Package 'PCpheno'

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Type Package

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Title Phenotypes and cellular organizational units

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Description Tools to integrate, annotate, and link phenotypes to cellular organizational units such as protein complexes and pathways.				
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Suggests KEGG.db, GO.db, org.Sc.sgd.db				
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Description

Tools to integrate, annotate and search for associations between phenotypes, protein complexes, and pathways.

Details

Package: PCpheno Type: Package Version: 1.3.1 Date: 2006-03-09

License: The Artistic License, Version 2.0

Author(s)

N. LeMeur and R. Gentleman

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References

Giaever G, et al. (2002) Functional profiling of the Saccharomyces cerevisiae genome. Nature. 418(6896):387-91. PMID: 12140549

Deutschbauer AM, Jaramillo DF, Proctor M, Kumm J, Hillenmeyer ME, Davis RW, Nislow C, Giaever G. (2005) Mechanisms of haploinsufficiency revealed by genome-wide profiling in yeast. Genetics. 169(4):1915-25.

Byrne KP, Wolfe KH. (2005) The Yeast Gene Order Browser: combining curated homology and syntenic context reveals gene fate in polyploid species. Genome Res. 15(10):1456-61. PMID: 16169922

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See Also

ScISI,SLGI

buildFDMat

Build fitness defect contingency matrix

Description

Function to build a fitness defect contingency matrix where rows correspond to tested genes and columns to experimental conditions.

Usage

buildFDMat(data, genenames, condition)

Arguments

List of 'significant' fitness defect scores and the associated genes at different data

experimental conditions.

Character vector of the different experimental conditions tested condition

Character vector of all the tested genes for fitness defect. genenames

Value

Contingency matrix of genes that present significant fitness defect in different experimental conditions.

Author(s)

N. LeMeur

Examples

data(GiaeverPheno) data(GiaeverExpCdt)

data(GiaeverGene)

fitnessData <- getFDgene (GiaeverPheno, condition=GiaeverExpCdt, cutoff=c(20,100,100), mode="generation", subset=getFDgene (GiaeverPheno, cutoff=c(20,100,100), mode=generation", subset=getFDgene (GiaeverPheno, cutoff=c(20,100,100), mode=generation", subset=getFDgene (GiaeverPheno, cutoff=c(20,100,100), mode=generation", subset=getFDgene (GiaeverPheno, cutoff=c(20,100,100), mode=generation", subset=getFDgene (GiaeverPheno, cutoff=c(20,100,100), mode=generation"), subset=getFDgene (GiaeverPheno, cutoff=c(20,100), mode=generation"), subset=getFDgene (GiaeverPheno, cutoff=c(20,100), mode=gene (GiaeverPheno, cutoff=Giaever Pheno M < -build FDMat (data=fitness Data, genenames = Giaever Gene, condition = Giaever ExpCdt[,3])

categoryToEntrezBuilder

Return a list mapping multi-protein complexes IDs to YEAST ids

Description

Return a list mapping multi-protein complexes (category) IDs to the YEAST ids annotated at the category id.

Usage

```
\#\# S4 method for signature 'CoHyperGParams' categoryToEntrezBuilder(p)
```

Arguments

р

A subclass of HyperGParams-class

Details

End users **should not** call this directly. This method gets called from hyperGTest. To add support for a new category, a new method for this generic must be defined. Its signature should match a subclass of HyperGParams-class appropriate for the new category.

Value

A list mapping category IDs to YEAST identifiers.

Author(s)

S. Falcon and N. LeMeur

See Also

hyperGTest CoHyperGParams-class

CoHyperGParams-class Class "CoHyperGParams"

Description

A parameter class for representing all parameters needed for running the hyperGTest method with multiprotein complexes.

Objects from the Class

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

Slots

- geneIds: Object of class "ANY": A vector of gene identifiers. Numeric and character vectors are probably the only things that make sense. These are the gene ids for the selected gene set.
- universeGeneIds: Object of class "ANY": A vector of gene ids in the same format as geneIds defining a subset of the gene ids on the chip that will be used as the universe for the hypergeometric calculation. If this is NULL or has length zero, then all gene ids on the chip will be used.
- annotation: A string giving the name of the annotation data package for the chip used to generate the data.
- cateogrySubsetIds: Object of class "ANY": If the test method supports it, can be used to specify a subset of category ids to include in the test instead of all possible category ids.
- categoryName: A string describing the category. Usually set automatically by subclasses. For example "ScISI".
- pvalueCutoff: The p-value to use as a cutoff for significance for testing methods that require it. This value will also be passed on to the result instance and used for display and counting of significant results. The default is 0.01.
- testDirection: A string indicating whether the test should be for overrepresentation ("over") or underrepresentation ("under").

Methods

- $\label{eq:hyperGTest} \begin{array}{ll} \textbf{hyperGTest} & \text{signature}(p = \text{"HyperGParams"}) \text{: Perform hypergeometric tests to assess} \\ \textbf{over-representation of category ids in the gene set.} & \textbf{See the documentation for the generic function for details.} \\ \textbf{This method must be called with a proper subclass of HyperGParams.} \end{array}$
- geneIds(r), $geneIds(r) \leftarrow value$ Accessors for the gene identifiers that will be used as the selected gene list.

codeannotation(object) Accessor for annotation

ontology(r) Accessor for GO ontology.

pvalueCutoff(r), pvalueCutoff(r) <- value Accessor for the p-value cutoff. When setting, value should be a numeric value between zero and one.

 $test Direction \ \ Accessor \ for \ the \ test \ direction. \ When setting, value \ must be either "over" \ or "under".$ $universe Gene Ids(r) \ \ accessor \ for \ vector \ of \ gene \ identifiers.$

isConditional(r) Returns TRUE if the instance has its conditional flag set

Author(s)

S. Falcon and N. LeMeur

See Also

 $HyperGResult-class\ CoHyperGResult-class\ hyperGTest$

CoHyperGResult-class Class "CoHyperGResult"

Description

This class represents the results of a test for over-representation of genes in a selected gene set based among protein complexes upon the Hypergeometric distribution.

Objects from the Class

Objects is created by calls to the funtion hyperGTest.

Slots

pvalues: "numeric" vector: the ordered p-values for each category term tested.

oddsRatios: Object of class "numeric" Odds ratio for each category term tested

expectedCounts: Object of class "numeric" The expected number of genes for each gene term tested

geneCounts: "integer" vector: for each category term tested, the number of genes from the gene set that are annotated at the term.

universeCounts: "integer" vector: for each category term tested, the number of genes from the gene universe that are annotated at the term.

catToGeneId: Object of class "list". The names of the list are category IDs. Each element is a vector of gene IDs annotated at the given category ID and in the specified gene universe.

Extends

Class "HyperGResultBase", directly.

Methods

```
\label{eq:geneCounts} \begin{tabular}{ll} geneCounts & signature (r = "CoHyperGResult"): return an "numeric" vector: for each category term tested, the number of genes from the gene set that are annotated at the term. \\ \end{tabular}
```

```
\begin{array}{ll} \textbf{p-values} \ \ \mathrm{signature}(r = \quad \text{"HyperGResult"}) \text{: return a "numeric" vector: the ordered} \\ \text{p-values for each category term tested.} \end{array}
```

```
\begin{tabular}{ll} \textbf{oddsRatios} & signature (r = "HyperGResult"): return a "numeric" vector: the odds ratio for each category term tested. \\ \end{tabular}
```

 $\begin{tabular}{ll} \textbf{universeCounts} & signature (r = "HyperGResult"): return an "numeric" vector: for each category term tested, the number of genes from the gene universe that are annotated at the term. \\ \end{tabular}$

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summary signature(r = "CoHyperGResult"): Returns a data.frame summarizing the test result. Optional arguments pvalue and categorySize allow specification of minimum p-value and categorySize, respectively. Optional argument htmlLinks is a logical value indicating whether to add HTML links (useful in conjunction with xtables print method with type set to "html").

Author(s)

S. Falcon and N. LeMeur

See Also

HyperGResultBase-class

Examples

```
data(DudleyPhenoM)
data(ScISIC)
## Select genes sensitive to paraquat
DudleyPhenoL <- apply(DudleyPhenoM,2,function(x) names(which(x==1)))
paraquat <- DudleyPhenoL[["Paraq"]]
## Apply a hypergeometric test
params <- new("CoHyperGParams",
          geneIds=paraquat,
          universeGeneIds=rownames(ScISIC),
          annotation="org.Sc.sgd.db",
          categoryName="ScISIC",
          pvalueCutoff=0.01,
          testDirection="over")
paraquat.complex <- hyperGTest(params)
\#\# access the p-values
pvalues(paraquat.complex)[1:5]
\#\# Display a summary of the results
summary(paraquat.complex)[,1:4]
```

complexStatus

Complex Status

Description

Categorize the complex whether or not a complex is composed of a significant number of genes involved in a particular phenotype than expected by chance.

Usage

```
complexStatus(data, phenotype, interactome, threshold=0.05)
```

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Arguments

data Output from CoHyperG test

phenotype List of gene names inducing an observed phenotype, e.g., list of essential gene

names (see package SLGI)

interactome A binary matrix composed of genes (rows) and biological complexes (columns)

(see package ScISI)

threshold pvalue threshold (default 0.05)

Details

We form four distinct categories from A to D to characterize how a complex might be involved in a particular phenotype (according to the number of genes it contains and that are involved in a particular phenotype - see also hyperGTest function)

Value

The returned value is a list with components:

A "interesting" complexes, complexes with a significant number of interesting

genes, i.e., genes that participate to a particular phenotype (at a given p-values

threshold)

B complexes with a NON significant number of interesting genes BUT that SHARE

genes with complexes from the A status

C complexes with a NON significant number of interesting genes AND that DON'T

SHARE interesting genes with complexes from cat A

D complexes WITHOUT interesting genes, i.e. the one involved in the studied

phenotype

Author(s)

N. LeMeur

densityEstimate 9

|--|

Description

Function to calculate the ratio of genes that characterize a phenotype (observed) among the genes that characterize a biological complex versus the ratio of a set of randomly sampled genes (expected) among the genes that characterize a biological complex.

Usage

densityEstimate(genename,interactome,perm)

Arguments

genename Character vector of the gene names that characterize a specific phenotype.

interactome Contingency matrix of genes (rows) and biological complexes (columns) (see

package ScISI)

perm Numeric vector indicating the number of simulations to run to compute the ex-

pected ratios.

Value

List of observed and simulated ratios.

Author(s)

N. LeMeur

Examples

```
data(ScISI)
data(essglist)
essential <- names(essglist)
ScISI <- as.matrix(ScISI)
ratio<- densityEstimate(genename=essential,interactome=ScISI,perm=50)
```

deResult-class

A class for representing the result of a densityEstimate test.

Description

A class for representing the result of a densityEstimate test.

Slots

Size: Object of class "numeric" representing the size of the cellular organizational unit tested

Observed: Return a "numeric" vector: the observed number of interactions between genes inducing a specific phenotype and each cellular organizational units

Expected: Return a matrix: the expected number of interactions between genes inducing a specific phenotype and each cellular organizational units

DudleyPheno DudleyPheno

Extends

```
Class "testResult", directly.
```

Methods

plot Graphical representation of the test result

Author(s)

N. LeMeur

See Also

testResult, gtResult, densityEstimate, plot

Examples

```
## apply a densityEstimate test
data( DudleyPhenoM)
data(ScISIC)

DudleyPhenoL <- apply(DudleyPhenoM,2,function(x) names(which(x==1)))
pH3 <- DudleyPhenoL[["pH3"]]

perm <- 20
pH3Density <- densityEstimate(genename=pH3, interactome=ScISIC, perm=perm)

## access results
pH3Density@Observed[1:5]

## use of the plot method
plot(pH3Density)
```

DudleyPheno

List of fitness defect score generated from Dudley et al 2005

Description

Dudley et al (2005) create a collection of gene-deletion mutants to determine genes that contribute to a particular phenotype in specific environmental conditions. This list is generated from a fitness analysis under 21 different experimental conditions.

Usage

```
data(DudleyPhenoFull)
data(DudleyPhenoFull)
data(DudleyPhenoM)
data(DudleySign)
```

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Format

DudleyGenelist is a character vector of length 814 genes. DudleyPhenoFull is a dataframe of 814 genes by 23 elements. The column contains the yeast gene name. The 22 remaining columns are the experimental conditions (see details). The values obtained in the different condition are the fitness defect scores for the 814 genes sensitive to the experimental condition, as defined by Dudley et al (2005). DudleyPhenoM is a 814 by 22 incidence matrix with rownames corresponding to the genes names and columns to an experimental condition. This matrix contains a 1 in the (i,j) position if the i-th gene is sensitive to the experimental condition of the j-th column; it contains a 0 otherwise.

Yname Yeast systematic gene name

benomyl 15ug/ml benomyl, microtubule function

CaCl2 0.7M calcium chloride, divalent cation

CAD 55uM Cadmium, heavy metal

Caff 2mg/ml Caffeine

cyclohex 0.18ug/ml cycloheximide, protein synthesis

DTT unknown

EtOH YPD + 6% Ethanol

FeLim iron limited, nutrient limited condition

HU 11.4mg/ml Hydroxyurea, DNA replication and repair

HygroB 50ug/ml hygromycin B, aminoglycosides

lowPO4 Low phosphate, nutrient limited condition

MPA 20ug/ml mycophenolic acid, transcriptional elongation

NaCl 1.2M sodium chloride, general stress condition

Paraq 1mM paraquat, oxidative stress

pH3 Low pH, general stress condition

rap 0.1ug/ml rapamycin, protein synthesis

Sorb 1.2M sorbitol, general stress condition

UV 100J/m2 ultra-violet, DNA replication and repair

YPGal 2% galactose, carbon source

YPGly 3% glycerol, carbon source

YPLac 2% lactate, carbon source

YPRaff 2% raffinose, carbon source

DudleySign is a list of dataframe that summaryzes in which complexes the gene related to the phenotype are found, the size of the complexes and the associated p-value. This is the result of applying a Hypergeometric test (see CoHyperGParams-class for more details) and the complexStatus function.

Dudleyresult is a data.frame that summaryzes the number of sensitive genes per condition, how many of those genes are present in the ScISI interactome and the associated p-value. This is the result of applying a Hypergeometric test (see CoHyperGParams-class for more details) and the complexStatus function.

Author(s)

N. LeMeur

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Source

Dudley et al (2005), supplementary information: http://arep.med.harvard.edu/pheno/default. htm

References

Aim\'ee Marie Dudley, Daniel Maarten Janse, Amos Tanay, Ron Shamir and George McDonald Church. (2005). A global view of pleiotropy and phenotypically derived gene function in yeast. *Molecular Systems Biology* doi:10.1038/msb4100004

Examples

```
data(DudleyPhenoFull)
data(DudleyPhenoM)
```

getDescr

Get formatted annotation data

Description

Function to retrieve the annotation of multi-protein complexes or pathways via GO, MIPS or KEGG.

Usage

```
getDescr(x, database="GO.db")
```

Arguments

x Vector of multi-protein complexes or pathways IDs to be described

KEGG.db

Author(s)

N. LeMeur

```
xx < -getDescr(c("MIPS-220","MIPS-260.20","04111"),c("MIPS","KEGG.db"))
```

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getFDgene	Get fitness defect genes	

Description

Function to select genes that present a significant growth defect according to the condition(media) or generation time.

Usage

```
getFDgene(data,condition,cutoff,mode="generation",subset)
```

Arguments

data	List of fitness defect scores for genes tested at different experimental conditions.
condition	Dataframe of experimental conditions
cutoff	Numerical vector of length one or more, defining the threshold of 'significance' for the fitness defect score
mode	Character string defining the base of the selection either 'condition' (media) or 'generation' time, Default=generation.
subset	Numerical vector or list to which apply the different cutoffs.

Value

Reduced list of gene fitness scores per experimental condition according to the experimental condition or the generation time.

Author(s)

N. LeMeur

References

```
Giaever G. et al. (2002) Functional profiling of the Saccharomyces cerevisiae genome. Nature. 418(6896):387-91. PMID: 12140549
```

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GiaeverPheno

List of fitness defect score generated from Giaever et al 2002

Description

Giaever et al (2002) create a collection of gene-deletion mutants to determine genes that contribute to a particular phenotype in specific environmental conditions. This list is generated from a fitness analysis under six different experimental conditions.

Usage

```
data(GiaeverPheno)
data(GiaeverGene)
data(GiaeverExpCdt)
```

Format

GiaeverPheno is a list with 31 elements. The name of each element is a experimental condition (see details). The value of each element are the fitness defect scores for the genes sensitive to the experimental condition, as defined by Giaever et al (2002).

GiaeverGene Vector of the systematic gene names of the 5898 tested genes. Note that some updates have been made for the list to be consistent with *Saccharomyces Genome Database*.

GiaeverExpCdt is a 3 columns dataframe with fileID from which the data were extracted, the generation time (growth time) and the condition (media).

```
gen. generations
rep. replicate
ypg5a,ypg5b yeast/peptone/galactose 5 gen. rep. a and b ==> carbone source
ypg15a ypg15b yeast/peptone/galactose 15 gen. rep. a and b ==> carbone source
sorbitol5a sorbitol5b 1.5M Sorbitol 5 gen. rep. a and b ==> sugar, osmotic stress
sorbitol20a sorbitol15b 1.5M Sorbitol 20 and 15 gen. rep. a and b respectively==> sugar, osmotic
NaCl5a NaCl5b 1M NaCl 5 gen. rep. a and b ==> salt, osmotic stress
NaCl15a NaCl15b 1M NaCl 15 gen. rep. a and b ==> salt, osmotic stress
lysM5a lysM5b lysine minus 5 gen. rep. a and b ==> lack of required AA
thM5a threonine minus 5 gen. rep. a ==> lack of required AA
trpM5a trpM5b tritophanee minus 5 gen. rep. a ==> lack of required AA
minimalPlus5a minimalPlus5b minimal + histidine/leuvine/uracile 5 gen. rep. a and b
minimalPlus15a minimalPlus15b minimal + histidine/leuvine/uracile 15 gen. rep. a and b
minimalC5a minimalC5b minimal complete 5 gen. rep. a and b
nystatin5a nystatin5b Nystatin 5 gen. rep. a and b ==> antifungal drug
nystatin15a nystatin15b Nystatin 5 gen. rep. a and b ==> antifungal drug
pH8g5a pH8g5b pH 8 5 gen. rep. a and b ==> alkali stress
pH8g15a pH8g20b pH 8 15 and 20 gen. rep. a and b respectively ==> alkali stress
```

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Note: in their study they confound the 15 and 20 generations.

Giaeverresult is a data frame that summaryzes the number of sensitive genes per condition, how many of those genes are present in the ScISI interactome and the associated p-value. This is the result of applying a Hypergeometric test (see CoHyperGParams-class for more details) and the complexStatus function.

Author(s)

N. LeMeur

Source

Giaever et al (2002), supplementary information: http://genomics.lbl.gov/YeastFitnessData/websitefiles/cel_index.html Saccharomyces Genome Database (last update 03/17/06): http://www.yeastgenome.org/

References

Giaever G et al (2002) Functional profiling of the Saccharomyces cerevisiae genome. Nature. 418(6896):387-91. PMID: 12140549

Examples

```
data(GiaeverPheno)
data(GiaeverExpCdt)
data(GiaeverGene)
```

graphTheory

Graph theory to test associations between two or more relationships

Description

Graph theory approach associated with a permutation test to evaluate whether the number of associations is unexpectedly large.

Usage

graphTheory(genename, interactome, perm)

Arguments

genename A vector a gene names that are associated with a particular phenotype

interactome A binary matrix composed of genes (rows) and biological complexes (columns)

(see package ScISI)

perm Numeric, number of permutation run

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Details

We form two distinct graphs where the set of nodes are the proteins (genes) in the organism. In one graph we create edges between genes if the two genes are members of one, or more, protein complexes. In the second graph we create an edge between all genes that are associated to a particular phenotype. We then construct a third graph on the same node set, but where there is an edge in this graph only if there is an edge in both of the first to graphs. We count the number of edges in the third and test by permutation whether the number of edges is unexpectedly large.

Value

The returned value is a list with components:

edgeCount Number of associations observed between the genes that are linked to a particu-

lar phenotype and the given interactome.

edgeSimul Number of associations if the genes that are linked to a particular phenotype are

randomly distributed across the given interactome

p.value Returned p.value

Author(s)

R. Gentleman and N. LeMeur

References

Balasubramanian, R., LaFramboise, T., Scholtens, D., Gentleman, R. (2004) A graph-theoretic approach to testing associations between disparate sources of functional genomics data. Bioinformatics, 20(18), 3353-3362.

Examples

```
data(ScISI)
data(essglist)
ans <- graphTheory(names(essglist), ScISI, perm=3)
```

gtResult-class

A class for representing the result of a graphTheory test.

Description

A class for representing the result of a graphTheory test.

Slots

Pvalue: Object of class "numeric"

Observed: Return a "numeric" vector: the observed number of interactions between genes inducing a specific phenotype and each cellular organizational units

Expected: Return a matrix: the expected number of interactions between genes inducing a specific phenotype and each cellular organizational units

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Extends

```
Class "testResult", directly.
```

Methods

plot Graphical representation of the test result

Author(s)

N. LeMeur

See Also

testResult,deResult,plot

Examples

```
## apply a densityEstimate test
data(DudleyPhenoM)
data(ScISIC)

DudleyPhenoL <- apply(DudleyPhenoM,2,function(x) names(which(x==1)))
NaCl <- DudleyPhenoL[["NaCl"]]

perm <- 20
NaClGraph <- graphTheory(genename=NaCl, interactome=ScISIC, perm=perm)

## access results
slotNames(NaClGraph)
NaClGraph@Pvalue[1:5]

## use of the plot method
plot(NaClGraph)
```

HI

Data from Deutshbauer et al. (2005)

Description

Mechanisms of Haploinsufficiency revealed by Genome-Wide Profiling in Yeast (Deutshbauer et al.,2005)

Usage

```
data(HI)
```

Details

```
HI stands for haploinsufficient. The dataframe is composed of:

orf Yeast ORF, systematic name

gene Yeast common gene name of the corresponding ORF

go GO terms
```

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Source

http://www.sciencemag.org/cgi/data/303/5659/808/DC1/1

References

Deutschbauer AM, Jaramillo DF, Proctor M, Kumm J, Hillenmeyer ME, Davis RW, Nislow C, Giaever G. (2005) Mechanisms of haploinsufficiency revealed by genome-wide profiling in yeast. Genetics. 169(4):1915-25.

Examples

data(HI)

KastenmayerRaw

Data from Kastenmayer et al. 2006

Description

Kastenmayer et al. (2006) undertook the first functional studies of small open reading frames (sORFs) in any system, using the model eukaryote Saccharomyces cerevisiae. Phenotypic analyses of the new gene-deletion strains identified 22 sORFs required for haploid growth, growth at high temperature, growth in the presence of a non-fermentable carbon source, or growth in the presence of DNA damage and replication-arrest agents.

Usage

data(KastenmayerRaw)

Format

Kastenmayer is a 5 columns dataframe.

SYSTEMATIC Systematic name of the sORF.

COMMUN Commun name of the sORF.

Length Length of the small ORF sequence in number of amino acids.

Evidence Experimental source of the data

refHomology Bibliographical evidence of reported homology.

Kocollection Bibliographical evidence of reported homology.

ESSENTIAL Indicates if the sORF knockout is essential. A blank in this column indicates that the knockout is not-essential, if available.

GFPTAP "GFP" or "TAP" signifies that sORF was detected by the indicated technique. "both" indicates that sORF was detected both as a TAP-tagged and GFP-tagged protein. "None" indicates that sORF was not detected by either method. Empty field indicates that sORF was not tested

UPTAG Sequence of the upstream primer.

DOWNTAG Sequence of the downstream primer.

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Author(s)

N. LeMeur

Source

Kastenmayer et al. (2006), supplementary information: http://www.genomeresearch.org or ftp://genome-ftp.stanford.edu/pub/yeast/systematic results/phenotypes

References

Kastenmayer JP, Ni L, Chu A, Kitchen LE, Au WC, Yang H, Carter CD, Wheeler D, Davis RW, Boeke JD, Snyder MA, Basrai MA. (2006) Functional genomics of genes with small open reading frames (sORFs) in S. cerevisiae. Genome Res. 16(3):365-73. PMID: 16510898

Examples

```
data(KastenmayerRaw)
str(KastenmayerRaw)
```

KEGG2SCISI

Mapping between KEGG and ScISI

Description

Count the number of genes shared between a KEGG pathway and a protein complex from the ScISI interactome.

Usage

```
KEGG2SCISI(pw, pc, pcMat, pwMat)
```

Arguments

pw list of pathway names
pc list of complex names
pwMat pathway incidence matrix
pcMat complex incidence matrix

Value

matrix

Author(s)

N. LeMeur

See Also

ScISI KEGG

20 LesageRaw

Examples

```
data(ScISIC)
## Mapping from Yeast genes to KEGG pathways.

KeggMat <- PWAmat("org.Sc.sgd")

KEGG2SCISI(pw = colnames(KeggMat)[1:5], pc = colnames(ScISIC)[1:5], pwMat = KeggMat, pcMat = ScISIC)
```

LesageRaw

Data from Lesage et al. 2005

Description

Lesage et al. (2005) assembled a network of 316 interactions among 163 genes using deletion mutants in CHS1, CHS3, CHS4, CHS5, CHS6, CHS7 and BNI4 in a synthetic genetic array analysis.

Usage

```
data(LesageRaw)
```

Format

LesageRaw is a 5 column dataframe.

SYSTEMATIC Systematic gene names. NOTE: All mutants are isogenic to BY4741 (MATa his3\u0394 leu2\u0394 met15\u0394 ura3\u0394) except anp1\u0394 and mnn9\u0394 that are isogenic to BY4742 (MAT\u03b1 his3\u0394 leu2\u0394 lys2\u0394 ura3\u0394).

COMMUN Commun gene names.

CFW Mutants showing increased, decreased or wild type sensitivity to Calcofluor white are scored s, r, or wt, respectively.

ChitinLevel Chitin level (nmole GlcNAc/mg dry weight). Values are an average of at least three independent determinations. Values statistically higher and lower than wild type (p < 0.01) are highlighted in red and green, respectively.

ChitinLevel.SD Standard deviation of the average of at least three independent determinations of Chitin level.

Author(s)

N. LeMeur

Source

Lesage et al. (2005), supplementary information: http://www.biomedcentral.com/1471-2156/6/8/suppl/S2 or $ftp://genome-ftp.stanford.edu/pub/yeast/systematic_results/phenotypes$

References

Lesage G, Shapiro J, Specht CA, Sdicu AM, Menard P, Hussein S, Tong AH, Boone C, Bussey H. (2005) An interactional network of genes involved in chitin synthesis in Saccharomyces cerevisiae.BMC Genet.6(1):8. PMID: 15715908

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Examples

data(LesageRaw) str(LesageRaw)

OsterbergRaw

Data from Osterberg at al. 2006

Description

Osterberg at al. (2006) report growth phenotypes in yeast for a strain collection over-expression ~600 C-terminal tagged integral membrane proteins growth both under normal and three different stress conditions.

Usage

data(OsterbergRaw)

Format

OsterbergRaw is a 17 columns dataframe.

SYSTEMATIC Systematic gene names of the studied membrane protein

COMMUN Commun gene names of the studied membrane protein

TMHMM.C The topology predicted by TMHMM (TransMembrane prediction using Hidden Markov Models) using the experimentally assigned C-terminal location for the protein as a constraint. The topology is represented in the format Location of N-terminus TMhelices Location of C-terminus (i and o stand for in and out respectively)

WesternBlot Protein expression levels (arbitrary units), estimated from the band intensity and normalized to the internal standard on each Western blot.

Bands Proteins detected as two distinct bands with different molecular mass on the Western blot analysis. category 1 indicates that both bands were insensitive to Endo H digestion, 2 indicates the higher molecular mass was shifted down upon Endo H digestion on SDS/PAGE and one band was predominant compared to the other, and 3 indicates that a higher molecular mass band shifted down upon Endo H digestion on SDS/PAGE and both bands were equal intensity on Western blot

Toxicity Toxicity index from Spoko et al. (2006). The index varies between 1 and 5, where 1 means the strain is dead, and 5 indicates no difference in growth rate compared with the wild type strain.

- **sign.norm** Over-expression strains that show a significant (P<0.001) growth rate phenotype (LSCrate) in synthetic defined medium conditions (Warringer et al., 2003). An average of two replicates is given. Strains that do not show a significant difference in doubling time compared with the wild-type strain are indicated by 0.
- **all.norm** Phenotypes (significant or not) of over-expression strains (LSCrate) in synthetic defined medium conditions (Warringer et al., 2003). An average of two replicates is given.
- **sign.NaCl** Over-expression strains that show a significant (P<0.001) growth rate phenotype (LPI-rate) in NaCl. An average of two replicates is given. Strains that do not show a significant difference in doubling time compared with the wild-type strain under NACL stress are indicated by 0.

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all.NaCl.LSC Phenotypes (significant or not) of over-expression strains (LSCrate) in NaCl (Warringer et al., 2003). An average of two replicates is given.

- **all.NaCl.LPI** Phenotypes (significant or not) of over-expression strains (LPIrate) in NaCl (Warringer et al., 2003). An average of two replicates is given.
- **sign.caff** Over-expression strains that show a significant (P<0.001) growth rate phenotype (LPI-rate) in caffeine. An average of two replicates is given. Strains that do not show a significant difference in doubling time compared with the wild-type strain under caffeine stress are indicated by 0.
- **all.caff.LSC** Phenotypes (significant or not) of over-expression strains (LSCrate) in caffeine (Warringer et al., 2003). An average of two replicates is given
- **all.caff.LPI** Phenotypes (significant or not) of over-expression strains (LPIrate) in caffeine (Warringer et al., 2003). An average of two replicates is given.
- **sign.paraq.** Over-expression strains that show a significant (P<0.001) growth rate phenotype (LPI-rate) in paraquat. An average of two replicates is given. Strains that do not show a significant difference in doubling time compared with the wild-type strain under paraquat stress are indicated by 0.
- **all.paraq.LSC** Phenotypes (significant or not) of overexpression strains (LSCrate) in paraquat (Warringer et al., 2003). An average of two replicates is given.
- **all.paraq.LPI** Phenotypes (significant or not) of overexpression strains (LPIrate) in paraquat (Warringer et al., 2003). An average of two replicates is given.

Author(s)

N. LeMeur

Source

Osterberg et al (2006), supplementary information: $http://www.pnas.org/content/vol0/issue2006/images/data/0604078103/DC1/04078Table1.xls or ftp://genome-ftp.stanford.edu/pub/yeast/systematic_results/phenotypes$

References

Osterberg M, Kim H, Warringer J, Melen K, Blomberg A, von Heijne G. (2006) Phenotypic effects of membrane protein overexpression in Saccharomyces cerevisiae. PNAS. 103(30):11148-53. PMID: 16847257

Examples

 $\frac{\mathrm{data}(\mathrm{OsterbergRaw})}{\mathrm{str}(\mathrm{OsterbergRaw})}$

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overlap

Count the number of proteins shared by protein complexes

Description

Count the number of proteins shared by protein complexes

Usage

```
overlap(interactome)
```

Arguments

interactome Binary matrix composed of genes (rows) and biological complexes (columns)

(see package ScISI)

Value

The returned value is a data frame with components:

C1 Name of the first biological complex
 C2 Name of the second biological complex

nbSharedProt Number of proteins in common

Author(s)

N. LeMeur

See Also

ScISI

Examples

```
xx = cbind("a"=c(0,1,1,1),"b"=c(1,1,0,1))
overlap(xx)
```

plot

Graphical method to represent the result of the density or graph test.

Description

a plot method for deResult and gtResult objects.

Usage

```
\#\# S4 method for signature 'deResult' plot(x,...) \#\# S4 method for signature 'gtResult' plot(x,...)
```

24 ppiInteract

Arguments

x the deResult or gtResult object to plot.... general commands to be sent to plot.

Details

The plot generated from a deResult object is a set of density plots.

The plot generated from a gtResult object is a histogram.

Author(s)

N. LeMeur

See Also

ScISI

Examples

```
data(ScISI)
data(essglist)
essential <- names(essglist)
ScISI <- as.matrix(ScISI)
ratio<- densityEstimate(genename=essential,interactome=ScISI,perm=50)
plot(ratio)
```

ppiInteract

Test the association between AP-MS data and phenotype

Description

Test the association between AP-MS data and phenotype data via a graph and permutation model.

Usage

```
ppiInteract(genename, expGraph, bait, prey, perm=10)
```

Arguments

genename Genes associated to a phenotype

expGraph A graphNEL object (a direct graph instance of classgraph). The nodes are the

union of viable baits (VB) and viable prey (VP) of the experiment (see package

ScISI)

bait Proteins which was sampled as a bait in the binary relationship prey Proteins which was sampled as a prey in the binary relationship

perm Number of permutation

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Value

The returned value is a list:

Observed Values

Expected Expected values after each permutation

Author(s)

R. Gentleman and N. LeMeur

See Also

ScISI

Examples

```
\begin{aligned} & \text{data(ScISI)} \\ & \text{data(essglist)} \\ & \text{s1} <- \text{ppiInteract(names(essglist), Gavin2002BPGraph, viableBaits[[8]],} \\ & \text{viablePrey}[[8]], \text{ perm}{=}10) \end{aligned}
```

reduceM

Resize a matrix

Description

Resize a matrix to the number of rows commun to a vector.

Usage

```
reduceM(x, mat, threshold=0)
```

Arguments

x Character or numeric vector.

mat Matrix sharing rownames with the supplied vextor x.

threshold Threshold upon column. Only the columns with a colSums above the threshold

are kept.

Value

Resized matrix.

Author(s)

N. LeMeur

```
mat<-matrix(c(1:25),\,nrow=5,\,ncol=5,\,dimnames=list(c(LETTERS[1:5]),\,c(1:5))) xx <- LETTERS[c(2, 4, 5)] reduceM(xx, mat)
```

26 testResult-class

SGDphenoL

Saccharomyces Genome Database list of phenotypic data

Description

Saccharomyces Genome Database list of phenotypes and associated genes from several published experiments (last update 2006).

Usage

```
data(SGDphenoL)
```

Format

SGDphenoL is a list of phenotypes. Under each phenotype is listed the genes that potentially induce that phenotype. A binary matrix can be built from that list using the list2Matrix function from the Rintact package.

Author(s)

N. LeMeur

Source

SGD, supplementary information: http://www.yeastgenome.org/

Examples

data(SGDphenoL)

testResult-class

A virtual class for representing the result of a test.

Description

The testResult class is the virtual base class for all result objects of the densityEstimate and graph-Theory tests proposed in PCpheno.

Objects from the Class

A virtual Class: No objects may be created from it.

Slots

Observed: Return a "numeric" vector: the observed number of genes or interactions within each cellular organizational units

Expected: Return a numeric or a matrix: the expected number of genes or interactions within each cellular organizational units

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Methods

No methods defined with class "testResult" in the signature.

Author(s)

N. LeMeur

See Also

gtResult,deResult

truncName

Truncate character strings

Description

Truncate character strings

Usage

truncName(x,n)

Arguments

x Character string

n Maximum length (in characters)

Value

Character string

Author(s)

N. LeMeur

```
xx< "Anticonstitution
nelement is a family name" trunc
Name(xx,5)
```

28 YEASTOHNOLOG

YEASTOHNOLOG

List of ohnolog gene pairs from Byrne, K.P and Wolfe, K.H (2005)

Description

List of 551 paralogous *Saccharomyces cerevisiae* gene pairs formed by Whole Genome Duplication (WGD) or ohnolog pairs.

Usage

data(YEASTOHNOLOG)

Format

YEASTOHNOLOG is a dataframe of 551 paired genes. The first two columns are the ohnolog gene pairs (systematic gene names). The third column is an index (numeric) of the rate of sequence evolution. The last two columns define the chromosome location.

Details

Gene1 Gene2 Systematic gene names of the ohnolog pairs

Ka Coefficient that represents the extent of non-synonymous sequence divergence between each ohnolog pairs (Yang and Nielsen, 2000). The highest is the coefficient the fastest the 2 elements of a pair have diverged.

ChrG1 ChrG2 Chromosome location of the each element of a pair. Note that repeat of the same chromosome locations shared by a set of pairs define a block of duplication.

Author(s)

N. LeMeur

Source

Byrne,KP and Wolfe KH (2005), Table2 of supplementary information and Scerevisiae_genome.tab file, chromosome location, from YGOB http://wolfe.gen.tcd.ie/ygob/ (last update 03/20/06)

References

Byrne KP, Wolfe KH. (2005) The Yeast Gene Order Browser: combining curated homology and syntenic context reveals gene fate in polyploid species. Genome Res. 15(10):1456-61. PMID: 16169922

Examples

 $\begin{array}{l} \operatorname{data}(\operatorname{YEASTOHNOLOG}) \\ \operatorname{str}(\operatorname{YEASTOHNOLOG}) \end{array}$

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