

Package ‘mQTL’

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Type Package

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Description mQTL provides a complete QTL analysis pipeline for metabolomic data.

Distinctive features include normalisation using PQN approach, peak alignment using RSPA approach, dimensionality reduction using SRV approach and finally QTL mapping using R/qtl package.

License GPL

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Description

mQTL provides a complete QTL analysis pipeline for 1H NMR data. Distinctive features include normalisation using QPN approach, peak alignment using RSPA approach, dimensionality reduction using SRV approach and finally QTL mapping using R/qtl package.

Details

Package:	<i>mQTL</i>
Type:	Package
Version:	1.0
Date:	2013-09-18
License:	GPL (>= 3)

Main fucntions:

- `format_mQTL`: creates the proper format of data
- `align_mQTL`: peak alignment and normalisation
- `pre_mQTL`: dimension reduction by statistical recoupling of variables
- `process_mQTL`: computes LODs using extended Haley-Knott method
- `post_mQTL`: plots the results of a given run
- `summary_mQTL`: provides the results as a table

Author(s)

Lyamine Hedjazi and Jean-Baptiste Cazier

Maintainer: Lyamine Hedjazi <<l.hedjazi@fondation-ican.com>>

alignSp	<i>Base function for Spectrum Alignment</i>
---------	---

Description

Alignment of spectrum segment to a spectrum of interest

Usage

```
alignSp(refSp, refSegments, intSp, intSegments, recursion, MAX_DIST_FACTOR, MIN_RC)
```

Arguments

refSp	reference spectrum
refSegments	reference segments
intSp	spectrum of interest
intSegments	segments of interest
recursion	parameters for recursive alignment
MAX_DIST_FACTOR	the distance matching parameter (0.5*peak width)
MIN_RC	minimum resemblance coefficient

Value

alignedSpectrum	aligned spectrum
extendedSegments	extended segments

Author(s)

Lyamine Hedjazi
Maintainer: Lyamine Hedjazi <<1.hedjazi@fondation-ican.com>>

See Also

[align_mQTL](#)

Examples

```
## Not run:  
## Data  
  
Sp=matrix(rnorm(10*13454,mean=0,sd=1), nrow=10,ncol=13454)
```

```

##Segmentation parameters

peakParam=list()
peakParam$ppmDist <- 0.03 #(ppm) # distance to concatenate adjacent peaks #default 0.03#
peakParam$ampThr <- 0.3 # amplitude value to threshold small peaks
peakParam$minPeakWidth <- 0.005 # min peak width in ppm scale
peakParam$iFrameLen<-11 # Savitzky-Golay frame length in ppm scale
peakParam$iOrder<-3 # polynomial order of Savitzky - Golay filter
peakParam$peakEdgeMax<-0.2 # maximal peak edge

##Recusrson alignment parameters

recursion=list()
recursion$resemblance<-0.95# Stop criterium of the recursion indicating
#the complete alignment of segment peaks
recursion$segShift<-0.02#(ppm) max peak shift for large peaks
recursion$inbetweenShift<-0.02 #(ppm) max shift for small peaks
recursion$acceptance<-0.5 # if resemblance after the alignment between modified test
recursion$minSegWidth<-0.01 #(ppm) Stop criteria of the recursion - the size of the smallest peak
recursion$step<-0.02 # Recursion step (default 0.02)

## Normalisation

normD<-normalise(Sp,'prob')

## Reference spectrum selection

index<-selectRefSp(normD$Sp,recursion$step)
refSp<-normD$Sp[index,] # reference spectrum picked-up

##segmentate a reference spectrum

refSegments<- segmentateSp(refSp, peakParam) # segmentate reference spectrum

##segmentate a test spectrum

spectrum<-normD$Sp[10,]
testSegments<- segmentateSp(spectrum, peakParam) # segmentate test spectrum (10th sample)

#match test and reference segments

attachedSegs<-attachSegments(refSegments,testSegments)

refSegments<-attachedSegs$refSegmentsNew
testSegments<-attachedSegs$testSegmentsNew

##matching parameters

MAX_DIST_FACTOR<-0.5 # The distance matching parameter (0.5*peak_width)
MIN_RC<-0.25 # Minimum resamblance coefficient

Segs<-matchSegments(refSp,spectrum, testSegments,refSegments,MAX_DIST_FACTOR,MIN_RC)

```

```
#align a test spectrum  
  
refSgs<-Segs$refSegs  
tstSgs<-Segs$testSegs  
  
SpAlg<- alignSp(refSp,refSgs,spectrum,tstSgs,recursion,MAX_DIST_FACTOR,MIN_RC)  
  
## End(Not run)
```

align_mQTL

Peak alignment and normalisation of metabolomic data

Description

Recursive Segment-Wise Peak Alignment (RSPA) for accounting peak position variation across metabolomic data

Usage

```
align_mQTL(datafile, outdat)
```

Arguments

datafile	raw spectra
outdat	aligned spectra

Details

The algorithm is based on the following workflow:

1. Quotient probabilistic normalisation of metabolomic data.
2. Automatic selection of a reference spectrum.
3. Segmentate a reference spectrum.
4. Then for each test spectrum:
 - segmentate a test spectrum.
 - match test and reference segments.
 - align a test spectrum.

Value

It returns aligned data.

Author(s)

Lyamine Hedjazi

References

Veselkov,K. et al (2009) Recursive Segment-Wise Peak Alignment of Biological 1H NMR Spectra for Improved Metabolic Biomarker Recovery, Anal. Chem., 81(1), 56-66.

See Also

[alignSp](#), [attachSegments](#), [matchSegments](#), [segmentateSp](#), [format_mQTL](#), [format_mQTL](#)

Examples

```
## Not run:
## Align metabolomic data profiles

cleandat<-"CleanMetaboFile.dat" ## Metabolomic data file in csvs format
aligdat<-"AlignData.dat" ## Aligned metabolomic profiles in csvs format

align_mQTL(cleandat,aligdat)

## End(Not run)
```

attachSegments	<i>Concatenation of test and reference segments</i>
-----------------------	---

Description

Concatenation of test and reference segments to ensure one-to-one correspondence.

Usage

```
attachSegments(refSegments,testSegments)
```

Arguments

refSegments	segments of the reference spectrum
testSegments	segments of the test spectrum

Details

The algorithm:

1. For each reference segment within segment boundaries, i.e. between initial and final positions, find all centre (middle) positions of test segments and merge those segments, if more than one centre position is found
2. Apply the same procedure for each test segment

Value

A list:

```
segments$start  start of each concatenated test segment
segments$PeakLeftBoundary
                  peak left boundary of each concatenated test segment
segments$PeakRightBoundary
                  peak right boundary of each concatenated test segment
segments$Peaks  peaks of each concatenated test segment
segments$end    end of each concatenated test segment
segments$center center of each concatenated test segment
```

Author(s)

Lyamine Hedjazi

References

Veselkov,K. et al (2009) Recursive Segment-Wise Peak Alignment of Biological 1H NMR Spectra for Improved Metabolic Biomarker Recovery, Anal. Chem., 81(1), 56-66.

See Also

[matchSegments](#)

Examples

```
## Not run:

## Data
Sp=matrix(rnorm(10*13454,mean=0,sd=1), nrow=10,ncol=13454)

##Segmentation parameters

peakParam=list()
peakParam$ppmDist <- 0.03# (ppm) # distance to concatenate adjacent peaks #default 0.03#
peakParam$ampThr <- 0.3 # amplitude value to threshold small peaks #
peakParam$minPeakWidth <- 0.005 #min peak width in ppm scale
peakParam$iFrameLen<-11 #Savitzky-Golay frame length in ppm scale
peakParam$iOrder<-3 #polynomial order of Savitzky - Golay filter
peakParam$peakEdgeMax<-0.2

##Reference spectrum selection

step<-0.02 # Recursion step (default 0.02)
index<-selectRefSp(Sp,step)
refSp<-Sp[index,]
```

```
#segmentate a reference spectrum
refSegments<- segmentateSp(refSp, peakParam) # segmentate reference spectrum

#segmentate a test spectrum
spectrum<-Sp[10,]
testSegments<- segmentateSp(spectrum, peakParam) # segmentate test spectrum (10th sample)

# match test and reference segments
attachedSegs<-attachSegments(refSegments,testSegments)

## End(Not run)
```

format_mQTL*Routine to reformat the data into the required format***Description**

This function enables to reformat data into the proper format. The user should provides in input metabolomic dataset, Genotype dataset and a dataset containing sex and pgm (parental grandmother).

Usage

```
format_mQTL(datafile, genofile, physdat, outdat, outgeno)
```

Arguments

datafile	metabolomic data file
genofile	genotype data file
physdat	a file containing sex and pgm
outdat	phenotype data (metabolomic data + sex + pgm) in the format csvs
outgeno	genotype data

Value

It returns phynotype and genotype files in the proper format

Author(s)

Lyamine Hedjazi

See Also

[align_mQTL](#),

Examples

```

## Not run:
## Clean the raw data to match the genotype and phenotype and create the proper format

rawfile<-"MetaboFile.dat" ## Metabolomic data file
genofile<-"GenoFile.dat" ## Genotype data file
physiodat="physiodat.dat" ## data file containing sex and pgm
cleandat<-"CleanMetaboFile.dat" ## Metabolomic data file in csvs format
cleangen<-"CleanGenoFile.dat" ## Genotype data file in csvs format

format_mQTL(rawfile,genofile,physiodat, cleandat,cleangen)

## End(Not run)

```

matchSegments

Matching of the segment of interest to the corresponding reference

Description

The algorithm makes use of a fuzzy logic approach to match the segment of interest to the corresponding reference

Usage

```
matchSegments(refSp, intSp, intSegments, refSegments, MAX_DIST_FACTOR, MIN_RC)
```

Arguments

refSp	spectrum of reference
intSp	spetcrum of interest (test spectrum)
intSegments	segments of spectrum of interest
refSegments	segments of reference spectrum
MAX_DIST_FACTOR	the distance matching parameter (0.5*peak_width)
MIN_RC	minimum resamblance coefficient

Details

Algorithm:

1. take segment of interest
2. take reference segments
3. calculate relative distance between them
4. calculate relative resamblance between them
5. find min value of relative distance and resamblance
6. use it as representative of similiarity between target and reference segments
7. find the segment that has the highest value of both relative distance and resamblance

Value

A list:

testSegs	matched test segments
refSegs	matched reference segments

Author(s)

Lyamine Hedjazi

References

Veselkov,K. et al (2009) Recursive Segment-Wise Peak Alignment of Biological 1H NMR Spectra for Improved Metabolic Biomarker Recovery, Anal. Chem., 81(1), 56-66.

See Also

[attachSegments](#)

Examples

```
## Not run:

# Data

Sp=matrix(rnorm(10*13454,mean=0,sd=1), nrow=10,ncol=13454)

##Segmentation parameters

peakParam=list()
peakParam$ppmDist <- 0.03# (ppm) # distance to concatenate adjacent peaks #default 0.03#
peakParam$ampThr <- 0.3 # amplitude value to threshold small peaks #
peakParam$minPeakWidth <- 0.005 #min peak width in ppm scale
peakParam$iFrameLen<-11 #Savitzky-Golay frame length in ppm scale
peakParam$iOrder<-3 #polynomial order of Savitzky - Golay filter
peakParam$peakEdgeMax<-0.2

##reference spectrum selection

step=0.02 # Recursion step (default 0.02)
index<-selectRefSp(Sp,step)
refSp<-Sp[index,]

#segmentate a reference spectrum

refSegments<- segmentateSp(refSp, peakParam) # segmentate reference spectrum

#segmentate a test spectrum

spectrum<-Sp[10,]
```

```
testSegments<- segmentateSp(spectrum, peakParam) # segmentate test spectrum (10th sample)

#attach test and reference segments
attachedSegs<-attachSegments(refSegments,testSegments)

##Matching parameters

MAX_DIST_FACTOR<-0.5 # The distance matching parameter (0.5*peak_width)
MIN_RC<-0.25 # Minimum resamblance coefficient

refSegments<-attachedSegs$refSegmentsNew
testSegments<-attachedSegs$testSegmentsNew
Segs<-matchSegments(refSp,spectrum, testSegments,refSegments,MAX_DIST_FACTOR, MIN_RC)

## End(Not run)
```

normalise*Normalisation of metabolomic data*

Description

Removing dilutions between biofluid samples (normalisation of spectra)

Usage

```
normalise(X, method)
```

Arguments

X	metabolomic data
method	total area (method<- "total") or quotient probabilistic method (method<- "prob")

Value

normalised spectrum

Author(s)

Lyamine Hedjazi

References

Dieterle,F., et al (2006) Probabilistic Quotient Normalization as Robust Method to Account for Dilution of Complex Biological Mixtures. Application in ¹H NMR Metabonomics, Anal. Chem., 78(13), 42814290.

See Also

[SRV](#)

Examples

```
## Data
Sp=matrix(rnorm(10*13454,mean=0,sd=1), nrow=10,ncol=13454)

## Quotient probabilistic normalisation
NormDat<-normalise(abs(Sp),'prob')
```

peakPeaks

Peak picking algorithm

Description

Identification of peaks in metabolomic data based on the calculation of smoothed derivates using Savitzky-Golay filter. The peak is identified if derivative crosses zero, i.e. $\text{sign}(X'(i)) > \text{sign}(X'(i+1))$.

Usage

```
peakPeaks(SpSmooth, dpDerivs, Sp)
```

Arguments

SpSmooth	smoothed spectrum
dpDerivs	smoothed derivative of the spectrum
Sp	Spectrum of interest

Value

identified peaks

Author(s)

Lyamine Hedjazi

References

Veselkov,K. et al (2009) Recursive Segment-Wise Peak Alignment of Biological 1H NMR Spectra for Improved Metabolic Biomarker Recovery, Anal. Chem., 81(1), 56-66.

See Also

[sgolayDeriv](#)

Examples

```
## Data
Sp=matrix(rnorm(10*13454,mean=0,sd=1), nrow=10,ncol=13454)

## Peak picking
Spectrum<-Sp[,]
iOrder <- 3
iFrameLen<- 11

SpDerivs=sgolayDeriv(Spectrum,iOrder,iFrameLen,2)
SpSmooth = sgolayDeriv(Spectrum,iOrder,iFrameLen,1)
peaks=peakPeaks(SpSmooth,SpDerivs,Spectrum)
```

post_mQTL

Plot top LOD results

Description

Function to plot the results of a given run

Usage

```
post_mQTL(results, probs = c(0.95, 0.99, 0.999, 0.9999))
```

Arguments

- | | |
|---------|---|
| results | results of mQTL analysis. |
| probs | numeric vector of probabilities with values in [0,1]. (Values up to 2e-14 outside that range are accepted and moved to the nearby endpoint.). |

Details

This function plots different results corresponding to top LOD marker

Value

It returns graphs and summaries

Author(s)

Jean-Baptiste Cazier

See Also

[pre_mQTL](#)

Examples

```

## Not run:
## Pre-process data

infile<-"ReducedData.dat" ## Reduced data by SRV
cleangen<-"CleanGenoFile.dat" ## Genotype data file in csvs format
nperm <- 0 ## Number of permutations
mQTL_results<-process_mQTL(infile, cleangen, nperm)

post_mQTL(results)## Plot mQTL results

## End(Not run)

```

ppersp

Plot a 3-D profile of LODs

Description

Plot 3-D profile of LODs as function of genomic position and chemical shift

Usage

```
ppersp(z, ppm, titre, theta=-15, phi=15, r=50)
```

Arguments

<code>z</code>	table of results
<code>ppm</code>	chemical shift
<code>titre</code>	title
<code>theta</code>	angle defining the viewing direction (azimuthal direction)
<code>phi</code>	angle defining the viewing direction (colatitude direction)
<code>r</code>	the distance of the eyepoint from the centre of the plotting box.

Value

plot 2D-profile

Author(s)

Jean-Baptiste Cazier

See Also

[pplot](#)

Examples

```
## Plot 3D profile  
  
## Not run:  
x11(width=5,height=5,pointsize=5)  
titel<-"Example plot"  
ppersp(results, ppm, title)  
  
## End(Not run)
```

pplot

Plot a color scale layer

Description

Plot the results with a color scale y layer over 3 in 2D

Usage

```
pplot(z, titre, ppm, res, LT = c(5,10,15,20))
```

Arguments

<code>z</code>	mQTL's whole results
<code>titre</code>	figure title
<code>ppm</code>	chemical shift
<code>res</code>	results to be plotted
<code>LT</code>	<code>quantil(res,probs)</code> , res: results and probs: vector of probabilities

Value

2-D profile

Author(s)

Jean-Baptiste Cazier

See Also

[ppersp](#)

Examples

```
## Not run:

## Plot 3D profile

x11(width=5,height=5,pointsize=5)
title<-"Example plot"

probs=c(0.95,0.99,0.999,0.9999) ## probabilities

pplot(res,"Full 2D Profile", ppm, best, quantile(res,probs=probs))

## End(Not run)
```

pre_mQTL

Statistical Recoupling of variables for mQTL analysis

Description

Makes use of SRV to preprocess metabolomic data for dimensionality reduction by statistical re-coupling of variables

Usage

```
pre_mQTL(infile, outfile, met, corrT = 0.9)
```

Arguments

<code>infile</code>	metabolomic datafile
<code>outfile</code>	reduced metabolomic datafile
<code>met</code>	used statistical summary
<code>corrT</code>	correlation threshold

Details

The SRV algorithm forms clusters of variables using a measure of a local spectral dependency. Then tests whether consecutive clusters are correlated to aggregate them into a single supercluster.

Value

The algorithm:

1. variables are associated into a series of clusters.
2. integration of clusters into superclusters.

Author(s)

Jean-Baptiste Cazier and Lyamine Hedjazi

References

Blaise,B. et al (2009) Statistical recoupling prior to significance testing in nuclear magnetic resonance based metabonomics, Anal. Chem., 81(15), 6242-6251.

See Also

[SRV](#),[post_mQTL](#)

Examples

```
## Not run:
## Pre-process data

infile<-"AlignData.dat" ## Aligned metabolomic profiles in csvs format
outfile<-"ReducedData.dat" ## Reduced data by SRV
met<- "rectangle" ## Summary measure (mean, max,...)
corrT<- 0.9 ## Correlation threshold (default 0.9)
(pre_mQTL(infile, outfile, met, corrT)

## End(Not run)
```

process_mQTL

mQTL mapping

Description

Function to process the tissue extract of the individuals for QTL analysis

Usage

```
process_mQTL(datfile, genfile, nperm = 0)
```

Arguments

datfile	phenotype data
genfile	genotype data
nperm	nperm

Details

This function makes use of metabolomic and genotype data to perform QTL analysis based on the R/QTL package, for mapping quantitative trait loci. In particular, it makes use of the extended Haley-Knott method to optimize the LOD score evaluation and avoid problems with missing genotypes.

Value

2D LOD score table

Author(s)

Jean-Baptiste Cazier and Hedjazi Lyamine

References

Broman,K., et al (2006) R/qtl: QTL mapping in experimental crosses, Bioinformatics, 19(7), 889-890.

See Also

[post_mQTL](#)

Examples

```
## Not run:
## Pre-process data
infile<-"ReducedData.dat" ## Reduced data by SRV
cleangen<-"CleanGenoFile.dat" ## Genotype data file in csvs format
nperm <- 0 ## Number of permutations
MQTL_results<-process_mQTL(infile, cleangen, nperm)

## End(Not run)
```

segmentateSp

Segmentation of a spectrum of interest

Description

Determination of highly intensive peaks in the spectrum of interest and subsequent concatenation of closely located peaks into larger segments

Usage

`segmentateSp(Sp, peakParam)`

Arguments

Sp	spectrum
peakParam	a list:
	<ul style="list-style-type: none"> • ampThr: amplitude threshold [default 2*median(peaksMaxValues)] • iFrameLen: Savitzky-Golay frame length • iOrder: polynomial order of Savitzky - Golay filter • iFrameLen: Savitzky-Golay frame length • minPeakWidth: min peak size • ppmDist: distance to concatenate adjacent peaks

Value

A list:

```
testSegmentsNew  
    new test segments  
refSegmentsNew new reference segments
```

Author(s)

Lyamine Hedjazi

References

Veselkov,K. et al (2009) Recursive Segment-Wise Peak Alignment of Biological 1H NMR Spectra for Improved Metabolic Biomarker Recovery, Anal. Chem., 81(1), 56-66.

See Also

[attachSegments](#), [matchSegments](#)

Examples

```
# Data  
  
Sp=matrix(rnorm(10*13454,mean=0,sd=1), nrow=10,ncol=13454)  
  
##Segmentation parameters  
  
peakParam=list()  
peakParam$ppmDist <- 0.03# (ppm) # distance to concatenate adjacent peaks #default 0.03#  
peakParam$ampThr <- 0.3 # amplitude value to threshold small peaks #  
peakParam$minPeakWidth <- 0.005 #min peak width in ppm scale  
peakParam$iFrameLen<-11 #Savitzky-Golay frame length in ppm scale  
peakParam$iOrder<-3 #polynomial order of Savitzky - Golay filter  
peakParam$peakEdgeMax<-0.2  
  
#segmentate a test spectrum (10th sample)  
  
Spectr<-Sp[10,]  
testSegments<- segmentateSp(Spectr, peakParam)
```

selectRefSp

Automated selection of a reference spectrum

Description

The selection of reference spectrum among all spectra is based on the highest similarity to all other spectra

Usage

```
selectRefSp(X, step)
```

Arguments

X	spectra
step	used to scale spectral regions down to specific bin size

Value

returns the index of selected spectrum

Author(s)

Lyamine Hedjazi

See Also

[alignSp](#)

Examples

```
# Data
Sp=matrix(rnorm(10*13454,mean=0,sd=1), nrow=10,ncol=13454)

# Reference spectrum selection

step=0.02 # Recursion step (default 0.02)
index<-selectRefSp(Sp,step)
```

sgolay

Find the matrix of differentiation filters

Description

designs a Savitzky-Golay (polynomial) FIR smoothing filter. The polynomial order must be less than the frame size which must be odd.

Usage

```
sgolay(k,F,W)
```

Arguments

k	polynomial order
F	frame size
W	weighting matrix

Value

matrix of differentiators

Author(s)

Lyamine Hedjazi

References

Sophocles J. Orfanidis, INTRODUCTION TO SIGNAL PROCESSING, Prentice-Hall, 1995, Chapter 8

See Also

[sgolayDeriv](#)

Examples

```
k <- 3  
F <- 11
```

```
Sg=sgolay(k,F)
```

`sgolayDeriv` *Calculate smoothed derivates*

Description

Calculate smoothed derivates using Savitzky-Golay filter

Usage

```
sgolayDeriv(dpSpectr, iOrder, iFrameLen, j)
```

Arguments

dpSpectr	input spectrum
iOrder	polynomial order of Savitzky - Golay filter
iFrameLen	Savitzky-Golay frame length in ppm scale
j	order of derivative

Value

jth derivative of the spectrum

Author(s)

Lyamine Hedjazi

See Also

[sgolay](#)

Examples

```
## Data
Sp=matrix(rnorm(10*13454,mean=0,sd=1), nrow=10,ncol=13454)

## Peak picking
Spectrum<-Sp[10,]
iOrder <- 3
iFrameLen<- 11
j<-2

SpDerivs=sgolayDeriv(Spectrum,iOrder,iFrameLen,j)
```

Description

Base function for dimensionality reduction by statistical recoupling of variables

Usage

```
SRV(X, minsize, correl, clustf = median)
```

Arguments

X	data matrix
minsize	singlet size
correl	bucketting resolution
clustf	correlation threshold

Value

A list:

indicesdebf	starting border of superclusters
indicesfinf	ending border of superclusters

Author(s)

Jean-Baptiste Cazier

References

Blaise,B. et al (2009) Statistical recoupling prior to significance testing in nuclear magnetic resonance based metabonomics, Anal. Chem., 81(15), 6242-6251.

See Also

[pre_mQTL](#)

Examples

```
## Not run:

## Statistical recoupling of variables

corrT=0.9 # correlation threshold
minsize=10 # singlet size
met="rectangle" # summary measure

#Perform the SRV analysis to reduce the number of dimension of Spectra data (Sp)

SRV<-SRV(Sp, minsize, corrT,clustf=met)

## End(Not run)
```

Description

Correlation generation for Statistical Recoupling of Variables.

Usage

SRV_Corr(X, Y, minsize, correl)

Arguments

X	data matrix
Y	class matrix
minsize	bucketting resolution
correl	correlation threshold

Value

A list:

Pfinal	pvalue vector
indicesdebft	starting border of cluster
indicesfinfl	ending border of cluster
Correlation	Correlation of superclusters/clusters

Author(s)

Jean-Baptiste Cazier

References

Blaise,B. et al (2009) Statistical recoupling prior to significance testing in nuclear magnetic resonance based metabonomics, Anal. Chem., 81(15), 6242-6251.

See Also

[SRV](#), [pre_mQTL](#)

summary_mQTL

Function to summarize the results of all the runs and their differences

Description

This function generates a table containing the genetic markers and their associated metabolomic variables and estimated LOD score.

Usage

```
summary_mQTL(results, Th = 5)
```

Arguments

results	mQTL mapping results
Th	Threshold of the top accepted LOD

Details

Generates a text file containing a table of results

Value

returns Summaries

Author(s)

Jean-Baptiste Cazier and Lyamine Hedjazi

See Also

[pre_mQTL](#)

Examples

```
## Not run:  
  
## Pre-process data  
  
infile<-"ReducedData.dat" ## Reduced data by SRV  
cleangen<-"CleanGenoFile.dat" ## Genotype data file in csvs format  
nperm <- 0 ## Number of permutations  
MQTL_results<-process_mQTL(infile, cleangen, nperm))  
  
T=10 ## LOD threshold  
summary_mQTL(results,T=8)## summarizes mQTL results in a table  
  
## End(Not run)
```

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