

# Package ‘MSML’

November 20, 2023

**Title** Model selection based on Machine Learning (ML)  
**Version** 1.0.0.0  
**Description** Model selection based on Machine Learning (ML) following modified Recursive feature elimination (RFE).  
**License** ``use_mit_license()`, `use_gpl3_license()`` or friends to pick a license  
**Encoding** UTF-8  
**Roxygen** `list(markdown = TRUE)`  
**RoxygenNote** 7.1.2  
**Depends** R ( $\geq 2.10$ )  
**LazyData** true

## R topics documented:

data_test . . . . .	1
data_train . . . . .	2
data_valid . . . . .	3
model_combination . . . . .	3
model_evaluation . . . . .	4
<b>Index</b>	<b>5</b>

---

data_test	<i>7 sets of PRSs for test dataset and target phenotype</i>
-----------	---

---

## Description

A dataset containing 7 sets of PRSs for test dataset and target phenotype

## Usage

data\_test

**Format**

A data frame for test dataset:

**V1** PRS1, for bin1

**V2** PRS2, for bin1

**V3** PRS3, for bin1

**V4** PRS4, for bin1

**V5** PRS5, for bin1

**V6** PRS6, for bin1

**V7** PRS7, for bin1

**target** Target Phenotype, value

---

data\_train

*7 sets of PRSs for training data set and target phenotype*

---

**Description**

A dataset containing 7 sets of PRSs for training data set and target phenotype

**Usage**

data\_train

**Format**

A data frame for training dataset:

**V1** PRS1, for bin1

**V2** PRS2, for bin1

**V3** PRS3, for bin1

**V4** PRS4, for bin1

**V5** PRS5, for bin1

**V6** PRS6, for bin1

**V7** PRS7, for bin1

**target** Target Phenotype, value

data\_valid

7 sets of PRSs for validation dataset and target phenotype

**Description**

A dataset containing 7 sets of PRSs for validation dataset and target phenotype

**Usage**

```
data_valid
```

**Format**

A data frame for validation dataset:

**V1** PRS1, for bin1

**V2** PRS2, for bin1

**V3** PRS3, for bin1

**V4** PRS4, for bin1

**V5** PRS5, for bin1

**V6** PRS6, for bin1

**V7** PRS7, for bin1

**target** Target Phenotype, value

---

`model_combination` *model\_combination function This function will generate PRS based on all possible combinations of model. The total number of models required to explore the combinations of these 'n' features can be calculated by summing the combinations for each possible number of features, ranging from 1 to 'n' ( $\sum_{i=1}^n C(n,i)$ ). where  $C(n,k)$  represents the binomial coefficient or "n choose k," with n denoting the total number of features and k indicating the number of features to include in each model.*

---

**Description**

`model_combination` function This function will generate PRS based on all possible combinations of model. The total number of models required to explore the combinations of these 'n' features can be calculated by summing the combinations for each possible number of features, ranging from 1 to 'n' ( $\sum_{i=1}^n C(n,i)$ ). where  $C(n,k)$  represents the binomial coefficient or "n choose k," with n denoting the total number of features and k indicating the number of features to include in each model.

**Usage**

```
model_combination(data_train, data_valid, data_test, mv)
```

**Arguments**

data_train	This is the matrix for training dataset
data_valid	This is the matrix for validation dataset
data_test	This is the matrix for test dataset
mv	The total number of columns in data_train/data_valid/data_test

**Value**

This function will generate all possible model outcomes for validation and test dataset

**Examples**

```
data_train <- data_train
data_train <- data_train
data_test <- data_test
mv=8
model_combination(data_train,data_valid,data_test,mv)
```

---

model\_evaluation     *model\_evaluation function*

---

**Description**

This function will identify best model in validation and test dataset.

**Usage**

```
model_evaluation(dat, mv, tn, prev)
```

**Arguments**

dat	This is the matrix for all the combinations of model
mv	The total number of columns in data_train/data_valid/data_test
tn	The total no of best models to be identified
prev	The prevalence of disease in the data

**Value**

This function will generate all possible model outcomes for validation and test dataset

**Examples**

```
data_train <- data_train
mv=8
tn=15
prev=0.047
model_evaluation(dat,mv,tn,prev)
```

# Index

- \*Topic **All**
  - model\_combination, 3
- \*Topic **Identify**
  - model\_evaluation, 4
- \*Topic **best**
  - model\_evaluation, 4
- \*Topic **combinations**
  - model\_combination, 3
- \*Topic **datasets**
  - data\_test, 1
  - data\_train, 2
  - data\_valid, 3
- \*Topic **models**
  - model\_evaluation, 4
- \*Topic **model**
  - model\_combination, 3
- \*Topic **possible**
  - model\_combination, 3

$C(n, i)$ , 3

data\_test, 1  
data\_train, 2  
data\_valid, 3

model\_combination, 3  
model\_evaluation, 4