Package 'kerDAA'

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| Type Package |
|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Title New Kernel-Based Test for Differential Association Analysis |
| Version 0.1.1 |
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| Imports mvtnorm |
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| Description A new practical method to evaluate whether relationships between two sets of high- dimensional variables are different or not across two conditions. Song, H. and Wu, M.C. (2023) <arxiv:2307.15268>.</arxiv:2307.15268> |
| License GPL (≥ 2) |
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| |

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R topics documented:

| kerDAA | | 1 |
|--------|------|---|
| kerdaa | | 3 |

Index

```
kerDAA
```

New kernel-based test for differential association analysis

5

Description

This package can be used to determine whether two high-dimensional samples have similar dependence relationships across two conditions.

Author(s)

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References

Song, H. and Wu, M.C. (2023). Multivariate differential association analysis. arXiv:2307.15268

See Also

kerdaa

Examples

```
# Dimension of variables.
d = 100
# The first covariance matrix
SIG = matrix(0, d, d)
for (i in 1:d) {
  for (j in 1:d) {
    SIG[i,j] = 0.4^{(abs(i-j))}
  }
}
# The second covariance matrix
SIG1 = matrix(0, d, d)
for (i in 1:d) {
  for (j in 1:d) {
    SIG1[i,j] = (0.4+0.5)^(abs(i-j))
  }
}
set.seed(500)
# We use 'rmvnorm' in 'mvtnorm' package to generate multivariate normally distributed samples
require(mvtnorm)
Z = rmvnorm(100, mean = rep(0, 100), sigma = SIG)
X1 = Z[, 1:50]
Y1 = Z[,51:100]
Z = rmvnorm(100, mean = rep(0, 100), sigma = SIG1)
X2 = Z[, 1:50]
Y2 = Z[,51:100]
a = kerdaa(X1, Y1, X2, Y2, perm=1000)
# output results based on the permutation and the asymptotic results
# the test statistic values can be found in a$stat_g and a$stat_l
# p-values can be found in a$pval and a$pval_perm
```

kerdaa

Description

This function provides the kernel-based differential association test.

Usage

kerdaa(X1, Y1, X2, Y2, perm=0)

Arguments

| X1 | The first multivariate data in the first condition. |
|------|---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Y1 | The second multivariate data in the first condition. |
| X2 | The first multivariate data in the second condition. |
| Y2 | The second multivariate data in the second condition. |
| perm | The number of permutations performed to calculate the p-value of the test. The default value is 0, which means the permutation is not performed and only approximated p-value based on the asymptotic theory is provided. Doing permutation could be time consuming, so be cautious if you want to set this value to be larger than 10,000. |

Value

Returns a list with test statistic values and p-values of the test. See below for more details.

| stat_g | The value of the test statistic using the Gaussian kernel. |
|-----------|------------------------------------------------------------------------------------------------------------|
| stat_l | The value of the test statistic using the linear kernel. |
| pval | The omnibus p-value using the approximated p-values of the test statistic based on asymptotic theory. |
| pval_perm | The omnibus p-value using the permutation p-values of the test statistic when argument 'perm' is positive. |

See Also

kerDAA-package

Examples

```
# Dimension of variables.
d = 100
# The first covariance matrix
SIG = matrix(0, d, d)
```

```
kerdaa
```

```
for (i in 1:d) {
  for (j in 1:d) {
    SIG[i,j] = 0.4^(abs(i-j))
  }
}
# The second covariance matrix
SIG1 = matrix(0, d, d)
for (i in 1:d) {
  for (j in 1:d) {
    SIG1[i,j] = (0.4+0.5)^(abs(i-j))
  }
}
set.seed(500)
# We use 'rmvnorm' in 'mvtnorm' package to generate multivariate normally distributed samples
require(mvtnorm)
Z = rmvnorm(100, mean = rep(0, 100), sigma = SIG)
X1 = Z[, 1:50]
Y1 = Z[,51:100]
Z = rmvnorm(100, mean = rep(0,100), sigma = SIG1)
X2 = Z[, 1:50]
Y2 = Z[,51:100]
a = kerdaa(X1, Y1, X2, Y2, perm=1000)
# output results based on the permutation and the asymptotic results
# the test statistic values can be found in a$stat_g and a$stat_l
# p-values can be found in a$pval and a$pval_perm
```

4

Index

kerDAA, 1 kerdaa, 2, 3 kerDAA-package (kerDAA), 1