

Package: MLFDR (via r-universe)

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Type Package

Title High Dimensional Mediation Analysis using Local False Discovery Rates

Version 0.1.0

Description Implements a high dimensional mediation analysis algorithm using Local False Discovery Rates. The methodology is described in Roy and Zhang (2024) <[doi:10.48550/arXiv.2402.13933](https://doi.org/10.48550/arXiv.2402.13933)>.

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Encoding UTF-8

LazyData true

Imports NMOF, stats

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Repository <https://asmita112358.r-universe.dev>

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Contents

localFDR	2
MLFDR	4
Index	6

localFDR	<i>localFDR: A function that fits a Gaussian Mixture model to the mediation coefficients and returns the localFDR estimated from the mixture model</i>
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Description

localFDR: A function that fits a Gaussian Mixture model to the mediation coefficients and returns the localFDR estimated from the mixture model

Usage

```
localFDR(
  alpha,
  beta,
  var_alpha,
  var_beta,
  lambda.init = NULL,
  kappa.init = 1,
  psi.init = 1,
  psi_int = NULL,
  kappa_int = NULL,
  twostep = FALSE,
  k = 4,
  d1 = NULL,
  d2 = NULL,
  eps = 0.01,
  verbose = TRUE,
  method = "unicore"
)
```

Arguments

alpha	a vector of estimated alpha coefficients from the first equation of mediation analysis
beta	a vector of estimated beta coefficients from the second equation of mediation analysis
var_alpha	a vector of estimated variances for alpha coefficients
var_beta	a vector of estimated variances for beta coefficients
lambda.init	initial values of the proportion of mixture, must sum to 1
kappa.init	initial value of kappa, the variance of the prior of alpha under the alternative
psi.init	initial value of psi, the variance of the prior of beta under the alternative
psi_int	gridSearch interval for psi, to be used in optimize function
kappa_int	gridSearch interval for kappa, to be used in optimize function

twostep	logical, whether to use two-step MLFDR
k	number of mixture components, default is 4. Used in one-step MLFDR.
d1	number of non-null components for alpha in two-step MLFDR.
d2	number of non-null components for beta in two-step MLFDR.
eps	stopping criteria for EM algorithm
verbose	logical, whether to print the log-likelihood at each iteration. Default is TRUE.
method	either "unicore" or "multicore", specifies whether parallelization should be used when estimating variances in two step EM. Is not used unless twostep = TRUE

Value

A list consisting of two elements - lfd, a vector of local false discovery rates and pi, the estimated mixture proportions

Examples

```

n = 100
m = 1000
pi = c(0.4, 0.1, 0.3, 0.2)
X = rbinom(n, 1, 0.1)
M = matrix(nrow = m, ncol = n)
Y = matrix(nrow = m, ncol = n)
gamma = sample(1:4, m, replace = TRUE, prob = pi)
alpha = vector()
beta = vector()
vec1 = rnorm(m, 0.05, 1)
vec2 = rnorm(m, -0.5, 2)
alpha = ((gamma==2) + (gamma == 4))*vec1
beta = ((gamma ==3) + (gamma == 4))*vec2
alpha_hat = vector()
beta_hat = vector()
var_alpha = c()
var_beta = c()
p1 = vector()
p2 = vector()
for(i in 1:m)
{
  M[i,] = alpha[i]*X + rnorm(n)
  Y[i,] = beta[i]*M[i,] + rnorm(1,0.5)*X + rnorm(n)
  obj1 = lm(M[i,] ~ X)
  obj2 = lm(Y[i,] ~ M[i,] + X)
  table1 = coef(summary(obj1))
  table2 = coef(summary(obj2))
  alpha_hat[i] = table1["X",1]
  beta_hat[i] = table2["M[i, ]",1]
  p1[i] = table1["X",4]
  p2[i] = table2["M[i, ]",4]
  var_alpha[i] = table1["X",2]^2
  var_beta[i] = table2["M[i, ]",2]^2
}
lfd <- localFDR(alpha_hat, beta_hat, var_alpha, var_beta, twostep = FALSE)

```

MLFDR

*MLFDR: Mediation analysis using localFDR***Description**

MLFDR: Mediation analysis using localFDR

Usage

```
MLFDR(lfdr, size = 0.05)
```

Arguments

`lfdr` a vector of local false discovery rates, as output from the function `localFDR`
`size` Target False Discovery rate, default is 0.05

Value

A vector of logicals indicating which hypotheses are rejected

Examples

```
n = 100
m = 1000
pi = c(0.4, 0.1, 0.3, 0.2)
X = rbinom(n, 1, 0.1)
M = matrix(nrow = m, ncol = n)
Y = matrix(nrow = m, ncol = n)
gamma = sample(1:4, m, replace = TRUE, prob = pi)
alpha = vector()
beta = vector()
vec1 = rnorm(m, 0.05, 1)
vec2 = rnorm(m, -0.5, 2)
alpha = ((gamma==2) + (gamma == 4))*vec1
beta = ((gamma ==3) + (gamma == 4))*vec2
alpha_hat = vector()
beta_hat = vector()
var_alpha = c()
var_beta = c()
p1 = vector()
p2 = vector()
for(i in 1:m)
{
  M[i,] = alpha[i]*X + rnorm(n)
  Y[i,] = beta[i]*M[i,] + rnorm(1,0.5)*X + rnorm(n)
  obj1 = lm(M[i,] ~ X)
  obj2 = lm(Y[i,] ~ M[i,] + X)
  table1 = coef(summary(obj1))
  table2 = coef(summary(obj2))
}
```

```
alpha_hat[i] = table1["X",1]
beta_hat[i] = table2["M[i, ]",1]
p1[i] = table1["X",4]
p2[i] = table2["M[i, ]",4]
var_alpha[i] = table1["X",2]^2
var_beta[i] = table2["M[i, ]",2]^2
}
lfdr <- localFDR(alpha_hat, beta_hat, var_alpha, var_beta, twostep = FALSE)
print(lfdr$pi)
MLFDR(lfdr$lfdr, size = 0.05)
```

Index

localFDR, [2](#)

MLFDR, [4](#)