

# Package ‘easyml’

June 26, 2017

**Version** 0.1.0

**Title** Easily Build and Evaluate Machine Learning Models

**Description** Easily build and evaluate machine learning models on a dataset. Machine learning models supported include penalized linear models, penalized linear models with interactions, random forest, support vector machines, neural networks, and deep neural networks.

**License** MIT + file LICENSE

**LazyData** true

**Depends** R (>= 3.3.1)

**Imports** caret, corrplot, darch, dummies, e1071, futile.logger, ggplot2, glinternet, glmnet, parallel, pbapply, pbmccapply, pROC, nnet, randomForest, scales, scorer

**Suggests** covr, lintr, testthat, knitr, rmarkdown

**URL** <https://github.com/CCS-Lab/easyml>

**BugReports** <https://github.com/CCS-Lab/easyml/issues>

**RoxygenNote** 6.0.1

**NeedsCompilation** no

**Author** Woo-Young Ahn [aut, cre],  
Paul Hendricks [aut],  
OSU-CCSL [cph]

**Maintainer** Woo-Young Ahn <ahn.280@osu.edu>

**Repository** CRAN

**Date/Publication** 2017-06-26 06:13:47 UTC

## R topics documented:

cocaine_dependence . . . . .	3
correlation_test . . . . .	4
easyml . . . . .	5
easy_analysis . . . . .	5

easy_avNNet . . . . .	8
easy_deep_neural_network . . . . .	10
easy_glinternet . . . . .	12
easy_glmnet . . . . .	15
easy_neural_network . . . . .	17
easy_random_forest . . . . .	19
easy_support_vector_machine . . . . .	21
extract_coefficients . . . . .	23
extract_coefficients.easy_glmnet . . . . .	24
extract_variable_importances . . . . .	24
extract_variable_importances.easy_random_forest . . . . .	25
fit_model . . . . .	25
fit_model.easy_avNNet . . . . .	26
fit_model.easy_deep_neural_network . . . . .	26
fit_model.easy_glinternet . . . . .	27
fit_model.easy_glmnet . . . . .	27
fit_model.easy_neural_network . . . . .	28
fit_model.easy_random_forest . . . . .	28
fit_model.easy_support_vector_machine . . . . .	29
generate_coefficients . . . . .	29
generate_model_performance . . . . .	30
generate_predictions . . . . .	30
generate_variable_importances . . . . .	31
measure_auc_score . . . . .	31
measure_correlation_score . . . . .	32
measure_mse_score . . . . .	32
measure_r2_score . . . . .	33
plot_coefficients_processed . . . . .	34
plot_model_performance_binomial_auc_score . . . . .	34
plot_model_performance_gaussian_correlation_score . . . . .	35
plot_model_performance_gaussian_mse_score . . . . .	36
plot_model_performance_gaussian_r2_score . . . . .	36
plot_model_performance_histogram . . . . .	37
plot_predictions_binomial . . . . .	38
plot_predictions_gaussian . . . . .	38
plot_roc_curve . . . . .	39
plot_variable_importances_processed . . . . .	40
predict_model . . . . .	40
predict_model.easy_avNNet . . . . .	41
predict_model.easy_deep_neural_network . . . . .	42
predict_model.easy_glinternet . . . . .	42
predict_model.easy_glmnet . . . . .	43
predict_model.easy_neural_network . . . . .	43
predict_model.easy_random_forest . . . . .	44
predict_model.easy_support_vector_machine . . . . .	44
preprocess_identity . . . . .	45
preprocess_scale . . . . .	45
process_coefficients . . . . .	46

process_variable_importances . . . . .	47
prostate . . . . .	47
reduce_cores . . . . .	48
remove_variables . . . . .	49
resample_fold_train_test_split . . . . .	49
resample_simple_train_test_split . . . . .	50
resample_stratified_class_train_test_split . . . . .	51
resample_stratified_simple_train_test_split . . . . .	52
set_categorical_variables . . . . .	52
set_column_names . . . . .	53
set_cores . . . . .	54
set_dependent_variable . . . . .	54
set_independent_variables . . . . .	55
set_looper . . . . .	56
set_looper_ . . . . .	56
set_measure . . . . .	57
set_parallel . . . . .	58
set_plot_model_performance . . . . .	58
set_plot_predictions . . . . .	59
set_preprocess . . . . .	60
set_random_state . . . . .	60
set_resample . . . . .	61

## Index 62

---

cocaine_dependence	<i>Cocaine data.</i>
--------------------	----------------------

---

### Description

Cocaine data.

### Usage

cocaine\_dependence

### Format

Data frame with columns

**subject** Subject ID

**diagnosis** Diagnosis status

**age** Age of Subject

**male** Gender of subject

**edu\_yrs** Education of subject in years

**imt\_comm\_errors** Immediate Memory Task commission errors

**imt\_omis\_errors** Immediate Memory Task omission errors

**a\_imt** Immediate Memory Task non-parametric discriminability  
**b\_d\_imt** Immediate Memory Task response bias  
**stop\_ssrt** Stop-Signal Task stop-signal reaction time  
**lnk\_adjdd** log(Adjusting Delay-Discounting Task discounting rate)  
**lkhat\_kirby** log(Kirby Monetary-Choice Delay-Discounting Questionnaire)  
**revlr\_per\_errors** Probabilistic Reversal-Learning Task errors  
**bis\_attention** Barratt Impulsiveness Scale attentional impulsivity  
**bis\_motor** Barratt Impulsiveness Scale motor impulsivity  
**bis\_nonpL** Barratt Impulsiveness Scale nonplanning impulsivity  
**igt\_total** Iowa Gambling Task total score

### Source

[https://github.com/CCS-Lab/easym1/blob/master/Python/datasets/cocaine\\_dependence.csv](https://github.com/CCS-Lab/easym1/blob/master/Python/datasets/cocaine_dependence.csv)

### See Also

Other data: [prostate](#)

### Examples

```
data("cocaine_dependence", package = "easym1")
```

---

correlation_test	<i>Compute the matrix of p-value.</i>
------------------	---------------------------------------

---

### Description

See here for source: <http://www.sthda.com/english/wiki/visualize-correlation-matrix-using-correlogram>.

### Usage

```
correlation_test(x, confidence_level = 0.95, ...)
```

### Arguments

**x** An object of class "DataFrame" or "matrix".  
**confidence\_level** confidence level for the returned confidence interval. Currently only used for the Pearson product moment correlation coefficient if there are at least 4 complete pairs of observations.  
**...** further arguments to be passed to or from `cor.test`.

**Value**

A list containing three matrices; `p_value`, `lower_bound`, and `upper_bound`.

**See Also**

Other utils: [process\\_coefficients](#), [process\\_variable\\_importances](#), [reduce\\_cores](#), [remove\\_variables](#)

---

easym1	<i>easym1: Easily build and evaluate machine learning models.</i>
--------	---

---

**Description**

easym1: Easily build and evaluate machine learning models.

---

easy_analysis	<i>The core recipe of easym1.</i>
---------------	-----------------------------------

---

**Description**

This recipe is the workhorse behind all of the `easy_*` functions.

**Usage**

```
easy_analysis(.data, dependent_variable, algorithm, family = "gaussian",
  resample = NULL, preprocess = NULL, measure = NULL,
  exclude_variables = NULL, categorical_variables = NULL,
  train_size = 0.667, foldid = NULL, survival_rate_cutoff = 0.05,
  n_samples = 1000, n_divisions = 1000, n_iterations = 10,
  random_state = NULL, progress_bar = TRUE, n_core = 1,
  coefficients = NULL, variable_importances = NULL, predictions = NULL,
  model_performance = NULL, model_args = list())
```

**Arguments**

<code>.data</code>	A <code>data.frame</code> ; the data to be analyzed.
<code>dependent_variable</code>	A character vector of length one; the dependent variable for this analysis.
<code>algorithm</code>	A character vector of length one; the algorithm to run on the data. Choices are currently one of <code>c("deep_neural_network", "glinternet", "glmnet", "neural_network", "random_forest", "support_vector_machine")</code> .
<code>family</code>	A character vector of length one; the type of regression to run on the data. Choices are one of <code>c("gaussian", "binomial")</code> . Defaults to <code>"gaussian"</code> .
<code>resample</code>	A function; the function for resampling the data. Defaults to <code>NULL</code> .
<code>preprocess</code>	A function; the function for preprocessing the data. Defaults to <code>NULL</code> .

measure	A function; the function for measuring the results. Defaults to NULL.
exclude_variables	A character vector; the variables from the data set to exclude. Defaults to NULL.
categorical_variables	A character vector; the variables that are categorical. Defaults to NULL.
train_size	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
foldid	A vector with length equal to length(y) which identifies cases belonging to the same fold.
survival_rate_cutoff	A numeric vector of length one; for <a href="#">easy_glmnet</a> , specifies the minimal threshold (as a percentage) a coefficient must appear out of n_samples. Defaults to 0.05.
n_samples	An integer vector of length one; specifies the number of times the coefficients and predictions should be generated. Defaults to 1000.
n_divisions	An integer vector of length one; specifies the number of times the data should be divided when replicating the measures of model performance. Defaults to 1000.
n_iterations	An integer vector of length one; during each division, specifies the number of times the predictions should be generated. Defaults to 10.
random_state	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.
progress_bar	A logical vector of length one; specifies whether to display a progress bar during calculations. Defaults to TRUE.
n_core	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to 1.
coefficients	A logical vector of length one; whether or not to generate coefficients for this analysis.
variable_importances	A logical vector of length one; whether or not to generate variable importances for this analysis.
predictions	A logical vector of length one; whether or not to generate predictions for this analysis.
model_performance	A logical vector of length one; whether or not to generate measures of model performance for this analysis.
model_args	A list; the arguments to be passed to the algorithm specified.

### Value

A list of class `easy_*`, where `*` is the name of the algorithm.

**call** An object of class `call`; the original function call.

**data** A `data.frame`; the original data.

**dependent\_variable** A character vector of length one; the dependent variable for this analysis.

**algorithm** A character vector of length one; the algorithm to run on the data.

**class** A character vector of length one; the class of the object.

**family** A character vector of length one; the type of regression to run on the data. Choices are one of c("gaussian", "binomial"). Defaults to "gaussian".

**resample** A function; the function for resampling the data.

**preprocess** A function; the function for preprocessing the data.

**measure** A function; the function for measuring the results.

**exclude\_variables** A character vector; the variables from the data set to exclude.

**train\_size** A numeric vector of length one; specifies what proportion of the data should be used for the training data set.

**survival\_rate\_cutoff** A numeric vector of length one; for `easy_glmnet`, specifies the minimal threshold (as a percentage) a coefficient must appear out of `n_samples`.

**n\_samples** An integer vector of length one; specifies the number of times the coefficients and predictions should be generated.

**n\_divisions** An integer vector of length one; specifies the number of times the data should be divided when generating measures of model performance.

**n\_iterations** An integer vector of length one; during each division, specifies the number of times the predictions should be generated.

**random\_state** An integer vector of length one; specifies the seed to be used for the analysis.

**progress\_bar** A logical vector of length one; specifies whether to display a progress bar during calculations.

**n\_core** An integer vector of length one; specifies the number of cores to use for this analysis.

**generate\_coefficients** A logical vector of length one; whether or not to generate coefficients for this analysis.

**generate\_variable\_importances** A logical vector of length one; whether or not to generate variable importances for this analysis.

**generate\_predictions** A logical vector of length one; whether or not to generate predictions for this analysis.

**generate\_model\_performance** A logical vector of length one; whether or not to generate measures of model performance for this analysis.

**model\_args** A list; the arguments to be passed to the algorithm specified.

**column\_names** A character vector; the column names.

**categorical\_variables** A logical vector; the variables that are categorical.

**X** A data.frame; the full dataset to be used for modeling.

**y** A vector; the full response variable to be used for modeling.

**coefficients** A (n\_variables, n\_samples) matrix; the generated coefficients.

**coefficients\_processed** A data.frame; the coefficients after being processed.

**plot\_coefficients\_processed** A ggplot object; the plot of the processed coefficients.

**X\_train** A data.frame; the train dataset to be used for modeling.

**X\_test** A data.frame; the test dataset to be used for modeling.

**y\_train** A vector; the train response variable to be used for modeling.

**y\_test** A vector; the test response variable to be used for modeling.

**predictions\_train** A (nrow(X\_train), n\_samples) matrix; the train predictions.

**predictions\_test** A (nrow(X\_test), n\_samples) matrix; the test predictions.

**predictions\_train\_mean** A vector; the mean train predictions.

**predictions\_test\_mean** A vector; the mean test predictions.

**plot\_predictions** A function; the function for plotting predictions generated by the model.

**plot\_predictions\_train\_mean** A ggplot object; the plot of the mean train predictions.

**plot\_predictions\_test\_mean** A ggplot object; the plot of the mean test predictions.

**model\_performance\_train** A vector of length n\_divisions; the measures of model performance on the train datasets.

**model\_performance\_test** A vector of length n\_divisions; the measures of model performance on the test datasets.

**plot\_model\_performance** A function; the function for plotting the measures of model performance.

**plot\_model\_performance\_train** A ggplot object; the plot of the measures of model performance on the train datasets.

**plot\_model\_performance\_test** A ggplot object; the plot of the measures of model performance on the test datasets.

### See Also

Other recipes: [easy\\_avNNet](#), [easy\\_deep\\_neural\\_network](#), [easy\\_glinternet](#), [easy\\_glmnet](#), [easy\\_neural\\_network](#), [easy\\_random\\_forest](#), [easy\\_support\\_vector\\_machine](#)

---

easy\_avNNet

*Easily build and evaluate an average neural network model.*

---

### Description

This function wraps the easymml core framework, allowing a user to easily run the easymml methodology for an average neural network model.

### Usage

```
easy_avNNet(.data, dependent_variable, family = "gaussian", resample = NULL,
  preprocess = preprocess_scale, measure = NULL, exclude_variables = NULL,
  categorical_variables = NULL, train_size = 0.667, foldid = NULL,
  survival_rate_cutoff = 0.05, n_samples = 1000, n_divisions = 1000,
  n_iterations = 10, random_state = NULL, progress_bar = TRUE,
  n_core = 1, coefficients = FALSE, variable_importances = FALSE,
  predictions = TRUE, model_performance = TRUE, model_args = list())
```



**Arguments**

<code>.data</code>	A data.frame; the data to be analyzed.
<code>dependent_variable</code>	A character vector of length one; the dependent variable for this analysis.
<code>family</code>	A character vector of length one; the type of regression to run on the data. Choices are one of c("gaussian", "binomial"). Defaults to "gaussian".
<code>resample</code>	A function; the function for resampling the data. Defaults to NULL.
<code>preprocess</code>	A function; the function for preprocessing the data. Defaults to NULL.
<code>measure</code>	A function; the function for measuring the results. Defaults to NULL.
<code>exclude_variables</code>	A character vector; the variables from the data set to exclude. Defaults to NULL.
<code>categorical_variables</code>	A character vector; the variables that are categorical. Defaults to NULL.
<code>train_size</code>	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
<code>foldid</code>	A vector with length equal to length(y) which identifies cases belonging to the same fold.
<code>survival_rate_cutoff</code>	A numeric vector of length one; for <code>easy_glmnet</code> , specifies the minimal threshold (as a percentage) a coefficient must appear out of n_samples. Defaults to 0.05.
<code>n_samples</code>	An integer vector of length one; specifies the number of times the coefficients and predictions should be generated. Defaults to 1000.
<code>n_divisions</code>	An integer vector of length one; specifies the number of times the data should be divided when replicating the measures of model performance. Defaults to 1000.
<code>n_iterations</code>	An integer vector of length one; during each division, specifies the number of times the predictions should be generated. Defaults to 10.
<code>random_state</code>	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.
<code>progress_bar</code>	A logical vector of length one; specifies whether to display a progress bar during calculations. Defaults to TRUE.
<code>n_core</code>	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to 1.
<code>coefficients</code>	A logical vector of length one; whether or not to generate coefficients for this analysis.
<code>variable_importances</code>	A logical vector of length one; whether or not to generate variable importances for this analysis.
<code>predictions</code>	A logical vector of length one; whether or not to generate predictions for this analysis.
<code>model_performance</code>	A logical vector of length one; whether or not to generate measures of model performance for this analysis.
<code>model_args</code>	A list; the arguments to be passed to the algorithm specified.

**Value**

A list of class `easy_avNNet`.

**See Also**

Other recipes: [easy\\_analysis](#), [easy\\_deep\\_neural\\_network](#), [easy\\_glinternet](#), [easy\\_glmnet](#), [easy\\_neural\\_network](#), [easy\\_random\\_forest](#), [easy\\_support\\_vector\\_machine](#)

**Examples**

```
## Not run:
library(easym1) # https://github.com/CCS-Lab/easym1

# Gaussian
data("prostate", package = "easym1")
results <- easy_avNNet(prostate, "lpsa",
                      n_samples = 10, n_divisions = 10,
                      n_iterations = 2, random_state = 12345,
                      n_core = 1)

# Binomial
data("cocaine_dependence", package = "easym1")
model_args <- list(size = 5, linout = TRUE, trace = FALSE)
results <- easy_avNNet(cocaine_dependence, "diagnosis",
                      family = "binomial",
                      exclude_variables = c("subject"),
                      categorical_variables = c("male"),
                      preprocess = preprocess_scale,
                      n_samples = 10, n_divisions = 10,
                      n_iterations = 2, random_state = 12345,
                      n_core = 1, model_args = model_args)

## End(Not run)
```

---

easy\_deep\_neural\_network

*Easily build and evaluate a deep neural network.*

---

**Description**

This function wraps the easym1 core framework, allowing a user to easily run the easym1 methodology for a deep neural network model.

**Usage**

```
easy_deep_neural_network(.data, dependent_variable, family = "gaussian",
                        resample = NULL, preprocess = preprocess_scale, measure = NULL,
                        exclude_variables = NULL, categorical_variables = NULL,
                        train_size = 0.667, foldid = NULL, survival_rate_cutoff = 0.05,
```

```
n_samples = 1000, n_divisions = 1000, n_iterations = 10,
random_state = NULL, progress_bar = TRUE, n_core = 1,
coefficients = FALSE, variable_importances = FALSE, predictions = TRUE,
model_performance = TRUE, model_args = list())
```

## Arguments

<code>.data</code>	A data.frame; the data to be analyzed.
<code>dependent_variable</code>	A character vector of length one; the dependent variable for this analysis.
<code>family</code>	A character vector of length one; the type of regression to run on the data. Choices are one of <code>c("gaussian", "binomial")</code> . Defaults to "gaussian".
<code>resample</code>	A function; the function for resampling the data. Defaults to NULL.
<code>preprocess</code>	A function; the function for preprocessing the data. Defaults to NULL.
<code>measure</code>	A function; the function for measuring the results. Defaults to NULL.
<code>exclude_variables</code>	A character vector; the variables from the data set to exclude. Defaults to NULL.
<code>categorical_variables</code>	A character vector; the variables that are categorical. Defaults to NULL.
<code>train_size</code>	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
<code>foldid</code>	A vector with length equal to <code>length(y)</code> which identifies cases belonging to the same fold.
<code>survival_rate_cutoff</code>	A numeric vector of length one; for <code>easy_glmnet</code> , specifies the minimal threshold (as a percentage) a coefficient must appear out of <code>n_samples</code> . Defaults to 0.05.
<code>n_samples</code>	An integer vector of length one; specifies the number of times the coefficients and predictions should be generated. Defaults to 1000.
<code>n_divisions</code>	An integer vector of length one; specifies the number of times the data should be divided when replicating the measures of model performance. Defaults to 1000.
<code>n_iterations</code>	An integer vector of length one; during each division, specifies the number of times the predictions should be generated. Defaults to 10.
<code>random_state</code>	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.
<code>progress_bar</code>	A logical vector of length one; specifies whether to display a progress bar during calculations. Defaults to TRUE.
<code>n_core</code>	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to 1.
<code>coefficients</code>	A logical vector of length one; whether or not to generate coefficients for this analysis.
<code>variable_importances</code>	A logical vector of length one; whether or not to generate variable importances for this analysis.

**predictions**     A logical vector of length one; whether or not to generate predictions for this analysis.  
**model\_performance**     A logical vector of length one; whether or not to generate measures of model performance for this analysis.  
**model\_args**     A list; the arguments to be passed to the algorithm specified.

### Value

A list of class `easy_deep_neural_network`.

### See Also

Other recipes: [easy\\_analysis](#), [easy\\_avNNet](#), [easy\\_glinternet](#), [easy\\_glmnet](#), [easy\\_neural\\_network](#), [easy\\_random\\_forest](#), [easy\\_support\\_vector\\_machine](#)

### Examples

```
## Not run:
library(easym1) # https://github.com/CCS-Lab/easym1

# Gaussian
data("prostate", package = "easym1")
results <- easy_deep_neural_network(prostate, "lpsa",
                                   n_samples = 10, n_divisions = 10,
                                   n_iterations = 2, random_state = 12345,
                                   n_core = 1)

# Binomial
data("cocaine_dependence", package = "easym1")
results <- easy_deep_neural_network(cocaine_dependence, "diagnosis",
                                   family = "binomial",
                                   exclude_variables = c("subject"),
                                   categorical_variables = c("male"),
                                   preprocess = preprocess_scale,
                                   n_samples = 10, n_divisions = 10,
                                   n_iterations = 2, random_state = 12345,
                                   n_core = 1)

## End(Not run)
```

---

easy_glinternet	<i>Easily build and evaluate a penalized regression model with interactions.</i>
-----------------	--

---

### Description

This function wraps the `easym1` core framework, allowing a user to easily run the `easym1` methodology for a `glinternet` model.

**Usage**

```
easy_glinternet(.data, dependent_variable, family = "gaussian",
  resample = NULL, preprocess = preprocess_scale, measure = NULL,
  exclude_variables = NULL, categorical_variables = NULL,
  train_size = 0.667, foldid = NULL, survival_rate_cutoff = 0.05,
  n_samples = 1000, n_divisions = 1000, n_iterations = 10,
  random_state = NULL, progress_bar = TRUE, n_core = 1,
  coefficients = FALSE, variable_importances = FALSE, predictions = TRUE,
  model_performance = TRUE, model_args = list())
```

**Arguments**

<code>.data</code>	A data.frame; the data to be analyzed.
<code>dependent_variable</code>	A character vector of length one; the dependent variable for this analysis.
<code>family</code>	A character vector of length one; the type of regression to run on the data. Choices are one of c("gaussian", "binomial"). Defaults to "gaussian".
<code>resample</code>	A function; the function for resampling the data. Defaults to NULL.
<code>preprocess</code>	A function; the function for preprocessing the data. Defaults to NULL.
<code>measure</code>	A function; the function for measuring the results. Defaults to NULL.
<code>exclude_variables</code>	A character vector; the variables from the data set to exclude. Defaults to NULL.
<code>categorical_variables</code>	A character vector; the variables that are categorical. Defaults to NULL.
<code>train_size</code>	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
<code>foldid</code>	A vector with length equal to length(y) which identifies cases belonging to the same fold.
<code>survival_rate_cutoff</code>	A numeric vector of length one; for <a href="#">easy_glmnet</a> , specifies the minimal threshold (as a percentage) a coefficient must appear out of n_samples. Defaults to 0.05.
<code>n_samples</code>	An integer vector of length one; specifies the number of times the coefficients and predictions should be generated. Defaults to 1000.
<code>n_divisions</code>	An integer vector of length one; specifies the number of times the data should be divided when replicating the measures of model performance. Defaults to 1000.
<code>n_iterations</code>	An integer vector of length one; during each division, specifies the number of times the predictions should be generated. Defaults to 10.
<code>random_state</code>	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.
<code>progress_bar</code>	A logical vector of length one; specifies whether to display a progress bar during calculations. Defaults to TRUE.
<code>n_core</code>	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to 1.



easy\_glmnet

*Easily build and evaluate a penalized regression model.***Description**

This function wraps the easymml core framework, allowing a user to easily run the easymml methodology for a glmnet model.

**Usage**

```
easy_glmnet(.data, dependent_variable, family = "gaussian", resample = NULL,
  preprocess = preprocess_scale, measure = NULL, exclude_variables = NULL,
  categorical_variables = NULL, train_size = 0.667, foldid = NULL,
  survival_rate_cutoff = 0.05, n_samples = 1000, n_divisions = 1000,
  n_iterations = 10, random_state = NULL, progress_bar = TRUE,
  n_core = 1, coefficients = TRUE, variable_importances = FALSE,
  predictions = TRUE, model_performance = TRUE, model_args = list())
```

**Arguments**

<code>.data</code>	A data.frame; the data to be analyzed.
<code>dependent_variable</code>	A character vector of length one; the dependent variable for this analysis.
<code>family</code>	A character vector of length one; the type of regression to run on the data. Choices are one of <code>c("gaussian", "binomial")</code> . Defaults to "gaussian".
<code>resample</code>	A function; the function for resampling the data. Defaults to NULL.
<code>preprocess</code>	A function; the function for preprocessing the data. Defaults to NULL.
<code>measure</code>	A function; the function for measuring the results. Defaults to NULL.
<code>exclude_variables</code>	A character vector; the variables from the data set to exclude. Defaults to NULL.
<code>categorical_variables</code>	A character vector; the variables that are categorical. Defaults to NULL.
<code>train_size</code>	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
<code>foldid</code>	A vector with length equal to <code>length(y)</code> which identifies cases belonging to the same fold.
<code>survival_rate_cutoff</code>	A numeric vector of length one; for <a href="#">easy_glmnet</a> , specifies the minimal threshold (as a percentage) a coefficient must appear out of <code>n_samples</code> . Defaults to 0.05.
<code>n_samples</code>	An integer vector of length one; specifies the number of times the coefficients and predictions should be generated. Defaults to 1000.
<code>n_divisions</code>	An integer vector of length one; specifies the number of times the data should be divided when replicating the measures of model performance. Defaults to 1000.





```

preprocess = preprocess_scale,
n_samples = 10, n_divisions = 10,
n_iterations = 2, random_state = 12345,
n_core = 1, model_args = list(alpha = 1.0))

## End(Not run)

```

---

easy\_neural\_network    *Easily build and evaluate a neural network.*

---

## Description

This function wraps the easymml core framework, allowing a user to easily run the easymml methodology for a neural network model.

## Usage

```

easy_neural_network(.data, dependent_variable, family = "gaussian",
  resample = NULL, preprocess = preprocess_scale, measure = NULL,
  exclude_variables = NULL, categorical_variables = NULL,
  train_size = 0.667, foldid = NULL, survival_rate_cutoff = 0.05,
  n_samples = 1000, n_divisions = 1000, n_iterations = 10,
  random_state = NULL, progress_bar = TRUE, n_core = 1,
  coefficients = FALSE, variable_importances = FALSE, predictions = TRUE,
  model_performance = TRUE, model_args = list())

```

## Arguments

.data	A data.frame; the data to be analyzed.
dependent_variable	A character vector of length one; the dependent variable for this analysis.
family	A character vector of length one; the type of regression to run on the data. Choices are one of c("gaussian", "binomial"). Defaults to "gaussian".
resample	A function; the function for resampling the data. Defaults to NULL.
preprocess	A function; the function for preprocessing the data. Defaults to NULL.
measure	A function; the function for measuring the results. Defaults to NULL.
exclude_variables	A character vector; the variables from the data set to exclude. Defaults to NULL.
categorical_variables	A character vector; the variables that are categorical. Defaults to NULL.
train_size	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
foldid	A vector with length equal to length(y) which identifies cases belonging to the same fold.



```

n_iterations = 2, random_state = 12345,
n_core = 1)

# Binomial
data("cocaine_dependence", package = "easym1")
results <- easy_neural_network(cocaine_dependence, "diagnosis",
                             family = "binomial",
                             exclude_variables = c("subject"),
                             categorical_variables = c("male"),
                             preprocess = preprocess_scale,
                             n_samples = 10, n_divisions = 10,
                             n_iterations = 2, random_state = 12345,
                             n_core = 1)

## End(Not run)

```

---

easy\_random\_forest      *Easily build and evaluate a random forest regression model.*

---

## Description

This function wraps the easym1 core framework, allowing a user to easily run the easym1 methodology for a random forest model.

## Usage

```

easy_random_forest(.data, dependent_variable, family = "gaussian",
  resample = NULL, preprocess = preprocess_identity, measure = NULL,
  exclude_variables = NULL, categorical_variables = NULL,
  train_size = 0.667, foldid = NULL, n_samples = 1000,
  n_divisions = 1000, n_iterations = 10, random_state = NULL,
  progress_bar = TRUE, n_core = 1, coefficients = FALSE,
  variable_importances = TRUE, predictions = TRUE,
  model_performance = TRUE, model_args = list())

```

## Arguments

<code>.data</code>	A data.frame; the data to be analyzed.
<code>dependent_variable</code>	A character vector of length one; the dependent variable for this analysis.
<code>family</code>	A character vector of length one; the type of regression to run on the data. Choices are one of <code>c("gaussian", "binomial")</code> . Defaults to "gaussian".
<code>resample</code>	A function; the function for resampling the data. Defaults to NULL.
<code>preprocess</code>	A function; the function for preprocessing the data. Defaults to NULL.
<code>measure</code>	A function; the function for measuring the results. Defaults to NULL.
<code>exclude_variables</code>	A character vector; the variables from the data set to exclude. Defaults to NULL.

categorical_variables	A character vector; the variables that are categorical. Defaults to NULL.
train_size	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
foldid	A vector with length equal to length(y) which identifies cases belonging to the same fold.
n_samples	An integer vector of length one; specifies the number of times the coefficients and predictions should be generated. Defaults to 1000.
n_divisions	An integer vector of length one; specifies the number of times the data should be divided when replicating the measures of model performance. Defaults to 1000.
n_iterations	An integer vector of length one; during each division, specifies the number of times the predictions should be generated. Defaults to 10.
random_state	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.
progress_bar	A logical vector of length one; specifies whether to display a progress bar during calculations. Defaults to TRUE.
n_core	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to 1.
coefficients	A logical vector of length one; whether or not to generate coefficients for this analysis.
variable_importances	A logical vector of length one; whether or not to generate variable importances for this analysis.
predictions	A logical vector of length one; whether or not to generate predictions for this analysis.
model_performance	A logical vector of length one; whether or not to generate measures of model performance for this analysis.
model_args	A list; the arguments to be passed to the algorithm specified.

**Value**

A list of class `easy_random_forest`.

**See Also**

Other recipes: [easy\\_analysis](#), [easy\\_avNNet](#), [easy\\_deep\\_neural\\_network](#), [easy\\_glinternet](#), [easy\\_glmnet](#), [easy\\_neural\\_network](#), [easy\\_support\\_vector\\_machine](#)

**Examples**

```
## Not run:
library(easym1) # https://github.com/CCS-Lab/easym1

# Gaussian
```

```

data("prostate", package = "easym1")
results <- easy_random_forest(prostate, "lpsa",
                             n_samples = 10L,
                             n_divisions = 10,
                             n_iterations = 2,
                             random_state = 12345, n_core = 1)

# Binomial
data("cocaine_dependence", package = "easym1")
results <- easy_random_forest(cocaine_dependence, "diagnosis",
                             family = "binomial",
                             exclude_variables = c("subject"),
                             categorical_variables = c("male"),
                             n_samples = 10,
                             n_divisions = 10,
                             n_iterations = 2,
                             random_state = 12345, n_core = 1)

## End(Not run)

```

---

easy\_support\_vector\_machine

*Easily build and evaluate a support vector machine regression model.*

---

## Description

This function wraps the easym1 core framework, allowing a user to easily run the easym1 methodology for a support vector machine model.

## Usage

```

easy_support_vector_machine(.data, dependent_variable, family = "gaussian",
                             resample = NULL, preprocess = preprocess_scale, measure = NULL,
                             exclude_variables = NULL, categorical_variables = NULL,
                             train_size = 0.667, foldid = NULL, n_samples = 1000,
                             n_divisions = 1000, n_iterations = 10, random_state = NULL,
                             progress_bar = TRUE, n_core = 1, coefficients = FALSE,
                             variable_importances = FALSE, predictions = TRUE,
                             model_performance = TRUE, model_args = list())

```

## Arguments

<code>.data</code>	A data.frame; the data to be analyzed.
<code>dependent_variable</code>	A character vector of length one; the dependent variable for this analysis.
<code>family</code>	A character vector of length one; the type of regression to run on the data. Choices are one of c("gaussian", "binomial"). Defaults to "gaussian".
<code>resample</code>	A function; the function for resampling the data. Defaults to NULL.

preprocess	A function; the function for preprocessing the data. Defaults to NULL.
measure	A function; the function for measuring the results. Defaults to NULL.
exclude_variables	A character vector; the variables from the data set to exclude. Defaults to NULL.
categorical_variables	A character vector; the variables that are categorical. Defaults to NULL.
train_size	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
foldid	A vector with length equal to length(y) which identifies cases belonging to the same fold.
n_samples	An integer vector of length one; specifies the number of times the coefficients and predictions should be generated. Defaults to 1000.
n_divisions	An integer vector of length one; specifies the number of times the data should be divided when replicating the measures of model performance. Defaults to 1000.
n_iterations	An integer vector of length one; during each division, specifies the number of times the predictions should be generated. Defaults to 10.
random_state	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.
progress_bar	A logical vector of length one; specifies whether to display a progress bar during calculations. Defaults to TRUE.
n_core	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to 1.
coefficients	A logical vector of length one; whether or not to generate coefficients for this analysis.
variable_importances	A logical vector of length one; whether or not to generate variable importances for this analysis.
predictions	A logical vector of length one; whether or not to generate predictions for this analysis.
model_performance	A logical vector of length one; whether or not to generate measures of model performance for this analysis.
model_args	A list; the arguments to be passed to the algorithm specified.

**Value**

A list of class `easy_support_vector_machine`.

**See Also**

Other recipes: [easy\\_analysis](#), [easy\\_avNNet](#), [easy\\_deep\\_neural\\_network](#), [easy\\_glinternet](#), [easy\\_glmnet](#), [easy\\_neural\\_network](#), [easy\\_random\\_forest](#)

## Examples

```
## Not run:
library(easym1) # https://github.com/CCS-Lab/easym1

# Gaussian
data("prostate", package = "easym1")
results <- easy_support_vector_machine(prostate, "lpsa",
                                       n_samples = 10,
                                       n_divisions = 10,
                                       n_iterations = 2,
                                       random_state = 1, n_core = 1)

# Binomial
data("cocaine_dependence", package = "easym1")
results <- easy_support_vector_machine(cocaine_dependence, "diagnosis",
                                       family = "binomial",
                                       preprocesss = preprocess_scale,
                                       exclude_variables = c("subject"),
                                       categorical_variables = c("male"),
                                       n_samples = 10,
                                       n_divisions = 10,
                                       n_iterations = 2,
                                       random_state = 1, n_core = 1)

## End(Not run)
```

---

extract\_coefficients *Extract coefficients.*

---

## Description

The generic function for extracting coefficients from a model within the easym1 core framework, if such an operation is applicable for that model. Users can create their own `extract_coefficients` function for their own class of model.

## Usage

```
extract_coefficients(object)
```

## Arguments

`object` A list of class `easy_*`, where `*` is the name of the algorithm.

## Value

A `data.frame`, the generated coefficients.

---

```
extract_coefficients.easy_glmnet
```

*Extract coefficients from a penalized regression model.*

---

### Description

This function wraps the procedure for extracting coefficients from a glmnet model and makes it accessible to the easymml core framework.

### Usage

```
## S3 method for class 'easy_glmnet'  
extract_coefficients(object)
```

### Arguments

object            A list of class easy\_glmnet.

### Value

A data.frame, the replicated penalized regression coefficients.

---

```
extract_variable_importances
```

*Extract variable importances.*

---

### Description

The generic function for extracting variable importances from a model within the easymml core framework, if such an operation is applicable for that model. Users can create their own extract\_variable\_importances function for their own class of model.

### Usage

```
extract_variable_importances(object)
```

### Arguments

object            A list of class easy\_\*, where \* is the name of the algorithm.

### Value

A data.frame, the generated random forest variable importance scores.



---

```
extract_variable_importances.easy_random_forest
```

*Extract variable importance scores from a random forest model.*

---

### Description

This function wraps the procedure for extracting variable importances from a random forest model and makes it accessible to the easymml core framework.

### Usage

```
## S3 method for class 'easy_random_forest'
extract_variable_importances(object)
```

### Arguments

object            A list of class `easy_random_forest`.

### Value

A data.frame, the replicated random forest variable importance scores.

---

```
fit_model
```

*Fit model.*

---

### Description

The generic function for fitting a model within the easymml core framework. Users can create their own `fit_model` function for their own class of model.

### Usage

```
fit_model(object)
```

### Arguments

object            A list of class `easy_*`, where `*` is the name of the algorithm.

### Value

A list of class `easy_*`, where `*` is the name of the algorithm.

fit\_model.easy\_avNNet *Fit an average neural network model.*

---

**Description**

This function wraps the procedure for fitting an average neural network model and makes it accessible to the easymf core framework.

**Usage**

```
## S3 method for class 'easy_avNNet'  
fit_model(object)
```

**Arguments**

object            A list of class easy\_avNNet.

**Value**

A list of class easy\_avNNet.

---

fit\_model.easy\_deep\_neural\_network  
*Fit a deep neural network model.*

---

**Description**

This function wraps the procedure for fitting a deep neural network model and makes it accessible to the easymf core framework.

**Usage**

```
## S3 method for class 'easy_deep_neural_network'  
fit_model(object)
```

**Arguments**

object            A list of class easy\_deep\_neural\_network.

**Value**

A list of class easy\_deep\_neural\_network.

---

`fit_model.easy_glinternet`*Fit a penalized regression model with interactions.*

---

**Description**

This function wraps the procedure for fitting a glinternet model and makes it accessible to the easymf core framework.

**Usage**

```
## S3 method for class 'easy_glinternet'  
fit_model(object)
```

**Arguments**

`object`            A list of class `easy_glinternet`.

**Value**

A list of class `easy_glinternet`.

---

`fit_model.easy_glmnet` *Fit a penalized regression model.*

---

**Description**

This function wraps the procedure for fitting a glmnet model and makes it accessible to the easymf core framework.

**Usage**

```
## S3 method for class 'easy_glmnet'  
fit_model(object)
```

**Arguments**

`object`            A list of class `easy_glmnet`.

**Value**

A list of class `easy_glmnet`.

---

```
fit_model.easy_neural_network
```

*Fit a neural network model.*

---

### Description

This function wraps the procedure for fitting a neural network model and makes it accessible to the easymml core framework.

### Usage

```
## S3 method for class 'easy_neural_network'  
fit_model(object)
```

### Arguments

object            A list of class easy\_neural\_network.

### Value

A list of class easy\_neural\_network.

---

```
fit_model.easy_random_forest
```

*Fit a random forest model.*

---

### Description

This function wraps the procedure for fitting a random forest model and makes it accessible to the easymml core framework.

### Usage

```
## S3 method for class 'easy_random_forest'  
fit_model(object)
```

### Arguments

object            A list of class easy\_random\_forest.

### Value

A list of class easy\_random\_forest.

---

```
fit_model.easy_support_vector_machine
```

*Fit a support vector machine regression model.*

---

**Description**

This function wraps the procedure for fitting a support vector machine model and makes it accessible to the easymml core framework.

**Usage**

```
## S3 method for class 'easy_support_vector_machine'  
fit_model(object)
```

**Arguments**

object            A list of class easy\_support\_vector\_machine.

**Value**

A list of class easy\_support\_vector\_machine.

---

```
generate_coefficients
```

*Generate coefficients for a model (if applicable).*

---

**Description**

Generate coefficients for a model (if applicable).

**Usage**

```
generate_coefficients(object)
```

**Arguments**

object            A list of class easy\_\*, where \* is the name of the algorithm.

**Value**

A data.frame, the generated penalized regression model coefficients.

**See Also**

Other generate: [generate\\_model\\_performance](#), [generate\\_predictions](#), [generate\\_variable\\_importances](#)

generate\_model\_performance

*Generate measures of model performance for a model.*

---

**Description**

Generate measures of model performance for a model.

**Usage**

```
generate_model_performance(object)
```

**Arguments**

object            A list of class easy\_\*, where \* is the name of the algorithm.

**Value**

A list of matrixes, the generated measures of model performance.

**See Also**

Other generate: [generate\\_coefficients](#), [generate\\_predictions](#), [generate\\_variable\\_importances](#)

---

generate\_predictions    *Generate predictions for a model.*

---

**Description**

Generate predictions for a model.

**Usage**

```
generate_predictions(object)
```

**Arguments**

object            A list of class easy\_\*, where \* is the name of the algorithm.

**Value**

A list of matrixes, the generated predictions.

**See Also**

Other generate: [generate\\_coefficients](#), [generate\\_model\\_performance](#), [generate\\_variable\\_importances](#)

---

`generate_variable_importances`*Generate variable importances for a model (if applicable).*

---

**Description**

Generate variable importances for a model (if applicable).

**Usage**

```
generate_variable_importances(object)
```

**Arguments**

`object` A list of class `easy_*`, where `*` is the name of the algorithm.

**Value**

A `data.frame`, the generated variable importance scores.

**See Also**

Other generate: [generate\\_coefficients](#), [generate\\_model\\_performance](#), [generate\\_predictions](#)

---

`measure_auc_score`*Measure area under the curve.*

---

**Description**

Given the ground truth (correct) target values and the estimated target values, calculates the the AUC metric.

**Usage**

```
measure_auc_score(y_true, y_pred)
```

**Arguments**

`y_true` A numeric vector; the ground truth (correct) target values.

`y_pred` A numeric vector; the estimated target values.

**Value**

A numeric vector of length one; the AUC metric.

**See Also**

Other measure: [measure\\_correlation\\_score](#), [measure\\_mse\\_score](#), [measure\\_r2\\_score](#)

---

measure\_correlation\_score

*Measure Pearsons Correlation Coefficient.*

---

### Description

Given the ground truth (correct) target values and the estimated target values, calculates the correlation metric.

### Usage

```
measure_correlation_score(y_true, y_pred)
```

### Arguments

`y_true` A numeric vector; the ground truth (correct) target values.  
`y_pred` A numeric vector; the estimated target values.

### Details

See here for more information: [https://en.wikipedia.org/wiki/Pearson\\_correlation\\_coefficient](https://en.wikipedia.org/wiki/Pearson_correlation_coefficient)

### Value

A numeric vector of length one; the correlation metric.

### See Also

Other measure: [measure\\_auc\\_score](#), [measure\\_mse\\_score](#), [measure\\_r2\\_score](#)

---

measure\_mse\_score

*Measure mean squared error.*

---

### Description

Given the ground truth (correct) target values and the estimated target values, calculates the mean squared error metric.

### Usage

```
measure_mse_score(y_true, y_pred)
```

### Arguments

`y_true` A numeric vector; the ground truth (correct) target values.  
`y_pred` A numeric vector; the estimated target values.



**Value**

A numeric vector of length one; the mean squared error metric.

**See Also**

Other measure: [measure\\_auc\\_score](#), [measure\\_correlation\\_score](#), [measure\\_r2\\_score](#)

---

measure_r2_score	<i>Measure Coefficient of Determination (R<sup>2</sup> Score).</i>
------------------	--

---

**Description**

Given the ground truth (correct) target values and the estimated target values, calculates the the R<sup>2</sup> metric.

**Usage**

```
measure_r2_score(y_true, y_pred)
```

**Arguments**

y\_true            A numeric vector; the ground truth (correct) target values.

y\_pred            A numeric vector; the estimated target values.

**Details**

See here for more information: [https://en.wikipedia.org/wiki/Coefficient\\_of\\_determination](https://en.wikipedia.org/wiki/Coefficient_of_determination)

**Value**

A numeric vector of length one; the R<sup>2</sup> metric.

**See Also**

Other measure: [measure\\_auc\\_score](#), [measure\\_correlation\\_score](#), [measure\\_mse\\_score](#)

---

`plot_coefficients_processed`*Plot penalized regression coefficients.*

---

**Description**

When calling `easy_glmnet`, coefficients from the `generate_coefficients` output are processed by the `process_coefficients` function and generated into a plot. This plot tells us the direction, magnitude, and statistical significance of each coefficient. Be careful using this plotting method with datasets containing more than 20 variables as the plot may not render as nicely.

**Usage**

```
plot_coefficients_processed(coefficients_processed)
```

**Arguments**`coefficients_processed`

A data.frame, the output of the function `process_coefficients`.

**Value**

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

**See Also**

Other plot: `plot_model_performance_binomial_auc_score`, `plot_model_performance_gaussian_correlation_score`, `plot_model_performance_gaussian_mse_score`, `plot_model_performance_gaussian_r2_score`, `plot_model_performance_histogram`, `plot_predictions_binomial`, `plot_predictions_gaussian`, `plot_roc_curve`, `plot_variable_importances_processed`

---

`plot_model_performance_binomial_auc_score`*Plot histogram of the area under the curve (AUC) metrics.*

---

**Description**

This function plots a histogram of the area under the curve (AUC) metrics.

**Usage**

```
plot_model_performance_binomial_auc_score(x)
```

### Arguments

`x` A vector, the area under the curve (AUC) metrics to be plotted as a histogram.

### Value

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

### See Also

Other plot: [plot\\_coefficients\\_processed](#), [plot\\_model\\_performance\\_gaussian\\_correlation\\_score](#), [plot\\_model\\_performance\\_gaussian\\_mse\\_score](#), [plot\\_model\\_performance\\_gaussian\\_r2\\_score](#), [plot\\_model\\_performance\\_histogram](#), [plot\\_predictions\\_binomial](#), [plot\\_predictions\\_gaussian](#), [plot\\_roc\\_curve](#), [plot\\_variable\\_importances\\_processed](#)

---

`plot_model_performance_gaussian_correlation_score`

*Plot histogram of the correlation coefficient metrics.*

---

### Description

This function plots a histogram of the correlation coefficient metrics.

### Usage

```
plot_model_performance_gaussian_correlation_score(x)
```

### Arguments

`x` A vector, the correlation coefficient metrics to be plotted as a histogram.

### Value

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

### See Also

Other plot: [plot\\_coefficients\\_processed](#), [plot\\_model\\_performance\\_binomial\\_auc\\_score](#), [plot\\_model\\_performance\\_gaussian\\_mse\\_score](#), [plot\\_model\\_performance\\_gaussian\\_r2\\_score](#), [plot\\_model\\_performance\\_histogram](#), [plot\\_predictions\\_binomial](#), [plot\\_predictions\\_gaussian](#), [plot\\_roc\\_curve](#), [plot\\_variable\\_importances\\_processed](#)

plot\_model\_performance\_gaussian\_mse\_score

*Plot histogram of the mean squared error metrics.*

---

### Description

This function plots a histogram of the mean squared error metrics.

### Usage

```
plot_model_performance_gaussian_mse_score(x)
```

### Arguments

x                    A vector, the mean squared error metrics to be plotted as a histogram.

### Value

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

### See Also

Other plot: [plot\\_coefficients\\_processed](#), [plot\\_model\\_performance\\_binomial\\_auc\\_score](#), [plot\\_model\\_performance\\_gaussian\\_correlation\\_score](#), [plot\\_model\\_performance\\_gaussian\\_r2\\_score](#), [plot\\_model\\_performance\\_histogram](#), [plot\\_predictions\\_binomial](#), [plot\\_predictions\\_gaussian](#), [plot\\_roc\\_curve](#), [plot\\_variable\\_importances\\_processed](#)

---

plot\_model\_performance\_gaussian\_r2\_score

*Plot histogram of the coefficient of determination (R<sup>2</sup>) metrics.*

---

### Description

This function plots a histogram of the coefficient of determination (R<sup>2</sup>) metrics.

### Usage

```
plot_model_performance_gaussian_r2_score(x)
```

### Arguments

x                    A vector, the coefficient of determination (R<sup>2</sup>) metrics to be plotted as a histogram.

**Value**

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

**See Also**

Other plot: [plot\\_coefficients\\_processed](#), [plot\\_model\\_performance\\_binomial\\_auc\\_score](#), [plot\\_model\\_performance\\_gaussian\\_correlation\\_score](#), [plot\\_model\\_performance\\_gaussian\\_mse\\_score](#), [plot\\_model\\_performance\\_histogram](#), [plot\\_predictions\\_binomial](#), [plot\\_predictions\\_gaussian](#), [plot\\_roc\\_curve](#), [plot\\_variable\\_importances\\_processed](#)

---

plot\_model\_performance\_histogram

*Plot histogram of measures of model performance.*

---

**Description**

Given a numeric vector will plot a histogram for any number of measures of model performance.

**Usage**

```
plot_model_performance_histogram(x, name)
```

**Arguments**

x	A vector, the mean squared error metrics to be plotted as a histogram.
name	A character vector of length one, the name of the metric.

**Value**

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

**See Also**

Other plot: [plot\\_coefficients\\_processed](#), [plot\\_model\\_performance\\_binomial\\_auc\\_score](#), [plot\\_model\\_performance\\_gaussian\\_correlation\\_score](#), [plot\\_model\\_performance\\_gaussian\\_mse\\_score](#), [plot\\_model\\_performance\\_gaussian\\_r2\\_score](#), [plot\\_predictions\\_binomial](#), [plot\\_predictions\\_gaussian](#), [plot\\_roc\\_curve](#), [plot\\_variable\\_importances\\_processed](#)

plot\_predictions\_binomial

*Plot binomial predictions.*

---

### Description

Plots a logistic plot of the ground truth (correct) target values and the estimated target values.

### Usage

```
plot_predictions_binomial(y_true, y_pred)
```

### Arguments

y_true	Ground truth (correct) target values.
y_pred	Estimated target values.

### Value

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

### See Also

Other plot: [plot\\_coefficients\\_processed](#), [plot\\_model\\_performance\\_binomial\\_auc\\_score](#), [plot\\_model\\_performance\\_gaussian\\_correlation\\_score](#), [plot\\_model\\_performance\\_gaussian\\_mse\\_score](#), [plot\\_model\\_performance\\_gaussian\\_r2\\_score](#), [plot\\_model\\_performance\\_histogram](#), [plot\\_predictions\\_gaussian](#), [plot\\_roc\\_curve](#), [plot\\_variable\\_importances\\_processed](#)

---

plot\_predictions\_gaussian

*Plot gaussian predictions.*

---

### Description

Plots a scatter plot of the ground truth (correct) target values and the estimated target values.

### Usage

```
plot_predictions_gaussian(y_true, y_pred)
```

### Arguments

y_true	Ground truth (correct) target values.
y_pred	Estimated target values.

**Value**

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

**See Also**

Other plot: [plot\\_coefficients\\_processed](#), [plot\\_model\\_performance\\_binomial\\_auc\\_score](#), [plot\\_model\\_performance\\_gaussian\\_correlation\\_score](#), [plot\\_model\\_performance\\_gaussian\\_mse\\_score](#), [plot\\_model\\_performance\\_gaussian\\_r2\\_score](#), [plot\\_model\\_performance\\_histogram](#), [plot\\_predictions\\_binomial](#), [plot\\_roc\\_curve](#), [plot\\_variable\\_importances\\_processed](#)

---

plot_roc_curve	<i>Plot ROC Curve.</i>
----------------	------------------------

---

**Description**

Given the ground truth (correct) target values and the estimated target values will plot an ROC curve.

**Usage**

```
plot_roc_curve(y_true, y_pred)
```

**Arguments**

y_true	Ground truth (correct) target values.
y_pred	Estimated target values.

**Value**

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

**See Also**

Other plot: [plot\\_coefficients\\_processed](#), [plot\\_model\\_performance\\_binomial\\_auc\\_score](#), [plot\\_model\\_performance\\_gaussian\\_correlation\\_score](#), [plot\\_model\\_performance\\_gaussian\\_mse\\_score](#), [plot\\_model\\_performance\\_gaussian\\_r2\\_score](#), [plot\\_model\\_performance\\_histogram](#), [plot\\_predictions\\_binomial](#), [plot\\_predictions\\_gaussian](#), [plot\\_variable\\_importances\\_processed](#)

---

plot\_variable\_importances\_processed

*Plot random forest variable importances scores.*

---

### Description

When calling `easy_random_forest`, variable importances scores from the `generate_variable_importances` output are processed by the `process_variable_importances` function and generated into a plot. Importance scores for each predictor were estimated using the increase in node impurity. Node impurity measures the change in residual squared error that is attributable to the predictor across all trees. Unlike the `easy_glmnet` coefficients, random forest importance scores do not indicate directional effects, but instead represent the magnitude of the effect that the predictor has on overall prediction accuracy. Be careful using this plotting method with datasets containing more than 20 variables as the plot may not render as nicely.

### Usage

```
plot_variable_importances_processed(variable_importances_processed)
```

### Arguments

variable\_importances\_processed

A data.frame, the output of the function `process_variable_importances`.

### Value

A ggplot object. This plot may be rendered by outputting it to the command line or modified using ggplot semantics.

### See Also

Other plot: `plot_coefficients_processed`, `plot_model_performance_binomial_auc_score`, `plot_model_performance_gaussian_correlation_score`, `plot_model_performance_gaussian_mse_score`, `plot_model_performance_gaussian_r2_score`, `plot_model_performance_histogram`, `plot_predictions_binomial`, `plot_predictions_gaussian`, `plot_roc_curve`

---

predict\_model

*Predict model.*

---

### Description

The generic function for generating predictions from a model within the easymml core framework. Users can create their own `predict_model` function for their own class of model.



**Usage**

```
predict_model(object, newx = NULL)
```

**Arguments**

<code>object</code>	A list of class <code>easy_*</code> , where <code>*</code> is the name of the algorithm.
<code>newx</code>	A <code>data.frame</code> , the new data to use for predictions.

**Value**

A vector, the predicted values using the new data.

---

`predict_model.easy_avNNet`

*Predict values for an average neural network model.*

---

**Description**

This function wraps the procedure for predicting values from a average neural network model and makes it accessible to the easymf core framework.

**Usage**

```
## S3 method for class 'easy_avNNet'  
predict_model(object, newx = NULL)
```

**Arguments**

<code>object</code>	A list of class <code>easy_avNNet</code> .
<code>newx</code>	A <code>data.frame</code> , the new data to use for predictions.

**Value**

A vector, the predicted values using the new data.

---

```
predict_model.easy_deep_neural_network
```

*Predict values for a deep neural network model.*

---

**Description**

This function wraps the procedure for predicting values from a deep neural network model and makes it accessible to the easymml core framework.

**Usage**

```
## S3 method for class 'easy_deep_neural_network'  
predict_model(object, newx = NULL)
```

**Arguments**

object	A list of class <code>easy_deep_neural_network</code> .
newx	A <code>data.frame</code> , the new data to use for predictions.

**Value**

A vector, the predicted values using the new data.

---

```
predict_model.easy_glinternet
```

*Predict values for a penalized regression model with interactions.*

---

**Description**

This function wraps the procedure for predicting values from a glinternet model and makes it accessible to the easymml core framework.

**Usage**

```
## S3 method for class 'easy_glinternet'  
predict_model(object, newx = NULL)
```

**Arguments**

object	A list of class <code>easy_glinternet</code> .
newx	A <code>data.frame</code> , the new data to use for predictions.

**Value**

A vector, the predicted values using the new data.

---

`predict_model.easy_glmnet`*Predict values for a penalized regression model.*

---

**Description**

This function wraps the procedure for predicting values from a glmnet model and makes it accessible to the easymml core framework.

**Usage**

```
## S3 method for class 'easy_glmnet'  
predict_model(object, newx = NULL)
```

**Arguments**

<code>object</code>	A list of class <code>easy_glmnet</code> .
<code>newx</code>	A <code>data.frame</code> , the new data to use for predictions.

**Value**

A vector, the predicted values using the new data.

---

`predict_model.easy_neural_network`*Predict values for a neural network model.*

---

**Description**

This function wraps the procedure for predicting values from a neural network model and makes it accessible to the easymml core framework.

**Usage**

```
## S3 method for class 'easy_neural_network'  
predict_model(object, newx = NULL)
```

**Arguments**

<code>object</code>	A list of class <code>easy_neural_network</code> .
<code>newx</code>	A <code>data.frame</code> , the new data to use for predictions.

**Value**

A vector, the predicted values using the new data.

---

`predict_model.easy_random_forest`*Predict values for a random forest regression model.*

---

**Description**

This function wraps the procedure for predicting values from a random forest model and makes it accessible to the easymml core framework.

**Usage**

```
## S3 method for class 'easy_random_forest'  
predict_model(object, newx = NULL)
```

**Arguments**

<code>object</code>	A list of class <code>easy_random_forest</code> .
<code>newx</code>	A <code>data.frame</code> , the new data to use for predictions.

**Value**

A vector, the predicted values using the new data.

---

`predict_model.easy_support_vector_machine`*Predict values for a support vector machine regression model.*

---

**Description**

This function wraps the procedure for predicting values from a support vector machine model and makes it accessible to the easymml core framework.

**Usage**

```
## S3 method for class 'easy_support_vector_machine'  
predict_model(object, newx = NULL)
```

**Arguments**

<code>object</code>	A list of class <code>easy_support_vector_machine</code> .
<code>newx</code>	A <code>data.frame</code> , the new data to use for predictions.

**Value**

A vector, the predicted values using the new data.

---

preprocess\_identity     *Preprocess data by leaving it exactly the way it is.*

---

### Description

This function is the same as the identity function. Anything passed to it is returned untouched.

### Usage

```
preprocess_identity(.data, categorical_variables = NULL)
```

### Arguments

`.data`                    A data.frame; the data to be analyzed.  
`categorical_variables`     A logical vector; each value TRUE indicates that column in the data.frame is a categorical variable. Defaults to NULL.

### Value

A list, containing one or two data.frames.

### See Also

Other preprocess: [preprocess\\_scale](#)

---

preprocess\_scale             *Preprocess data by scaling it.*

---

### Description

This function takes either a data.frame or a list of data.frames. In the event of the first, this function takes the dataset and will scale each column that is not categorical such that that column has zero mean and unit variance. In the event of the second, this function takes the training dataset and will identify the parameters needed to scale the training dataset such that each column that is not categorical has zero mean and unit variance, and then will apply those parameters to each column in the testing dataset that is not categorical.

### Usage

```
preprocess_scale(.data, categorical_variables = NULL)
```

**Arguments**

`.data` A data.frame; the data to be analyzed.

`categorical_variables` A logical vector; each value TRUE indicates that column in the data.frame is a categorical variable. Defaults to NULL.

**Value**

A list, containing one or two data.frames.

**See Also**

Other preprocess: [preprocess\\_identity](#)

---

`process_coefficients` *Process coefficients.*

---

**Description**

Takes the coefficients returned by the `generate_coefficients` function and prepares the coefficients for plotting.

**Usage**

```
process_coefficients(coefficients, survival_rate_cutoff = 0.05)
```

**Arguments**

`coefficients` A data.frame, the replicated coefficients.

`survival_rate_cutoff` A numeric vector of length one; for [easy\\_glmnet](#), specifies the minimal threshold (as a percentage) a coefficient must appear out of `n_samples`. Defaults to 0.05.

**Value**

A data.frame; the replicated coefficients processed for easy plotting.

**See Also**

Other utils: [correlation\\_test](#), [process\\_variable\\_importances](#), [reduce\\_cores](#), [remove\\_variables](#)

---

process\_variable\_importances  
*Process variable importances.*

---

**Description**

Takes the variable importances returned by the generate\_variable\_importances function and prepares the variable importances for plotting.

**Usage**

```
process_variable_importances(variable_importances)
```

**Arguments**

variable\_importances  
A data.frame, the replicated coefficients.

**Value**

A data.frame; the replicated variable importances processed for easy plotting.

**See Also**

Other utils: [correlation\\_test](#), [process\\_coefficients](#), [reduce\\_cores](#), [remove\\_variables](#)

---

prostate *Prostate data.*

---

**Description**

Prostate data.

**Usage**

```
prostate
```

**Format**

Data frame with columns

**lcavol** log(cancer volume)

**lweight** log(prostate weight)

**age** age

**lbph** log(benign prostatic hyperplasia amount)

**svi** seminal vesicle invasion  
**lcp** log(capsular penetration)  
**gleason** Gleason score  
**pgg45** percentage Gleason scores 4 or 5  
**lpsa** log(prostate specific antigen)

### Source

Stamey, T.A., Kabalin, J.N., McNeal, J.E., Johnstone, I.M., Freiha, F., Redwine, E.A. and Yang, N. (1989) Prostate specific antigen in the diagnosis and treatment of adenocarcinoma of the prostate: II. radical prostatectomy treated patients, *Journal of Urology* 141(5), 1076-1083.

### See Also

Other data: [cocaine\\_dependence](#)

### Examples

```
data("prostate", package = "easym1")
```

---

reduce_cores	<i>Reduce number of cores.</i>
--------------	--------------------------------

---

### Description

This function takes the number of cores specified by the user and reduces it to the maximum number of cores supported by the computer.

### Usage

```
reduce_cores(n_core, cpu_count = NULL)
```

### Arguments

n_core	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to 1L.
cpu_count	An integer vector of length one; specifies the number of cores potentially available to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to 1L.

### Value

An integer vector of length one; specifies the number of cores to use for this analysis.

### See Also

Other utils: [correlation\\_test](#), [process\\_coefficients](#), [process\\_variable\\_importances](#), [remove\\_variables](#)



---

remove_variables	<i>Remove variables from a dataset.</i>
------------------	---

---

**Description**

This utility function removes variables from a data.frame.

**Usage**

```
remove_variables(.data = NULL, exclude_variables = NULL)
```

**Arguments**

.data                   A data.frame; the data to be analyzed.  
exclude\_variables       A character vector; the variables from the data set to exclude. Defaults to NULL.

**Value**

A data.frame; the data to be analyzed.

**See Also**

Other utils: [correlation\\_test](#), [process\\_coefficients](#), [process\\_variable\\_importances](#), [reduce\\_cores](#)

---

resample_fold_train_test_split	<i>Sample with respect to an identification vector</i>
--------------------------------	--

---

**Description**

This will sample the training and test sets so that case identifiers (e.g. subject ID's) are not shared across training and test sets.

**Usage**

```
resample_fold_train_test_split(X, y, train_size = 0.667, foldid = NULL,  
  random_state = NULL)
```

**Arguments**

<code>X</code>	A data.frame, the data to be resampled.
<code>y</code>	A numeric vector with two classes, 0 and 1.
<code>train_size</code>	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
<code>foldid</code>	A vector with length equal to <code>length(y)</code> which identifies cases belonging to the same fold.
<code>random_state</code>	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.

**Value**

A boolean vector of length `n_obs` where TRUE represents that observation should be in the train set.

**See Also**

Other resample: [resample\\_simple\\_train\\_test\\_split](#), [resample\\_stratified\\_class\\_train\\_test\\_split](#), [resample\\_stratified\\_simple\\_train\\_test\\_split](#)

---

`resample_simple_train_test_split`  
*Train test split.*

---

**Description**

This will split the data into train and test.

**Usage**

```
resample_simple_train_test_split(X, y, train_size = 0.667, foldid = NULL,
  random_state = NULL)
```

**Arguments**

<code>X</code>	A data.frame, the data to be resampled.
<code>y</code>	A numeric vector with two classes, 0 and 1.
<code>train_size</code>	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
<code>foldid</code>	Not currently supported in this function.
<code>random_state</code>	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.

**Value**

A boolean vector of length `n_obs` where `TRUE` represents that observation should be in the train set.

**See Also**

Other resample: [resample\\_fold\\_train\\_test\\_split](#), [resample\\_stratified\\_class\\_train\\_test\\_split](#), [resample\\_stratified\\_simple\\_train\\_test\\_split](#)

---

`resample_stratified_class_train_test_split`  
*Sample in equal proportion.*

---

**Description**

This will sample in equal proportion.

**Usage**

```
resample_stratified_class_train_test_split(X, y, train_size = 0.667,  
foldid = NULL, random_state = NULL)
```

**Arguments**

<code>X</code>	A <code>data.frame</code> , the data to be resampled.
<code>y</code>	A numeric vector with two classes, 0 and 1.
<code>train_size</code>	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
<code>foldid</code>	Not currently supported in this function.
<code>random_state</code>	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to <code>NULL</code> .

**Value**

A boolean vector of length `n_obs` where `TRUE` represents that observation should be in the train set.

**See Also**

Other resample: [resample\\_fold\\_train\\_test\\_split](#), [resample\\_simple\\_train\\_test\\_split](#), [resample\\_stratified\\_si](#)

---

```
resample_stratified_simple_train_test_split
    Sample in equal proportion.
```

---

**Description**

This will sample in equal proportion.

**Usage**

```
resample_stratified_simple_train_test_split(X, y, train_size = 0.667,
    foldid = NULL, random_state = NULL)
```

**Arguments**

X	A data.frame, the data to be resampled.
y	A numeric vector with two classes, 0 and 1.
train_size	A numeric vector of length one; specifies what proportion of the data should be used for the training data set. Defaults to 0.667.
foldid	A vector with length equal to length(y) which identifies cases belonging to the same fold.
random_state	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.

**Value**

A boolean vector of length n\_obs where TRUE represents that observation should be in the train set.

**See Also**

Other resample: [resample\\_fold\\_train\\_test\\_split](#), [resample\\_simple\\_train\\_test\\_split](#), [resample\\_stratified\\_cl](#)

---

```
set_categorical_variables
    Set categorical variables.
```

---

**Description**

This helper functions determines a logical boolean vector based on the column names and the designation for which ones are categorical variables.

**Usage**

```
set_categorical_variables(column_names, categorical_variables = NULL)
```

**Arguments**

`column_names` A character vector; the column names of the data for this analysis.  
`categorical_variables`  
 A character vector; the variables that are categorical. Defaults to NULL.

**Value**

NULL, or if `categorical_variables` is not NULL, then a logical vector of length `length(column_names)` where TRUE represents that column is a categorical variable.

**See Also**

Other setters: [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_column_names	<i>Set column names.</i>
------------------	--------------------------

---

**Description**

This functions helps decide what the updated column names of a data.frame should be within the easym1 framework based on the dependent variable, preprocessing function, exclusionary variables, and categorical variables.

**Usage**

```
set_column_names(column_names, dependent_variable, preprocess = NULL,
  exclude_variables = NULL, categorical_variables = NULL)
```

**Arguments**

`column_names` A character vector; the column names of the data for this analysis.  
`dependent_variable`  
 A character vector of length one; the dependent variable for this analysis.  
`preprocess` A function; the function for preprocessing the data. Defaults to NULL.  
`exclude_variables`  
 A character vector; the variables from the data set to exclude. Defaults to NULL.  
`categorical_variables`  
 A character vector; the variables that are categorical. Defaults to NULL.

**Value**

The updated columns, in the correct order for preprocessing.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_cores	<i>Set cores.</i>
-----------	-------------------

---

**Description**

Please note this affects global state and sets the number of cores by running `options(mc.cores = n_core)`.

**Usage**

```
set_cores(n_core)
```

**Arguments**

n_core	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSx and Unix/Linux systems. Defaults to 1L.
--------	--

**Value**

NULL.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_dependent_variable	<i>Set dependent variable.</i>
------------------------	--------------------------------

---

**Description**

This helper functions isolates the dependent variable in a data.frame.

**Usage**

```
set_dependent_variable(.data, dependent_variable)
```

**Arguments**

`.data` A data.frame; the data to be analyzed.  
`dependent_variable` A character vector of length one; the dependent variable for this analysis.

**Value**

A vector, the dependent variable of the analysis.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set\_independent\_variables

*Set independent variables.*

---

**Description**

This helper functions isolates the independent variables in a data.frame.

**Usage**

```
set_independent_variables(.data, dependent_variable)
```

**Arguments**

`.data` A data.frame; the data to be analyzed.  
`dependent_variable` A character vector of length one; the dependent variable for this analysis.

**Value**

A data.frame, the independent variables of the analysis.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_looper	<i>Set looper.</i>
------------	--------------------

---

### Description

This function decides which looper (a functional like `lapply`) to run. Please note this affects global state and sets the number of cores by ultimately running `options(mc.cores = n_core)`.

### Usage

```
set_looper(progress_bar = FALSE, n_core = 1)
```

### Arguments

<code>progress_bar</code>	A logical vector of length one; specifies whether to display a progress bar during calculations. Defaults to <code>TRUE</code> .
<code>n_core</code>	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to <code>1L</code> .

### Value

The looper to use depending on progress bar and whether to run in parallel or not.

### See Also

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_looper_	<i>Set looper.</i>
-------------	--------------------

---

### Description

This function decides which looper (a functional like `lapply`) to run. This function does not affect global state.

### Usage

```
set_looper_(progress_bar = FALSE, parallel = FALSE)
```



**Arguments**

- `progress_bar` A logical vector of length one; specifies whether to display a progress bar during calculations. Defaults to FALSE.
- `parallel` A logical vector of length one; specifies whether to run calculations in parallel. Defaults to FALSE.

**Value**

The looper to use depending on progress bar and whether to run in parallel or not.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_measure	<i>Set measure function.</i>
-------------	------------------------------

---

**Description**

Sets the function responsible for measuring the results.

**Usage**

```
set_measure(measure = NULL, algorithm, family)
```

**Arguments**

- `measure` A function; the function for measuring the results. Defaults to NULL.
- `algorithm` A character vector of length one; the algorithm to run on the data. Choices are one of `c("glmnet", "random_forest", "support_vector_machine")`.
- `family` A character vector of length one; the type of regression to run on the data. Choices are one of `c("gaussian", "binomial")`. Defaults to "gaussian".

**Value**

A function; the function for measuring the results.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper](#), [set\\_looper](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_parallel	<i>Set parallel.</i>
--------------	----------------------

---

**Description**

This helper function decides whether the analysis should be run in parallel based on the number of cores specified.

**Usage**

```
set_parallel(n_core)
```

**Arguments**

n_core	An integer vector of length one; specifies the number of cores to use for this analysis. Currently only works on Mac OSX and Unix/Linux systems. Defaults to 1L.
--------	--

**Value**

A logical vector of length one; whether analysis should be run in parallel or not.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_plot_model_performance	<i>Set plot model performance function.</i>
----------------------------	---

---

**Description**

Sets the function responsible for plotting the measures of model performance generated from the predictions generated from a fitted model.

**Usage**

```
set_plot_model_performance(measure)
```

**Arguments**

measure	A function; the function for measuring the results. Defaults to NULL.
---------	---

**Value**

TA function; the function for plotting the measures of model performance generated from the predictions generated from a fitted model.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set\_plot\_predictions    *Set plot predictions function.*

---

**Description**

Sets the function responsible for plotting the predictions generated from a fitted model.

**Usage**

```
set_plot_predictions(algorithm, family)
```

**Arguments**

algorithm	A character vector of length one; the algorithm to run on the data. Choices are one of c("glmnet", "random_forest", "support_vector_machine").
family	A character vector of length one; the type of regression to run on the data. Choices are one of c("gaussian", "binomial"). Defaults to "gaussian".

**Value**

A function; the function for plotting the predictions generated from a fitted model.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_preprocess](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_preprocess	<i>Set preprocess function.</i>
----------------	---------------------------------

---

**Description**

Sets the function responsible for preprocessing the data.

**Usage**

```
set_preprocess(preprocess = NULL, algorithm)
```

**Arguments**

preprocess	A function; the function for preprocessing the data. Defaults to NULL.
algorithm	A character vector of length one; the algorithm to run on the data. Choices are one of <code>c("glmnet", "random_forest", "support_vector_machine")</code> .

**Value**

A function; the function for preprocessing the data.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_random\\_state](#), [set\\_resample](#)

---

set_random_state	<i>Set random state.</i>
------------------	--------------------------

---

**Description**

Sets the random state to a specific seed. Please note this function affects global state.

**Usage**

```
set_random_state(random_state = NULL)
```

**Arguments**

random_state	An integer vector of length one; specifies the seed to be used for the analysis. Defaults to NULL.
--------------	--

**Value**

NULL.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_resample](#)

---

set_resample	<i>Set resample function.</i>
--------------	-------------------------------

---

**Description**

Sets the function responsible for resampling the data.

**Usage**

```
set_resample(resample = NULL, family = NULL)
```

**Arguments**

resample	A function; the function for resampling the data. Defaults to NULL.
family	A character vector of length one; the type of regression to run on the data. Choices are one of c("gaussian", "binomial"). Defaults to "gaussian".

**Value**

A function; the function for resampling the data.

**See Also**

Other setters: [set\\_categorical\\_variables](#), [set\\_column\\_names](#), [set\\_cores](#), [set\\_dependent\\_variable](#), [set\\_independent\\_variables](#), [set\\_looper\\_](#), [set\\_looper](#), [set\\_measure](#), [set\\_parallel](#), [set\\_plot\\_model\\_performance](#), [set\\_plot\\_predictions](#), [set\\_preprocess](#), [set\\_random\\_state](#)

# Index

## \*Topic **datasets**

- cocaine\_dependence, 3
- prostate, 47
  
- cocaine\_dependence, 3, 48
- correlation\_test, 4, 46–49
  
- easy\_analysis, 5, 10, 12, 14, 16, 18, 20, 22
- easy\_avNNet, 8, 8, 12, 14, 16, 18, 20, 22
- easy\_deep\_neural\_network, 8, 10, 10, 14, 16, 18, 20, 22
- easy\_glinternet, 8, 10, 12, 12, 16, 18, 20, 22
- easy\_glmnet, 6–15, 15, 18, 20, 22, 34, 40, 46
- easy\_neural\_network, 8, 10, 12, 14, 16, 17, 20, 22
- easy\_random\_forest, 8, 10, 12, 14, 16, 18, 19, 22, 40
- easy\_support\_vector\_machine, 8, 10, 12, 14, 16, 18, 20, 21
- easym1, 5
- easym1-package (easym1), 5
- extract\_coefficients, 23
- extract\_coefficients.easy\_glmnet, 24
- extract\_variable\_importances, 24
- extract\_variable\_importances.easy\_random\_forest, 25
  
- fit\_model, 25
- fit\_model.easy\_avNNet, 26
- fit\_model.easy\_deep\_neural\_network, 26
- fit\_model.easy\_glinternet, 27
- fit\_model.easy\_glmnet, 27
- fit\_model.easy\_neural\_network, 28
- fit\_model.easy\_random\_forest, 28
- fit\_model.easy\_support\_vector\_machine, 29
  
- generate\_coefficients, 29, 30, 31, 34
- generate\_model\_performance, 29, 30, 30, 31
  
- generate\_predictions, 29, 30, 30, 31
- generate\_variable\_importances, 29, 30, 31, 40
  
- measure\_auc\_score, 31, 32, 33
- measure\_correlation\_score, 31, 32, 33
- measure\_mse\_score, 31, 32, 32, 33
- measure\_r2\_score, 31–33, 33
  
- plot\_coefficients\_processed, 34, 35–40
- plot\_model\_performance\_binomial\_auc\_score, 34, 34, 35–40
- plot\_model\_performance\_gaussian\_correlation\_score, 34, 35, 35, 36–40
- plot\_model\_performance\_gaussian\_mse\_score, 34, 35, 36, 37–40
- plot\_model\_performance\_gaussian\_r2\_score, 34–36, 36, 37–40
- plot\_model\_performance\_histogram, 34–37, 37, 38–40
- plot\_predictions\_binomial, 34–37, 38, 39, 40
- plot\_predictions\_gaussian, 34–38, 38, 39, 40
- plot\_roc\_curve, 34–39, 39, 40
- plot\_variable\_importances\_processed, 34–39, 40
- predict\_model, 40
- predict\_model.easy\_avNNet, 41
- predict\_model.easy\_deep\_neural\_network, 42
- predict\_model.easy\_glinternet, 42
- predict\_model.easy\_glmnet, 43
- predict\_model.easy\_neural\_network, 43
- predict\_model.easy\_random\_forest, 44
- predict\_model.easy\_support\_vector\_machine, 44
  
- preprocess\_identity, 45, 46
- preprocess\_scale, 45, 45
- process\_coefficients, 5, 34, 46, 47–49

`process_variable_importances`, [5](#), [40](#), [46](#),  
[47](#), [48](#), [49](#)  
`prostate`, [4](#), [47](#)

`reduce_cores`, [5](#), [46](#), [47](#), [48](#), [49](#)  
`remove_variables`, [5](#), [46–48](#), [49](#)  
`resample_fold_train_test_split`, [49](#), [51](#),  
[52](#)  
`resample_simple_train_test_split`, [50](#),  
[50](#), [51](#), [52](#)  
`resample_stratified_class_train_test_split`,  
[50](#), [51](#), [51](#), [52](#)  
`resample_stratified_simple_train_test_split`,  
[50](#), [51](#), [52](#)

`set_categorical_variables`, [52](#), [54–61](#)  
`set_column_names`, [53](#), [53](#), [54–61](#)  
`set_cores`, [53](#), [54](#), [54](#), [55–61](#)  
`set_dependent_variable`, [53](#), [54](#), [54](#), [55–61](#)  
`set_independent_variables`, [53–55](#), [55](#),  
[56–61](#)  
`set_looper`, [53–55](#), [56](#), [57–61](#)  
`set_looper_`, [53–56](#), [56](#), [57–61](#)  
`set_measure`, [53–57](#), [57](#), [58–61](#)  
`set_parallel`, [53–57](#), [58](#), [59–61](#)  
`set_plot_model_performance`, [53–58](#), [58](#),  
[59–61](#)  
`set_plot_predictions`, [53–59](#), [59](#), [60](#), [61](#)  
`set_preprocess`, [53–59](#), [60](#), [61](#)  
`set_random_state`, [53–60](#), [60](#), [61](#)  
`set_resample`, [53–61](#), [61](#)