Package 'tidysynthesis'

November 11, 2025

Title A Common API for Synthesizing Data

Version 0.1.2

Maintainer Aaron R. Williams <awilliams@urban.org>

Description A system built on 'tidymodels' for generating synthetic tabular data. We provide tools for ordering a sequential synthesis, feature and target engineering, sampling, hyperparameter tuning, enforcing constraints, and adding extra noise during a synthesis.

URL https://ui-research.github.io/tidysynthesis-documentation/

BugReports https://github.com/UrbanInstitute/tidysynthesis/issues

Depends R (>= 4.1.0)

Imports dplyr, forcats, parsnip, pillar, purrr, progressr, recipes, rlang, rsample, stringr, tibble, tidyr (>= 1.0.0), tune, vctrs, workflows, yardstick, ExtDist, dapper

Suggests hardhat, palmerpenguins, poissonreg, randomForest, ranger, rpart, rpart.LAD (>= 0.1.2), testthat (>= 2.1.0), usethis, VGAM

License GPL-3

Encoding UTF-8

LazyData true

RoxygenNote 7.3.2

NeedsCompilation no

Author Aaron R. Williams [aut, cre] (ORCID:
<https://orcid.org/0000-0001-5564-1938>),
Kyle Ueyama [aut],
Gabe Morrison [aut] (ORCID: <https://orcid.org/0009-0008-1815-5920>),
Jeremy Seeman [aut] (ORCID: <https://orcid.org/0000-0003-3526-3209>),
Elyse McFalls [ctb],
Claire Morton [ctb],
Livia Mucciolo [ctb],
Madeline Pickens [ctb],
Noah Zwiefel [ctb],
The Urban Institute [cph]

Repository CRAN

Date/Publication 2025-11-11 21:20:07 UTC

Contents

`.name_to_inspect` *constant mapping between component names and inspections*

Description

constant mapping between component names and inspections

Usage

```
.name_to_inspect
```

Format

An object of class `list` of length 7.

acs_conf *American Community Survey confidential microdata (with weights)*

Description

An extract constructed from the 2019 American Community Survey containing a survey sample of n = 1500 Nebraska respondents, with survey weights included.

Usage

```
acs_conf
```

Format

`acs_conf`**:**

A data frame with 1,500 rows and 12 columns:

county fct, county

gq fct, group quarter kind

sex fct, sex

marst fct, marital status

hcovany fct, health insurance status

empstat fct, employment status; contains empty levels.

classwkr fct, employment kind (ex: self-employed, etc.); contains "N/A" levels.

age dbl, age (in years)

famsize dbl, household/family size

transit_time dbl, transit time to work (in minutes)

inctot dbl, annual income; contains missing values

wgt dbl, survey weight

Details

Original data source: Steven Ruggles, Sarah Flood, Matthew Sobek, Daniel Backman, Annie Chen, Grace Cooper, Stephanie Richards, Renae Rogers, and Megan Schouweiler. IPUMS USA: Version 15.0 [dataset]. Minneapolis, MN: IPUMS, 2024. https://doi.org/10.18128/D010.V15.0

Source

https://usa.ipums.org/usa/

`acs_conf_nw` *American Community Survey confidential microdata (without weights)*

Description

An extract constructed from the 2019 American Community Survey containing a survey sample of n = 1500 Nebraska respondents, with survey weights included.

Usage

```
acs_conf_nw
```

Format

`acs_conf_nw`**:**

A data frame with 1,500 rows and 11 columns:

county fct, county

gq fct, group quarter kind

sex fct, sex

marst fct, marital status

hcovany fct, health insurance status

empstat fct, employment status; contains empty levels.

classwkr fct, employment kind (ex: self-employed, etc.); contains "N/A" levels.

age dbl, age (in years)

famsize dbl, household/family size

transit_time dbl, transit time to work (in minutes)

inctot dbl, annual income; contains missing values

Details

Original data source: Steven Ruggles, Sarah Flood, Matthew Sobek, Daniel Backman, Annie Chen, Grace Cooper, Stephanie Richards, Renae Rogers, and Megan Schouweiler. IPUMS USA: Version 15.0 [dataset]. Minneapolis, MN: IPUMS, 2024. https://doi.org/10.18128/D010.V15.0

Source

`https://usa.ipums.org/usa/`

`acs_start` *American Community Survey starting microdata (with weights)*

Description

An extract constructed from the 2019 American Community Survey containing a survey sample of n = 500 Nebraska respondents, with survey weights included.

Usage

```
acs_start
```

Format

`acs_start`:

A data frame with 500 rows and 5 columns:

county fct, county

gq fct, group quarter kind

sex fct, sex

marst fct, marital status

wgt dbl, survey weight

Details

Original data source: Steven Ruggles, Sarah Flood, Matthew Sobek, Daniel Backman, Annie Chen, Grace Cooper, Stephanie Richards, Renae Rogers, and Megan Schouweiler. IPUMS USA: Version 15.0 [dataset]. Minneapolis, MN: IPUMS, 2024. https://doi.org/10.18128/D010.V15.0

Source

`https://usa.ipums.org/usa/`

`acs_start_nw` *American Community Survey starting microdata (without weights)*

Description

An extract constructed from the 2019 American Community Survey containing a survey sample of n = 500 Nebraska respondents, with survey weights included.

Usage

```
acs_start_nw
```

Format

`acs_start_nw`**:**

A data frame with 500 rows and 4 columns:

county fct, county

gq fct, group quarter kind

sex fct, sex

marst fct, marital status

Details

Original data source: Steven Ruggles, Sarah Flood, Matthew Sobek, Daniel Backman, Annie Chen, Grace Cooper, Stephanie Richards, Renae Rogers, and Megan Schouweiler. IPUMS USA: Version 15.0 [dataset]. Minneapolis, MN: IPUMS, 2024. https://doi.org/10.18128/D010.V15.0

Source

`https://usa.ipums.org/usa/`

`add_noise_cat_unif` *Inject noise into a categorical random variable by mixing a sample of uniform records into the predictions.*

Description

Inject noise into a categorical random variable by mixing a sample of uniform records into the predictions.

Usage

```
add_noise_cat_unif(
  model,
  new_data,
  conf_model_data,
  outcome_var,
  col_schema,
  pred,
  unif_prop,
  resample_props = NULL,
  observed_levels = FALSE
)
```

Arguments

`model` A `model_spec` or a list of `model_specs` from `library(parsnip)`

`new_data` A data frame used to generate predictions

`conf_model_data` A data frame for estimating the predictive model

`outcome_var` A string name representing the outcome variable

`col_schema` A list of column schema specifications for the new variable

`pred` A vector of values predicted by the model

`unif_prop` A proportion of records to resample with uniform noise

`resample_props` An optional named vector of probabilities for resampling, defaults to uniform over all levels supplied in `col_schema`.

`observed_levels` An optional Boolean to only resample from observed levels in the confidential data.

Value

A numeric vector with noise added to each prediction

Examples

```
conf_model_data <- mtcars|>
  dplyr::mutate(gear = factor(.data[["gear"]]))

col_schema <- list(
  "dtype" = "fct",
  "levels" = c("3", "4", "5"),
  "na_prop" = 0
)

add_noise_cat_unif(
  model = conf_model_data,
  new_data = NULL,
  conf_model_data = NULL,
  outcome_var = "gear",
  col_schema = col_schema,
  pred = factor(c(rep("3", 10), rep("4", 10), rep("5", 10))),
  unif_prop = 0.5
)
```

`add_noise_disc_gaussian` *Add discrete normal noise with mean 0 to predicted values with constant variance*

Description

Add discrete normal noise with mean 0 to predicted values with constant variance

Usage

```
add_noise_disc_gaussian(
  model,
  new_data,
  conf_model_data,
  outcome_var,
  col_schema,
  pred,
  variance = NULL,
  rho = NULL,
  sensitivity = NULL,
  increment = 1
)
```

Arguments

`model` A `model_spec` or a list of `model_specs` from `library(parsnip)`

`new_data` A data frame used to generate predictions

`conf_model_data` A data frame for estimating the predictive model

`outcome_var` A string name representing the outcome variable

`col_schema` A list of column schema specifications for the new variable

`pred` A vector of values predicted by the model

`variance` float, sampling variance for additive noise

`rho` float, alternative privacy loss budget prescribed by the Gaussian mechanism under rho-zero-concentrated differential privacy.

`sensitivity` float, alternative sample sensitivity prescribed by the Gaussian mechanism under rho-zero-concentrated differential privacy.

`increment` Numeric indicating space between discrete noise samples, defaults to 1. Note that this does not impact the noise sampling variance, as the increment rescales noise distributions specified by sampling variance.

Value

A numeric vector with noise added to each prediction

Examples

```
add_noise_disc_gaussian(
  model = NULL,
  new_data = NULL,
  conf_model_data = NULL,
  outcome_var = NULL,
  col_schema = NULL,
  pred = 1:100,
  variance = 3
)
```

add_noise_disc_laplace

Add discrete Laplace noise with mean 0 to predicted values with constant variance

Description

Add discrete Laplace noise with mean 0 to predicted values with constant variance

Usage

```
add_noise_disc_laplace(
  model,
  new_data,
  conf_model_data,
  outcome_var,
  col_schema,
  pred,
  variance = NULL,
  epsilon = NULL,
  sensitivity = NULL,
  increment = 1
)
```

Arguments

`model`	A `model_spec` or a list of `model_specs` from `library(parsnip)`
`new_data`	A data frame used to generate predictions
`conf_model_data`	A data frame for estimating the predictive model
`outcome_var`	A string name representing the outcome variable
`col_schema`	A list of column schema specifications for the new variable
`pred`	A vector of values predicted by the model
`variance`	float, sampling variance for additive noise

`epsilon` float, alternative privacy loss budget prescribed by the Laplace mechanism under epsilon differential privacy.

`sensitivity` float, alternative sample sensitivity prescribed by the Laplace mechanism under epsilon differential privacy.

`increment` Numeric indicating space between discrete noise samples, defaults to 1. Note that this does not impact the noise sampling variance, as the increment rescales noise distributions specified by sampling variance.

Value

A numeric vector with noise added to each prediction

Examples

```
add_noise_disc_laplace(
  model = NULL,
  new_data = NULL,
  conf_model_data = NULL,
  outcome_var = NULL,
  col_schema = NULL,
  pred = 1:100,
  variance = 3
)
```

`add_noise_gaussian` *Add normal noise with mean 0 to predicted values with constant variance*

Description

Add normal noise with mean 0 to predicted values with constant variance

Usage

```
add_noise_gaussian(
  model,
  new_data,
  conf_model_data,
  outcome_var,
  col_schema,
  pred,
  variance = NULL,
  rho = NULL,
  sensitivity = NULL
)
```

Arguments

`model`	A `model_spec` or a list of `model_specs` from `library(parsnip)`
`new_data`	A data frame used to generate predictions
`conf_model_data`	A data frame for estimating the predictive model
`outcome_var`	A string name representing the outcome variable
`col_schema`	A list of column schema specifications for the new variable
`pred`	A vector of values predicted by the model
`variance`	Sampling variance for additive noise
`rho`	Alternative privacy loss budget prescribed by the Gaussian mechanism under rho-zero-concentrated differential privacy.
`sensitivity`	Alternative sample sensitivity prescribed by the Gaussian mechanism under rho-zero-concentrated differential privacy.

Value

A numeric vector with noise added to each prediction

Examples

```
add_noise_gaussian(
  model = NULL,
  new_data = NULL,
  conf_model_data = NULL,
  outcome_var = NULL,
  col_schema = NULL,
  pred = 1:100,
  variance = 3
)
```

`add_noise_kde` *Add normal noise to predicted values with variances calculated for ntiles using Gaussian kernel density estimators*

Description

Add normal noise to predicted values with variances calculated for ntiles using Gaussian kernel density estimators

Usage

```
add_noise_kde(
  model,
  new_data,
  conf_model_data,
  outcome_var,
  col_schema,
  pred,
  exclusions = NULL,
  n_ntiles = NULL,
  obs_per_ntile = NULL,
  ties_method = "collapse",
  sd_scale = 1
)
```

Arguments

`model` A `model_spec` or a list of `model_specs` from `library(parsnip)`

`new_data` A data frame used to generate predictions

`conf_model_data` A data frame for estimating the predictive model

`outcome_var` A string name representing the outcome variable

`col_schema` A list of column schema specifications for the new variable

`pred` A vector of values predicted by the model

`exclusions` Numeric values that should not receive extra noise

`n_ntiles` The number of ntiles

`obs_per_ntile` A numeric for the minimum number of observations to be in an ntile. Cannot be used in conjunction with the `n_ntiles` argument.

`ties_method` The ntiles approach to adding noise requires a one-to-one mapping from model-generated values to ntiles in the original data. The methods "collapse", "random", and "exclusions" deal with situations where the ntiles lack unique bounds. "collapse" collapses ntile breaks to preserve the one-to-one relationship; "random" adds a small random perturbation to the derived boundaries; finally, "exclusions" treats ntile tie values as derived exclusions.

`sd_scale` float, a positive number to scale the estimated KDE variance. Defaults to 1.0

Value

A numeric vector with noise added to each prediction

Examples

```
add_noise_kde(
  model = NULL,
  new_data = tibble::tibble(x = 1:100),
  conf_model_data = tibble::tibble(x = 1:100),
  outcome_var = "x",
  col_schema = NULL,
  pred = 1:100,
  n_ntiles = 4
)
```

add_noise_laplace — *Add Laplace noise with mean 0 to predicted values with constant variance*

Description

Add Laplace noise with mean 0 to predicted values with constant variance

Usage

```
add_noise_laplace(
  model,
  new_data,
  conf_model_data,
  outcome_var,
  col_schema,
  pred,
  variance = NULL,
  epsilon = NULL,
  sensitivity = NULL
)
```

Arguments

`model`	A `model_spec` or a list of `model_specs` from `library(parsnip)`
`new_data`	A data frame used to generate predictions
`conf_model_data`	A data frame for estimating the predictive model
`outcome_var`	A string name representing the outcome variable
`col_schema`	A list of column schema specifications for the new variable
`pred`	A vector of values predicted by the model
`variance`	Sampling variance for additive noise
`epsilon`	Alternative privacy loss budget prescribed by the Laplace mechanism under epsilon differential privacy.
`sensitivity`	Alternative sample sensitivity prescribed by the Laplace mechanism under epsilon differential privacy.

Value

A numeric vector with noise added to each prediction

Examples

```
add_noise_laplace(
  model = NULL,
  new_data = NULL,
  conf_model_data = NULL,
  outcome_var = NULL,
  col_schema = NULL,
  pred = 1:100,
  variance = 3
)
```

add_sequence_factor *Add to visit sequence for factor variables*

Description

Add to visit sequence for factor variables

Usage

```
add_sequence_factor(roadmap, ..., method = c("entropy", "gini"))
```

Arguments

roadmap	A roadmap object
...	<tidy-select> One or more unquoted expressions separated by commas. Variable names can be used as if they were positions in the data frame, so expressions like x:y can be used to select a range of variables.
method	A quoted name for the method used to sort the visit_sequence. Current methods include "entropy" and "gini".

Value

An updated visit_sequence

Examples

```
roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
) |>
  add_sequence_factor(dplyr::where(is.factor), method = "gini")
```

`add_sequence_manual` *Add to visit sequence using a manual method*

Description

Add to visit sequence using a manual method

Usage

```
add_sequence_manual(roadmap, ...)
```

Arguments

`roadmap` A `roadmap` object.

`...` `<tidy-select>` One or more unquoted expressions separated by commas. Variable names can be used as if they were positions in the data frame, so expressions like x:y can be used to select a range of variables.

Value

An updated `roadmap` object.

Examples

```
roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
) |>
  add_sequence_manual(
    c("inctot", "hcovany", "empstat",  "classwkr",  "age",   "famsize",
      "transit_time")
  )
```

`add_sequence_numeric` *Add to visit sequence for numeric variables*

Description

Add to visit sequence for numeric variables

Usage

```
add_sequence_numeric(
  roadmap,
  ...,
  method = c("correlation", "proportion", "weighted total", "absolute weighted total",
    "weighted absolute total"),
  cor_var = NULL,
  na.rm = FALSE,
  cor_use = "everything"
)
```

Arguments

`roadmap` A `roadmap` object

`...` `<tidy-select>` One or more unquoted expressions separated by commas. Variable names can be used as if they were positions in the data frame, so expressions like x:y can be used to select a range of variables.

`method` A quoted name for the method used to sort the visit_sequence

`cor_var` A numeric variable for the correlation method

`na.rm` Boolean that if TRUE, removes `NA` values from computations

`cor_use` A string correlation data method passed to `stats::cor` if using. If `na.rm == TRUE` then defaults to `complete.obs`. See `?stats::cor` for more options.

Value

An updated visit_sequence

Examples

```
roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
) |>
  add_sequence_numeric(
    dplyr::where(is.numeric),
    method = "correlation",
    cor_var = "age",
    na.rm = TRUE
  )
```

collapse_na *Collapse data frames with _NA variables to coerce related variables to include NA*

Description

Collapse data frames with _NA variables to coerce related variables to include NA

Usage

```
collapse_na(data)
```

Arguments

data	A data frame with columns ending in _NA

Value

A data frame with no _NA columns and NA values

Examples

```
example_na_expanded <- expand_na(data = example_na)

collapse_na(data = example_na_expanded)
```

constraints *Create a constraints object*

Description

Create a constraints object

Usage

```
constraints(
  schema,
  constraints_df_num = NULL,
  constraints_df_cat = NULL,
  max_z_num = 0,
  max_z_cat = 0
)
```

Arguments

`schema` A `schema` object

`constraints_df_num`
A specially formatted data frame with constraints to be imposed during the synthesis process. See examples for formatting.

`constraints_df_cat`
A specifically formatted data frame with constraints to be imposed during the synthesis process.

`max_z_num` Numeric vector(s) for the number of times a value should be resampled before hardbounding if it violates a constraint.

`max_z_cat` Numeric vector(s) for the number of times a value should be resampled before hardbounding if it violates a constraint.

Value

A `constraints` object.

Examples

```
constraints(
  schema = schema(
    conf_data = mtcars |> dplyr::mutate(vs = factor(vs)),
    start_data = dplyr::select(mtcars, cyl)
  ),
  constraints_df_num = tibble::tribble(
    ~var, ~min, ~max, ~conditions,
    # ensure all mpg values are greater than 0
    "mpg", 0, Inf, "TRUE",
    # ensure when cyl == 6, mpg is less than 15
    "mpg", -Inf, 15, "cyl == 6",
    # ensure disp is always between 0 and 150
    "disp", 0, 150, "TRUE"
  ),
  constraints_df_cat = tibble::tribble(
    ~var, ~allowed, ~forbidden, ~conditions,
    # ensure vs != 1 when gear >= 4
    "vs", NA,  1, "gear >= 5",
    # ensure vs == 1 when gear >= 4
    "vs", 0,  NA, "gear == 4"
  )
)
```

`constraints_api` *Add, update, or reset a* `constraints` *object within an existing* `roadmap`.

Description

Add, update, or reset a `constraints` object within an existing `roadmap`.

Usage

```
add_constraints(roadmap, constraints)

update_constraints(roadmap, ...)

reset_constraints(roadmap)
```

Arguments

`roadmap`	A `roadmap` object
`constraints`	A `constraints` object.
`...`	Optional named parameters passed to `constraints()`.

Value

A new `roadmap` object.

A roadmap object with added constraints.

A roadmap object with updated constraints.

A roadmap object with reset constraints.

Examples

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

constraints_df_num <-
  tibble::tribble(~var, ~min, ~max, ~conditions,
                  ”transit_time”, 0, 300, ”TRUE”)

constraints <- constraints(
  schema = rm[[”schema”]],
  constraints_df_num = constraints_df_num,
  max_z_num = 0
)

rm |>
  add_constraints(constraints)


rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)
```

```
constraints_df_num <-
  tibble::tribble(~var, ~min, ~max, ~conditions,
                 "transit_time", 0, 300, "TRUE")

constraints <- constraints(
  schema = rm[["schema"]],
  constraints_df_num = constraints_df_num,
  max_z_num = 0
)

rm |>
  update_constraints(constraints)


rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

constraints_df_num <-
  tibble::tribble(~var, ~min, ~max, ~conditions,
                 "transit_time", 0, 300, "TRUE")

constraints <- constraints(
  schema = rm[["schema"]],
  constraints_df_num = constraints_df_num,
  max_z_num = 0
)

rm <- rm |>
  add_constraints(constraints)

reset_constraints(rm)
```

construct_extractors *Construct a list of extractors for parsnip models*

Description

Construct a list of extractors for parsnip models

Usage

```
construct_extractors(
  roadmap,
  default_extractor = NULL,
  custom_extractors = NULL
)
```

Arguments

`roadmap`	A roadmap object
`default_extractor`	An extractor from library(parsnip)
`custom_extractors`	A formatted list of extractors

Value

A named list of extractors

Examples

```
# construct_extractors() can create a sequence of extractors using a fully-default
# approach, a hybrid approach, or a fully-customized approach. All approaches
# require a roadmap and extractors.

rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

# Fully-default approach

construct_extractors(
  roadmap = rm,
  default_extractor = parsnip::extract_fit_engine
)

# Hybrid approach

construct_extractors(
  roadmap = rm,
  default_extractor = parsnip::extract_fit_engine,
  custom_extractors = list(
    list(vars = "hcovany", extractor = parsnip::extract_parameter_dials)
  )
)

# Fully-customized approach

construct_extractors(
  roadmap = rm,
  custom_extractors = list(
    list(
      vars = c("hcovany", "empstat", "classwkr"),
      extractor = parsnip::extract_fit_engine
    ),
    list(
      vars = c("age", "famsize", "transit_time", "inctot"),
      extractor = parsnip::extract_parameter_dials
    )
  )
)
```

construct_models *Construct a list of models for synthesis*

Description

Construct a list of models for synthesis

Usage

```
construct_models(
  roadmap,
  default_regression_model = NULL,
  default_classification_model = NULL,
  custom_models = NULL
)
```

Arguments

`roadmap` A roadmap object

`default_regression_model`
A `parsnip` model object used for regression in numeric outcome variables

`default_classification_model`
A `parsnip` model object used for classification in categorical outcome variables

`custom_models` A formatted list with `parsnip` model objects explicitly paired with every variable in the `visit_sequence`

Value

A named list of models

Examples

```
# construct_models() can create a sequence of models using a fully-default
# approach, a hybrid approach, or a fully-customized approach. All approaches
# require a roadmap and model objects.

rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rpart_mod_reg <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "regression")

rpart_mod_class <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "classification")

lm_mod <- parsnip::linear_reg() |>
  parsnip::set_engine("lm") |>
  parsnip::set_mode(mode = "regression")

# Fully-default approach

construct_models(
  roadmap = rm,
  default_regression_model = lm_mod,
  default_classification_model = rpart_mod_class
)

# Hybrid approach

construct_models(
  roadmap = rm,
  default_regression_model = lm_mod,
  default_classification_model = rpart_mod_class,
  custom_models = list(
    list(vars = "age", model = lm_mod)
  )
)

# Fully-customized approach

construct_models(
  roadmap = rm,
  custom_models = list(
    list(vars = c("hcovany", "empstat", "classwkr"), model = rpart_mod_class),
    list(vars = c("age", "famsize", "transit_time", "inctot"), model = rpart_mod_reg)
  )
)
```

construct_noise *Construct a list of noise objects for synthesis*

Description

Construct a list of noise objects for synthesis

Usage

```
construct_noise(
    roadmap,
    default_regression_noise = NULL,
    default_classification_noise = NULL,
    custom_noise = NULL
  )
```

Arguments

roadmap	A roadmap object
default_regression_noise	A noise function for regression models
default_classification_noise	A noise function for classification models
custom_noise	A formatted list of noise functions

Value

A named list of noise

Examples

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

noise_defaults <- construct_noise(
  roadmap = rm,
  default_regression_noise = noise(),
  default_classification_noise = noise()
)


# construct_noise() can create a sequence of noise objects using a
# fully-default approach, a hybrid approach, or a fully-customized approach.
# All approaches require a roadmap and noise objects.

rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

noise_reg <- noise(
  add_noise = TRUE,
  mode = "regression",
  noise_fun = add_noise_gaussian
)

noise_class <- noise(
  add_noise = TRUE,
  mode = "classification",
  noise_fun = add_noise_cat_unif
)

# Fully-default approach

construct_noise(
  roadmap = rm,
  default_regression_noise = noise_reg,
  default_classification_noise = noise_class
)

# Hybrid approach

noise_reg2 <- noise(
  add_noise = TRUE,
  mode = "regression",
  noise_fun = add_noise_disc_gaussian
)

construct_noise(
  roadmap = rm,
  default_regression_noise = noise_reg,
  default_classification_noise = noise_class,
  custom_noise = list(
    list(vars = "age", noise = noise_reg2)
  )
)

# Fully-customized approach

construct_noise(
  roadmap = rm,
  custom_noise = list(
    list(vars = c("hcovany", "empstat", "classwkr"), noise = noise_class),
    list(vars = c("age", "famsize", "transit_time", "inctot"), noise = noise_reg)
  )
)
```

construct_recipes *Construct a sequence of model recipes for sequential synthesis*

Description

Construct a sequence of model recipes for sequential synthesis

Usage

```
construct_recipes(
  roadmap,
  default_regression_steps = NULL,
  default_classification_steps = NULL,
  custom_steps = NULL
)
```

Arguments

`roadmap` A roadmap object

`default_regression_steps`
A list containing one or more recipes::step_*()

`default_classification_steps`
A list containing one or more recipes::step_*()

`custom_steps` A list of lists containing one or more recipes::step_*()

Value

A list of formulas

Examples

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

construct_recipes(rm)


# construct_recipes() can create a sequence of recipes using a fully-default
# approach, a hybrid approach, or a fully-customized approach. All approaches
# require a roadmap and steps.

rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

step1 <- function(x) {
x |>
  recipes::step_center(recipes::all_predictors(), id = "center")
}

# Fully-default approach

construct_recipes(
  roadmap = rm,
  default_regression_steps = step1,
  default_classification_steps = step1
)

# Hybrid approach
```

```
step2 <- function(x) {
  x |>
    recipes::step_scale(recipes::all_predictors(), id = "scale")
}

construct_recipes(
  roadmap = rm,
  default_regression_steps = step1,
  default_classification_steps = step1,
  custom_steps = list(
    list(vars = "age", step = step2)
  )
)

# Fully-customized approach

construct_recipes(
  roadmap = rm,
  custom_steps = list(
    list(vars = c("hcovany", "empstat", "classwkr"), step = step1),
    list(vars = c("age", "famsize", "transit_time", "inctot"), step = step1)
  )
)
```

`construct_samplers` *Construct a list of samplers for synthesis*

Description

Construct a list of samplers for synthesis

Usage

```
construct_samplers(
  roadmap,
  default_regression_sampler = NULL,
  default_classification_sampler = NULL,
  custom_samplers = NULL
)
```

Arguments

`roadmap` A roadmap object

`default_regression_sampler`
A sampler function for regression models

`default_classification_sampler`
A sampler function for classification models

`custom_samplers`
A formatted list of sampler functions

Value

A named list of samplers

Examples

```
# construct_samplers() can create a sequence of samplers using a fully-default
# approach, a hybrid approach, or a fully-customized approach. All approaches
# require a roadmap and samplers.

rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

# Fully-default approach

construct_samplers(
  roadmap = rm,
  default_regression_sampler = sample_lm,
  default_classification_sampler = sample_rpart
)

# Hybrid approach

construct_samplers(
  roadmap = rm,
  default_regression_sampler = sample_lm,
  default_classification_sampler = sample_rpart,
  custom_samplers = list(
    list(vars = "hcovany", sampler = sample_rpart)
  )
)

# Fully-customized approach

construct_samplers(
  roadmap = rm,
  custom_samplers = list(
    list(vars = c("hcovany", "empstat", "classwkr"), sampler = sample_rpart),
    list(vars = c("age", "famsize", "transit_time", "inctot"), sampler = sample_lm)
  )
)
```

construct_tuners — *Construct a list of tuning grids for hyperparameter tuning predictive models*

Description

Construct a list of tuning grids for hyperparameter tuning predictive models

Usage

```
construct_tuners(
  roadmap,
  default_regression_tuner = NULL,
  default_classification_tuner = NULL,
  custom_tuners = NULL
)
```

Arguments

`roadmap`	A roadmap object
`default_regression_tuner`	A tuner.
`default_classification_tuner`	A tuner.
`custom_tuners`	A formatted list of tuners.

Value

A named list of tuners

Examples

```
# construct_tuners() can create a sequence of tuners using a fully-default
# approach, a hybrid approach, or a fully-customized approach. All approaches
# require a roadmap and tuners.

rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

tuner_reg <- list(
  v = 3,
  grid = 3,
  metrics = yardstick::metric_set(yardstick::rmse)
)

tuner_cat <- list(
  v = 3,
  grid = 3,
  metrics = yardstick::metric_set(yardstick::roc_auc)
)

# Fully-default approach

construct_tuners(
  roadmap = rm,
  default_regression_tuner = tuner_reg,
  default_classification_tuner = tuner_cat
)

# Hybrid approach

tuner_cat2 <- list(
  v = 3,
  grid = 3,
  metrics = yardstick::metric_set(yardstick::precision)
)

construct_tuners(
  roadmap = rm,
  default_regression_tuner = tuner_reg,
  default_classification_tuner = tuner_cat,
  custom_tuners = list(
    list(vars = "hcovany", tuner = tuner_cat2)
  )
)

# Fully-customized approach

construct_tuners(
  roadmap = rm,
  custom_tuners = list(
    list(vars = c("hcovany", "empstat", "classwkr"), tuner = tuner_reg),
    list(vars = c("age", "famsize", "transit_time", "inctot"), tuner = tuner_cat)
  )
)
```

`convert_level_to_na` *Convert* `"NA"` *values to* `NA` *for categorical variables*

Description

Convert `"NA"` values to `NA` for categorical variables

Usage

```
convert_level_to_na(data)
```

Arguments

data A data frame or tibble

Value

A data frame or tibble with "NA" converted to NA

Examples

```
data <- data.frame(
  x1 = c(1, 2, NA),
  x2 = c("1", "2", "NA"),
  x3 = factor(c("1", "2", "NA")),
  x4 = factor(c("b", "NA", "a"), ordered = TRUE)
)

convert_level_to_na(data)
```

convert_na_to_level *Convert NA values to "NA" for categorical variables*

Description

Convert NA values to "NA" for categorical variables

Usage

```
convert_na_to_level(data)
```

Arguments

data A data frame or tibble

Value

A data frame or tibble with NA converted to "NA"

Examples

```
data <- data.frame(
  x1 = c(1, 2, NA),
  x2 = c("1", "2", NA),
  x3 = factor(c("1", "2", NA)),
  x4 = factor(c("b", NA, "a"), levels = c("b", NA, "a"), ordered = TRUE)
)

convert_na_to_level(data)
```

enforce_custom_na *Redefine NA value for a dataset.*

Description

Redefine NA value for a dataset.

Usage

```
enforce_custom_na(data, col_schema)
```

Arguments

data	A data.frame object
col_schema	A col_schema from a schema object

Value

A data.frame

Examples

```
# create custom NA filter
example_na_custom <- example_na |>
  tidyr::replace_na(
    list("wages" = -999)
  )

example_na_expanded_custom <- enforce_custom_na(
  data = example_na_custom,
  col_schema = list(
    "wages" = list(
      dtype = "dbl",
      na_value = -999
     )
   )
 )
```

enforce_na *Add missing values where values should be missing according to _NA variables*

Description

Add missing values where values should be missing according to _NA variables

Usage

```
enforce_na(data)
```

Arguments

data	A synthetic data frame with _NA columns

Value

A synthetic data frame with _NA columns that converts values that are labelled missing in an _NA variable to missing in the corresponding variable

Examples

```
example_na_expanded <- expand_na(data = example_na)

enforce_na(data = example_na_expanded)
```

---

enforce_schema *Enforce a* roadmap*'s* schema *on its existing data*

Description

Enforce a roadmap's schema on its existing data

Usage

```
enforce_schema(roadmap)
```

Arguments

roadmap	A roadmap object

Value

A roadmap object with modified conf_data, start_data, and schema information.

Examples

```
rm <- roadmap(conf_data = acs_conf, start_data = acs_start) |>
  update_schema(na_numeric_to_ind = TRUE)

enforce_schema(rm)
```

example_na *A df with different types of missingness*

Description

A df with different types of missingness

Usage

```
example_na
```

Format

A tibble with 200 observations and 6 variables:

age Age of respondent

sex Sex of respondent with missingness at random

labor_force Labor force status of respondent with structural missingness

hours Hours work of respondent with missingness at random

wages Wages earned with structural missingness

expand_na *Add new variables that indicate if a value is "missing" or "not missing" for original variables that contain NA*

Description

Add new variables that indicate if a value is "missing" or "not missing" for original variables that contain NA

Usage

```
expand_na(
  data,
  types = c(”chr”, ”dbl”, ”fct”, ”lgl”, ”int”, ”ord”),
  skip_vars = NULL
)
```

Arguments

data	A data frame
types	A vector of variables types to expand
skip_vars	A character vector of variables that shouldn’t be expanded

Value

An augmented data frame with the original variables and new variables that contain the missingness patterns of variables with NA

Examples

```
expand_na(data = example_na, type = c("dbl", "int"))
```

invert — *An S3 method for inverting a step*

Description

An S3 method for inverting a step

Usage

```
invert(object, predictions, ...)
```

Arguments

`object`	A recipe after fitting a model
`predictions`	A data frame with .pred
`...`	Other arguments

Value

A tibble with inverted model-generated values

Examples

```
data <- tibble::tibble(
  y = rlnorm(n = 1000, meanlog = 0, sdlog = 1),
  x = rnorm(n = 1000)
)

adj <- recipes::recipe(y ~ x, data = data) |>
  recipes::step_BoxCox(recipes::all_outcomes()) |>
  recipes::prep()

invert(
  object = adj$steps[[1]],
  predictions = tibble::tibble(.pred = adj[["template"]][["y"]])
)
```

`invert.step_BoxCox` *Invert a Box-Cox transformation*

Description

Invert a Box-Cox transformation

Usage

```
## S3 method for class 'step_BoxCox'
invert(object, predictions, ...)
```

Arguments

`object`	A recipe after fitting a model
`predictions`	A data frame with .pred
`...`	Other arguments

Value

A tibble with the Box-Cox transformation inverted for .pred

Examples

```
data <- tibble::tibble(
  y = rlnorm(n = 1000, meanlog = 0, sdlog = 1),
  x = rnorm(n = 1000)
)

adj <- recipes::recipe(y ~ x, data = data) |>
  recipes::step_BoxCox(recipes::all_outcomes()) |>
  recipes::prep()

invert(
  object = adj$steps[[1]],
  predictions = tibble::tibble(.pred = adj[[”template”]][[”y”]])
)
```

invert.step_log *Invert a log transformation*

Description

Invert a log transformation

Usage

```
## S3 method for class 'step_log'
invert(object, predictions, ...)
```

Arguments

object	A recipe after fitting a model
predictions	A data frame with .pred
...	Other arguments

Value

A tibble with the log transformation inverted for .pred

Examples

```
data <- tibble::tibble(
  y = rlnorm(n = 1000, meanlog = 0, sdlog = 1),
  x = rnorm(n = 1000)
)

adj <- recipes::recipe(y ~ x, data = data) |>
  recipes::step_log(recipes::all_outcomes()) |>
  recipes::prep()

invert(
  object = adj$steps[[1]],
  predictions = tibble::tibble(.pred = adj[["template"]][["y"]])
)
```

invert.step_YeoJohnson

Invert a Yeo-Johnson transformation

Description

Invert a Yeo-Johnson transformation

Usage

```
## S3 method for class 'step_YeoJohnson'
invert(object, predictions, ...)
```

Arguments

object	A recipe after fitting a model
predictions	A data frame with .pred
...	Other arguments

Value

A tibble with the Yeo_johnson transformation inverted for .pred

Examples

```
data <- tibble::tibble(
  y = rlnorm(n = 1000, meanlog = 0, sdlog = 1),
  x = rnorm(n = 1000)
)

adj <- recipes::recipe(y ~ x, data = data) |>
  recipes::step_YeoJohnson(recipes::all_outcomes()) |>
  recipes::prep()

invert(
  object = adj$steps[[1]],
  predictions = tibble::tibble(.pred = adj[["template"]][["y"]])
)
```

ks_distance *Kolmogorov-Smirnov distance*

Description

Kolmogorov-Smirnov distance

Usage

```
ks_distance(data, ...)

## S3 method for class 'data.frame'
ks_distance(data, truth, estimate, na_rm = TRUE, case_weights = NULL, ...)

ks_distance_vec(truth, estimate, na_rm = TRUE, case_weights = NULL, ...)
```

Arguments

`data`	A `data.frame` containing the columns specified by the truth and estimate arguments.
`...`	Not currently used.
`truth`	The column identifier for the true results (that is numeric). This should be an unquoted column name although this argument is passed by expression and supports quasiquotation (you can unquote column names). For `_vec()` functions, a numeric vector.
`estimate`	The column identifier for the predicted results (that is also numeric). As with truth this can be specified different ways but the primary method is to use an unquoted variable name. For `_vec()` functions, a numeric vector.
`na_rm`	A `logical` value indicating whether `NA` values should be stripped before the computation proceeds.
`case_weights`	This is a placeholder for now and will be added when case_weights are added to tidysynthesis.

Value

For `ks_distance_vec()`, a single numeric value (or `NA`).

A single numeric value (or `NA`).

A single numeric value (or `NA`).

Examples

```
ks1 <- data.frame(x = 1:100, y = 101:200)

ks_distance(data = ks1, truth = x, estimate = y)
```

```
ks1 <- data.frame(x = 1:100, y = 101:200)

ks_distance(data = ks1, truth = x, estimate = y)


ks1 <- data.frame(x = 1:100, y = 101:200)

ks_distance_vec(truth = ks1$x, estimate = ks1$y)
```

---

noise *Create a noise object*

Description

Create a noise object

Usage

```
noise(add_noise = FALSE, mode = "regression", noise_func = NULL, ...)
```

Arguments

add_noise	Boolean, TRUE if adding noise
mode	String, one of "regression" or "classification"
noise_func	A function that adds noise to
...	Optional named additional arguments to pass to noise_func(...)

Value

A noise object

Examples

```
# create default noise object
noise()

# create noise object for classification
noise(
  add_noise = TRUE,
  mode = "classification",
  noise_func = add_noise_cat_unif
)

# create noise object for regression
noise(
  add_noise = TRUE,
  mode = "regression",
  noise_func = add_noise_kde,
  n_ntiles = 10
)
```

presynth *Create a presynth object*

Description

Create a presynth object

Usage

```
presynth(roadmap, synth_spec)
```

Arguments

roadmap	A roadmap object from roadmap().
synth_spec	A synth_spec object from synth_spec().

Value

A presynth object.

Examples

```
# create roadmap
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rpart_mod_reg <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "regression")

rpart_mod_class <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "classification")

synth_spec1 <- synth_spec(
  default_regression_model = rpart_mod_reg,
  default_regression_sampler = sample_rpart,
  default_classification_model = rpart_mod_class,
  default_classification_sampler = sample_rpart
)

# create a presynth object
# use defaults for noise, constraints, and replicates
presynth(
  roadmap = rm,
  synth_spec = synth_spec1
)
```

`print.constraints` *Print the constraints object to the console with formatting*

Description

Print the constraints object to the console with formatting

Usage

```
## S3 method for class 'constraints'
print(x, ...)
```

Arguments

`x` A `constraints` object

`...` further arguments passed to or from other methods (not currently used).

Value

Invisibly returns the input `constraints` object.

Examples

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

constraints_df_num <-
  tibble::tribble(~var, ~min, ~max, ~conditions,
                  "transit_time", 0, 300, "TRUE")

constraints <- constraints(
  schema = rm[["schema"]],
  constraints_df_num = constraints_df_num,
  max_z_num = 0
)

print(constraints)
```

print.noise — *Print the noise object to the console with formatting*

Description

Print the noise object to the console with formatting

Usage

```
## S3 method for class 'noise'
print(x, ...)
```

Arguments

x	A noise object
...	further arguments passed to or from other methods (not currently used).

Value

Invisibly returns the input noise object.

Examples

```
print(noise())
```

print.postsynth — *Print the postsynth object to the console with formatting*

Description

Print the postsynth object to the console with formatting

Usage

```
## S3 method for class 'postsynth'
print(x, ...)
```

Arguments

x	A postsynth object
...	further arguments passed to or from other methods (not currently used).

Value

Invisibly returns the input postsynth object.

Examples

```
# create roadmap
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rpart_mod_reg <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "regression")

rpart_mod_class <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "classification")

synth_spec1 <- synth_spec(
  default_regression_model = rpart_mod_reg,
  default_regression_sampler = sample_rpart,
  default_classification_model = rpart_mod_class,
  default_classification_sampler = sample_rpart
)

# create a presynth object
# use defaults for noise, constraints, and replicates
presynth1 <- presynth(
  roadmap = rm,
  synth_spec = synth_spec1
)

# synthesize!
set.seed(1)
postsynth1 <- synthesize(presynth = presynth1)

print(postsynth1)
```

`print.presynth` *print method for presynth objects*

Description

print method for presynth objects

Usage

```
## S3 method for class 'presynth'
print(x, ...)
```

Arguments

`x` A `presynth` object

`...` further arguments passed to or from other methods (not currently used).

Value

A `presynth` object

Examples

```
# create roadmap
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rpart_mod_reg <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "regression")

rpart_mod_class <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "classification")

synth_spec1 <- synth_spec(
  default_regression_model = rpart_mod_reg,
  default_regression_sampler = sample_rpart,
  default_classification_model = rpart_mod_class,
  default_classification_sampler = sample_rpart
)

# create a presynth object
# use defaults for noise, constraints, and replicates
presynth <- presynth(
  roadmap = rm,
  synth_spec = synth_spec1
)

print(presynth)
```

`print.replicates` *Print the replicates object to the console with formatting*

Description

Print the replicates object to the console with formatting

Usage

```
## S3 method for class 'replicates'
print(x, ...)
```

Arguments

x	A replicates object
...	further arguments passed to or from other methods (not currently used).

Value

Invisibly returns the input replicates object.

Examples

```
rep <- replicates(
  start_data_replicates = 2,
  model_sample_replicates = 2,
  end_to_end_replicates = 2
)

print(rep)
```

print.schema *Print the schema object to the console with formatting*

Description

Print the schema object to the console with formatting

Usage

```
## S3 method for class 'schema'
print(x, ...)
```

Arguments

x	A schema object
...	further arguments passed to or from other methods (not currently used).

Value

Invisibly returns the input schema object.

Examples

```
# default inferred schema
schema1 <- schema(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

print(schema1)
```

`print.start_method` *Print the start_method object to the console with formatting*

Description

Print the start_method object to the console with formatting

Usage

```
## S3 method for class 'start_method'
print(x, ...)
```

Arguments

`x`	A `start_method` object
`...`	further arguments passed to or from other methods (not currently used).

Value

A `start_method` object

Examples

```
print(start_method())
```

print.synth_spec *Print the replicates object to the console with formatting*

Description

Print the replicates object to the console with formatting

Usage

```
## S3 method for class 'synth_spec'
print(x, ...)
```

Arguments

x	A replicates object
...	further arguments passed to or from other methods (not currently used).

Value

A synth_spec object

Examples

```
synth_spec <- synth_spec()

print(synth_spec)
```

print.visit_sequence *Print method for* visit_sequence *objects*

Description

Print method for visit_sequence objects

Usage

```
## S3 method for class 'visit_sequence'
print(x, ...)
```

Arguments

x	A visit_sequence object
...	further arguments passed to or from other methods (not currently used).

Value

Invisibly returns the input visit_sequence object.

Examples

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

print(rm[["visit_sequence"]])
```

replicates *Create a replicates object*

Description

Create a replicates object

Usage

```
replicates(
  start_data_replicates = 1,
  model_sample_replicates = 1,
  end_to_end_replicates = 1
)
```

Arguments

start_data_replicates
: The number of starting data replicates to use. Note that if no start_method is provided, all start data replicates will be identical.

model_sample_replicates
: The number of replicates for the conditional modeling process, including modeling and sampling new synthetic values.

end_to_end_replicates
: The number of replicates for the entire synthesis process, including all previously specified steps.

Value

A new replicates object.

Examples

```
replicates(
  start_data_replicates = 2,
  model_sample_replicates = 2,
  end_to_end_replicates = 2
)
```

replicates_api *Add, update, or reset a* `replicates` *object within an existing* `roadmap`*.*

Description

Add, update, or reset a `replicates` object within an existing `roadmap`.

Usage

```
add_replicates(roadmap, replicates)

update_replicates(roadmap, ...)

reset_replicates(roadmap)
```

Arguments

`roadmap`	A `roadmap` object
`replicates`	A `replicates` object.
`...`	Optional named parameters passed to `replicates()`.

Value

A new `roadmap` object.

A new `roadmap` object with the added `replicates`.

A new `roadmap` object with updated `replicates`.

A new `roadmap` object with reset `replicates`.

Examples

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

new_replicates <- replicates(end_to_end_replicates = 2)

rm |>
  add_replicates(new_replicates)
```

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rm |>
  update_replicates(start_data_replicates = 3)


rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rm <- rm |>
  add_replicates(replicates(start_data_replicates = 3))

 reset_replicates(roadmap = rm)
```

roadmap *Create a roadmap*

Description

A roadmap is a container object that aggregates information required to specify the order of operations for synthesis modeling and sampling steps.

Usage

```
roadmap(
  conf_data,
  start_data,
  start_method = NULL,
  schema = NULL,
  visit_sequence = NULL,
  replicates = NULL,
  constraints = NULL
)
```

Arguments

conf_data	A data.frame of confidential data.
start_data	A data.frame of starting data used to initialize the process.
start_method	An optional start_method object.
schema	An optional schema object.

visit_sequence	An optional visit_sequence object.
replicates	An optional replicates object.
constraints	An optional constraints object.

Details

Users initiate a roadmap object with conf_data and start_data. All other objects will either be completed with defaults or specified interactively via the provided API.

Value

A new roadmap object.

Examples

```
roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw,
  start_method = start_method(
    start_func = start_resample, n = 1000
  )
)
```

sample_glm *Sample the conditional distribution created by a generalized linear model*

Description

Currently, logistic and poisson regression are supported using parsnip and the standard glm engine. Note that poisson regression requires the suggested poissonreg library.

Usage

```
sample_glm(model, new_data, conf_data)
```

Arguments

model	A "model_fit" object created by parsnip
new_data	A data frame with predictors
conf_data	A data frame with original confidential predictors

Value

A numeric vector of predictions

Examples

```
acs_conf <- acs_conf |>
  tidyr::drop_na()

logistic_mod <- parsnip::logistic_reg() |>
  parsnip::set_engine("glm") |>
  parsnip::set_mode(mode = "classification")

classification_rec <- recipes::recipe(hcovany ~ ., data = acs_conf)

model_class <- workflows::workflow() |>
  workflows::add_model(spec = logistic_mod) |>
  workflows::add_recipe(recipe = classification_rec) |>
  parsnip::fit(data = acs_conf)

set.seed(1)
sample1 <- sample_glm(
  model = model_class,
  new_data = acs_conf[1:3, ],
  conf_data = acs_conf
)
```

---

sample_lm *Sample the conditional distribution created by a linear model*

Description

Sample the conditional distribution created by a linear model

Usage

```
sample_lm(model, new_data, conf_data)
```

Arguments

model	A "model_fit" object created by parsnip::linear_reg()
new_data	A data frame with predictors
conf_data	A data frame with original confidential predictors

Value

A numeric vector of predictions

Examples

```
lm_mod <- parsnip::linear_reg() |>
  parsnip::set_engine("lm") |>
  parsnip::set_mode(mode = "regression")

regression_rec <- recipes::recipe(inctot ~ ., data = acs_conf)

model_reg <- workflows::workflow() |>
  workflows::add_model(spec = lm_mod) |>
  workflows::add_recipe(recipe = regression_rec) |>
  parsnip::fit(data = acs_conf)

set.seed(1)
sample1 <- sample_lm(
  model = model_reg,
  new_data = acs_conf[1:3, ],
  conf_data = acs_conf
)
```

sample_ranger *Sample the conditional distribution created by a ranger rf model*

Description

Sample the conditional distribution created by a ranger rf model

Usage

```
sample_ranger(model, new_data, conf_data)
```

Arguments

model	A "model_fit" object created by parsnip::ranger()
new_data	A data frame with predictors
conf_data	A data frame with original confidential predictors

Value

A numeric vector of predictions

Examples

```
rf_mod_regression <- parsnip::rand_forest(trees = 500, min_n = 1) |>
  parsnip::set_engine(engine = "ranger") |>
  parsnip::set_mode(mode = "regression") |>
  parsnip::set_args(quantreg = TRUE)
```

```
regression_rec <- recipes::recipe(age ~ ., data = acs_conf)

model_reg <- workflows::workflow() |>
  workflows::add_model(spec = rf_mod_regression) |>
  workflows::add_recipe(recipe = regression_rec) |>
  parsnip::fit(data = acs_conf)

set.seed(1)
sample1 <- sample_ranger(
  model = model_reg,
  new_data = acs_conf[1:3, ],
  conf_data = acs_conf
)
```

`sample_rpart` *Sample the conditional distribution created by a CART model*

Description

Sample the conditional distribution created by a CART model

Usage

```
sample_rpart(model, new_data, conf_data, ignore_zeros = TRUE)
```

Arguments

`model`	A "model_fit" object created by rpart
`new_data`	A data frame with predictors
`conf_data`	A data frame with original confidential predictors
`ignore_zeros`	Should a vector of all 0 observations return NA for the l-diversity calculation. Defaults to `TRUE`.

Value

A numeric vector of predictions

Examples

```
rpart_mod_reg <- parsnip::decision_tree() |>
  parsnip::set_engine("rpart") |>
  parsnip::set_mode(mode = "regression")

 regression_rec <- recipes::recipe(inctot ~ ., data = acs_conf)

 model_reg <- workflows::workflow() |>
   workflows::add_model(spec = rpart_mod_reg) |>
   workflows::add_recipe(recipe = regression_rec) |>
   parsnip::fit(data = acs_conf)

 set.seed(1)
 sample1 <- sample_rpart(
   model = model_reg,
   new_data = acs_conf[1:3, ],
   conf_data = acs_conf
 )


rpart_mod_class <- parsnip::decision_tree() |>
  parsnip::set_engine("rpart") |>
  parsnip::set_mode(mode = "classification")

classification_rec <- recipes::recipe(hcovany ~ ., data = acs_conf)

model_reg <- workflows::workflow() |>
  workflows::add_model(spec = rpart_mod_class) |>
  workflows::add_recipe(recipe = classification_rec) |>
  parsnip::fit(data = acs_conf)

set.seed(1)
sample1 <- sample_rpart(
  model = model_reg,
  new_data = acs_conf[1:10, ],
  conf_data = acs_conf
)
```

schema *Generate a* schema *object.*

Description

Generate a schema object.

Usage

```
schema(
  conf_data,
  start_data,
  col_schema = NULL,
  enforce = TRUE,
  coerce_to_factors = FALSE,
  coerce_to_doubles = FALSE,
  na_factor_to_level = TRUE,
  na_numeric_to_ind = TRUE
)
```

Arguments

conf_data
: A data frame to be synthesized.

start_data
: A data frame with starting variables.

col_schema
: An optional named list of columns in the confidential data with their properties, including data type and factor levels. If NULL or only partially specified, col_schema will be inferred from the confidential data. See example code for formatting.

enforce
: Boolean that if true, will preprocess both conf_data and start_data to align with col_schema and the arguments below.

coerce_to_factors
: Boolean that if true, coerces categorical data types (chr, fct, ord) to base R factors when enforce_schema is called.

coerce_to_doubles
: Boolean that if true, coerces columns specified as dbl in col_schema to base R doubles when enforce_schema is called.

na_factor_to_level
: Boolean that if true, applies convert_level_to_na() to factor variables when enforce_schema is called.

na_numeric_to_ind
: Boolean that if true, applies expand_na() to numeric data to create logical missingness indicators when enforce_schema is called.

Value

A schema object.

Examples

```
conf_data <- data.frame(
  var1 = c("1", "1", "2"),
  var2 = c(1L, 2L, 3L),
  var3 = c(1.1, 2.2, 3.3)
)

start_data <- dplyr::select(conf_data, var1)

# default inferred schema
schema(
  conf_data = conf_data,
  start_data = start_data
)

# overwriting factor levels
schema(
  conf_data = conf_data,
  start_data = start_data,
  col_schema = list(
    "var1" = list(
      "dtype" = "fct",
      "levels" = c("1", "2", "3")
    )
  ),
  coerce_to_factors = TRUE
)
```

schema_api *Add, update, or reset a* schema *object within an existing* roadmap.

Description

Add, update, or reset a schema object within an existing roadmap.

Usage

```
add_schema(roadmap, schema)

update_schema(roadmap, ...)

reset_schema(roadmap)
```

Arguments

roadmap	A roadmap object
schema	A schema object.
...	Optional named parameters passed to schema().

Value

A new roadmap object.

A roadmap object with added schema.

A roadmap object with updated schema.

A roadmap object with reset schema.

Examples

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

acs_schema <- schema(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw,
  na_numeric_to_ind = TRUE
)

rm |>
  add_schema(schema = acs_schema)


rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rm |>
  update_schema(na_numeric_to_ind = TRUE)


rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rm <- rm |>
  update_schema(na_numeric_to_ind = TRUE)

reset_schema(roadmap = rm)
```

---

start_method *Create a* start_method *object.*

Description

A start_method gets executed prior to running a synthesis. This modifies the start_data, typically randomly, to provide greater disclosure risk protections.

Usage

```
start_method(start_func = NULL, ...)
```

Arguments

start_func	A function that accepts and returns a data.frame. If none provided .identity_start() is used.
...	Optional keyword arguments passed to start_func(...)

Value

A start_method object

Examples

```
# basic usage
start_method(start_func = start_resample)

# adjust the number of observations
start_method(
  start_func = start_resample,
  start_data = acs_start_nw,
  n = 10
)

# adjust the number of observations and use all combinations as support
start_method(
  start_func = start_resample,
  start_data = acs_start_nw,
  n = 10,
  inv_noise_scale = 1,
  support = "all"
)
```

start_method_api *Add, update, or reset a start method within an existing* roadmap.

Description

Add, update, or reset a start method within an existing roadmap.

Usage

```
add_start_method(roadmap, start_method)

update_start_method(roadmap, ...)

remove_start_method(roadmap)
```

Arguments

roadmap	A roadmap object
start_method	A start_method object.
...	Optional named parameters passed to start_method()

Value

A new roadmap object.

A new roadmap object with added start_method.

A new roadmap object with updated start_method.

A new roadmap object with removed start_method.

Examples

```
rm <- roadmap(
 conf_data = acs_conf_nw,
 start_data = acs_start_nw,
)

add_start_method(
  roadmap = rm,
  start_method = start_method()
)


rm <- roadmap(
 conf_data = acs_conf_nw,
 start_data = acs_start_nw
)

update_start_method(
  roadmap = rm,
  start_method = start_method()
)

rm <- roadmap(
 conf_data = acs_conf_nw,
 start_data = acs_start_nw,
 start_method = start_method()
)

remove_start_method(
  roadmap = rm
)
```

start_resample *Specify a resampling scheme for start_data*

Description

Specify a resampling scheme for start_data

Usage

```
start_resample(
  start_data,
  n = NULL,
  inv_noise_scale = NULL,
  support = c("observed", "all")
)
```

Arguments

`start_data` A `data.frame`

`n` An optional integer sample size. If unspecified, `n` = `nrow(start_data)`

`inv_noise_scale`
An optional parameter to set randomized noise to the proportions of records with different `start_data` characteristics. Corresponds to a privacy loss budget under epsilon differential privacy.

`support` A string that specifies the method of resampling from the `start_data` domain.

Value

A `start_method` object for resampling starting data

Examples

```
start_method(
  start_func = start_resample, n = 1000
)
```

`synthesize` *Synthesize a data set*

Description

Synthesize a data set

Usage

```
synthesize(presynth, progress = FALSE)
```

Arguments

`presynth` A `presynth` object created by `presynth()`.

`progress` A single logical. Should a progress be displayed?

Value

A postsynth object.

Examples

```
# create roadmap
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rpart_mod_reg <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "regression")

rpart_mod_class <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "classification")

synth_spec1 <- synth_spec(
  default_regression_model = rpart_mod_reg,
  default_regression_sampler = sample_rpart,
  default_classification_model = rpart_mod_class,
  default_classification_sampler = sample_rpart
)

# create a presynth object
# use defaults for noise, constraints, and replicates
presynth1 <- presynth(
  roadmap = rm,
  synth_spec = synth_spec1
)

# synthesize!
set.seed(1)
postsynth1 <- synthesize(presynth = presynth1)
```

---

synth_spec *Create a* synth_spec *object*

Description

The synth_spec object holds specifications for modeling and sampling components for sequential synthetic data generation. Each component has an associated construct_* function called when creating a presynth object.

Usage

```
synth_spec(
  default_regression_model = NULL,
  default_classification_model = NULL,
  custom_models = NULL,
  default_regression_steps = NULL,
  default_classification_steps = NULL,
  custom_steps = NULL,
  default_regression_sampler = NULL,
  default_classification_sampler = NULL,
  custom_samplers = NULL,
  default_regression_noise = NULL,
  default_classification_noise = NULL,
  custom_noise = NULL,
  default_regression_tuner = NULL,
  default_classification_tuner = NULL,
  custom_tuners = NULL,
  default_extractor = NULL,
  custom_extractors = NULL,
  invert_transformations = TRUE,
  enforce_na = TRUE
)
```

Arguments

`default_regression_model`
A `model_spec` object from `library(parsnip)` for use in regression models.

`default_classification_model`
A `model_spec` object from `library(parsnip)` for use in classification models.

`custom_models`
A list of named lists each with two elements: `vars` for variable names, and `model` for their associated model. from `library(parsnip)`.

`default_regression_steps`
A list of `recipe::step_` function(s) from `library(recipes)` for use in regression models.

`default_classification_steps`
A list of `recipe::step_` function(s) from `library(recipes)` for use in classification models.

`custom_steps`
A list of named lists each with two elements: `vars` for variable names, and `steps` for their associated recipe.

`default_regression_sampler`
A sampling function for drawing new values from regression models.

`default_classification_sampler`
A sampling function for drawing new values from classification models.

`custom_samplers`
A list of named lists each with two elements: `vars` for variable names, and `sampler` for their associated sampler

`default_regression_noise`
A noise function for adding noise to numeric values.

`default_classification_noise`
A noise function for adding noise to classification values.

`custom_noise`
A list of named lists each with two elements: `vars` for variable names, and `noise` for their associated noise

`default_regression_tuner`
A `tuner` from `library(tune)` for use in regression models.

`default_classification_tuner`
A `tuner` from `library(tune)` for use in classification models.

`custom_tuners` A list of named lists each with two elements: `vars` for variable names, and `tuner` for their associated tuner

`default_extractor`
An optional method for extracting workflows or extracts from workflows.

`custom_extractors`
A list of named lists each with two elements: `vars` for variable names, and `extractor` for their associated extractor

`invert_transformations`
A Boolean for if outcome variable transformations applied through recipes should be inverted during synthesis. recipes need ids that begin with "outcome".

`enforce_na` A Boolean for if NA values should be added into the synthetic data with `enforce_na()` during synthesis. An alternative approach is to add the NA values after synthesis

Value

A `synth_spec` object

Examples

```
rpart_mod <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "regression")

lm_mod <- parsnip::linear_reg() |>
  parsnip::set_engine("lm") |>
  parsnip::set_mode(mode = "regression")

step1 <- function(x) {
 x |>
   recipes::step_center(recipes::all_predictors(), id = "center")
}

step2 <- function(x) {
  x |>
    recipes::step_scale(recipes::all_predictors(), id = "scale")
}

step3 <- function(x) { x |> step1() |> step2() }


synth_spec(
 default_regression_model = rpart_mod,
 custom_models = list(
   list("vars" = c("var1", "var2"),
        "model" = lm_mod)
 ),
 default_regression_steps = step1,
 custom_steps = list(
   list("vars" = c("var2", "var3"),
        "steps" = step2),
   list("vars" = c("var4"),
        "steps" = step3)
 ),
 default_regression_sampler = sample_rpart,
 custom_samplers = list(
   list("vars" = c("var1", "var2"),
        "sampler" = sample_lm)
 )
)
```

---

synth_spec_extractor_api

Add, update, or remove extractors from a `synth_spec` *object*

Description

Add, update, or remove extractors from a `synth_spec` object

Usage

```
add_custom_extractors(synth_spec, ...)

update_custom_extractors(synth_spec, ...)

remove_custom_extractors(synth_spec)
```

Arguments

`synth_spec`	A `synth_spec` object
`...`	Optional named lists with two elements, `vars` and `extractor`, mapping variable names to extractors.

Value

A new `synth_spec` object.

A new `synth_spec` object with added custom extractors.

A new `synth_spec` object with updated custom extractors.

A new `synth_spec` object with removed custom extractors.

Examples

```
synth_spec <- synth_spec()

add_custom_extractors(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "extractor" = parsnip::extract_fit_engine)
)


synth_spec <- synth_spec()

update_custom_extractors(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "extractor" = parsnip::extract_fit_engine)
)


synth_spec <- synth_spec()

synth_spec <- add_custom_extractors(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "extractor" = parsnip::extract_fit_engine)
)

remove_custom_extractors(synth_spec = synth_spec)
```

`synth_spec_is_component`

Inspections for `synth_spec` *components*

Description

Inspections for `synth_spec` components

Arguments

`z`	Object

Value

Boolean if matches class type

synth_spec_model_api *Add, update, or remove custom models from a synth_spec object*

Description

Add, update, or remove custom models from a synth_spec object

Usage

```
add_custom_models(synth_spec, ...)

update_custom_models(synth_spec, ...)

remove_custom_models(synth_spec)
```

Arguments

synth_spec	A synth_spec object
...	Optional named lists with two elements, vars and model, mapping variable names to model_spec objects from library(parsnip).

Value

A new synth_spec object.

A new synth_spec object with added custom models.

A new synth_spec object with updated custom models.

A new synth_spec object with removed custom models.

Examples

```
synth_spec <- synth_spec()

dt_reg_mod <- parsnip::decision_tree() |>
  parsnip::set_engine("rpart") |>
  parsnip::set_mode("regression")

add_custom_models(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "model" = dt_reg_mod)
)


synth_spec <- synth_spec()

dt_reg_mod <- parsnip::decision_tree() |>
  parsnip::set_engine("rpart") |>
  parsnip::set_mode("regression")
```

```
update_custom_models(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "model" = dt_reg_mod)
)


synth_spec <- synth_spec()

dt_reg_mod <- parsnip::decision_tree() |>
  parsnip::set_engine("rpart") |>
  parsnip::set_mode("regression")

synth_spec <- update_custom_models(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "model" = dt_reg_mod)
)

remove_custom_models(synth_spec = synth_spec)
```

`synth_spec_noise_api` *Add, update, or remove noise from a* `synth_spec` *object*

Description

Add, update, or remove noise from a `synth_spec` object

Usage

```
add_custom_noise(synth_spec, ...)
```

Arguments

`synth_spec`	A `synth_spec` object
`...`	Optional named lists with two elements, `vars` and `noise`, mapping variable names to samplers.

Value

A new `synth_spec` object.

A new `synth_spec` object with added custom noise.

Examples

```
synth_spec <- synth_spec()

noise1 <- noise(
  add_noise = TRUE,
  noise_func = add_noise_kde,
  noise_params = list(
    n_ntiles = 2
  )
)

add_custom_noise(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "noise" = noise1)
)
```

---

synth_spec_recipes_api

Add, update, or remove recipe recipes from a `synth_spec` *object*

Description

Add, update, or remove recipe recipes from a `synth_spec` object

Usage

```
add_custom_steps(synth_spec, ...)

update_custom_steps(synth_spec, ...)

remove_custom_steps(synth_spec)
```

Arguments

`synth_spec`	A `synth_spec` object
`...`	Optional named arguments mapping variables to lists of `recipe::recipe_` function(s) from `library(recipes)`.

Value

A new `synth_spec` object.

A new `synth_spec` object with added custom steps.

A new `synth_spec` object with updated custom steps.

A new `synth_spec` object with removed custom steps.

Examples

```
synth_spec <- synth_spec()

step1 <- function(x) {
  x |> recipes::step_center(recipes::all_predictors(), id = "center")
}

add_custom_steps(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "steps" = step1)
)


synth_spec <- synth_spec()

step1 <- function(x) {
  x |> recipes::step_center(recipes::all_predictors(), id = "center")
}

update_custom_steps(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "steps" = step1)
)


synth_spec <- synth_spec()

step1 <- function(x) {
  x |> recipes::step_center(recipes::all_predictors(), id = "center")
}

synth_spec <- add_custom_steps(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "steps" = step1)
)

remove_custom_steps(synth_spec = synth_spec)
```

---

synth_spec_sampler_api *Add, update, or remove samplers from a* synth_spec *object*

Description

Add, update, or remove samplers from a synth_spec object

Usage

```
add_custom_samplers(synth_spec, ...)

update_custom_samplers(synth_spec, ...)

remove_custom_samplers(synth_spec)

update_custom_noise(synth_spec, ...)

remove_custom_noise(synth_spec)
```

Arguments

synth_spec	A synth_spec object
...	Optional named lists with two elements, vars and sampler, mapping variable names to samplers.

Value

A new synth_spec object.

A new synth_spec object with added custom samplers.

A new synth_spec object with updated custom samplers.

A new synth_spec object with removed custom samplers.

A new synth_spec object with updated custom noise.

A new synth_spec object with removed custom noise.

Examples

```
synth_spec <- synth_spec()

add_custom_samplers(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "sampler" = sample_rpart)
)


synth_spec <- synth_spec()

update_custom_samplers(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "sampler" = sample_rpart)
)


synth_spec <- synth_spec()

synth_spec <- add_custom_samplers(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "sampler" = sample_rpart)
)

remove_custom_samplers(synth_spec = synth_spec)


synth_spec <- synth_spec()

noise1 <- noise(
  add_noise = TRUE,
  noise_func = add_noise_kde,
  noise_params = list(
    n_ntiles = 2
  )
)

update_custom_noise(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "noise" = noise1)
)


synth_spec <- synth_spec()

noise1 <- noise(
  add_noise = TRUE,
  noise_func = add_noise_kde,
  noise_params = list(
    n_ntiles = 2
  )
)

synth_spec <- add_custom_noise(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "noise" = noise1)
)

remove_custom_noise(synth_spec = synth_spec)
```

synth_spec_tuner_api *Add, update, or remove tuners from a* synth_spec *object*

Description

Add, update, or remove tuners from a synth_spec object

Usage

```
add_custom_tuners(synth_spec, ...)
```

```
update_custom_tuners(synth_spec, ...)

remove_custom_tuners(synth_spec)
```

Arguments

`synth_spec` A `synth_spec` object

`...` Optional named lists with two elements, `vars` and `tuner`, mapping variable names to tuners.

Value

A new `synth_spec` object.

A new `synth_spec` object with added custom tuners.

A new `synth_spec` object with updated custom tuners.

A new `synth_spec` object with removed custom tuners.

Examples

```
synth_spec <- synth_spec()

tuner1 <- list(
  v = 3,
  grid = 3,
  metrics = yardstick::metric_set(yardstick::rmse)
)

add_custom_tuners(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "tuner" = tuner1)
)


synth_spec <- synth_spec()

tuner1 <- list(
  v = 3,
  grid = 3,
  metrics = yardstick::metric_set(yardstick::rmse)
)

update_custom_tuners(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "tuner" = tuner1)
)


synth_spec <- synth_spec()
```

```
tuner1 <- list(
  v = 3,
  grid = 3,
  metrics = yardstick::metric_set(yardstick::rmse)
)

synth_spec <- add_custom_tuners(
  synth_spec = synth_spec,
  list("vars" = c("a", "b", "c"), "tuner" = tuner1)
)

remove_custom_tuners(synth_spec = synth_spec)
```

---

tune_synthesis *Generate syntheses from multiple* presynth *objects.*

Description

Generate syntheses from multiple presynth objects.

Usage

```
tune_synthesis(
  presynths,
  postprocessing_func,
  metadata_func = NULL,
  simplify_post = FALSE,
  seed = NULL
)
```

Arguments

presynths A list of presynth objects

postprocessing_func
A function with arguments "synth_id", "synth_name", and "postsynth" that performs any desired postprocessing operations, like writing

metadata_func An optional function with argument "presynth" that extracts specified information from each presynth object and returns a list. Each list element becomes an additional column in the output metadata.

simplify_post Boolean that, if true, expects postprocessing_func to return a list corresponding to the row of the output dataframe (one per synthesis).

seed A RNG seed to pass to set.seed()

Value

A post_tunesynth object.

Examples

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

dt_mod_reg <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "regression")

dt_mod_reg_cp <- parsnip::decision_tree(cost_complexity = 0.01) |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "regression")

dt_mod_class <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "classification")

# synth specs
synth_spec1 <- synth_spec(
  default_regression_model = dt_mod_reg,
  default_regression_sampler = sample_rpart,
  default_classification_model = dt_mod_class,
  default_classification_sampler = sample_rpart
)

synth_spec2 <- synth_spec(
  default_regression_model = dt_mod_reg_cp,
  default_regression_sampler = sample_rpart,
  default_classification_model = dt_mod_class,
  default_classification_sampler = sample_rpart
)


presynth1 <- presynth(
  roadmap = rm,
  synth_spec = synth_spec1
)

presynth2 <- presynth(
  roadmap = rm,
  synth_spec = synth_spec2
)

postproc_f_null <- function(synth_id, synth_name, postsynth) {
  return(postsynth[["synthetic_data"]])
}

tune_synthesis(
  presynths = list(presynth1, presynth2),
  postprocessing_func = postproc_f_null,
  seed = 12345
)
```

update_presynth *Update* presynth *object*

Description

Update presynth object

Usage

```
update_presynth(presynth, roadmap = NULL, synth_spec = NULL)
```

Arguments

presynth	A presynth object
roadmap	An optional roadmap object
synth_spec	An optional synth_spec object

Value

A presynth object.

Examples

```
# create roadmap
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rpart_mod_reg <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "regression")

rpart_mod_class <- parsnip::decision_tree() |>
  parsnip::set_engine(engine = "rpart") |>
  parsnip::set_mode(mode = "classification")

synth_spec1 <- synth_spec(
  default_regression_model = rpart_mod_reg,
  default_regression_sampler = sample_rpart,
  default_classification_model = rpart_mod_class,
  default_classification_sampler = sample_rpart
)
```

```
# create a presynth object
# use defaults for noise, constraints, and replicates
presynth <- presynth(
  roadmap = rm,
  synth_spec = synth_spec1
)

lm_mod <- parsnip::linear_reg() |>
  parsnip::set_engine(engine = "lm") |>
  parsnip::set_mode(mode = "regression")

synth_spec2 <- synth_spec(
  default_regression_model = lm_mod,
  default_regression_sampler = sample_lm,
  default_classification_model = rpart_mod_class,
  default_classification_sampler = sample_rpart
)
```

update_synth_spec *Tidy API calls* ————————————————————-

Description

Update non-custom synth_spec arguments

Usage

```
update_synth_spec(synth_spec, ...)
```

Arguments

synth_spec	A synth_spec object
...	Optional named keywords in synth_spec, with the exception of any custom_*arguments

Value

A synth_spec

Examples

```
synth_spec <- synth_spec()

lm_mod <- parsnip::linear_reg() |>
  parsnip::set_engine("lm") |>
  parsnip::set_mode(mode = "regression")

update_synth_spec(
  synth_spec,
  default_regression_model = lm_mod
)
```

visit_sequence *Generate a visit sequence.*

Description

Generate a visit sequence.

Usage

```
visit_sequence(schema, weight_var = NULL, synthesize_weight = TRUE)
```

Arguments

schema	A schema object.
weight_var	A numeric weight for the weighted total ordering.
synthesize_weight	Boolean for if weight_var should be included in the visit sequence.

Value

A `visit_sequence` object.

Examples

```
df <- data.frame(
  factor_var = c("1", "1", "2"),
  vara = c(10000, 20000, 100000),
  varb = c(300, 200, 100),
  var_loss = c(1999999, 0, -1000000),
  weight = c(1000, 1000, 2000)
)

start_df <- dplyr::select(df, factor_var)

schema1 <- schema(
  conf_data = dplyr::select(df, -weight),
  start_data = start_df
)

vs1 <- visit_sequence(
  schema = schema1
)

schema2 <- schema(
  conf_data = df,
  start_data = start_df
)

vs2 <- visit_sequence(
  schema = schema2,
  weight_var = weight,
  synthesize_weight = TRUE
)
```

visit_sequence_api *Add or reset a* visit_sequence *object within an existing* roadmap.

Description

Add or reset a visit_sequence object within an existing roadmap.

Usage

```
update_visit_sequence(roadmap, ...)

reset_visit_sequence(roadmap)
```

Arguments

roadmap	A roadmap object
...	Optional additional parameters.

Value

A new roadmap object.

A roadmap with an updated visit_sequence.

A new roadmap object with reset visit_sequence.

Examples

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rm |>
  update_visit_sequence(
    weight_var = wgt,
    synthesize_weight = TRUE
  )
```

```
rm <- roadmap(
  conf_data = acs_conf_nw,
  start_data = acs_start_nw
)

rm <- rm |>
  update_visit_sequence(
    weight_var = wgt,
    synthesize_weight = TRUE
  )

reset_visit_sequence(roadmap = rm)
```

Index