

Package ‘PepSetTest’

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Title Peptide Set Test

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Description Peptide Set Test (PepSetTest) is a peptide-centric strategy to infer differentially expressed proteins in LC-MS/MS proteomics data. This test detects coordinated changes in the expression of peptides originating from the same protein and compares these changes against the rest of the peptidome. Compared to traditional aggregation-based approaches, the peptide set test demonstrates improved statistical power, yet controlling the Type I error rate correctly in most cases. This test can be valuable for discovering novel biomarkers and prioritizing drug targets, especially when the direct application of statistical analysis to protein data fails to provide substantial insights.

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PepSetTest-package *PepSetTest: Peptide Set Test*

Description

Peptide Set Test (PepSetTest) is a peptide-centric strategy to infer differentially expressed proteins in LC-MS/MS proteomics data. This test detects coordinated changes in the expression of peptides originating from the same protein and compares these changes against the rest of the peptidome. Compared to traditional aggregation-based approaches, the peptide set test demonstrates improved statistical power, yet controlling the Type I error rate correctly in most cases. This test can be valuable for discovering novel biomarkers and prioritizing drug targets, especially when the direct application of statistical analysis to protein data fails to provide substantial insights.

Author(s)

Maintainer: Junmin Wang <jmwang.bio@gmail.com>

See Also

Useful links:

- <https://github.com/JmWangBio/PepSetTest>
- Report bugs at <https://github.com/JmWangBio/PepSetTest/issues>

AggLimmaWorkflow *Aggregation-based LIMMA workflow*

Description

Given peptide abundance and assignment of peptide sequences to proteins, execute the aggregation-based LIMMA workflow to compute the log₂ fold change, p-value, and adjusted p-value of all proteins identified.

Usage

```
AggLimmaWorkflow(
  dat,
  contrasts.par,
  group,
  pep_mapping_tbl,
  covar = NULL,
  method = c("sum", "robreg"),
  logged = c(TRUE, FALSE),
  npep.trend = FALSE,
  eb = TRUE
)
```

Arguments

<code>dat</code>	a dataframe or matrix of peptide abundance, or a SummarizedExperiment object where grouping and peptide-protein mapping are provided in <code>colData</code> and <code>rowData</code> , respectively.
<code>contrasts.par</code>	group levels to be compared separated by dash (e.g., "B-A" if group B is to be compared against group A)
<code>group</code>	a vector of group levels corresponding to each sample. Alternatively, it can be the column name of the group in <code>colData</code> if <code>dat</code> is a SummarizedExperiment object.
<code>pep_mapping_tbl</code>	a table mapping peptides to proteins. Alternatively, it can be the column name of the protein in <code>rowData</code> if <code>dat</code> is a SummarizedExperiment object.
<code>covar</code>	covariate matrix. Alternatively, it can be the column names of the covariates in <code>colData</code> if <code>dat</code> is a SummarizedExperiment object.
<code>method</code>	method of aggregation. Options including "sum" (summed peptide intensity) and "robreg" (robust regression with M-Estimation).
<code>logged</code>	Boolean variable indicating whether abundance data have been log-transformed
<code>npep.trend</code>	logical, should a number-of-peptide-trend be allowed for the prior variance? Default is constant prior variance.
<code>eb</code>	logical, whether to output the result from the empirical Bayes or ordinary approach.

Value

AggLimmaWorkflow returns a dataframe containing the following columns

feature	unique protein identifier
logFC	log2 fold change
t	t-statistic
P.Value	raw p-value
adj.P.Val	p-value adjusted via the Benjamini-Hochberg method
B	B-statistic (empirical Bayes only)

Author(s)

Junmin Wang

References

Ritchie, ME, Phipson, B, Wu, D, Hu, Y, Law, CW, Shi, W, and Smyth, GK (2015). limma powers differential expression analyses for RNA-sequencing and microarray studies. *Nucleic Acids Research* 43, e47.

Examples

```
# Generate random peptide data
dat <- 2^matrix(rnorm(3000), ncol = 6)
colnames(dat) <- paste0("Sample", 1:6)
rownames(dat) <- paste0("Peptide", 1:500)

# Generate peptide mapping table
pep_mapping_tbl <- data.frame(peptide = paste0("Peptide", 1:500),
protein = paste0("Protein", rep(1:100, each = 5)))

# Generate groups and contrasts
group <- c(rep("A", 3), rep("B", 3))
contrasts.par <- "B-A"

AggLimmaWorkflow(dat, contrasts.par = contrasts.par,
group = group,
pep_mapping_tbl = pep_mapping_tbl,
method = "sum",
logged = FALSE)

# Store data as a SummarizedExperiment object; add covariates
library(tibble)
library(SummarizedExperiment)
colData <- data.frame(sample = LETTERS[seq_along(group)], group = group,
sex = c("M", "F", "M", "F", "M"), age = 1:6) |>
column_to_rownames(var = "sample")
rowData <- pep_mapping_tbl |> column_to_rownames(var = "peptide")
dat.nn <- dat
rownames(dat.nn) <- NULL
```

```

colnames(dat.nn) <- NULL
dat.se <- SummarizedExperiment(assays = list(int = dat.nn), colData = colData, rowData = rowData)

AggLimmaWorkflow(dat.se, contrasts.par = contrasts.par,
group = "group",
covar = c("sex", "age"),
pep_mapping_tbl = "protein",
method = "sum",
logged = FALSE)

```

AggPeps

Aggregate peptide abundance values

Description

Given peptide abundance and assignment of peptide sequences to proteins, aggregate peptide abundance values into protein abundance values.

Usage

```

AggPeps(
  dat,
  pep_mapping_tbl,
  method = c("sum", "robreg"),
  logged = c(TRUE, FALSE)
)

```

Arguments

<code>dat</code>	a dataframe or matrix of peptide abundance, or a SummarizedExperiment object where grouping and peptide-protein mapping are provided in <code>colData</code> and <code>rowData</code> , respectively.
<code>pep_mapping_tbl</code>	a table mapping peptides to proteins. Alternatively, it can be the column name of the protein in <code>rowData</code> if <code>dat</code> is a SummarizedExperiment object.
<code>method</code>	method of aggregation. Options including "sum" (summed peptide intensity) and "robreg" (robust regression with M-Estimation).
<code>logged</code>	Boolean variable indicating whether abundance data have been log-transformed

Value

AggPeps returns a list containing a matrix of protein abundance values and a vector of number of peptides

Author(s)

Junmin Wang

Examples

```

# Generate random peptide data
dat <- 2^matrix(rnorm(3000), ncol = 6)
colnames(dat) <- paste0("Sample", 1:6)
rownames(dat) <- paste0("Peptide", 1:500)

# Generate peptide mapping table
pep_mapping_tbl <- data.frame(peptide = paste0("Peptide", 1:500),
  protein = paste0("Protein", rep(1:100, each = 5)))

AggPeps(dat, pep_mapping_tbl, method = "sum",
  logged = FALSE)

# Store data as a SummarizedExperiment object
library(tibble)
library(SummarizedExperiment)
rowData <- pep_mapping_tbl |> column_to_rownames(var = "peptide")
dat.nn <- dat
rownames(dat.nn) <- NULL
colnames(dat.nn) <- NULL
dat.se <- SummarizedExperiment(assays = list(int = dat.nn), rowData = rowData)

AggPeps(dat.se, pep_mapping_tbl = "protein", method = "sum",
  logged = FALSE)

```

 CompPepSetTest

Competitive peptide set test

Description

Given peptide-wise t-statistics and assignment of peptides to proteins, conduct peptide set tests to assess differential protein expression.

Usage

```

CompPepSetTest(
  result,
  pep_mapping_tbl,
  stat = c("t", "logFC"),
  cor_coef = 0,
  pepC.estim = c("sd", "mad")
)

```

Arguments

result output from EnframeContrastsRes

pep_mapping_tbl	a table mapping peptides to proteins. Alternatively, it can be the column name of the protein in rowData if dat is a SummarizedExperiment object.
stat	statistics to be used in the peptide set test. Options include "t" (t-statistic) and "logFC" (log2 fold change).
cor_coef	inter-peptide correlation coefficient(s)
pepC.estim	estimator of the variance of peptide-wise t-statistics not belonging to the protein of interest, i.e., test set. Options include "sd" and "mad". "sd" represents sample standard deviation. "mad" represents sample median absolute deviation.

Value

CompPepSetTest returns a dataframe containing the following columns

protein	unique protein identifier
NPeps	number of peptides
Correlation	inter-peptide correlation coefficient
Direction	direction of change
PValue	raw p-value
adj.P.Val	p-value adjusted via the Benjamini-Hochberg method
logFC	average log2 fold change of peptides
Up	number of upregulated peptides
Down	number of downregulated peptides

Author(s)

Junmin Wang

Examples

```
# Generate random peptide data
dat <- 2^matrix(rnorm(3000), ncol = 6)
colnames(dat) <- paste0("Sample", 1:6)
rownames(dat) <- paste0("Peptide", 1:500)

# Generate peptide mapping table
pep_mapping_tbl <- data.frame(peptide = paste0("Peptide", 1:500),
  protein = paste0("Protein", rep(1:100, each = 5)))

# Generate groups and contrasts
group <- c(rep("A", 3), rep("B", 3))
contrasts.par <- "B-A"

fit.cont <- FitContrasts(dat, contrasts.par, group)
cont.res <- EnframeContrastsRes(fit.cont)

# Run peptide set test based on t-statistics and standard deviation
CompPepSetTest(cont.res, pep_mapping_tbl, stat = "t",
```

```

cor_coef = 0, pepC.estim = "sd")

# Run peptide set test based on log2 fold change and median absolute deviation
CompPepSetTest(cont.res, pep_mapping_tbl, stat = "logFC",
cor_coef = 0, pepC.estim = "mad")

```

CompPepSetTestWorkflow

Competitive Peptide Set Test Workflow

Description

Given peptide abundance and assignment of peptide sequences to proteins, execute the competitive peptide set test workflow to compute the log₂ fold change, p-value, and adjusted p-value of all proteins identified.

Usage

```

CompPepSetTestWorkflow(
  dat,
  contrasts.par,
  group,
  pep_mapping_tbl,
  covar = NULL,
  stat = c("t", "logFC"),
  correlated = FALSE,
  equal.correlation = FALSE,
  pepC.estim = c("sd", "mad"),
  logged = FALSE
)

```

Arguments

dat	a dataframe or matrix of peptide abundance (row names should be peptide sequences or peptide IDs), or a SummarizedExperiment object where grouping and peptide-protein mapping are provided in colData and rowData, respectively.
contrasts.par	group levels to be compared separated by dash (e.g., "B-A" if group B is to be compared against group A)
group	a vector of group levels corresponding to each sample. Alternatively, it can be the column name of the group in colData if dat is a SummarizedExperiment object.
pep_mapping_tbl	a table mapping peptides to proteins (it should include two columns named "peptide" and "protein"). Alternatively, it can be the column name of the protein in rowData if dat is a SummarizedExperiment object.

covar	covariate matrix. Alternatively, it can be the column names of the covariates in colData if dat is a SummarizedExperiment object.
stat	statistics to be used in the peptide set test. Options include "t" (t-statistic) and "logFC" (log2 fold change).
correlated	Boolean variable indicating whether peptides are assumed to be correlated. If correlated, inter-peptide correlation will be estimated.
equal.correlation	Boolean variable indicating whether all pairwise inter-peptide correlation coefficients are assumed to be equal within a protein. If true, the mixed model approach will be applied; otherwise, the approach described in Wu and Smyth (2012), <i>Nucleic Acids Research</i> will be applied. Note that this parameter matters only if "correlated" is set to true.
pepC.estim	estimator of the variance of peptide-wise t-statistics not belonging to the protein of interest, i.e., test set. Options include "sd" and "mad". "sd" represents sample standard deviation. "mad" represents sample median absolute deviation.
logged	Boolean variable indicating whether abundance data have been log-transformed

Value

CompPepSetTestWorkflow returns a dataframe containing the following columns

protein	unique protein identifier
NPeps	number of peptides
Correlation	inter-peptide correlation coefficient
Direction	direction of change
PValue	raw p-value
adj.P.Val	p-value adjusted via the Benjamini-Hochberg method
logFC	average log2 fold change of peptides
Up	number of upregulated peptides
Down	number of downregulated peptides

Author(s)

Junmin Wang

References

Wu, D, and Smyth, GK (2012). Camera: a competitive gene set test accounting for inter-gene correlation. *Nucleic Acids Research* 40, e133

Examples

```
# Generate random peptide data
dat <- 2^matrix(rnorm(3000), ncol = 6)
colnames(dat) <- paste0("Sample", 1:6)
rownames(dat) <- paste0("Peptide", 1:500)
```

```

# Generate peptide mapping table
pep_mapping_tbl <- data.frame(peptide = paste0("Peptide", 1:500),
protein = paste0("Protein", rep(1:100, each = 5)))

# Generate groups and contrasts
group <- c(rep("A", 3), rep("B", 3))
contrasts.par <- "B-A"

CompPepSetTestWorkflow(dat, contrasts.par = contrasts.par,
group = group,
pep_mapping_tbl = pep_mapping_tbl,
stat = "t",
correlated = TRUE,
equal.correlation = TRUE,
pepC.estim = "mad",
logged = FALSE)

# Store data as a SummarizedExperiment object; add covariates
library(tibble)
library(SummarizedExperiment)
colData <- data.frame(sample = LETTERS[seq_along(group)], group = group,
sex = c("M", "F", "M", "F", "F", "M"), age = 1:6) |>
column_to_rownames(var = "sample")
rowData <- pep_mapping_tbl |> column_to_rownames(var = "peptide")
dat.nn <- dat
rownames(dat.nn) <- NULL
colnames(dat.nn) <- NULL
dat.se <- SummarizedExperiment(assays = list(int = dat.nn), colData = colData, rowData = rowData)

CompPepSetTestWorkflow(dat.se, contrasts.par = contrasts.par,
group = "group",
pep_mapping_tbl = "protein",
covar = c("sex", "age"),
stat = "t",
correlated = TRUE,
equal.correlation = TRUE,
pepC.estim = "mad",
logged = FALSE)

```

EnframeContrastsRes *Enframe result of LIMMA analysis*

Description

Convert result of LIMMA analysis into a dataframe.

Usage

```
EnframeContrastsRes(eBayes.fit, eb = TRUE)
```

Arguments

eBayes.fit	output from FitContrasts. See ?limma::eBayes for details.
eb	logical, whether to output the result from the empirical Bayes or ordinary approach.

Value

EnframeContrastsRes returns a dataframe containing the following columns

feature	unique feature identifier
logFC	log2 fold change
t	t-statistic
P.Value	raw p-value
adj.P.Val	p-value adjusted via the Benjamini-Hochberg method
B	B-statistic (empirical Bayes only)

Author(s)

Junmin Wang

Examples

```
# Generate random peptide data
dat <- 2^matrix(rnorm(3000), ncol = 6)
colnames(dat) <- paste0("Sample", 1:6)
rownames(dat) <- paste0("Peptide", 1:500)

# Generate groups and contrasts
group <- c(rep("A", 3), rep("B", 3))
contrasts.par <- "B-A"

fit.cont <- FitContrasts(dat, contrasts.par, group)
EnframeContrastsRes(fit.cont)
```

EstimInterPepCor

Estimation of inter-peptide correlation

Description

Given peptide abundance and assignment of peptide sequences to proteins, estimate inter-peptide correlation coefficient for each protein via the mixed model approach or approach described in Wu and Smyth (2012), *Nucleic Acids Research*.

Usage

```
EstimInterPepCor(
  dat,
  design,
  pep_mapping_tbl,
  equal.correlation = FALSE,
  logged = FALSE
)
```

Arguments

<code>dat</code>	a dataframe or matrix of peptide abundance, or a SummarizedExperiment object where grouping and peptide-protein mapping are provided in <code>colData</code> and <code>rowData</code> , respectively.
<code>design</code>	design matrix
<code>pep_mapping_tbl</code>	a table mapping peptides to proteins. Alternatively, it can be the column name of the protein in <code>rowData</code> if <code>dat</code> is a SummarizedExperiment object.
<code>equal.correlation</code>	Boolean variable indicating whether all pairwise inter-peptide correlation coefficients are assumed to be equal within a protein. If true, the mixed model approach will be applied; otherwise, the approach described in Wu and Smyth (2012), <i>Nucleic Acids Research</i> will be applied. In either case, only non-negative mean correlations are allowed.
<code>logged</code>	Boolean variable indicating whether abundance data have been log-transformed

Value

EstimInterPepCor returns a numeric vector of inter-peptide correlation coefficients (one value for each protein).

Author(s)

Junmin Wang and Steven Novick

References

Wu, D, and Smyth, GK (2012). Camera: a competitive gene set test accounting for inter-gene correlation. *Nucleic Acids Research* 40, e133

Examples

```
# Generate random peptide data
dat <- 2^matrix(rnorm(540), ncol = 6)
colnames(dat) <- paste0("Sample", 1:6)
rownames(dat) <- paste0("Peptide", 1:90)

# Generate peptide mapping table
pep_mapping_tbl <- data.frame(peptide = paste0("Peptide", 1:90),
```

```

protein = paste0("Protein", rep(1:30, each = 3))

# Generate design matrix
group <- c(rep("A", 3), rep("B", 3))
group <- factor(group)
design <- stats::model.matrix(~ 0 + group)

EstimInterPepCor(dat, design, pep_mapping_tbl,
equal.correlation = TRUE, logged = FALSE)

```

FitContrasts

Empirical Bayes moderated t-test

Description

Fit a linear model to feature abundance and compute moderated t-statistics via the empirical Bayes method.

Usage

```

FitContrasts(
  dat,
  contrasts.par,
  group,
  covar = NULL,
  logged = FALSE,
  NPeptide = NULL
)

```

Arguments

<code>dat</code>	a dataframe or matrix of feature (e.g., peptide, protein) abundance
<code>contrasts.par</code>	group levels to be compared separated by dash (e.g., "B-A" if group B is to be compared against group A)
<code>group</code>	list of group levels corresponding to each sample. The order of group levels needs to match that of samples in the feature abundance table.
<code>covar</code>	covariate matrix
<code>logged</code>	Boolean variable indicating whether data have been log-transformed
<code>NPeptide</code>	numeric vector indicating number of peptides aggregated for each protein. <code>logN-Peptide</code> will be passed to <code>limma-trend</code> . Constant prior variance if null.

Value

`FitContrasts` returns an object of class `MArrayLM`. See `?limma::eBayes` for details.

Author(s)

Junmin Wang

References

Ritchie, ME, Phipson, B, Wu, D, Hu, Y, Law, CW, Shi, W, and Smyth, GK (2015). limma powers differential expression analyses for RNA-sequencing and microarray studies. *Nucleic Acids Research* 43, e47.

Examples

```
# Generate random peptide data
dat <- 2^matrix(rnorm(3000), ncol = 6)
colnames(dat) <- paste0("Sample", 1:6)
rownames(dat) <- paste0("Peptide", 1:500)

# Generate groups and contrasts
group <- c(rep("A", 3), rep("B", 3))
contrasts.par <- "B-A"

# Run moderated t-test without covariates
FitContrasts(dat, contrasts.par, group)

# Run moderated t-test with covariates
covar <- matrix(c(1:6, 0, 1, 0, 1, 1, 0), nrow = 6, ncol = 2, byrow = FALSE)
FitContrasts(dat, contrasts.par, group, covar = covar)
```

FitLmerBySample

Fit a linear mixed model

Description

Fit a linear mixed model to the abundance of peptides belonging to one protein and compute the correlation coefficient based on variance components. Sample is treated as a random effect in the mixed model.

Usage

```
FitLmerBySample(y, design)
```

Arguments

y	a matrix of log2-transformed peptide abundance for one protein
design	design matrix

Value

FitLmerBySample returns the estimated inter-peptide correlation coefficient.

Author(s)

Junmin Wang and Steven Novick

Examples

```
y <- matrix(rnorm(1000*6), 1000, 6)
design <- cbind(Intercept = 1, Group = c(0, 0, 0, 1, 1, 1))

FitLmerBySample(y, design)
```

RobustReg

Robust Regression

Description

Estimate protein abundance by fitting a linear model to peptide abundance via robust regression.

Usage

```
RobustReg(dat, logged = FALSE)
```

Arguments

dat	a dataframe or matrix of peptide abundance, or a SummarizedExperiment object where grouping and peptide-protein mapping are provided in colData and rowData, respectively.
logged	Boolean variable indicating whether abundance data have been log-transformed

Value

RobustReg returns a numeric vector of estimated protein abundance.

Author(s)

Junmin Wang

References

Sticker, A, Goeminne, L, Martens, L, and Clement, L (2020). Robust Summarization and Inference in Proteome-wide Label-free Quantification. *Molecular & Cellular Proteomics* 19, 1209-19.

Gatto, L, Rainer, J, and Gibb, S (2021). MsCoreUtils: Core Utils for Mass Spectrometry Data. R package version 1.4.0. <https://github.com/RforMassSpectrometry/MsCoreUtils>

Examples

```
# Generate random peptide data
dat <- 2^matrix(rnorm(600), ncol = 6)
colnames(dat) <- paste0("Sample", 1:6)
rownames(dat) <- paste0("Peptide", 1:100)
```

```
RobustReg(dat, logged = FALSE)
```

SelfContPepSetTestWorkflow

Self-contained Peptide Set Test Workflow

Description

Given peptide abundance and assignment of peptide sequences to proteins, execute the self-contained peptide set test workflow to compute the log₂ fold change, p-value, and adjusted p-value of all proteins identified.

Usage

```
SelfContPepSetTestWorkflow(
  dat,
  contrasts.par,
  group,
  pep_mapping_tbl,
  covar = NULL,
  logged = FALSE
)
```

Arguments

dat	a dataframe or matrix of peptide abundance (row names should be peptide sequences or peptide IDs), or a SummarizedExperiment object where grouping and peptide-protein mapping are provided in colData and rowData, respectively.
contrasts.par	group levels to be compared separated by dash (e.g., "B-A" if group B is to be compared against group A)
group	a vector of group levels corresponding to each sample. Alternatively, it can be the column name of the group in colData if dat is a SummarizedExperiment object.
pep_mapping_tbl	a table mapping peptides to proteins (it should include two columns named "peptide" and "protein"). Alternatively, it can be the column name of the protein in rowData if dat is a SummarizedExperiment object.
covar	covariate matrix. Alternatively, it can be the column names of the covariates in colData if dat is a SummarizedExperiment object.
logged	Boolean variable indicating whether abundance data have been log-transformed

Value

SelfContPepSetTestWorkflow returns a dataframe containing the following columns

protein	unique protein identifier
NPeps	number of peptides
Direction	direction of change
PValue	raw p-value
adj.P.Val	adjusted p-value
logFC	average log2 fold change of peptides
Up	number of upregulated peptides
Down	number of downregulated peptides

Author(s)

Junmin Wang

References

Wu, D, Lim, E, Francois Vaillant, F, Asselin-Labat, M-L, Visvader, JE, and Smyth, GK (2010). ROAST: rotation gene set tests for complex microarray experiments. *Bioinformatics* 26, 2176-2182

Examples

```
# Generate random peptide data
dat <- 2^matrix(rnorm(3000), ncol = 6)
colnames(dat) <- paste0("Sample", 1:6)
rownames(dat) <- paste0("Peptide", 1:500)

# Generate peptide mapping table
pep_mapping_tbl <- data.frame(peptide = paste0("Peptide", 1:500),
protein = paste0("Protein", rep(1:100, each = 5)))

# Generate groups and contrasts
group <- c(rep("A", 3), rep("B", 3))
contrasts.par <- "B-A"

SelfContPepSetTestWorkflow(dat, contrasts.par = contrasts.par,
group = group,
pep_mapping_tbl = pep_mapping_tbl,
logged = FALSE)

# Store data as a SummarizedExperiment object; add covariates
library(tibble)
library(SummarizedExperiment)
colData <- data.frame(sample = LETTERS[seq_along(group)], group = group,
sex = c("M", "F", "M", "F", "F", "M"), age = 1:6) |>
column_to_rownames(var = "sample")
rowData <- pep_mapping_tbl |> column_to_rownames(var = "peptide")
dat.nn <- dat
```

```

rownames(dat.nn) <- NULL
colnames(dat.nn) <- NULL
dat.se <- SummarizedExperiment(assays = list(int = dat.nn), colData = colData, rowData = rowData)

SelfContPepSetTestWorkflow(dat.se, contrasts.par = contrasts.par,
group = "group",
pep_mapping_tbl = "protein",
covar = c("sex", "age"),
logged = FALSE)

```

TTestwCor

Two-sample t-test accounting for inter-peptide correlation

Description

Test whether peptides belonging to the same protein are differentially expressed relative to the rest of the peptidome, accounting for inter-peptide correlation. This function is adapted from `cameraPR()` in LIMMA R package (Wu and Smyth (2012), *Nucleic Acids Research*).

Usage

```
TTestwCor(statistic, index, inter.pep.cor, pepC.estim = c("sd", "mad"))
```

Arguments

<code>statistic</code>	a numeric vector of peptide-wise t-statistics.
<code>index</code>	an index vector or a list of index vectors. <code>statistic[index]</code> selects corresponding rows in the protein of interest, i.e., test set(s).
<code>inter.pep.cor</code>	a numeric vector of inter-peptide correlation coefficients within the protein of interest, i.e., test set(s).
<code>pepC.estim</code>	estimator of the variance of peptide-wise t-statistics not belonging to the protein of interest, i.e., test set. Options include "sd" and "mad". "sd" represents sample standard deviation. "mad" represents sample median absolute deviation.

Value

TTestwCor returns a dataframe in which each row represents a protein of interest, i.e., test set. Columns include

<code>NPeps</code>	number of peptides
<code>Correlation</code>	inter-peptide correlation coefficient
<code>Direction</code>	direction of change
<code>PValue</code>	raw p-value
<code>adj.P.Val</code>	p-value adjusted via the Benjamini-Hochberg method

Author(s)

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References

Wu, D, and Smyth, GK (2012). Camera: a competitive gene set test accounting for inter-gene correlation. *Nucleic Acids Research* 40, e133

Examples

```
y <- matrix(rnorm(1000 * 6), 1000, 6)
design <- cbind(Intercept = 1, Group = c(0, 0, 0, 1, 1, 1))

# First set of 20 genes are genuinely differentially expressed
index1 <- 1:20
y[index1, 4:6] <- y[index1, 4:6]+1

fit <- limma::eBayes(limma::lmFit(y, design))
TTestwCor(fit$t[, 2], index = index1,
inter.pep.cor = 0,
pepC.estim = "sd")
```

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