Package 'dfphase1'

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Title Phase I Control Charts (with Emphasis on Distribution-Free Methods)

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Description Statistical methods for retrospectively detecting changes in location and/or dispersion of univariate and multivariate variables. Data values are assumed to be independent, can be individual (one observation at each instant of time) or subgrouped (more than one observation at each instant of time). Control limits are computed, often using a permutation approach, so that a prescribed false alarm probability is guaranteed without making any parametric assumptions on the stable (in-control) distribution. See G. Capizzi and G. Masarotto (2018)
<doi:10.1007/978-3-319-75295-2_1> for an introduction to the package.

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dfphase1-package Phase I Control Charts (with Emphasis on Distribution-Free Methods)

Description

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Statistical methods for retrospectively detecting changes in location and/or dispersion of univariate and multivariate variables. Data values are assumed to be independent, can be individual (one observation at each instant of time) or subgrouped (more than one observation at each instant of time). Control limits are computed, often using a permutation approach, so that a prescribed false alarm probability is guaranteed without making any parametric assumptions on the stable (in-control) distribution.

Details

The main functions are:

- shewhart and mshewhart: univariate and multivariate Shewhart-type control charts based either on the original observations or on a rank transformation. These functions are particularly useful for detecting isolated shifts in the mean and/or variance of subgrouped observations. Functions shewhart and mshewhart also allow the simultaneously use of two control charts originally designed to detect separately location and scale shifts. In particular, note that when more than one critical values are needed, the false alarm probability is "balanced" between the separate control charts as discussed by Capizzi (2015).
- changepoint and mchangepoint: univariate or multivariate control charts useful for detecting sustained (and other patterned) mean and/or variance shifts. The control statistic is based on a generalized likelihood ratio test computed under a Gaussian assumption. However, the control limits are computed by permutation. An optional preliminary rank transformation can be used to improve the performance in the case of nonnormal data.
- rsp and mphase1: the univariate and multivariate methods introduced by Capizzi and Masarotto (2013) and (2017) to detect multiple isolated or step shifts in individual or subgrouped data.

The use of distribution-free control limits is emphasized. However, the package also includes some functions for computing normal-based control limits. As noted in the individual help pages, these limits can also be suitable for some non-normal distributions (e.g., applying a multivariate rank.-transformation, normal-based control limits mantain the desired false alarm probability in the class of the multivariate elliptical distributions). Nevertheless, their use is not generally recommended.

The data should be organized as follows:

changepoint

- Univariate control charts: an nxm matrix, where n and m are the size of each subgroup and the number of subgroups, respectively. A vector of length m is accepted in the case of individual data, i.e., when n=1.
- *Multivariate control charts:* a pxnxm array, where p denotes the number of monitored variables. A p x m matrix is accepted in the case of individual data.

Functions phase1Plot and mphase1Plot can be used for plotting the data.

Author(s)

Giovanna Capizzi and Guido Masarotto (maintainer: Giovanna Capizzi <giovanna.capizzi@unipd.it>).

References

G. Capizzi (2015) "Recent advances in process monitoring: Nonparametric and variable-selection methods for Phase I and Phase II (with discussion)". *Quality Engineering*, **27**, pp. 44–80, doi:10.1080/08982112.2015.968046.

G. Capizzi and G. Masarotto (2013), "Phase I Distribution-Free Analysis of Univariate Data". *Journal of Quality Technology*, **45**, pp. 273–284, doi:10.1080/00224065.2013.11917938.

G. Capizzi and G. Masarotto (2017), Phase I Distribution-Free Analysis of Multivariate Data, *Technometrics*, **59**, pp. 484–495, doi:10.1080/00401706.2016.1272494.

G. Capizzi and G. Masarotto (2018), "Phase I Distribution-Free Analysis with the R Package dfphase1". *Frontiers in Statistical Quality Control 12*, eds. S. Knoth and W. Schmid, pp. 3–19, Springer, doi:10.1007/9783319752952_1

See Also

shewhart, shewhart.normal.limits, mshewhart, mshewhart.normal.limits, changepoint, changepoint.normal.limits, mchangepoint, mchangepoint.normal.limits, rsp, mphase1, phase1Plot, mphase1Plot.

changepoint Detection of a sustained change-point in univariate and multivariate data

Description

changepoint (univariate data) and mchangepoint (multivariate data) test for the presence of a sustained location and/or dispersion shift. Both functions can be applied to individual and subgrouped observations.

changepoint.normal.limits and mchangepoint.normal.limits precompute the corresponding control limits when the in-control distribution is normal.

Usage

```
changepoint(x, subset, score = c("Identity", "Ranks"), only.mean = FALSE,
  plot = TRUE, FAP = 0.05, seed = 11642257, L = 10000, limits = NA)
mchangepoint(x, subset, score = c("Identity", "Signed Ranks", "Spatial Signs",
  "Spatial Ranks", "Marginal Ranks"), only.mean = FALSE,
  plot = TRUE, FAP = 0.05, seed = 11642257, L = 10000, limits = NA)
changepoint.normal.limits(n, m, score = c("Identity", "Ranks"),
  only.mean = FALSE, FAP = 0.05, seed = 11642257, L = 100000)
mchangepoint.normal.limits(p, n, m, score = c("Identity", "Signed Ranks", "Spatial Signs",
  "Spatial Ranks", "Marginal Ranks"), only.mean = FALSE,
  FAP = 0.05, seed = 11642257, L = 100000)
```

Arguments

х	changepoint: a nxm numeric matrix or a numeric vector of length m.
	mchangepoint: a pxnxm data numeric array or a pxm numeric vector.
	See below, for the meaning of p, n and m.
р	integer: number of monitored variables.
n	integer: size of each subgroup (number of observations gathered at each time point).
m	integer: number of subgroups (time points).
subset	an optional vector specifying a subset of subgroups/time points to be used
score	character: the transformation to use; see mshewhart.
only.mean	logical; if TRUE only a location change-point is searched.
plot	logical; if TRUE, the control statistic is displayed.
FAP	numeric (between 0 and 1): the desired false alarm probability.
seed	positive integer; if not NA, the RNG's state is resetted using seed. The current .Random.seed will be preserved. Unused by mshewhart when limits is not NA.
L	positive integer: the number of Monte Carlo replications used to compute the control limits. Unused by changepoint and mchangepoint when limits is not NA.
limits	numeric: a precomputed vector of length m containing the control limits.

Details

After an optional rank transformation (argument score), changepoint and mchangepoint compute, for $\tau = 2, ..., m$, the normal likelihood ratio test statistics for verifying whether the mean and dispersion (or only the mean when only.mean=TRUE) are the same before and after τ . See Sullivan and Woodall (1999, 2000) and Qiu (2013), Chapter 6 and Section 7.5.

Note that the control statistic is equivalent to that proposed by Lung-Yut-Fong *et al.* (2011) when score="Marginal Ranks" and only.mean=TRUE.

changepoint

As suggested by Sullivan and Woodall (1999, 2000), control limits proportional to the in-control mean of the likelihood ratio test statistics are used. Further, when plot=TRUE, the control statistics divided by the time-varying control limits are plotted with a "pseudo-limit" equal to one.

When only.mean=FALSE, the decomposition of the likelihood ratio test statistic suggested by Sullivan and Woodall (1999, 2000) for diagnostic purposes is also computed, and optionally plotted.

Value

changepoint and mchangepoint return an invisible list with elements

glr	control statistics.				
mean, dispersion					
	decomposition of the control statistics in the two parts due to changes in the mean and dispersion, respectively. These elements are present only when only.mean=FALSE.				
limits	control limits.				
score, only.mean, FAP, L, seed					
	input arguments.				

changepoint.normal.limits and mchangepoint.normal.limits return a numeric vector containing the control limits.

Note

- 1. When limits is NA, changepoint and mchangepoint compute the control limits by permutation. The resulting control charts are distribution-free.
- 2. Pre-computed limits, like those computed using changepoint.normal.limits and mchangepoint.normal.limits, are recommended *only* for univariate data when score=Ranks. Indeed, in all the other cases, the resulting control chart will not be distribution-free.
- 3. However, note that, when score is Signed Ranks, Spatial Signs, Spatial Ranks the normalbased control limits are distribution-free in the class of all multivariate elliptical distributions.

Author(s)

Giovanna Capizzi and Guido Masarotto.

References

A. Lung-Yut-Fong, C. Lévy-Leduc, O. Cappé O (2011) "Homogeneity and change-point detection tests for multivariate data using rank statistics". arXiv:11071971, https://arxiv.org/abs/1107. 1971.

P. Qiu (2013) Introduction to Statistical Process Control. Chapman & Hall/CRC Press.

J. H. Sullivan, W. H. Woodall (1996) "A control chart for preliminary analysis of individual observations". *Journal of Quality Technology*, **28**, pp. 265–278, doi:10.1080/00224065.1996.11979677.

J. H. Sullivan, W. H. Woodall (2000) "Change-point detection of mean vector or covariance matrix shifts using multivariate individual observations". *IIE Transactions*, **32**, pp. 537–549 doi:10.1080/07408170008963929.

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Examples

```
data(gravel)
changepoint(gravel[1,,])
mchangepoint(gravel)
mchangepoint(gravel, score="Signed Ranks")
```

colonscopy

Colonscopy Times

Description

This data set contains the colonscopy times (minutes) for 30 subgroups of 5 patients given in Allison Jones-Farmer *et al.* (2009).

Usage

data(colonscopy)

Format

A 5x30 matrix.

References

L. A. Jones-Farmer, V. Jordan, C. W. Champs (2009) "Distribution-free Phase I control charts for subgroup location", *Journal of Quality Technology*, **41**, pp. 304–316, doi:10.1080/00224065.2009.11917784.

Examples

```
data(colonscopy)
phase1Plot(colonscopy)
```

fe

Ferric Oxide data

Description

This data set contains 189 ferric-oxide individual measurement collected in an aluminum smelter.

Usage

data(fe)

Format

A vector of length 189.

gravel

References

M. D. Holland, D. M. Hawkins (2014) "A Control Chart Based on a Nonparametric Multivariate Change-Point Model", *Journal of Quality Technology*, **46**, pp 63–77, doi:10.1080/00224065.2014.11917954.

Examples

data(fe)
phase1Plot(fe)

gravel

Gravel data

Description

This data set contains 56 individual bivariate observations from a gravel-producing plant given by Holmes and Mergen (1993). There are two variables measuring the percentage of the particles (by weight) that are large or medium in size, respectively.

Usage

data(gravel)

Format

A 2x56 matrix.

References

D. S. Holmes, A. Mergen (1993) "Improving the Performance of the T^2 Control Chart", *Quality Engineering*, **5**, pp. 619–625, doi:10.1080/08982119308919004.

Examples

```
data(gravel)
mphase1Plot(gravel)
```

mphase1

Description

Retrospective change point detection using the method described by Capizzi and Masarotto (2017).

Usage

Arguments

x	a pxnxm array containing the observations; x[r,j,i] is the jth observation on the rth variable of the ith subgroup.
plot	logical; if FALSE the diagnostic plot is not displayed.
post.signal	logical; if FALSE the diagnostic LASSO-based analysis is not performed.
isolated	logical; if FALSE isolated shifts are not detected.
step	logical; if FALSE step shifts are not detected.
alpha	real; the acceptable false alarm probability; if the observed p-value is greater than alpha, then the estimated mean function is a constant.
gamma	real; the extra penalization for the extended BIC criteria.
К	integer; the maximum number of shifts which the procedure tries to detect.
lmin	integer; the minimum length of a step shift.
L	integer; the number of random permutations used to compute the p-values.
seed	integer; if not NA, the RNG's state is re-setted using seed. The current .Random.seed will be preserved.

Value

Functions mphase1 returns an object of class mphase1 containing

p.value	The p-value.
Wobs	The overall test statistic.
alasso	A data-frame containing the result of the post-signal diagnosis analysis, i.e., the times and types of shifts and the involved variables identified using the adaptive LASSO.
forward	A data frame containing the result of the forward search analysis, i.e., the times and types of the possible shifts as well as the elementary test statistics and the estimates of their (conditional) means and standard deviations.
center, scatter	
	The location vector and dispersion matrix used to standardize the original data.

signed.ranks A pxnxm array containing the signed ranks.

fitted, residuals

Two pxnxm arrays containing the fitted means and the residuals, i.e., the difference between the observations and the fitted values.

Author(s)

Giovanna Capizzi and Guido Masarotto.

References

G. Capizzi and G. Masarotto (2017), Phase I Distribution-Free Analysis of Multivariate Data, *Technometrics*, **59**, pp. 484–495, doi:10.1080/00401706.2016.1272494.

See Also

postsignal.

Examples

```
# A simulated in-control data from a Student's t distribution
# with 3 degrees of freedom
set.seed(123)
x <- sweep(array(rnorm(5*5*50),c(5,5,50)),c(2,3),sqrt(rchisq(5*50,3)/3),"/")
mphase1(x)
# Reproduction of the two examples given in Capizzi and Masarotto (2016)
data(ryan)
mphase1(ryan)
data(gravel)
mphase1(gravel)
```

mphase1.methods *Methods for objects of class* mphase1

Description

Methods print and plot allow to write to the console and plot (optionally changing the layout) the result of the Phase I analysis performed with function mphase1.

Method postsignal implements the post-signal Phase I analysis based on the adaptive LASSO described in Capizzi and Masarotto (2016). It uses the p-value and the results on the forward search contained in its first argument. Hence, it is useful for re-running the analysis with different values of alpha and/or gamma.

Usage

```
## S3 method for class 'mphase1'
print(x,...)
## S3 method for class 'mphase1'
plot(x,layout=c(1,p),...)
## S3 method for class 'mphase1'
postsignal(x, plot = TRUE, alpha = 0.05, gamma = 0.5,...)
```

Arguments

х	an object returned by function mphase1.
layout	an integer vector describing the multi-panel (and possible multi-page) layout.
plot	logical; if TRUE the diagnostic plot is displayed.
alpha	real; the acceptable false alarm probability; if the observed p-value is greater than alpha, then the estimated mean function is a constant.
gamma	real; the extra penalization for the extended BIC criteria.
	ignored.

Value

An object of class mphase1. See mphase1 for the description.

Author(s)

Giovanna Capizzi and Guido Masarotto.

References

G. Capizzi and G. Masarotto (2017), Phase I Distribution-Free Analysis of Multivariate Data, *Technometrics*, **59**, pp. 484–495, doi:10.1080/00401706.2016.1272494.

Examples

```
data(gravel)
u <- mphase1(gravel,plot=FALSE)
print(u)
plot(u,layout=c(2,1))
postsignal(u,plot=FALSE,gamma=1)
```

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mshewhart

Description

mshewhart computes, and, optionally, plots, several Shewhart-type Phase I control charts for detecting location and scale changes in multivariate subgrouped data.

mshewhart.normal.limits pre-computes the corresponding control limits when the in-control distribution is multivariate normal.

Usage

```
mshewhart(x, subset, stat = c("T2Var", "T2", "Var", "Depth Ranks"), score = c("Identity",
    "Signed Ranks", "Spatial Signs", "Spatial Ranks", "Marginal Ranks"),
    loc.scatter = c("Classic", "Robust"), plot = TRUE, FAP = 0.05,
    seed = 11642257, L = 1000, limits = NA)
mshewhart.normal.limits(p, n, m, stat = c("T2Var", "T2", "Var", "Depth Ranks"),
    score = c("Identity", "Signed Ranks", "Spatial Signs", "Spatial Ranks",
    "Marginal Ranks"), loc.scatter = c("Classic", "Robust"),
    FAP = 0.05, seed = 11642257, L = 100000)
```

Arguments

x	a pxnxm data numeric array (n observations gathered at m time points on p variables).
р	integer: number of monitored variables.
n	integer: size of each subgroup (number of observations gathered at each time point).
m	integer: number of subgroups (time points).
subset	an optional vector specifying a subset of subgroups/time points to be used
stat	character: control statistic[s] to use; see Details.
score	character: transformation to use; unused when stat=Depth Ranks; see Details.
loc.scatter	character: estimates of the multivariate location and scatter to use when no pre- liminary rank transformation is applied. Unused when stat is equal to Depth
	Ranks or score is Marginal Ranks. See Details.
plot	Ranks or score is Marginal Ranks. See Details. logical; if TRUE, control statistic[s] is[are] displayed.
plot FAP	
	logical; if TRUE, control statistic[s] is[are] displayed.

limits numeric: pre-computed vector of control limits. This vector should contain (A, B) when stat=T2Var, (A) when stat=T2, (B) when stat=Var and (C) when stat=Depth Ranks. See Details for the definition of the critical values A, B and C.

Details

The implemented control statistics are

- T2Var: combination of the T2 and Var statistics described below.
- T2: Hotelling's T^2 control statistics (see Montgomery, 2009, equation 11.19, or Qiu, 2013, equation 7.7) with control limit equal to A.
- Var: normal likelihood ratio control statistics for detecting changes in the multivariate dispersion (see Montgomery, 2009, equation 11.34), with control limit equal to *B*.
- Depth Ranks: control statistics based on the rank of the Mahalanobis depths, proposed by Bell *et. al.*. As suggested Bell *et al.*, the Mahalanobis depths are computed using the BACON estimates of the multivariate mean vector and the mean of the subgroups sample covariance matrices. An alarm is signalled if any of the statistics is greater than a positive control limit *C*.

The T2 and Var control statistics are computed

- score=Identical: from the original data standardized using either the classical pooled estimates of the mean vector and dispersion matrix (Montgomery, 2009, equations 11.14–11.18; Qiu, 2013, equations at page 269) or the highly robust minimum covariance determinant (MCD) estimate when argument loc.scatter is equal to Classic or Robust, respectively.
- score=Signed Ranks, Spatial Signs, Spatial Ranks, Marginal Ranks: from a "rank" transformation of the original data. In particular, see Hallin and Paindaveine (2005) for the definition of the multivariate signed ranks and Oja (2010) for those of the spatial signs, spatial ranks, and marginal ranks. Multivariate signed ranks, spatial signs and ranks are "inner" standardized while marginal ranks are "outer" standardized (see Oja (2010) for the definition of "inner" and "outer" standardization). When loc.scatter is equal to Classic, inner standardization takes into account the subgroup structure of the data imposing that the average of the **within-group** covariances of the transformed data is proportional to the identity matrix. Otherwise, i.e., when loc.scatter is equal to Robust, it is based on a standard Hettmansperger-Randles-like scatter estimate. Note that the T^2 control statistics based on the spatial signs corresponds to the control charts suggested by Cheng and Shiau (2015) when loc.scatter is equal to Robust.

Value

mshewhart returns an invisible list with elements:

Т2	T^2 control statistic; this element is present only if stat is T2Var or T2.
Var	Var control statistic; this element is present only if stat is T2Var or Var.
DepthRanks	control statistic based on the rank of the Mahalanobis depths; this element is present only if stat is Depth Ranks.

mshewhart

center, scatter	
	estimates of the multivariate location and scatter used to standardized the observations.
limits	control limits.
stat, score, loc	.scatter, FAP, L, seed
	input arguments.

mshewhart.normal.limits returns a numeric vector containing the control limits.

Note

- 1. When limits is NA, mshewhart computes the control limits by permutation. Then, the resulting control chart is distribution-free.
- 2. Pre-computed limits, such as those computed by using mshewhart.normal.limits, are not recommended. Indeed, the resulting control chart will not be distribution-free.
- 3. However, when score is Signed Ranks, Spatial Signs, Spatial Ranks or stat is Depth Ranks, the computed control limits are distribution-free in the class of all multivariate elliptical distributions.

Author(s)

Giovanna Capizzi and Guido Masarotto.

References

R. C. Bell, L. A. Jones-Farmer, N. Billor (2014) "A distribution-free multivariate Phase I location control chart for subgrouped data from elliptical distributions". *Technometrics*, **56**, pp. 528–538, doi:10.1080/00401706.2013.879264.

C. R. Cheng, J. J. H. Shiau JJH (2015) "A distribution-free multivariate control chart for Phase I applications". *Quality and Reliability Engineering International*, **31**, pp. 97–111, doi:10.1002/ qre.1751.

M. Hallin and D. Paindaveine (2005) "Affine-Invariant Aligned Rank Tests for the Multivariate General Linear Model with VARMA Errors". *Journal of Multivariate Analysis*, **93**, pp. 122–163, doi:10.1016/j.jmva.2004.01.005.

D. C. Montgomery (2009) Introduction to Statistical Quality Control, 6th edn. Wiley.

H. Oja (2010) Multivariate Nonparametric Methods with R. An Approach Based on Spatial Signs and Ranks. Springer.

P. Qiu (2013) Introduction to Statistical Process Control. Chapman & Hall/CRC Press.

Examples

```
data(ryan)
mshewhart(ryan)
mshewhart(ryan, subset=-10)
mshewhart(ryan, subset=-c(10, 20))
mshewhart(ryan, score="Signed Ranks")
mshewhart(ryan, subset=-10, score="Signed Ranks")
mshewhart(ryan, subset=-c(10, 20), score="Signed Ranks")
```

phase1Plot

Description

phase1Plot and mphase1Plot plot univariate or multivariate Phase 1 observations, organized as required by the dfphase1 package.

Usage

phase1Plot(x)

mphase1Plot(x, layout = c(1, p))

Arguments

х	phase1Plot: a nxm numeric matrix or a numeric vector of length m.
	mphase1Plot: a pxnxm data numeric array or a pxm numeric matrix.
	Here, p denotes the number of variables, n the size of each subgroup and m the number of subgroups.
layout	an integer vector describing the multi-panel (and possible multi-page) layout. See the third example below.

Author(s)

Giovanna Capizzi and Guido Masarotto.

Examples

```
x <- matrix(rt(5*20,5),5)
x[,10] <- x[,10]+3
phase1Plot(x)
# a data set with many variables
x <- array(rnorm(20*5*50),c(20,5,50))+10*(1:20)
mphase1Plot(x)
# it is better to organize the plot on two pages
if (interactive()) old <- grDevices::devAskNewPage(TRUE)
mphase1Plot(x,c(2,5,2))
if (interactive()) grDevices::devAskNewPage(old)
```

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Distribution-Free Phase I Analysis of Univariate Data based on Recursive Segmentation and Permutation

Description

rsp implements the Phase I method described in Capizzi and Masarotto (2013).

Usage

```
rsp(y, plot = TRUE, L = 1000, seed = 11642257, alpha = 0.05,
maxsteps = min(50, round(NROW(y)/15)), lmin = max(5, min(10, round(NROW(y)/10))))
```

Arguments

у	Phase I data; y can be either (i) a vector or a 1xm matrix in the case of individual observations or (ii) a nxm matrix for subgrouped data (n observations gathered at m time points).
plot	logical; if TRUE, the diagnostic plot is displayed.
L	integer; the number of random permutations used to compute the p-values.
seed	positive integer; if not NA, the RNG's state is resetted using seed. The current .Random.seed will be preserved.
alpha	real; the significance level used to compute the level and scale estimates; if one of the p-values is greater than alpha, the corresponding estimate is a constant.
maxsteps	integer; the maximum number of step shifts which the procedure tries to detect.
lmin	integer; the minimum length of a step.

Value

A list with elements

р	the adjusted p-values
stat	the summary statistics (a mx2 matrix)
fitted	the (possibly time-variant) estimates of the process level and scale (a mx2 ma- trix).

Author(s)

Giovanna Capizzi and Guido Masarotto.

References

G. Capizzi, G. Masarotto (2013), "Phase I Distribution-Free Analysis of Univariate Data". *Journal of Quality Technology*, **45**, pp. 273-284, doi:10.1080/00224065.2013.11917938.

rsp

Examples

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```
# Individual observations with a transient level change
set.seed(112233)
level <- c(rep(0,20),rep(3,10),rep(0,20))
x <- level+rt(50,4)
rsp(x)
# Individual observations with a scale step change
scale <- c(rep(1,25),rep(3,25))
x <- scale*rt(50,4)
rsp(x)
data(fe)
rsp(fe)
data(colonscopy)
rsp(colonscopy)
```

ryan

Ryan data

Description

This data set contains the data given in Table 9.2 by Ryan (2011, p. 323). The sample comprises 20 subgroups, each with 4 observations, on two quality characteristics X_1 and X_2 . According to Ryan (2011), the 10th and 20th subgroups are out-of-control.

Usage

data(ryan)

Format

A 2x4x20 array.

References

T. P. Ryan (2011), Statistical Methods for Quality Improvement, 3rd ed., Wiley.

Examples

data(ryan)
mphase1Plot(ryan)

ryan

Description

shewhart computes, and, optionally, plots, Shewhart-type Phase I control charts for detecting changes in location and scale of univariate subgrouped data.

shewhart.normal.limits pre-computes the corresponding control limits when the in-control distribution is normal.

Usage

```
shewhart(x, subset,
         stat = c("XbarS", "Xbar", "S",
                   "Rank", "lRank", "sRank",
"Lepage", "Cucconi"),
         aggregation = c("mean", "median"),
         plot = TRUE,
         FAP = 0.05,
         seed = 11642257,
         L = 1000,
         limits = NA)
shewhart.normal.limits(n, m,
                         stat = c("XbarS", "Xbar", "S",
                                   "Rank", "lRank", "sRank",
                                   "Lepage", "Cucconi"),
                         aggregation = c("mean", "median"),
                         FAP = 0.05,
                         seed = 11642257,
                         L = 100000)
```

Arguments

x	a nxm data numeric matrix (n observations gathered at m time points).
subset	an optional vector specifying a subset of subgroups/time points to be used
stat	character: the control statistic[s] to use; see Details.
aggregation	character: it specify how to aggregate the subgroup means and standard devia- tions. Used only when stat is XbarS, Xbar or S.
plot	logical; if TRUE, control statistic[s] is[are] displayed.
FAP	numeric (between 0 and 1): desired false alarm probability. Unused by shewhart when limits is not NA.
seed	positive integer; if not NA, the RNG's state is resetted using seed. The current .Random.seed will be preserved. Unused by shewhart when limits is not NA.

L	positive integer: number of random permutations used to compute the control limits. Unused by shewhart when limits is not NA.
limits	numeric: a precomputed vector of control limits. The vector should contain (A,B_1,B_2) when stat=XbarS, (A) when stat=Xbar, (B_1,B_2) when stat=S, (C,D) when stat=Rank, (C) when stat=lRank, (D) when stat=sRank, and (E) when stat=Lepage or stat=Cucconi. See Details for the definition of the critical values A,B_1,B_2,C,D and $E.$
n	integer: size of each subgroup (number of observations gathered at each time point).
m	integer: number of subgroups (time points).

Details

The implemented control charts are:

- XbarS: combination of the Xbar and S control charts described in the following.
- Xbar: chart based on plotting the subgroup means with control limits

$$\hat{\mu} \pm A \frac{\hat{\sigma}}{\sqrt{n}}$$

where $\hat{\mu}(\hat{\sigma})$ denotes the estimate of the in-control mean (standard deviation) computed as the mean or median of the subgroup means (standard deviations).

- S: chart based on plotting the (unbiased) subgroup standard deviations with lower control limit $B_1\hat{\sigma}$ and upper control limit $B_2\hat{\sigma}$.
- Rank: combination of the 1Rank and sRank control charts described in the following.
- 1Rank: control chart based on the standardized rank-sum control statistic suggested by Jones-Farmer *et al.* (2009) for detecting changes in the location parameter. Control limits are of the type $\pm C$.
- sRank: chart based on the standardized rank-sum control statistic suggested by Jones-Farmer and Champ (2010) for detecting changes in the scale parameter. Control limits are of the type $\pm D$.
- Lepage: chart based on the Lepage control statistic suggested by Li *et al.* (2019) for detecting changes in location and/or scale. There is only a upper control limit equal to *E*.
- Cucconi: chart based on the Cucconi control statistic suggested by Li *et al.* (2020) for detecting changes in location and/or scale. There is only a upper control limit equal to *E*.

Value

shewhart returns an invisible list with elements

Xbar	subgroup means; this element is present only if stat is XbarS or Xbar.
S	subgroup standard deviation; this element is present only if stat is XbarS or S.
lRank	rank-based control statistics for detecting changes in location; this element is present only if stat is Rank or lRank.
sRank	rank-based control-statistics for detecting changes in scale; this element is present only if stat is Rank or sRank.

shewhart

Lepage, W2, AB2	
	Lepage, squared Wilcoxon and squared Ansari-Bradley statistics; these elements are present only if stat is Lepage.
Cucconi, 1Cucco	ni, sCucconi
	Cucconi control statistic and its location and scale components; these elements are present only if stat is Cucconi.
limits	control limits.
center, scale	estimates $\hat{\mu}$ and $\hat{\sigma}$ of the in-control mean and standard deviation; these elements are present only if stat is XbarS, Xbar and S.
stat, L, aggregation, FAP, seed	
	input arguments.

shewhart.normal.limits returns a numeric vector containing the limits.

Note

- 1. If argument limits is NA, shewhart computes the control limits by permutation. The resulting control chart are distribution-free.
- Pre-computed limits, such as those computed using shewhart.normal.limits, are not recommended when stat is XbarS, Xbar or S. Indeed, the resulting control chart will not be distribution-free.
- 3. When stat is Rank, 1Rank, sRank, Lepage or Cucconi the control limits computed by shewhart.normal.limits are distribution-free in the class of all univariate continuous distributions. So, if user plan to apply rank-based control charts on a repeated number of samples of the same size, pre-computing the control limits using mshewhart.normal.limits can reduce the overall computing time.

Author(s)

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References

L. A. Jones-Farmer, V. Jordan, C. W. Champs (2009) "Distribution-free Phase I control charts for subgroup location", *Journal of Quality Technology*, **41**, pp. 304–316, doi:10.1080/00224065.2009.11917784.

L. A. Jones-Farmer, C. W. Champ (2010) "A distribution-free Phase I control chart for subgroup scale". *Journal of Quality Technology*, **42**, pp. 373–387, doi:10.1080/00224065.2010.11917834

C. Li, A. Mukherjee, Q. Su (2019) "A distribution-free Phase I monitoring scheme for subgroup location and scale based on the multi-sample Lepage statistic", *Computers & Industrial Engineering*, **129**, pp. 259–273, doi:10.1016/j.cie.2019.01.013

C. Li, A. Mukherjee, M. Marozzi (2020) "A new distribution-free Phase-I procedure for bi-aspect monitoring based on the multi-sample Cucconi statistic", *Computers & Industrial Engineering*, **149**, doi:10.1016/j.cie.2020.106760

D. C. Montgomery (2009) Introduction to Statistical Quality Control, 6th edn. Wiley.

P. Qiu (2013) Introduction to Statistical Process Control. Chapman & Hall/CRC Press.

Student

Examples

```
# A simulated example
set.seed(12345)
y <- matrix(rt(100,3),5)</pre>
y[,20] <- y[,20]+3
shewhart(y)
shewhart(y, stat="Rank")
shewhart(y, stat="Lepage")
shewhart(y, stat="Cucconi")
# Reproduction of the control chart shown
# by Jones-Farmer et. al. (2009)
data(colonscopy)
u <- shewhart.normal.limits(NROW(colonscopy),NCOL(colonscopy),</pre>
                            stat="lRank", FAP=0.1, L=10000)
# In Jones-Farmer et al. (2009) is estimated as 2.748
u
shewhart(colonscopy,stat="lRank",limits=u)
# Examples of control limits for comparisons
# with Li et al. (2019) and (2020) but
# using a limited number of Monte Carlo
# replications
# Lepage: in Li et al. (2019) is estimated as 11.539
shewhart.normal.limits(5, 25, stat="Lepage", L=10000)
# Cucconi: in Li et al. (2020) is estimated as 0.266
shewhart.normal.limits(5, 25, stat="Cucconi", L=10000)
```

Student

A simulated dataset

Description

This simulated data set consists in 50 subgroups, each with 5 observations, on 4 variables.

There is an isolated location shift involving only the first variable at time t = 10 and a step shift, involving the third and fourth variables, starting from t = 31. The in-control distribution is Student's t with 3 degrees of freedom, zero mean and such that $cov(X_i, X_j) = 0.8^{|i-j|}$.

See the example for the exact code used to simulate the data.

Usage

```
data(Student)
```

Format

A 4x5x50 array.

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Student

Examples

```
data(Student)
mphase1(Student)
#
# Replication of the simulation
#
# Generation of the in-control observations
set.seed(1)
m <- 50
n <- 5
p <- 4
df <- 3
Sigma <- outer(1:p,1:p,function(i,j) 0.8^abs(i-j))</pre>
Sigma
xnorm <- crossprod(chol(Sigma),matrix(rnorm(p*n*m),p))</pre>
xchisq <- sqrt(rchisq(n*m,df)/(df-2))</pre>
x <- array(sweep(xnorm,2,xchisq,"/"),c(p,n,m))</pre>
# Then, we add an isolated shift at time 10
# (only for the first variable)
x[1,,10] <- x[1,,10]+1
# and, a step shift starting at time 31
# (only for the third and fourth variable)
x[3:4,,31:50] <- x[3:4,,31:50] + c(0.50,-0.25)
dimnames(x)<-list(paste("X",1:4,sep=""),NULL,NULL)</pre>
identical(x,Student)
```

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