NGS2 course

Making Sense of Gene Lists

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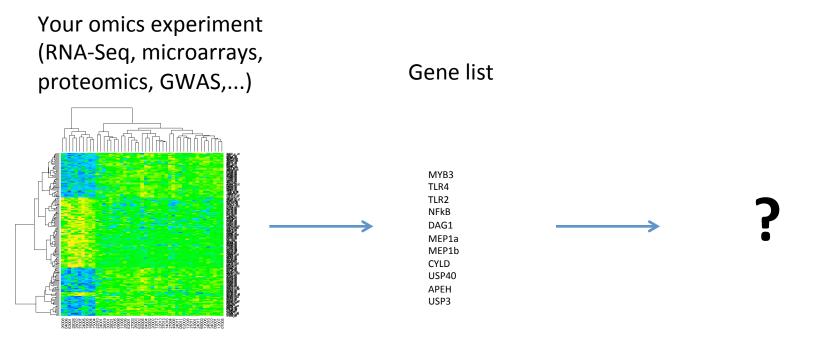




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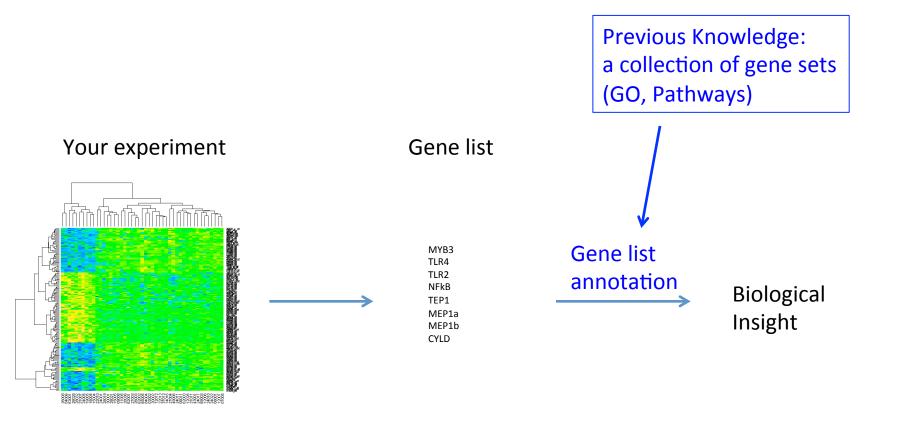
Gene List Annotation

- You performed a genomic experiment and obtained a gene list
- Who wants to work through a list of hundreds of genes?
- What's next?

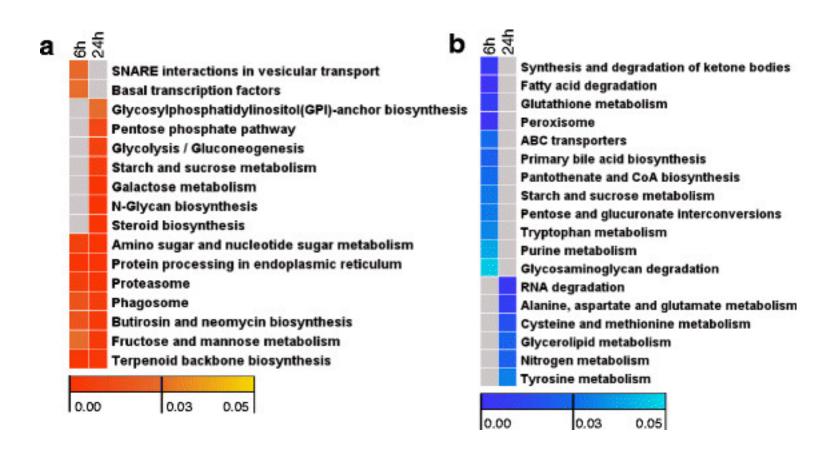


Gene List Annotation

- We test whether the differentially expressed genes in our experiment are enriched in some predefined gene lists.
- Based on previous knowledge



Gene List Annotation



Obtaining Biological Insight

- to summarize gene lists
- to help and speed up the interpretation of an experiment
- to gain mechanical insight
- to find regulated processes/pathways
- to find involved regulatory elements (TF, miRNA)
- to identify new members of a pathway
- to find similar experiments
- •

Analysis based on gene lists is expected to be more robust and reproducible than single-gene analysis.

Enrichment Analysis

Over-Representation Analysis

- hypergeometric aka Fisher's exact test
- input: 4 counts
- we need to set a cut-off a priori
- different results at different thresholds!

| Gene Set | Enrichment | Analysis | (GSEA) |
|-----------------|-------------------|-----------------|--------|
|-----------------|-------------------|-----------------|--------|

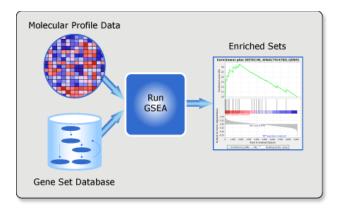
Subramanian et al. (2005) PNAS and many follow-up papers

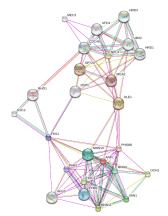
- bypasses the need for a cut-off
- input: list of all measured genes ranked by some statistics / effect size
- weak but consistent regulation of several members of a gene set can be detected

Network Analysis

- covers also the less well understood portion of gene interactions
- often inferred from co-expression data
- example: STRING (http://www.string-db.org/)
- combines info from co-expression, co-citation, PPI,

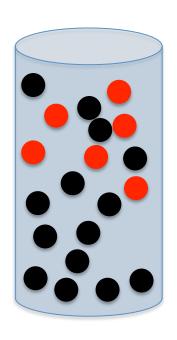
| 8 | 12 |
|---|------|
| 2 | 2412 |



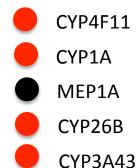


Over-Representation Analysis

5000 black and 10 red balls in an urn each ball represents 1 gene
10 red balls ("Cytochromes")



Our list of differentially expr. genes: 4/5 balls are red

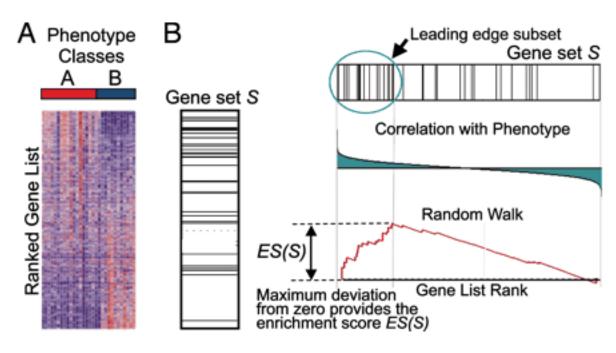


What is the probability? 2x2 contigency table

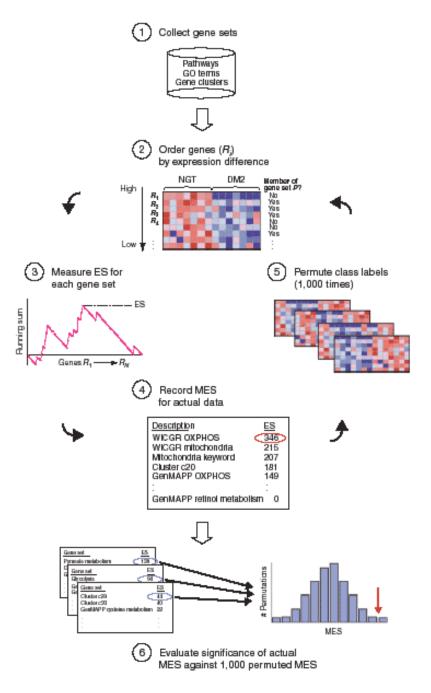
| | Selected | Not |
|-----------------|----------|------|
| in category | 4 | 6 |
| not in category | 1 | 4989 |

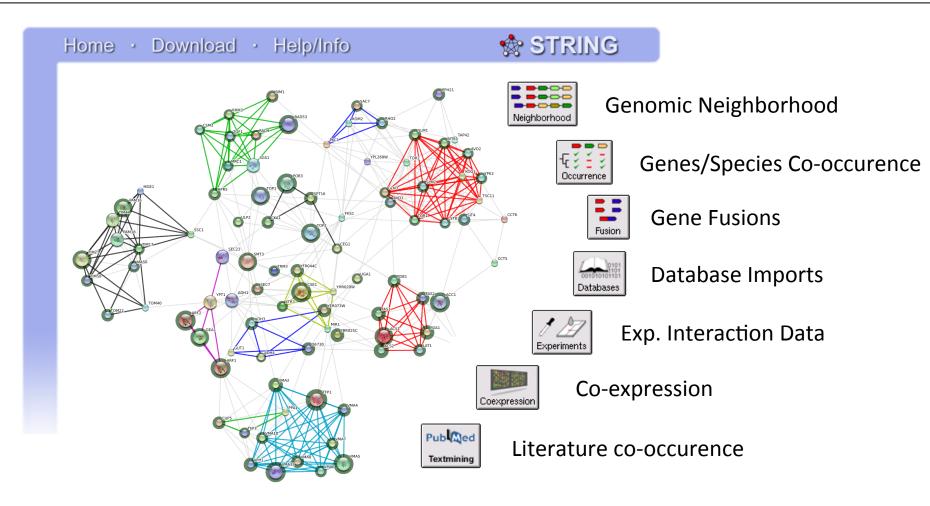
one-sided Fisher's exact test p-value = 4.03e-11

Gene Set Enrichment Analysis



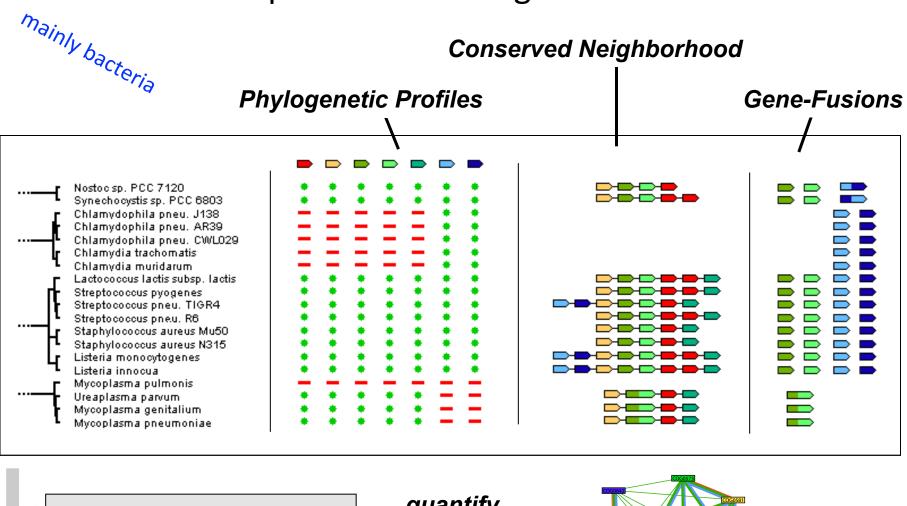
Subramanian et al. (2005) PNAS



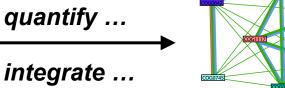


- functional association networks (physical or functional interactions)
- focus on useability and speed
- integrated scoring scheme (each interaction has confidence score)
- information transfer between species (>2000 species: Animals, Bact, Plants,...)

Interaction prediction from genome information

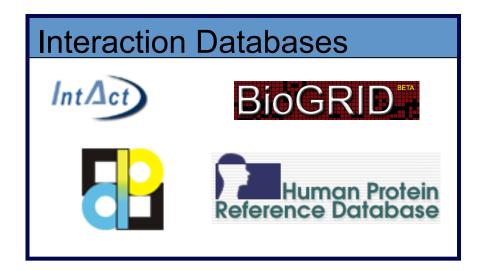


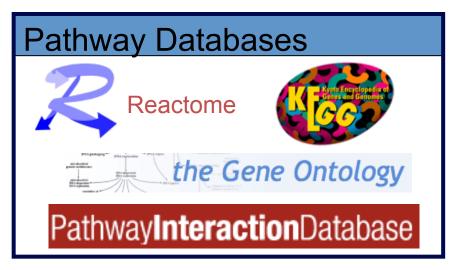
"genomic context"

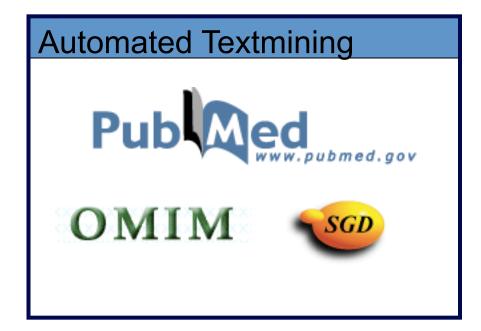


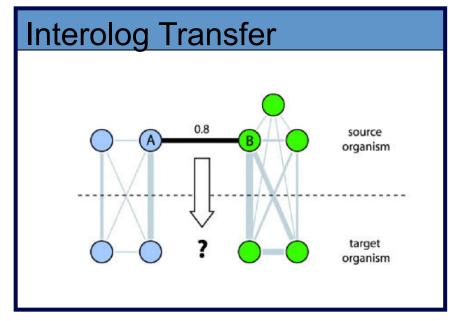
networks

Other Interaction Sources

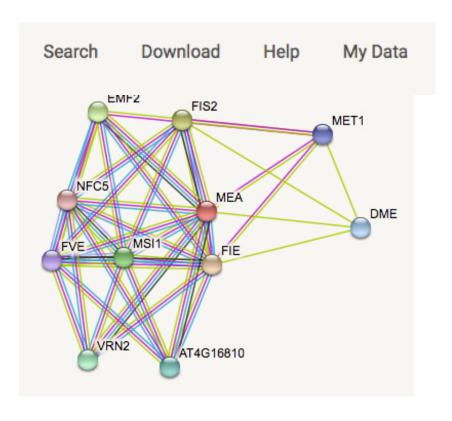


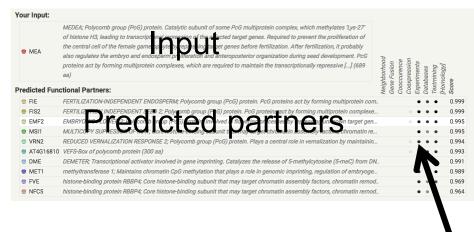






Output





Experiments

Textmining



✓ Neighborhood Gene Fusion ✓ Co-occurrence **Switch** On/off channel

✓ Databases

Clickable

evidence

✓ Co-expression

STRING

- can do more than gene list annotation:
 - Predicting gene function
 - Identifying candidates for an unknown enzyme in a pathway
 - Identifying new member genes of a biological process
 - Finding relevant literature
- ID mapper engine understands a large number of gene formats
- STRING performs well compared with single-species databases
- R package to access STRING functionality from R
- available for download

Annotation Sources

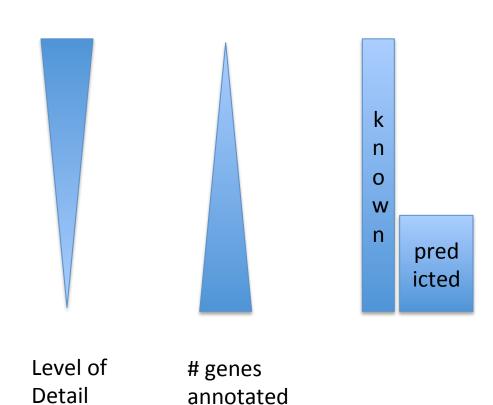
Pathways

KEGG, Reactome, BioCyc, ...

Gene Ontology (GO)

Gene/Protein Networks

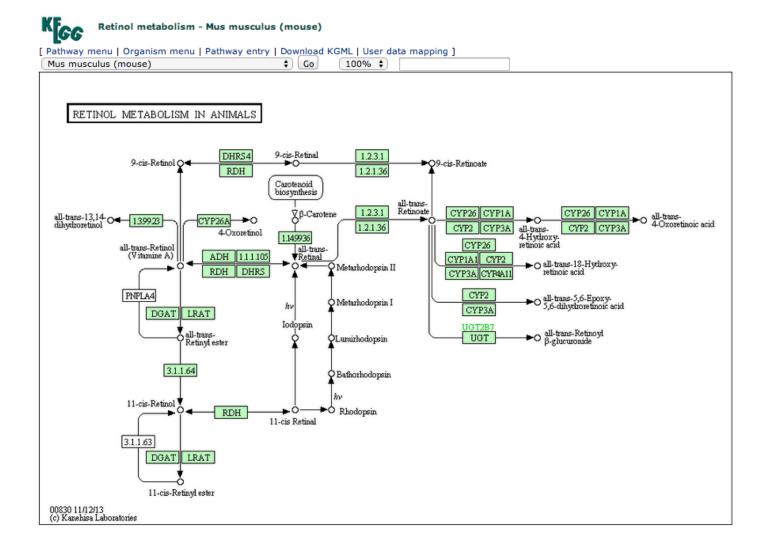
e.g. STRING



Pathways

- pathway maps (aka reaction networks / wiring diagrams) represent experimental knowledge on metabolism and various other functions of the cell and the organism
- manually curated
- the main databases are KEGG and Reactome
- KEGG is free to use over the web but file download requires subscription
- KEGG covers >3'800 species (Archae, Bacteria, Plants, Animals) and Reactome covers 20 species (mostly mammals + fly + plants + E.coli) as of May 2015.

Example KEGG Pathway



Gene Ontology

Gene Ontology (GO)

http://www.geneontology.org/

- describes how gene products behave in a cellular context (BP, MF, C)
- controlled vocabulary of terms
- transparent (sources)
- manually curated lists for model species
- transfer to orthologs in other species (inferred annotation)

Example

murine ADAM10

Molecular function

GO:0008237 metallopeptidase activity

GO:0042169 SH2 domain binding

. .

Biological Process

GO:0007220 Notch receptor processing

GO:0001701 in utero embryonic development

GO:0008284 positive regulation of cell proliferation

..

Cellular Compartment

GO:0005794 Golgi apparatus

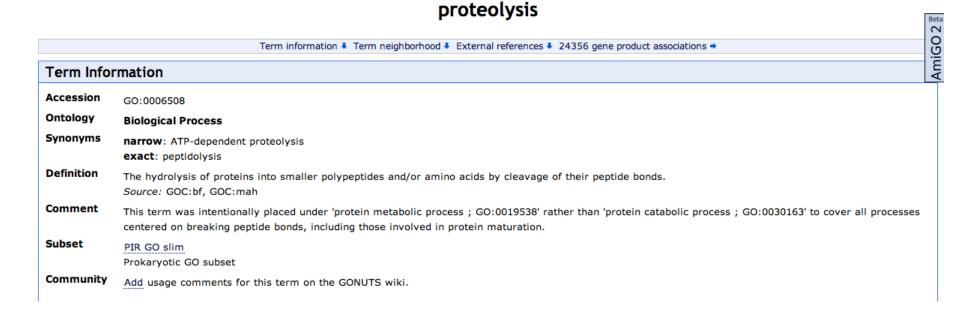
GO:0009986 cell surface

• •

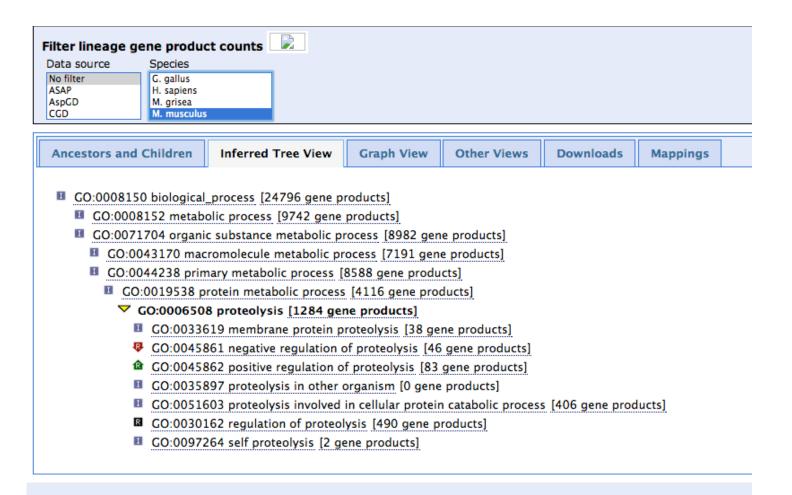
Lookup of GO terms

AmiGO http://amigo.geneontology.org

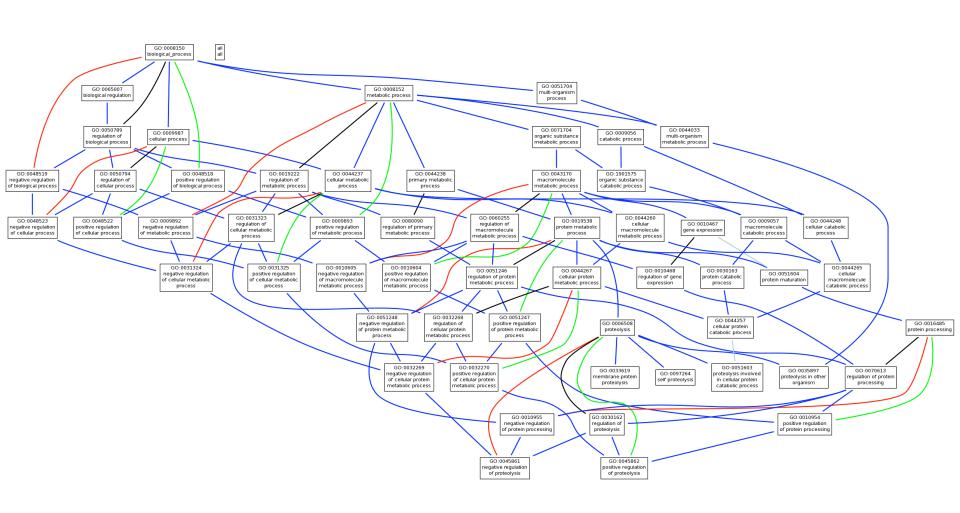




GO Table View GO:0006508 Proteolysis



Graphical View GO:0006508 Proteolysis



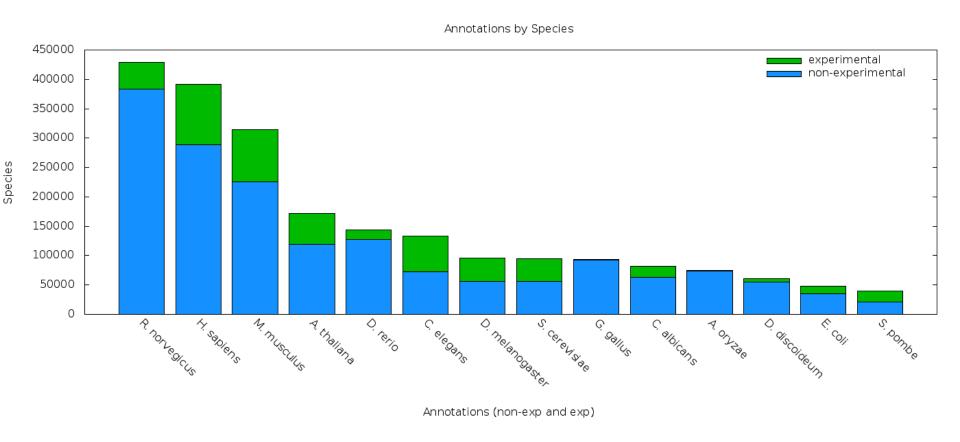
Ancestors and Children

AmiGO http://amigo.geneontology.org

| Ancestors and Children | Inferred Tree View | Graph View | Other Views | Downloads | Mappings | | | |
|--|----------------------------|-------------|---------------------|--------------------|------------------|------------------------|--------|-------------|
| | | | | | | | | |
| | | | Ancestors of pr | oteolysis (GO:0006 | 508) | | | |
| s | subject ¢ | | relation \$ | | o | bject ¢ | ann | otations ¢ |
| proteolysis | | ■ is | _a (inferred) | biological_p | ocess (GO:000 | 8150) | 665024 | |
| proteolysis | | ■ is | _a (inferred) | metabolic pr | ocess (GO:0008 | 8152) | 368913 | |
| proteolysis | | ■ is | _a (inferred) | organic subs | tance metaboli | c process (GO:0071704) | 300256 | |
| proteolysis | | ■ is | _a (inferred) | macromolec | ule metabolic p | rocess (GO:0043170) | 202070 | |
| proteolysis | | ■ is | _a (inferred) | primary met | abolic process (| (GO:0044238) | 277534 | |
| proteolysis | | ■ is | _a | protein meta | bolic process (| GO:0019538) | 105597 | |
| Children of proteolysis (GO:0006508) | | | | | | | | |
| s | subject ø | | relation ¢ | | o | bject ¢ | ann | otations \$ |
| membrane protein proteoly | ysis (GO:0033619) | Ⅱ is | _a | proteolysis | | | 387 | |
| negative regulation of prot | teolysis (GO:0045861) | ₽ ne | egatively_regulate: | s proteolysis | | | 502 | |
| positive regulation of prote | eolysis (GO:0045862) | ⊉ p | ositively_regulates | proteolysis | | | 696 | |
| proteolysis in other organi | sm (GO:0035897) | ■ is | _a | proteolysis | | | 83 | |
| proteolysis involved in cell (GO:0051603) | ular protein catabolic pro | ocess I is | _a | proteolysis | | | 8312 | |
| regulation of proteolysis (C | GO:0030162) | ⊠ r€ | egulates | proteolysis | | | 4093 | |
| self proteolysis (GO:00972 | .64) | ■ is | _a | proteolysis | | | 38 | |
| | | | | | | | | |

GO statistics

Even in model organisms only a minority of genes has experimental GO annotation



http://beta.geneontology.org/user-story-tags/everybody?page=5

False Discovery Rate (FDR)

Significance (alpha) level: probability of rejecting the null hypothesis given that it is true

Therefore at 5% significance level: for 100 tests where all null hypotheses are true, the expected number of incorrect rejections is 5

| tests | incorrect rejections |
|--------|-------------------------|
| 100 | 5 |
| 10,000 | 500 |

Multiple Testing Correction

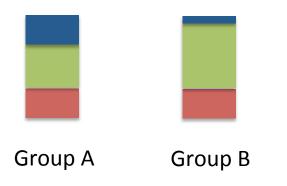
- Bonferroni
- False Discovery Rate (FDR): If we find 100 genes significantly differentially expressed at a 5% FDR, we expect at most 5 false discoveries in the list.

Experimental design

Interpretability depends mostly on appropriate experimental design!

Randomize samples/treatments across lanes / flow cells

Multiple tissues/cell types/stages pooled in a sample -> complex and difficult to understand the ongoing processes (e.g. observed changes can simply be due to changes in relative abundance of different cell types independent of regulation)



Blood example during pregnancy

Summary

- Gene list annotation with Pathways and Gene Ontology can help to obtain biological insight.
- 3 main methods: 1. Over-Representation Analysis, 2. Gene Set Enrichment Analysis (GSEA), 3. Network Analysis
- Biological interpretation requires broad knowledge of physiology & biochemistry and is often the most difficult and time-consuming step of an experiment.
- Even experts can usually not make sense of all the significantly enriched processes/pathways in well understood biological systems.
- Good experiments start with good experimental design! Think of possible confounders

Fastest way to lose innovative finding is pathway analysis! Lose top new significant genes with no pathway info!

Atul Butte

URLs & Tips

Main general Annotation Sources

- Gene Ontology (http://www.geneontology.org/)
 - AmiGO: http://amigo.geneontology.org
 - QuickGO: http://www.ebi.ac.uk/QuickGO/
 - Compilation of GO Tools: http://www.geneontology.org/GO.tools.shtml
- KEGG (http://www.genome.jp/kegg)
- Reactome (http://www.reactome.org)

- Most pathway databases offer also tools to colorize genes of interest on pathways
- Pathway analysis can also be done in R/bioconductor, see
 http://www.bioconductor.org/packages/release/BiocViews.html#___Pathways