

# Package ‘mcc’

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**Type** Package

**Title** Moment Corrected Correlation

**Version** 1.0

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**Author** Yi-Hui Zhou

**Description** A number of biomedical problems involve performing many hypothesis tests, with an attendant need to apply stringent thresholds. Often the data take the form of a series of predictor vectors, each of which must be compared with a single response vector, perhaps with nuisance covariates. Parametric tests of association are often used, but can result in inaccurate type I error at the extreme thresholds, even for large sample sizes. Furthermore, standard two-sided testing can reduce power compared to the doubled p-value, due to asymmetry in the null distribution. Exact (permutation) testing approaches are attractive, but can be computationally intensive and cumbersome. MCC is an approximation to exact association testing of two vectors that is accurate and fast enough for standard use in high-throughput settings, and can easily provide standard two-sided or doubled p-values.

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**License** GPL (>= 2)

**LazyLoad** yes

**NeedsCompilation** no

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MCC-package

*Moment Correction Correlation*

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**Description**

MCC is a method which can measure association between rows of a matrix with a single response vector. The method uses a parametric approximation to permutation of the correlation coefficient. It is very accurate, often to p-values of  $10^{-8}$  or smaller.

**Details**

Package: MCC  
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License: GPL (>= 2)

Please prepare a matrix format of the data, with each row representing the feature and each column for each sample. Please also prepare the clinical response/phenotype in a vector format.

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**References**

Yi-Hui Zhou, Fred Wright, 2013, *Fast And Robust Association Testing For High-Throughput Testing*, Submitted.

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getAkmoment*Four moments of Pearson correlation under permutation*

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**Description**

The Pearson Correlation Coefficient can be derived by the correlation of scaled feature matrix and clinical response. This function provides the first four moments under permutation of the scaled pearson correlation coefficient. In order to simplify the computation of moments summed over strata, an internal offset is applied to center each stratum levels.

**Usage**

```
getAkmoment(x,y)
```

**Arguments**

x                    matrix ( $m \times n$ ) format.  
y                    clinical/experimental  $n$ -vector.

**Details**

Generates the first 4 moments of pearson correlation under permutation of  $A_k = \sum_{i \in I_k} x_i y_i$ . These steps are performed simultaneously for all  $m$  features.

**Author(s)**

Yi-Hui Zhou: <yihui\_zhou@ncsu.edu>

**References**

Yi-Hui Zhou, Fred Wright, 2013, *Fast And Robust Association Testing For High-Throughput Testing*, Submitted.

See also the vignette included with this package.

**See Also**

[getAmoment](#).

**Examples**

```
set.seed(1)
## simulate a data matrix with 300 rows and 30 columns
m=300
n=30
## assume x is a gene expression matrix with 300 genes, 30 samples
x=matrix(rnorm(m*n),m,n)
## y is the clinical response (continuous)
y=rnorm(n)
out=getAkmoment(x,y)
names(out)
out$final2[1:10]    ## the second moments for the first 10 genes

## y can also be dichotomous
y=c(rep(1,15),rep(2,15))
out.dia=getAkmoment(x,y)
out.dia$final4[1:10]
```

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getAmoment	<i>Four moments of Pearson correlation under permutation given covariates</i>
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### Description

The Pearson Correlation Coefficient can be derived by the correlation of scaled feature matrix and clinical response. This function provides the first four moments under permutation of the scaled pearson correlation coefficient. Different from function [getAkmoment](#), this function can handle covariate. For convenience in later functions, both central and noncentral moments are outputted.

### Usage

```
getAmoment(x, y, z=NULL)
```

### Arguments

x	matrix ( $m \times n$ ) of predictors.
y	clinical/experimental $n$ -vector.
z	covariate $n$ -vector, assumed discrete with at least two observations per value of $z$ . If $z$ is not provided, the function assumes no covariates. Generates the first 4 moments of pearson correlation under permutation of $A = \sum_k^K \sum_i (x_{ik} y_{\pi[i]k})$ , given $K$ covariate classes defined by $z$ . <a href="#">getAkmoment</a> provides the results for the samples in stratum $k$ .

### Details

Generatate the first moments of pearson correlation under permutation if we are given the covariate  $z$ . In practice, we first divide into  $K$  strata, [getAkmoment](#) actually provides the results for the samples in stratum  $k$ . [getAmoment](#) uses covariate  $z$  to know which sample belong to the  $k$ th stratum, and obtains the 4 moments for each of the  $K$  strata (using [getAkmoment](#)), and finally computes the 4 moments for  $A = \sum_k (A_k)$ . The actual  $x$  and  $y$  objects are  $m \times n$  matrices, so [getAkmoment](#) computes the results simultaneously for all  $m$  features.

### Author(s)

Yi-Hui Zhou: <yihui\_zhou@ncsu.edu>

### References

Yi-Hui Zhou, Fred Wright, 2013, *Fast And Robust Association Testing For High-Throughput Testing*, Submitted.

See also the vignette included with this package.

### See Also

[getAkmoment](#).

**Examples**

```

set.seed(1)
#####
## simulate a data matrix with 300 rows and 30 columns      ##
## assume x is a gene expression matrix with 300 genes, 30 samples ##
## y is the clinical response (continuous)                ##
#####
m=300
n=30
x=matrix(rnorm(m*n),m,n)
y=rnorm(n)
out=getAmoment(x,y)
names(out)
out$EA2[1:10]  ## the second moments of A=sum(x*(centered phenotype))
               ## under permutation for the the first 10 genes

## y can also be dichotomuous
y=c(rep(1,15),rep(2,15))
out.dia=getAmoment(x,y)
length(out.dia$EA4)

## z is covariate. It can either be continuous or dichotomous
## if z is dichotomous
z=c(rep(1,25),rep(2,5))
out.cov=getAmoment(x,y,z)
out.cov$EA2

## if z is continuous, then we need to splict z into K strata
## for example, suppose I sign K=4 strata for z
K=4
z=rexp(n)
cutpoint=as.vector(quantile(z, seq(0,1,b=1/K)))
newz=rep(0,n)
for (k in (1:K)){
  if (k<K) which.group=which((z>=cutpoint[k] & z<cutpoint[k+1])==TRUE)
  if (k==K) which.group=which((z>=cutpoint[k] & z<=cutpoint[k+1])==TRUE)
  newz[which.group]=k
}
out.cov.con=getAmoment(x,y,newz)
out.cov.con$EA3[1:5]

```

getbetap.A

*Momoent Corrected Correlation***Description**

Computes association statistics, the first four moments of the trend statistic  $A$  under permutation, and  $p$ -values based on the Beta density approximation.

**Usage**

```
getbetap.A(getAmoment.list,A=NULL,fix.obs=F)
```

**Arguments**

<code>getAmoment.list</code>	A list which contains all the items generated by <code>getAmoment</code> function.
<code>A</code>	The vector $A$ is composed of elements $A_i = \sum_j x_{ij}y_j$ . In other words, each $A_i$ is the observed statistic of interest. <code>getbetap.A</code> also allows the user to input the $A$ vector directly. This is useful mainly for debugging purposes, or for obtaining approximate $p$ -values across a grid of hypothetical $A$ values.
<code>fix.obs</code>	this option identifies and improves the $p$ -values using for $MCC_1$ features with implausibly small $MCC$ $p$ -values.

**Details**

`mcc` utilizes a general framework for testing the association between features and clinical variables. It simultaneously performs up to millions of tests by using simple parametric approximations applied to exact permutation moments.

**Value**

The function returns an object of class `mcc`.

**Author(s)**

Yi-Hui Zhou: <yihui\_zhou@ncsu.edu>

**References**

Yi-Hui Zhou, Fred Wright, 2013, *Fast And Robust Association Testing For High-Throughput Testing*, Submitted.

See also the vignette included with this package.

**See Also**

[getAmoment](#), [getAkmoment](#), [getbetap.A.2](#).

**Examples**

```
set.seed(1)
m=300
n=30
x=matrix(rnorm(m*n),m,n)
y=c(rep(1,15),rep(2,15))
z=c(rep(1,25),rep(2,5))
output=getbetap.A(getAmoment(x,y,z),A=NULL,fix.obs=FALSE)
#names(output)
#[1] "twosidedp"      "rightp"          "leftp"           "pdouble"         "chebyshev.p"
#[6] "pt"              "lowest.alpha"
```

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`getbetap.A.2`*Momoment Corrected Correlation One*

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**Description**

Perform  $MCC_1$  by successively considering each of the  $n$  samples as a potential outlier. Otherwise the syntax and output are the same as [getbetap.A](#).

**Usage**

```
getbetap.A.2(x,y,z=NULL)
```

**Arguments**

<code>x</code>	matrix ( $m \times n$ ) of predictors.
<code>y</code>	clinical/experimental $n$ -vector.
<code>z</code>	covariate $n$ -vector, assumed discrete with at least two observations per value of $z$ . If $z$ is not provided, the function assumes no covariates. Generates the first 4 moments of pearson correlation under permutation of $A = \sum_k^K \sum_i (x_{ik} y_{\pi[i]k})$ , given $K$ covariate classes defined by $z$ . <code>getAkmoment</code> provides the results for the samples in stratum $k$ .

**Author(s)**

Yi-Hui Zhou: <yihui\_zhou@ncsu.edu>

**References**

Yi-Hui Zhou, Fred Wright, 2013, *Fast And Robust Association Testing For High-Throughput Testing*, Submitted.

See also the vignette included with this package.

**See Also**

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