

# Package ‘fcros’

March 23, 2018

**Version** 1.5.6

**Date** 2018-03-21

**Title** A Method to Search for Differentially Expressed Genes and to Detect Recurrent Chromosomal Copy Number Aberrations

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**Depends** R (>= 3.1.0)

**Description** A fold change rank based method is presented to search for genes with changing expression and to detect recurrent chromosomal copy number aberrations. This method may be useful for high-throughput biological data (micro-array, sequencing, ...). Probabilities are associated with genes or probes in the data set and there is no problem of multiple tests when using this method. For array-based comparative genomic hybridization data, segmentation results are obtained by merging the significant probes detected.

**License** GPL (>= 2)

**NeedsCompilation** yes

**Repository** CRAN

**Date/Publication** 2018-03-23 16:14:36 UTC

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fcros-package	<i>A Method to Search for Differentially Expressed Genes and to Detect Recurrent Chromosomal Copy Number Aberrations</i>
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## Description

Implementation of a method based on fold change rank ordering statistics to search for differentially expressed genes or to detect recurrent chromosomal copy number aberrations. This package can be used for two biological conditions high-throughput dataset (microarray, RNA-seq, ...), for expression profiling dataset over time without replicates or for cytogenetics dataset (aCGH, Sequencing).

## Details

Package:	fcros
Type:	Package
Version:	1.5.6
Date:	2018-03-21
License:	GPL (>= 2)

Package fcros has the following functions:

fcros():	The function to use with a dataset from two biological condition samples. The
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	dataset should be in a single table. The function <code>fcros()</code> performs a pairwise comparison of samples to obtain a matrix of fold changes. The fold changes are sorted and their rank values are used to associate statistic with genes/probes.
<code>fcros2()</code> :	The function to use with datasets from two biological biological conditions. The datasets should be in two separate tables as inputs. The function <code>fcros2()</code> performs a pairwise comparison of samples from each table to obtain fold changes. The fold changes are sorted, their rank values are combined and then used to associate statistic with genes/probes.
<code>pfco()</code> :	The function to use with a dataset from two biological condition samples. The dataset should be in a single table. The function <code>pfco()</code> performs a pairwise comparison of samples to obtain a matrix of fold changes. The fold changes are sorted and their rank values are used to associate statistic with genes/probes using a singular value decomposition.
<code>fcrosMod()</code> :	This function uses fold changes or ratios matrix as input to associate statistic with genes/probes.
<code>pfcoMod()</code> :	This function uses fold changes or ratios matrix as input to associate statistic with genes/probes using a singular value decomposition.
<code>fcrosFCmat()</code> :	This function allows to compute a matrix of fold changes using pairwise comparisons of the two biological condition samples in a dataset.
<code>fcrosTtest()</code> :	This function allows to use the Student t-test to calculate p-values for the genes in a dataset.
<code>fcrosRead()</code> :	This function allows to read a tab delimited text file to be use as an input for the function <code>fcros()</code> , <code>fcros2()</code> or <code>fcrosMod()</code> .
<code>fcrosWrite()</code> :	This function allows to save the results obtained using the function <code>fcros()</code> , <code>fcros2()</code> or <code>fcrosMod()</code> in a tab delimited text file.
<code>fcrosTopN()</code> :	This function allows to search for the top N down- and/or up-regulated genes from the results obtained using the function <code>fcros()</code> , <code>fcros2()</code> , <code>pfco()</code> , <code>fcrosMod()</code> or <code>pfcoMod()</code> .
<code>fvalTopN()</code> :	This function allows to search for the top N down- and/or up-regulated genes from the results obtained using the function <code>fcros()</code> , <code>fcros2()</code> , <code>pfco()</code> , <code>fcrosMod()</code> or <code>pfcoMod()</code> .
<code>pvalTopN()</code> :	This function allows to search for the top N down- and/or up-regulated genes from the results obtained using the function <code>fcros()</code> , <code>fcros2()</code> , <code>pfco()</code> , <code>fcrosMod()</code> or <code>pfcoMod()</code> .
<code>histoPlot()</code> :	This function plots on the screen the histogram of the FCROS statistics obtained using the results of the function <code>fcros()</code> , <code>fcros2()</code> , <code>pfco()</code> , <code>fcrosMod()</code> or <code>pfcoMod()</code> .
<code>fvalVolcanoPlot()</code> :	This function performs a volcano plot of the results obtained using the function <code>fcros()</code> , <code>fcros2()</code> , <code>pfco()</code> , <code>fcrosMod()</code> or <code>pfcomod()</code> .
<code>pvalVolcanoPlot()</code> :	This function performs a volcano plot of the results obtained using the function <code>fcros()</code> , <code>fcros2()</code> , <code>pfco()</code> , <code>fcrosMod()</code> or <code>pfcoMod()</code> .
<code>chrSummary()</code> :	This function summarizes detection results by chromosome
<code>chrSegment()</code> :	This function segments a chromosome data
<code>chrPlot()</code> :	This function performs a plot of the chromosome probes data
<code>chrPlot2()</code> :	This function performs a plot of the chromosome segmentation results
<code>voomReads()</code> :	This function performs a transformation of the read counts
<code>tcnReads()</code> :	This function performs a total count normalization of reads
<code>rankReads()</code> :	The function to use with a dataset from two biological condition samples. The

dataset should be in a single table. The function rankReads() performs a pairwise comparison of samples to obtain a matrix of fold changes. Small uniform values are added to read counts. This is repeated nrun time. The fold changes are sorted and their rank values are used to associate statistic with genes/probes.

### Author(s)

Doulaye Dembele Maintainer: Doulaye Dembele doulaye@igbmc.fr

### References

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC Bioinformatics, 2014, 15:14

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

### Examples

```
data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;
trim.opt <- 0.25;

# perform fcros()
af <- fcros(fdata, cont, test, log2.opt, trim.opt);

# perform Volcano plot
fvalVolcanoPlot(af, thr = 0.01)

# save fcros values in a file
fcrosWrite(af, file = "test2delete_values.txt");

# save fcros parameters in a file
fcrosWrite(af, file = "test2delete_params.txt", values = FALSE);

# now select top 20 down and/or up regulated genes
top20 <- fcrosTopN(af, 20);
alpha1 <- top20$alpha[1];
alpha2 <- top20$alpha[2];
id.down <- matrix(c(0,11), ncol = 1);
id.up <- matrix(c(rep(0,11)), ncol = 1);
n <- length(af$FC);
f.value <- af$f.value;

idown <- 1;
iup <- 1;
for (i in 1:n) {
  if (f.value[i] <= alpha1) { id.down[idown] <- i; idown <- idown+1; }
  if (f.value[i] >= alpha2) { id.up[iup] <- i; iup <- iup+1; }
```

```

}

data.down <- fdata[id.down[1:(idown-1)], ];
ndown <- nrow(data.down);
data.up <- fdata[id.up[1:(iup-1)], ];
nup <- nrow(data.up);

# now plot down regulated genes
t <- 1:20;
op = par(mfrow = c(2,1));
plot(t, data.down[1,2:21], type = "l", col = "blue", xlim = c(1,20),
      ylim = c(0,18), main = "Top down-regulated genes");
for (i in 2:ndown) {
  lines(t, data.down[i,2:21], type = "l", col = "blue")
}

# now plot down and up regulated genes
plot(t, data.up[1,2:21], type = "l", col = "red", xlim = c(1,20),
      ylim = c(0,18), main = "Top up-regulated genes");
for (i in 2:nup) {
  lines(t, data.up[i,2:21], type = "l", col = "red")
}
par(op)

```

---

cghData

*Example of aCGH "data file" for the fcros package.*


---

### Description

This dataset is part of the study performed and published by Sircoulomb et al. 2010, BMC Cancer, 10:539. For our illustration, we used the first 10 patients' data and for only 3 chromosomes: 7, 8 and 9. The complete data are available from the Gene Expression Omnibus website under the accession number GSE17907.

### Usage

```
data(cghData)
```

### Format

A data frame with 33,3613 observations for 10 samples.

Probes: a text for the unique probe ID

GSM447252: a numeric vector with log2 change values

GSM447253: a numeric vector with log2 change values

GSM447254: a numeric vector with log2 change values

GSM447255: a numeric vector with log2 change values

GSM447256: a numeric vector with log2 change values

GSM447257: a numeric vector with log2 change values  
GSM447258: a numeric vector with log2 change values  
GSM447259: a numeric vector with log2 change values  
GSM447260: a numeric vector with log2 change values  
GSM447261: a numeric vector with log2 change values

### Details

This dataset is obtained using array Comparative Genomic Hybridization technology. Agilent 244K design arrays have been used to monitor breast cancer patients, see reference below for details.

### References

F Sircoulomb, I Bekhouche, P Finetti, J Adelaide, AB Hamida, J Bonansea, S Raynaud, C Innocenti, E Charafe-Jauffret, C Tarpin, FB Ayed, P Viens, J Jacquemier, F Bertucci, D Birnbaum and M Chaffanet; Genome profiling of ERBB2-amplified breast cancers. BMC Cancer, 2010, 10:539

### Examples

```
data(cghData)

summary(cghData)

# perform boxplot of data sample values
boxplot(cghData[,2:11])
```

---

cghInfo

*Example of aCGH "info file" for the fcros package.*

---

### Description

This dataset is part of the Agilent 244K design array probes description file which is used in the publication of Sircoulomb et al. 2010, BMC Cancer, 10:539. For our illustration, we used only 3 chromosomes (7, 8 and 9) data. The complete data are available in the file "GPL9158-3352.txt", see the Gene Expression Omnibus website and accession number GSE17907.

### Usage

```
data(cghInfo)
```

### Format

A data frame with 33,3613 rows for 7 columns with information on probes.

Index: a numeric used for the probe

ProbeName: a text for the unique probe ID

GeneSymbol: a text with the gene symbol associated with the probe

**Chromosome:** a text with the chromosome index associated with the probe

**Start:** a start position value for the sequence associated with the probe

**End:** an end position value for the sequence associated with the probe

**Cytoband:** a text for the cytoband associated with the probe

### Details

These information data are obtained from the Agilent 244K design array probes description file, see reference below for details.

### References

F Sircoulomb, I Bekhouche, P Finetti, J Adelaide, AB Hamida, J Bonansea, S Raynaud, C Innocenti, E Charafe-Jauffret, C Tarpin, FB Ayed, P Viens, J Jacquemier, F Bertucci, D Birnbaum and M Chaffanet; Genome profiling of ERBB2-amplified breast cancers. *BMC Cancer*, 2010, 10:539

### Examples

```
data(cghInfo)
```

```
summary(cghInfo)
```

---

 chrMerge

*Using a C code for merging chromosome segments*

---

### Description

This is an internal function for using a C code while merging chromosome segments in the segmentation step.

### Usage

```
chrMerge(nbSeg, idStart, idEnd, lBound, uBound, segVal, segProba,
         fcall, L2R, nd, dm, sigma)
```

### Arguments

nbSeg	Number of current segments
idStart	Position indexes of the first probes for segments
idEnd	Positions indexes of the last probes for segments
lBound	Lower bound position for segments
uBound	Upper position for segments
segVal	Change values associated with segments
segProba	Probabilities associated with segments
fcall	Detection status associated with probes

L2R	Change values associated with probes
nd	Number of acceptable non-detection between two significant of a segment
dm	Average distance between two consecutive probes of the chromosome
sigma	Standard deviation of the residual observations, see reference

**Author(s)**

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**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# data(fdata);
```

---

chrPlot *Plot a chromosome data*

---

**Description**

This function generates a picture using a chromosome data.

**Usage**

```
chrPlot(chrData, thr = 0.05, deb = 100, fin = 1e10)
```

**Arguments**

chrData	A chromosome data obtained from an output of the function dataSummary(): <pre>xinfo2 = dataSummary(af, xinfo, chromosomes, alpha) idx = which(xinfo2\$xinfo.s\$Chromosome == "chr1") chrData = xinfo2\$xinfo.s[idx, ]</pre>
thr	The probability threshold leading to the selection of the significant probes: thr = 0.05
deb	This parameter allows to specify the start position of the chromosome region for plotting. It can be used for zooming. Negative value will lead to the plot of all chromosome data. deb = 100
fin	This parameter allows to specify the end position of the chromosome region for plot. It can be used for zooming. Negative value will lead to the plot of all chromosome data. thr = 1e7

**Value**

This function generates a picture on the screen



**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# load CGH data and info files
data(cghData)
data(cghInfo)
noms = colnames(cghData)
m = length(noms)
samp <- noms[2:m]

# associate statistics with probes
af <- fcrosMod(cghData, samp, log2.opt = 0, trim.opt = 0.25)

chromosomes = c(7:9)
alpha = 0.05

# summarize results for each chromosome
xinfo2 = dataSummary(af, cghInfo, chromosomes, alpha)

# focused on chromosome 7 data
idx = which(xinfo2$xinfo.s$Chromosome == "7")
chrData = xinfo2$xinfo.s[idx, ]

# Plot chromosome 7 data
chrPlot(chrData, thr = alpha)
```

---

chrPlot2

*Plot a chromosome segmentation results*

---

**Description**

This function generates a picture. It uses a chromosome data and the output results of the segmentation function chrSegment().

**Usage**

```
chrPlot2(chrData, chrSeg, deb = 100, fin = 1e10)
```

**Arguments**

chrData	A chromosome data obtained from an output of the function dataSummary(): <pre>xinfo2 = dataSummary(af, xinfo, chromosomes, alpha) idx = which(xinfo2\$xinfo.s\$Chromosome == "chr1") chrData = xinfo2\$xinfo.s[idx, ]</pre>
chrSeg	An output object of the function chrSegment(): chrSeg = chrSegment(chrData, nd = 10)
deb	This parameter allows to specify the start position of the chromosome region for plotting. It can be used for zooming. Negative value will lead to the plot of all chromosome data. deb = 100
fin	This parameter allows to specify the end position of the chromosome region for plot. It can be used for zooming. Negative value will lead to the plot of all chromosome data. thr = 1e7

**Value**

This function generates a picture on the screen

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# load CGH data and info files
data(cghData)
data(cghInfo)
noms = colnames(cghData)
m = length(noms)
samp <- noms[2:m]

# associate statistics with probes in the dataset
af <- pfcoMod(cghData, samp, log2.opt = 0, trim.opt = 0.25)

chromosomes = c(7:9)
alpha = 0.05

# summarize results for each chromosome
xinfo2 = dataSummary(af, cghInfo, chromosomes, alpha)

# focused on chromosome 7 data
idx = which(xinfo2$xinfo.s$Chromosome == "7")
chrData = xinfo2$xinfo.s[idx, ]

# segment chromosome 7 data
chrSeg = chrSegment(chrData, nd = 15)
```

```
# plot chromosome 7 results
op = par(mfrow = c(2,1))
chrPlot(chrData, thr = alpha, deb == -1, fin = 3.5e7)
chrPlot2(chrData, chrSeg, -1, fin = 3.5e7)
par(op)
```

chrSegment

*Segmentation of a chromosome data***Description**

This function allows to segment a chromosome data

**Usage**

```
chrSegment(chrData, nd = 10)
```

**Arguments**

chrData	A chromosome data obtained from an output of the function dataSummary(): <pre>xinfo2 = dataSummary(af, xinfo, chromosomes, alpha) idx = which(xinfo2\$xinfo.s\$Chromosome == "chr1") chrData = xinfo2\$xinfo.s[idx, ]</pre>
nd	The acceptable number of non-detected probes which can separate two significant probes in a segment. Default setting value is 10: nd = 10

**Value**

This function returns a data frame containing 6 information for each segment

idStart	The start position indexes associated with segments
idEnd	The End position indexes associated with segments
lBounds	The lower bound positions associated with segments
uBounds	The upper bound positions associated with segments
segL2R	The change values associated with segments
segProba	The probabilities associated with segments

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```

# load CGH data and info files
data(cghData)
data(cghInfo)
noms = colnames(cghData)
m = length(noms)
samp <- noms[2:m]

# associate statistics with probes in the dataset
af <- pfcoMod(cghData, samp, log2.opt = 0, trim.opt = 0.25)

chromosomes = c(7:9)
alpha = 0.05

# summarize results for each chromosome
xinfo2 = dataSummary(af, cghInfo, chromosomes, alpha)

# focused on chromosome 7 data
idx = which(xinfo2$xinfo.s$Chromosome == "7")
chrData = xinfo2$xinfo.s[idx, ]

# segment chromosome 7 data
chrSeg = chrSegment(chrData, nd = 15)

# show first 10 segment results
chrSeg[1:10,]

```

---

dataSummary

*Summarization of the detection results for a list of chromosomes*


---

**Description**

From an output object of the function `fcrosMod()` or `pfcoMod()`, the chromosomes information object, the list of chromosomes and a threshold, this function creates two objects containing ordered chromosome data and summary results.

**Usage**

```
dataSummary(af, xinfo, chromosomes = c(1:22,"X","Y"), alpha = 0.05)
```

**Arguments**

af	An output object of the function <code>fcrosMod()</code> or <code>pfcoMod()</code> : <code>af = fcrosMod(xdata, samp, log2.opt, trim.opt)</code> <code>af = pfcoMod(xdata, samp, log2.opt, trim.opt)</code>
xinfo	A data frame containing chromosomes information (probe name, gene symbol, chromosome index, start position, end position and the cytoband). These information should appear with the labels <code>ProbeName</code> , <code>GeneSymbol</code> , <code>Chromosome</code> ,

	Start, End, Cytoband. Additional information may be used. Only labels Chromosome, Start and End are mandatory.
chromosomes	A list of chromosomes. Default setting is a list with all chromosomes: chromosomes = (1:22,"X","Y")
alpha	A threshold allowing to select significant probes based on probabilities. Default setting is 0.05 (5% of error) thr = 0.05

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# load CGH data and info files
data(cghData)
data(cghInfo)
noms = colnames(cghData)
m = length(noms)
samp <- noms[2:m]

# associate statistics with probes in the dataset
af <- pfcoMod(cghData, samp, log2.opt = 0, trim.opt = 0.25)

chromosomes = c(7:9)
alpha = 0.05

# summarize results for each chromosome
xinfo2 = dataSummary(af, cghInfo, chromosomes, alpha)

# display the number of significant probes for each chromosome
xinfo2$chrSumm
```

---

fc2Calc

---

*Calculation of fold change using pairwise comparison values*


---

**Description**

This is an internal function for using a C code to calculate fold changes using pairwise comparison of samples.

**Usage**

```
fc2Calc(rvect, n, m, idx, m2)
```

**Arguments**

rvect	Vector containing the full or reduced matrix with the pairwise comparison of samples results
n	Number of genes or probes in the dataset
m	Number of columns of the full or reduced matrix of pairwise comparison of samples results
idx	Indexes of the columns to keep in the pairwise comparison of samples
m2	Number of columns in the full or reduced matrix of comparison of samples

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# data(fdata);
```

---

fcros	<i>Search for differentially expressed genes/probes</i>
-------	---

---

**Description**

Implementation of a method based on fold change rank ordering statistics for detecting differentially expressed genes in a dataset. This function should be used with two biological conditions dataset (microarray or RNA-seq, ...). Using pairwise combinations of samples from the two biological conditions, fold changes (FC) are calculated. For each combination, the FC obtained are sorted in increasing order and corresponding rank values are associated with genes. Then, a statistic is assigned to the robust average ordered rank values for each gene/probe.

**Usage**

```
fcros(xdata, cont, test, log2.opt = 0, trim.opt = 0.25)
```

**Arguments**

xdata	A matrix or a table containing two biological conditions dataset to process for detecting differentially expressed genes: xdata. The first column of the matrix "xdata" should contain the gene IDs or their names.
cont	A vector containing the label names of the control samples: cont = c("cont01", "cont02", ...).
test	A vector containing the label names of the test samples: test = c("test01", "test02", "test03", ...).

log2.opt	A scalar equals to 0 or 1. The value 0 (default) means that data in the matrix "xdata" are expressed in a log2 scale: log2.opt = 0
trim.opt	A scalar between 0 and 0.5. The value 0.25 (default) means that 25% of the lower and the upper rank values of each gene are not used for computing its statistics "ri", i.e. the inter-quartile range rank values are averaged: trim.opt = 0.25

### Details

Label names appearing in the parameters "cont" and "test" should match with some label names in the columns of the data matrix "xdata". It is not necessary to use all label names appearing in the columns of the dataset matrix.

### Value

This function returns a data frame containing 9 components

idnames	A vector containing the list of IDs or symbols associated with genes
ri	The average of rank values associated with genes. These values are rank values statistics leading to f-values and p-values.
FC	The fold changes for genes in the dataset. These fold changes are calculated as a ratio of averages from the test and the control samples. Non log scale values are used in the calculation.
FC2	The robust fold changes for genes. These fold changes are calculated as a trimmed mean of the fold changes or ratios obtained from the dataset samples. Non log scale values are used in the calculation.
f.value	The f-values are probabilities associated with genes using the "mean" and the "standard deviation" ("sd") of the statistics "ri". The "mean" and "sd" are used as a normal distribution parameters.
p.value	The p-values associated with genes. These values are obtained from the fold change rank values and one sample t-test.
bounds	Two values, which are the lower and the upper bounds or the minimum and the maximum values of the non standardized "ri".
params	Three values, which are the estimates for the parameters "delta" (average difference between consecutive ordered average of rank values) "mean" (mean value of "ri") and the standard deviation ("sd") of "ri".
params_t	Three values which are theoretical levels for parameters "delta", "mean" and "sd".

### Author(s)

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

### References

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC Bioinformatics, 2014, 15:14

**Examples**

```

data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;
trim.opt <- 0.25;

# perform fcros()
af <- fcros(fdata, cont, test, log2.opt, trim.opt);

# now select top 20 down and/or up regulated genes
top20 <- fcrosTopN(af, 20);
alpha1 <- top20$alpha[1];
alpha2 <- top20$alpha[2];
id.down <- matrix(0, 1);
id.up <- matrix(0, 1);
n <- length(af$FC);
f.value <- af$f.value;

idown <- 1;
iup <- 1;
for (i in 1:n) {
  if (f.value[i] <= alpha1) { id.down[idown] <- i; idown <- idown + 1; }
  if (f.value[i] >= alpha2) { id.up[iup] <- i; iup <- iup + 1; }
}

data.down <- fdata[id.down[1:(idown-1)], ];
ndown <- nrow(data.down);
data.up <- fdata[id.up[1:(iup-1)], ];
nup <- nrow(data.up);

# now plot down regulated genes
t <- 1:20;
op = par(mfrow = c(2,1));
plot(t, data.down[1,2:21], type = "l", col = "blue", xlim = c(1,20),
      ylim = c(0,18), main = "Top down-regulated genes");
for (i in 2:ndown) {
  lines(t,data.down[i,2:21], type = "l", col = "blue")
}

# now plot down and up regulated genes
plot(t, data.up[1,2:21], type = "l", col = "red", xlim = c(1,20),
      ylim = c(0,18), main = "Top up-regulated genes");
for (i in 2:nup) {
  lines(t, data.up[i,2:21], type = "l", col = "red")
}
par(op)

```



fcros2

*Search for differentially expressed genes/probes***Description**

Implementation of a method based on fold change or ratio rank ordering statistics for detecting differentially expressed genes. This function should be used with dataset in two separate tables and from two biological conditions datasets (microarray, RNA-seq, ...). Pairwise combinations of samples from the two biological conditions are used to obtain a matrix of fold changes. For each combination, the FCs obtained are sorted in an increasing order and the corresponding rank values are associated with genes/probes. Then, a statistic is associated with each gene/probe.

**Usage**

```
fcros2(xdata1, xdata2, cont, test, log2.opt = 0, trim.opt = 0.25)
```

**Arguments**

xdata1	A matrix or a table containing two biological conditions dataset to process for detecting differentially expressed genes: xdata1. The first column of the matrix "xdata1" should contain the gene IDs or symbols.
xdata2	A matrix or a table containing two biological conditions dataset to process for detecting differentially expressed genes: xdata2. The first column of the matrix "xdata2" should contain the gene IDs or symbols.
cont	A vector containing the label names of the control samples: cont = c("cont01", "cont02", ...)
test	A vector containing the label names of the test samples: test = c("test01", "test02", "test03", ...)
log2.opt	A scalar equals to 0 or 1. The value 0 (default) means that data in the tables "xdata1" and "xdata2" are expressed in a log2 scale: log2.opt = 0
trim.opt	A scalar between 0 and 0.5. The value 0.25 (default) means that 25% of the lower and the upper rank values for a gene are not used for computing the statistics "ri", i.e. the inter-quartile range rank values are averaged: trim.opt = 0.25

**Details**

The label names appearing in the parameters "cont" and "test" should match some label names in the columns of the data matrices "xdata1" and "xdata2". It is not necessary to use all column label names appearing in matrices "xdata1" and "xdata2". However, it is assumed that the same genes (same IDs or symbol) are used in xdata1 and xdata2.

**Value**

This function returns a data frame containing 9 components

idnames	A vector containing the list of the IDs or symbols associated with genes
---------	--

ri	The average of rank values associated with genes in the datasets. These values are rank statistics leading to f-values and p-values.
FC	The fold changes for genes. These fold changes are calculated as a ratio of averages from the test and control samples. Non log scale values are used in the calculation.
FC2	The robust fold changes for genes. These fold changes are calculated as a trimmed mean of the fold changes or ratios obtained from the dataset samples using the pairwise comparisons. Non log scale values are used in the calculation.
f.value	The f-values are probabilities associated with genes. These values are obtained using the "mean" and the "standard deviation" ("sd") of the statistics "ri". The "mean" and "sd" are used as a normal distribution parameters.
p.value	The p-values associated with genes. These values are obtained from the f-values.
bounds	Two values which are the lower and the upper bound or the minimum and the maximum values of non standardized "ri".
params	Three values, which are the estimates for the parameters "delta" (average difference between consecutive ordered average of rank) "mean" (mean value of variable "ri") and the standard deviation ("sd") of variable "ri".
params_t	Three values, which are the theoretical levels for parameters "delta", "mean" and "sd".

### Author(s)

Doulaye Dembele doulaye@igbmc.fr

### References

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC Bioinformatics, 2014, 15:14

### Examples

```

data(fdata);

cont <- c("cont01", "cont02", "cont03", "cont04", "cont05",
         "cont06", "cont07", "cont08", "cont09", "cont10");
test <- c("test01", "test02", "test03", "test04", "test05",
         "test06", "test07", "test08", "test09", "test10");
log2.opt <- 0;
trim.opt <- 0.25;

# perform fcros2()
xdata1 <- fdata[,c(2:5, 12:17)];
xdata2 <- fdata[,c(6:11, 18:21)];
af2 <- fcros2(xdata1, xdata2, cont, test, log2.opt, trim.opt);

# now select top 20 down and/or up regulated genes
top20 <- fcrosTopN(af2, 20);
alpha1 <- top20$alpha[1];
alpha2 <- top20$alpha[2];

```

```

id.down <- matrix(0,1);
id.up <- matrix(0,1);
n <- length(af2$FC);
f.value <- af2$f.value;

idown <- 1;
iup <- 1;
for (i in 1:n) {
  if (f.value[i] <= alpha1) { id.down[idown] <- i; idown <- idown + 1; }
  if (f.value[i] >= alpha2) { id.up[iup] <- i; iup <- iup + 1; }
}

data.down <- fdata[id.down[1:(idown-1)], ];
ndown <- nrow(data.down);
data.up <- fdata[id.up[1:(iup-1)], ];
nup <- nrow(data.up);

# now plot down regulated genes
t <- 1:20;
op = par(mfrow = c(2,1));
plot(t, data.down[1,2:21], type = "l", col = "blue", xlim = c(1,20),
      ylim = c(0,18), main = "Top down-regulated genes");
for (i in 2:ndown) {
  lines(t,data.down[i,2:21], type = "l", col = "blue")
}

# now plot down and up regulated genes
plot(t, data.up[1,2:21], type = "l", col = "red", xlim = c(1,20),
      ylim = c(0,18), main = "Top up-regulated genes");
for (i in 2:nup) {
  lines(t, data.up[i,2:21], type = "l", col = "red")
}
par(op)

```

---

fcrosFCmat

*Calculation of a matrix of fold changes using pairwise comparisons*


---

### Description

This function allows to compute a matrix of fold changes using pairwise comparisons of a two biological conditions dataset. This function is used internally by fcros2()

### Usage

```
fcrosFCmat(xdata, cont, test, log2.opt=0, trim.opt=0.25)
```

### Arguments

**xdata** A table containing a two biological conditions dataset to process for obtaining a matrix of fold changes that results from pairwise comparisons of samples:

	xdata. The first column of the matrix "xdata" should contain the gene IDs or symbols.
cont	A vector containing label names of the control samples: <code>cont = c("cont01", "cont02", ...)</code>
test	A vector containing label names of the test samples: <code>test = c("test01", "test02", "test03", ...)</code>
log2.opt	A scalar equals to 0 or 1. The value 0 (default) means that the data values in matrix "xdata" are expressed in a log2 scale: <code>log2.opt = 0</code>
trim.opt	A scalar between 0 and 0.5. The value 0.25 (default) means that 25% of the lower and the upper rank values of each gene are not used for computing the statistic "ri" or "u1", i.e. the inter-quartile rank values are averaged: <code>trim.opt = 0.25</code>

### Details

The label names appearing in the parameters "cont" and "test" should match some label names of the columns in the table "xdata". The dataset "xdata" can contain other label names not used.

### Value

This function returns a data frame containing 3 components

idnames	A vector containing the list of IDs or symbols associated with genes
fMat	A matrix of fold changes associated with genes. This matrix is obtained using pairwise comparisons of samples in the dataset "xdata".
FC	The fold changes for genes in the dataset "xdata". These fold changes are calculated as a ratio of averages from the test and the control samples. Non log scale values are used in this calculation.
FC2	The robust fold changes for genes in the dataset "xdata". These fold changes are calculated as a trimmed mean of the fold changes obtained using pairs of samples. Non log scale values are used in this calculation.

### Author(s)

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

### References

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC Bioinformatics, 2014, 15:14

### Examples

```
data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;
```

```
# perform fcrosFCmat()
fc <- fcrosFCmat(fdata, cont, test, log2.opt);

# plot histogram of the fold change (log2 scale) in dataset
hist(log2(fc$FC),nclass = 50);
```

---

fcrosMod	<i>Search for differentially expressed genes or to detect recurrent copy number aberration probes</i>
----------	---

---

## Description

Implementation of a method based on fold change rank ordering statistics to search for differentially expressed genes or to detect chromosomal recurrent copy number aberration probes. This function should be used with a matrix of fold changes or ratios from biological dataset (microarray, RNA-seq, ...). The function fcrosMod() is an extension of the function fcros() to a dataset which does not contain replicate samples or to a dataset with one biological condition dataset. Statistics are associated with genes/probes to characterize their change levels.

## Usage

```
fcrosMod(fcMat, samp, log2.opt = 0, trim.opt = 0.25)
```

## Arguments

fcMat	A matrix containing fold changes or ratios from a biological dataset to process for searching differentially expressed genes or for detecting recurrent copy number aberrations regions: fcMat. The first column of the matrix "fcMat" should contain the gene IDs or symbols.
samp	A vector of sample label names which should appear in the columns of the matrix fcMat: samp.
log2.opt	A scalar equals to 0 or 1. The value 0 (default) means that values in the matrix "fcMat" are expressed in a log2 scale: log2.opt = 0
trim.opt	A scalar between 0 and 0.5. The value 0.25 (default) means that 25% of the lower and the upper rank values for each gene are not used for computing the statistic "ri", i.e. the inter-quartile range rank values are averaged: trim.opt = 0.25

## Details

The label names appearing in the parameter "samp" should match some label names of the columns in the data matrix "xdata". It is not necessary to use all label names appearing in the columns of the dataset matrix.

**Value**

This function returns a data frame containing 8 components

idnames	A vector containing the list of IDs or symbols associated with genes
ri	The average of ordered rank values associated with genes in the dataset. These values are rank statistics leading to the f-values and the p-values.
FC2	The robust fold changes for genes in matrix "fcMat". These fold changes are calculated as a trimmed mean of the values in "fcMat". Non log scale values are used in this calculation.
f.value	The f-values are probabilities associated with genes using the "mean" and the "standard deviation" ("sd") of values in "ri". The "mean" and "sd" are used as a normal distribution parameters.
p.value	The p-values associated with genes. The p-values are obtained using a one sample t-test on the fold change rank values.
bounds	Two values, which are the lower and the upper bounds or the minimum and the maximum values of the non standardized "ri".
params	Three values, which are the estimates for the parameters "delta" (average difference between consecutive ordered average of rank) "mean" (mean value of the "ri") and the standard deviation ("sd") of the "ri".
params_t	Three values, which are theoretical levels for the parameters "delta", "mean" and "sd".

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;
trim.opt <- 0.25;

# perform fcrosMod()
fc <- fcrosFCmat(fdata, cont, test, log2.opt, trim.opt);
m <- ncol(fc$fcMat)
samp <- paste("Col", as.character(1:m), sep = "");
fc.val <- cbind(fdata[,1], data.frame(fc$fcMat))
colnames(fc.val) <- c("ID", samp)

af <- fcrosMod(fc.val, samp, log2.opt, trim.opt);
```

```

# now select top 20 down and/or up regulated genes
top20 <- fcrosTopN(af, 20);
alpha1 <- top20$alpha[1];
alpha2 <- top20$alpha[2];
id.down <- matrix(0,1);
id.up <- matrix(0,1);
n <- length(af$FC);
f.value <- af$f.value;

idown <- 1;
iup <- 1;
for (i in 1:n) {
  if (f.value[i] <= alpha1) { id.down[idown] <- i; idown <- idown + 1; }
  if (f.value[i] >= alpha2) { id.up[iup] <- i; iup <- iup + 1; }
}

data.down <- fdata[id.down[1:(idown-1)], ];
ndown <- nrow(data.down);
data.up <- fdata[id.up[1:(iup-1)], ];
nup <- nrow(data.up);

# now plot down regulated genes
t <- 1:20;
op = par(mfrow = c(2,1));
plot(t, data.down[1, 2:21], type = "l", col = "blue", xlim = c(1,20),
      ylim = c(0,18), main = "Top down-regulated genes");
for (i in 2:ndown) {
  lines(t,data.down[i, 2:21], type = "l", col = "blue")
}

# now plot down and up regulated genes
plot(t, data.up[1,2:21], type = "l", col = "red", xlim = c(1,20),
      ylim = c(0,18), main = "Top up-regulated genes");
for (i in 2:nup) {
  lines(t, data.up[i,2:21], type = "l", col = "red")
}
par(op)

```

---

fcrosRead

*Read a tab delimited text file*


---

### Description

This function can be used to read a tab delimited text file to be used as input for the functions `fcros()`, `fcros2()`, `fcrosMod()` `pfco()` or `pfcoMod()`. You may also use other functions (`read.csv()`, `read.csv2()`, `read.delim()` or `read.delim2()`) to read your dataset.

### Usage

```
fcrosRead(filename)
```

**Arguments**

filename "filename" is a tab delimited text file to read. The first line of this file should contain labels.

**Value**

Output is a matrix or a table containing data.

**Author(s)**

Doulaye Dembele, doulaye@igbmc.fr

**Examples**

```
# generate random dataset of size 100 x 5
xdata = matrix(c(rep(0,600)), ncol = 6);

xdata[,2:6] <- matrix(rnorm(500,0,1), ncol = 5);

x0 <- matrix(NA, 100, 0);

for (i in 1:100) {
  x0[i] <- paste("Obs_",i, sep = "");
}

# set column names
colnames(xdata) <- c("Index", "Col_1", "Col_2", "Col_3", "Col_4", "Col_5");

# save data in a text file
write.table(xdata, file = "test2delete.txt", quote = FALSE, sep = "\t",
  eol = "\n", col.names = TRUE, row.names = FALSE);

# now used fcros.read() to open the file
mydata <- fcrosRead(file = "test2delete.txt");

summary(mydata)
```

---

fcrosTopN

*Search for the top N changed genes or probes*

---

**Description**

This function allows to search for the top N differentially expressed genes or changed probes. It uses the output results obtained using one of the following functions `fcros()`, `fcros2()`, `fcrosMod()`, `pfco()` or `pfcoMod()`.

**Usage**

```
fcrosTopN(af, topN)
```



**Arguments**

af	This is an output object obtained using the functions fcros(), fcros2(), fcrosMod(), pfco() or pfcoMod(): f.value
topN	The expected number of the top DE genes in the dataset used: topN

**Value**

This function returns a data frame containing 2 components

alpha	Two threshold values for the down- and the up-regulated allowing to have the top N DE genes
index	The indexes of the top N DE genes

**Author(s)**

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

**References**

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC Bioinformatics, 2014, 15:14

**Examples**

```
data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;

# perform fcros()
af <- fcros(fdata, cont, test, log2.opt);

# now select top 10 down and/or up regulated genes
top10 <- fcrosTopN(af, 10);

# display thresholds
top10$alpha

# display index of top10 genes
fdata[top10$index, 1]

# display fvalue of the top10 genes
(af$f.value)[top10$index]
```

---

fcrosTtest	<i>Student t-test for detecting differentially expressed genes</i>
------------	--

---

### Description

The function uses the basic R `t.test()` function to perform the Student t-test. It should be used for two biological conditions dataset (microarray, or RNA-seq). The Fold changes, statistics and p-values are returned for each gene in the dataset.

### Usage

```
fcrosTtest(xdata, cont, test, log2.opt = 0)
```

### Arguments

xdata	A table containing a two biological conditions dataset to process for detecting differentially expressed genes: xdata. The first column of the dataset "xdata" should be gene IDs or symbols
cont	A vector containing the label names of the control samples: <code>cont = c("cont01", "cont02", ...)</code>
test	A vector containing the label names of the test samples: <code>test = c("test01", "test02", "test03", ...)</code>
log2.opt	A scalar equals to 0 or 1. The value 0 (default) means that data in the matrix "xdata" are expressed in a log2 scale: <code>log2.opt = 0</code>

### Details

Label names appearing in the parameters "cont" and "test" should match column label names of the data matrix "xdata". It is not necessary to use all column label names of the dataset "xdata".

### Value

FC	The fold changes for the genes in the dataset.
stat	The Student t-test statistics associated with genes.
p.value	The Student t-test p-values associated with genes.

### Author(s)

Doulaye Dembele

**Examples**

```

data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;

# perform fcrosTtest()
at <- fcrosTtest(fdata, cont, test, log2.opt);

# now select some differentially expressed genes
id.de <- matrix(0, 1);
n <- length(at$FC);
for (i in 1:n) {
  if ((at$p.value)[i] <= 0.0005) { id.de <- rbind(id.de, i); }
}

data.de <- fdata[id.de, ];
nde <- nrow(data.de);

# now plot the DE genes
t <- 1:20;
plot(t, data.de[1, 2:21], type = "l", col = "blue", xlim = c(1,20),
      ylim = c(0,18), main = "Down- and up-regulated genes");
for (i in 2:nde) {
  lines(t, data.de[i,2:21], type = "l", col = "blue")
}

```

---

fcrosWrite

*Writing the fcros() or pfco() results in a tab delimited text file*


---

**Description**

This function creates a tab-delimited text file with the results of the function fcros(), fcros2(), fcrosMod(), pfco() or pfcoMod(). Two kind of files can be obtained. In the first, the results of the parameters "idnames", "ri" ("u1"), "FC", "FC2", "f.value" and "p.value" are saved. In the second, the results of the parameters "lb", "ub", "delta", "meand" and "sd" (fcros) or "comp", "comp.w" and "comp.wcum" (pfco()) are saved.

**Usage**

```
fcrosWrite(af, file = "fcrosResults.txt", values = TRUE, thr = 1)
```

**Arguments**

af	An output object of the functions fcros(), fcros2(), fcrosMod(), pfco() or pfcoMod(): af = fcros(xdata, cont, test, log2.opt, trim.opt) af = pfco(xdata, cont, test, log2.opt, trim.opt)
file	The output file name: file = "fcrosResults.txt"

values	A flag allowing to create a file with the parameters "ri" (or "u1"), "FC", "FC2", "f.value" and "p.value" (default) or with the parameters "lb", "ub", "delta", "meand" and "sd" (fcros() case) or with the parameters "comp", "comp.w" and "comp.wcum" (pfco() case): values = TRUE
thr	A threshold allowing to filter data based on p-values. Default setting is 1 (no filtering) thr = 1

### Value

This function creates and saves a tab-delimited text file on the disk.

### Author(s)

Doulaye Dembele doulaye@igbmc.fr

### Examples

```
data(fdata);
cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;
trim.opt <- 0.25;
af <- fcros(fdata, cont, test, log2.opt, trim.opt);
af2 <- pfco(fdata, cont, test, log2.opt, trim.opt);

fcrosWrite(af, file = "test2delete_values.txt");
fcrosWrite(af, file = "test2delete_params.txt", values = FALSE);
fcrosWrite(af2, file = "test2delete2_values.txt");
fcrosWrite(af2, file = "test2delete2_params.txt", values = FALSE);
```

---

fdata

*Example of dataset for the fcros package.*

---

### Description

This is a synthetic dataset generated using the R package "madsim" (available from the CRAN web site). Two biological conditions data matrix using 10,000 genes and 20 samples (10 control and 10 test samples) were simulated. 108 and 102 genes of this dataset are down- and up-regulated, respectively.

### Usage

```
data(fdata)
```

**Format**

A data frame with 10,000 observations on the following 22 variables.

**index:** a text for the genes, unique ID

**cont01:** a numeric vector with log2 intensities for control sample 01

**cont02:** a numeric vector with log2 intensities for control sample 02

**cont03:** a numeric vector with log2 intensities for control sample 03

**cont04:** a numeric vector with log2 intensities for control sample 04

**cont05:** a numeric vector with log2 intensities for control sample 05

**cont06:** a numeric vector with log2 intensities for control sample 06

**cont07:** a numeric vector with log2 intensities for control sample 07

**cont08:** a numeric vector with log2 intensities for control sample 08

**cont09:** a numeric vector with log2 intensities for control sample 09

**cont10:** a numeric vector with log2 intensities for control sample 10

**test01:** a numeric vector with log2 intensities for test sample 01

**test02:** a numeric vector with log2 intensities for test sample 02

**test03:** a numeric vector with log2 intensities for test sample 03

**test04:** a numeric vector with log2 intensities for test sample 04

**test05:** a numeric vector with log2 intensities for test sample 05

**test06:** a numeric vector with log2 intensities for test sample 06

**test07:** a numeric vector with log2 intensities for test sample 07

**test08:** a numeric vector with log2 intensities for test sample 08

**test09:** a numeric vector with log2 intensities for test sample 09

**test10:** a numeric vector with log2 intensities for test sample 10

**DE\_status:** a numeric vector with values -1, 0 and 1. Value 0 is used for no change genes, while -1 and 1 are used for down- and up-regulated genes, respectively.

**Details**

This dataset is obtained using the microarray data simulation model implemented in the package "madsim". A real microarray data, "madsim\_test", was used as seed and the number of the control and the test samples were set to 10. The parameter "sdn" was set to 0.3 and all the other parameters in the madsim package were set to their default settings.

**References**

Dembele D, A flexible microarray data simulation model. Microarrays, 2013, v.2, n.2, pp.115-130

**Examples**

```

data(fdata)

op = par(mfrow = c(2,1));

# perform MA plot using samples "cont01" and "cont05"
A1 = 0.5*(fdata$cont01 + fdata$cont05);
M1 = fdata$cont05 - fdata$cont01;

plot(A1, M1, col="red", xlim=c(2,18), ylim=c(-5,5),
      main="MA plot with two control samples");
lines(x = c(2,18), y = c(0,0), col = "blue")
lines(x = c(2,18), y = c(-1,-1), col = "blue")
lines(x = c(2,18), y = c(1,1), col = "blue")

# perform MA plot using samples "cont01" and "test05"
A2 = 0.5*(fdata$cont01 + fdata$test05);
M2 = fdata$test05 - fdata$cont01;

plot(A2, M2, col="red", xlim=c(2,18), ylim=c(-5,5),
      main="MA plot with one control and one test samples");
lines(x = c(2,18), y = c(0,0), col = "blue")
lines(x = c(2,18), y = c(-1,-1), col = "blue")
lines(x = c(2,18), y = c(1,1), col = "blue")
par(op)

```

---

fvalTopN

*Search for the top N changed genes or probes using f-values*


---

**Description**

This function allows to search for the top N differentially expressed genes or changed probes. It uses the f-values obtained using one of the following functions `fcros()`, `fcros2()`, `fcrosMod()`, `pfco()` or `pfcoMod()`.

**Usage**

```
fvalTopN(fval, topN)
```

**Arguments**

fval	This is a f-values vector obtained using the functions <code>fcros()</code> , <code>fcros2()</code> , <code>fcrosMod()</code> , <code>pfco()</code> or <code>pfcoMod()</code> : <code>fval = af\$f.value</code>
topN	The expected number of the top DE genes/probes in the dataset used: <code>topN</code>

**Value**

This function returns a data frame containing 2 components

alpha	Two threshold values for the down- and the up-regulated allowing to have the top N DE genes
index	The indexes of the top N DE genes / probes

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC Bioinformatics, 2014, 15:14

**Examples**

```
data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;

# perform pfco()
af <- pfco(fdata, cont, test, log2.opt);

# now select top 10 down and/or up regulated genes
top10 <- fvalTopN(af$f.value, 10);

# display thresholds
top10$alpha

# display index of top10 genes
fdata[top10$index, 1]

# display fvalue of the top10 genes
(af$f.value)[top10$index]
```

---

fvalVolcanoPlot

*Performs a volcano plot of the FCROS/PFCO statistics*

---

**Description**

This function allows to have a volcano like plot using the output results of the function fcros(), fcros2(), fcrosMod(), pfco() or pfcoMod(): p-values versus robust fold changes (FC2). The p-value are transformed using  $-\log_{10}()$ , while FC2 are transformed using  $\log_2()$ .

**Usage**

```
fvalVolcanoPlot(af, thr = 0.05)
```

**Arguments**

**af** This is an object obtained using the functions `fcros()`, `fcros2()`, `fcrosMod()`, `pfco()` or `pfcoMod()`: `af = fcros(xdata, cont, test)`

**thr** The threshold to obtain the DE genes in the dataset (red plots): `thr = 0.05`

**Value**

This function displays on the screen a volcano like plot using the f-values and the robust fold changes (FC2).

**Author(s)**

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

**References**

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC bioinformatics, 2014, 15:14

**Examples**

```
data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;

# perform fcros()
af <- fcros(fdata, cont, test, log2.opt);

# Volcano plot
fvalVolcanoPlot(af, thr = 0.01);
```

---

histoPlot

*Histogram plot function of the fcros package results*

---

**Description**

This function allows to have a histogram plot. It uses the statistics "ri" or "u1" obtained using one of the following functions: `fcros()`, `fcros2()`, `fcrosMod()`, `pfco()` or `pfcoMod()`.

**Usage**

```
histoPlot(af, nbins = 50)
```



**Arguments**

af	This is an object obtained using the function fcros(), fcros2(), fcrosMod(), pfco() or pfcoMod()
nbins	This parameter is used for the number of bins in the histogram. Default setting is 50: nbins = 50

**Value**

This function plots a histogram on the screen.

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC Bioinformatics, 2014, 15:14

**Examples**

```
data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;

# perform fcros() and pfco()
af <- fcros(fdata, cont, test, log2.opt);
af2 <- pfco(fdata, cont, test, log2.opt);

# Histogram plots
op <- par(mfrow = c(1,2))
  histoPlot(af);
  histoPlot(af2);
par(op);
```

---

moyStdCalc

*Calculation of the mean and the standard deviation of the full or reduced matrix of sorted ranks*

---

**Description**

This is an internal function for using a C code in the calculation of the mean and the standard deviation of the full or reduced matrix with sorted rank values. The calculations are performed for each row.

**Usage**

```
moyStdCalc(rvect, n, m)
```

**Arguments**

<code>rvect</code>	Vector containing the full or reduced matrix with sorted rank values
<code>n</code>	Number of genes or probes in the dataset
<code>m</code>	Number of columns of the full or reduced matrix of sorted rank values

**Author(s)**

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# data(fdata);
```

---

pfco	<i>Searching for differentially expressed genes/probes using an approach based on the Perron-Frobenius theorem</i>
------	--

---

**Description**

Implementation of a method based on fold change and the Perron theorem for detecting differentially expressed genes in a dataset. This function should be used with two biological conditions dataset (microarray or RNA-seq, ...). Using pairwise combinations of samples from the two biological conditions, fold changes (FC) are calculated. For each combination, the FC obtained are sorted in increasing order and corresponding rank values are associated with genes. Then, a statistic is assigned to the robust average ordered rank values for each gene/probe.

**Usage**

```
pfco(xdata, cont, test, log2.opt = 0, trim.opt = 0.25)
```

**Arguments**

<code>xdata</code>	A matrix or a table containing two biological conditions dataset to process for detecting differentially expressed genes: <code>xdata</code> . The first column of the matrix " <code>xdata</code> " should contain the gene IDs or their names.
<code>cont</code>	A vector containing the label names of the control samples: <code>cont = c("cont01", "cont02", ...)</code> .
<code>test</code>	A vector containing the label names of the test samples: <code>test = c("test01", "test02", "test03", ...)</code> .

log2.opt	A scalar equals to 0 or 1. The value 0 (default) means that data in the matrix "xdata" are expressed in a log2 scale: log2.opt = 0
trim.opt	A scalar between 0 and 0.5. The value 0.25 (default) means that 25% of the lower and the upper rank values of each gene are not used for computing its statistics "ri", i.e. the inter-quartile range rank values are averaged: trim.opt = 0.25

### Details

Label names appearing in the parameter "samp" should match with some label names in the columns of the data matrix "xdata". It is not necessary to use all label names appearing in the columns of the dataset matrix.

### Value

This function returns a data frame containing 9 components

idnames	A vector containing the list of IDs or symbols associated with genes
u1	The average of rank values associated with genes. These values are rank values statistics leading to f-values and p-values.
FC	The fold changes for genes in the dataset. These fold changes are calculated as a ratio of averages from the test and the control samples. Non log scale values are used in the calculation.
FC2	The robust fold changes for genes. These fold changes are calculated as a trimmed mean of the fold changes or ratios obtained from the dataset samples. Non log scale values are used in the calculation.
f.value	The f-values are probabilities associated with genes using the "mean" and the "standard deviation" ("sd") of the statistics "ri". The "mean" and "sd" are used as a normal distribution parameters.
p.value	The p-values associated with genes. These values are obtained using a one sample Student t-test on the fold change rank values.
comp	Singular values.
comp.w	Singular values weights.
comp.wcum	Cumulative sum of the singular values weights.

### Author(s)

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

### References

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```

data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;
trim.opt <- 0.25;

# perform pfco()
af <- pfco(fdata, cont, test, log2.opt, trim.opt);

# now select top 20 down and/or up regulated genes
top20 <- fcrosTopN(af, 20);
alpha1 <- top20$alpha[1];
alpha2 <- top20$alpha[2];
id.down <- matrix(0, 1);
id.up <- matrix(0, 1);
n <- length(af$FC);
f.value <- af$f.value;

idown <- 1;
iup <- 1;
for (i in 1:n) {
  if (f.value[i] <= alpha1) { id.down[idown] <- i; idown <- idown + 1; }
  if (f.value[i] >= alpha2) { id.up[iup] <- i; iup <- iup + 1; }
}

data.down <- fdata[id.down[1:(idown-1)], ];
ndown <- nrow(data.down);
data.up <- fdata[id.up[1:(iup-1)], ];
nup <- nrow(data.up);

# now plot down regulated genes
t <- 1:20;
op = par(mfrow = c(2,1));
plot(t, data.down[1,2:21], type = "l", col = "blue", xlim = c(1,20),
      ylim = c(0,18), main = "Top down-regulated genes");
for (i in 2:ndown) {
  lines(t,data.down[i,2:21], type = "l", col = "blue")
}

# now plot down and up regulated genes
plot(t, data.up[1,2:21], type = "l", col = "red", xlim = c(1,20),
      ylim = c(0,18), main = "Top up-regulated genes");
for (i in 2:nup) {
  lines(t, data.up[i,2:21], type = "l", col = "red")
}
par(op)

```

---

pfcoMod	<i>Searching for differentially expressed genes or detecting recurrent copy number aberration probes using an approach based on the Perron-Frobenius theorem</i>
---------	--

---

## Description

Implementation of a method based on fold change rank and the Perron theorem to search for differentially expressed genes or to detect chromosomal recurrent copy number aberration probes. This function should be used with a matrix of fold changes or ratios from biological dataset (microarray, RNA-seq, ...). The function pfcoMod() is an extension of the function pfco() to a dataset which does not contain replicate samples or to a dataset with one biological condition dataset. Statistics are associated with genes/probes to characterize their change levels.

## Usage

```
pfcoMod(fcMat, samp, log2.opt = 0, trim.opt = 0.25)
```

## Arguments

fcMat	A matrix containing fold changes or ratios from a biological dataset to process for searching differentially expressed genes or for detecting recurrent copy number aberrations regions: fcMat. The first column of the matrix "fcMat" should contain the gene IDs or symbols.
samp	A vector of sample label names which should appear in the columns of the matrix fcMat: samp.
log2.opt	A scalar equals to 0 or 1. The value 0 (default) means that values in the matrix "fcMat" are expressed in a log2 scale: log2.opt = 0
trim.opt	A scalar between 0 and 0.5. The value 0.25 (default) means that 25% of the lower and the upper rank values for each gene are not used for computing the statistic "ri", i.e. the inter-quartile range rank values are averaged: trim.opt = 0.25

## Details

The label names appearing in the parameter "samp" should match some label names of the columns in the data matrix "xdata". It is not necessary to use all label names appearing in the columns of the dataset matrix.

## Value

This function returns a data frame containing 8 components

idnames	A vector containing the list of IDs or symbols associated with genes
u1	The average of ordered rank values associated with genes in the dataset. These values are rank statistics leading to the f-values and the p-values.

FC2	The robust fold changes for genes in matrix "fcMat". These fold changes are calculated as a trimmed mean of the values in "fcMat". Non log scale values are used in this calculation.
f.value	The f-values are probabilities associated with genes using the "mean" and the "standard deviation" ("sd") of values in "ri". The "mean" and "sd" are used as a normal distribution parameters.
p.value	The p-values associated with genes. The p-values are obtained from the fold change ranks using a one sample t-test.
comp	Singular values.
comp.w	Singular values weights.
comp.wcum	Cumulative sum of the singular values weights.

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```

data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;
trim.opt <- 0.25;

# perform pfcoMod()
fc <- fcrosFCmat(fdata, cont, test, log2.opt, trim.opt);
m <- ncol(fc$fcMat)
samp <- paste("Col", as.character(1:m), sep = "");
fc.val <- cbind(fdata[,1], data.frame(fc$fcMat))
colnames(fc.val) <- c("ID", samp)

af <- pfcoMod(fc.val, samp, log2.opt, trim.opt);

# now select top 20 down and/or up regulated genes
top20 <- fcrosTopN(af, 20);
alpha1 <- top20$alpha[1];
alpha2 <- top20$alpha[2];
id.down <- matrix(0,1);
id.up <- matrix(0,1);
n <- length(af$FC);
f.value <- af$f.value;

idown <- 1;
iup <- 1;
for (i in 1:n) {

```

```

    if (f.value[i] <= alpha1) { id.down[idown] <- i; idown <- idown + 1; }
    if (f.value[i] >= alpha2) { id.up[iup] <- i; iup <- iup + 1; }
  }

  data.down <- fdata[id.down[1:(idown-1)], ];
  ndown <- nrow(data.down);
  data.up <- fdata[id.up[1:(iup-1)], ];
  nup <- nrow(data.up);

  # now plot down regulated genes
  t <- 1:20;
  op = par(mfrow = c(2,1));
  plot(t, data.down[1, 2:21], type = "l", col = "blue", xlim = c(1,20),
       ylim = c(0,18), main = "Top down-regulated genes");
  for (i in 2:ndown) {
    lines(t,data.down[i, 2:21], type = "l", col = "blue")
  }

  # now plot down and up regulated genes
  plot(t, data.up[1,2:21], type = "l", col = "red", xlim = c(1,20),
       ylim = c(0,18), main = "Top up-regulated genes");
  for (i in 2:nup) {
    lines(t, data.up[i,2:21], type = "l", col = "red")
  }
  par(op)

```

---

pvalTopN

*Search for the top N changed genes or probes using p-values*

---

## Description

This function allows to search for the top N differentially expressed genes or changed probes. It uses the p-value. These p-values can be obtained using one of the following functions `fcros()`, `fcros2()`, `fcrosMod()`, `pfco()`, `pfcoMod()` or another statistical method.

## Usage

```
pvalTopN(pval, topN)
```

## Arguments

pval	This is vector with p-values obtained using the functions <code>fcros()</code> , <code>fcros2()</code> , <code>fcrosMod()</code> , <code>pfco()</code> , <code>pfcoMod()</code> or another statistical method: <code>pval = af\$p.value</code>
topN	The expected number of the top DE genes/probes in the dataset used: <code>topN</code>

**Value**

This function returns a data frame containing 2 components

alpha	Two threshold values for the down- and the up-regulated allowing to have the top N DE genes
index	The indexes of the top N DE genes

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC Bioinformatics, 2014, 15:14

**Examples**

```
data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;

# perform fcros()
af <- fcros(fdata, cont, test, log2.opt);

# now select top 10 down and/or up regulated genes
top10 <- pvalTopN(af$p.value, 12);

# display thresholds
top10$alpha

# display index of top10 genes
fdata[top10$index, 1]

# display fvalue of the top10 genes
(af$f.value)[top10$index]
```

---

pvalVolcanoPlot

*Performs a volcano plot of the FCROS/PFCO statistics*

---

**Description**

This function allows to have a volcano like plot using the output results of the function fcros(), fcros2(), fcrosMod(), pfco() or pfcomod(): p-values versus robust fold changes (FC2). The p-value are transformed using  $-\log_{10}()$ , while FC2 are transformed using  $\log_2()$ .



**Usage**

```
pvalVolcanoPlot(af, thr = 0.05)
```

**Arguments**

af This is an object obtained using the functions `fcros()`, `fcros2()`, `fcrosMod()`, `pfco()` or `pfcoMod()`: `af = fcros(xdata, cont, test)`

thr The threshold to obtain the DE genes in the dataset (red plots): `thr = 0.05`

**Value**

This function displays on the screen a volcano like plot using the p-values and the robust fold changes (FC2).

**Author(s)**

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

**References**

Dembele D and Kastner P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC bioinformatics, 2014, 15:14

**Examples**

```
data(fdata);

cont <- c("cont01", "cont07", "cont03", "cont04", "cont08");
test <- c("test01", "test02", "test08", "test09", "test05");
log2.opt <- 0;

# perform fcros()
af <- fcros(fdata, cont, test, log2.opt);

# Volcano plot
pvalVolcanoPlot(af, thr = 1e-6);
```

---

rankReads

*Use of rank ordering statistics for count reads data*

---

**Description**

Implementation of a method based on fold change rank ordering statistics for detecting differentially expressed genes in a count reads data. This function is used with two biological conditions RNA-seq dataset. Using pairwise combinations of samples from the two biological conditions, fold changes (FC) are calculated. For each combination, the FC obtained are sorted in increasing order and corresponding rank values are associated with genes. Then, a statistic is assigned to the robust average ordered rank values for each gene/probe. Uniform small values are added to reads to overcome ties. A score is computed to localize genes with almost zero count reads for all samples.

**Usage**

```
rankReads(xdata, cont, test, nrun=10, err=0.01, trim.opt=0.25,
          rseed=57)
```

**Arguments**

xdata	A matrix or a table containing two biological conditions dataset (count reads) to process for detecting differentially expressed genes: xdata. The first column of the matrix "xdata" should contain the gene IDs or their names.
cont	A vector containing the label names of the control samples: cont = c("cont01", "cont02", ...).
test	A vector containing the label names of the test samples: test = c("test01", "test02", "test03", ...).
nrun	An integer used to associate a score with each gene, that allows to localize genes with almost zero count value for all samples. The value 20 (default) means that the dataset is used 20 times and small uniform values are added at each time: nrun = 20
err	This is the amount of the uniform values to add to count values. The value 0.01 (default) is used: err = 0.01
trim.opt	A scalar between 0 and 0.5. The value 0.25 (default) means that 25% of the lower and the upper rank values of each gene are not used for computing its statistics "ri", i.e. the inter-quartile range rank values are averaged: trim.opt = 0.25
rseed	This value allow to set the computer random generation seed value in order to be able to have the same results for runs performed at different times: rseed = 57

**Details**

Label names appearing in the parameters "cont" and "test" should match with some label names in the columns of the data matrix "xdata". It is not necessary to use all label names appearing in the columns of the dataset matrix.

**Value**

This function returns a data frame containing 10 components

idnames	A vector containing the list of IDs or symbols associated with genes
ri	The average of rank values associated with genes. These values are rank values statistics leading to f-values and p-values.
FC	The fold changes for genes in the dataset. These fold changes are calculated as a ratio of averages from the test and the control samples. Non log scale values are used in the calculation.
FC2	The robust fold changes for genes. These fold changes are calculated as a trimmed mean of the fold changes or ratios obtained from the dataset samples. Non log scale values are used in the calculation.

f.value	The f-values are probabilities associated with genes using the "mean" and the "standard deviation" ("sd") of the statistics "ri". The "mean" and "sd" are used as a normal distribution parameters.
p.value	The p-values associated with genes. These values are obtained from the fold change rank values and one sample t-test.
bounds	Two values, which are the lower and the upper bounds or the minimum and the maximum values of the non standardized "ri".
params	Three values, which are the estimates for the parameters "delta" (average difference between consecutive ordered average of rank values) "mean" (mean value of "ri") and the standard deviation ("sd") of "ri".
params_t	Three values which are theoretical levels for parameters "delta", "mean" and "sd".
score	probability of the Fisher test on variances. Higher values are associated with genes with almost zero count reads for all samples.

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# data(fdata);
```

---

rmatCalc                      *Calculation of the FC matrix using pairwise comparisons*

---

**Description**

This is an internal function for using a C code to calculate a vector form of the matrix of fold changes using pairwise comparison of data samples.

**Usage**

```
rmatCalc(fvect, n, m1, m2)
```

**Arguments**

fvect	Two biological conditions dataset matrix
n	Number of genes or probes in the dataset
m1	Number of samples in the first biological condition
m2	Number of samples in the second biological condition

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D and Kaster P, Fold change rank ordering statistics: a new method for detecting differentially expressed genes, BMC bioinformatics, 2014, 15:14

**Examples**

```
# data(fdata);
```

---

rmatTrim

*Calculation of the reduced matrix containing sorted rank values*

---

**Description**

This is an internal function for using a C code for calculating the reduced matrix of sorted rank values.

**Usage**

```
rmatTrim(rvect, n, m, idx, m2)
```

**Arguments**

rvect	Vector containing the full matrix with sorted rank values
n	Number of genes or probes in the dataset
m	Number of columns of the full matrix of sorted rank values
idx	Indexes of the columns to keep in the reduced matrix
m2	Number of columns of the reduced matrix

**Author(s)**

Doulaye Dembele doulaye@igbmc.fr

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# data(fdata);
```

---

tcnReads	<i>Performs a total count normalization of reads</i>
----------	--

---

**Description**

Given a data table `x` with count reads, one column for each sample, this function adjust values in such a way they become comparable between samples. User can specify the maximum value for total count reads in each sample.

**Usage**

```
tcnReads(x, maxVal=0)
```

**Arguments**

<code>x</code>	Data table with count reads, one column for one sample
<code>maxVal</code>	Target total number reads for each sample. The value 0 (default) means that the median total of samples is used: <code>maxVal = 0</code>

**Value**

This function returns a data table of the same size as input

<code>x2</code>	Data table of the same size as <code>x</code>
-----------------	---

**Author(s)**

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# data(fdata);
```

---

tprobaCalc                      *Calculation of the Student one sample test probabilities*

---

**Description**

This is an internal function for using a C code for performing a Student one sample test for each row of the full or reduced matrix with sorted rank values.

**Usage**

```
tprobaCalc(moy, std, n, dl, em)
```

**Arguments**

moy	Vector containing average of rank values for rows
std	Vector containing standard deviation of rank values for rows
n	Number of genes or probes in the dataset
dl	Degree of freedom in the test. This is equal to the number of the columns in the full or reduced matrix of sorted rank values minus one
em	Expected average rank values for each row

**Author(s)**

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

**References**

Dembele D, Analysis of high biological data using their rank values, Stat Methods Med Res, 2018

**Examples**

```
# data(fdata);
```

---

voomReads                      *Transformation of read count values*

---

**Description**

This function allows to transform the count values associated with Sequencing reads. The purpose of this transformation is to allow applying normal-based microarray-like statistical methods to RNA-seq read counts. Log2 values are returned.

**Usage**

```
voomReads(x, Rm=1e+06)
```

**Arguments**

x	This is a read counts matrix
Rm	A constant used in the transformation

**Value**

This function returns a data matrix of the same size as input matrix x. The values of this matrix are expressed in log2 scale.

**Author(s)**

Doulaye Dembele [doulaye@igbmc.fr](mailto:doulaye@igbmc.fr)

**References**

Charity W Law, Yunshun Chen, Wei Shi and Gordon K Smyth, voom: precision weights unlock linear model analysis tools for RNA-seq read counts, *Genome Biology*, 2014, 15R29

**Examples**

```
# data(fdata);
```

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