Package 'ngboostForecast'

October 13, 2022

Title Probabilistic Time Series Forecasting

Version 0.1.1

Description
Probabilistic time series forecasting via Natural Gradient Boosting for Probabilistic Prediction.

License Apache License (>= 2)

URL https://github.com/Akai01/ngboostForecast

BugReports https://github.com/Akai01/ngboostForecast/issues

Encoding UTF-8

LazyData true

SystemRequirements Python (>= 3.6)

RoxygenNote 7.2.0

Imports dplyr (>= 1.0.7), forecast (>= 8.15), magrittr (>= 2.0.1), R6 (>= 2.5.1)

Suggests ggplot2 (>= 3.3.5), testthat (>= 3.0.0)

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Depends R (>= 3.6), reticulate (>= 1.20)

NeedsCompilation no

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R topics documented:

`Dist` *NGBoost distributions*

Description

NGBoost distributions

Usage

```
Dist(
 dist = c("Normal", "Bernoulli", "k_categorical", "StudentT", "Laplace", "Cauchy",
    "Exponential", "LogNormal", "MultivariateNormal", "Poisson"),
  k
)
```

Arguments

`dist`	NGBoost distributions. One of the following: • Bernoulli • k_categorical • StudentT • Poisson • Laplace • Cauchy • Exponential • LogNormal • MultivariateNormal • Normal
`k`	Used only with k_categorical and MultivariateNormal

Value

An NGBoost Distribution object

is_exists_conda *Is conda installed?*

Description

Only for internal usage.

Usage

```
is_exists_conda()
```

Value

Logical, TRUE if conda is installed.

Author(s)

Resul Akay

NGBforecast *NGBoost forecasting class*

Description

The main forecasting class.

Value

An NGBforecast class

Methods

Public methods:

- `NGBforecast$new()`
- `NGBforecast$fit()`
- `NGBforecast$forecast()`
- `NGBforecast$feature_importances()`
- `NGBforecast$plot_feature_importance()`
- `NGBforecast$get_params()`
- `NGBforecast$clone()`

Method `new()`**:** Initialize an NGBforecast model.

Usage:

```
NGBforecast$new(
  Dist = NULL,
  Score = NULL,
  Base = NULL,
  natural_gradient = TRUE,
  n_estimators = as.integer(500),
  learning_rate = 0.01,
  minibatch_frac = 1,
  col_sample = 1,
  verbose = TRUE,
  verbose_eval = as.integer(100),
  tol = 1e-04,
  random_state = NULL
)
```

Arguments:

`Dist` Assumed distributional form of `Y|X=x`. An output of `Dist` function, e.g. `Dist('Normal')`

`Score` Rule to compare probabilistic predictions to the observed data. A score from `Scores` function, e.g. `Scores(score = "LogScore")`.

`Base` Base learner. An output of `sklearner` function, e.g. `sklearner(module = "tree", class = "DecisionTreeRegressor", ...)`

`natural_gradient` Logical flag indicating whether the natural gradient should be used

`n_estimators` The number of boosting iterations to fit

`learning_rate` The learning rate

`minibatch_frac` The percent subsample of rows to use in each boosting iteration

`col_sample` The percent subsample of columns to use in each boosting iteration

`verbose` Flag indicating whether output should be printed during fitting. If TRUE it will print logs.

`verbose_eval` Increment (in boosting iterations) at which output should be printed

`tol` Numerical tolerance to be used in optimization

`random_state` Seed for reproducibility.

Returns: An NGBforecast object that can be fit.

Method `fit()`**:** Fit the initialized model.

Usage:

```
NGBforecast$fit(
  y,
  max_lag = 5,
  xreg = NULL,
  test_size = NULL,
  seasonal = TRUE,
  K = frequency(y)/2 - 1,
  train_loss_monitor = NULL,
  val_loss_monitor = NULL,
  early_stopping_rounds = NULL
)
```

Arguments:

`y` A time series (ts) object

`max_lag` Maximum number of lags

`xreg` Optional. A numerical matrix of external regressors, which must have the same number of rows as y.

`test_size` The length of validation set. If it is NULL, then, it is automatically specified.

`seasonal` Boolean. If `seasonal = TRUE` the fourier terms will be used for modeling seasonality.

`K` Maximum order(s) of Fourier terms, used only if `seasonal = TRUE`.

`train_loss_monitor` A custom score or set of scores to track on the training set during training. Defaults to the score defined in the NGBoost constructor. Please do not modify unless you know what you are doing.

`val_loss_monitor` A custom score or set of scores to track on the validation set during training. Defaults to the score defined in the NGBoost constructor. Please do not modify unless you know what you are doing.

`early_stopping_rounds` The number of consecutive boosting iterations during which the loss has to increase before the algorithm stops early.

Returns: NULL

Method `forecast()`**:** Forecast the fitted model

Usage:

```
NGBforecast$forecast(h = 6, xreg = NULL, level = c(80, 95), data_frame = FALSE)
```

Arguments:

`h` Forecast horizon

`xreg` A numerical vector or matrix of external regressors

`level` Confidence level for prediction intervals

`data_frame` Bool. If TRUE, forecast will be returned as a data.frame object, if FALSE it will return a forecast class. If TRUE, `autoplot` will function.

Method `feature_importances()`**:** Return the feature importance for all parameters in the distribution (the higher, the more important the feature).

Usage:

```
NGBforecast$feature_importances()
```

Returns: A data frame

Method `plot_feature_importance()`**:** Plot feature importance

Usage:

```
NGBforecast$plot_feature_importance()
```

Returns: A ggplot object

Method `get_params()`**:** Get parameters for this estimator.

Usage:

```
NGBforecast$get_params(deep = TRUE)
```

Arguments:

deep bool, default = TRUE If True, will return the parameters for this estimator and contained subobjects that are estimators.

Returns: A named list of parameters.

Method `clone()`**:** The objects of this class are cloneable with this method.

Usage:

```
NGBforecast$clone(deep = FALSE)
```

Arguments:

deep Whether to make a deep clone.

Author(s)

Resul Akay

References

Duan, T et. al. (2019), NGBoost: Natural Gradient Boosting for Probabilistic Prediction.

Examples

```
## Not run:

library(ngboostForecast)

model <- NGBforecast$new(Dist = Dist("Normal"),
                         Base = sklearner(module = "linear_model",
                         class = "Ridge"),
                         Score = Scores("LogScore"),
                         natural_gradient = TRUE,
                         n_estimators = 200,
                         learning_rate =  0.1,
                         minibatch_frac = 1,
                         col_sample = 1,
                         verbose = TRUE,
                         verbose_eval = 100,
                         tol = 1e-5)
model$fit(y = AirPassengers, seasonal = TRUE, max_lag = 12, xreg = NULL,
early_stopping_rounds = 10L)
fc <- model$forecast(h = 12, level = c(90, 80), xreg = NULL)

autoplot(fc)

## End(Not run)
```

NGBforecastCV *NGBoost forecasting model selection class*

Description

It is a wrapper for the sklearn GridSearchCV with TimeSeriesSplit.

Methods

Public methods:

- NGBforecastCV$new()
- NGBforecastCV$tune()
- NGBforecastCV$clone()

Method `new()`**:** Initialize an NGBforecastCV model.

Usage:

```
NGBforecastCV$new(
  Dist = NULL,
  Score = NULL,
  Base = NULL,
  natural_gradient = TRUE,
  n_estimators = as.integer(500),
  learning_rate = 0.01,
  minibatch_frac = 1,
  col_sample = 1,
  verbose = TRUE,
  verbose_eval = as.integer(100),
  tol = 1e-04,
  random_state = NULL
)
```

Arguments:

`Dist` Assumed distributional form of `Y|X=x`. An output of `Dist` function, e.g. `Dist('Normal')`

`Score` Rule to compare probabilistic predictions to the observed data. A score from `Scores` function, e.g. `Scores(score = "LogScore")`.

`Base` Base learner. An output of `sklearner` function, e.g. `sklearner(module = "tree", class = "DecisionTreeRegressor", ...)`

`natural_gradient` Logical flag indicating whether the natural gradient should be used

`n_estimators` The number of boosting iterations to fit

`learning_rate` The learning rate

`minibatch_frac` The percent subsample of rows to use in each boosting iteration

`col_sample` The percent subsample of columns to use in each boosting iteration

`verbose` Flag indicating whether output should be printed during fitting. If TRUE it will print logs.

`verbose_eval` Increment (in boosting iterations) at which output should be printed

`tol` Numerical tolerance to be used in optimization

`random_state` Seed for reproducibility.

Returns: An NGBforecastCV object that can be fit.

Method `tune()`**:** Tune ngboosForecast.

Usage:

```
NGBforecastCV$tune(
  y,
  max_lag = 5,
  xreg = NULL,
  seasonal = TRUE,
  K = frequency(y)/2 - 1,
  n_splits = NULL,
  train_loss_monitor = NULL,
  val_loss_monitor = NULL,
  early_stopping_rounds = NULL
)
```

Arguments:

`y` A time series (ts) object

`max_lag` Maximum number of lags

`xreg` Optional. A numerical matrix of external regressors, which must have the same number of rows as y.

`seasonal` Boolean. If `seasonal = TRUE` the fourier terms will be used for modeling seasonality.

`K` Maximum order(s) of Fourier terms, used only if `seasonal = TRUE`.

`n_splits` Number of splits. Must be at least 2.

`train_loss_monitor` A custom score or set of scores to track on the training set during training. Defaults to the score defined in the NGBoost constructor. Please do not modify unless you know what you are doing.

`val_loss_monitor` A custom score or set of scores to track on the validation set during training. Defaults to the score defined in the NGBoost constructor. Please do not modify unless you know what you are doing.

`early_stopping_rounds` The number of consecutive boosting iterations during which the loss has to increase before the algorithm stops early.

`test_size` The length of validation set. If it is NULL, then, it is automatically specified.

Returns: A named list of best parameters.

Method `clone()`**:** The objects of this class are cloneable with this method.

Usage:

```
NGBforecastCV$clone(deep = FALSE)
```

Arguments:

`deep` Whether to make a deep clone.

Author(s)

Resul Akay

References

https://stanfordmlgroup.github.io/ngboost/2-tuning.html

Examples

```
## Not run:

library(ngboostForecast)

dists <- list(Dist("Normal"))

base_learners <- list(sklearner(module = "tree", class = "DecisionTreeRegressor",
                                max_depth = 1),
                      sklearner(module = "tree", class = "DecisionTreeRegressor",
                                max_depth = 2),
                      sklearner(module = "tree", class = "DecisionTreeRegressor",
                                max_depth = 3),
                      sklearner(module = "tree", class = "DecisionTreeRegressor",
                                max_depth = 4),
                      sklearner(module = "tree", class = "DecisionTreeRegressor",
                                max_depth = 5),
                      sklearner(module = "tree", class = "DecisionTreeRegressor",
                                max_depth = 6),
                      sklearner(module = "tree", class = "DecisionTreeRegressor",
                                max_depth = 7))

scores <-  list(Scores("LogScore"))

model <- NGBforecastCV$new(Dist = dists,
                           Base = base_learners,
                           Score = scores,
                           natural_gradient = TRUE,
                           n_estimators = list(10, 100),
                           learning_rate = list(0.1, 0.2),
                           minibatch_frac = list(0.1, 1),
                           col_sample = list(0.3),
                           verbose = FALSE,
                           verbose_eval = 100,
                           tol = 1e-5)

params <- model$tune(y = AirPassengers,
seasonal = TRUE,
max_lag = 12,
xreg = NULL,
early_stopping_rounds = NULL,
n_splits = 4L)

params


## End(Not run)
```

ngboostForecast *Probabilistic Time Series Forecasting*

Description

Probabilistic time series forecasting via Natural Gradient Boosting for Probabilistic Prediction.

References

Duan, T et. al. (2019), NGBoost: Natural Gradient Boosting for Probabilistic Prediction.

Examples

```
## Not run:

library(ngboostForecast)

model <- NGBforecast$new(Dist = Dist("Normal"),
                         Base = sklearner(module = "linear_model",
                         class = "Ridge"),
                         Score = Scores("LogScore"),
                         natural_gradient = TRUE,
                         n_estimators = 200,
                         learning_rate =  0.1,
                         minibatch_frac = 1,
                         col_sample = 1,
                         verbose = TRUE,
                         verbose_eval = 100,
                         tol = 1e-5)
model$fit(y = AirPassengers, seasonal = TRUE, max_lag = 12, xreg = NULL,

early_stopping_rounds = 10L)

fc <- model$forecast(h = 12, level = c(90, 80), xreg = NULL)

autoplot(fc)


## End(Not run)
```

Scores *Select a rule to compare probabilistic predictions to the observed data.*

Description

Select a rule to compare probabilistic predictions to the observed data. A score from ngboost.scores, e.g. LogScore.

Usage

```
Scores(score = c("LogScore", "CRPS", "CRPScore", "MLE"))
```

Arguments

score	A string. can be one of the following: • LogScore : Generic class for the log scoring rule. • CRPS : Generic class for the continuous ranked probability scoring rule. • CRPScore : Generic class for the continuous ranked probability scoring rule. • MLE : Generic class for the log scoring rule.

Value

A score class from ngboost.scores

Author(s)

Resul Akay

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seatbelts *Road Casualties in Great Britain 1969-84*

Description

The Seatbelts dataset from the datasets package.

Usage

```
seatbelts
```

Format

An object of class `mts` (inherits from `ts`) with 192 rows and 8 columns.

Source

Harvey, A.C. (1989). Forecasting, Structural Time Series Models and the Kalman Filter. Cambridge University Press, pp. 519–523.

Durbin, J. and Koopman, S. J. (2001). Time Series Analysis by State Space Methods. Oxford University Press.

https://stat.ethz.ch/R-manual/R-devel/library/datasets/html/UKDriverDeaths.html

References

Harvey, A. C. and Durbin, J. (1986). The effects of seat belt legislation on British road casualties: A case study in structural time series modelling. Journal of the Royal Statistical Society series A, 149, 187–227.

sklearner *Scikit-Learn interface*

Description

Scikit-Learn interface

Usage

```
sklearner(module = "tree", class = "DecisionTreeRegressor", ...)
```

Arguments

module	scikit-learn module name, default is 'tree'.
class	scikit-learn's module class, default is 'DecisionTreeRegressor'
...	Other arguments passed to model class

Author(s)

Resul Akay

Examples

```
## Not run:

sklearner(module = "tree", class = "DecisionTreeRegressor",
criterion="friedman_mse", min_samples_split=2)


## End(Not run)
```

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