

# Package ‘EMSNM’

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**Type** Package

**Title** EM Algorithm for Sigmoid Normal Model

**Version** 1.0

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**Description** It provides a method based on EM algorithm to estimate the parameter of a mixture model, Sigmoid-Normal Model, where the samples come from several normal distributions (also call them subgroups) whose mean is determined by co-variable Z and coefficient alpha while the variance are homogeneous. Meanwhile, the subgroup each item belongs to is determined by co-variables X and coefficient eta through Sigmoid link function which is the extension of Logistic Link function. It uses bootstrap to estimate the standard error of parameters. When sample is indeed separable, removing estimation with abnormal sigma, the estimation of alpha is quite well. I used this method to explore the subgroup structure of HIV patients and it can be used in other domains where exists subgroup structure.

**License** GPL (>= 2)

**NeedsCompilation** no

**Repository** CRAN

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| EMSNM-package | <i>EM Algorithm for Sigmoid Normal Model</i> |
|---------------|--|

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## Description

It provides a method based on EM algorithm to estimate the parameter of a mixture model, Sigmoid-Normal Model, where the samples come from several normal distributions (also call them subgroups) whose mean is determined by co-variable Z and coefficient alpha while the variance are homogeneous. Meanwhile, the subgroup each item belongs to is determined by co-variables X and coefficient eta through Sigmoid link function which is the extension of Logistic Link function. It uses bootstrap to estimate the standard error of parameters. When sample is indeed separable, removing estimation with abnormal sigma, the estimation of alpha is quite well. I used this method to explore the subgroup structure of HIV patients and it can be used in other domains where exists subgroup structure.

## Details

The DESCRIPTION file:

```

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License:      GPL(>=2)

```

Index of help topics:

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| EM_result_sort  | Sort Parameter                        |
| EMalgorithm     | Parameter Estimation                  |
| EMbootstrap     | Bootstrap Method                      |
| EMsimulation    | Simulation For Estimation             |
| Wgenerate       | Subgroup Determination                |

|               |                                  |
|---------------|----------------------------------|
| Wgenerate     | Sigmoid Logistic Data Generation |
| fnorm         | Density Value                    |
| softmax       | Softmax Value                    |
| standard      | Data Standardlization            |
| update_eta    | Udata Eta                        |
| update_gamma  | Udata Alpha and Sigma            |
| weight_matrix | Weighted Inner Product           |

The EMalgorithm is used to estimate the parameters, EMbootstrap is used to estimate the parameters with bootstrap method. In EMSimulation we can simulate the situation with given parameters, so parameter estimation can be verified.

### Author(s)

Linsui Deng <denglinsui@gmail.com>

Maintainer: Linsui Deng <denglinsui@gmail.com>

### Examples

```
#parameter initialization
etasize <- 2
classsize <- 2
alphasize <- 3
samplesize <- 100
experiments <- 30

etatest <- matrix(c(1,1,
                   0,0),etasize,classsize)

alphatest <- matrix(c(1,0,2,
                     4,3,5),alphasize,classsize)

sigmatest <- 0.5

#test of EMSimulation
EMsimulation_result <- EMSimulation(eta=etatest,alpha=alphatest,sigma=sigmatest,
                                   samplesize=samplesize,experiments=experiments,
                                   compact_flag=TRUE,C0=5,C1=0.5,C2=5)
index <- which(EMsimulation_result$sigma<0.8)
EMsimulation_result_sort <- EM_result_sort(EMsimulation_result$alpha[index,,],
                                           EMsimulation_result$eta[index,,])
EM_parameter <- EM_parameter_sd(EMsimulation_result_sort$alpha,
                               EMsimulation_result_sort$eta,
                               EMsimulation_result$sigma[index])

#test of EMbootstrap
samplesize <- 1000
X <- matrix(c(matrix(1,samplesize),
              rnorm(samplesize*(etasize-1))+1),samplesize,etasize)
Z <- matrix(c(matrix(1,samplesize),rbinom(prob=1/2,size=1,n=samplesize),
              rnorm(samplesize*(alphasize-2))+1),samplesize,alphasize)
```

```

Wtest <- Wgenerate(alpha=alphatest,eta=etatest,sigma=sigmat, X=X,Z=Z)

boots_samplesize <- 100
boots_experiments <- 30
samplesize <- dim(Wtest$X)[1]
EMbootstrap_theta <- EMbootstrap(Wtest$X,Wtest$Y,Wtest$Z,samplesize,
                                boots_samplesize,boots_experiments,
                                classsize=2,compact_flag=TRUE,C0=5,C1=0.2,C2=5)
index <- which(EMbootstrap_theta$sigma<0.8)
EMsimulation_result_sort <- EM_result_sort(EMbootstrap_theta$alpha[index,,],
                                           EMbootstrap_theta$eta[index,,])
EM_parameter <- EM_parameter_sd(EMsimulation_result_sort$alpha,
                               EMsimulation_result_sort$eta,
                               EMbootstrap_theta$sigma[index])

```

---

Ccompute

*Ccompute*


---

### Description

Compute the probability of Y in given parameters alphas, sigmas, etas and variables X, Z by the Bayesian Formula under the assumption of Sigmoid-Normal Model.

### Usage

```
Ccompute(alphas, sigmas, etas, X, Z, Y)
```

### Arguments

|        |   |
|--------|---|
| alphas | the coefficients of the mean of each subgroup |
| sigmas | the variance of Y                             |
| etas   | the coefficients determining subgroup         |
| X      | the covariables of the mean of each subgroup  |
| Z      | the covariables determining subgroup          |
| Y      | the respond variable                          |

### Value

the probability of Y in given parameters alphas, sigmas, etas and variables X, Z under the assumption of Sigmoid-Normal Model.

### Author(s)

Linsui Deng

**Examples**

```

#some variables
samplesize <- 1000
classsize <- 6
etasize <- 3
alphasize <- 2

Xtest <- data.frame(matrix(rnorm(samplesize*etasize),samplesize,etasize))
Ztest <- matrix(rnorm(samplesize*alphasize),samplesize,alphasize)

etatest <- matrix(seq(1.15,1,length=etasize*classsize),etasize,classsize)
alphatest <- matrix(seq(1.15,1,length=alphasize*classsize),alphasize,classsize)

Wtest <- Wgenerate(alpha=alphatest,eta=etatest,X=Xtest,Z=Ztest)

#test of Ccompute
sigmatest <- 1
Ctest <-
  Ccompute(alphat=alphatest,sigmat=sigmatest,
           etat=etatest,X=Wtest$X,Z=Wtest$Z,Y=Wtest$Y)

```

EMalgorithm

*Parameter Estimation***Description**

Estimate paramters value by EM algorithm under the assumption of Sigmoid-Normal Model.

**Usage**

```

EMalgorithm(X, Y, Z, etat, alphat, sigmat, classsize = 2, learning_rate = 0.1,
           regular_parameter_eta = 0.001, max_iteration = 10000,
           max_iteration_eta = 10000, compact_flag = FALSE, C0 = 5, C1 = 2, C2 = 9)

```

**Arguments**

|                       |  |
|-----------------------|--|
| X                     | the covariables of the mean of each subgroup                 |
| Y                     | the respond variable   |
| Z                     | the covariables determining subgroup                         |
| etat                  | the coeffients determining subgroup                          |
| alphat                | the coeffients of the mean of each subgroup                  |
| sigmat                | the variance of Y  |
| classsize             | the number of subgroup types in your model assumption        |
| learning_rate         | learning rate of updating eta                                |
| regular_parameter_eta | regular value of updating eta by gradiant descending method. |

`max_iteration` maximum steps of iteration to avoid unlimited looping.  
`max_iteration_eta` maximal steps of eta iteration to avoid unlimited looping.  
`compact_flag` if the value of eta is limited in a compact set, set it TRUE  
`C0` the maximum of intercept of eta.  
`C1` the minimum of the norm of slope of eta  
`C2` the maximum of the norm of slope of eta

**Value**

`alpha` alpha estimation  
`eta` eta estimation  
`sigma` sigma estimation

**Author(s)**

Linsui Deng

**Examples**

```

#data generation
samplesize <- 1000
classsize <- 2
etasize <- 3
alphasize <- 3

set.seed(1)
Xtest <- data.frame(matrix(rnorm(samplesize*etasize), samplesize, etasize))
etatest <- matrix(c(1,2,-1,
                   0,0,0), etasize, classsize)

Ztest <- matrix(rnorm(samplesize*alphasize), samplesize, alphasize)
alphatest <- matrix(c(1,0,2,
                    5,0,7), alphasize, classsize)

sigmatest <- 5

Wtest <- Wgenerate(alpha=alphatest, eta=etatest, X=Xtest, Z=Ztest, sigma=sigmatest)

eta_initial <- matrix(c(rnorm(3),0,0,0), etasize, classsize)
alpha_initial<- matrix(rnorm(alphasize*classsize)*3, alphasize, classsize)
sigma_initial <- 1

EMtheta <- EMalgorithm(X=Wtest$X, Z=Wtest$Z, Y=Wtest$Y, classsize=2,
                      etat=eta_initial, alphas=alpha_initial, sigma=sigma_initial,
                      learning_rate=0.01, regular_parameter_eta=0.001,
                      max_iteration=1000, max_iteration_eta=10000,
                      compact_flag = TRUE, C0 = 5, C1 = 2, C2 = 9)
  
```

EMbootstrap

*Bootstrap Method***Description**

Estimate the value of parameters several times by bootstrap method where the parameters are estimated by EM algorithm. In this way, we can observe the distribution of parameter.

**Usage**

```
EMbootstrap(X, Y, Z, samplesize, boots_samplesize, boots_experiments, classsize = 2,
            learning_rate = 0.1, regular_parameter_eta = 0.001, max_iteration = 10000,
            max_iteration_eta = 10000, compact_flag = FALSE, C0 = 5, C1 = 2, C2 = 9)
```

**Arguments**

|                       |  |
|-----------------------|--|
| X                     | the co-variables of the mean of each subgroup                |
| Y                     | the respond variable   |
| Z                     | the co-variables determining subgroup                        |
| samplesize            | the size of this sample.                                     |
| boots_samplesize      | the size of chosen sample in one step of bootstrap           |
| boots_experiments     | the steps of bootstrap                                       |
| classsize             | the number of subgroup types in your model assumption        |
| learning_rate         | learning rate of updating eta                                |
| regular_parameter_eta | regular value of updating eta by gradient descending method. |
| max_iteration         | maximum steps of iteration to avoid unlimited looping.       |
| max_iteration_eta     | maximal steps of eta iteration to avoid unlimited looping.   |
| compact_flag          | if the value of eta is limited in a compact set, set it TRUE |
| C0                    | the maximum of intercept of eta.                             |
| C1                    | the minimum of the norm of slope of eta                      |
| C2                    | the maximum of the norm of slope of eta                      |

**Details**

Actually, the method can be extended to other parameter estimation where the standard error of parameter can't be calculated in a simple way.

**Value**

alpha            alpha estimated by bootstrap method.  
eta                eta estimated by bootstrap method.  
sigma             sigma estimated by bootstrap method.

**Author(s)**

Linsui Deng

**Examples**

```
#parameter initialization
etasize <- 2
classsize <- 2
alphasize <- 3
samplesize <- 1000

etatest <- matrix(c(1,1,
                   0,0),etasize,classsize)

alphatest <- matrix(c(1,0,2,
                    4,3,5),alphasize,classsize)

sigmatest <- 0.5

#test of EMbootstrap
X <- matrix(c(matrix(1,samplesize),
              rnorm(samplesize*(etasize-1))+1),samplesize,etasize)
Z <- matrix(c(matrix(1,samplesize),rbinom(prob=1/2,size=1,n=samplesize),
              rnorm(samplesize*(alphasize-2))+1),samplesize,alphasize)

Wtest <- Wgenerate(alpha=alphatest,eta=etatest,sigma=sigmatest,X=X,Z=Z)

boots_samplesize <- 100
boots_experiments <- 30
samplesize <- dim(Wtest$X)[1]
EMbootstrap_theta <- EMbootstrap(Wtest$X,Wtest$Y,Wtest$Z,samplesize,
                                boots_samplesize,boots_experiments,
                                classsize=2,compact_flag=TRUE,C0=5,C1=0.2,C2=5)
index <- which(EMbootstrap_theta$sigma<0.8)
EMsimulation_result_sort <- EM_result_sort(EMbootstrap_theta$alpha[index,,],
                                           EMbootstrap_theta$eta[index,,])
EM_parameter <- EM_parameter_sd(EMsimulation_result_sort$alpha,
                                EMsimulation_result_sort$eta,
                                EMbootstrap_theta$sigma[index])
```



**Description**

It simulates the experiments with given alpha, eta and sigma to verify the EMalgorithm

**Usage**

```
EMsimulation(eta, alpha, sigma, samplesize, experiments,
             compact_flag = FALSE, C0 = 5, C1 = 2, C2 = 9)
```

**Arguments**

|              |  |
|--------------|--|
| eta          | the true value of eta  |
| alpha        | the true value of alpha                                      |
| sigma        | the true value of sigma                                      |
| samplesize   | the size of sample   |
| experiments  | the times of experiments                                     |
| compact_flag | if the value of eta is limited in a compact set, set it TRUE |
| C0           | the maximum of intercept of eta.                             |
| C1           | the minimum of the norm of slope of eta                      |
| C2           | the maximum of the norm of slope of eta                      |

**Value**

|       |                                |
|-------|--------------------------------|
| alpha | alpha estimated in simulation. |
| eta   | eta estimated in simulation.   |
| sigma | sigma estimated in simulation. |

**Author(s)**

Linsui Deng

**Examples**

```
#parameter initialization
etasize <- 2
classsize <- 2
alphasize <- 3
samplesize <- 100
experiments <- 30

etatest <- matrix(c(1,1,
                   0,0),etasize,classsize)

alphatest <- matrix(c(1,0,2,
                     4,3,5),alphasize,classsize)

sigmatest <- 0.5

#test of EMsimulation
```

```
EMsimulation_result <- EMsimulation(eta=etatest,alpha=alphatest,sigma=sigmatest,
                                   samplesize=samplesize,experiments=experiments,
                                   compact_flag=TRUE,C0=5,C1=0.5,C2=5)
```

---

EM\_parameter\_sd      *Bootstrap Parameter Inference*

---

### Description

Estimating the parameters and their stand error through the sorted parameters estimated by bootstrap method.

### Usage

```
EM_parameter_sd(EMsimulation_sort_alpha, EMsimulation_sort_eta, EMsimulation_sort_sigma)
```

### Arguments

EMsimulation\_sort\_alpha  
sorted alpha estimated by bootstrap method.

EMsimulation\_sort\_eta  
sorted eta estimated by bootstrap method.

EMsimulation\_sort\_sigma  
sorted sigma estimated by bootstrap method.

### Value

|          |   |
|----------|---|
| sigma    | the estimation of sigma                   |
| sigma_sd | the estimation of standard error of sigma |
| alpha    | the estimation of alpha                   |
| alpha_sd | the estimation of standard error of alpha |
| eta      | the estimation of eta                     |
| eta_sd   | the estimation of standard error of eta   |

### Author(s)

Linsui Deng

### Examples

```
#parameter initialization
etasize <- 2
classsize <- 2
alphasize <- 3
samplesize <- 100
experiments <- 30
```

```

etatest <- matrix(c(1,1,
                   0,0),etasize,classsize)

alphatest <- matrix(c(1,0,2,
                    4,3,5),alphasize,classsize)
sigmatest <- 0.5

EMsimulation_result <- EMsimulation(eta=etatest,alpha=alphatest,sigma=sigmatest,
                                   samplesize=samplesize,experiments=experiments,
                                   compact_flag=TRUE,C0=5,C1=0.5,C2=5)
index <- which(EMsimulation_result$sigma<0.8)
EMsimulation_result_sort <- EM_result_sort(EMsimulation_result$alpha[index,,],
                                           EMsimulation_result$eta[index,,])

#test of EM_parameter_sd
EM_parameter <- EM_parameter_sd(EMsimulation_result_sort$alpha,
                               EMsimulation_result_sort$eta,
                               EMsimulation_result$sigma[index])

```

---

|                |                       |
|----------------|-----------------------|
| EM_result_sort | <i>Sort Parameter</i> |
|----------------|-----------------------|

---

### Description

Since the number of subgroup types is always beyond 1, the order of subgroup may be different, we can sort them in this function.

### Usage

```
EM_result_sort(EMsimulation_result_alpha, EMsimulation_result_eta)
```

### Arguments

```
EMsimulation_result_alpha
    alpha estimated by bootstrap method.
EMsimulation_result_eta
    eta estimated by bootstrap method.
```

### Value

```
alpha    sorted alpha estimated by bootstrap method.
eta      sorted eta estimated by bootstrap method.
```

### Author(s)

Linsui Deng

**Examples**

```

#parameter initialization
etasize <- 2
classsize <- 2
alphasize <- 3
samplesize <- 100
experiments <- 30

etatest <- matrix(c(1,1,
                   0,0),etasize,classsize)

alphatest <- matrix(c(1,0,2,
                     4,3,5),alphasize,classsize)

sigmatest <- 0.5

#test of EMsimulation
EMsimulation_result <- EMsimulation(eta=etatest,alpha=alphatest,sigma=sigmatest,
                                   samplesize=samplesize,experiments=experiments,
                                   compact_flag=TRUE,C0=5,C1=0.5,C2=5)

#test of EM_result_sort
EMsimulation_result_sort <- EM_result_sort(EMsimulation_result$alpha,
                                           EMsimulation_result$eta)

```

---

 fnorm

*Density Value*


---

**Description**

Calculate the density value of respond value Y under each mean and homogeneous variance.

**Usage**

```
fnorm(Y, mu, sigma)
```

**Arguments**

|       |                                 |
|-------|---------------------------------|
| Y     | the respond variable            |
| mu    | different mean of each subgroup |
| sigma | standard error                  |

**Value**

the density value of Y under different mu and common sigma.

**Author(s)**

Linsui Deng

**Examples**

```
fnormtest <- fnorm(matrix(1:6,3,2),matrix(seq(1,3,length=6),3,2),1)
```

---

Ggenerate

*Subgroup Determination*


---

**Description**

In data generation, determining the subgroup of each item belonging to through random number and Sigmoid Link function.

**Usage**

```
Ggenerate(eta, X, seed = 0)
```

**Arguments**

|      |                                       |
|------|---------------------------------------|
| eta  | the coefficients determining subgroup |
| X    | the covariables determining subgroup  |
| seed | random seed                           |

**Value**

the classes items belonging to, it's a vector. If X1 belongs to class 3, then the 1st row 3rd column is 1 and the rest of 1st row are 0.

**Author(s)**

Linsui Deng

**Examples**

```
#some variables
samplesize <- 1000
classsize <- 6
etasize <- 3
alphasize <- 2

#test of Ggenerate
Xtest <- data.frame(matrix(rnorm(samplesize*etasize),samplesize,etasize))
etatest <- matrix(seq(1.15,1,length=etasize*classsize),etasize,classsize)

Gtest1 <- Ggenerate(etatest,Xtest)
Gtest2 <- Ggenerate(etatest,Xtest,1)
```

softmax

*Softmax Value*

---

**Description**

Calculate the Softmax Value of each subgroup to represent the probability of items belonging to specific class.

**Usage**

```
softmax(eta, X)
```

**Arguments**

|     |                                       |
|-----|---------------------------------------|
| eta | the coefficients determining subgroup |
| X   | the covariables determining subgroup  |

**Value**

Softmax Value of each subgroup

**Author(s)**

Linsui Deng

**Examples**

```
#some variables
samplesize <- 1000
classsize <- 6
etasize <- 3
alphasize <- 2

#test of softmax
Xtest <- data.frame(matrix(rnorm(samplesize*etasize), samplesize, etasize))
etatest <- matrix(seq(1.15, 1, length=etasize*classsize), etasize, classsize)
softmax_value <- softmax(etatest, Xtest)
```

---

|          |                              |
|----------|------------------------------|
| standard | <i>Data Standardlization</i> |
|----------|------------------------------|

---

**Description**

Standardize the data to elimilate the effect of scale

**Usage**

```
standard(X)
```

**Arguments**

X                    the original data

**Value**

standarlized data

**Author(s)**

Linsui Deng

**Examples**

```
#data generata
Y <- rnorm(100)*2+5
Y_sta <- standard(Y)
```

---

|            |                   |
|------------|-------------------|
| update_eta | <i>Updata Eta</i> |
|------------|-------------------|

---

**Description**

Updata eta in step t+1 with given data and coefficients estimated in step t.

**Usage**

```
update_eta(fun, alphas, sigmas, etas, X, Y, Z, learning_rate_eta = 0.001,
           regular_parameter_eta = 0.001, max_iteration_eta = 10000)
```

**Arguments**

|                       |   |
|-----------------------|---|
| fun                   | the function updata eta   |
| alphat                | the estimated coefficients of the mean of each subgroup in step t |
| sigmat                | the estimated standard error of Y in step t                       |
| etat                  | the estimated coefficients determining subgroup in step t         |
| X                     | the covariables of the mean of each subgroup                      |
| Z                     | the covariables determining subgroup                              |
| Y                     | the respond variable  |
| learning_rate_eta     | learning rate of updating eta                                     |
| regular_parameter_eta | regular value of updating eta by gradiant descending method.      |
| max_iteration_eta     | maximal steps of eta interation to avoid unlimited looping.       |

**Value**

|       |                            |
|-------|----------------------------|
| alpha | alpha estimated in step t. |
| eta   | eta estimated in step t+1. |
| sigma | sigma estimated in step t. |

**Author(s)**

Linsui Deng

**Examples**

```
#some variables
samplesize <- 1000
classsize <- 6
etasize <- 3
alphasize <- 2

Xtest <- data.frame(matrix(rnorm(samplesize*etasize),samplesize,etasize))
Ztest <- matrix(rnorm(samplesize*alphasize),samplesize,alphasize)

etatest <- matrix(seq(1.15,1,length=etasize*classsize),etasize,classsize)
alphatest <- matrix(seq(1.15,1,length=alphasize*classsize),alphasize,classsize)
sigmatest <- 0.1

Wtest <- Wgenerate(alpha=alphatest,eta=etatest,X=Xtest,Z=Ztest)

#test of update_eta
thetainfo <- update_eta(fun=eta_gradient_fun,alphat=alphatest,sigmat=sigmatest,
  etat=etatest,X=Wtest$X,Z=Wtest$Z,Y=Wtest$Y,
  learning_rate=0.1,regular_parameter=0.001,max_iteration=10000)
```



---

|              |                               |
|--------------|-------------------------------|
| update_gamma | <i>Update Alpha and Sigma</i> |
|--------------|-------------------------------|

---

**Description**

Update alpha and sigma in t+1 step with given data and coefficients estimated in step t.

**Usage**

```
update_gamma(alphat, sigmat, etat, X, Z, Y)
```

**Arguments**

|        |   |
|--------|---|
| alphat | the estimated coefficients of the mean of each subgroup in step t |
| sigmat | the estimated standard error of Y in step t                       |
| etat   | the estimated coefficients determining subgroup in step t         |
| X      | the covariables of the mean of each subgroup                      |
| Z      | the covariables determining subgroup                              |
| Y      | the respond variable  |

**Value**

|       |                              |
|-------|------------------------------|
| alpha | alpha estimated in step t+1. |
| eta   | eta estimated in step t.     |
| sigma | sigma estimated in step t+1. |

**Author(s)**

Linsui Deng

**Examples**

```
#some variables
samplesize <- 1000
classsize <- 6
etasize <- 3
alphasize <- 2

Xtest <- data.frame(matrix(rnorm(samplesize*etasize), samplesize, etasize))
Ztest <- matrix(rnorm(samplesize*alphasize), samplesize, alphasize)

etatest <- matrix(seq(1.15, 1, length=etasize*classsize), etasize, classsize)
alphatest <- matrix(seq(1.15, 1, length=alphasize*classsize), alphasize, classsize)
sigmatest <- 0.1

Wtest <- Wgenerate(alpha=alphatest, eta=etatest, X=Xtest, Z=Ztest)
```

```
#test of updata_gamma
thetaupdate_gamma <- update_gamma(alphat=alphatest, sigmat=sigmatest,
                                   etat=etatest, X=Wtest$X, Z=Wtest$Z, Y=Wtest$Y)
```

---

|               |                               |
|---------------|-------------------------------|
| weight_matrix | <i>Weighted Inner Product</i> |
|---------------|-------------------------------|

---

### Description

Calculate the weighted inner product of A and B with the weight W. This result is useful in Logistic Regression.

### Usage

```
weight_matrix(A, W, B)
```

### Arguments

|   |                  |
|---|------------------|
| A | the left matrix  |
| W | the weight       |
| B | the right matrix |

### Value

the wighted inner product, noticing it can be vector when A or B is 2 dimension.

### Author(s)

Linsui Deng

### Examples

```
#data generation
A <- matrix(rnorm(200),100,2)
W <- rnorm(100)
B <- matrix(rnorm(100),100,1)
weighted <- weight_matrix(A,W,B)
```

---

Wgenerate

*Sigmoid Logistic Data Generation*


---

**Description**

Generate data satisfies Sigmoid Logistic Model to check EMalgorithm.

**Usage**

```
Wgenerate(alpha, sigma = 1, eta, samplesize = 0, X, Z, seed1 = 0, seed2 = 0)
```

**Arguments**

|            |   |
|------------|---|
| alpha      | the coefficients of the mean of each subgroup |
| sigma      | the variance of Y                             |
| eta        | the coefficients determining subgroup         |
| samplesize | the size of sample you wanna generate         |
| X          | the covariables of the mean of each subgroup  |
| Z          | the covariables determining subgroup          |
| seed1      | random seed of generating Y                   |
| seed2      | random seed of generating G                   |

**Value**

|   |  |
|---|--|
| X | the covariables of the mean of each subgroup |
| Z | the covariables determining subgroup         |
| Y | the generated respond variable               |
| G | the classes items belonging to               |

**Author(s)**

Linsui Deng

**Examples**

```
#some variables
samplesize <- 1000
classsize <- 6
etasize <- 3
alphasize <- 2

#test of Wgenerate
Xtest <- data.frame(matrix(rnorm(samplesize*etasize), samplesize, etasize))
Ztest <- matrix(rnorm(samplesize*alphasize), samplesize, alphasize)
```

```
etatest <- matrix(seq(1.15,1,length=etasize*classsize),etasize,classsize)
alphatest <- matrix(seq(1.15,1,length=alphasize*classsize),alphasize,classsize)

Wttest <- Wgenerate(alpha=alphatest,eta=etatest,X=Xtest,Z=Ztest)
```

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