

Package ‘rBayesianOptimization’

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Type Package

Title Bayesian Optimization of Hyperparameters

Version 1.2.0

Description A Pure R implementation of Bayesian Global Optimization with Gaussian Processes.

URL <https://github.com/yanyachen/rBayesianOptimization>

BugReports <https://github.com/yanyachen/rBayesianOptimization/issues>

Depends R (>= 2.10)

Imports stats, utils, data.table (>= 1.9.6), magrittr, foreach, GPfit

Suggests xgboost

License GPL-2

RoxygenNote 7.1.1

NeedsCompilation no

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Repository CRAN

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 BayesianOptimization *Bayesian Optimization*

Description

Bayesian Optimization of Hyperparameters.

Usage

```
BayesianOptimization(
  FUN,
  bounds,
  init_grid_dt = NULL,
  init_points = 0,
  n_iter,
  acq = "ucb",
  kappa = 2.576,
  eps = 0,
  kernel = list(type = "exponential", power = 2),
  verbose = TRUE,
  ...
)
```

Arguments

FUN	The function to be maximized. This Function should return a named list with 2 components. The first component "Score" should be the metrics to be maximized, and the second component "Pred" should be the validation/cross-validation prediction for ensembling/stacking.
bounds	A named list of lower and upper bounds for each hyperparameter. The names of the list should be identical to the arguments of FUN. All the sample points in <code>init_grid_dt</code> should be in the range of bounds. Please use "L" suffix to indicate integer hyperparameter.
init_grid_dt	User specified points to sample the target function, should be a <code>data.frame</code> or <code>data.table</code> with identical column names as bounds. User can add one "Value" column at the end, if target function is pre-sampled.
init_points	Number of randomly chosen points to sample the target function before Bayesian Optimization fitting the Gaussian Process.
n_iter	Total number of times the Bayesian Optimization is to repeated.
acq	Acquisition function type to be used. Can be "ucb", "ei" or "poi". <ul style="list-style-type: none"> • ucb GP Upper Confidence Bound • ei Expected Improvement • poi Probability of Improvement

kappa	tunable parameter kappa of GP Upper Confidence Bound, to balance exploitation against exploration, increasing kappa will make the optimized hyperparameters pursuing exploration.
eps	tunable parameter epsilon of Expected Improvement and Probability of Improvement, to balance exploitation against exploration, increasing epsilon will make the optimized hyperparameters are more spread out across the whole range.
kernel	Kernel (aka correlation function) for the underlying Gaussian Process. This parameter should be a list that specifies the type of correlation function along with the smoothness parameter. Popular choices are square exponential (default) or matern 5/2
verbose	Whether or not to print progress.
...	Other arguments passed on to GP_fit .

Value

a list of Bayesian Optimization result is returned:

- Best_Par a named vector of the best hyperparameter set found
- Best_Value the value of metrics achieved by the best hyperparameter set
- History a data.table of the bayesian optimization history
- Pred a data.table with validation/cross-validation prediction for each round of bayesian optimization history

References

Jasper Snoek, Hugo Larochelle, Ryan P. Adams (2012) *Practical Bayesian Optimization of Machine Learning Algorithms*

Examples

```
# Example 1: Optimization
## Set Pred = 0, as placeholder
Test_Fun <- function(x) {
  list(Score = exp(-(x - 2)^2) + exp(-(x - 6)^2/10) + 1/ (x^2 + 1),
       Pred = 0)
}
## Set larger init_points and n_iter for better optimization result
OPT_Res <- BayesianOptimization(Test_Fun,
                               bounds = list(x = c(1, 3)),
                               init_points = 2, n_iter = 1,
                               acq = "ucb", kappa = 2.576, eps = 0.0,
                               verbose = TRUE)

## Not run:
# Example 2: Parameter Tuning
library(xgboost)
data(agaricus.train, package = "xgboost")
dtrain <- xgb.DMatrix(agaricus.train$data,
                     label = agaricus.train$label)
cv_folds <- KFold(agaricus.train$label, nfolds = 5,
```

```

        stratified = TRUE, seed = 0)
xgb_cv_bayes <- function(max_depth, min_child_weight, subsample) {
  cv <- xgb.cv(params = list(booster = "gbtree", eta = 0.01,
    max_depth = max_depth,
    min_child_weight = min_child_weight,
    subsample = subsample, colsample_bytree = 0.3,
    lambda = 1, alpha = 0,
    objective = "binary:logistic",
    eval_metric = "auc"),
    data = dtrain, nround = 100,
    folds = cv_folds, prediction = TRUE, showsd = TRUE,
    early_stopping_rounds = 5, maximize = TRUE, verbose = 0)
  list(Score = cv$evaluation_log$test_auc_mean[cv$best_iteration],
    Pred = cv$pred)
}
OPT_Res <- BayesianOptimization(xgb_cv_bayes,
  bounds = list(max_depth = c(2L, 6L),
    min_child_weight = c(1L, 10L),
    subsample = c(0.5, 0.8)),
  init_grid_dt = NULL, init_points = 10, n_iter = 20,
  acq = "ucb", kappa = 2.576, eps = 0.0,
  verbose = TRUE)

## End(Not run)

```

 KFold

K-Folds cross validation index generator

Description

Generates a list of indices for K-Folds Cross-Validation.

Usage

```
KFold(target, nfolds = 10, stratified = FALSE, seed = 0)
```

Arguments

target	Samples to split in K folds.
nfolds	Number of folds.
stratified	whether to apply Stratified KFold.
seed	random seed to be used.

Value

a list of indices for K-Folds Cross-Validation

rBayesianOptimization *rBayesianOptimization: Bayesian Optimization of Hyperparameters*

Description

A Pure R implementation of bayesian global optimization with gaussian processes.

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