

Package ‘bdots’

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Type Package

Title Bootstrapped Differences of Time Series

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Imports doParallel, doRNG, foreach

LazyData FALSE

Description Analyze differences among time series curves with p-value adjustment for multiple comparisons introduced in Oleson et al (2015) <DOI:10.1177/0962280215607411>.

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`bdots.write.csv`*Write to CSV*

Description

Write bootstrapped estimates and confidence intervals to csv

Usage

```
bdots.write.csv(part1.list, part2.list, file, agg = TRUE, ...)
```

Arguments

<code>part1.list</code>	list. Output from <code>doubleGauss.fit</code>
<code>part2.list</code>	list. Output from <code>doubleGauss.boot</code>
<code>file</code>	Name of file to write to
<code>agg</code>	Whether to aggregate the data.
<code>...</code>	Further arguments to <code>write.csv</code>

Details

Write raw group averages, bootstrapped estimates, and confidence intervals to csv

Value

NULL

Note

There are no further notes

Examples

```
## Not run:
data(ci)
ci.1 <- subset(ci, ci$LookType == "Target")
ci.1$Group <- ci.1$protocol
out.1 <- logistic.fit(ci.1, 4)
out.2 <- logistic.boot(out.1)
bdots.write.csv(out.1, out.2, "CIOutput.csv", row.names = FALSE)

## End(Not run)
```

Description

Bootstrap on the fitted parameters, plot the estimates, and highlight the significant regions

Usage

```
doubleGauss.boot(part1.list, seed = new.seed(), alpha = 0.05, paired = FALSE,
N.iter = 1000, cores = 1, p.adj = "oleson", test.spots = NULL,
time.test = NULL, test.params = FALSE)
logistic.boot(part1.list, seed = new.seed(), alpha = 0.05, paired = FALSE,
N.iter = 1000, cores = 1, p.adj = "oleson", test.spots = NULL,
time.test = NULL, test.params = FALSE)
```

Arguments

part1.list	list. Output from doubleGauss.fit
seed	integer. What to set seed at
alpha	numeric (Between 0 and 1). Probability of familywise Type I Error
paired	boolean. Whether the same subjects are in both data sets
N.iter	numeric (positive integer). Number of bootstrap iterations to run
cores	integer. Number of cores on the localhost to use
p.adj	Options: oleson, fdr, none
test.spots	numeric. Specify specific x-values for testing at
time.test	numeric. Specify individual time points to conduct t-tests at without any p-value correction
test.params	boolean. Whether to test for significant differences in group mean parameter estimates. Performs a 2-sample t-test with equal variance assumption if paired = FALSE. If paired = TRUE, performs a paired t-test.

Details

Bootstrap on the fitted parameters, plot the estimates, and highlight the significant regions

Value

List: for input in replot

Note

There are no further notes

Examples

```
## Not run:
data(ci)
ci.1 <- subset(ci, ci$LookType == "Target")
ci.1$Group <- ci.1$protocol
out.1 <- logistic.fit(ci.1, 4)
out.2 <- logistic.boot(out.1)
replot(out.2, bucket.lim = c(0, 1))

ci.2 <- subset(ci, ci$LookType == "Cohort" | ci$LookType == "Unrelated")
ci.2$Group <- ci.2$protocol
ci.2$Curve <- ifelse(ci.2$LookType == "Cohort", 1, 2)
out.1 <- doubleGauss.fit(ci.2, 4, diffs = TRUE)
out.1 <- doubleGauss.refit(out.1, subj = c(13, 23), group = c(2, 2),
  curves = c(2, 2), cor = c(FALSE, FALSE))
out.2 <- doubleGauss.boot(out.1)
replot(out.2, ylim = c(-0.01, 0.1), bucket.lim = c(0, 0.08))

## End(Not run)
```

ci	<i>Eyetracking Data from Normal Hearing Individuals and those with Cochlear Implants</i>
----	--

Description

This data set has stuff on CIs and NHs. Provide more info here.

Usage

```
ci
```

Format

A matrix of values

References

Farris-Trimble, A., McMurray, B., Cigrand, N. and Tomblin, J.B. (2014) The process of spoken word recognition in the face of signal degradation: Cochlear implant users and normal-hearing listeners. *Journal of Experimental Psychology: Human Perception and Performance*, 40(1), 308-327

`ests.plot`*Plot Parameter Estimates*

Description

Plots of a histogram of the parameter estimates

Usage

```
ests.plot(part1.list)
```

Arguments

`part1.list` Output from `doubleGauss.fit` or `logistic.fit`

Details

Plots of a histogram of the parameter estimates

Value

NULL

Note

There are no further notes

Examples

```
## Not run:
data(ci)
ci.1 <- subset(ci, ci$LookType == "Target")
ci.1$Group <- ci.1$protocol
out.1 <- logistic.fit(ci.1, 4)ests.plot(out.1)
ests.plot(out.1)

## End(Not run)
```

 Fitting Step

Fit Subjects Individual Curves

Description

Fit Subjects from 2 groups with the 6-parameter Double Gaussian or the 4-parameter Logistic

Usage

```
doubleGauss.fit(data, col, concave = TRUE, diffs = FALSE,
rho.0 = 0.9, cor = TRUE, cores = 1, verbose = TRUE)
logistic.fit(data, col, diffs = FALSE, rho.0 = 0.9, cor = TRUE, cores = 1, verbose = TRUE)
```

Arguments

data	data.frame. A data.frame with the columns 'Subject', 'Time', and 'Group'. 'Subject' designates the subject number (numeric), 'Time' designates the time (numeric, should be ordered from low to high), and 'Group' designates the group (should be 2 unique groups)
col	numeric. The column in the data.frame that corresponds to the eyetracking
concave	boolean. TRUE indicates concave UP, FALSE indicates concave DOWN. Only for Double Gaussian.
diffs	boolean. If the each group is calculating the difference of 2 logistic curves, set to TRUE. In this case, there needs to be a numeric 'Curve' column (with only 1s and 2s) designating the secondary curve (2) to subtract from the primary curve (1).
rho.0	numeric (Between 0 and 1). Assumed autocorrelation of errors for individual subject's curve
cor	boolean. If TRUE assumes an AR1 autocorrelation structure among the residuals for fitting
cores	integer. Number of cores on the localhost to use
verbose	boolean. Whether to include estimates for each curve

Details

Fit Subjects from 2 groups with the 6-parameter Double Gaussian or the 4-parameter Logistic

Value

List: for input into refit, boot, plot.ests, and plot.subjs functions

Note

There are no further notes

Examples

```
## Not run:
data(ci)
ci.1 <- subset(ci, ci$LookType == "Target")
ci.1$Group <- ci.1$protocol
out.1 <- logistic.fit(ci.1, 4)
out.2 <- logistic.boot(out.1)
replot(out.2, bucket.lim = c(0, 1))

ci.2 <- subset(ci, ci$LookType == "Cohort" | ci$LookType == "Unrelated")
ci.2$Group <- ci.2$protocol
ci.2$Curve <- ifelse(ci.2$LookType == "Cohort", 1, 2)
out.1 <- doubleGauss.fit(ci.2, 4, diffs = TRUE)
out.1 <- doubleGauss.refit(out.1, subj = c(13, 23), group = c(2, 2),
  curves = c(2, 2), cor = c(FALSE, FALSE))
out.2 <- doubleGauss.boot(out.1)
replot(out.2, ylim = c(-0.01, 0.1), bucket.lim = c(0, 0.08))

## End(Not run)
```

Print Fits

Print summary of quality of curve fits

Description

Print summary of quality of curve fits

Usage

```
printFits(part1.list)
```

Arguments

part1.list list. Output from doubleGauss.fit

Details

Print summary of quality of curve fits

Note

There are no further notes

Examples

```
## Not run:
data(ci)
ci.2 <- subset(ci, ci$LookType == "Cohort" | ci$LookType == "Unrelated")
ci.2$Group <- ci.2$protocol
ci.2$Curve <- ifelse(ci.2$LookType == "Cohort", 1, 2)
out.1 <- doubleGauss.fit(ci.2, 4, diffs = TRUE)
out.1 <- doubleGauss.refit(out.1, subj = c(13, 23), group = c(2, 2),
  curves = c(2, 2), cor = c(FALSE, FALSE))
printFits(out.1)
out.2 <- doubleGauss.boot(out.1)
replot(out.2, ylim = c(-0.01, 0.1), bucket.lim = c(0, 0.08))

## End(Not run)
```

 Refitting Step

Refit Subjects Individual Curves

Description

Refit Subjects from 2 groups with the 6-parameter Double Gaussian or the 4-parameter Logistic. Can specify starting parameters

Usage

```
doubleGauss.refit(part1.list, subj = NULL, group = NULL, curves = NULL,
  params = NULL, cor=NULL, rho.0 = NULL, info.matrix = NULL, get.cov.only = FALSE)
logistic.refit(part1.list, subj = NULL, group = NULL, curves = NULL,
  params = NULL, cor = NULL, rho.0 = NULL, info.matrix = NULL, get.cov.only = FALSE)
```

Arguments

<code>part1.list</code>	Output from fitting step
<code>subj</code>	numeric vector. Subject numbers (within their group) that you want to refit
<code>group</code>	numeric vector. Group numbers corresponding to subject vector
<code>curves</code>	numeric vector. Curve numbers if it's a difference of fits (i.e. <code>diffs = TRUE</code>)
<code>params</code>	list of numeric vectors (length 4 or 6). Parameter estimates for the 6 parameter DoubleGauss in the order μ , height, sd 1, sd 2, base 1, base 2. Parameter estimate for the 4 parameter Logistic in the order min, peak, slope, crossover
<code>cor</code>	logical vector. If <code>TRUE</code> assumes an correlation structure of AR(ρ) and if <code>FALSE</code> assumes no correlation structure
<code>rho.0</code>	numeric. assumed autocorrelation of errors for subject's curve
<code>info.matrix</code>	numeric matrix. Can put subj/group/curves/params information here for convenience. Columns should have the following order: subj, group, curves, params. This matrix cannot contain any NA/NaN values. If <code>diffs=FALSE</code> , this information is not used – simply use filler values.

`get.cov.only` logical. If TRUE, calculates covariance matrix at given parameter estimates. Does not perform any parameter fitting

Details

Refit Subjects from 2 groups with the 6-parameter Double Gaussian or the 4-parameter Logistic. Can specify starting parameters

Value

List

Note

There are no further notes

Examples

```
## Not run:
data(ci)
ci.2 <- subset(ci, ci$LookType == "Cohort" | ci$LookType == "Unrelated")
ci.2$Group <- ci.2$protocol
ci.2$Curve <- ifelse(ci.2$LookType == "Cohort", 1, 2)
out.1 <- doubleGauss.fit(ci.2, 4, diffs = TRUE)
out.1 <- doubleGauss.refit(out.1, subj = c(13, 23), group = c(2, 2),
  curves = c(2, 2), cor = c(FALSE, FALSE))
out.2 <- doubleGauss.boot(out.1)
replot(out.2, ylim = c(-0.01, 0.1), bucket.lim = c(0, 0.08))

## End(Not run)
```

replot

Replot Bootstrapped Output

Description

Plot the bootstrapped output with different parameters than the default ones.

Usage

```
replot(part2.list, xlim = NULL, ylim = c(0, 1), main = "Curve",
  legend.location = "topleft", bucket.lim = c(0, .9))
```

Arguments

part2.list	list. Output from doubleGauss.boot or logistic.boot
xlim	numeric vector (length = 2). Start and end point of x-axis. If NULL, takes the full time course
ylim	numeric vector (length = 2). Start and end point of y-axis
main	string. Title
legend.location	string. Location of the legend
bucket.lim	numeric vector (length = 2). How far the yellow significant region goes on the y axis

Details

Plot the bootstrapped output with different parameters than the default ones.

Value

NULL

Note

Options for legend.location include "topleft", "top", "topright", "right", "bottomright", "bottom", "bottomleft", "left"

Examples

```
## Not run:
data(ci)
ci.1 <- subset(ci, ci$LookType == "Target")
ci.1$Group <- ci.1$protocol
out.1 <- logistic.fit(ci.1, 4)
out.2 <- logistic.boot(out.1)
replot(out.2, bucket.lim = c(0, 1))

## End(Not run)
```

subs.delete

Delete individual subjects from both data and fits

Description

Delete individual subjects from both data and fits

Usage

```
subs.delete(part1.list, subjs, groups)
```

Arguments

part1.list Output from doubleGauss.fit or logistic.fit
 subjs numeric. Subject numbers to remove.
 groups numeric. Corresponding group numbers to subjects.

Details

Delete individual subjects from both data and fits

Value

part1.list

Examples

```
## Not run:
data(ci)
ci.1 <- subset(ci, ci$LookType == "Target")
ci.1$Group <- ci.1$protocol
out.1 <- logistic.fit(ci.1, 4)ests.plot(out.1)
#Remove subject 1 from group 1 and subject 10 from group 2
out.1 <- subjs.delete(out.1, subj = c(1, 10), group = c(1, 2))

## End(Not run)
```

subs.plot

Plot subjects raw data along with function fits

Description

Plot subjects raw data along with function fits

Usage

```
subs.plot(part1.list, legend.spot = "topright", ylim = NULL, groups = NA,
subjs = NA, curves = NA)
```

Arguments

part1.list Output from doubleGauss.fit or logistic.fit
 legend.spot string. Location of the legend
 ylim If NULL, takes the min and max of all observed curves
 groups Vector of groups corresponding to subjects to plot. Leave as NA for all
 subjs Vector of subjects to plot. Leave as NA for all
 curves Vector of curves to plot. Leave as NA for all

Details

Plot subjects raw data along with function fits

Value

NULL

Note

Options for legend.location include "topleft", "top", "topright", "right", "bottomright", "bottom", "bottomleft", "left"

Examples

```
## Not run:
data(ci)
ci.1 <- subset(ci, ci$LookType == "Target")
ci.1$Group <- ci.1$protocol
out.1 <- logistic.fit(ci.1, 4)ests.plot(out.1)
subs.plot(out.1)

## End(Not run)
```

 tsmultcomp

Calculate a corrected alpha

Description

Calculates an adjusted alpha to test against for timeseries datasets

Usage

```
tsmultcomp(rho, alpha, N, df, maxItr = 1000, tol = 1e-8, verbose = FALSE)
```

Arguments

rho	Output from doubleGauss.fit or logistic.fit
alpha	Familywise error rate.
N	Total number of tests being performed (i.e. length of the time series)
df	Degrees of freedom used for the individual t-tests.
maxItr	Maximum number of iterations to reach convergence for adjusted alpha estimate.
tol	Tolerance level for assessing convergence of the adjusted alpha.
verbose	If TRUE, prints adjusted alpha estimate for each iteration.

Details

Calculates an adjusted alpha to test against for timeseries datasets. This is intended for use with t-statistics. An AR1 autocorrelation structure is assumed.

Value

Numeric.

Note

There are no further notes

Examples

```
# Sample Data
x <- numeric(500)
x[1] <- rnorm(1)
for(i in 2:500) x[i] <- rnorm(1, x[i - 1] * .9, 1 - .9 ^ 2)

# Estimate autocorrelation
# Assume the x values are t-statistics based on 50 subjects
rho.est <- ar(x, FALSE, order.max = 1)$ar
alphastar <- tsmultcomp(rho.est, alpha = .05, N = 500, df = 49)
alphastar
```

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