# Package 'sgboost'

February 17, 2025

Title Sparse-Group Boosting								
Version 0.2.0								
Description  Sparse-group boosting to be used in conjunction with the 'mboost' for modeling grouped data.  Applicable to all sparse-group lasso type problems where within-group and betweengroup sparsity is desired.  Interprets and visualizes individual variables and groups.								
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balance

Balances selection frequencies for unequal groups

# **Description**

Returns optimal degrees of freedom for group boosting to achieve more balanced variables selection. Groups should be defined through group\_df. Each base\_learner

# Usage

```
balance(
  df = NULL,
  group_df = NULL,
 blearner = "bols",
  outcome_name = "y",
  group_name = "group_name",
 var_name = "var_name",
  n_{reps} = 3000,
  iterations = 15,
  nu = 0.5,
  red_fact = 0.9,
 min_weights = 0.01,
 max_weights = 0.99,
  intercept = TRUE,
  verbose = F
)
```

# **Arguments**

df	data.frame to be analyzed
group_df	input data.frame containing variable names with group structure. All variables in df to used in the analysis must be present in this data.frame.
blearner	Type of baselearner. Default is 'bols'.
outcome_name	String indicating the name of dependent variable. Default is "y"
group_name	Name of column in group_df indicating the group structure of the variables. Default is "group_name.
var_name	Name of column in group_df containing the variable names to be used as predictors. Default is "var_name". should not contain categorical variables with more than two categories, as they are then treated as a group only.

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n_reps	Number of samples to be drawn in each iteration
iterations	Number of iterations performed in the algorithm. Default is "20"
nu	Learning rate as the step size to move away from the current estimate. Default is $\emptyset.5$
red_fact	Factor by which the learning rate is reduced if the algorithm overshoots, meaning the loss increases. Default is $0.9$
min_weights	The minimum weight size to be used. Default is 0.01
max_weights	The maximum weight size to be used. Default is 0.99
intercept	Logical, should intercept be used?
verbose	Logical, should iteration be printed?

#### Value

Character containing the formula to be passed to mboost::mboost() yielding the sparse-group boosting for a given value mixing parameter alpha.

#### **Examples**

```
library(mboost)
library(dplyr)
set.seed(1)
df <- data.frame(</pre>
  x1 = rnorm(100), x2 = rnorm(100), x3 = rnorm(100),
  x4 = rnorm(100), x5 = runif(100)
df <- df %>%
  mutate_all(function(x) {
    as.numeric(scale(x))
  })
df$y <- df$x1 + df$x4 + df$x5
group_df <- data.frame(</pre>
  group_name = c(1, 1, 1, 2, 2),
  var_name = c("x1", "x2", "x3", "x4", "x5")
)
sgb_formula <- create_formula(alpha = 0.3, group_df = group_df)</pre>
sgb_model <- mboost(formula = sgb_formula, data = df)</pre>
summary(sgb_model)
```

create\_formula

Create a sparse-group boosting formula

# **Description**

Creates a mboost formula that allows to fit a sparse-group boosting model based on boosted Ridge Regression with mixing parameter alpha. The formula consists of a group baselearner part with degrees of freedom 1-alpha and individual baselearners with degrees of freedom alpha. Groups should be defined through group\_df. The corresponding modeling data should not contain categorical variables with more than two categories, as they are then treated as a group only.

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#### Usage

```
create_formula(
  alpha = 0.3,
  group_df = NULL,
  blearner = "bols",
  outcome_name = "y",
  group_name = "group_name",
  var_name = "var_name",
  group_weights = "group_weights",
  intercept = FALSE
)
```

#### **Arguments**

alpha Numeric mixing parameter. For alpha = 0 only group baselearners and for alpha

= 1 only individual baselearners are defined.

group\_df input data.frame containing variable names with group structure.

blearner Type of baselearner. Default is 'bols'.

outcome\_name String indicating the name of dependent variable. Default is "y"

group\_name Name of column in group\_df indicating the group structure of the variables.

Default is "group\_name.

var\_name Name of column in group\_df containing the variable names to be used as pre-

dictors. Default is "var\_name". should not contain categorical variables with

more than two categories, as they are then treated as a group only.

group\_weights Optional name of the column in group\_df indication the group weights.

intercept Logical, should intercept be used?

#### Value

Character containing the formula to be passed to mboost::mboost() yielding the sparse-group boosting for a given value mixing parameter alpha.

```
library(mboost)
library(dplyr)
set.seed(1)
df <- data.frame(
    x1 = rnorm(100), x2 = rnorm(100), x3 = rnorm(100),
    x4 = rnorm(100), x5 = runif(100)
)
df <- df %>%
    mutate_all(function(x) {
        as.numeric(scale(x))
    })
df$y <- df$x1 + df$x4 + df$x5
group_df <- data.frame(
    group_name = c(1, 1, 1, 2, 2),</pre>
```

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```
var_name = c("x1", "x2", "x3", "x4", "x5")
)

sgb_formula <- create_formula(alpha = 0.3, group_df = group_df)
sgb_model <- mboost(formula = sgb_formula, data = df)
summary(sgb_model)</pre>
```

get\_coef

Aggregated and raw coefficients in a sparse group boosting model

#### **Description**

Computes the aggregated coefficients from group and individual baselearners. Also returns the raw coefficients associated with each baselearner.

# Usage

```
get_coef(sgb_model)
```

#### **Arguments**

sgb\_model

Model of type mboost to compute the coefficients for.

#### **Details**

in a sparse group boosting models a variable in a dataset can be selected as an individual variable or as a group. Therefore there can be two associated effect sizes for the same variable. This function aggregates both and returns it in a data.frame.

# Value

List of data.frames containing the a data.frame '\$raw' with the variable and the raw (Regression) coefficients and the data.frame '\$aggregated' with the aggregated (Regression) coefficients.

```
library(mboost)
library(dplyr)
set.seed(1)
df <- data.frame(
    x1 = rnorm(100), x2 = rnorm(100), x3 = rnorm(100),
    x4 = rnorm(100), x5 = runif(100)
)
df <- df %>%
    mutate_all(function(x) {
        as.numeric(scale(x))
    })
df$y <- df$x1 + df$x4 + df$x5
group_df <- data.frame(</pre>
```

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```
group_name = c(1, 1, 1, 2, 2),
  var_name = c("x1", "x2", "x3", "x4", "x5")
)

sgb_formula <- create_formula(alpha = 0.3, group_df = group_df)
sgb_model <- mboost(formula = sgb_formula, data = df)
sgb_coef <- get_coef(sgb_model)</pre>
```

get\_coef\_path

Path of aggregated and raw coefficients in a sparse-group boosting model

# **Description**

Computes the aggregated coefficients from group and individual baselearners for each boosting iteration.

# Usage

```
get_coef_path(sgb_model)
```

#### **Arguments**

sgb\_model

Model of type mboost to compute the coefficient path for .

# **Details**

in a sparse-group boosting models a variable in a dataset can be selected as an individual variable or as a group. Therefore there can be two associated effect sizes for the same variable. This function aggregates both and returns it in a data.frame for each boosting iteration

#### Value

List of data.frames containing the a data.frame \$raw with the variable and the raw (Regression) coefficients and the data.frame \$aggregated with the aggregated (Regression) coefficients.

#### See Also

```
get_coef()
```

```
library(mboost)
library(dplyr)
set.seed(1)
df <- data.frame(
    x1 = rnorm(100), x2 = rnorm(100), x3 = rnorm(100),
    x4 = rnorm(100), x5 = runif(100)
)</pre>
```

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```
df <- df %>%
  mutate_all(function(x) {
    as.numeric(scale(x))
  })
df$y <- df$x1 + df$x4 + df$x5
group_df <- data.frame(
  group_name = c(1, 1, 1, 2, 2),
    var_name = c("x1", "x2", "x3", "x4", "x5")
)
sgb_formula <- create_formula(alpha = 0.3, group_df = group_df)
sgb_model <- mboost(formula = sgb_formula, data = df)
sgb_coef_path <- get_coef_path(sgb_model)</pre>
```

get\_varimp

Variable importance of a sparse-group boosting model

#### **Description**

Variable importance is computed as relative reduction of loss-function attributed to each predictor (groups and individual variables). Returns a list of two data.frames. The first contains the variable importance of a sparse-group model in a data.frame for each predictor. The second one contains the aggregated relative importance of all groups vs. individual variables.

#### **Usage**

```
get_varimp(sgb_model)
```

#### **Arguments**

sgb\_model

Model of type mboost to compute the variable importance for.

#### Value

List of two data.frames. \$raw contains the name of the variables, group structure and variable importance on both group and individual variable basis. \$group\_importance contains the the aggregated relative importance of all group baselearners and of all individual variables.

#### See Also

```
mboost::varimp() which this function uses.
```

```
library(mboost)
library(dplyr)
set.seed(1)
df <- data.frame(
    x1 = rnorm(100), x2 = rnorm(100), x3 = rnorm(100),</pre>
```

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```
x4 = rnorm(100), x5 = runif(100)
)
df <- df %>%
  mutate_all(function(x) {
    as.numeric(scale(x))
})
df$y <- df$x1 + df$x4 + df$x5
group_df <- data.frame(
    group_name = c(1, 1, 1, 2, 2),
    var_name = c("x1", "x2", "x3", "x4", "x5")
)
sgb_formula <- as.formula(create_formula(alpha = 0.3, group_df = group_df))
sgb_model <- mboost(formula = sgb_formula, data = df)
sgb_varimp <- get_varimp(sgb_model)</pre>
```

plot\_effects

Visualizing a sparse-group boosting model

# **Description**

Radar or scatter/lineplot visualizing the effects sizes relative to the variable importance in a sparse-group boosting model. Works also for a regular mboost model.

#### **Usage**

```
plot_effects(
   sgb_model,
   plot_type = "radar",
   prop = 0,
   n_predictors = 30,
   max_char_length = 5,
   base_size = 8
)
```

# Arguments

sgb\_model

Model of type mboost to be used.

plot\_type

String indicating the type of visualization to use. 'radar' refers to a radar plot using polar coordinates. Here the angle is relative to the cumulative relative importance of predictors and the radius is proportional to the effect size. "clock" does the same as "radar" but uses clock coordinates instead of polar coordinates. "scatter" uses the effect size as y-coordinate and the cumulative relative importance as x-axis in a classical Scatter plot.

prop

Numeric value indicating the minimal importance a predictor/baselearner has to have to be plotted. Default value is zero, meaning all predictors are plotted. By increasing prop the number of plotted variables can be reduced. One can also use n\_predictors for limiting the number of variables to be plotted directly.

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n\_predictors The maximum number of predictors to be plotted. Default is 30. Alternative to prop.

max\_char\_length

The maximum character length of a predictor to be printed. Default is 5. For long variable names one may adjust this number.

base\_size

The base\_size argument to be passed to the ggplot2:theme\_ggplot2::theme\_classic to be used to control the overall size of the figure. Default value is 8.

#### Value

ggplot2 object mapping the effect sizes and variable importance.

#### See Also

get\_coef(), get\_varimp() which this function uses.

# **Examples**

```
library(mboost)
library(dplyr)
set.seed(1)
df <- data.frame(</pre>
  x1 = rnorm(100), x2 = rnorm(100), x3 = rnorm(100),
  x4 = rnorm(100), x5 = runif(100)
df <- df %>%
  mutate_all(function(x) {
    as.numeric(scale(x))
df$y <- df$x1 + df$x4 + df$x5
group_df <- data.frame(</pre>
  group_name = c(1, 1, 1, 2, 2),
  var_name = c("x1", "x2", "x3", "x4", "x5")
)
sgb_formula <- as.formula(create_formula(alpha = 0.3, group_df = group_df))</pre>
sgb_model <- mboost(formula = sgb_formula, data = df)</pre>
plot_effects(sgb_model)
```

plot\_path

Coefficient path of a sparse-group boosting model

#### **Description**

Shows how the effect sizes change throughout the boosting iterations in a sparse-group boosting model. Works also for a regular mboost models. Color indicates the selection of group or individual variables within a boosting iteration.

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#### Usage

```
plot_path(sgb_model, max_char_length = 5, base_size = 8)
```

#### Arguments

 $\begin{tabular}{ll} {\tt sgb\_model} & {\tt Model} \ of \ type \ {\tt mboost} \ to \ be \ used. \\ {\tt max\_char\_length} & \end{tabular}$ 

The maximum character length of a predictor to be printed. Default is 5. For long variable names one may adjust this number.

base\_size The base\_size argument to be passed to the ggplot2 theme ggplot2::theme\_bw to be used to control the overall size of the figure. Default value is 8.

#### Value

ggplot2 object mapping the effect sizes and variable importance.

#### See Also

get\_coef\_path() which this function uses.

```
library(mboost)
library(dplyr)
set.seed(1)
df <- data.frame(</pre>
  x1 = rnorm(100), x2 = rnorm(100), x3 = rnorm(100),
  x4 = rnorm(100), x5 = runif(100)
df <- df %>%
  mutate_all(function(x) {
    as.numeric(scale(x))
df$y <- df$x1 + df$x4 + df$x5
group_df <- data.frame(</pre>
  group_name = c(1, 1, 1, 2, 2),
  var_name = c("x1", "x2", "x3", "x4", "x5")
)
sgb_formula <- as.formula(create_formula(alpha = 0.4, group_df = group_df))</pre>
sgb_model <- mboost(formula = sgb_formula, data = df)</pre>
plot_path(sgb_model)
```

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plot\_varimp

Variable importance bar plot of a sparse group boosting model

# **Description**

Visualizes the variable importance of a sparse-group boosting model. Color indicates if a predictor is an individual variable or a group.

#### Usage

```
plot_varimp(
   sgb_model,
   prop = 0,
   n_predictors = 30,
   max_char_length = 15,
   base_size = 8
)
```

#### **Arguments**

sgb\_model Model of type mboost to plot the variable importance.

prop Numeric value indicating the minimal importance a predictor/baselearner has

to have. Default value is zero, meaning all predictors are plotted. By increasing prop the number of plotted variables can be reduced. One can also use

'n\_predictors' for limiting the number of variables to be plotted directly.

n\_predictors The maximum number of predictors to be plotted. Default is 30. Alternative to

'prop'.

max\_char\_length

The maximum character length of a predictor to be printed. Default is 15. For larger groups or long variable names one may adjust this number to differentiate

variables from groups.

base\_size The base\_size argument to be passed to the ggplot2 theme ggplot2::theme\_bw

to be used to control the overall size of the figure. Default value is 8.

#### **Details**

Note that aggregated group and individual variable importance printed in the legend is based only on the plotted variables and not on all variables that were selected in the sparse-group boosting model.

#### Value

object of type ggplot2.

#### See Also

get\_varimp which this function uses.

plot\_varimp

```
library(mboost)
library(dplyr)
set.seed(1)
df <- data.frame(</pre>
  x1 = rnorm(100), x2 = rnorm(100), x3 = rnorm(100),
  x4 = rnorm(100), x5 = runif(100)
)
df <- df %>%
  mutate_all(function(x) {
    as.numeric(scale(x))
  })
df$y <- df$x1 + df$x4 + df$x5
group_df <- data.frame(</pre>
  group_name = c(1, 1, 1, 2, 2),
  var_name = c("x1", "x2", "x3", "x4", "x5")
)
sgb\_formula \leftarrow as.formula(create\_formula(alpha = 0.3, group\_df = group\_df))
sgb_model <- mboost(formula = sgb_formula, data = df)</pre>
sgb_varimp <- plot_varimp(sgb_model)</pre>
```

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