Package 'SIPI'

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Description Testing single nucleotide polymorphism (SNP) interactions is considered as a key for overcoming bottlenecks of genetic association studies. SNP Interaction Pattern Identifier (SIPI) evaluates SNP-SNP interactions associated with a binary or continuous outcome. The primary strengths of SIPI are (1) taking non-hierarchical models, reverse coding and inheritance modes (dominant, recessive and additive mode) into consideration and (2) using the Bayesian information criterion (BIC) to search for a best interaction pattern. For each SNP pair, the SIPI evaluates 45 interaction models. The best interaction pattern is the one with the lowest BIC value.

Reference (1) Lin HY, Chen DT, Huang PY, Liu YH, Ochoa A, Zabaleta J, Mercante DE, Fang Z, Sellers TA, Pow-Sang JM, Cheng CH, Eeles R, Easton D, Kote-Jarai Z, Amin Al Olama A, Benlloch S, Muir K, Giles GG, Wiklund F, Gronberg H, Haiman CA, Schleutker J, Nordestgaard BG, Travis RC, Hamdy F, Pashayan N, Khaw KT, Stanford JL, Blot WJ, Thibodeau SN, Maier C, Kibel AS, Cybulski C, Cannon-Albright L, Brenner H, Kaneva R, Batra J, Teixeira MR, Pandha H, Lu YJ, Consortium P, Park JY. SNP interaction pattern identifier (SIPI): an intensive search for SNP-SNP interaction patterns. Bioinformatics. 2017;33(6):822-33. PubMed PMID: 28039167. (2) Lin HY, Huang PY, Chen DT, Tung HY, Sellers TA, Pow-Sang J, Eeles R, Easton D, Kote-Jarai Z, Amin Al Olama A, Benlloch S, Muir K, Giles GG, Wiklund F, Gronberg H, Haiman CA, Schleutker J, Nordestgaard BG, Travis RC, Hamdy F, Neal DE, Pashayan N, Khaw KT, Stanford JL, Blot WJ, Thibodeau SN, Maier C, Kibel AS, Cybulski C, Cannon-Albright L, Brenner H, Kaneva R, Batra J, Teixeira MR, Pandha H, Lu YJ, consortium P, Park JY. AA9int: SNP Interaction Pattern Search Using Non-Hierarchical Additive Model Set. Bioinformatics. 2018. doi: 10.1093/bioinformatics/bty461. PubMed PMID: 29878078.

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AA9int

R topics documented:

AA9int		AA9int (Additive-Additive 9 interaction models): Detect SNP-SNP interactions through testing the 9 models						
Index						16		
	SIPI					12		
	plot3by3					10		
	parSIPI					8		
	1							
	MAFinfo					6		
	Grid3by3					4		
	AA9int					2		

Description

AA9int (Additive-Additive 9 interaction models), a mini version of SIPI, evaluates SNP-SNP interactions associated with a binary or continuous outcome through testing nine interaction models. AA9int treats both SNPs as an additive mode and takes reverse coding and non-hierarchical models into consideration. The best interaction pattern is the one with the lowest value of the Bayesian information criterion (BIC). The details of the nine models/patterns are listed in the reference.

Usage

```
AA9int(Outcome, SNPdata, PairInfo, X=NULL, categXNames=NULL, TestType="WaldTest", ModelType="binomial", OR=FALSE)
```

Arguments

Outcome	Binary (1: event of interest; 0: reference) or continuous variable.
SNPdata	SNP data: All SNP variables should have a character variable attribute and contain two of four letters (C, T, A, and G). No other letters or numbers should be used. An invalid character or blank field are considered to be missing values.
PairInfo	3 types of PairInfo:
	(1) 2d-vector: names of the given SNP pair for one-pair analyses. ex: c('SNP1', 'SNP2')
	(2) 2d-matrix or 2d-dataframe: names for candidate SNP pair.
	(3) "all": for pairwise analyses.
X	Covariate(s) to be adjusted in the model (for missing values, keep the field blank), NULL=without covariate. Default is NULL.
categXNames	The variable names of categorical variables, NULL=without categorical covariates. Default is NULL.
TestType	Specify the statistical test type: "WaldTest" = the Wald test; "LRT" = the likelihood ratio test. Default is "WaldTest".
ModelType	Model type: "binomial"=logistic regression; "gaussian"=linear regression. Default is "binomial".
OR	If TRUE print the odds ratios, 95% confidence intervals and corresponding p-values.

AA9int 3

Value

Returns a list with the following attributes:

selectedModel The results of the best model with the lowest BIC value among the 9 models.

res9Models

For one-pair analyses only: Detailed results with all 9 models sorted by BIC (lowest first). Output variables: Var1: SNP 1; Var2: SNP 2 (the pattern/model labels are based on this order); Model: interaction model/pattern; Wald_Chisq: the Wald chi-square value of the interaction term; Wald_p: the Wald p-value of the interaction term; LRT_Chisq: the chi-square value of likelihood ratio test (LRT) for the interaction term; LRT_p: the LRT p-value of the interaction term; BIC: the Bayesian information criterion. The model with the lowest BIC value

is preferred.

OR For the "all" pairs analyses, only results of the best model will show. The group

coding, please see suppl. Figures 1-3 in the end of this manual.

Author(s)

Hui-Yi Lin and Po-Yu Huang

References

Lin HY, Chen DT, Huang PY, Liu YH, Ochoa A, Zabaleta J, Mercante DE, Fang Z, Sellers TA, Pow-Sang JM, Cheng CH, Eeles R, Easton D, Kote-Jarai Z, Amin Al Olama A, Benlloch S, Muir K, Giles GG, Wiklund F, Gronberg H, Haiman CA, Schleutker J, Nordestgaard BG, Travis RC, Hamdy F, Pashayan N, Khaw KT, Stanford JL, Blot WJ, Thibodeau SN, Maier C, Kibel AS, Cybulski C, Cannon-Albright L, Brenner H, Kaneva R, Batra J, Teixeira MR, Pandha H, Lu YJ, Consortium P, Park JY. SNP interaction pattern identifier (SIPI): an intensive search for SNP-SNP interaction patterns. Bioinformatics. 2017;33(6):822-33. PubMed PMID: 28039167.

Lin HY, Huang PY, Chen DT, Tung HY, Sellers TA, Pow-Sang J, Eeles R, Easton D, Kote-Jarai Z, Amin Al Olama A, Benlloch S, Muir K, Giles GG, Wiklund F, Gronberg H, Haiman CA, Schleutker J, Nordestgaard BG, Travis RC, Hamdy F, Neal DE, Pashayan N, Khaw KT, Stanford JL, Blot WJ, Thibodeau SN, Maier C, Kibel AS, Cybulski C, Cannon-Albright L, Brenner H, Kaneva R, Batra J, Teixeira MR, Pandha H, Lu YJ, consortium P, Park JY. AA9int: SNP Interaction Pattern Search Using Non-Hierarchical Additive Model Set. Bioinformatics. 2018. doi: 10.1093/bioinformatics/bty461. PubMed PMID: 29878078.

See Also

```
parAA9int, SIPI, parSIPI
```

4 Grid3by3

```
res_snp_2v8 = AA9int(simData$D, SNPdata, c('SNP2', 'SNP8'))
## print out the best model
res_snp_2v8$selectedModel
## list of 9 models for one SNP pair
res_snp_2v8$res9Models
### run AA9int for a list of multiple SNP pairs <Wald test>
############################
## For 5 SNP pairs(1 v.s. 6, 2 v.s. 7, 3 v.s. 8, 4 v.s. 9, and 5 v.s. 10)
pairMatrix = c("SNP1", "SNP2", "SNP3", "SNP4", "SNP5", "SNP6", "SNP7", "SNP8", "SNP9", "SNP9")
pairMatrix = matrix(pairMatrix,5)
pairMatrix = as.data.frame(pairMatrix)
res_snp = AA9int(simData$D, SNPdata, pairMatrix)
############################
### run AA9int pairwise analyses <Wald test>
res_all = AA9int(simData$D, SNPdata, "all")
############################
### run AA9int pairwise analyses <Wald test> adjusted for covariates
### [age(numeric), gender (binary), and group (categorical)]
X1 = simData[,c('age','gender','group')]
res_all_cov = AA9int(simData$D, SNPdata, "all", X1, c("gender", "group"))
############################
## export ORs and p-values
##########################
try = AA9int(simData$D,SNPdata,"all", OR=TRUE)
write_OR_csv(try$OR, 'C:/OR.csv')
write\_OR\_csv(try\$selectedModel, 'C:/p.csv')
```

Grid3by3

Description

Outcome proportions by the 3-by-3 genotype combinations of a give SNP pair.

Usage

```
Grid3by3(Outcome, SNPdata, PairInfo)
```

Arguments

Outcome Binary outcome variable name: a binary variable with "1" as the event of inter-

est and "0" as the reference.

SNP data. All SNP variables should have a character variable attribute and con-

tain two of four letters (C, T, A, and G). No other letters or numbers should be used. An invalid character or blank field are considered to be missing values.

PairInfo c('SNP1', 'SNP2'): names of the given SNP pairs for one-pair analyses

Value

maj_min Major and minor allele

table3by3 Present outcome proportions by genotype combinations

table3by3Freq Sample size by genotype combinations

Author(s)

Hui-Yi Lin and Po-Yu Huang

See Also

```
plot3by3
```

6 parAA9int

MAFinfo

SNP minor allele frequency (MAF)

Description

Obtain minor allele frequency (MAF) and major and minor allele.

Usage

```
MAFinfo(SNPdata)
```

Arguments

SNP data

Value

maj/min major and minor allele
MAF minor allele frequency

Missing (%) missing vales

Author(s)

Hui-Yi Lin and Po-Yu Huang

References

Gonzalez JR, Armengol L, Sole X, Guino E, Mercader JM, Estivill X, Moreno V. SNPassoc: an R package to perform whole genome association studies. Bioinformatics, 2007;23(5):654-5.

Examples

```
data(simData)
SNPdata = simData[,3:12]
MAFinfo(SNPdata)
```

parAA9int

Parallel computing for AA9int

Description

parAA9int is a parallel computing version of AA9int. This function can decrease computing time, which is useful for large-scale data.

Usage

parAA9int 7

Arguments

Outcome Binary (1: event of interest; 0: reference) or continuous variable.

SNP data: All SNP variables should have a character variable attribute and con-

tain two of four letters (C, T, A, and G). No other letters or numbers should be used. No other letters or numbers should be used. Invalid character or field

blank is considered to be missing values.

PairInfo 3 types of PairInfo:

(1) 2d-vector: names of the given SNP pair for one-pair analyses. ex: c('SNP1', 'SNP2')

(2) 2d-matrix or 2d-dataframe: names for candidate SNP pair.

(3) "all": for pairwise analyses.

X Covariate(s) to be adjusted in the model (for missing values, keep the field

blank), NULL=without covariate. Default is NULL.

categXNames The variable names of categorical variables, NULL=without categorical covari-

ates. Default is NULL.

TestType Specify the statistical test type: "WaldTest" = the Wald test; "LRT" = the likeli-

hood ratio test. Default is "WaldTest".

Model Type Model type: "binomial"=logistic regression; "gaussian"=linear regression.

Default is "binomial".

core_ratio The ratio of total cores for parallel computing. Default is 0.9.

OR If TRUE print the odds ratios, 95% confidence intervals and corresponding p-

values.

Value

Returns a data frame of the results of the best model with the lowest BIC value among the 9 models. For the "all" pairs analyses, only OR results of the best model will show. The group coding, please see suppl. Figures 1-3 in the end of this manual.)

Author(s)

Hui-Yi Lin and Po-Yu Huang

References

Lin HY, Chen DT, Huang PY, Liu YH, Ochoa A, Zabaleta J, Mercante DE, Fang Z, Sellers TA, Pow-Sang JM, Cheng CH, Eeles R, Easton D, Kote-Jarai Z, Amin Al Olama A, Benlloch S, Muir K, Giles GG, Wiklund F, Gronberg H, Haiman CA, Schleutker J, Nordestgaard BG, Travis RC, Hamdy F, Pashayan N, Khaw KT, Stanford JL, Blot WJ, Thibodeau SN, Maier C, Kibel AS, Cybulski C, Cannon-Albright L, Brenner H, Kaneva R, Batra J, Teixeira MR, Pandha H, Lu YJ, Consortium P, Park JY. SNP interaction pattern identifier (SIPI): an intensive search for SNP-SNP interaction patterns. Bioinformatics. 2017;33(6):822-33. PubMed PMID: 28039167.

Lin HY, Huang PY, Chen DT, Tung HY, Sellers TA, Pow-Sang J, Eeles R, Easton D, Kote-Jarai Z, Amin Al Olama A, Benlloch S, Muir K, Giles GG, Wiklund F, Gronberg H, Haiman CA, Schleutker J, Nordestgaard BG, Travis RC, Hamdy F, Neal DE, Pashayan N, Khaw KT, Stanford JL, Blot WJ, Thibodeau SN, Maier C, Kibel AS, Cybulski C, Cannon-Albright L, Brenner H, Kaneva R, Batra J, Teixeira MR, Pandha H, Lu YJ, consortium P, Park JY. AA9int: SNP Interaction Pattern Search Using Non-Hierarchical Additive Model Set. Bioinformatics. 2018. doi: 10.1093/bioinformatics/bty461. PubMed PMID: 29878078.

8 parSIPI

See Also

```
AA9int, SIPI, and parSIPI
```

Examples

```
##load data
data(simData)
#### define SNP data
SNPdata = simData[,3:12]
##########################
### run parAA9int pairwise analyses <Wald test>
############################
res_all = parAA9int(simData$D, SNPdata, "all")
###run parAA9int pairwise analyses <Wald test> adjusted for covariates using SIPI with parallel computing
###[age(numeric), gender (binary), and group (categorical)]
###########################
X1 = simData[,c('age','gender','group')]
res_all_X = parAA9int(simData$D,SNPdata,"all",X1,c("gender","group"))
## export ORs and p-values
try = parAA9int(simData$D,SNPdata,"all", OR=TRUE)
write_OR_csv(try$OR, 'C:/OR.csv')
write_OR_csv(try$selectedModel, 'C:/p.csv')
```

parSIPI

Parallel computing for SIPI

Description

parSIPI is a parallel computing version of SIPI. This function can decrease computing time, which is useful for large-scale data.

Usage

Arguments

Outcome

Binary (1: event of interest; 0: reference) or continuous variable.

parSIPI 9

SNPdata SNP data: All SNP variables should have a character variable attribute and contain two of four letters (C, T, A, and G). No other letters or numbers should be used. Invalid character or field blank is considered to be missing values. PairInfo 3 types of PairInfo: (1) 2d-vector: names of the given SNP pair for one-pair analyses. ex: c('SNP1', 'SNP2') (2) 2d-matrix or 2d-dataframe: names for candidate SNP pair. (3) "all": for pairwise analyses. Χ Covariate(s) to be adjusted in the model (for missing values, keep the field blank), NULL=without covariate. Default is NULL. The variable names of categorical variables, NULL=without categorical covaricategXNames ates. Default is NULL. Specify the statistical test type: "WaldTest" = the Wald test; "LRT" = the likeli-TestType hood ratio test. Default is "WaldTest". Model type: "binomial"=logistic regression; "gaussian"=linear regression. ModelType Default is "binomial". core_ratio The ratio of total cores for parallel computing. Default is 0.9.

Value

ΩR

Returns a data frame of the results of the best model with the lowest BIC value among the 45 models for each SNP pair. For the "all" pairs analyses, only OR results of the best model will show. The group coding, please see suppl. Figures 1-3 in the end of this manual.)

If TRUE print the odds ratios, 95% confidence intervals and corresponding p-

Author(s)

Hui-Yi Lin and Po-Yu Huang

values.

References

Lin HY, Chen DT, Huang PY, Liu YH, Ochoa A, Zabaleta J, Mercante DE, Fang Z, Sellers TA, Pow-Sang JM, Cheng CH, Eeles R, Easton D, Kote-Jarai Z, Amin Al Olama A, Benlloch S, Muir K, Giles GG, Wiklund F, Gronberg H, Haiman CA, Schleutker J, Nordestgaard BG, Travis RC, Hamdy F, Pashayan N, Khaw KT, Stanford JL, Blot WJ, Thibodeau SN, Maier C, Kibel AS, Cybulski C, Cannon-Albright L, Brenner H, Kaneva R, Batra J, Teixeira MR, Pandha H, Lu YJ, Consortium P, Park JY. SNP interaction pattern identifier (SIPI): an intensive search for SNP-SNP interaction patterns. Bioinformatics. 2017;33(6):822-33. PubMed PMID: 28039167.

See Also

SIPI

```
##load data
data(simData)

#### define SNP data
SNPdata = simData[,3:12]
```

10 plot3by3

```
#############################
### run parSIPI pairwise analyses <Wald test>
##########################
res_all = parSIPI(simData$D, SNPdata, "all")
###run parSIPI pairwise analyses <Wald test> adjusted for covariates using SIPI with parallel computing
###[age(numeric), gender (binary), and group (categorical)]
############################
X1 = simData[,c('age','gender','group')]
res_all_X = parSIPI(simData$D,SNPdata,"all",X1,c("gender","group"))
############################
## export ORs and p-values
try = parSIPI(simData$D, SNPdata, "all", OR=TRUE)
write_OR_csv(try$OR, 'C:/OR.csv')
write_OR_csv(try$selectedModel, 'C:/p.csv')
```

plot3by3

Heatmap plot of outcome proportions by genotype combinations

Description

Create a heatmap plot of sample sizes and outcome proportions by the 3-by-3 genotype combinations for a given SNP pair.

Usage

Arguments

x	List object output from function Grid3by3.
SNP_info	Put SNP information (SNP name and major/minor allele) on plot. Default is TRUE.
outcome	Include outcome proportions in each cell of plot. Default is TRUE. When no observations is in the given cell, 'NaN' will be shown. If there is no observation and outcome=FALSE, a warning will be shown.
freq	Include frequency in each cell. Default is TRUE.
legend	Include legend. Default is TRUE.
monochrome	Output monochrome plot. Default is FALSE

A character string specifying the colour gradient scale type. "fixed" will lend color to heatmap with fixed color gradient scale from 0 to 1. "sliding" will lend

color to heatmap with fixed color gradient scale from 0 to 1, "sliding" will lend color to heatmap with sliding gradient scale between minimum and maximum

outcome proportion. Default is "fixed".

plot3by3 11

axis_fs	Axis font size. Adjusted both axis title font size and axis label font size. Number greater than 1 will enlarge the font size, less than 1 will reduce the font size. Default is 1.
outcome_fs	Outcome font size. Number greater than 1 will enlarge the font size, less than 1 will reduce the font size. Default is 1.
freq_fs	Frequency font sizr. Number greater than 1 will enlarge the font size, less than 1 will reduce the font size. Default is 1.
lgd_fs	Legend font size. Number greater than 1 will enlarge the font size, less than 1 will reduce the font size. Default is 1.
marginal	Show details of the genotypes for a SNP.

Details

This function creates a heatmap plot based on the output of Grid3by3, which generates outcome proportions by genotype combinations of a given SNP pair.

Value

A heatmap plot of outcome proportions, which is a ggplot object.

Author(s)

Hui-Yi Lin and Heng-Yuan Tung

References

H. Wickham. ggplot2: Elegant Graphics for Data Analysis. Springer-Verlag New York, 2009.

See Also

Grid3by3

```
##load data
data(simData)

#### define SNP data
SNPdata = simData[,3:12]

#############################

### run plot3by3
######################

x = Grid3by3(simData$D, SNPdata, c('SNP1', 'SNP2'))
plot3by3(x, SNP_info = F, outcome = F, legend = T, scale = "fixed", monochrome = T, lgd_fs = 1.2, marginal=T)

x = Grid3by3(simData$D, SNPdata, c('SNP4', 'SNP6'))
plot3by3(x, scale = "sliding", axis_fs = 1.2, outcome_fs = 0.9, freq_fs = 0.8, marginal = T)
plot3by3(x, scale = "sliding", freq = F, axis_fs = 1.2, outcome_fs = 0.9, marginal = T)
```

simData An example data set

Description

simData is an example dataset with one binary outcome variable (D), the 10 SNPs, and three covariates [age (numeric), gender (binary), and group (categorical)].

Usage

```
data(simData)
```

Format

A data frame with 1000 observations on the following 12 variables.

id A numeric vector for identification.

D A numeric vector for a binary outcome with "1" as the event of interest and "0" as the reference.

SNP1 to SNP10 A factor has three genotypes, which are composed of two of the four letters (C, T, A, and G), such as CC, TC and TT.

```
age numeric age.
```

```
gender 0: female, 1: male.
group 1: Group 1; 2: Group 2; and 3: Group 3.
```

Examples

```
data(simData)
```

SIPI

SNP Interaction Pattern Identifier (SIPI): Detect SNP-SNP interactions through testing the 45 models

Description

SNP Interaction Pattern Identfier (SIPI) evaluates SNP-SNP interactions associated with a binary or continuous outcome. The primary strengths of SIPI are (1) taking non-hierarchical models, reverse coding and inheritance modes (dominant, recessive and additive mode) into consideration and (2) using BIC to search for a best interaction pattern. For each SNP pair, the SIPI evaluates 45 interaction models. The best interaction pattern is the one with the lowest value of the Bayesian information criterion (BIC). The details of the 45 models/patterns are listed in the SIPI published paper.

Usage

```
SIPI(Outcome, SNPdata, PairInfo,
    X=NULL, categXNames=NULL,
    TestType="WaldTest", ModelType="binomial", OR=FALSE)
```

Arguments

Outcome Binary (1: event of interest; 0: reference) or continuous variable.

SNP data: All SNP variables should have a character variable attribute and con-

tain two of four letters (C, T, A, and G). No other letters or numbers should be

used. Invalid character or field blank is considered to be missing values.

PairInfo 3 types of PairInfo:

(1) 2d-vector: names of the given SNP pair for one-pair analyses. ex: c('SNP1', 'SNP2')

(2) 2d-matrix or 2d-dataframe: names for candidate SNP pair.

(3) "all": for pairwise analyses.

X Covariate(s) to be adjusted in the model (for missing values, keep the field

blank), NULL=without covariate. Default is NULL.

categXNames The variable names of categorical variables, NULL=without categorical covari-

ates. Default is NULL.

TestType Specify the statistical test type: "WaldTest" = the Wald test; "LRT" = the likeli-

hood ratio test. Default is "WaldTest".

Model type: "binomial"=logistic regression; "gaussian"=linear regression.

Default is "binomial".

OR If TRUE print the odds ratios, 95% confidence intervals and corresponding p-

values.

Value

Returns a list with the following attributes:

selectedModel The results of the best model with the lowest BIC value among the 45 models.

res45Models For one-pair analyses only: Detailed results with all 45 models sorted by BIC

(lowest first). Output variables: Var1: SNP 1; Var2: SNP 2 (the pattern/model labels are based on this order); Model: interaction model/pattern; Wald_Chisq: the Wald chi-square value of the interaction term; Wald_p: the Wald p-value of the interaction term; LRT_Chisq: the chi-square value of likelihood ratio test (LRT) for the interaction term; LRT_p: the LRT p-value of the interaction term; BIC: the Bayesian information criterion. The model with the lowest BIC value

is preferred.

OR For the "all" pairs analyses, only results of the best model will show. The group

coding, please see suppl. Figures 1-3 in the end of this manual.

Author(s)

Hui-Yi Lin and Po-Yu Huang

References

Lin HY, Chen DT, Huang PY, Liu YH, Ochoa A, Zabaleta J, Mercante DE, Fang Z, Sellers TA, Pow-Sang JM, Cheng CH, Eeles R, Easton D, Kote-Jarai Z, Amin Al Olama A, Benlloch S, Muir K, Giles GG, Wiklund F, Gronberg H, Haiman CA, Schleutker J, Nordestgaard BG, Travis RC, Hamdy F, Pashayan N, Khaw KT, Stanford JL, Blot WJ, Thibodeau SN, Maier C, Kibel AS, Cybulski C, Cannon-Albright L, Brenner H, Kaneva R, Batra J, Teixeira MR, Pandha H, Lu YJ, Consortium P, Park JY. SNP interaction pattern identifier (SIPI): an intensive search for SNP-SNP interaction patterns. Bioinformatics. 2017;33(6):822-33. PubMed PMID: 28039167.

See Also

```
parSIPI, AA9int, and parAAint
```

```
##load data
data(simData)
#### define SNP data
SNPdata = simData[,3:12]
###########################
### run SIPI one-pair analyses <Wald test>
############################
## For a SNP pair of SNP2 and SNP8
res_snp_2v8 = SIPI(simData$D,SNPdata,c('SNP2','SNP8'))
## print out the best model
res_snp_2v8$selectedModel
## list of 45 models for one SNP pair
res_snp_2v8$res45Models
#############################
### run SIPI for a list of multiple SNP pairs <Wald test>
############################
## For 5 SNP pairs(1 v.s. 6, 2 v.s. 7, 3 v.s. 8, 4 v.s. 9, and 5 v.s. 10)
pairMatrix = c("SNP1","SNP2", "SNP3", "SNP4","SNP5","SNP6","SNP7","SNP8","SNP9","SNP9","SNP10")
pairMatrix = matrix(pairMatrix,5)
pairMatrix = as.data.frame(pairMatrix)
res_snp = SIPI(simData$D, SNPdata, pairMatrix)
### run SIPI pairwise analyses <Wald test>
############################
res_all = SIPI(simData$D, SNPdata, "all")
############################
### run SIPI pairwise analyses <Wald test> adjusted for covariates
### [age(numeric), gender (binary), and group (categorical)]
##########################
X1 = simData[,c('age','gender','group')]
res_all_cov = SIPI(simData$D,SNPdata,"all",X1,c("gender","group"))
```

```
try = SIPI(simData$D,SNPdata,"all", OR=TRUE)
write_OR_csv(try$OR,'C:/OR.csv')
write_OR_csv(try$selectedModel,'C:/p.csv')
```

Index

```
*Topic datasets
simData, 12

AA9int, 2, 8, 14

ggplot, 11
Grid3by3, 4, 11

MAFinfo, 6

parAA9int, 3, 6
parAAint, 14
parSIPI, 3, 8, 8, 14
plot3by3, 5, 10

simData, 12
SIPI, 3, 8, 9, 12
```