

Package ‘mlsauce’

July 7, 2020

Title Miscellaneous Statistical/Machine Learning stuff

Version 0.4.0

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Description Miscellaneous Statistical/Machine Learning stuff.

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Imports reticulate, R6, Rcpp

Suggests reticulate, R6, Rcpp

Collate 'zzz.R' 'adaopt.R' 'lsboost.R'

Encoding UTF-8

LazyData true

Roxygen list(markdown = TRUE)

RoxygenNote 7.1.1

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AdaOpt	<i>AdaOpt classifier</i>
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Description

AdaOpt classifier

Usage

```
AdaOpt(
  n_iterations = 50L,
  learning_rate = 0.3,
  reg_lambda = 0.1,
  reg_alpha = 0.5,
  eta = 0.01,
  gamma = 0.01,
  k = 3L,
  tolerance = 0,
  n_clusters = 0,
  batch_size = 100L,
  row_sample = 1,
  type_dist = "euclidean-f",
  cache = TRUE,
  seed = 123L
)
```

Arguments

<code>n_iterations</code>	number of iterations of the optimizer at training time
<code>learning_rate</code>	controls the speed of the optimizer at training time
<code>reg_lambda</code>	L2 regularization parameter for successive errors in the optimizer (at training time)
<code>reg_alpha</code>	L1 regularization parameter for successive errors in the optimizer (at training time)
<code>eta</code>	controls the slope in gradient descent (at training time)
<code>gamma</code>	controls the step size in gradient descent (at training time)
<code>k</code>	number of nearest neighbors selected at test time for classification
<code>tolerance</code>	controls early stopping in gradient descent (at training time)
<code>n_clusters</code>	number of clusters, if MiniBatch k-means is used at test time (for faster prediction)
<code>batch_size</code>	size of the batch, if MiniBatch k-means is used at test time (for faster prediction)
<code>row_sample</code>	percentage of rows chosen from training set (by stratified subsampling, for faster prediction)
<code>type_dist</code>	distance used for finding the nearest neighbors; currently euclidean-f (euclidean distances calculated as whole), euclidean (euclidean distances calculated row by row), cosine (cosine distance)
<code>cache</code>	if the nearest neighbors are cached or not, for faster retrieval in subsequent calls
<code>seed</code>	reproducibility seed for initial weak learner and clustering

Value

An object of class `AdaOpt`

Examples

```

library(datasets)

X <- as.matrix(iris[, 1:4])
y <- as.integer(iris[, 5]) - 1L

n <- dim(X)[1]
p <- dim(X)[2]
set.seed(21341)
train_index <- sample(x = 1:n, size = floor(0.8*n), replace = TRUE)
test_index <- -train_index
X_train <- as.matrix(iris[train_index, 1:4])
y_train <- as.integer(iris[train_index, 5]) - 1L
X_test <- as.matrix(iris[test_index, 1:4])
y_test <- as.integer(iris[test_index, 5]) - 1L

obj <- mlsauce::AdaOpt()

print(obj$get_params())

obj$fit(X_train, y_train)

print(obj$score(X_test, y_test))

```

LSBoostClassifier

LSBoost classifier

Description

LSBoost classifier

Usage

```

LSBoostClassifier(
  n_estimators = 100L,
  learning_rate = 0.1,
  n_hidden_features = 5L,
  reg_lambda = 0.1,
  row_sample = 1,
  col_sample = 1,
  dropout = 0,
  tolerance = 1e-04,
  direct_link = 1L,
  verbose = 1L,
  seed = 123L
)

```

Arguments

n_estimators: int, number of boosting iterations.
learning_rate: float, controls the learning speed at training time.
n_hidden_features:
 int
 number of nodes in successive hidden layers.
reg_lambda: float, L2 regularization parameter for successive errors in the optimizer (at training time).
row_sample: float, percentage of rows chosen from the training set.
col_sample: float, percentage of columns chosen from the training set.
dropout: float, percentage of nodes dropped from the training set.
tolerance: float, controls early stopping in gradient descent (at training time).
direct_link: bool, indicates whether the original features are included (True) in model's fitting or not (False).
verbose: int, progress bar (yes = 1) or not (no = 0) (currently).
seed: int, reproducibility seed for nodes_sim=='uniform', clustering and dropout.

Value

An object of class LSBoostClassifier

Examples

```

library(datasets)

X <- as.matrix(iris[, 1:4])
y <- as.integer(iris[, 5]) - 1L

n <- dim(X)[1]
p <- dim(X)[2]
set.seed(21341)
train_index <- sample(x = 1:n, size = floor(0.8*n), replace = TRUE)
test_index <- -train_index
X_train <- as.matrix(iris[train_index, 1:4])
y_train <- as.integer(iris[train_index, 5]) - 1L
X_test <- as.matrix(iris[test_index, 1:4])
y_test <- as.integer(iris[test_index, 5]) - 1L

obj <- mlsauce::LSBoostClassifier()

print(obj$get_params())

obj$fit(X_train, y_train)

print(obj$score(X_test, y_test))

```

LSBoostRegressor	<i>LSBoost Regressor</i>
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Description

LSBoost Regressor

Usage

```
LSBoostRegressor(
    n_estimators = 100L,
    learning_rate = 0.1,
    n_hidden_features = 5L,
    reg_lambda = 0.1,
    row_sample = 1,
    col_sample = 1,
    dropout = 0,
    tolerance = 1e-04,
    direct_link = 1L,
    verbose = 1L,
    seed = 123L
)
```

Arguments

n_estimators: int, number of boosting iterations.

learning_rate: float, controls the learning speed at training time.

n_hidden_features:
int
number of nodes in successive hidden layers.

reg_lambda: float, L2 regularization parameter for successive errors in the optimizer (at training time).

row_sample: float, percentage of rows chosen from the training set.

col_sample: float, percentage of columns chosen from the training set.

dropout: float, percentage of nodes dropped from the training set.

tolerance: float, controls early stopping in gradient descent (at training time).

direct_link: bool, indicates whether the original features are included (True) in model's fitting or not (False).

verbose: int, progress bar (yes = 1) or not (no = 0) (currently).

seed: int, reproducibility seed for nodes_sim=='uniform', clustering and dropout.

Value

An object of class LSBoostRegressor

Examples

```
library(datasets)

X <- as.matrix(datasets::mtcars[, -1])
y <- as.integer(datasets::mtcars[, 1])

n <- dim(X)[1]
p <- dim(X)[2]
set.seed(21341)
train_index <- sample(x = 1:n, size = floor(0.8*n), replace = TRUE)
test_index <- -train_index
X_train <- as.matrix(X[train_index, ])
y_train <- as.integer(y[train_index, ])
X_test <- as.matrix(X[test_index, ])
y_test <- as.integer(y[test_index])

obj <- mlsauce::LSBoostRegressor()

print(obj$get_params())

obj$fit(X_train, y_train)

print(obj$score(X_test, y_test))
```

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