## Package 'hJAM'

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Title Hierarchical Joint Analysis of Marginal Summary Statistics

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Author Lai Jiang <jian848@usc.edu>

Maintainer Lai Jiang <jian848@usc.edu>

**Description** Provides functions to implement a hierarchical approach which is designed to perform joint analysis of summary statistics using the framework of Mendelian Randomization or transcriptome analysis. Reference: Lai Jiang, Shujing Xu, Nicholas Mancuso, Paul J. Newcombe, David V. Conti (2020). ``A Hierarchical Approach Using Marginal Summary Statistics for Multiple Intermediates in a Mendelian Randomization or Transcriptome Analysis." <br/><br/>doi:10.1101/2020.02.03.924241>.

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LazyData true

RoxygenNote 6.1.1

Suggests knitr, rmarkdown

VignetteBuilder knitr

URL https://github.com/lailylajiang/hJAM

BugReports https://github.com/lailylajiang/hJAM/issues

Imports ggplot2, ggpubr, dplyr, reshape2

NeedsCompilation no

**Repository** CRAN

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## **R** topics documented:

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## conditional\_A

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betas.Gy

Example beta list of hJAM

### Description

Example beta list of hJAM

## Usage

betas.Gy

## Format

The betas.Gy is the beta vector in the hJAM model: the association estimates between 210 SNPs and myocardial infarction. The summary data was collected from UK Biobank (n=459,324).

## References

Sudlow C, Gallacher J, Allen N, et al. UK biobank: an open access resource for identifying the causes of a wide range of complex diseases of middle and old age. PLoS Med 2015; 12: e1001779.

conditional\_A Example conditional A matrix of hJAM

## Description

Example conditional A matrix of hJAM

#### Usage

conditional\_A

#### get\_cond\_A

#### Format

The conditional\_A is the conditional estimates alpha matrix in the hJAM model: the association estimates between 210 SNPs and body mass index (BMI) and type 2 diabetes (T2D). The summary data was collected from GIANT consortium (n=339,224) and DIAGRAM+GERA+UKB (n=659316) for BMI and T2D, respectively. We converted it from marginal\_A, using get\_cond\_A function in hJAM package.

#### References

1. Locke AE, Kahali B, Berndt SI, et al. Genetic studies of body mass index yield new insights for obesity biology. Nature 2015; 518: 197-206. 2. Xue A, Wu Y, Zhu Z, et al. Genome-wide association analyses identify 143 risk variants and putative regulatory mechanisms for type 2 diabetes. Nat Commun 2018; 9: 2941.

get\_cond\_A

Compute conditional Z matrix

#### Description

The get\_cond\_A function is to get the conditional A matrix by using marginal A matrix

#### Usage

```
get_cond_A(marginal_A, Gl, N.Gx, ridgeTerm = FALSE)
```

#### Arguments

| marginal_A | the marginal effects of SNPs on the exposures (Gx).  |
|------------|--|
| Gl         | the reference panel (Gl), such as 1000 Genome  |
| N.Gx       | the sample size of each Gx. It can be a scalar or a vector. If there are multiple X's from different Gx, it should be a vector including the sample size of each Gx. If all alphas are from the same Gx, it could be a scalar. |
| ridgeTerm  | ridgeTerm = TRUE when the matrix L is singular. Matrix L is obtained from the cholesky decomposition of G0'G0. Default as FALSE.   |

## Value

A matrix with conditional estimates which are converted from marginal estimates using the JAM model.

#### Author(s)

Lai Jiang

### Examples

```
data(G1)
data(betas.Gy)
data(marginal_A)
get_cond_A(marginal_A = marginal_A, G1 = G1, N.Gx = c(339224, 659316), ridgeTerm = TRUE)
```

get\_cond\_alpha Compute conditional alphas

#### Description

The get\_cond\_alpha function is to compute the conditional alpha vector for each X If only one X in the model, please use get\_cond\_alpha instead of get\_cond\_A A sub-step in the get\_cond\_A function

#### Usage

```
get_cond_alpha(alphas, Gl, N.Gx, ridgeTerm = FALSE)
```

### Arguments

| alphas    | the marginal effects of SNPs on one exposure (Gx).                            |
|-----------|---|
| Gl        | the reference panel (Gl), such as 1000 Genome                                 |
| N.Gx      | the sample size of the Gx. It can be a scalar.                                |
| ridgeTerm | ridgeTerm = TRUE when the matrix L is singular. Matrix L is obtained from the |
|           | cholesky decomposition of G0'G0. Default as FALSE                             |

#### Value

A vector with conditional estimates which are converted from marginal estimates using the JAM model.

#### Author(s)

Lai Jiang

## References

Lai Jiang, Shujing Xu, Nicholas Mancuso, Paul J. Newcombe, David V. Conti (2020). A Hierarchical Approach Using Marginal Summary Statistics for Multiple Intermediates in a Mendelian Randomization or Transcriptome Analysis. *bioRxiv* https://doi.org/10.1101/2020.02.03. 924241.

#### Examples

```
data(G1)
data(betas.Gy)
data(marginal_A)
get_cond_alpha(alphas = marginal_A[, 1], G1 = G1, N.Gx = 339224, ridgeTerm = TRUE)
```

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## Description

The real data example from hJAM paper

## Usage

Gl

## Format

The Gl object is a data matrix with 2467 individual of 210 SNPs from 1000 Genome project.

## References

Consortium GP. A global reference for human genetic variation. Nature 2015; 526: 68.

hJAM\_egger Fit hJAM with Egger regression

## Description

The hJAM\_egger function is to get the results from the hJAM model with Egger regression. It is for detecting potential pleiotropy

#### Usage

```
hJAM_egger(betas.Gy, N.Gy, Gl, A, ridgeTerm = FALSE)
```

## Arguments

| betas.Gy  | The betas in the paper: the marginal effects of SNPs on the phenotype (Gy)   |
|-----------|--|
| N.Gy      | The sample size of Gy  |
| Gl        | The reference panel (Gl), such as 1000 Genome  |
| A         | The A matrix in the paper: the marginal/conditional effects of SNPs on the exposures (Gx) $% \left( G_{x}\right) =0$             |
| ridgeTerm | ridgeTerm = TRUE when the matrix L is singular. Matrix L is obtained from the cholesky decomposition of G0'G0. Default as FALSE. |

Gl

#### Value

An object of the hJAM with egger regression results.

- **Exposure** The intermediates, such as the modifiable risk factors in Mendelian Randomization and gene expression in transcriptome analysis.
- numSNP The number of SNPs that the user use in the instrument set.
- Estimate The conditional estimates of the associations between intermediates and the outcome.
- **StdErr** The standard error of the conditional estimates of the associations between intermediates and the outcome.

Lower.CI The lower bound of the 95% confidence interval of the estimates.

Upper.CI The upper bound of the 95% confidence interval of the estimates.

**Pvalue** The p value of the estimates with a type-I error equals 0.05.

Est.Int The intercept of the regression of intermediates on the outcome.

StdErr.Int The standard error of the intercept of the regression of intermediates on the outcome.

Lower.CI.Int The lower bound of the 95% confidence interval of the intercept.

Upper.CI.Int The upper bound of the 95% confidence interval of the intercept.

**Pvalue.Int** The p value of the intercept with a type-I error equals 0.05.

An object of hJAM with egger regression results.

#### Author(s)

Lai Jiang

#### References

Lai Jiang, Shujing Xu, Nicholas Mancuso, Paul J. Newcombe, David V. Conti (2020). A Hierarchical Approach Using Marginal Summary Statistics for Multiple Intermediates in a Mendelian Randomization or Transcriptome Analysis. *bioRxiv* https://doi.org/10.1101/2020.02.03. 924241.

#### Examples

```
data(G1)
data(betas.Gy)
data(conditional_A)
hJAM_egger(betas.Gy = betas.Gy, G1 = G1, N.Gy = 459324, A = conditional_A, ridgeTerm = TRUE)
```

hJAM\_lnreg

#### Description

The hJAM function is to get the results from the hJAM model using input data

#### Usage

hJAM\_lnreg(betas.Gy, N.Gy, Gl, A, ridgeTerm = FALSE)

#### Arguments

| betas.Gy  | The betas in the paper: the marginal effects of SNPs on the phenotype (Gy)   |
|-----------|--|
| N.Gy      | The sample size of Gy  |
| Gl        | The reference panel (Gl), such as 1000 Genome  |
| A         | The A matrix in the paper: the marginal/conditional effects of SNPs on the exposures (Gx)  |
| ridgeTerm | ridgeTerm = TRUE when the matrix L is singular. Matrix L is obtained from the cholesky decomposition of G0'G0. Default as FALSE. |

#### Value

An object of the hJAM with linear regression results.

- **Exposure** The intermediates, such as the modifiable risk factors in Mendelian Randomization and gene expression in transcriptome analysis.
- numSNP The number of SNPs that the user use in the instrument set.
- Estimate The conditional estimates of the associations between intermediates and the outcome.
- **StdErr** The standard error of the conditional estimates of the associations between intermediates and the outcome.
- Lower.CI The lower bound of the 95% confidence interval of the estimates.

**Upper.CI** The upper bound of the 95% confidence interval of the estimates.

**Pvalue** The p value of the estimates with a type-I error equals 0.05.

#### Author(s)

Lai Jiang

#### References

Lai Jiang, Shujing Xu, Nicholas Mancuso, Paul J. Newcombe, David V. Conti (2020). A Hierarchical Approach Using Marginal Summary Statistics for Multiple Intermediates in a Mendelian Randomization or Transcriptome Analysis. *bioRxiv* https://doi.org/10.1101/2020.02.03. 924241.

### Examples

```
data(G1)
data(betas.Gy)
data(conditional_A)
hJAM_lnreg(betas.Gy = betas.Gy, Gl = Gl, N.Gy = 459324, A = conditional_A, ridgeTerm = TRUE)
```

marginal\_A

Example marginal A matrix of hJAM

#### Description

Example marginal A matrix of hJAM

#### Usage

marginal\_A

#### Format

The marginal\_A is the marginal estimates alpha matrix in the hJAM model: the association estimates between 210 SNPs and body mass index (BMI) and type 2 diabetes (T2D). The summary data was collected from GIANT consortium (n=339,224) and DIAGRAM+GERA+UKB (n=659316) for BMI and T2D, respectively.

## References

1. Locke AE, Kahali B, Berndt SI, et al. Genetic studies of body mass index yield new insights for obesity biology. Nature 2015; 518: 197-206. 2. Xue A, Wu Y, Zhu Z, et al. Genome-wide association analyses identify 143 risk variants and putative regulatory mechanisms for type 2 diabetes. Nat Commun 2018; 9: 2941.

output.format Keep the output as three digits

#### Description

Keep the output as three digits

## Usage

output.format(x, ...)

#### Arguments

| Х | input                            |
|---|----------------------------------|
|   | other options you want to put in |

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## print.hJAM\_egger

## Author(s)

Lai Jiang

print.hJAM\_egger Print out for hJAM\_egger

## Description

Print out for hJAM\_egger

## Usage

## S3 method for class 'hJAM\_egger'
print(x, ...)

## Arguments

| Х | input                            |
|---|----------------------------------|
|   | other options you want to put in |

## Author(s)

Lai Jiang

print.hJAM\_lnreg Print out for hJAM\_lnreg

## Description

Print out for hJAM\_lnreg

#### Usage

## S3 method for class 'hJAM\_lnreg'
print(x, ...)

## Arguments

| х | input                            |
|---|----------------------------------|
|   | other options you want to put in |

## Author(s)

Lai Jiang

SNPs\_heatmap

## Description

To generate the heatmap of all the SNPs that the user use in the analysis

#### Usage

SNPs\_heatmap(G1)

### Arguments

Gl

The reference panel (Gl) of the SNPs that the user use in the analysis, such as 1000 Genome

## Author(s)

Lai Jiang

## Examples

```
data(Gl)
t = SNPs_heatmap(Gl = Gl)
t
```

SNPs\_info

Example SNPs' information of hJAM

#### Description

Example SNPs' information of hJAM

#### Usage

SNPs\_info

#### Format

The SNPs\_info is the information of the 210 SNPs that we used in this data example. It includes three columns: the rsID, major allele, and minor allele frequency of each SNP. The minor allele frequencies were calculated in the 503 European-ancestry subjects in 1000 Genome project.

#### References

Consortium GP. A global reference for human genetic variation. Nature 2015; 526: 68.

SNPs\_scatter\_plot Scatter plot for all the SNPs used in the analysis

## Description

To generate the scatter plot of all the SNPs that the user use in the analysis

## Usage

```
SNPs_scatter_plot(A, betas.Gy, num_X)
```

## Arguments

| A        | The effects of SNPs on the exposures (Gx).                                 |
|----------|--|
| betas.Gy | The betas in the paper: the marginal effects of SNPs on the phenotype (Gy) |
| num_X    | The number of intermediates in the research question.                      |

### Value

A set of scatter plots with x-axis being the conditional  $\alpha$  estimates for each intermediate and y-axis being the  $\beta$  estimates.

## Author(s)

Lai Jiang

## Examples

```
data(conditional_A)
data(betas.Gy)
t = SNPs_scatter_plot(A = conditional_A, betas.Gy = betas.Gy, num_X = 2)
t
```

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