlna(m, layers, layout = "circle")

## End(Not run)
## Not run:
nodes <- paste0("N", 1:9)
m <- matrix(runif(81, 0, 0.3), 9, 9)
diag(m) <- 0
colnames(m) <- rownames(m) <- nodes
layers <- list(L1 = nodes[1:3], L2 = nodes[4:6], L3 = nodes[7:9])
mlna(m, layers)

## End(Not run)

```

plot_ml_heatmap

Multilayer Network Heatmap

Description

Visualizes multiple network layers as heatmaps on tilted 3D-perspective planes, similar to the plot_mlna network visualization style.

Usage

```

plot_ml_heatmap(
  x,
  layer_list = NULL,
  colors = "viridis",
  layer_spacing = 2.5,
  skew = 0.4,
  compress = 0.6,
  show_connections = FALSE,
  connection_color = "#E63946",
  connection_style = "dashed",
  show_borders = TRUE,
  border_color = "black",
  border_width = 1,
  cell_border_color = "white",
  cell_border_width = 0.2,
  show_labels = TRUE,
  label_size = 5,
  show_legend = TRUE,
  legend_title = "Weight",
  title = NULL,
  limits = NULL,

```

```

    na_color = "grey90",
    threshold = 0
  )

```

Arguments

<code>x</code>	A list of matrices (one per layer), a <code>group_tna</code> object, <code>cograph_network</code> , or a single matrix with <code>layer_list</code> specified.
<code>layer_list</code>	Optional list defining layers, column name string, or NULL for auto-detection from <code>cograph_network</code> nodes.
<code>colors</code>	Color palette: "viridis", "heat", "blues", "reds", "inferno", "plasma", or a vector of colors. Default "viridis".
<code>layer_spacing</code>	Vertical spacing between layers. Default 2.5.
<code>skew</code>	Horizontal skew for perspective effect (0-1). Default 0.4.
<code>compress</code>	Vertical compression for perspective (0-1). Default 0.6.
<code>show_connections</code>	Show inter-layer connection lines? Default FALSE.
<code>connection_color</code>	Color for inter-layer connections. Default "#E63946".
<code>connection_style</code>	Line style: "dashed", "solid", "dotted". Default "dashed".
<code>show_borders</code>	Show layer outline borders? Default TRUE.
<code>border_color</code>	Color for layer borders. Default "black".
<code>border_width</code>	Width of layer borders. Default 1.
<code>cell_border_color</code>	Color for cell borders. Default "white".
<code>cell_border_width</code>	Width of cell borders. Default 0.2.
<code>show_labels</code>	Show layer name labels? Default TRUE.
<code>label_size</code>	Size of layer labels. Default 5.
<code>show_legend</code>	Show color legend? Default TRUE.
<code>legend_title</code>	Title for legend. Default "Weight".
<code>title</code>	Plot title. Default NULL.
<code>limits</code>	Color scale limits <code>c(min, max)</code> . NULL for auto.
<code>na_color</code>	Color for NA values. Default "grey90".
<code>threshold</code>	Minimum absolute value to display. Cells with <code>abs(value) < threshold</code> are set to NA (rendered as background). Default 0.

Value

A `ggplot2` object.

Examples

```
## Not run:
# List of matrices
layers <- list(
  Layer1 = matrix(runif(16), 4, 4),
  Layer2 = matrix(runif(16), 4, 4),
  Layer3 = matrix(runif(16), 4, 4)
)
plot_ml_heatmap(layers)

# With connections
plot_ml_heatmap(layers, show_connections = TRUE)

# Custom styling
plot_ml_heatmap(layers,
  colors = "plasma",
  layer_spacing = 3,
  skew = 0.5
)

## End(Not run)
```

plot_mtna

Multi-Cluster TNA Network Plot

Description

Visualizes multiple network clusters with summary edges between clusters and individual edges within clusters. Each cluster is displayed as a shape (circle, square, diamond, triangle) containing its nodes.

Usage

```
plot_mtna(
  x,
  cluster_list = NULL,
  community = NULL,
  layout = "circle",
  spacing = 4,
  shape_size = 1.8,
  node_spacing = 0.5,
  colors = NULL,
  shapes = NULL,
  edge_colors = NULL,
  bundle_edges = TRUE,
  bundle_strength = 0.8,
  summary_edges = TRUE,
```

```
aggregation = c("sum", "mean", "max", "min", "median", "density"),
within_edges = TRUE,
show_border = TRUE,
legend = TRUE,
legend_position = "topright",
curvature = 0.3,
node_size = 3,
layout_margin = 0.15,
scale = 1,
show_labels = FALSE,
nodes = NULL,
label_size = NULL,
label_abbrev = NULL,
cluster_shape = NULL,
...
)

mtna(
  x,
  cluster_list = NULL,
  community = NULL,
  layout = "circle",
  spacing = 4,
  shape_size = 1.8,
  node_spacing = 0.5,
  colors = NULL,
  shapes = NULL,
  edge_colors = NULL,
  bundle_edges = TRUE,
  bundle_strength = 0.8,
  summary_edges = TRUE,
  aggregation = c("sum", "mean", "max", "min", "median", "density"),
  within_edges = TRUE,
  show_border = TRUE,
  legend = TRUE,
  legend_position = "topright",
  curvature = 0.3,
  node_size = 3,
  layout_margin = 0.15,
  scale = 1,
  show_labels = FALSE,
  nodes = NULL,
  label_size = NULL,
  label_abbrev = NULL,
  cluster_shape = NULL,
  ...
)
```

Arguments

x	A tna object, weight matrix, cograph_network, or cluster_summary object.
cluster_list	Clusters can be specified as: <ul style="list-style-type: none"> • A list of character vectors (node names per cluster) • A string column name from nodes data (e.g., "groups") • NULL with community specified for auto-detection
community	Community detection method to use for auto-clustering. If specified, overrides cluster_list. See detect_communities for available methods: "louvain", "walktrap", "fast_greedy", "label_prop", "infomap", "leiden".
layout	How to arrange the clusters: "circle" (default), "grid", "horizontal", "vertical".
spacing	Distance between cluster centers. Default 4.
shape_size	Size of each cluster shape (shell radius). Default 1.8.
node_spacing	Radius for node placement within shapes (0-1 relative to shape_size). Default 0.5.
colors	Vector of colors for each cluster. Default auto-generated.
shapes	Vector of shapes for each cluster: "circle", "square", "diamond", "triangle". Default cycles through these.
edge_colors	Vector of edge colors by source cluster. Default auto-generated.
bundle_edges	Logical. Bundle inter-cluster edges through channels. Default TRUE.
bundle_strength	How tightly to bundle edges (0-1). Default 0.8.
summary_edges	Logical. Show aggregated summary edges between clusters instead of individual node edges. Default TRUE.
aggregation	Method for aggregating edge weights between clusters: "sum" (total flow), "mean" (average strength), "max" (strongest link), "min" (weakest link), "median", or "density" (normalized by possible edges). Default "sum". Only used when summary_edges = TRUE.
within_edges	Logical. When summary_edges is TRUE, also show individual edges within each cluster. Default TRUE.
show_border	Logical. Draw a border around each cluster. Default TRUE.
legend	Logical. Whether to show legend. Default TRUE.
legend_position	Position for legend. Default "topright".
curvature	Edge curvature. Default 0.3.
node_size	Size of nodes inside shapes. Default 3.
layout_margin	Margin around the layout as fraction of range. Default 0.15.
scale	Scaling factor for spacing parameters. Use scale > 1 for high-resolution output (e.g., scale = 4 for 300 dpi). This multiplies spacing and shape_size to maintain proper proportions at higher resolutions. Default 1.
show_labels	Logical. Show node labels inside clusters. Default FALSE.

nodes	Node metadata. Can be: <ul style="list-style-type: none"> • NULL (default): Use existing nodes data from <code>cograph_network</code> • Data frame: Must have <code>label</code> column for matching; if <code>labels</code> column exists, uses it for display text Display priority: <code>labels</code> column > <code>label</code> column (identifiers).
label_size	Label text size. Default NULL (auto-scaled).
label_abbrev	Label abbreviation: NULL (none), integer (max chars), or "auto" (adaptive based on node count).
cluster_shape	Shape for cluster summary nodes when using summary view. Can be single value or vector. Overrides shapes. Default NULL (use shapes).
...	Additional parameters passed to <code>plot_tna()</code> .

Value

Invisibly returns a `cluster_summary` object for summary mode, or the `plot_tna` result otherwise.

See [plot_mtna](#).

See Also

[cluster_summary](#), [plot_mcml](#)

Examples

```
## Not run:
# Create network with 4 clusters
nodes <- paste0("N", 1:20)
m <- matrix(runif(400, 0, 0.3), 20, 20)
diag(m) <- 0
colnames(m) <- rownames(m) <- nodes

clusters <- list(
  North = paste0("N", 1:5),
  East = paste0("N", 6:10),
  South = paste0("N", 11:15),
  West = paste0("N", 16:20)
)

# Summary edges between clusters + individual edges within
plot_mtna(m, clusters, summary_edges = TRUE)

# With node labels
plot_mtna(m, clusters, show_labels = TRUE, label_abbrev = 3)

# Control spacing and sizes
plot_mtna(m, clusters, spacing = 4, shape_size = 1.5, node_spacing = 0.6)

## End(Not run)
## Not run:
nodes <- paste0("N", 1:12)
```

```
m <- matrix(runif(144, 0, 0.3), 12, 12)
diag(m) <- 0
colnames(m) <- rownames(m) <- nodes
clusters <- list(C1 = nodes[1:4], C2 = nodes[5:8], C3 = nodes[9:12])
mtna(m, clusters)

## End(Not run)
```

plot_netobject_group *Plot a Group of Nestimate netobjects*

Description

Creates a multi-panel plot for a netobject_group list, one panel per group. Mirrors plot_group_permutation() in structure.

Usage

```
plot_netobject_group(
  x,
  nrow = NULL,
  ncol = NULL,
  common_scale = TRUE,
  title_prefix = NULL,
  ...
)

## S3 method for class 'netobject_group'
plot(x, ...)
```

Arguments

x	A netobject_group object (named list of netobjects).
nrow	Integer: number of rows in the panel grid. Auto-computed if NULL.
ncol	Integer: number of columns in the panel grid. Auto-computed if NULL.
common_scale	Logical: use the same maximum weight across all panels? Default TRUE.
title_prefix	Character: optional prefix added before each group name in panel titles.
...	Additional arguments passed to splot().

Value

Invisibly returns x.

Examples

```

mat <- matrix(c(0, .5, .3, .5, 0, .4, .3, .4, 0), 3, 3)
colnames(mat) <- rownames(mat) <- c("A", "B", "C")
net1 <- as_cograph(mat)
net2 <- as_cograph(mat * 0.5)
grp <- structure(list(G1 = net1, G2 = net2), class = c("netobject_group", "list"))
plot_netobject_group(grp)

```

plot_netobject_ml *Plot a Multilevel Nestimate netobject*

Description

Creates a side-by-side plot for a netobject_ml object, showing the between-person and within-person networks.

Usage

```

plot_netobject_ml(
  x,
  layout = NULL,
  common_scale = TRUE,
  titles = c("Between-person", "Within-person"),
  ...
)

## S3 method for class 'netobject_ml'
plot(x, ...)

```

Arguments

x	A netobject_ml object with \$between and \$within networks.
layout	Character: layout algorithm. Default "oval" (deterministic).
common_scale	Logical: use the same maximum weight for both panels? Default TRUE.
titles	Character vector of length 2: panel titles. Default c("Between-person", "Within-person").
...	Additional arguments passed to splot().

Value

Invisibly returns x.

Examples

```
mat <- matrix(c(0, .5, .3, .5, 0, .4, .3, .4, 0), 3, 3)
colnames(mat) <- rownames(mat) <- c("A", "B", "C")
btw <- as_cograph(mat)
wth <- as_cograph(mat * 0.6)
ml <- structure(list(between = btw, within = wth), class = c("netobject_ml", "list"))
plot_netobject_ml(ml)
```

plot_network_evolution

Plot Network Evolution (Small Multiples)

Description

Displays a network at different time points side by side. Accepts an edge list data frame with a time column, or a pre-built list of networks. All panels share the same node layout for visual comparison.

Usage

```
plot_network_evolution(
  x,
  time = NULL,
  slices = NULL,
  cumulative = FALSE,
  labels = NULL,
  layout = "spring",
  ncol = NULL,
  node_size = 5,
  seed = 42,
  ...
)
```

Arguments

x	An edge list data frame with columns from, to, and a time column, OR a list of network objects (matrices, igraph, etc.).
time	Character. Name of the time/group column in x. Ignored if x is a list.
slices	Integer or NULL. Number of equal-width time bins. Default NULL uses unique values of the time column.
cumulative	Logical. If TRUE, each panel shows all edges up to that time point (growing network). If FALSE (default), each panel shows only edges from that period.
labels	Character vector of panel labels. Default NULL (auto from time values).
layout	Layout specification. Default "spring".
ncol	Integer. Grid columns. Default auto.
node_size	Numeric. Default 5.
seed	Integer or NULL. Default 42.
...	Additional arguments passed to splot .

Value

Invisible list of edge-list data frames (one per panel).

Examples

```
## Not run:
# Edge list with time column
edges$week <- sample(1:4, nrow(edges), replace = TRUE)
cograph::plot_network_evolution(edges, time = "week")

# Cumulative: edges accumulate over time
cograph::plot_network_evolution(edges, time = "week", cumulative = TRUE)

## End(Not run)
```

plot_net_bootstrap_group

Plot a Group Bootstrap Result

Description

Plots each cluster's net_bootstrap in a grid, routing every panel through `splot.net_bootstrap` so significance styling (solid vs dashed edges) is preserved. Earlier versions extracted `bs$original` per cluster and handed plain netobjects to `splot()`, which dispatches to `splot.netobject` — that path has no concept of significance, so every edge rendered identically.

Usage

```
plot_net_bootstrap_group(x, nrow = NULL, ncol = NULL, common_scale = TRUE, ...)
```

```
## S3 method for class 'net_bootstrap_group'
plot(x, ...)
```

Arguments

<code>x</code>	A <code>net_bootstrap_group</code> object (list of <code>net_bootstrap</code>).
<code>nrow, ncol</code>	Grid dimensions. Defaults to auto-computed square layout.
<code>common_scale</code>	Logical: use the same maximum weight across panels? Default TRUE.
<code>...</code>	Additional arguments passed to <code>splot.net_bootstrap</code> (e.g. <code>display = "significant"</code> , <code>show_stars = FALSE</code>).

Value

Invisibly returns `x`.

Examples

```
## Not run:
grp <- Nestimate::cluster_network(data, k = 2)
gbs <- Nestimate::bootstrap_network(grp, iter = 100)
plot_net_bootstrap_group(gbs)

## End(Not run)
```

plot_net_stability *Plot Centrality Stability Results*

Description

Visualizes the centrality stability analysis from a `net_stability` object. Shows how centrality correlations drop as cases are removed.

Usage

```
plot_net_stability(x, ...)
```

```
## S3 method for class 'net_stability'
plot(x, ...)
```

Arguments

`x` A `net_stability` object (from `Nestimate::centrality_stability`).

`...` Additional graphical arguments.

Value

Invisibly returns `x`.

Examples

```
## Not run:
net <- Nestimate::build_network(data, method = "tna")
cs <- Nestimate::centrality_stability(net, iter = 100)
plot_net_stability(cs)

## End(Not run)
```

plot_robustness *Plot Network Robustness*

Description

Creates a visualization of network robustness showing the fraction of remaining nodes in the largest connected component during sequential node/edge removal. Supports comparison of multiple attack strategies.

Usage

```
plot_robustness(
  ...,
  x = NULL,
  measures = c("betweenness", "degree", "random"),
  colors = NULL,
  title = "Network Robustness: sequential removal of nodes",
  xlab = "Fraction of removed nodes",
  ylab = "Fraction of remaining nodes",
  lwd = 1.5,
  legend_pos = "topright",
  n_iter = 1000,
  seed = NULL,
  type = "vertex"
)
```

Arguments

...	One or more robustness results from robustness , or named arguments to pass networks for on-the-fly computation.
x	Network for computing robustness on-the-fly.
measures	Character vector of attack strategies to compare. Default c("betweenness", "degree", "random").
colors	Named vector of colors. Default: green=Degree, red=Betweenness, blue=Random (matching Nature paper style).
title	Plot title. Default "Network Robustness: sequential removal of nodes".
xlab	X-axis label. Default "Fraction of removed nodes".
ylab	Y-axis label. Default "Fraction of remaining nodes".
lwd	Line width. Default 1.5.
legend_pos	Legend position. Default "topright".
n_iter	Number of iterations for random. Default 1000.
seed	Random seed. Default NULL.
type	Removal type. Default "vertex".

Value

Invisibly returns combined data frame of all robustness results.

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::sample_pa(50, m = 2, directed = FALSE)

  # Quick comparison of all strategies
  plot_robustness(x = g, n_iter = 20)

  # Or compute separately
  rob1 <- robustness(g, measure = "betweenness")
  rob2 <- robustness(g, measure = "degree")
  rob3 <- robustness(g, measure = "random", n_iter = 20)
  plot_robustness(rob1, rob2, rob3)
}
```

plot_simplicial

Simplicial Complex Visualization

Description

Visualize higher-order pathways as smooth blobs overlaid on a network layout. Source nodes are blue, target nodes are red.

Usage

```
plot_simplicial(
  x = NULL,
  pathways = NULL,
  method = "hon",
  max_pathways = 10L,
  layout = "circle",
  labels = NULL,
  node_color = "#4A7FB5",
  target_color = "#E8734A",
  ring_color = "#F5A623",
  node_size = 22,
  label_size = 5,
  blob_alpha = 0.25,
  blob_colors = NULL,
  blob_linetype = NULL,
  blob_linewidth = 0.7,
  blob_line_alpha = 0.8,
  shadow = TRUE,
  title = NULL,
  dismantled = FALSE,
```

```

    ncol = NULL,
    ...
)

```

Arguments

x	A network object: tna, netobject, matrix, igraph, cograph_network, net_hon, or net_hypa. When x is a tna or netobject with sequence data and pathways is NULL, higher-order pathways are built automatically using the method parameter.
pathways	Character vector of pathway strings, a list of character vectors, or a net_hon / net_hypa object. String separators: "A B -> C", "A, B, C", "A - B - C", "A B C". Last state is the target. When NULL and x is a model with sequence data, pathways are built automatically.
method	Pathway source when auto-building from a tna/netobject: "hon" (default, higher-order network), "hypa" (anomalous paths via hypergeometric null), or "rules" (association-rule itemsets via Nestimate::association_rules; rules are rendered as single-colored blobs because itemsets are undirected).
max_pathways	Maximum number of pathways to display. HON pathways are ranked by count, HYPA by anomaly ratio. NULL shows all. Default 10.
layout	"circle" (default) or a coordinate matrix.
labels	Display labels. NULL uses state names.
node_color	Source node fill color.
target_color	Target node fill color.
ring_color	Donut ring color.
node_size	Node point size.
label_size	Label text size.
blob_alpha	Blob fill transparency.
blob_colors	Blob fill colors (recycled).
blob_linetype	Blob border line styles (recycled).
blob_linewidth	Blob border line width.
blob_line_alpha	Blob border line transparency.
shadow	Draw soft drop shadows?
title	Plot title.
dismantled	If TRUE, one panel per pathway arranged in a grid layout.
ncol	Number of columns in the grid when dismantled = TRUE. Default NULL auto-selects based on the number of pathways.
...	Additional arguments passed to Nestimate::build_hon() or Nestimate::build_hypa() when auto-building.

Details

Supports direct use with tna and netobject models: when x has sequence data, HON or HYPY pathways are built automatically (requires the **Nestimate** package). Pathways can also be passed as net_hon or net_hypa objects, with labels auto-translated when x is a tna/netobject.

Value

A ggplot object (or combined grid if dismantled), invisibly.

Examples

```
## Not run:
mat <- matrix(runif(16), 4, 4,
              dimnames = list(LETTERS[1:4], LETTERS[1:4]))
diag(mat) <- 0
plot_simplicial(mat, c("A B -> C", "B C -> D"))

# Direct from tna model (requires Nestimate):
# model <- tna::tna(tna::group_regulation)
# plot_simplicial(model, dismantled = TRUE)
# plot_simplicial(model, method = "hypa")

# With pre-built HON + tna for label translation:
# hon <- Nestimate::build_hon(as.data.frame(model$data))
# plot_simplicial(model, hon, dismantled = TRUE)

## End(Not run)
```

plot_temporal

Temporal Network Prism (3D Glass Box)

Description

Displays a network at different time points as vertical planes inside a 3D oblique-projection box, with time flowing left to right. Each network plane extends into the depth of the box.

Usage

```
plot_temporal(
  x,
  time = NULL,
  slices = NULL,
  cumulative = FALSE,
  labels = NULL,
  layout = "spring",
  node_size = 2.5,
  node_color = "steelblue",
```

```

node_shape = 21,
node_border = "gray30",
edge_color = "#E41A1C",
edge_width = 1.5,
edge_alpha = 0.35,
plane_color = "gray92",
plane_alpha = 0.2,
plane_border = "gray60",
plane_lty = 2,
box = TRUE,
box_color = "gray40",
connections = FALSE,
connection_color = "gray50",
connection_alpha = 0.15,
minimum = 0,
show_labels = FALSE,
label_size = 0.4,
title = NULL,
angle = c(1, 0.7),
seed = 42,
...
)

```

Arguments

x	An edge list data frame with columns from, to, and a time column, OR a cograph_network (reads time from stored data), OR a named list of network objects.
time	Character. Name of the time column.
slices	Integer or NULL. Number of equal-width time bins. Default NULL uses unique time values.
cumulative	Logical. If TRUE, edges accumulate. Default FALSE.
labels	Character vector of layer labels. Default auto.
layout	Character or matrix. Shared node layout. Default "spring".
node_size	Numeric. Node size. Default 2.5.
node_color	Character or vector. Node fill color. A single color applies to all layers, or a vector of length n_layers for per-layer colors. Default "steelblue".
node_shape	Integer. Point shape (pch). Default 21 (filled circle).
node_border	Character. Node border color. Default "gray30".
edge_color	Character or vector. Edge color (single or per-layer). Default "#E41A1C".
edge_width	Numeric. Base edge width. Actual width scales by weight. Default 1.5.
edge_alpha	Numeric. Edge transparency (0-1). Default 0.35.
plane_color	Character or vector. Plane fill color (single or per-layer). Default "gray92".
plane_alpha	Numeric. Plane fill transparency (0-1). Default 0.2.

plane_border	Character. Plane border color. Default "gray60".
plane_lty	Integer. Plane border line type. Default 2 (dashed).
box	Logical. Draw 3D bounding box. Default TRUE.
box_color	Character. Box edge color. Default "gray40".
connections	Logical. Draw lines connecting same nodes across planes. Default FALSE.
connection_color	Character. Default "gray50".
connection_alpha	Numeric. Default 0.15.
minimum	Numeric. Minimum edge weight to display. Default 0.
show_labels	Logical. Default FALSE.
label_size	Numeric. Label text size. Default 0.4.
title	Character or NULL. Plot title. Default NULL.
angle	Numeric vector of length 2: c(dz_x, dz_y) controlling the oblique projection shear. Default c(1.0, 0.7).
seed	Integer or NULL. Default 42.
...	Additional arguments (currently unused).

Value

Invisible list of adjacency matrices per layer.

See Also

[plot_network_evolution](#), [plot_mlna](#)

Examples

```
## Not run:
edges$week <- sample(1:3, nrow(edges), replace = TRUE)
cograph::plot_temporal(edges, time = "week")

# Customized
cograph::plot_temporal(edges, time = "week",
  node_color = "coral", edge_color = "gray30",
  node_size = 3, edge_width = 2, plane_lty = 3,
  angle = c(1.2, 0.8), title = "Classroom Evolution")

## End(Not run)
```

`plot_tna`*TNA-Style Network Plot (qgraph Compatible)*

Description

A drop-in replacement for `qgraph::qgraph()` that uses `cograph`'s `splot` engine. Accepts `qgraph` parameter names for seamless migration from `qgraph` to `cograph`.

Usage

```
plot_tna(  
  x,  
  color = NULL,  
  labels = NULL,  
  layout = "oval",  
  theme = "colorblind",  
  mar = c(0.1, 0.1, 0.1, 0.1),  
  cut = NULL,  
  edge.labels = TRUE,  
  edge.label.position = 0.7,  
  edge.label.cex = 0.6,  
  edge.color = COGRAPH_SCALE$tna_edge_color,  
  vsize = 7,  
  pie = NULL,  
  pieColor = NULL,  
  lty = NULL,  
  directed = NULL,  
  minimum = NULL,  
  posCol = NULL,  
  negCol = NULL,  
  arrowAngle = NULL,  
  title = NULL,  
  ...  
)  
  
tplot(  
  x,  
  color = NULL,  
  labels = NULL,  
  layout = "oval",  
  theme = "colorblind",  
  mar = c(0.1, 0.1, 0.1, 0.1),  
  cut = NULL,  
  edge.labels = TRUE,  
  edge.label.position = 0.7,  
  edge.label.cex = 0.6,  
  edge.color = COGRAPH_SCALE$tna_edge_color,
```

```

    vsize = 7,
    pie = NULL,
    pieColor = NULL,
    lty = NULL,
    directed = NULL,
    minimum = NULL,
    posCol = NULL,
    negCol = NULL,
    arrowAngle = NULL,
    title = NULL,
    ...
)

```

Arguments

x	A weight matrix (adjacency matrix) or tna object
color	Node fill colors
labels	Node labels
layout	Layout: "circle", "spring", "oval", or a coordinate matrix
theme	Plot theme ("colorblind", "gray", etc.)
mar	Plot margins (numeric vector of length 4)
cut	Edge emphasis threshold
edge.labels	Show edge weight labels
edge.label.position	Position of edge labels along edge (0-1)
edge.label.cex	Edge label size multiplier
edge.color	Edge colors
vsize	Node size
pie	Pie/donut fill values (e.g., initial probabilities)
pieColor	Pie/donut segment colors
lty	Line type for edges (1=solid, 2=dashed, 3=dotted)
directed	Logical, is the graph directed?
minimum	Minimum edge weight to display
posCol	Color for positive edges
negCol	Color for negative edges
arrowAngle	Arrow head angle in radians. Default pi/6 (30 degrees).
title	Plot title
...	Additional arguments passed to splot()

Value

Invisibly returns the `cograph_network` object from `splot()`.

Invisibly returns the `cograph_network` object from `splot()`.

Examples

```

# Simple usage
m <- matrix(runif(25), 5, 5)
plot_tna(m)

# With qgraph-style parameters
plot_tna(m, vsize = 15, edge.label.cex = 2, layout = "circle")

# With custom colors
plot_tna(m, color = rainbow(5), vsize = 10)

## Not run:
m <- matrix(runif(25), 5, 5)
tplot(m)

## End(Not run)

```

plot_trajectories *Plot Individual Trajectories*

Description

Creates an alluvial-style diagram where each individual's trajectory is shown as a separate line. This is an alias for `plot_transitions()` with `track_individuals = TRUE`.

Usage

```

plot_trajectories(
  x,
  from_title = NULL,
  title = NULL,
  from_colors = NULL,
  flow_color_by = "first",
  node_width = 0.08,
  node_border = NA,
  node_spacing = 0.02,
  label_size = 3.5,
  label_position = c("beside", "inside", "above", "below", "outside"),
  mid_label_position = NULL,
  label_halo = TRUE,
  label_color = "black",
  label_fontface = "plain",
  label_nudge = 0.02,
  title_size = 5,
  title_color = "black",
  title_fontface = "bold",
  curve_strength = 0.6,
  line_alpha = 0.3,

```

```

    line_width = 0.5,
    jitter_amount = 0.8,
    show_totals = FALSE,
    total_size = 4,
    total_color = "white",
    total_fontface = "bold",
    show_values = FALSE,
    value_position = c("center", "origin", "destination"),
    value_size = 3,
    value_color = "black",
    value_halo = NULL,
    value_fontface = "bold",
    value_nudge = 0.03,
    value_min = 0,
    value_digits = 2,
    column_gap = 1,
    proportional_nodes = TRUE,
    node_label_format = NULL,
    bundle_size = NULL,
    bundle_legend = TRUE,
    bundle_legend_size = 3,
    bundle_legend_color = "grey50",
    bundle_legend_fontface = "italic",
    bundle_legend_position = c("bottom", "top")
  )

```

Arguments

x	Input data in one of several formats: <ul style="list-style-type: none"> • A transition matrix (rows = from, cols = to, values = counts) • Two vectors: pass before as x and after as second argument (contingency table computed automatically, like chi-square) • A 2-column data frame (raw observations; table computed automatically) • A data frame with columns: from, to, count • A list of matrices for multi-step transitions
from_title	Title for the left column. Default "From". For multi-step, use a vector of titles (e.g., c("T1", "T2", "T3", "T4")).
title	Optional plot title. Applied via <code>ggplot2::labs(title = title)</code> .
from_colors	Colors for left-side nodes. Default uses palette.
flow_color_by	Color flows by "source", "destination", or NULL (use <code>flow_fill</code>). Default NULL.
node_width	Width of node rectangles (0-1 scale). Default 0.08.
node_border	Border color for nodes. Default NA (no border).
node_spacing	Vertical spacing between nodes (0-1 scale). Default 0.02.
label_size	Size of node labels. Default 3.5.

label_position	Position of node labels: "beside" (default), "inside", "above", "below", "outside". Applied to first and last columns. See mid_label_position for middle columns.
mid_label_position	Position of labels for intermediate (middle) columns. Same options as label_position. Default NULL uses label_position value.
label_halo	Logical: add white halo around labels for readability? Default TRUE.
label_color	Color of state name labels. Default "black".
label_fontface	Font face of state name labels ("plain", "bold", "italic", "bold.italic"). Default "plain".
label_nudge	Distance between node edge and label (in plot units). Default 0.02. Increase for more spacing.
title_size	Size of column titles. Default 5.
title_color	Color of column title text. Default "black".
title_fontface	Font face of column titles. Default "bold".
curve_strength	Controls bezier curve shape (0-1). Default 0.6.
line_alpha	Alpha for individual tracking lines. Default 0.3.
line_width	Width of individual tracking lines. Default 0.5.
jitter_amount	Vertical jitter for individual lines (0-1). Default 0.8.
show_totals	Logical: show total counts on nodes? Default FALSE.
total_size	Size of total labels. Default 4.
total_color	Color of total labels. Default "white".
total_fontface	Font face of total labels. Default "bold".
show_values	Logical: show transition counts on flows? Default FALSE.
value_position	Position of flow values: "center", "origin", "destination", "outside_origin", "outside_destination". Default "center".
value_size	Size of value labels on flows. Default 3.
value_color	Color of value labels. Default "black".
value_halo	Logical: add halo around flow value labels? Default NULL (inherits from label_halo).
value_fontface	Font face of flow value labels. Default "bold".
value_nudge	Distance of value labels from node edge when using "origin" or "destination" positions. Default 0.03.
value_min	Minimum count to show a flow value label. Default 0 (show all). Use to hide small flows (e.g., value_min = 100).
value_digits	Number of decimal places for flow value labels and node totals. Default 2.
column_gap	Horizontal spread of columns (0-1). Default 1 uses full width. Use smaller values (e.g., 0.6) to bring columns closer together.
proportional_nodes	Logical: size nodes proportionally to counts? Default TRUE.

node_label_format	Format string for node labels with {state} and {count} placeholders. Default NULL (plain state name). Example: "{state} (n={count})".
bundle_size	Controls line bundling for large datasets. Default NULL (no bundling). Integer >= 2: each drawn line represents that many cases. Numeric in (0,1): reduce to this fraction of original lines (e.g., 0.15 keeps about 15 percent of lines).
bundle_legend	Logical or character: show annotation when bundling is active? Default TRUE shows "Each line ~ N cases" below the plot. Pass a string to use custom text (with {n} placeholder for count).
bundle_legend_size	Size of the bundle legend text. Default 3.
bundle_legend_color	Color of the bundle legend text. Default "grey50".
bundle_legend_fontface	Font face of the bundle legend text. Default "italic".
bundle_legend_position	Position of the bundle legend: "bottom" (default) or "top".

Value

A ggplot2 object.

See Also

[plot_transitions](#), [plot_alluvial](#)

Examples

```
## Not run:  
# Track individual trajectories across time points  
df <- data.frame(  
  Baseline = c("Light", "Light", "Intense", "Resource"),  
  Week4 = c("Light", "Intense", "Intense", "Light"),  
  Week8 = c("Resource", "Intense", "Light", "Light")  
)  
plot_trajectories(df, flow_color_by = "first")  
  
## End(Not run)
```

plot_transitions

Plot Transitions Between States

Description

Creates an elegant alluvial/Sankey diagram showing how items flow from one set of categories to another. Useful for visualizing cluster transitions, state changes, or any categorical mapping.

Usage

```
plot_transitions(  
  x,  
  from_title = "From",  
  to_title = "To",  
  title = NULL,  
  from_colors = NULL,  
  to_colors = NULL,  
  flow_fill = "#888888",  
  flow_alpha = 0.4,  
  flow_color_by = NULL,  
  flow_border = NA,  
  flow_border_width = 0.5,  
  node_width = 0.08,  
  node_border = NA,  
  node_spacing = 0.02,  
  label_size = 3.5,  
  label_position = c("beside", "inside", "above", "below", "outside"),  
  mid_label_position = NULL,  
  label_halo = TRUE,  
  label_color = "black",  
  label_fontface = "plain",  
  label_nudge = 0.02,  
  title_size = 5,  
  title_color = "black",  
  title_fontface = "bold",  
  curve_strength = 0.6,  
  show_values = FALSE,  
  value_position = c("center", "origin", "destination", "outside_origin",  
    "outside_destination"),  
  value_size = 3,  
  value_color = "black",  
  value_halo = NULL,  
  value_fontface = "bold",  
  value_nudge = 0.03,  
  value_min = 0,  
  show_totals = FALSE,  
  total_size = 4,  
  total_color = "white",  
  total_fontface = "bold",  
  conserve_flow = TRUE,  
  min_flow = 0,  
  threshold = 0,  
  value_digits = 2,  
  column_gap = 1,  
  track_individuals = FALSE,  
  line_alpha = 0.3,  
  line_width = 0.5,  
)
```

```

    jitter_amount = 0.8,
    proportional_nodes = TRUE,
    node_label_format = NULL,
    bundle_size = NULL,
    bundle_legend = TRUE,
    bundle_legend_size = 3,
    bundle_legend_color = "grey50",
    bundle_legend_fontface = "italic",
    bundle_legend_position = c("bottom", "top")
  )

```

Arguments

x	Input data in one of several formats: <ul style="list-style-type: none"> • A transition matrix (rows = from, cols = to, values = counts) • Two vectors: pass before as x and after as second argument (contingency table computed automatically, like chi-square) • A 2-column data frame (raw observations; table computed automatically) • A data frame with columns: from, to, count • A list of matrices for multi-step transitions
from_title	Title for the left column. Default "From". For multi-step, use a vector of titles (e.g., c("T1", "T2", "T3", "T4")).
to_title	Title for the right column. Default "To". Ignored for multi-step.
title	Optional plot title. Applied via <code>ggplot2::labs(title = title)</code> .
from_colors	Colors for left-side nodes. Default uses palette.
to_colors	Colors for right-side nodes. Default uses palette.
flow_fill	Fill color for flows. Default "#888888" (grey). Ignored if <code>flow_color_by</code> is set.
flow_alpha	Alpha transparency for flows. Default 0.4.
flow_color_by	Color flows by "source", "destination", or NULL (use <code>flow_fill</code>). Default NULL.
flow_border	Border color for flows. Default NA (no border).
flow_border_width	Line width for flow borders. Default 0.5.
node_width	Width of node rectangles (0-1 scale). Default 0.08.
node_border	Border color for nodes. Default NA (no border).
node_spacing	Vertical spacing between nodes (0-1 scale). Default 0.02.
label_size	Size of node labels. Default 3.5.
label_position	Position of node labels: "beside" (default), "inside", "above", "below", "outside". Applied to first and last columns. See <code>mid_label_position</code> for middle columns.
mid_label_position	Position of labels for intermediate (middle) columns. Same options as <code>label_position</code> . Default NULL uses <code>label_position</code> value.
label_halo	Logical: add white halo around labels for readability? Default TRUE.

label_color	Color of state name labels. Default "black".
label_fontface	Font face of state name labels ("plain", "bold", "italic", "bold.italic"). Default "plain".
label_nudge	Distance between node edge and label (in plot units). Default 0.02. Increase for more spacing.
title_size	Size of column titles. Default 5.
title_color	Color of column title text. Default "black".
title_fontface	Font face of column titles. Default "bold".
curve_strength	Controls bezier curve shape (0-1). Default 0.6.
show_values	Logical: show transition counts on flows? Default FALSE.
value_position	Position of flow values: "center", "origin", "destination", "outside_origin", "outside_destination". Default "center".
value_size	Size of value labels on flows. Default 3.
value_color	Color of value labels. Default "black".
value_halo	Logical: add halo around flow value labels? Default NULL (inherits from label_halo).
value_fontface	Font face of flow value labels. Default "bold".
value_nudge	Distance of value labels from node edge when using "origin" or "destination" positions. Default 0.03.
value_min	Minimum count to show a flow value label. Default 0 (show all). Use to hide small flows (e.g., value_min = 100).
show_totals	Logical: show total counts on nodes? Default FALSE.
total_size	Size of total labels. Default 4.
total_color	Color of total labels. Default "white".
total_fontface	Font face of total labels. Default "bold".
conserve_flow	Logical: should left and right totals match? Default TRUE. When FALSE, each side scales independently (allows for "lost" or "gained" items).
min_flow	Minimum flow value to display. Default 0 (show all).
threshold	Minimum edge weight to display. Flows below this value are removed. Combined with min_flow: effective minimum is max(threshold, min_flow). Default 0.
value_digits	Number of decimal places for flow value labels and node totals. Default 2.
column_gap	Horizontal spread of columns (0-1). Default 1 uses full width. Use smaller values (e.g., 0.6) to bring columns closer together.
track_individuals	Logical: draw individual lines instead of aggregated flows? Default FALSE. When TRUE, each row in the data frame becomes a separate line.
line_alpha	Alpha for individual tracking lines. Default 0.3.
line_width	Width of individual tracking lines. Default 0.5.
jitter_amount	Vertical jitter for individual lines (0-1). Default 0.8.

proportional_nodes	Logical: size nodes proportionally to counts? Default TRUE.
node_label_format	Format string for node labels with {state} and {count} placeholders. Default NULL (plain state name). Example: "{state} (n={count})".
bundle_size	Controls line bundling for large datasets. Default NULL (no bundling). Integer ≥ 2 : each drawn line represents that many cases. Numeric in (0,1): reduce to this fraction of original lines (e.g., 0.15 keeps about 15 percent of lines).
bundle_legend	Logical or character: show annotation when bundling is active? Default TRUE shows "Each line ~ N cases" below the plot. Pass a string to use custom text (with {n} placeholder for count).
bundle_legend_size	Size of the bundle legend text. Default 3.
bundle_legend_color	Color of the bundle legend text. Default "grey50".
bundle_legend_fontface	Font face of the bundle legend text. Default "italic".
bundle_legend_position	Position of the bundle legend: "bottom" (default) or "top".

Details

The function creates smooth bezier curves connecting nodes from the left column to the right column. Flow width is proportional to the transition count. Nodes are sized proportionally to their total flow.

Value

A ggplot2 object.

Examples

```
## Not run:
# From a transition matrix
mat <- matrix(c(50, 10, 5, 15, 40, 10, 5, 20, 30), 3, 3, byrow = TRUE)
rownames(mat) <- c("Light", "Resource", "Intense")
colnames(mat) <- c("Light", "PBL", "Resource")
plot_transitions(mat, from_title = "Time 1", to_title = "Time 2")

# From a 2-column data frame - auto-computes contingency table
before <- c("A", "A", "B", "B", "A", "C", "B", "C")
after <- c("X", "Y", "X", "Z", "X", "Y", "Z", "X")
df <- data.frame(time1 = before, time2 = after)
plot_transitions(df, from_title = "Time 1", to_title = "Time 2")

# Custom colors
plot_transitions(mat,
  from_colors = c("#FFD166", "#06D6A0", "#9D4EDD"),
  to_colors = c("#FFD166", "#EF476F", "#06D6A0"))
```

```

)

## End(Not run)

## Not run:
# Multi-step transitions (list of matrices)
mat1 <- matrix(c(40, 10, 5, 15, 35, 5, 5, 15, 25), 3, 3, byrow = TRUE,
              dimnames = list(c("A","B","C"), c("A","B","C")))
mat2 <- matrix(c(35, 15, 5, 10, 30, 10, 10, 10, 30), 3, 3, byrow = TRUE,
              dimnames = list(c("A","B","C"), c("A","B","C")))
mat3 <- matrix(c(30, 20, 5, 5, 25, 15, 15, 5, 35), 3, 3, byrow = TRUE,
              dimnames = list(c("A","B","C"), c("A","B","C")))
plot_transitions(list(mat1, mat2, mat3),
                 from_title = c("T1", "T2", "T3", "T4"),
                 show_totals = TRUE
)

## End(Not run)

```

```
print.cograph_communities
```

Print Community Structure

Description

Print Community Structure

Usage

```
## S3 method for class 'cograph_communities'
print(x, ...)
```

Arguments

x	A cograph_communities object.
...	Ignored.

Value

Invisibly returns the original object.

Examples

```
g <- igraph::make_graph("Zachary")
comm <- community_louvain(g)
print(comm)
```

```
print.cograph_degree_fit
      Print method for cograph_degree_fit
```

Description

Displays the comparison table of fitted distributions sorted by AIC.

Usage

```
## S3 method for class 'cograph_degree_fit'
print(x, digits = 4, ...)
```

Arguments

x	A cograph_degree_fit object from fit_degree_distribution .
digits	Number of decimal places. Default 4.
...	Additional arguments passed to print.data.frame.

Value

Invisible x.

Examples

```
adj <- matrix(c(0, 1, 1, 0, 0,
               1, 0, 1, 1, 0,
               1, 1, 0, 1, 1,
               0, 1, 1, 0, 1,
               0, 0, 1, 1, 0), 5, 5, byrow = TRUE)
fit <- cograph::fit_degree_distribution(adj,
  distributions = c("exponential", "poisson"))
print(fit)
```

```
print.cograph_network Print cograph_network Object
```

Description

Print cograph_network Object

Usage

```
## S3 method for class 'cograph_network'
print(x, ...)
```

Arguments

x A cograph_network object.
 ... Ignored.

Value

The input object x, invisibly.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- cograph(adj)
print(net)
```

project_bipartite *Project Bipartite Network to One-Mode*

Description

Projects a two-mode (bipartite/incidence) network into a one-mode adjacency matrix. Row-mode projection yields a matrix of shared-column connections among row nodes; column-mode projection does the converse.

Usage

```
project_bipartite(x, mode = "rows", method = "sum", ...)
```

Arguments

x An incidence matrix (rows = type 1 nodes, columns = type 2 nodes) where non-zero entries indicate connections. Can also be a data.frame with columns type1, type2, and optionally weight.

mode Character. "rows" (default) projects onto row nodes (result: n_rows x n_rows). "columns" projects onto column nodes (result: n_cols x n_cols).

method Character. Projection method:
 "sum" Weighted projection: $A \%*\% t(A)$ (rows) or $t(A) \%*\% A$ (columns). Edge weight equals sum of shared connection weights.
 "binary" Co-occurrence count: binarize A first, then compute overlap. Edge weight equals number of shared connections.
 "jaccard" Jaccard similarity: $\text{shared} / (\text{total}_i + \text{total}_j - \text{shared})$ for each pair.
 "cosine" Cosine similarity: dot product of row (or column) vectors divided by the product of their norms.
 "newman" Newman's weighted projection (Newman 2001): each shared affiliation contributes $1 / (d_k - 1)$ where d_k is the degree of the shared node. Gives more weight to connections through exclusive affiliations.

... Additional arguments (currently unused).

Details

For the Newman projection, affiliations shared with only one node of the focal type ($d_k = 1$) are skipped, since $1 / (d_k - 1)$ is undefined. This follows the convention in Newman (2001).

Value

A square adjacency matrix with row and column names preserved from the input. Diagonal is set to 0 (no self-loops).

References

Newman, M. E. J. (2001). Scientific collaboration networks. II. Shortest paths, weighted networks, and centrality. *Physical Review E*, 64(1), 016132.

See Also

[is_bipartite](#), [plot_heatmap](#)

Examples

```
# Incidence matrix: 4 students x 3 courses
inc <- matrix(c(1, 1, 0,
               1, 0, 1,
               0, 1, 1,
               1, 1, 1), 4, 3, byrow = TRUE)
rownames(inc) <- paste0("S", 1:4)
colnames(inc) <- paste0("C", 1:3)

# Student co-enrollment (weighted)
cograph::project_bipartite(inc, mode = "rows", method = "sum")

# Course overlap (Jaccard similarity)
cograph::project_bipartite(inc, mode = "columns", method = "jaccard")

# Newman's weighted projection
cograph::project_bipartite(inc, mode = "rows", method = "newman")
```

reaching_global

Global Reaching Centrality (Mones, Vicsek & Vicsek 2012)

Description

A graph-level hierarchy measure computed from per-node local reaching centralities:

$$GRC(G) = \frac{1}{N-1} \sum_v \left(\max_u LRC(u) - LRC(v) \right)$$

Usage

```
reaching_global(x, mode = "all", ...)
```

Arguments

x	Network input (matrix, igraph, network, cograph_network, tna object).
mode	For directed networks: "all" (default), "in", or "out".
...	Additional arguments passed to centrality_reaching_local .

Details

Values close to 0 indicate a flat network (all nodes reach equal proportions of the graph); values close to 1 indicate strong hierarchical structure. Matches `networkx.global_reaching_centrality` exactly.

Value

A single numeric value in $[0, 1]$.

References

Mones, E., Vicsek, L., & Vicsek, T. (2012). Hierarchy measure for complex networks. *PLoS ONE*, 7(3), e33799.

See Also

[centrality_reaching_local](#), [summarize_network](#).

Examples

```
# Star graph: highly hierarchical (directed out from center)
adj <- matrix(0, 5, 5)
adj[1, 2:5] <- 1
rownames(adj) <- colnames(adj) <- LETTERS[1:5]
reaching_global(adj, mode = "out")
```

register_layout

Register a Custom Layout

Description

Register a new layout algorithm that can be used for network visualization.

Usage

```
register_layout(name, layout_fn)
```

Arguments

name	Character. Name of the layout.
layout_fn	Function. A function that computes node positions. Should accept a Cograph-Network object and return a matrix with x, y columns.

Value

Invisible NULL.

Examples

```
# Register a simple random layout
register_layout("random", function(network, ...) {
  n <- network$n_nodes
  cbind(x = runif(n), y = runif(n))
})
```

register_shape	<i>Register a Custom Shape</i>
----------------	--------------------------------

Description

Register a new shape that can be used for node rendering.

Usage

```
register_shape(name, draw_fn)
```

Arguments

name	Character. Name of the shape.
draw_fn	Function. A function that draws the shape. Should accept parameters: x, y, size, fill, border_color, border_width, ...

Value

Invisible NULL.

Examples

```
# Register a custom hexagon shape
register_shape("hexagon", function(x, y, size, fill, border_color, border_width, ...) {
  angles <- seq(0, 2 * pi, length.out = 7)
  grid::polygonGrob(
    x = x + size * cos(angles),
    y = y + size * sin(angles),
    gp = grid::gpar(fill = fill, col = border_color, lwd = border_width)
  )
})
```

register_svg_shape *Register Custom SVG Shape*

Description

Register an SVG file or string as a custom node shape.

Usage

```
register_svg_shape(name, svg_source)
```

Arguments

name	Character: unique name for this shape (used in node_shape parameter).
svg_source	Character: path to SVG file OR inline SVG string.

Value

Invisible NULL. The shape is registered for use with sn_nodes().

Examples

```
## Not run:
# Register from file
register_svg_shape("custom_icon", "path/to/icon.svg")

# Register from inline SVG
register_svg_shape("simple_star",
  '<svg viewBox="0 0 100 100">
  <polygon points="50,5 20,99 95,39 5,39 80,99" fill="currentColor"/>
  </svg>')

# Use in network
cograph(adj) |> sn_nodes(shape = "custom_icon")

## End(Not run)
```

register_theme *Register a Custom Theme*

Description

Register a new theme for network visualization.

Usage

```
register_theme(name, theme)
```

Arguments

name	Character. Name of the theme.
theme	A CographTheme object or a list of theme parameters.

Value

Invisible NULL.

Examples

```
# Register a custom theme
register_theme("custom", list(
  background = "white",
  node_fill = "steelblue",
  node_border = "navy",
  edge_color = "gray50"
))
```

render-ggplot *ggplot2 Conversion*

Description

Convert Cograph network to ggplot2 object.

Value

A ggplot2 object representing the network.

Examples

```
## Not run:
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
p <- sn_ggplot(adj)

## End(Not run)
```

`render-grid`*Grid Rendering*

Description

Main grid-based rendering functions.

Value

See individual functions: `soplot` returns a `cograph_network` object invisibly; `sn_ggplot` returns a `ggplot2` object.

Examples

```
## Not run:
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
soplot(adj)

## End(Not run)
```

`rich_club`*Rich Club Coefficient*

Description

Computes the rich club curve across all prominence thresholds, measuring whether prominent nodes preferentially direct their strongest ties toward each other. Supports both unweighted (Colizza et al. 2006) and weighted (Opsahl et al. 2008) formulations.

Usage

```
rich_club(
  x,
  rich = c("k", "s"),
  weighted = TRUE,
  normalized = TRUE,
  n_random = 100,
  directed = NULL,
  seed = NULL,
  digits = NULL,
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object.
rich	Character. Prominence definition: "k" (degree, default) or "s" (strength / weighted degree).
weighted	Logical. If TRUE (default), compute the weighted rich club coefficient. If FALSE, compute the unweighted version (density among rich nodes).
normalized	Logical. If TRUE (default), normalize against degree-preserving random graphs and include confidence intervals.
n_random	Integer. Number of random graphs for normalization. Default 100.
directed	Logical or NULL. Default NULL (auto-detect).
seed	Integer or NULL. Random seed for reproducibility. Default NULL.
digits	Integer or NULL. Round numeric output. Default NULL.
...	Additional arguments passed to to_igraph .

Details

Unweighted: $\phi(k) = 2E_{>k} / (N_{>k}(N_{>k} - 1))$

Weighted: $\phi^w(k) = W_{>k} / \sum_{l=1}^{E_{>k}} w_l^{ranked}$

Normalization: $\phi_{norm} = \phi_{obs} / \bar{\phi}_{rand}$. A value > 1 indicates rich club ordering beyond what the degree sequence alone explains.

Value

A data frame with class "cograph_rich_club" and columns: threshold, n_rich, phi, and if normalized: phi_norm, phi_rand, ci_lo, ci_hi.

References

Opsahl, T., Colizza, V., Panzarasa, P. & Ramasco, J.J. (2008). Prominence and control: The weighted rich-club effect. *Physical Review Letters*, 101, 168702.

Colizza, V., Flammini, A., Serrano, M.A. & Vespignani, A. (2006). Detecting rich-club ordering in complex networks. *Nature Physics*, 2, 110-115.

See Also

[rich_club_local](#), [robustness](#), [centrality](#)

Examples

```
## Not run:
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::sample_pa(100, m = 3, directed = FALSE)
  rc <- cograph::rich_club(g)
  plot(rc)
}

## End(Not run)
```

rich_club_local *Local Rich Club Score*

Description

For each node, measures whether it preferentially directs its strongest ties toward prominent nodes. A score > 1 means the node's ties to prominent nodes are stronger than average.

Usage

```
rich_club_local(
  x,
  prominence = NULL,
  rich = c("k", "s"),
  directed = NULL,
  digits = NULL,
  sort_by = "score",
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object.
prominence	Integer or logical vector indicating which nodes are prominent (1/TRUE = prominent), OR a numeric threshold. If NULL, nodes above median degree (or strength) are prominent.
rich	Character. "k" (degree, default) or "s" (strength). Used when prominence is NULL or a threshold.
directed	Logical or NULL. Default NULL (auto-detect).
digits	Integer or NULL. Round scores. Default NULL.
sort_by	Character or NULL. Column to sort by (descending). Default "score".
...	Additional arguments passed to to_igraph .

Details

For each node i : $r_i = \bar{w}_{i \rightarrow rich} / \bar{w}_i$

Value

A data frame with columns node and score, sorted by score descending. Values > 1 indicate the node directs disproportionately strong ties to prominent nodes.

References

Opsahl, T., Colizza, V., Panzarasa, P. & Ramasco, J.J. (2008). Prominence and control: The weighted rich-club effect. *Physical Review Letters*, 101, 168702.

See Also

[rich_club](#), [centrality](#)

Examples

```
adj <- matrix(c(0,5,3,1, 5,0,4,2, 3,4,0,1, 1,2,1,0), 4, 4)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
cograph::rich_club_local(adj, prominence = c(1, 1, 0, 0))
```

robustness

Network Robustness Analysis

Description

Performs a targeted attack or random failure analysis on a network, calculating the size of the largest connected component after sequential vertex or edge removal.

In a targeted attack, vertices are sorted by degree or betweenness centrality (or edges by betweenness), and successively removed from highest to lowest. In a random failure analysis, vertices/edges are removed in random order.

Usage

```
robustness(
  x,
  type = c("vertex", "edge"),
  measure = c("betweenness", "degree", "random"),
  strategy = c("sequential", "static"),
  n_iter = 1000,
  mode = "all",
  seed = NULL,
  ...
)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
type	Character string; either "vertex" or "edge" removals. Default: "vertex"
measure	Character string; sort by "betweenness", "degree", or "random". Default: "betweenness"
strategy	Character string; "sequential" (default) recalculates centrality after each removal. "static" computes centrality once on the original network and removes nodes in that fixed order (brainGraph-style). Only affects targeted attacks; random removal is unaffected.
n_iter	Integer; number of iterations for random analysis. Default: 1000 (matching brainGraph convention)

mode	For directed networks: "all", "in", or "out". Default "all".
seed	Random seed for reproducibility. Default NULL.
...	Additional arguments passed to <code>to_igraph</code>

Details

Three attack strategies are available:

Targeted Attack - Betweenness (default): Vertices/edges are sorted by betweenness centrality and removed from highest to lowest. This targets nodes that bridge different network regions.

Targeted Attack - Degree: Vertices are sorted by degree and removed from highest to lowest. This targets highly connected hub nodes. Note: for edge attacks, degree is not available; use betweenness instead.

Random Failure: Vertices/edges are removed in random order, averaged over `n_iter` iterations. This simulates random component failures.

Strategy: The `strategy` parameter controls how targeted attacks work:

- "sequential" (default): Recalculates centrality after each removal. This is a stronger attack because removing a hub changes which nodes become the new bridges/hubs.
- "static": Computes centrality once on the original network and removes nodes in that fixed order (as in brainGraph). This matches the original Albert et al. (2000) method.

Scale-free networks are typically robust to random failures but vulnerable to targeted attacks, while random networks degrade more uniformly.

Value

A data frame (class "cograph_robustness") with columns:

removed_pct Fraction of vertices/edges removed (0 to 1)

comp_size Size of largest component after removal

comp_pct Ratio of component size to original maximum

measure Attack strategy used

type Type of analysis (vertex or edge removal)

References

Albert, R., Jeong, H., & Barabasi, A.L. (2000). Error and attack tolerance of complex networks. *Nature*, 406, 378-381. doi:10.1038/35019019

See Also

[plot_robustness](#), [robustness_auc](#)

Examples

```
# Create a scale-free network
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::sample_pa(50, m = 2, directed = FALSE)

  # Targeted attack by betweenness
  rob_btw <- robustness(g, measure = "betweenness")

  # Targeted attack by degree
  rob_deg <- robustness(g, measure = "degree")

  # Random failure
  rob_rnd <- robustness(g, measure = "random", n_iter = 50)

  # View results
  head(rob_btw)
}
```

robustness_auc

Calculate Area Under Robustness Curve (AUC)

Description

Computes the area under the robustness curve using trapezoidal integration. Higher AUC indicates a more robust network. Maximum AUC is 1.0.

Usage

```
robustness_auc(x)
```

Arguments

x A robustness result from [robustness](#).

Value

Numeric AUC value between 0 and 1.

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::sample_pa(30, m = 2, directed = FALSE)

  rob_btw <- robustness(g, measure = "betweenness")
  rob_rnd <- robustness(g, measure = "random", n_iter = 20)

  cat("Betweenness attack AUC:", round(robustness_auc(rob_btw), 3), "\n")
  cat("Random failure AUC:", round(robustness_auc(rob_rnd), 3), "\n")
}
```

robustness_summary	<i>Summary of Robustness Analysis</i>
--------------------	---------------------------------------

Description

Provides a summary comparing robustness metrics across attack strategies.

Usage

```
robustness_summary(..., x = NULL, measures = NULL, n_iter = 1000)
```

Arguments

...	Robustness results to summarize.
x	Network for on-the-fly computation.
measures	Measures to compute if x provided.
n_iter	Iterations for random. Default 1000.

Value

Data frame with AUC and critical points for each measure.

Examples

```
## Not run:
if (requireNamespace("igraph", quietly = TRUE)) {
  g <- igraph::sample_pa(30, m = 2, directed = FALSE)
  robustness_summary(x = g, measures = c("degree", "random"), n_iter = 10)
}

## End(Not run)
```

select_bridges	<i>Select Bridge Edges</i>
----------------	----------------------------

Description

Select edges whose removal would disconnect the graph.

Usage

```
select_bridges(
  x,
  ...,
  .keep_isolates = FALSE,
  keep_format = FALSE,
  directed = NULL
)
```

Arguments

x Network input.
 ... Additional filter expressions.
 .keep_isolates Keep nodes with no edges? Default FALSE.
 keep_format Keep input format? Default FALSE.
 directed Auto-detect if NULL.

Value

A `cograph_network` with bridge edges only.

See Also

[select_edges](#), [select_nodes](#)

Examples

```
# Create network with bridge
adj <- matrix(0, 5, 5)
adj[1, 2] <- adj[2, 1] <- 1
adj[2, 3] <- adj[3, 2] <- 1 # Bridge
adj[3, 4] <- adj[4, 3] <- 1
adj[4, 5] <- adj[5, 4] <- 1
adj[3, 5] <- adj[5, 3] <- 1
rownames(adj) <- colnames(adj) <- LETTERS[1:5]

select_bridges(adj)
```

select_component	<i>Select Connected Component</i>
------------------	-----------------------------------

Description

Select nodes belonging to a specific connected component.

Usage

```
select_component(
  x,
  which = "largest",
  ...,
  .keep_edges = c("internal", "none"),
  keep_format = FALSE,
  directed = NULL
)
```

Arguments

x	Network input.
which	Component selection: "largest" (default) The largest connected component Integer Component by ID Character Component containing the named node
...	Additional filter expressions to apply after component selection.
.keep_edges	How to handle edges. Default "internal".
keep_format	Logical. Keep input format? Default FALSE.
directed	Logical or NULL. Auto-detect if NULL.

Value

A `cograph_network` with nodes in the selected component.

See Also

[select_nodes](#), [select_neighbors](#)

Examples

```
# Create disconnected network
adj <- matrix(0, 6, 6)
adj[1, 2] <- adj[2, 1] <- 1
adj[1, 3] <- adj[3, 1] <- 1
adj[4, 5] <- adj[5, 4] <- 1
adj[5, 6] <- adj[6, 5] <- 1
adj[4, 6] <- adj[6, 4] <- 1
rownames(adj) <- colnames(adj) <- LETTERS[1:6]

# Largest component
select_component(adj, which = "largest")

# Component containing node "A"
select_component(adj, which = "A")
```

select_edges

Select Edges with Lazy Computation

Description

A powerful edge selection function with lazy computation (only computes metrics actually referenced), multiple selection modes, and structural awareness (bridges, communities, reciprocity).

Usage

```

select_edges(
  x,
  ...,
  top = NULL,
  by = "weight",
  involving = NULL,
  between = NULL,
  bridges_only = FALSE,
  mutual_only = FALSE,
  community = "louvain",
  .keep_isolates = FALSE,
  keep_format = FALSE,
  directed = NULL
)

```

Arguments

x	Network input: <code>cograph_network</code> , <code>matrix</code> , <code>igraph</code> , <code>network</code> , or <code>tna</code> object.
...	Filter expressions using edge columns or computed metrics. Available variables: Edge columns <code>from</code> , <code>to</code> , <code>weight</code> , plus any custom Computed metrics <code>abs_weight</code> , <code>from_degree</code> , <code>to_degree</code> , <code>from_strength</code> , <code>to_strength</code> , <code>edge_betweenness</code> , <code>is_bridge</code> , <code>is_mutual</code> , <code>same_community</code>
top	Integer. Select top N edges by a metric.
by	Character. Metric for top selection. Default <code>"weight"</code> . Options: <code>"weight"</code> , <code>"abs_weight"</code> , <code>"edge_betweenness"</code> .
involving	Character or integer. Select edges involving these nodes (by name or index). An edge is selected if either endpoint matches.
between	List of two character/integer vectors. Select edges between two node sets. Example: <code>between = list(c("A", "B"), c("C", "D"))</code> .
bridges_only	Logical. Select only bridge edges (edges whose removal disconnects the graph). Default <code>FALSE</code> .
mutual_only	Logical. For directed networks, select only mutual (reciprocated) edges. Default <code>FALSE</code> .
community	Character. Community detection method for <code>same_community</code> variable. One of <code>"louvain"</code> , <code>"walktrap"</code> , <code>"fast_greedy"</code> , <code>"label_prop"</code> , <code>"infomap"</code> , <code>"leiden"</code> . Default <code>"louvain"</code> .
.keep_isolates	Logical. Keep nodes with no remaining edges? Default <code>FALSE</code> .
keep_format	Logical. If <code>TRUE</code> , return the same format as input. Default <code>FALSE</code> returns <code>cograph_network</code> .
directed	Logical or <code>NULL</code> . If <code>NULL</code> (default), auto-detect.

Details

Selection modes are combined with AND logic:

- `select_edges(x, top = 10, involving = "A")` selects top 10 edges **among those involving node A**
- All criteria must be satisfied for an edge to be selected

Edge metrics are computed lazily - only those actually referenced in expressions or required by selection modes are computed.

Value

A `cograph_network` object with selected edges. If `keep_format = TRUE`, returns the same type as input.

See Also

[filter_edges](#), [select_nodes](#), [select_bridges](#), [select_top_edges](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")

# Expression-based (lazy - only computes what's needed)
select_edges(adj, weight > 0.5)
select_edges(adj, abs_weight > 0.4)

# Top N edges by weight
select_edges(adj, top = 3)
select_edges(adj, top = 3, by = "edge_betweenness")

# Edges involving specific nodes
select_edges(adj, involving = "A")
select_edges(adj, involving = c("A", "B"))

# Edges between two node sets
select_edges(adj, between = list(c("A", "B"), c("C", "D")))

# Bridge edges only
select_edges(adj, bridges_only = TRUE)

# Combined: top 3 edges involving A
select_edges(adj, involving = "A", top = 3)

# Using endpoint degrees
select_edges(adj, from_degree >= 3 | to_degree >= 3)

# Within-community edges
```

```
select_edges(adj, same_community)
```

select_edges_between *Select Edges Between Node Sets*

Description

Select edges connecting two specified node sets.

Usage

```
select_edges_between(  
  x,  
  set1,  
  set2,  
  ...,  
  .keep_isolates = FALSE,  
  keep_format = FALSE,  
  directed = NULL  
)
```

Arguments

x	Network input.
set1	Character or integer. First node set (names or indices).
set2	Character or integer. Second node set (names or indices).
...	Additional filter expressions.
.keep_isolates	Keep nodes with no edges? Default FALSE.
keep_format	Keep input format? Default FALSE.
directed	Auto-detect if NULL.

Value

A `cograph_network` with edges between the two node sets.

See Also

[select_edges](#), [select_edges_involving](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")

# Edges between {A, B} and {C, D}
select_edges_between(adj, set1 = c("A", "B"), set2 = c("C", "D"))
```

```
select_edges_involving
```

Select Edges Involving Nodes

Description

Select edges where at least one endpoint is in the specified node set.

Usage

```
select_edges_involving(
  x,
  nodes,
  ...,
  .keep_isolates = FALSE,
  keep_format = FALSE,
  directed = NULL
)
```

Arguments

x	Network input.
nodes	Character or integer. Node names or indices.
...	Additional filter expressions.
.keep_isolates	Keep nodes with no edges? Default FALSE.
keep_format	Keep input format? Default FALSE.
directed	Auto-detect if NULL.

Value

A `cograph_network` with edges involving the specified nodes.

See Also

[select_edges](#), [select_edges_between](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
```

```
# Edges involving A
select_edges_involving(adj, nodes = "A")
```

```
# Edges involving A or B
select_edges_involving(adj, nodes = c("A", "B"))
```

select_neighbors	<i>Select Node Neighbors (Ego Network)</i>
------------------	--

Description

Select nodes within a specified distance from focal nodes.

Usage

```
select_neighbors(
  x,
  of,
  order = 1L,
  ...,
  .keep_edges = c("internal", "none"),
  keep_format = FALSE,
  directed = NULL
)
```

Arguments

x	Network input.
of	Character or integer. Focal node(s) by name or index.
order	Integer. Neighborhood order (1 = direct neighbors). Default 1.
...	Additional filter expressions to apply after neighborhood selection.
.keep_edges	How to handle edges. Default "internal".
keep_format	Logical. Keep input format? Default FALSE.
directed	Logical or NULL. Auto-detect if NULL.

Value

A `graph_network` with nodes in the neighborhood.

See Also

[select_nodes](#), [select_component](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")

# Direct neighbors of A
select_neighbors(adj, of = "A")

# Neighbors up to 2 hops
select_neighbors(adj, of = "A", order = 2)
```

select_nodes

Select Nodes with Lazy Centrality Computation

Description

A more nuanced node selection function that improves upon `filter_nodes()` with lazy centrality computation (only computes measures actually referenced), multiple selection modes, and global context variables for structural awareness.

Usage

```
select_nodes(
  x,
  ...,
  name = NULL,
  index = NULL,
  top = NULL,
  by = "degree",
  neighbors_of = NULL,
  order = 1L,
  component = NULL,
  .keep_edges = c("internal", "none"),
  keep_format = FALSE,
  directed = NULL
)
```

Arguments

`x` Network input: `cograph_network`, `matrix`, `igraph`, `network`, or `tna` object.

...	Filter expressions using node columns, centrality measures, or global context variables. Centrality measures are computed lazily (only those actually referenced). Available variables: Node columns All columns in the nodes dataframe: id, label, name, x, y, inits, color, plus any custom Centrality measures degree, indegree, outdegree, strength, instrength, outstrength, betweenness, closeness, eigenvector, pagerank, hub, authority, coreness Global context component, component_size, is_largest_component, neighborhood_size, k_core, is_articulation, is_bridge_endpoint
name	Character vector. Select nodes by name/label.
index	Integer vector. Select nodes by index (1-based).
top	Integer. Select top N nodes by centrality measure.
by	Character. Centrality measure for top selection. Default "degree".
neighbors_of	Character or integer. Select neighbors of these nodes (by name or index).
order	Integer. Neighborhood order (1 = direct neighbors, 2 = neighbors of neighbors, etc.). Default 1.
component	Selection mode for connected components: "largest" Select nodes in the largest connected component Integer Select nodes in component with this ID Character Select component containing node with this name
.keep_edges	How to handle edges. One of: "internal" (default) Keep only edges between remaining nodes "none" Remove all edges
keep_format	Logical. If TRUE, return the same format as input. Default FALSE returns <code>cograph_network</code> .
directed	Logical or NULL. If NULL (default), auto-detect.

Details

Selection modes are combined with AND logic (like tidygraph/dplyr):

- `select_nodes(x, top = 10, component = "largest")` selects top 10 nodes **within** the largest component
- All criteria must be satisfied for a node to be selected

Centrality measures are computed lazily - only measures actually referenced in expressions or the `by` parameter are computed. This makes `select_nodes()` faster than `filter_nodes()` for large networks.

For networks with negative edge weights, `betweenness` and `closeness` will return NA with a warning (igraph cannot compute these with negative weights).

Value

A `cograph_network` object with selected nodes. If `keep_format = TRUE`, returns the same type as input.

See Also

[filter_nodes](#), [select_neighbors](#), [select_component](#), [select_top](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")

# Lazy - only computes degree
select_nodes(adj, degree >= 3)

# Global context - computes component info
select_nodes(adj, is_largest_component & degree >= 2)

# By name
select_nodes(adj, name = c("A", "B", "C"))

# Top 2 by PageRank
select_nodes(adj, top = 2, by = "pagerank")

# Neighborhood of "A" up to 2 hops
select_nodes(adj, neighbors_of = "A", order = 2)

# Largest connected component
select_nodes(adj, component = "largest")

# Combined: top 2 in largest component
select_nodes(adj, component = "largest", top = 2, by = "degree")

# Articulation points with high degree
# select_nodes(adj, is_articulation & degree >= 2)
```

select_top

Select Top N Nodes by Centrality

Description

Select the top N nodes ranked by a centrality measure.

Usage

```
select_top(
  x,
  n,
  by = "degree",
  ...,
```

```

    .keep_edges = c("internal", "none"),
    keep_format = FALSE,
    directed = NULL
  )

```

Arguments

x	Network input.
n	Integer. Number of top nodes to select.
by	Character. Centrality measure for ranking. One of: "degree", "indegree", "outdegree", "strength", "instrength", "outstrength", "betweenness", "closeness", "eigenvector", "pagerank", "hub", "authority", "coreness". Default "degree".
...	Additional filter expressions to apply.
.keep_edges	How to handle edges. Default "internal".
keep_format	Logical. Keep input format? Default FALSE.
directed	Logical or NULL. Auto-detect if NULL.

Value

A `cograph_network` with the top N nodes.

See Also

[select_nodes](#), [select_component](#)

Examples

```

adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")

# Top 2 by degree
select_top(adj, n = 2)

# Top 2 by PageRank
select_top(adj, n = 2, by = "pagerank")

```

select_top_edges	<i>Select Top N Edges</i>
------------------	---------------------------

Description

Select the top N edges ranked by weight or another metric.

Usage

```
select_top_edges(
  x,
  n,
  by = "weight",
  ...,
  .keep_isolates = FALSE,
  keep_format = FALSE,
  directed = NULL
)
```

Arguments

x	Network input.
n	Integer. Number of top edges to select.
by	Character. Metric for ranking. One of: "weight", "abs_weight", "edge_betweenness". Default "weight".
...	Additional filter expressions.
.keep_isolates	Keep nodes with no edges? Default FALSE.
keep_format	Keep input format? Default FALSE.
directed	Auto-detect if NULL.

Value

A `cograph_network` with the top N edges.

See Also

[select_edges](#), [select_top](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
```

```
# Top 3 edges by weight
select_top_edges(adj, n = 3)

# Top 2 by edge betweenness
select_top_edges(adj, n = 2, by = "edge_betweenness")
```

set_edges	<i>Set Edges in Cograph Network</i>
-----------	-------------------------------------

Description

Replaces the edges in a `cograph_network` object. Expects a data frame with `from`, `to`, and optionally `weight` columns.

Usage

```
set_edges(x, edges_df)
```

Arguments

<code>x</code>	A <code>cograph_network</code> object.
<code>edges_df</code>	A data frame with columns: <code>from</code> , <code>to</code> , and optionally <code>weight</code> .

Value

The modified `cograph_network` object.

See Also

[as_cograph](#), [get_edges](#), [set_nodes](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
new_edges <- data.frame(from = c(1, 2), to = c(2, 3), weight = c(0.5, 0.8))
net <- set_edges(net, new_edges)
get_edges(net)
```

 set_groups

Set Node Groups

Description

Assigns node groupings to a `cograph_network` object. Groups are stored as metadata with a type column ("layer", "cluster", or "group") for use by specialized plot functions.

Usage

```
set_groups(
  x,
  groups = NULL,
  type = c("group", "cluster", "layer"),
  nodes = NULL,
  layers = NULL,
  clusters = NULL
)
```

Arguments

<code>x</code>	A <code>cograph_network</code> object.
<code>groups</code>	Node groupings in one of these formats: <ul style="list-style-type: none"> • Character string: Community detection method ("louvain", "walktrap", "fast_greedy", "label_prop", "infomap", "leiden") • Named list: Group name -> node vector mapping (e.g., <code>list(A = c("N1", "N2"), B = c("N3", "N4"))</code>) • Unnamed vector: Group assignment per node (same order as nodes) • Data frame: Must have "node"/"nodes" column plus one of "layer"/"layers", "cluster"/"clusters", or "group"/"groups" (plural forms are automatically normalized to singular) • NULL: Use nodes + one of layers/clusters/groups vectors
<code>type</code>	Group type. One of "group" (default), "cluster", or "layer". Ignored when using vector arguments (layers, clusters, groups) since the type is inferred from which argument is provided.
<code>nodes</code>	Character vector of node labels. Use with layers, clusters, or groups to specify groupings via vectors instead of a data frame.
<code>layers</code>	Character/factor vector of layer assignments (same length as nodes).
<code>clusters</code>	Character/factor vector of cluster assignments (same length as nodes).

Value

The modified `cograph_network` object with `node_groups` set.

See Also

[get_groups](#), [splot](#), [detect_communities](#)

Examples

```
# Create network (symmetric for community detection)
mat <- matrix(runif(100), 10, 10)
mat <- (mat + t(mat)) / 2 # Make symmetric (undirected)
diag(mat) <- 0
rownames(mat) <- colnames(mat) <- paste0("N", 1:10)
net <- as_cograph(mat)

# Using vectors (recommended)
net <- set_groups(net,
  nodes = paste0("N", 1:10),
  layers = c(rep("Macro", 3), rep("Meso", 4), rep("Micro", 3))
)

# Named list -> layers
net <- set_groups(net, list(
  Macro = paste0("N", 1:3),
  Meso = paste0("N", 4:7),
  Micro = paste0("N", 8:10)
), type = "layer")

# Vector -> clusters
net <- set_groups(net, c("A", "A", "A", "B", "B", "B", "C", "C", "C", "C"),
  type = "cluster")

# Community detection -> groups
net <- set_groups(net, "louvain", type = "group")

# Data frame with explicit columns
df <- data.frame(nodes = paste0("N", 1:10),
  layers = rep(c("Top", "Bottom"), each = 5))
net <- set_groups(net, df)

# Check groups
get_groups(net)
```

 set_layout

Set Layout in Cograph Network

Description

Sets the layout coordinates in a `cograph_network` object. Updates the `x` and `y` columns in the nodes data frame.

Usage

```
set_layout(x, layout_df)
```

Arguments

x A `cograph_network` object.
layout_df A data frame with x and y columns, or a matrix with 2 columns.

Value

The modified `cograph_network` object.

See Also

[as_cograph](#), [get_nodes](#), [sn_layout](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
layout <- data.frame(x = c(0, 1, 0.5), y = c(0, 0, 1))
net <- set_layout(net, layout)
get_nodes(net)
```

set_nodes

Set Nodes in Cograph Network

Description

Replaces the nodes data frame in a `cograph_network` object.

Usage

```
set_nodes(x, nodes_df)
```

Arguments

x A `cograph_network` object.
nodes_df A data frame with node information (id, label columns expected).

Value

The modified `cograph_network` object.

See Also

[as_cograph](#), [get_nodes](#), [set_edges](#)

Examples

```
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- as_cograph(mat)
new_nodes <- data.frame(id = 1:3, label = c("A", "B", "C"))
net <- set_nodes(net, new_nodes)
get_labels(net)
```

shortest_paths *Compute Shortest Path Distances*

Description

Computes shortest path distances between nodes in a network. Supports all-pairs, single-source, and point-to-point queries.

Usage

```
shortest_paths(x, from = NULL, to = NULL, weights = NULL, directed = NULL, ...)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object
from	Character or numeric node identifier(s) for the source. If NULL (default), compute distances from all nodes.
to	Character or numeric node identifier(s) for the target. If NULL (default), compute distances to all nodes.
weights	Edge weight handling: NULL (default) auto-detects from edge attributes, NA forces unweighted distances, or a numeric vector of custom weights.
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.
...	Additional arguments passed to to_igraph

Details

Uses `igraph::distances()` internally. For weighted networks, edge weights are used as distances by default. Pass `weights = NA` to ignore weights and treat all edges as having unit distance.

Note: `igraph::distances()` with `weights = NULL` automatically uses edge weight attributes if present. To force unweighted computation, pass `weights = NA` explicitly.

Value

Depends on the query:

- If both `from` and `to` are NULL: a full distance matrix (all pairs)
- If `from` is a single node and `to` is NULL: a named numeric vector of distances from that node to all others

- If from is multiple nodes and to is NULL: a matrix with rows for each source
- If both from and to are single nodes: a single numeric value
- Otherwise: a matrix of distances between the specified node sets

See Also

[k_shortest_paths](#), [network_summary](#)

Examples

```
# All-pairs distances
adj <- matrix(c(
  0, 1, 0, 0,
  1, 0, 1, 0,
  0, 1, 0, 1,
  0, 0, 1, 0
), 4, 4)
rownames(adj) <- colnames(adj) <- LETTERS[1:4]
cograph::shortest_paths(adj)

# Single source to all
cograph::shortest_paths(adj, from = "A")

# Point-to-point
cograph::shortest_paths(adj, from = "A", to = "D")
```

simmelian_strength *Simmelian Strength (Triangle Count per Edge)*

Description

Convenience wrapper around [edge centrality](#) that returns only the triangle count per edge, sorted descending.

Usage

```
simmelian_strength(x, top = NULL, directed = NULL, digits = NULL, ...)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object.
top	Integer or NULL. Return only the top N edges. Default NULL.
directed	Logical or NULL. Default NULL (auto-detect).
digits	Integer or NULL. Round numeric columns. Default NULL.
...	Additional arguments passed to edge centrality .

Value

A data frame sorted by triangles (descending) with columns: from, to, weight (if weighted), triangles.

See Also

[edge centrality](#), [neighborhood overlap](#)

Examples

```
k4 <- matrix(1, 4, 4); diag(k4) <- 0
rownames(k4) <- colnames(k4) <- c("A", "B", "C", "D")
cograph::simmelian_strength(k4)
```

simplify

Simplify a Network

Description

Removes self-loops and merges duplicate (multi-)edges, similar to `igraph::simplify()`. Works on matrices, `cograph_network`, `igraph`, and `tna` objects.

Usage

```
simplify(x, remove_loops, remove_multiple, edge_attr_comb, ...)
```

```
## S3 method for class 'matrix'
simplify(
  x,
  remove_loops = TRUE,
  remove_multiple = TRUE,
  edge_attr_comb = "mean",
  ...
)
```

```
## S3 method for class 'cograph_network'
simplify(
  x,
  remove_loops = TRUE,
  remove_multiple = TRUE,
  edge_attr_comb = "mean",
  ...
)
```

```
## S3 method for class 'igraph'
simplify(
  x,
```

```

    remove_loops = TRUE,
    remove_multiple = TRUE,
    edge_attr_comb = "mean",
    ...
)

## S3 method for class 'tna'
simplify(
  x,
  remove_loops = TRUE,
  remove_multiple = TRUE,
  edge_attr_comb = "mean",
  ...
)

## Default S3 method:
simplify(
  x,
  remove_loops = TRUE,
  remove_multiple = TRUE,
  edge_attr_comb = "mean",
  ...
)

```

Arguments

<code>x</code>	Network input (matrix, <code>cograph_network</code> , <code>igraph</code> , <code>tna</code> object).
<code>remove_loops</code>	Logical. Remove self-loops (diagonal entries)?
<code>remove_multiple</code>	Logical. Merge duplicate edges?
<code>edge_attr_comb</code>	How to combine weights of duplicate edges: "sum", "mean", "max", "min", "first", or a custom function.
<code>...</code>	Additional arguments (currently unused).

Value

The simplified network in the same format as the input.

See Also

[filter_edges](#) for conditional edge removal, [centrality](#) which has its own `simplify` parameter

Examples

```

# Matrix with self-loops
mat <- matrix(c(0.5, 0.3, 0, 0.3, 0.2, 0.4, 0, 0.4, 0.1), 3, 3)
rownames(mat) <- colnames(mat) <- c("A", "B", "C")
simplify(mat)

```

```
# Edge list with duplicates
edges <- data.frame(from = c(1, 1, 2), to = c(2, 2, 3), weight = c(0.3, 0.7, 0.5))
net <- cograph(edges, layout = NULL)
simplify(net)
simplify(net, edge_attr_comb = "sum")
```

sn_edges

Set Edge Aesthetics

Description

Customize the visual appearance of edges in a network plot.

Usage

```
sn_edges(
  network,
  width = NULL,
  edge_size = NULL,
  esize = NULL,
  edge_width_range = NULL,
  edge_scale_mode = NULL,
  edge_cutoff = NULL,
  cut = NULL,
  color = NULL,
  edge_positive_color = NULL,
  positive_color = NULL,
  edge_negative_color = NULL,
  negative_color = NULL,
  alpha = NULL,
  style = NULL,
  curvature = NULL,
  arrow_size = NULL,
  show_arrows = NULL,
  maximum = NULL,
  width_scale = NULL,
  labels = NULL,
  label_size = NULL,
  label_color = NULL,
  label_position = NULL,
  label_offset = NULL,
  label_bg = NULL,
  label_bg_padding = NULL,
  label_fontface = NULL,
  label_border = NULL,
  label_border_color = NULL,
  label_underline = NULL,
```

```

label_shadow = NULL,
label_shadow_color = NULL,
label_shadow_offset = NULL,
label_shadow_alpha = NULL,
bidirectional = NULL,
loop_rotation = NULL,
curve_shape = NULL,
curve_pivot = NULL,
curves = NULL,
ci = NULL,
ci_scale = NULL,
ci_alpha = NULL,
ci_color = NULL,
ci_style = NULL,
ci_arrows = NULL,
ci_lower = NULL,
ci_upper = NULL,
label_style = NULL,
label_template = NULL,
label_digits = NULL,
label_ci_format = NULL,
label_p = NULL,
label_p_digits = NULL,
label_p_prefix = NULL,
label_stars = NULL
)

```

Arguments

network	A <code>cograph_network</code> object, matrix, data.frame, or igraph object. Matrices and other inputs are auto-converted.
width	Edge width. Can be a single value, vector (per-edge), or "weight".
edge_size	Base edge size for weight scaling. NULL (default) uses adaptive sizing based on network size: $15 * \exp(-n_nodes/90) + 1$. Larger values = thicker edges.
esize	Deprecated. Use <code>edge_size</code> instead.
edge_width_range	Output width range as <code>c(min, max)</code> for weight-based scaling. Default <code>c(0.5, 4)</code> . Edges are scaled to fit within this range.
edge_scale_mode	Scaling mode for edge weights: "linear" (default), "log" (for wide weight ranges), "sqrt" (moderate compression), or "rank" (equal visual spacing).
edge_cutoff	Two-tier cutoff for edge width scaling. NULL (default) = auto 75th percentile. 0 = disabled. Positive number = manual threshold.
cut	Deprecated. Use <code>edge_cutoff</code> instead.
color	Edge color. Can be a single color, vector, or "weight" for automatic coloring based on edge weights.

edge_positive_color	Color for positive edge weights.
positive_color	Deprecated. Use edge_positive_color instead.
edge_negative_color	Color for negative edge weights.
negative_color	Deprecated. Use edge_negative_color instead.
alpha	Edge transparency (0-1).
style	Line style: "solid", "dashed", "dotted", "longdash", "twodash".
curvature	Edge curvature amount (0 = straight).
arrow_size	Size of arrow heads for directed networks.
show_arrows	Logical. Show arrows? Default TRUE for directed networks.
maximum	Maximum edge weight for scaling width. Weights above this are capped. Similar to qgraph's maximum parameter.
width_scale	Scale factor for edge widths. Values > 1 make edges thicker, values < 1 make them thinner. Applied after all other width calculations.
labels	Edge labels. Can be TRUE (show weights), a vector, or column name.
label_size	Edge label text size.
label_color	Edge label text color.
label_position	Position along edge (0 = source, 0.5 = middle, 1 = target).
label_offset	Perpendicular offset from edge line.
label_bg	Background color for edge labels (default "white"). Set to NA for transparent.
label_bg_padding	Padding around label text as proportion of text size (default 0.3).
label_fontface	Font face: "plain", "bold", "italic", "bold.italic" (default "plain").
label_border	Border style: NULL (none), "rect", "rounded", "circle" (default NULL).
label_border_color	Border color for label border (default "gray50").
label_underline	Logical. Underline the label text? (default FALSE).
label_shadow	Logical. Enable drop shadow for labels? (default FALSE).
label_shadow_color	Color for label shadow (default "gray40").
label_shadow_offset	Offset distance for shadow in points (default 0.5).
label_shadow_alpha	Transparency for shadow (0-1, default 0.5).
bidirectional	Logical. Show arrows at both ends of edges?
loop_rotation	Angle in radians for self-loop direction (default: pi/2 = top).
curve_shape	Spline tension for curved edges (-1 to 1, default: 0).
curve_pivot	Pivot position along edge for curve control point (0-1, default: 0.5).

curves	Curve mode: FALSE (straight edges), "mutual" (only curve reciprocal pairs), or "force" (curve all edges). Default FALSE.
ci	Numeric vector of CI widths (0-1 scale). Larger values = more uncertainty.
ci_scale	Width multiplier for CI underlay thickness. Default 2.
ci_alpha	Transparency for CI underlay (0-1). Default 0.15.
ci_color	CI underlay color. NA (default) uses main edge color.
ci_style	Line type for CI underlay: 1=solid, 2=dashed, 3=dotted. Default 2.
ci_arrows	Logical: show arrows on CI underlay? Default FALSE.
ci_lower	Numeric vector of lower CI bounds for labels.
ci_upper	Numeric vector of upper CI bounds for labels.
label_style	Preset style: "none", "estimate", "full", "range", "stars".
label_template	Template with placeholders: {est}, {range}, {low}, {up}, {p}, {stars}.
label_digits	Decimal places for estimates in template. Default 2.
label_ci_format	CI format: "bracket" for [low, up] or "dash" for low-up.
label_p	Numeric vector of p-values for edges.
label_p_digits	Decimal places for p-values. Default 3.
label_p_prefix	Prefix for p-values. Default "p=".
label_stars	Stars for labels: character vector, TRUE (compute from p), or numeric (treated as p-values).

Details

Vectorization:

Most aesthetic parameters can be specified as:

- **Single value:** Applied to all edges
- **Vector:** Per-edge values (must match edge count)
- **"weight":** Special value for width and color that auto-maps from edge weights

Weight-Based Styling:

When color = "weight", edges are colored by sign:

- Positive weights use edge_positive_color (default: green)
- Negative weights use edge_negative_color (default: red)

When width = "weight", edge widths scale with absolute weight values, respecting the maximum parameter if set.

Edge Label Templates:

For statistical output (e.g., regression coefficients with CIs), use templates:

- label_template = "{est}": Show estimate only
- label_template = "{est} [{low}, {up}]": Estimate with CI
- label_template = "{est}\\{stars}": Estimate with significance

Preset styles via `label_style`:

- "estimate": Weight/estimate only
- "full": Estimate + CI in brackets
- "range": CI range only
- "stars": Significance stars

CI Underlays:

Visualize uncertainty by drawing a wider, semi-transparent edge behind:

- `ci`: Vector of CI widths (0-1 scale)
- `ci_scale`: Width multiplier (default 2)
- `ci_alpha`: Transparency (default 0.15)

Value

Modified `cograph_network` object that can be piped to further customization functions or plotting functions.

See Also

[sn_nodes](#) for node customization, [cograph](#) for network creation, [splot](#) and [soplot](#) for plotting, [sn_layout](#) for layout algorithms, [sn_theme](#) for visual themes

Examples

```
adj <- matrix(c(0, 1, -0.5, 1, 0, 1, -0.5, 1, 0), nrow = 3)

# Basic: auto-style by weight
cograph(adj) |>
  sn_edges(width = "weight", color = "weight")

# Direct matrix input (auto-converted)
adj |> sn_edges(width = 2, color = "gray50")

# Custom positive/negative colors
cograph(adj) |>
  sn_edges(
    color = "weight",
    edge_positive_color = "darkblue",
    edge_negative_color = "darkred"
  ) |>
  splot()

# Edge labels showing weights
cograph(adj) |>
  sn_edges(labels = TRUE, label_size = 0.8) |>
  splot()

# Statistical output with CI template
# Suppose we have estimates, lower/upper CI bounds
estimates <- c(0.5, -0.3, 0.8)
```

```

ci_lo <- c(0.2, -0.6, 0.5)
ci_hi <- c(0.8, -0.1, 1.1)

cograph(adj) |>
  sn_edges(
    label_template = "{est} [{low}, {up}]",
    ci_lower = ci_lo,
    ci_upper = ci_hi,
    label_digits = 2
  ) |>
  plot()

# Curved edges for reciprocal pairs
cograph(adj) |>
  sn_edges(curves = "mutual", curvature = 0.3) |>
  plot()

```

sn_ggplot

Convert Network to ggplot2

Description

Convert a Cograph network visualization to a ggplot2 object for further customization and composability.

Usage

```
sn_ggplot(network, title = NULL)
```

Arguments

network	A cograph_network object, matrix, data.frame, or igraph object. Matrices and other inputs are auto-converted.
title	Optional plot title.

Value

A ggplot2 object.

Examples

```

adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
# With cograph()
p <- cograph(adj) |> sn_ggplot()
print(p)

# Direct matrix input
p <- adj |> sn_ggplot()

# Further customization
p + ggplot2::labs(title = "My Network")

```

sn_layout	<i>Apply Layout to Network</i>
-----------	--------------------------------

Description

Apply a layout algorithm to compute node positions.

Usage

```
sn_layout(network, layout, seed = 42, ...)
```

Arguments

network	A cograph_network object, matrix, data.frame, or igraph object. Matrices and other inputs are auto-converted.
layout	Layout algorithm name or a CographLayout object.
seed	Random seed for deterministic layouts. Default 42. Set NULL for random.
...	Additional arguments passed to the layout function.

Details

Built-in Layouts:

spring Force-directed layout (Fruchterman-Reingold style). Good general-purpose layout. Default.

circle Nodes arranged in a circle. Good for small networks or when structure is less important.

groups Circular layout with grouped nodes clustered together.

grid Nodes in a regular grid.

random Random positions. Useful as starting point.

star Central node with others arranged around it.

bipartite Two-column layout for bipartite networks.

igraph Layouts:

Two-letter codes for igraph layouts: "kk" (Kamada-Kawai), "fr" (Fruchterman-Reingold), "drl", "mds", "ni" (nicely), "tr" (tree), "ci" (circle), etc.

You can also pass igraph layout functions directly or use full names like "layout_with_kk".

Value

Modified cograph_network object.

See Also

[cograph](#) for network creation, [sn_nodes](#) for node customization, [sn_edges](#) for edge customization, [sn_theme](#) for visual themes, [splot](#) and [soplot](#) for plotting

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)

# Built-in layouts
cograph(adj) |> sn_layout("circle") |> splot()
cograph(adj) |> sn_layout("spring") |> splot()

# igraph layouts (if igraph installed)
if (requireNamespace("igraph", quietly = TRUE)) {
  cograph(adj) |> sn_layout("kk") |> splot()
  cograph(adj) |> sn_layout("fr") |> splot()
}

# Custom coordinates
coords <- matrix(c(0, 0, 1, 0, 0.5, 1), ncol = 2, byrow = TRUE)
cograph(adj) |> sn_layout(coords) |> splot()

# Direct matrix input (auto-converts)
adj |> sn_layout("circle")
```

sn_nodes

Set Node Aesthetics

Description

Customize the visual appearance of nodes in a network plot.

Usage

```
sn_nodes(
  network,
  size = NULL,
  shape = NULL,
  node_svg = NULL,
  svg_preserve_aspect = NULL,
  fill = NULL,
  border_color = NULL,
  border_width = NULL,
  alpha = NULL,
  label_size = NULL,
  label_color = NULL,
  label_position = NULL,
  show_labels = NULL,
  pie_values = NULL,
  pie_colors = NULL,
  pie_border_width = NULL,
  donut_fill = NULL,
  donut_values = NULL,
```

```

    donut_color = NULL,
    donut_colors = NULL,
    donut_border_width = NULL,
    donut_inner_ratio = NULL,
    donut_bg_color = NULL,
    donut_shape = NULL,
    donut_show_value = NULL,
    donut_value_size = NULL,
    donut_value_color = NULL,
    donut_value_fontface = NULL,
    donut_value_fontfamily = NULL,
    donut_value_digits = NULL,
    donut_value_prefix = NULL,
    donut_value_suffix = NULL,
    donut_value_format = NULL,
    donut2_values = NULL,
    donut2_colors = NULL,
    donut2_inner_ratio = NULL,
    label_fontface = NULL,
    label_fontfamily = NULL,
    label_hjust = NULL,
    label_vjust = NULL,
    label_angle = NULL,
    node_names = NULL
)

```

Arguments

network	A cograph_network object, matrix, data.frame, or igraph object. Matrices and other inputs are auto-converted.
size	Node size. Can be a single value, vector (per-node), or column name.
shape	Node shape. Options: "circle", "square", "triangle", "diamond", "pentagon", "hexagon", "ellipse", "heart", "star", "pie", "donut", "cross", "rectangle", or any custom SVG shape registered with register_svg_shape().
node_svg	Custom SVG for node shape: path to SVG file OR inline SVG string. Overrides shape parameter when provided.
svg_preserve_aspect	Logical: maintain SVG aspect ratio? Default TRUE.
fill	Node fill color. Can be a single color, vector, or column name.
border_color	Node border color.
border_width	Node border width.
alpha	Node transparency (0-1).
label_size	Label text size.
label_color	Label text color.
label_position	Label position: "center", "above", "below", "left", "right".

show_labels	Logical. Show node labels? Default TRUE.
pie_values	For pie shape: list or matrix of values for pie segments. Each element corresponds to a node and contains values for its segments.
pie_colors	For pie shape: colors for pie segments.
pie_border_width	Border width for pie chart nodes.
donut_fill	For donut shape: numeric value (0-1) specifying fill proportion. 0.1 = 10% filled, 0.5 = 50% filled, 1.0 = fully filled ring. Can be a single value (all nodes) or vector (per-node values).
donut_values	Deprecated. Use donut_fill for simple fill proportion. Still works for backwards compatibility.
donut_color	For donut shape: fill color(s) for the donut ring. Single color sets fill for all nodes. Two colors set fill and background for all nodes. More than 2 colors set per-node fill colors (recycled to n_nodes). Default: "lightgray" fill, "gray90" background when shape="donut".
donut_colors	Deprecated. Use donut_color instead.
donut_border_width	Border width for donut chart nodes.
donut_inner_ratio	For donut shape: inner radius ratio (0-1). Default 0.5.
donut_bg_color	For donut shape: background color for unfilled portion.
donut_shape	For donut: base shape for ring ("circle", "square", "hexagon", "triangle", "diamond", "pentagon"). Default "circle".
donut_show_value	For donut shape: show value in center? Default FALSE.
donut_value_size	For donut shape: font size for center value.
donut_value_color	For donut shape: color for center value text.
donut_value_fontface	For donut shape: font face for center value ("plain", "bold", "italic", "bold.italic"). Default "bold".
donut_value_fontfamily	For donut shape: font family for center value ("sans", "serif", "mono"). Default "sans".
donut_value_digits	For donut shape: decimal places for value display. Default 2.
donut_value_prefix	For donut shape: text before value (e.g., "\$"). Default "".
donut_value_suffix	For donut shape: text after value (e.g., "%"). Default "".
donut_value_format	For donut shape: custom format function (overrides digits).
donut2_values	For double donut: list of values for inner donut ring.

donut2_colors	For double donut: colors for inner donut ring segments.
donut2_inner_ratio	For double donut: inner radius ratio for inner donut ring. Default 0.4.
label_fontface	Font face for node labels: "plain", "bold", "italic", "bold.italic". Default "plain".
label_fontfamily	Font family for node labels: "sans", "serif", "mono", or system font. Default "sans".
label_hjust	Horizontal justification for node labels (0=left, 0.5=center, 1=right). Default 0.5.
label_vjust	Vertical justification for node labels (0=bottom, 0.5=center, 1=top). Default 0.5.
label_angle	Text rotation angle in degrees for node labels. Default 0.
node_names	Alternative names for legend (separate from display labels).

Details

Vectorization:

All aesthetic parameters can be specified as:

- **Single value:** Applied to all nodes (e.g., fill = "blue")
- **Vector:** Per-node values, recycled if shorter than node count
- **Column name:** String referencing a column in the node data frame

Parameters are validated for correct length; providing a vector with length other than 1 or n_nodes will produce a warning about recycling.

Donut Charts:

Donut charts are ideal for showing a single proportion (0-1) per node:

- Set donut_fill to a numeric value or vector (0 = empty, 1 = full)
- Use donut_color to set fill color(s)
- Use donut_shape for non-circular donuts ("square", "hexagon", etc.)
- Enable donut_show_value = TRUE to display the value in the center

Value

Modified cograph_network object that can be piped to further customization functions or plotting functions.

See Also

[sn_edges](#) for edge customization, [cograph](#) for network creation, [splot](#) and [soplot](#) for plotting, [sn_layout](#) for layout algorithms, [sn_theme](#) for visual themes

Examples

```

adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)

# Basic usage with cograph()
cograph(adj) |>
  sn_nodes(size = 0.08, fill = "steelblue", shape = "circle")

# Direct matrix input (auto-converted)
adj |> sn_nodes(fill = "coral", size = 0.1)

# Per-node customization with vectors
cograph(adj) |>
  sn_nodes(
    size = c(0.08, 0.06, 0.1),
    fill = c("red", "blue", "green"),
    label_position = c("above", "below", "center")
  ) |>
  splot()

# Donut chart nodes showing proportions
cograph(adj) |>
  sn_nodes(
    donut_fill = c(0.25, 0.75, 0.5),
    donut_color = "steelblue",
    donut_show_value = TRUE,
    donut_value_suffix = "%"
  ) |>
  splot()

# Mixed shapes per node
cograph(adj) |>
  sn_nodes(
    shape = c("circle", "square", "triangle"),
    fill = c("#E41A1C", "#377EB8", "#4DAF4A")
  ) |>
  splot()

```

 sn_palette

Apply Color Palette to Network

Description

Apply a color palette for node and/or edge coloring.

Usage

```
sn_palette(network, palette, target = "nodes", by = NULL)
```

Arguments

network	A <code>cograph_network</code> object, matrix, data.frame, or igraph object. Matrices and other inputs are auto-converted.
palette	Palette name or function.
target	What to apply the palette to: "nodes", "edges", or "both".
by	Variable to map colors to (for nodes: column name or "group").

Details

Available Palettes:

Use `list_palettes()` to see all available palettes. Common options:

viridis Perceptually uniform, colorblind-friendly.

colorblind Optimized for color vision deficiency.

pastel Soft, muted colors.

bright Saturated, vivid colors.

grayscale Shades of gray.

You can also pass a custom palette function that takes `n` and returns `n` colors.

Value

Modified `cograph_network` object.

See Also

[cograph](#) for network creation, [sn_theme](#) for visual themes, [sn_nodes](#) for node customization, [list_palettes](#) to see available palettes, [splot](#) and [soplot](#) for plotting

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)

# Apply palette to nodes
cograph(adj) |> sn_palette("viridis") |> splot()

# Apply to edges
cograph(adj) |> sn_palette("colorblind", target = "edges") |> splot()

# Apply to both
cograph(adj) |> sn_palette("pastel", target = "both") |> splot()

# Custom palette function
my_pal <- function(n) rainbow(n, s = 0.7)
cograph(adj) |> sn_palette(my_pal) |> splot()

# Direct matrix input
adj |> sn_palette("viridis")
```

sn_save	<i>Save Network Visualization</i>
---------	-----------------------------------

Description

Save a Cograph network visualization to a file.

Usage

```
sn_save(network, filename, width = 7, height = 7, dpi = 300, title = NULL, ...)
```

Arguments

network	A cograph_network object, matrix, data.frame, or igraph object. Matrices and other inputs are auto-converted.
filename	Output filename. Format is detected from extension.
width	Width in inches (default 7).
height	Height in inches (default 7).
dpi	Resolution for raster formats (default 300).
title	Optional plot title.
...	Additional arguments passed to the graphics device.

Value

The output filename, invisibly.

Examples

```
## Not run:  
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)  
net <- cograph(adj)  
sn_save(net, file.path(tempdir(), "network.pdf"))  
  
## End(Not run)
```

sn_save_ggplot	<i>Save as ggplot2</i>
----------------	------------------------

Description

Save network as a ggplot2 object to file using ggsave.

Usage

```
sn_save_ggplot(  
  network,  
  filename,  
  width = 7,  
  height = 7,  
  dpi = 300,  
  title = NULL,  
  ...  
)
```

Arguments

network	A cograph_network object.
filename	Output filename.
width	Width in inches.
height	Height in inches.
dpi	Resolution for raster formats.
title	Optional plot title.
...	Additional arguments passed to ggsave.

Value

The output filename, invisibly.

Examples

```
## Not run:  
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)  
net <- cograph(adj)  
sn_save_ggplot(net, file.path(tempdir(), "network.pdf"))  
  
## End(Not run)
```

sn_theme	<i>Apply Theme to Network</i>
----------	-------------------------------

Description

Apply a visual theme to the network.

Usage

```
sn_theme(network, theme, ...)
```

Arguments

network	A <code>cograph_network</code> object, matrix, data.frame, or igraph object. Matrices and other inputs are auto-converted.
theme	Theme name (string) or <code>CographTheme</code> object.
...	Additional theme parameters to override.

Details

Available Themes:

classic Default theme with white background, blue nodes, gray edges.

dark Dark background with light nodes. Good for presentations.

minimal Subtle styling with thin edges and muted colors.

colorblind Optimized for color vision deficiency.

grayscale Black and white only.

vibrant Bold, saturated colors.

Use `list_themes()` to see all available themes.

Value

Modified `cograph_network` object.

See Also

[cograph](#) for network creation, [sn_palette](#) for color palettes, [sn_nodes](#) for node customization, [sn_edges](#) for edge customization, [list_themes](#) to see available themes, [splot](#) and [soplot](#) for plotting

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)

# Apply different themes
cograph(adj) |> sn_theme("dark") |> splot()
cograph(adj) |> sn_theme("minimal") |> splot()

# Override specific theme properties
cograph(adj) |> sn_theme("classic", background = "lightgray") |> splot()

# Direct matrix input
adj |> sn_theme("dark")
```

soplot

Plot Cograph Network

Description

Main plotting function for Cograph networks. Renders the network visualization using grid graphics. Accepts all node and edge aesthetic parameters.

Usage

```
soplot(
  network,
  title = NULL,
  title_size = 14,
  margins = c(0.05, 0.05, 0.1, 0.05),
  layout_margin = 0.15,
  newpage = TRUE,
  background = "white",
  layout = NULL,
  theme = NULL,
  seed = 42,
  labels = NULL,
  threshold = NULL,
  maximum = NULL,
  node_size = NULL,
  node_shape = NULL,
  node_fill = NULL,
  node_border_color = NULL,
  node_border_width = NULL,
  node_alpha = NULL,
  label_size = NULL,
  label_color = NULL,
  label_position = NULL,
  show_labels = NULL,
```

```
pie_values = NULL,
pie_colors = NULL,
pie_border_width = NULL,
donut_values = NULL,
donut_border_width = NULL,
donut_inner_ratio = NULL,
donut_bg_color = NULL,
donut_show_value = NULL,
donut_value_size = NULL,
donut_value_color = NULL,
donut_fill = NULL,
donut_color = NULL,
donut_colors = NULL,
donut_shape = "circle",
donut_value_fontface = "bold",
donut_value_fontfamily = "sans",
donut_value_digits = 2,
donut_value_prefix = "",
donut_value_suffix = "",
donut2_values = NULL,
donut2_colors = NULL,
donut2_inner_ratio = 0.4,
edge_width = NULL,
edge_size = NULL,
esize = NULL,
edge_width_range = NULL,
edge_scale_mode = "linear",
edge_cutoff = NULL,
cut = NULL,
edge_width_scale = NULL,
edge_color = NULL,
edge_alpha = NULL,
edge_style = NULL,
curvature = NULL,
arrow_size = NULL,
show_arrows = NULL,
edge_positive_color = NULL,
positive_color = NULL,
edge_negative_color = NULL,
negative_color = NULL,
edge_duplicates = NULL,
edge_labels = NULL,
edge_label_size = NULL,
edge_label_color = NULL,
edge_label_position = NULL,
edge_label_offset = NULL,
edge_label_bg = NULL,
edge_label_fontface = NULL,
```

```
    edge_label_border = NULL,  
    edge_label_border_color = NULL,  
    edge_label_underline = NULL,  
    bidirectional = NULL,  
    loop_rotation = NULL,  
    curve_shape = NULL,  
    curve_pivot = NULL,  
    curves = NULL,  
    node_names = NULL,  
    legend = FALSE,  
    legend_position = "topright",  
    scaling = "default",  
    weight_digits = 2  
  )  
  
sn_render(  
  network,  
  title = NULL,  
  title_size = 14,  
  margins = c(0.05, 0.05, 0.1, 0.05),  
  layout_margin = 0.15,  
  newpage = TRUE,  
  background = "white",  
  layout = NULL,  
  theme = NULL,  
  seed = 42,  
  labels = NULL,  
  threshold = NULL,  
  maximum = NULL,  
  node_size = NULL,  
  node_shape = NULL,  
  node_fill = NULL,  
  node_border_color = NULL,  
  node_border_width = NULL,  
  node_alpha = NULL,  
  label_size = NULL,  
  label_color = NULL,  
  label_position = NULL,  
  show_labels = NULL,  
  pie_values = NULL,  
  pie_colors = NULL,  
  pie_border_width = NULL,  
  donut_values = NULL,  
  donut_border_width = NULL,  
  donut_inner_ratio = NULL,  
  donut_bg_color = NULL,  
  donut_show_value = NULL,  
  donut_value_size = NULL,
```

```
donut_value_color = NULL,  
donut_fill = NULL,  
donut_color = NULL,  
donut_colors = NULL,  
donut_shape = "circle",  
donut_value_fontface = "bold",  
donut_value_fontfamily = "sans",  
donut_value_digits = 2,  
donut_value_prefix = "",  
donut_value_suffix = "",  
donut2_values = NULL,  
donut2_colors = NULL,  
donut2_inner_ratio = 0.4,  
edge_width = NULL,  
edge_size = NULL,  
esize = NULL,  
edge_width_range = NULL,  
edge_scale_mode = "linear",  
edge_cutoff = NULL,  
cut = NULL,  
edge_width_scale = NULL,  
edge_color = NULL,  
edge_alpha = NULL,  
edge_style = NULL,  
curvature = NULL,  
arrow_size = NULL,  
show_arrows = NULL,  
edge_positive_color = NULL,  
positive_color = NULL,  
edge_negative_color = NULL,  
negative_color = NULL,  
edge_duplicates = NULL,  
edge_labels = NULL,  
edge_label_size = NULL,  
edge_label_color = NULL,  
edge_label_position = NULL,  
edge_label_offset = NULL,  
edge_label_bg = NULL,  
edge_label_fontface = NULL,  
edge_label_border = NULL,  
edge_label_border_color = NULL,  
edge_label_underline = NULL,  
bidirectional = NULL,  
loop_rotation = NULL,  
curve_shape = NULL,  
curve_pivot = NULL,  
curves = NULL,  
node_names = NULL,
```

```

    legend = FALSE,
    legend_position = "topright",
    scaling = "default",
    weight_digits = 2
)

```

Arguments

network	A <code>cograph_network</code> object, matrix, data.frame, or igraph object. Matrices and other inputs are auto-converted.
title	Optional plot title.
title_size	Title font size.
margins	Plot margins as <code>c(bottom, left, top, right)</code> .
layout_margin	Margin around the network layout (proportion of viewport). Default 0.15.
newpage	Logical. Start a new graphics page? Default TRUE.
background	Background color for the plot. Default "white".
layout	Layout algorithm. Built-in: "circle", "spring", "groups", "grid", "random", "star", "bipartite". igraph (2-letter): "kk" (Kamada-Kawai), "fr" (Fruchterman-Reingold), "drl", "mds", "ni" (nicely), "tr" (tree), etc. Can also pass a coordinate matrix or igraph layout function directly.
theme	Theme name: "classic", "dark", "minimal", etc.
seed	Random seed for deterministic layouts. Default 42. Set NULL for random.
labels	Node labels. Can be a character vector to set custom labels.
threshold	Minimum absolute edge weight to display. Edges with <code>abs(weight) < threshold</code> are hidden. Similar to <code>qgraph</code> 's <code>threshold</code> .
maximum	Maximum edge weight for width scaling. Weights above this are capped. Similar to <code>qgraph</code> 's <code>maximum</code> parameter.
node_size	Node size.
node_shape	Node shape: "circle", "square", "triangle", "diamond", "ellipse", "heart", "star", "pie", "donut", "cross".
node_fill	Node fill color.
node_border_color	Node border color.
node_border_width	Node border width.
node_alpha	Node transparency (0-1).
label_size	Node label text size.
label_color	Node label text color.
label_position	Label position: "center", "above", "below", "left", "right".
show_labels	Logical. Show node labels?
pie_values	For pie/donut/donut_pie nodes: list or matrix of values for segments. For donut with single value (0-1), shows that proportion filled.

<code>pie_colors</code>	For pie/donut/donut_pie nodes: colors for pie segments.
<code>pie_border_width</code>	Border width for pie chart segments.
<code>donut_values</code>	For donut_pie nodes: vector of values (0-1) for outer ring proportion.
<code>donut_border_width</code>	Border width for donut ring.
<code>donut_inner_ratio</code>	For donut nodes: inner radius ratio (0-1). Default 0.5.
<code>donut_bg_color</code>	For donut nodes: background color for unfilled portion.
<code>donut_show_value</code>	For donut nodes: show value in center? Default FALSE.
<code>donut_value_size</code>	For donut nodes: font size for center value.
<code>donut_value_color</code>	For donut nodes: color for center value text.
<code>donut_fill</code>	Numeric value (0-1) for donut fill proportion. This is the simplified API for creating donut charts. Can be a single value or vector per node.
<code>donut_color</code>	Fill color(s) for the donut ring. Simplified API: single color for fill, or c(fill, background) for both.
<code>donut_colors</code>	Deprecated. Use donut_color instead.
<code>donut_shape</code>	Base shape for donut: "circle", "square", "hexagon", "triangle", "diamond", "pentagon". Default inherits from node_shape.
<code>donut_value_fontface</code>	Font face for donut center value: "plain", "bold", "italic", "bold.italic". Default "bold".
<code>donut_value_fontfamily</code>	Font family for donut center value. Default "sans".
<code>donut_value_digits</code>	Decimal places for donut center value. Default 2.
<code>donut_value_prefix</code>	Text before donut center value (e.g., "\$"). Default "".
<code>donut_value_suffix</code>	Text after donut center value (e.g., "%"). Default "".
<code>donut2_values</code>	List of values for inner donut ring (for double donut).
<code>donut2_colors</code>	List of color vectors for inner donut ring segments.
<code>donut2_inner_ratio</code>	Inner radius ratio for inner donut ring. Default 0.4.
<code>edge_width</code>	Edge width. If NULL, scales by weight using edge_size and edge_width_range.
<code>edge_size</code>	Base edge size for weight scaling. NULL (default) uses adaptive sizing based on network size: $15 * \exp(-n_nodes/90) + 1$. Larger values = thicker edges.
<code>esize</code>	Deprecated. Use edge_size instead.
<code>edge_width_range</code>	Output width range as c(min, max) for weight-based scaling. Default c(0.5, 4). Edges are scaled to fit within this range.

edge_scale_mode	Scaling mode for edge weights: "linear" (default), "log" (for wide weight ranges), "sqrt" (moderate compression), or "rank" (equal visual spacing).
edge_cutoff	Two-tier cutoff for edge width scaling. NULL (default) = auto 75th percentile. 0 = disabled. Positive number = manual threshold.
cut	Deprecated. Use edge_cutoff instead.
edge_width_scale	Scale factor for edge widths. Values > 1 make edges thicker.
edge_color	Edge color.
edge_alpha	Edge transparency (0-1).
edge_style	Line style: "solid", "dashed", "dotted".
curvature	Edge curvature amount.
arrow_size	Size of arrow heads.
show_arrows	Logical. Show arrows?
edge_positive_color	Color for positive edge weights.
positive_color	Deprecated. Use edge_positive_color instead.
edge_negative_color	Color for negative edge weights.
negative_color	Deprecated. Use edge_negative_color instead.
edge_duplicates	How to handle duplicate edges in undirected networks. NULL (default) = stop with error listing duplicates. Options: "sum", "mean", "first", "max", "min", or a custom aggregation function.
edge_labels	Edge labels. Can be TRUE to show weights, or a vector.
edge_label_size	Edge label text size.
edge_label_color	Edge label text color.
edge_label_position	Position along edge (0 = source, 0.5 = middle, 1 = target).
edge_label_offset	Perpendicular offset from edge line.
edge_label_bg	Background color for edge labels (default "white"). Set to NA for transparent.
edge_label_fontface	Font face: "plain", "bold", "italic", "bold.italic".
edge_label_border	Border style: NULL, "rect", "rounded", "circle".
edge_label_border_color	Border color for label border.
edge_label_underline	Logical. Underline the label text?

<code>bidirectional</code>	Logical. Show arrows at both ends of edges?
<code>loop_rotation</code>	Angle in radians for self-loop direction (default: $\pi/2$ = top).
<code>curve_shape</code>	Spline tension for curved edges (-1 to 1, default: 0).
<code>curve_pivot</code>	Pivot position along edge for curve control point (0-1, default: 0.5).
<code>curves</code>	Curve mode: TRUE (default) = single edges straight, reciprocal edges curve as ellipse (two opposing curves); FALSE = all straight; "force" = all curved.
<code>node_names</code>	Alternative names for legend (separate from display labels).
<code>legend</code>	Logical. Show legend?
<code>legend_position</code>	Legend position: "topright", "topleft", "bottomright", "bottomleft".
<code>scaling</code>	Scaling mode: "default" for qgraph-matched scaling where <code>node_size=6</code> looks similar to <code>qgraph vsize=6</code> , or "legacy" to preserve pre-v2.0 behavior.
<code>weight_digits</code>	Number of decimal places to round edge weights to before plotting. Edges that round to zero are automatically removed. Default 2. Set NULL to disable rounding.

Details

soplot vs splot:

`soplot()` uses grid graphics while `splot()` uses base R graphics. Both accept the same parameters and produce visually similar output. Choose based on:

- **soplot**: Better for integration with `ggplot2`, combining plots, and publication-quality vector graphics.
- **splot**: Better for large networks (faster rendering), interactive exploration, and traditional R workflows.

Edge Curve Behavior:

Edge curving is controlled by the `curves` and `curvature` parameters:

curves = FALSE All edges are straight lines.

curves = TRUE (Default) Reciprocal edge pairs (A->B and B->A) curve in opposite directions to form a visual ellipse. Single edges remain straight.

curves = "force" All edges curve inward toward the network center.

Weight Scaling Modes (`edge_scale_mode`):

Controls how edge weights map to visual widths:

linear Width proportional to weight. Best for similar-magnitude weights.

log Logarithmic scaling. Best for weights spanning orders of magnitude.

sqrt Square root scaling. Moderate compression for skewed data.

rank Rank-based scaling. Equal visual spacing regardless of values.

Donut Visualization:

The donut system visualizes proportions (0-1) as filled rings around nodes:

donut_fill Proportion filled (0-1). Can be scalar or per-node vector.

donut_color Fill color. Single color, `c(fill, bg)`, or per-node vector.

donut_shape Base shape: "circle", "square", "hexagon", etc.

donut_show_value Show numeric value in center.

Value

Invisible NULL. Called for side effect of drawing.

Invisible NULL. Called for side effect of drawing.

See Also

[splot](#) for base R graphics rendering (alternative engine), [cograph](#) for creating network objects, [sn_nodes](#) for node customization, [sn_edges](#) for edge customization, [sn_layout](#) for layout algorithms, [sn_theme](#) for visual themes, [from_qgraph](#) and [from_tna](#) for converting external objects

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
# With cograph()
cograph(adj) |> splot()

# Direct matrix input with all options
adj |> splot(
  layout = "circle",
  node_fill = "steelblue",
  node_size = 0.08,
  edge_width = 2
)
## Not run:
mat <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
sn_render(mat)

## End(Not run)
```

```
splot.group_tna_permutation
```

Plot Group Permutation Test Results

Description

Visualizes all pairwise permutation test results from a `group_tna` object. Creates a multi-panel plot with one panel per comparison.

Usage

```
splot.group_tna_permutation(x, ...)
```

```
plot_group_permutation(x, i = NULL, ...)
```

Arguments

<code>x</code>	A <code>group_tna_permutation</code> object (from <code>tna::permutation_test</code> on <code>group_tna</code>).
<code>...</code>	Additional arguments passed to <code>plot_permutation()</code> .
<code>i</code>	Index or name of specific comparison to plot. <code>NULL</code> for all.

Value

Invisibly returns NULL.

Examples

```
## Not run:
# Create a mock group_tna_permutation object
set.seed(42)
d1 <- matrix(c(0, 0.2, -0.1, -0.2, 0, 0.1, 0.1, -0.1, 0), 3, 3)
rownames(d1) <- colnames(d1) <- c("A", "B", "C")
d1_sig <- d1
d1_sig[abs(d1) < 0.15] <- 0
perm1 <- list(edges = list(diffs_true = d1, diffs_sig = d1_sig, stats = NULL))
attr(perm1, "labels") <- c("A", "B", "C")
class(perm1) <- c("tna_permutation", "list")
gperm <- list("G1 vs. G2" = perm1)
class(gperm) <- c("group_tna_permutation", "list")
plot_group_permutation(gperm)

## End(Not run)
```

splot.net_bootstrap *Plot Nestimate Bootstrap Results*

Description

Visualizes net_bootstrap objects from the Nestimate package. Mirrors splot.tna_bootstrap but adapts to Nestimate's field layout: weights live under \$original\$weights, directed is not always TRUE, and there are no donut/inits.

Plots the original tna model with nodes colored by community membership. The original model is retrieved from attr(x, "tna"), which tna::communities() sets automatically. Uses walktrap if present in x\$assignments; otherwise falls back to the first available algorithm column.

Plots the original network with nodes colored by community membership. The network is retrieved from attr(x, "network"), which detect_communities() / .wrap_communities() sets automatically.

Applies TNA-compatible styling defaults before delegating to splot(): directed networks get oval layout, coloured nodes, and sized arrows; undirected networks get spring layout with no arrows or dashes. All parameters can be overridden by the caller.

Visualizes boot_glasso objects from the Nestimate package. Plots a partial-correlation network with edge inclusion probabilities mapped to edge transparency.

Plot a wtna_mixed object either as a single overlaid network or as two separate group panels.

Visualizes net_permutation objects from the Nestimate package. Differs from plot_permutation: p_values and effect_size are already p x p matrices (no edge-name parsing needed), and directed comes from x\$x\$directed.

Network visualization using base R graphics (similar to qgraph).

Creates a network visualization using base R graphics functions (polygon, lines, xspline, etc.) instead of grid graphics. This provides better performance for large networks and uses the same snake_case parameter names as splot() for consistency.

Usage

```
splot.net_bootstrap(  
  x,  
  display = c("styled", "significant", "full"),  
  show_ci = FALSE,  
  show_stars = TRUE,  
  inherit_style = TRUE,  
  ...  
)  
  
splot.tna_communities(x, ...)  
  
splot.cograph_communities(x, ...)  
  
splot.net_mlvar(x, type = "temporal", ...)  
  
splot.netobject(x, ...)  
  
splot.boot_glasso(  
  x,  
  use_thresholded = TRUE,  
  show_inclusion = TRUE,  
  inclusion_threshold = NULL,  
  edge_positive_color = "#2E7D32",  
  edge_negative_color = "#C62828",  
  ...  
)  
  
splot.wtna_mixed(x, type = c("overlay", "group"), ...)  
  
splot.net_permutation(  
  x,  
  show_nonsig = FALSE,  
  show_effect = FALSE,  
  edge_positive_color = "#009900",  
  edge_negative_color = "#C62828",  
  edge_nonsig_color = "#888888",  
  edge_nonsig_style = 2L,  
  show_stars = TRUE,  
  ...  
)  
  
splot(  
  x,
```

```
layout = "oval",
directed = NULL,
seed = 42,
theme = NULL,
node_size = NULL,
node_size2 = NULL,
scale_nodes_by = NULL,
node_size_range = c(2, 8),
scale_nodes_scale = 1,
node_shape = "circle",
node_svg = NULL,
svg_preserve_aspect = TRUE,
node_fill = NULL,
node_border_color = NULL,
node_border_width = 1,
node_alpha = 1,
labels = TRUE,
label_size = NULL,
label_color = "black",
label_position = "center",
label_fontface = "plain",
label_fontfamily = "sans",
label_hjust = 0.5,
label_vjust = 0.5,
label_angle = 0,
pie_values = NULL,
pie_colors = NULL,
pie_border_width = NULL,
donut_fill = NULL,
donut_values = NULL,
donut_color = NULL,
donut_colors = NULL,
donut_border_color = NULL,
donut_border_width = NULL,
donut_inner_border_color = NULL,
donut_inner_border_width = NULL,
donut_outer_border_color = NULL,
donut_line_type = "solid",
donut_border_lty = NULL,
donut_inner_ratio = 0.8,
donut_bg_color = "gray90",
donut_shape = "circle",
donut_show_value = FALSE,
donut_value_size = 0.8,
donut_value_color = "black",
donut_value_fontface = "bold",
donut_value_fontfamily = "sans",
donut_value_digits = 2,
```

```
donut_value_prefix = "",
donut_value_suffix = "",
donut_empty = TRUE,
donut2_values = NULL,
donut2_colors = NULL,
donut2_inner_ratio = 0.4,
edge_color = NULL,
edge_width = NULL,
edge_size = NULL,
esize = NULL,
edge_width_range = c(0.1, 4),
edge_scale_mode = "linear",
edge_cutoff = NULL,
cut = NULL,
edge_alpha = 0.8,
edge_labels = FALSE,
edge_label_size = 0.8,
edge_label_color = "gray30",
edge_label_bg = NA,
edge_label_position = 0.5,
edge_label_offset = 0,
edge_label_fontface = "plain",
edge_label_shadow = FALSE,
edge_label_shadow_color = "gray40",
edge_label_shadow_offset = 0.5,
edge_label_shadow_alpha = 0.5,
edge_label_halo = TRUE,
edge_style = 1,
curvature = 0,
curve_scale = TRUE,
curve_shape = 0,
curve_pivot = 0.5,
curves = TRUE,
arrow_size = 1,
arrow_angle = pi/6,
show_arrows = TRUE,
bidirectional = FALSE,
loop_rotation = NULL,
edge_start_style = "solid",
edge_start_length = 0.15,
edge_start_dot_density = "12",
edge_ci = NULL,
edge_ci_scale = 2,
edge_ci_alpha = 0.15,
edge_ci_color = NA,
edge_ci_style = 2,
edge_ci_arrows = FALSE,
edge_priority = NULL,
```

```
edge_label_style = "none",
edge_label_template = NULL,
edge_label_digits = 2,
edge_label_online = TRUE,
edge_label_ci_format = "bracket",
edge_label_leading_zero = TRUE,
edge_ci_lower = NULL,
edge_ci_upper = NULL,
edge_label_p = NULL,
edge_label_p_digits = 3,
edge_label_p_prefix = "p=",
edge_label_stars = NULL,
weight_digits = 2,
threshold = 0,
minimum = 0,
maximum = NULL,
edge_positive_color = "#2E7D32",
positive_color = NULL,
edge_negative_color = "#C62828",
negative_color = NULL,
edge_duplicates = NULL,
title = NULL,
title_size = 1.2,
margins = c(0.1, 0.1, 0.1, 0.1),
background = "white",
rescale = TRUE,
layout_scale = 1,
layout_margin = 0.15,
aspect = TRUE,
use_pch = FALSE,
usePCH = NULL,
scaling = "default",
legend = FALSE,
legend_position = "topright",
legend_size = 0.8,
legend_edge_colors = TRUE,
legend_node_sizes = FALSE,
groups = NULL,
node_names = NULL,
tna_styling = NULL,
psych_styling = NULL,
i = NULL,
filetype = "default",
filename = file.path(tempdir(), "splot"),
width = 7,
height = 7,
res = 600,
...
```

)

Arguments

x	Network input. Can be: <ul style="list-style-type: none"> • A square numeric matrix (adjacency/weight matrix) • A data frame with edge list (from, to, optional weight columns) • An igraph object • A cograph_network object • A tna object (from tna package) • A group_tna object (list of tna objects from tna package). Use parameter <code>i</code> to select a specific group, or omit to plot all groups.
display	Display mode: "styled" (default), "significant", or "full".
show_ci	Logical: overlay CI bounds on edge labels? Default FALSE.
show_stars	Logical: show significance stars? Default TRUE.
inherit_style	Logical: inherit labels/layout/colors from network? Default TRUE.
...	Additional arguments passed to layout functions.
type	Character. "overlay" (default) renders both networks on a single canvas via plot_mixed_network — co-occurrence as straight undirected edges, transitions as curved directed arrows. "group" plots each component as a separate panel.
use_thresholded	Logical: use <code>\$thresholded_pcor</code> ? If FALSE, uses <code>\$original_pcor</code> . Default TRUE.
show_inclusion	Logical: scale edge alpha by inclusion probability? Default TRUE.
inclusion_threshold	Numeric: minimum inclusion probability to show an edge. Default $1 - x\alpha$ (i.e. the complement of the alpha level).
edge_positive_color	Color for positive weights.
edge_negative_color	Color for negative weights.
show_nonsig	Logical: show non-significant edges? Default FALSE.
show_effect	Logical: show effect size in parentheses? Default FALSE.
edge_nonsig_color	Color for non-significant edges. Default "#888888".
edge_nonsig_style	Line style for non-significant edges. Default 2L.
layout	Layout algorithm: "circle", "spring", "groups", or a matrix of x,y coordinates, or an igraph layout function. Also supports igraph two-letter codes: "kk", "fr", "drl", "mds", "ni", etc.
directed	Logical. Force directed interpretation. NULL for auto-detect.
seed	Random seed for deterministic layouts. Default 42.

theme	Theme name: "classic", "dark", "minimal", "colorblind", etc.
node_size	Node size(s). Single value or vector. Default 3.
node_size2	Secondary node size for ellipse/rectangle height.
scale_nodes_by	Scale node sizes by a centrality measure. Can be: <ul style="list-style-type: none"> • A measure name: "degree", "strength", "betweenness", "closeness", "eigenvector", "pagerank", "authority", "hub", "harmonic", etc. • A directional shorthand: "indegree", "outdegree", "instrength", "outstrength", "incloseness", "outcloseness", "inharmonic", "outharmonic", "ineccentricity", "outeccentricity". • A list with measure and parameters: list("pagerank", damping = 0.9) When used, node_size is ignored. Use node_size_range to control the min/max size. Default NULL (no centrality scaling).
node_size_range	Size range for centrality-based scaling. Numeric vector c(min_size, max_size). Default c(2, 8).
scale_nodes_scale	Dampening exponent for centrality-based sizing. Values < 1 compress differences (e.g., 0.5 applies square root), values > 1 exaggerate differences. Default 1 (linear).
node_shape	Node shape(s): "circle", "square", "triangle", "diamond", "pentagon", "hexagon", "star", "heart", "ellipse", "cross", or any custom SVG shape registered with register_svg_shape().
node_svg	Custom SVG for nodes: path to SVG file OR inline SVG string.
svg_preserve_aspect	Logical: maintain SVG aspect ratio? Default TRUE.
node_fill	Node fill color(s).
node_border_color	Node border color(s).
node_border_width	Node border width(s).
node_alpha	Node transparency (0-1). Default 1.
labels	Node labels: TRUE (use node names/indices), FALSE (none), or character vector.
label_size	Label character expansion factor.
label_color	Label text color.
label_position	Label position: "center", "above", "below", "left", "right".
label_fontface	Font face for labels: "plain", "bold", "italic", "bold.italic". Default "plain".
label_fontfamily	Font family for labels: "sans", "serif", "mono". Default "sans".
label_hjust	Horizontal justification (0=left, 0.5=center, 1=right). Default 0.5.
label_vjust	Vertical justification (0=bottom, 0.5=center, 1=top). Default 0.5.
label_angle	Text rotation angle in degrees. Default 0.

pie_values	List of numeric vectors for pie chart nodes. Each element corresponds to a node and contains values for pie segments. If a simple numeric vector with values between 0 and 1 is provided (e.g., centrality scores), it is automatically converted to donut_fill for convenience.
pie_colors	List of color vectors for pie segments.
pie_border_width	Border width for pie slice dividers. NULL uses node_border_width.
donut_fill	Numeric value (0-1) for donut fill proportion. This is the qgraph-style API: 0.1 = 10% filled, 0.5 = 50% filled, 1.0 = fully filled. Can be a single value (all nodes) or vector (per-node values).
donut_values	Deprecated. Use donut_fill for simple fill proportion.
donut_color	Fill color(s) for the donut ring. Single color sets fill for all nodes. Two colors set fill and background for all nodes. More than 2 colors set per-node fill colors (recycled to n_nodes). Default: "maroon" fill, "gray90" background when node_shape="donut".
donut_colors	Deprecated. Use donut_color instead.
donut_border_color	Border color for donut rings. NULL uses node_border_color.
donut_border_width	Border width for donut rings. NULL uses node_border_width.
donut_inner_border_color	Color for the inner boundary (where the donut meets its hole). NULL (default) uses donut_border_color. Can be scalar or per-node vector.
donut_inner_border_width	Width for the inner boundary border. NULL (default) uses donut_border_width. Can be scalar or per-node vector.
donut_outer_border_color	Color for outer boundary border (enables double border). NULL (default) shows single border. Set to a color for double border effect. Can be scalar or per-node vector.
donut_line_type	Line type for donut borders: "solid", "dashed", "dotted", or numeric (1=solid, 2=dashed, 3=dotted). Can be scalar or per-node vector.
donut_border_lty	Deprecated. Use donut_line_type instead.
donut_inner_ratio	Inner radius ratio for donut (0-1). Default 0.5.
donut_bg_color	Background color for unfilled donut portion.
donut_shape	Base shape for donut: "circle", "square", "hexagon", "triangle", "diamond", "pentagon". Can be a single value or per-node vector. Default inherits from node_shape (e.g., hexagon nodes get hexagon donuts). Set explicitly to override (e.g., donut_shape = "hexagon" for hexagon donuts on all nodes regardless of node_shape).
donut_show_value	Logical: show value in donut center? Default FALSE.

donut_value_size	Font size for donut center value.
donut_value_color	Color for donut center value.
donut_value_fontface	Font face for donut center value: "plain", "bold", "italic", "bold.italic". Default "bold".
donut_value_fontfamily	Font family for donut center value: "sans", "serif", "mono". Default "sans".
donut_value_digits	Decimal places for donut center value. Default 2.
donut_value_prefix	Text before donut center value (e.g., "\$"). Default "".
donut_value_suffix	Text after donut center value (e.g., "%"). Default "".
donut_empty	Logical: render empty donut rings for NA values? Default TRUE.
donut2_values	List of values for inner donut ring (for double donut).
donut2_colors	List of color vectors for inner donut ring segments.
donut2_inner_ratio	Inner radius ratio for inner donut ring. Default 0.4.
edge_color	Edge color(s). If NULL, uses edge_positive_color/edge_negative_color based on weight.
edge_width	Edge width(s). If NULL, scales by weight using edge_size and edge_width_range.
edge_size	Base edge size for weight scaling. NULL (default) uses adaptive sizing based on network size: $15 * \exp(-n_nodes/90) + 1$. For directed networks, this is halved. Larger values = thicker edges overall.
esize	Deprecated. Use edge_size instead.
edge_width_range	Output width range as c(min, max) for weight-based scaling. Default c(0.5, 4). Edges are scaled to fit within this range.
edge_scale_mode	Scaling mode for edge weights: "linear" (default, qgraph-style), "log" (logarithmic for wide weight ranges), "sqrt" (moderate compression), or "rank" (equal visual spacing regardless of weight distribution).
edge_cutoff	Two-tier cutoff for edge width scaling. NULL (default) = auto-calculate as 75th percentile of weights (qgraph behavior). 0 = disabled (continuous scaling). Positive number = manual threshold. Edges below cutoff get minimal width variation.
cut	Deprecated. Use edge_cutoff instead.
edge_alpha	Edge transparency (0-1). Default 0.8.
edge_labels	Edge labels: TRUE (show weights), FALSE (none), or character vector.
edge_label_size	Edge label size.

edge_label_color	Edge label text color.
edge_label_bg	Edge label background color.
edge_label_position	Position along edge (0-1).
edge_label_offset	Perpendicular offset for edge labels (0 = on line, positive = above).
edge_label_fontface	Font face: "plain", "bold", "italic", "bold.italic".
edge_label_shadow	Logical: enable drop shadow for edge labels? Default FALSE.
edge_label_shadow_color	Color for edge label shadow. Default "gray40".
edge_label_shadow_offset	Offset distance for shadow in points. Default 0.5.
edge_label_shadow_alpha	Transparency for shadow (0-1). Default 0.5.
edge_label_halo	Logical: enable white halo/outline around edge labels for readability over dark edges? Default FALSE. When TRUE, overrides shadow settings.
edge_style	Line type(s): 1=solid, 2=dashed, 3=dotted, etc.
curvature	Edge curvature. 0 for straight, positive/negative for curves.
curve_scale	Logical: auto-curve reciprocal edges?
curve_shape	Spline tension (-1 to 1). Default 0.
curve_pivot	Position along edge for curve control point (0-1).
curves	Curve mode: TRUE (default) = single edges straight, reciprocal edges curve as ellipse (two opposing curves); FALSE = all straight; "force" = all curved.
arrow_size	Arrow head size.
arrow_angle	Arrow head angle in radians. Default pi/6 (30 degrees).
show_arrows	Logical or vector: show arrows on directed edges?
bidirectional	Logical or vector: show arrows at both ends?
loop_rotation	Angle(s) in radians for self-loop direction.
edge_start_style	Style for the start segment of edges: "solid" (default), "dashed", or "dotted". Use dashed/dotted to indicate edge direction (source node).
edge_start_length	Fraction of edge length for the styled start segment (0-0.5). Default 0.15 (15% of edge). Only applies when edge_start_style is not "solid".
edge_start_dot_density	Pattern for dotted start segments. A two-character string where the first digit is dot length and second is gap length (in line width units). Default "12" (1 unit dot, 2 units gap). Use "11" for tighter dots, "13" for more spacing. Only applies when edge_start_style = "dotted".

<code>edge_ci</code>	Numeric vector of CI widths (0-1 scale). Larger values = more uncertainty.
<code>edge_ci_scale</code>	Width multiplier for underlay thickness. Default 2.
<code>edge_ci_alpha</code>	Transparency for underlay (0-1). Default 0.15.
<code>edge_ci_color</code>	Underlay color. NA (default) uses main edge color.
<code>edge_ci_style</code>	Line type for underlay: 1=solid, 2=dashed, 3=dotted. Default 2.
<code>edge_ci_arrows</code>	Logical: show arrows on underlay? Default FALSE.
<code>edge_priority</code>	Numeric vector of edge priorities. Higher values render on top. Useful for ensuring significant edges appear above non-significant ones.
<code>edge_label_style</code>	<p>Preset style: "none", "estimate", "full", "range", "stars".</p>
<code>edge_label_template</code>	<p>Template with placeholders: {est}, {range}, {low}, {up}, {p}, {stars}. Overrides <code>edge_label_style</code> if provided.</p>
<code>edge_label_digits</code>	Decimal places for estimates. Default 2.
<code>edge_label_online</code>	Logical: single line format? Default TRUE.
<code>edge_label_ci_format</code>	CI format: "bracket" for [low, up] or "dash" for low-up.
<code>edge_label_leading_zero</code>	Logical: show leading zero for values < 1? Default TRUE. Set to FALSE to display ".5" instead of "0.5".
<code>edge_ci_lower</code>	Numeric vector of lower CI bounds for labels.
<code>edge_ci_upper</code>	Numeric vector of upper CI bounds for labels.
<code>edge_label_p</code>	Numeric vector of p-values for edges.
<code>edge_label_p_digits</code>	Decimal places for p-values. Default 3.
<code>edge_label_p_prefix</code>	Prefix for p-values. Default "p=".
<code>edge_label_stars</code>	Stars for labels: character vector, TRUE (compute from p), or numeric (treated as p-values).
<code>weight_digits</code>	Number of decimal places to round edge weights to before plotting. Edges that round to zero are automatically removed. Default 2. Set NULL to disable rounding.
<code>threshold</code>	Minimum absolute weight to display.
<code>minimum</code>	Alias for threshold (qgraph compatibility). Uses max of threshold and minimum.
<code>maximum</code>	Maximum weight for scaling. NULL for auto.
<code>positive_color</code>	Deprecated. Use <code>edge_positive_color</code> instead.
<code>negative_color</code>	Deprecated. Use <code>edge_negative_color</code> instead.

edge_duplicates	How to handle duplicate edges in undirected networks. NULL (default) = stop with error listing duplicates. Options: "sum", "mean", "first", "max", "min", or a custom aggregation function.
title	Plot title.
title_size	Title font size.
margins	Margins as c(bottom, left, top, right).
background	Background color.
rescale	Logical: rescale layout to -1 to 1 range?
layout_scale	Scale factor for layout. >1 expands (spreads nodes apart), <1 contracts (brings nodes closer). Use "auto" to automatically scale based on node count (compact for small networks, expanded for large). Default 1.
layout_margin	Margin around the layout as fraction of range. Default 0.15. Set to 0 for no extra margin (tighter fit). Affects white space around nodes.
aspect	Logical: maintain aspect ratio?
use_pch	Logical: use points() for simple circles (faster). Default FALSE.
usePCH	Deprecated. Use use_pch instead.
scaling	Scaling mode: "default" for qgraph-matched scaling where node_size=6 looks similar to qgraph vsize=6, or "legacy" to preserve pre-v2.0 behavior.
legend	Logical: show legend?
legend_position	Position: "topright", "topleft", "bottomright", "bottomleft".
legend_size	Legend text size.
legend_edge_colors	Logical: show positive/negative edge colors in legend?
legend_node_sizes	Logical: show node size scale in legend?
groups	Group assignments for node coloring/legend.
node_names	Alternative names for legend (separate from labels).
tna_styling	Logical or NULL. If TRUE, applies TNA visual defaults (oval layout, TNA color palette, edge labels as estimates, dotted edge starts, etc.) as a base layer. Any explicitly provided argument overrides the TNA default. If FALSE, no TNA styling is applied. If NULL (default), automatically set to TRUE when x is a tna object, FALSE otherwise. Can be used with any input type (matrix, igraph, co-graph_network).
psych_styling	Logical or NULL. Undirected counterpart of tna_styling. If TRUE, applies psychometric-network defaults (spring layout, Okabe-Ito palette, no arrows, thin edges) as a base layer. If NULL (default), splot.netobject auto-enables it on correlation-family input (glasso, cor, pcor, ising) and on the undirected constituents of net_m1var. Explicit user args always win.
i	Group index or name when x is a group_tna object. If NULL (default), plots all groups in a grid. If specified (e.g., i = 1 or i = "Treatment"), plots only that group.

filetype	Output format: "default" (screen), "png", "pdf", "svg", "jpeg", "tiff".
filename	Output filename (without extension).
width	Output width in inches.
height	Output height in inches.
res	Resolution in DPI for raster outputs (PNG, JPEG, TIFF). Default 600.

Details

Edge Curve Behavior:

Edge curving is controlled by three parameters that interact:

curves Mode for automatic curving. FALSE = all straight, TRUE (default) = curve only reciprocal edge pairs as an ellipse, "force" = curve all edges inward toward network center.

curvature Manual curvature amount (0-1 typical). Sets the magnitude of curves. Default 0 uses automatic 0.175 for curved edges. Positive values curve edges; the direction is automatically determined.

curve_scale Not currently used; reserved for future scaling.

For reciprocal edges (A->B and B->A both exist), the edges curve in opposite directions to form a visual ellipse, making bidirectional relationships clear.

Weight Scaling Modes (edge_scale_mode):

Controls how edge weights are mapped to visual widths:

linear (default) Width proportional to weight. Best when weights are similar in magnitude.

log Logarithmic scaling. Best when weights span multiple orders of magnitude (e.g., 0.01 to 100).

sqrt Square root scaling. Moderate compression, good for moderately skewed distributions.

rank Rank-based scaling. Ignores actual values; uses relative ordering. All edges get equal visual spacing regardless of weight distribution.

Donut vs Pie vs Double Donut:

Three ways to show additional data on nodes:

Donut (donut_fill) Single ring showing a proportion (0-1). Ideal for completion rates, probabilities, or any single metric per node. Use donut_color for fill color and donut_bg_color for unfilled portion.

Pie (pie_values) Multiple colored segments showing category breakdown. Ideal for composition data. Values are normalized to sum to 1. Use pie_colors for segment colors.

Double Donut (donut2_values) Two concentric rings for comparing two metrics per node. Outer ring uses donut_fill/donut_color, inner ring uses donut2_values/donut2_colors.

CI Underlay System:

Confidence interval underlays draw a wider, semi-transparent edge behind the main edge to visualize uncertainty:

edge_ci Vector of CI widths (0-1 scale). Larger = more uncertainty.

edge_ci_scale Multiplier for underlay width relative to main edge. Default 2 means underlay is twice as wide as main edge at CI=1.

edge_ci_alpha Transparency of underlay (0-1). Default 0.15.

edge_ci_style Line type: 1=solid, 2=dashed (default), 3=dotted.

Edge Label Templates:

For statistical output, use templates to format complex labels:

edge_label_template Template string with placeholders: {est} for estimate/weight, {low}/{up} for CI bounds, {range} for formatted range, {p} for p-value, {stars} for significance stars.

edge_label_style Preset styles: "estimate" (weight only), "full" (estimate + CI), "range" (CI only), "stars" (significance).

Value

Invisibly returns the plot.

Invisibly, the plot result.

Invisibly, the plot result.

Invisibly returns x.

Invisibly returns the plot.

Invisibly returns the plot.

Invisibly returns x.

Invisibly returns the plot.

Invisibly returns the `cograph_network` object.

See Also

[soplot](#) for grid graphics rendering (alternative engine), [cograph](#) for creating network objects, [sn_nodes](#) for node customization, [sn_edges](#) for edge customization, [sn_layout](#) for layout algorithms, [sn_theme](#) for visual themes, [from_qgraph](#) and [from_tna](#) for converting external objects

Examples

```
# Basic network from adjacency matrix
adj <- matrix(c(0, 1, 1, 0,
               0, 0, 1, 1,
               0, 0, 0, 1,
               0, 0, 0, 0), 4, 4, byrow = TRUE)
splot(adj)

# With curved edges
splot(adj, curvature = 0.2)

# Weighted network with colors
w_adj <- matrix(c(0, 0.5, -0.3, 0,
                 0.8, 0, 0.4, -0.2,
                 0, 0, 0, 0.6,
                 0, 0, 0, 0), 4, 4, byrow = TRUE)
splot(w_adj, edge_positive_color = "darkgreen", edge_negative_color = "red")

# Pie chart nodes
splot(adj, pie_values = list(c(1,2,3), c(2,2), c(1,1,1,1), c(3,1)))
```

```
# Circle layout with labels
splot(adj, layout = "circle", labels = c("A", "B", "C", "D"))
```

splot.tna_disparity *Plot Disparity Results with splot*

Description

Plot Disparity Results with splot

Usage

```
splot.tna_disparity(
  x,
  show = c("styled", "backbone", "full"),
  edge_style_sig = 1,
  edge_style_nonsig = 2,
  alpha_nonsig = 0.3,
  ...
)
```

Arguments

x	A tna_disparity object.
show	What to display: "styled" (default), "backbone", "full".
edge_style_sig	Line style for backbone edges. Default 1 (solid).
edge_style_nonsig	Line style for non-backbone edges. Default 2 (dashed).
alpha_nonsig	Alpha for non-backbone edges. Default 0.3.
...	Additional arguments passed to splot.

Value

Invisibly returns NULL. Called for the side effect of producing a plot.

Examples

```
## Not run:
mat <- matrix(c(
  0.0, 0.5, 0.1, 0.0,
  0.3, 0.0, 0.4, 0.1,
  0.1, 0.2, 0.0, 0.5,
  0.0, 0.1, 0.3, 0.0
), nrow = 4, byrow = TRUE)
rownames(mat) <- colnames(mat) <- c("A", "B", "C", "D")
```

```

disp <- disparity_filter(cograph(mat), level = 0.05)
splot.tna_disparity(disp)
splot.tna_disparity(disp, show = "backbone")

## End(Not run)

```

splot.tna_permutation *Plot Permutation Test Results*

Description

Visualizes permutation test results with styling to distinguish significant from non-significant edge differences. Works with tna_permutation objects from the tna package.

Usage

```

splot.tna_permutation(x, ...)

plot_permutation(
  x,
  show_nonsig = FALSE,
  edge_positive_color = "#009900",
  edge_negative_color = "#C62828",
  edge_nonsig_color = "#888888",
  edge_nonsig_style = 2,
  show_stars = TRUE,
  show_effect = FALSE,
  edge_nonsig_alpha = 0.4,
  ...
)

```

Arguments

x	A tna_permutation object (from tna::permutation_test).
...	Additional arguments passed to splot().
show_nonsig	Logical: show non-significant edges? Default FALSE (only significant shown).
edge_positive_color	Color for positive differences ($x > y$). Default "#009900" (green).
edge_negative_color	Color for negative differences ($x < y$). Default "#C62828" (red).
edge_nonsig_color	Color for non-significant edges. Default "#888888" (grey).
edge_nonsig_style	Line style for non-significant edges (2=dashed). Default 2.
show_stars	Logical: show significance stars (*, **, ***) on edges? Default TRUE.

show_effect Logical: show effect size in parentheses for significant edges? Default FALSE.
 edge_nonsig_alpha Alpha for non-significant edges. Default 0.4.

Details

The function expects a tna_permutation object containing:

- edges\$diffs_true: Matrix of actual edge differences (x - y)
- edges\$diffs_sig: Matrix of significant differences only
- edges\$stats: Data frame with edge_name, diff_true, effect_size, p_value

Edge styling:

- Significant positive: solid green, bold labels with stars
- Significant negative: solid red, bold labels with stars
- Non-significant (when show_nonsig=TRUE): dashed grey, plain labels, lower alpha

Value

Invisibly returns the plot.

Examples

```
## Not run:
# Create a mock tna_permutation object with synthetic data
set.seed(42)
diffs <- matrix(c(0, 0.15, -0.1, -0.2, 0, 0.05, 0.1, -0.05, 0), 3, 3)
rownames(diffs) <- colnames(diffs) <- c("A", "B", "C")
diffs_sig <- diffs
diffs_sig[abs(diffs) < 0.1] <- 0
perm <- list(
  edges = list(
    diffs_true = diffs,
    diffs_sig = diffs_sig,
    stats = data.frame(
      edge_name = c("A -> B", "A -> C", "B -> A", "B -> C", "C -> A", "C -> B"),
      diff_true = c(0.15, -0.1, -0.2, 0.05, 0.1, -0.05),
      effect_size = c(2.1, -1.5, -2.8, 0.4, 1.2, -0.3),
      p_value = c(0.01, 0.04, 0.001, 0.3, 0.02, 0.5)
    )
  )
)
attr(perm, "level") <- 0.05
attr(perm, "labels") <- c("A", "B", "C")
class(perm) <- c("tna_permutation", "list")
plot_permutation(perm)

## End(Not run)
```

student_interactions *Student Interaction Edge List*

Description

An edge list of observed interactions between 34 students during collaborative learning sessions. Each row represents one observed interaction between two students. The same pair may appear multiple times, reflecting repeated interactions.

Usage

```
student_interactions
```

Format

A data frame with 389 rows and 2 columns:

from Character. Anonymized two-letter student code (e.g., "Ac", "Bd")

to Character. Anonymized two-letter student code (e.g., "Ce", "Df")

Details

The dataset includes self-loops (34 rows where `from == to`), which may represent self-directed actions. These can be removed with `student_interactions[student_interactions$from != student_interactions$to,]`.

Because interactions repeat, this edge list naturally represents a multigraph when loaded into `igraph` with `igraph::graph_from_data_frame()`.

Value

A data frame with 389 rows and 2 columns:

from Character. Anonymized two-letter student code.

to Character. Anonymized two-letter student code.

Source

Anonymized collaborative learning interaction data.

Examples

```
# Load and build network
data(student_interactions)
head(student_interactions)

# Remove self-loops and build igraph
el <- student_interactions[student_interactions$from != student_interactions$to, ]
if (requireNamespace("igraph", quietly = TRUE)) {
```

```

g <- igraph::graph_from_data_frame(e1, directed = FALSE)
cat("Students:", igraph::vcount(g), "\n")
cat("Interactions:", igraph::ecount(g), "\n")
}

```

subgraphs

Extract Specific Motif Instances (Subgraphs)

Description

Convenience wrapper for `motifs(x, named_nodes = TRUE, ...)`. Returns specific node triples forming each MAN pattern.

Usage

```
subgraphs(...)
```

Arguments

... Additional arguments passed to internal plot helpers.

Value

A `cograph_motif_result` object with `named_nodes = TRUE`. Contains `$results` (data frame with columns `triad`, `type`, `observed`, and optionally `z`, `p`, `sig`), `$type_summary`, `$level`, `$n_units`, and `$params`.

See Also

[motifs\(\)](#)

Other motifs: [extract_motifs\(\)](#), [extract_triads\(\)](#), [get_edge_list\(\)](#), [motif_census\(\)](#), [motifs\(\)](#), [plot.cograph_motif_analysis\(\)](#), [plot.cograph_motifs\(\)](#), [triad_census\(\)](#)

Examples

```

mat <- matrix(c(0,3,2,0, 0,0,5,1, 0,0,0,4, 2,0,0,0), 4, 4, byrow = TRUE)
rownames(mat) <- colnames(mat) <- c("Plan", "Execute", "Monitor", "Adapt")
subgraphs(mat, significance = FALSE)

```

summarize_network *Summarize Network by Clusters*

Description

Creates a summary network where each cluster becomes a single node. Edge weights are aggregated from the original network using the specified method. Returns a `cograph_network` object ready for plotting.

Usage

```
summarize_network(
  x,
  cluster_list = NULL,
  method = c("sum", "mean", "max", "min", "median", "density", "geomean"),
  directed = TRUE
)

cluster_network(
  x,
  cluster_list = NULL,
  method = c("sum", "mean", "max", "min", "median", "density", "geomean"),
  directed = TRUE
)

cnet(
  x,
  cluster_list = NULL,
  method = c("sum", "mean", "max", "min", "median", "density", "geomean"),
  directed = TRUE
)
```

Arguments

<code>x</code>	A weight matrix, tna object, or <code>cograph_network</code> .
<code>cluster_list</code>	Cluster specification: <ul style="list-style-type: none"> • Named list of node vectors (e.g., <code>list(A = c("n1", "n2"), B = c("n3", "n4"))</code>) • String column name from nodes data (e.g., "clusters", "groups") • NULL to auto-detect from common column names
<code>method</code>	Aggregation method for edge weights: "sum", "mean", "max", "min", "median", "density", "geomean". Default "sum".
<code>directed</code>	Logical. Treat network as directed. Default TRUE.

Value

A `cograph_network` object with:

- One node per cluster (named by cluster)
- Edge weights = aggregated between-cluster weights
- `nodes$size` = cluster sizes (number of original nodes)

See [summarize_network](#).

See [summarize_network](#).

See Also

[cluster_summary](#), [plot_mcml](#)

Examples

```
# Create a network with clusters
mat <- matrix(runif(100), 10, 10)
diag(mat) <- 0
rownames(mat) <- colnames(mat) <- LETTERS[1:10]

# Define clusters
clusters <- list(
  Group1 = c("A", "B", "C"),
  Group2 = c("D", "E", "F"),
  Group3 = c("G", "H", "I", "J")
)

# Create summary network
summary_net <- summarize_network(mat, clusters)
plot(summary_net)

# With cograph_network (auto-detect clusters column)
Net <- cograph(mat)
Net$nodes$clusters <- rep(c("A", "B", "C"), c(3, 3, 4))
summary_net <- summarize_network(Net) # Auto-detects 'clusters'
```

summary.cograph_network

Summary of cograph_network Object

Description

Summary of `cograph_network` Object

Usage

```
## S3 method for class 'cograph_network'
summary(object, ...)
```

Arguments

object A cograph_network object.
 ... Ignored.

Value

A list with network summary information (invisibly), containing elements n_nodes, n_edges, directed, weighted, and has_layout.

Examples

```
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), nrow = 3)
net <- cograph(adj)
summary(net)
```

supra_adjacency	<i>Supra-Adjacency Matrix</i>
-----------------	-------------------------------

Description

Builds the supra-adjacency matrix for multilayer networks. Diagonal blocks = intra-layer, off-diagonal = inter-layer.

Usage

```
supra_adjacency(
  layers,
  omega = 1,
  coupling = c("diagonal", "full", "custom"),
  interlayer_matrices = NULL
)

supra(
  layers,
  omega = 1,
  coupling = c("diagonal", "full", "custom"),
  interlayer_matrices = NULL
)
```

Arguments

layers List of adjacency matrices (same dimensions)
 omega Inter-layer coupling coefficient (scalar or L x L matrix)
 coupling Coupling type: "diagonal", "full", or "custom"
 interlayer_matrices
 For coupling="custom", list of inter-layer matrices

Value

Supra-adjacency matrix of dimension $(NL) \times (NL)$

Examples

```
# layers <- list(L1 = mat1, L2 = mat2)
# S <- supra_adjacency(layers, omega = 0.5)
# dim(S) # (2*n) x (2*n)
```

supra_interlayer	<i>Extract Inter-Layer Block</i>
------------------	----------------------------------

Description

Extract Inter-Layer Block

Usage

```
supra_interlayer(x, from, to)

extract_interlayer(x, from, to)
```

Arguments

x	Supra-adjacency matrix
from	Source layer index
to	Target layer index

Value

Inter-layer adjacency matrix

Examples

```
L1 <- matrix(c(0,.5,.3,.5,0,.4,.3,.4,0), 3, 3)
L2 <- matrix(c(0,.2,.6,.2,0,.1,.6,.1,0), 3, 3)
S <- supra_adjacency(list(L1 = L1, L2 = L2), omega = 0.5)
supra_interlayer(S, 1, 2)
L1 <- matrix(c(0,.5,.3,.5,0,.4,.3,.4,0), 3, 3)
L2 <- matrix(c(0,.2,.6,.2,0,.1,.6,.1,0), 3, 3)
S <- supra_adjacency(list(L1 = L1, L2 = L2), omega = 0.5)
extract_interlayer(S, 1, 2)
```

supra_layer	<i>Extract Layer from Supra-Adjacency Matrix</i>
-------------	--

Description

Extract Layer from Supra-Adjacency Matrix

Usage

```
supra_layer(x, layer)
extract_layer(x, layer)
```

Arguments

x	Supra-adjacency matrix
layer	Layer index to extract

Value

Intra-layer adjacency matrix

Examples

```
L1 <- matrix(c(0,.5,.3,.5,0,.4,.3,.4,0), 3, 3)
L2 <- matrix(c(0,.2,.6,.2,0,.1,.6,.1,0), 3, 3)
S <- supra_adjacency(list(L1 = L1, L2 = L2), omega = 0.5)
supra_layer(S, 1)
L1 <- matrix(c(0,.5,.3,.5,0,.4,.3,.4,0), 3, 3)
L2 <- matrix(c(0,.2,.6,.2,0,.1,.6,.1,0), 3, 3)
S <- supra_adjacency(list(L1 = L1, L2 = L2), omega = 0.5)
extract_layer(S, 2)
```

themes-builtin	<i>Built-in Themes</i>
----------------	------------------------

Description

Pre-defined themes for network visualization.

Value

A CographTheme object.

Examples

```
theme_cograph_classic()
theme_cograph_dark()
```

theme_cograph_classic *Classic Theme*

Description

Traditional network visualization style with blue nodes and gray edges.

Usage

```
theme_cograph_classic()
```

Value

A CographTheme object.

Examples

```
theme <- theme_cograph_classic()
```

theme_cograph_colorblind
Colorblind-friendly Theme

Description

Theme using colors distinguishable by people with color vision deficiency.

Usage

```
theme_cograph_colorblind()
```

Value

A CographTheme object.

Examples

```
theme <- theme_cograph_colorblind()
```

theme_cograph_dark *Dark Theme*

Description

Dark background theme for presentations.

Usage

```
theme_cograph_dark()
```

Value

A CographTheme object.

Examples

```
theme <- theme_cograph_dark()
```

theme_cograph_gray *Grayscale Theme*

Description

Black and white theme suitable for print.

Usage

```
theme_cograph_gray()
```

Value

A CographTheme object.

Examples

```
theme <- theme_cograph_gray()
```

theme_cograph_minimal *Minimal Theme*

Description

Clean, minimal style with thin borders.

Usage

```
theme_cograph_minimal()
```

Value

A CographTheme object.

Examples

```
theme <- theme_cograph_minimal()
```

theme_cograph_nature *Nature Theme*

Description

Earth tones theme inspired by nature.

Usage

```
theme_cograph_nature()
```

Value

A CographTheme object.

Examples

```
theme <- theme_cograph_nature()
```

theme_cograph_viridis *Viridis Theme*

Description

Theme using viridis color palette.

Usage

```
theme_cograph_viridis()
```

Value

A CographTheme object.

Examples

```
theme <- theme_cograph_viridis()
```

to_data_frame *Export Network as Edge List Data Frame*

Description

Converts a network to an edge list data frame with columns for source, target, and weight.

Usage

```
to_data_frame(x, directed = NULL)
```

```
to_df(x, directed = NULL)
```

Arguments

x Network input: matrix, igraph, network, cograph_network, or tna object.

directed Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.

Value

A data frame with columns:

- from: Source node name/label
- to: Target node name/label
- weight: Edge weight

See Also

[to_df](#), [to_igraph](#), [as_cograph](#)

Examples

```
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")

# Convert to edge list
to_data_frame(adj)

# Use alias
to_df(adj)
```

to_igraph

Convert Network to igraph Object

Description

Converts various network representations to an igraph object. Supports matrices, igraph objects, network objects, cograph_network, and tna objects.

Usage

```
to_igraph(x, directed = NULL)
```

Arguments

x	Network input. Can be: <ul style="list-style-type: none"> • A square numeric matrix (adjacency/weight matrix) • An igraph object (returned as-is or converted if directed differs) • A statnet network object • A cograph_network object • A tna object
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.

Value

An igraph object.

See Also

[to_data_frame](#), [as_cograph](#)

Examples

```
# From matrix
adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
g <- to_igraph(adj)

# Force directed
g_dir <- to_igraph(adj, directed = TRUE)
```

to_matrix

*Convert Network to Adjacency Matrix***Description**

Converts any supported network format to an adjacency matrix.

Usage

```
to_matrix(x, directed = NULL)
```

Arguments

x Network input: matrix, cograph_network, igraph, network, tna, etc.
 directed Logical or NULL. If NULL (default), auto-detect from input.

Value

A square numeric adjacency matrix with row/column names.

See Also

[to_igraph](#), [to_df](#), [as_cograph](#), [to_network](#)

Examples

```
# From matrix
adj <- matrix(c(0, .5, .8, 0,
               .5, 0, .3, .6,
               .8, .3, 0, .4,
               0, .6, .4, 0), 4, 4, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C", "D")
to_matrix(adj)

# From cograph_network
net <- as_cograph(adj)
to_matrix(net)

# From igraph (weighted graph)
if (requireNamespace("igraph", quietly = TRUE)) {
```

```
g <- igraph::graph_from_adjacency_matrix(adj, mode = "undirected", weighted = TRUE)
to_matrix(g)
}
```

to_network*Convert Network to statnet network Object*

Description

Converts any supported network format to a statnet network object.

Usage

```
to_network(x, directed = NULL)
```

Arguments

x Network input: matrix, cograph_network, igraph, tna, etc.
directed Logical or NULL. If NULL (default), auto-detect from input.

Value

A network object from the network package.

See Also

[to_igraph](#), [to_matrix](#), [to_df](#), [as_cograph](#)

Examples

```
if (requireNamespace("network", quietly = TRUE)) {
  adj <- matrix(c(0, 1, 1, 1, 0, 1, 1, 1, 0), 3, 3)
  rownames(adj) <- colnames(adj) <- c("A", "B", "C")
  net <- to_network(adj)
}
```

trophic_incoherence *Trophic Incoherence Parameter*

Description

The trophic incoherence parameter q is a measure of how "vertically ordered" a directed network is (Johnson et al. 2014). For each edge (u, v) , the trophic difference is $x_{uv} = s_v - s_u$ where s_i is the trophic level of node i . The trophic incoherence parameter is the (population) standard deviation of these differences:

$$q = \sqrt{\frac{1}{|E|} \sum_{(u,v) \in E} (x_{uv} - \bar{x})^2}$$

Usage

```
trophic_incoherence(x, cannibalism = TRUE)
```

Arguments

<code>x</code>	Directed network input.
<code>cannibalism</code>	Logical. If FALSE, self-loops are removed before computing trophic differences. Default TRUE.

Details

Low values ($q \approx 0$) indicate a perfectly coherent network (e.g., a pure food web where every edge goes up one level). High values indicate an incoherent network with many level-skipping or downward edges. Johnson et al. 2014 showed that low- q food webs are dynamically more stable.

Matches `networkx.trophic_incoherence_parameter` at machine epsilon. Directed-only; requires at least one basal node (node with no incoming edges) for trophic levels to be well-defined.

Value

A single numeric value (NA_real_ for empty edge sets or undirected input).

References

Johnson, S., Dominguez-Garcia, V., Donetti, L., & Munoz, M. A. (2014). Trophic coherence determines food-web stability. *PNAS*, 111(50), 17923-17928.

See Also

[centrality](#) (the `trophic_level` measure) for the per-node levels used in the incoherence calculation.

Examples

```
# Small directed 3-node chain: 1 -> 2 -> 3 (perfectly coherent, q = 0)
adj <- matrix(c(0,1,0, 0,0,1, 0,0,0), 3, 3, byrow = TRUE)
rownames(adj) <- colnames(adj) <- c("A", "B", "C")
trophic_incoherence(adj)
```

unregister_svg_shape *Unregister SVG Shape*

Description

Remove a custom SVG shape from the registry.

Usage

```
unregister_svg_shape(name)
```

Arguments

name Shape name to remove.

Value

Invisible TRUE if removed, FALSE if not found.

Examples

```
# Attempt to unregister a non-existent shape (returns FALSE)
unregister_svg_shape("nonexistent")
```

verify_with_igraph *Verify Against igraph*

Description

Confirms numerical match with igraph's contract_vertices + simplify.

Usage

```
verify_with_igraph(x, clusters, method = "sum", type = "raw")

verify_igraph(x, clusters, method = "sum", type = "raw")
```

Arguments

x	Adjacency matrix
clusters	Cluster specification
method	Aggregation method
type	Normalization type. Defaults to "raw" for igraph compatibility.

Value

List with comparison results

Examples

```
if (requireNamespace("igraph", quietly = TRUE)) {
  mat <- matrix(runif(100), 10, 10)
  diag(mat) <- 0
  rownames(mat) <- colnames(mat) <- LETTERS[1:10]
  clusters <- c(1,1,1,2,2,2,3,3,3,3)
  verify_igraph(mat, clusters)
}
```

vulnerability	<i>Node Vulnerability</i>
---------------	---------------------------

Description

Computes the vulnerability of each node, defined as the relative drop in global efficiency when that node is removed from the network.

Usage

```
vulnerability(x, directed = NULL, normalized = TRUE, digits = NULL, ...)
```

Arguments

x	Network input: matrix, igraph, network, cograph_network, or tna object.
directed	Logical or NULL. If NULL (default), auto-detect from matrix symmetry. Set TRUE to force directed, FALSE to force undirected.
normalized	Logical. If TRUE (default), return the proportional drop. If FALSE, return the raw efficiency difference.
digits	Integer or NULL. Round scores to this many decimal places. Default NULL (no rounding).
...	Additional arguments passed to to_igraph .

Details

$$V(i) = \frac{E_{global} - E_{global \setminus i}}{E_{global}}$$

where E_{global} is the global efficiency of the full network and $E_{global \setminus i}$ is the global efficiency after removing node i and all its edges.

Global efficiency is defined as:

$$E_{global} = \frac{1}{n(n-1)} \sum_{i \neq j} \frac{1}{d(i,j)}$$

Nodes with high vulnerability are critical to the network's communication efficiency. Removing them causes the greatest drop in global efficiency.

Performance note: This function computes all-pairs shortest paths once for the full graph and once per node removal, giving $O(n)$ calls to the shortest-path algorithm. A warning is issued for networks with more than 500 nodes.

Value

An object of class "cograph_vulnerability" with components:

scores Named numeric vector of vulnerability scores, sorted descending.

network The original input network.

normalized Logical flag indicating normalization mode.

References

Latora, V. & Marchiori, M. (2007). A measure of centrality based on network efficiency. *New Journal of Physics*, 9(6), 188. doi:10.1088/13672630/9/6/188

See Also

[network_global_efficiency](#), [robustness](#), [centrality](#)

Examples

```
# Star network: hub is most vulnerable
star <- matrix(c(0,1,1,1, 1,0,0,0, 1,0,0,0, 1,0,0,0), 4, 4)
rownames(star) <- colnames(star) <- c("hub", "a", "b", "c")
cograph::vulnerability(star)

# Complete graph: all nodes equally vulnerable
k4 <- matrix(1, 4, 4); diag(k4) <- 0
rownames(k4) <- colnames(k4) <- c("A", "B", "C", "D")
cograph::vulnerability(k4)
```

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