# Package 'yeastExpData'

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The Teast Experimental Data	
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<b>Description</b> A collection of different sets of experimental data from yeast.	
biocViews ExperimentData, Genome, Saccharomyces_cerevisiae_Data	
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cicense GPL	
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Suggests Biobase, GO.db, RBGL, org.Sc.sgd.db	
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ccyclered

~~ data name/kind ... ~~

# **Description**

The data are 2885 yeast genes, common to a number of different experiments. The original data were reported by Cho et al. and were processed to

# Usage

data(ccyclered)

#### **Format**

A data frame with 2885 observations on the following 11 variables.

Cluster The cluster number the gene was assigned to.

**Distance** The distance from the cluster center?

Y.name The name of the gene, using standard yeast nomenclature.

SGDID The Stanford yeast genome database identifier for the gene.

**GENE** The common name for the gene.

**Chromsome** The chromosome the gene is located on.

**Start** The start of the gene, in bases, most likely from the 3' end.

**End** The end of the gene.

**Introns** The number of introns.

**Exons** The location of the exons.

**Description** A description of the gene.

### **Details**

Cho, et al. discuss the k means clustering of 2885 Saccharomyces genes into 30 clusters with measurements taken over two synchronized cell cycles.

#### Source

http://arep.med.harvard.edu/network\_discovery

# References

Cho, et al. (1998) A genome-wide transcriptional analysis of the mitotic cell cycle. Molecular Cell, 2, 65-73.

Tavazoie, et al. (1999) Systematic determination of genetic network architecture. Nature Genetics, 22, 281-285.

### **Examples**

data(ccyclered)

fcabundance 3

fcabundance

Yeast protein abundance

# **Description**

Abundance of yeast proteins measured using flow cytometry and GFP tagging

# Usage

```
data(fcabundance)
```

### **Format**

A data frame with 4159 observations on the following 6 variables.

```
yORF a factor denoting yeast Open Reading Frames with levels YAL001C, YAL002W, YAL005C, etc gene_name a factor denoting corresponding gene names, with levels 37164, AAC1, AAC3, etc YEPD.mean a numeric vector, giving average abundance in rich (YEPD) media YEPD.error a numeric vector, giving corresponding (standard?) error SD.mean a numeric vector, giving average abundance in minimal (SD) media SD.error a numeric vector, giving corresponding error
```

#### Source

http://www.nature.com/nature/journal/v441/n7095/extref/nature04785-s03.xls

# References

"Single-cell proteomic analysis of S. cerevisiae reveals the architecture of biological noise" (15 June 2006). John R. S. Newman, Sina Ghaemmaghami, Jan Ihmels, David K. Breslow, Matthew Noble, Joseph L. DeRisi and Jonathan S. Weissman Nature 441, 840-846

# **Examples**

```
data(fcabundance)
plot(YEPD.mean ~ SD.mean, fcabundance, log = "xy")
plot(SD.error ~ SD.mean, fcabundance, log = "x")
```

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gfp

Yeast GFP Fusion Data

### **Description**

This data frame contains data concerning the localization and abundance of various yeast proteins.

#### Usage

data(gfp)

#### **Format**

A data frame with 6234 observations on the following 33 variables.

orfid a numeric vector of identifiers

yORF a factor representing yeast ORF names, with levels YAL001C, YAL002W, etc. These are also the row names of the data frame.

gene\_name a factor representing corresponding yeast gene names, with levels AAC1, AAC3, etc.

GFP\_tagged a factor with levels not tagged and tagged, indicating whether or not the ORF was GFP tagged

GFP\_visualized a factor with levels not visualized and visualized, indicating whether or not GFP fluoresence was visualized

TAP\_visualized a factor with levels TAP visualized and not TAP visualized, indicating success of TAP tag

abundance a numeric vector, giving estimated abundance in units of molecules per cell

error a numeric vector of estimated errors in abundance for a subset of proteins, in the same units as abundance (see details below)

localization\_summary a factor with levels, ER, ER to Golgi, ER, ambiguous, ER, ambiguous, bud, etc. Summarizes the information contained in the subsequent columns.

The following columns indicate whether or not the protein was localized in the specific region of the cell. A protein can be localized in more than one region.

ambiguous a logical vector
mitochondrion a logical vector
vacuole a logical vector
spindle\_pole a logical vector
cell\_periphery a logical vector
punctate\_composite a logical vector
vacuolar\_membrane a logical vector
ER a logical vector
nuclear\_periphery a logical vector

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endosome a logical vector
bud\_neck a logical vector
microtubule a logical vector
Golgi a logical vector
late\_Golgi a logical vector
peroxisome a logical vector
actin a logical vector
nucleolus a logical vector
cytoplasm a logical vector
ER\_to\_Golgi a logical vector
early\_Golgi a logical vector
lipid\_particle a logical vector
nucleus a logical vector
bud a logical vector

Explanation for missing abundance values are given by

missingAbundance a factor with levels low signal, not visualized and technical problem

#### **Details**

The information on abundance is available in three columns. abundance gives (where available) absolute protein abundances determined by quantitative Western blot analysis of TAP-tagged strains. Abundances that have a non-NA error value were done in triplicate with serial dilutions of purified TAP-tagged standards included in each gel, which substantially reduces the measurement error. In addition, for these strains, the tagged genes were confirmed to rescue the loss of function phenotype of the corresponding deletion strain. For rows where abundance is missing (NA), the missingAbundance column gives the reason. Possible reasons are:

"not visualized" Either the tagging was unsuccessful or no signal was detected.

"low signal" The tagging was successful, but the signal was not sufficiently high above background to permit accurate quantitation (about 50 molecules/cell).

"technical problem" The protein was detectable but could not be quantitated because it did not migrate as a single band or comigrated with the internal standards in the gel.

Replicate analysis for a subset of tagged strains found a linear correlation coefficient of R = 0.94, with the pairs of proteins having a median variation of a factor of 2.0. This error analysis does not account for potential alterations in the endogenous levels of the proteins caused by the the fused tag, which may be particularly disruptive for small proteins.

#### Source

The data were obtained from http://yeastgfp.ucsf.edu/, which contains a lot more information as well as raw image data. This data frame was specifically generated from http://yeastgfp.ucsf.edu/allorfData.txt

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### References

For the Localization data: Huh, et al., Nature 425, 686-691 (2003) – http://www.ncbi.nlm.nih.gov/entrez/query.fcgi?cmd=Retrieve&db=PubMed&list\_uids=14562095&dopt=Abstract

For the Protein abundance data: Ghaemmaghami, et al., Nature 425, 737-741 (2003) — http://www.ncbi.nlm.nih.gov/entrez/query.fcgi?cmd=Retrieve&db=PubMed&list\_uids=14562106&dopt=Abstract

# **Examples**

litG

Literature and Y2H interaction graphs

### **Description**

These two data objects represent protein/gene interactions as reported in Ge et al.

# Usage

#### **Format**

The data are stored as instances of the graphNEL class. Each has 2885 nodes, named using yeast standard names. Interactions either represent literature curated interactions, or Y2H interactions.

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# **Details**

The data were reported and used in Correlation between transcriptome and interactome mapping data from Saccharomyces cerevisiae, Nature Genetics, 2001, H. Ge and Z. Liu and G. M. Church and M. Vidal.

See the package vignette for more details.

# **Examples**

data(litG)
data(y2hG)

nPdist

Counts from a node permutation experiment.

# **Description**

In comparing the yeast cell cycle data to the protein-protein interaction data a node permutation distribution is suggested. The output here are the counts of the number of common edges for 500 permutations.

# Usage

data(nPdist)

# **Format**

Five hundred counts of nodes in the intersection of two graphs.

# **Details**

The seed was 123 and we called the function nodePerm with litG and cg1 as arguments. See the vignette in this package for the explicit computations.

# **Examples**

data(nPdist)

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proteinProperties

Properties of Yeast proteins

## **Description**

A data frame which details 33 properties of proteins in the Yeast Genome

## Usage

data(proteinProperties)

#### **Format**

A data frame with 6718 observations on the following 33 variables.

yORF a factor representing yeast ORF names, with levels Q0010, Q0017, etc.

SGDID a factor representing SGD IDs

molwt a numeric vector giving Molecular Weight in Daltons

pi a numeric vector denoting the theoretical isoelectric point(pI), the pH at which the protein carries no net charge

cai a numeric vector denoting Codon Adaptation Index

length a numeric vector denoting length of the protein (number of amino acids)

nterm a factor representing N Term Sequence with levels MAAACIC MAAAPWY, etc.

cterm a factor representing N Term Sequence with levels AAAAMLL AAADKKT, etc.

codonBias a numeric vector denoting Codon Bias

The next set of columns, designated by amino acids, is the number of times that particular residue appears in the protein sequence. For example, if the ALA column is 2, then the protein contains 2 alanines. These columns (should) add up to the length column.

ALA a numeric vector

ARG a numeric vector

ASN a numeric vector

ASP a numeric vector

CYS a numeric vector

GLN a numeric vector

GLU a numeric vector

GLY a numeric vector

HIS a numeric vector

ILE a numeric vector

LEU a numeric vector

LYS a numeric vector

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```
MET a numeric vector
PHE a numeric vector
PRO a numeric vector
SER a numeric vector
THR a numeric vector
TRP a numeric vector
TYR a numeric vector
VAL a numeric vector
The remaining columns are:
```

fop FOP score, a numeric vector, denoting Frequency of Optimal Codons

gravy Gravy score, a numeric vector denoting Hydropathicity of Protein

aromaticity Aromaticity score, a numeric vector denoting Frequency of aromatic amino acids: Phe, Tyr, Trp

type Feature type, a factor with levels ORF | Dubious ORF | Uncharacterized ORF | Verified ORF | Verified | silenced\_gen pseudogene transposable\_element\_gene

#### **Details**

This data frame is downloaded directly from SGD. It contains 33 characteristics for 6714 open reading frames (ORFS). From the SGD README:

"Contains basic protein information about each ORF in SGD. This file does not include information on deleted or merged ORFs. Note, however, that it includes ORFs of all other classifications (Verified, Uncharacterized, and Dubious)."

For more details see http://www.yeastgenome.org/help/protein\_page.html.

#### Source

ftp://genome-ftp.stanford.edu/pub/yeast/protein\_info/protein\_properties.tab. This file is updated weekly (Saturday). The version used here was downloaded on 2009-12-03.

### **Examples**

```
data(proteinProperties)
pairs(proteinProperties[, c("molwt", "pi", "cai", "gravy", "aromaticity")],
     pch = ".", col = as.numeric(proteinProperties$type))
```

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