

# Package ‘CorrectedFDR’

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**Type** Package

**Title** Correcting False Discovery Rates

**Version** 1.1

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## Description

There are many estimators of false discovery rate. In this package we compute the Nonlocal False Discovery Rate (NFDR) and the estimators of local false discovery rate: Corrected False discovery Rate (CFDR), Re-ranked False Discovery rate (RFDR) and the blended estimator.  
Bickel, D.R., Rahal, A. (2019) <<https://tinyurl.com/kkdc9rk8>>.

**Depends** R(>= 2.14.2)

**Suggests** LFDR.MLE, LFDREmpiricalBayes, ProData

**biocViews** Bayesian Statistics, MathematicalBiology

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CorrectedFDR-package *Correcting False Discovery Rates*

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## Description

There are many estimators of false discovery rate. In this package we compute the Nonlocal False Discovery Rate (NFDR) and the estimators of local false discovery rate: Corrected False discovery Rate (CFDR), Re-ranked False Discovery rate (RFDR) and the blended estimator. Bickel, D.R., Rahal, A. (2019) <<https://tinyurl.com/kkdc9rk8>>.

## Details

The DESCRIPTION file:

```
Package: CorrectedFDR
Type: Package
Version: 1.1
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License: GPL-3
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Suggests: LFDR.MLE, LFDREmpiricalBayes, ProData
```

Two functions in CorrectedFDR package to compute the LFDR estimators. The function `EstimatorsFDR` computes the nonlocal false discovery rate (NFDR), the CFDR and the RFDR. The function `BlendedLFDR` uses a Benchmark of FDR, and other estimators of LFDR in order to get an estimate of LFDR.

## Author(s)

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## References

Bickel, D.R., Rahal, A. (2019). Correcting false discovery rates for their bias toward false positives. *Communications in Statistics - Simulation and Computation*, <https://tinyurl.com/kkdc9rk8>.

Bickel, D. R. (2015). Corrigendum to: Simple estimators of false discovery rates given as few as one or two p-values without strong parametric assumptions. *Statistical Applications in Genetics and Molecular Biology*, 2015, 14, 225.

Bickel, D. R. (2015). Blending Bayesian and frequentist methods according to the precision of prior information with applications to hypothesis testing. *Statistical Methods and Applications*, 24(4), pp. 523-546.

Bickel, D. R. (2013). Simple estimators of false discovery rates given as few as one or two p-values without strong parametric assumptions. *Statistical Applications in Genetics and Molecular Biology*, 2013, 12, 529-543.

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BlendedLFDR

*Blended Estimator of Local False Discovery Rate (LFDR)*

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### Description

BlendedLFDR is a function used to compute the blended estimator based on a benchmark estimator, usually the nonlocal false discovery rate (NFDR), and a set of estimators of local false discovery rates (LFDR).

### Usage

```
BlendedLFDR(Benchmark, EstLFDR)
```

### Arguments

Benchmark	Input numeric vector for benchmark estimator (often NFDR).
EstLFDR	Input a matrix containing two or more sets of LFDR estimators.

### Details

Benchmark is an estimator of the FDR. This is usually the nonlocal false discovery rate (NFDR). EstLFDR is a matrix of several LFDR estimators such as corrected FDR (CFDR), re-ranked FDR (RFDR), MLE (Maximum Likelihood Estimator), BBE1 (Binomial Based Estimator), etc. The output returns a single numeric vector containing the blended estimator of the LFDR.

### Value

The value of the blended estimator is an estimator of the LFDR.

### Note

The number of rows for the Benchmark and EstLFDR must have equal lengths.

### Author(s)

Code: Abbas Rahal.  
Documentation: Anna Akpawu, Justin Chitpin and Abbas Rahal.  
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### References

Bickel, D. R. (2015). Blending Bayesian and frequentist methods according to the precision of prior information with applications to hypothesis testing. *Statistical Methods and Applications*, 24(4), pp. 523-546.

**Examples**

```

#The data used to compute the LFDR estimators (CFDR, RFDR, MLE, and BBE1)
#comes from the ER/PR breast cancer data from the "ProData" package.
#To read more about the data, visit the website: https://www.bioconductor.org/
#Test statistics were first obtain, then the estimators for the FDR and LFDR were estimated.
#Benchmark vector
NFDR<-c(0.5661106448, 0.6897735492, 0.0000288516, 0.1549745113, 0.1305508970, 0.2421032979,
0.1482335568, 1, 1, 1, 0.6602562820, 0.7034682859, 0.7036332234, 0.0071192090,
0.8204536037, 0.9757716498, 0.7379329991, 1, 0.6333245479, 0.9904389701)
#Estimators of LFDR
CFDR<- c(1, 1, 0.0000288516, 0.2841199373, 0.2980912149, 0.5931530799, 0.3088199101,
1, 1, 1, 1, 1, 0.0106788135, 1, 1, 1, 1, 1)

RFDR<- c(0.689773549, 1, 0.007119209, 0.130550897, 0.703633223, 0.660256282, 0.242103298,
1, 1, 1, 0.820453604, 1, 0.703468286, 0.154974511, 1, 1, 1, 1, 0.975771650,1)

MLE<- c(0.9865479126, 0.9969935995, 0.0002372158, 0.6531633437, 0.7611453549, 0.9187425383,
0.7359259207, 0.9996548155, 0.9997310453, 0.9997437131, 0.9944712582, 0.9981685029,
0.9937604664, 0.0215892618, 0.9990504315, 0.9997493086, 0.9967673540, 0.9997016985,
0.9970142319, 0.9997625673)

BBE1<- c(1,1, 0.0003169812, 0.1138333734, 1, 1, 1, 1, 1, 0.3279109564, 1, 0.0504755806,
0.0091823115, 0.0182614994, 0.0165386682, 1, 0.6964403713, 0.1001337298, 0.8415641198 )

#Matrix of LFDR Estimators
Est.LFDR<- matrix(c(CFDR,RFDR,MLE,BBE1), ncol=4)
output<-BlendedLFDR(Benchmark = NFDR, EstLFDR = Est.LFDR)
output$Blended

```

---

EstimatorsFDR

*Estimators of Local False Discovery Rate (LFDR)*


---

**Description**

EstimatorsFDR is an R function that computes the Nonlocal False Discovery Rate (NFDR) and the estimators of local false discovery rate: Corrected False discovery Rate (CFDR) and Re-ranked False Discovery rate (RFDR).

**Usage**

```
EstimatorsFDR(pvalue)
```

**Arguments**

pvalue            Input numeric vector of pvalues.

## Details

The input is a list of pvalues. The pvalues can be obtained for example by performing Student's t-test between two datasets. The two groups can be data from healthy and disease states. Let  $i = 1, 2, \dots, N$ , where  $i$  represents the  $i$ th feature (SNP or gene, for example). Then, for each  $i$ , the hypothesis indicator  $A_i$  can have two possible values.

$A_i = 0$ , if the  $i$ th null hypothesis is true, or

$A_i = 1$ , if the  $i$ th null hypothesis is not true,

where the null hypothesis is defined by: the  $i$ th feature is unaffected by a treatment, unassociated with a disease, etc. The values for each estimator (NFDR, CFDR, RFDR) indicate the probability that the null hypothesis of the  $i$ th feature is true ( $A_i = 0$ ) given the statistics  $T_i$ . The alternative hypothesis is true if  $A_i = 1$ . For example, in gene expression data analysis, if the null hypothesis is true, this would mean that the genes are not differentially expressed.

## Value

The output returns three lists. It returns the NFDR, CFDR, and RFDR estimators:

NFDR	nonlocal FDR
CFDR	corrected FDR
RFDR	re-ranked FDR

## Author(s)

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## References

Bickel, D.R., Rahal, A. (2019). Correcting false discovery rates for their bias toward false positives. Communications in Statistics - Simulation and Computation, <https://tinyurl.com/kkdc9rk8>.

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Bickel, D. R. (2013). Simple estimators of false discovery rates given as few as one or two p-values without strong parametric assumptions. Statistical Applications in Genetics and Molecular Biology, 2013, 12, 529-543.

## Examples

```
#The examples below are from the "ProData" package.  
#In order to use the "Prodata" input you would first need to install the ProData package.  
#You will also need the function exprs in this package.  
#First, make sure that the ProData package is properly installed:  
#source("https://bioconductor.org/biocLite.R")  
#biocLite("ProData")  
#library(ProData)  
#data("f45cbmk")
```

```
#q1<- quantile(as(exprs(f45cbmk[, pData(f45cbmk)$GROUP == "B"]), "numeric"), probs = 0.25)
#logish<- function(x){log(x + q1)}
#Vectors of proteins for 20 patients ER/PR-positive and Healthy
#Y<- logish(exprs(f45cbmk[, pData(f45cbmk)$GROUP == "B"])) # Control (Healthy)
#X.ER<- logish(exprs(f45cbmk[, pData(f45cbmk)$GROUP == "C"])) # Case ER/PR-positive
#pvalue<- NULL
#for (i in 1:nrow(X.ER))
#{
# t<-t.test(x=X.ER[i,], y=Y[i,], alternative = "two.sided")
# pvalue[i]<- t$p.value
#}

#The pvalues obtained from the t-test:
pvalue<- c(0.1981, 0.3794, 0.000001443, 0.02325, 0.03264, 0.07263, 0.02965, 0.8016, 0.8888,
          0.9133, 0.2971, 0.4573, 0.2815, 0.0007119, 0.5743, 0.927, 0.369, 0.8478, 0.38, 0.9904)
output<- EstimatorsFDR(pvalue)
#Three lists
output$NFDR
output$CFDR
output$RFDR
```

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