# Systematic Annotation

#### Mark Voorhies

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- RTFM
- PNAS 95:14863

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Three directed acyclic graphs (aspects):

- Biological Process
- Molecular <u>Function</u>
- Subcellular Component

## The Gene Ontology



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### The Gene Ontology



# The AmiGO browser



Tools

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AmiGO browser

OBO-Edit ontology editor

Ontology downloads

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representation of gene and gene product attributes across species and databases. The project provides a controlled vocabulary of terms, for describing gene product characteristics and gene product annotation data from GO Consortium members, as well as tools to access and process this data. Read more about the Gene Ontology...

	Annotation downloads
Search the Cone Ontology Database	Database downloads
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	GO FAQ
Search for genes, proteins or GO terms using AmiGO:	GO on SourceForge 🦉
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	Finding updates
AmiGO is the official GO browser and search engine. Browse the Gene Ontology with AmiGO.	G0 newsdesk 🛛 🗟 🖕
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- How might we annotate genes with GO terms?
- How do we calculate the significance of the GO terms associated with a particular group of genes?

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- By sequence homology (e.g., BLAST)
- By domain homology (e.g., InterProScan)
- Mapping from an annotated relative (e.g., INPARANOID)
- Human curation of the literature (e.g., SGD)

# Associating GO terms: Evidence codes

#### Experimental

- EXP: Inferred from Experiment
- IDA: Inferred from Direct Assay
- IPI: Inferred from Physical Interaction
- IMP: Inferred from Mutant Phenotype
- IGI: Inferred from Genetic Interaction
- IEP: Inferred from Expression Pattern

#### Computational Analysis

- ISS: Inferred from Sequence or Structural Similarity
- ISO: Inferred from Sequence Orthology
- ISA: Inferred from Sequence Alignment
- ISM: Inferred from Sequence Model
- IGC: Inferred from Genomic Context
- RCA: inferred from Reviewed Computational Analysis

#### Author Statement

- TAS: Traceable Author Statement
- NAS: Non-traceable Author Statement
- Curator Statement Evidence Codes
- IC: Inferred by Curator
- ND: No biological Data available

#### Automatically-assigned

• IEA: Inferred from Electronic Annotation

#### Obsolete

NR: Not Recorded

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$$p_{u,n} = \left(1 - \frac{p_m}{N_g}\right)^n \tag{4}$$

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This is also the expected genome coverage.



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What about exactly *k* events? Binomial distribution:

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What if there is more than one type of event? Multinomial distribution:

$$p_{k_1,k_2,...,n} = \frac{n!}{\prod k_i!} \prod p_i^{k_i}$$
 (8)

The binomial distribution assumes that event probabilities are constant:

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$$p_{k,m,n} = \frac{\binom{m}{k}\binom{N-m}{n-k}}{\binom{N}{n}}$$
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More than one *disjoint* type of label:

$$p_{k_1,k_2,...,m_1,m_2,...,n} = \frac{\prod {\binom{m_i}{k_i}}}{{\binom{N}{n}}}$$
 (11)

## Extracting gene lists from JavaTreeView

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						YHL016C	DUR3	TRANSPORT	UREA PERMEASE
						YFL053W	DAK2	CARBOHYDRATE ME	TABOLISM; DIHYDRO
						YBR294W YDR439W	LRS4	TRANSPORT TRANSCRIPTION/R	SULFATE PERMEAS
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	Part of					YGL192W	IME4 RPM2	MEIOSIS TRNA PROCESSING	MITOCHO RNASE E
						YMR056C	AAC1	TRANSPORT	MITOCHONDRIAL /
						YPL119C	DBP1	MRNA PROCESSING	RNA HELICASE
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إلاح						YIL170W YJL219W	HXT12 HXT9	TRANSPORT	HEXOSE PERMEASE HEXOSE PERMEASE
						YOL156W	HXT11 PTC1	TRANSPORT	HEXOSE PERMEASE
						YGL163C	RAD54	DNA REPAIR	DNA - DEPENDENT
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	in and					YGL203C YLR439W	KEX1 CAR2	SECRETION ARGININE METABO	CARBOXYPEPTIDAS LISM ORNITHINE
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Mark Voorhies Systematic

Systematic Annotation

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## The SGD GO Slim Mapper



#### SGD Gene Ontology Slim Mapper

The GO Slim Mapper maps annotations of a group of genes to more general terms and/or bins them into broad categories, ie. GO Slim terms,

Three GO Slim sets are available at SGD:

- 1. Macromolecular complex terms: protein complex terms from the Cellular Component ontology
- 2. Super GO-Slim: very broad, high level GO terms
- 3. Yeast GO-Slim: high level GO terms that represent the major biological processes, functions, and cellular components in S. cerevisiae

To find significant shared GO terms, or parents of those GO terms, used to describe the genes in your list, use the GO Term Finder.

Step 1: Choose Gene/ORF names		
Either Enter Gene/ORF names (separated by a return or space) YQL1550 YQL022C YQL023C	a OR	Upload a file of Gene/ORF names: (.txt or .tab format) Browse
Step 2: Choose GO SLIM Terms(s) by choosing a GO Set Ferms from the selected GO Set will be automatically entered in the	box in S	itep 3
Yeast GO-Slim: Process		
Step 3: Refine your list of GO Slim Terms		
GO Slim Terms		You must choose at least one term from the list
Done		

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#### Multiple Hypothesis Testing



http://xkcd.com/882/

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# Alternatives to Hierarchical Clustering

• GORDER and pre-clustering by SOM

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- Pre-calling number of clusters: k-means and k-medians

- GORDER and pre-clustering by SOM
- Pre-calling number of clusters: k-means and k-medians
- Principal Component Analysis (PCA)

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