Package: rBayesianOptimization (via r-universe)

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Type Package		
Title Bayesian Optimization of Hyperparameters		
Version 1.2.1		
Description A Pure R implementation of Bayesian Global Optimization with Gaussian Processes.		
<pre>URL https://github.com/yanyachen/rBayesianOptimization</pre>		
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		
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Repository https://yanyachen.r-universe.dev		
RemoteUrl https://github.com/yanyachen/rbayesianoptimization		
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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 init_grid_dt 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 data.frame or

data.table 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".

- ucb GP Upper Confidence Bound
- ei Expected Improvement
- poi Probability of Improvement

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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) {</pre>
  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,</pre>
                                 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$data,</pre>
                      label = agaricus.train$label)
cv_folds <- KFold(agaricus.train$label, nfolds = 5,</pre>
```

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```
stratified = TRUE, seed = 0)
xgb_cv_bayes <- function(max_depth, min_child_weight, subsample) {</pre>
  cv <- xgb.cv(params = list(booster = "gbtree", eta = 0.01,</pre>
                             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,</pre>
                                 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

 $rBayesian Optimization \ \ rBayesian Optimization: Bayesian \ Optimization \ \ of \ Hyperparameters$

Description

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

Author(s)

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See Also

Useful links:

- https://github.com/yanyachen/rBayesianOptimization
- Report bugs at https://github.com/yanyachen/rBayesianOptimization/issues

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