

# Package ‘TcGSA’

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**Type** Package

**Title** Time-course Gene Set Analysis

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**Suggests** doSNOW

**Description** This package implements the TcGSA, an algorithm to analyze longitudinal gene-expression data at the gene set level.

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## R topics documented:

TcGSA-package . . . . .	2
clustTrend . . . . .	3
data_simu_TcGSA . . . . .	6
multtest.TcGSA . . . . .	8
plot.TcGSA . . . . .	9
plot1GS . . . . .	15
plotFit.GS . . . . .	19
plotPat.1GS . . . . .	22
plotPat.TcGSA . . . . .	26

plotSelect.GS . . . . .	32
pval_simu . . . . .	36
rmixchisq . . . . .	37
signifLRT.TcGSA . . . . .	38
summary.TcGSA . . . . .	39
TcGSA.LR . . . . .	41
TcGSA.LR.parallel . . . . .	44

<b>Index</b>	<b>48</b>
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TcGSA-package	<i>Time-course Gene Set Analysis</i>
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## Description

This package implements TcGSA, an algorithm to analyze longitudinal gene-expression data at the gene set level.

## Details

Package: TcGSA  
 Type: Package  
 Version: 0.9.8  
 Date: 2014-04-28  
 License: **LGPL-3**

The main function in this package is [TcGSA.LR](#) which performs Time-course Gene Set Analysis, and provide nice representations of its results (see [plot.TcGSA](#) and [plot1GS](#)).

## Author(s)

Boris P. Hejblum — Maintainer: Boris P. Hejblum

## References

Hejblum, B.P., Skinner, J., Thiebaut, R., 2014, TcGSA: a gene set approach for longitudinal gene expression data analysis, **submitted**.

## See Also

[GSA](#)

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clustTrend	<i>Cluster the genes dynamics into different dominant trends.</i>
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## Description

This function clusters the genes dynamics of one gene sets into different dominant trends. The optimal number of custers is computed thanks to the gap statistics. See [clusGap](#).

## Usage

```
clustTrend(x, expr, Subject_ID, TimePoint, threshold = 0.05, myproc = "BY",
  nbsimu_pval = 1e+06, baseline = NULL, only.signif = TRUE,
  group.var = NULL, Group_ID_paired = NULL, ref = NULL,
  group_of_interest = NULL, FUNcluster = NULL,
  clustering_metric = "euclidian", clustering_method = "ward", B = 100,
  max_trends = 4, aggreg.fun = "median", trend.fun = "median",
  methodOptiClust = "firstSEmax", indiv = "genes", verbose = TRUE)
```

```
## S3 method for class 'ClusteredTrends'
print(x, ...)
```

```
## S3 method for class 'ClusteredTrends'
plot(x, ...)
```

## Arguments

x	a <b>tcgsa</b> object for <code>clustTrend</code> , or a <b>ClusteredTrends</b> object for <code>print.ClusteredTrends</code> and <code>plot.ClusteredTrends</code> .
expr	either a matrix or dataframe of gene expression upon which dynamics are to be calculated, or a list of gene sets estimation of gene expression. In the case of a matrix or dataframe, its dimension are $n \times p$ , with the $p$ sample in column and the $n$ genes in row. In the case of a list, its length should correspond to the number of gene sets under scrutiny and each element should be an 3 dimension array of estimated gene expression, such as for the list returned in the 'Estimations' element of <a href="#">TcGSA.LR</a> . See details.
Subject_ID	a factor of length $p$ that is in the same order as the columns of <code>expr</code> (when it is a dataframe) and that contains the patient identifier of each sample.
TimePoint	a numeric vector or a factor of length $p$ that is in the same order as <code>Subject_ID</code> and the columns of <code>expr</code> (when it is a dataframe), and that contains the time points at which gene expression was measured.
threshold	the threshold at which the FDR or the FWER should be controlled. #'
myproc	a vector of character strings containing the names of the multiple testing procedures for which adjusted p-values are to be computed. This vector should include any of the following: "Bonferroni", "Holm", "Hochberg", "SidakSS",

	"SidakSD", "BH", "BY", "ABH", "TSBH". See <a href="#">mt.rawp2adjp</a> for details. Default is "BY", the Benjamini & Yekutieli (2001) step-up FDR-controlling procedure (general dependency structures). In order to control the FWER (in case of an analysis that is more a hypothesis confirmation than an exploration of the expression data), we recommend to use "Holm", the Holm (1979) step-down adjusted p-values for strong control of the FWER.
nbsimu_pval	the number of observations under the null distribution to be generated in order to compute the p-values. Default is 1e+06.
baseline	a character string which is the value of TimePoint that can be used as a baseline. Default is NULL, in which case no timepoint is used as a baseline value for gene expression. Has to be NULL when comparing two treatment groups.
only.signif	logical flag for analysing the trends in only the significant gene sets. If FALSE, all the gene sets from the <b>gmt</b> object contained in <b>x</b> are clustered. Default is TRUE.
group.var	in the case of several treatment groups, this is a factor of length <i>p</i> that is in the same order as Timepoint, Subject_ID and the columns of expr. It indicates to which treatment group each sample belongs to. Default is NULL, which means that there is only one treatment group.
Group_ID_paired	a character vector of length <i>p</i> that is in the same order as Timepoint, Subject_ID, group.var and the columns of expr. This argument must not be NULL in the case of a paired analysis, and must be NULL otherwise. Default is NULL.
ref	the group which is used as reference in the case of several treatment groups. Default is NULL, which means that reference is the first group in alphabetical order of the labels of group.var.
group_of_interest	the group of interest, for which dynamics are to be computed in the case of several treatment groups. Default is NULL, which means that group of interest is the second group in alphabetical order of the labels of group.var.
FUNcluster	the clustering function used to agglomerate genes in trends. Default is NULL, in which a hierarchical clustering is performed via the function <a href="#">agnes</a> , using the metric <code>clustering_metric</code> and the method <code>clustering_method</code> . See <a href="#">clusGap</a>
clustering_metric	character string specifying the metric to be used for calculating dissimilarities between observations in the hierarchical clustering when FUNcluster is NULL. The currently available options are "euclidean" and "manhattan". Default is "euclidean". See <a href="#">agnes</a> . Also, a "sts" option is available in TcGSA. It implements the 'Short Time Series' distance [Moller-Levet et al., Fuzzy Clustering of short time series and unevenly distributed sampling points, <i>Advances in Intelligent Data Analysis V:330-340 Springer, 2003</i> ] designed specifically for clustering time series.
clustering_method	character string defining the agglomerative method to be used in the hierarchical clustering when FUNcluster is NULL. The six methods implemented are "average" ([unweighted pair-]group average method, UPGMA), "single" (single linkage), "complete" (complete linkage), "ward" (Ward's method), "weighted" (weighted average linkage). Default is "ward". See <a href="#">agnes</a> .

B	integer specifying the number of Monte Carlo ("bootstrap") samples used to compute the gap statistics. Default is 500. See <a href="#">clusGap</a> .
max_trends	integer specifying the maximum number of different clusters to be tested. Default is 4.
aggreg.fun	a character string such as "mean", "median" or the name of any other defined statistics function that returns a single numeric value. It specifies the function used to aggregate the observations before the clustering. Default is to median. Default is "median".
trend.fun	a character string such as "mean", "median" or the name of any other function that returns a single numeric value. It specifies the function used to calculate the trends of the identified clustered. Default is to median.
methodOptiClust	character string indicating how the "optimal" number of clusters is computed from the gap statistics and their standard deviations. Possible values are "globalmax", "firstmax", "Tibs2001SEmax", "firstSEmax" and "globalSEmax". Default is "firstSEmax". See 'method' in <a href="#">clusGap</a> , Details and <i>Tibshirani et al., 2001</i> in References.
indiv	a character string indicating by which unit observations are aggregated (through <code>aggreg.fun</code> ) before the clustering. Possible values are "genes" or "patients". Default is "genes".
verbose	logical flag enabling verbose messages to track the computing status of the function. Default is TRUE.
...	further arguments passed to or from other methods.

## Details

If `expr` is a matrix or a dataframe, then the genes dynamics are clustered on the "original" data. On the other hand, if `expr` is a list returned in the 'Estimations' element of `TcGSA.LR`, then the dynamics are computed on the estimations made by the `TcGSA.LR` function.

This function uses the Gap statistics to determine the optimal number of clusters in the plotted gene set. See [clusGap](#).

## Value

An object of class `ClusteredTrends` which is a list with the 4 following components:

- `NbClust` a vector that contains the optimal number of clusters for each analysed gene sets.
- `ClustsMeds` a list of the same length as `NsClust` (the number of analysed gene sets). Each element of the list is a data frame, in which there is as many column as the optimal number of clusters for the corresponding gene sets for each cluster. Each column of the data frame contains the median trend values for the corresponding cluster.
- `GenesPartition` a list of the same length as `NsClust` (the number of analysed gene sets). Each element of the list is a vector which gives the partition of the genes inside the corresponding gene set.
- `MaxNbClust` an integer storing the maximum number of different clusters tested, as given by the argument 'max\_trends'.

**Author(s)**

Boris P. Hejblum

**References**

Tibshirani, R., Walther, G. and Hastie, T., 2001, Estimating the number of data clusters via the Gap statistic, *Journal of the Royal Statistical Society, Series B (Statistical Methodology)*, **63**, 2: 41–423.

**See Also**

[plot1GS](#), [TcGSA.LR](#), [clusGap](#)

**Examples**

```
data(data_simu_TcGSA)

tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                        subject_name="Patient_ID", time_name="TimePoint",
                        time_func="linear", crossedRandom=FALSE)

CT <- clustTrend(tcgsa_sim_1grp,
                expr=expr_1grp, Subject_ID=design$Subject_ID, TimePoint=design$TimePoint)
CT
plot(CT)

CT$NbClust
CT$NbClust["Gene set 5"]
CT$ClustMeds[["Gene set 4"]]
CT$ClustMeds[["Gene set 5"]]
```

---

data\_simu\_TcGSA

*Simulated Data for TcGSA*

---

**Description**

Simulated data for 5 genesets of 50 genes each. Gene expression is simulated at 5 time points for 10 patients.

**Usage**

```
data(data_simu_TcGSA)
```

**Details**

In `expr_1grp` all patients belong to the same unique treatment group. The first 2 gene sets are simulated under the null hypothesis. The gene sets 3 and 4 are simulated under the alternative hypothesis that there is a significant homogeneous time trend within the gene set. The gene set 5 is simulated under the alternative hypothesis that there are significant heterogeneous time trends within the gene

set.

In `expr_2grp` all patients belong to 2 treatment groups. The 5 first patients belong to the treatment group 'T', The 5 other patients belong to the treatment group 'C'. The first 2 gene sets are simulated under the null hypothesis that there is no difference in the time trend between the 2 treatment groups. The gene sets 3 and 4 are simulated under the alternative hypothesis that there are significantly different homogeneous time trends within the gene set between the 2 treatment groups. The gene set 5 is simulated under the alternative hypothesis that there are significantly different heterogeneous time trends between the 2 treatment groups within the gene set.

### Value

<code>expr_1grp</code>	See Details.
<code>expr_2grp</code>	See Details.
<code>design</code>	a data frame with 5 variables: <ul style="list-style-type: none"><li>• <code>Patient_ID</code>: a factor that contains the patient ID.</li><li>• <code>TimePoint</code>: a numeric vector or a factor that contains the time points at which gene expression was measured.</li><li>• <code>sample_name</code>: a character vector with the names of the sample (corresponding to the names of the columns of <code>expr_1grp</code> and of <code>expr_2grp</code>).</li><li>• <code>group.var</code>: a factor that indicates to which of the 2 treatment groups each sample belongs to.</li><li>• <code>Group_paired_ID</code> a random paired identifier for paired couples (one in each of the 2 treatment groups) of patients.</li></ul>
<code>gmt_sim</code>	a <b>gmt</b> object containing the gene sets definition. See <a href="#">GSA.read.gmt</a> and <a href="#">GMT definition on www.broadinstitute.org</a> .

### Author(s)

Boris P. Hejblum

### Source

This is simulated data.

### See Also

[TcGSA.LR](#)

### Examples

```
data(data_simu_TcGSA)
summary(expr_1grp)
summary(design)
gmt_sim
```

---

multtest.TcGSA	<i>Computing the P-value of the Likelihood Ratios Applying a Multiple Testing Correction</i>
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---

## Description

This function computes the p-value of the likelihood ratios and apply a multiple testing correction.

## Usage

```
multtest.TcGSA(tcgsa, threshold = 0.05, myproc = "BY",
  nbsimu_pval = 1e+06)
```

## Arguments

tcgsa	a TcGSA object.
threshold	the threshold at which the FDR or the FWER should be controlled.
myproc	a vector of character strings containing the names of the multiple testing procedures for which adjusted p-values are to be computed. This vector should include any of the following: "Bonferroni", "Holm", "Hochberg", "SidakSS", "SidakSD", "BH", "BY", "ABH", "TSBH" or "none". "none" indicates no adjustment for multiple testing. See <a href="#">mt.rawp2adjp</a> for details. Default is "BY", the Benjamini & Yekutieli (2001) step-up FDR-controlling procedure (general dependency structures). In order to control the FWER (in case of an analysis that is more a hypothesis confirmation than an exploration of the expression data), we recommend to use "Holm", the Holm (1979) step-down adjusted p-values for strong control of the FWER.
nbsimu_pval	the number of observations under the null distribution to be generated in order to compute the p-values. Default is 1e+06

## Value

multtest.TcGSA returns an dataframe with 5 variables. The rows correspond to the gene sets under scrutiny. The 1st column is the likelihood ratios LR, the 2nd column is the convergence status of the model under the null hypothesis CVG\_H0, the 3rd column is the convergence status of the model under the alternative hypothesis CVG\_H1, the 4th column is the raw p-value of the mixed likelihood ratio test raw\_pval, the 5th column is the adjusted p-value of the mixed likelihood ratio test adj\_pval.

## Author(s)

Boris P. Hejblum

## See Also

[TcGSA.LR](#), [mt.rawp2adjp](#), [signifLRT.TcGSA](#)



**Examples**

```

data(data_simu_TcGSA)

tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
  subject_name="Patient_ID", time_name="TimePoint",
  time_func="linear", crossedRandom=FALSE)

mtt <- multttest.TcGSA(tcgsa_sim_1grp, threshold = 0.05,
  myproc = "BY", nbsimu_pval = 1000)

mtt

```

---

plot.TcGSA                      *Plot a Gene Set Trends Heatmap.*

---

**Description**

This function plots a gene sets dynamic trends heatmap.

**Usage**

```

## S3 method for class 'TcGSA'
plot(x, threshold = 0.05, myproc = "BY",
  nbsimu_pval = 1e+06, expr, Subject_ID, TimePoint, baseline = NULL,
  only.signif = TRUE, group.var = NULL, Group_ID_paired = NULL,
  ref = NULL, group_of_interest = NULL, ranking = FALSE,
  FUNcluster = NULL, clustering_metric = "euclidian",
  clustering_method = "ward", B = 500, max_trends = 4,
  aggreg.fun = "median", methodOptiClust = "firstSEmax", indiv = "genes",
  verbose = TRUE, clust_trends = NULL, N_clusters = NULL,
  myclusters = NULL, label.clusters = NULL, prev_rowCL = NULL,
  descript = TRUE, plot = TRUE, color.vec = c("darkred", "#D73027",
  "#FC8D59", "snow", "#91BFDB", "#4575B4", "darkblue"), legend.breaks = NULL,
  label.column = NULL, time_unit = "", cex.label.row = 1,
  cex.label.column = 1, margins = c(5, 25), heatKey.size = 1,
  dendrogram.size = 1, heatmap.height = 1, heatmap.width = 1,
  cex.clusterKey = 1, cex.main = 1, horiz.clusterKey = TRUE,
  main = NULL, subtitle = NULL, ...)

```

**Arguments**

x	an object of class 'TcGSA'.
threshold	the threshold at which the FDR or the FWER should be controlled.
myproc	a vector of character strings containing the names of the multiple testing procedures for which adjusted p-values are to be computed. This vector should include any of the following: "Bonferroni", "Holm", "Hochberg", "SidakSS", "SidakSD", "BH", "BY", "ABH", "TSBH" or "none". "none" indicates no adjustment for multiple testing. See <a href="#">mt.rawp2adjp</a> for details. Default is "BY", the

Benjamini & Yekutieli (2001) step-up FDR-controlling procedure (general dependency structures). In order to control the FWER (in case of an analysis that is more a hypothesis confirmation than an exploration of the expression data), we recommend to use "Holm", the Holm (1979) step-down adjusted p-values for strong control of the FWER.

nbsimu_pval	the number of observations under the null distribution to be generated in order to compute the p-values. Default is 1e+06.
expr	either a matrix or dataframe of gene expression upon which dynamics are to be calculated, or a list of gene sets estimation of gene expression. In the case of a matrix or dataframe, its dimension are $n \times p$ , with the $p$ sample in column and the $n$ genes in row. In the case of a list, its length should correspond to the number of gene sets under scrutiny and each element should be an 3 dimension array of estimated gene expression, such as for the list returned in the 'Estimations' element of <code>TcGSA.LR</code> . See details.
Subject_ID	a factor of length $p$ that is in the same order as the columns of <code>expr</code> (when it is a dataframe) and that contains the patient identifier of each sample. Ignored if <code>expr</code> is a list of estimations.
TimePoint	a numeric vector or a factor of length $p$ that is in the same order as <code>Subject_ID</code> and the columns of <code>expr</code> (when it is a dataframe), and that contains the time points at which gene expression was measured. Ignored if <code>expr</code> is a list of estimations.
baseline	a character string which is the value of <code>TimePoint</code> used as baseline. See Details.
only.signif	logical flag for plotting only the significant gene sets. If FALSE, all the gene sets from the <code>gmt</code> object contained in <code>x</code> are plotted. Default is TRUE.
group.var	in the case of several treatment' groups, this is a factor of length $p$ that is in the same order as <code>Timepoint</code> , <code>Subject_ID</code> , <code>sample_name</code> and the columns of <code>expr</code> . It indicates to which treatment group each sample belongs to. Default is NULL, which means that there is only one treatment group. See Details.
Group_ID_paired	a character vector of length $p$ that is in the same order as <code>Timepoint</code> , <code>Subject_ID</code> , <code>sample_name</code> , <code>group.var</code> and the columns of <code>expr</code> . This argument must not be NULL in the case of a paired analysis, and must be NULL otherwise. Default is NULL. See Details.
ref	the group which is used as reference in the case of several treatment groups. Default is NULL, which means that reference is the first group in alphabetical order of the labels of <code>group.var</code> . See Details.
group_of_interest	the group of interest, for which dynamics are to be computed in the case of several treatment groups. Default is NULL, which means that group of interest is the second group in alphabetical order of the labels of <code>group.var</code> . See Details. here~~
ranking	a logical flag. If TRUE, the gene set trends are not hierarchically classified, but ordered by decreasing Likelihood ratios. Default is FALSE.
FUNcluster	the clustering function used to agglomerate genes in trends. Default is NULL, in which a hierachical clustering is performed via the function <code>agnes</code> , using the metric <code>clustering_metric</code> and the method <code>clustering_method</code> . See <code>clusGap</code>

clustering_metric	character string specifying the metric to be used for calculating dissimilarities between observations in the hierarchical clustering when FUNcluster is NULL. The currently available options are "euclidean" and "manhattan". Default is "euclidean". See <a href="#">agnes</a> . Also, a "sts" option is available in TcGSA. It implements the 'Short Time Series' distance [Moller-Levet et al., Fuzzy CLustering of short time series and unevenly distributed sampling points, <i>Advances in Intelligent Data Analysis V:330-340</i> Springer, 2003] designed specifically for clustering time series.
clustering_method	character string defining the agglomerative method to be used in the hierarchical clustering when FUNcluster is NULL. The six methods implemented are "average" ([unweighted pair-]group average method, UPGMA), "single" (single linkage), "complete" (complete linkage), "ward" (Ward's method), "weighted" (weighted average linkage). Default is "ward". See <a href="#">agnes</a> .
B	integer specifying the number of Monte Carlo ("bootstrap") samples used to compute the gap statistics. Default is 500. See <a href="#">clusGap</a> .
max_trends	integer specifying the maximum number of different clusters to be tested. Default is 4.
aggreg.fun	a character string such as "mean", "median" or the name of any other statistics function defined that returns a single numeric value. It specifies the function used to aggregate the observations before the clustering. Default is to median. Default is "median".
methodOptiClust	character string indicating how the "optimal" number of clusters is computed from the gap statistics and their standard deviations. Possible values are "globalmax", "firstmax", "Tibs2001SEmax", "firstSEmax" and "globalSEmax". Default is "firstSEmax". See 'method' in <a href="#">clusGap</a> , Details and <i>Tibshirani et al., 2001</i> in References.
indiv	a character string indicating by which unit observations are aggregated (through <code>aggreg.fun</code> ) before the clustering. Possible values are "genes" or "patients". Default is "genes".
verbose	logical flag enabling verbose messages to track the computing status of the function. Default is TRUE.
clust_trends	object of class <b>ClusteredTrends</b> containing already computed trends for the plotted gene sets. Default is NULL.
N_clusters	an integer that is the number of clusters in which the dynamics should be re-grouped. The cutoff of the clustering tree is automatically calculated accordingly. Default is NULL, in which case the dendrogram is not cut and no clusters are identified.
myclusters	a character vector of colors for predefined clusters of the represented genesets, with as many levels as the value of <code>N_clusters</code> . Default is NULL, in which case the clusters are automatically identified and colored via the <a href="#">cutree</a> function and the <code>N_clusters</code> argument only.
label.clusters	if <code>N_clusters</code> is not NULL, a character vector of length <code>N_clusters</code> ss. Default is NULL, in which case if <code>N_clusters</code> is not NULL, clusters are simply labelled with numbers.

prev_rowCL	a <b>hclust</b> object, such as the one return by the present plotting function (see Value) for instance. If not NULL, no clustering is calculated by the present plotting function and this tree is used to represent the gene sets dynamics. Default is NULL.
descript	logical flag indicating that the description of the gene sets should appear after their name on the right side of the plot if TRUE. Default is TRUE. See Details.
plot	logical flag indicating wether the heatmap should be plotted or not. Default is TRUE.
color.vec	a character strings vector used to define the color <a href="#">palette</a> used in the plot. Default is <code>c("#D73027", "#FC8D59", "lightyellow", "#91BFDB", "#4575B4")</code> .
legend.breaks	a numeric vector indicating the splitting points for coloring. Default is NULL, in which case the break points will be spaced equally and symmetrically about 0.
label.column	a vector of character strings with the labels to be displayed for the columns (i.e. the time points). Default is NULL.
time_unit	the time unit to be displayed (such as "Y", "M", "W", "D", "H", etc) next to the values of TimePoint in the columns labels when label.column is NULL. Default is "".
cex.label.row	a numerical value giving the amount by which row labels text should be magnified relative to the default 1.
cex.label.column	a numerical value giving the amount by which column labels text should be magnified relative to the default 1.
margins	numeric vector of length 2 containing the margins (see <a href="#">par</a> (mar=*)) for column and row names, respectively. Default is <code>c(15, 100)</code> . See Details.
heatKey.size	the size of the color key for the heatmap fill. Default is 1.
dendrogram.size	the horizontal size of the dendrogram. Default is 1
heatmap.height	the height of the heatmap. Default is 1
heatmap.width	the width of the heatmap. Default is 1
cex.clusterKey	a numerical value giving the amount by which the clusters legend text should be magnified relative to the default 1, when N_clusters is not NULL.
cex.main	a numerical value giving the amount by which title text should be magnified relative to the default 1.
horiz.clusterKey	a logical flag; if TRUE, set the legend for clusters horizontally rather than vertically. Only used if the N_clusters argument is not NULL. Default is TRUE.
main	a character string for an optionnal title. Default is NULL.
subtitle	a character string for an optionnal subtitle. Default is NULL.
...	other parameters to be passed through to plotting functions.

## Details

On the heatmap, each line corresponds to a gene set, and each column to a timepoint.

If `expr` is a matrix or a dataframe, then the "original" data are plotted. On the other hand, if `expr` is a list returned in the 'Estimations' element of `TcGSA.LR`, then it is those "estimations" made by the `TcGSA.LR` function that are plotted.

If `descript` is `FALSE`, the second element of `margins` can be reduced (for instance use `margins = c(5, 10)`), as there is not so much need for space in order to display only the gene set names, without their description.

If there is a large number of significant gene sets, the hierarchical clustering step repeated for each of them can take a few minutes. To speed things up (especially) when playing with the plotting parameters for having a nice plot, one can run the `clustTrend` function beforehand, and plug its results in the `plot.TcGSA` function via the `clust_trends` argument.

## Value

An object of class `hclust` which describes the tree produced by the clustering process. The object is a list with components:

- `merge` an  $n - 1$  by 2 matrix. Row  $i$  of `merge` describes the merging of clusters at step  $i$  of the clustering. If an element  $j$  in the row is negative, then observation  $-j$  was merged at this stage. If  $j$  is positive then the merge was with the cluster formed at the (earlier) stage  $j$  of the algorithm. Thus negative entries in `merge` indicate agglomerations of singletons, and positive entries indicate agglomerations of non-singletons.
- `height` a set of  $n - 1$  real values (non-decreasing for ultrametric trees). The clustering height: that is, the value of the criterion associated with the Ward clustering method.
- `order` a vector giving the permutation of the original observations suitable for plotting, in the sense that a cluster plot using this ordering and matrix `merge` will not have crossings of the branches.
- `labels` the gene set trends name.
- `call` the call which produced the result clustering:  
`hclust(d = dist(map2heat, method = "euclidean"), method = "ward.D2")`
- `method` "ward.D2", as it is the clustering method that has been used for clustering the gene set trends.
- `dist.method` "euclidean", as it is the distance that has been used for clustering the gene set trends.
- `legend.breaks` a numeric vector giving the splitting points used for coloring the heatmap. If `plot` is `FALSE`, then it is `NULL`.
- `myclusters` a character vector of colors for the dynamic clusters of the represented gene set trends, with as many levels as the value of `N_clusters`. If no dynamic clusters were represented, then this is `NULL`.
- `ddr` a **dendrogram** object with the reordering used for the heatmap. See [heatmap.2](#).
- `geneset.names` character vector with the names of the gene sets used in the heatmap.
- `clust.trends` a **ClusteredTrends** object.
- `clustersExport` a data frame with 2 variables containing the two following variables :

- GeneSet: the gene set trends clustered.
- Cluster: the dynamic cluster they belong to.

The data frame is order by the variable Cluster.

- data\_plotted: the data matrix represented by the heatmap

### Author(s)

Boris P. Hejblum

### References

Hejblum, B.P., Skinner, J., Thiebaut, R., 2014, TcGSA: a gene set approach for longitudinal gene expression data analysis, **submitted**.

### See Also

[heatmap.2](#), [TcGSA.LR](#), [hclust](#)

### Examples

```
data(data_simu_TcGSA)

tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                        subject_name="Patient_ID", time_name="TimePoint",
                        time_func="linear", crossedRandom=FALSE)
summary(tcgsa_sim_1grp)

plot(x=tcgsa_sim_1grp, expr=tcgsa_sim_1grp$Estimations,
     Subject_ID=design$Patient_ID, TimePoint=design$TimePoint,
     baseline=1,
     B=100,
     time_unit="H",
     dendrogram.size=0.4, heatmap.width=0.8, heatmap.height=2, cex.main=0.7
     )

## Not run:
tcgsa_sim_2grp <- TcGSA.LR(expr=expr_2grp, gmt=gmt_sim, design=design,
                        subject_name="Patient_ID", time_name="TimePoint",
                        time_func="linear", crossedRandom=FALSE,
                        group_name="group.var")
summary(tcgsa_sim_2grp)
plot(x=tcgsa_sim_2grp, expr=expr_2grp,
     Subject_ID=design$Patient_ID, TimePoint=design$TimePoint,
     B=100,
     time_unit="H",
     )

## End(Not run)
```

plot1GS

*Plotting a Specific Gene Set***Description**

This function can plot different representations of the gene expression in a specific gene set.

**Usage**

```
plot1GS(expr, gmt, Subject_ID, TimePoint, geneset.name, baseline = NULL,
  group.var = NULL, Group_ID_paired = NULL, ref = NULL,
  group_of_interest = NULL, FUNcluster = NULL,
  clustering_metric = "euclidian", clustering_method = "ward", B = 500,
  max_trends = 4, aggreg.fun = "median", trend.fun = "median",
  methodOptiClust = "firstSEmax", indiv = "genes", verbose = TRUE,
  clustering = TRUE, showTrend = TRUE, smooth = TRUE, time_unit = "",
  title = NULL, y.lab = NULL, desc = TRUE, lab.cex = 1, axis.cex = 1,
  main.cex = 1, y.lab.angle = 90, x.axis.angle = 45, y.lim = NULL,
  x.lim = NULL, gg.add = list(theme()), plot = TRUE)
```

**Arguments**

expr	either a matrix or dataframe of gene expression upon which dynamics are to be calculated, or a list of gene sets estimation of gene expression. In the case of a matrix or dataframe, its dimension are $n \times p$ , with the $p$ sample in column and the $n$ genes in row. In the case of a list, its length should correspond to the number of gene sets under scrutiny and each element should be an 3 dimension array of estimated gene expression, such as for the list returned in the 'Estimations' element of <a href="#">TcGSA.LR</a> . See details.
gmt	a <b>gmt</b> object containing the gene sets definition. See <a href="#">GSA.read.gmt</a> and definition on <a href="http://www.broadinstitute.org">www.broadinstitute.org</a> .
Subject_ID	a factor of length $p$ that is in the same order as the columns of expr (when it is a dataframe) and that contains the patient identifier of each sample.
TimePoint	a numeric vector or a factor of length $p$ that is in the same order as TimePoint and the columns of expr (when it is a dataframe), and that contains the time points at which gene expression was measured.
geneset.name	a character string containing the name of the gene set to be plotted, that must appear in the "geneset.names" element of gmt.
baseline	a character string which is the value of TimePoint that can be used as a baseline. Default is NULL, in which case no timepoint is used as a baseline value for gene expression. Has to be NULL when comparing two treatment groups.
group.var	in the case of several treatment groups, this is a factor of length $p$ that is in the same order as Timepoint, Subject_ID and the columns of expr. It indicates to which treatment group each sample belongs to. Default is NULL, which means that there is only one treatment group.

Group_ID_paired	a character vector of length $p$ that is in the same order as Timepoint, Subject_ID, group.var and the columns of expr. This argument must not be NULL in the case of a paired analysis, and must be NULL otherwise. Default is NULL.
ref	the group which is used as reference in the case of several treatment groups. Default is NULL, which means that reference is the first group in alphabetical order of the labels of group.var. See Details.
group_of_interest	the group of interest, for which dynamics are to be computed in the case of several treatment groups. Default is NULL, which means that group of interest is the second group in alphabetical order of the labels of group.var.
FUNcluster	a function which accepts as first argument a matrix $x$ and as second argument the number of clusters desired $k$ , and which returns a list with a component named 'cluster' which is a vector of length $n = \text{nrow}(x)$ of integers in $1:k$ , determining the clustering or grouping of the $n$ observations. Default is NULL, in which case a hierarchical clustering is performed via the function <a href="#">agnes</a> , using the metric clustering_metric and the method clustering_method. See 'FUNcluster' in <a href="#">clusGap</a> and Details.
clustering_metric	character string specifying the metric to be used for calculating dissimilarities between observations in the hierarchical clustering when FUNcluster is NULL. The currently available options are "euclidean" and "manhattan". Default is "euclidean". See <a href="#">agnes</a> . Also, a "sts" option is available in TcGSA. It implements the 'Short Time Series' distance [Moller-Levet et al., Fuzzy CLustering of short time series and unevenly distributed sampling points, <i>Advances in Intelligent Data Analysis V:330-340</i> Springer, 2003] designed specifically for clustering time series.
clustering_method	character string defining the agglomerative method to be used in the hierarchical clustering when FUNcluster is NULL. The six methods implemented are "average" ([unweighted pair-]group average method, UPGMA), "single" (single linkage), "complete" (complete linkage), "ward" (Ward's method), "weighted" (weighted average linkage). Default is "ward". See <a href="#">agnes</a> .
B	integer specifying the number of Monte Carlo ("bootstrap") samples used to compute the gap statistics. Default is 500. See <a href="#">clusGap</a> .
max_trends	integer specifying the maximum number of different clusters to be tested. Default is 4.
aggreg.fun	a character string such as "mean", "median" or the name of any other defined statistics function that returns a single numeric value. It specifies the function used to aggregate the observations before the clustering. Default is to median.
trend.fun	a character string such as "mean", "median" or the name of any other function that returns a single numeric value. It specifies the function used to calculate the trends of the identified clustered. Default is to median.
methodOptiClust	character string indicating how the "optimal" number of clusters is computed from the gap statistics and their standard deviations. Possible values are "globalmax",



	"firstmax", "Tibs2001SEmax", "firstSEmax" and "globalSEmax". Default is "firstSEmax". See 'method' in <a href="#">clusGap</a> , Details and <i>Tibshirani et al., 2001</i> in References.
indiv	a character string indicating by which unit observations are aggregated (through <code>aggred.fun</code> ) before the clustering. Possible values are "genes" or "patients". Default is "genes". See Details.
verbose	logical flag enabling verbose messages to track the computing status of the function. Default is TRUE.
clustering	logical flag. If FALSE, there is no clustering representation; if TRUE, the lines are colored according to which cluster they belong to. Default is TRUE. See Details.
showTrend	logical flag. If TRUE, a black line is added for each cluster, representing the corresponding trend.fun. Default is TRUE.
smooth	logical flag. If TRUE and <code>showTrend</code> is also TRUE, the representation of each cluster trend.fun is smoothed using cubic polynoms (see <a href="#">stat_smooth</a> ). Default is TRUE.
time_unit	the time unit to be displayed (such as "Y", "M", "W", "D", "H", etc) next to the values of <code>TimePoint</code> on the x-axis. Default is "", in which case the time scale on the x-axis is proportional to the time values.
title	character specifying the title of the plot. If NULL, a title is automatically generated, if "", no title appears. Default is NULL.
y.lab	character specifying the annotation of the y axis. If NULL, an annotation is automatically generated, if "", no annotation appears. Default is NULL.
desc	a logical flag. If TRUE, a line is added to the title of the plot with the description of the gene set plotted (from the <code>gmt</code> file). Default is TRUE.
lab.cex	a numerical value giving the amount by which lab labels text should be magnified relative to the default 1.
axis.cex	a numerical value giving the amount by which axis annotation text should be magnified relative to the default 1.
main.cex	a numerical value giving the amount by which title text should be magnified relative to the default 1.
y.lab.angle	a numerical value (in [0, 360]) giving the orientation by which y-label text should be turned (anti-clockwise). Default is 90. See <a href="#">element_text</a> .
x.axis.angle	a numerical value (in [0, 360]) giving the orientation by which x-axis annotation text should be turned (anti-clockwise). Default is 45.
y.lim	a numeric vector of length 2 giving the range of the y-axis. See <a href="#">plot.default</a> .
x.lim	if numeric, will create a continuous scale, if factor or character, will create a discrete scale. Observations not in this range will be dropped. See <a href="#">xlim</a> .
gg.add	A list of instructions to add to the <code>ggplot2</code> instruction. See <a href="#">+gg</a> . Default is <code>list(theme())</code> , which adds nothing to the plot.
plot	logical flag. If FALSE, no plot is drawn. Default is TRUE.

**Details**

If `expr` is a matrix or a dataframe, then the "original" data are plotted. On the other hand, if `expr` is a list returned in the 'Estimations' element of `TcGSA.LR`, then it is those "estimations" made by the `TcGSA.LR` function that are plotted.

If `indiv` is 'genes', then each line of the plot is the median of a gene expression over the patients. On the other hand, if `indiv` is 'patients', then each line of the plot is the median of a patient genes expression in this gene set.

This function uses the Gap statistics to determine the optimal number of clusters in the plotted gene set. See [clusGap](#).

**Value**

A dataframe the 2 following variables:

- ProbeID which contains the IDs of the probes of the plotted gene set.
- Cluster which to which cluster the probe belongs to.

If clustering is FALSE, then Cluster is NA for all the probes.

**Author(s)**

Boris P. Hejblum

**References**

Tibshirani, R., Walther, G. and Hastie, T., 2001, Estimating the number of data clusters via the Gap statistic, *Journal of the Royal Statistical Society, Series B (Statistical Methodology)*, **63**, 2: 411–423.

**See Also**

[ggplot2](#), [clusGap](#)

**Examples**

```
data(data_simu_TcGSA)
tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                          subject_name="Patient_ID", time_name="TimePoint",
                          time_func="linear", crossedRandom=FALSE)

plot1GS(expr=expr_1grp, TimePoint=design$TimePoint,
        Subject_ID=design$Patient_ID, gmt=gmt_sim,
        geneset.name="Gene set 4",
        indiv="genes", clustering=FALSE,
        time_unit="H",
        lab.cex=0.7)

## Not run:
plot1GS(expr=expr_1grp, TimePoint=design$TimePoint,
        Subject_ID=design$Patient_ID, gmt=gmt_sim,
```

```

        geneset.name="Gene set 5",
        indiv="patients", clustering=FALSE, baseline=1,
        time_unit="H",
        lab.cex=0.7)

## End(Not run)
## Not run:
plot1GS(expr=tcgsa_sim_1grp$Estimations, TimePoint=design$TimePoint,
        Subject_ID=design$Patient_ID, gmt=gmt_sim,
        geneset.name="Gene set 5",
        indiv="genes",
        time_unit="H",
        lab.cex=0.7
)

## End(Not run)

## Not run:
colval <- c(hsv(0.56, 0.9, 1),
            hsv(0, 0.27, 1),
            hsv(0.52, 1, 0.5),
            hsv(0, 0.55, 0.97),
            hsv(0.66, 0.15, 1),
            hsv(0, 0.81, 0.55),
            hsv(0.7, 1, 0.7),
            hsv(0.42, 0.33, 1)
)
n <- length(colval); y <- 1:n
op <- par(mar=rep(1.5,4))
plot(y, axes = FALSE, frame.plot = TRUE,
     xlab = "", ylab = "", pch = 21, cex = 8,
     bg = colval, ylim=c(-1,n+1), xlim=c(-1,n+1),
     main = "Color scale"
)
par(op)

require(ggplot2)
plot1GS(expr=expr_1grp, TimePoint=design$TimePoint,
        Subject_ID=design$Patient_ID, gmt=gmt_sim,
        geneset.name="Gene set 5",
        indiv="genes",
        time_unit="H",
        title="",
        gg.add=list(scale_color_manual(values=colval),
                    guides(colour = guide_legend(reverse=TRUE))),
        lab.cex=0.7
)

## End(Not run)

```

---

plotFit.GS

*Plotting function for exploring the fitness of the mixed modeling used in TcGSA*

---

## Description

This function plots graphs informing on the fit of the mixed modeling of the gene expression performed in TcGSA, for 1 or several gene sets.

## Usage

```
plotFit.GS(x, expr, design, subject_name = "Patient_ID",
           time_name = "TimePoint", colnames_ID, plot_type = c("Fit",
           "Residuals Obs", "Residuals Est", "Histogram Obs"), GeneSetsList,
           color = c("genes", "time", "subjects"), marginal_hist = TRUE,
           gg.add = list(theme()))
```

## Arguments

x	a <b>tgsa</b> object for <code>clustTrend</code> , or a <b>ClusteredTrends</b> object for <code>print.ClusteredTrends</code> and <code>plot.ClusteredTrends</code> .
expr	a matrix or dataframe of gene expression. Its dimension are $n \times p$ , with the $p$ samples in column and the $n$ genes in row.
design	a matrix or dataframe containing the experimental variables that used in the model, namely <code>subject_name</code> , <code>time_name</code> , and <code>covariates_fixed</code> and <code>time_covariates</code> if applicable. Its dimension are $p \times m$ and its row are is in the same order as the columns of <code>expr</code> .
subject_name	the name of the factor variable from <code>design</code> that contains the information on the repetition units used in the mixed model, such as the patient identifiers for instance. Default is 'Patient_ID'. See Details.
time_name	the name of a numeric variable from <code>design</code> that contains the information on the time replicates (the time points at which gene expression was measured). Default is 'TimePoint'. See Details.
colnames_ID	the name of the variable from <code>design</code> that contains the columnnames of the <code>expr</code> expression data matrix. See Details.
plot_type	a character string indicating the type of plot to be drawn. The options are 'Fit', 'Residuals Obs', 'Residuals Est' or 'Histogram Obs'.
GeneSetsList	a character string containing the names of the gene set whose fit is being checked. If several gene sets are being checked, can be a character list or vector of the names of those gene sets.
color	a character string indicating which color scale should be used. One of the 3 : 'genes', 'time', 'subjects', otherwise, no coloring is used.
marginal_hist	a logical flag indicating whether marginal histograms should be drawn. Only used for 'Fit' plot type. Default is 'TRUE'
gg.add	A list of instructions to add to the <code>ggplot2</code> instruction. See <code>+gg</code> . Default is <code>list(theme())</code> , which adds nothing to the plot.

**Details**

#'

**Author(s)**

Boris P. Hejblum

**References**

Hejblum, B.P., Skinner, J., Thiebaut, R., 2014, TcGSA: a gene set approach for longitudinal gene expression data analysis, **submitted**.

**See Also**

[plot1GS](#), [plotSelect.GS](#)

**Examples**

```
data(data_simu_TcGSA)

tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                        subject_name="Patient_ID", time_name="TimePoint",
                        time_func="linear", crossedRandom=FALSE)

require(ggplot2)
plotFit.GS(x=tcgsa_sim_1grp, expr=expr_1grp, design=design,
          subject_name="Patient_ID", time_name="TimePoint",
          colnames_ID="Sample_name",
          plot_type="Residuals Obs",
          GeneSetsList=c("Gene set 1", "Gene set 2", "Gene set 3",
                        "Gene set 4", "Gene set 5"),
          color="genes", gg.add=list(guides(color=FALSE))
)

## Not run:
plotFit.GS(x=tcgsa_sim_1grp, expr=expr_1grp, design=design,
          subject_name="Patient_ID", time_name="TimePoint",
          colnames_ID="Sample_name",
          plot_type="Histogram Obs",
          GeneSetsList=c("Gene set 1", "Gene set 5"),
          color="genes", gg.add=list(guides(fill=FALSE))
)

plotFit.GS(x=tcgsa_sim_1grp, expr=expr_1grp, design=design,
          subject_name="Patient_ID", time_name="TimePoint",
          colnames_ID="Sample_name",
          plot_type="Histogram Obs",
          GeneSetsList=c("Gene set 1", "Gene set 2", "Gene set 3",
                        "Gene set 4", "Gene set 5"),
          color="genes")
)

## End(Not run)
```

plotPat.1GS

*Plotting a Specific Gene Set Stratifying on Patients***Description**

This function can plot different representations of the gene expression in a specific gene set, stratified on all subjects.

**Usage**

```
plotPat.1GS(expr, gmt, Subject_ID, TimePoint, geneset.name, baseline = NULL,
  group.var = NULL, Group_ID_paired = NULL, ref = NULL,
  group_of_interest = NULL, FUNcluster = NULL,
  clustering_metric = "euclidian", clustering_method = "ward", B = 500,
  max_trends = 4, aggreg.fun = "median", trend.fun = "median",
  methodOptiClust = "firstSEmax", verbose = TRUE, clustering = TRUE,
  time_unit = "", title = NULL, y.lab = NULL, desc = TRUE,
  lab.cex = 1, axis.cex = 1, main.cex = 1, y.lab.angle = 90,
  x.axis.angle = 45, y.lim = NULL, x.lim = NULL, gg.add = list(theme()))
```

**Arguments**

expr	either a matrix or dataframe of gene expression upon which dynamics are to be calculated, or a list of gene sets estimation of gene expression. In the case of a matrix or dataframe, its dimension are $n \times p$ , with the $p$ sample in column and the $n$ genes in row. In the case of a list, its length should correspond to the number of gene sets under scrutiny and each element should be an 3 dimension array of estimated gene expression, such as for the list returned in the 'Estimations' element of <a href="#">TcGSA.LR</a> . See details.
gmt	a <b>gmt</b> object containing the gene sets definition. See <a href="#">GSA.read.gmt</a> and definition on <a href="http://www.broadinstitute.org">www.broadinstitute.org</a> .
Subject_ID	a factor of length $p$ that is in the same order as the columns of expr (when it is a dataframe) and that contains the patient identifier of each sample.
TimePoint	a numeric vector or a factor of length $p$ that is in the same order as TimePoint and the columns of expr (when it is a dataframe), and that contains the time points at which gene expression was measured.
geneset.name	a character string containing the name of the gene set to be plotted, that must appear in the "geneset.names" element of gmt.
baseline	a character string which is the value of TimePoint that can be used as a baseline. Default is NULL, in which case no timepoint is used as a baseline value for gene expression. Has to be NULL when comparing two treatment groups.
group.var	in the case of several treatment groups, this is a factor of length $p$ that is in the same order as Timepoint, Subject_ID and the columns of expr. It indicates to which treatment group each sample belongs to. Default is NULL, which means that there is only one treatment group. See Details.

Group_ID_paired	a character vector of length $p$ that is in the same order as Timepoint, Subject_ID, group.var and the columns of expr. This argument must not be NULL in the case of a paired analysis, and must be NULL otherwise. Default is NULL.
ref	the group which is used as reference in the case of several treatment groups. Default is NULL, which means that reference is the first group in alphabetical order of the labels of group.var. See Details.
group_of_interest	the group of interest, for which dynamics are to be computed in the case of several treatment groups. Default is NULL, which means that group of interest is the second group in alphabetical order of the labels of group.var.
FUNcluster	a function which accepts as first argument a matrix $x$ and as second argument the number of clusters desired $k$ , and which returns a list with a component named 'cluster' which is a vector of length $n = \text{nrow}(x)$ of integers in $1:k$ , determining the clustering or grouping of the $n$ observations. Default is NULL, in which case a hierarchical clustering is performed via the function <a href="#">agnes</a> , using the metric clustering_metric and the method clustering_method. See 'FUNcluster' in <a href="#">clusGap</a> and Details.
clustering_metric	character string specifying the metric to be used for calculating dissimilarities between observations in the hierarchical clustering when FUNcluster is NULL. The currently available options are "euclidean" and "manhattan". Default is "euclidean". See <a href="#">agnes</a> . Also, a "sts" option is available in TcGSA. It implements the 'Short Time Series' distance [Moller-Levet et al., Fuzzy CLustering of short time series and unevenly distributed sampling points, <i>Advances in Intelligent Data Analysis V:330-340</i> Springer, 2003] designed specifically for clustering time series.
clustering_method	character string defining the agglomerative method to be used in the hierarchical clustering when FUNcluster is NULL. The six methods implemented are "average" ([unweighted pair-]group average method, UPGMA), "single" (single linkage), "complete" (complete linkage), "ward" (Ward's method), "weighted" (weighted average linkage). Default is "ward". See <a href="#">agnes</a> .
B	integer specifying the number of Monte Carlo ("bootstrap") samples used to compute the gap statistics. Default is 500. See <a href="#">clusGap</a> .
max_trends	integer specifying the maximum number of different clusters to be tested. Default is 4.
aggreg.fun	a character string such as "mean", "median" or the name of any other defined statistics function that returns a single numeric value. It specifies the function used to aggregate the observations before the clustering. Default is to median.
trend.fun	a character string such as "mean", "median" or the name of any other function that returns a single numeric value. It specifies the function used to calculate the trends of the identified clustered. Default is to median.
methodOptiClust	character string indicating how the "optimal" number of clusters is computed from the gap statistics and their standard deviations. Possible values are "globalmax",

	"firstmax", "Tibs2001SEmax", "firstSEmax" and "globalSEmax". Default is "firstSEmax". See 'method' in <a href="#">clusGap</a> , Details and <i>Tibshirani et al., 2001</i> in References.
verbose	logical flag enabling verbose messages to track the computing status of the function. Default is TRUE.
clustering	logical flag. If FALSE, there is no clustering representation; if TRUE, the lines are colored according to which cluster they belong to. Default is TRUE. See Details.
time_unit	the time unit to be displayed (such as "Y", "M", "W", "D", "H", etc) next to the values of TimePoint on the x-axis. Default is "".
title	character specifying the title of the plot. If NULL, a title is automatically generated, if "", no title appears. Default is NULL.
y.lab	character specifying the annotation of the y axis. If NULL, an annotation is automatically generated, if "", no annotation appears. Default is NULL.
desc	a logical flag. If TRUE, a line is added to the title of the plot with the description of the gene set plotted (from the gmt file). Default is TRUE.
lab.cex	a numerical value giving the amount by which lab labels text should be magnified relative to the default 1.
axis.cex	a numerical value giving the amount by which axis annotation text should be magnified relative to the default 1.
main.cex	a numerical value giving the amount by which title text should be magnified relative to the default 1.
y.lab.angle	a numerical value (in [0, 360]) giving the orientation by which y-label text should be turned (anti-clockwise). Default is 90. See <a href="#">element_text</a> .
x.axis.angle	a numerical value (in [0, 360]) giving the orientation by which x-axis annotation text should be turned (anti-clockwise). Default is 45.
y.lim	a numeric vector of length 2 giving the range of the y-axis. See <a href="#">plot.default</a> .
x.lim	if numeric, will create a continuous scale, if factor or character, will create a discrete scale. Observations not in this range will be dropped. See <a href="#">xlim</a> .
gg.add	A list of instructions to add to the ggplot2 instruction. See <a href="#">+gg</a> . Default is <code>list(theme())</code> , which adds nothing to the plot.

## Details

If `expr` is a matrix or a dataframe, then the "original" data are plotted. On the other hand, if `expr` is a list returned in the 'Estimations' element of `TcGSA.LR`, then it is those "estimations" made by the `TcGSA.LR` function that are plotted.

If `indiv` is 'genes', then each line of the plot is the median of a gene expression over the patients. On the other hand, if `indiv` is 'patients', then each line of the plot is the median of a patient genes expression in this gene set.

This function uses the Gap statistics to determine the optimal number of clusters in the plotted gene set. See [clusGap](#).



**Value**

A dataframe the 2 following variables:

- ProbeID which contains the IDs of the probes of the plotted gene set.
- Cluster which to which cluster the probe belongs to.

If clustering is FALSE, then Cluster is NA for all the probes.

**Author(s)**

Boris P. Hejblum

**References**

Tibshirani, R., Walther, G. and Hastie, T., 2001, Estimating the number of data clusters via the Gap statistic, *Journal of the Royal Statistical Society, Series B (Statistical Methodology)*, **63**, 2: 411–423.

**See Also**

[ggplot2](#), [clusGap](#)

**Examples**

```
data(data_simu_TcGSA)
tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                        subject_name="Patient_ID", time_name="TimePoint",
                        time_func="linear", crossedRandom=FALSE)

plotPat.1GS(expr=expr_1grp, TimePoint=design$TimePoint,
            Subject_ID=design$Patient_ID, gmt=gmt_sim,
            geneset.name="Gene set 4",
            clustering=FALSE,
            time_unit="H",
            lab.cex=0.7)

## Not run:
plotPat.1GS(expr=expr_1grp, TimePoint=design$TimePoint,
            Subject_ID=design$Patient_ID, gmt=gmt_sim,
            geneset.name="Gene set 4",
            clustering=FALSE, baseline=1,
            time_unit="H",
            lab.cex=0.7)

## End(Not run)

## Not run:
colval <- c(hsv(0.56, 0.9, 1),
            hsv(0, 0.27, 1),
            hsv(0.52, 1, 0.5),
            hsv(0, 0.55, 0.97),
            hsv(0.66, 0.15, 1),
```

```

        hsv(0, 0.81, 0.55),
        hsv(0.7, 1, 0.7),
        hsv(0.42, 0.33, 1)
    )
n <- length(colval); y <- 1:n
op <- par(mar=rep(1.5,4))
plot(y, axes = FALSE, frame.plot = TRUE,
     xlab = "", ylab = "", pch = 21, cex = 8,
     bg = colval, ylim=c(-1,n+1), xlim=c(-1,n+1),
     main = "Color scale"
    )
par(op)

require(ggplot2)
plotPat.1GS(expr=expr_1grp, TimePoint=design$TimePoint,
           Subject_ID=design$Patient_ID, gmt=gmt_sim,
           geneset.name="Gene set 5",
           time_unit="H",
           title="",
           gg.add=list(scale_color_manual(values=colval)),
           lab.cex=0.7
    )

## End(Not run)

## Not run:
plotPat.1GS(expr=tcgsa_sim_1grp$Estimations, TimePoint=design$TimePoint,
           Subject_ID=design$Patient_ID, gmt=gmt_sim,
           geneset.name="Gene set 3",
           time_unit="H",
           lab.cex=0.7
    )

## End(Not run)

```

---

plotPat.TcGSA

*Plot a Gene Set Trends Heatmap for each Patient.*


---

### Description

This function plots a series of gene sets dynamic trends heatmaps. One heatmap is drawn for each patient. NOT IMPLEMENTED YET (TODO)

### Usage

```

plotPat.TcGSA(x, threshold = 0.05, myproc = "BY", nbsimu_pval = 1e+06,
             expr, Subject_ID, TimePoint, baseline = NULL, only.signif = TRUE,
             group.var = NULL, Group_ID_paired = NULL, ref = NULL,
             group_of_interest = NULL, FUNcluster = NULL,
             clustering_metric = "euclidian", clustering_method = "ward", B = 500,

```

```

max_trends = 4, aggreg.fun = "median", methodOptiClust = "firstSEmax",
verbose = TRUE, clust_trends = NULL, N_clusters = NULL,
myclusters = NULL, label.clusters = NULL, prev_rowCL = NULL,
descript = TRUE, plotAll = TRUE, color.vec = c("darkred", "#D73027",
"#FC8D59", "snow", "#91BFDB", "#4575B4", "darkblue"), legend.breaks = NULL,
label.column = NULL, time_unit = "", cex.label.row = 1,
cex.label.column = 1, margins = c(5, 25), heatKey.size = 1,
dendrogram.size = 1, heatmap.height = 1, heatmap.width = 1,
cex.clusterKey = 1, cex.main = 1, horiz.clusterKey = TRUE,
main = NULL, subtitle = NULL, ...)

```

## Arguments

x	a <b>tgsa</b> object.
threshold	the threshold at which the FDR or the FWER should be controlled.
myproc	a vector of character strings containing the names of the multiple testing procedures for which adjusted p-values are to be computed. This vector should include any of the following: "Bonferroni", "Holm", "Hochberg", "SidakSS", "SidakSD", "BH", "BY", "ABH", "TSBH" or "none". "none" indicates no adjustment for multiple testing. See <a href="#">mt.rawp2adjp</a> for details. Default is "BY", the Benjamini & Yekutieli (2001) step-up FDR-controlling procedure (general dependency structures). In order to control the FWER(in case of an analysis that is more a hypothesis confirmation than an exploration of the expression data), we recommend to use "Holm", the Holm (1979) step-down adjusted p-values for strong control of the FWER.
nbsimu_pval	the number of observations under the null distribution to be generated in order to compute the p-values. Default is 1e+06.
expr	either a matrix or dataframe of gene expression upon which dynamics are to be calculated, or a list of gene sets estimation of gene expression. In the case of a matrix or dataframe, its dimension are $n \times p$ , with the $p$ sample in column and the $n$ genes in row. In the case of a list, its length should correspond to the number of gene sets under scrutiny and each element should be an 3 dimension array of estimated gene expression, such as for the list returned in the 'Estimations' element of <a href="#">TcGSA.LR</a> . See Details.
Subject_ID	a factor of length $p$ that is in the same order as the columns of expr (when it is a dataframe) and that contains the patient identifier of each sample.
TimePoint	a numeric vector or a factor of length $p$ that is in the same order as Subject_ID and the columns of expr (when it is a dataframe), and that contains the time points at which gene expression was measured.
baseline	a character string which is the value of TimePoint used as baseline.
only.signif	logical flag for plotting only the significant gene sets. If FALSE, all the gene sets from the <b>gmt</b> object contained in x are plotted. Default is TRUE.
group.var	in the case of several treatment groups, this is a factor of length $p$ that is in the same order as Timepoint, Subject_ID, sample_name and the columns of expr. It indicates to which treatment group each sample belongs to. Default is NULL, which means that there is only one treatment group. See Details.

Group_ID_paired	a character vector of length $p$ that is in the same order as Timepoint, Subject_ID, sample_name, group.var and the columns of expr. This argument must not be NULL in the case of a paired analysis, and must be NULL otherwise. Default is NULL.
ref	the group which is used as reference in the case of several treatment groups. Default is NULL, which means that reference is the first group in alphabetical order of the labels of group.var. See Details.
group_of_interest	the group of interest, for which dynamics are to be computed in the case of several treatment groups. Default is NULL, which means that group of interest is the second group in alphabetical order of the labels of group.var.
FUNcluster	the clustering function used to agglomerate genes in trends. Default is NULL, in which a hierarchical clustering is performed via the function <a href="#">agnes</a> , using the metric clustering_metric and the method clustering_method. See <a href="#">clusGap</a>
clustering_metric	character string specifying the metric to be used for calculating dissimilarities between observations in the hierarchical clustering when FUNcluster is NULL. The currently available options are "euclidean" and "manhattan". Default is "euclidean". See <a href="#">agnes</a> . Also, a "sts" option is available in TcGSA. It implements the 'Short Time Series' distance [Moller-Levet et al., Fuzzy CLustering of short time series and unevenly distributed sampling points, <i>Advances in Intelligent Data Analysis V:330-340</i> Springer, 2003] designed specifically for clustering time series.
clustering_method	character string defining the agglomerative method to be used in the hierarchical clustering when FUNcluster is NULL. The six methods implemented are "average" ([unweighted pair-]group average method, UPGMA), "single" (single linkage), "complete" (complete linkage), "ward" (Ward's method), "weighted" (weighted average linkage). Default is "ward". See <a href="#">agnes</a> .
B	integer specifying the number of Monte Carlo ("bootstrap") samples used to compute the gap statistics. Default is 500. See <a href="#">clusGap</a> .
max_trends	integer specifying the maximum number of different clusters to be tested. Default is 4.
aggreg.fun	a character string such as "mean", "median" or the name of any other statistics function defined that returns a single numeric value. It specifies the function used to aggregate the observations before the clustering. Default is to median. Default is "median".
methodOptiClust	character string indicating how the "optimal" number of clusters is computed from the gap statistics and their standard deviations. Possible values are "globalmax", "firstmax", "Tibs2001SEmax", "firstSEmax" and "globalSEmax". Default is "firstSEmax". See 'method' in <a href="#">clusGap</a> , Details and <a href="#">Tibshirani et al., 2001</a> in References.
verbose	logical flag enabling verbose messages to track the computing status of the function. Default is TRUE.

<code>clust_trends</code>	object of class <b>ClusteredTrends</b> containing already computed trends for the plotted gene sets. Default is NULL.
<code>N_clusters</code>	an integer that is the number of clusters in which the dynamics should be re-grouped. The cutoff of the clustering tree is automatically calculated accordingly. Default is NULL, in which case the dendrogram is not cut and no clusters are identified.
<code>myclusters</code>	a character vector of colors for predefined clusters of the represented genesets, with as many levels as the value of <code>N_clusters</code> . Default is NULL, in which case the clusters are automatically identified and colored via the <code>cutree</code> function and the <code>N_clusters</code> argument only.
<code>label.clusters</code>	if <code>N_clusters</code> is not NULL, a character vector of length <code>N_clusters</code> ss. Default is NULL, in which case if <code>N_clusters</code> is not NULL, clusters are simply labelled with numbers.
<code>prev_rowCL</code>	a <b>hclust</b> object, such as the one return by the present plotting function (see Value) for instance. If not NULL, no clustering is calculated by the present plotting function and this tree is used to represent the gene sets dynamics. Default is NULL.
<code>descript</code>	logical flag indicating that the description of the gene sets should appear after their name on the right side of the plot if TRUE. Default is TRUE. See Details.
<code>plotAll</code>	logical flag indicating wether a first heatmap with the median over all the patients should be plotted, or not. Default is TRUE.
<code>color.vec</code>	a character strings vector used to define the color <b>palette</b> used in the plot. Default is <code>c("#D73027", "#FC8D59", "lightyellow", "#91BFDB", "#4575B4")</code> .
<code>legend.breaks</code>	a numeric vector indicating the splitting points for coloring. Default is NULL, in which case the break points will be spaced equally and symetrically about 0.
<code>label.column</code>	a vector of character strings with the labels to be displayed for the columns (i.e. the time points). Default is NULL.
<code>time_unit</code>	the time unit to be displayed (such as "Y", "M", "W", "D", "H", etc) next to the values of <code>TimePoint</code> in the columns labels when <code>label.column</code> is NULL. Default is "".
<code>cex.label.row</code>	a numerical value giving the amount by which row labels text should be magnified relative to the default 1.
<code>cex.label.column</code>	a numerical value giving the amount by which column labels text should be magnified relative to the default 1.
<code>margins</code>	numeric vector of length 2 containing the margins (see <code>par(mar=*)</code> ) for column and row names, respectively. Default is <code>c(15, 100)</code> . See Details.
<code>heatKey.size</code>	the size of the color key for the heatmap fill. Default is 1.
<code>dendrogram.size</code>	the horizontal size of the dendrogram. Default is 1
<code>heatmap.height</code>	the height of the heatmap. Default is 1
<code>heatmap.width</code>	the width of the heatmap. Default is 1
<code>cex.clusterKey</code>	a numerical value giving the amount by which the clusters legend text should be magnified relative to the default 1, when <code>N_clusters</code> is not NULL.

<code>cex.main</code>	a numerical value giving the amount by which title text should be magnified relative to the default 1.
<code>horiz.clusterKey</code>	a logical flag; if TRUE, set the legend for clusters horizontally rather than vertically. Only used if the <code>N_clusters</code> argument is not NULL. Default is TRUE.
<code>main</code>	a character string for an optionnal title. Default is NULL.
<code>subtitle</code>	a character string for an optionnal subtitle. Default is NULL.
<code>...</code>	other parameters to be passed through to plotting functions.

### Details

On the heatmap, each line corresponds to a gene set, and each column to a timepoint.

First a heatmap is computed on all the patients (see `plot.TcGSA` and `clustTrend`) to define the clustering. Then, the clustering and coloring thus defined on all the patients are consistently used in the separate heatmaps that are plotted by patient.

If `expr` is a matrix or a dataframe, then the "original" data are plotted. On the other hand, if `expr` is a list returned in the 'Estimations' element of `TcGSA.LR`, then it is those "estimations" made by the `TcGSA.LR` function that are plotted.

If `descript` is FALSE, the second element of `margins` can be reduced (for instance use `margins = c(5, 10)`), as there is not so much need for space in order to display only the gene set names, without their description.

The median shown in the heatmap uses the respectively standardized (reduced and centered) expression of the genes over the patients.

### Value

An object of class `hclust` which describes the tree produced by the clustering process. The object is a list with components:

- merge an  $n - 1$  by 2 matrix. Row  $i$  of merge describes the merging of clusters at step  $i$  of the clustering. If an element  $j$  in the row is negative, then observation  $-j$  was merged at this stage. If  $j$  is positive then the merge was with the cluster formed at the (earlier) stage  $j$  of the algorithm. Thus negative entries in merge indicate agglomerations of singletons, and positive entries indicate agglomerations of non-singletons.
- height a set of  $n - 1$  real values (non-decreasing for ultrametric trees). The clustering height: that is, the value of the criterion associated with the Ward clustering method.
- order a vector giving the permutation of the original observations suitable for plotting, in the sense that a cluster plot using this ordering and matrix merge will not have crossings of the branches.
- labels the gene sets name.
- call the call which produced the result clustering:  
`hclust(d = dist(map2heat,method = "euclidean"), method = "ward.D2")`
- method "ward.D2", as it is the clustering method that has been used for clustering the gene set trends.
- dist.method "euclidean", as it is the distance that has been used for clustering the gene set trends.

- legend.breaks a numeric vector giving the splitting points used for coloring the heatmap. If plot is FALSE, then it is NULL.
- myclusters a character vector of colors for clusters of the represented genesets, with as many levels as the value of N\_clusters. If no clusters were represented, than this is NULL.
- ddr a **dendrogram** object with the reordering used for the heatmap. See [heatmap.2](#).
- clustersExport a data frame with 2 variables containing the two following variables :
  - GeneSet: the gene sets clustered.
  - Cluster: the cluster they belong to.

The data frame is order by the variable Cluster.

### Author(s)

Boris P. Hejblum

### References

Hejblum, B.P., Skinner, J., Thiebaut, R., 2014, TcGSA: a gene set approach for longitudinal gene expression data analysis, **submitted**.

### See Also

[plot.TcGSA](#), [heatmap.2](#), [TcGSA.LR](#), [hclust](#)

### Examples

```
data(data_simu_TcGSA)

tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                        subject_name="Patient_ID", time_name="TimePoint",
                        time_func="linear", crossedRandom=FALSE)

plotPat.TcGSA(x=tcgsa_sim_1grp, expr=expr_1grp,
             Subject_ID=design$Patient_ID, TimePoint=design$TimePoint,
             B=100,
             time_unit="H"
             )

plotPat.TcGSA(x=tcgsa_sim_1grp, expr=tcgsa_sim_1grp$Estimations,
             Subject_ID=design$Patient_ID, TimePoint=design$TimePoint,
             baseline=1,
             B=100,
             time_unit="H"
             )
```

plotSelect.GS

*Plotting (several) Selected Gene Set(s) in some Subjects***Description**

This function can plot different representations of the gene expression in selected gene sets, among a subset of selected subjects.

**Usage**

```
plotSelect.GS(expr, gmt, Subject_ID, TimePoint, geneset.names.select,
  Subject_ID.select, display = "one subject per page", baseline = NULL,
  group.var = NULL, Group_ID_paired = NULL, ref = NULL,
  group_of_interest = NULL, FUNcluster = NULL,
  clustering_metric = "euclidian", clustering_method = "ward", B = 500,
  max_trends = 4, aggreg.fun = "median", trend.fun = "median",
  methodOptiClust = "firstSEmax", verbose = TRUE, clustering = TRUE,
  time_unit = "", title = NULL, y.lab = NULL, desc = TRUE,
  lab.cex = 1, axis.cex = 1, main.cex = 1, y.lab.angle = 90,
  x.axis.angle = 45, y.lim = NULL, x.lim = NULL, gg.add = list(theme()))
```

**Arguments**

expr	either a matrix or dataframe of gene expression upon which dynamics are to be calculated, or a list of gene sets estimation of gene expression. In the case of a matrix or dataframe, its dimension are $n \times p$ , with the $p$ sample in column and the $n$ genes in row. In the case of a list, its length should correspond to the number of gene sets under scrutiny and each element should be an 3 dimension array of estimated gene expression, such as for the list returned in the 'Estimations' element of <a href="#">TcGSA.LR</a> . See Details.
gmt	a <b>gmt</b> object containing the gene sets definition. See <a href="#">GSA.read.gmt</a> and definition on <a href="http://www.broadinstitute.org">www.broadinstitute.org</a> .
Subject_ID	a factor of length $p$ that is in the same order as the columns of expr (when it is a dataframe) and that contains the patient identifier of each sample.
TimePoint	a numeric vector or a factor of length $p$ that is in the same order as TimePoint and the columns of expr (when it is a dataframe), and that contains the time points at which gene expression was measured.
geneset.names.select	a character vector containing the names of the gene sets to be plotted, that must appear in the "geneset.names" element of gmt.
Subject_ID.select	a character vector containing the names of the subjects to be plotted, that must appear in the Subject_ID vector.
display	How to display the resulting graphs. One of the following: "one GS per page", "one subject per page", "median over selected patients". Default is "one subject per page".



baseline	a character string which is the value of TimePoint that can be used as a baseline. Default is NULL, in which case no timepoint is used as a baseline value for gene expression. Has to be NULL when comparing two treatment groups.
group.var	in the case of several treatment groups, this is a factor of length $p$ that is in the same order as Timepoint, Subject_ID and the columns of expr. It indicates to which treatment group each sample belongs to. Default is NULL, which means that there is only one treatment group.
Group_ID_paired	a character vector of length $p$ that is in the same order as Timepoint, Subject_ID, group.var and the columns of expr. This argument must not be NULL in the case of a paired analysis, and must be NULL otherwise. Default is NULL.
ref	the group which is used as reference in the case of several treatment groups. Default is NULL, which means that reference is the first group in alphabetical order of the labels of group.var. See Details.
group_of_interest	the group of interest, for which dynamics are to be computed in the case of several treatment groups. Default is NULL, which means that group of interest is the second group in alphabetical order of the labels of group.var.
FUNcluster	a function which accepts as first argument a matrix $x$ and as second argument the number of clusters desired $k$ , and which returns a list with a component named 'cluster' which is a vector of length $n = nrow(x)$ of integers in $1:k$ , determining the clustering or grouping of the $n$ observations. Default is NULL, in which case a hierarchical clustering is performed via the function <a href="#">agnes</a> , using the metric clustering_metric and the method clustering_method. See 'FUNcluster' in <a href="#">clusGap</a> and Details.
clustering_metric	character string specifying the metric to be used for calculating dissimilarities between observations in the hierarchical clustering when FUNcluster is NULL. The currently available options are "euclidean" and "manhattan". Default is "euclidean". See <a href="#">agnes</a> . Also, a "sts" option is available in TcGSA. It implements the 'Short Time Series' distance [Moller-Levet et al., Fuzzy CLustering of short time series and unevenly distributed sampling points, <i>Advances in Intelligent Data Analysis V:330-340</i> Springer, 2003] designed specifically for clustering time series.
clustering_method	character string defining the agglomerative method to be used in the hierarchical clustering when FUNcluster is NULL. The six methods implemented are "average" ([unweighted pair-]group average method, UPGMA), "single" (single linkage), "complete" (complete linkage), "ward" (Ward's method), "weighted" (weighted average linkage). Default is "ward". See <a href="#">agnes</a> .
B	integer specifying the number of Monte Carlo ("bootstrap") samples used to compute the gap statistics. Default is 500. See <a href="#">clusGap</a> .
max_trends	integer specifying the maximum number of different clusters to be tested. Default is 4.
aggreg.fun	a character string such as "mean", "median" or the name of any other defined statistics function that returns a single numeric value. It specifies the function used to aggregate the observations before the clustering. Default is to median.

trend.fun	a character string such as "mean", "median" or the name of any other function that returns a single numeric value. It specifies the function used to calculate the trends of the identified clustered. Default is to median.
methodOptiClust	character string indicating how the "optimal" number of clusters is computed from the gap statistics and their standard deviations. Possible values are "globalmax", "firstmax", "Tibs2001SEmax", "firstSEmax" and "globalSEmax". Default is "firstSEmax". See 'method' in <a href="#">clusGap</a> , Details and <i>Tibshirani et al., 2001</i> in References.
verbose	logical flag enabling verbose messages to track the computing status of the function. Default is TRUE.
clustering	logical flag. If FALSE, there is no clustering representation; if TRUE, the lines are colored according to which cluster they belong to. Default is TRUE. See Details.
time_unit	the time unit to be displayed (such as "Y", "M", "W", "D", "H", etc) next to the values of TimePoint on the x-axis. Default is "".
title	character specifying the title of the plot. If NULL, a title is automatically generated, if "", no title appears. Default is NULL.
y.lab	character specifying the annotation of the y axis. If NULL, an annotation is automatically generated, if "", no annotation appears. Default is NULL.
desc	a logical flag. If TRUE, a line is added to the title of the plot with the description of the gene set plotted (from the gmt file). Default is TRUE.
lab.cex	a numerical value giving the amount by which lab labels text should be magnified relative to the default 1.
axis.cex	a numerical value giving the amount by which axis annotation text should be magnified relative to the default 1.
main.cex	a numerical value giving the amount by which title text should be magnified relative to the default 1.
y.lab.angle	a numerical value (in [0, 360]) giving the orientation by which y-label text should be turned (anti-clockwise). Default is 90. See <a href="#">element_text</a> .
x.axis.angle	a numerical value (in [0, 360]) giving the orientation by which x-axis annotation text should be turned (anti-clockwise). Default is 45.
y.lim	a numeric vector of length 2 giving the range of the y-axis. See <a href="#">plot.default</a> .
x.lim	if numeric, will create a continuous scale, if factor or character, will create a discrete scale. Observations not in this range will be dropped. See <a href="#">xlim</a> .
gg.add	A list of instructions to add to the ggplot2 instruction. See <a href="#">+gg</a> . Default is <code>list(theme())</code> , which adds nothing to the plot.

## Details

If `expr` is a matrix or a dataframe, then the "original" data are plotted. On the other hand, if `expr` is a list returned in the 'Estimations' element of [TcGSA.LR](#), then it is those "estimations" made by the [TcGSA.LR](#) function that are plotted.

If `indiv` is 'genes', then each line of the plot is the median of a gene expression over the patients. On the other hand, if `indiv` is 'patients', then each line of the plot is the median of a patient genes expression in this gene set.

This function uses the Gap statistics to determine the optimal number of clusters in the plotted gene set. See [clusGap](#).

### Value

A dataframe the 2 following variables:

- ProbeID which contains the IDs of the probes of the plotted gene set.
- Cluster which to which cluster the probe belongs to.

If clustering is FALSE, then Cluster is NA for all the probes.

### Author(s)

Boris P. Hejblum

### References

Tibshirani, R., Walther, G. and Hastie, T., 2001, Estimating the number of data clusters via the Gap statistic, *Journal of the Royal Statistical Society, Series B (Statistical Methodology)*, **63**, 2: 411–423.

### See Also

[ggplot2](#), [clusGap](#)

### Examples

```
## Not run:
data(data_simu_TcGSA)
tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                        subject_name="Patient_ID", time_name="TimePoint",
                        time_func="linear", crossedRandom=FALSE)

## End(Not run)

## Not run:
plotSelect.GS(expr=tcgsa_sim_1grp$Estimations, TimePoint=design$TimePoint,
             Subject_ID=design$Patient_ID, gmt=gmt_sim,
             geneset.names.select=c("Gene set 3", "Gene set 4", "Gene set 5"),
             Subject_ID.select=c("P1", "P2"),
             display="one GS per page",
             time_unit="H",
             lab.cex=0.7
)

## End(Not run)

## Not run:
plotSelect.GS(expr=tcgsa_sim_1grp$Estimations, TimePoint=design$TimePoint,
             Subject_ID=design$Patient_ID, gmt=gmt_sim,
             geneset.names.select=c("Gene set 3", "Gene set 4", "Gene set 5"),
```

```
      Subject_ID.select=c("P1", "P2"),
      display="one subject per page",
      time_unit="H",
      lab.cex=0.7
    )

## End(Not run)
```

---

pval\_simu

*Computing P-values with a Simulated Sample from the Null Distribution*

---

### Description

This function computes the p-value of a statistic using a simulated sample from its theoretical null distribution.

### Usage

```
pval_simu(s, theo_dist)
```

### Arguments

`s` the observation whose p-value is computed. For instance a Likelihood Ratio.  
`theo_dist` the sample of the distribution under the null hypothesis.

### Value

The p-value associated to the observation `s`.

### Author(s)

Boris P. Hejblum

### See Also

[rmixchisq](#)

### Examples

```
theo_dist <- rnorm(n=10000, mean=0, sd=1)
pval_simu(s=1.96, theo_dist)
1-pnorm(q=1.96, mean = 0, sd = 1, lower.tail = TRUE, log.p = FALSE)
```

**Description**

rmixchisq is used to simulate a mixture of chi-square distributions that corresponds to the null distribution of the Likelihood Ratio between 2 nested mixed models.

**Usage**

```
rmixchisq(n, s, q)
```

**Arguments**

n	number of observations.
s	number of fixed effects to be tested.
q	number of random effects to be tested.

**Details**

The approximate null distribution of a likelihood ratio for 2 nested mixed models, where both fixed and random effects are tested simultaneously, is a very specific mixture of  $\chi^2$  distributions [Self & Liang (1987), Stram & Lee (1994) and Stram & Lee (1995)]. It depends on both the number of random effects and the number of fixed effects to be tested simultaneously:

$$LRT_{H_0} \sim \sum_{k=q}^{q+r} \binom{r}{k-q} 2^{-r} \chi_{(k)}^2$$

**Value**

A vector of random independent observations of the chisquare mixture identified by the values of s and q.

**Author(s)**

Boris P. Hejblum

**References**

- Self, S. G. and Liang, K., 1987, Asymptotic properties of maximum likelihood estimators and likelihood ratio tests under nonstandard conditions, *Journal of the American Statistical Association* 82: 605–610.
- Stram, D. O. and Lee, J. W., 1994, Variance components testing in the longitudinal mixed effects model, *Biometrics* 50: 1171–1177.
- Stram, D. O. and Lee, J. W., 1995, Corrections to "Variance components testing in the longitudinal mixed effects model" by Stram, D. O. and Lee, J. W.; 50: 1171–1177 (1994), *Biometrics* 51: 1196.

**See Also**[pval\\_simu](#)**Examples**

```
sample_mixt <- rmixchisq(n=1000, s=3, q=3)
plot(density(sample_mixt))
```

---

 signifLRT.TcGSA

*Identifying the Significant Gene Sets*


---

**Description**

A function that identifies the significant gene sets in an object of class 'TcGSA'.

**Usage**

```
signifLRT.TcGSA(tcgsa, threshold = 0.05, myproc = "BY",
  nbsimu_pval = 1e+06, write = F, txtfilename = NULL, directory = NULL)
```

**Arguments**

tcgsa	a tcgsa object.
threshold	the threshold at which the FDR or the FWER should be controlled.
myproc	a vector of character strings containing the names of the multiple testing procedures for which adjusted p-values are to be computed. This vector should include any of the following: "Bonferroni", "Holm", "Hochberg", "SidakSS", "SidakSD", "BH", "BY", "ABH", "TSBH" or "none". "none" indicates no adjustment for multiple testing. See <a href="#">mt.rawp2adjp</a> for details. Default is "BY", the Benjamini & Yekutieli (2001) step-up FDR-controlling procedure (general dependency structures). In order to control the FWER(in case of an analysis that is more a hypothesis confirmation than an exploration of the expression data), we recommend to use "Holm", the Holm (1979) step-down adjusted p-values for strong control of the FWER.
nbsimu_pval	the number of observations under the null distribution to be generated in order to compute the p-values. Default is 1e+06.
write	logical flag enabling the export of the results as a table in a .txt file. Default is FALSE.
txtfilename	a character string with the name of the .txt file in which the results table is to be written, if write is TRUE. Default is NULL.
directory	if write is TRUE, a character string with the directory of the .txt file in which the results table is to be written, if write is TRUE. Default is NULL.

**Value**

signifLRT.TcGSA returns a list.

The first element `mixedLRTadjRes` is data frame with  $p$  rows (one row for each significant gene set) and the 3 following variables:

- `GeneSet` the significant gene set name from the `gmt` object.
- `AdjPval` the adjusted p-value corresponding to the significant gene set.
- `desc` the significant gene set description from the `gmt` object.

The second element `multCorProc` passes along the multiple testing procedure used (from the argument `myproc`).

The third element `threshold` passes along the significance threshold used (from the argument `threshold`).

**Author(s)**

Boris P. Hejblum

**References**

Hejblum, B.P., Skinner, J., Thiebaut, R., 2014, TcGSA: a gene set approach for longitudinal gene expression data analysis, **submitted**.

**See Also**

[multtest.TcGSA](#), [TcGSA.LR](#)

**Examples**

```
data(data_simu_TcGSA)

tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                        subject_name="Patient_ID", time_name="TimePoint",
                        time_func="linear", crossedRandom=FALSE)

sgnifs <- signifLRT.TcGSA(tcgsa_sim_1grp, threshold = 0.05, myproc = "BY",
                        nbsimu_pval = 1000, write=FALSE)

sgnifs
```

---

summary.TcGSA

*Summarizing TcGSA*

---

**Description**

summary method for class 'TcGSA'

**Usage**

```
## S3 method for class 'TcGSA'  
summary(object, ...)  
  
## S3 method for class 'summary.TcGSA'  
print(x, ...)
```

**Arguments**

object	an object of class 'TcGSA'.
...	further arguments passed to or from other methods.
x	an object of class 'summary.TcGSA'.

**Value**

The function `summary.TcGSA` returns a list with the following components (list elements):

- `time_func` the chosen form for the time trend.
- `separateSubjects` a logical flag indicating whether gene sets tested for discriminating among patients, or for time trends over time.
- `ntg` the number of treatment groups.
- `ngs` the number of tested gene sets.
- `nsignif` the number of significant gene sets at a 5% FDR (using the default Benjamini & Yekutieli step-up procedure).

**Author(s)**

Boris P. Hejblum

**See Also**

[TcGSA.LR](#)

**Examples**

```
data(data_simu_TcGSA)  
  
tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,  
                        subject_name="Patient_ID", time_name="TimePoint",  
                        time_func="linear", crossedRandom=FALSE)  
summary(tcgsa_sim_1grp)  
  
## Not run:  
tcgsa_sim_2grp <- TcGSA.LR(expr=expr_2grp, gmt=gmt_sim, design=design,  
                        subject_name="Patient_ID", time_name="TimePoint",  
                        time_func="linear", crossedRandom=FALSE,  
                        group_name="group.var")  
summary(tcgsa_sim_2grp)  
  
## End(Not run)
```



**Description**

This function computes the Likelihood Ratios for the gene sets under scrutiny, as well as estimations of genes dynamics inside those gene sets through mixed models.

**Usage**

```
TcGSA.LR(expr, gmt, design, subject_name = "Patient_ID",
         time_name = "TimePoint", crossedRandom = FALSE, covariates_fixed = "",
         time_covariates = "", time_func = "linear", group_name = "",
         separateSubjects = FALSE, minGSsize = 10, maxGSsize = 500)
```

```
## S3 method for class 'TcGSA'
print(x, ...)
```

**Arguments**

<code>expr</code>	a matrix or dataframe of gene expression. Its dimension are $n \times p$ , with the $p$ samples in column and the $n$ genes in row.
<code>gmt</code>	a <b>gmt</b> object containing the gene sets definition. See <a href="#">GSA.read.gmt</a> and definition on <a href="http://www.broadinstitute.org">www.broadinstitute.org</a> .
<code>design</code>	a matrix or dataframe containing the experimental variables that used in the model, namely <code>subject_name</code> , <code>time_name</code> , and <code>covariates_fixed</code> and <code>time_covariates</code> if applicable. Its dimension are $p \times m$ and its row are is in the same order as the columns of <code>expr</code> .
<code>subject_name</code>	the name of the factor variable from <code>design</code> that contains the information on the repetition units used in the mixed model, such as the patient identifiers for instance. Default is 'Patient_ID'. See Details.
<code>time_name</code>	the name of a numeric variable from <code>design</code> that contains the information on the time replicates (the time points at which gene expression was measured). Default is 'TimePoint'. See Details.
<code>crossedRandom</code>	logical flag indicating whether the random effects of the subjects and of the time points should be modeled as one crossed random effect or as two separated random effects. Default is FALSE. See details.
<code>covariates_fixed</code>	a character vector with the names of numeric or factor variables from the design matrix that should appear as fixed effects in the model. See details. Default is "", which corresponds to no covariates in the model.
<code>time_covariates</code>	a character vector with the names of numeric or factor variables from the design matrix that should appear as fixed effects interaction with the <code>time_name</code> variable in the model. See details. Default is "", which corresponds to no covariates in the model.

time_func	the form of the time trend. Can be either one of "linear", "cubic", "splines" or specified by the user, or the column name of a factor variable from design. If specified by the user, it must be as an expression using only names of variables from the design matrix with only the three following operators: +, *, / . The "splines" form corresponds to the natural cubic B-splines (see also <a href="#">ns</a> ). If there are only a few timepoints, a "linear" form should be sufficient. Otherwise, the "cubic" form is more parsimonious than the "splines" form, and should be sufficiently flexible. If the column name of a factor variable from design is supplied, then time is considered as discrete in the analysis. If the user specify a formula using column names from design, both factor and numeric variables can be used.
minGSsize	the minimum number of genes in a gene set. If there are less genes than this number in one of the gene sets under scrutiny, the Likelihood Ratio of this gene set is not computed (the mixed model are not fitted). Default is 10 genes as the minimum.
maxGSsize	the maximum number of genes in a gene set. If there are more genes than this number in one of the gene sets under scrutiny, the Likelihood Ratio of this gene set is not computed (the mixed model are not fitted). This is to avoid very long computation times. Default is 500 genes as the maximum.
group_name	in the case of several treatment groups, the name of a factor variable from the design matrix. It indicates to which treatment group each sample belongs to. Default is "", which means that there is only one treatment group. See Details.
separateSubjects	logical flag indicating that the analysis identifies gene sets that discriminates patients rather than gene sets that have a significant trend over time. Default is FALSE. See Details.
x	an object of class 'TcGSA'.
...	further arguments passed to or from other methods.

## Details

This Time-course Gene Set Analysis aims at identifying gene sets that are not stable over time, either homogeneously or heterogeneously (see *Hejblum et al, 2012*) in terms of their probes. And when the argument `separateSubjects` is TRUE, instead of identifying gene sets that have a significant trend over time, *TcGSA* identifies gene sets that have significantly different trends over time depending on the patient.

## Value

`TcGSA.LR` returns a `tcgsa` object, which is a list with the 5 following elements:

- fit a data frame that contains the 3 following variables:
  - LR: the likelihood ratio between the model under the null hypothesis and the model under the alternative hypothesis.
  - CVG\_H0: convergence status of the model under the null hypothesis.
  - CVG\_H1: convergence status of the model under the alternative hypothesis.

- `time_func`: a character string passing along the value of the `time_func` argument used in the call.
- `GeneSets_gmt`: a `gmt` object passing along the value of the `gmt` argument used in the call.
- `group.var`: a factor passing along the `group_name` variable from the design matrix.
- `separateSubjects`: a logical flag passing along the value of the `separateSubjects` argument used in the call.
- `Estimations`: a list of 3 dimensions arrays. Each element of the list (i.e. each array) corresponds to the estimations of gene expression dynamics for each of the gene sets under scrutiny (obtained from mixed models). The first dimension of those arrays is the genes included in the concerned gene set, the second dimension is the `Patient_ID`, and the third dimension is the `TimePoint`. The values inside those arrays are estimated gene expressions.
- `time_DF`: the degree of freedom of the natural splines functions

### Author(s)

Boris P. Hejblum

### References

Hejblum, B.P., Skinner, J., Thiebaut, R., 2014, TcGSA: a gene set approach for longitudinal gene expression data analysis, **submitted**.

### See Also

[summary.TcGSA](#), [plot.TcGSA](#), and [TcGSA.LR.parallel](#) for an implementation using parallel computing

### Examples

```
data(data_simu_TcGSA)

tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                        subject_name="Patient_ID", time_name="TimePoint",
                        time_func="linear", crossedRandom=FALSE)

tcgsa_sim_1grp
summary(tcgsa_sim_1grp)

## Not run:
plot(x=tcgsa_sim_1grp, expr=expr_1grp,
     Subject_ID=design$Patient_ID, TimePoint=design$TimePoint,
     baseline=1,
     B=100,
     time_unit="H"
     )

## End(Not run)

## Not run:
tcgsa_sim_2grp <- TcGSA.LR(expr=expr_2grp, gmt=gmt_sim, design=design,
```

```

subject_name="Patient_ID", time_name="TimePoint",
time_func="linear", crossedRandom=FALSE,
group_name="group.var")

tcgsa_sim_2grp

## End(Not run)

```

---

TcGSA.LR.parallel      *Parallel computing the Likelihood Ratios for the Gene Sets under Scrutiny*

---

## Description

A parallel version of the function `TcGSA.LR` to be used on a cluster of computing processors. This function computes the Likelihood Ratios for the gene sets under scrutiny, as well as estimations of genes dynamics inside those gene sets through mixed models.

## Usage

```

TcGSA.LR.parallel(Ncpus, type_connec, expr, gmt, design,
  subject_name = "Patient_ID", time_name = "TimePoint",
  crossedRandom = FALSE, covariates_fixed = "", time_covariates = "",
  time_func = "linear", group_name = "", separateSubjects = FALSE,
  minGSsize = 10, maxGSsize = 500, monitorfile = "")

```

## Arguments

<code>Ncpus</code>	The number of processors available on the cluster.
<code>type_connec</code>	The type of connection between the processors. Supported cluster types are "SOCK", "PVM", "MPI", and "NWS". See also <a href="#">makeCluster</a> .
<code>expr</code>	a matrix or dataframe of gene expression. Its dimension are $n \times p$ , with the $p$ samples in column and the $n$ genes in row.
<code>gmt</code>	a <b>gmt</b> object containing the gene sets definition. See <a href="#">GSA.read.gmt</a> and definition on <a href="http://www.broadinstitute.org">www.broadinstitute.org</a> .
<code>design</code>	a matrix or dataframe containing the experimental variables that used in the model, namely <code>subject_name</code> , <code>time_name</code> , and <code>covariates_fixed</code> and <code>time_covariates</code> if applicable. Its dimension are $p \times m$ and its row are is in the same order as the columns of <code>expr</code> .
<code>subject_name</code>	the name of the factor variable from <code>design</code> that contains the information on the repetition units used in the mixed model, such as the patient identifiers for instance. Default is 'Patient_ID'. See Details.
<code>time_name</code>	the name of the numeric or factor variable from <code>design</code> contains the information on the time replicates (the time points at which gene expression was measured). Default is 'TimePoint'. See Details.

<code>crossedRandom</code>	logical flag indicating whether the random effects of the subjects and of the time points should be modeled as one crossed random effect or as two separated random effects. Default is FALSE. See details.
<code>covariates_fixed</code>	a character vector with the names of numeric or factor variables from the design matrix that should appear as fixed effects in the model. See details. Default is "", which corresponds to no covariates in the model.
<code>time_covariates</code>	the name of a numeric variable from design that contains the information on the time replicates (the time points at which gene expression was measured). Default is 'TimePoint'. See Details.
<code>time_func</code>	the form of the time trend. Can be either one of "linear", "cubic", "splines" or specified by the user, or the column name of a factor variable from design. If specified by the user, it must be as an expression using only names of variables from the design matrix with only the three following operators: +, *, /. The "splines" form corresponds to the natural cubic B-splines (see also <a href="#">ns</a> ). If there are only a few timepoints, a "linear" form should be sufficient. Otherwise, the "cubic" form is more parsimonious than the "splines" form, and should be sufficiently flexible. If the column name of a factor variable from design is supplied, then time is considered as discrete in the analysis. If the user specifies a formula using column names from design, both factor and numeric variables can be used.
<code>minGSsize</code>	the minimum number of genes in a gene set. If there are less genes than this number in one of the gene sets under scrutiny, the Likelihood Ratio of this gene set is not computed (the mixed model are not fitted). Default is 10 genes as the minimum.
<code>maxGSsize</code>	the maximum number of genes in a gene set. If there are more genes than this number in one of the gene sets under scrutiny, the Likelihood Ratio of this gene set is not computed (the mixed model are not fitted). This is to avoid very long computation times. Default is 500 genes as the maximum.
<code>group_name</code>	in the case of several treatment groups, the name of a factor variable from the design matrix. It indicates to which treatment group each sample belongs to. Default is "", which means that there is only one treatment group. See Details.
<code>separateSubjects</code>	logical flag indicating that the analysis identifies gene sets that discriminates patients rather than gene sets that have a significant trend over time. Default is FALSE. See Details.
<code>monitorfile</code>	a writable <a href="#">connections</a> or a character string naming a file to write into, to monitor the progress of the analysis. Default is "" which is no monitoring. See Details.

## Details

This Time-course Gene Set Analysis aims at identifying gene sets that are not stable over time, either homogeneously or heterogeneously (see *Hejblum et al, 2012*) in terms of their probes. And when the argument `separatePatients` is TRUE, instead of identifying gene sets that have a significant trend over time (possibly with probes heterogeneity of this trend), *TcGSA* identifies gene sets that have significantly different trends over time depending on the patient.

If the `monitorfile` argument is a character string naming a file to write into, in the case of a new file that does not exist yet, such a new file will be created. A line is written each time one of the gene sets under scrutiny has been analysed (i.e. the two mixed models have been fitted, see `TcGSA.LR`) by one of the parallelized processors.

### Value

`TcGSA.LR` returns a `tcgsa` object, which is a list with the 5 following elements:

- `fit` a data frame that contains the 3 following variables:
  - `LR`: the likelihood ratio between the model under the null hypothesis and the model under the alternative hypothesis.
  - `CVG_H0`: convergence status of the model under the null hypothesis.
  - `CVG_H1`: convergence status of the model under the alternative hypothesis.
- `time_func`: a character string passing along the value of the `time_func` argument used in the call.
- `GeneSets_gmt`: a `gmt` object passing along the value of the `gmt` argument used in the call.
- `group.var`: a factor passing along the `group_name` variable from the design matrix.
- `separateSubjects`: a logical flag passing along the value of the `separateSubjects` argument used in the call.
- `Estimations`: a list of 3 dimensions arrays. Each element of the list (i.e. each array) corresponds to the estimations of gene expression dynamics for each of the gene sets under scrutiny (obtained from mixed models). The first dimension of those arrays is the genes included in the concerned gene set, the second dimension is the `Patient_ID`, and the third dimension is the `TimePoint`. The values inside those arrays are estimated gene expressions.
- `time_DF`: the degree of freedom of the natural splines functions

### Author(s)

Boris P. Hejblum

### References

Hejblum, B.P., Skinner, J., Thiebaut, R., 2014, TcGSA: a gene set approach for longitudinal gene expression data analysis, **submitted**.

### See Also

[summary.TcGSA](#), [plot.TcGSA](#)

### Examples

```
data(data_simu_TcGSA)
```

```
tcgsa_sim_1grp <- TcGSA.LR(expr=expr_1grp, gmt=gmt_sim, design=design,
                          subject_name="Patient_ID", time_name="TimePoint",
                          time_func="linear", crossedRandom=FALSE)
```

```
## Not run:
require(doSNOW)
tcgsa_sim_1grp <- TcGSA.LR.parallel(Ncpus = 2, type_connec = 'SOCK',
                                  expr=expr_1grp, gmt=gmt_sim, design=design,
                                  subject_name="Patient_ID", time_name="TimePoint",
                                  time_func="linear", crossedRandom=FALSE,
                                  separateSubjects=TRUE)

## End(Not run)
tcgsa_sim_1grp
summary(tcgsa_sim_1grp)
```

# Index

`+.gg`, [17](#), [20](#), [24](#), [34](#)  
`agnes`, [4](#), [10](#), [11](#), [16](#), [23](#), [28](#), [33](#)  
`clusGap`, [3–6](#), [10](#), [11](#), [16–18](#), [23–25](#), [28](#), [33–35](#)  
`ClusteredTrends`, [5](#), [11](#), [13](#), [29](#)  
`ClusteredTrends (clustTrend)`, [3](#)  
`clustTrend`, [3](#), [30](#)  
`connections`, [45](#)  
`cutree`, [11](#), [29](#)  
  
`data_simu_TcGSA`, [6](#)  
`design (data_simu_TcGSA)`, [6](#)  
  
`element_text`, [17](#), [24](#), [34](#)  
`expr_1grp (data_simu_TcGSA)`, [6](#)  
`expr_2grp (data_simu_TcGSA)`, [6](#)  
  
`ggplot2`, [18](#), [25](#), [35](#)  
`gmt_sim (data_simu_TcGSA)`, [6](#)  
`GSA`, [2](#)  
`GSA.read.gmt`, [7](#), [15](#), [22](#), [32](#), [41](#), [44](#)  
  
`hclust`, [13](#), [14](#), [30](#), [31](#)  
`heatmap.2`, [13](#), [14](#), [31](#)  
  
`makeCluster`, [44](#)  
`mt.rawp2adjp`, [4](#), [8](#), [9](#), [27](#), [38](#)  
`multtest.TcGSA`, [8](#), [39](#)  
  
`ns`, [42](#), [45](#)  
  
`palette`, [12](#), [29](#)  
`par`, [12](#), [29](#)  
`plot.ClusteredTrends (clustTrend)`, [3](#)  
`plot.default`, [17](#), [24](#), [34](#)  
`plot.TcGSA`, [2](#), [9](#), [30](#), [31](#), [43](#), [46](#)  
`plot1GS`, [2](#), [6](#), [15](#), [21](#)  
`plotFit.GS`, [19](#)  
`plotPat.1GS`, [22](#)  
`plotPat.TcGSA`, [26](#)  
  
`plotSelect.GS`, [21](#), [32](#)  
`print.ClusteredTrends (clustTrend)`, [3](#)  
`print.summary.TcGSA (summary.TcGSA)`, [39](#)  
`print.TcGSA (TcGSA.LR)`, [41](#)  
`pval_simu`, [36](#), [38](#)  
  
`rmixchisq`, [36](#), [37](#)  
  
`signifLRT.TcGSA`, [8](#), [38](#)  
`stat_smooth`, [17](#)  
`summary.TcGSA`, [39](#), [43](#), [46](#)  
  
`TcGSA (TcGSA-package)`, [2](#)  
`TcGSA-package`, [2](#)  
`TcGSA.LR`, [2](#), [3](#), [5–8](#), [10](#), [13–15](#), [18](#), [22](#), [24](#), [27](#),  
[30–32](#), [34](#), [39](#), [40](#), [41](#), [44](#), [46](#)  
`TcGSA.LR.parallel`, [43](#), [44](#)  
  
`xlim`, [17](#), [24](#), [34](#)