

Package ‘vivo’

September 7, 2020

Title Variable Importance via Oscillations

Version 0.2.1

Description

Provides an easy to calculate local variable importance measure based on Ceteris Paribus profile and global variable importance measure based on Partial Dependence Profiles.

Depends R (>= 3.0)

License GPL-2

Encoding UTF-8

LazyData true

Imports ggplot2, DALEX

Suggests knitr, rmarkdown, mlbench, randomForest, gridExtra, grid,
lattice, testthat, ingredients

VignetteBuilder knitr

RoxygenNote 7.1.0

URL <https://github.com/ModelOriented/vivo>

BugReports <https://github.com/ModelOriented/vivo/issues>

NeedsCompilation no

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Repository CRAN

Date/Publication 2020-09-07 11:00:02 UTC

R topics documented:

calculate_variable_split	2
calculate_weight	2
global_variable_importance	3
local_variable_importance	4
plot.global_importance	6
plot.local_importance	7

Index**9**

calculate_variable_split

*Internal Function for Split Points for Selected Variables***Description**

This function calculate candidate splits for each selected variable. For numerical variables splits are calculated as percentiles (in general uniform quantiles of the length `grid_points`). For all other variables splits are calculated as unique values.

Usage

```
calculate_variable_split(data, variables = colnames(data), grid_points = 101)
```

Arguments

<code>data</code>	validation dataset. Is used to determine distribution of observations.
<code>variables</code>	names of variables for which splits shall be calculated
<code>grid_points</code>	number of points used for response path

Value

A named list with splits for selected variables

Note

This function is a copy of `calculate_variable_split()` from `ingredients` package with small change.

Author(s)

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calculate_weight

*Calculated empirical density and weight based on variable split.***Description**

This function calculate an empirical density of raw data based on variable split from Ceteris Paribus profiles. Then calculated weight for values generated by `DALEX::predict_profile()`, `DALEX::individual_profile()` or `ingredients::ceteris_paribus()`.

Usage

```
calculate_weight(profiles, data, variable_split)
```

Arguments

profiles data.frame generated by DALEX::predict_profile(), DALEX::individual_profile()
 or ingredients::ceteris_paribus()
data data.frame with raw data to model
variable_split list generated by vivo::calculate_variable_split()

Value

Return an weight based on empirical density.

Examples

```
library("DALEX", warn.conflicts = FALSE, quietly = TRUE)
data(apartments)

split <- vivo::calculate_variable_split(apartments,
                                       variables = colnames(apartments),
                                       grid_points = 101)

library("randomForest", warn.conflicts = FALSE, quietly = TRUE)
apartments_rf_model <- randomForest(m2.price ~ construction.year + surface +
                                   floor + no.rooms, data = apartments)

explainer_rf <- explain(apartments_rf_model, data = apartmentsTest[,2:5],
                       y = apartmentsTest$m2.price)

new_apartment <- data.frame(construction.year = 1998, surface = 88, floor = 2L, no.rooms = 3)

profiles <- predict_profile(explainer_rf, new_apartment)

library("vivo")
calculate_weight(profiles, data = apartments[, 2:5], variable_split = split)
```

global_variable_importance

Global Variable Importance measure based on Partial Dependence profiles.

Description

This function calculate global importance measure.

Usage

```
global_variable_importance(profiles)
```

Arguments

profiles data.frame generated by DALEX::model_profile() or DALEX::variable_profile()

Value

A data.frame of the class global_variable_importance. It's a data.frame with calculated global variable importance measure.

Examples

```
library("DALEX")
data(apartments)

library("randomForest")
apartments_rf_model <- randomForest(m2.price ~ construction.year + surface +
                                   floor + no.rooms, data = apartments)

explainer_rf <- explain(apartments_rf_model, data = apartmentsTest[,2:5],
                       y = apartmentsTest$m2.price)

profiles <- model_profile(explainer_rf)

library("vivo")
global_variable_importance(profiles)
```

local_variable_importance

Local Variable Importance measure based on Ceteris Paribus profiles.

Description

This function calculate local importance measure in eight variants. We obtain eight variants measure through the possible options of three parameters such as absolute_deviation, point and density.

Usage

```
local_variable_importance(
  profiles,
  data,
  absolute_deviation = TRUE,
  point = TRUE,
  density = TRUE,
  grid_points = 101
)
```

Arguments

profiles	data.frame generated by DALEX::predict_profile(), DALEX::individual_profile() or ingredients::ceteris_paribus()
data	data.frame with raw data to model
absolute_deviation	logical parameter, if absolute_deviation = TRUE then measure is calculated as absolute deviation, else is calculated as a root from average squares
point	logical parameter, if point = TRUE then measure is calculated as a distance from f(x), else measure is calculated as a distance from average profiles
density	logical parameter, if density = TRUE then measure is weighted based on the density of variable, else is not weighted
grid_points	maximum number of points for profile calculations, the default values is 101, the same as in ingredients::ceteris_paribus(), if you use a different one, you should also change here

Value

A data.frame of the class local_variable_importance. It's a data.frame with calculated local variable importance measure.

Examples

```
library("DALEX")
data(apartments)

library("randomForest")
apartments_rf_model <- randomForest(m2.price ~ construction.year + surface +
                                   floor + no.rooms, data = apartments)

explainer_rf <- explain(apartments_rf_model, data = apartmentsTest[,2:5],
                      y = apartmentsTest$m2.price)

new_apartment <- data.frame(construction.year = 1998, surface = 88, floor = 2L, no.rooms = 3)

profiles <- predict_profile(explainer_rf, new_apartment)

library("vivo")
local_variable_importance(profiles, apartments[,2:5],
                          absolute_deviation = TRUE, point = TRUE, density = TRUE)

local_variable_importance(profiles, apartments[,2:5],
                          absolute_deviation = TRUE, point = TRUE, density = FALSE)

local_variable_importance(profiles, apartments[,2:5],
                          absolute_deviation = TRUE, point = FALSE, density = TRUE)
```

`plot.global_importance`*Plot Global Variable Importance measure*

Description

Function `plot.global_importance` plots global importance measure based on Partial Dependence profiles.

Usage

```
## S3 method for class 'global_importance'  
plot(x, ..., variables = NULL, type = NULL, title = "Variable importance")
```

Arguments

<code>x</code>	object returned from <code>global_variable_importance()</code> function
<code>...</code>	other object returned from <code>global_variable_importance()</code> function that shall be plotted together
<code>variables</code>	if not NULL then only variables will be presented
<code>type</code>	a character. How variables shall be plotted? Either "bars" (default) or "lines".
<code>title</code>	the plot's title, by default 'Variable importance'

Value

a `ggplot2` object

Examples

```
library("DALEX")  
data(apartments)  
  
library("randomForest")  
apartments_rf_model <- randomForest(m2.price ~ construction.year + surface +  
                                   floor + no.rooms, data = apartments)  
  
explainer_rf <- explain(apartments_rf_model, data = apartmentsTest[,2:5],  
                       y = apartmentsTest$m2.price)  
  
profiles <- model_profile(explainer_rf)  
  
library("vivo")  
measure <- global_variable_importance(profiles)
```



```
explainer_rf <- explain(apartments_rf_model, data = apartmentsTest[,2:5],
                       y = apartmentsTest$m2.price)

new_apartment <- data.frame(construction.year = 1998, surface = 88, floor = 2L, no.rooms = 3)

profiles <- predict_profile(explainer_rf, new_apartment)

library("vivo")
measure1 <- local_variable_importance(profiles, apartments[,2:5],
                                     absolute_deviation = TRUE, point = TRUE, density = FALSE)

plot(measure1)

measure2 <- local_variable_importance(profiles, apartments[,2:5],
                                     absolute_deviation = TRUE, point = TRUE, density = TRUE)
plot(measure1, measure2, color = "_label_method_", type = "lines")
```


Index

`calculate_variable_split`, [2](#)
`calculate_weight`, [2](#)

`global_variable_importance`, [3](#)

`local_variable_importance`, [4](#)

`plot.global_importance`, [6](#)
`plot.local_importance`, [7](#)