# Package 'tidyrules'

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**Type** Package

**Title** Utilities to Retrieve Rulelists from Model Fits, Filter, Prune,
Reorder and Predict on Unseen Data

**Version** 0.2.7

**Maintainer** Srikanth Komala Sheshachala `<sri.teach@gmail.com>`

**Depends** R (>= 3.6.0),

**Imports** stringr (>= 1.3.1), magrittr (>= 1.5), purrr (>= 0.3.2),
partykit (>= 1.2.2), rlang (>= 1.1.3), generics (>= 0.1.3),
checkmate (>= 2.3.1), tidytable (>= 0.11.0), data.table (>=
1.14.6), DescTools (>= 0.99.54), MetricsWeighted (>= 1.0.3),
cli (>= 3.6.2), glue (>= 1.7.0), pheatmap (>= 1.0.12), proxy
(>= 0.4.27), tibble (>= 3.2.1),

**Suggests** AmesHousing (>= 0.0.3), dplyr (>= 0.8), C50 (>= 0.1.2),
Cubist (>= 0.2.2), rpart (>= 1.2.2), rpart.plot (>= 3.0.7),
modeldata (>= 0.0.1), testthat (>= 2.0.1), MASS (>= 7.3.50),
mlbench (>= 2.1.1), rmarkdown (>= 1.13), palmerpenguins (>=
0.1.1),

**Description** Provides a framework to work with decision rules. Rules can be extracted from sup-
ported models, augmented with (custom) metrics using validation data, manipulated using stan-
dard dataframe operations, reordered and pruned based on a metric, predict on un-
seen (test) data. Utilities include; Creating a rulelist manually, Export-
ing a rulelist as a SQL case statement and so on. The package of-
fers two classes; rulelist and ruleset based on dataframe.

**URL** https://github.com/talegari/tidyrules,

https://talegari.github.io/tidyrules/

**BugReports** https://github.com/talegari/tidyrules/issues

**License** GPL-3

**Encoding** UTF-8

**RoxygenNote** 7.3.1

**NeedsCompilation** no

# Contents

---

as_rulelist *as_rulelist generic from [tidyrules](#) package*

---

### Description

as_rulelist generic

### Usage

```
as_rulelist(x, ...)
```

### Arguments

| | |
|---|---|
| x | object to be coerced to a [rulelist](#) |
| ... | for methods to use |

### Value

A [rulelist](#)

### See Also

[rulelist](#), [tidy](#), [augment](#), [predict](#), [calculate](#), [prune](#), [reorder](#)

---

as_rulelist.data.frame

*as_rulelist method for a data.frame*

---

### Description

Convert a set of rules in a dataframe to a [rulelist](#)

### Usage

```
## S3 method for class 'data.frame'
as_rulelist(x, keys = NULL, model_type = NULL, estimation_type, ...)
```

### Arguments

| | |
|---|---|
| x | dataframe to be coerced to a [rulelist](#) |
| keys | (character vector, default: NULL) column names which form the key |
| model_type | (string, default: NULL) Name of the model which generated the rules |
| estimation_type | |
| | (string) One among: 'regression', 'classification' |
| ... | currently unused |

## Details

Input dataframe should contain these columns: rule_nbr, LHS, RHS. Providing other inputs helps augment better.

## Value

[rulelist](#) object

## See Also

[rulelist](#), [tidy](#), [augment](#), [predict](#), [calculate](#), [prune](#), [reorder](#)

## Examples

```
rules_df = tidytable::tidytable(rule_nbr = 1:2,
                                LHS      = c("var_1 > 50", "var_2 < 30"),
                                RHS      = c(2, 1)
                                )
as_rulelist(rules_df, estimation_type = "regression")
```

---

as_ruleset                  *Get a ruleset from a rulelist*

---

## Description

Returns a ruleset object

## Usage

```
as_ruleset(rulelist)
```

## Arguments

rulelist          A [rulelist](#)

## Value

A [ruleset](#)

## See Also

[rulelist](#)

## Examples

```
model_class_party = partykit::ctree(species ~ .,
                                    data = palmerpenguins::penguins
                                    )
as_ruleset(tidy(model_class_party))
```

---

augment                 augment *is re-export of [generics::augment](#) from [tidyrules](#) package*

---

### Description

See [augment.rulelist](#)

### Usage

```
augment(x, ...)
```

### Arguments

| | |
|---|---|
| x | A [rulelist](#) |
| ... | For methods to use |

### See Also

[rulelist](#), [tidy](#), [augment](#), [predict](#), [calculate](#), [prune](#), [reorder](#)

---

augment.rulelist         *Augment a [rulelist](#)*

---

### Description

augment outputs a [rulelist](#) with an additional column named augmented_stats based on summary statistics calculated using attribute validation_data.

### Usage

```
## S3 method for class 'rulelist'
augment(x, ...)
```

### Arguments

| | |
|---|---|
| x | A [rulelist](#) |
| ... | (expressions) To be send to [tidytable::summarise](#) for custom aggregations. See examples. |

### Details

The dataframe-column augmented_stats will have these columns corresponding to the estimation_type:

- For regression: support, IQR, RMSE
- For classification: support, confidence, lift

along with custom aggregations.

**Value**

A [rulelist](#) with a new dataframe-column named augmented_stats.

**See Also**

[rulelist](#), [tidy](#), [augment](#), [predict](#), [calculate](#), [prune](#), [reorder](#)

**Examples**

```
# Examples for augment ----------------------------------------------------
library("magrittr")

# C5 ----
att = modeldata::attrition
set.seed(100)
train_index = sample(c(TRUE, FALSE), nrow(att), replace = TRUE)

model_c5 = C50::C5.0(Attrition ~., data = att[train_index, ], rules = TRUE)
tidy_c5  =
  model_c5 %>%
  tidy() %>%
  set_validation_data(att[!train_index, ], "Attrition")

tidy_c5

augment(tidy_c5) %>%
  tidytable::unnest(augmented_stats, names_sep = "__") %>%
  tidytable::glimpse()

# augment with custom aggregator
augment(tidy_c5,output_counts = list(table(Attrition))) %>%
  tidytable::unnest(augmented_stats, names_sep = "__") %>%
  tidytable::glimpse()

# rpart ----
set.seed(100)
train_index = sample(c(TRUE, FALSE), nrow(iris), replace = TRUE)

model_class_rpart = rpart::rpart(Species ~ ., data = iris[train_index, ])
tidy_class_rpart  = tidy(model_class_rpart) %>%
  set_validation_data(iris[!train_index, ], "Species")
tidy_class_rpart

model_regr_rpart = rpart::rpart(Sepal.Length ~ ., data = iris[train_index, ])
tidy_regr_rpart  = tidy(model_regr_rpart) %>%
  set_validation_data(iris[!train_index, ], "Sepal.Length")
tidy_regr_rpart

# augment (classification case)
augment(tidy_class_rpart) %>%
  tidytable::unnest(augmented_stats, names_sep = "__") %>%
  tidytable::glimpse()
```

```
# augment (regression case)
augment(tidy_regr_rpart) %>%
  tidytable::unnest(augmented_stats, names_sep = "__") %>%
  tidytable::glimpse()

# party ----
pen = palmerpenguins::penguins %>%
  tidytable::drop_na(bill_length_mm)
set.seed(100)
train_index = sample(c(TRUE, FALSE), nrow(pen), replace = TRUE)

model_class_party = partykit::ctree(species ~ ., data = pen[train_index, ])
tidy_class_party  = tidy(model_class_party) %>%
  set_validation_data(pen[!train_index, ], "species")
tidy_class_party

model_regr_party =
  partykit::ctree(bill_length_mm ~ ., data = pen[train_index, ])
tidy_regr_party  = tidy(model_regr_party) %>%
  set_validation_data(pen[!train_index, ], "bill_length_mm")
tidy_regr_party

# augment (classification case)
augment(tidy_class_party) %>%
  tidytable::unnest(augmented_stats, names_sep = "__") %>%
  tidytable::glimpse()

# augment (regression case)
augment(tidy_regr_party) %>%
  tidytable::unnest(augmented_stats, names_sep = "__") %>%
  tidytable::glimpse()

# cubist ----
att       = modeldata::attrition
set.seed(100)
train_index = sample(c(TRUE, FALSE), nrow(att), replace = TRUE)
cols_att    = setdiff(colnames(att), c("MonthlyIncome", "Attrition"))

model_cubist = Cubist::cubist(x = att[train_index, cols_att],
                              y = att[train_index, "MonthlyIncome"]
                              )

tidy_cubist = tidy(model_cubist) %>%
  set_validation_data(att[!train_index, ], "MonthlyIncome")
tidy_cubist

augment(tidy_cubist) %>%
  tidytable::unnest(augmented_stats, names_sep = "__") %>%
  tidytable::glimpse()
```

---

| calculate | calculate *is re-export of [generics::calculate](#) from [tidyrules](#) package* |

---

### Description

See [calculate.rulelist](#)

### Usage

```
calculate(x, ...)
```

### Arguments

| x | A [rulelist](#) |
|---|---|
| ... | See [calculate.rulelist](#) |

### See Also

[rulelist,](#) [tidy,](#) [augment,](#) [predict,](#) [calculate,](#) [prune,](#) [reorder](#)

---

| calculate.rulelist | calculate *metrics for a [rulelist](#)* |

---

### Description

Computes some metrics (based on estimation_type) in cumulative window function style over the rulelist (in the same order) ignoring the keys.

### Usage

```
## S3 method for class 'rulelist'
calculate(x, metrics_to_exclude = NULL, ...)
```

### Arguments

| x | A [rulelist](#) |
|---|---|
| metrics_to_exclude | |
| | (character vector) Names of metrics to exclude |
| ... | Named list of custom metrics. See 'details'. |

## Details

### Default Metrics:

These metrics are calculated by default:

- cumulative_coverage: For nth rule in the rulelist, number of distinct `row_nbrs` (of `new_data`) covered by nth and all preceding rules (in order). In weighted case, we sum the weights corresponding to the distinct `row_nbrs`.
- cumulative_overlap: Up til nth rule in the rulelist, number of distinct `row_nbrs` (of `new_data`) already covered by some preceding rule (in order). In weighted case, we sum the weights corresponding to the distinct `row_nbrs`.

For classification:

- cumulative_accuracy: For nth rule in the rulelist, fraction of `row_nbrs` such that RHS matches the `y_name` column (of `new_data`) by nth and all preceding rules (in order). In weighted case, weighted accuracy is computed.

For regression:

- cumulative_RMSE: For nth rule in the rulelist, weighted RMSE of all predictions (RHS) predicted by nth rule and all preceding rules.

### Custom metrics:

Custom metrics to be computed should be passed a named list of function(s) in `....` The custom metric function should take these arguments in same order: `rulelist`, `new_data`, `y_name`, `weight`. The custom metric function should return a numeric vector of same length as the number of rows of rulelist.

## Value

A dataframe of metrics with a `rule_nbr` column.

## See Also

rulelist, tidy, augment, predict, calculate, prune, reorder

## Examples

```
library("magrittr")
model_c5  = C50::C5.0(Attrition ~., data = modeldata::attrition, rules = TRUE)
tidy_c5   = tidy(model_c5) %>%
            set_validation_data(modeldata::attrition, "Attrition") %>%
            set_keys(NULL)

# calculate default metrics (classification)
calculate(tidy_c5)

model_rpart = rpart::rpart(MonthlyIncome ~., data = modeldata::attrition)
tidy_rpart  =
  tidy(model_rpart) %>%
  set_validation_data(modeldata::attrition, "MonthlyIncome") %>%
  set_keys(NULL)
```

```
# calculate default metrics (regression)
calculate(tidy_rpart)

# calculate default metrics with a custom metric
#' custom function to get cumulative MAE
library("tidytable")
get_cumulative_MAE = function(rulelist, new_data, y_name, weight){

  priority_df =
    rulelist %>%
    select(rule_nbr) %>%
    mutate(priority = 1:nrow(rulelist)) %>%
    select(rule_nbr, priority)

  pred_df =
    predict(rulelist, new_data) %>%
    left_join(priority_df, by = "rule_nbr") %>%
    mutate(weight = local(weight)) %>%
    select(rule_nbr, row_nbr, weight, priority)

  new_data2 =
    new_data %>%
    mutate(row_nbr = 1:n()) %>%
    select(all_of(c("row_nbr", y_name)))

  rmse_till_rule = function(rn){

    if (is.character(rulelist$RHS)) {
      inter_df =
        pred_df %>%
        tidytable::filter(priority <= rn) %>%
        left_join(mutate(new_data, row_nbr = 1:n()), by = "row_nbr") %>%
        left_join(select(rulelist, rule_nbr, RHS), by = "rule_nbr") %>%
        nest(.by = c("RHS", "rule_nbr", "row_nbr", "priority", "weight")) %>%
        mutate(RHS = purrr::map2_dbl(RHS,
                                     data,
                                     ~ eval(parse(text = .x), envir = .y)
                                     )
               ) %>%
        unnest(data)
    } else {

      inter_df =
        pred_df %>%
        tidytable::filter(priority <= rn) %>%
        left_join(new_data2, by = "row_nbr") %>%
        left_join(select(rulelist, rule_nbr, RHS), by = "rule_nbr")
    }

    inter_df %>%
      summarise(rmse = MetricsWeighted::mae(RHS,
                                            .data[[y_name]],
                                            weight,
```

```
                                           na.rm = TRUE
                                           )
               ) %>%
      `[[`("rmse")
  }

  res = purrr::map_dbl(1:nrow(rulelist), rmse_till_rule)
  return(res)
}

calculate(tidy_rpart,
          metrics_to_exclude = NULL,
          list("cumulative_mae" = get_cumulative_MAE)
          )
```

---

convert_rule_flavor     *Convert a R parsable rule to python/sql parsable rule*

---

### Description

Convert a R parsable rule to python/sql parsable rule

### Usage

```
convert_rule_flavor(rule, flavor)
```

### Arguments

rule             (chr vector) R parsable rule(s)

flavor           (string) One among: 'python', 'sql'

### Value

(chr vector) of rules

### See Also

rulelist, tidy, augment, predict, to_sql_case

Other Auxiliary Rulelist Utility: `to_sql_case`()

| package_tidyrules | tidyrules |
|---|---|

## Description

`tidyrules` package provides a framework to work with decision rules. Rules can be extracted from supported models using tidy, augmented using validation data by augment, manipulated using standard dataframe operations, (modified) rulelists can be used to predict on unseen (test) data. Utilities include: Create a rulelist manually (as_rulelist), Export a rulelist to SQL (to_sql_case) and so on. The package offers two classes; rulelist and ruleset based on dataframe.

## Author(s)

**Maintainer**: Srikanth Komala Sheshachala <sri.teach@gmail.com>

Authors:

• Amith Kumar Ullur Raghavendra <amith54@gmail.com>

## See Also

rulelist, tidy, augment, predict

| plot.prune_rulelist | *Plot method for* prune_rulelist *class* |
|---|---|

## Description

Plot method for `prune_rulelist` class

## Usage

```
## S3 method for class 'prune_rulelist'
plot(x, ...)
```

## Arguments

| x | A 'prune_rulelist' object |
|---|---|
| ... | unused |

## Value

ggplot2 object (invisibly)

plot.rulelist *Plot method for rulelist*

#### Description

Plots a heatmap with `rule_nbr`'s on x-side and clusters of `row_nbr`'s on y-side of a binary matrix with 1 if a rule is applicable for a row.

#### Usage

```
## S3 method for class 'rulelist'
plot(x, thres_cluster_rows = 1000, dist_metric = "jaccard", ...)
```

#### Arguments

| | |
|---|---|
| x | A [rulelist] |
| thres_cluster_rows | |
| | (positive integer) Maximum number of rows beyond which a x-side dendrogram is not computed |
| dist_metric | (string or function, default: "jaccard") Distance metric for y-side (`rule_nbr`) passed to `method` argument of [proxy::dist] |
| ... | Arguments to be passed to [pheatmap::pheatmap] |

#### Details

Number of clusters is set to min(number of unique rows in the row_nbr X rule_nbr matrix and thres_cluster_rows)

#### Examples

```
library("magrittr")
att = modeldata::attrition
tidy_c5 =
  C50::C5.0(Attrition ~., data = att, rules = TRUE) %>%
  tidy() %>%
  set_validation_data(att, "Attrition") %>%
  set_keys(NULL)

plot(tidy_c5)
```

---

predict.rulelist          predict *method for a [rulelist](rulelist)*

---

### Description

Predicts `rule_nbr` applicable (as per the order in rulelist) for a `row_nbr` (per key) in new_data

### Usage

```
## S3 method for class 'rulelist'
predict(object, new_data, multiple = FALSE, ...)
```

### Arguments

| | |
|---|---|
| object | A [rulelist](rulelist) |
| new_data | (dataframe) |
| multiple | (flag, default: FALSE) Whether to output all rule numbers applicable for a row. If FALSE, the first satisfying rule is provided. |
| ... | unused |

### Details

If a `row_nbr` is covered more than one `rule_nbr` per 'keys', then `rule_nbr` appearing earlier (as in row order of the [rulelist](rulelist)) takes precedence.

#### Output Format:

- When multiple is `FALSE`(default), output is a dataframe with three or more columns: `row_number` (int), columns corresponding to 'keys', `rule_nbr` (int).
- When multiple is `TRUE`, output is a dataframe with three or more columns: `row_number` (int), columns corresponding to 'keys', `rule_nbr` (list column of integers).
- If a row number and 'keys' combination is not covered by any rule, then `rule_nbr` column has missing value.

### Value

A dataframe. See **Details**.

### See Also

[rulelist](rulelist), [tidy](tidy), [augment](augment), [predict](predict), [calculate](calculate), [prune](prune), [reorder](reorder)

## Examples

```
model_c5 = C50::C5.0(species ~.,
                     data = palmerpenguins::penguins,
                     trials = 5,
                     rules = TRUE
                     )
tidy_c5 = tidy(model_c5)
tidy_c5

output_1 = predict(tidy_c5, palmerpenguins::penguins)
output_1 # different rules per 'keys' (`trial_nbr` here)

output_2 = predict(tidy_c5, palmerpenguins::penguins, multiple = TRUE)
output_2 # `rule_nbr` is a list-column of integer vectors
```

---

| predict.ruleset | predict *method for a [ruleset](#)* |
|---|---|

---

### Description

Predicts multiple rule_nbr(s) applicable for a row_nbr (per key) in new_data

### Usage

```
## S3 method for class 'ruleset'
predict(object, new_data, ...)
```

### Arguments

| object | A [ruleset](#) |
|---|---|
| new_data | (dataframe) |
| ... | unused |

### Value

A dataframe with three or more columns: row_number (int), columns corresponding to 'keys', rule_nbr (list column of integers). If a row number and 'keys' combination is not covered by any rule, then rule_nbr column has missing value.

### See Also

[predict.rulelist](#)

## Examples

```
model_c5 = C50::C5.0(species ~.,
                     data = palmerpenguins::penguins,
                     trials = 5,
                     rules = TRUE
                     )
tidy_c5_ruleset = as_ruleset(tidy(model_c5))
tidy_c5_ruleset

predict(tidy_c5_ruleset, palmerpenguins::penguins)
```

---

print.prune_rulelist     *Print method for* prune_rulelist *class*

---

### Description

Print method for prune_rulelist class

### Usage

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

### Arguments

| | |
|---|---|
| x | A 'prune_rulelist' object |
| ... | unused |

---

print.rulelist     *Print method for* rulelist *class*

---

### Description

Prints rulelist attributes and first few rows.

### Usage

```
## S3 method for class 'rulelist'
print(x, banner = TRUE, ...)
```

### Arguments

| | |
|---|---|
| x | A rulelist object |
| banner | (flag, default: TRUE) Should the banner be displayed |
| ... | Passed to tidytable::print |

## Value

input [rulelist](invisibly)

## See Also

[rulelist](rulelist), [tidy](tidy), [augment](augment), [predict](predict), [calculate](calculate), [prune](prune), [reorder](reorder)

---

print.ruleset *Print method for ruleset class*

---

## Description

Prints the ruleset object

## Usage

```
## S3 method for class 'ruleset'
print(x, banner = TRUE, ...)
```

## Arguments

| | |
|---|---|
| x | A [rulelist](rulelist) |
| banner | (flag, default: TRUE) Should the banner be displayed |
| ... | Passed to print.rulelist |

## Value

(invisibly) Returns the ruleset object

## See Also

[print.rulelist](print.rulelist)

## Examples

```
model_class_party = partykit::ctree(species ~ .,
                                    data = palmerpenguins::penguins
                                    )
as_ruleset(tidy(model_class_party))
```

---

prune                          prune *is re-export of [generics::prune](#) from [tidyrules](#) package*

---

## Description

See [prune.rulelist](#)

## Usage

```
prune(tree, ...)
```

## Arguments

| | |
|---|---|
| tree | A [rulelist](#) |
| ... | See [prune.rulelist](#) |

## See Also

[rulelist](#), [tidy](#), [augment](#), [predict](#), [calculate](#), [prune](#), [reorder](#)

---

prune.rulelist            prune *rules of a [rulelist](#)*

---

## Description

Prune the [rulelist](#) by suggesting to keep first 'k' rules based on metrics computed by [calculate](#)

## Usage

```
## S3 method for class 'rulelist'
prune(
  tree,
  metrics_to_exclude = NULL,
  stop_expr_string = "relative__cumulative_coverage >= 0.9",
  min_n_rules = 1,
  ...
)
```

## Arguments

| | |
|---|---|
| tree | A [rulelist](#) |
| metrics_to_exclude | |
| | (character vector or NULL) Names of metrics not to be calculated. See [calculate](#) for the list of default metrics. |
| stop_expr_string | |
| | (string default: "relative__cumulative_coverage >= 0.9") Parsable condition |
| min_n_rules | (positive integer) Minimum number of rules to keep |
| ... | Named list of custom metrics passed to [calculate](#) |

**Details**

1. Metrics are computed using calculate. 2. Relative metrics (prepended by 'relative__') are calculated by dividing each metric by its max value. 3. The first rule in rulelist order which meets the 'stop_expr_string' criteria is stored (say 'pos'). Print method suggests to keep rules until pos.

**Value**

Object of class 'prune_ruleslist' with these components: 1. pruned: ruleset keeping only first 'pos' rows. 2. n_pruned_rules: pos. If stop criteria is never met, then pos = nrow(ruleset) 3. n_total_rules: nrow(ruleset), 4. metrics_df: Dataframe with metrics and relative metrics 5. stop_expr_string

**See Also**

rulelist, tidy, augment, predict, calculate, prune, reorder

**Examples**

```
library("magrittr")
model_c5  = C50::C5.0(Attrition ~., data = modeldata::attrition, rules = TRUE)
tidy_c5   = tidy(model_c5) %>%
              set_validation_data(modeldata::attrition, "Attrition") %>%
              set_keys(NULL)

#' prune with defaults
prune_obj = prune(tidy_c5)
#' note that all other metrics are visible in the print output
prune_obj
plot(prune_obj)
prune_obj$pruned

#' prune with a different stop_expr_string threshold
prune_obj = prune(tidy_c5,
                    stop_expr_string = "relative__cumulative_coverage >= 0.2"
                    )
prune_obj #' as expected, has smaller then 10 rules as compared to default args
plot(prune_obj)
prune_obj$pruned

#' prune with a different stop_expr_string metric
st = "relative__cumulative_overlap <= 0.7 & relative__cumulative_overlap > 0"
prune_obj = prune(tidy_c5, stop_expr_string = st)
prune_obj #' as expected, has smaller then 10 rules as compared to default args
plot(prune_obj)
prune_obj$pruned
```

---

reorder *reorder generic*

---

### Description

reorder generic for rulelist

### Usage

```
reorder(x, ...)
```

### Arguments

| | |
|---|---|
| x | A rulelist |
| ... | See reorder.rulelist |

### See Also

rulelist, tidy, augment, predict, calculate, prune, reorder

---

reorder.rulelist *Reorder the rules/rows of a rulelist*

---

### Description

Implements a greedy strategy to add one rule at a time which maximizes/minimizes a metric.

### Usage

```
## S3 method for class 'rulelist'
reorder(x, metric = "cumulative_coverage", minimize = FALSE, init = NULL, ...)
```

### Arguments

| | |
|---|---|
| x | A rulelist |
| metric | (character vector or named list) Name of metrics or a custom function(s). See calculate. The 'n+1'th metric is used when there is a match at 'nth' level, similar to base::order. If there is a match at final level, row order of the rulelist comes into play. |
| minimize | (logical vector) Whether to minimize. Either TRUE/FALSE or a logical vector of same length as metric |
| init | (positive integer) Initial number of rows after which reordering should begin |
| ... | passed to calculate |

## See Also

rulelist, tidy, augment, predict, calculate, prune, reorder

## Examples

```
library("magrittr")
att = modeldata::attrition
tidy_c5 =
  C50::C5.0(Attrition ~., data = att, rules = TRUE) %>%
  tidy() %>%
  set_validation_data(att, "Attrition") %>%
  set_keys(NULL) %>%
  head(5)

# with defaults
reorder(tidy_c5)

# use 'cumulative_overlap' to break ties (if any)
reorder(tidy_c5, metric = c("cumulative_coverage", "cumulative_overlap"))

# reorder after 2 rules
reorder(tidy_c5, init = 2)
```

---

| rulelist | *Rulelist* |

---

## Description

### Structure:

A `rulelist` is ordered list of rules stored as a dataframe. Each row, specifies a rule (LHS),
expected outcome (RHS) and some other details.

It has these mandatory columns:

- rule_nbr: (integer vector) Rule number
- LHS: (character vector) A rule is a string that can be parsed using base::parse()
- RHS: (character vector or a literal)

### Example:

```
| rule_nbr|LHS                                                            |RHS       | support| confidence|    lift|
|--------:|:--------------------------------------------------------------|:---------|-------|------
|        1|( island %in% c('Biscoe') ) & ( flipper_length_mm > 203 )      |Gentoo    |    122| 1.0000000
|        2|( island %in% c('Biscoe') ) & ( flipper_length_mm <= 203 )     |Adelie    |     46| 0.9565211
|        3|( island %in% c('Dream', 'Torgersen') ) & ( bill_length_mm > 44.1 ) |Chinstrap |     65| 0.953
|        4|( island %in% c('Dream', 'Torgersen') ) & ( bill_length_mm <= 44.1 )|Adelie    |    111| 0.945
```

### Create a rulelist:

A `rulelist` can be created using tidy() on some supported model fits (run: utils::methods(tidy)).
It can also be created manually from a existing dataframe using as_rulelist.

**Keys and attributes:**

Columns identified as 'keys' along with `rule_nbr` form a unique combination – a group of rules. For example, rule-based C5 model with multiple trials creates rules per each `trial_nbr`. `predict` method understands 'keys', thereby provides/predicts a rule number (for each row in new data / test data) within the same `trial_nbr`.

A rulelist has these mandatory attributes:

- `estimation_type`: One among `regression`, `classification`
  A rulelist has these optional attributes:
- keys: (character vector)Names of the column that forms a key.
- `model_type`: (string) Name of the model

*Set Validation data:*

This helps a few methods like [augment](#), [calculate](#), [prune](#), [reorder](#) require few additional attributes which can be set using [set_validation_data](#).

*Methods for rulelist:*

1. [Predict](#): Given a dataframe (possibly without a dependent variable column aka 'test data'), predicts the first rule (as ordered in the rulelist) per 'keys' that is applicable for each row. When `multiple = TRUE`, returns all rules applicable for a row (per key).
2. [Augment](#): Outputs summary statistics per rule over validation data and returns a rulelist with a new dataframe-column.
3. [Calculate](#): Computes metrics for a rulelist in a cumulative manner such as `cumulative_coverage`, `cumulative_overlap`, `cumulative_accuracy`.
4. [Prune](#): Suggests pruning a rulelist such that some expectation are met (based on metrics). Example: cumulative_coverage of 80% can be met with a first few rules.
5. [Reorder](#): Reorders a rulelist in order to maximize a metric.

*Manipulating a rulelist:*

Rulelists are essentially dataframes. Hence, any dataframe operations which preferably preserve attributes will output a rulelist. [as_rulelist](#) and [as.data.frame](#) will help in moving back and forth between rulelist and dataframe worlds.

*Utilities for a rulelist:*

1. [as_rulelist](#): Create a `rulelist` from a dataframe with some mandatory columns.
2. [set_keys](#): Set or Unset 'keys' of a `rulelist`.
3. [to_sql_case](#): Outputs a SQL case statement for a `rulelist`.
4. [convert_rule_flavor](#): Converts R-parsable rule strings to python/SQL parsable rule strings.

## See Also

[rulelist](#), [tidy](#), [augment](#), [predict](#), [calculate](#), [prune](#), [reorder](#)

---

| ruleset | *Ruleset* |
|---------|-----------|

---

## Description

ruleset class is a piggyback class that inherits [rulelist](#) class for convenience of [print](#) and [predict](#) methods.

---

set_keys                          *Set keys for a rulelist*

---

### Description

'keys' are a set of column(s) which identify a group of rules in a rulelist. Methods like predict, augment produce output per key combination.

### Usage

```
set_keys(x, keys, reset = FALSE)
```

### Arguments

| | |
|---|---|
| x | A rulelist |
| keys | (character vector or NULL) |
| reset | (flag) Whether to reset the keys to sequential numbers starting with 1 when keys is set to NULL |

### Details

A new rulelist is returned with attr keys is modified. The input rulelist object is unaltered.

### Value

A rulelist object

### See Also

rulelist, tidy, augment, predict, calculate, prune, reorder

Other Core Rulelist Utility: `set_validation_data`()

### Examples

```
model_c5 = C50::C5.0(Attrition ~., data = modeldata::attrition, rules = TRUE)
tidy_c5 = tidy(model_c5)
tidy_c5 # keys are: "trial_nbr"

tidy_c5[["rule_nbr"]] = 1:nrow(tidy_c5)
new_tidy_c5 = set_keys(tidy_c5, NULL) # remove all keys
new_tidy_c5

new_2_tidy_c5 = set_keys(new_tidy_c5, "trial_nbr") # set "trial_nbr" as key
new_2_tidy_c5

# Note that `tidy_c5` and `new_tidy_c5` are not altered.
tidy_c5
new_tidy_c5
```

---

set_validation_data          *Add* validation_data *to a rulelist*

---

### Description

Returns a rulelist with three new attributes set: validation_data, y_name and weight. Methods such as augment, calculate, prune, reorder require this to be set.

### Usage

```
set_validation_data(x, validation_data, y_name, weight = 1)
```

### Arguments

| | |
|---|---|
| x | A rulelist |
| validation_data | |
| | (dataframe) Data to used for computing some metrics. It is expected to contain y_name column. |
| y_name | (string) Name of the dependent variable column. |
| weight | (non-negative numeric vector, default: 1) Weight per observation/row of validation_data. This is expected to have same length as the number of rows in validation_data. Only exception is when it is a single positive number, which means that all rows have equal weight. |

### Value

A rulelist with some extra attributes set.

### See Also

rulelist, tidy, augment, predict, calculate, prune, reorder

Other Core Rulelist Utility: `set_keys()`

### Examples

```
att = modeldata::attrition
set.seed(100)
index = sample(c(TRUE, FALSE), nrow(att), replace = TRUE)
model_c5 = C50::C5.0(Attrition ~., data = att[index, ], rules = TRUE)

tidy_c5 = tidy(model_c5)
tidy_c5

tidy_c5_2 = set_validation_data(tidy_c5,
                                validation_data = att[!index, ],
                                y_name = "Attrition",
                                weight = 1 # default
                                )
```

```
tidy_c5_2
tidy_c5 # not altered
```

---

tidy                    tidy *is re-export of [generics::tidy](#) from [tidyrules](#) package*

---

### Description

tidy applied on a supported model fit creates a [rulelist](#). **See Also** section links to documentation of specific methods.

### Usage

```
tidy(x, ...)
```

### Arguments

x               A supported model object

...             For model specific implementations to use

### See Also

[rulelist](#), [tidy](#), [augment](#), [predict](#), [calculate](#), [prune](#), [reorder](#)

Other Core Tidy Utility: `tidy.C5.0()`, `tidy.cubist()`, `tidy.rpart()`

---

tidy.C5.0               *Get the [rulelist](#) from a [C5](#) model*

---

### Description

Each row corresponds to a rule per `trial_nbr`

### Usage

```
## S3 method for class 'C5.0'
tidy(x, ...)
```

### Arguments

x               [C50::C5.0](#) model fitted with `rules = TRUE`

...             Other arguments (See details)

## Details

- The output columns are: rule_nbr, trial_nbr, LHS, RHS, support, confidence, lift.

- Rules per trial_nbr are sorted in this order: desc(confidence), desc(lift), desc(support).

Optional named arguments:

- laplace (flag, default: TRUE) is supported. This computes confidence with laplace correction as documented under 'Rulesets' here: C5 doc.

## Value

A rulelist object

## See Also

rulelist, tidy, augment, predict, calculate, prune, reorder

Other Core Tidy Utility: `tidy()`, `tidy.cubist()`, `tidy.rpart()`

## Examples

```
model_c5 = C50::C5.0(Attrition ~., data = modeldata::attrition, rules = TRUE)
tidy(model_c5)
```

---

tidy.constparty *Get the rulelist from a party model*

---

## Description

Each row corresponds to a rule

## Usage

```
## S3 method for class 'constparty'
tidy(x, ...)
```

## Arguments

| | |
|---|---|
| x | partykit::party model typically built using partykit::ctree |
| ... | Other arguments (currently unused) |

## Details

These types of [party](#) models are supported: regression (y is numeric), classification (y is factor)

For [party](#) classification model:

- Output columns are: rule_nbr, LHS, RHS, support, confidence, lift, terminal_node_id.
- Rules are sorted in this order: desc(confidence), desc(lift), desc(support).

For [party](#) regression model:

- Output columns are: rule_nbr, LHS, RHS, support, IQR, RMSE, terminal_node_id.
- Rules are sorted in this order: RMSE, desc(support).

## Value

A [rulelist](#) object

## See Also

[rulelist](#), [tidy](#), [augment](#), [predict](#), [calculate](#), [prune](#), [reorder](#)

## Examples

```
pen = palmerpenguins::penguins
model_class_party = partykit::ctree(species ~ ., data = pen)
tidy(model_class_party)
model_regr_party = partykit::ctree(bill_length_mm ~ ., data = pen)
tidy(model_regr_party)
```

---

tidy.cubist *Get the [rulelist](#) from a [cubist](#) model*

---

## Description

Each row corresponds to a rule per committee

## Usage

```
## S3 method for class 'cubist'
tidy(x, ...)
```

## Arguments

| | |
|---|---|
| x | [Cubist::cubist](#) model |
| ... | Other arguments (currently unused) |

## Details

- The output columns are: rule_nbr, committee, LHS, RHS, support, mean, min, max, error.

- Rules are sorted in this order per committee: error, desc(support)

## Value

A [rulelist](#) object

## See Also

[rulelist](#), [tidy](#), [augment](#), [predict](#), [calculate](#), [prune](#), [reorder](#)

Other Core Tidy Utility: [tidy](#)(), [tidy.C5.0](#)(), [tidy.rpart](#)()

## Examples

```
att = modeldata::attrition
cols_att    = setdiff(colnames(att), c("MonthlyIncome", "Attrition"))
model_cubist = Cubist::cubist(x = att[, cols_att],
                              y = att[["MonthlyIncome"]]
                              )
tidy(model_cubist)
```

---

tidy.rpart                     *Get the [rulelist](#) from a [rpart](#) model*

---

## Description

Each row corresponds to a rule

## Usage

```
## S3 method for class 'rpart'
tidy(x, ...)
```

## Arguments

| | |
|---|---|
| x | [rpart::rpart](#) model |
| ... | Other arguments (currently unused) |

## Details

For rpart rules, one should build the model without ordered factor variable. We recommend you to convert ordered factor to factor or *integer* class.

For rpart::rpart classification model:

- Output columns are: rule_nbr, LHS, RHS, support, confidence, lift.
- The rules are sorted in this order: desc(confidence), desc(lift), desc(support).

For rpart::rpart regression(anova) model:

- Output columns are: rule_nbr, LHS, RHS, support.
- The rules are sorted in this order: desc(support).

## Value

A rulelist object

## See Also

rulelist, tidy, augment, predict, calculate, prune, reorder

Other Core Tidy Utility: `tidy()`, `tidy.C5.0()`, `tidy.cubist()`

## Examples

```
model_class_rpart = rpart::rpart(Species ~ ., data = iris)
tidy(model_class_rpart)

model_regr_rpart = rpart::rpart(Sepal.Length ~ ., data = iris)
tidy(model_regr_rpart)
```

---

to_sql_case *Extract SQL case statement from a rulelist*

---

## Description

Extract SQL case statement from a rulelist

## Usage

```
to_sql_case(rulelist, rhs_column_name = "RHS", output_colname = "output")
```

## Arguments

| | |
|---|---|
| rulelist | A rulelist object |
| rhs_column_name | |
| | (string, default: "RHS") Name of the column in the rulelist to be used as RHS (WHEN some_rule THEN rhs) in the sql case statement |
| output_colname | (string, default: "output") Name of the output column created by the SQL statement (used in case ... AS output_column) |

## Details

As a side-effect, the SQL statement is cat to stdout. The output contains newline character.

## Value

(string invisibly) SQL case statement

## See Also

rulelist, tidy, augment, predict, convert_rule_flavor

Other Auxiliary Rulelist Utility: `convert_rule_flavor()`

## Examples

```
model_c5 = C50::C5.0(Attrition ~., data = modeldata::attrition, rules = TRUE)
tidy(model_c5)
to_sql_case(tidy(model_c5))
```

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