Package 'live'

January 15, 2020

Type Package

Title Local Interpretable (Model-Agnostic) Visual Explanations

Version 1.5.13

Description Interpretability of complex machine learning models is a growing concern. This package helps to understand key factors that drive the decision made by complicated predictive model (so called black box model). This is achieved through local approximations that are either based on additive regression like model or CART like model that allows for higher interactions. The methodology is based on Tulio Ribeiro, Singh, Guestrin (2016) <doi:10.1145/2939672.2939778>. More details can be found in Staniak, Biecek (2018) <doi:10.32614/RJ-2018-072>.

URL https://github.com/ModelOriented/live

BugReports https://github.com/ModelOriented/live/issues

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Encoding UTF-8

LazyData true

RoxygenNote 7.0.2

Depends R (>= 3.0.2)

Suggests knitr, rmarkdown, testthat, glmnet, covr, DALEX, RWeka, mda, modeltools

VignetteBuilder knitr

Imports mlr, dplyr, breakDown, data.table, forestmodel, shiny, MASS, ggplot2, gower, e1071

NeedsCompilation no

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Repository CRAN

Date/Publication 2020-01-15 06:30:17 UTC

R topics documented:

`add_predictions` *Add black box predictions to generated dataset*

Description

Add black box predictions to generated dataset

Usage

```
add_predictions(
  to_explain,
  black_box_model,
  data = NULL,
  predict_fun = predict,
  hyperparams = list(),
  ...
)
```

Arguments

`to_explain`	List return by sample_locally function.
`black_box_model`	String with mlr signature of a learner or a model with predict interface.
`data`	Original data frame used to generate new dataset. Need not be provided when a trained model is passed in black_box_model argument.

predict_fun	Either a "predict" function that returns a vector of the same type as response or custom function that takes a model as a first argument, and data used to calculate predictions as a second argument and returns a vector of the same type as respone. Will be used only if a model object was provided in the black_box argument.
hyperparams	Optional list of (hyper)parameters to be passed to mlr::makeLearner.
...	Additional parameters to be passed to predict function.

Value

list of class "live_explorer" consisting of

data	Dataset generated by sample_locally function with response variable.
target	Name of the response variable.
model	Black box model which is being explained.
explained_instance	Instance that is being explained.
sampling_method	Name of used sampling method
fixed_variables	Names of variables which were not sampled
sdevations	Standard deviations of numerical variables

Examples

```
## Not run:
# Train a model inside add_predictions call.
local_exploration1 <- add_predictions(dataset_for_local_exploration,
                                      black_box_model = "regr.svm",
                                      data = wine)
# Pass trained model to the function.
svm_model <- svm(quality ~., data = wine)
local_exploration2 <- add_predictions(dataset_for_local_exploration,
                                      black_box_model = svm_model)

## End(Not run)
```

---

euclidean_kernel *LIME kernel equal to the inverse of euclidean distance.*

Description

LIME kernel equal to the inverse of euclidean distance.

Usage

```
euclidean_kernel(explained_instance, simulated_instance)
```

Arguments

explained_instance	explained instance
simulated_instance	new observation

Value

numeric

`fit_explanation` *Fit white box model to the simulated data.*

Description

Fit white box model to the simulated data.

Usage

```
fit_explanation(
  live_object,
  white_box = "regr.lm",
  kernel = gaussian_kernel,
  standardize = FALSE,
  selection = FALSE,
  response_family = "gaussian",
  predict_type = "response",
  hyperpars = list()
)
```

Arguments

live_object	List return by add_predictions function.
white_box	String, learner name recognized by mlr package.
kernel	function which will be used to calculate distance between simulated observations and explained instance.
standardize	If TRUE, numerical variables will be scaled to have mean 0, variance 1 before fitting explanation model.
selection	If TRUE, variable selection based on glmnet implementation of LASSO will be performed.
response_family	family argument to glmnet (and then glm) function. Default value is "gaussian"

predict_type	Argument passed to mlr::makeLearner() argument "predict.type". Defaults to "response".
hyperpars	Optional list of values of hyperparameteres of a model.

Value

List of class "live_explainer" that consists of

data	Dataset used to fit explanation model (may have less column than the original)
model	Fitted explanation model
explained_instance	Instance that is being explained
weights	Weights used in model fitting
selected_variables	Names of selected variables

Examples

```
## Not run:
fitted_explanation <- fit_explanation(local_exploration1, "regr.lm", selection = TRUE)

## End(Not run)
```

---

gaussian_kernel *LIME kernel from the original article with sigma = 1.*

Description

LIME kernel from the original article with sigma = 1.

Usage

```
gaussian_kernel(explained_instance, simulated_instance)
```

Arguments

explained_instance	explained instance
simulated_instance	new observation

Value

numeric

`identity_kernel` *LIME kernel that treats all observations as equally similar to observation of interest.*

Description

LIME kernel that treats all observations as equally similar to observation of interest.

Usage

```
identity_kernel(explained_instance, simulated_instance)
```

Arguments

`explained_instance`
: explained instance

`simulated_instance`
: new observation

Value

numeric

`live` *live: visualizing interpretable models to explain black box models.*

Description

This package aims to help locally fit and visualize interpretable models similarly to LIME methodology. Interface provided by mlr package is used. Tools are provided to create a simulated dataset of similar observations, fit chosen white box models (GLM and CART in particular) and visualize them. The methodology is based on Tulio Ribeiro, Singh, Guestrin (2016) <doi:10.1145/2939672.2939778>. More details can be found in Staniak, Biecek (2018) <doi:10.32614/RJ-2018-072>.

Important functions

`sample_locally` generates a dataset that will be used for local exploration. `add_predictions` adds black box model predictions to simulated dataset. `fit_explanation` fits a chosen white box model to simulated dataset. generic `plot` function visualizes fitted model. `local_approximation` function can be used with DALEX explainers to perform all the steps of local model exploration.

Example datasets

`wine` Data on wine quality taken from Modeling wine preferences by data mining from physicochemical properties

`live_shiny` *Function that starts a Shiny app which helps use LIVE.*

Description

Function that starts a Shiny app which helps use LIVE.

Usage

```
live_shiny(train_data, black_box_model, target, explained_data = train_data)
```

Arguments

`train_data`	dataset from which observations will be sampled.
`black_box_model`	Pre-trained model with predict interface.
`target`	character, name of the response variable.
`explained_data`	Data frame with predictions to explain.

Value

shiny app

`local_approximation` *Fit local model around the observation: shortcut for DALEX explainer objects*

Description

Fit local model around the observation: shortcut for DALEX explainer objects

Usage

```
local_approximation(
  explainer,
  observation,
  target_variable_name,
  n_new_obs,
  local_model = "regr.lm",
  select_variables = F,
  predict_type = "response",
  kernel_type = gaussian_kernel,
  ...
)
```

Arguments

explainer	a model to be explained, preprocessed by the DALEX::explain function
observation	a new observation for which predictions need to be explained
target_variable_name	name of the response variablea as a character
n_new_obs	Number of observation in the simulated dataset
local_model	Character specyfing mlr learner to be used as a local model
select_variables	If TRUE, variable selection will be performed while fitting the local linear model
predict_type	Argument passed to mlr::makeLearner() argument "predict.type" while fitting the local model. Defaults to "response"
kernel_type	Function which will be used to calculate distances from simulated observation to explained instance
...	Arguments to be passed to sample_locally function

Value

object of class live_explainer. More details in fit_explanation function help.

Examples

```
## Not run:
data('wine')
library(randomForest)
library(DALEX)
rf <- randomForest(quality~., data = wine)
expl <- explain(rf, wine, wine$quality)
live_expl <- local_approximation(expl, wine[5, ], "quality", 500)

## End(Not run)
```

---

local_permutation_importance

Local permutation variable importance

Description

This function calculates local variable importance (variable drop-out) by finding top_n observations closest to the explained instance, performing permutation variable importance and using weighted mean square error as loss function with weights equal to 1 - Gower distances of the closest observations to the explainedi instance.

Usage

```
local_permutation_importance(
  explained_instance,
  data,
  explained_var,
  model,
  top_n = nrow(data)
)
```

Arguments

explained_instance	Data frame with one observation for which prediction will be explained
data	Data from with the same columns as explained_instance
explained_var	Character with the names of response variable
model	Model to be explained
top_n	Number of observation that will be used to calculate local variable importance

Value

list of class "local_permutation_importance" that consists of

residuals	Data frame with names of variables in the dataset ("label") and values of dropout loss ("dropout_loss")
weighted_local_mse	Value of weighted MSE for the whole dataset with weights given by 1 - Gower distance from the explained instance
explained_instance	Explained instance as a data frame

Examples

```
## Not run:
local_permutation_importance(wine[5, ], wine,
                             randomForest(quality~., data = wine),
                             top_n = 1000)

## End(Not run)
```

`plot.live_explainer` *Plotting white box models.*

Description

Plotting white box models.

Usage

```
## S3 method for class 'live_explainer'
plot(x, type = "waterfall", ...)
```

Arguments

`x`	List returned by fit_explanation function.
`type`	Chr, "forest" or "waterfall" depending on which type of plot is to be created. if lm/glm model is used as interpretable approximation.
`...`	Additional parameters that will be passed to plot.broken or plot method. In particular, when number of features is large, top_features argument can be set in plot.broken.

Value

plot (ggplot2 or base)

Examples

```
## Not run:
# Forest plot for regression
plot(fitted_explanation1, type = "forest")
# Waterfall plot
plot(fitted_explanation1, type = "waterfall")
# Plot decision tree
plot(fitted_explanation2)

## End(Not run)
```

plot.local_permutation_importance *Plot local permutation importance*

Description

Plot local permutation importance

Usage

```
## S3 method for class 'local_permutation_importance'
plot(x, ...)
```

Arguments

x	Object of class local_permutation_importance
...	Optional arguments, currently ignored

Value

ggplot2 object

print.live_explainer *Generic print function for live explainer*

Description

Generic print function for live explainer

Usage

```
## S3 method for class 'live_explainer'
print(x, ...)
```

Arguments

x	Object created using fit_explanation function
...	other arguments

print.live_explorer *Generic print function for class live_explorer*

Description

Generic print function for class live_explorer

Usage

```
## S3 method for class 'live_explorer'
print(x, ...)
```

Arguments

x	Object created by sample_locally function or add_predictions function
...	Other arguments

print.local_permutation_importance *Print method for local_permutation_importance class*

Description

Print method for local_permutation_importance class

Usage

```
## S3 method for class 'local_permutation_importance'
print(x, ...)
```

Arguments

x	Object of class local_permutation_importance
...	Optional arguments, currently ignored

`sample_locally` *Generate dataset for local exploration.*

Description

Generate dataset for local exploration.

Usage

```
sample_locally(
  data,
  explained_instance,
  explained_var,
  size,
  method = "live",
  fixed_variables = NULL,
  seed = NULL,
  ...
)
```

Arguments

`data`	Data frame from which new dataset will be simulated.
`explained_instance`	One row data frame with the same variables as in data argument. Local exploration will be performed around this observation.
`explained_var`	Name of a column with the variable to be predicted.
`size`	Number of observations is a simulated dataset.
`method`	If "live", new observations will be created by changing one value per observation. If "permute", new observation will be created by permuting all columns of data. If "normal", numerical features will be sampled from multivariate normal distribution specified by ... arguments mu and Sigma.
`fixed_variables`	names or numeric indexes of columns which will not be changed while sampling.
`seed`	Seed to set before sampling. If NULL, results will not be reproducible.
`...`	Mean and covariance matrix for normal sampling method.

Value

list of class "live_explorer" consisting of

`data`	Dataset generated by sample_locally function with response variable.
`target`	Name of the response variable.
`explained_instance`	Instance that is being explained.

`sampling_method`
: Name of used sampling method

`fixed_variables`
: Names of variables which were not sampled

`sdevations`
: Standard deviations of numerical variables

Examples

```
## Not run:
dataset_for_local_exploration <- sample_locally(data = wine,
                                                explained_instance = wine[5, ],
                                                explained_var = "quality",
                                                size = 50)

## End(Not run)
```

`wine` *Red wine characteristics and quality.*

Description

Popular dataset related to wine samples from north Portugal.

Usage

```
wine
```

Format

Data frame with 1599 rows and 12 columns.

References

P. Cortez, A. Cerdeira, F. Almeida, T. Matos and J. Reis. Modeling wine preferences by data mining from physicochemical properties. In Decision Support Systems, Elsevier, 47(4):547-553, 2009.

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