

Departement Biomedizin Basel

Bioinfo Seminar, DBM

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### Outline

- 1. Overrepresentation analysis (ORA), mathematical formulation
- 2. enrichR: online tool for ORA
- 3. Critical evaluation of enrichR, pros and cons
- 4. Alternative: clusterProfiler

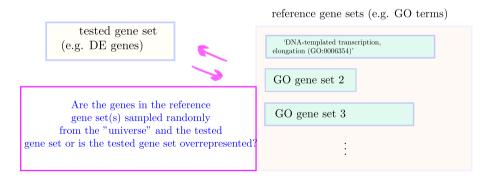
presentation inspired by
https://www.pathwaycommons.org/guide/primers/statistics/fishers\_exact\_test/

## Task:

Check if a tested gene set shares an 'unusual large number of genes' with a reference gene set.

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- a simple case:
  - 30 genes expressed in total (background)
  - 15 genes are DE (tested gene set)
  - 12 genes overlap with a reference gene set (transcription/elongation GO term)

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	Differential Expression	NO Differential Expression	Total
IN Transcription Elongation	12	3	15
NOT IN Transcription Elongation	Transcription 3		15
Total	15	15	30

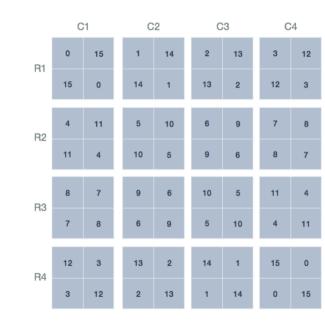
Fisher's exact test

null hypothesis  $H_0$ : The two gene sets are independent

Establish the p value for our observation.

# Fisher's exact test

The number of possible configurations (contingency tables) is finite but not all are equally probable



### Fisher's exact test

$$\mathcal{P} = \frac{\binom{15}{1}\binom{15}{14}}{\binom{30}{15}} = \frac{\binom{15}{1}^2}{\binom{30}{15}} = 1.45 \times 10^{-6}$$

## Hypergeometric distribution

$$f(x; N, n, r) = \frac{\binom{r}{x} \binom{N-r}{n-x}}{\binom{N}{n}} \text{ for } x = 0, 1 \dots r$$

x: random variable (overlap), N: background size, r: size of the tested gene set, n: size of the reference gene set

	Differential Expression	NO Differential Expression	Total
IN Transcription Elongation	1	14	15
NOT IN Transcription 14 Elongation		1	15
Total	15	15	30

probabilities: C2 C3 C4 C1 Fisher's exact test 15 0 14 2 13 3 12 R1 12 15 Λ 14 13 2 3 p = 6.45E-09p = 1.45E-06p = 7.11E-05p = 1.33E-03One-sided (directional) test: 11 5 10 9 7 8 R2 11 10 5 9 8  $pval = 1.33 \times 10^{-2} + 7.11 \times 10^{-5}$ p = 1.20E-02p = 5.81E-02p = 1.61E-02p = 2.67E-02 $+ 1.45 \times 10^{-6} + 6.45 \times 10^{-9}$ 8 9 6 10 11 4 R3 = 0.001407159Q 10 4 11 p = 2.67E-02p = 1.61E-02p = 5.81E-02p = 1.20E-0212 13 14 15 R4 13 15

p = 1.33E-03

p = 7.11E-05

p = 1.45E-06

p = 6.45E-09



50,441,107 sets analyzed 403,883 terms

Analyze

What's new?

Libraries

Gene search

Term search

About

Help

199 libraries

# Input data

Expand a gene, a term, or a variant into a gene set:

e.g. STAT3, breast cancer, or rs28897756



Try an example STAT3 breast cancer rs28897756

Include the top 100 most relevant genes

Paste a set of valid Entrez gene symbols on each row in the text-box below. Try a gene set example.

Paste a set of valid Entrez gene symbols (e.g. STAT3) on each row in the text-box

0 gene(s) entered

In order to enable others to search your set please enter a brief description of it.

☐ Contribute your set so it can be searched by others

Submit



Transcription

Pathways

Ontologies

Diseases/Drugs

Cell Types

a

a

Misc

Legacy

Crowd

A

Description No description available (234 genes)



#### **BioPlanet 2019**

a

Beta-1 integrin cell surface interactions

ECM-receptor interaction

Collagen biosynthesis and modifying enzym

Extracellular matrix organization

#### WikiPathway 2021 Human

miRNA targets in ECM and membrane recep

Focal Adhesion WP306

EGFA-VEGFR2 Signaling Pathway WP3888

ocal Adhesion-PI3K-Akt-mTOR-signaling par

pe I collagen synthesis in the context of O

KEGG 2021 Human

ECM-receptor interaction

Focal adhesion

rotein digestion and absorption

PI3K-Akt signaling pathway

roteoglycans in cancer

### ARCHS4 Kinases Coexp

DDR2 human kinase ARCHS4 coexpression

PDGFRB human kinase ARCHS4 coexpressio

PDGFRA human kinase ARCHS4 coexpressio

PS6KA2 human kinase ARCHS4 coexpression AYLK human kinase ARCHS4 coexpression

Elsevier Pathway Collection

Proteins with Altered Expression in Cancer N

Proteins Involved in Gliobiastoma

vadopodia Formation in Cancer Cells

oma Invasion Signaling

novial Fibroblast Proliferation in Rheumat

### MSigDB Hallmark 2020

Epithelial Mesenchymal Transition

### MSigDB Hallmark 2020

Bar Graph Table

.

Clustergram

Appyter



Hover each row to see the overlapping genes.

10 \$ entries per page

Search:

Index	Name	P-value	Adjusted p- value	Odds Ratio	Combined score
1	Epithelial Mesenchymal Transition	3.880e-51	1.707e-49	34.41	3993.73
2	Angiogenesis	1.715e-7	0.000001510	20.99	326.95
3	Apical Junction	2.478e-11	5.453e-10	8.97	218.98
4	Coagulation	8.946e-9	1.312e-7	9.24	171.29
5	Complement	1.052e-7	0.000001158	6.70	107.63
6	IL-2/STAT5 Signaling	6.680e-7	0.000004899	6.19	88.05
7	UV Response Dn	0.00005096	0.0003203	5.82	57.49
8	Myogenesis	0.0001284	0.0007060	4.60	41.22
9	Protein Secretion	0.0009280	0.004083	5.75	40.17
10	Apoptosis	0.0006305	0.003082	4.54	33.44

Showing 1 to 10 of 44 entries | Export entries to table Terms marked with an \* have an overlap of less than 5

Previous Next

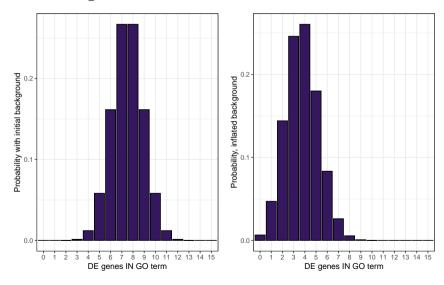
enrichR: not adjustable

genes, which has advantages and disadvantages. Enrichr does not have an ID conversion tool, which is highly desired by many users. Enrichr also does not have the ability to upload a background list, and it does not have implementation of parametric tests such as Gene Set Enrichment Analysis (GSEA) (40), Parametric Analysis of Gene set Enrichment (PAGE) (9), and our own Principal Angle Enrichment Analysis (PAEA) (41). These features are planned.

## Effect of the background

```
129
130
     N <- 30 #universe
132
     n <- 15 #reference gene set
133
134
     probabilities <- dhyper(x, r, N - r, n, log = FALSE)</pre>
     pvalue <- sum(probabilities[13:16])</pre>
135
136
     pvalue
137
                                  dhyper(x, m, n, k, log = FALSE)
138
139
     probabilities.BG <- dhyper(x, r, 2*N - r, n, log = FALSE)</pre>
140
     pvalue.BG <- sum(probabilities.BG[13:16])</pre>
141
     pvalue.BG
142
```

## Effect of the background



### Alternative: clusterProfiler

calling the function in R

```
> head(m_tZg.2)

# A tlbble: 6 x 2
gs_name
<chr>
<chr>
A tlbble: 6 x 2
gs_name
<chr>
A tlbble: 7 x 3
grading via NF-kB 18081

A tlbble: 7 x 3
grading via NF-kB 18081

A tlbble: 7 x 3
grading via NF-kB 17691

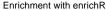
B tlbble: 7 x 3
grading via NF-kB 17691

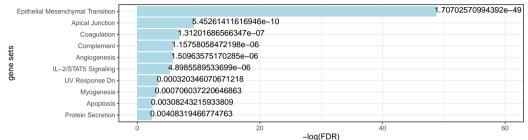
A tlbble: 7 x 3
grading via NF-kB 17691

B tlbble: 7 x 3
grading via NF-kB 11910
```

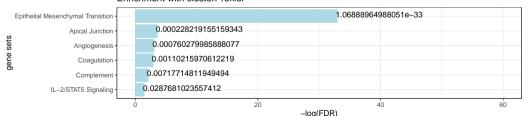
By defauls, the background is formed by the union of genes from the tested data sets

## Comparison





#### Enrichment with clusterProfiler



### Summary

- enrichR is quite widely used mainly for its userfriendliness
- Impressive collection of databases, updated
- Biggest problem: not adjustable (background!!!), apparently very broad

   → false positives
- There are alternatives but not so easy to use. However, more contralable (clusterProfiler)
- Despite the test being exact, different implementations (packages) give somewhat different results

# Summary

