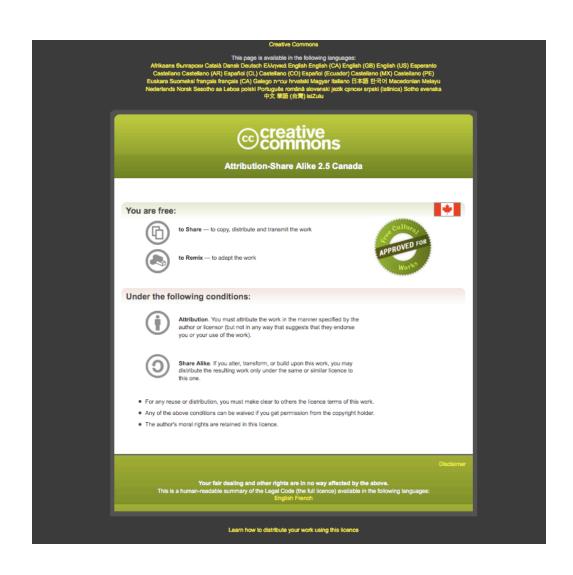


Canadian Bioinformatics Workshops

www.bioinformatics.ca

bioinformaticsdotca.github.io



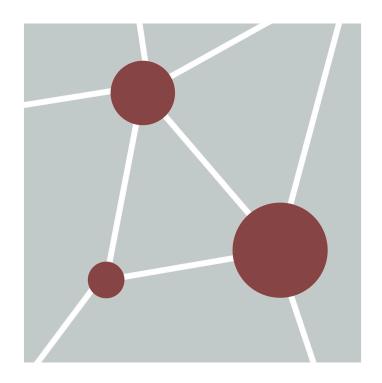


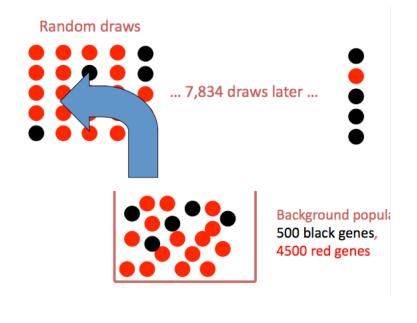
Finding over-represented pathways in gene lists

Veronique Voisin

Pathway and Network Analysis of –omics Data
July 27-29, 2020



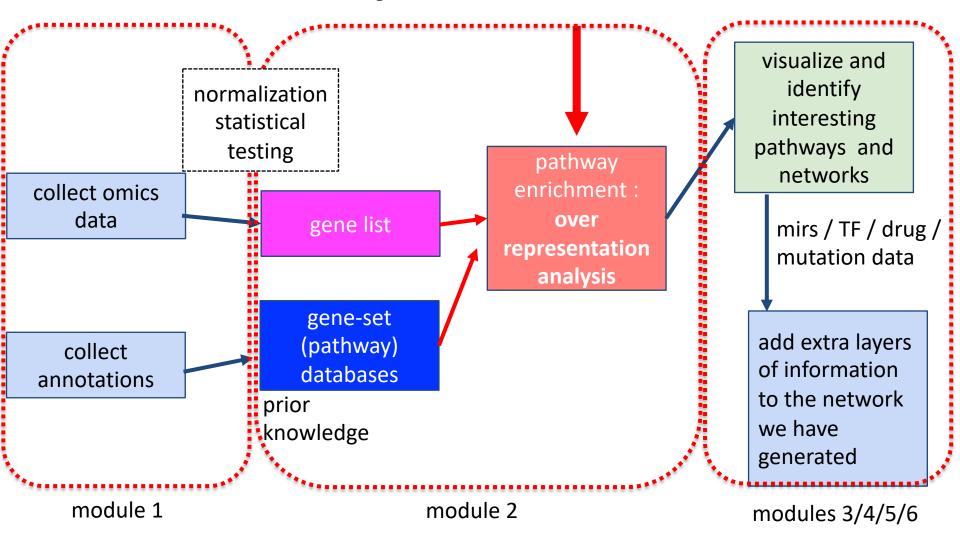




Learning Objectives

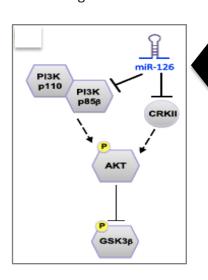
- Be able to understand the differences between a defined gene list and a ranked gene list and which enrichment test to apply.
- Be able to understand the result of an enrichment test and how to interpret it
- Be able to understand the concept of pvalue and corrected pvalue (FDR) in the context of enrichment analysis.
- Presentation of 2 enrichment tools

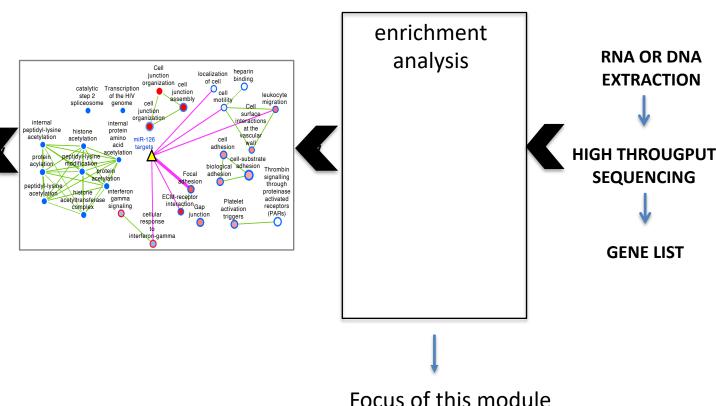
Analysis workflow



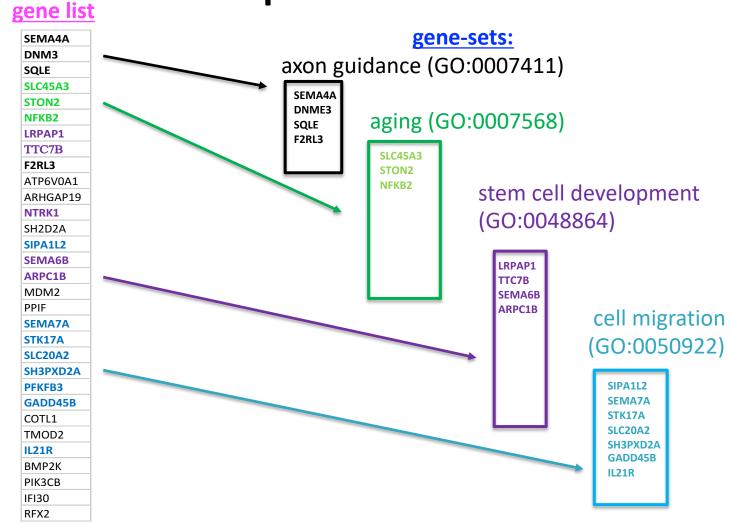
pathway analysis workflow...rewind

"In HSC/early progenitors, miR-126 regulates multiple targets within the PI3K/AKT/GSK3β pathway, attenuating signal transduction in response to extrinsic signals."

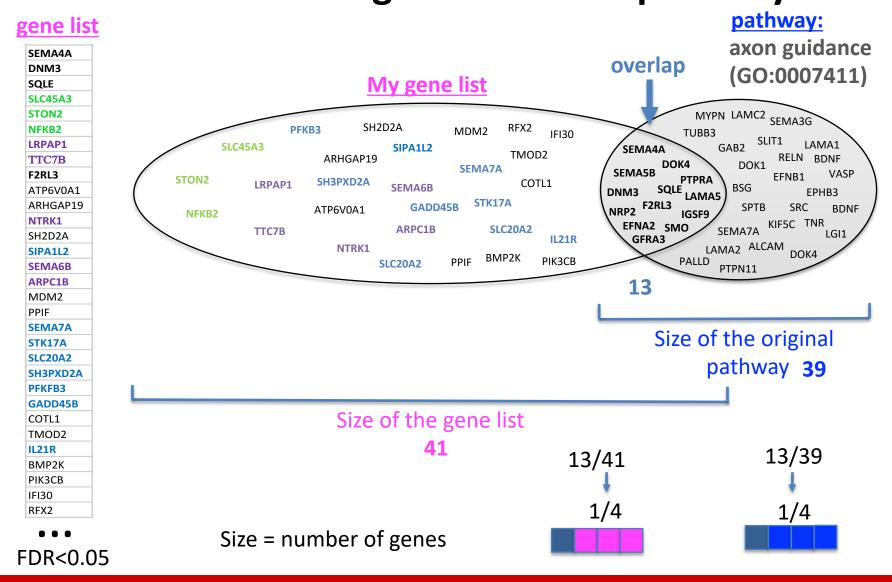




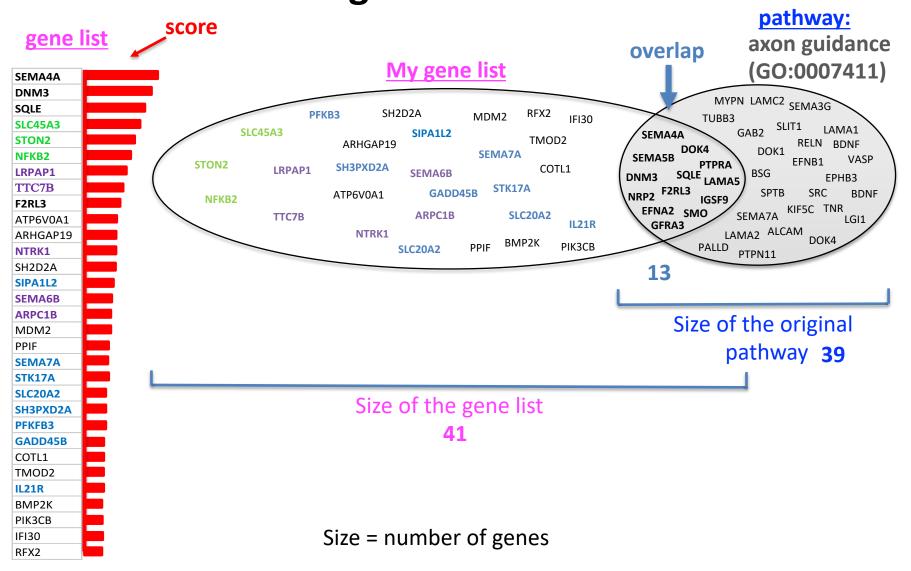
Gene set enrichment analysis is a way to summarize your gene list into pathways to ease biological interpretation of the data



Gene set enrichment analysis calculates the overlap between our gene list and a pathway



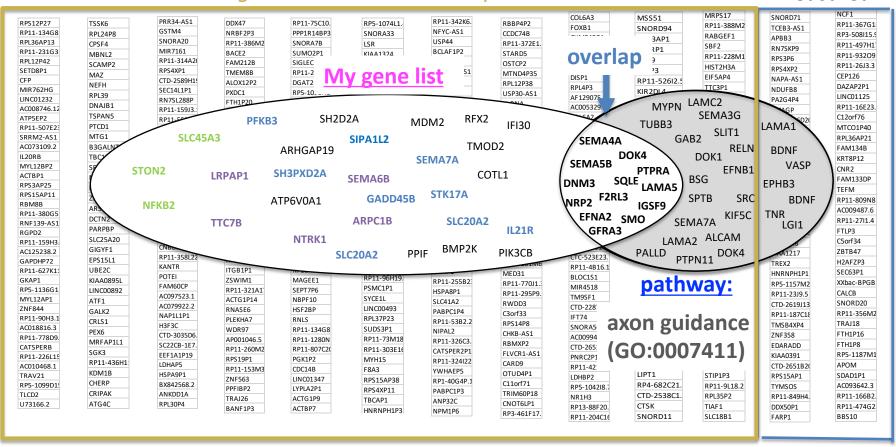
Can we add a score associated with the genes when calculating the enrichment score?



The background represents the genes that could have been captured in my omics experiment

genes measured in the experiment

genes not measured



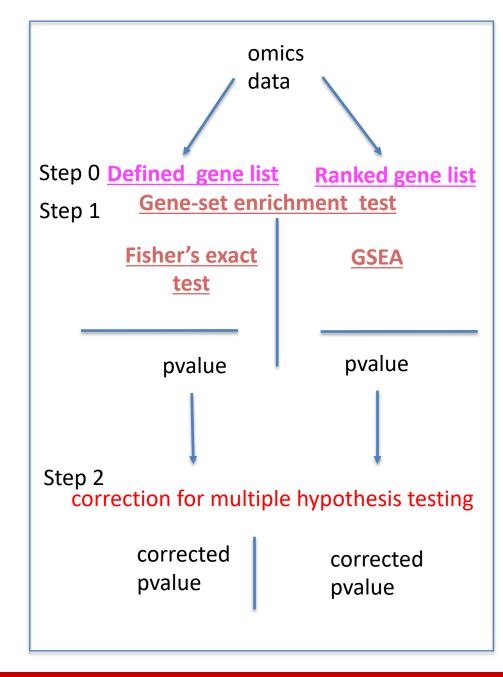
estimated 20,000-25,000 human protein-coding genes
How many genes could have been captured in your experiment?

We are testing many pathways at the same time

correction for multiple hypothesis testing

Outline

- Two types of gene lists (ranked or not)
- Introduction to enrichment analysis
- Fisher's Exact Test, aka Hypergeometric Test
- GSEA for ranked lists.
- Multiple test corrections:
 - Bonferroni correction
 - False Discovery Rate computation using Benjamini-Hochberg procedure



Types of enrichment analysis

- Defined gene list (e.g. expression change > 2fold)
 - Answers the question: Are any pathways (gene sets) surprisingly enriched (or depleted) in my gene list?
 - Statistical test: Fisher's Exact Test (aka Hypergeometric test)
- Ranked gene list (e.g. by differential expression)
 - Answers the question: Are any pathways (gene sets) ranked surprisingly high or low in my ranked list of genes?
 - Statistical test: GSEA, Wilcoxon rank sum test (+ others we won't discuss)

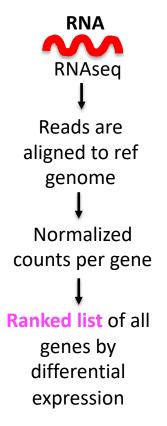
Why test enrichment in ranked gene lists?

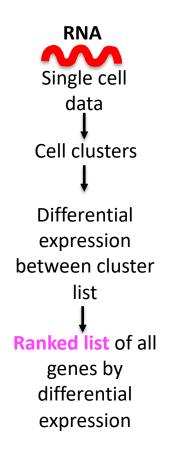
- Possible problems with gene list test
 - No "natural" value for the threshold
 - Different results at different threshold settings
 - Possible loss of statistical power due to thresholding
 - No resolution between significant signals with different strengths
 - Weak signals neglected

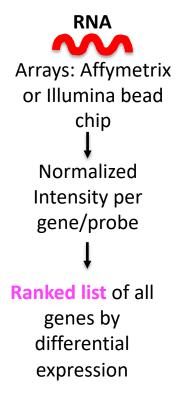
OMICS gene lists: ranked or not ranked? a few examples

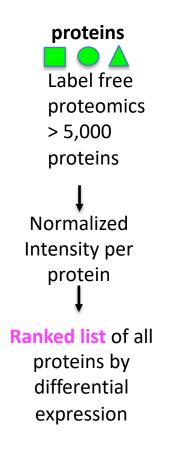
Experimental design: 2 class-design, treated versus control

Starting point:

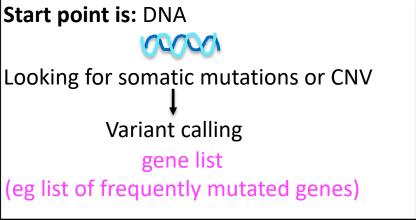


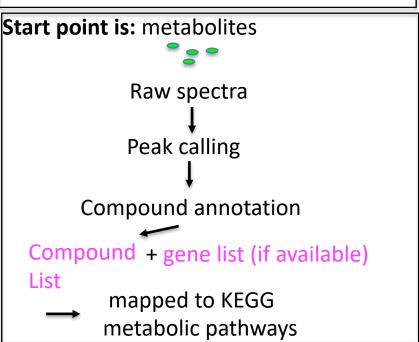


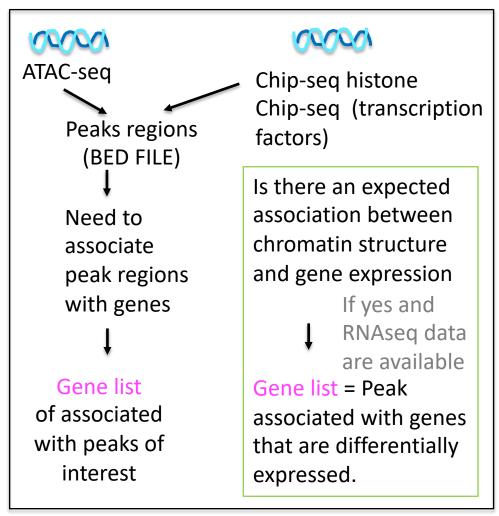




OMICS gene lists: ranked or not ranked? a few examples, cont.

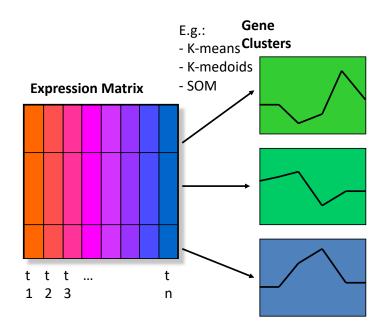






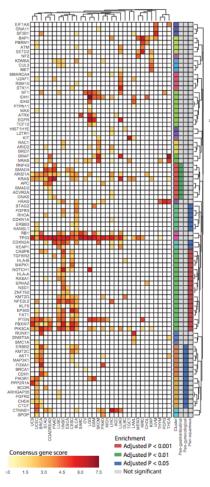
Example of defined gene lists

RNA: Time course or cluster analysis



Each cluster is a separate gene list

DNA: Gene list of frequently mutated genes



Comprehensive Characterization of Cancer Driver Genes and Mutations

Bailey et al. Cell 2018.

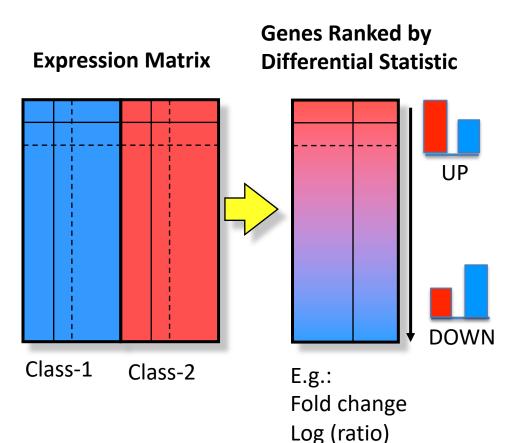
PMID: 30096302

They reported a PanCancer and PanSoftware analysis spanning **9,423 tumor exomes** and using 26 computational tools to catalog driver genes and mutations.

They identified 299 driver genes with implications regarding their anatomical sites and cancer/cell types

Two-class design: ranked gene list

t values from t-test



Ranking score = sign(logFC)*-log10(pvalue)

	LogFC	Pvalue	score
BGN	+1	1.73E-33	32.76
ANTXR1	+1	4.39E-31	30.36
FZD1	+1	4.41E-30	29.36
COL16A1	+1	1.33E-29	28.88
KLF3	+1	8.32E-02	1.08
RASEF	+1	9.01E-01	0.05
ISOC1	+1	9.01E-01	0.05
ANO1	+1	9.01E-01	0.04
CBWD3	-1	8.18E-02	-1.09
GBP4	-1	2.45E-16	-15.61
TAP1	-1	1.04E-19	-18.98
PSMB9	-1	1.84E-20	-19.73

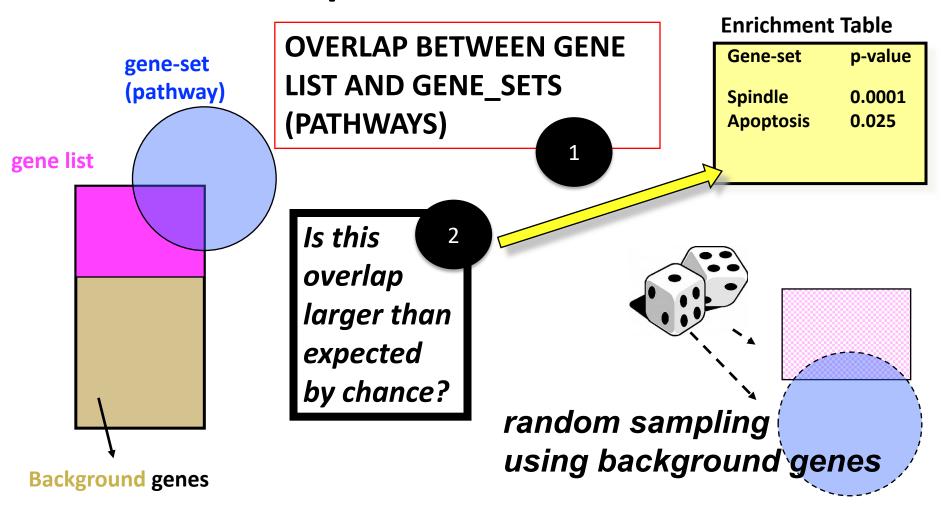
UP

Gene list enrichment test

Gene list enrichment analysis

- Given:
 - 1. Gene list: e.g. RRP6, MRD1, RRP7, RRP43, RRP42 (yeast)
 - Gene sets (pathways) or annotations: e.g. The Gene Ontology, transcription factor binding sites in promoter
- Question: Are any of the gene sets (pathways) surprisingly enriched in the gene list?
- Details:
 - Where do the gene lists come from?
 - How to assess "surprisingly" (statistics)
 - How to correct for repeating the tests

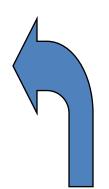
How do simple enrichment tests work?



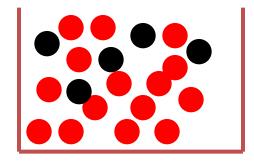
The Fisher's exact test

Gene list

- RRP6
- MRD1
- RRP7
- RRP43
- RRP42



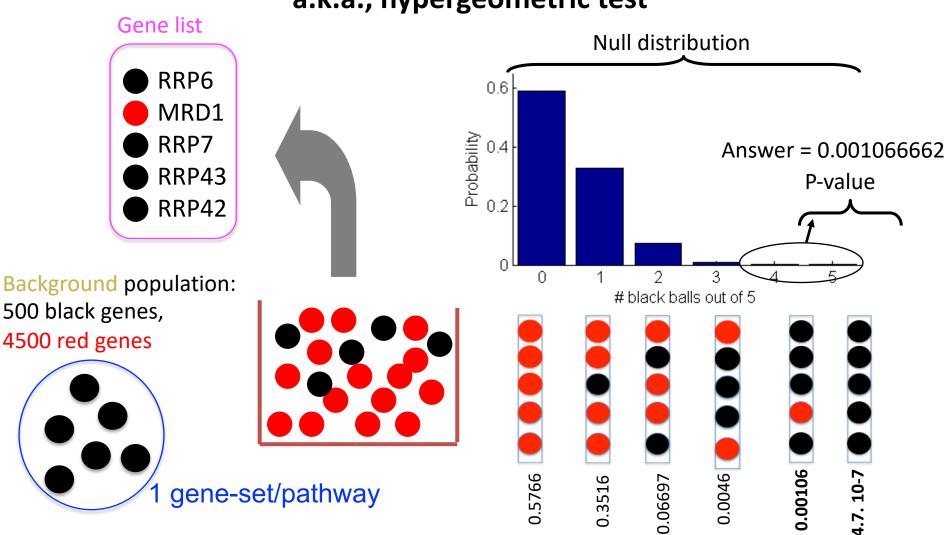
Null hypothesis: List is a random sample from population
Alternative hypothesis: More black genes than expected in my list



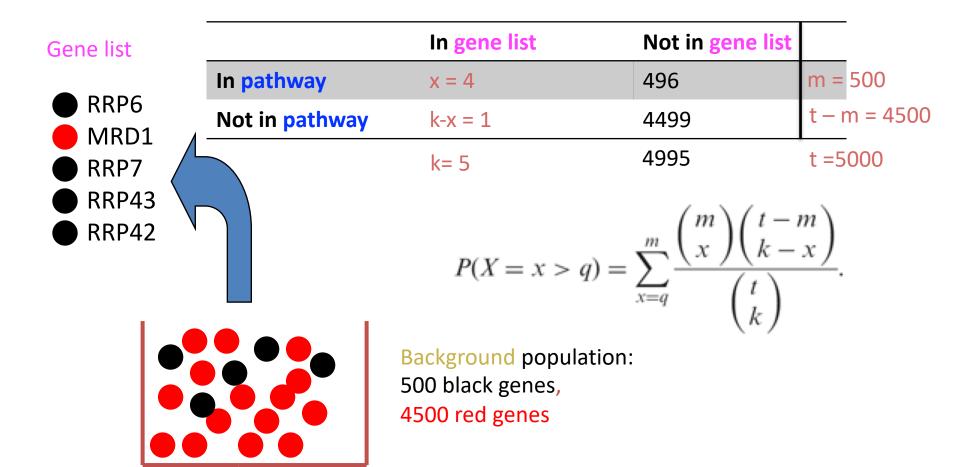
Background population: 500 black genes, 4500 red genes

The Fisher's exact test

a.k.a., hypergeometric test



2x2 contingency table for Fisher's Exact Test



Do you need to learn more about Fisher's exact test?

VIDEO the M&M's examples: https://www.youtube.com/watch?v=udyAvvaMjfM

StatQuest with Josh Starmer



gene sets



gene list















I'm going to use the histogram of the "ideal" bag of m&m's, based on proportions I got off the internet, and my "sample", my handful of m&m's, to determine if my bag is special



Background

Pathway Commons Guide:

https://www.pathwaycommons.org/guide/primers/statistics/fishers_exact_test/

And

Important points

- We usually test over-enrichment of "black". To test for underenrichment of "black", test for over-enrichment of "red".
- Need to choose "background population" appropriately, e.g.,
 if only portion of the total gene complement is queried (or
 available for annotation), only use that population as
 background.
- To test for enrichment of more than one independent types of annotation (red vs black and circle vs square), apply Fisher's exact test separately for each type.

g:Profiler

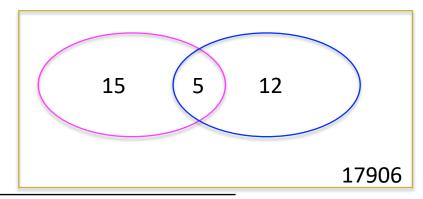
GO:BP		stats					<u> </u>
☐ Term name	Term ID	padj	_log10(padj) _{≤16}	Т	Q	T∩Q	U ↑
pulmonary valve morphogenesis	GO:0003184	1.034×10 ⁻⁸		17	20	5	17906
pulmonary valve development	GO:0003177	3.392×10 ⁻⁸		21	20	5	17906
regulation of myeloid leukocyte differentiation	GO:0002761	6.876×10 ⁻⁸		122	20	7	17906
regulation of osteoclast differentiation	GO:0045670	1.353×10 ⁻⁷		67	20	6	17906
		7					

T (term): pathway that is being tested

Q (query): my gene list

TnQ: overlap between pathway and gene list

U (universe): background



2x2 contingency table

	In gene list	Not in gene list
In pathway	5	12
Not in pathway	15	17894
	20	17906

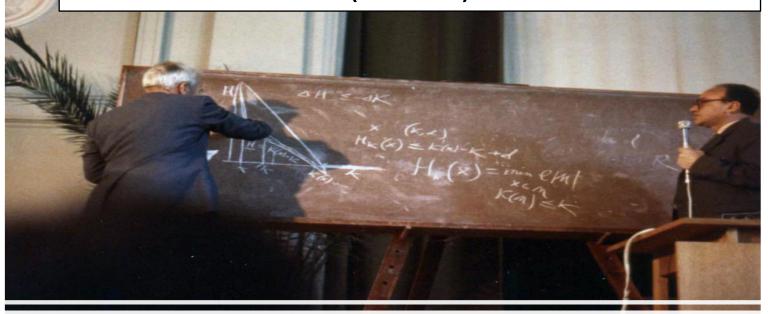
Other enrichment tests for defined gene lists (not covered in this lecture)

Note: Fisher's Exact Test is often called the hypergeometric test

- Approximation of the Fisher's Exact Test (Monte Carlo simulation)
- Binomial test
- Chi-squared test

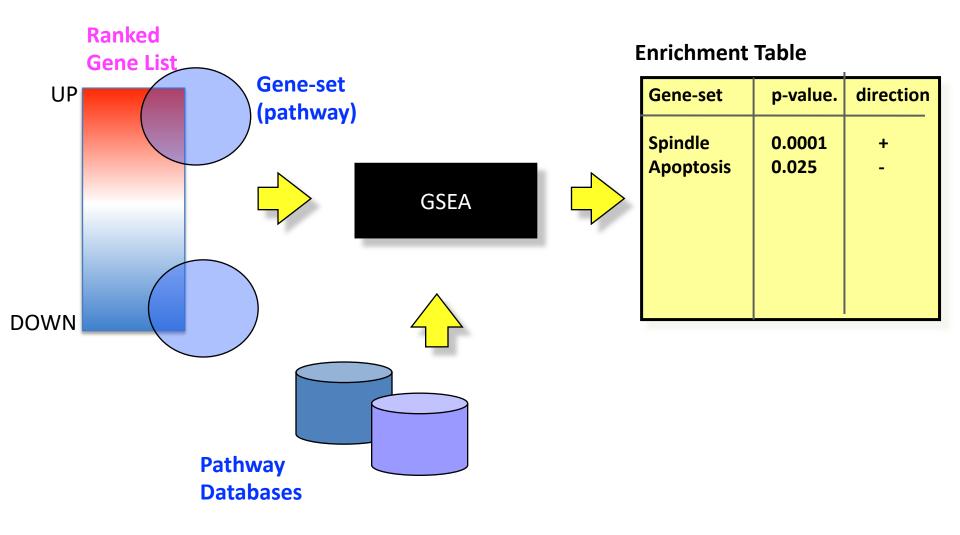
Ranked gene list enrichment test

GSEA → modified Kolmogorov Smirnov test (KS test)



https://en.wikipedia.org/wiki/Andrey_Kolmogorov#/media/File:Kolm_complexity_lect.jpg

Example of a ranked list enrichment test





- In their original paper, Mootha et al (2003) studied diabetes and identified that their gene list was significantly enriched in a pathway called "oxidative phosphorylation".
- The particularity of this finding was that individual genes in this
 pathway were only down-regulated by a small amount but the
 addition of all these subtle decreases had a great impact on the
 pathway.
- They validated their finding experimentally.

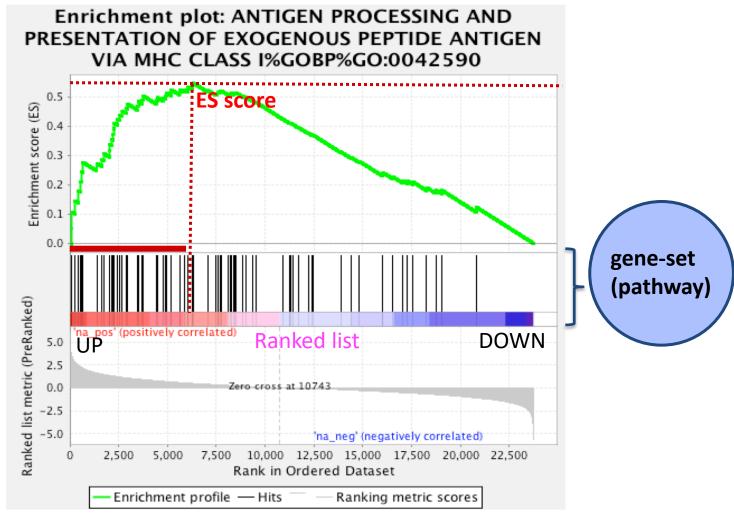
http://www.people.vcu.edu/~mreimers/HTDA/Mootha%20-%20GSEA.pdf

GSEA score calculation

Ranked gene list

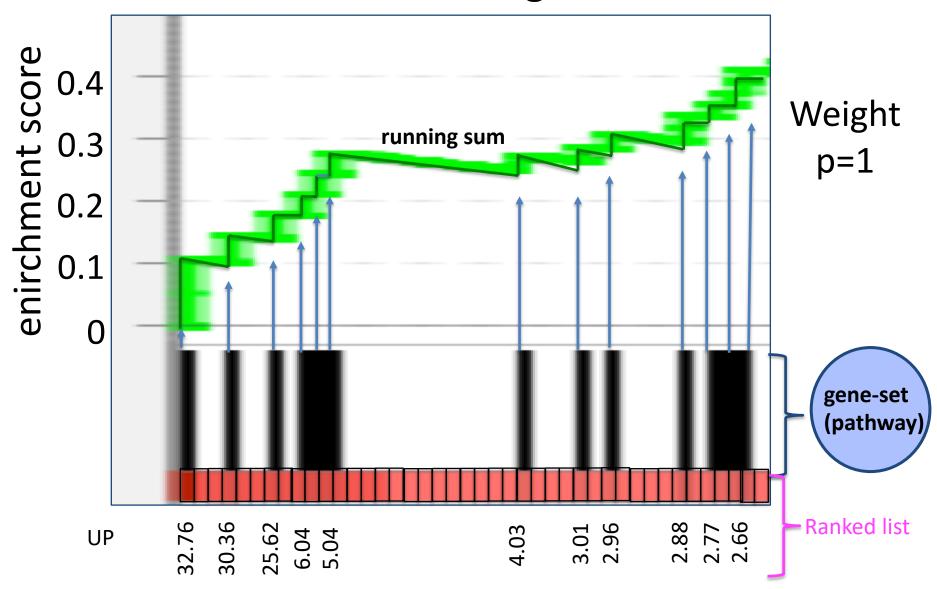
	UP
BGN	32.76
ANTXR1	30.36
FZD1	29.36
COL16A1	28.88
KLF3	1.08
RASEF	0.05
	•••
	•••
ISOC1	0.05
ANO1	0.04
CBWD3	-1.09
GBP4	-15.6
TAP1	-19
PSMB9	-19.7

DOWN

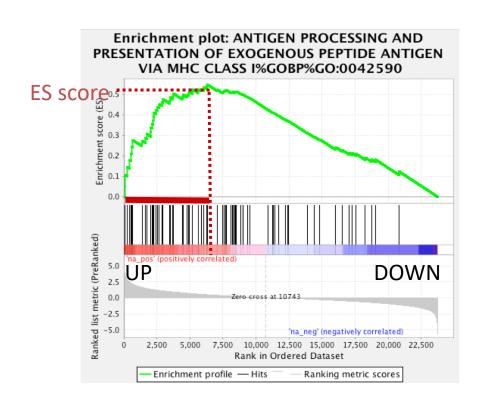


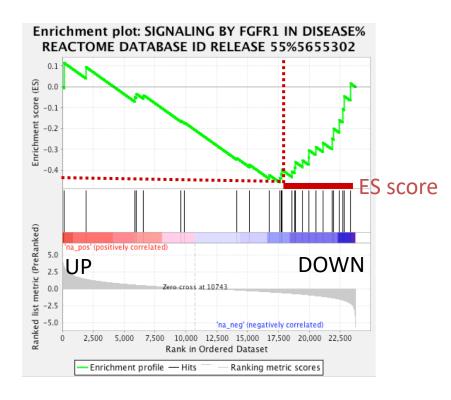
- 1. Maximum (or minimum) ES score is the final **ES score** for the gene set
- 2. Can define "leading edge subset" as all those genes ranked as least as high as the enriched set.

GSEA running sum



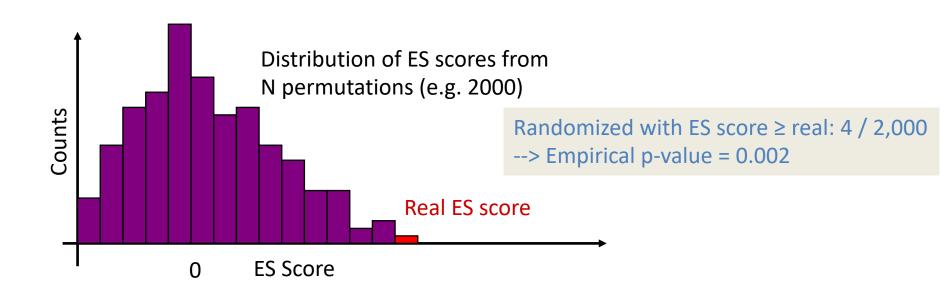
Positive and negative enrichment scores





Going from ES score → P-value

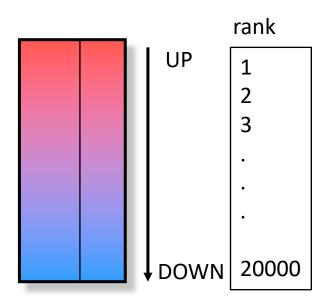
- 1.Generate null-hypothesis distribution from randomized data (see permutation settings)
- 2.Estimate empirical p-value by comparing observed ES score to null-hypothesis distribution from randomized data (for every gene-set)

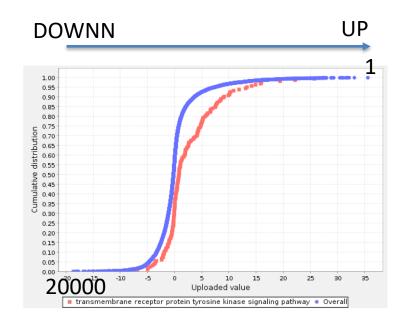


Other enrichment tests for

a ranked gene list

Wilcoxon ranksum test





Panther

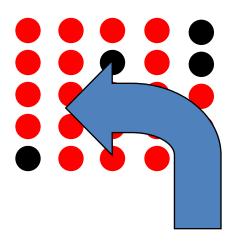
Outline of theory component

- Fisher's exact test (or binomial) for calculating enrichment P-values for defined gene lists
- GSEA, wilcoxon rank sum test for computing enrichment P-values for ranked gene lists

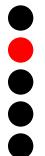
Multiple test corrections

How to win the p-value lottery

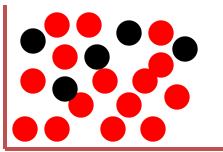
Random draws



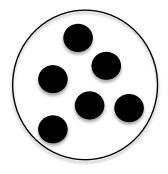
... 7,834 draws later ...



Expect a random draw with observed enrichment once every 1 / P-value draws



Background population: 500 black genes, 4500 red genes



1 gene-set (apoptosis)

Simple P-value correction: Bonferroni

If M = # of gene-sets (pathways) tested:

Corrected P-value = M x original P-value

Corrected P-value is greater than or equal to the probability that **one or more** of the observed enrichments could be due to random draws. The jargon for this correction is "controlling for the Family-Wise Error Rate (FWER)"

Bonferroni correction caveats

- Bonferroni correction is very stringent and can "wash away" real enrichments leading to false negatives,
- Often one is willing to accept a less stringent condition, the "false discovery rate" (FDR), which leads to a gentler correction when there are real enrichments.

False discovery rate (FDR)

- FDR is the expected proportion of the observed enrichments due to random chance.
- Compare to Bonferroni correction which is a bound on the probability that any one of the observed enrichments could be due to random chance.
- Typically FDR corrections are calculated using the Benjamini-Hochberg procedure.
- FDR threshold is often called the "q-value"

False discovery rate (FDR)

- 1. Sort P-values of all tests in increasing order
- 2. Adjusted P-value is "nominal" P-value times # of tests divided by the rank of the P-value in sorted list: P-value x [# of tests] / Rank
- 3. Q-value (or FDR) corresponding to a nominal P-value is the smallest adjusted P-value assigned to P-values with the same or larger ranks.
- 4. Look at which gene-sets have a **FDR of 0.05 or less** and report them as **significantly enriched**.

Benjamini-Hochberg example

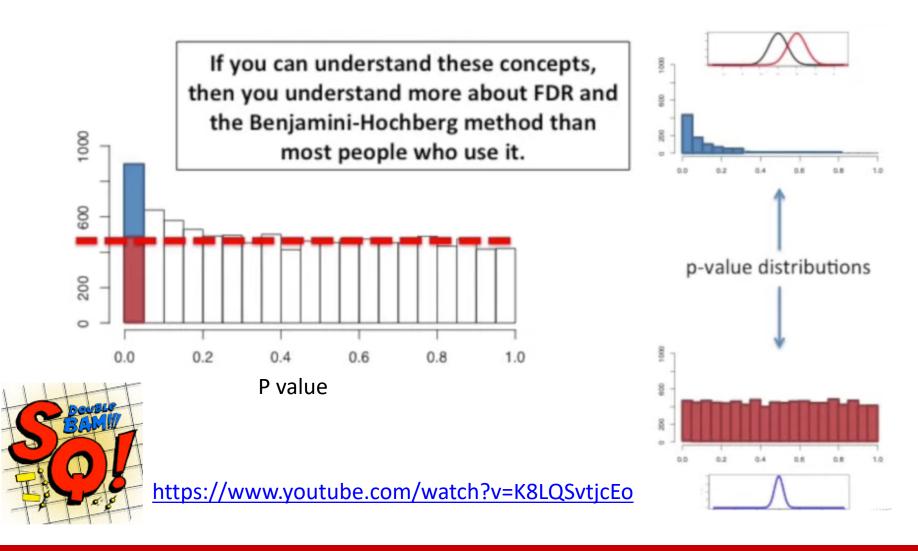
Rank	Category	(Nominal) P-value	Adjusted P-value	FDR / Q-value
1	Transcriptional regulation	0.001	$0.001 \times 53/1 = 0.053$	0.040
2	Transcription factor	0.002	$0.002 \times 53/2 = 0.053$	0.040
3	Initiation of transcription	0.003	$0.003 \times 53/3 = 0.053$	0.040
4	Nuclear localization	0.0031	$0.0031 \times 53/4 = 0.040$	0.040
5	Chromatin modification	0.005	$0.005 \times 53/5 = 0.053$	0.053
	•••		•••	
52	Cytoplasmic localization	0.97	$0.985 \times 53/52 = 1.004$	0.99
53	Translation	0.99	$0.99 \times 53/53 = 0.99$	0.99

Q-value (or FDR) corresponding to a nominal P-value is the smallest adjusted P-value assigned to P-values with the same or larger ranks.

Gene set enrichment significant at FDR < 0.05

How to win the p-value lottery, part 2

Keep the gene list the same, evaluate different gene-sets(pathways)



Reducing multiple test correction stringency

- The correction to the P-value threshold α depends on the # of tests that you do, so, no matter what, the more tests you do, the more sensitive the test needs to be
- Can control the stringency by reducing the number of tests: e.g. use GO slim; restrict testing to the appropriate GO annotations; or filter gene sets by size.

Summary

Multiple test correction

- Bonferroni: stringent, controls probability of at least one false positive*
- FDR: more forgiving, controls expected proportion of false positives* -- typically uses Benjamini-Hochberg

^{*} Type 1 error, aka probability that observed enrichment if no association

What Have We Learned?

Typical output of an enrichment analysis is:

Pathway name	Number of overlapping genes	Number of genes in pathway	P-value	Adjusted p- value

Typical output

gene-set name (pathway)

number of overlapping genes
... corrected for gene-set size

p-value

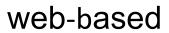
... corrected for multiple hypothesi

♦		+	+	
RNA HELICASE ACTIVITY%GO%GO:0003724	28	1.77	0.0041	0.0164386
MRNA SURVEILLANCE PATHWAY%KEGG%HSA03015	82	1.77	0	0.0466167
UBIQUITIN-DEPENDENT DEGRADATION OF CYCLIN D1%REACTOME%REACT_4.1	50	1.77	0.0021	0.0486015
BIOCARTA_CD40_PATHWAY%MSIGDB_C2%BIOCARTA_CD40_PATHWAY	15	1.77	0.0048	0.0483781
IGF1 PATHWAY%PATHWAY INTERACTION DATABASE NCI-NATURE CURATED DATA%IGF1 PATHWAY	29	1.76	0.003	0.0489742
UBIQUITIN-DEPENDENT PROTEIN CATABOLIC PROCESS%GO%GO:0006511	204	1.76	0	0.0488442
PHAGOSOME%KEGG%HSA04145	147	1.76	0	0.0486164
PROTEASOME COMPLEX%GO%GO:0000502	29	1.76	0.007	0.0490215
ANTIGEN PRESENTATION: FOLDING, ASSEMBLY AND PEPTIDE LOADING OF CLASS I MHC%REACTOME%REACT_7	24	1.76	0.0041	0.0505599
ABORTIVE ELONGATION OF HIV-1 TRANSCRIPT IN THE ABSENCE OF TAT%REACTOME%REACT_6261.3	23	1.75	0	0.0529242
DNA DAMAGE RESPONSE, SIGNAL TRANSDUCTION BY PSACLASS MEDIATOR RESULTING IN CELL CYCLE ARREST9	67	1.75	0	0.052886
REGULATION OF MACROPHAGE ACTIVATION%GO%GC 004 020-	11	1.75	0.003	0.0534709
PROTEIN FOLDING%REACTOME%REACT_16952.2	52	1.75	0.002	0.0537717
ENDOPLASMIC RETICULUM UNFOLDED PROTEIN SEPONSE GO%GO:0030968	73	1.75	0	0.0546052
PROTEIN EXPORT%KEGG%HSA03060	24	1.75	9.75E-04	0.0548699
TRANSCRIPTION INITIATION FROM RNA POLYMERASI. \ PROMO7 \ \%GO%GO:0006367	64	1.75	0.001	0.0545783
S PHASE%REACTOME%REACT_899.4	110	1.75	0	0.0546003
PROTEASOMAL PROTEIN CATABOLIC PRO SSS% 6GO:001	163	1.75	0	0.0550066
ATP-DEPENDENT RNA HELICASE ACTIVITY%GO GO:0004004	20	1.74	0.0059	0.0556722
ACID-AMINO ACID LIGASE ACTIVITY%GO SO: 016881	217	1.74	0	0.0560217
G0%G0:0072474	67	1.74	0.002	0.0565978
GO%GO:0035966	107	1.74	0	0.0562957
GO%GO:0072413	67	1.74	9.81E-04	0.05761
BIOCARTA_IL4_PATHWAY%N 162%BIOCA TAATHWAY	11	1.74	0.0082	0.0581508
ASSOCIATION OF TRIC CC WITH ARGET PROTEINS VRING BIOSYNTHESIS%REACTOME%REACT_16907.2	28	1.74	0.0039	0.0581298
UBIQUITIN-DEPENDENT S. RADATION OF LYCLIN 1%REACTOME%REACT_938.4	50	1.74	0.0029	0.057876
MODIFICATION-DEPENDENT PROTEIN CATA OLIC ROCESS%GO%GO:0019941	207	1.74	0	0.0576579
TRANSLATION INITIATION COMPLEX F RMATION%REACTOME%REACT_1979.1	55	1.74	0.0021	0.0575181
GO%GO:0001906	13	1.74	0.0117	0.0572877
G1 S TRANSITION%REACTOME%REACT_17s2	107	1.74	0	0.0572618
GO%GO:0034620	73	1.73	0.0021	0.0576606
SIGNALING BY NOTCH%REACTOME%REACT_299.2	19	1.73	0.0069	0.0578565
RESPONSE TO UNFOLDED PROTEIN%GO%GO:0006986	102	1.73	0	0.0583864
SIGNAL TRANSDUCTION INVOLVED IN G1 S TRANSITION CHECKPOINT%GO%GO:0072404	68	1.73	0.002	0.0582213
GO%GO:0072431	67	1.73	0	0.058551
BIOCARTA_PROTEASOME_PATHWAY%MSIGDB_C2%BIOCARTA_PROTEASOME_PATHWAY	19	1.73	0.0099	0.0586655
HOST INTERACTIONS OF HIV FACTORS%REACTOME%REACT_6288.4	117	1.73	0	0.0586888
AUTOPHAGIC VACUOLE ASSEMBLY%GO%GO:0000045	13	1.73	0.0122	0.0588271
CYCLIN A:CDK2-ASSOCIATED EVENTS AT S PHASE ENTRY%REACTOME%REACT_9029.2	66	1.73	0	0.0610099
2000 1000 200 200 200 200 200 200 200 20				

NETWORK VISUALIZATION

Many available enrichment analysis tools







Cytoscape app



Standalone



R package

How to choose a tool?

- Does it cover your model organism?
- Is there a good choice of gene-sets (pathway database)
- Are the pathway databases up to date?
- Which statistics (for gene list or ranked gene list)?
- Is the description of statistics clear enough?
- Do you like the output style?
- Can you connect it with network visualization tools like Cytoscape?

Defined gene list (Fisher's exact test)

	g:Profiler	PANTHER	biNGO	Cluego
Updated database	yes	yes	no? *1	yes
Choice of database (more than 1)	yes	yes	no (GO) *1	yes
Do we test database individually or together	together	individually	individually	together
Multiple model organisms?	yes	yes	yes	yes
Possibility to upload your own custom database	yes	no?	yes	no?
Statistics: possibility to use the Fisher's exact test (ORA) (thresholded gene list)	yes	yes	yes	yes
Multiple hypothesis correction; possibility to use B-H FDR	yes	yes	yes	yes
Possibility to upload reference genes (background)	yes	yes	yes	yes
Website (Web) or Cytoscape App (App)	Web	Web	Арр	Арр
Possibility to visualize with Cytoscape EnrichmentMap	YES	no	YES	Cytoscape

^{*1:} can still be used with custom database ;

Ranked list

	GSEA	PANTHER
Rank test	Modified KS test	Wilcoxon Rank Sum test
Correction for multiple hypothesis testing	yes	yes
Possibility to visualize results with Cytoscape enrichment map	yes	no

Recipe for defined gene list enrichment test

- Step 1: Define your gene list and your background list,
- **Step 2:** Select your gene sets (pathways) to test for enrichment,
- Step 3: Run enrichment tests using the Fisher's exact test and correct for multiple testing if you test more than one gene set (pathway)
- **Step 4:** Interpret your enrichments
- Step 5: Publish! ;)

Recipe for ranked list enrichment test

- Step 1: Rank your genes,
- **Step 2:** Select your gene sets (pathways) to test for enrichment,
- Step 3: Run enrichment tests and correct for multiple testing, if necessary,
- Step 4: Interpret your enrichments
- Step 5: Publish! ;)

Advanced topics (not covered in this lecture)

- Issues with tests: correlation between gene-sets, dependency of genes.
- Other types of tools: topology aware.
- Modern tools are starting to include some network visualization.

Go to: Pathway enrichment analysis and visualization of omics data using g:Profiler, GSEA, Cytoscape and EnrichmentMap

https://www.nature.com/articles/s41596-018-0103-9

Tips

Be precise at each step of your analysis

 Try to answer one biological question at a time

We are on a Coffee Break & Networking Session



Workshop Sponsors:





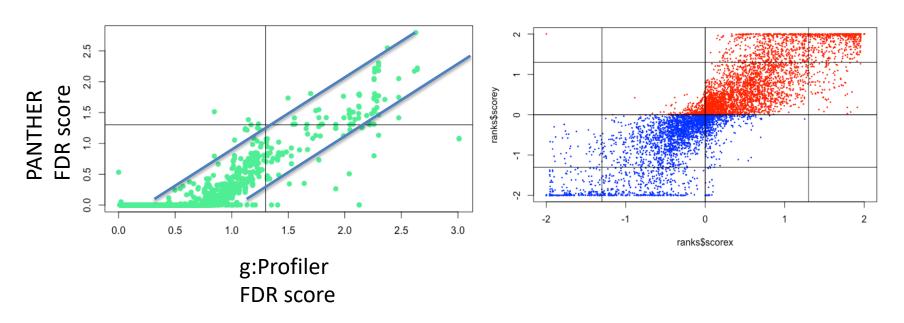


Additional slides

Comparison of results

gene list

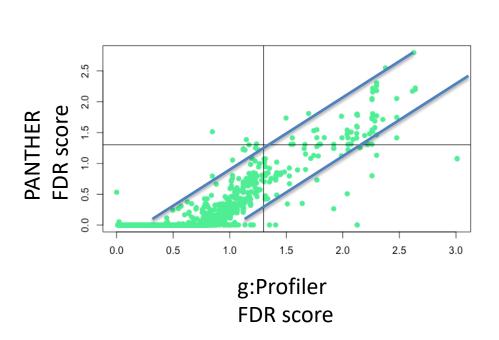
Ranked list



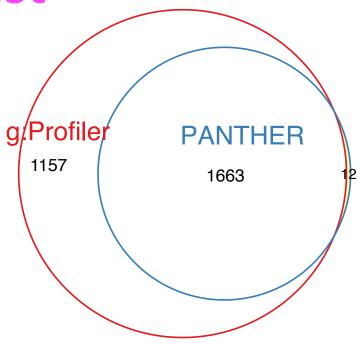
Similar results are obtained between g:Profiler and PANTHER

Comparison of results

gene list

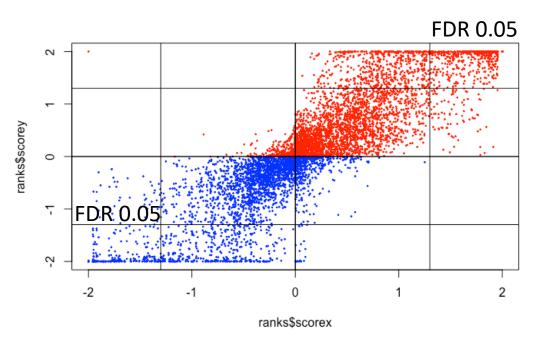


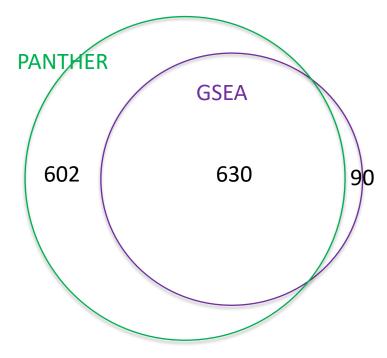
g:Profiler 16132 gene-sets PANTHER 15815 gene-sets



Number of gene-sets significant under FDR < 0.05

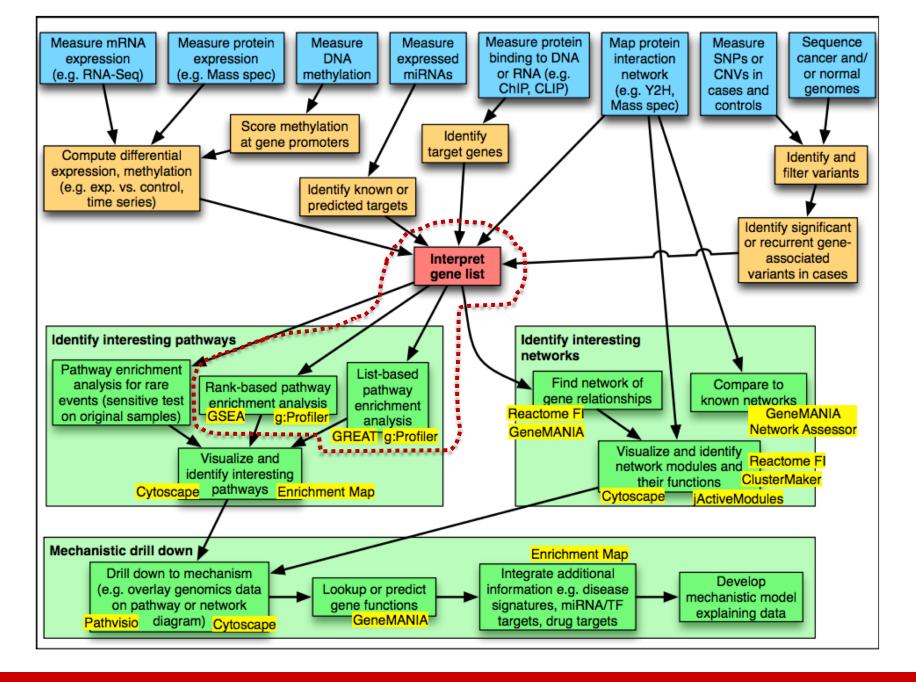
Ranked list



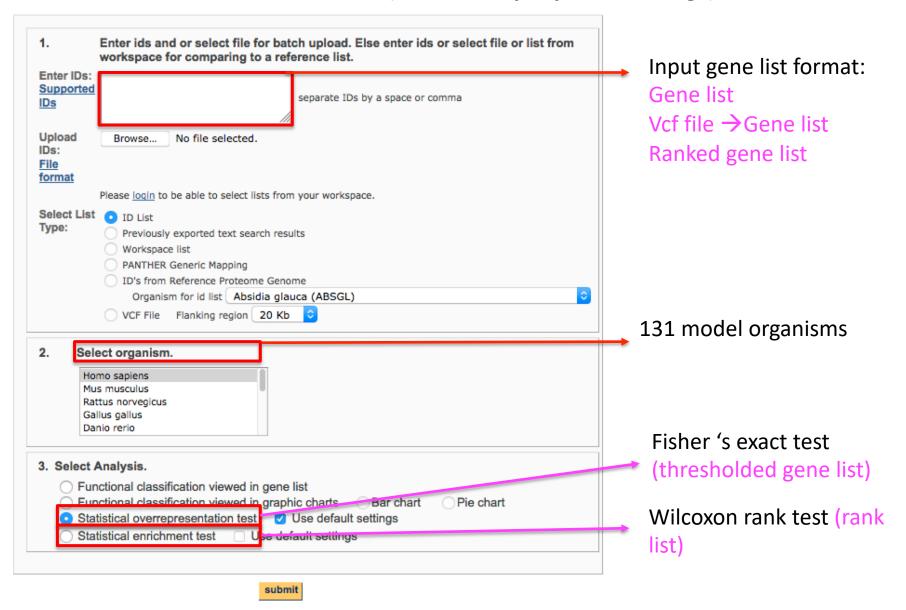


Number of gene-sets significant under FDR < 0.05

- gene-set enriched in genes up-regulated
- gene-set enriched in genes down-regulated

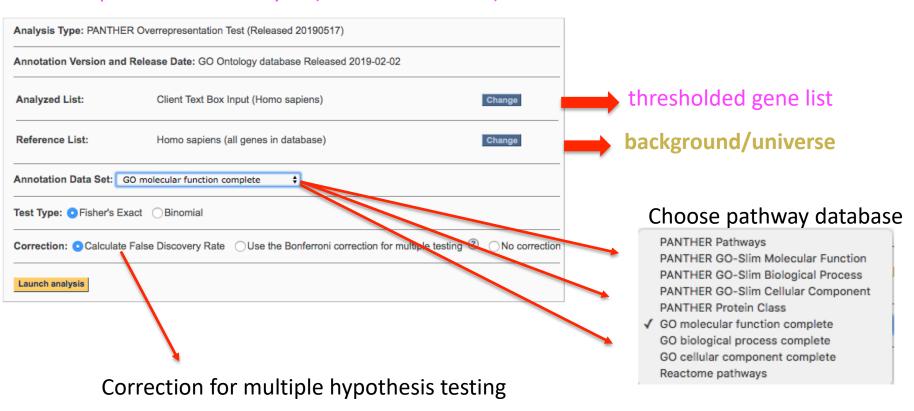


PANTHER (website: http://pantherdb.org/)



PANTHER (website)

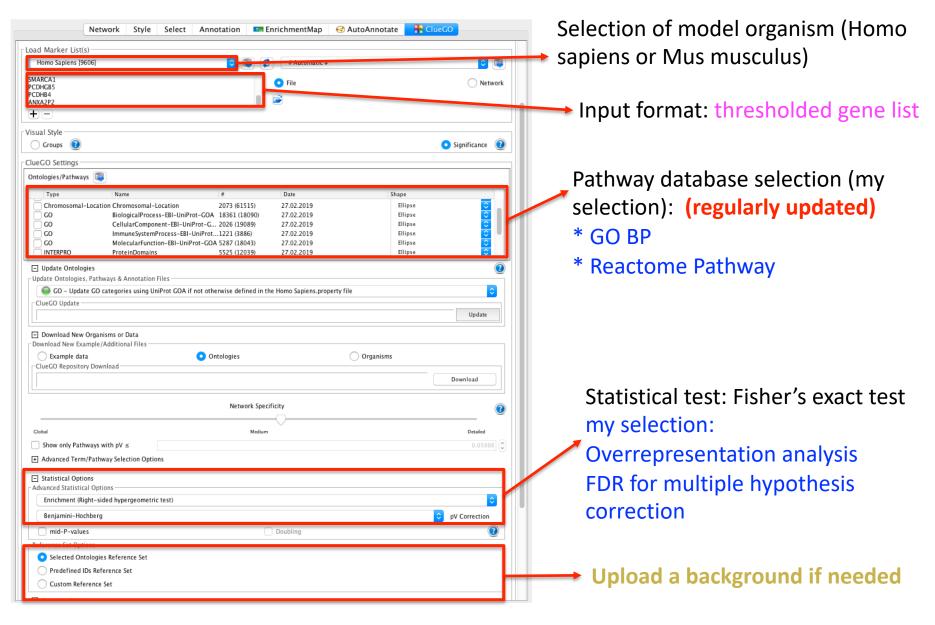
Over-representation Analysis (Fisher's exact test)



Updated frequently!!

Only text output visualization?

ClueGO (Cytoscape app)



Input

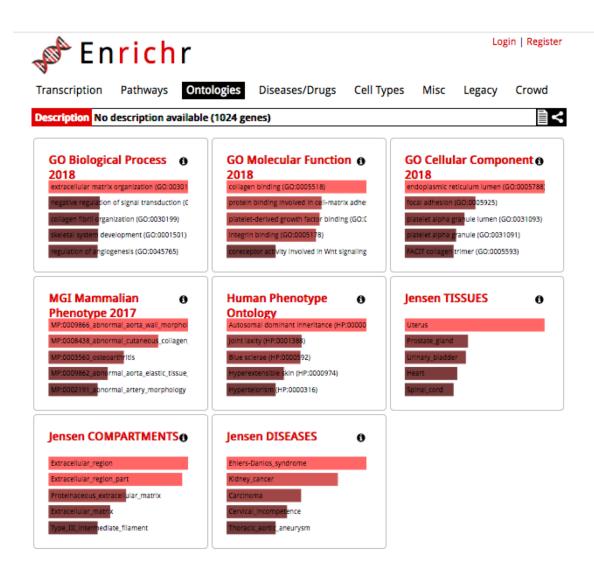
Gene list or Bed file

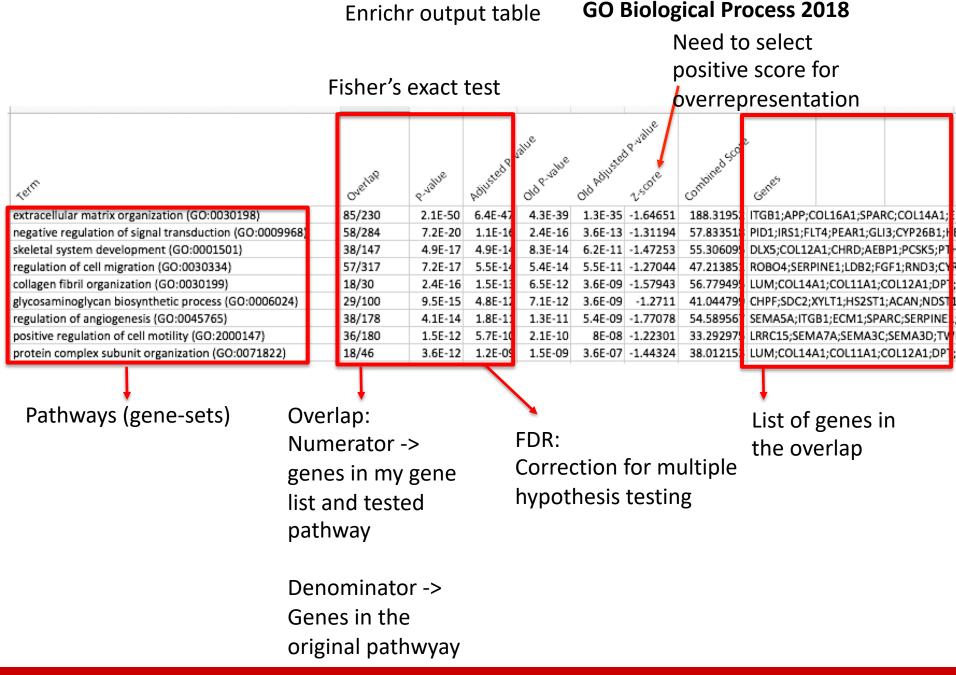
Test:
Fisher's exact test,
pvalue corrected
for multiple hypothesis

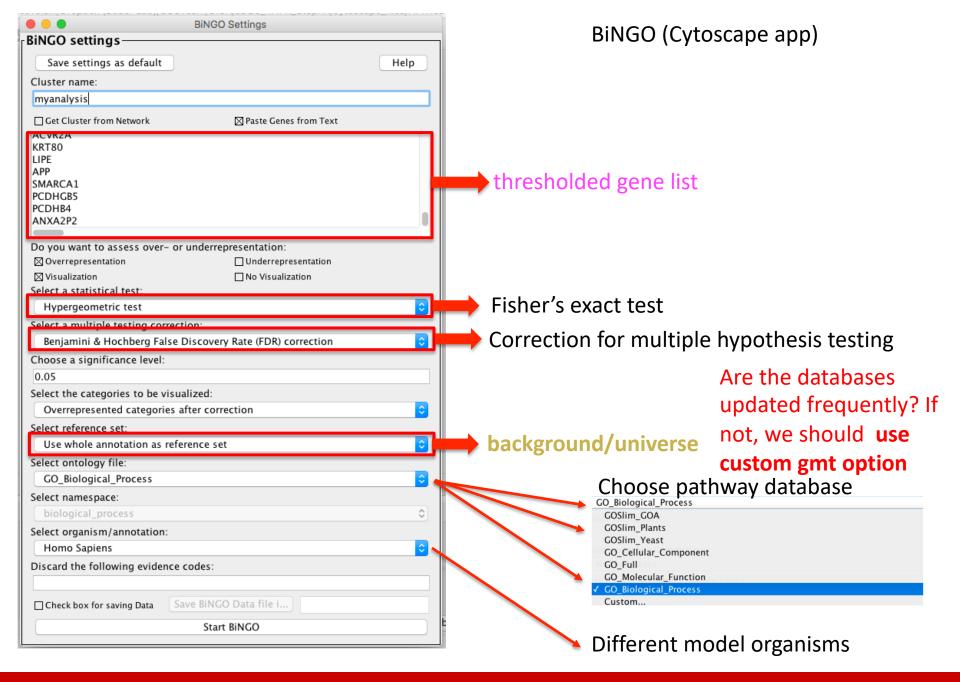
Output: Table or grahs

testing

? No option to put Reference background (use only if you are doing a whole genome study)







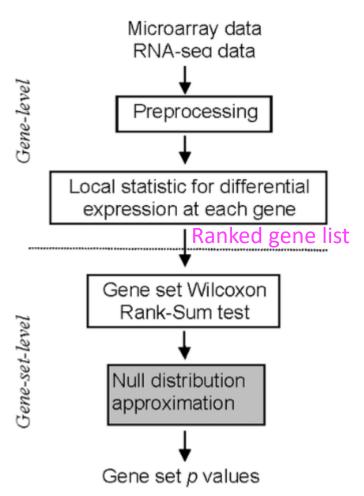
BiNGO output

	_,			J .	••	5.
GO-ID	Description	p-val	corr p-val	cluster freq	total freq	genes
48731	system development	8.7490E-55	3.1041E-51	316/805 39.2%	2416/14265 16.9%	SPON2 ERRFI1 APP SPARC SERPINE1 COL12A1 XYLT1 STMN3 ELK3 AQP1 NDST1
48856	anatomical structure development	1.1175E-52	1.9824E-49	330/805 40.9%	2649/14265 18.5%	SPON2 ERRFI1 APP SPARC SERPINE1 COL12A1 XYLT1 STMN3 ANTXR1 ELK3 AQP1
	multicellular organismal development	6.5566E-52	7.7543E-49	352/805 43.7%	2965/14265 20.7%	SPON2 ERRFI1 APP SPARC SERPINE1 COL12A1 DIXDC1 XYLT1 STMN3 ELK3 AQP1
32502	abelFor=,text=developmental process,verticalAlignment=CENTER,verticalTextPosition=CENTER	5.2369E-49	4.6452E-46	365/805 45.3%	3227/14265 22.6%	SPON2 ERRFI1 APP SPARC SERPINE1 COL12A1 DIXDC1 XYLT1 STMN3 ANTXR1 EL
7155	cell adhesion	1.9208E-47	1.3630E-44	149/805 18.5%	711/14265 4.9%	SEMA5A SPON2 APP COL16A1 COL12A1 ANTXR1 CTGF LOXL2 COMP CDH5 ISLR
22610	biological adhesion	2.3126E-47	1.3675E-44	149/805 18.5%	712/14265 4.9%	SEMA5A SPON2 APP COL16A1 COL12A1 ANTXR1 CTGF LOXL2 COMP CDH5 ISLR
9653	anatomical structure morphogenesis	7.9139E-38	4.0112E-35	184/805 22.8%	1214/14265 8.5%	SEMA5A SPON2 ERRFI1 APP SERPINE1 FGF1 ANTXR1 CTGF ELK3 AQP1 COMP NDS
48513	organ development	9.3874E-37	4.1633E-34	231/805 28.6%	1788/14265 12.5%	ERRFI1 APP SERPINE1 ELK3 AQP1 NDST1 GJA1 EDNRA KDR HOXA3 HOXA1 SOX7 S
1944	vasculature development	6.7888E-32	2.6763E-29	75/805 9.3%	273/14265 1.9%	SEMA5A ERRFI1 ROBO4 TCF21 SHB FGF1 GLI3 CYR61 CTGF ELK3 CDH5 GJA1 EDN
1568	blood vessel development	2.6225E-30	9.3046E-28	72/805 8.9%	265/14265 1.8%	SEMA5A ROBO4 SHB FGF1 GLI3 CYR61 CTGF ELK3 CDH5 GJA1 EDNRA PLAU KDR
32501	multicellular organismal process	6.4334E-30	2.0750E-27	396/805 49.1%	4368/14265 30.6%	SERPINE1 COL12A1 XYLT1 STMN3 ELK3 NDST1 GPR176 GJA1 DPYSL4 DPYSL3 HO

5.

- 1. each row is a pathway (gene-set) that was in the original pathway database that we selected (GO Biological Process)
- 2. corr pval: the most important column as it is the pvalue corrected for multiple hypothesis.
- 3 .information about the size of the overlap between my gene list and and the pathway (gene-set).
- 4. information about the size of the original pathway (gene-set) in the chosen pathway database (GO biological process).
- 5. Genes in the my gene list and in the pathway (gene-set).

Wilcoxon rank sum test (Mann-Whitney U Test)



- Rank based non parametric test for comparing two groups of observations without the assumption of certain distributions.
- It has been implemented in many packages and software for gene-set testing (limma R function geneSetTest, R package SAFE using safe(), Gostat(), Panther.

https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0031505

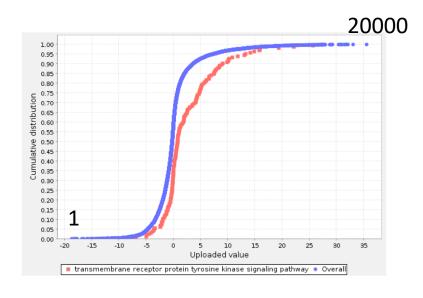
Wilcoxon rank sum test

(as used and described in PANTHER tool -pantherdb.org/)

Step 1 out of 3

All genes rank sum:

- All genes ordered by expression values to create a rank.
- Genes with the smallest values get a rank of 1.
- A rank sum is calculated (summing up the ranks for all genes)
- The average rank, R2, is calculated by dividing the rank sum by the total number of genes uploaded, n2.



Rank sum = 1 + 2 +3 +....+ 20000 n2 = 20000 R2 = rank sum / n2

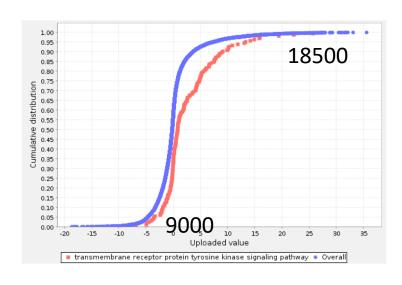
Wilcoxon rank sum test

(as used and described in PANTHER tool -pantherdb.org/)

Step 2 out of 3

Gene set rank sum:

- A rank sum calculated for genes in the tested gene-set: sum up the ranks for all genes that map the gene-set.
- The average rank, R1 is then calculated by dividing the rank sum by the number of genes, n1, that map to the category.



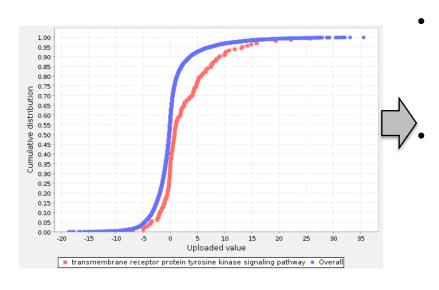
Rank sum = 9000 + 9005 + ...18500 n1 = 250 (size of gene set) R1 = rank sum / n1

Wilcoxon rank sum test

(as used and described in PANTHER tool -pantherdb.org/)

Step 3 out of 3

- The Mann Whitney U statistic is calculated :
 - U1 = n1* n2 + (n1 * (n1 + 1)) / 2 R1 (gene-set)
 - U2 = n2* n2 + (n1 * (n2 + 1)) / 2 R2 (all genes)
 - U: The larger of these two values is the Mann Whitney U-statistic,
 - Pvalue associated with U, whose distribution for small sample sizes can be found in most statistic books or use the normal approximation (Z-score = (U- (n1* n2)/2)/sqrt(n1*n2*(n1+n2+1)/12).



- The distribution of values for your uploaded list is shifted towards greater values than the overall distribution of all genes that were uploaded.
- A small, significant p-value indicates that the distribution for this category is non-random and different than the overall distribution. A cutoff of 0.05 is recommended as a starting point.

Minimum hypergeometric test (mHG)

(used in g:Profiler, ordered query)

Steps

1. Calculate p-value at multiple thresholds

1. Correct for multiple testing (or compute empirical p-values using permutations)

Eden E, Lipson D, Yogev S, Yakhini Z. Discovering motifs in ranked lists of DNA sequences. PLoS Comput Biol. 2007 Mar 23;3(3):e39

mHG Method

mHG score calculation

gene-set
(pathway)

thresholded gene list

Where are the gene-set genes located in the ranked list? Is there distribution random, or is there an enrichment in either end?

Gene list (threshold) chosen is the one associated with the lowest pvalue