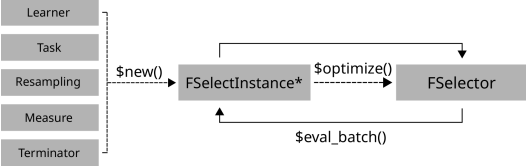


Feature Selection with mlr3fselect::CHEAT SHEET

Class Overview

The package provides a set of R6 classes which allow to (a) define general feature selection instances and (b) run algorithms which optimize on these. (a) is called a `FSelectInstanceSingleCrit` or `FSelectInstanceMultiCrit`, which define a blackbox optimization function that maps feature subsets to resampled performance values for arbitrary performance measures.



Terminators - When to stop

Construction: `trm(key, ...)`

- `evals (n_evals)`
After a given amount of iterations.
- `clock_time (secs, stop_time)`
After a given absolute time.
- `model_time (secs)`
After a given training time.
- `perf_reached (level)`
After a specific performance was reached.
- `stagnation (iters, threshold)`
After the performance stagnated for given iterations.
- `stagnation_batch (n, threshold)`
After the performance stagnated for given batches.

```
as.data.table(mlr_terminators)
```

Lists all available terminators.

FSelectInstance* - Search Scenario

Evaluator and container for resampled performances of feature subsets. The main (internal) function `eval_batch(xdt)` calls `benchmark()` to evaluate a table of feature subsets. Also stores archive of all evaluated feature subsets and the final result.

```
instance = FSelectInstanceSingleCrit$new(
  task, learner, resampling, measure,
  terminator)
```

Set `store_benchmark_result = TRUE` to store resamplings of evaluations and `store_models = TRUE` to store associated models.

Example

```
instance = FSelectInstanceSingleCrit$new(
  task("iris"), lrn("classif.rpart"), rsmp("cv"), msr("classif.ce"),
  trm("evals", n_evals = 10))
fselector = fs("random_search")
fselector$optimize(instance)
```

Use `FSelectInstanceMultiCrit` for multi-criteria tuning.

FSelector - Search Strategy

Feature Selection strategy. Generates feature subsets and passes these to `FSelectInstance*` for evaluation until termination.

Creation: `fs(.key, ...)`

- `random_search (batch_size)`
Random search.
- `exhaustive_search (max_features)`
Exhaustive Search.
- `sequential (strategy)`
Sequential Selection.
- `rfe (feature_fraction, recursive)`
Recursive Feature Elimination.
- `design_points (batch_size, design)`
User supplied feature subsets.

```
as.data.table(mlr_fselectors)
```

Lists all available feature selection algorithms.

Executing the Feature Selection

```
fselector$optimize(instance)
```

Starts the feature selection. `FSelector` generates feature subsets and passes these to the `$eval_batch()` method of the `FSelectInstance*` until the budget of the `Terminator` is exhausted.

```
as.data.table(instance$archive)
```

Returns all evaluated feature subsets and their resampling results.

Example

```
as.data.table(instance$archive)
## >   Petal.Length Petal.Width Sepal.Length Sepal.Width classif.ce
## > 1:      FALSE      FALSE      TRUE      FALSE 0.26000000
## > 2:      TRUE      FALSE      TRUE      FALSE 0.08000000
```

`uhash` refers to `instance$archive$benchmark_result`.

```
instance$result
```

Returns `data.table` with optimal feature subset and estimated performance.

```
task$select(instance$result_feature_set)
```

Set optimized feature subset in `Task`.

AutoFSelector - Select before Train

Wraps learner and performs integrated feature selection.

```
at = AutoFSelector$new(
  learner, resampling, measure, terminator,
  fselector)
```

Inherits from class `Learner`. Training starts feature selection on the training set. After completion the learner is trained with the "optimal" feature subset on the given task.

```
at$train(task)
at$predict(task, row_ids)
```

Nested Resampling

Resampling the `AutoFSelector` results in nested resampling with an inner and outer loop.

Example

```
resampling_inner = rsmp("holdout")
evals20 = trm("evals", n_evals = 20)

at = AutoFSelector$new(learner, resampling_inner, measure, evals20, fselector)
at$store_fselect_instance = TRUE

resampling_outer = rsmp("cv", folds = 2)
rr = resample(task, at, resampling_outer, store_models = TRUE)

as.data.table(rr)
## >   learner      resampling iteration      prediction
## > 1: <AutoFSelector[38]> <ResamplingCV[19]>      1 <PredictionClassif[19]>
## > 2: <AutoFSelector[38]> <ResamplingCV[19]>      2 <PredictionClassif[19]>
```

```
rr$aggregate()
```

Aggregates performances of outer folds.

```
as.data.table(rr)$learner[[1]]$fselect_result
```

Retrieves inner feature selection results.

Logging and Parallelization

```
lgr::get_logger("bbotk")$set_threshold("<level>")`
```

Change log-level only for `mlr3fselect`.

```
future::plan(strategy)
```

Sets the parallelization backend. Speeds up feature selection by running iterations in parallel.