Package: GGPA (via r-universe)

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Type Package

Title graph-GPA: A graphical model for prioritizing GWAS results and investigating pleiotropic architecture

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Description Genome-wide association studies (GWAS) is a widely used tool for identification of genetic variants associated with phenotypes and diseases, though complex diseases featuring many genetic variants with small effects present difficulties for traditional these studies. By leveraging pleiotropy, the statistical power of a single GWAS can be increased. This package provides functions for fitting graph-GPA, a statistical framework to prioritize GWAS results by integrating pleiotropy. 'GGPA' package provides user-friendly interface to fit graph-GPA models, implement association mapping, and generate a phenotype graph.

License GPL (>= 2)

URL https://github.com/dongjunchung/GGPA/

Depends R (>= 4.0.0), stats, methods, graphics, GGally, network, sna, scales, matrixStats

Suggests BiocStyle

Imports Rcpp (>= 0.11.3)

LinkingTo Rcpp, RcppArmadillo

RcppModules cGGPAmodule

NeedsCompilation yes

biocViews Software, StatisticalMethod, Classification, GenomeWideAssociation, SNP, Genetics, Clustering, MultipleComparison, Preprocessing, GeneExpression, DifferentialExpression

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`GGPA-package` *A graphical model to investigate genetic relationship among multiple phenotypes (short line)*

Description

More about what it does (maybe more than one line)

Details

The DESCRIPTION file:

Package:	GGPA
Type:	Package
Title:	A graphical model to investigate genetic relationship among multiple phenotypes (short line)
Version:	2.1.0
Date:	2016-07-02
Author:	Hang J. Kim, Dongjun Chung
Maintainer:	Hang J. Kim <hang.kim@uc.edu>
Description:	More about what it does (maybe more than one line)
License:	What license is it under?
Imports:	Rcpp (>= 0.11.3)
LinkingTo:	Rcpp, RcppArmadillo
RcppModules:	cGGPAmodule
NeedsCompilation:	yes
Packaged:	2015-08-20 04:11:34 UTC; hangkim

Index of help topics:

GGPA-package	What the package does (short line)
rcpparma_hello_world	Set of functions in example RcppArmadillo package

~~ An overview of how to use the package, including the most important functions ~~

Author(s)

Hang J. Kim, Dongjun Chung

Maintainer: Hang J. Kim <hang.kim@uc.edu>

References

~~ Literature or other references for background information ~~

See Also

~~ Optional links to other man pages, e.g. ~~ ~~ <pkg> ~~

Examples

```
~~ simple examples of the most important functions ~~
```

---

assoc *Association mapping*

Description

Association mapping.

Usage

```
assoc( object, ... )
## S4 method for signature 'GGPA'
assoc( object, FDR=0.05, fdrControl="global", i=NULL, j=NULL )
```

Arguments

object	A GGPA model fit as obtained by GGPA().
FDR	The desired FDR level.
fdrControl	Method to control FDR. Possible values are "global" (global FDR control) and "local" (local FDR control). Default is "global".
i	Index for the first phenotype used in association mapping. See the details about how users can specify the pattern.
j	Index for the second phenotype used in association mapping. See the details about how users can specify the pattern.
...	Other parameters to be passed through to generic assoc.

Details

assoc uses the direct posterior probability approach of Newton et al. (2004) to control global FDR in association mapping.

By default (i.e., i=NULL, j=NULL), assoc implements association mapping for each phenotype. If users are interested in identifying SNPs associated with a pair of phenotypes, users can specify indices of phenotypes of interest using the arguments i and j. Note that both i and j should be either NULL or numeric.

Value

If i=NULL, j=NULL, returns a binary matrix indicating association of SNPs for each phenotype, where its rows and columns match those of input p-value matrix for function GGPA. Otherwise, returns a binary vector indicating association of SNPs for i-th and j-th phenotype pair.

Author(s)

Hang J. Kim and Dongjun Chung

References

Chung D, Kim H, and Zhao H (2016), "graph-GPA: A graphical model for prioritizing GWAS results and investigating pleiotropic architecture," 13(2): e1005388

Kim H, Yu Z, Lawson A, Zhao H, and Chung D (2017), "Improving SNP prioritization and pleiotropic architecture estimation by incorporating prior knowledge using graph-GPA."

Newton MA, Noueiry A, Sarkar D, and Ahlquist P (2004), "Detecting differential gene expression with a semiparametric hierarchical mixture method," *Biostatistics*, Vol. 5, pp. 155-176.

See Also

GGPA, GGPA.

Examples

```
# Load the included simulation data
data(simulation)

# fit GGPA model with 200 iterations and a burn-in of 200 iterations
# Note that we recommend more than 200 iterations in practice
fit <- GGPA( simulation$pmat, nMain = 200, nBurnin = 200)

# Association mapping with FDR of 0.1 and global control
head(assoc( fit, FDR=0.1, fdrControl="global" ))

# We may specift i = 1 and j = 2 if we are interested in that specific phenotype
head(assoc( fit, FDR=0.1, fdrControl="global", i=1, j=2 ))
```

GGPA *Fit graph-GPA model*

Description

Fit graph-GPA model.

Usage

```
GGPA( gwasPval, pgraph=NULL, nBurnin=10000, nMain=40000, lbPval=1e-10, verbose=1 )
```

Arguments

gwasPval	p-value matrix from GWAS data, where row and column correspond to SNP and phenotype, respectively.
pgraph	A binary matrix representing the prior phenotype graph, where its rows and columns match the columns of gwasPval.
nBurnin	Number of burn-in iterations. Default is 10000.
nMain	Number of main MCMC iterations. Default is 40000.
lbPval	Lower bound for GWAS p-value. Any GWAS p-values smaller than lbPval are set to lbPval. Default is 1e-10.
verbose	Amount of progress report during the fitting procedure. Possible values are 0 (minimal output), 1, 2, or 3 (maximal output). Default is 1.

Details

GGPA fits the graph-GPA model. It requires to provide GWAS p-value to gwasPval. If a phenotype graph is provided in pgraph, it is utilized to guide the phenotype graph estimation. Based on this GGPA fit, assoc implements association mapping and plot provides a phenotype graph.

Value

Construct GGPA class object.

Author(s)

Hang J. Kim and Dongjun Chung

References

Chung D, Kim H, and Zhao H (2016), "graph-GPA: A graphical model for prioritizing GWAS results and investigating pleiotropic architecture," 13(2): e1005388

Kim H, Yu Z, Lawson A, Zhao H, and Chung D (2018), "Improving SNP prioritization and pleiotropic architecture estimation by incorporating prior knowledge using graph-GPA," Bioinformatics, bty061.

See Also

assoc, GGPA.

Examples

```
# Load the included simulation data
data(simulation)

# fit GGPA model with 200 iterations and a burn-in of 200 iterations
# Note that we recommend more than 200 iterations in practice
fit <- GGPA( simulation$pmat, nMain = 200, nBurnin = 200)


# Association mapping with FDR of 0.1 and global control
head(assoc( fit, FDR=0.1, fdrControl="global" ))

# We may specift i = 1 and j = 2 if we are interested in that specific phenotype
head(assoc( fit, FDR=0.1, fdrControl="global", i=1, j=2 ))

# plot the GGPA model fit
plot(fit)
```

GGPA-class *Class "GGPA"*

Description

This class represents graph-GPA model fit.

Objects from the Class

Objects can be created by calls of the form `new("GGPA", ...)`.

Slots

`fit`: Object of class `"list"`, representing the MCMC draws.

`summary`: Object of class `"list"`, representing the summary statistics.

`setting`: Object of class `"list"`, representing the setting for graph-GPA model fitting.

`gwasPval`: Object of class `"matrix"`, representing the p-value matrix from GWAS data.

`pgraph`: Object of class `"matrix"`, representing the prior phenotype graph.

Methods

show `signature(object = "GGPA")`: provide brief summary of the object.

plot `signature(x = "GGPA", y = "missing", pCutoff = 0.5, betaCI = 0.95)`: plot a phenotype graph. Nodes i and j are connected if the posterior probability of E_ij > `pCutoff` and the posterior probability of beta_ij > `betaCI`.

fdr `signature(object = "GGPA", i=NULL, j=NULL)`: provide local FDR. By default (i.e., `i=NULL`, `j=NULL`), it returns a matrix of local FDR that a SNP is not associated with each phenotype (i.e., marginal FDR), where the order of columns is same as that in input GWAS data. If phenotype indices i and j are specified, a vector of corresponding local FDR is provided.

estimates `signature(object = "GGPA")`: extract parameter estimates from graph-GPA model fit.

Author(s)

Hang J. Kim, Dongjun Chung

References

Chung D, Kim H, and Zhao H (2016), "graph-GPA: A graphical model for prioritizing GWAS results and investigating pleiotropic architecture," 13(2): e1005388

Kim H, Yu Z, Lawson A, Zhao H, and Chung D (2018), "Improving SNP prioritization and pleiotropic architecture estimation by incorporating prior knowledge using graph-GPA," Bioinformatics, bty061.

See Also

GGPA.

Examples

```
showClass("GGPA")

# Load the included simulation data
data(simulation)

# fit GGPA model with 200 iterations and a burn-in of 200 iterations
# Note that we recommend more than 200 iterations in practice
fit <- GGPA( simulation$pmat, nMain = 200, nBurnin = 200)

# Plot GGPA model fit
plot(fit)

head(fdr( fit ))
head(fdr( fit, i=1, j=2 ))
str(estimates( fit ))
```

simulation *Simulation dataa for graph-GPA*

Description

This is an simulation dataset.

Usage

```
data(simulation)
```

Format

`simulation` list object containing simulation data (element Y) and its simulation setting (the remaining elements).

Author(s)

Hang J. Kim, Dongjun Chung

References

Chung D, Kim H, and Zhao H (2016), "graph-GPA: A graphical model for prioritizing GWAS results and investigating pleiotropic architecture," 13(2): e1005388

Kim H, Yu Z, Lawson A, Zhao H, and Chung D (2017), "Improving SNP prioritization and pleiotropic architecture estimation by incorporating prior knowledge using graph-GPA."

Examples

```
# The simulation data set is included with the GGPA package
data(simulation)
head(t(simulation$pmat))
```

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