# Package 'MSML'

November 21, 2023
Title Model selection based on Machine Learning (ML)
<b>Version</b> 1.0.0.0
<b>Description</b> The MSML package is designed to determine the optimal model(s) by leveraging all available features.
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LazyData true
data_test       1         data_train       2         data_valid       3         model_configuration       3         model_evaluation       4
Index
data_test 7 sets of PRSs for test dataset and target phenotype
<b>Description</b> A dataset containing 7 sets of PRSs for test dataset and target phenotype
11 dataset containing 7 sets of 1 feet for test dataset and target phenotype

## Usage

data\_test

2 data\_train

## **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 3

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\_configuration

model\_configuration 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' (?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\_configuration 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' (?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_configuration(data_train, data_valid, data_test, mv)
```

4 model\_evaluation

#### **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_valid <- data_valid
data_test <- data_test
mv=8
model_configuration(data_train,data_valid,data_test,mv)</pre>
```

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

my The total number of columns in data\_train/data\_valid/data\_test

tn The total no of best models to be identified prev The prevalance of disease in the data

#### Value

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

## **Examples**

```
dat <- read.table("models_test_all")
mv=8
tn=15
prev=0.047
model_evaluation(dat,mv,tn,prev)</pre>
```

## **Index**

```
*Topic All
    model\_configuration, 3
*Topic Identify
   model_evaluation, 4
*Topic best
    model\_evaluation, 4
*Topic combinations
   model_configuration, 3
*Topic datasets
    data\_test, 1
    data\_train, 2
    data_valid, 3
*Topic models
   model\_evaluation, 4
*Topic model
    model\_configuration, 3
*Topic possible
   model_configuration, 3
C(n,i), 3
{\tt data\_test}, 1
data_train, 2
data\_valid, 3
{\tt model\_configuration}, 3
model_evaluation, 4
```