

The background of the slide features a cluster of colorful, semi-transparent spheres in shades of purple, blue, yellow, and orange, arranged in a way that suggests a molecular or cellular structure. The spheres are of varying sizes and are set against a dark, teal-colored background.

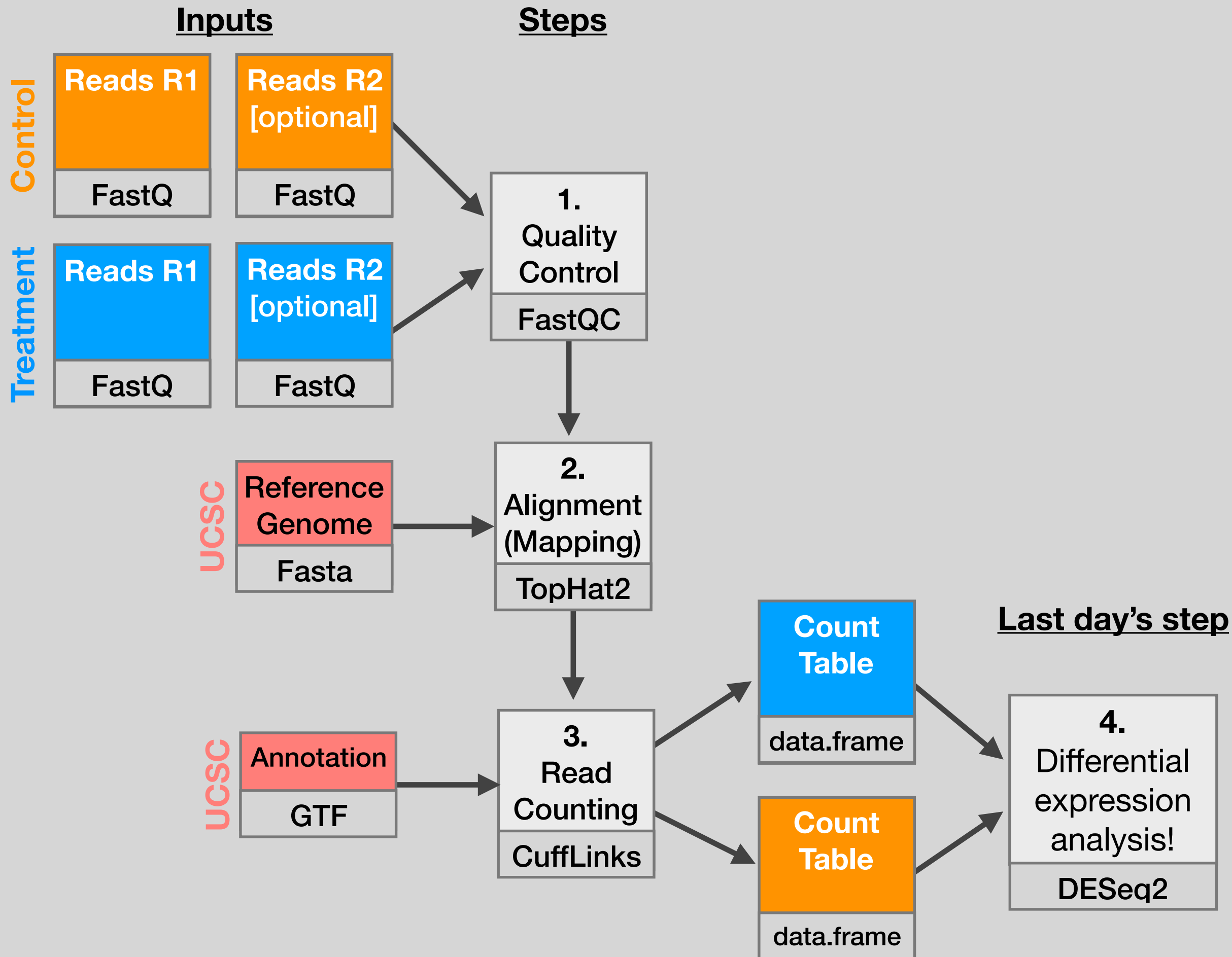
BGGN 213

Pathway Analysis and the Interpretation of Gene Lists

Lecture 15

Barry Grant
UC San Diego

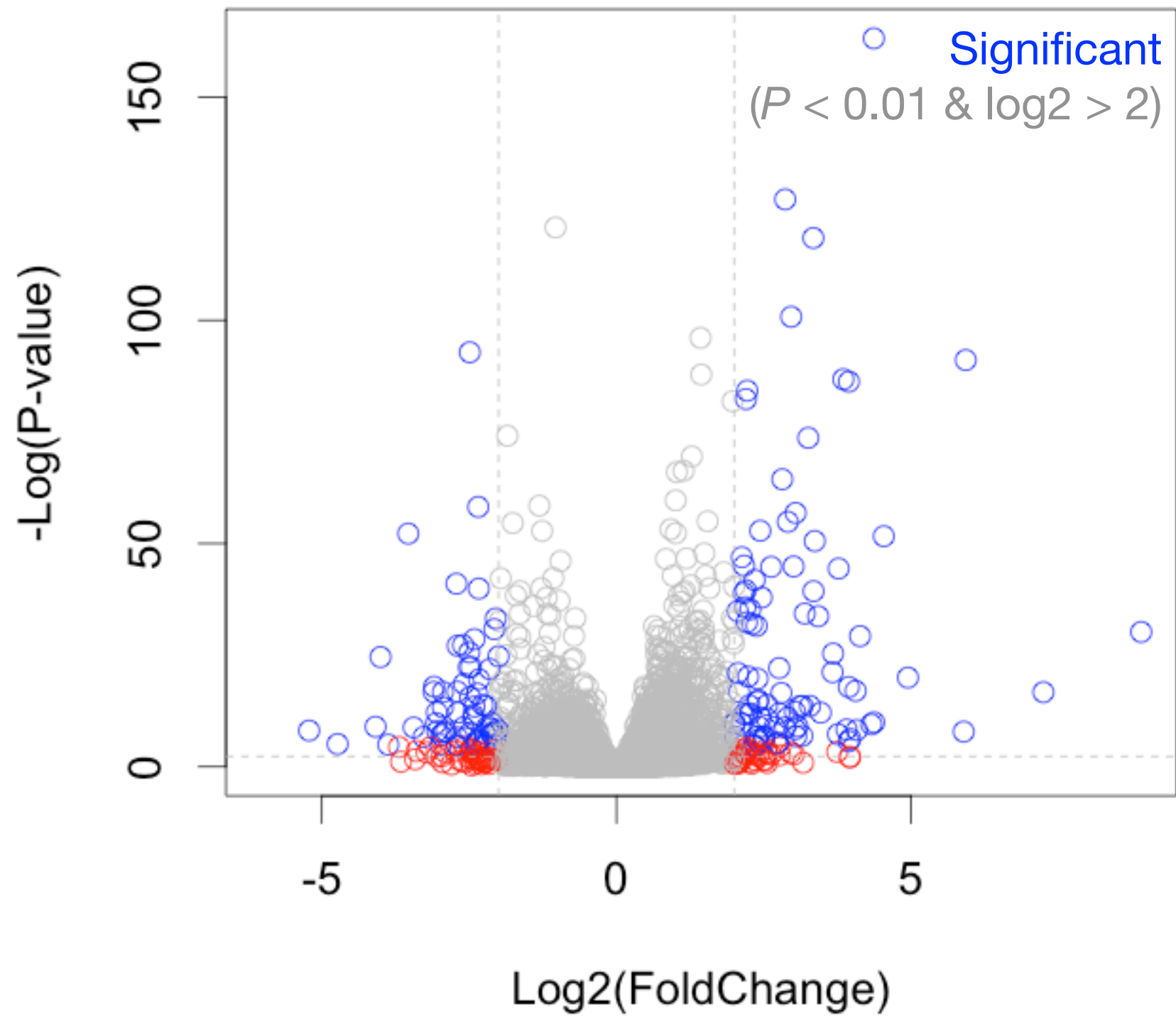
<http://thegrantlab.org/bggn213>



X	baseMean	log2FoldChange	lfcSE	stat	pvalue	padj	symbol
ENSG00000152583	954.77093	4.3683590	0.23713648	18.421286	8.867079e-76	1.342919e-71	SPARCL1
ENSG00000179094	743.25269	2.8638885	0.17555825	16.313039	7.972621e-60	6.037267e-56	PER1
ENSG00000116584	2277.91345	-1.0347000	0.06505273	-15.905557	5.798513e-57	2.927283e-53	ARHGEF2
ENSG00000189221	2383.75371	3.3415441	0.21241508	15.731200	9.244206e-56	3.500088e-52	MAOA
ENSG00000120129	3440.70375	2.9652108	0.20370277	14.556557	5.306416e-48	1.607313e-44	DUSP1
ENSG00000148175	13493.92037	1.4271683	0.10036663	14.219550	6.929711e-46	1.749175e-42	STOM
ENSG00000178695	2685.40974	-2.4890689	0.17806407	-13.978501	2.108817e-44	4.562576e-41	KCTD12
ENSG00000109906	439.54152	5.9275950	0.42819442	13.843233	1.397758e-43	2.646131e-40	ZBTB16
ENSG00000134686	2933.64246	1.4394898	0.10582729	13.602255	3.882769e-42	6.533838e-39	PHC2
ENSG00000101347	14134.99177	3.8504143	0.28490701	13.514635	1.281894e-41	1.941428e-38	SAMHD1
ENSG00000096060	2630.23049	3.9450524	0.29291821	13.468102	2.409807e-41	3.317866e-38	FKBP5
ENSG00000166741	7542.25287	2.2195906	0.16673544	13.312050	1.970000e-40	2.486304e-37	NNMT
ENSG00000125148	3695.87946	2.1985636	0.16700546	13.164621	1.402400e-39	1.633797e-36	MT2A
ENSG00000162614	5646.18314	1.9711402	0.15020631	13.122885	2.434854e-39	2.633990e-36	NEXN
ENSG00000106976	989.04683	-1.8501713	0.14778657	-12.519211	5.861471e-36	5.918132e-33	DNM1
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ENSG00000177666	2639.57020	1.1399947	0.09606884	11.866436	1.768422e-32	1.487930e-29	PNPLA2
ENSG00000164125	7257.00808	1.0248523	0.08657600	11.837603	2.494830e-32	1.988642e-29	FAM198B
ENSG00000198624	2020.04495	2.8141014	0.24063429	11.694515	1.359615e-31	1.029569e-28	CCDC69
ENSG00000123562	5008.55294	1.0045453	0.08901501	11.285123	1.554241e-29	1.120904e-26	MORF4L2
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ENSG00000196517	241.91536	-2.3456877	0.21047366	-11.144804	7.591120e-29	4.998588e-26	SLC6A9
ENSG00000135821	19973.40000	3.0413943	0.27601796	11.018828	3.100706e-28	1.956675e-25	GLUL

Volcano Plot

Fold change vs P-value



My high-throughput
experiment generated a
long list of genes/proteins...

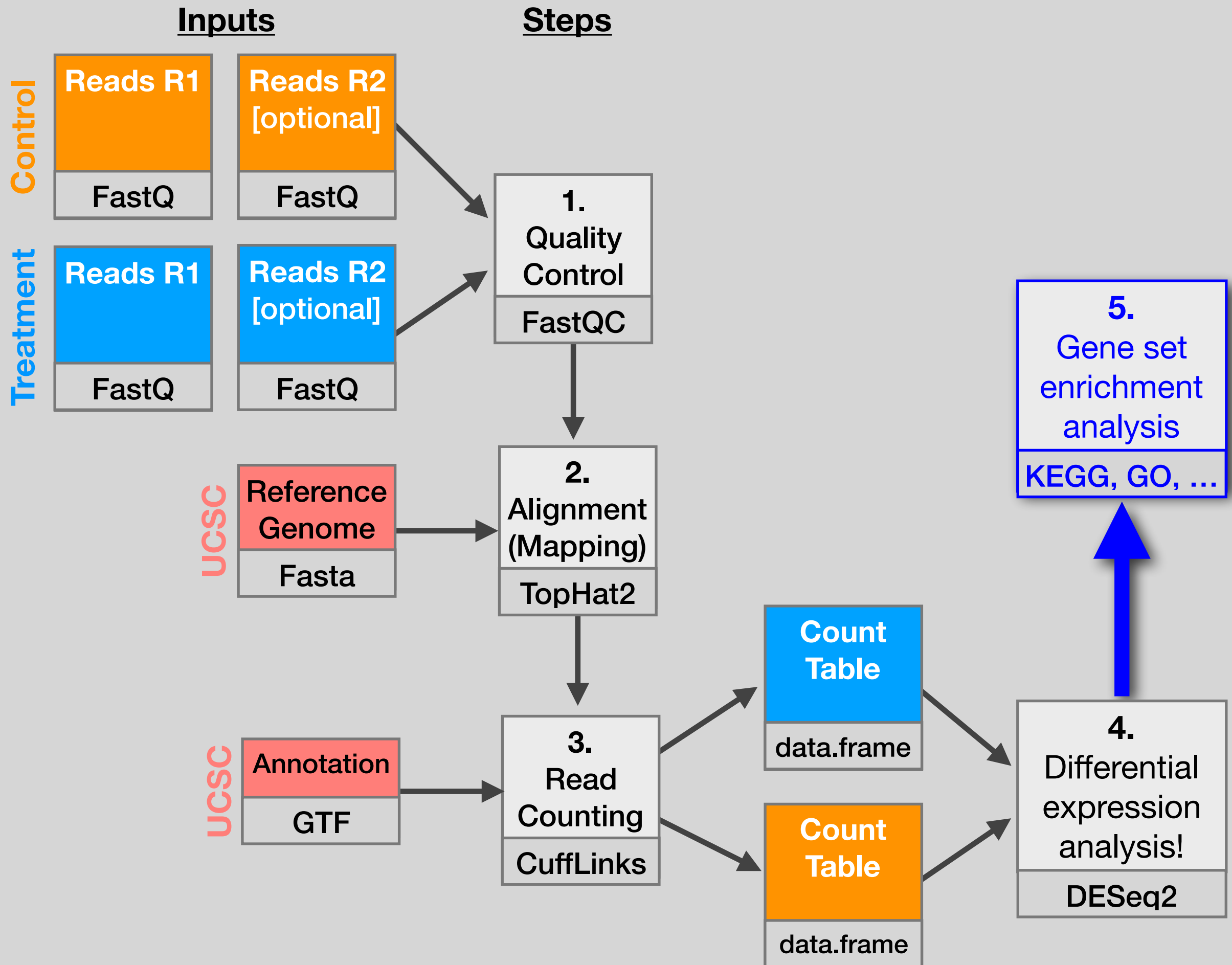
What do I do now?



Pathway analysis!

(a.k.a. geneset enrichment)

Use bioinformatics methods to help extract biological meaning from such lists...



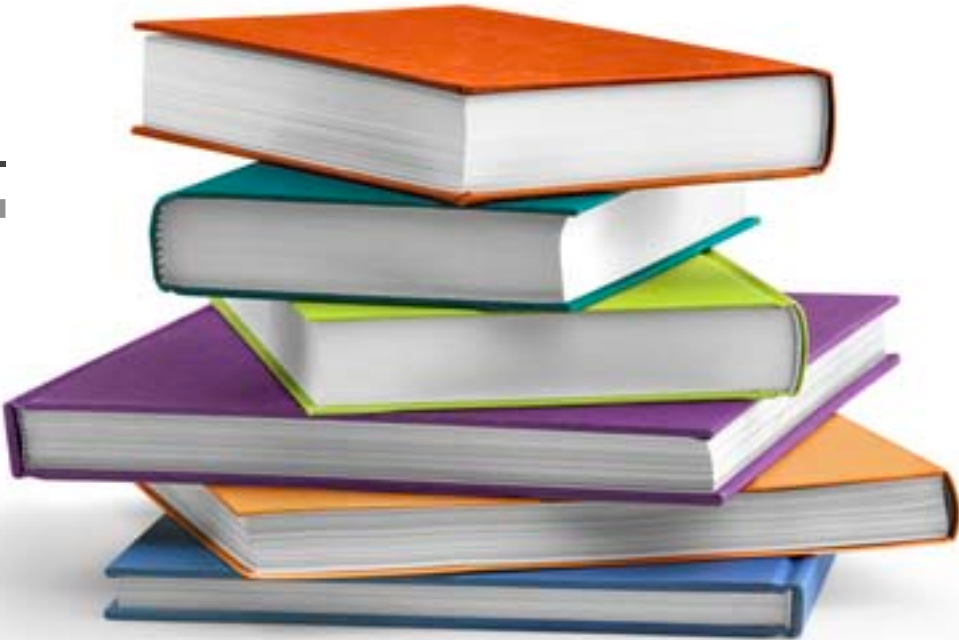
Basic idea

Differentially Expressed Genes (DEGs)

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Annotate...

Gene-sets (Pathways, annotations, etc...)



Basic idea

Differentially Expressed Genes (DEGs)

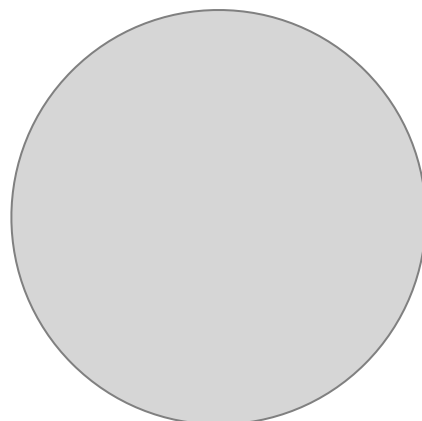
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Gene-sets (Pathways, annotations, etc...)

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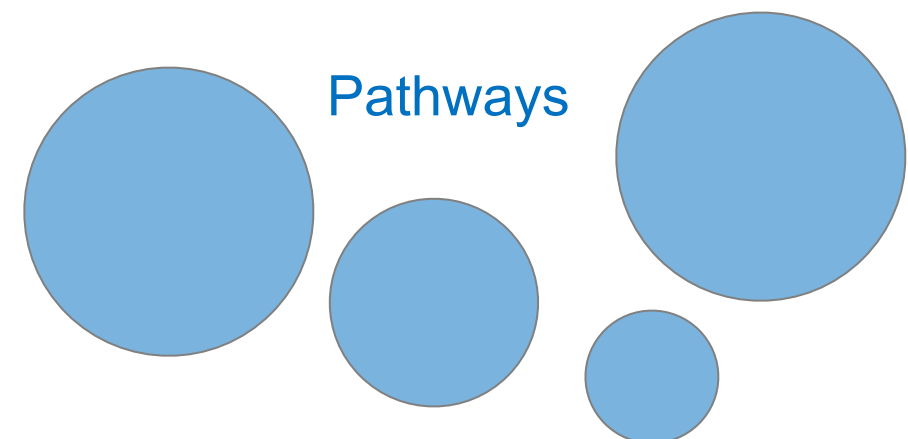


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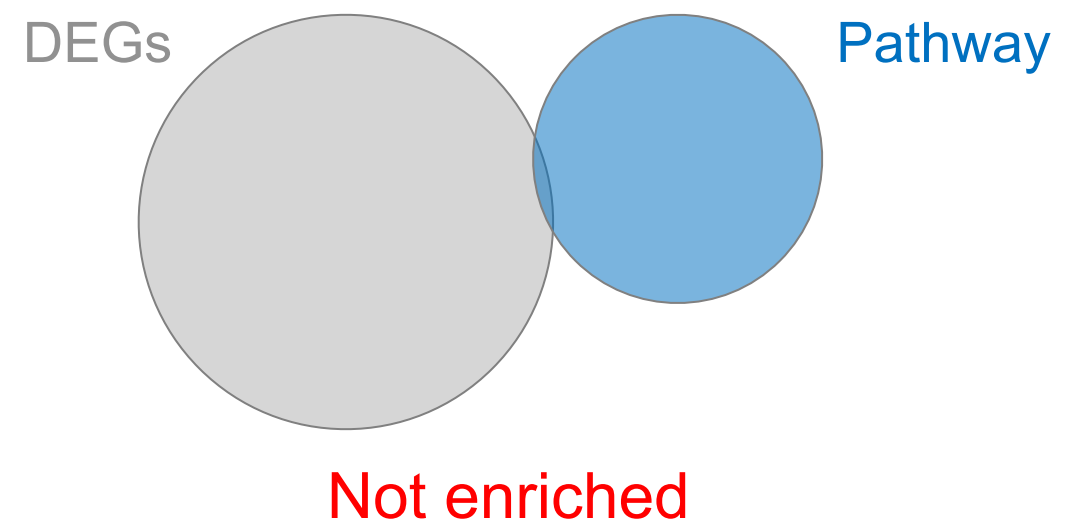
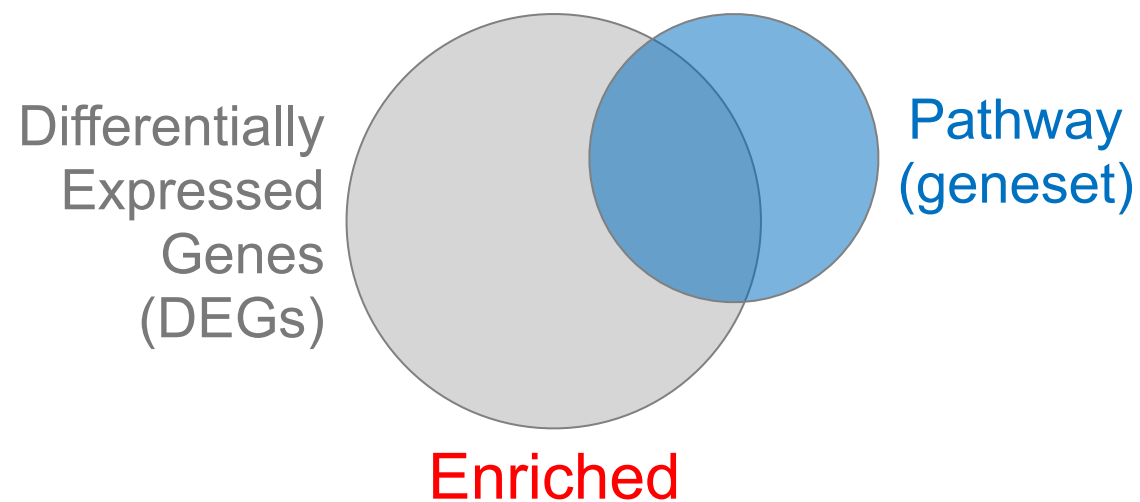
Overlap...

Pathway analysis (geneset enrichment)



Pathway analysis (a.k.a. geneset enrichment)

Principle



-
- DEGs come from your experiment
 - *Critical, needs to be as clean as possible*
 - Pathway genes (“geneset”) come from annotations
 - *Important, but typically not a competitive advantage*
 - Variations of the math: overlap, ranking, networks...
 - *Not critical, different algorithms show similar performances*

Pathway analysis (a.k.a. geneset enrichment)

Limitations

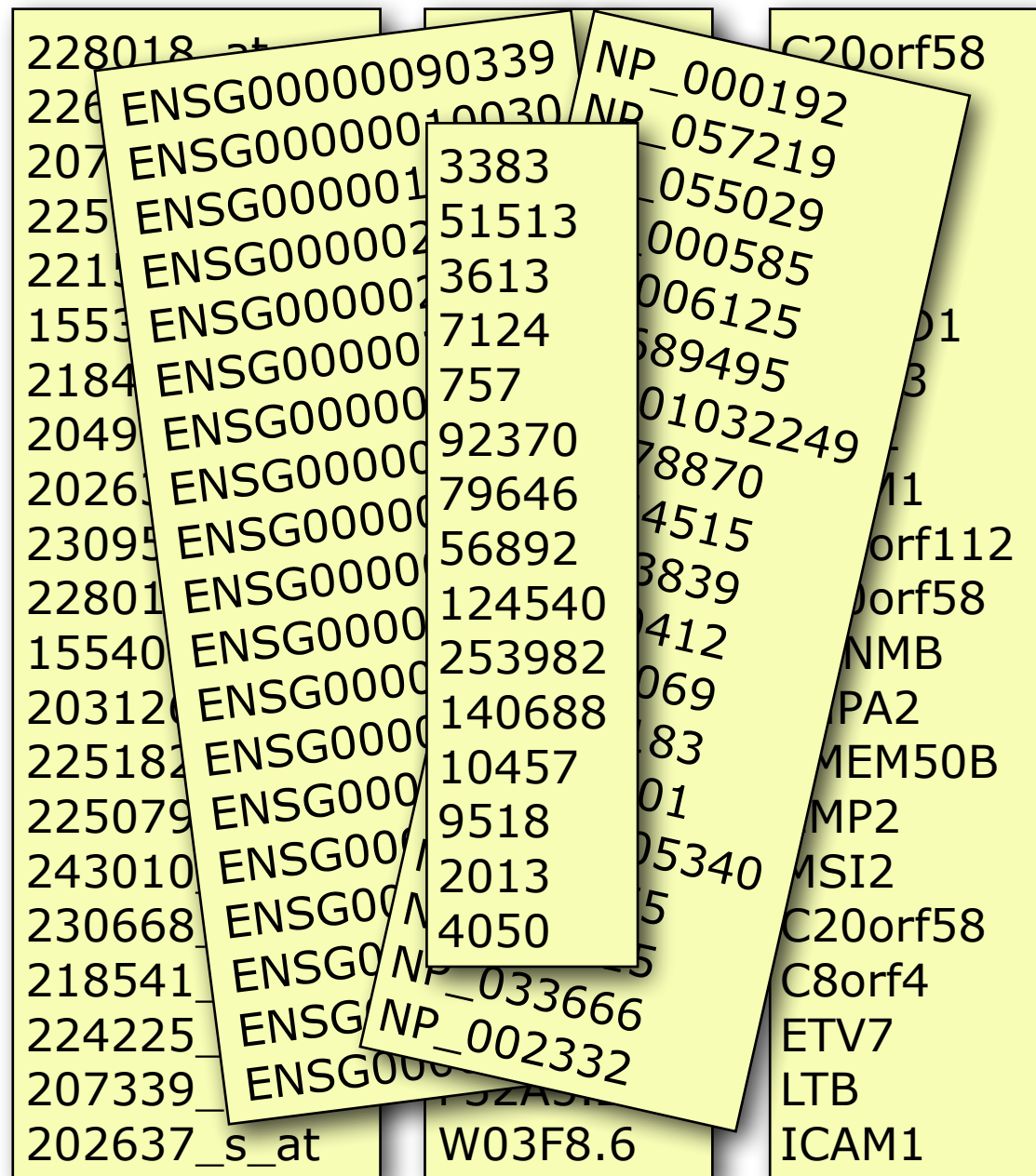
- **Geneset annotation bias:** can only discover what is already known
- **Non-model organisms:** no high-quality genesets available
- **Post-transcriptional regulation** is neglected
- **Tissue-specific** variations of pathways are not annotated
 - e.g. NF- κ B regulates metabolism, not inflammation, in adipocytes
- **Size bias:** stats are influenced by the size of the pathway
 - Many pathways/receptors **converge** to few regulators
e.g. Tens of innate immune receptors activate four TFs:
NF- κ B, AP-1, IRF3/7, NFAT

Starting point for pathway analysis:

Your gene list

- You have a list of genes/proteins of interest
- You have quantitative data for each gene/protein

- Fold change
- p-value
- Spectral counts
- Presence/absence



Translating between identifiers

- Many different identifiers exist for genes and proteins, e.g. UniProt, Entrez, etc.
- Often you will have to translate one set of ids into another
 - A program might only accept certain types of ids
 - You might have a list of genes with one type of id and info for genes with another type of id


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Translating between identifiers: UniProt < www.uniprot.org >

UniProt [Downloads](#) · [Contact](#) · [Documentation/Help](#)

Search in **Protein Knowledgebase (UniProtKB)** Query [Fields »](#)

WELCOME [NEWS](#) 

Identifiers

From

To

or no file selected

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- **VLOOKUP in Excel - *good if you are an excel whizz - I am not!***
 - **Download flat file from Entrez, Uniprot, etc; Open in Excel; Find columns that correspond to the 2 IDs you want to convert between; Sort by ID; Use vlookup to translate your list**

Translating between identifiers: Excel VLOOKUP

VLOOKUP(lookup_value, table_array, col_index_num)

B3 =VLOOKUP(A3,\$G\$3:\$O\$30490,2,FALSE)

	A	B	C	D	E	F	G	H	I	J	K
1	Data Table						Annotation Table				
2	RefSeq	Symbol	Exp1	Exp2	Exp3		RefSeq	Symbol	Entrez ID	Unigene	RefSeq
3	NM_153103	Kif1c	2.31975457	1.24558927	2.78816871		NM_001001	Zfp85-rs1	22746	Mm.288396	NM_001
4	NM_146017	Gabrp	4.15029735	3.08055836	1.18919962		NM_001001	Scap	235623	Mm.288741	NM_001
5	NM_018883	Camkk1	3.83282512	0.0522951	0.64684259		NM_001001	Scap	235623	Mm.288741	NM_001
6	NM_145936	Tspyl2	0.45449369	1.62761318	7.59770627		NM_001001	Fbxo41	330369	Mm.38777	NM_001
7	NM_026599	Cgnl1	4.84541871	2.84751796	1.61595768		NM_001001	Taf9b	407786	Mm.19440	NM_001
8	NM_013926	Cbx8	1.22903318	0.2863077	0.02952665		NM_001001	Taf9b	407786	Mm.19440	NM_001
9	NR_015566	A330023F24	1.44695053	0.98809479	1.59330144		NM_001001	BC051142	407788	Mm.73205	NM_001
10	NM_008623	Mpz	0.50749263	0.94350028	6.10581569		NM_001001	BC051142	407788	Mm.73205	NM_001
11	NM_183127	Fate1	2.45672795	4.87960794	3.60759511		NM_001001	BC048546	232400	Mm.259234	NM_001
12	NM_008943		4.78701069	4.15302647	0.85432314		NM_001001	Zfp941	407812	Mm.359154	NM_001
13	NM_025382		0.66397344	1.40664187	3.09539802		NM_001001	BC031181	407819	Mm.29866	NM_001
14	NM_182841		1.25528938	0.20505996	2.76879488		NM_001001	Baz2b	407823	Mm.486364	NM_001
15	NM_030061		0.17670108	2.75415469	2.98900691		NM_001001	Tmem204	407831	Mm.34379	NM_001
16	NM_133216		6.572343	0.59671282	3.84650536		NM_001001	Ccdc111	408022	Mm.217385	NM_001
17	NM_030063		7.05132762	0.65043627	1.68111836		NM_001001	BC048507	408058	Mm.177840	NM_001

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- Use the **merge()** or **mapIDs()** functions in **R** - fast, versatile & reproducible!
 - Also **clusterProfiler::bitr()** function and many others... [[Link to clusterProfiler vignette](#)]

Using the merge() function

> anno <- read.csv("data/annotables_grch38.csv")

This is an annotation file

> merge(mygenes, anno, by.x="row.names", by.y="ensgene")

This is our differential expressed genes

2. class-material (bash)

Using the merge() function

```
> anno <- read.csv("data/annotables_grch38.csv")
```

```
> merge(mygenes, anno, by.x="row.names", by.y="ensgene")
```

Using mapIds() function from bioconductor

```
> library("AnnotationDbi")
```

```
> library("org.Hs.eg.db")
```

Load the required Bioconductor packages

```
> mygenes$symbol <- mapIds( org.Hs.eg.db,
                           column="SYMBOL",
                           keys=row.names(mygenes),
                           keytype="ENSEMBL" )
```

Annotation we want to add

Our vector of gene names & their format

bitr: Biological Id Translator

clusterProfiler provides `bitr` and `bitr_kegg` for converting ID types. Both `bitr` and `bitr_kegg` support many species including model and many non-model organisms.

```
x <- c("GPX3", "GLRX", "LBP", "CRYAB", "DEFB1", "HCLS1", "SOD2", "HSPA2",
      "ORM1", "IGFBP1", "PTHLH", "GPC3", "IGFBP3", "TOB1", "MITF", "NDRG1",
      "NR1H4", "FGFR3", "PVR", "IL6", "PTPRM", "ERBB2", "NID2", "LAMB1",
      "COMP", "PLS3", "MCAM", "SPP1", "LAMC1", "COL4A2", "COL4A1", "MYOC",
      "ANXA4", "TFPI2", "CST6", "SLPI", "TIMP2", "CPM", "GGT1", "NNMT",
      "MAL", "EEF1A2", "HGD", "TCN2", "CDA", "PCCA", "CRYM", "PDXK",
      "STC1", "WARS", "HMOX1", "FXVD2", "RBP4", "SLC6A12", "KDEL3", "ITM2B")
eg = bitr(x, fromType="SYMBOL", toType="ENTREZID", OrgDb="org.Hs.eg.db")
head(eg)
```

##	SYMBOL	ENTREZID
## 1	GPX3	2878
## 2	GLRX	2745
## 3	LBP	3929
## 4	CRYAB	1410
## 5	DEFB1	1672
## 6	HCLS1	3059

See package vignette:

<https://bioconductor.org/packages/release/bioc/html/clusterProfiler.html>

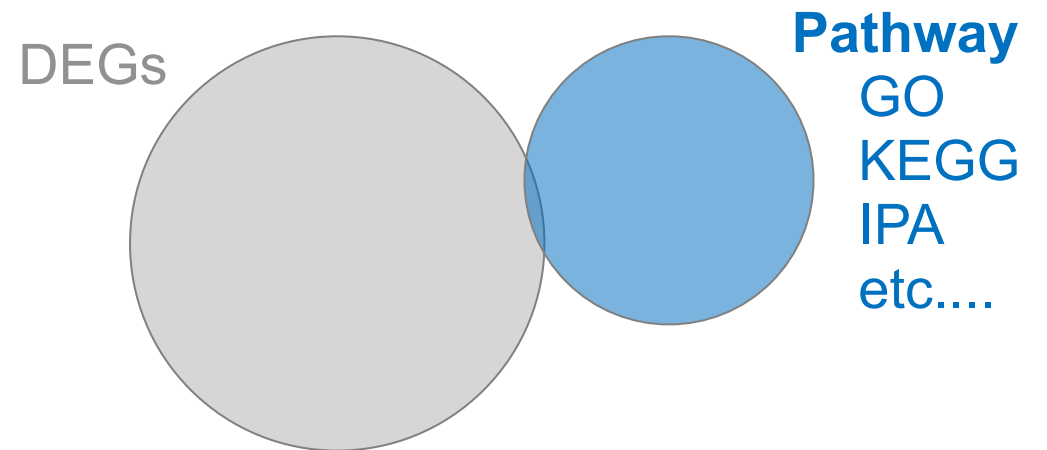
What functional set databases do you want?

- **Most commonly used:**

- **Gene Ontology (GO)**
- **KEGG Pathways** (mostly metabolic)
- **GeneGO MetaBase**
- **Ingenuity Pathway Analysis (IPA)**



INGENUITY
S Y S T E M S



- **Many others...**

- **Enzyme Classification, PFAM, Reactome,**
- Disease Ontology, MSigDB, Chemical Entities of Biological Interest, Network of Cancer Genes etc...
- See: Open Biomedical Ontologies (www.obofoundry.org)

GO < www.geneontology.org >

- **What function does HSF1 perform?**
 - *response to heat; sequence-specific DNA binding; transcription; etc*
- **Ontology** => a structured and controlled vocabulary that allows us to annotate gene products consistently, interpret the relationships among annotations, and can easily be *handled by a computer*
- GO database consists of 3 ontologies that describe gene products in terms of their associated **biological processes**, **cellular components** and **molecular functions**

GO Annotations

- GO is not a stand-alone database of genes/proteins or sequences
- Rather gene products get annotated with **GO terms** by UniProt and other organism specific databases, such as Flybase, Wormbase, MGI, ZFIN, etc.
- Annotations are available through AmiGO < amigo.geneontology.org >

The screenshot shows the AmiGO web interface. At the top, there is a header with a diagram of GO terms on the left, the text "the Gene Ontology" in the center, and "AmiGO" on the right. Below this is a blue navigation bar with links: Search, Browse, BLAST, Homolog Annotations, Tools & Resources, and Help. The main content area has the heading "Search the Gene Ontology database" followed by a search input box. Below the box are three radio buttons: "GO terms", "genes or proteins" (which is selected), and "exact match". A "Submit" button is located below the radio buttons. On the right side of the search area, there is a vertical label "AmiGO 2" with "Beta" above it. The footer contains the text "AmiGO version: 1.8" and "Try AmiGO Labs" on the left, and "GO database release 2013-10-05", "Cite this data • Terms of use • GO helpdesk", and "Copyright © 1999-2010 the Gene Ontology" on the right.

the Gene Ontology

AmiGO

Search Browse BLAST Homolog Annotations Tools & Resources Help

Search the Gene Ontology database

☐ GO terms ☒ genes or proteins ☐ exact match

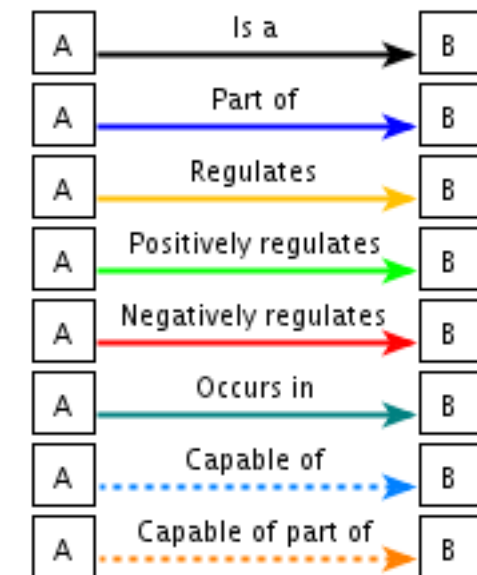
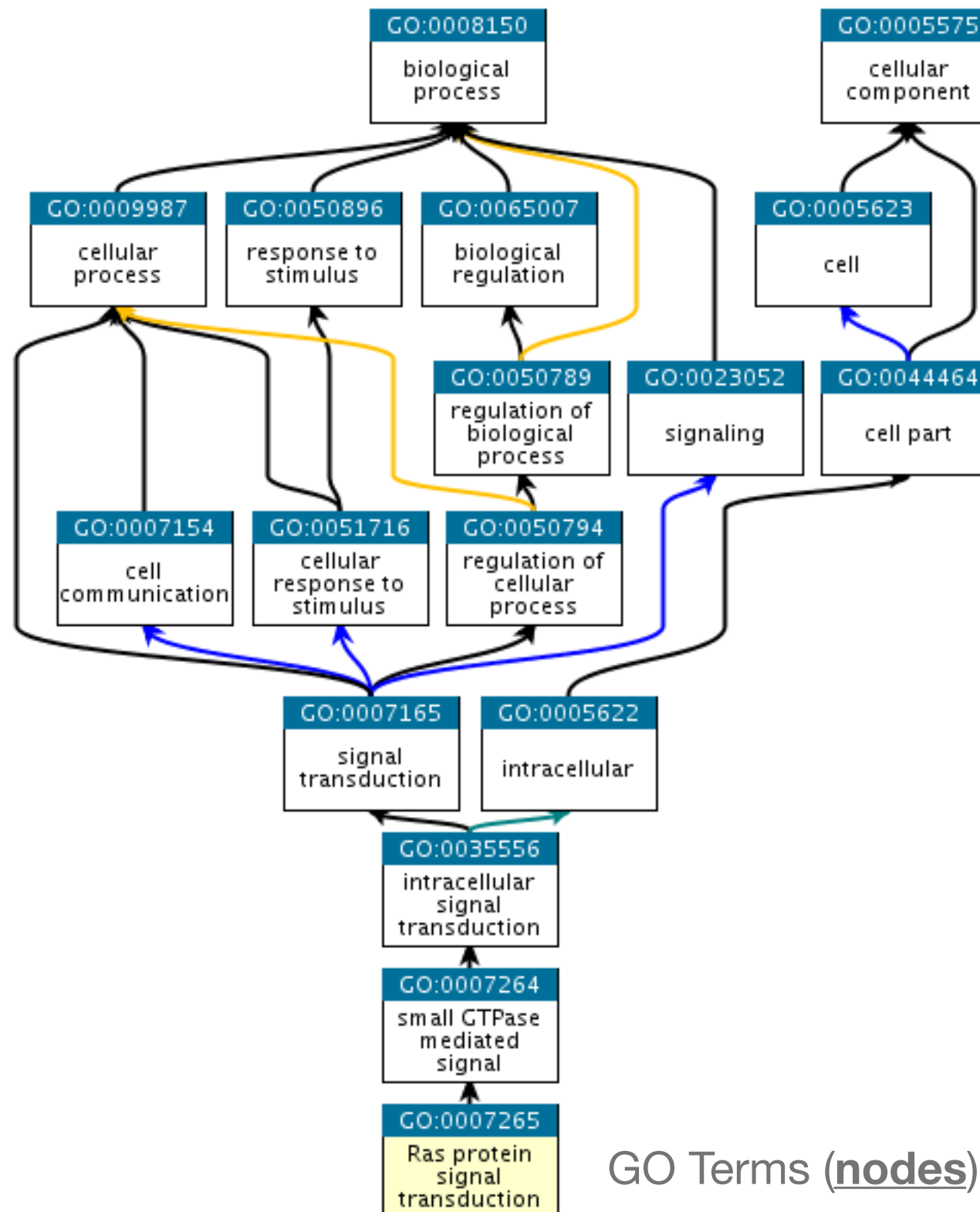
Submit

AmiGO 2 Beta

AmiGO version: 1.8
Try AmiGO Labs

GO database release 2013-10-05
Cite this data • Terms of use • GO helpdesk
Copyright © 1999-2010 the Gene Ontology

GO is structured as a “directed graph”



Relationships (edges)

Parent terms are more general & child terms more specific

GO Terms (nodes)

GO evidence codes

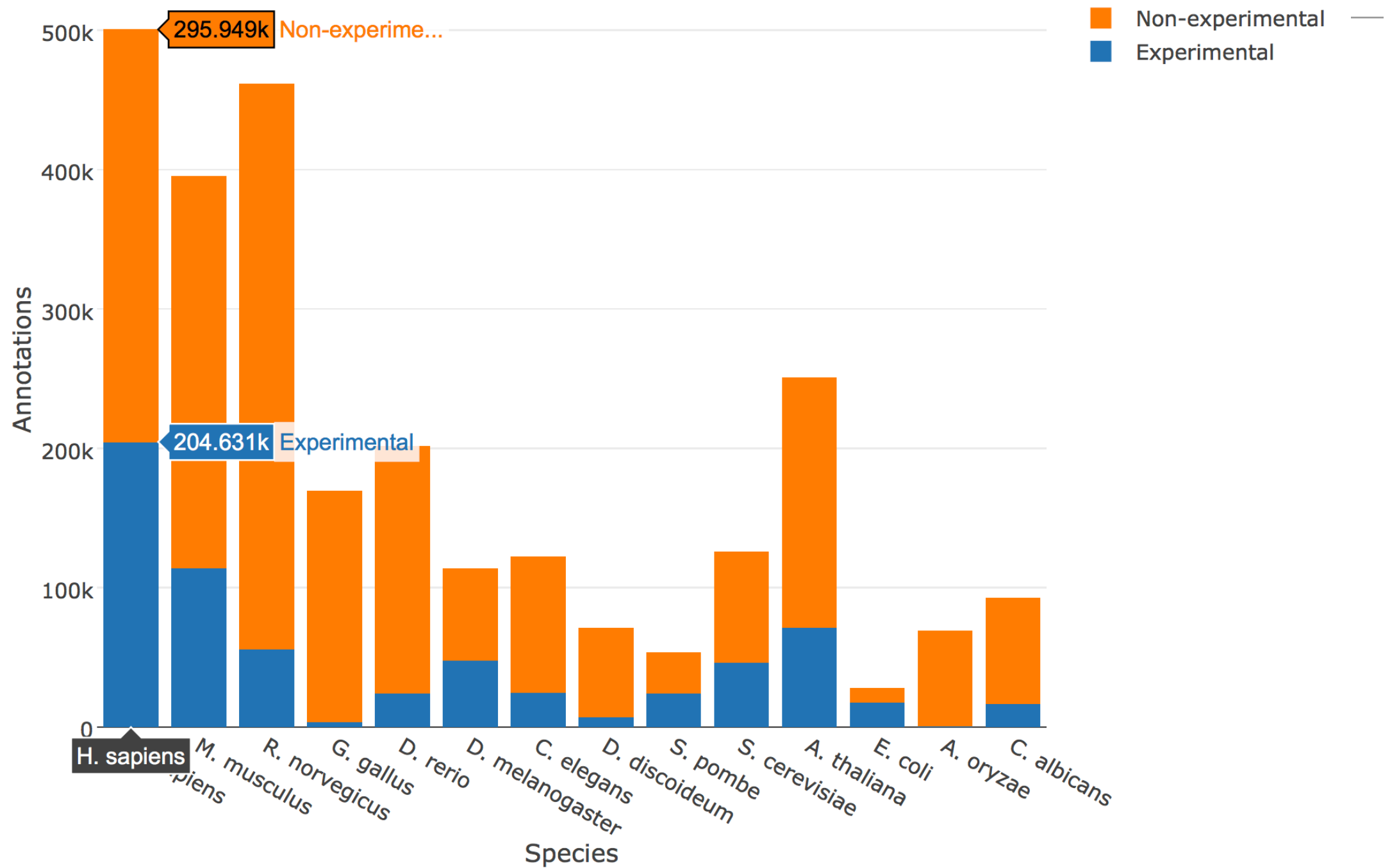
Evidence code	Evidence code description	Source of evidence	Manually checked	Current number of annotations*
IDA	Inferred from direct assay	Experimental	Yes	71,050
IEP	Inferred from expression pattern	Experimental	Yes	4,598
IGI	Inferred from genetic interaction	Experimental	Yes	8,311
IMP	Inferred from mutant phenotype	Experimental	Yes	61,549
IPI	Inferred from physical interaction	Experimental	Yes	17,043
ISS	Inferred from sequence or structural similarity	Computational	Yes	196,643
RCA	Inferred from reviewed computational analysis	Computational	Yes	103,792
IGC	Inferred from genomic context	Computational	Yes	4
IEA	Inferred from electronic annotation	Computational	No	15,687,382
IC	Inferred by curator	Indirectly derived from experimental or computational evidence made by a curator	Yes	5,167
TAS	Traceable author statement	Indirectly derived from experimental or computational evidence made by the author of the published article	Yes	44,564
NAS	Non-traceable author statement	No 'source of evidence' statement given	Yes	25,656
ND	No biological data available	No information available	Yes	132,192
NR	Not recorded	Unknown	Yes	1,185

*October 2007 release

Use and misuse of the gene ontology annotations

Seung Yon Rhee, Valerie Wood, Kara Dolinski & Sorin Draghici
Nature Reviews Genetics 9, 509-515 (2008)

Experimental annotations by species



- See AmiGO for details: http://amigo.geneontology.org/amigo/base_statistics

Can now do gene list analysis with GeneGO online!


The screenshot shows the PANTHER Classification System web interface. The browser address bar displays `pantherdb.org/webservices/go/overrep.jsp`. The page header includes the Geneontology logo and the PANTHER Classification System logo. A navigation bar contains links: Home, About, PANTHER Data, PANTHER Tools, Workspace, Downloads, and Help/Tutorial. A secondary navigation bar includes LOGIN, REGISTER, and CONTACT US. A banner below the navigation bar states "New! PANTHER13.1 released."

The main content area is titled "Gene List Analysis" and contains a message: "Please refer to our article in [Nature Protocols](#) for detailed instructions on how to use this page." Below this message is a red error message: "Error parsing request, no input specified".

The interface is divided into several sections:

- Search:** A search bar with a dropdown menu set to "All" and a "Go" button.
- Quick links:** A list of links including "Whole genome function views", "Genome statistics", "Data Version", "How to cite PANTHER", and "NEW! Recent publication describing PANTHER".
- News:** A section titled "PANTHER13.1 Released" with a link "Click for additional info."
- Newsletter subscription:** A form to enter an email address and a "Subscribe" button.
- Help Tips Steps:** A list of steps: 1. Select list and list type to analyze, 2. Select Organism, 3. Select operation.
- 1. Enter IDs and or select file for batch upload. Else enter ids or select file or list from workspace for comparing to a reference list.**
 - Enter IDs:** A text input field with a note "separate IDs by a space or comma".
 - Upload IDs:** A "Choose File" button and the text "no file selected".
 - Select List Type:** Radio buttons for "ID List" (selected), "Previously exported text search results", "Workspace list", "PANTHER Generic Mapping File", and "VCF File". A "Flanking region" dropdown menu is set to "20 Kb".
- 2. Select organism.** A list of organisms: Homo sapiens, Mus musculus, Rattus norvegicus, Gallus gallus, and Danio rerio.
- 3. Select Analysis.** Radio buttons for "Functional classification viewed in gene list" (selected) and "Functional classification viewed in pie chart".

Another popular online tool: **DAVID** at NIAID < david.abcc.ncifcrf.gov >

**Analysis Wizard**
DAVID Bioinformatics Resources 2008, NIAID/NIH

HomeStart AnalysisShortcut to DAVID ToolsTechnical CenterDownloads & APIs
Term of ServiceWhy DAVID?About Us

UploadListBackground

Upload Gene List
[Demolist 1](#) [Demolist 2](#)
[Upload Help](#)
Step 1: Enter Gene List
A: Paste a list

Clear

Or
B: Choose From a File

Choose Fileno file selected

Step 2: Select Identifier

AFFY_ID

Step 3: List Type
Gene List ☐
Background ☐

Step 4: Submit List

Submit List

Analysis Wizard

[Tell us how you like the tool](#)
[Contact us for questions](#)

← Step 1. Submit your gene list through left panel.

new! Note: Affy Exon IDs and Affy Gene Array IDs are now supported in DAVID, as "affy_id" type.

An example:

Copy/paste IDs to "box A" -> Select Identifier as "Affy_ID" -> List Type as "Gene List" -> Click "Submit" button

1007_s_at
1053_at
117_at
121_at
1255_g_at
1294_at
1316_at
1320_at
1405_i_at
1431_at
1438_at
1487_at
1494_f_at
1598_g_at

DAVID

- *Functional Annotation Chart*

Functional Annotation Chart									
<p>Current Gene List: Uploaded List_1 Current Background: Homo sapiens 2316 DAVID IDs</p> <p>Options</p> <p>Rerun Using Options Create Sublist</p> <p>Download File</p>									
Sublist	Category	Term	RT	Genes	Count	%	P-Value	Benjamini	
<input type="checkbox"/>	GOTERM_BP_5	regulation of progression through cell cycle	RT		98	4.2	3.3E-7	8.6E-4	
<input type="checkbox"/>	GOTERM_BP_5	apoptosis	RT		131	5.7	1.6E-6	2.1E-3	
<input type="checkbox"/>	GOTERM_BP_5	cell death	RT		136	5.9	3.8E-6	3.3E-3	
<input type="checkbox"/>	GOTERM_BP_5	regulation of transcription from RNA polymerase II promoter	RT		83	3.6	3.7E-5	2.4E-2	
<input type="checkbox"/>	GOTERM_BP_5	protein kinase cascade	RT		71	3.1	4.7E-5	2.4E-2	
<input type="checkbox"/>	GOTERM_BP_5	regulation of kinase activity	RT		48	2.1	5.4E-5	2.3E-2	
<input type="checkbox"/>	GOTERM_BP_5	negative regulation of cell proliferation	RT		48	2.1	1.0E-4	3.7E-2	
<input type="checkbox"/>	GOTERM_BP_5	regulation of cell size	RT		41	1.8	1.2E-4	3.9E-2	
<input type="checkbox"/>	GOTERM_BP_5	monocarboxylic acid metabolic process	RT		48	2.1	1.3E-4	3.6E-2	
<input type="checkbox"/>	GOTERM_BP_5	positive regulation of nucleobase, nucleoside, nucleotide and nucleic acid metabolic process	RT		61	2.6	1.5E-4	3.8E-2	
<input type="checkbox"/>	GOTERM_BP_5	positive regulation of cellular metabolic process	RT		72	3.1	1.7E-4	3.8E-2	

Systematic and integrative analysis of large gene lists using DAVID bioinformatics resources

Da Wei Huang, Brad T Sherman & Richard A Lempicki

Nature Protocols **4**, 44 - 57 (2009)

Overlapping functional sets

- **Many functional sets overlap**

- In particular those from databases that are hierarchical in nature (e.g. GO)

- **Hierarchy enables:**

- Annotation flexibility (e.g. allow different degrees of annotation completeness based on what is known)
- Computational methods to “understand” function relationships (e.g. ATPase function is a subset of enzyme function)

- **Unfortunately, this also makes functional profiling trickier**

- Clustering of functional sets can be helpful in these cases

DAVID

- DAVID now offers functional annotation clustering:

Annotation Summary Results


[Help and Tool Manual](#)

Current Gene List: Uploaded List_3 **2320 DAVID IDs**

Current Background: HOMO SAPIENS **Check Defaults** ☒

- ☒ **Main Accessions** (0 selected)
- ☒ **Other Accessions** (0 selected)
- ☒ **Gene Ontology** (4 selected)
- ☒ **Protein Domains** (3 selected)
- ☒ **Pathways** (3 selected)
- ☒ **General Annotations** (0 selected)
- ☒ **Functional Categories** (3 selected)
- ☒ **Protein Interactions** (0 selected)
- ☒ **Literature** (0 selected)
- ☒ **Disease** (1 selected)
- ☒ **Tissue Expression**

Combined View for Selected Annotation



DAVID Functional Annotation Clustering

- Based on shared genes between functional sets

Functional Annotation Clustering						
Help and Manual Current Gene List: Uploaded List_3 2320 DAVID IDs						
Options Classification Stringency Medium						
Rerun using options Create Sublist Download File						
Annotation Cluster 1	Enrichment Score: 3.72	G		Count	P_Value	Benjamini
<input type="checkbox"/> GOTERM_BP_5	regulation of transcription from RNA polymerase II promoter	RT		83	3.7E-5	2.4E-2
<input type="checkbox"/> GOTERM_BP_5	positive regulation of nucleobase, nucleoside, nucleotide and nucleic acid metabolic process	RT		61	1.5E-4	3.8E-2
<input type="checkbox"/> GOTERM_BP_5	positive regulation of cellular metabolic process	RT		72	1.7E-4	3.8E-2
<input type="checkbox"/> GOTERM_BP_5	positive regulation of transcription	RT		58	3.8E-4	5.0E-2
<input type="checkbox"/> GOTERM_BP_5	positive regulation of transcription, DNA-dependent	RT		48	7.4E-4	7.6E-2
Annotation Cluster 2	Enrichment Score: 3.54	G		Count	P_Value	Benjamini
<input type="checkbox"/> GOTERM_BP_5	regulation of cell size	RT		41	1.2E-4	3.9E-2
<input type="checkbox"/> GOTERM_BP_5	regulation of cell growth	RT		33	3.7E-4	5.1E-2
<input type="checkbox"/> GOTERM_BP_5	cell morphogenesis	RT		81	5.2E-4	5.7E-2
Annotation Cluster 3	Enrichment Score: 3.37	G		Count	P_Value	Benjamini
<input type="checkbox"/> GOTERM_BP_5	apoptosis	RT		131	1.6E-6	2.1E-3
<input type="checkbox"/> GOTERM_BP_5	cell death	RT		136	3.8E-6	3.3E-3
<input type="checkbox"/> GOTERM_BP_5	regulation of programmed cell death	RT		88	3.2E-4	5.8E-2
<input type="checkbox"/> GOTERM_BP_5	positive regulation of apoptosis	RT		48	3.3E-4	5.6E-2
<input type="checkbox"/> GOTERM_BP_5	regulation of apoptosis	RT		87	3.5E-4	5.2E-2
<input type="checkbox"/> GOTERM_BP_5	positive regulation of programmed cell death	RT		48	4.0E-4	5.0E-2

Want more?

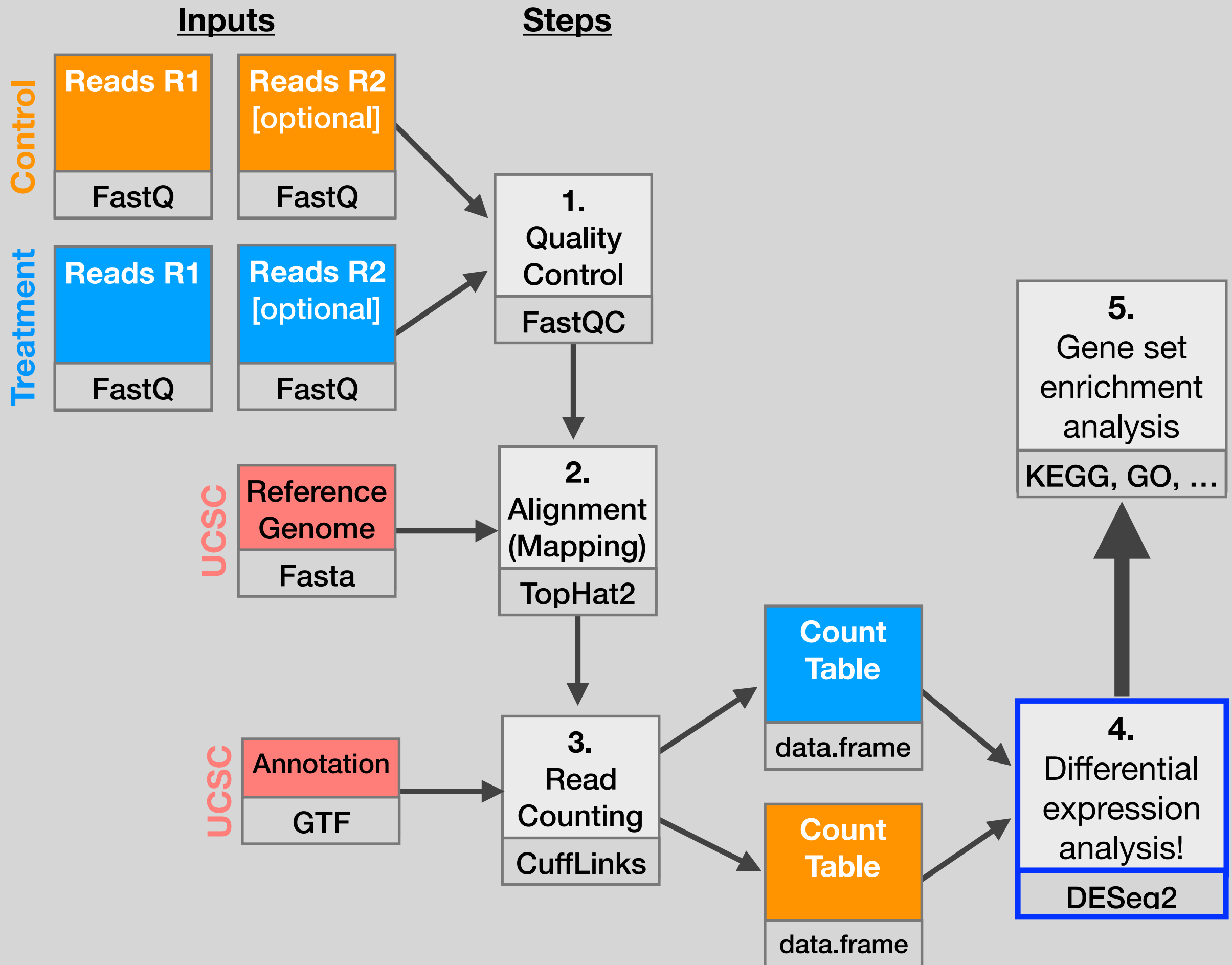


- **GeneGO** < portal.genego.com >
 - MD/PhD curated annotations, great for certain domains (eg, Cystic Fibrosis)
 - Nice network analysis tools
 - Email us for access
- **Oncomine** < www.oncomine.org >
 - Extensive cancer related expression datasets
 - Nice concept analysis tools
 - Research edition is free for academics, Premium edition \$\$\$
- **Lots and lots other R/Bioconductor packages in this area!!!**

Do it Yourself!

Hands-on time!

<http://thegrantlab.org/bgggn213>



counts + metadata

1

countData

gene	ctrl_1	ctrl_2	exp_1	exp_2
geneA	10	11	56	45
geneB	0	0	128	54
geneC	42	41	59	41
geneD	103	122	1	23
geneE	10	23	14	56
geneF	0	1	2	0
...

2

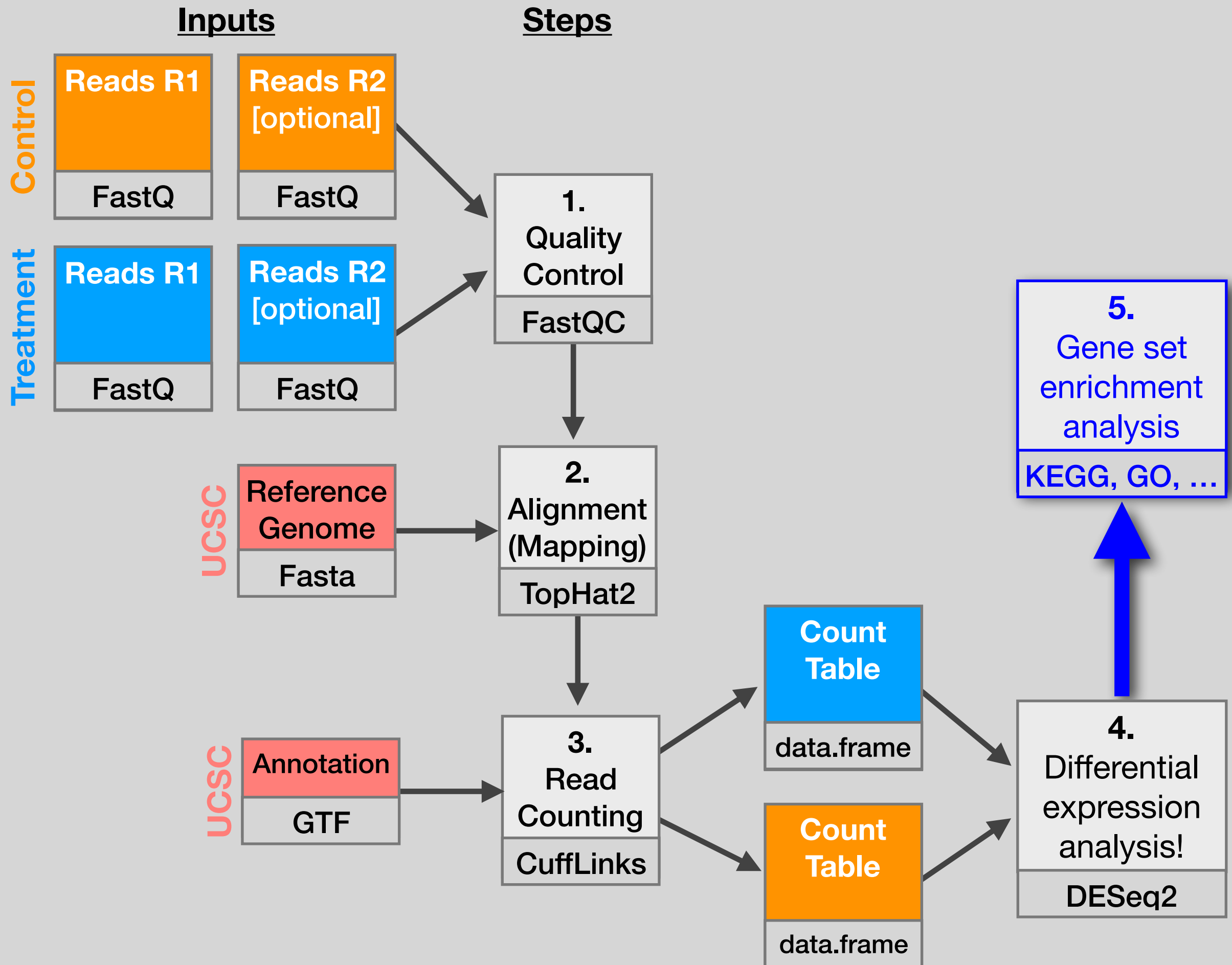
colData

id	treatment	sex	...
ctrl_1	control	male	...
ctrl_2	control	female	...
exp_1	treated	male	...
exp_2	treated	female	...

colData describes metadata about the *columns* of countData

countData is the count matrix
(Number of reads coming from each
gene for each sample)

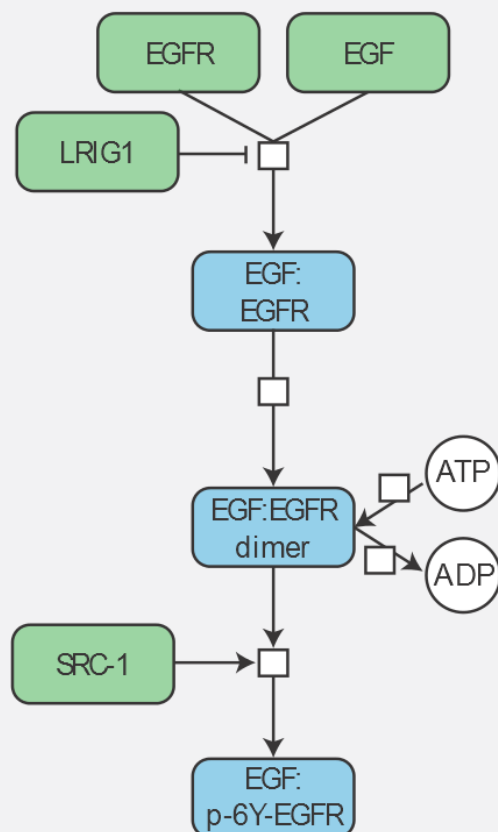
N.B. First column of colData must match column names (i.e. **sample names**) of countData (-1st)



Pathways vs Networks

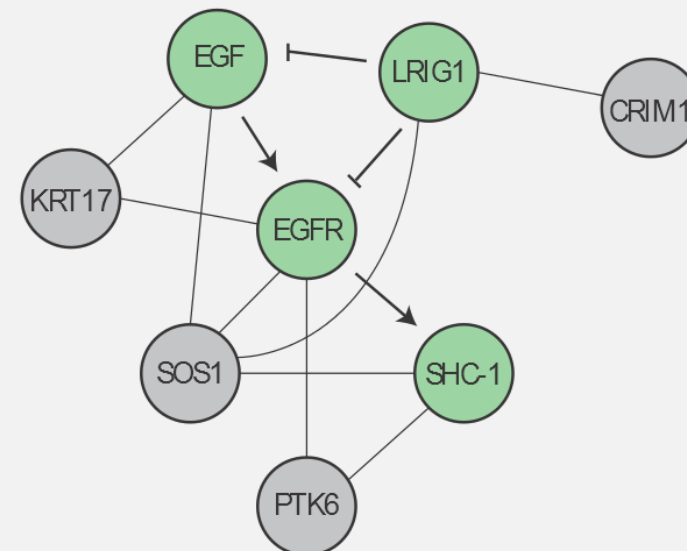
Next Class

EGFR-centered
Pathway

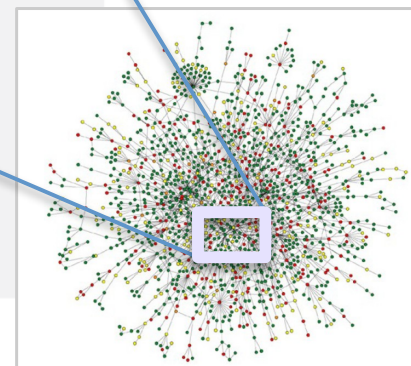


- Detailed, high-confidence consensus
- Biochemical reactions
- Small-scale, fewer genes
- Concentrated from decades of literature

EGFR-centered
Network



- Simplified cellular logic, noisy
- Abstractions: directed, undirected
- Large-scale, genome-wide
- Constructed from *omics* data integration



Goal

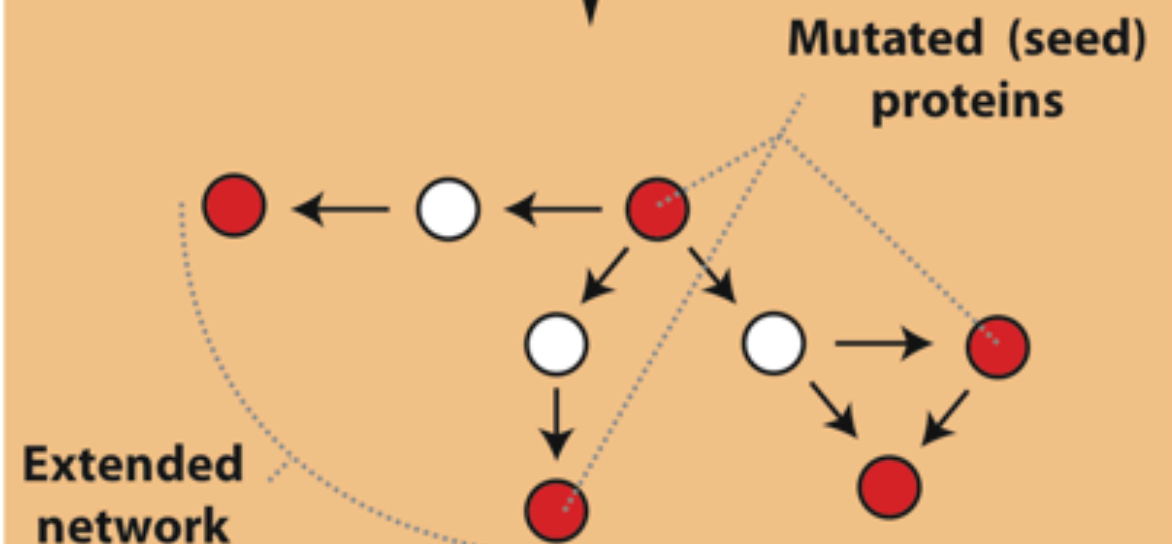
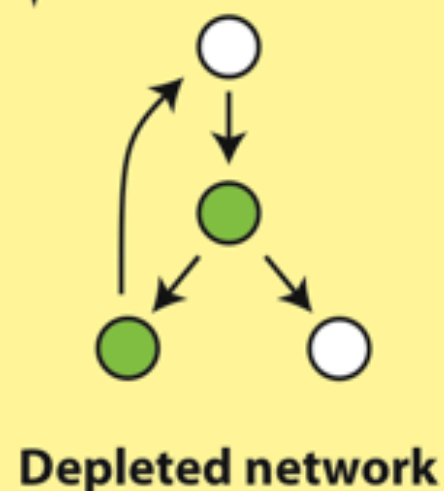
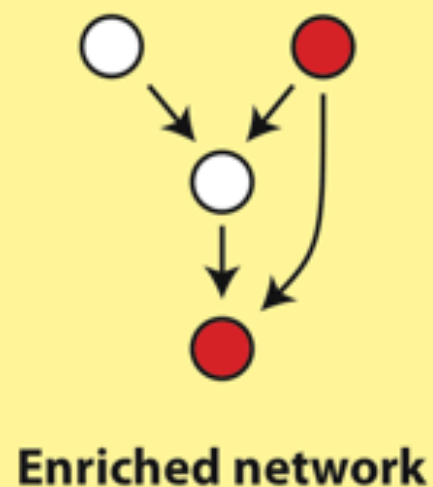
1 Enrichment of fixed gene sets

Identification of pre-built pathways or networks that are enriched in a set of mutated or differentially expressed genes

2 De novo sub-network construction and clustering

Construction of specific sub-networks from the set of mutated or differentially expressed genes to identify an extended list of putative cancer genes

Output



Goal

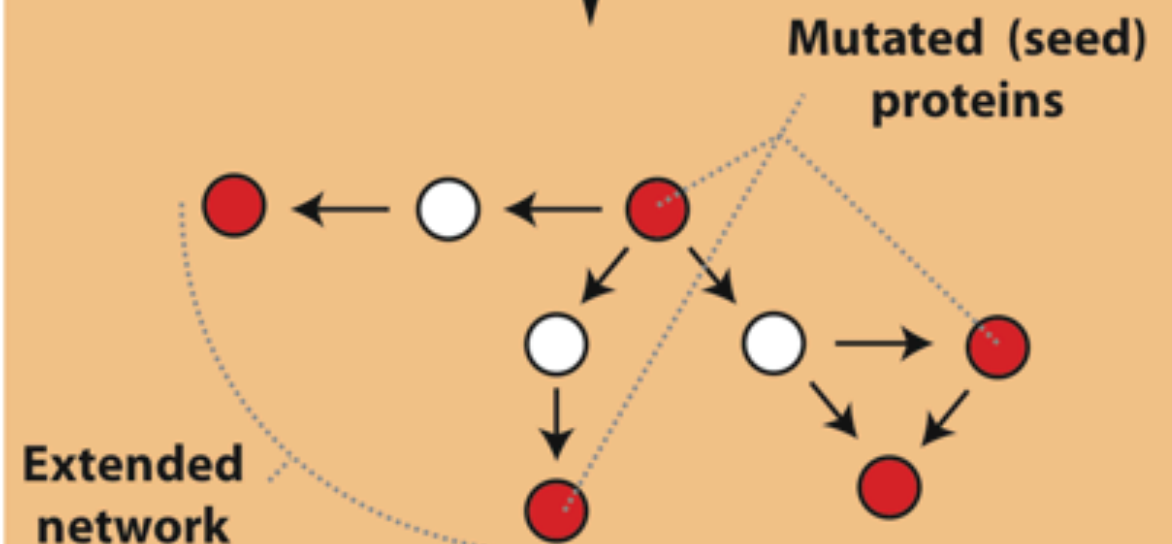
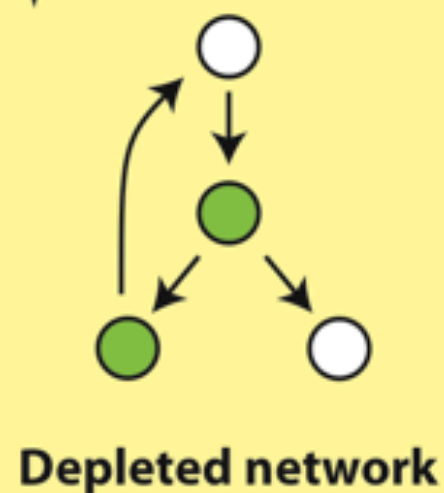
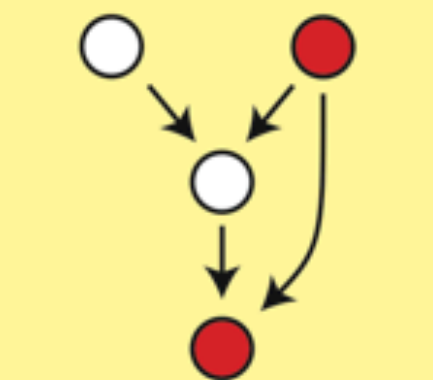
1 Enrichment of fixed gene sets

Identification of pre-built pathways or networks that are enriched in a set of mutated or differentially expressed genes

2 De novo sub-network construction and clustering

Construction of specific sub-networks from the set of mutated or differentially expressed genes to identify an extended list of putative cancer genes

Output



What biological process is altered in this cancer?

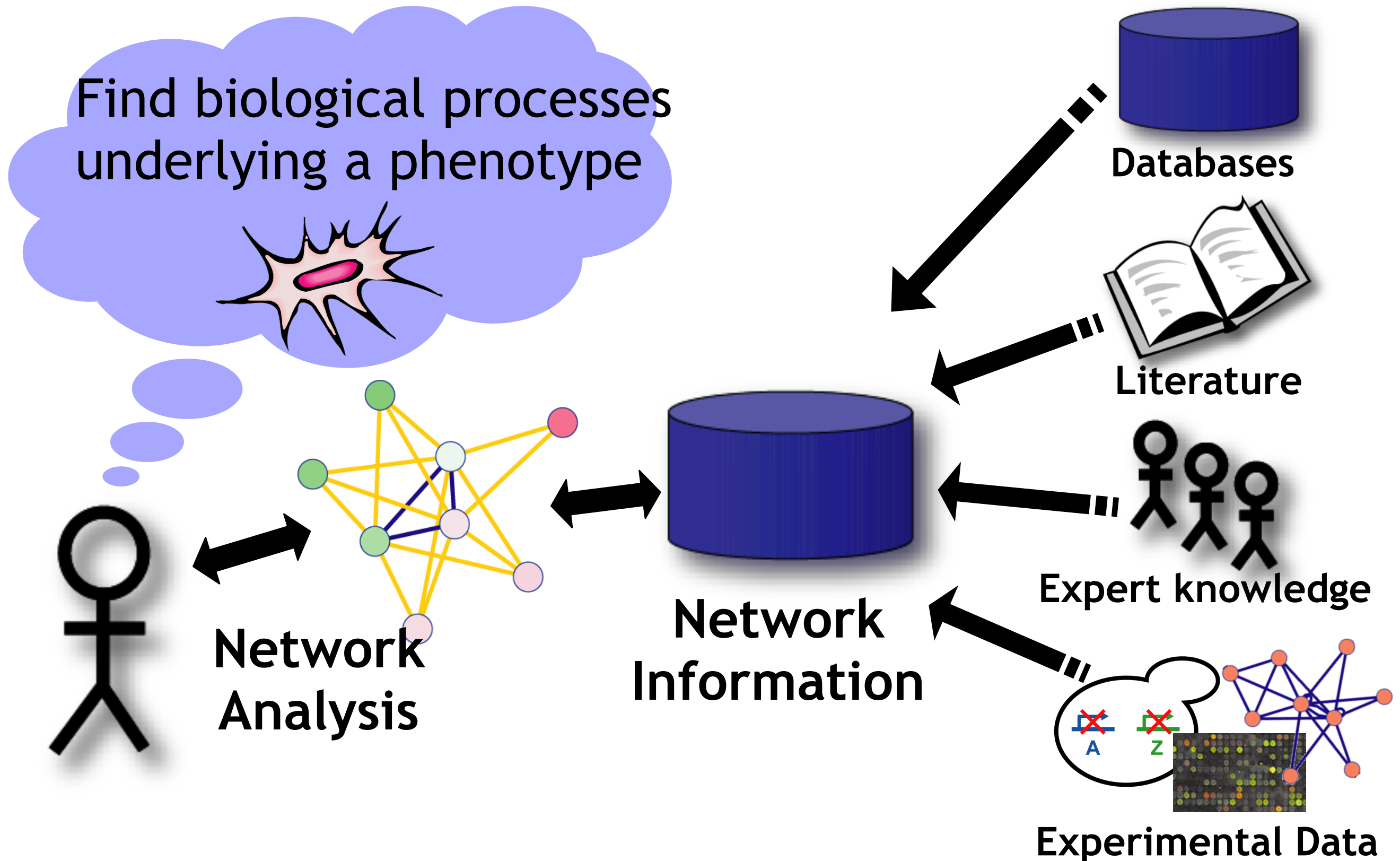
Are NEW pathways altered in this cancer? Are there clinically relevant tumor subtypes?

Pathway analysis (a.k.a. geneset enrichment)

Limitations

- **Geneset annotation bias:** can only discover what is already known
- **Non-model organisms:** no high-quality genesets available
- **Post-transcriptional regulation** is neglected
- **Tissue-specific** variations of pathways are not annotated
 - e.g. NF- κ B regulates metabolism, not inflammation, in adipocytes
- **Size bias:** stats are influenced by the size of the pathway
 - Many pathways/receptors **converge** to few regulators
e.g. Tens of innate immune receptors activate four TFs:
NF- κ B, AP-1, IRF3/7, NFAT

Pathway & Network Analysis Overview



R Knowledge Check For BGGN-213 Quiz

This will be marked but not graded
(*i.e.* will not factor into your course grade)

Time Limit: 30 mins

Your results will be e-mailed to you upon quiz close

Thank you!