

Package ‘combinIT’

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Type Package

Title A Combined Interaction Test for Unreplicated Two-Way Tables

Version 2.0.0

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Description There are several non-functional-form-based interaction tests for testing interaction in unreplicated two-way layouts. However, no single test can detect all patterns of possible interaction and the tests are sensitive to a particular pattern of interaction. This package combines six non-functional-form-based interaction tests for testing additivity. These six tests were proposed by Boik (1993) <doi:10.1080/02664769300000004>, Piepho (1994), Kharrati-Kopaei and Sadooghi-Alvandi (2007) <doi:10.1080/03610920701386851>, Franck et al. (2013) <doi:10.1016/j.csda.2013.05.002>, Malik et al. (2016) <doi:10.1080/03610918.2013.870196> and Kharrati-Kopaei and Miller (2016) <doi:10.1080/00949655.2015.1057821>. The p-values of these six tests are combined by Bonferroni, Sidak, Jacobi polynomial expansion, and the Gaussian copula methods to provide researchers with a testing approach which leverages many existing methods to detect disparate forms of non-additivity. This package is based on the following published paper: Shenavari and Kharrati-Kopaei (2018) “A Method for Testing Additivity in Unreplicated Two-Way Layouts Based on Combining Multiple Interaction Tests”. In addition, several sentences in help files or descriptions were copied from that paper.

URL <https://github.com/haghbinh/combinIT>

BugReports <https://github.com/haghbinh/combinIT/issues>

License GPL (>= 2)

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Boik_test	<i>Boik's (1993) Locally Best Invariant (LBI) Test</i>
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Description

This function calculates the LBI test statistic for testing the null hypothesis H_0 : There is no interaction. It returns an exact p-value when $p = 2$ where $p = \min\{a - 1, b - 1\}$. It returns an exact Monte Carlo p-value when $p > 2$. It also provides an asymptotic chi-squared p-value. Note that the p-value of the Boik.test is always one when $p = 1$.

Usage

```
Boik_test(x, nsim = 10000, alpha = 0.05, report = TRUE)
```

Arguments

x	a numeric matrix, $a \times b$ data matrix where the number of row and column is corresponding to the number of factor levels.
nsim	a numeric value, the number of Monte Carlo samples for calculating an exact Monte Carlo p-value. The default value is 10000.
alpha	a numeric value, the level of the test. The default value is 0.05.
report	logical: if TRUE the result of the test is reported at the alpha level.

Details

The LBI test statistic is $T_{B93} = (tr(R'R))^2 / (ptr((R'R)^2))$ where $p = \min\{a - 1, b - 1\}$ and R is the residual matrix of the input data matrix, x , under the null hypothesis H_0 : There is no interaction. This test rejects the null hypothesis of no interaction when T_{B93} is small. Boik (1993) provided the exact distribution of T_{B93} when $p = 2$ under H_0 . In addition, he provided an asymptotic distribution of T_{B93} under H_0 when q tends to infinity where $q = \max\{a - 1, b - 1\}$. Note that the LBI test is powerful when the $a \times b$ matrix of interaction terms has small rank and one singular value dominates the remaining singular values or in practice, if the largest eigenvalue of RR' is expected to dominate the remaining eigenvalues.

Value

An object of the class ITtest, which is a list inducing following components:

pvalue_exact	An exact Monte Carlo p-value when $p > 2$. For $p = 2$, an exact p-value is calculated.
pvalue_appro	An chi-squared asymptotic p-value.
statistic	The value of test statistic.
Nsim	The number of Monte Carlo samples that are used to estimate p-value.
data_name	The name of the input dataset.
test	The name of the test.
Level	The level of test.
Result	The result of the test at the alpha level with some descriptions on the type of significant interaction.

References

Boik, R.J. (1993). Testing additivity in two-way classifications with no replications: the locally best invariant test. *Journal of Applied Statistics* 20(1): 41-55.

Shenavari, Z., Kharrati-Kopaei, M. (2018). A Method for Testing Additivity in Unreplicated Two-Way Layouts Based on Combining Multiple Interaction Tests. *International Statistical Review* 86(3): 469-487.

Examples

```
data(MVGH)
Boik_test(MVGH, nsim = 1000)
```

 CI_test

 Combined Interaction Test

Description

This function reports the p-values of the tests for non-additivity developed by Boik (1993), Piepho (1994), Kharrati-Kopaei and Sadooghi-Alvandi (2007), Franck et al. (2013), Malik et al. (2016) and Kharrati-Kopaei and Miller (2016). In addition, it combines the p-values of these six tests (and some other available p-values) into a single p-value as a test statistic for testing interaction. There are four combination methods: Bonferroni, Sidak, Jacobi expansion, and Gaussian Copula. The results of these four combined tests are also reported. If there is a significant interaction, the type of interaction is also provided.

Usage

```
CI_test(
  x,
  nsim = 10000,
  nc0 = 10000,
  opvalue = NULL,
  alpha = 0.05,
  report = TRUE,
  Elapsed_time = TRUE
)
```

Arguments

x	numeric matrix, $a \times b$ data matrix where the number of row and column is corresponding to the number of factor levels.
nsim	a numeric value, the number of Monte Carlo samples for computing an exact Monte Carlo p-value. The default value is 10000.
nc0	a numeric value, the number of Monte Carlo samples for computing the unbiased constant c_0 in KKM. test. The default value is 10000.
opvalue	a numeric vector, other p-values (in addition to the six considered p-values) that are going to be combined.
alpha	a numeric value, the level of the test. The default value is 0.05.
report	logical: if TRUE the result of the test is reported at the alpha level.
Elapsed_time	logical: if TRUE the progress will be printed in the console.

Details

The data matrix is divided based on the row of the data matrix for KKSA_test and Franck_test. Note that KKSA_test is not applicable when a is less than four. Franck_test and Piepho_test are not applicable when a is less than three. This function needs mvtnorm package.

Value

An object of the class `combttest`, which is a list inducing following components:

<code>nsim</code>	The number of Monte Carlo samples that are used to estimate p-value.
<code>Piepho_pvalue</code>	The p-value of Piepho's (1994) test.
<code>Piepho_Stat</code>	The value of Piepho's (1994) test statistic.
<code>Boik_pvalue</code>	The p-value of Boik's (1993) test.
<code>Boik_Stat</code>	The value of Boik's (1993) test statistic.
<code>Malik_pvalue</code>	The p-value of Malik's (2016) et al. test.
<code>Malik_Stat</code>	The value of Malik's (2016) et al. test statistic.
<code>KKM_pvalue</code>	The p-value of Kharrati-Kopaei and Miller's (2016) test.
<code>KKM_Stat</code>	The value of Kharrati-Kopaei and Miller's (2016) test statistic.
<code>KKSA_pvalue</code>	The p-value of Kharrati-Kopaei and Sadooghi-Alvandi's (2007) test.
<code>KKSA_Stat</code>	The value of Kharrati-Kopaei and Sadooghi-Alvandi's (2007) test statistic.
<code>Franck_pvalue</code>	The p-value of Franck's (2013) et al. test.
<code>Franck_Stat</code>	The value of Franck's (2013) et al. test statistic.
<code>Bonferroni</code>	The combined p-value by using the Bonferroni method.
<code>Sidak</code>	The combined p-value by using the Sidak method.
<code>Jacobi</code>	The combined p-value by using the Jacobi method.
<code>GC</code>	The combined p-value by using the Gaussian copula.
<code>data_name</code>	The name of the input dataset.
<code>test</code>	The name of the test.
<code>Level</code>	The level of test.
<code>Result</code>	The result of the combined test at the alpha level with some descriptions on the type of significant interaction.

References

Shenavari, Z., Kharrati-Kopaei, M. (2018). A Method for Testing Additivity in Unreplicated Two-Way Layouts Based on Combining Multiple Interaction Tests. *International Statistical Review* 86(3): 469-487.

Examples

```
data(CNV)
CI_test(CNV, nsim = 1000, Elapsed_time = FALSE)
```

CNV

Copy number variation (CNV).

Description

This data set are about copy number variation (CNV) between normal and tumor tissue samples among six dogs. In this data set, the value of CNV was measured as a signal intensity obtained from a comparative genomic hybridization (CGH) array, with higher signals corresponding to higher copy numbers; see Franck et al. (2013) and Franck and Osborne (2016). The data set was selected from 5899 sets (the full data have been made available as the supplementary material of the paper published by Franck et al. (2013)). The test of interaction between the dogs and tissues is of interest.

Format

A matrix with six rows (Dogs) and two columns (Tissues):

Row1 Dog1

Row2 Dog2

Row3 Dog3

Row4 Dog4

Row5 Dog5

Row6 Dog6

Column1 Normal tissue

Column2 Tumor

References

1. Franck, C., Nielsen, D., Osborne, J.A. (2013). A method for detecting hidden additivity in two-factor unreplicated experiments. *Computational Statistics and Data Analysis* 67:95-104.
2. Franck, C., Osborne, J.A. (2016). Exploring Interaction Effects in Two-Factor Studies using the hidden Package in R. *R Journal* 8 (1):159-172.

Franck_test

Franck's (2013) et al. test for Interaction

Description

This function calculates Franck's (2013) et al. test statistic, ACMIF, and corresponding p-value.

Usage

```

Franck_test(
  x,
  nsim = 10000,
  alpha = 0.05,
  report = TRUE,
  plot = FALSE,
  vecolor = c("blue", "red"),
  linetype = c(1, 2),
  Elapsed_time = TRUE
)

```

Arguments

<code>x</code>	numeric matrix, $a \times b$ data matrix where the number of row and column is corresponding to the number of factor levels.
<code>nsim</code>	a numeric value, the number of Monte Carlo samples for computing an exact Monte Carlo p-value. The default value is 10000.
<code>alpha</code>	a numeric value, the level of the test. The default value is 0.05.
<code>report</code>	logical: if TRUE the result of the test is reported at the alpha level.
<code>plot</code>	logical: if TRUE an interaction plot will be plotted.
<code>vecolor</code>	character vector of length two, for visualizing the colors of lines in interaction plot. The default colors are blue and red.
<code>linetype</code>	numeric vector of length two, for visualizing the line types in interaction plot. The default line types are 1 and 2.
<code>Elapsed_time</code>	logical: if TRUE the progress will be printed in the console.

Details

Franck et al. (2013) derived a test statistic based on the “hidden additivity” structure. They defined this structure as “the levels of one factor belong in two or more groups such that within each group the effects of the two factors are additive but the groups may interact with the ungrouped factor”. To detect hidden additivity, Franck et al. (2013) divided the table of data into two sub-tables (based on the rows of the data matrix) and an interaction F-test was developed. Then, they performed a search over all possible configurations of data and used the maximum of the interaction F-test as a test statistic. The hypothesis of no interaction is rejected when the maximum interaction F-test is large. If `plot` is TRUE an interaction plot will be plotted by displaying levels of column factor on the horizontal axis, levels of row factor using lines that are visually distinguished by line type and color, and the observed values on the vertical axis. Color and line type are used to display which levels of row factor are assigned to which groups based on the maximum F-values among all possible configurations. Note that the grouping colors and line types appear whether or not the `Franck.test` detects a significant non-additivity. The default colors are blue and red, and the default line types are one and two for the two groups. They can be customized by supplying arguments called `vecolor` and `linetype`. Note that the number of rows should be greater than two to perform the `Franck.test`. This test is powerful when there is a hidden additivity structure in the data set.

Value

An object of the class `ITtest`, which is a list inducing following components:

<code>pvalue_exact</code>	The calculated exact Monte Carlo p-value.
<code>pvalue_appro</code>	The Bonferroni-adjusted p-value is calculated.
<code>statistic</code>	The value of the test statistic.
<code>Nsim</code>	The number of Monte Carlo samples that are used to estimate p-value.
<code>data_name</code>	The name of the input dataset.
<code>test</code>	The name of the test.
<code>Level</code>	The level of test.
<code>Result</code>	The result of the test at the alpha level with some descriptions on the type of significant interaction.

References

Franck, C., Nielsen, D., Osborne, J.A. (2013). A method for detecting hidden additivity in two-factor unreplicated experiments. *Computational Statistics and Data Analysis* 67:95-104.

Franck, C., Osborne, J.A. (2016). Exploring Interaction Effects in Two-Factor Studies using the hidden Package in R. *R Journal* 8 (1):159-172.

Shenavari, Z., Kharrati-Kopaei, M. (2018). A Method for Testing Additivity in Unreplicated Two-Way Layouts Based on Combining Multiple Interaction Tests. *International Statistical Review* 86(3): 469-487.

Examples

```
data(CNV)
Franck_test(CNV, nsim = 1000, Elapsed_time = FALSE)
```

 IDCP

Impurity data in a chemical product (IDCP).

Description

This data were collected in an experiment to assess the impurity present in a chemical product. The impurity is affected by two factors: pressure and temperature. Montgomery (2001, p. 193) analyzed the data by using the Tukey single-degree-of-freedom test and concluded that there is no evidence of interaction.

Format

A matrix with five rows (Pressures) and three columns (Temperatures):

Row1 Pressure 25

Row2 Pressure 30

Row3 Pressure 35

Row4 Pressure 40

Row5 Pressure45

Column1 Temperature 100

Column2 Temperature 125

Column3 Temperature 150

References

1. Montgomery, D. C. (2001). Design and analysis of experiments, 5th Edition, p 193. John Wiley & Sons.

interaction_plot	<i>Interaction Plot</i>
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Description

Interaction Plot

Usage

```
interaction_plot(x, ...)
```

Arguments

x	numeric matrix, $a \times b$ data matrix where the number of row and column is corresponding to the number of factor levels.
...	plot parameters

Value

Plots an interaction plot for input.

Author(s)

Shenavari, Z.; Haghbin, H.; Kharrati-Kopaei, M.; Najibi, S.M.

Examples

```
## Not run: this is an example  
data(CNV)  
interaction_plot(CNV)
```

 KKM_test

Kharrati-Kopaei and Miller's (2016) Test for Interaction

Description

This function calculates the test statistic for testing H_0 : There is no interaction, and corresponding Monte Carlo p-value proposed by Kharrati-Kopaei and Miller (2016).

Usage

```
KKM_test(x, nsim = 1000, alpha = 0.05, report = TRUE, nc0 = 10000)
```

Arguments

x	a numeric matrix, $a \times b$ data matrix where the number of row and column is corresponding to the number of factor levels.
nsim	a numeric value, the number of Monte Carlo samples for computing an exact Monte Carlo p-value. The default value is 10000.
alpha	a numeric value, the level of the test. The default value is 0.05.
report	logical: if TRUE the result of the test is reported at the alpha level.
nc0	a numeric value, the number of Monte Carlo samples for computing the unbiased constant c_0 . The default value is 10000.

Details

Kharrati-Kopaei and Miller (2016) proposed a test statistic for testing interaction based on inspecting all pairwise interaction contrasts (PIC). This test depends on an unbiased constant c_0 that is calculated by a Monte Carlo simulation. In addition, the null distribution of the test statistic is calculated by a Monte Carlo simulation. This test is not applicable when both a and b are less than three. Note that this test procedure is powerful when significant interactions are caused by some data cells.

Value

An object of the class ITtest, which is a list inducing following components:

pvalue_exact	The calculated exact Monte Carlo p-value.
pvalue_appro	is not available for KKM_test.
Nsim	The number of Monte Carlo samples that are used to estimate p-value.
statistic	The value of the test statistic.
data_name	The name of the input dataset.
test	The name of the test.
Level	The level of test.
Result	The result of the test at the alpha level with some descriptions on the type of significant interaction.

References

Kharrati-Kopaei, M., Miller, A. (2016). A method for testing interaction in unreplicated two-way tables: using all pairwise interaction contrasts. *Statistical Computation and Simulation* 86(6):1203-1215.

Shenavari, Z., Kharrati-Kopaei, M. (2018). A Method for Testing Additivity in Unreplicated Two-Way Layouts Based on Combining Multiple Interaction Tests. *International Statistical Review* 86(3): 469-487.

Examples

```
data(RDWW)
KKM_test(RDWW, nsim = 1000, nc0 = 1000)
```

KKSA_test	<i>Kharrati-Kopaei and Sadooghi-Alvandi's (2007) test for interaction</i>
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Description

This function calculates Kharrati-Kopaei and Sadooghi-Alvandi's test statistic and corresponding p-value for testing interaction.

Usage

```
KKSA_test(
  x,
  nsim = 10000,
  alpha = 0.05,
  report = TRUE,
  plot = FALSE,
  vecolor = c("blue", "red"),
  linetype = c(1, 2),
  Elapsed_time = TRUE
)
```

Arguments

x	numeric matrix, $a \times b$ data matrix where the number of row and column is corresponding to the number of factor levels.
nsim	a numeric value, the number of Monte Carlo samples for computing an exact Monte Carlo p-value. The default value is 10000.
alpha	a numeric value, the level of the test. The default value is 0.05.
report	logical: if TRUE the result of the test is reported at the alpha level.
plot	logical: if TRUE an interaction plot will be plotted.
vecolor	character vector with length two, for visualizing the colors of lines in interaction plot. The default colors are blue and red.

linetype	numeric vector with length two, for visualizing the line types in interaction plot. The default line types are 1 and 2.
Elapsed_time	logical: if TRUE the progress will be printed in the console.

Details

Suppose that $a \geq b$ and $b \geq 4$. Consider the l -th division of the data table into two sub-tables, obtained by putting a_1 ($2 \leq a_1 \leq a - 2$) rows in the first sub-table and the remaining a_2 rows in the second sub-table ($a_1 + a_2 = a$). Let $RSS1$ and $RSS2$ denote the residual sum of squares for these two sub-tables, respectively. For a particular division l , let $F_l = \max\{F_l, 1/F_l\}$ where $F_l = (a_2 - 1)RSS1 / ((a_1 - 1)RSS2)$ and let P_l denote the corresponding p-value. Kharrati-Kopaei and Sadooghi-Alvandi (2007) proposed their test statistic as the minimum value of P_l over $l = 1, \dots, 2^{(a-1)} - a - 1$ all possible divisions of the table. If `plot` is TRUE an interaction plot will be plotted by displaying levels of column factor on the horizontal axis, levels of row factor using lines that are visually distinguished by line type and color, and the observed values on the vertical axis. Color and line type are used to display which levels of row factor are assigned to which sub-tables based on the minimum p-values among all possible configurations. Note that the grouping colors and line types appear whether or not the `KKSA.test` detects a significant non-additivity. The default colors are blue and red, and the default line types are one and two for the two sub-tables. They can be customized by supplying arguments called `vecolor` and `linetype`. Note that this method of testing requires that the data matrix has more than three rows. This test procedure is powerful for detecting interaction when the magnitude of interaction effects is heteroscedastic across the sub-tables of observations.

Value

An object of the class `ITtest`, which is a list inducing following components:

<code>pvalue_exact</code>	The calculated exact Monte Carlo p-value.
<code>pvalue_appro</code>	The Bonferroni-adjusted p-value is calculated.
<code>statistic</code>	The value of the test statistic.
<code>Nsim</code>	The number of Monte Carlo samples that are used to estimate p-value.
<code>data_name</code>	The name of the input dataset.
<code>test</code>	The name of the test.
<code>Level</code>	The level of test.
<code>Result</code>	The result of the test at the alpha level with some descriptions on the type of significant interaction.

References

- Kharrati-Kopaei, M., Sadooghi-Alvandi, S.M. (2007). A New Method for Testing Interaction in Unreplicated Two-Way Analysis of Variance. *Communications in Statistics-Theory and Methods* 36:2787–2803.
- Shenavari, Z., Kharrati-Kopaei, M. (2018). A Method for Testing Additivity in Unreplicated Two-Way Layouts Based on Combining Multiple Interaction Tests. *International Statistical Review* 86(3): 469-487.

Examples

```
data(IDCPC)
KKSA_test(IDCPC, nsim = 1000, Elapsed_time = FALSE)
```

Malik_test

Malik's (2016) et al. Test for Interaction

Description

The Malik's (2016) et al. test statistic is calculated and the corresponding exact p-value is calculated by a Monte Carlo simulation.

Usage

```
Malik_test(x, nsim = 10000, alpha = 0.05, report = TRUE, Elapsed_time = TRUE)
```

Arguments

x	numeric matrix, $a \times b$ data matrix where the number of row and column is corresponding to the number of factor levels.
nsim	a numeric value, the number of Monte Carlo samples for computing an exact Monte Carlo p-value. The default value is 10000.
alpha	a numeric value, the level of the test. The default value is 0.05.
report	logical: if TRUE the result of the test is reported at the alpha level.
Elapsed_time	logical: if TRUE the progress will be printed in the console.

Details

Malik (2016) et al. proposed to partition the residuals into three clusters using a suitable clustering method like “k-means clustering”. The hypothesis of no interaction can be interpreted as the effect of the three clusters are equal. Therefore, the result of the test may depend on the method of clustering. In this package, clustering is done by kmeans function in RcppArmadillo. The speed_mode parameter on the kmeans clustering was set as static_subset. Note that the Malik's et al. test performs well when there are some outliers in the residuals; i.e. some cells produce large negative or positive residuals due to the significant interaction. Further, the distribution of the Malik's et al. test statistic is not known under additivity and the corresponding p-value is calculated by a Monte Carlo simulation.

Value

An object of the class ITtest, which is a list inducing following components:

pvalue_exact	The calculated exact Monte Carlo p-value.
pvalue_appro	is not available for Malik_test.
statistic	The value of the test statistic.

Nsim	The number of Monte Carlo samples that are used to estimate p-value.
data_name	The name of the input dataset.
test	The name of the test.
Level	The level of test.
Result	The result of the test at the alpha level with some descriptions on the type of significant interaction.

References

Malik, W.A., Moring, J., Piepho, H.P. (2016). A clustering-based test for non-additivity in an unreplicated two-way layout. *Communications in Statistics-Simulation and Computation* 45(2):660-670.

Shenavari, Z., Kharrati-Kopaei, M. (2018). A Method for Testing Additivity in Unreplicated Two-Way Layouts Based on Combining Multiple Interaction Tests. *International Statistical Review* 86(3): 469-487.

Examples

```
data(IDCPC)
Malik_test(IDCPC, nsim = 1000, Elapsed_time = FALSE)
```

MVGH

The mean values of growth hormone (MVGH).

Description

This data set are about the mean values of growth hormone for the levels of zinc and thyroid hormone obtained by Freake et al. (2001). This data set has been previously analyzed by Alin and Kurt (2006). There three levels of zinc: Zinc deficient, Pair-fed, and Control. There are also three levels of thyroid hormone: Hypothyroid, Euthyroid, and Hyperthyroid. The test of interaction between the zinc and thyroid hormone is of interest.

Format

A matrix with three rows (Thyroid levels) and three columns (Zinc levels):

Row1 Hypothyroid

Row2 Euthyroid

Row3 Hyperthyroid.

Column1 Zinc deficient

Column2 Pair-fed

Column3 Control

References

1. Alin, A., Kurt, S. (2006). Testing non-additivity (interaction) in two-way ANOVA tables with no replication, *Statistical Methods in Medical Research* **15**: 63–85.
2. Freake, H. C., Govoni, K. E., Guda, K., Huang, C, Zinn, S. A. (2001). Actions and interactions of thyroid hormone and zinc status in growing rats. *Journal of Nutrition* 131:1135–41.

Piepho_test

Piepho's (1994) Test for Interaction

Description

This function tests the interaction based on a statistic proposed by Piepho (1994). This function reports Piepho's test statistic, an asymptotic p-value, and a Monte Carlo p-value.

Usage

```
Piepho_test(x, nsim = 10000, alpha = 0.05, report = TRUE)
```

Arguments

x	numeric matrix, $a \times b$ data matrix where the number of row and column is corresponding to the number of factor levels.
nsim	a numeric value, the number of Monte Carlo samples for computing an exact Monte Carlo p-value. The default value is 10000.
alpha	a numeric value, the level of the test. The default value is 0.05.
report	logical: if TRUE the result of the test is reported at the alpha level.

Details

Piepho (1994) proposed three test statistics. The third one is based on Grubbs' (1948) type estimator of variance for the level of the row factor. This type of estimator is used in this function. Piepho (1994) proposed an asymptotic distribution of test statistic; however, a Monte Carlo method is used to calculate the p-value. The Piepho test is not applicable when the row number of the data matrix is less than three. Note that Piepho's test is powerful for detecting interactions when the Grubbs' type estimators of variances are heterogeneous across the levels of one factor.

Value

An object of the class ITtest, which is a list inducing following components:

pvalue_exact	The calculated exact Monte Carlo p-value.
pvalue_appro	The asymptotic p-value.
statistic	The value of the test statistic.
Nsim	The number of Monte Carlo samples that are used to estimate p-value.
data_name	The name of the input dataset.

test	The name of the test.
Level	The level of test.
Result	The result of the test at the alpha level with some descriptions on the type of significant interaction.

References

Piepho, H. P. (1994). On Tests for Interaction in a Nonreplicated Two-Way Layout. *Australian Journal of Statistics* 36:363-369.

Shenavari, Z., Kharrati-Kopaei, M. (2018). A Method for Testing Additivity in Unreplicated Two-Way Layouts Based on Combining Multiple Interaction Tests. *International Statistical Review* 86(3): 469-487.

Grubbs, F.E. (1948). On Estimating Precision of Measuring Instruments and Product Variability. *Journal of the American Statistical Association* 43(242): 243-264.

Examples

```
data(MVGH)
Piepho_test(MVGH, nsim = 1000)
```

RDWW

Ratio of dry to wet wheat (RDWW).

Description

This data set are about the ratio of dry to wet wheat of four different blocks and four times of nitrogen applied: None, Early, Middle, and Late. The test of interaction between the blocks and the level of nitrogen applied is of interest.

Format

A matrix with four rows (Blocks) and four columns (Nitrogen Applied):

Row1 Block1

Row2 Block2

Row3 Block3

Row4 Block4

Column1 None

Column2 Early

Column3 Middle

Column4 Late

References

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