# SLGI

# October 5, 2010

Atong

*Systematic genetic analysis with ordered arrays of yeast deletion Tong et. al. (2004).* 

# Description

Data from Tong et. al. (2004) buffering experiments using ordered arrays of yeast deletion design by Tong et. al. (2001).

## Usage

```
data(Atong)
data(tong2004raw)
```

## Format

tong2004Raw is dataframe extracted from Table S1 of Tong et al. (2004) online supporting material. We added an extra column, queryGene.sysName, which is the systematic names of the query genes.

- **queryGene.geneName** Column indicates the gene used as query in the synthetic genetic array screen (SGA).
- Int.geneName Column indicates the gene identified as an interactor with a particular query.
- **Int.sysName** Column indicates the systematic name of the open reading frame (ORF) that corresponds to the interactor gene.
- **Score** An interaction scored three times in the three runs by visual inspection received a scored of 3. An interaction scored twice in the three in the runs by visual inspection received a scored of 2. An interaction scored by the computer-based image analysis but not visual inspection received a scored of 1. For interactions that scored once in the three runs by visual inspection confirmation was attempted only for those genes pairs related functions. Such confirmed interactions received a score of 0.
- RSA Column identifies an interaction that was confirmed by random spore analysis.

Tetrad Column identifies an interaction confirmed by tetrad analysis.

- SS Refers to synthetic sick interaction.
- SL Refers to synthetic lethal interaction.

- **Functional.Role** Column indicates the assigned GO functional annotation from their defined subset of annotations.All the interactions are identified in this study unless otherwise stated.
- References Genetic Interactions that have been previously described.
- **queryGene.sysName** Column indicates the systematic (ORF) name of the gene used as query in a SGA screen.

At ong is a 132 by 1008 adjacency matrix of the systematic genetic interactions identified between 132 query genes and the deletion gene set (Tong et al. 2001; see SGA for more details). The row names correspond to the systematic (ORF) names for the 132 query genes. The column names correspond to the systematic (ORF) names of the 1011 reporter genes, which showed a synthetic lethal or synthetic sick interaction with at least one query genes. Values are 0 or 1, with a 1 indicating the occurrence of the genetic interaction between the gene pairs.

#### Source

Tong et al, Science. Vol.303, 2004.

## References

Global Mapping of the Yeast Genetic Interaction Network, Tong et al, Science Vol.303, 2004. http://www.sciencemag.org/cgi/data/303/5659/808/DC1/1

#### See Also

SGA

#### Examples

data(Atong) dim(Atong)

AtongFnDomain	The functional domains shared by the tested pairs in Tong et al exper-
	iment.

## Description

Data developed from Tong et. al. buffering experiments.

#### Usage

```
data(AtongFnDomain)
```

# Format

A list containing 3 items.

pairs Dataframe of all the gene pairs and their synthetic lethality status.

- **SharedPfam** List of the Pfam domains shared by each pair. The order of this list is the same as the order of the pairs.
- **SharedSMART** List of the SMART domains shared by each pair. The order of this list is also the same as the order of the pairs.

#### AtongPair

## Author(s)

Z. Jiang

# Source

Created from the association matrix reported by Tong et al. and the Pfam (Protein family database http://pfam.janelia.org/) and SMART database of yeast.

#### References

Global Mapping of the Yeast Genetic Interaction Network, Tong et al, Science Vol.303, 2004 Saccharomyces Genome Database

## Examples

```
data(AtongFnDomain)
names(AtongFnDomain)
```

AtongPair

Data frame the pair of yeast gene tested in Tong et al. 2004.

## Description

Data from Tong et. al. buffering experiments (2004) using synthetic genetic arrays (SGA) (Tong et al. 2001).

## Usage

```
data(AtongPair)
```

#### Format

A data frame with 3 columns and 607881 rows.

## Details

AtongPair stores the yeast gene names for each tested pairs in Tong buffering experiment. Each row represents one pair.

query Query gene name

array Array gene name

interact Logical indicating the synthetic lethal status, if TRUE the genetic interaction is lethal.

#### Author(s)

N. LeMeur

#### Source

Created from the association matrix reported by Tong et al (2004) and the genes from the SGA array developed by Tong et al. (2001).

#### References

Systematic genetic analysis with ordered arrays of yeast deletion mutants, Tong et al, *Science*. Vol. 294, 2001

Global Mapping of the Yeast Genetic Interaction Network, Tong et al, Science Vol.303, 2004

#### Examples

```
data(AtongPair)
dim(AtongPair)
```

Boeke

Incidence matrix of Synthetic Lethal interaction from the Boeke Lab

#### Description

Data from Pan et al. experiments on DNA integrity Network in the Yeast S. cerevisiae.

#### Usage

```
data(Boeke2006raw)
data(Boeke2006)
```

#### Format

Boeke2006raw is a data frame with 5775 observations on the following 6 variables.

Query.ORF ORF associated with the query gene.

Query.Gene Common name of the query gene.

Target.ORF ORF for the array gene.

Target.Gene Common name of the array gene.

RSA Random spore analysis

Tetrad Tetrad dissection

Boeke2006 is an incidence matrix is a 74 by 843 adjacency matrix of the systematic genetic interactions identified between 74 query genes and the deletion gene set in Pan et al.(2004). The row names correspond to the systematic (ORF) names for the 74 query genes. The column names correspond to the systematic (ORF) names of the 843 reporter genes, which showed a synthetic lethal or synthetic sick interaction with at least one query genes. Values are 0 or 1, with a 1 indicating the occurrence of the genetic interaction between the gene pairs.

## Details

In Pan et al (2006), the authors provide this note. Note: SL - synthetically lethal; SF/SL-very severe synthetic fitness defects; SF-obvious but modest synthetic fitness defects; SF (slight) - slight synthetic fitness defect. Approximately 10% of the positive interactions presented here were not scored as positive in the dSLAM screens. These were individually tested because we wanted to make sure that they were indeed false negatives in the dSLAM screens. We also note that there is a small chance that the interactions scored as positive in RSA (random spore analysis) might not reflect direct growth defects of the double mutants but rather, the double mutants are defective in expressing the MFA1pr-HIS3 reporter.

SDL

## Source

The data were extracted from Pan et al (2004) and Table S1 of Pan et al. (2006).

#### References

Pan X, Ye P, Yuan DS, Wang X, Bader JS, Boeke JD. A DNA integrity network in the yeast Saccharomyces cerevisiae. Cell. 2006 Mar 10;124(5):1069-81

Pan X, Yuan DS, Xiang D, Wang X, Sookhai-Mahadeo S, Bader JS, Hieter P, Spencer F, Boeke JD. A robust toolkit for functional profiling of the yeast genome. Mol Cell. 2004 Nov 5;16(3):487-96.

#### See Also

dSLAM.GPL1444, and dSLAM

SDL

The Association matrix for the synthetic dosage lethal screens in Yeast.

#### Description

The data reported in Table 6 of the supplementary data of Measday et. al.

#### Usage

data(SDL) data(SLchr)

## Format

SDL is a matrix with 141 rows and 9 columns. The columns represent 3 genes at each of 3 temperatures (16, 25, 37 Celsius). The gene names and temperatures are combined in the column names. The row names are yeast standard names. The values are NA, no effect, SDS for synthetic dosage sick, SL for synthetic lethal and SDL for synthetic dosage lethal.

SLChr is a matrix with 84 rows and 14 columns. Each column represents a query strain which was tested against the genome wide set of deletion strains. The entries can be NA for no effect, SL for synthetic lethal and SS for synthetic sick.

## Source

Supplementary Table 6 of the reference given below.

## References

Systematic yeast synthetic lethal and synthetic dosage lethal screens identify genes required for chromosome segregation. Measday et al, PNAS, 2005, 13956-13961.

## Examples

data(SDL) table(SDL) SGA

#### Description

Listed of yeast deletion genes used as array probes in the Systematic Genetic Analysis (SGA) of yeast deletion Tong et. al. (2001).

#### Usage

data(SGAraw) data(SGA)

## Details

SGAraw is a character vector of length 4672, corresponding to the original yeast deletion genes set on the array. Note that some of those genes correspond to ORFS that have subsequently been rejected.

SGA is a character vector of length 4655, corresponding to the updated yeast deletion genes set on the array. The gene names have been updated from common gene name or alias to systemantic names (last update Feb. 2006).

#### Source

Table S1 from Tong et al. (2001) online supporting material. http://www.sciencemag. org/cgi/content/full/294/5550/2364/DC1

## References

Systematic genetic analysis with ordered arrays of yeast deletion mutants, Tong et al, Science. Vol. 294, 2001

#### Examples

```
data(SGAraw)
length(SGAraw)
if(require("YEAST"))
updateSGA <- mget(SGAraw, YEASTCOMMON2SYSTEMATIC, ifnotfound = SGAraw)</pre>
```

SGD.SL

Interaction data from the Saccharomyces Genome Database

#### Description

The Saccharomyces Genome Database (SGD) provides, for download a table listing all known interactions in yeast. This table was downloaded on Jan 25, 2007 and three subsets were extracted. The synthetic lethal interactions, SGD.SL, the synthetic grow defect interactions, SGD.SynGrowthDefect and the synthetic rescue interactions, SGD.SynRescue. No other processing has been done.

## SGD.SL

#### Usage

```
data(SGD.SL)
data(SGD.SynRescue)
data(SGD.SynGrowthDefect)
```

# Format

Each data set is a data frame with the following 7 variables.

- V1 Factor, indicating the type of data.
- V2 Factor describing the interaction, in particular naming bait and prey and interactors.
- V3 Factor indicating whether the cells were viable.
- V4 Factor which is always NA for these data.
- V5 Factor naming the reference for the interaction.
- V6 Factor with levels indicating the PubMed ID for the publication in V5.
- V7 Factor with level BioGRID, probably indicating the source.

# Details

SGD says this about the file:

```
Contains interaction data. Tab-separated columns are:
1) interaction_type (mandatory)
2) genes involved and their mutation type, in the format: ORF
(mutation_type, action), with multiples separated by a |
3) phenotype (optional, multiples separated by |)
4) description (optional)
5) citation (multiples separated by |)
6) PubMed ID (optional, multiples separated by |)
```

This file is updated weekly.

## Author(s)

Z. Jiang

## Source

The file can be downloaded from, ftp://genome-ftp.stanford.edu/pub/yeast/literature\_curation.

# Examples

data(SGD.SL)

TFmat

#### Description

The data are from Lee et al, the rows of the matrix represent genes in S. cerevisiae, the columns known transcription factor. The value in each entry represents the p-value, as reported by Lee et al, for the transcription factor (TF) binding upstream of the gene.

## Usage

data(TFmat)

## Format

TFmat is a matrix, rows represent genes, columns transcription factors and the elements are p-values representing some notion of the likelihood that the transcription factor binds up stream of the gene.

# Author(s)

Z. Jiang

## Source

Supplementary material from http://web.wi.mit.edu/young/regulator\_network/

#### References

Transcriptional Regulatory Networks in Saccharomyces cerevisiae, Lee et al. Science 298:799-804 (2002).

## Examples

data(TFmat)

byComplex

Evaluate protein co-membership within cellular organizational units

## Description

Count the protein co-members of one (or more) cellular organizational units such as complex(es). This co-membership can be characterized by a synthetic lethal interaction if bpL is the list of observed synthetic lethal interactions or it can be characterized by the number of all the expected interactions within that complexes if bpL is all the interactions tested.

# Usage

byComplex(bpL, interactome)

#### comemberIn

## Arguments

bpL	List of tested genes (or reported as synthetic lethal) per bait.
interactome	Adjacency matrix where the rows are the genes and the columns represent the cellular organizational units, e.g., $Scisi$

# Value

Vector of the number of genes(proteins) co-member in one or more biological complexes or pathways.

## Author(s)

N. LeMeur and R. Gentleman

# See Also

withinComplex

# Examples

```
data(ScISIC)
data(AtongPair)
pairSL <- AtongPair[ AtongPair[,3],]
SLlist <- split(as.character(pairSL[,2]),as.character(pairSL[,1]))
##Number of synthetic lethal pairs within the same complexe
bySL <-byComplex(SLlist, ScISIC)</pre>
```

comemberIn *Retrieve the biological complexes.* 

# Description

Retrieve the biological complexes within which two proteins are comembers.

# Usage

```
comemberIn(iMat, interactome)
```

## Arguments

iMat	Comembership matrix of genes(proteins) that linked to other genes(proteins) by any biological experiment
interactome	Adjacency matrix composed of genes (rows) and biological complexes (columns) ScISI

# Value

Dataframe of pairs of genes(proteins) and their common biological complexes.

compare

## Author(s)

N. LeMeur

## See Also

withinComplex

# Examples

```
data(Atong)
data(ScISI)
coMember<-withinComplex(Atong,ScISI)
SLpairWithinComplex <- comemberIn(coMember,ScISI)</pre>
```

compare

Compare observed data to expected in permutation models

# Description

This method summarizes the result of the modelSLGI function.

#### Usage

```
## S4 method for signature 'siResult':
compare(x)
```

## Arguments

x a siResult object to summarize

## Details

This compares the number of observed interactions to the number of expected interactions in each permutation model. It counts how many times the number of observed interactions is greater than the number of expected interactions (from the permutations) and divides by the number of permutations applied.

## Value

Numerical vector

# Author(s)

N. LeMeur

## See Also

modelSLGI

#### congruence

## Examples

```
data(ScISIC)
data(Atong)
data(SGA)
model <- modelSLGI(Atong, universe= SGA, interactome=ScISIC, type="intM", perm=2)
ans <- compare(model)</pre>
```

```
congruence
```

Calculate congruence score between pairs of of genes sharing pattern of synthetic genetic interactions (Ye et al. (2005)).

## Description

The congruence score represents the number of common synthetic genetic interacting partners between two genes. The higher is the score the more overlap there is between the synthetic genetic partners of those genes.

# Usage

congruence(iMat, sharedInt, mode="query", universe, padjust=FALSE)

## Arguments

iMat	Adjacency matrix reporting genetic Interactions. Each entry has value 0 or 1, representing positive or negative interaction of corresponding pairs of row and column, respectively.
sharedInt	numeric vector representing the number of common genetic interactions be- tween a pair of query or target genes. See getSharedInteraction for more details
mode	character vector of value "query" or "target"
universe	total number of genes tested
padjust	adjust by the number of genes tested that show at least one synthetic genetic interaction.

#### Value

A numeric vector of the congruence score values.

## Author(s)

N. LeMeur

#### References

Ye P. et al. (2005). Gene function prediction from congruent synthetic lethal interactions in yeast. Molecular Systems Biology 1:2005.0026.

# See Also

getSharedInteraction

# Examples

createSquareMatrix Create a square matrix

# Description

Create a square matrix based on row and column names. The new matrix is created so that the row and column names are a perfect match and the added values are zero.

In the case of genetic interactions, for example it could be useful that the matrix of all the interactions tested and not tested.

# Usage

createSquareMatrix(data)

# Arguments

data Matrix

# Value

matrix.

## Author(s)

N. LeMeur

## Examples

```
data(Atong)
dim(Atong)
```

Tong<- createSquareMatrix(Atong)
dim(Tong)</pre>

dSLAM.GPL1444 dSLAM platform used for Synthetic Lethal screens in the Boeke Lab

#### Description

These data are the 21991 probes spotted on the dSLAM array (heterozygote diploid-based synthetic lethality analyzed by microarray) used to test synthetic lethal interactions by Pan et al (2006).

## Usage

```
data(dSLAM.GPL1444)
```

data(dSLAM)

## Format

dSLAM.GPL1444 is a data frame with 21991 observations on the following 10 variables.

ID Serial identifier for probe.

ROW Row number in the array as scanned with GenePix scanner.

COLUMN Column number in the array as scanned with GenePix scanner.

- TAGTYPE Code for whether tag is 5' (Up) or 3' (Dn) relative to the open reading frame (ORF).
- PROBE Code for singleton probes arrayed in ORF order (ArrA, ArrB), five-fold replicate probes arrayed in randomized order (Rpts), systematic mutations arrayed across the center of the array (Muts), negative controls (NegT), or probes peripheral to the array as specified by the manufacturer (Edge)
- ORF Systematic ORF name (from SGD, Feb 2003)(http://genome-www4.stanford.edu/ cgi-bin/SGD/locus.pl?locus=
- GENE Standard gene name (SGD) (or ORF if not available)
- SEQUENCE DNA sequence of probe (includes custom-designed sequences for 193 YA\* and YM\* ORFs missing DnTags)
- SGDID Unique ORF identifier from SGD; 'S000000000' denotes missing value
- SPOT\_ID spot identifier; ('YQL' ORFs denote custom-designed sequences; 'NegA', 'NegB', 'PosA', 'PosB' denote proprietary sequences specified by the manufacturer)

dSLAM is a character vector of length 5641 that contains the unique and valid systematic ORF names.

#### Details

The dSLAM.GPL1444 were directly obtain from parsing the GPL1444\\_family.soft.gz available at http://www.ncbi.nlm.nih.gov/projects/geo/query/acc.cgi?acc=GPL1444

dSLAM is a vector of length 5641, extracted from the dSLAM.GPL1444 ORF, and that contains the unique and valid systematic ORF names. This vector was built in three steps. First the ORFs with SGDID equals to S000000000 in the dSLAM.GPL1444 data frame were removed as some correspond to custom sequences and other were dubious ORFs that have been deleted from SGD or merged with other ORFs. Secondly, the duplicated names were removed. Then, the systematic ORF names were verified against the YEAST data package.

#### Source

The data were extracted from the Gene Expression Omnibus (GEO) website: http://www.ncbi.nlm.nih.gov/projects/geo/query/acc.cgi?acc=GPL1444

# References

Pan X, Ye P, Yuan DS, Wang X, Bader JS, Boeke JD. A DNA integrity network in the yeast Saccharomyces cerevisiae. Cell. 2006 Mar 10;124(5):1069-81

## See Also

Boeke2006raw, and Boeke2006

domainDist Finds the number of gene sets for each shared domain

# Description

domainDist takes a list of shared domains, and compute for each distinct domain how many gene sets share it.

# Usage

```
domainDist(domainL)
```

#### Arguments

domainL Each element of the list is a vector of functional domains.

## Details

For each domain that appears in the domain list, domainDist counts the number of elements that have this domain.

## Value

Returns a frequency table with descending order.

# Author(s)

Z. Jiang

## See Also

getSharedDomains, sharedBy

```
data(AtongFnDomain)
domainDist(AtongFnDomain$SharedPfam[1:20])
```

essglist

#### Description

List of systematic names and common names of the yeast essential genes.

## Usage

```
data(essglist)
```

#### Format

essglist is a list with 1103 elements (last download 03/17/2006). The name of each element is the systematic gene name. The value of each element is its corresponding common (standard) name.

#### Details

The aliases of the yeast gene names can be retrieved with the YEASTALIAS environment of the YEAST package.

# Source

Saccharomyces Genome Database ftp://genome-ftp.stanford.edu/pub/yeast/data\_ download/literature\_curation/phenotypes.tab (last download 03/17/2006)

## References

Saccharomyces Genome Database http://www.yeastgenome.org/

## Examples

```
data(essglist)
essglist[[1]]
names(essglist)
```

getFASTAname Obtain sequence name from FASTA object

#### Description

Extract the name of a sequence from a FASTA object that created by readFASTA function from Biostrings package.

## Usage

```
getFASTAname(Fobj)
```

# Arguments

Fobj	is a FASTA object created by re	adFASTA function from <b>Biostrings</b> pa	ickage.
------	---------------------------------	--	---------

## Details

The function gets the first string between ">" and space the "desc" element of the Fobj, which is the names of the sequence.

# Value

A character string.

## Author(s)

Z. Jiang

## See Also

readFASTA

## Examples

```
f <- gzfile(file.path(.path.package("SLGI"),
                      "extdata/orf_trans.fasta.gz"), open = "rt")
library(Biostrings)
yeastF <- readFASTA(f)
sapply(yeastF[1:5], getFASTAname)
```

getInteraction	Count genetic interactions within and between cellular organizational
	units

## Description

Count the number of genetic interactions within and between the elements of the interactome.

## Usage

```
getInteraction(iMat, universe, interactome)
```

## Arguments

iMat	Interaction matrix. Each entry has value 0 or 1, representing positive or negative interaction of corresponding pairs of row and column, respectively.
universe	Character vector of gene names, e.g., array genes used in synthetic genetic array experiments (SGA)
interactome	Adjacency matrix where row are gene names and columns are cellular organizational units.

#### getSharedDomains

## Value

The returned value is a list of 2 matrices:

bwMat	A interaction matrix that corresponds to the cellular organizational units interac-
	tion matrix where row and columns a organizational units names and the value
	inside the matrix are the number of genetic interactions they share.
CDs	Subset of the input interactome that shares interactions.

# Author(s)

N. LeMeur

#### Examples

```
##Create the genetic interaction matrix
gInt <- sample(c(0, 1), 25, TRUE)
iMat <- matrix(gInt, nrow=5, ncol=5, dimnames=list(letters[1:5],letters[4:8]))
##Create the interactome
cInt <- sample(c(0,1),30, TRUE)
interactome <- matrix(cInt, nrow=6, ncol=5,dimnames=list(letters[2:7],LETTERS[1:5]))
## Reduce the genetic interaction matrix to match the gene present in
## the interactome
reducediMat <- gi2Interactome(iMat, interactome)
## Create the interaction matrix</pre>
```

```
## Get the interaction
prey <- letters[1:20]
myInteraction <- getInteraction(reducediMat, prey, interactome)</pre>
```

getSharedDomains Find domains shared by a given list of gene names.

## Description

getSharedDomains finds domains in the provided environment that are shared by a list of genes.

#### Usage

```
getSharedDomains(geneNameV, env)
```

# Arguments

geneNameV (	Character vector of gene names.
	R object that provides mappings between an entrez gene identifier and the asso- tiated Pfam identifiers.

# Value

getSharedDomains returns a vector of the names of the shared domains.

## Author(s)

Z. Jiang

## See Also

domainDist, sharedBy

# Examples

```
library("org.Sc.sgd.db")
getSharedDomains(c("YEL003W","YLR200W"), org.Sc.sgdPFAM)
```

```
getSharedInteraction
```

Calculate the number of shared synthetic genetic interactions between pairs of genes.

# Description

The number of common synthetic genetic interacting partners between two genes.

#### Usage

```
getSharedInteraction(iMat, mode="query")
```

# Arguments

iMat	Adjacency matrix reporting genetic Interactions. Each entry has value 0 or 1,
	representing positive or negative interaction of corresponding pairs of row and
	column, respectively.
mode	Character vector of value "query" or "target"

## Value

A numeric vector of the number of common genetic interactions between a pair of query or target genes.

## Author(s)

N. LeMeur

## See Also

congruence

## Examples

sharedInt <- getSharedInteraction(intM)</pre>

getTestedPairs Find interacting and non-interacting tested pairs from an genetic interaction matrix.

# Description

getTestedPairs find all the pairs from an interaction matrix and a list of tested genes.

# Usage

```
getTestedPairs(iMat, respV)
```

## Arguments

iMat	Adjacency matrix reporting genetic Interactions. Each entry has value 0 or 1, representing positive or negative interaction of corresponding pairs of row and column, respectively.
respV	Character vector of all gene names that were tested (found to interact or not)

# Value

A data.frame with 4 columns:

query	gene names of the query genes
array	gene names og the tested genes (e.g., array genes)
interact	numeric vector of the number of observed interactions (0: no interaction; 1: one interaction; 2: two interactions when the query genes were also on the array)
recip	logical to indicate whether the reported genes were both query and array genes (TRUE: both genes were query and array genes)

# Author(s)

•

N. LeMeur

# See Also

getSharedDomains getUniquePairs

```
intM <- c(0,1,0,0,1,1,0,0,1,0,0,1,1,0,1,0)
dim(intM) <- c(4,4)
dimnames(intM) <- list(c("p1","p2","p3","p4"),c("p1","p3","p5","p7") )
respV <- c("p6","p8")
intM
getTestedPairs(intM,respV)</pre>
```

getUniquePairs

#### Description

getUniquePairs can find all the unique pairs from an interaction matrix and supplementary array genes, or finds only the unique pairs that shows positive interaction.

## Usage

```
getUniquePairs(iMat, respV = character(0), only = FALSE)
```

## Arguments

iMat	Adjacency matrix reporting genetic Interactions. Each entry has value 0 or 1, representing positive or negative interaction of corresponding pairs of row and column, respectively.
respV	Character vector of all gene names that were tested (found to interact or not)
only	has default value FALSE, if TRUE, then only reports the positively interacted pairs.

# Value

A data.frame with two or three columns. The first two columns are the query gene name and the array gene name, respectively. If only is TRUE, the third column shows the interaction status.

## Author(s)

Z. Jiang

## See Also

getSharedDomains

```
intM <- c(0,1,0,0,1,1,0,0,1,1,0,1,0)
dim(intM) <- c(4,4)
dimnames(intM) <- list(c("p1","p2","p3","p4"),c("p1","p3","p5","p7") )
respV <- c("p6","p8")
intM
getUniquePairs(intM,respV,only=FALSE)
getUniquePairs(intM,respV,only=TRUE)
getUniquePairs(intM,only=FALSE)
getUniquePairs(intM,only=FALSE)
```

gi2005

# Description

The data are in the form of a 424 by 424 array which contains the scores from using the EMAP procedure on yeast strains which are ideally double mutants, each strain with a different pair of genes knocked out. For each row, the gene named in the row label is knocked out in all pairs, and the same holds true for each column.

# Usage

data(gi2005) data(gi2005.metadata)

## Format

gi2005 is a 424 by 424 array of real values. gi2005.metadata is a vector of length 424 which contains the common names for the genes that were knocked out. The row and column names of gi2005 are standard names.

# Details

NA values in gi2005 are interactions that were not scored.

## Source

Data were obtained as supplementary material from the publication listed below.

#### References

Schuldiner et al, Exploration of the function and organization of the yeast early secretory pathway through an epistatic miniarray profile. Cell, 2005, 123:507-519.

Collins et al, A strategy for extracting and analyzing large-scale quantitative epistatic interaction data. Genome Biology, 2006, 7:R63.

```
data(gi2005)
data(gi2005.metadata)
```

gi2007

## Description

The data gi2007 are a 754 by 754 set of genetic interactions that were tested pairwise by either deletion or decreased abundance messenger RNA perturbation.

#### Usage

```
data(gi2007)
data(gi2007.metadata)
```

## Format

The gi2007 data are a 754 by 754 matrix where values indicate a score for a synthetic genetic interaction. An NA indicates that the genetic interaction was not measured.

gi2007.metadata is a data.frame of dimensions 754 rows and two columns. The columns are the systematic names and the mutation (which is typically either DAMP, DELETION or the name of the alternate allele that was tested. In 11 cases an alternative allele was tested.

#### References

Collins et al. Nature, 2007, Vol 446, p. 806-810. Data are available as supplementary material.

#### Examples

```
data(gi2007)
data(gi2007.metadata)
```

gi2Interactome *Reduce genetic interactions matrix* 

## Description

Reduce genetic interactions matrix to the pairs that genetically interact and that are present in the interactome of interest.

#### Usage

```
gi2Interactome(iMat, interactome, threshold=0)
```

## Arguments

iMat	Genetic interaction matrix. Each entry has usually a value of 0 or 1, represent- ing positive or negative interaction of corresponding pairs of row and column, respectively.
interactome	Interactome matrix, e.g. ScISIC.
threshold	Integer

#### hyperG

## Value

The returned value is the genetic interaction matrix reduced to the row and column (genes) names that are present in the interactome and where the row and column sums are higher than the specified threshold.

## Author(s)

N. LeMeur

# Examples

```
##Create the genetic interaction matrix
gInt <- sample(c(0, 1), 25, TRUE)
iMat <- matrix(gInt, nrow=5, ncol=5, dimnames=list(letters[1:5],letters[4:8]))
##Create the interactome
cInt <- sample(c(0,1),30, TRUE)
interactome <- matrix(cInt, nrow=6, ncol=5,dimnames=list(letters[2:7],LETTERS[1:5]))
## Reduce the genetic interaction matrix to match the gene present in
## the interactome
reducediMat <- gi2Interactome(iMat, interactome)</pre>
```

hyperG

Hypergeometric test

### Description

a hypergeometric test for genetic interaction data.

# Usage

hyperG(data, nbTested, universe)

## Arguments

data Matrix with 2 columns the first one corresponds to the number of interactions per pair of interacting complexes and the second one to number of tested interactions. This could be the first two columns resulting from a call to the test2Interact function.

nbTested	Number of interacting pairs
universe	Total Number of tested pairs

## Author(s)

N. LeMeur

## See Also

phyper

#### iSummary

## Examples

```
## Create matrix interaction x tested matrix
interact <- c(1, 3, 2, 2, 6, 5, 2, 4, 1, 3)
tested <- c(3, 3, 5, 4, 8, 5, 3, 4, 2, 3)
mat <- cbind(interact, tested)
## Perform test
res <- hyperG(mat, 1000, 10000)
summary(res$P)
```

iSummary Summarize cellular organizational units sharing genetic interaction

## Description

Summarize the cellular organizational units sharing genetic interactions and display their GO annotation if available

# Usage

iSummary(iMat, n=10, reverse=FALSE)

## Arguments

iMat	Comembership matrix of genes(proteins) that linked to other genes(proteins) by any biological experiment, e.g., output of the getInteraction function.
n	Numeric threshold indicating the minimum number of genetic interactions that a pair of cellular organizational unit must share.
reverse	Logical, by default the function return a list of pair of cellular organizational units where the name of each element is the number of genetic interactions they share. If reverse is TRUE, the output is a vector where the values are the number of interactions and the names are the combination of the 2 cellular organizational units.

## Value

The function print the result in the standard output but can also save it in variable.

If reverse is FALSE the output is a list of pairs of cellular organizational units where the name of each element is the number of genetic interactions they share.

If reverse is TRUE the output is a vector where the values are the number of interactions and the names are the combination of the 2 cellular organizational units.

## Author(s)

N. LeMeur

## modelSLGI

## Examples

```
data(Atong)
data(ScISIC)
data(SGA)
SLa2 <- gi2Interactome(Atong, ScISIC)
## Search for synthetic lethal interaction
compM <- getInteraction(SLa2, SGA, ScISIC)
## Display the tightly interacting pairs
largeInt <- iSummary(compM$bwMat,n=15)</pre>
```

modelSLGI

Permutation model for assessing synthetic genetic interactions in cellular organizational units.

# Description

Permutation model for assessing synthetic genetic interactions within and between cellular organizational units such as multi-protein complexes.

## Usage

modelSLGI(iMat, universe, interactome, type="intM", perm=50)

## Arguments

iMat	Adjacency matrix reporting genetic interactions. Each entry has value 0 or 1, representing positive or negative interaction of corresponding pairs of row and column.
universe	character vector of the names of the tested genes, e.g., names of the genes on the synthetic genetic array (SGA) used by Tong et al.
interactome	Adjacency matrix where row are genes and columns are cellular organizational units. Each entry has value 0 or 1, for absence or presence of a gene in a complex.
type	Character vector of value "intM" (Default) or "interactome" to either perform the test based on to the genetic interaction matrix or the interactome, respectively.
perm	Number of permutations to apply. Default is 50.

## Value

Interaction matrix between cellular organizational units.

## Author(s)

N. LeMeur

## See Also

getInteraction

# Examples

```
data(ScISIC)
data(Atong)
data(SGA)
model <- modelSLGI(Atong, universe= SGA, interactome=ScISIC,
type="intM", perm=2)</pre>
```

normInteraction Normalize a matrix of biological interactions

## Description

Normalize a square matrix of biological interactions according to the number of possible interactions between each biological complex.

# Usage

normInteraction(data, genename, interactome)

## Arguments

data	Square Matrix of biological complexes that shares one or more genes(proteins)
genename	Character vector of the gene names that possibly create interactions between complexes
interactome	Adjacency matrix where row are genes and columns are cellular organizational units. Each entry has value 0 or 1, for absence or presence of a gene in a complex, e.g., ScISI

## Value

Square matrix of biological complexes linked by one or more interacting proteins and normalized by the possible number of interactions between each complex.

## Author(s)

N. LeMeur

# See Also

getInteraction

## Examples

```
data(Atong)
data(ScISIC)
data(SGA)
SLa2 <- gi2Interactome(Atong, ScISIC)
## Search for synthetic lethal interaction
compM <- getInteraction(SLa2, SGA, ScISIC)
## Normalize
normIntComplex<- normInteraction(compM$bwMat, SGA, ScISIC)</pre>
```

plot

# Description

a plot method for siResult.

# Usage

```
## S4 method for signature 'siResult':
plot(x,...)
```

## Arguments

Х	the siResult object to plot.
	general commands to be sent to plot.

# Details

The plot generated from a siResult object is a dotplot with the observed and expected data average of interaction represented in 2 different colors.

## Author(s)

N. LeMeur

## See Also

ScISI

## Examples

```
data(ScISIC)
data(Atong)
data(SGA)
model <- modelSLGI(Atong, universe= SGA, interactome=ScISIC,
type="intM", perm=2)
plot(model)</pre>
```

seqMatcherAlign	Functions to do local alignment of two sequences using EMBOSS	5
	matcher	

# Description

seqMatcherAlign matches two sequences using the EMBOSS matcher program.
getAlignStats extract the statistics from the alignment result data.

## Usage

```
seqMatcherAlign(pairNameV,BankIDV,seqBank)
getAlignStats(alignRes)
```

## Arguments

pairNameV	a vector of gene pair names
BankIDV	a vector of the sequence IDs in the sequence Bank.
seqBank	a database of all the sequences
alignRes	object returned by seqMatcherAlign

## Details

seqMatcherAlign matches the gene pair names with the sequence bank IDs and export the two sequences in to two files: seq1.new and seq2.new. Then uses system calls to run EMBOSS matcher program to align the two sequences. The result from matcher is store in file "out.matcher". seqMatcherAlign read in this file and create a R object summarize the alignment results.

getAlignStats takes the alignment result data and extract the statistics of the result in to data.frame.

## Value

names	contains the names of the gene pair
results	contains the alignment statistics: the aligned total length, the number of identical match, the number of similar match, the number of gaps, and the alignment score
seq	displays the aligned sequences

## Note

pairMatcherAlign use system calls to run EMBOSS matcher program. You must have EM-BOSS matcher installed on your computer.

#### Author(s)

Z. Jiang

## References

EMBOSS: The European Molecular Biology Open Software Suite (2000) Rice, P. Longden, I. and Bleasby, A. Trends in Genetics 16, (6) pp276–277

# Examples

```
seq1 <- "RPHEDEKEAIDEAKKMKVPGENEDESKEEEKSQELEEAIDSKEKSTDARDEQGDEGDNEEENNEEDNENENHTAPPALV
seq2 <- "QKYLLKKAIRNFSEYPFYAQNKLIHQQATGLILTEEEKSQELEEKIISKIKKEEHLKKINLKHDYFDLQKKYEKECEILT
seq3 <- "IHQQATGLILTKIISKIKKEEHVPGENEDLKKINLKHDYFDLQKKYEKECEILTKLSENLRKEEIENKRKEHELMEQKRE
seqBank <- list(seq1=list(seq=seq1),seq2=list(seq=seq2),seq3=list(seq=seq3))
bid <- names(seqBank)
pnames <- c("seq1","seq3")
## Not run:
ar <- seqMatcherAlign(pnames, bid, seqBank)</pre>
```

## sharedBy

```
ar
getAlignStats(ar)
## End(Not run)
```

sharedBy

#### Find the gene pairs that share a domain.

## Description

sharedBy finds whether the given domain is in each of the elements of the domain list.

## Usage

```
sharedBy(domainL)
```

## Arguments

domainL is a list, each element of the list is a vector of domains.

# Details

sharedBy first remove all the elements with length 0 or have value 'NA'. Then apply the reverseSplit on the remaining list.

## Value

A list with each element represent a domain, and the values of the element are the pairs that share this domain.

## Author(s)

Z. Jiang

## See Also

reverseSplit,domainDist,getSharedDomains

```
## Load PFAM and SMART domains shared between Tong's Synthetic lethal data
data(AtongFnDomain)
## Find pair that share identical domain
sharedBy(AtongFnDomain$SharedPfam[1:20])
```

sharedInt

## Description

List shared interactions and cellular organizational units names between genes.

#### Usage

sharedInt(pairL, interactome, threshold=0)

## Arguments

pairL	Dataframe with 3 columns. The first columns are the pair of genes tested i.e., the query and array genes. The third columns in a logical: TRUE when the 2 genes genetically interact and FALSE when they do not.(see AtongPair dataset as example)
interactome	Adjacency matrix where row are gene names and columns are cellular organi- zational units names. Each entry has value 0 or 1, for absence or presence of a gene in the complex.
threshold	Numeric. Indicate the minimum number of interactions that 2 genes must share

# Value

The return value is a list. Each element of the list has for name 2 genes that genetically interact. Each element of the list corresponds to the list of cellular organizational units where the interacting genes are found (independently or together).

## Author(s)

N. LeMeur

```
## Synthetic genetic interactions
dat <- data.frame("query" = LETTERS[1:5], "array" = LETTERS[2:6], "interact" = as.logical
## interactome
interA <- matrix(sample(c(0, 1), 30,TRUE), nrow=6, ncol=5,dimnames = list(LETTERS[1:6], 1
sharedInt(dat, interA, threshold=1)</pre>
```

siResult-class A class for representing the result of the SLGI graph permutation model.

# Description

A class for representing the result of the modelSLGI function.

# Slots

- Observed: Return a "numeric" vector: the observed number of synthetic genetic interactions between components of one or two cellular organizational units
- Expected: Return a matrix: the expected number of synthetic genetic interactions between components of one or two cellular organizational units

## Methods

plot Graphical representation of the permutation model result compare Summarizes the result of the modelSLGI function

## Author(s)

N. LeMeur

#### See Also

modelSLGI,plot

### Examples

```
## apply a permutation model
data(ScISIC)
data(Atong)
data(SGA)
model <- modelSLGI(Atong, universe= SGA, interactome=ScISIC,
   type="intM", perm=2)
```

model

test2Interact	Summarize genetic interactions within or between cellular organiza-
	tional units

## Description

Summarize the genetic interactions within one cellular organizational unit or between 2 cellular organizational units.

## Usage

test2Interact(iMat, tMat, interactome)

# Arguments

iMat	Genetic interaction matrix. Each entry has value 0 or 1, representing positive or negative interaction of corresponding pairs of row and column, respectively.
tMat	Adjacency matrix of tested object. Each entry has value 0 or 1, representing the fact that the corresponding pairs of row and column have been tested for interaction or not.
interactome	Adjacency matrix where row are gene names and columns are cellular organi- zational units names. Each entry has value 0 or 1, for absence or presence of a gene in the complex.

## Value

the return value is a data.frame with 6 columns.

unit1,	unit2	cellular organizational units tested and interacting
tested		Number of interactions tested between unit1 and unit2
intera	ct	Number of interactions found between unit1 and unit2
sizeC1, sizeC2		
Number of genes in unit1 and unit2		

## Author(s)

N. LeMeur

## Examples

```
set.seed(123)
##Create the interactome
cInt <- sample(c(0,1),30, TRUE)
interactome <- matrix(cInt, nrow=6, ncol=5,dimnames=list(letters[2:7],LETTERS[1:5]))
## Create cellular organizational units interaction matrix
gInt <- sample(c(1:8), 25, TRUE)
gInt <- matrix(gInt, nrow=5, ncol=5, dimnames=list(LETTERS[1:5],LETTERS[1:5]))
## All interactome tested
gTest <- matrix(sample(c(0:3), 25, TRUE), nrow=5, ncol=5)
gTested <- gInt+gTest
val <- test2Interact(iMat=gInt, tMat=gTested, interactome=interactome)</pre>
```

topInteraction Extract interacting biological complexes

## Description

Extract the top X interacting biological complexes.

# Usage

topInteraction(data,top=10)

#### twoWayTable

## Arguments

data	Square matrix of biological complexes that shares one or more genes(proteins)
top	Interger that represents the percentage of interacting complexe

# Value

Data frame of biological complexes that interact. The first two columns are the cellular organizational units names and the third column indicates the number of interactions.

## Author(s)

N. LeMeur

## Examples

```
data(Atong)
data(ScISIC)
data(SGA)
SLa2 <- gi2Interactome(Atong, ScISIC)
## Search for synthetic lethal interaction
compM <- getInteraction(SLa2, SGA, ScISIC)
top10Interaction<- topInteraction(compM$bwMat,top=10)</pre>
```

twoWayTable	Generate two-way table for	genetic interaction data
-------------	----------------------------	--------------------------

## Description

Generate two-way table from a vector of genetic interaction status and a vector of the pairs that share a functional domain.

#### Usage

```
twoWayTable(var1, var2idx)
```

## Arguments

varl	Vector of the status of the first property.
var2idx	Vector of the index in var1 that have the second property.

## Details

Calculates the count numbers from the given vectors. Then put them into a matrix format.

#### Value

A two-way contingency table of genetic interaction and whether sharing a functional domain.

# Author(s)

Z. Jiang

## See Also

sharedBy,getUniquePairs

## Examples

```
var1 <- c(0,1,1,0,0,0,1,0,1,1)
var2idx <- c(3,5,7)
twoWayTable(var1,var2idx)</pre>
```

```
data("AtongFnDomain")
pf <- Biobase::reverseSplit(AtongFnDomain$SharedPfam)
idx <- which(rownames(AtongFnDomain$pairs) %in% pf$PF00478)
twoWayTable(AtongFnDomain$pairs[,"interact"],idx)</pre>
```

withinComplex Search for protein co-membership within complexes.

## Description

Search for protein co-membership within one (or more) complex(es).

# Usage

```
withinComplex(data, interactome)
```

## Arguments

data	Binary matrix of genes(proteins) linked to other genes(protein) by any biological experiment
interactome	Binary matrix composed of genes (rows) and biological complexes (columns) ScISI

# Value

Matrix of genes(proteins) co-member of one or more biological complexes.

## Author(s)

N. LeMeur

## See Also

byComplex

# Examples

```
data(Atong)
data(ScISIC)
coMember <- withinComplex(Atong, ScISIC)
table(coMember)</pre>
```

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