

# Package ‘ROCSI’

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**Type** Package

**Title** Receiver Operating Characteristic Based Signature Identification

**Version** 0.1.0

**Description** Optimal linear combination predictive signatures for maximizing the area between two Receiver Operating Characteristic (ROC) curves (treatment vs. control).

**License** GPL (>= 2)

**Encoding** UTF-8

**RoxygenNote** 7.1.1

**Imports** glmnet, MASS

**NeedsCompilation** no

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AUC	<i>AUC</i>
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**Description**

Empirical AUC estimate

**Usage**

```
AUC(outcome, predict)
```

**Arguments**

outcome	binary outcome (1: desired outcome; 0: otherwise)
predict	prediction score

**Details**

Function for AUC when input is X and Y.

**Value**

a numeric value of empirical estimation of area under the ROC curves

**Examples**

```
# no run
```

beta2theta	<i>beta2theta</i>
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**Description**

Function to translate beta into theta, the n-sphere constrain

**Usage**

```
beta2theta(beta)
```

**Arguments**

beta	estimates of coefficient beta
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**Details**

Function to translate beta into theta, the n-sphere constrain

**Value**

a numeric vector for theta (dimension-1)

**Examples**

```
# no run
```

---

*C.index**C.index*

---

**Description**

Empirical c-index estimate

**Usage**

```
C.index(yvar, score, censorvar, data)
```

**Arguments**

yvar	column name for observed time
score	column name for marker value
censorvar	column name for censor (1 is event, 0 is censored)
data	input data matrix

**Details**

Function for c-index when input is X and Y.

**Value**

a numeric value of empirical estimation of c-index

**Examples**

```
# no run
```

**cvfolds0***cvfolds0***Description**

internal function for generating CV fold index

**Usage**

```
cvfolds0(X, Y, idx, nfolds = 5)
```

**Arguments**

X	marker matrix for non-responders
Y	marker matrix for responders
idx	m*n by 2 matrix for row index of marker matrix, first column is row index in X; second column is for Y
nfolds	the cross-validation folds

**Details**

Function for generate CV fold index

**Value**

a vector containing CV fold index for each row in Z

**Examples**

```
# no run
```

**data.gen***data.gen***Description**

Function for simulated data generation

**Usage**

```
data.gen(
  n,
  k,
  prevalence = sqrt(0.5),
  prog.eff = 1,
  sig2,
  y.sig2,
  rho,
  rhos.bt.real,
  a.constant
)
```

**Arguments**

n	Total sample size
k	Number of markers
prevalence	prevalence of predictive biomarkers with values above the cutoff
prog.eff	effect size <i>beta</i> for prognostic biomarker
sig2	standard deviation of each marker
y.sig2	Standard Deviation of the error term in the linear component
rho	$\rho \times \text{sig2}$ is the entries for covariance matrix between pairs of different k markers
rhos.bt.real	correlation between each prognostic and predictive markers
a.constant	a constant is set such that there is no overall treatment effect

**Details**

Function for simulated data generation

**Value**

A list of simulated clinical trial data with heterogeneous prognostic and predictive biomarkers

**Examples**

```
n <- 500
k <- 10
prevalence <- sqrt(0.5)
rho<-0.2
sig2 <- 2
rhos.bt.real <- c(0, rep(0.1, (k-3)))*sig2
y.sig2 <- 1
prog.eff <- 0.5
effect.size <- 1
a.constant <- effect.size/(2*(1-prevalence))
ObsData <- data.gen(n=n, k=k, prevalence=prevalence, prog.eff=prog.eff,
                     sig2=sig2, y.sig2=y.sig2, rho=rho,
                     rhos.bt.real=rhos.bt.real, a.constant=a.constant)
```

grad.sub

*grad.sub***Description**

Internal function of grad\_square in the GCV

**Usage**

```
grad.sub(z, beta)
```

**Arguments**

<b>z</b>	(m x n) x p data matrix as prepared for ROCSI
<b>beta</b>	estimates of coefficient beta

**Details**

Internal function of grad\_square in the GCV

**Value**

grad\_square in the GCV

**Examples**

```
# no run
```

gradsqr

*gradsqr***Description**

Internal function for HIC calculation

**Usage**

```
gradsqr(beta, Z0, index, w = 1)
```

**Arguments**

<b>beta</b>	estimates of coefficient beta
<b>Z0</b>	(m x n) x p Z matrix as prepared for ROCSI
<b>index</b>	m*n by 2 matrix for the subindex for the pair difference in Z
<b>w</b>	a vector of weights Z (can be used for inverse probability weighting for missing data, default is 1)

**Details**

Internal function for HIC calculation

**Value**

gradient square for the GCV.

**Examples**

```
# no run
```

---

*hessAUC**hessAUC*

---

**Description**

function for Hessian matrix of AUC

**Usage**

```
hessAUC(beta, Z, w = 1)
```

**Arguments**

- |      |  |
|------|--|
| beta | estimates of coefficient beta  |
| Z    | (m x n) x p data matrix as prepared for ROCSI  |
| w    | a vector of weights Z (can be used for inverse probability weighting for missing data, default is 1) |

**Details**

function for Hessian matrix of AUC

**Value**

Hessian matrix of AUC.

**Examples**

```
# no run
```

`hessAUC.sub`      *hessAUC.sub*

### Description

Internal function for hessAUC

### Usage

```
hessAUC.sub(z, beta)
```

### Arguments

<code>z</code>	(m x n) x p data matrix as prepared for ROCSI
<code>beta</code>	estimates of coefficient beta

### Details

Internal function for hessAUC

### Value

Hessian matrix components.

### Examples

```
# no run
```

`HIC`      *HIC*

### Description

function for HIC calculation

### Usage

```
HIC(beta, Z, index, w = 1)
```

### Arguments

<code>beta</code>	estimates of coefficient beta
<code>Z</code>	matrix prepared for ROCSI
<code>index</code>	m*n by 2 matrix for the subindex for the pair difference in Z
<code>w</code>	a vector of weights Z (can be used for inverse probability weighting for missing data, default is 1)

**Details**

Function for HIC calculation

**Value**

A numeric value with corresponding HIC

**Examples**

```
# no run
```

---

*MClogit**MClogit*

---

**Description**

function for modified covariate methods based on glmnet

**Usage**

```
MClogit(  
  dataset,  
  yvar,  
  xvars,  
  trtvar,  
  cvar = NULL,  
  nfolds = 5,  
  type = "binary",  
  newx = NULL,  
  bestsub = "lambda.1se",  
  type.measure = "auc"  
)
```

**Arguments**

dataset	data matrix for training dataset
yvar	column name for outcome
xvars	a string vector of column names for input markers
trtvar	column name for treatment (the column should contain binary code with 1 being treatment and 0 being control)
cvar	column name for censor (the column should contain binary code with 1 being event and 0 being censored)
nfolds	n fold CV used for cv.glmnet
type	outcome type ("binary" for binary outcome and "survival" for time-to-event outcome)

newx	data matrix for testing dataset X
bestsub	criteria for best lambda, used by glmnet
type.measure	type of measure used by glmnet

## Details

function for ROCSI

## Value

A list with ROCSI output

**x.logit** final beta estimated from MClogit

**predScore** a data.frame of testing data and its predictive signature scores (based on beta.aABC) for each subjects

**abc** ABC in testing dataset based on optimal beta

**fit.cv** the fitted glmnet object

## Examples

```
n <- 100
k <- 5
prevalence <- sqrt(0.5)
rho<-0.2
sig2 <- 2
rhos.bt.real <- c(0, rep(0.1, (k-3)))*sig2
y.sig2 <- 1
yvar="y.binary"
xvars=paste("x", c(1:k), sep="")
trtvar="treatment"
prog.eff <- 0.5
effect.size <- 1
a.constant <- effect.size/(2*(1-prevalence))
ObsData <- data.gen(n=n, k=k, prevalence=prevalence, prog.eff=prog.eff,
                     sig2=sig2, y.sig2=y.sig2, rho=rho,
                     rhos.bt.real=rhos.bt.real, a.constant=a.constant)
TestData <- data.gen(n=n, k=k, prevalence=prevalence, prog.eff=prog.eff,
                      sig2=sig2, y.sig2=y.sig2, rho=rho,
                      rhos.bt.real=rhos.bt.real, a.constant=a.constant)
bst.mod <- MClogit(dataset=ObsData$data, yvar=yvar, xvars=xvars,
                     trtvar=trtvar, nfolds = 5, newx=TestData$data,
                     type="binary", bestsub="lambda.1se")
bst.mod$abc
bst.mod$x.logit[-1,1]
```

---

*pair.diff**pair.diff*

---

**Description**

internal function for generating Z matrix (binary endpoint)

**Usage**

```
pair.diff(X, Y, A)
```

**Arguments**

X	marker matrix for non-responders
Y	marker matrix for responders
A	Treatment arm indicator (1 is treatment, 0 is control)

**Details**

Function for generate Z matrix for binary endpoint

**Value**

A list of prepared data input for ROCSI

**Examples**

```
# no run
```

---

*pair.diff.surv**pair.diff.surv*

---

**Description**

internal function for generating Z matrix (time-to-event endpoint)

**Usage**

```
pair.diff.surv(X, Y, A, C)
```

**Arguments**

X	marker matrix
Y	a vector for observed time
A	a vector for Treatment arm indicator (1 is treatment, 0 is control)
C	a vector for censor (1 is event, 0 is censored)

**Details**

Function for generate Z matrix for time-to-event endpoint

**Value**

A list of prepared data input for ROCSI

**Examples**

```
# no run
```

ROCSI

*ROCSI*

**Description**

function for ROCSI

**Usage**

```
ROCSI(  
  Dtrain,  
  Dtest = NULL,  
  yvar,  
  xvars,  
  trtvar,  
  cvar = NULL,  
  nfolds = 5,  
  type = "binary"  
)
```

**Arguments**

Dtrain	data matrix for training dataset
Dtest	optional data matrix for testing dataset
yvar	column name for outcome
xvars	a string vector of column names for input markers
trtvar	column name for treatment (the column should contain binary code with 1 being treatment and 0 being control)
cvar	column name for censor (the column should contain binary code with 1 being event and 0 being censored)
nfolds	n fold CV used for cv.glmnet
type	outcome type ("binary" for binary outcome and "survival" for time-to-event outcome)

## Details

function for ROCSI

## Value

A list with ROCSI output

**beta.aABC** final beta estimated from ROCSI based on  $ABC^{(acv)}$   
**beta.1se** final beta estimated from lambda.1se based on nfold CV  
**lambda.aABC** optimal lambda selected by optimizing  $ABC^{(acv)}$   
**fit.cv** fitted cv.glmnet model  
**log** log matrix of all lambdas and ABCs  
**abc.test** ABC in testing dataset based on optimal beta  
**abc.test1se** ABC in testing dataset based on 1se beta  
**predScore** a data.frame of testing data and its predictive signature scores (based on beta.aABC) for each subjects  
**predScore.1se** a data.frame of testing data and its predictive signature scores (based on beta.1se) for each subjects

## Examples

```

n <- 100
k <- 5
prevalence <- sqrt(0.5)
rho<-0.2
sig2 <- 2
rhos.bt.real <- c(0, rep(0.1, (k-3)))*sig2
y.sig2 <- 1
yvar="y.binary"
xvars=paste("x", c(1:k), sep="")
trtvar="treatment"
prog.eff <- 0.5
effect.size <- 1
a.constant <- effect.size/(2*(1-prevalence))
ObsData <- data.gen(n=n, k=k, prevalence=prevalence, prog.eff=prog.eff,
                     sig2=sig2, y.sig2=y.sig2, rho=rho,
                     rhos.bt.real=rhos.bt.real, a.constant=a.constant)
TestData <- data.gen(n=n, k=k, prevalence=prevalence, prog.eff=prog.eff,
                      sig2=sig2, y.sig2=y.sig2, rho=rho,
                      rhos.bt.real=rhos.bt.real, a.constant=a.constant)
bst.aabc <- ROCSI(Dtrain=ObsData$data, Dtest = TestData$data, yvar=yvar,
                     xvars=xvars, trtvar=trtvar, cvar=NULL, nfolds=5, type="binary")
bst.aabc$beta.aABC
bst.aabc$log
bst.aabc$abc.test
bst.aabc$beta.1se
bst.aabc$abc.test1se
  
```

---

theta2beta	<i>theta2beta</i>
------------	-------------------

---

**Description**

Function to translate theta into beta

**Usage**

```
theta2beta(theta)
```

**Arguments**

theta	n-sphere coordination
-------	-----------------------

**Details**

Function to translate beta into theta, the n-sphere constrain

**Value**

a numeric vector for beta (dimension+1)

**Examples**

```
# no run
```

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