

# Package ‘vivo’

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**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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## 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**

data	validation dataset. Is used to determine distribution of observations.
variables	names of variables for which splits shall be calculated
grid_points	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)**

Przemyslaw Biecek

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)
```

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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

x	object returned from global_variable_importance() function
...	other object returned from global_variable_importance() function that shall be plotted together
variables	if not NULL then only variables will be presented
type	a character. How variables shall be plotted? Either "bars" (default) or "lines".
title	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](#)