

Package ‘mlr3’

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Title Machine Learning in R - Next Generation

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Description Efficient, object-oriented programming on the building blocks of machine learning. Provides 'R6' objects for tasks, learners, resamplings, and measures. The package is geared towards scalability and larger datasets by supporting parallelization and out-of-memory data-backends like databases. While 'mlr3' focuses on the core computational operations, add-on packages provide additional functionality.

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URL <https://mlr3.ml-org.com>, <https://github.com/mlr-org/mlr3>

BugReports <https://github.com/mlr-org/mlr3/issues>

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'DataBackendCbind.R' 'DataBackendDataTable.R'
'DataBackendMatrix.R' 'DataBackendRbind.R'
'DataBackendRename.R' 'Learner.R' 'LearnerClassif.R'
'mlr_learners.R' 'LearnerClassifDebug.R'
'LearnerClassifFeatureless.R' 'LearnerClassifRpart.R'
'LearnerRegr.R' 'LearnerRegrFeatureless.R' 'LearnerRegrRpart.R'

'Measure.R' 'MeasureClassif.R' 'mlr_measures.R'
 'MeasureClassifCosts.R' 'MeasureDebug.R' 'MeasureElapsedTime.R'
 'MeasureOOBError.R' 'MeasureRegr.R' 'MeasureSelectedFeatures.R'
 'MeasureSimple.R' 'Prediction.R' 'PredictionClassif.R'
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 'assertions.R' 'benchmark.R' 'benchmark_grid.R'
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 'mlr_coercions.R' 'mlr_sugar.R' 'predict.R' 'reexports.R'
 'resample.R' 'worker.R' 'zzz.R'

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Description

Efficient, object-oriented programming on the building blocks of machine learning. Provides 'R6' objects for tasks, learners, resamplings, and measures. The package is geared towards scalability and larger datasets by supporting parallelization and out-of-memory data-backends like databases. While 'mlr3' focuses on the core computational operations, add-on packages provide additional functionality.

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References

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See Also

Useful links:

- <https://mlr3.mlr-org.com>
- <https://github.com/mlr-org/mlr3>
- Report bugs at <https://github.com/mlr-org/mlr3/issues>

as_benchmark_result *Convert to BenchmarkResult*

Description

Simple S3 method to convert objects to a [BenchmarkResult](#).

Usage

```
as_benchmark_result(x, ...)

## S3 method for class 'ResampleResult'
as_benchmark_result(x, ...)
```

Arguments

x :: any
Object to dispatch on, e.g. a [ResampleResult](#).

... :: any
Currently not used.

Value

([BenchmarkResult](#)).

as_data_backend *Create a Data Backend*

Description

Wraps a [DataBackend](#) around data.

Usage

```
as_data_backend(data, ...)
```

Arguments

data :: any
Data to create a [DataBackend](#) for. For a `data.frame()` (this includes `tibble()` from [tibble](#) and `data.table::data.table()`) this function creates a [DataBackendDataTable](#). See `methods("as_data_backend")` for possible input formats. Note that third-party packages may extend this functionality.

... :: any
Additional arguments passed to the respective [DataBackend](#) method.

Value

[DataBackend](#).

See Also

Other DataBackend: [DataBackendDataTable](#), [DataBackendMatrix](#), [DataBackend](#)

Examples

```
# create a new backend using the iris data:
as_data_backend(iris)
```

as_task.character *Object Coercion*

Description

S3 generics and methods to coerce to (lists of) [Task](#), [Learner](#), [Resampling](#), and [Measure](#).

Usage

```
## S3 method for class 'character'
as_task(x, clone = FALSE)

## S3 method for class 'character'
as_tasks(x, clone = FALSE)

## S3 method for class 'character'
as_learner(x, clone = FALSE)

## S3 method for class 'character'
as_learners(x, clone = FALSE)

## S3 method for class 'character'
as_resampling(x, clone = FALSE)

## S3 method for class 'character'
as_resamplings(x, clone = FALSE)

## S3 method for class 'character'
as_measure(x, task_type = NULL, clone = FALSE)

## S3 method for class 'character'
as_measures(x, task_type = NULL, clone = FALSE)

as_task(x, clone = FALSE)
```

```
## S3 method for class 'Task'
as_task(x, clone = FALSE)

as_tasks(x, clone = FALSE)

## S3 method for class 'list'
as_tasks(x, clone = FALSE)

## S3 method for class 'Task'
as_tasks(x, clone = FALSE)

as_learner(x, clone = FALSE)

## S3 method for class 'Learner'
as_learner(x, clone = FALSE)

as_learners(x, clone = FALSE)

## S3 method for class 'list'
as_learners(x, clone = FALSE)

## S3 method for class 'Learner'
as_learners(x, clone = FALSE)

as_resampling(x, clone = FALSE)

## S3 method for class 'Resampling'
as_resampling(x, clone = FALSE)

as_resamplings(x, clone = FALSE)

## S3 method for class 'list'
as_resamplings(x, clone = FALSE)

## S3 method for class 'Resampling'
as_resamplings(x, clone = FALSE)

as_measure(x, task_type = NULL, clone = FALSE)

## S3 method for class ``NULL``
as_measure(x, task_type = NULL, clone = FALSE)

## S3 method for class 'Measure'
as_measure(x, task_type = NULL, clone = FALSE)

as_measures(x, task_type = NULL, clone = FALSE)

## S3 method for class ``NULL``
```



```

as_measures(x, task_type = NULL, clone = FALSE)

## S3 method for class 'list'
as_measures(x, task_type = NULL, clone = FALSE)

## S3 method for class 'Measure'
as_measures(x, task_type = NULL, clone = FALSE)

```

Arguments

x	:: any Object to coerce.
clone	:: logical(1) If TRUE, ensures that the returned object is not the same as the input x, e.g. by cloning it or constructing it from a mlr3misc::Dictionary .
task_type	:: character(1) Used if x is NULL to construct a default measure for the respective task type. The default measures are stored in mlr_reflections\$default_measures .

Value

Coerced object. The default method will return the object as-is. Failed coercions have to be handled by one of the assertions in [mlr_assertions](#).

Examples

```

# convert single measure to list of measures
measure = msr("classif.ce")
as_measures(measure)

```

 benchmark

Benchmark Multiple Learners on Multiple Tasks

Description

Runs a benchmark on arbitrary combinations of tasks ([Task](#)), learners ([Learner](#)), and resampling strategies ([Resampling](#)), possibly in parallel.

Usage

```
benchmark(design, store_models = FALSE)
```

Arguments

`design` `:: data.frame()`
 Data frame (or `data.table::data.table()`) with three columns: "task", "learner", and "resampling". Each row defines a resampling by providing a [Task](#), [Learner](#) and an instantiated [Resampling](#) strategy. The helper function `benchmark_grid()` can assist in generating an exhaustive design (see examples) and instantiate the [Resamplings per Task](#).

`store_models` `:: logical(1)`
 Keep the fitted model after the test set has been predicted? Set to TRUE if you want to further analyse the models or want to extract information like variable importance.

Value

[BenchmarkResult](#).

Parallelization

This function can be parallelized with the [future](#) package. One job is one resampling iteration, and all jobs are sent to an apply function from [future.apply](#) in a single batch. To select a parallel backend, use `future::plan()`.

Logging

The [mlr3](#) uses the [lgr](#) package for logging. [lgr](#) supports multiple log levels which can be queried with `getOption("lgr.log_levels")`.

To suppress output and reduce verbosity, you can lower the log from the default level "info" to "warn":

```
lgr::get_logger("mlr3")$set_threshold("warn")
```

To get additional log output for debugging, increase the log level to "debug" or "trace":

```
lgr::get_logger("mlr3")$set_threshold("debug")
```

To log to a file or a data base, see the documentation of [lgr::lgr-package](#).

Note

The fitted models are discarded after the predictions have been scored in order to reduce memory consumption. If you need access to the models for later analysis, set `store_models` to TRUE.

Examples

```
# benchmarking with benchmark_grid()
tasks = lapply(c("iris", "sonar"), tsk)
learners = lapply(c("classif.featureless", "classif.rpart"), lrn)
resamplings = rsmpl("cv", folds = 3)
```

```

design = benchmark_grid(tasks, learners, resamplings)
print(design)

set.seed(123)
bmr = benchmark(design)

## data of all resamplings
head(as.data.table(bmr))

## aggregated performance values
aggr = bmr$aggregate()
print(aggr)

## Extract predictions of first resampling result
rr = aggr$resample_result[[1]]
as.data.table(rr$prediction())

# benchmarking with a custom design:
# - fit classif.featureless on iris with a 3-fold CV
# - fit classif.rpart on sonar using a holdout
tasks = list(tsk("iris"), tsk("sonar"))
learners = list(lrn("classif.featureless"), lrn("classif.rpart"))
resamplings = list(rsm("cv", folds = 3), rsm("holdout"))

design = data.table::data.table(
  task = tasks,
  learner = learners,
  resampling = resamplings
)

## instantiate resamplings
design$resampling = Map(
  function(task, resampling) resampling$clone()$instantiate(task),
  task = design$task, resampling = design$resampling
)

## run benchmark
bmr = benchmark(design)
print(bmr)

## get the training set of the 2nd iteration of the featureless learner on iris
rr = bmr$aggregate()[learner_id == "classif.featureless"]$resample_result[[1]]
rr$resampling$train_set(2)

```

BenchmarkResult

Container for Results of benchmark()

Description

This is the result container object returned by `benchmark()`. A `BenchmarkResult` consists of the data row-binded data of multiple `ResampleResults`, which can easily be re-constructed.

Note that all stored objects are accessed by reference. Do not modify any object without cloning it first.

Format

`R6::R6Class` object.

Construction

```
bmr = BenchmarkResult$new(data = data.table())
```

- `data :: data.table::data.table()`
Table with data for one resampling iteration per row: `Task`, `Learner`, `Resampling`, iteration (`integer(1)`), `Prediction`, and the unique hash `uhash` (`character(1)`) of the corresponding `ResampleResult`. Additional columns are kept in the resulting object.

Fields

- `data :: data.table::data.table()`
Internal data storage with one row per resampling iteration. Can be joined with `$rr_data` by joining on column "hash". We discourage users to directly work with this table.
Package developers on the other hand may opt to add additional columns here. These columns are preserved in all mutators.
- `rr_data :: data.table::data.table()`
Internal data storage with one row per `ResampleResult`. Can be joined with `$data` by joining on column "hash". Not used in `mlr3` directly, but can be exploited by add-on packages.
Package developers may opt to add additional columns here. These columns are preserved in all mutators.
- `task_type :: character(1)`
Task type of objects in the `BenchmarkResult`. All stored objects (`Task`, `Learner`, `Prediction`) in a single `BenchmarkResult` are required to have the same task type, e.g., "classif" or "regr".
- `tasks :: data.table::data.table()`
Table of used tasks with three columns: "task_hash" (`character(1)`), "task_id" (`character(1)`) and "task" (`Task`).
- `learners :: data.table::data.table()`
Table of used learners with three columns: "learner_hash" (`character(1)`), "learner_id" (`character(1)`) and "learner" (`Learner`). Note that it is not feasible to access learnt models via this getter, as the training task would be ambiguous. Instead, select a row from the table returned by `$score()`.
- `resamplings :: data.table::data.table()`
Table of used resamplings with three columns: "resampling_hash" (`character(1)`), "resampling_id" (`character(1)`) and "resampling" (`Resampling`).
- `n_resample_results :: integer(1)`
Returns the number of stored `ResampleResults`.
- `uhashes :: character()`
Vector of unique hashes of all included `ResampleResults`.

Methods

- `aggregate(measures = NULL, ids = TRUE, uhashes = FALSE, params = FALSE, conditions = FALSE)`
 (list of [Measure](#), logical(1), logical(1), logical(1), logical(1)) -> `data.table::data.table()`
 Returns a result table where resampling iterations are combined into [ResampleResults](#). A column with the aggregated performance score is added for each [Measure](#), named with the id of the respective measure.
 For convenience, the following parameters can be set to extract more information from the returned [ResampleResult](#):
 - `uhashes :: logical(1)`
 Adds the uhash values of the [ResampleResult](#) as extra character column "uhash".
 - `ids :: logical(1)`
 Adds object ids ("task_id", "learner_id", "resampling_id") as extra character columns.
 - `params :: logical(1)`
 Adds the hyperparameter values as extra list column "params". You can unnest them with `mlr3misc::unnest()`.
 - `conditions :: logical(1)`
 Adds the number of resampling iterations with at least one warning as extra integer column "warnings", and the number of resampling iterations with errors as extra integer column "errors".
- `score(measures = NULL, ids = TRUE)`
 (list of [Measure](#), logical(1)) -> `data.table::data.table()`
 Returns a table with one row for each resampling iteration, including all involved objects: [Task](#), [Learner](#), [Resampling](#), iteration number (integer(1)), and [Prediction](#). If `ids` is set to TRUE, character column of extracted ids are added to the table for convenient filtering: "task_id", "learner_id", and "resampling_id". Additionally calculates the provided performance measures and binds the performance as extra columns. These columns are named using the id of the respective [Measure](#).
- `resample_result(i = NULL, uhash = NULL)`
 (integer(1), character(1)) -> [ResampleResult](#)
 Retrieve the i-th [ResampleResult](#), by position or by unique hash uhash. `i` and `uhash` are mutually exclusive.
- `combine(bmr)`
 ([BenchmarkResult](#) | NULL) -> self
 Fuses a second [BenchmarkResult](#) into itself, mutating the [BenchmarkResult](#) in-place. If `bmr` is NULL, simply returns self.
- `filter(task_ids = NULL, learner_ids = NULL, resampling_ids = NULL)`
 (character(), character(), character()) -> self
 Subsets the benchmark result. If `task_ids` is not NULL, keeps all tasks with provided task ids while discards all others. Same procedure for `learner_ids` and `resampling_ids`.
- `help()`
 () -> NULL
 Opens the help page for this object.

S3 Methods

- `as.data.table(bmr)`

`BenchmarkResult` -> `data.table::data.table()`
Returns a copy of the internal data.

Examples

```
set.seed(123)
learners = list(
  lrn("classif.featureless", predict_type = "prob"),
  lrn("classif.rpart", predict_type = "prob")
)

design = benchmark_grid(
  tasks = list(tsk("sonar"), tsk("spam")),
  learners = learners,
  resamplings = rsmpl("cv", folds = 3)
)
print(design)

bmr = benchmark(design)
print(bmr)

bmr$tasks
bmr$learners

# first 5 individual resamplings
head(as.data.table(bmr, measures = c("classif.acc", "classif.auc")), 5)

# aggregate results
bmr$aggregate()

# aggregate results with hyperparameters as separate columns
mlr3misc::unnest(bmr$aggregate(params = TRUE), "params")

# extract resample result for classif.rpart
rr = bmr$aggregate()[learner_id == "classif.rpart", resample_result][[1]]
print(rr)

# access the confusion matrix of the first resampling iteration
rr$predictions()[[1]]$confusion
```

benchmark_grid

Generate a Benchmark Grid Design

Description

Takes a list of `Task`, a list of `Learner` and a list of `Resampling` to generate a design in an `expand.grid()` fashion (a.k.a. cross join or Cartesian product).

Resampling strategies are not allowed to be instantiated when passing the argument, and instead will be instantiated per task internally. The only exception to this rule applies if all tasks have exactly the same number of rows, and the resamplings are all instantiated for such tasks.

Usage

```
benchmark_grid(tasks, learners, resamplings)
```

Arguments

```
tasks          :: list of Task.
learners       :: list of Learner.
resamplings    :: list of Resampling.
```

Value

([data.table::data.table\(\)](#)) with the cross product of the input vectors.

Examples

```
tasks = list(tsk("iris"), tsk("sonar"))
learners = list(lrn("classif.featureless"), lrn("classif.rpart"))
resamplings = list(rsm("cv"), rsm("subsampling"))
benchmark_grid(tasks, learners, resamplings)
```

DataBackend

DataBackend

Description

This is the abstract base class for data backends.

Data Backends provide a layer of abstraction for various data storage systems. It is not recommended to work directly with the DataBackend. Instead, all data access is handled transparently via the [Task](#).

To connect to out-of-memory database management systems such as SQL servers, see [mlr3db](#).

The required set of fields and methods to implement a custom DataBackend is listed in the respective sections. See [DataBackendDataTable](#) or [DataBackendMatrix](#) for exemplary implementations of the interface.

Format

[R6::R6Class](#) object.

Construction

Note: This object is typically constructed via a derived classes, e.g. [DataBackendDataTable](#) or [DataBackendMatrix](#), or via the S3 method [as_data_backend\(\)](#).

```
DataBackend$new(data, primary_key = NULL, data_formats = "data.table")
```

- `data :: any`
The format of the input data depends on the specialization. E.g., `DataBackendDataTable` expects a `data.table::data.table()` and `DataBackendMatrix` expects a `Matrix::Matrix()` constructed with the **Matrix** package.
- `primary_key :: character(1)`
Each `DataBackend` needs a way to address rows, which is done via a column of unique values, referenced here by `primary_key`. The use of this variable may differ between backends.
- `data_formats (character())`
Set of supported formats, e.g. "data.table" or "Matrix".

Fields

- `nrow :: integer(1)`
Number of rows (observations).
- `ncol :: integer(1)`
Number of columns (variables), including the primary key column.
- `colnames :: character()`
Returns vector of all column names, including the primary key column.
- `rownames :: (integer() | character())`
Returns vector of all distinct row identifiers, i.e. the primary key column.
- `hash :: character(1)`
Returns a unique hash for this backend. This hash is cached.
- `data_formats :: character()`
Vector of supported data formats. A specific format can be chosen in the `$data()` method.

Methods

- `data(rows = NULL, cols = NULL, format = "data.table")`
`(integer() | character(), character()) -> any`
Returns a slice of the data in the specified format. Currently, the only supported formats are "data.table" and "Matrix". The rows must be addressed as vector of primary key values, columns must be referred to via column names. Queries for rows with no matching row id and queries for columns with no matching column name are silently ignored. Rows are guaranteed to be returned in the same order as rows, columns may be returned in an arbitrary order. Duplicated row ids result in duplicated rows, duplicated column names lead to an exception.
- `distinct(rows, cols, na_rm = TRUE)`
`(integer() | character(), character(), logical(1)) -> named list()`
Returns a named list of vectors of distinct values for each column specified. If `na_rm` is `TRUE`, missing values are removed from the returned vectors of distinct values. Non-existing rows and columns are silently ignored.
If `rows` is `NULL`, all possible distinct values will be returned, even if the value is not present in the data. This affects factor-like variables with empty levels, if supported by the backend.
- `head(n = 6)`
`integer(1) -> data.table::data.table()`
Returns the first up-to `n` rows of the data as `data.table::data.table()`.

- `missings(rows, cols)`
(integer() | character(), character()) -> named integer()
Returns the number of missing values per column in the specified slice of data. Non-existing rows and columns are silently ignored.

See Also

Extension Packages: [mlr3db](#)

Other DataBackend: [DataBackendDataTable](#), [DataBackendMatrix](#), [as_data_backend\(\)](#)

Examples

```
data = data.table::data.table(id = 1:5, x = runif(5), y = sample(letters[1:3], 5, replace = TRUE))

b = DataBackendDataTable$new(data, primary_key = "id")
print(b)
b$head(2)
b$data(rows = 1:2, cols = "x")
b$distinct(rows = b$rownames, "y")
b$missings(rows = b$rownames, cols = names(data))
```

DataBackendDataTable *DataBackend for data.table*

Description

[DataBackend](#) for [data.table](#) as an in-memory data base.

Format

[R6::R6Class](#) object inheriting from [DataBackend](#).

Construction

```
DataBackendDataTable$new(data, primary_key = NULL)
as_data_backend(data, primary_key = NULL, ...)
```

- `data` :: [data.table::data.table\(\)](#)
The input [data.table::data.table\(\)](#).
- `primary_key` :: `character(1)`
Name of the primary key column.

[DataBackendDataTable](#) does not copy the input data, while [as_data_backend](#) calls [data.table::copy\(\)](#).
[as_data_backend](#) creates a primary key column as integer column if `primary_key` is `NULL`.

Fields

See [DataBackend](#).

Methods

See [DataBackend](#).

See Also

Other DataBackend: [DataBackendMatrix](#), [DataBackend](#), [as_data_backend\(\)](#)

Examples

```
data = as.data.table(iris)
data$id = seq_len(nrow(iris))
b = DataBackendDataTable$new(data = data, primary_key = "id")
print(b)
b$head()
b$data(rows = 100:101, cols = "Species")

b$nrow
head(b$rownames)

b$ncol
b$colnames

# alternative construction
as_data_backend(iris)
```

DataBackendMatrix *DataBackend for Matrix*

Description

[DataBackend](#) for **Matrix**. Data is stored as (sparse) matrix.

Format

[R6::R6Class](#) object inheriting from [DataBackend](#).

Construction

```
DataBackendMatrix$new(data, primary_key = NULL)
as_data_backend(data, primary_key = NULL, ...)
```

- `data` :: [Matrix::Matrix\(\)](#).
- `primary_key` :: `character(1)`
Not supported by this backend. Rows are addresses by their [rownames\(\)](#). If the matrix does not have row names, integer row indices are used.

Fields

See [DataBackend](#).

Methods

See [DataBackend](#).

See Also

Other DataBackend: [DataBackendDataTable](#), [DataBackend](#), [as_data_backend\(\)](#)

Examples

```
requireNamespace("Matrix")
data = Matrix::Matrix(sample(0:1, 20, replace = TRUE), ncol = 2)
colnames(data) = c("x1", "x2")
rownames(data) = paste0("row_", 1:10)

b = as_data_backend(data)
b$head()
b$data(b$rownames[1:3], b$colnames, data_format = "Matrix")
```

default_measures	<i>Get a Default Measure</i>
------------------	------------------------------

Description

Gets the default measures using the information in [mlr_reflections\\$default_measures](#):

- ["classif.ce"](#) for classification (["classif"](#)).
- ["regr.mse"](#) for regression (["regr"](#)).
- Add-on package may register additional default measures for their own task types.

Usage

```
default_measures(task_type)
```

Arguments

`task_type` :: character(1)
 Get the default measure for the task type `task_type`, e.g., ["classif"](#) or ["regr"](#).
 If `task_type` is NULL, an empty list is returned.

Value

list of [Measure](#).

Examples

```
default_measures("classif")
default_measures("regr")
```

Learner

*Learner Class***Description**

This is the abstract base class for learner objects like [LearnerClassif](#) and [LearnerRegr](#).

Learners consist of the following parts:

- Methods `train()` and `predict()` which call `train_internal()` and `predict_internal()`.
- The fitted model, after calling `train()`.
- A [paradox::ParamSet](#) which stores meta-information about available hyperparameters, and also stores hyperparameter settings.
- Meta-information about the requirements and capabilities of the learner.

Predefined learners are stored in the [mlr3misc::Dictionary mlr_learners](#), e.g. `classif.rpart` or `regr.rpart`. A guide on how to extend **mlr3** with custom learners can be found in the [mlr3book](#).

Format

[R6::R6Class](#) object.

Construction

Note: This object is typically constructed via a derived classes, e.g. [LearnerClassif](#) or [LearnerRegr](#).

```
l = Learner$new(id, task_type, param_set = ParamSet$new(), predict_types = character(),
  feature_types = character(), properties = character(), packages = character())
```

- `id` :: `character(1)`
Identifier for the learner.
- `task_type` :: `character(1)`
Type of the task the learner can operator on. E.g., "classif" or "regr".
- `param_set` :: [paradox::ParamSet](#)
Set of hyperparameters.
- `predict_types` :: `character()`
Supported predict types. Must be a subset of `mlr_reflections$learner_predict_types`.
- `predict_sets` :: `character()`
Sets to predict on during `resample()/benchmark()`. Creates and stores a separate [Prediction](#) object for each set. The individual sets can be combined via getters in [ResampleResult/BenchmarkResult](#), or [Measures](#) can be set to operate on subsets of the calculated prediction sets. Must be a non-empty subset of ("train", "test"). Default is "test".
- `feature_types` :: `character()`
Feature types the learner operates on. Must be a subset of `mlr_reflections$task_feature_types`.
- `properties` :: `character()`
Set of properties of the learner. Must be a subset of `mlr_reflections$learner_properties`. The following properties are currently standardized and understood by learners in **mlr3**:

- "missings": The learner can handle missing values in the data.
 - "weights": The learner supports observation weights.
 - "importance": The learner supports extraction of importance scores, i.e. comes with a `importance()` extractor function (see section on optional extractors).
 - "selected_features": The learner supports extraction of the set of selected features, i.e. comes with a `selected_features()` extractor function (see section on optional extractors).
 - "oob_error": The learner supports extraction of estimated out of bag error, i.e. comes with a `oob_error()` extractor function (see section on optional extractors).
- `data_formats` :: `character()`
Vector of supported data formats which can be processed during `$train()` and `$predict()`. Defaults to "data.table".
 - `packages` :: `character()`
Set of required packages. Note that these packages will be loaded via `requireNamespace()`, and are not attached.
 - `man` :: `character(1)`
String in the format `[pkg>::[topic]` pointing to a manual page for this object.

Fields

- `id` :: `character(1)`
Identifier of the learner.
- `task_type` :: `character(1)`
Stores the type of class this learner can operate on, e.g. "classif" or "regr". A complete list of task types is stored in `mlr_reflections$task_types`.
- `param_set` :: `paradox::ParamSet`
Description of available hyperparameters and hyperparameter settings.
- `predict_types` :: `character()`
Stores the possible predict types the learner is capable of. A complete list of candidate predict types, grouped by task type, is stored in `mlr_reflections$learner_predict_types`.
- `predict_type` :: `character(1)`
Stores the currently selected predict type. Must be an element of `l$predict_types`.
- `feature_types` :: `character()`
Stores the feature types the learner can handle, e.g. "logical", "numeric", or "factor". A complete list of candidate feature types, grouped by task type, is stored in `mlr_reflections$task_feature_types`.
- `properties` :: `character()`
Stores a set of properties/capabilities the learner has. A complete list of candidate properties, grouped by task type, is stored in `mlr_reflections$learner_properties`.
- `packages` :: `character()`
Stores the names of required packages.
- `state` :: `NULL | named list()`
Current (internal) state of the learner. Contains all information learnt during `train()` and `predict()`. It is not recommended to access elements from `state` directly, this is an internal data structure which may change in the future.

- `encapsulate(named character())`
How to call the code in `train_internal()` and `predict_internal()`. Must be a named character vector with names "train" and "predict". Possible values are "none", "evaluate" and "callr". See `mlr3misc::encapsulate()` for more details.
- `fallback(Learner)`
Learner which is fitted to impute predictions in case that either the model fitting or the prediction of the top learner is not successful. Requires you to enable encapsulation, otherwise errors are not caught and the execution is terminated before the fallback learner kicks in.
- `hash :: character(1)`
Hash (unique identifier) for this object.
- `model :: any`
The fitted model. Only available after `$train()` has been called.
- `timings :: numeric(2)`
Elapsed time in seconds for the steps "train" and "predict". Measured via `mlr3misc::encapsulate()`.
- `log :: data.table::data.table()`
Returns the output (including warning and errors) as table with columns "stage" (train or predict), "class" (output, warning, error) and "msg" (character()).
- `warnings :: character()`
Returns the logged warnings as vector.
- `errors :: character()`
Returns the logged errors as vector.

Methods

- `train(task, row_ids = NULL)`
(`Task`, `integer()` | `character()`) -> `self`
Train the learner on the row ids of the provided `Task`. Mutates the learner by reference, i.e. stores the model alongside other objects in field `$state`.
- `predict(task, row_ids = NULL)`
(`Task`, `integer()` | `character()`) -> `Prediction`
Uses the data stored during `$train()` in `$state` to create a new `Prediction` based on the provided `row_ids` of the task.
- `predict_newdata(newdata, task = NULL)`
(`data.frame()`, `Task`) -> `Prediction`
Uses the model fitted during `$train()` in to create a new `Prediction` based on the new data in `newdata`. Object `task` is the task used during `$train()` and required for conversions of `newdata`. If the learner's `$train()` method has been called, there is a (size reduced) version of the training task stored in the learner. If the learner has been fitted via `resample()` or `benchmark()`, you need to pass the corresponding task stored in the `ResampleResult` or `BenchmarkResult`, respectively.
- `reset()`
() -> `self`
Reset the learner, i.e. un-train by resetting the state.
- `help()`
() -> `NULL`
Opens the corresponding help page referenced by `$man`.

Optional Extractors

Specific learner implementations are free to implement additional getters to ease the access of certain parts of the model in the inherited subclasses.

For the following operations, extractors are standardized:

- `importance(...)`: Returns the feature importance score as numeric vector. The higher the score, the more important the variable. The returned vector is named with feature names and sorted in decreasing order. Note that the model might omit features it has not used at all. The learner must be tagged with property "importance". To filter variables using the importance scores, use package **mlr3filters**.
- `selected_features(...)`: Returns a subset of selected features as `character()`. The learner must be tagged with property "selected_features".
- `oob_error(...)`: Returns the out-of-bag error of the model as `numeric(1)`. The learner must be tagged with property "oob_error".

Setting Hyperparameters

All information about hyperparameters is stored in the slot `param_set` which is a [paradox::ParamSet](#). The printer gives an overview about the ids of available hyperparameters, their storage type, lower and upper bounds, possible levels (for factors), default values and assigned values. To set hyperparameters, assign a named list to the subplot values:

```
lrn = lrn("classif.rpart")
lrn$param_set$values = list(minsplit = 3, cp = 0.01)
```

Note that this operation replaces all previously set hyperparameter values. If you only intend to change one specific hyperparameter value and leave the others as-is, you can use the helper function [mlr3misc::insert_named\(\)](#):

```
lrn$param_set$values = mlr3misc::insert_named(lrn$param_set$values, list(cp = 0.001))
```

If the learner has additional hyperparameters which are not encoded in the [ParamSet](#), you can easily extend the learner. Here, we add a hyperparameter with id "foo" possible levels "a" and "b":

```
lrn$param_set$add(paradox::ParamFct$new("foo", levels = c("a", "b")))
```

See Also

Other Learner: [LearnerClassif](#), [LearnerRegr](#), [mlr_learners](#)

LearnerClassif

Classification Learner

Description

This Learner specializes [Learner](#) for classification problems.

Many predefined learners can be found in the `mlr3misc::Dictionary mlr_learners` after loading the **mlr3learners** package.

Format

`R6::R6Class` object inheriting from [Learner](#).

Construction

```
l = LearnerClassif$new(id, param_set = ParamSet$new(), predict_types = character(), feature_types = character(),  
  properties = character(), data_formats = "data.table", packages = character(), man = NA_character_)
```

For a description of the arguments, see [Learner](#). `task_type` is set to "classif".

Possible values for `predict_types` are passed to and converted by [PredictionClassif](#):

- "response": Predicts a class label for each observation in the test set.
- "prob": Predicts the posterior probability for each class for each observation in the test set.

Additional learner properties include:

- "twoclass": The learner works on binary classification problems.
- "multiclass": The learner works on multiclass classification problems.

Fields

See [Learner](#).

Methods

See [Learner](#).

See Also

Example classification learners: [classif.rpart](#)

Other Learner: [LearnerRegr](#), [Learner](#), [mlr_learners](#)

Examples

```
# get all classification learners from mlr_learners:
lrns = mlr_learners$mget(mlr_learners$keys("^classif"))
names(lrns)

# get a specific learner from mlr_learners:
lrn = lrn("classif.rpart")
print(lrn)

# train the learner:
task = tsk("iris")
lrn$train(task, 1:120)

# predict on new observations:
lrn$predict(task, 121:150)$confusion
```

LearnerRegr

Regression Learner

Description

This Learner specializes [Learner](#) for regression problems.

Many predefined learners can be found in the [mlr3misc::Dictionary mlr_learners](#) after loading the [mlr3learners](#) package.

Format

[R6::R6Class](#) object inheriting from [Learner](#).

Construction

```
l = LearnerRegr$new(id, param_set = ParamSet$new(), predict_types = character(), feature_types = character(),
  properties = character(), data_formats = "data.table", packages = character(), man = NA_character_)
```

For a description of the arguments, see [Learner](#). `task_type` is set to "regr".

Possible values for `predict_types` are passed to and converted by [PredictionRegr](#):

- "response": Predicts a numeric response for each observation in the test set.
- "se": Predicts the standard error for each value of response for each observation in the test set.

Fields

See [Learner](#).

Methods

See [Learner](#).

See Also

Example regression learners: [regr.rpart](#)

Other Learner: [LearnerClassif](#), [Learner](#), [mlr_learners](#)

Examples

```
# get all regression learners from mlr_learners:
lrns = mlr_learners$mget(mlr_learners$keys("^regr"))
names(lrns)
```

```
# get a specific learner from mlr_learners:
mlr_learners$get("regr.rpart")
lrn("classif.featureless")
```

Measure

Measure Class

Description

This is the abstract base class for measures like [MeasureClassif](#) and [MeasureRegr](#).

Measures are classes around tailored around two functions:

1. A function score which quantifies the performance by comparing true and predicted response.
2. A function aggregator which combines multiple performance scores returned by calculate to a single numeric value.

In addition to these two functions, meta-information about the performance measure is stored.

Predefined measures are stored in the [mlr3misc::Dictionary mlr_measures](#), e.g. [classif.auc](#) or [time_train](#). A guide on how to extend [mlr3](#) with custom measures can be found in the [mlr3book](#).

Format

[R6::R6Class](#) object.

Construction

Note: This object is typically constructed via a derived classes, e.g. [MeasureClassif](#) or [MeasureRegr](#).

```
m = Measure$new(id, task_type = NA, range = c(-Inf, Inf), minimize = NA, average = "macro",
  aggregator = NULL, properties = character(), predict_type = "response", predict_sets = "test",
  task_properties = character(), packages = character(), man = NA_character_)
```

- `id` :: `character(1)`
Identifier for the measure.
- `task_type` :: `character(1)`
Type of the task the measure can operator on. E.g., "classif" or "regr".

- `range :: numeric(2)`
Feasible range for this measure as `c(lower_bound, upper_bound)`. Both bounds may be infinite.
- `minimize :: logical(1)`
Set to TRUE if good predictions correspond to small values, and to FALSE if good predictions correspond to large values. If set to NA (default), tuning this measure is not possible.
- `average :: character(1)`
How to average multiple [Predictions](#) from a [ResampleResult](#).
The default, "macro", calculates the individual performances scores for each [Prediction](#) and then uses the function defined in `aggregator` to average them to a single number.
If set to "micro", the individual [Prediction](#) objects are first combined into a single new [Prediction](#) object which is then used to assess the performance. The function `aggregator` is not used in this case.
- `aggregator :: function(x)`
Function to aggregate individual performance scores `x` where `x` is a numeric vector. If NULL, defaults to `mean()`.
- `properties :: character()`
Properties of the measure. Must be a subset of `mlr_reflections$measure_properties`. Supported by `mlr3`:
 - "requires_task" (requires the complete [Task](#)),
 - "requires_learner" (requires the trained [Learner](#)),
 - "requires_train_set" (requires the training indices from the [Resampling](#)), and
 - "na_score" (the measure is expected to occasionally return NA).
- `predict_type :: character(1)`
Required predict type of the [Learner](#). Possible values are stored in `mlr_reflections$learner_predict_types`.
- `predict_sets :: character()`
Prediction sets to operate on, used in `aggregate()` to extract the matching `predict_sets` from the [ResampleResult](#). Multiple predict sets are calculated by the respective [Learner](#) during `resample()/benchmark()`. Must be a non-empty subset of `c("train", "test")`. If multiple sets are provided, these are first combined to a single prediction object. Default is "test".
- `task_properties :: character()`
Required task properties, see [Task](#).
- `packages :: character()`
Set of required packages. Note that these packages will be loaded via `requireNamespace()`, and are not attached.
- `man :: character(1)`
String in the format `[pkg]::[topic]` pointing to a manual page for this object.

Fields

All variables passed to the constructor.

Methods

- `aggregate(rr)`
[ResampleResult](#) -> numeric(1)
 Aggregates multiple performance scores into a single score using the aggregator function of the measure. Operates on the [Predictions](#) of [ResampleResult](#) with matching `predict_sets`.
- `score(prediction, task = NULL, learner = NULL, train_set = NULL)`
 ((named list of) [Prediction](#), [Task](#), [Learner](#), integer() | character()) -> numeric(1)
 Takes a [Prediction](#) (or a list of [Prediction](#) objects named with valid `predict_sets`) and calculates a numeric score. If the measure is flagged with the properties "requires_task", "requires_learner" or "requires_train_set", you must additionally pass the respective [Task](#), the trained [Learner](#) or the training set indices. This is handled internally during [resample\(\)/benchmark\(\)](#).
- `help()`
 () -> NULL
 Opens the corresponding help page referenced by `$man`.

See Also

Other Measure: [MeasureClassif](#), [MeasureRegr](#), [mlr_measures](#)

MeasureClassif

Classification Measure

Description

This measure specializes [Measure](#) for classification problems. Predefined measures can be found in the [mlr3misc::Dictionary mlr_measures](#).

Format

[R6::R6Class](#) object inheriting from [Measure](#).

Construction

```
m = MeasureClassif$new(id, range, minimize = NA, average = "macro", aggregator = NULL, properties = character(),
  predict_sets = "test", task_properties = character(), packages = character(), man = NA_character_)
```

For a description of the arguments, see [Measure](#). The `task_type` is set to "classif". Possible values for `predict_type` are "response" and "prob".

Fields

See [Measure](#).

Methods

See [Measure](#).

See Also

Example classification measures: [classif.ce](#)

Other Measure: [MeasureRegr](#), [Measure](#), [mlr_measures](#)

MeasureRegr

Regression Measure

Description

This measure specializes [Measure](#) for regression problems. Predefined measures can be found in the [mlr3misc::Dictionary mlr_measures](#).

Format

[R6::R6Class](#) object inheriting from [Measure](#).

Construction

```
m = MeasureRegr$new(id, range, minimize = NA, average = "macro", aggregator = NULL, properties = character(),  
  predict_sets = "test", task_properties = character(), packages = character())
```

For a description of the arguments, see [Measure](#). The `task_type` is set to "regr". Possible values for `predict_type` are "response" and "se".

Fields

See [Measure](#).

Methods

See [Measure](#).

See Also

Example regression measures: [regr.mse](#)

Other Measure: [MeasureClassif](#), [Measure](#), [mlr_measures](#)

`mlr_learners`*Dictionary of Learners*

Description

A simple `mlr3misc::Dictionary` storing objects of class `Learner`. Each learner has an associated help page, see `mlr_learners_[id]`.

This dictionary can get populated with additional learners by add-on packages. For more classification and regression learners, load the `mlr3learners` package.

For a more convenient way to retrieve and construct learners, see `lrn()`.

Format

`R6::R6Class` object inheriting from `mlr3misc::Dictionary`.

Methods

See `mlr3misc::Dictionary`.

S3 methods

- `as.data.table(dict)`
`mlr3misc::Dictionary` -> `data.table::data.table()`
Returns a `data.table::data.table()` with fields "key", "feature_types", "packages", "properties" and "predict_types" as columns.

See Also

Example learners: `classif.rpart`, `regr.rpart`, `classif.featureless`, `regr.featureless`, `classif.debug`

Sugar function: `lrn()`

Extension Packages: `mlr3learners`

Other Dictionary: `mlr_measures`, `mlr_resamplings`, `mlr_task_generators`, `mlr_tasks`

Other Learner: `LearnerClassif`, `LearnerRegr`, `Learner`

Examples

```
as.data.table(mlr_learners)
mlr_learners$get("classif.featureless")
lrn("classif.rpart")
```

`mlr_learners_classif.debug`*Classification Learner for Debugging*

Description

A simple [LearnerClassif](#) used primarily in the unit tests and for debugging purposes. If no hyperparameter is set, it simply constantly predicts a randomly selected label. The following hyperparameters trigger the following actions:

message_train: Outputs a message during train if the parameter value exceeds `runif(1)`.

message_predict: Outputs a message during predict if the parameter value exceeds `runif(1)`.

warning_train: Signals a warning during train if the parameter value exceeds `runif(1)`.

warning_predict: Signals a warning during predict if the parameter value exceeds `runif(1)`.

error_train: Raises an exception during train if the parameter value exceeds `runif(1)`.

error_predict: Raises an exception during predict if the parameter value exceeds `runif(1)`.

segfault_train: Provokes a segfault during train if the parameter value exceeds `runif(1)`.

segfault_predict: Provokes a segfault during predict if the parameter value exceeds `runif(1)`.

predict_missing Ratio of predictions which will be NA.

save_tasks: Saves input task in `model` slot during training and prediction.

x: Numeric parameter. Has no effect.

Note that segfaults may not work on your operating system. Also note that if they work, they will tear down your R session immediately!

Format

[R6::R6Class](#) inheriting from [LearnerClassif](#).

Construction

```
LearnerClassifDebug$new()  
mlr_learners$get("classif.debug")  
lrn("classif.debug")
```

See Also

Dictionary of [Learners](#): [mlr_learners](#)

`as.data.table(mlr_learners)` for a complete table of all (also dynamically created) [Learner](#) implementations.

Examples

```

learner = lrn("classif.debug")
learner$param_set$values = list(message_train = 1, save_tasks = TRUE)

# this should signal a message
task = tsk("iris")
learner$train(task)
learner$predict(task)

# task_train and task_predict are the input tasks for train() and predict()
names(learner$model)

```

```

mlr_learners_classif.featureless
      Featureless Classification Learner

```

Description

A simple [LearnerClassif](#) which only analyses the labels during train, ignoring all features. Hyperparameter method determines the mode of operation during prediction:

mode: Predicts the most frequent label. If there are two or more labels tied, randomly selects one per prediction.

sample: Randomly predict a label uniformly.

weighed.sample: Randomly predict a label, with probability estimated from the training distribution.

Format

[R6::R6Class](#) inheriting from [LearnerClassif](#).

Construction

```

LearnerClassifFeatureless$new()
mlr_learners$get("classif.featureless")
lrn("classif.featureless")

```

See Also

[Dictionary of Learners: mlr_learners](#)

as.data.table(mlr_learners) for a complete table of all (also dynamically created) [Learner](#) implementations.

mlr_learners_classif.rpart
Classification Tree Learner

Description

A [LearnerClassif](#) for a classification tree implemented in `rpart::rpart()` in package **rpart**. Parameter `xval` is set to 0 in order to save some computation time.

Format

[R6::R6Class](#) inheriting from [LearnerClassif](#).

Construction

```
LearnerClassifRpart$new()  
mlr_learners$get("classif.rpart")  
lrn("classif.rpart")
```

References

Breiman L, Friedman JH, Olshen RA, Stone CJ (2017). *Classification And Regression Trees*. Routledge. doi: [10.1201/9781315139470](https://doi.org/10.1201/9781315139470).

See Also

Dictionary of Learners: [mlr_learners](#)

`as.data.table(mlr_learners)` for a complete table of all (also dynamically created) [Learner](#) implementations.

mlr_learners_regr.featureless
Featureless Regression Learner

Description

A simple [LearnerRegr](#) which only analyses the response during train, ignoring all features. If hyperparameter `robust` is `FALSE` (default), constantly predicts `mean(y)` as response and `sd(y)` as standard error. If `robust` is `TRUE`, `median()` and `madn()` are used instead of `mean()` and `sd()`, respectively.

Format

[R6::R6Class](#) inheriting from [LearnerRegr](#).

Construction

```
LearnerRegrFeatureless$new()  
mlr_learners$get("regr.featureless")  
lrn("regr.featureless")
```

See Also

Dictionary of Learners: [mlr_learners](#)

`as.data.table(mlr_learners)` for a complete table of all (also dynamically created) [Learner](#) implementations.

`mlr_learners_regr.rpart`

Regression Tree Learner

Description

A [LearnerRegr](#) for a regression tree implemented in `rpart::rpart()` in package **rpart**. Parameter `xval` is set to 0 in order to save some computation time.

Format

[R6::R6Class](#) inheriting from [LearnerRegr](#).

Construction

```
LearnerRegrRpart$new()  
mlr_learners$get("regr.rpart")  
lrn("regr.rpart")
```

References

Breiman L, Friedman JH, Olshen RA, Stone CJ (2017). *Classification And Regression Trees*. Routledge. doi: [10.1201/9781315139470](https://doi.org/10.1201/9781315139470).

See Also

Dictionary of Learners: [mlr_learners](#)

`as.data.table(mlr_learners)` for a complete table of all (also dynamically created) [Learner](#) implementations.

mlr_measures	<i>Dictionary of Performance Measures</i>
--------------	---

Description

A simple [mlr3misc::Dictionary](#) storing objects of class [Measure](#). Each measure has an associated help page, see `mlr_measures_[id]`.

This dictionary can get populated with additional measures by add-on packages.

For a more convenient way to retrieve and construct measures, see [msr\(\)](#).

Format

[R6::R6Class](#) object inheriting from [mlr3misc::Dictionary](#).

Methods

See [mlr3misc::Dictionary](#).

S3 methods

- `as.data.table(dict)`
[mlr3misc::Dictionary](#) -> `data.table::data.table()`
Returns a `data.table::data.table()` with fields "key", "task_type", "predict_type", and "packages" as columns.

See Also

Example measures: [classif.auc](#), [time_train](#) Sugar function: [msr\(\)](#)

Other Dictionary: [mlr_learners](#), [mlr_resamplings](#), [mlr_task_generators](#), [mlr_tasks](#)

Other Measure: [MeasureClassif](#), [MeasureRegr](#), [Measure](#)

Examples

```
as.data.table(mlr_measures)
mlr_measures$get("classif.ce")
msr("regr.mse")
```

mlr_measures_classif.acc
Classification Accuracy

Description

Classification measure defined as

$$\frac{1}{n} \sum_{i=1}^n (t_i = r_i).$$

Format

R6: `:R6Class()` inheriting from `Measure`.

Construction

This measures can be retrieved from the dictionary `mlr_measures`:

```
mlr_measures$get("classif.acc")
msr("classif.acc")
```

Meta Information

- Type: "classif"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::acc()` from package `mlr3measures`.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: `mlr_measures`

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) `Measure` implementations.

Other classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other multiclass classification measures: [mlr_measures_classif.bacc](#), [mlr_measures_classif.ce](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.logloss](#)

mlr_measures_classif.auc

Area Under the ROC Curve

Description

Computes the area under the Receiver Operator Characteristic (ROC) curve. The AUC can be interpreted as the probability that a randomly chosen positive observation has a higher predicted probability than a randomly chosen negative observation.

Format

[R6::R6Class\(\)](#) inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.auc")
msr("classif.auc")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: prob

Note

The score function calls [mlr3measures::auc\(\)](#) from package [mlr3measures](#).

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: [mlr_measures_classif.acc](#), [mlr_measures_classif.bacc](#), [mlr_measures_classif.ce](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fbeta](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fnr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.logloss](#), [mlr_measures_classif.n](#)

mlr_measures_classif.npv, mlr_measures_classif.ppv, mlr_measures_classif.precision, mlr_measures_classif.recall, mlr_measures_classif.sensitivity, mlr_measures_classif.specificity, mlr_measures_classif.tnr, mlr_measures_classif.tn, mlr_measures_classif.tpr, mlr_measures_classif.tp

Other binary classification measures: mlr_measures_classif.dor, mlr_measures_classif.fbeta, mlr_measures_classif.fdr, mlr_measures_classif.fnr, mlr_measures_classif.fn, mlr_measures_classif.fomr, mlr_measures_classif.fpr, mlr_measures_classif.fp, mlr_measures_classif.mcc, mlr_measures_classif.npv, mlr_measures_classif.ppv, mlr_measures_classif.precision, mlr_measures_classif.recall, mlr_measures_classif.sensitivity, mlr_measures_classif.specificity, mlr_measures_classif.tnr, mlr_measures_classif.tn, mlr_measures_classif.tpr, mlr_measures_classif.tp

mlr_measures_classif.bacc

Balanced Accuracy

Description

Computes the weighted balanced accuracy, suitable for imbalanced data sets. It is defined analogously to the definition in [sklearn](#).

First, the sample weights w are normalized per class:

$$\hat{w}_i = \frac{w_i}{\sum_j 1(y_j = y_i)w_i}.$$

The balanced accuracy is calculated as

$$\frac{1}{\sum_i \hat{w}_i} \sum_i 1(r_i = t_i)\hat{w}_i.$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.bacc")
msr("classif.bacc")
```

Meta Information

- Type: "classif"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::bacc()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other multiclass classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.logloss`

`mlr_measures_classif.ce`

Classification Error

Description

Classification measure defined as

$$\frac{1}{n} \sum_{i=1}^n (t_i \neq r_i).$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.ce")
msr("classif.ce")
```

Meta Information

- Type: "classif"
- Range: [0, 1]
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::ce()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: [mlr_measures_classif.acc](#), [mlr_measures_classif.auc](#), [mlr_measures_classif.bacc](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fbeta](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fnr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.logloss](#), [mlr_measures_classif.npv](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#), [mlr_measures_classif.sensitivity](#), [mlr_measures_classif.specificity](#), [mlr_measures_classif.tnr](#), [mlr_measures_classif.tn](#), [mlr_measures_classif.tpr](#), [mlr_measures_classif.tp](#)

Other multiclass classification measures: [mlr_measures_classif.acc](#), [mlr_measures_classif.bacc](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.logloss](#)

`mlr_measures_classif.costs`

Cost-sensitive Classification Measure

Description

Uses a cost matrix to create a classification measure. True labels must be arranged in columns, predicted labels must be arranged in rows. The cost matrix is stored as slot `$costs`.

For calculation of the score, the confusion matrix is multiplied element-wise with the cost matrix. The costs are then summed up (and potentially divided by the number of observations if `normalize` is set to `TRUE`).

This measure requires the [Task](#) during scoring to ensure that the rows and columns of the cost matrix are in the same order as in the confusion matrix.

Format

`R6::R6Class()` inheriting from [MeasureClassif](#).

Construction

```
MeasureClassifCosts$new(costs = NULL, normalize = TRUE)
mlr_measures$get("classif.costs")
msr("classif.costs")
```


- `costs` :: `matrix()`
Numeric matrix of costs (truth in columns, predicted response in rows).
- `normalize` :: `logical(1)`
If TRUE, calculate the mean cost per observation instead of the total costs.

Meta Information

- Type: "classif"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: 'response'

See Also

Dictionary of Measures: [mlr_measures](#)

as.data.table(mlr_measures) for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.n`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other multiclass classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.logloss`

Examples

```
# get a cost sensitive task
task = tsk("german_credit")

# cost matrix as given on the UCI page of the german credit data set
# https://archive.ics.uci.edu/ml/datasets/statlog+(german+credit+data)
costs = matrix(c(0, 5, 1, 0), nrow = 2)
dimnames(costs) = list(truth = task$class_names, predicted = task$class_names)
print(costs)

# mlr3 needs truth in columns, predictions in rows
costs = t(costs)

# create measure which calculates the absolute costs
m = msr("classif.costs", id = "german_credit_costs", costs = costs, normalize = FALSE)

# fit models and calculate costs
learner = lrn("classif.rpart")
rr = resample(task, learner, rsmpl("cv", folds = 3))
rr$aggregate(m)
```

mlr_measures_classif.dor

Diagnostic Odds Ratio

Description

Binary classification measure defined as

$$\frac{TP/FP}{FN/TN}$$

Format

R6: `:R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.dor")
msr("classif.dor")
```

Meta Information

- Type: "binary"
- Range: $[0, \infty)$
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::dor()` from package [mlr3measures](#).

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: [mlr_measures_classif.acc](#), [mlr_measures_classif.auc](#), [mlr_measures_classif.bacc](#), [mlr_measures_classif.ce](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.fbeta](#), [mlr_measures_classif.f](#), [mlr_measures_classif.fnr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.logloss](#), [mlr_measures_classif.mcc](#), [mlr_measures_classif.n](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#),

mlr_measures_classif.sensitivity, mlr_measures_classif.specificity, mlr_measures_classif.tnr,
mlr_measures_classif.tn, mlr_measures_classif.tpr, mlr_measures_classif.tp

Other binary classification measures: mlr_measures_classif.auc, mlr_measures_classif.fbeta,
mlr_measures_classif.fdr, mlr_measures_classif.fnr, mlr_measures_classif.fn, mlr_measures_classif.fomr,
mlr_measures_classif.fpr, mlr_measures_classif.fp, mlr_measures_classif.mcc, mlr_measures_classif.npv,
mlr_measures_classif.ppv, mlr_measures_classif.precision, mlr_measures_classif.recall,
mlr_measures_classif.sensitivity, mlr_measures_classif.specificity, mlr_measures_classif.tnr,
mlr_measures_classif.tn, mlr_measures_classif.tpr, mlr_measures_classif.tp

mlr_measures_classif.fbeta

F-beta Score

Description

Binary classification measure defined with P as `precision()` and R as `recall()` as

$$(1 + \beta^2) \frac{P \cdot R}{(\beta^2 P) + R}.$$

It measures the effectiveness of retrieval with respect to a user who attaches β times as much importance to recall as precision. For $\beta = 1$, this measure is called "F1" score.

Format

`R6::R6Class()` inheriting from `Measure`.

Construction

This measures can be retrieved from the dictionary `mlr_measures`:

```
mlr_measures$get("classif.fbeta")
msr("classif.fbeta")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::fbeta()` from package `mlr3measures`.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.fdr`

False Discovery Rate

Description

Binary classification measure defined as

$$\frac{FP}{TP + FP}$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.fdr")
msr("classif.fdr")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::fdr()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: `mlr_measures`

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) **Measure** implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fb`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.n`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.fn`

False Negatives

Description

Classification measure counting the false negatives (type 2 error), i.e. the number of predictions indicating a negative class label while in fact it is positive. This is sometimes also called a "false alarm".

Format

`R6::R6Class()` inheriting from **Measure**.

Construction

This measures can be retrieved from the dictionary `mlr_measures`:

```
mlr_measures$get("classif.fn")
msr("classif.fn")
```

Meta Information

- Type: "binary"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::fn()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

[Dictionary of Measures: mlr_measures](#)

as `data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fb`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.n`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.fnr`

False Negative Rate

Description

Binary classification measure defined as

$$\frac{\text{FN}}{\text{TP} + \text{FN}}$$

Also know as "miss rate".

Format

[R6::R6Class\(\)](#) inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.fnr")
msr("classif.fnr")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: TRUE
- Required prediction: response

Note

The score function calls [mlr3measures::fnr\(\)](#) from package [mlr3measures](#).

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: [mlr_measures_classif.acc](#), [mlr_measures_classif.auc](#), [mlr_measures_classif.bacc](#), [mlr_measures_classif.ce](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fb](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.logloss](#), [mlr_measures_classif.mcc](#), [mlr_measures_classif.n](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#), [mlr_measures_classif.sensitivity](#), [mlr_measures_classif.specificity](#), [mlr_measures_classif.tnr](#), [mlr_measures_classif.tn](#), [mlr_measures_classif.tpr](#), [mlr_measures_classif.tp](#)

Other binary classification measures: [mlr_measures_classif.auc](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fbeta](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.mcc](#), [mlr_measures_classif.npv](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#), [mlr_measures_classif.sensitivity](#), [mlr_measures_classif.specificity](#), [mlr_measures_classif.tnr](#), [mlr_measures_classif.tn](#), [mlr_measures_classif.tpr](#), [mlr_measures_classif.tp](#)

 mlr_measures_classif.fomr

False Omission Rate

Description

Binary classification measure defined as

$$\frac{FN}{FN + TN}$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.fomr")
msr("classif.fomr")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::fomr()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: [mlr_measures_classif.acc](#), [mlr_measures_classif.auc](#), [mlr_measures_classif.bacc](#), [mlr_measures_classif.ce](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fbc](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fnr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.logloss](#), [mlr_measures_classif.mcc](#), [mlr_measures_classif.n](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#),


```
mlr_measures_classif.sensitivity,mlr_measures_classif.specificity,mlr_measures_classif.tnr,  
mlr_measures_classif.tn,mlr_measures_classif.tpr,mlr_measures_classif.tp
```

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`,
`mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fpr`,
`mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`,
`mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`,
`mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`,
`mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

```
mlr_measures_classif.fp
```

False Positives

Description

Classification measure counting the false positives (type 1 error), i.e. the number of predictions indicating a positive class label while in fact it is negative.

Format

`R6::R6Class()` inheriting from `Measure`.

Construction

This measures can be retrieved from the dictionary `mlr_measures`:

```
mlr_measures$get("classif.fp")  
msr("classif.fp")
```

Meta Information

- Type: "binary"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::fp()` from package `mlr3measures`.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fb`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.fpr`

False Positive Rate

Description

Binary classification measure defined as

$$\frac{FP}{FP + TN}$$

Also know as fall out or probability of false alarm.

Format

`R6: :R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.fpr")
msr("classif.fpr")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::fpr()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

as `data.table(mlr_measures)` for a complete table of all (also dynamically created) **Measure** implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fb`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.logloss`
Log Loss

Description

Classification measure defined as

$$-\frac{1}{n} \sum_{i=1}^n \log(p_i)$$

where p_i is the probability for the true class of observation i .

Format

`R6::R6Class()` inheriting from **Measure**.

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.logloss")
msr("classif.logloss")
```

Meta Information

- Type: "classif"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: prob

Note

The score function calls `mlr3measures::logloss()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fb`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other multiclass classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`

`mlr_measures_classif.mcc`

Matthews Correlation Coefficient

Description

Binary classification measure defined as

$$\frac{TP \cdot TN - FP \cdot FN}{\sqrt{(TP + FP)(TP + FN)(TN + FP)(TN + FN)}}$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measure can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.mcc")
msr("classif.mcc")
```

Meta Information

- Type: "binary"
- Range: $[-1, 1]$
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::mcc()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

 mlr_measures_classif.npv

Negative Predictive Value

Description

Binary classification measure defined as

$$\frac{TN}{FN + TN}$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.npv")
msr("classif.npv")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls [mlr3measures::npv\(\)](#) from package [mlr3measures](#).

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: [mlr_measures_classif.acc](#), [mlr_measures_classif.auc](#), [mlr_measures_classif.bacc](#), [mlr_measures_classif.ce](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fb](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fnr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.logloss](#), [mlr_measures_classif.n](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#),

mlr_measures_classif.sensitivity, mlr_measures_classif.specificity, mlr_measures_classif.tnr,
mlr_measures_classif.tn, mlr_measures_classif.tpr, mlr_measures_classif.tp

Other binary classification measures: mlr_measures_classif.auc, mlr_measures_classif.dor,
mlr_measures_classif.fbeta, mlr_measures_classif.fdr, mlr_measures_classif.fnr, mlr_measures_classif.fpr,
mlr_measures_classif.fomr, mlr_measures_classif.fpr, mlr_measures_classif.fp, mlr_measures_classif.mcc,
mlr_measures_classif.ppv, mlr_measures_classif.precision, mlr_measures_classif.recall,
mlr_measures_classif.sensitivity, mlr_measures_classif.specificity, mlr_measures_classif.tnr,
mlr_measures_classif.tn, mlr_measures_classif.tpr, mlr_measures_classif.tp

mlr_measures_classif.ppv

Positive Predictive Value

Description

Binary classification measure defined as

$$\frac{TP}{TP + FP}$$

Also know as "precision".

Format

R6: `:R6Class()` inheriting from `Measure`.

Construction

This measures can be retrieved from the dictionary `mlr_measures`:

```
mlr_measures$get("classif.ppv")
msr("classif.ppv")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::ppv()` from package `mlr3measures`.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fb`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.n`, `mlr_measures_classif.npv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.precision`

Positive Predictive Value

Description

Binary classification measure defined as

$$\frac{TP}{TP + FP}$$

Also know as "precision".

Format

`R6: :R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.precision")
msr("classif.precision")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::precision()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.recall`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.recall`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.recall`

True Positive Rate

Description

Binary classification measure defined as

$$\frac{TP}{TP + FN}$$

Also know as "recall" or "sensitivity".

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.recall")
msr("classif.recall")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::recall()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.sensitivity`

True Positive Rate

Description

Binary classification measure defined as

$$\frac{TP}{TP + FN}$$

Also know as "recall" or "sensitivity".

Format

[R6::R6Class\(\)](#) inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.sensitivity")
msr("classif.sensitivity")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls [mlr3measures::sensitivity\(\)](#) from package [mlr3measures](#).

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: [mlr_measures_classif.acc](#), [mlr_measures_classif.auc](#), [mlr_measures_classif.bacc](#), [mlr_measures_classif.ce](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fbeta](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fnr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.logloss](#), [mlr_measures_classif.n](#), [mlr_measures_classif.npv](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#), [mlr_measures_classif.specificity](#), [mlr_measures_classif.tnr](#), [mlr_measures_classif.tn](#), [mlr_measures_classif.tpr](#), [mlr_measures_classif.tp](#)

Other binary classification measures: [mlr_measures_classif.auc](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fbeta](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fnr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.mcc](#), [mlr_measures_classif.npv](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#), [mlr_measures_classif.specificity](#), [mlr_measures_classif.tnr](#), [mlr_measures_classif.tn](#), [mlr_measures_classif.tpr](#), [mlr_measures_classif.tp](#)

mlr_measures_classif.specificity
True Negative Rate

Description

Binary classification measure defined as

$$\frac{TN}{FP + TN}$$

Also know as "specificity".

Format

R6: `:R6Class()` inheriting from `Measure`.

Construction

This measures can be retrieved from the dictionary `mlr_measures`:

```
mlr_measures$get("classif.specificity")
msr("classif.specificity")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::specificity()` from package `mlr3measures`.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: `mlr_measures`

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) `Measure` implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbc`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.n`

```
mlr_measures_classif.npv, mlr_measures_classif.ppv, mlr_measures_classif.precision,
mlr_measures_classif.recall, mlr_measures_classif.sensitivity, mlr_measures_classif.tnr,
mlr_measures_classif.tn, mlr_measures_classif.tpr, mlr_measures_classif.tp
```

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`,
`mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fr`,
`mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`,
`mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`,
`mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.tnr`,
`mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

```
mlr_measures_classif.tn
```

True Negatives

Description

Classification measure counting the true negatives, i.e. the number of predictions correctly indicating a negative class label.

Format

`R6::R6Class()` inheriting from `Measure`.

Construction

This measure can be retrieved from the dictionary `mlr_measures`:

```
mlr_measures$get("classif.tn")
msr("classif.tn")
```

Meta Information

- Type: "binary"
- Range: $[0, \infty)$
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::tn()` from package `mlr3measures`.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fb`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.tnr`

True Negative Rate

Description

Binary classification measure defined as

$$\frac{TN}{FP + TN}$$

Also know as "specificity".

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.tnr")
msr("classif.tnr")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::tnr()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

as.data.table(mlr_measures) for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.n`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`, `mlr_measures_classif.tp`

`mlr_measures_classif.tp`

True Positives

Description

Binary classification measure counting the true positives, i.e. the number of predictions correctly indicating a positive class label.

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.tp")
msr("classif.tp")
```

Meta Information

- Type: "binary"
- Range: $[0, \infty)$
- Minimize: FALSE
- Required prediction: response

Note

The score function calls `mlr3measures::tp()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

[Dictionary of Measures: mlr_measures](#)

as `data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: `mlr_measures_classif.acc`, `mlr_measures_classif.auc`, `mlr_measures_classif.bacc`, `mlr_measures_classif.ce`, `mlr_measures_classif.costs`, `mlr_measures_classif.dor`, `mlr_measures_classif.fb`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.logloss`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`

Other binary classification measures: `mlr_measures_classif.auc`, `mlr_measures_classif.dor`, `mlr_measures_classif.fb`, `mlr_measures_classif.fbeta`, `mlr_measures_classif.fdr`, `mlr_measures_classif.fnr`, `mlr_measures_classif.fn`, `mlr_measures_classif.fomr`, `mlr_measures_classif.fpr`, `mlr_measures_classif.fp`, `mlr_measures_classif.mcc`, `mlr_measures_classif.npv`, `mlr_measures_classif.ppv`, `mlr_measures_classif.precision`, `mlr_measures_classif.recall`, `mlr_measures_classif.sensitivity`, `mlr_measures_classif.specificity`, `mlr_measures_classif.tnr`, `mlr_measures_classif.tn`, `mlr_measures_classif.tpr`

`mlr_measures_classif.tpr`

True Positive Rate

Description

Binary classification measure defined as

$$\frac{TP}{TP + FN}$$

Also know as "recall" or "sensitivity".

Format

[R6::R6Class\(\)](#) inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.tpr")
msr("classif.tpr")
```

Meta Information

- Type: "binary"
- Range: [0, 1]
- Minimize: FALSE
- Required prediction: response

Note

The score function calls [mlr3measures::tpr\(\)](#) from package [mlr3measures](#).

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other classification measures: [mlr_measures_classif.acc](#), [mlr_measures_classif.auc](#), [mlr_measures_classif.bacc](#), [mlr_measures_classif.ce](#), [mlr_measures_classif.costs](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fbeta](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fnr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.logloss](#), [mlr_measures_classif.n](#), [mlr_measures_classif.npv](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#), [mlr_measures_classif.sensitivity](#), [mlr_measures_classif.specificity](#), [mlr_measures_classif.tnr](#), [mlr_measures_classif.tn](#), [mlr_measures_classif.tp](#)

Other binary classification measures: [mlr_measures_classif.auc](#), [mlr_measures_classif.dor](#), [mlr_measures_classif.fbeta](#), [mlr_measures_classif.fdr](#), [mlr_measures_classif.fnr](#), [mlr_measures_classif.fn](#), [mlr_measures_classif.fomr](#), [mlr_measures_classif.fpr](#), [mlr_measures_classif.fp](#), [mlr_measures_classif.mcc](#), [mlr_measures_classif.npv](#), [mlr_measures_classif.ppv](#), [mlr_measures_classif.precision](#), [mlr_measures_classif.recall](#), [mlr_measures_classif.sensitivity](#), [mlr_measures_classif.specificity](#), [mlr_measures_classif.tnr](#), [mlr_measures_classif.tn](#), [mlr_measures_classif.tp](#)

mlr_measures_debug *Debug Measure*

Description

This measure returns the number of observations in the [Prediction](#) object. Its main purpose is debugging.

Format

[R6::R6Class\(\)](#) inheriting from [Measure](#).

Construction

```
MeasureDebug$new(na_ratio = 0)
mlr_measures$get("debug")
msr("debug")
```

- `na_ratio` :: `numeric(1)`
Ratio of scores which should be NA. Default is 0.

Fields

- `na_ratio` :: `numeric(1)`.

Meta Information

- Type: NA
- Range: $[0, \infty)$
- Minimize: NA
- Required prediction: 'response'

See Also

[Dictionary of Measures: mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Examples

```
task = tsk("wine")
learner = lrn("classif.featureless")
measure = msr("debug")
rr = resample(task, learner, rsmp("cv", folds = 3))
rr$score(measure)
```

mlr_measures_elapsed_time
Elapsed Time Measure

Description

Measures the elapsed time during train ("time_train"), predict ("time_predict"), or both ("time_both").

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

```
MeasureElapsedTime$new(id, stages)
```

```
mlr_measures$get("time_train")  
mlr_measures$get("time_predict")  
mlr_measures$get("time_both")
```

```
msr$get("time_train")  
msr$get("time_predict")  
msr$get("time_both")
```

- `id` :: character(1)
Id for the created measure.
- `stages` :: character()
Subset of ("train", "predict"). The runtime of all stages will be summed.

Meta Information

- Type: NA
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: 'response'

See Also

[Dictionary of Measures](#): [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

 mlr_measures_oob_error

Out-of-bag Error Measure

Description

Returns the out-of-bag error of the [Learner](#) for learners that support it (learners with property "oob_error"). Returns NA for unsupported learners.

Format

[R6](#): `R6Class()` inheriting from [Measure](#).

Construction

```
MeasureOOBError$new()
mlr_measures$get("oob_error")
msr("oob_error")
```

Meta Information

- Type: NA
- Range: $(-\infty, \infty)$
- Minimize: NA
- Required prediction: 'response'

See Also

[Dictionary of Measures](#): [mlr_measures](#)

as.data.table(mlr_measures) for a complete table of all (also dynamically created) [Measure](#) implementations.

mlr_measures_regr.bias

Bias

Description

Regression measure defined as

$$\frac{1}{n} \sum_{i=1}^n (t_i - r_i).$$

Good predictions score close to 0.

Format

`R6::R6Class()` inheriting from `Measure`.

Construction

This measures can be retrieved from the dictionary `mlr_measures`:

```
mlr_measures$get("classif.bias")
msr("classif.bias")
```

Meta Information

- Type: "regr"
- Range: $(-\infty, \infty)$
- Minimize: NA
- Required prediction: response

Note

The score function calls `mlr3measures::bias()` from package `mlr3measures`.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: `mlr_measures`

as `data.table(mlr_measures)` for a complete table of all (also dynamically created) `Measure` implementations.

Other regression measures: `mlr_measures_regr.ktau`, `mlr_measures_regr.mae`, `mlr_measures_regr.mape`, `mlr_measures_regr.maxae`, `mlr_measures_regr.medae`, `mlr_measures_regr.medse`, `mlr_measures_regr.mse`, `mlr_measures_regr.msle`, `mlr_measures_regr.pbias`, `mlr_measures_regr.rae`, `mlr_measures_regr.rmse`, `mlr_measures_regr.rmsle`, `mlr_measures_regr.rrse`, `mlr_measures_regr.rse`, `mlr_measures_regr.rsq`, `mlr_measures_regr.sae`, `mlr_measures_regr.smape`, `mlr_measures_regr.srho`, `mlr_measures_regr.sse`

`mlr_measures_regr.ktau`

Kendall's tau

Description

Regression measure defined as Kendall's rank correlation coefficient between truth and response. Calls `stats::cor()` with method set to "kendall".

Format

`R6::R6Class()` inheriting from `Measure`.

Construction

This measure can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.ktau")
msr("classif.ktau")
```

Meta Information

- Type: "regr"
- Range: $[-1, 1]$
- Minimize: FALSE
- Required prediction: response

Note

The score function calls [mlr3measures::ktau\(\)](#) from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

`mlr_measures_regr.mae` *Mean Absolute Errors*

Description

Regression measure defined as

$$\frac{1}{n} \sum_{i=1}^n |t_i - r_i|.$$

Format

[R6::R6Class\(\)](#) inheriting from [Measure](#).

Construction

This measure can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.mae")
msr("classif.mae")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::mae()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

`mlr_measures_regr.mape`

Mean Absolute Percent Error

Description

Regression measure defined as

$$\frac{1}{n} \sum_{i=1}^n \left| \frac{t_i - r_i}{t_i} \right|.$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measure can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.mape")
msr("classif.mape")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::mape()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: `mlr_measures_regr.bias`, `mlr_measures_regr.ktau`, `mlr_measures_regr.mae`, `mlr_measures_regr.maxae`, `mlr_measures_regr.medae`, `mlr_measures_regr.medse`, `mlr_measures_regr.mse`, `mlr_measures_regr.msle`, `mlr_measures_regr.pbias`, `mlr_measures_regr.rae`, `mlr_measures_regr.rmse`, `mlr_measures_regr.rmsle`, `mlr_measures_regr.rrse`, `mlr_measures_regr.rse`, `mlr_measures_regr.rsq`, `mlr_measures_regr.sae`, `mlr_measures_regr.smape`, `mlr_measures_regr.srho`, `mlr_measures_regr.sse`

`mlr_measures_regr.maxae`

Max Absolute Error

Description

Regression measure defined as

$$\max(|t_i - r_i|).$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.maxae")
msr("classif.maxae")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::maxae()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: `mlr_measures_regr.bias`, `mlr_measures_regr.ktau`, `mlr_measures_regr.mae`, `mlr_measures_regr.mape`, `mlr_measures_regr.medae`, `mlr_measures_regr.medse`, `mlr_measures_regr.mse`, `mlr_measures_regr.msle`, `mlr_measures_regr.pbias`, `mlr_measures_regr.rae`, `mlr_measures_regr.rmse`, `mlr_measures_regr.rmsle`, `mlr_measures_regr.rrse`, `mlr_measures_regr.rse`, `mlr_measures_regr.rsq`, `mlr_measures_regr.sae`, `mlr_measures_regr.smape`, `mlr_measures_regr.srho`, `mlr_measures_regr.sse`

`mlr_measures_regr.medae`

Median Absolute Errors

Description

Regression measure defined as

$$\operatorname{median}_i |t_i - r_i|.$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.medae")
msr("classif.medae")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::medae()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

mlr_measures_regr.medse

Median Squared Error

Description

Regression measure defined as

$$\operatorname{median}_i \left[(t_i - r_i)^2 \right].$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measure can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.medse")
msr("classif.medse")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::medse()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: `mlr_measures_regr.bias`, `mlr_measures_regr.ktau`, `mlr_measures_regr.mae`, `mlr_measures_regr.mape`, `mlr_measures_regr.maxae`, `mlr_measures_regr.medae`, `mlr_measures_regr.mse`, `mlr_measures_regr.msle`, `mlr_measures_regr.pbias`, `mlr_measures_regr.rae`, `mlr_measures_regr.rmse`, `mlr_measures_regr.rmsle`, `mlr_measures_regr.rrse`, `mlr_measures_regr.rse`, `mlr_measures_regr.rsq`, `mlr_measures_regr.sae`, `mlr_measures_regr.smape`, `mlr_measures_regr.srho`, `mlr_measures_regr.sse`

`mlr_measures_regr.mse` *Mean Squared Error*

Description

Regression measure defined as

$$\frac{1}{n} \sum_{i=1}^n (t_i - r_i)^2.$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.mse")
msr("classif.mse")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::mse()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

as.data.table(mlr_measures) for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

mlr_measures_regr.msle

Mean Squared Log Error

Description

Regression measure defined as

$$\frac{1}{n} \sum_{i=1}^n (\ln(1 + t_i) - \ln(1 + r_i))^2.$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.msle")
msr("classif.msle")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::msle()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: `mlr_measures_regr.bias`, `mlr_measures_regr.ktau`, `mlr_measures_regr.mae`, `mlr_measures_regr.mape`, `mlr_measures_regr.maxae`, `mlr_measures_regr.medae`, `mlr_measures_regr.medse`, `mlr_measures_regr.mse`, `mlr_measures_regr.pbias`, `mlr_measures_regr.rae`, `mlr_measures_regr.rmse`, `mlr_measures_regr.rmsle`, `mlr_measures_regr.rrse`, `mlr_measures_regr.rse`, `mlr_measures_regr.rsq`, `mlr_measures_regr.sae`, `mlr_measures_regr.smape`, `mlr_measures_regr.srho`, `mlr_measures_regr.sse`

`mlr_measures_regr.pbias`

Percent Bias

Description

Regression measure defined as

$$\frac{1}{n} \sum_{i=1}^n \frac{(t_i - r_i)}{|t_i|}.$$

Good predictions score close to 0.

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.pbias")
msr("classif.pbias")
```

Meta Information

- Type: "regr"
- Range: $(-\infty, \infty)$
- Minimize: NA
- Required prediction: response

Note

The score function calls `mlr3measures::pbias()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

[Dictionary of Measures: mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

`mlr_measures_regr.rae` *Relative Absolute Error*

Description

Regression measure defined as

$$\frac{\sum_{i=1}^n |t_i - r_i|}{\sum_{i=1}^n |t_i - \bar{t}|}$$

Can be interpreted as absolute error of the predictions relative to a naive model predicting the mean.

Format

[R6::R6Class\(\)](#) inheriting from [Measure](#).

Construction

This measure can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.rae")
msr("classif.rae")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::rae()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

`mlr_measures_regr.rmse`

Root Mean Squared Error

Description

Regression measure defined as

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (t_i - r_i)^2}.$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.rmse")
msr("classif.rmse")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::rmse()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

`mlr_measures_regr.rmsle`

Root Mean Squared Log Error

Description

Regression measure defined as

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (\ln(1 + t_i) - \ln(1 + r_i))^2}.$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measure can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.rmsle")
msr("classif.rmsle")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::rmsle()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

`mlr_measures_regr.rrse`

Root Relative Squared Error

Description

Regression measure defined as

$$\sqrt{\frac{\sum_{i=1}^n (t_i - r_i)^2}{\sum_{i=1}^n (t_i - \bar{t})^2}}$$

Can be interpreted as root of the squared error of the predictions relative to a naive model predicting the mean.

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.rrse")
msr("classif.rrse")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::rrse()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

[Dictionary of Measures: mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: `mlr_measures_regr.bias`, `mlr_measures_regr.ktau`, `mlr_measures_regr.mae`, `mlr_measures_regr.mape`, `mlr_measures_regr.maxae`, `mlr_measures_regr.medae`, `mlr_measures_regr.medse`, `mlr_measures_regr.mse`, `mlr_measures_regr.msle`, `mlr_measures_regr.pbias`, `mlr_measures_regr.rae`, `mlr_measures_regr.rmse`, `mlr_measures_regr.rmsle`, `mlr_measures_regr.rse`, `mlr_measures_regr.rsq`, `mlr_measures_regr.sae`, `mlr_measures_regr.smape`, `mlr_measures_regr.srho`, `mlr_measures_regr.sse`

`mlr_measures_regr.rse` *Relative Squared Error*

Description

Regression measure defined as

$$\frac{\sum_{i=1}^n (t_i - r_i)^2}{\sum_{i=1}^n (t_i - \bar{t})^2}.$$

Can be interpreted as squared error of the predictions relative to a naive model predicting the mean.

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.rse")
msr("classif.rse")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::rse()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

[Dictionary of Measures: mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: `mlr_measures_regr.bias`, `mlr_measures_regr.ktau`, `mlr_measures_regr.mae`, `mlr_measures_regr.mape`, `mlr_measures_regr.maxae`, `mlr_measures_regr.medae`, `mlr_measures_regr.medse`, `mlr_measures_regr.mse`, `mlr_measures_regr.msle`, `mlr_measures_regr.pbias`, `mlr_measures_regr.rae`, `mlr_measures_regr.rmse`, `mlr_measures_regr.rmsle`, `mlr_measures_regr.rrse`, `mlr_measures_regr.rsq`, `mlr_measures_regr.sae`, `mlr_measures_regr.smape`, `mlr_measures_regr.srho`, `mlr_measures_regr.sse`

`mlr_measures_regr.rsq` *R Squared*

Description

Regression measure defined as

$$1 - \frac{\sum_{i=1}^n (t_i - r_i)^2}{\sum_{i=1}^n (t_i - \bar{t})^2}.$$

Also known as coefficient of determination or explained variation. Subtracts the `rse()` from 1, hence it compares the squared error of the predictions relative to a naive model predicting the mean.

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.rsq")
msr("classif.rsq")
```

Meta Information

- Type: "regr"
- Range: $(-\infty, 1]$
- Minimize: FALSE
- Required prediction: response

Note

The score function calls [mlr3measures::rsq\(\)](#) from package [mlr3measures](#).

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

`mlr_measures_regr.sae` *Sum of Absolute Errors*

Description

Regression measure defined as

$$\sum_{i=1}^n |t_i - r_i|.$$

Format

[R6::R6Class\(\)](#) inheriting from [Measure](#).

Construction

This measure can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.sae")
msr("classif.sae")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::sae()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#), [mlr_measures_regr.sse](#)

`mlr_measures_regr.smape`

Symmetric Mean Absolute Percent Error

Description

Regression measure defined as

$$\frac{2}{n} \sum_{i=1}^n \frac{|t_i - r_i|}{|t_i| + |r_i|}$$

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.smape")
msr("classif.smape")
```

Meta Information

- Type: "regr"
- Range: [0, 2]
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::smape()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: `mlr_measures_regr.bias`, `mlr_measures_regr.ktau`, `mlr_measures_regr.mae`, `mlr_measures_regr.mape`, `mlr_measures_regr.maxae`, `mlr_measures_regr.medae`, `mlr_measures_regr.medse`, `mlr_measures_regr.mse`, `mlr_measures_regr.msle`, `mlr_measures_regr.pbias`, `mlr_measures_regr.rae`, `mlr_measures_regr.rmse`, `mlr_measures_regr.rmsle`, `mlr_measures_regr.rrse`, `mlr_measures_regr.rse`, `mlr_measures_regr.rsq`, `mlr_measures_regr.sae`, `mlr_measures_regr.srho`, `mlr_measures_regr.sse`

`mlr_measures_regr.srho`

Spearman's rho

Description

Regression measures defined as Spearman's rank correlation coefficient between truth and response. Calls `stats::cor()` with method set to "spearman".

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

This measure can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.srho")
msr("classif.srho")
```

Meta Information

- Type: "regr"
- Range: $[-1, 1]$
- Minimize: FALSE
- Required prediction: response

Note

The score function calls [mlr3measures::srho\(\)](#) from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.sse](#)

`mlr_measures_regr.sse` *Sum of Squared Errors*

Description

Regression measure defined as

$$\sum_{i=1}^n (t_i - r_i)^2.$$

Format

[R6::R6Class\(\)](#) inheriting from [Measure](#).

Construction

This measures can be retrieved from the dictionary [mlr_measures](#):

```
mlr_measures$get("classif.sse")
msr("classif.sse")
```

Meta Information

- Type: "regr"
- Range: $[0, \infty)$
- Minimize: TRUE
- Required prediction: response

Note

The score function calls `mlr3measures::sse()` from package **mlr3measures**.

If the measure is undefined for the input, NaN is returned. This can be customized by setting the field `na_value`.

See Also

Dictionary of Measures: [mlr_measures](#)

as.data.table(mlr_measures) for a complete table of all (also dynamically created) [Measure](#) implementations.

Other regression measures: [mlr_measures_regr.bias](#), [mlr_measures_regr.ktau](#), [mlr_measures_regr.mae](#), [mlr_measures_regr.mape](#), [mlr_measures_regr.maxae](#), [mlr_measures_regr.medae](#), [mlr_measures_regr.medse](#), [mlr_measures_regr.mse](#), [mlr_measures_regr.msle](#), [mlr_measures_regr.pbias](#), [mlr_measures_regr.rae](#), [mlr_measures_regr.rmse](#), [mlr_measures_regr.rmsle](#), [mlr_measures_regr.rrse](#), [mlr_measures_regr.rse](#), [mlr_measures_regr.rsq](#), [mlr_measures_regr.sae](#), [mlr_measures_regr.smape](#), [mlr_measures_regr.srho](#)

mlr_measures_selected_features

Selected Features Measure

Description

Measures the number of selected features by extracting it from learners with property "selected_features". If the learner does not support this, NA is returned.

This measure requires the [Task](#) and the [Learner](#) for scoring.

Format

`R6::R6Class()` inheriting from [Measure](#).

Construction

```
MeasureSelectedFeatures$new(normalize = FALSE)
mlr_measures$get("selected_features")
msr("selected_features")
```

- `normalize` :: `logical(1)`
If `normalize` is set to `TRUE`, divides the number of features by the total number of features.

Meta Information

- Type: NA
- Range: $[0, \infty)$
- Minimize: `TRUE`
- Required prediction: 'response'

See Also

[Dictionary of Measures: mlr_measures](#)

`as.data.table(mlr_measures)` for a complete table of all (also dynamically created) [Measure](#) implementations.

mlr_resamplings

Dictionary of Resampling Strategies

Description

A simple [mlr3misc::Dictionary](#) storing objects of class [Resampling](#). Each resampling has an associated help page, see `mlr_resamplings_[id]`.

This dictionary can get populated with additional resampling strategies by add-on packages.

For a more convenient way to retrieve and construct resampling strategies, see [rsmp\(\)](#).

Format

[R6::R6Class](#) object inheriting from [mlr3misc::Dictionary](#).

Methods

See [mlr3misc::Dictionary](#).

S3 methods

- `as.data.table(dict)`
[mlr3misc::Dictionary](#) -> [data.table::data.table\(\)](#)
Returns a [data.table::data.table\(\)](#) with columns "key", "params", and "iters".

See Also

Example resamplings: [cv](#), [bootstrap](#)

Sugar function: [rsmpl\(\)](#)

Other Dictionary: [mlr_learners](#), [mlr_measures](#), [mlr_task_generators](#), [mlr_tasks](#)

Other Resampling: [Resampling](#)

Examples

```
as.data.table(mlr_resamplings)
mlr_resamplings$get("cv")
rsmpl("subsampling")
```

mlr_resamplings_bootstrap
Bootstrap Resampling

Description

Splits data into bootstrap samples (sampling with replacement). Hyperparameters are the number of bootstrap iterations (`repeats`, default: 30) and the ratio of observations to draw per iteration (`ratio`, default: 1) for the training set.

Format

[R6::R6Class](#) inheriting from [Resampling](#).

Construction

```
ResamplingBootstrap$new()
mlr_resamplings$get("bootstrap")
rsmpl("bootstrap")
```

Fields

See [Resampling](#).

Methods

See [Resampling](#).

Parameters

- `repeats` :: integer(1)
Number of repetitions.
- `ratio` :: numeric(1)
Ratio of observations to put into the training set.

See Also

Dictionary of Resamplings: [mlr_resamplings](#)

`as.data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [Resampling](#) implementations.

Examples

```
# Create a task with 10 observations
task = tsk("iris")
task$filter(1:10)

# Instantiate Resampling
rb = rsmpl("bootstrap", repeats = 2, ratio = 1)
rb$instantiate(task)

# Individual sets:
rb$train_set(1)
rb$test_set(1)
intersect(rb$train_set(1), rb$test_set(1))

# Internal storage:
rb$instance$M # Matrix of counts
```

mlr_resamplings_custom

Custom Resampling

Description

Splits data into training and test sets using manually provided indices.

Format

[R6::R6Class](#) inheriting from [Resampling](#).

Construction

```
ResamplingCustom$new()
mlr_resamplings$get("custom")
rsmpl("custom")
```

Fields

See [Resampling](#).

Methods

See [Resampling](#).

See Also

[Dictionary of Resamplings: mlr_resamplings](#)

as.data.table(mlr_resamplings) for a complete table of all (also dynamically created) [Resampling](#) implementations.

Examples

```
# Create a task with 10 observations
task = tsk("iris")
task$filter(1:10)

# Instantiate Resampling
rc = rsmpl("custom")
train_sets = list(1:5, 5:10)
test_sets = list(5:10, 1:5)
rc$instantiate(task, train_sets, test_sets)

rc$train_set(1)
rc$test_set(1)
```

mlr_resamplings_cv *Cross Validation Resampling*

Description

Splits data using a folds-folds (default: 10 folds) cross-validation.

Format

[R6::R6Class](#) inheriting from [Resampling](#).

Construction

```
ResamplingCV$new()
mlr_resamplings$get("cv")
rsmpl("cv")
```

Fields

See [Resampling](#).

Methods

See [Resampling](#).

Parameters

- folds :: integer(1)
Number of folds.

See Also

Dictionary of Resamplings: [mlr_resamplings](#)

`as.data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [Resampling](#) implementations.

Examples

```
# Create a task with 10 observations
task = tsk("iris")
task$filter(1:10)

# Instantiate Resampling
rcv = rsmpl("cv", folds = 3)
rcv$instantiate(task)

# Individual sets:
rcv$train_set(1)
rcv$test_set(1)
intersect(rcv$train_set(1), rcv$test_set(1))

# Internal storage:
rcv$instance # table
```

mlr_resamplings_holdout

Holdout Resampling

Description

Splits data into a training set and a test set. Parameter `ratio` determines the ratio of observation going into the training set (default: 2/3).

Format

[R6::R6Class](#) inheriting from [Resampling](#).

Construction

```
ResamplingHoldout$new()
mlr_resamplings$get("holdout")
rsmpl("holdout")
```

Fields

See [Resampling](#).

Methods

See [Resampling](#).

Parameters

- `ratio` :: `numeric(1)`
Ratio of observations to put into the training set.

See Also

Dictionary of Resamplings: [mlr_resamplings](#)

`as.data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [Resampling](#) implementations.

Examples

```
# Create a task with 10 observations
task = tsk("iris")
task$filter(1:10)

# Instantiate Resampling
rho = rsm("holdout", ratio = 0.5)
rho$instantiate(task)

# Individual sets:
rho$train_set(1)
rho$test_set(1)
intersect(rho$train_set(1), rho$test_set(1))

# Internal storage:
rho$instance # simple list
```

`mlr_resamplings_repeated_cv`

Repeated Cross Validation Resampling

Description

Splits data repeats (default: 10) times using a folds-fold (default: 10) cross-validation.

The iteration counter translates to repeats blocks of folds cross-validations, i.e., the first folds iterations belong to a single cross-validation.

Format

`R6::R6Class()` inheriting from [Resampling](#).

Construction

```
ResamplingRepeatedCV$new()
mlr_resamplings$get("repeated_cv")
rsm("repeated_cv")
```

Fields

See [Resampling](#).

Methods

See [Resampling](#). Additionally, the class provides two helper function to translate iteration numbers to folds / repeats:

- `folds(iters)`
`integer() -> integer()`
Translates iteration numbers to fold number.
- `repeats(iters)`
`integer() -> integer()`
Translates iteration numbers to repetition number.

Parameters

- `repeats :: integer(1)`
Number of repetitions.
- `folds :: integer(1)`
Number of folds.

See Also

Dictionary of [Resamplings](#): [mlr_resamplings](#)

as `.data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [Resampling](#) implementations.

Examples

```
# Create a task with 10 observations
task = tsk("iris")
task$filter(1:10)

# Instantiate Resampling
rrcv = rsmpl("repeated_cv", repeats = 2, folds = 3)
rrcv$instantiate(task)
rrcv$iters
rrcv$folds(1:6)
rrcv$repeats(1:6)

# Individual sets:
rrcv$train_set(1)
rrcv$test_set(1)
intersect(rrcv$train_set(1), rrcv$test_set(1))

# Internal storage:
rrcv$instance # table
```

mlr_resamplings_subsampling
Subsampling Resampling

Description

Splits data repeats (default: 30) times into training and test set with a ratio of ratio (default: 2/3) observations going into the training set.

Format

[R6::R6Class](#) inheriting from [Resampling](#).

Construction

```
ResamplingSubsampling$new()  
mlr_resamplings$get("subsampling")  
rsmpl("subsampling")
```

Fields

See [Resampling](#).

Methods

See [Resampling](#).

Parameters

- `repeats` :: integer(1)
Number of repetitions.
- `ratio` :: numeric(1)
Ratio of observations to put into the training set.

See Also

Dictionary of [Resamplings](#): [mlr_resamplings](#)

`as.data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [Resampling](#) implementations.

Examples

```
# Create a task with 10 observations  
task = tsk("iris")  
task$filter(1:10)  
  
# Instantiate Resampling  
rss = rsmpl("subsampling", repeats = 2, ratio = 0.5)
```



```
rss$instantiate(task)

# Individual sets:
rss$train_set(1)
rss$test_set(1)
intersect(rss$train_set(1), rss$test_set(1))

# Internal storage:
rss$instance$train # list of index vectors
```

mlr_sugar

Syntactic Sugar for Object Construction

Description

Functions to retrieve objects, set hyperparameters and assign to fields in one go. Relies on `mlr3misc::dictionary_sugar_g` to extract objects from the respective `mlr3misc::Dictionary`:

- `tsk()` for a [Task](#) from `mlr_tasks`.
- `tsks()` for a list of [Tasks](#) from `mlr_tasks`.
- `tgen()` for a [TaskGenerator](#) from `mlr_task_generators`.
- `tgens()` for a list of [TaskGenerators](#) from `mlr_task_generators`.
- `lrn()` for a [Learner](#) from `mlr_learners`.
- `lrns()` for a list of [Learners](#) from `mlr_learners`.
- `rsmp()` for a [Resampling](#) from `mlr_resamplings`.
- `rsmps()` for a list of [Resamplings](#) from `mlr_resamplings`.
- `msr()` for a [Measure](#) from `mlr_measures`.
- `msrs()` for a list of [Measures](#) from `mlr_measures`.

Usage

```
tsk(.key, ...)
```

```
tsks(.keys, ...)
```

```
tgen(.key, ...)
```

```
tgens(.keys, ...)
```

```
lrn(.key, ...)
```

```
lrns(.keys, ...)
```

```
rsmp(.key, ...)
```

```
rsmps(.keys, ...)
```

```
msr(.key, ...)
```

```
msrs(.keys, ...)
```

Arguments

<code>.key</code>	:: character(1) Key passed to the respective mlr3misc::Dictionary to retrieve the object.
<code>...</code>	:: named list() Named arguments passed to the constructor, to be set as parameters in the paradox::ParamSet , or to be set as public field. See mlr3misc::dictionary_sugar_get() for more details.
<code>.keys</code>	:: character() Keys passed to the respective mlr3misc::Dictionary to retrieve multiple objects.

Value

[R6::R6Class](#) object of the respective type, or a list of [R6::R6Class](#) objects for the plural versions.

Examples

```
# iris task with new id
tsk("iris", id = "iris2")

# classification tree with different hyperparameters
# and predict type set to predict probabilities
lrn("classif.rpart", cp = 0.1, predict_type = "prob")

# multiple learners with predict type 'prob'
lapply(c("classif.featureless", "classif.rpart"), lrn, predict_type = "prob")
```

mlr_tasks

Dictionary of Tasks

Description

A simple [mlr3misc::Dictionary](#) storing objects of class [Task](#). Each task has an associated help page, see `mlr_tasks_[id]`.

This dictionary can get populated with additional tasks by add-on packages.

For a more convenient way to retrieve and construct tasks, see [tsk\(\)](#).

Format

[R6::R6Class](#) object inheriting from [mlr3misc::Dictionary](#).

Methods

See [mlr3misc::Dictionary](#).

S3 methods

- `as.data.table(dict)`
[mlr3misc::Dictionary](#) -> `data.table::data.table()`
Returns a `data.table::data.table()` with columns "key", "task_type", "measures", "nrow", "ncol" and the number of features of type "lgl", "int", "dbl", "chr", "fct" and "ord" as columns.

See Also

Example tasks: [iris](#) (multi-class classification), [spam](#) (binary classification), [boston_housing](#) (regression)

Sugar function: [tsk\(\)](#)

Other Dictionary: [mlr_learners](#), [mlr_measures](#), [mlr_resamplings](#), [mlr_task_generators](#)

Other Task: [TaskClassif](#), [TaskRegr](#), [TaskSupervised](#), [Task](#)

Examples

```
as.data.table(mlr_tasks)
task = mlr_tasks$get("iris") # same as tsk("iris")
head(task$data())

# Add a new task, based on a subset of iris:
data = iris
data$Species = factor(iffelse(data$Species == "setosa", "1", "0"))
task = TaskClassif$new("iris.binary", data, target = "Species", positive = "1")

# add to dictionary
mlr_tasks$add("iris.binary", task)

# list available tasks
mlr_tasks$keys()

# retrieve from dictionary
mlr_tasks$get("iris.binary")

# remove task again
mlr_tasks$remove("iris.binary")
```

mlr_tasks_boston_housing

Boston Housing Regression Task

Description

A regression task for the [mlbench::BostonHousing2](#) data set.

Format

[R6::R6Class](#) inheriting from [TaskRegr](#).

Construction

```
mlr_tasks$get("boston_housing")
tsk("boston_housing")
```

See Also

Dictionary of Tasks: [mlr_tasks](#)

`as.data.table(mlr_tasks)` for a complete table of all (also dynamically created) [Tasks](#).

mlr_tasks_german_credit

German Credit Classification Task

Description

A classification task for the German credit data set. The aim is to predict creditworthiness, labeled as "good" and "bad". Positive class is set to label "good".

See example for the creation of a [MeasureClassifCosts](#) as described misclassification costs.

Format

[R6::R6Class](#) inheriting from [TaskClassif](#).

Construction

```
mlr_tasks$get("german_credit")
tsk("german_credit")
```

Source

Data set originally published on [UCI](#). This is the preprocessed version taken from package [evtree](#).

Donor: Professor Dr. Hans Hofmann
Institut für Statistik und Ökonometrie
Universität Hamburg
FB Wirtschaftswissenschaften
Von-Melle-Park 5
2000 Hamburg 13

See Also

[Dictionary of Tasks: mlr_tasks](#)

`as.data.table(mlr_tasks)` for a complete table of all (also dynamically created) [Tasks](#).

Examples

```
task = tsk("german_credit")
costs = matrix(c(0, 1, 5, 0), nrow = 2)
dimnames(costs) = list(predicted = task$class_names, truth = task$class_names)
measure = msr("classif.costs", id = "german_credit_costs", costs = costs)
print(measure)
```

mlr_tasks_iris	<i>Iris Classification Task</i>
----------------	---------------------------------

Description

A classification task for the popular [datasets::iris](#) data set.

Format

[R6::R6Class](#) inheriting from [TaskClassif](#).

Construction

```
mlr_tasks$get("iris")
tsk("iris")
```

See Also

[Dictionary of Tasks: mlr_tasks](#)

`as.data.table(mlr_tasks)` for a complete table of all (also dynamically created) [Tasks](#).

mlr_tasks_mtcars	<i>Motor Trend Regression Task</i>
------------------	------------------------------------

Description

A regression task for the [datasets::mtcars](#) data set. Target variable is mpg (Miles/(US) gallon).

Format

[R6::R6Class](#) inheriting from [TaskRegr](#).

Construction

```
mlr_tasks$get("mtcars")  
tsk("mtcars")
```

See Also

[Dictionary of Tasks: mlr_tasks](#)

`as.data.table(mlr_tasks)` for a complete table of all (also dynamically created) [Tasks](#).

mlr_tasks_pima	<i>Pima Indian Diabetes Classification Task</i>
----------------	---

Description

A classification task for the [mlbench::PimaIndiansDiabetes2](#) data set. Positive class is set to "pos".

Format

[R6::R6Class](#) inheriting from [TaskClassif](#).

Construction

```
mlr_tasks$get("pima")  
tsk("pima")
```

See Also

[Dictionary of Tasks: mlr_tasks](#)

`as.data.table(mlr_tasks)` for a complete table of all (also dynamically created) [Tasks](#).

mlr_tasks_sonar	<i>Sonar Classification Task</i>
-----------------	----------------------------------

Description

A classification task for the [mlbench::Sonar](#) data set. Positive class is set to "M" (Mine).

Format

[R6::R6Class](#) inheriting from [TaskClassif](#).

Construction

```
mlr_tasks$get("sonar")  
tsk("sonar")
```

See Also

[Dictionary of Tasks: mlr_tasks](#)

`as.data.table(mlr_tasks)` for a complete table of all (also dynamically created) [Tasks](#).

mlr_tasks_spam	<i>Spam Classification Task</i>
----------------	---------------------------------

Description

Spam data set from the UCI machine learning repository (<http://archive.ics.uci.edu/ml/datasets/spambase>). Data set collected at Hewlett-Packard Labs to classify emails as spam or non-spam. 57 variables indicate the frequency of certain words and characters in the e-mail. The positive class is set to "spam".

Format

`R6::R6Class` inheriting from `TaskClassif`.

Construction

```
mlr_tasks$get("spam")
tsk("spam")
```

Source

Creators: Mark Hopkins, Erik Reeber, George Forman, Jaap Suermondt. Hewlett-Packard Labs, 1501 Page Mill Rd., Palo Alto, CA 94304

Donor: George Forman (gforman at nospam hpl.hp.com) 650-857-7835

Preprocessing: Columns have been renamed. Preprocessed data taken from the **kernlab** package.

References

Dua D, Graff C (2017). "UCI Machine Learning Repository." <http://archive.ics.uci.edu/ml>.

See Also

[Dictionary of Tasks: mlr_tasks](#)

`as.data.table(mlr_tasks)` for a complete table of all (also dynamically created) [Tasks](#).

mlr_tasks_wine	<i>Wine Classification Task</i>
----------------	---------------------------------

Description

Wine data set from the UCI machine learning repository (<https://archive.ics.uci.edu/ml/datasets/wine>). Results of a chemical analysis of three types of wines grown in the same region in Italy but derived from three different cultivars.

Format

`R6::R6Class` inheriting from `TaskClassif`.

Construction

```
mlr_tasks$get("wine")
tsk("wine")
```

Source

Original owners: Forina, M. et al, PARVUS - An Extendible Package for Data Exploration, Classification and Correlation. Institute of Pharmaceutical and Food Analysis and Technologies, Via Brigata Salerno, 16147 Genoa, Italy.

Donor: Stefan Aeberhard, email: stefan@coral.cs.jcu.edu.au

References

Dua D, Graff C (2017). "UCI Machine Learning Repository." <http://archive.ics.uci.edu/ml>.

See Also

Dictionary of Tasks: `mlr_tasks`

`as.data.table(mlr_tasks)` for a complete table of all (also dynamically created) `Tasks`.

mlr_tasks_zoo	<i>Zoo Classification Task</i>
---------------	--------------------------------

Description

A classification task for the `mlbench::Zoo` data set.

Format

`R6::R6Class` inheriting from `TaskClassif`.

Construction

```
mlr_tasks$get("zoo")
tsk("zoo")
```

See Also

[Dictionary of Tasks: mlr_tasks](#)

`as.data.table(mlr_tasks)` for a complete table of all (also dynamically created) [Tasks](#).

mlr_task_generators *Dictionary of Task Generators*

Description

A simple [mlr3misc::Dictionary](#) storing objects of class [TaskGenerator](#). Each task generator has an associated help page, see `mlr_task_generators_[id]`.

This dictionary can get populated with additional task generators by add-on packages.

For a more convenient way to retrieve and construct task generators, see [tgen\(\)](#).

Format

[R6::R6Class](#) object inheriting from [mlr3misc::Dictionary](#).

Methods

See [mlr3misc::Dictionary](#).

S3 methods

- `as.data.table(dict)`
[mlr3misc::Dictionary](#) -> `data.table::data.table()`
Returns a `data.table::data.table()` with fields "key" and "packages" as columns.

See Also

Example generators: [xor](#)

Sugar function: [tgen\(\)](#)

Other Dictionary: [mlr_learners](#), [mlr_measures](#), [mlr_resamplings](#), [mlr_tasks](#)

Other TaskGenerator: [TaskGenerator](#)

Examples

```
mlr_task_generators$get("smiley")
tgen("2dnormals")
```

```
mlr_task_generators_2dnormals
  2D Normals Classification Task Generator
```

Description

A [TaskGenerator](#) for the 2d normals task in `mlbench::mlbench.2dnormals()`.

Format

[R6::R6Class](#) inheriting from [TaskGenerator](#).

Construction

```
TaskGenerator2DNormals$new()
mlr_task_generators$get("2dnormals")
tgen("2dnormals")
```

See Also

[Dictionary of TaskGenerators: mlr_task_generators](#)

as `data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [TaskGenerator](#) implementations.

Examples

```
tgen("2dnormals")$generate(10)$data()
```

```
mlr_task_generators_friedman1
  Friedman1 Regression Task Generator
```

Description

A [TaskGenerator](#) for the friedman1 task in `mlbench::mlbench.friedman1()`.

Format

[R6::R6Class](#) inheriting from [TaskGenerator](#).

Construction

```
TaskGeneratorFriedman1$new()
mlr_task_generators$get("friedman1")
tgen("friedman1")
```

See Also

Dictionary of TaskGenerators: [mlr_task_generators](#)

`as.data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [TaskGenerator](#) implementations.

Examples

```
tgen("friedman1")$generate(10)$data()
```

```
mlr_task_generators_smiley  
Smiley Classification Task Generator
```

Description

A [TaskGenerator](#) for the smiley task in `mlbench::mlbench.smiley()`.

Format

[R6::R6Class](#) inheriting from [TaskGenerator](#).

Construction

```
TaskGeneratorSmiley$new()  
mlr_task_generators$get("smiley")  
tgen("smiley")
```

See Also

Dictionary of TaskGenerators: [mlr_task_generators](#)

`as.data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [TaskGenerator](#) implementations.

Examples

```
tgen("smiley")$generate(10)$data()
```

```
mlr_task_generators_xor
```

XOR Classification Task Generator

Description

A [TaskGenerator](#) for the xor task in `mlbench::mlbench.xor()`.

Format

[R6::R6Class](#) inheriting from [TaskGenerator](#).

Construction

```
TaskGeneratorXor$new()
mlr_task_generators$get("xor")
tgen("xor")
```

See Also

Dictionary of [TaskGenerators](#): [mlr_task_generators](#)

as `data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [TaskGenerator](#) implementations.

Examples

```
tgen("xor")$generate(10)$data()
```

```
predict.Learner
```

Predict Method for Learners

Description

Extends the generic `stats::predict()` with a method for [Learner](#). Note that this function is intended as glue code to be used in third party packages. We recommend to work with the [Learner](#) directly, i.e. calling `learner$predict()` or `learner$predict_newdata()` directly.

Performs the following steps:

- Sets additional hyperparameters passed to this function.
- Creates a [Prediction](#) object by calling `learner$predict_newdata()`.
- Returns (subset of) [Prediction](#).

Usage

```
## S3 method for class 'Learner'
predict(object, newdata, predict_type = NULL, ...)
```

Arguments

object	:: Learner Any Learner .
newdata	:: <code>data.frame()</code> New data to predict on.
predict_type	:: <code>character(1)</code> The predict type to return. Set to <code><Prediction></code> to retrieve the complete Prediction object. If set to <code>NULL</code> (default), the first predict type for the respective class of the Learner as stored in <code>mlr_reflections</code> is used.
...	:: any Hyperparameters to pass down to the Learner .

Examples

```

task = tsk("spam")

learner = lrn("classif.rpart", predict_type = "prob")
learner$train(task)
predict(learner, task$data(1:3), predict_type = "response")
predict(learner, task$data(1:3), predict_type = "prob")
predict(learner, task$data(1:3), predict_type = "<Prediction>")

```

Prediction	<i>Abstract Prediction Object</i>
------------	-----------------------------------

Description

This is the abstract base class for task objects like [PredictionClassif](#) or [PredictionRegr](#).

Prediction objects store the following information:

1. The row ids of the test set
2. The corresponding true (observed) response.
3. The corresponding predicted response.
4. Additional predictions based on the class and `predict_type`. E.g., the class probabilities for classification or the estimated standard error for regression.

Format

[R6::R6Class](#) object.

Construction

This object is constructed via a derived classes, e.g. [PredictionClassif](#) or [PredictionRegr](#).

Fields

- `row_ids` :: (integer() | character())
Vector of row ids for which predictions are stored.
- `truth` :: any
True (observed) outcome.
- `task_type` :: character(1)
Stores the type of the [Task](#).
- `task_properties` :: character()
Stores properties of the [Task](#).
- `predict_types` :: character()
Vector of predict types this object stores.
- `missing` :: logical()
Returns `row_ids` for which the predictions are missing or incomplete.

Methods

- `score(measures = NULL, task = NULL, learner = NULL)`
(list of [Measure](#), [Task](#), [Learner](#)) -> [Prediction](#)
Calculates the performance for all provided measures [Task](#) and [Learner](#) may be NULL for most measures, but some measures need to extract information from these objects.
- `help()`
() -> NULL
Opens the help page for this object.

S3 Methods

- `as.data.table(rr)`
[Prediction](#) -> `data.table::data.table()`
Converts the data to a `data.table::data.table()`.
- `c(..., keep_duplicates = TRUE)`
([Prediction](#), [Prediction](#), ...) -> [Prediction](#)
Combines multiple Predictions to a single Prediction. If `keep_duplicates` is FALSE and there are duplicated row ids, the data of the former passed objects get overwritten by the data of the later passed objects.

See Also

Other Prediction: [PredictionClassif](#), [PredictionRegr](#)

PredictionClassif *Prediction Object for Classification*

Description

This object wraps the predictions returned by a learner of class [LearnerClassif](#), i.e. the predicted response and class probabilities.

If the response is not provided during construction, but class probabilities are, the response is calculated from the probabilities: the class label with the highest probability is chosen. In case of ties, a label is selected randomly.

Format

[R6::R6Class](#) object inheriting from [Prediction](#).

Construction

```
p = PredictionClassif$new(task = NULL, row_ids = task$row_ids, truth = task$truth(), response = NULL, pr
```

- `task` :: [TaskClassif](#)
Task, used to extract defaults for `row_ids` and `truth`.
- `row_ids` :: (`integer()` | `character()`)
Row ids of the observations in the test set.
- `truth` :: `factor()`
True (observed) labels. See the note on manual construction.
- `response` :: (`character()` | `factor()`)
Vector of predicted class labels. One element for each observation in the test set. Character vectors are automatically converted to factors. See the note on manual construction.
- `prob` :: `matrix()`
Numeric matrix of posterior class probabilities with one column for each class and one row for each observation in the test set. Columns must be named with class labels, row names are automatically removed. If `prob` is provided, but `response` is not, the class labels are calculated from the probabilities using `max.col()` with `ties.method` set to "random".

Fields

All fields from [Prediction](#), and additionally:

- `response` :: `factor()`
Access to the stored predicted class labels.
- `prob` :: `matrix()`
Access to the stored probabilities.
- `confusion` :: `matrix()`
Confusion matrix resulting from the comparison of truth and response. Truth is in columns, predicted response is in rows.

The field `task_type` is set to "classif".

Methods

- `set_threshold(th)`
`numeric() -> self`
 Sets the prediction response based on the provided threshold. See the section on thresholding for more information.

Thresholding

If probabilities are stored, it is possible to change the threshold which determines the predicted class label. Usually, the label of the class with the highest predicted probability is selected. For binary classification problems, such an threshold defaults to 0.5. For cost-sensitive or imbalanced classification problems, manually adjusting the threshold can increase the predictive performance.

- For binary problems only a single threshold value can be set. If the probability exceeds the threshold, the positive class is predicted. If the probability equals the threshold, the label is selected randomly.
- For binary and multi-class problems, a named numeric vector of thresholds can be set. The length and names must correspond to the number of classes and class names, respectively. To determine the class label, the probabilities are divided by the threshold. This results in a ratio > 1 if the probability exceeds the threshold, and a ratio < 1 otherwise. Note that it is possible that either none or multiple ratios are greater than 1 at the same time. Anyway, the class label with maximum ratio is selected. In case of ties in the ratio, one of the tied class labels is selected randomly.

Note

If this object is constructed manually, make sure that the factor levels for `truth` have the same levels as the task, in the same order. In case of binary classification tasks, the positive class label must be the first level.

See Also

Other Prediction: [PredictionRegr](#), [Prediction](#)

Examples

```
task = tsk("iris")
learner = lrn("classif.rpart", predict_type = "prob")
learner$train(task)
p = learner$predict(task)
p$predict_types
head(as.data.table(p))

# confusion matrix
p$confusion

# change threshold
th = c(0.05, 0.9, 0.05)
names(th) = task$class_names
```



```
# new predictions
p$set_threshold(th)$response
p$score(measures = msr("classif.ce"))
```

PredictionRegr	<i>Prediction Object for Regression</i>
----------------	---

Description

This object wraps the predictions returned by a learner of class [LearnerRegr](#), i.e. the predicted response and standard error.

Format

[R6::R6Class](#) object inheriting from [Prediction](#).

Construction

```
p = PredictionRegr$new(task = NULL, row_ids = task$row_ids, truth = task$truth(), response = NULL, se = N
```

- `task :: TaskRegr`
Task, used to extract defaults for `row_ids` and `truth`.
- `row_ids :: (integer() | character())`
Row ids of the observations in the test set.
- `truth :: numeric()`
True (observed) response.
- `response :: numeric()`
Vector of numeric response values. One element for each observation in the test set.
- `se :: numeric()`
Numeric vector of predicted standard errors. One element for each observation in the test set.

Fields

All fields from [Prediction](#), and additionally:

- `response :: numeric()`
Access to the stored predicted response.
- `se :: numeric()`
Access to the stored standard error.

The field `task_type` is set to "regr".

See Also

Other Prediction: [PredictionClassif](#), [Prediction](#)

Examples

```
task = tsk("boston_housing")
learner = lrn("regr.featureless", predict_type = "se")
p = learner$train(task)$predict(task)
p$predict_types
head(as.data.table(p))
```

resample

*Resample a Learner on a Task***Description**

Runs a resampling (possibly in parallel).

Usage

```
resample(task, learner, resampling, store_models = FALSE)
```

Arguments

task	:: Task .
learner	:: Learner .
resampling	:: Resampling .
store_models	:: logical(1) Keep the fitted model after the test set has been predicted? Set to TRUE if you want to further analyse the models or want to extract information like variable importance.

Value

[ResampleResult](#).

Parallelization

This function can be parallelized with the [future](#) package. One job is one resampling iteration, and all jobs are sent to an apply function from [future.apply](#) in a single batch. To select a parallel backend, use [future::plan\(\)](#).

Logging

The [mlr3](#) uses the [lgr](#) package for logging. [lgr](#) supports multiple log levels which can be queried with `getOption("lgr.log_levels")`.

To suppress output and reduce verbosity, you can lower the log from the default level "info" to "warn":

```
lgr::get_logger("mlr3")$set_threshold("warn")
```

To get additional log output for debugging, increase the log level to "debug" or "trace":

```
lgr::get_logger("mlr3")$set_threshold("debug")
```

To log to a file or a data base, see the documentation of [lgr::lgr-package](#).

Note

The fitted models are discarded after the predictions have been scored in order to reduce memory consumption. If you need access to the models for later analysis, set `store_models` to TRUE.

Examples

```
task = tsk("iris")
learner = lrn("classif.rpart")
resampling = rsmpl("cv")

# explicitly instantiate the resampling for this task for reproducibility
set.seed(123)
resampling$instantiate(task)

rr = resample(task, learner, resampling)
print(rr)

# retrieve performance
rr$score(msr("classif.ce"))
rr$aggregate(msr("classif.ce"))

# merged prediction objects of all resampling iterations
pred = rr$prediction()
pred$confusion

# Repeat resampling with featureless learner
rr_featureless = resample(task, lrn("classif.featureless"), resampling)

# Convert results to BenchmarkResult, then combine them
bmr1 = as_benchmark_result(rr)
bmr2 = as_benchmark_result(rr_featureless)
print(bmr1$combine(bmr2))
```

ResampleResult

Container for Results of resample()

Description

This is the result container object returned by [resample\(\)](#).

Note that all stored objects are accessed by reference. Do not modify any object without cloning it first.

Format

[R6::R6Class](#) object.

Construction

```
rr = ResampleResult$new(data, uhash = NULL)
```

- `data` :: [data.table::data.table\(\)](#)
Table with data for one resampling iteration per row: [Task](#), [Learner](#), [Resampling](#), iteration (`integer(1)`), and [Prediction](#).
- `uhash` :: `character(1)`
Unique hash for this `ResampleResult`. If `NULL`, a new unique hash is generated. This unique hash is primarily needed to group information in [BenchmarkResults](#).

Fields

- `data` :: [data.table::data.table\(\)](#)
Internal data storage. We discourage users to directly work with this field.
- `task` :: [Task](#)

The task [resample\(\)](#) operated on.
- `learners` :: list of [Learner](#)
List of trained learners, sorted by resampling iteration.
- `resampling` :: [Resampling](#)
Instantiated [Resampling](#) object which stores the splits into training and test.
- `warnings` :: [data.table::data.table\(\)](#)
Returns a table with all warning messages. Column names are "iteration" and "msg". Note that there can be multiple rows per resampling iteration if multiple warnings have been recorded.
- `errors` :: [data.table::data.table\(\)](#)
Returns a table with all error messages. Column names are "iteration" and "msg". Note that there can be multiple rows per resampling iteration if multiple errors have been recorded.
- `uhash` :: `character(1)`
Unique hash for this object.

Methods

- `predictions(predict_sets = "test")`
`character()` -> list of [Prediction](#)
List of prediction objects, sorted by resampling iteration. If multiple sets are given, these are combined to a single one for each iteration.
- `prediction(predict_sets = "test")`
`character()` -> [Prediction](#)
Combined [Prediction](#) of all individual resampling iterations, and all provided predict sets. Note that performance measures do not operate on this object, but instead on each prediction object separately and then combine the performance scores with the aggregate function of the respective [Measure](#).

- `score(measures = NULL, ids = TRUE)`
(list of [Measure](#), `logical(1)`) -> `data.table::data.table()`
Returns a table with one row for each resampling iteration, including all involved objects: [Task](#), [Learner](#), [Resampling](#), iteration number (`integer(1)`), and [Prediction](#). A column with the individual (per resampling iteration) performance is added for each [Measure](#), named with the id of the respective measure. If `ids` is `TRUE`, extra columns with the ids of objects ("`task_id`", "`learner_id`", "`resampling_id`") are binded to the table to allow a more convenient subsetting. If `measures` is `NULL`, `measures` defaults to the return value of `default_measures()`.
- `aggregate(measures = NULL)`
list of [Measure](#) -> `named numeric()`
Calculates and aggregates performance values for all provided measures, according to the respective aggregation function in [Measure](#). If `measures` is `NULL`, `measures` defaults to the return value of `default_measures()`.
- `help()`
() -> `NULL`
Opens the help page for this object.

S3 Methods

- `as.data.table(rr)`
[ResampleResult](#) -> `data.table::data.table()`
Returns a copy of the internal data.

Examples

```
task = tsk("iris")
learner = lrn("classif.rpart")
resampling = rsmpl("cv", folds = 3)
rr = resample(task, learner, resampling)
print(rr)

rr$aggregate(msr("classif.acc"))
rr$prediction()
rr$prediction()$confusion
rr$warnings
rr$errors
```

Resampling

Resampling Class

Description

This is the abstract base class for resampling objects like [ResamplingCV](#) and [ResamplingBootstrap](#).

The objects of this class define how a task is partitioned for resampling (e.g., in `resample()` or `benchmark()`), using a set of hyperparameters such as the number of folds in cross-validation.

Resampling objects can be instantiated on a [Task](#), which applies the strategy on the task and manifests in a fixed partition of `row_ids` of the [Task](#).

Predefined resamplings are stored in the `mlr3misc::Dictionary mlr_resamplings`, e.g. `cv` or `bootstrap`.

Format

[R6::R6Class](#) object.

Construction

Note: This object is typically constructed via a derived classes, e.g. [ResamplingCV](#) or [ResamplingHoldout](#).

```
r = Resampling$new(id, param_set, duplicated_ids = FALSE, man = NA_character_)
```

- `id` :: `character(1)`
Identifier for the resampling strategy.
- `param_set` :: [paradox::ParamSet](#)
Set of hyperparameters.
- `duplicated_ids` :: `logical(1)`
Set to TRUE if this resampling strategy may have duplicated row ids in a single training set or test set.
- `man` :: `character(1)`
String in the format `[pkg]::[topic]` pointing to a manual page for this object.

Fields

All variables passed to the constructor, and additionally:

- `iters` :: `integer(1)`
Return the number of resampling iterations, depending on the values stored in the `param_set`.
- `instance` :: `any`
During `instantiate()`, the instance is stored in this slot. The instance can be in any arbitrary format.
- `is_instantiated` :: `logical(1)`
Is TRUE, if the resampling has been instantiated.
- `task_hash` :: `character(1)`
The hash of the [Task](#) which was passed to `r$instantiate()`.
- `task_nrow` :: `integer(1)`
The number of observations of the [Task](#) which was passed to `r$instantiate()`.
- `hash` :: `character(1)`
Hash (unique identifier) for this object.
E.g., this is TRUE for Bootstrap, and FALSE for cross validation. Only used internally.

Methods

- `instantiate(task)`
[Task](#) -> `self`
Materializes fixed training and test splits for a given task and stores them in `r$instance`.
- `train_set(i)`
`integer(1)` -> (`integer()` | `character()`)
Returns the row ids of the `i`-th training set.

- `test_set(i)`
integer(1) -> (integer() | character())
Returns the row ids of the i-th test set.
- `help()`
() -> NULL
Opens the corresponding help page referenced by `$man`.

Stratification

All derived classes support stratified sampling. The stratification variables are assumed to be discrete and must be stored in the [Task](#) with column role "stratum". In case of multiple stratification variables, each combination of the values of the stratification variables forms a strata.

First, the observations are divided into subpopulations based one or multiple stratification variables (assumed to be discrete), c.f. `task$strata`.

Second, the sampling is performed in each of the `k` subpopulations separately. Each subgroup is divided into `iter` training sets and `iter` test sets by the derived [Resampling](#). These sets are merged based on their iteration number: all training sets from all subpopulations with iteration 1 are combined, then all training sets with iteration 2, and so on. Same is done for all test sets. The merged sets can be accessed via `$train_set(i)` and `$test_set(i)`, respectively.

Grouping / Blocking

All derived classes support grouping of observations. The grouping variable is assumed to be discrete and must be stored in the [Task](#) with column role "group".

Observations in the same group are treated like a "block" of observations which must be kept together. These observations either all go together into the training set or together into the test set.

The sampling is performed by the derived [Resampling](#) on the grouping variable. Next, the grouping information is replaced with the respective row ids to generate training and test sets. The sets can be accessed via `$train_set(i)` and `$test_set(i)`, respectively.

See Also

Dictionary of Resamplings: [mlr_resamplings](#)

`as.data.table(mlr_resamplings)` for a complete table of all (also dynamically created) [Resampling](#) implementations.

Other Resampling: [mlr_resamplings](#)

Examples

```
r = rsmp("subsampling")

# Default parametrization
r$param_set$values

# Do only 3 repeats on 10% of the data
r$param_set$values = list(ratio = 0.1, repeats = 3)
r$param_set$values
```

```

# Instantiate on iris task
task = tsk("iris")
r$instantiate(task)

# Extract train/test sets
train_set = r$train_set(1)
print(train_set)
intersect(train_set, r$test_set(1))

# Another example: 10-fold CV
r = rsmp("cv")$instantiate(task)
r$train_set(1)

# Stratification
task = tsk("pima")
prop.table(table(task$truth())) # moderately unbalanced
task$col_roles$stratum = task$target_names

r = rsmp("subsampling")
r$instantiate(task)
prop.table(table(task$truth(r$train_set(1)))) # roughly same proportion

```

Task	<i>Task Class</i>
------	-------------------

Description

This is the abstract base class for task objects like [TaskClassif](#) and [TaskRegr](#).

Tasks serve two purposes:

1. Tasks wrap a [DataBackend](#), an object to transparently interface different data storage types.
2. Tasks store meta-information, such as the role of the individual columns in the [DataBackend](#). For example, for a classification task a single column must be marked as target column, and others as features.

Predefined (toy) tasks are stored in the [mlr3misc::Dictionary mlr_tasks](#), e.g. [iris](#) or [boston_housing](#).

Format

[R6::R6Class](#) object.

Construction

Note: This object is typically constructed via a derived classes, e.g. [TaskClassif](#) or [TaskRegr](#).

```
t = Task$new(id, task_type, backend)
```

- `id` :: `character(1)`
Identifier for the task.

- `task_type` :: `character(1)`
Set in the classes which inherit from this class. Must be an element of `mlr_reflections$task_types$type`.
- `backend` :: `DataBackend`
Either a `DataBackend`, or any object which is convertible to a `DataBackend` with `as_data_backend()`.
E.g., a `data.frame()` will be converted to a `DataBackendDataTable`.

Fields

- `backend` :: `DataBackend`.
- `col_info` :: `data.table::data.table()`
Table with with 3 columns:
 - "id" stores the name of the column.
 - "type" holds the storage type of the variable, e.g. `integer`, `numeric` or `character`.
 - "levels" stores a vector of distinct values (levels) for factor variables.
- `col_roles` :: `named list()`
Each column (feature) can have an arbitrary number of the following roles:
 - "feature": Regular feature used in the model fitting process.
 - "target": Target variable.
 - "name": Row names / observation labels. To be used in plots.
 - "order": Data returned by `$data()` is ordered by this column (or these columns).
 - "group": During resampling, observations with the same value of the variable with role "group" are marked as "belonging together". They will be exclusively assigned to be either in the training set or in the test set for each resampling iteration. Only up to one column may have this role.
 - "stratum": Stratification variables. Multiple discrete columns may have this role.
 - "weight": Observation weights. Only up to one column (assumed to be discrete) may have this role.

`col_roles` keeps track of the roles with a named list, the elements are named by column role and each element is a character vector of column names. To alter the roles, just modify the list, e.g. with R's set functions (`intersect()`, `setdiff()`, `union()`, ...).

- `row_roles` :: `named list()`
Each row (observation) can have an arbitrary number of roles in the learning task:
 - "use": Use in train / predict / resampling.
 - "validation": Hold the observations back unless explicitly requested. Validation sets are not yet completely integrated into the package.

`row_roles` keeps track of the roles with a named list, elements are named by row role and each element is a `integer()` or `character()` vector of row ids. To alter the roles, just modify the list, e.g. with R's set functions (`intersect()`, `setdiff()`, `union()`, ...).

- `feature_names` :: `character()`
Return all column names with `role == "feature"`.
- `feature_types` :: `data.table::data.table()`
Returns a table with columns `id` and `type` where `id` are the column names of "active" features of the task and `type` is the storage type.

- `hash :: character(1)`
Hash (unique identifier) for this object.
- `id :: character(1)`
Identifier of the Task.
- `ncol :: integer(1)`
Returns the total number of cols with role "target" or "feature".
- `nrow :: integer(1)`
Return the total number of rows with role "use".
- `row_ids :: (integer() | character())`
Returns the row ids of the [DataBackend](#) for observations with with role "use".
- `target_names :: character()`
Returns all column names with role "target".
- `task_type :: character(1)`
Stores the type of the [Task](#).
- `properties :: character()`
Set of task properties. Possible properties are are stored in `mlr_reflections$task_properties`. The following properties are currently standardized and understood by tasks in **mlr3**:
 - "strata": The task is resampled using one or more stratification variables (role "stratum").
 - "groups": The task comes with grouping/blocking information (role "group").
 - "weights": The task comes with observation weights (role "weight"). Note that above listed properties are calculated from the `$col_roles` and must not be set explicitly.
- `strata :: data.table::data.table()`
If the task has columns designated with role "stratum", returns a table with one subpopulation per row and two columns: `N` (`integer()`) with the number of observations in the subpopulation and `row_id` (list of `integer()` | list of `character()`) as list column with the row ids in the respective subpopulation. Returns NULL if there are is no stratification variable. See [Resampling](#) for more information on stratification.
- `groups :: data.table::data.table()`
If the task has a column with designated role "group", table with two columns: `row_id` (`integer()` | `character()`) and the grouping variable `group` (`vector()`). Returns NULL if there are is no grouping column. See [Resampling](#) for more information on grouping.
- `weights :: data.table::data.table()`
If the task has a column with designated role "weight", table with two columns: `row_id` (`integer()` | `character()`) and the observation weights `weight` (`numeric()`). Returns NULL if there are is no weight column.
- `man :: character(1)`
String in the format `[pkg]::[topic]` pointing to a manual page for this object.

Methods

- `data(rows = NULL, cols = NULL, data_format = NULL)`
(`integer()` | `character()`, `character()`, `character(1)`) -> any
Returns a slice of the data from the [DataBackend](#) in the data format specified by `data_format` (depending on the [DataBackend](#), but usually a `data.table::data.table()`).

Rows are additionally subsetted to only contain observations with role "use", and columns are filtered to only contain features with roles "target" and "feature". If invalid rows or cols are specified, an exception is raised.

- `formula(rhs = ".")`
`character() -> stats::formula()`
 Constructs a `stats::formula()`, e.g. `[target] ~ [feature_1] + [feature_2] + ... + [feature_k]`, using the features provided in argument `rhs` (defaults to all columns with role "feature", symbolized by ".").
- `levels(cols = NULL)`
`character() -> named list()`
 Returns the distinct values for columns referenced in `cols` with storage type "factor" or "ordered". Argument `cols` defaults to all such columns with role "target" or "feature".
 Note that this function ignores the row roles, it returns all levels available in the [DataBackend](#). To update the stored level information, e.g. after filtering a task, call `$droplevels()`.
- `droplevels(cols = NULL)`
`character() -> self`
 Updates the cache of stored factor levels, removing all levels not present in the current set of active rows. `cols` defaults to all columns with storage type "factor" or "ordered".
- `missings(cols = NULL)`
`character() -> named integer()`
 Returns the number of missing observations for columns referenced in `cols`. Considers only active rows with row role "use". Argument `cols` defaults to all columns with role "target" or "feature".
- `head(n = 6)`
`integer() -> data.table::data.table()`
 Get the first `n` observations with role "use".
- `set_row_role(rows, new_roles, exclusive = TRUE)`
`(character(), character(), logical(1)) -> self`
 Adds the roles `new_roles` to rows referred to by `rows`. If `exclusive` is `TRUE`, the referenced rows will be removed from all other roles.
 This function is deprecated and will be removed in the next version in favor of directly modifying `$row_roles`.
- `set_col_role(cols, new_roles, exclusive = TRUE)`
`(character(), character(), logical(1)) -> self`
 Adds the roles `new_roles` to columns referred to by `cols`. If `exclusive` is `TRUE`, the referenced columns will be removed from all other roles.
 This function is deprecated and will be removed in the next version in favor of directly modifying `$col_roles`.
- `filter(rows)`
`(integer() | character()) -> self`
 Subsets the task, reducing it to only keep the rows specified in `rows`.
 This operation mutates the task in-place. See the section on task mutators for more information.

- `select(cols)`
`character() -> self`
 Subsets the task, reducing it to only keep the features specified in `cols`. Note that you cannot deselect the target column, for obvious reasons.
 This operation mutates the task in-place. See the section on task mutators for more information.
- `cbind(data)`
`data.frame() -> self`
 Adds additional columns to the [DataBackend](#). The row ids must be provided as column in `data` (with column name matching the primary key name of the [DataBackend](#)). If this column is missing, it is assumed that the rows are exactly in the order of `t$row_ids`. In case of name clashes of column names in `data` and [DataBackend](#), columns in `data` have higher precedence and virtually overwrite the columns in the [DataBackend](#).
 This operation mutates the task in-place. See the section on task mutators for more information.
- `rbind(data)`
`data.frame() -> self`
 Adds additional rows to the [DataBackend](#). The new row ids must be provided as column in `data`. If this column is missing, new row ids are constructed automatically. In case of name clashes of row ids, rows in `data` have higher precedence and virtually overwrite the rows in the [DataBackend](#).
 This operation mutates the task in-place. See the section on task mutators for more information.
- `rename(from, to)`
`(character(), character()) -> self`
 Renames columns by mapping column names in `old` to new column names in `new`.
 This operation mutates the task in-place. See the section on task mutators for more information.
- `help()`
`() -> NULL`
 Opens the corresponding help page referenced by `$man`.

S3 methods

- `as.data.table(t)`
[Task](#) -> `data.table::data.table()`
 Returns the complete data as `data.table::data.table()`.

Task mutators

The following methods change the task in-place:

- Any modification to `$col_roles` and `row_roles`. This provides a different "view" on the data without altering the data itself.
- `filter()` and `select()` subset the set of active rows or features in `row_roles` or `col_roles`, respectively. This provides a different "view" on the data without altering the data itself.

- `rbind()` and `cbind()` change the task in-place by binding rows or columns to the data, but without modifying the original [DataBackend](#). Instead, the methods first create a new [DataBackendDataTable](#) from the provided new data, and then merge both backends into an abstract [DataBackend](#) which combines the results on-demand.
- `rename()` wraps the [DataBackend](#) of the Task in an additional [DataBackend](#) which deals with the renaming. Also updates `col_roles` and `col_info`.

See Also

Other Task: [TaskClassif](#), [TaskRegr](#), [TaskSupervised](#), [mlr_tasks](#)

Examples

```
# we use the inherited class TaskClassif here,
# Class Task is not intended for direct use
task = TaskClassif$new("iris", iris, target = "Species")

task$nrow
task$ncol
task$feature_names
task$formula()

# de-select "Petal.Width"
task$select(setdiff(task$feature_names, "Petal.Width"))

task$feature_names

# Add new column "foo"
task$cbind(data.frame(foo = 1:150))
task$head()
```

TaskClassif	<i>Classification Task</i>
-------------	----------------------------

Description

This task specializes [Task](#) and [TaskSupervised](#) for classification problems. The target column is assumed to be a factor. The `task_type` is set to `"classif"`.

Additional task properties include:

- `"twoclass"`: The task is a binary classification problem.
- `"multiclass"`: The task is a multiclass classification problem.

Predefined tasks are stored in the [mlr3misc::Dictionary mlr_tasks](#).

Format

[R6::R6Class](#) object inheriting from [Task/TaskSupervised](#).

Construction

```
t = TaskClassif$new(id, backend, target, positive = NULL)
```

- `id` :: character(1)
Identifier for the task.
- `backend` :: [DataBackend](#)
Either a [DataBackend](#), or any object which is convertible to a [DataBackend](#) with `as_data_backend()`.
E.g., a `data.frame()` will be converted to a [DataBackendDataTable](#).
- `target` :: character(1)
Name of the target column.
- `positive` :: character(1)
Only for binary classification: Name of the positive class. The levels of the target columns are reordered accordingly, so that the first element of `$class_names` is the positive class, and the second element is the negative class.

Fields

All methods from [TaskSupervised](#), and additionally:

- `class_names` :: character()
Returns all class labels of the target column.
- `positive` :: character(1)
Stores the positive class for binary classification tasks, and NA for multiclass tasks. To switch the positive class, assign a level to this field.
- `negative` :: character(1)
Stores the negative class for binary classification tasks, and NA for multiclass tasks.

Methods

See [TaskSupervised](#).

See Also

Example classification tasks: [iris](#)

Other Task: [TaskRegr](#), [TaskSupervised](#), [Task](#), [mlr_tasks](#)

Examples

```
data("Sonar", package = "mlbench")
task = TaskClassif$new("sonar", backend = Sonar, target = "Class", positive = "M")

task$task_type
task$formula()
task$truth()
task$class_names
task$positive

# possible properties:
mlr_reflections$task_properties$classif
```

TaskGenerator	<i>TaskGenerator Class</i>
---------------	----------------------------

Description

Creates a [Task](#) of arbitrary size. Predefined task generators are stored in the [mlr3misc::Dictionary mlr_task_generators](#), e.g. `xor`.

Format

[R6::R6Class](#) object.

Construction

```
g = TaskGenerator$new(id, task_type, packages = character(), param_set = ParamSet$new(), man = NA_character_)
```

- `id` :: `character(1)`
Identifier for the learner.
- `task_type` :: `character(1)`
Type of the task the learner can operator on. E.g., "classif" or "regr".
- `packages` :: `character()`
Set of required packages. Note that these packages will be loaded via [requireNamespace\(\)](#), and are not attached.
- `param_set` :: [paradox::ParamSet](#)
Set of hyperparameters.
- `man` :: `character(1)`
String in the format `[pkg]::[topic]` pointing to a manual page for this object.

Fields

All variables passed to the constructor, and additionally:

- `task_type` :: `character(1)`
Stores the type of class this learner can operate on, e.g. "classif" or "regr". A complete list of task types is stored in [mlr_reflections\\$task_types\\$type](#).

Methods

- `generate(n)`
`integer(1) -> Task`
Creates a task of type `task_type` with `n` observations, possibly using additional settings stored in `param_set`.
- `help()`
`() -> NULL`
Opens the corresponding help page referenced by `$man`.

See Also

Other TaskGenerator: [mlr_task_generators](#)

TaskRegr

Regression Task

Description

This task specializes [Task](#) and [TaskSupervised](#) for regression problems. The target column is assumed to be numeric. The `task_type` is set to "classif".

Predefined tasks are stored in the [mlr3misc::Dictionary mlr_tasks](#).

Format

[R6::R6Class](#) object inheriting from [Task/TaskSupervised](#).

Construction

```
t = TaskRegr$new(id, backend, target)
```

- `id` :: `character(1)`
Identifier for the task.
- `backend` :: ([DataBackend](#) | `data.frame()` | ...)
Either a [DataBackend](#), or any object which is convertible to a [DataBackend](#) with `as_data_backend()`.
E.g., a `data.frame()` will be converted to a [DataBackendDataTable](#).
- `target` :: `character(1)`
Name of the target column.

Fields

See [TaskSupervised](#).

Methods

See [TaskSupervised](#).

See Also

Example regression tasks: [boston_housing](#)

Other Task: [TaskClassif](#), [TaskSupervised](#), [Task](#), [mlr_tasks](#)

Examples

```
task = TaskRegr$new("iris", backend = iris, target = "Sepal.Length")
task$task_type
task$formula()
task$truth()

# possible properties:
mlr_reflections$task_properties$regr
```

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