Package 'CoGAPS'

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| Title Coordinated | Gene Activity in Pattern Sets |
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| implements a GAPS, and l process activ | dinated Gene Activity in Pattern Sets (CoGAPS) a Bayesian MCMC matrix factorization algorithm, inks it to gene set statistic methods to infer biological ity. It can be used to perform sparse matrix factorization on I when this data represents biomolecules, to do gene set |
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| Depends R ($>= 3$. gplots ($>= 2$. | 0.1), Rcpp (>= 0.11.2), RColorBrewer (>= 1.0.5), 8.0) |
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| binaryA calcCoGA calcZ . CoGAPS computeC gapsInter gapsMapi gapsMapi gapsRun gapsTestI | TestRun 13 16 Run 18 |
| GIS1.D | |

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| CoGA | PS-package | CoGAPS: Coordinated Gene Analysis Pattern Sets | |
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Description

CoGAPS implements a Bayesian MCMC matrix factorization algorithm, GAPS, and links it to gene set statistic methods to infer biological process activity. It can be used to perform sparse matrix factorization on any data, and when this data represents biomolecules, to do gene set analysis.

Package: CoGAPS Type: Package Version: 1.0

Date: 2014-07-23 License: LGPL

Author(s)

Maintainer: Elana J. Fertig <ejfertig@jhmi.edu>, Michael F. Ochs <ochsm@tcnj.edu>

References

Fertig EJ, Ding J, Favorov AV, Parmigiani G, Ochs MF. CoGAPS: an R/C++ package to identify patterns and biological process activity in transcriptomic data. Bioinformatics. 2010 Nov 1;26(21):2792-3

| binaryA | binaryA creates a binarized heatmap of the A matrix in which the value is 1 if the value in Amean is greater than threshold * Asd and 0 otherwise |
|---------|---|
| | |

Description

binaryA creates a binarized heatmap of the A matrix in which the value is 1 if the value in Amean is greater than threshold * Asd and 0 otherwise

calcCoGAPSStat 3

Usage

binaryA(Amean, Asd, threshold = 3)

Arguments

Amean the mean estimate for the A matrix
Asd the standard deviations on Amean

threshold the number of standard deviations above zero that an element of Amean must be

to get a value of 1

calcCoGAPSStat CoGAPS gene set statistic

Description

Computes the p-value for the association of underlying patterns from microarray data to activity in gene sets.

Usage

calcCoGAPSStat(Amean, Asd, GStoGenes, numPerm=500)

Arguments

Amean Sampled mean value of the amplitude matrix A. row.names(Amean) must cor-

respond to the gene names contained in GStoGenes.

Asd Sampled standard deviation of the amplitude matrix A.

GStoGenes List or data frame containing the genes in each gene set. If a list, gene set

names are the list names and corresponding elements are the names of genes contained in each set. If a data frame, gene set names are in the first column and corresponding gene names are listed in rows beneath each gene set name.

numPerm Number of permuations used for the null distribution in the gene set statistic.

(optional; default=500)

Details

This script links the patterns identified in the columns of P to activity in each of the gene sets specified in GStoGenes using a novel z-score based statistic developed in Ochs et al. (2009). Specifically, the z-score for pattern p and gene set G_i containing G total genes is given by

$$Z_{i,p} = \frac{1}{G} \sum_{ginG_i} A_{gp} / \sigma_{gp}$$

, where g indexes the genes in the set and σ_{gp} is the standard deviation of A_{gp} obtained from MCMC sampling. CoGAPS then uses the specified numPerm random sample tests to compute a consistent p value estimate from that z score.

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Value

A list containing:

GSUpreg p-values for upregulation of each gene set in each pattern.

GSDownreg p-values for downregulation of each gene set in each pattern.

GSActEst p-values for activity of each gene set in each pattern.

Author(s)

```
Elana J. Fertig <ejfertig@jhmi.edu>
```

References

M.F. Ochs, L. Rink, C. Tarn, S. Mburu, T. Taguchi, B. Eisenberg, and A.K. Godwin. (2009) Detection and treatment-induced changes in signaling pathways in gastrointestinal stromal tumors using transcriptomic data. Cancer Research, 69:9125-9132.

See Also

CoGAPS

| calcZ | calcZ calculates the Z-score for each element based on input mean |
|-------|---|
| | and standard deviation matrices |

Description

calcZ calculates the Z-score for each element based on input mean and standard deviation matrices

Usage

```
calcZ(meanMat, sdMat)
```

Arguments

meanMat matrix of mean values

sdMat matrix of standard deviation values

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| CoGAPS | Cogaps calls the C++ MCMC code through gapsRun and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix and then calls calcCoGAPSStat to estimate gene |
|--------|--|
| | set activity with nPerm set to 500 |

Description

CoGAPS calls the C++ MCMC code through gapsRun and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix and then calls calcCoGAPSStat to estimate gene set activity with nPerm set to 500

Usage

```
CoGAPS(data, unc, ABins = data.frame(), PBins = data.frame(), GStoGenes,
    nFactor = 7, simulation_id = "simulation", nEquil = 1000,
    nSample = 1000, nOutR = 1000, output_atomic = FALSE,
    fixedBinProbs = FALSE, fixedDomain = "N", sampleSnapshots = TRUE,
    numSnapshots = 100, plot = TRUE, nPerm = 500, alphaA = 0.01,
    nMaxA = 1e+05, max_gibbmass_paraA = 100, alphaP = 0.01, nMaxP = 1e+05,
    max_gibbmass_paraP = 100)
```

Arguments

alphaA

| Ę | guinenes | |
|---|-----------------|---|
| | data | data matrix |
| | unc | uncertainty matrix (std devs for chi-squared of Log Likelihood) |
| | ABins | a matrix of same size as A which gives relative probability of that element being non-zero |
| | PBins | a matrix of same size as P which gives relative probability of that element being non-zero |
| | GStoGenes | data.frame or list with gene sets |
| | nFactor | number of patterns (basis vectors, metagenes) |
| | simulation_id | name to attach to atoms files if created |
| | nEquil | number of iterations for burn-in |
| | nSample | number of iterations for sampling |
| | nOutR | how often to print status into R by iterations |
| | output_atomic | whether to write atom files (large) |
| | fixedBinProbs | Boolean for using relative probabilities given in Abins and Pbins |
| | fixedDomain | character to indicate whether A or P is domain for relative probabilities |
| | sampleSnapshots | |
| | | Boolean to indicate whether to capture individual samples from Markov chain during sampling |
| | numSnapshots | the number of individual samples to capture |
| | plot | Boolean to indicate whether to produce output graphics |
| | nPerm | number of permutations in gene set test |
| | | |

sparsity parameter for A domain

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nMaxA PRESENTLY UNUSED, future = limit number of atoms

max_gibbmass_paraA

limit truncated normal to max size

alphaP sparsity parameter for P domain

nMaxP PRESENTLY UNUSED, future = limit number of atoms

max_gibbmass_paraP

limit truncated normal to max size

Details

CoGAPS first decomposes the data matrix using GAPS, D, into a basis of underlying patterns and then determines the gene set activity in each of these patterns.

The GAPS decomposition is achieved by finding amplitude and pattern matrices (**A** and **P**, respectively) for which

$$\mathbf{D} = \mathbf{AP} + \Sigma$$
,

where Σ is the matrix of uncertainties given by unc. The matrices **A** and **P** are assumed to have the atomic prior described in Sibisi and Skilling (1997) and are found with MCMC sampling.

Then, the patterns identified in the columns of P are linked to activity in each of the gene sets specified in GStoGenes using a novel z-score based statistic developed in Ochs et al. (2009). Specifically, the z-score for pattern p and gene set G_i containing \$G\$ total genes is given by

$$Z_{i,p} = \frac{1}{G} \sum_{gin\mathcal{G}_{i}} \frac{\mathbf{A_{gp}}}{Asd_{gp}},$$

where g indexes the genes in the set and Asd_{gp} is the standard deviation of \mathbf{A}_{gp} obtained from MCMC sampling. CoGAPS then uses the specified nPerm random sample tests to compute a consistent p value estimate from that z score. Note that the data from Ochs et al. (2009) are provided with this package in GIST_TS_20084.RData and TFGSList.RData are also provided with this package for further validation.

Value

A list containing:

meanChi2 Value of chi^2 for Amean and Pmean.

D Data matrix **D** input to factorization.

Sigma uncertainty matrix (std devs for chi-squared of Log Likelihood)

Amean Sampled mean value of the amplitude matrix **A**.

Asd Sampled standard deviation of the amplitude matrix **A**.

Pmean Sampled mean value of the amplitude matrix **P**.

Psd Sampled standard deviation of the amplitude matrix **P**.

GSUpreg p-values for upregulation of each gene set in each pattern.

GSDownreg p-values for downregulation of each gene set in each pattern.

GSActEst p-values for activity of each gene set in each pattern.

See Also

gapsRun,calcCoGAPSStat

computeGeneGSProb 7

Examples

computeGeneGSProb

CoGAPS gene membership statistic

Description

Computes the p-value for gene set membership using the CoGAPS-based statistics developed in Fertig et al. (2012). This statistic refines set membership for each candidate gene in a set specified in GSGenes by comparing the inferred activity of that gene to the average activity of the set. Specifically, we compute the following summary statistic for each gene g that is a candidate member of gene set G:

$$S_{g,G} = (\sum_{p} -log(Pr_{G,p})Pw[p](A_{gp}/\sigma_{gp}))/\sum_{p} -log(Pr_{G,p})Pw[p],$$

where p indexes each of the patterns, $Pr_{G,p}$ is the probability that gene set G is upregulated computed with calcCoGAPSStat, A_{gp} is the mean amplitude matrix from the GAPS matrix factorization, Pw[p] is a prior weighting for each pattern based upon the context to which that pattern relates, and σ_{gp} is the standard deviation of the amplitude matrix. P-values are formulated from a permutation test comparing the value of $S_{g,G}$ for genes in GSGenes relative to the value of $S_{g,G}$ numPerm random gene sets with the same number of targets.

Usage

computeGeneGSProb(Amean, Asd, GSGenes, Pw=rep(1,ncol(Amean)),numPerm=500,PwNull=F)

Arguments

Amean Sampled mean value of the amplitude matrix A. row.names(Amean) must cor-

respond to the gene names contained in GSGenes.

Asd Sampled standard deviation of the amplitude matrix **A**.

GSGenes Vector containing the prior estimate of members of the gene set of interest.

Pw Vector containing the weight to assign each pattern in the gene statistic assumed

to be computed from the association of the pattern with samples in a given con-

text (optional: default=1 giving all patterns equal weight).

computeGeneGSProb

numPerm Number of permuations used for the null distribution in the gene set statistic.

(optional; default=500)

PwNull Logical value. If TRUE, use pattern weighting in Pw when computing the null

distribution for the statistic. If FALSE, do not use the pattern weighting so that

the null is context independent. (optional; default=F)

Value

A vector of length GSGenes containing the p-values of set membership for each gene containined in the set specified in GSGenes.

Author(s)

```
Elana J. Fertig <ejfertig@jhmi.edu>
```

References

E.J. Fertig, A.V. Favorov, and M.F. Ochs (2013) Identifying context-specific transcription factor targets from prior knowledge and gene expression data. 2012 IEEE Nanobiosciences.

See Also

```
calcCoGAPSStat
```

Examples

```
## Not run:
# Results for GIST data in Fertig et al. (2012) #
# load the data
data('GIST_TS_20084')
data('TFGSList')
\# define transcription factors of interest based on Ochs et al. (2009)
TFs <- c("c.Jun", 'NF.kappaB', 'Smad4', "STAT3", "Elk.1", "c.Myc", "E2F.1", "AP.1", "CREB", "F0X0", "p53", "Sp1")
# run the GAPS matrix factorization
nIter <- 10000
results <- CoGAPS(GIST.D, GIST.S, tf2ugFC,
                 nFactor=5,
                 nEquil=nIter, nSample=nIter,
                 plot=FALSE)
# set membership statistics
permTFStats <- list()</pre>
for (tf in TFs) {
     genes <- levels(tf2ugFC[,tf])</pre>
     genes <- genes[2:length(genes)]</pre>
     permTFStats[[tf]] <- computeGeneTFProb(Amean = GISTResults$Amean,</pre>
                                          Asd = GistResults$Asd, genes)
}
```

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End(Not run)

gapsInterPattern gapsInterPattern calculates statistics for measuring the distance between patterns based on genes associated with the patterns

Description

gapsInterPattern calculates the overlap in significant genes between patterns according to the set standard deviation threshold from the mean. Returns the names of significant genes, a matrix of overlap fractions, and the mean of this matrix. A warning is issued if a column of **A** does not have significant genes at a given threshold.

Usage

```
gapsInterPattern(Amean, Asd, sdThreshold = 3)
```

Arguments

Amean Sampled mean value of the amplitude matrix **A** from a run of CoGAPS.

Asd Sampled standard deviation of the amplitude matrix **A** from a run of CoGAPS.

sdThreshold How many standard deviations a gene's sampled mean needs to be above 0 to

be considered significantly expressed in a pattern.

Details

This calculates a statistic for comparing different patterns based on the genes associated with them.

Value

A list containing:

SignificantGeneNames

A list in which each element holds a vector of the name of the significant genes in each column of **A**.

 ${\tt SignificantGeneTotals}$

A vector of significant genes in each respective column.

SeparationMatrix

The matrix containing the overlap fractions for the significant genes in each column of ${\bf A}$.

InterPatternValue

The final value of the Interpattern measure.

See Also

CoGAPS

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Examples

Description

gapsIntraPatternObtains correlation matrices for significantly expressed genes in each column of A. Obtains the means of these matrices, then averages those to get a sense of how closely correlated genes in the patterns of CoGAPS are.

Usage

```
gapsIntraPattern(Amean, Asd, DMatrix, sdThreshold = 3)
```

Arguments

Amean Sampled mean value of the amplitude matrix **A** from a run of CoGAPS.

Asd Sampled standard deviation of the amplitude matrix **A** from a run of CoGAPS.

DMatrix data matrix

 $\operatorname{\mathsf{sdThreshold}}$ How many standard deviations a gene's sampled mean needs to be above 0 to

be considered significantly expressed in a pattern.

Details

This calculates a statistic for determining how tight a pattern is based on the genes associated with it and their distribution in the data.

Value

A list containing:

CorrelationMatrices

A list containing the correlation matrices between the significant genes in each column of ${\bf A}$.

CorrelationMatrixMeans

A list containing the means of the Correlation Matrices.

IntraPatternValue

The final value of the Intrapattern measure.

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See Also

CoGAPS

Examples

gapsMapRun

gapsMapRun calls the C++ MCMC code and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix; as opposed to gapsRun, this method takes an additional input specifying set patterns in the P matrix

Description

gapsMapRun calls the C++ MCMC code and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix; as opposed to gapsRun, this method takes an additional input specifying set patterns in the P matrix

Usage

```
gapsMapRun(D, S, FP, ABins = data.frame(), PBins = data.frame(),
    nFactor = 5, simulation_id = "simulation", nEquil = 1000,
    nSample = 1000, nOutR = 1000, output_atomic = "FALSE",
    fixedMatrix = "P", fixedBinProbs = "FALSE", fixedDomain = "N",
    sampleSnapshots = "TRUE", numSnapshots = 100, alphaA = 0.01,
    nMaxA = 1e+05, max_gibbmass_paraA = 100, alphaP = 0.01, nMaxP = 1e+05,
    max_gibbmass_paraP = 100)
```

Arguments

| D | data matrix |
|-------|--|
| S | uncertainty matrix (std devs for chi-squared of Log Likelihood) |
| FP | data.frame with rows giving fixed patterns for P |
| ABins | a matrix of same size as A which gives relative probability of that element being non-zero |
| PBins | a matrix of same size as P which gives relative probability of that element being non-zero |

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nFactor number of patterns (basis vectors, metagenes), which must be greater than or

equal to the number of rows of FP

simulation_id name to attach to atoms files if created

nEquil number of iterations for burn-in nSample number of iterations for sampling

nOutR how often to print status into R by iterations

output_atomic whether to write atom files (large)

fixedMatrix character indicating whether A or P matrix has fixed columns or rows respec-

tively

fixedBinProbs Boolean for using relative probabilities given in Abins and Pbins

fixedDomain character to indicate whether A or P is domain for relative probabilities

sampleSnapshots

Boolean to indicate whether to capture individual samples from Markov chain

during sampling

numSnapshots the number of individual samples to capture

alphaA sparsity parameter for A domain

nMaxA PRESENTLY UNUSED, future = limit number of atoms

max_gibbmass_paraA

limit truncated normal to max size

alphaP sparsity parameter for P domain

nMaxP PRESENTLY UNUSED, future = limit number of atoms

max_gibbmass_paraP

limit truncated normal to max size

Details

The decomposition in GAPS is achieved by finding amplitude and pattern matrices (A and P, respectively) for which

$$\mathbf{D} = \mathbf{AP} + \boldsymbol{\Sigma}$$

, where Σ is the matrix of uncertainties given by S. The matrices $\bf A$ and $\bf P$ are assumed to have the atomic prior described in Sibisi and Skilling (1997) and are found with MCMC sampling. However, some rows of $\bf P$ are fixed to be the values specified in the input argument FP after rescaling to have norm 1.

Value

A list containing:

Amean Sampled mean value of the amplitude matrix **A**.

Asd Sampled standard deviation of the amplitude matrix **A**.

Pmean Sampled mean value of the amplitude matrix \mathbf{P} .

Psd Sampled standard deviation of the amplitude matrix **P**.

atomsAEquil Number of atoms in **A** during each iteration of the equilibration phase.

Number of atoms in **A** during each iteration of the sampling phase.

Number of atoms in **P** during each iteration of the equilibration phase.

Number of atoms in **P** during each iteration of the sampling phase.

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chiSqValues Value of chi^2 at each step during equilibration and sampling.

meanChi2 Value of chi^2 for Amean and Pmean.

ASnapshots Samples of A matrices taken during sampling.
PSnapshots Samples of P matrices taken during sampling.

See Also

CoGAPS,gapsRun

Examples

gapsMapTestRun

gapsMapTestRun calls the C++ MCMC code and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix; as opposed to gapsRun, this method takes an additional input specifying set patterns in the P matrix. Test procedures allow for the returning of the matrix and atomic information for A and P during each step of the equilibration and sampling.

Description

gapsMapTestRun calls the C++ MCMC code and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix; as opposed to gapsRun, this method takes an additional input specifying set patterns in the P matrix. Test procedures allow for the returning of the matrix and atomic information for A and P during each step of the equilibration and sampling.

Usage

```
gapsMapTestRun(D, S, FP, ABins = data.frame(), PBins = data.frame(), nFactor = 7,
    simulation_id = "simulation", nEquil = 1000, nSample = 1000,
    nOutR = 1000, output_atomic = FALSE, fixedMatrix="P", fixedBinProbs = FALSE,
    fixedDomain = "N", alphaA = 0.01, nMaxA = 1e+05, max_gibbmass_paraA = 100,
    alphaP = 0.01, nMaxP = 1e+05, max_gibbmass_paraP = 100)
```

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Arguments

D data matrix

S uncertainty matrix (std devs for chi-squared of Log Likelihood)

FP data.frame with rows giving fixed patterns for P

ABins a matrix of same size as A which gives relative probability of that element being

non-zero

PBins a matrix of same size as P which gives relative probability of that element being

non-zero

nFactor number of patterns (basis vectors, metagenes), which must be greater than or

equal to the number of rows of FP

simulation_id name to attach to atoms files if created

nEquil number of iterations for burn-in nSample number of iterations for sampling

nOutR how often to print status into R by iterations

output_atomic whether to write atom files (large)

fixedMatrix character indicating whether A or P matrix has fixed columns or rows respec-

tively

fixedBinProbs Boolean for using relative probabilities given in Abins and Pbins

fixedDomain character to indicate whether A or P is domain for relative probabilities

alphaA sparsity parameter for A domain

nMaxA PRESENTLY UNUSED, future = limit number of atoms

max_gibbmass_paraA

limit truncated normal to max size

alphaP sparsity parameter for P domain

nMaxP PRESENTLY UNUSED, future = limit number of atoms

max_gibbmass_paraP

limit truncated normal to max size

Details

The decomposition in GAPS is achieved by finding amplitude and pattern matrices (\mathbf{A} and \mathbf{P} , respectively) for which

 $\mathbf{D} = \mathbf{AP} + \Sigma$

, where Σ is the matrix of uncertainties given by S. The matrices **A** and **P** are assumed to have the atomic prior described in Sibisi and Skilling (1997) and are found with MCMC sampling.

Value

A list containing:

Amean Sampled mean value of the amplitude matrix **A**.

Asd Sampled standard deviation of the amplitude matrix **A**.

Pmean Sampled mean value of the amplitude matrix **P**.

Psd Sampled standard deviation of the amplitude matrix **P**.

atomsAEquil Number of atoms in **A** during each iteration of the equilibration phase.

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| atomsASamp | Number of atoms in A during each iteration of the sampling phase. |
|----------------|--|
| atomsPEquil | Number of atoms in ${\bf P}$ during each iteration of the equilibration phase. |
| atomsPSamp | Number of atoms in ${\bf P}$ during each iteration of the sampling phase. |
| chiSqValues | Value of chi^2 at each step during equilibration and sampling. |
| matricesAEquil | State of the A matrix during each iteration of the equilibration phase. |
| matricesASamp | State of the A matrix during each iteration of the sampling phase. |
| matricesPEquil | State of the ${\bf P}$ matrix during each iteration of the equilibration phase. |
| matricesPSamp | State of the ${\bf P}$ matrix during each iteration of the sampling phase. |
| domainAEquil | A list containing the locations and magnitudes of the atoms of ${\bf A}$ during each iteration of the equilibration phase. |
| domainASamp | A list containing the locations and magnitudes of the atoms of ${\bf A}$ during each iteration of the sampling phase. |
| domainPEquil | A list containing the locations and magnitudes of the atoms of ${\bf P}$ during each iteration of the equilibration phase. |
| domainPSamp | A list containing the locations and magnitudes of the atoms of ${\bf P}$ during each iteration of the sampling phase. |
| meanChi2 | Value of chi^2 for Amean and Pmean. |

See Also

CoGAPS

Examples

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| gapsRun | gapsRun calls the C++ MCMC code and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix |
|---------|--|
|---------|--|

Description

gapsRun calls the C++ MCMC code and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix

Usage

```
gapsRun(D, S, ABins = data.frame(), PBins = data.frame(), nFactor = 7,
 simulation_id = "simulation", nEquil = 1000, nSample = 1000,
 nOutR = 1000, output_atomic = "FALSE", fixedBinProbs = "FALSE",
 fixedDomain = "N", sampleSnapshots = "TRUE", numSnapshots = 100,
 alphaA = 0.01, nMaxA = 1e+05, max_gibbmass_paraA = 100, alphaP = 0.01,
 nMaxP = 1e+05, max_gibbmass_paraP = 100)
```

Arg

nMaxP

max_gibbmass_paraP

| guments | | |
|--------------------|--|--|
| D | data matrix | |
| S | uncertainty matrix (std devs for chi-squared of Log Likelihood) | |
| ABins | a matrix of same size as A which gives relative probability of that element being non-zero (optional) | |
| PBins | a matrix of same size as P which gives relative probability of that element being non-zero (optional) | |
| nFactor | number of patterns (basis vectors, metagenes), which must be greater than or equal to the number of rows of FP (optional, defaults to 7) | |
| simulation_id | name to attach to atoms files if created | |
| nEquil | number of iterations for burn-in | |
| nSample | number of iterations for sampling | |
| nOutR | how often to print status into R by iterations | |
| output_atomic | whether to write atom files (large) | |
| fixedBinProbs | Boolean for using relative probabilities given in Abins and Pbins | |
| fixedDomain | character to indicate whether A or P is domain for relative probabilities | |
| sampleSnapshots | | |
| | Boolean to indicate whether to capture individual samples from Markov chain during sampling | |
| numSnapshots | the number of individual samples to capture | |
| alphaA | sparsity parameter for A domain | |
| nMaxA | PRESENTLY UNUSED, future = limit number of atoms | |
| max_gibbmass_paraA | | |
| | limit truncated normal to max size | |
| alphaP | sparsity parameter for P domain | |

PRESENTLY UNUSED, future = limit number of atoms

limit truncated normal to max size

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Details

The decomposition in GAPS is achieved by finding amplitude and pattern matrices (\mathbf{A} and \mathbf{P} , respectively) for which

$$\mathbf{D} = \mathbf{AP} + \Sigma$$

, where Σ is the matrix of uncertainties given by S. The matrices **A** and **P** are assumed to have the atomic prior described in Sibisi and Skilling (1997) and are found with MCMC sampling.

Value

A list containing:

Amean Sampled mean value of the amplitude matrix **A**.

Asd Sampled standard deviation of the amplitude matrix **A**.

Pmean Sampled mean value of the amplitude matrix \mathbf{P} .

Psd Sampled standard deviation of the amplitude matrix **P**.

atomsAEquil Number of atoms in **A** during each iteration of the equilibration phase.

Number of atoms in **A** during each iteration of the sampling phase.

Number of atoms in **P** during each iteration of the equilibration phase.

atomsPSamp Number of atoms in **P** during each iteration of the sampling phase.

chiSqValues Value of chi^2 at each step during equilibration and sampling.

meanChi2 Value of chi^2 for Amean and Pmean.

ASnapshots Samples of A matrices taken during sampling.
PSnapshots Samples of P matrices taken during sampling.

See Also

CoGAPS

Examples

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| gapsTestRun | gapsTestRun calls the C++ MCMC code and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix. Test procedures allow for the returning of the matrix and atomic information for A and B during each step of the equilibration |
|-------------|--|
| | atomic information for A and P during each step of the equilibration and sampling |
| | ana sampung |

Description

gapsTestRun calls the C++ MCMC code and performs Bayesian matrix factorization returning the two matrices that reconstruct the data matrix. Test procedures allow for the returning of the matrix and atomic information for A and P during each step of the equilibration and sampling.

Usage

```
gapsTestRun(D, S, ABins = data.frame(), PBins = data.frame(), nFactor = 7,
  simulation_id = "simulation", nEquil = 1000, nSample = 1000,
 nOutR = 1000, output_atomic = "FALSE", fixedBinProbs = "FALSE",
 fixedDomain = "N", alphaA = 0.01, nMaxA = 1e+05, max_gibbmass_paraA = 100,
 alphaP = 0.01, nMaxP = 1e+05, max_gibbmass_paraP = 100)
```

Arguments

max_gibbmass_paraP

| ξ | guments | |
|--------------------|---------------|--|
| | D | data matrix |
| | S | uncertainty matrix (std devs for chi-squared of Log Likelihood) |
| | ABins | a matrix of same size as A which gives relative probability of that element being non-zero |
| | PBins | a matrix of same size as P which gives relative probability of that element being non-zero |
| | nFactor | number of patterns (basis vectors, metagenes), which must be greater than or equal to the number of rows of FP |
| | simulation_id | name to attach to atoms files if created |
| | nEquil | number of iterations for burn-in |
| | nSample | number of iterations for sampling |
| | nOutR | how often to print status into R by iterations |
| | output_atomic | whether to write atom files (large) |
| | fixedBinProbs | Boolean for using relative probabilities given in Abins and Pbins |
| | fixedDomain | character to indicate whether A or P is domain for relative probabilities |
| | alphaA | sparsity parameter for A domain |
| | nMaxA | PRESENTLY UNUSED, future = limit number of atoms |
| max_gibbmass_paraA | | |
| | | limit truncated normal to max size |
| | alphaP | sparsity parameter for P domain |
| | nMaxP | PRESENTLY UNUSED, future = limit number of atoms |

limit truncated normal to max size

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Details

The decomposition in GAPS is achieved by finding amplitude and pattern matrices (\mathbf{A} and \mathbf{P} , respectively) for which

$$\mathbf{D} = \mathbf{AP} + \Sigma$$

, where Σ is the matrix of uncertainties given by S. The matrices ${\bf A}$ and ${\bf P}$ are assumed to have the atomic prior described in Sibisi and Skilling (1997) and are found with MCMC sampling.

Value

A list containing:

| Amean | an Sampled mean value of the amplitude matrix A . | |
|----------------|--|--|
| Asd | Sampled standard deviation of the amplitude matrix A . | |
| Pmean | Sampled mean value of the amplitude matrix P . | |
| Psd | Sampled standard deviation of the amplitude matrix P . | |
| atomsAEquil | Number of atoms in A during each iteration of the equilibration phase. | |
| atomsASamp | Number of atoms in A during each iteration of the sampling phase. | |
| atomsPEquil | Number of atoms in P during each iteration of the equilibration phase. | |
| atomsPSamp | Number of atoms in P during each iteration of the sampling phase. | |
| chiSqValues | Value of chi^2 at each step during equilibration and sampling. | |
| matricesAEquil | State of the A matrix during each iteration of the equilibration phase. | |
| matricesASamp | State of the A matrix during each iteration of the sampling phase. | |
| matricesPEquil | State of the ${\bf P}$ matrix during each iteration of the equilibration phase. | |
| matricesPSamp | State of the ${\bf P}$ matrix during each iteration of the sampling phase. | |
| domainAEquil | A list containing the locations and magnitudes of the atoms of ${\bf A}$ during each iteration of the equilibration phase. | |
| domainASamp | A list containing the locations and magnitudes of the atoms of ${\bf A}$ during each iteration of the sampling phase. | |
| domainPEquil | A list containing the locations and magnitudes of the atoms of ${\bf P}$ during each iteration of the equilibration phase. | |
| domainPSamp | A list containing the locations and magnitudes of the atoms of ${\bf P}$ during each iteration of the sampling phase. | |
| meanChi2 | Value of chi^2 for Amean and Pmean. | |
| | | |

See Also

CoGAPS

Examples

20 GIST.S

```
## Plot the results
plotGAPS(results$Amean, results$Pmean, 'GSFigs')
## End(Not run)
```

GIST.D

Sample GIST gene expression data from Ochs et al. (2009).

Description

Gene expression data from gastrointestinal stromal tumor cell lines treated with Gleevec.

Usage

GIST_TS_20084

Format

Matrix with 1363 genes by 9 samples of mean gene expression data.

References

Ochs, M., Rink, L., Tarn, C., Mburu, S., Taguchi, T., Eisenberg, B., and Godwin, A. (2009). Detection of treatment-induced changes in signaling pathways in gastrointestinal stromal tumors using transcriptomic data. Cancer Res, 69(23), 9125-9132.

GIST.S

Sample GIST gene expression data from Ochs et al. (2009).

Description

Standard deviation of gene expression data from gastrointestinal stromal tumor cell lines treated with Gleevec.

Usage

GIST_TS_20084

Format

Matrix with 1363 genes by 9 samples containing standard deviation (GIST.S) of the gene expression data.

References

Ochs, M., Rink, L., Tarn, C., Mburu, S., Taguchi, T., Eisenberg, B., and Godwin, A. (2009). Detection of treatment-induced changes in signaling pathways in gastrointestinal stromal tumors using transcriptomic data. Cancer Res, 69(23), 9125-9132.

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GSets

Simulated dataset to quantify gene set membership.

Description

Simulated gene sets used to generate amplitude matrix in SimpSim. A and corresponding data SimpSim. D.

Usage

GSets

Format

A list containing names of genes in two simulated gene sets used to generate the data in SimpSim.D.

plotAtoms

plotAtoms a simple plot of the number of atoms from one of the vectors returned with atom numbers

Description

plotAtoms a simple plot of the number of atoms during the sampling period or equilibration periodfrom one of either A or P as specified in type.

Usage

```
plotAtoms(gapsRes, type = "sampA")
```

Arguments

gapsRes

the list resulting from applying GAPS

type

the atoms to plot, values are sampA, sampP, equilA, or equilP to plot sampling

or equilibration teop atome numbers

plotDiag

plotDiag plots a series of diagnostic plots

Description

plotDiag plots a series of diagnostic plots

Usage

```
plotDiag(gapsRes)
```

Arguments

gapsRes

list returned by gapsRun, gapsMapRun, or CoGAPS

22 plotP

plotGAPS

Plotter for GAPS decomposition results

Description

Plots the A and P matrices obtained from the GAPS matrix decomposition.

Usage

```
plotGAPS(A, P, outputPDF="")
```

Arguments

A The amplitude matrix **A** obtained from GAPS.

P The pattern matrix **P** obtained from GAPS.

outputPDF Name of an pdf file to which the results will be output. (Optional; default=""

will output plots to screen).

Note

If the plot option is true in CoGAPS, this function will be called automatically to plot results to the screen.

Author(s)

Elana J. Fertig <efertig@jhmi.edu>

See Also

CoGAPS

plotP

plotP plots the P matrix in a line plot with error bars

Description

plotP plots the P matrix in a line plot with error bars

Usage

```
plotP(PMean_Mat, P_SD)
```

Arguments

PMean_Mat matrix of mean values of P

P_SD matrix of standard deviation values of P

plotSmoothPatterns 23

| tSmoothPatterns Plot loess smoothed CoGAPS patterns |
|---|
|---|

Description

Plots the sampled mean value of the pattern matrix ${\bf P}$ obtained from the CoGAPS matrix factorization vs. a specificed X value for each sample in the columns of ${\bf P}$. Lines plot loess normalized values of ${\bf P}$ vs specified X variables.

Usage

plotSmoothPatterns(P, x=NULL, breakStyle=T, orderP=!all(is.null(x)), plotPTS=F, policy and the plotSmoothPatterns(P, x=NULL, breakStyle=T, orderP=!all(is.null(x)), plotPTS=F, orderP=!all(is.null(x)), plotPTS=F, orderP=!all(is.null(x)), plotPTS=F, orderP=!all(is.null(x)), plotPTS=F, orderP=!all(is.null(x)), plotPTS=F, orderP=!all(is.

Arguments

| Р | A [p, M] pattern matrix (P.mean) obtained from the CoGAPS matrix factorization. |
|------------|--|
| X | A [M, 1] matrix of values for the X axis for each of the corresponding M columns of P. (Optional: Default: $x=1:M$) |
| breaks | A vector of X values at which breaks in plotting should occur. Loess lines fit to data will start and stop at breaks. (Optional: Default: no breaks). May also be specified as an integer to determine the number of equal groups into which to divide the data. |
| breakStyle | A logical vector. If TRUE, the corresponding break will start a new plot on the row for each pattern. If FALSE, a vertical line will demarcate the break point. (Optional: Defaults to all hard breaks). Note, if one logical value is used, that value will determine the break type at each break point. |
| orderP | A logical value. If TRUE, vertical ordering of patterns will be determined in order of the value of x at which they peak. If FALSE, vertical ordering will be determined by the rows in the P matrix. (Optional: Default: FALSE) |
| plotPTS | A logical value. If TRUE, plot will include points for each value of the P matrix in addition to the loess smoothed curve. If FALSE, only the loess smoothed values of P will be plotted. (Optional: Default: FALSE) |
| pointCol | Color of points of the P matrix plotted when plotPTS=TRUE. (Optional: Default: black) |
| lineCol | Color of loess smoothed values of the P matrix. (Optional: Default: grey) |
| add | A logical value. If TRUE, plot will be added to existing graphics device. If FALSE, will create a new graphics device. (Optional: Default: FALSE) |
| | Additional arguments to plotting functions. |

Author(s)

Genevieve Stein-O'Brien <gsteino1@jhmi.edu>

See Also

CoGAPS

24 residuals

Examples

```
## Not run:
# create simulated data
P <- rbind(1:10 + rnorm(10), seq(from=10,to=1) + rnorm(10))

# saved as PDF since figure margins are often too large for the null device with this function
# and the null device may also have trouble with the overlay
pdf('Test.pdf', width=10)
plotSmoothPatterns(P=P, x=rep(seq(from=1,to=10,by=2),each=2), breaks=3, breakStyle=c(F,T,T), plotPTS=T)

# demonstrating the overlay of the plot
plotSmoothPatterns(P=P, x=rep(seq(from=1,to=10,by=2),each=2), breaks=c(0.992,3.660,6.340,9.010), breakStyle=cv.off()

## End(Not run)</pre>
```

reorderByPatternMatch Match two sets of patterns found with CoGAPS

Description

Matches two sets of pattern matrices (of the same size) found with CoGAPS. Matches are identified by finding identifying subsequently decreasing correlations between patterns in the respective matrices.

Usage

```
reorderByPatternMatch(P, matchTo)
```

Arguments

P Pattern matrix for which rows will be arranged to match the matrix in matchTo matchTo Pattern matrix to which P is matched.

Value

Pattern matrix derived from reordering columns of P

residuals

residuals calculate residuals and produce heatmap

Description

residuals calculate residuals and produce heatmap

Usage

```
residuals(AMean_Mat, PMean_Mat, D, S)
```

SimpSim.A 25

Arguments

| AMean_Mat | matrix of mean values for A from GAPS |
|-----------|---------------------------------------|
| PMean_Mat | matrix of mean values for P from GAPS |
| D | original data matrix run through GAPS |
| | |

S original standard deviation matrix run through GAPS

| SimpSim.A | Simulated | data |
|-----------|-----------|------|
|-----------|-----------|------|

Description

True amplitude matrix generated from gene sets in GSets used to generate simulated data in SimpSim. D.

Usage

SimpSim.A

Format

Matrix with 30 genes by 3 patterns of true amplitude used to generate simulated data.

|--|--|

Description

Simulated gene expression data from true patterns in SimpSim.P and amplitude in SimpSim.A.

Usage

SimpSim.D

Format

Matrix with 30 genes by 20 samples of simulated gene expression data.

Description

True pattern matrix used to generate simulated data in SimpSim.D.

Usage

SimpSim.P

Format

Matrix with 3 patterns by 20 samples of true patterns used to generate simulated data.

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SimpSim.S

Simulated data

Description

Standard deviation of simulated gene expression data from true patterns in SimpSim.P and amplitude in SimpSim.A.

Usage

SimpSim.S

Format

Matrix with 30 genes by 20 samples of containing standard deviation of simulated gene expression data.

tf2ugFC

Gene sets defined by transcription factors defined from TRANSFAC.

Description

List of genes contained in gastrointestinal stromal tumor cell line measurements that are regulated by transcription factors in the TRANSFAC database. Used for the gene set analysis in Ochs et al. (2009).

Usage

TFGSList

Format

Data.frame containing genes (rows) regulated by each transcription factor (columns).

References

Ochs, M., Rink, L., Tarn, C., Mburu, S., Taguchi, T., Eisenberg, B., and Godwin, A. (2009). Detection of treatment-induced changes in signaling pathways in gastrointestinal stromal tumors using transcriptomic data. Cancer Res, 69(23), 9125-9132.

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