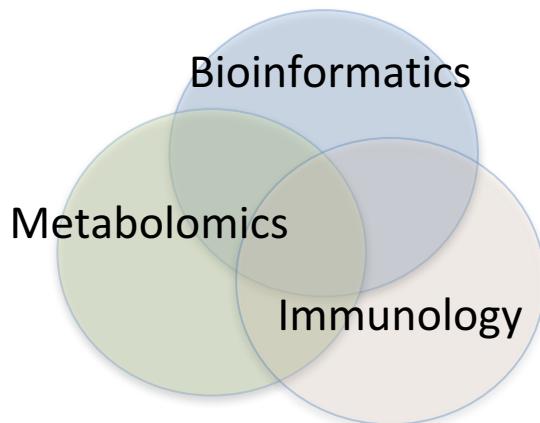




# Metabolomics Pathway Analysis



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**November 30, 2018**

# Outline

- What are pathways and metabolic models
- Pathway analysis for targeted data
- Untargeted data analysis and *mummichog*
- Application examples
  - Intracellular mechanisms
  - Population studies
  - Multi-omics integration
- Resources and issues

# Metabolic pathways we know today

→ C | 🔒 <https://www.genome.jp/kegg/pathway.html#global>

## **1.0 Global and overview maps**

- 01100 Metabolic pathways
  - 01110 Biosynthesis of secondary metabolites
  - 01120 Microbial metabolism in diverse environments
  - 01130 Biosynthesis of antibiotics
  - 01200 Carbon metabolism
  - 01210 2-Oxocarboxylic acid metabolism
  - 01212 Fatty acid metabolism
  - 01230 Biosynthesis of amino acids
  - 01220 Degradation of aromatic compounds

## 1.1 Carbohydrate metabolism

- 00010 Glycolysis / Gluconeogenesis
  - 00020 Citrate cycle (TCA cycle)
  - 00030 Pentose phosphate pathway
  - 00040 Pentose and glucuronate interconversions
  - 00051 Fructose and mannose metabolism
  - 00052 Galactose metabolism
  - 00053 Ascorbate and aldarate metabolism
  - 00050 Starch and sucrose metabolism
  - 00052 Amino sugar and nucleotide sugar metabolism
  - 00060 Pyruvate metabolism
  - 000630 Glyoxylate and dicarboxylate metabolism
  - 000640 Propanoate metabolism
  - 000650 Butanoate metabolism
  - 000660 C5-Branched dibasic acid metabolism
  - 000562 Inositol phosphate metabolism

## 1.2 Energy metabolism

- 00190 Oxidative phosphorylation
  - 00195 Photosynthesis
  - 00196 Photosynthesis - antenna proteins
  - 00710 Carbon fixation in photosynthetic organisms
  - 00720 Carbon fixation pathways in prokaryotes
  - 00680 Methane metabolism
  - 00910 Nitrogen metabolism
  - 00920 Sulfur metabolism

### 1.3 Lipid metabolism

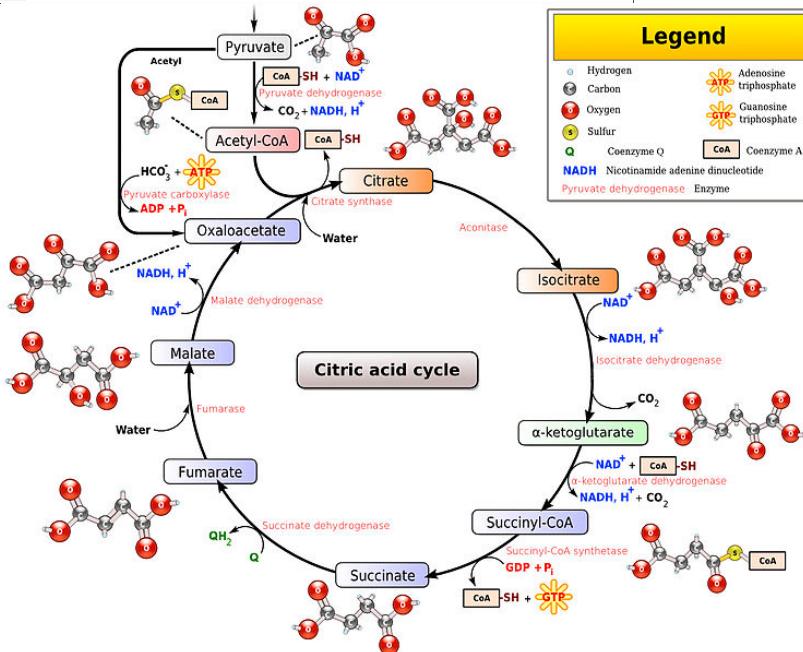
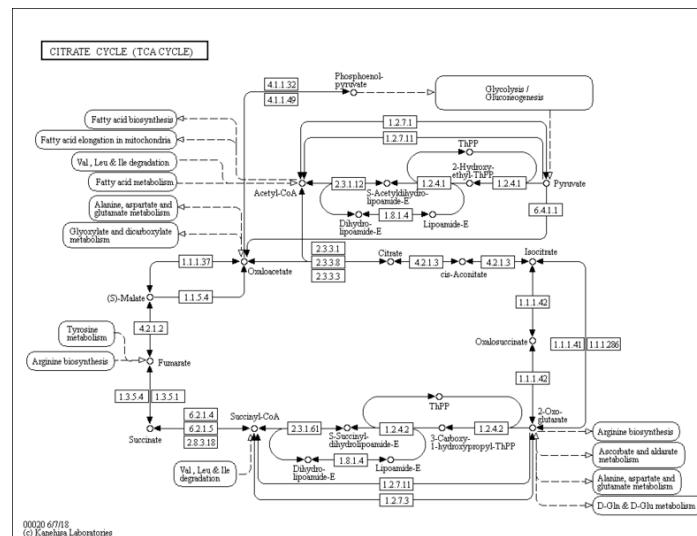
- 00061 Fatty acid biosynthesis
  - 00062 Fatty acid elongation
  - 00071 Fatty acid degradation
  - 00072 Synthesis and degradation of ketone bodies
  - 00073 Cutin, suberin and wax biosynthesis
  - 00100 Steroid biosynthesis
  - 00120 Primary bile acid biosynthesis
  - 00121 Secondary bile acid biosynthesis
  - 00140 Steroid hormone biosynthesis
  - 00561 Glycerolipid metabolism
  - 00564 Glycerophospholipid metabolism
  - 00565 Ether lipid metabolism
  - 00600 Sphingolipid metabolism
  - 00590 Arachidonic acid metabolism
  - 00591 Linoleic acid metabolism
  - 00592 alpha-Linolenic acid metabolism
  - 01040 Biosynthesis of unsaturated fatty acids

## 1.4 Nucleotide metabolism

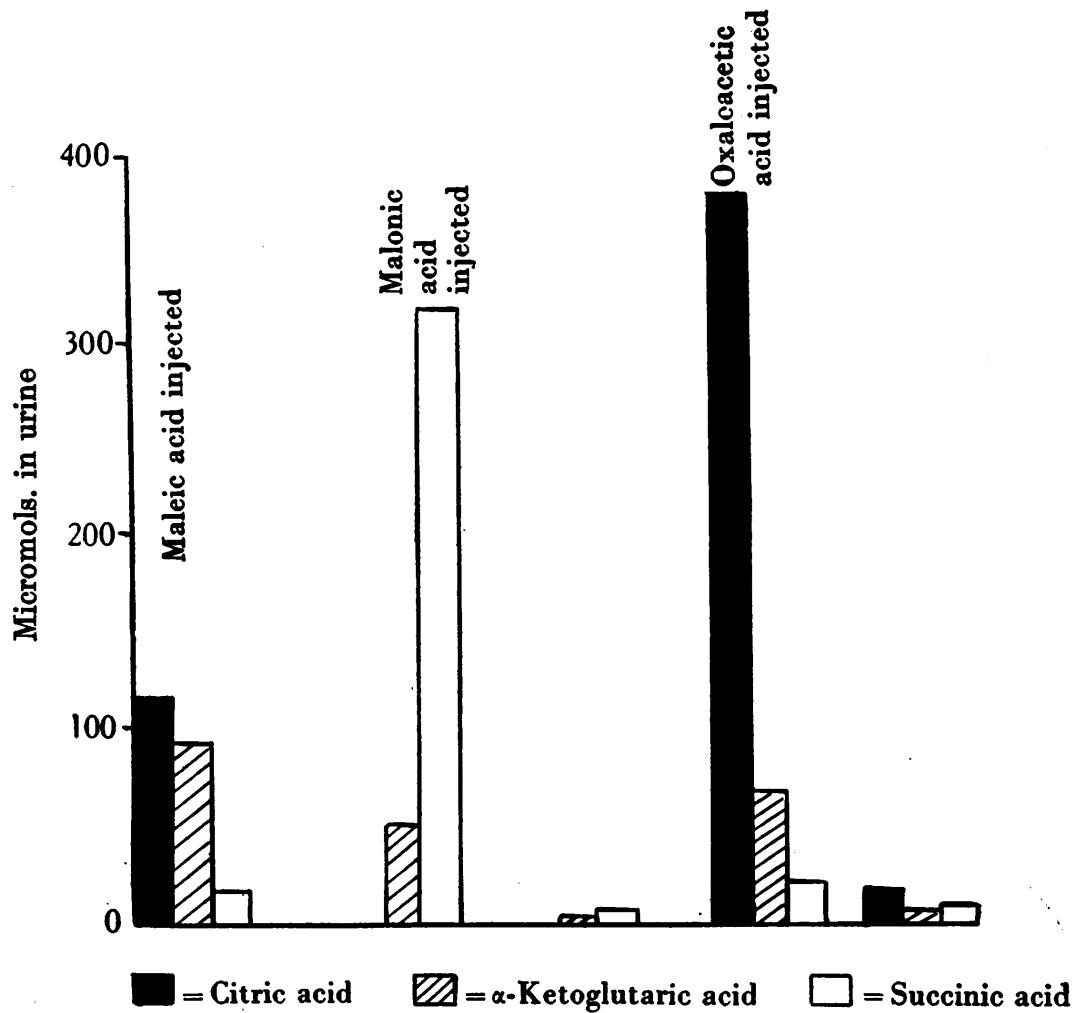
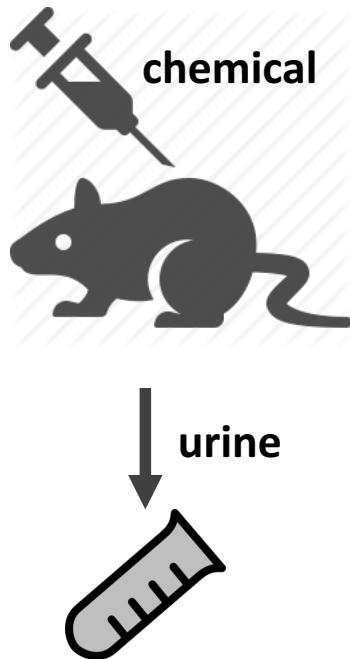
- 00230 Purine metabolism  
00240 Pyrimidine metabolism

## 1.5 Amino acid metabolism

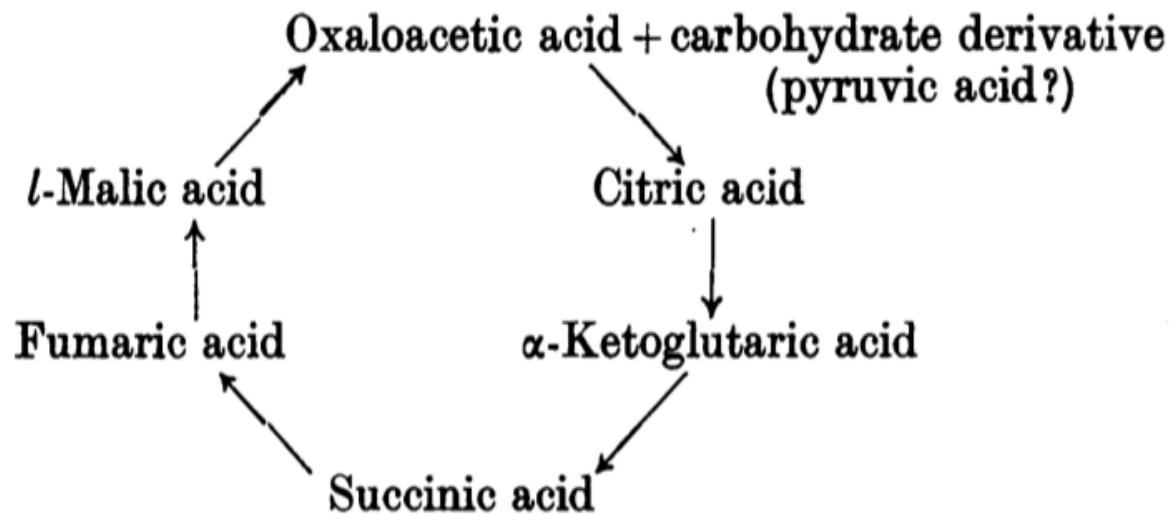
- 00250 Alanine, aspartate and glutamate metabolism
  - 00260 Glycine, serine and threonine metabolism
  - 00270 Cysteine and methionine metabolism
  - 00280 Valine, leucine and isoleucine degradation



# Krebs et al, 1937



# Krebs Cycle, 1938



Krebs et al. 1938. Biochem Journal. 32:113

# Beadle 1941

## Genetic Control of Biochemical Reactions in *Neurospora*

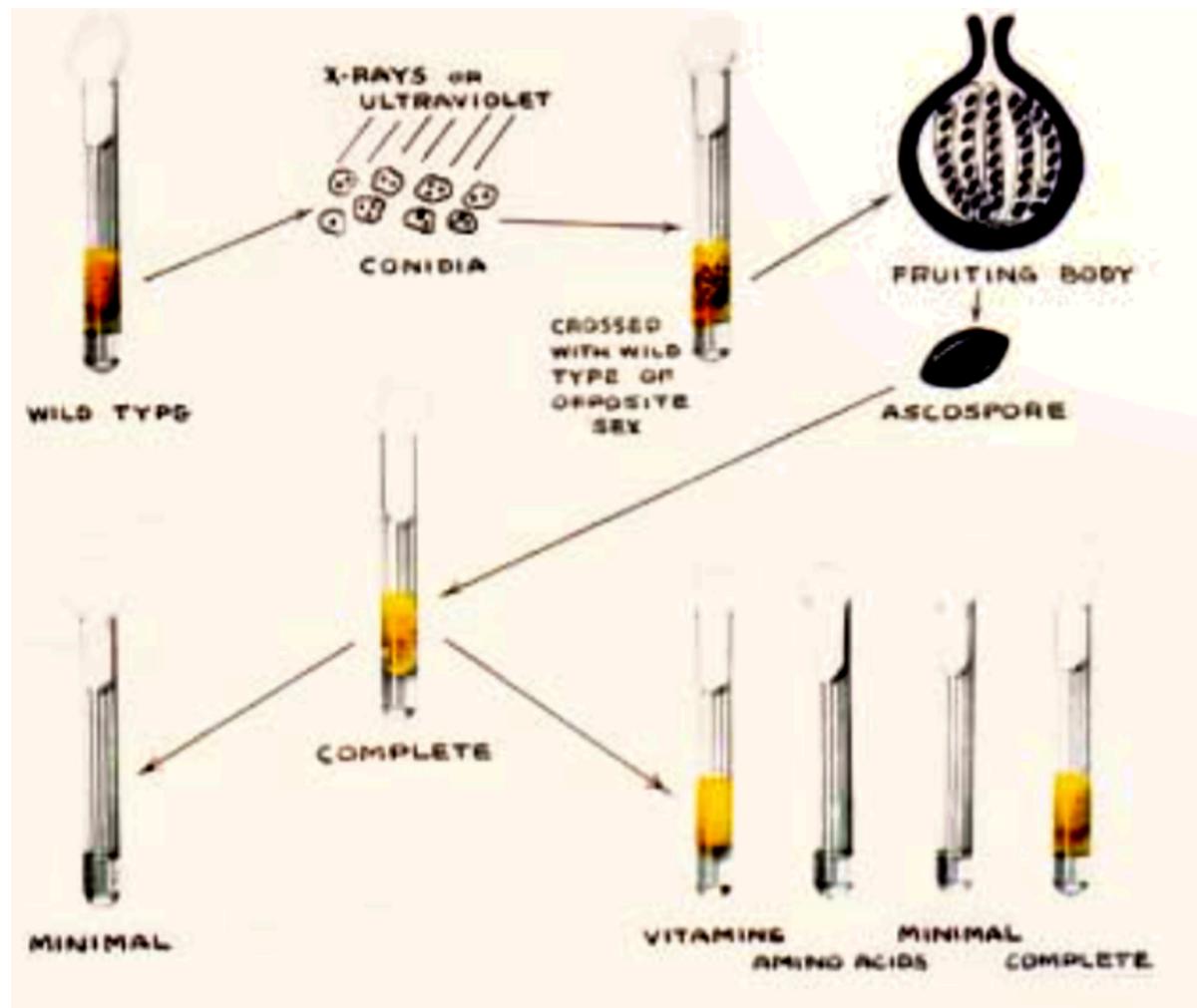
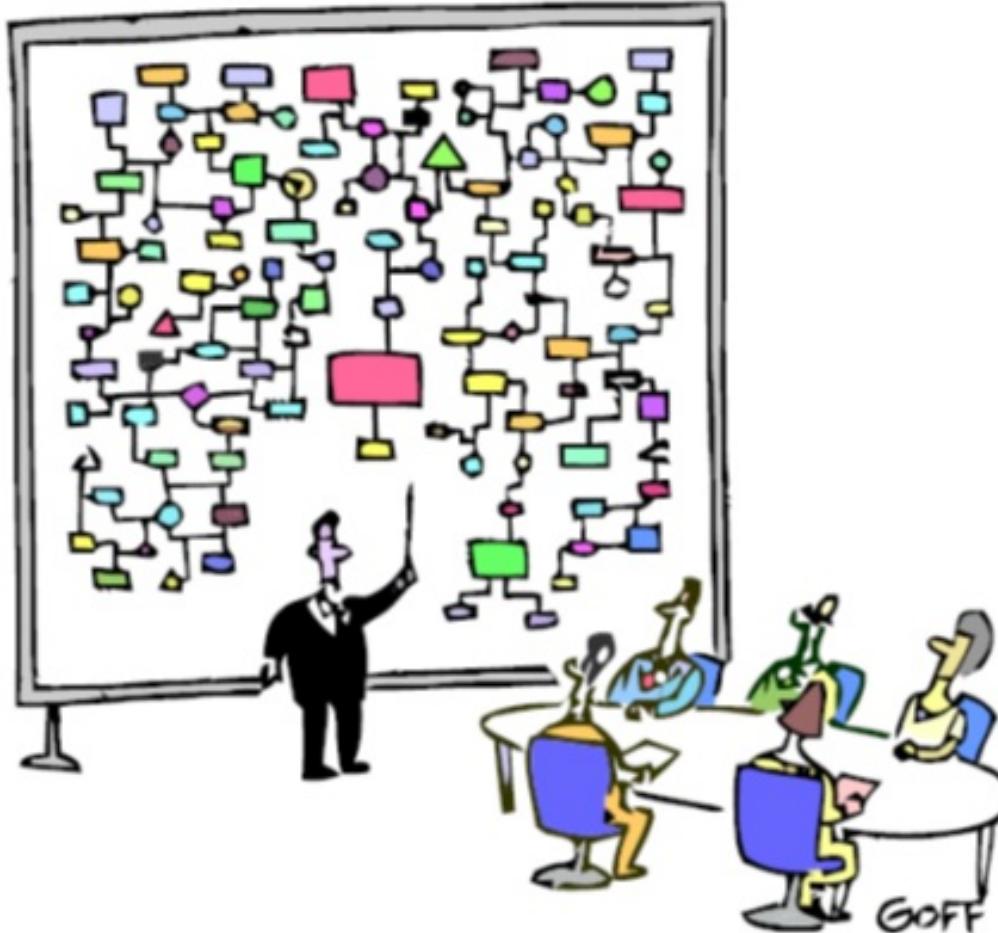


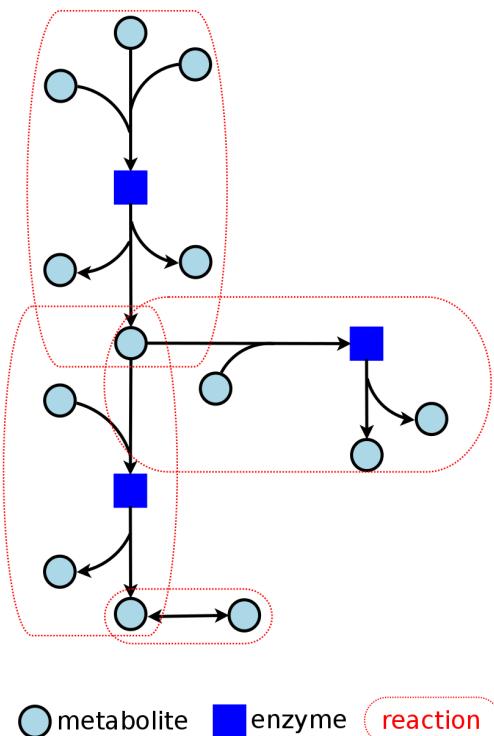
Figure from *Nature Reviews Genetics* 5, 949-954, 2004

# Metabolic pathways are curated human knowledge



"And that's why we need a computer."

# Metabolic models



- A metabolic model consists of metabolites, enzymes, reactions, pathways
- Reactions can be described by differential equations (mathematical models)

$$\begin{array}{c} \begin{array}{|c|} \hline v_1 & v_2 & b_1 & b_2 & b_3 & b_4 & b_5 \\ \hline A & -1 & 0 & 1 & 0 & 0 & 0 & 0 \\ B & 1 & -1 & 0 & -1 & 0 & 0 & 0 \\ C & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ D & -1 & 0 & 0 & 0 & 0 & 1 & 0 \\ E & 1 & 0 & 0 & 0 & 0 & 0 & 1 \\ \hline \end{array} & \xrightarrow{\hspace{1cm}} & \begin{array}{l} \text{Stoichiometric matrix} \\ \\ \frac{d}{dt} \begin{pmatrix} A \\ B \\ C \\ D \\ E \end{pmatrix} = S v = S \begin{pmatrix} v_1 \\ v_2 \\ b_1 \\ b_2 \\ b_3 \\ b_4 \\ b_5 \end{pmatrix} = 0 \\ \\ \text{Steady State Mass balance} \end{array} \end{array}$$

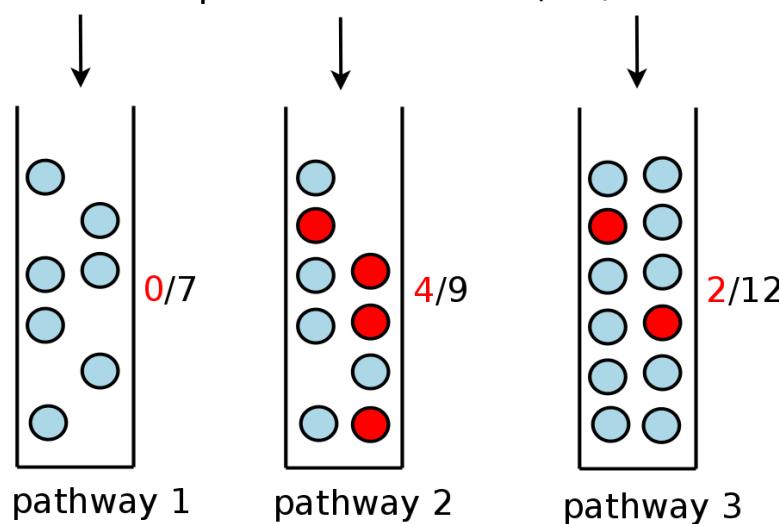
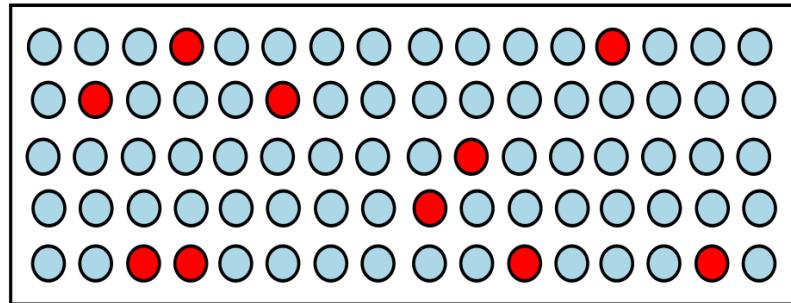
- We focus on statistical models; flux models are not covered here
- Pathways and networks are mathematically indistinguishable

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# Pathway enrichment test

If metabolites are known; red are significant metabolites



$$P = \frac{\binom{9}{4} \binom{71}{6}}{\binom{80}{10}} = 0.01$$

# MetaboAnalyst



## MetaboAnalyst - statistical, functional and integrative analysis of metabolomics data

[Home](#)

[Overview](#)

[Data Formats](#)

[FAQs](#)

[Tutorials](#)

[Troubleshooting](#)

[Resources](#)

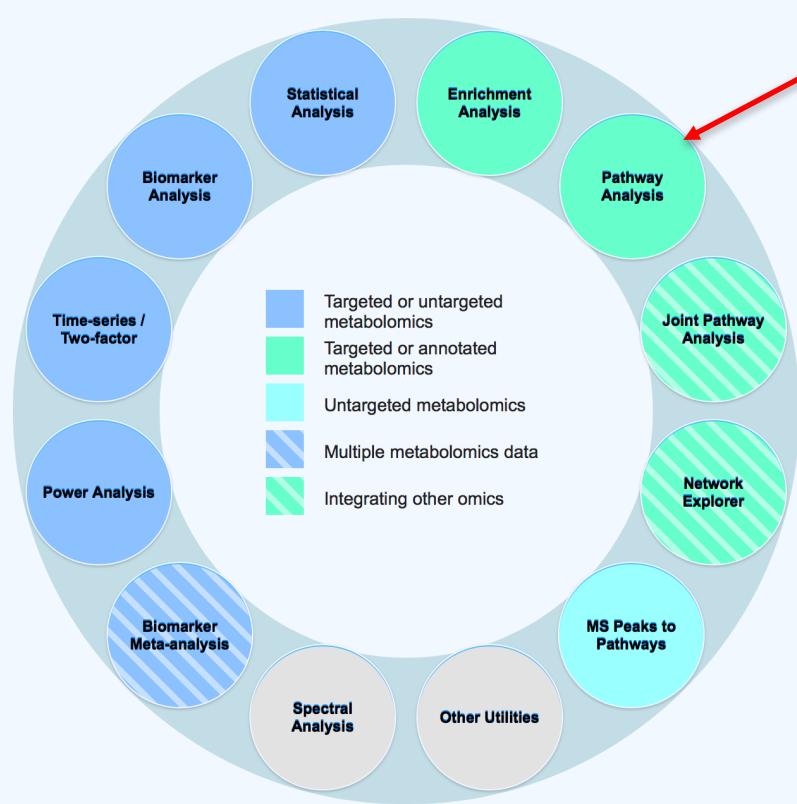
[Update History](#)

[User Stats](#)

[About](#)



Click a module to proceed, or [scroll down](#) for more details:



**Targeted**

# Exercise 1-3

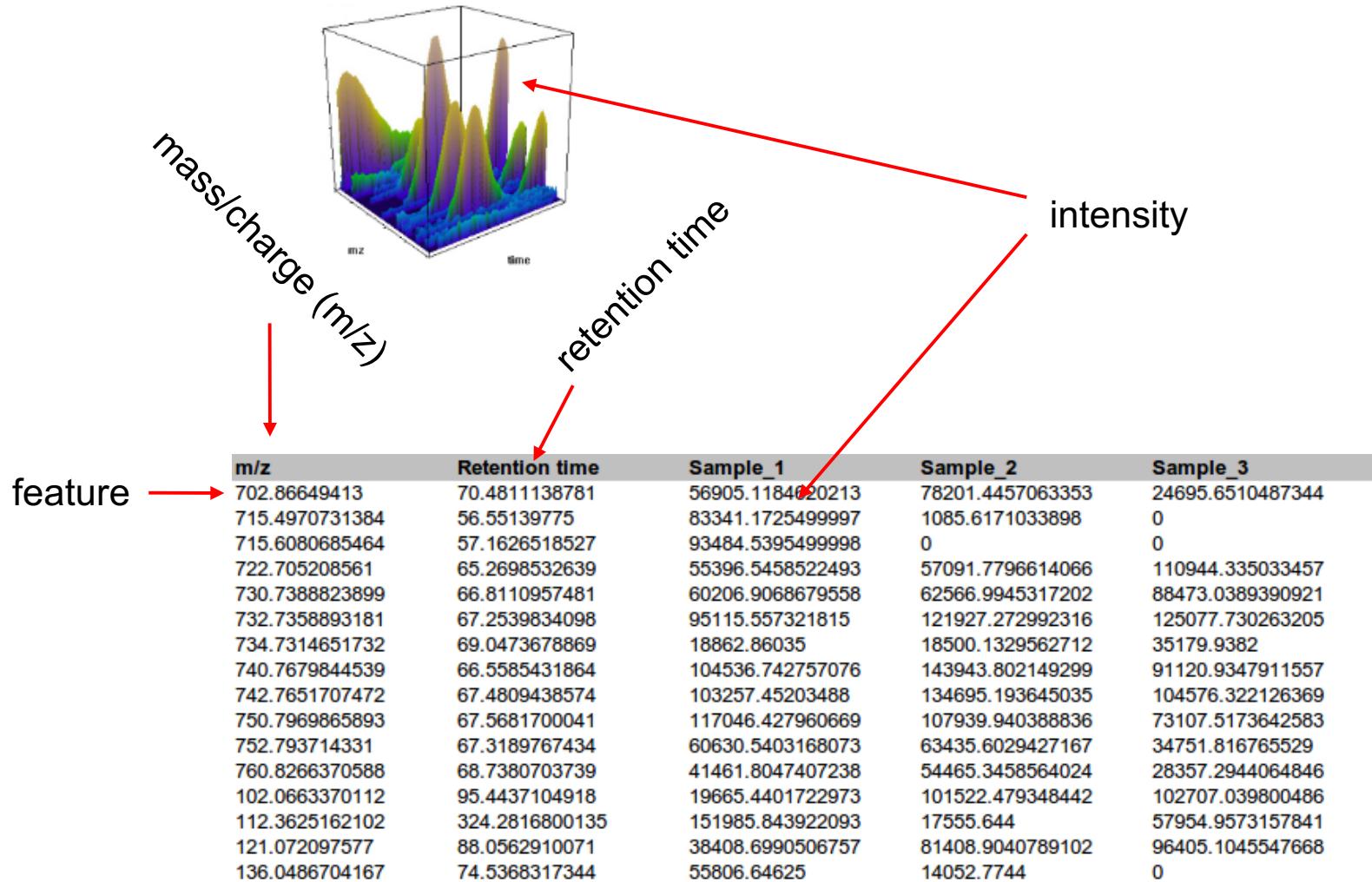
[Download link will be given at workshop]

1. Targeted pathway analysis using MetaboAnalyst  
<https://www.metaboanalyst.ca>
2. Enrichment analysis in command line  
[optional]
3. Pathway visualization – mapping, KEGG painting  
[https://www.genome.jp/kegg/tool/map\\_pathway2.html](https://www.genome.jp/kegg/tool/map_pathway2.html)

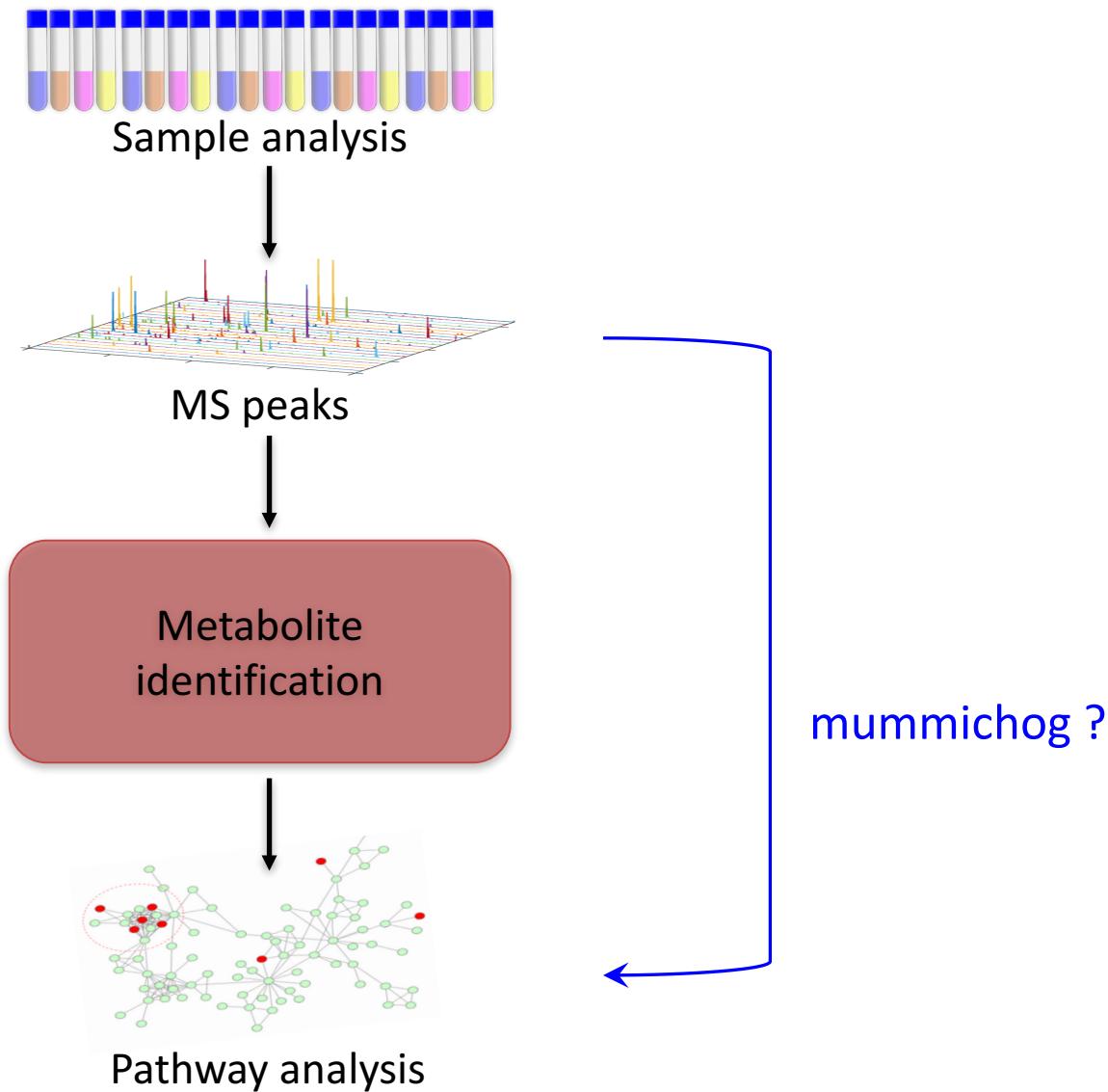
# Outline

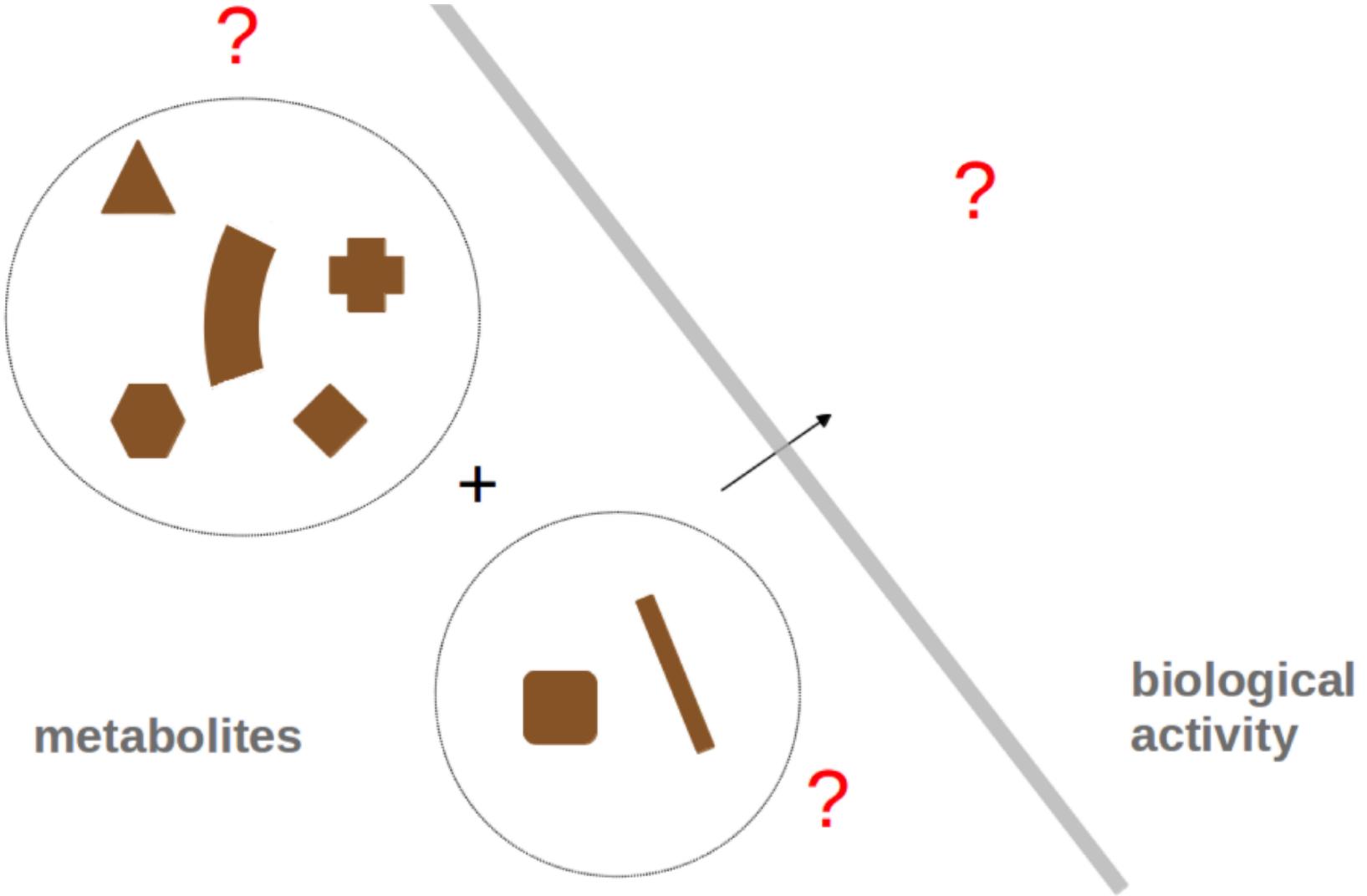
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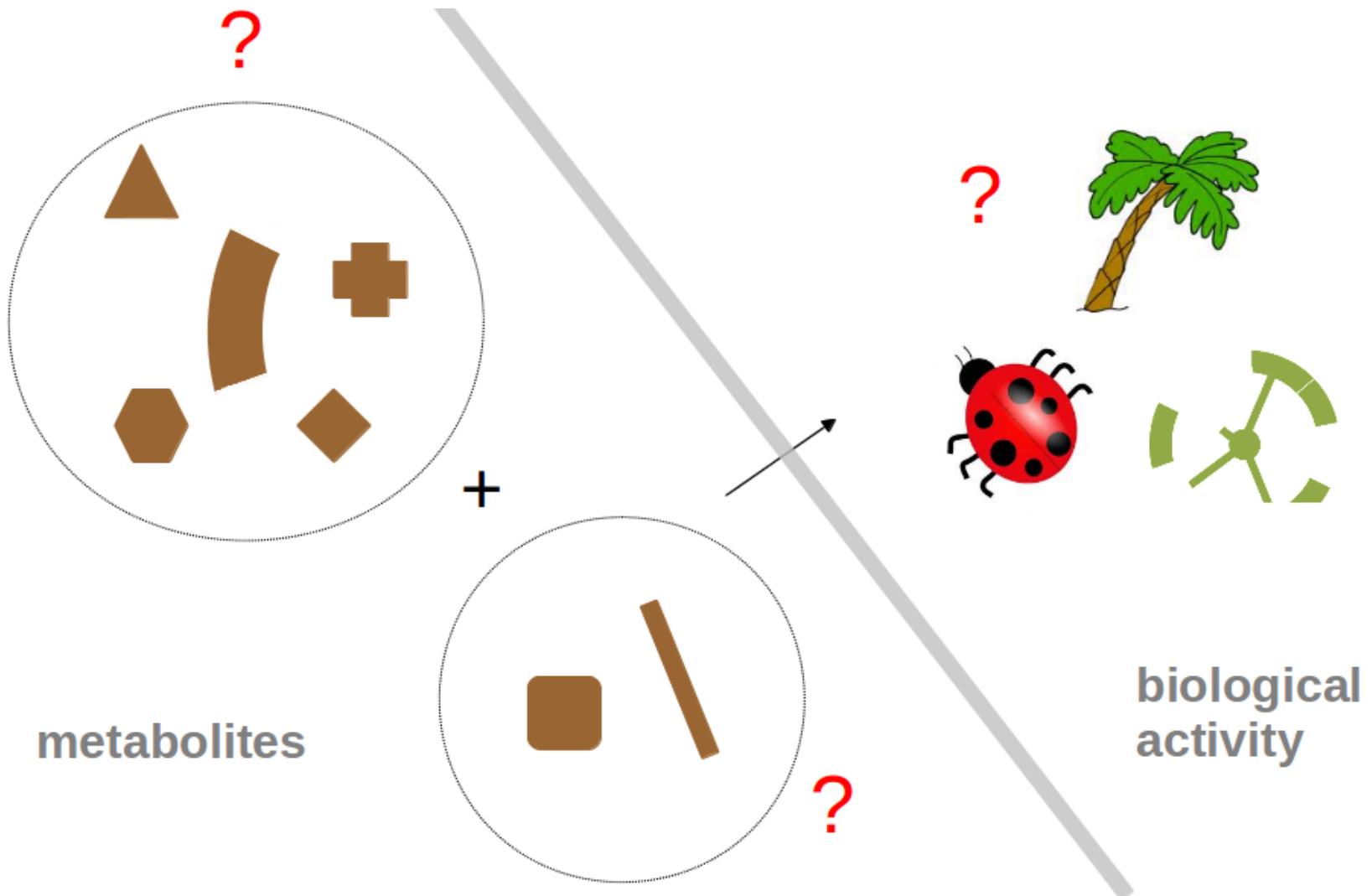
# Untargeted metabolomics data

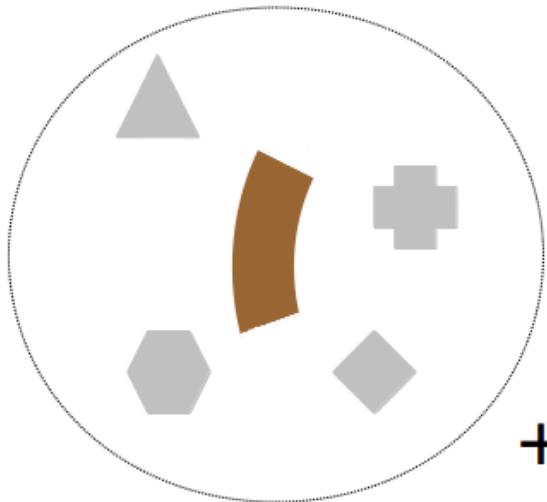


# Fastlane from chemistry to biology?





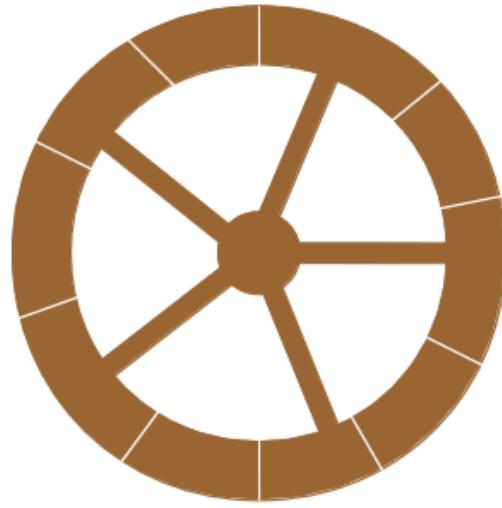




metabolites

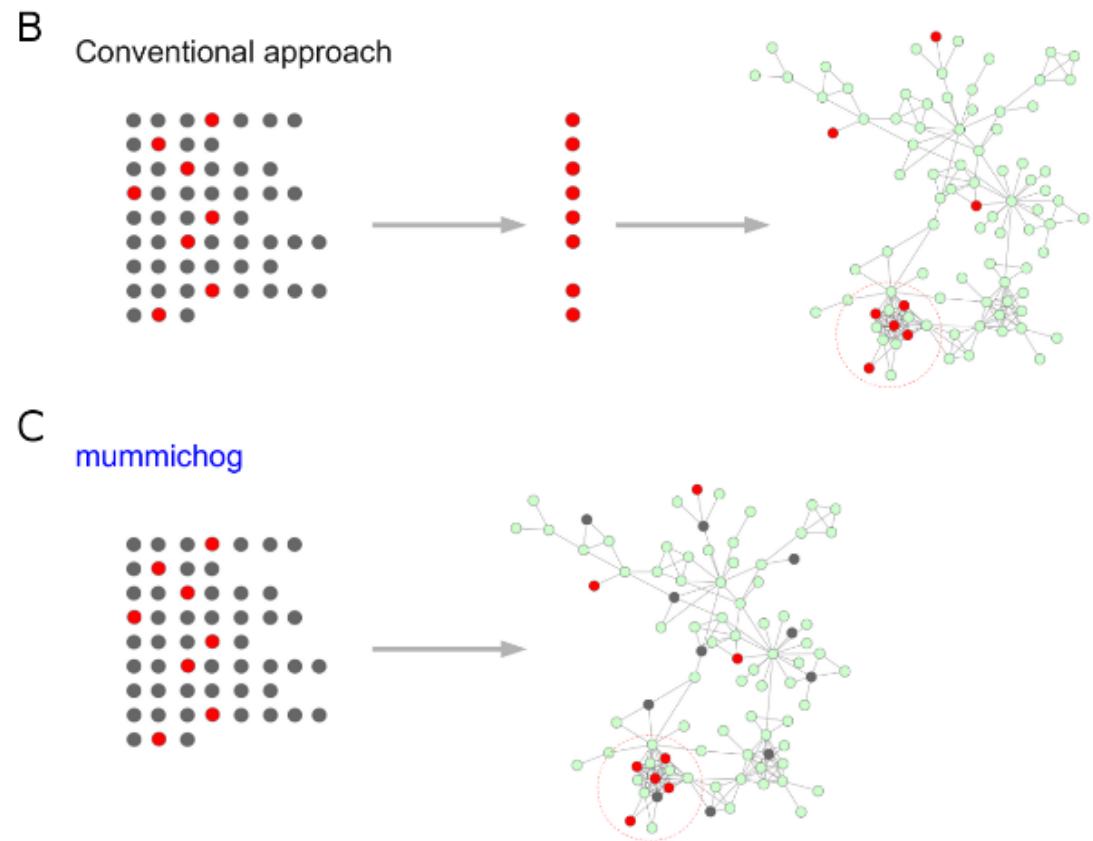
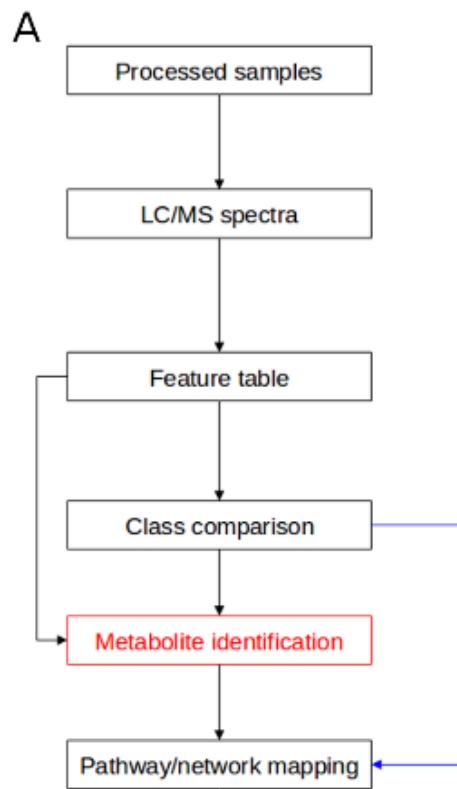


+

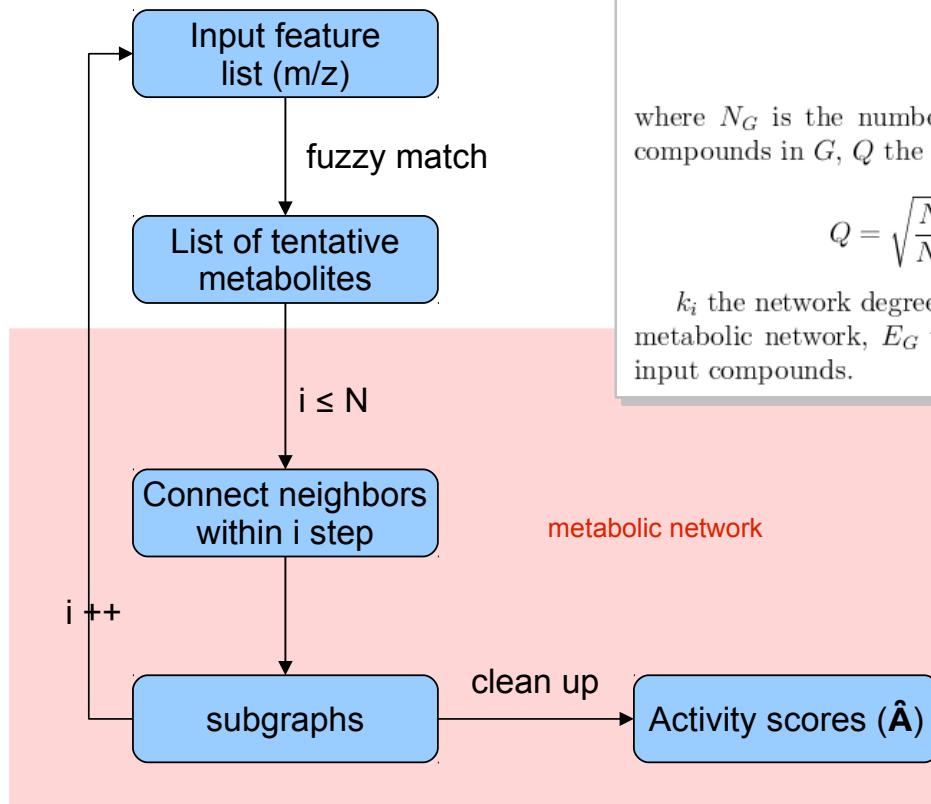


biological  
activity

# *Mummichog* tests metabolite grouping patterns



# Module analysis in *mummichog*



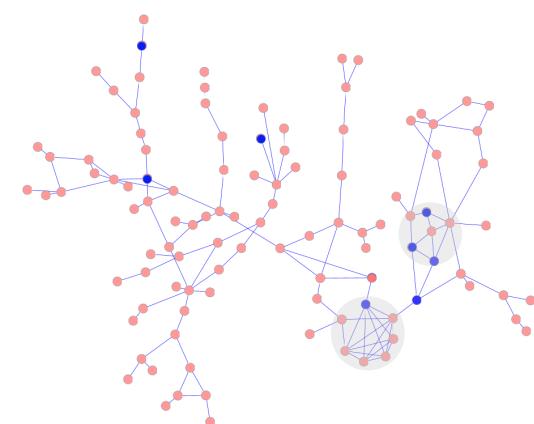
For a subgraph  $G$ , activity score

$$\hat{A} = Q \cdot \frac{N_{I,G}}{N_G}, \quad (1)$$

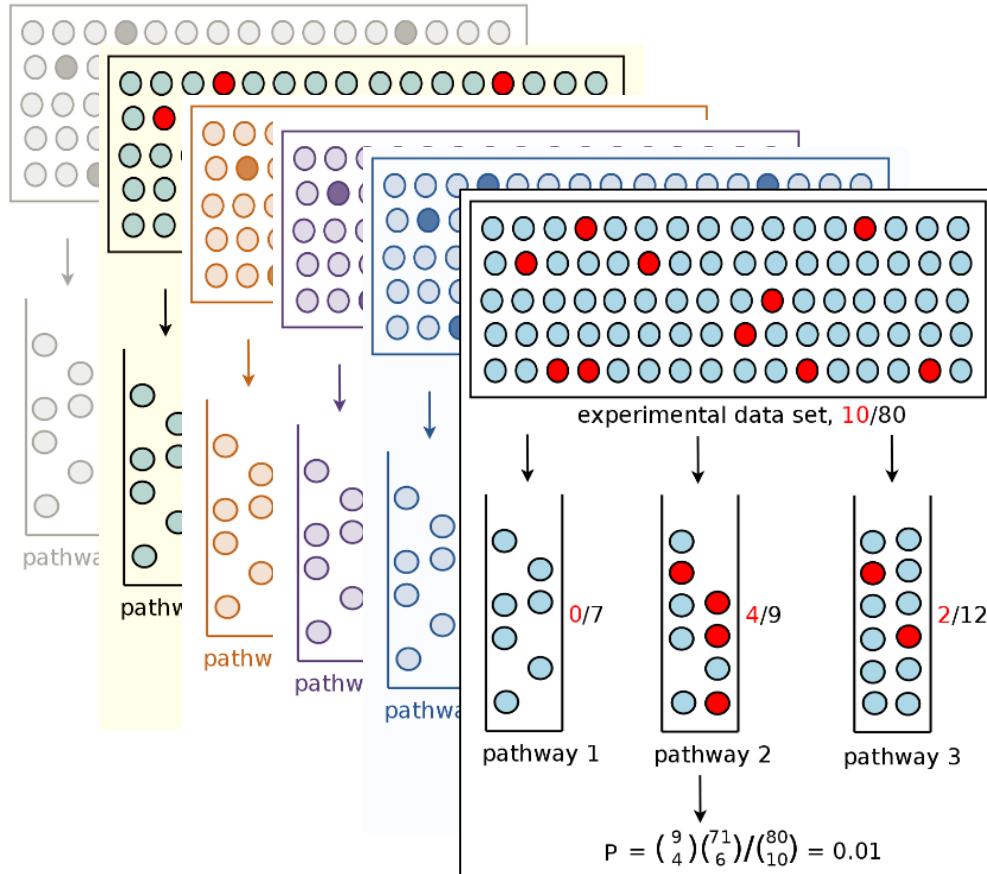
where  $N_G$  is the number of compounds in  $G$ ,  $N_{I,G}$  the number of input compounds in  $G$ ,  $Q$  the adjusted Newman-Girvan modularity:

$$Q = \sqrt{\frac{N_I}{N_G}} \cdot \left( \frac{E_G}{m} - \sum_{i,j} \frac{k_i k_j}{2m} \right), \quad i, j \in G \quad (2)$$

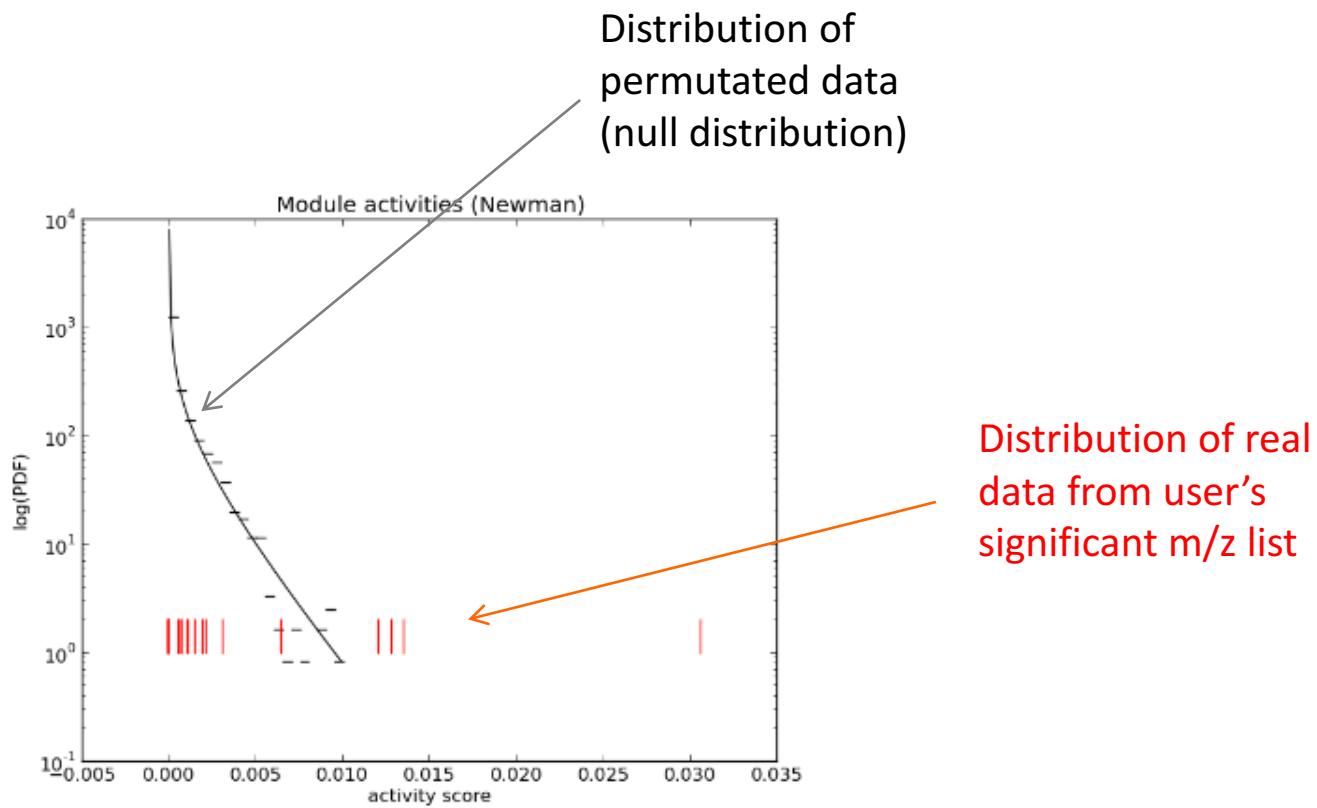
$k_i$  the network degree of compound  $i$ ,  $m$  the total number of edges in the metabolic network,  $E_G$  the total number of edges in  $G$ ,  $N_I$  the number of input compounds.



# Pathway analysis in *mummichog*



# Testing module/pathway significance in *mummichog*



# On the organization of pathways

- Pathway definition may be incomplete, different between databases and species
- Pathways may not be what you measured - compartmentalization, sampling, metabolite stability.
- Metabolic reactions connect to networks
- Modules are based on network topology, less biased by human definition but data dependent. A module can be within a pathway or in between several pathways.
- Other thoughts: metabolite sets, chemical similarity
- A major mission of metabolomics is to update biochemical pathways

## Mummichog pathway and network analysis for metabolomics

Home    Software    Publications    Notebooks    Version 1.0.9    mummichog-server

Mummichog is a free Python program for analyzing data from high throughput, untargeted metabolomics. It leverages the organization of metabolic networks to predict functional activity directly from feature tables, bypassing metabolite identification. Thus, high-quality hypotheses can be quickly generated from a LC-MS data table.

Download and use mummichog. Now via standard Python package index.

OS independent install (version 1):

```
pip install mummichog1
```

Or mummichog 2 (beta test):

```
pip install mummichog
```

Run it in command line:

```
mummichog -f myData --o myResult
```

More to [click "Software"](#).

Example publications that were supported by using mummichog

Li et al. (2017) Metabolic Phenotypes of Response to Vaccination in Humans. *Cell* 169(5): p862-877.

Huan, Tao, et al. "Systems biology guided by XCMS Online metabolomics." *Nature methods* 14.5 (2017): 461.

Xu et al. "Autophagy is essential for effector CD8+ T cell survival and memory formation." *Nature immunology* 15.12 (2014): 1152-1161.

More to [click "Publications"](#).

Jupyter notebooks for data analysis in metabolomics and systems biology

Jupyter notebooks are to data people like lab notebooks to bench scientists. They keep code and result in the same web browser. It's increasingly popular for data analysis and collaboration. We continue posting notebooks as tutorials and for record keeping.

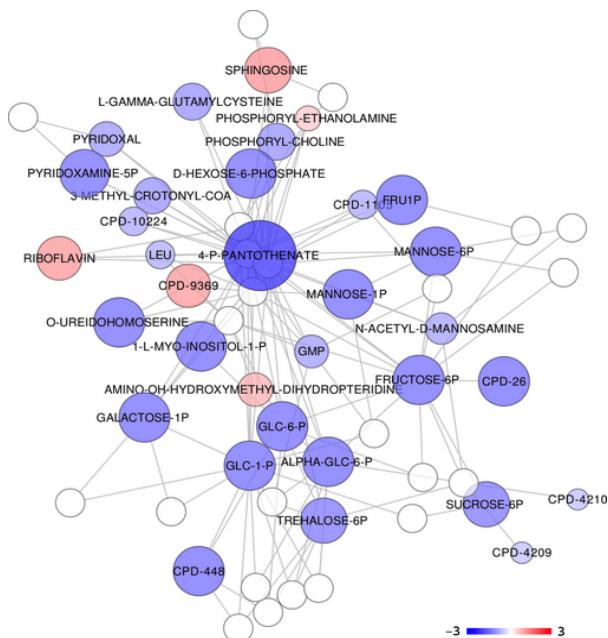
More to [click "Notebooks"](#).

## News

- Web version is now testing at [here](#).
- Mummichog 2 is now on [GitHub](#).
- Mummichog 1 with a web interface is now available on [MetaboAnalyst 4](#).
- A common error, "AttributeError: 'NodeView' object has no attribute 'sort'", is caused by Networkx 2.x, which is not backward compatible. This can be fixed by installing Networkx 1.x via pip in your terminal, "sudo pip install networkx==1.10".
- Mummichog 2 test version is available at [PyPi](#).
- Mummichog version 1.0.10 is available at [PyPi](#).
- Mummichog helped decipher metabolic phenotypes in human vaccination - Li et al. (2017) at [Cell 169\(5\): p862-877](#). Also see commentary "[Orthogonal Data Integration to Define Immunometabolic-Phenotypes](#)" at [Cell Systems](#).
- Emory University Sys/Bio/Info Group runs a monthly meeting on Systems Biology and Bioinformatics. These are casual seminars and discussions, on every first Wednesday of the month, 12 pm, Whitehead Biomedical Research Building, Rm 200. Email Dr. Shuzhao Li to join the mailing list.

# Exercise 4-6

4. Explore mummichog output
5. Using barplot to visualize pathway significance
6. Network visualization using Cytoscape:  
<http://cytoscape.org> [On your own]



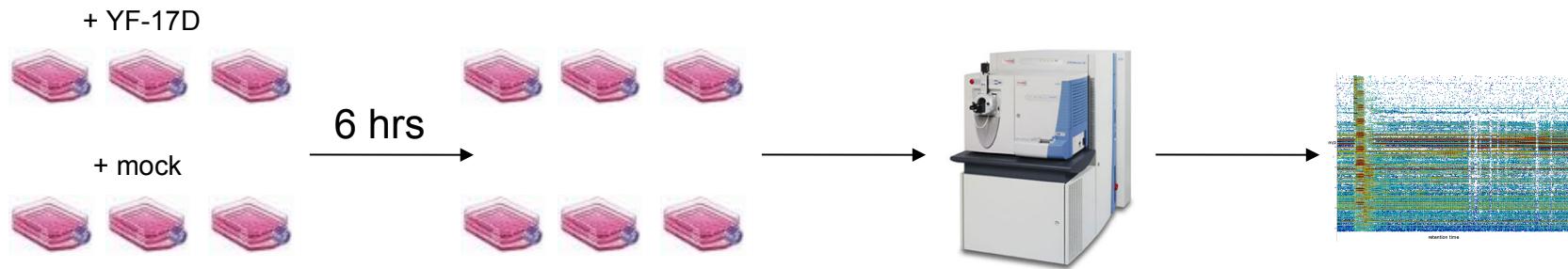
Example using Cytoscape to  
visualize mummichog  
metabolite module (Hoffman et  
al, 2014. Aging Cell, 13: 596 )

# Outline

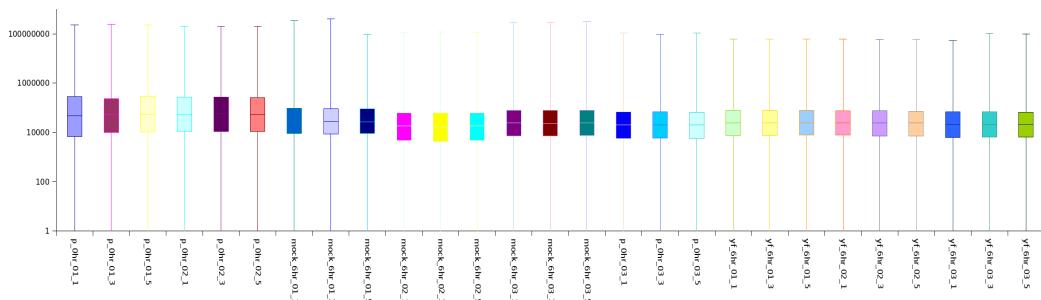
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# Case study: viral activation of immune cells

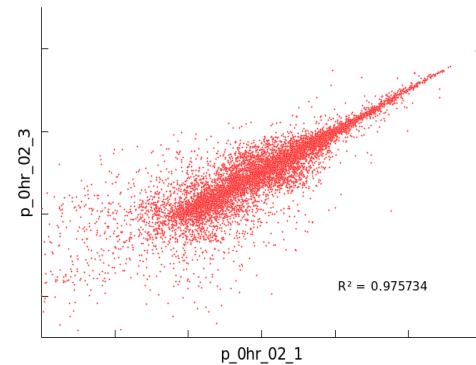
Monocyte derived dendritic cells (moDC)



QA: total ion counts are similar among samples

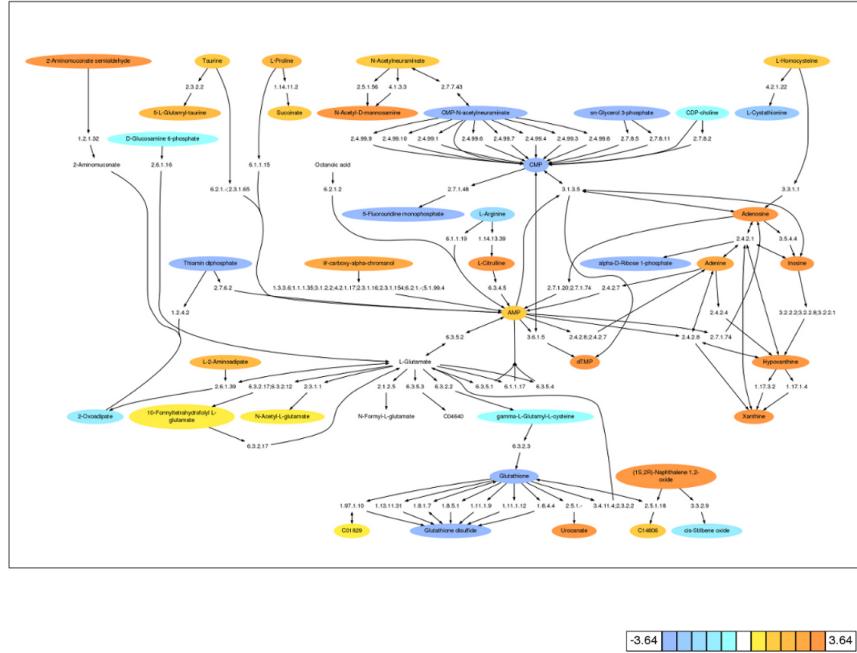


technical replicates, 10,000 features

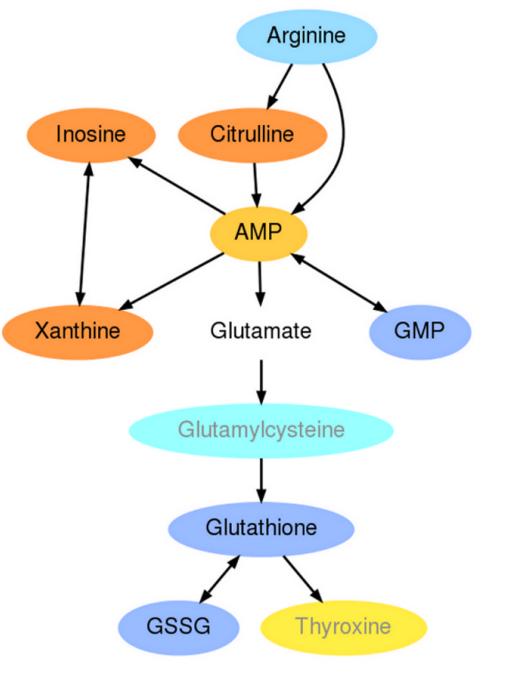


# Mummichog: viral activation of immune cells

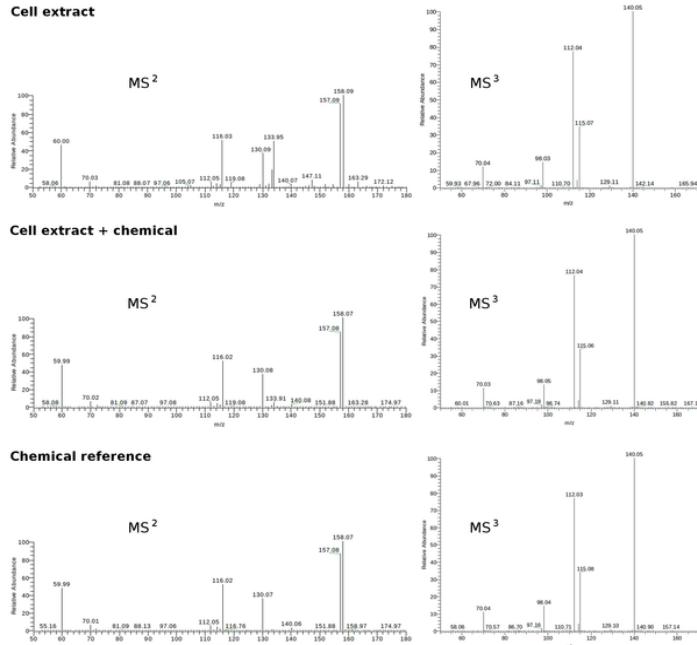
A



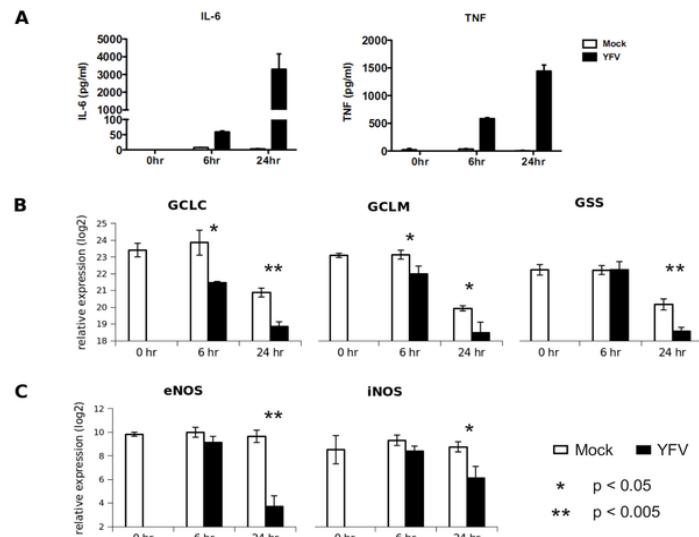
B



# Experimental validation of *mummichog* prediction

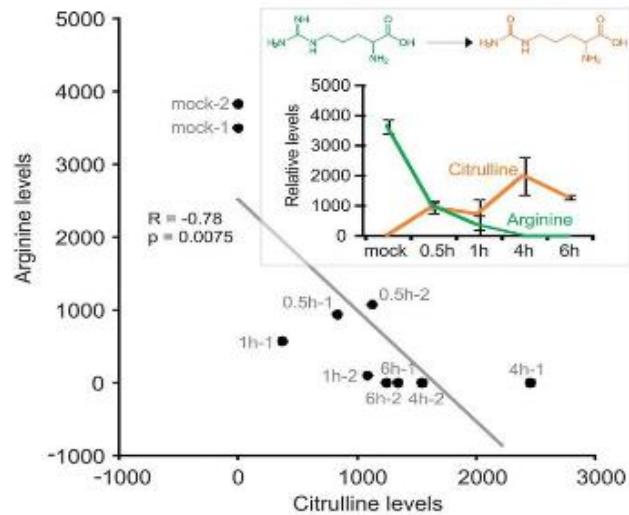


Tandem mass spectrometry confirmed 9/11 metabolites

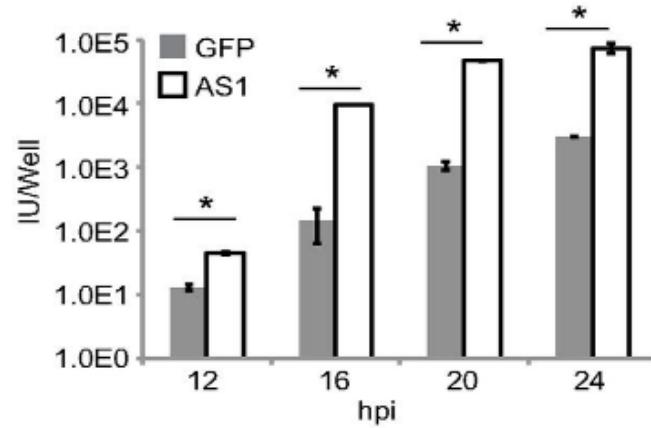


Gene expression supported GSH/GSSG depletion and Arg/Cit conversion

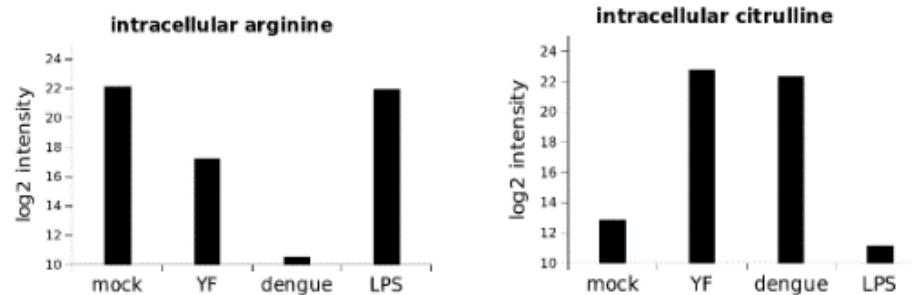
# Arginine as master regulator of viral response



Ravindran et al. 2014. Science 343:313

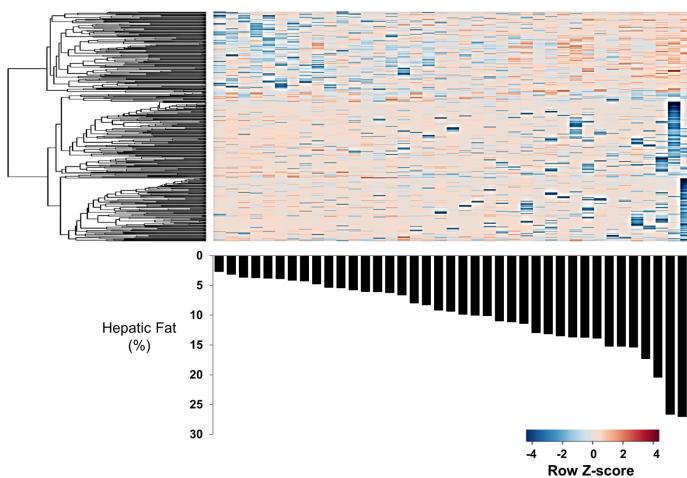


Argininosuccinate synthetase 1 knockdown led to increased replication of HSV-1.  
Grady, Purdy, Rabinowitz & Shenk. 2013. PNAS 110:E5006.



Li et al. 2013. PLoS Computational Biology. 9:e10031323

# MWAS + *mummichog* (NAFLD)



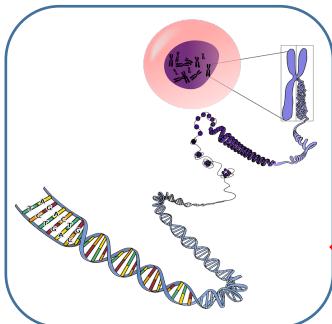
Jin, Banton, et al., 2016.

Amino Acid Metabolism is Altered in Adolescents  
with Nonalcoholic Fatty Liver Disease - An  
Untargeted, High Resolution Metabolomics Study.  
*The Journal of pediatrics* 172: 14

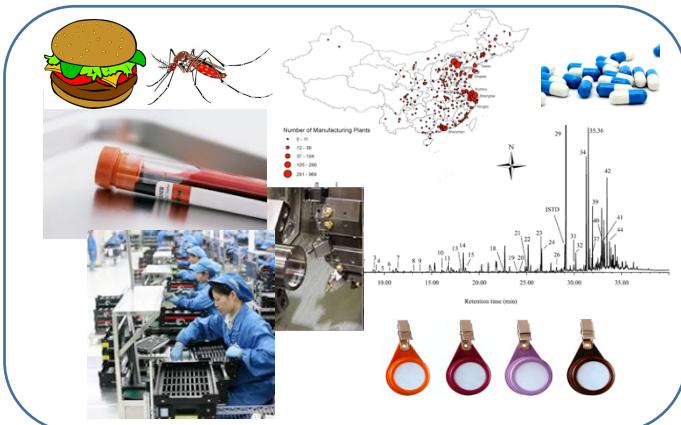
Pathway	Overlap_size	Pathway_size	Model p-value
Vitamin E metabolism	9	32	0.00095
Drug metabolism - cytochrome P450	8	34	0.00196
Tyrosine metabolism	15	79	0.00202
Vitamin B2 (riboflavin) metabolism	3	6	0.00229
Purine metabolism	10	51	0.00332
Ascorbate (Vitamin C) and Aldarate Metabolism	4	16	0.00773
Vitamin B9 (folate) metabolism	4	18	0.01307
Glutamate metabolism	3	12	0.01834
Methionine and cysteine metabolism	7	42	0.02026
Alanine and Aspartate Metabolism	4	20	0.02159
Biopterin metabolism	3	13	0.02493
Di-unsaturated fatty acid beta-oxidation	3	13	0.02493
Histidine metabolism	4	22	0.03449
Glycine, serine, alanine and threonine metabolism	8	53	0.03499
Valine, leucine and isoleucine degradation	7	46	0.03894

# G × M × E

## Genome



## Environment



## Gene function



### Endogenous metabolome

Core Biological Metabolome

Microbiome-related Chemicals

Non-nutritive Chemicals in Diet

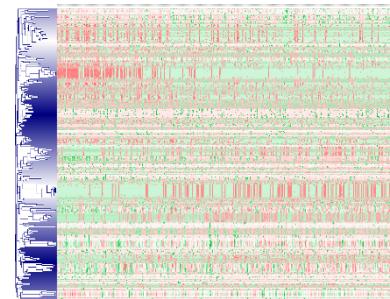
Supplements and Pharmaceuticals

### Environmental metabolome

Commercial Products

Environmental Chemicals

## Molecular response

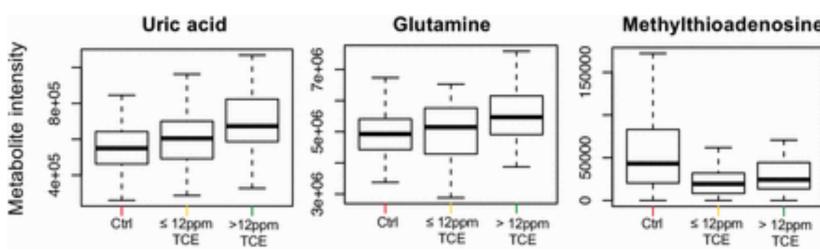
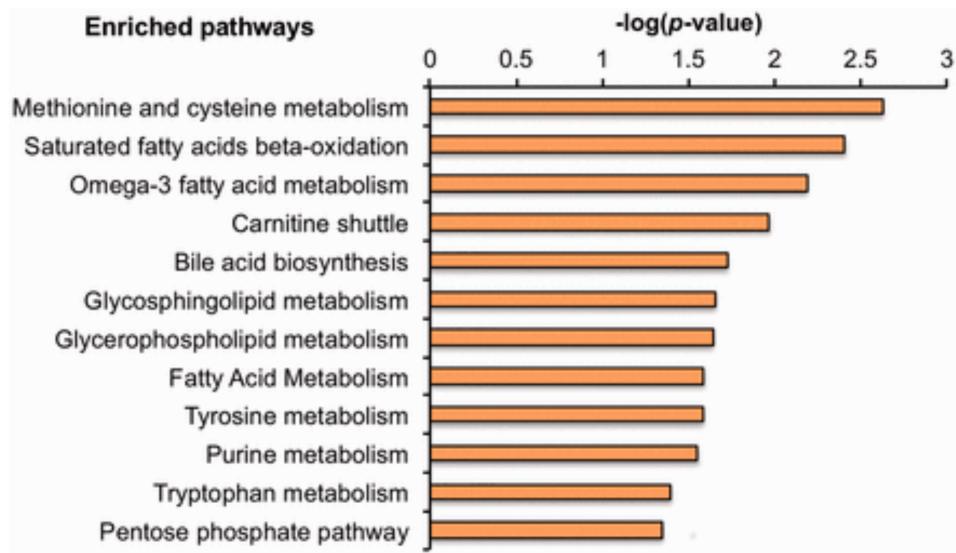
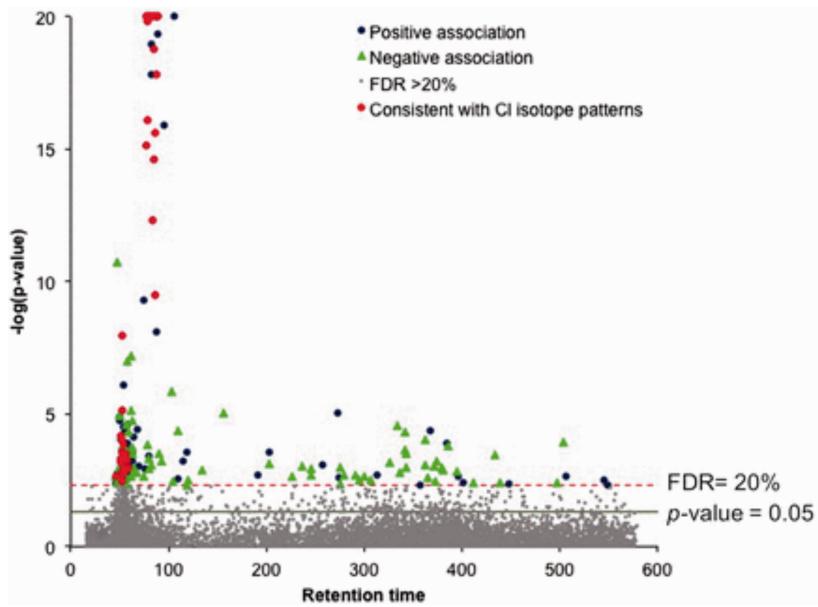


## Metabolome

## Body burden

Modified based on figures by Doug Walker & Dean Jones

# MWAS of occupational exposure to trichloroethylene



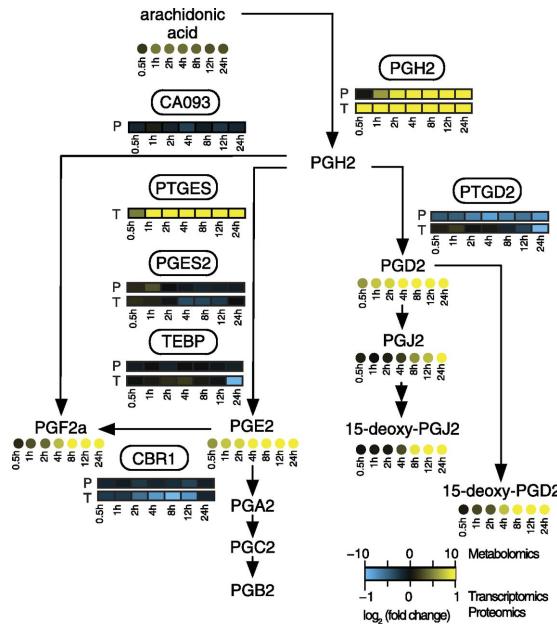
Walker DI. et al (2016), High-resolution metabolomics of occupational exposure to trichloroethylene.  
*Int J Epidemiol.* 45 (5): 1517-1527

# Exercise 7-9

7. Importance of experiment design and biological controls

8. MWAS + mummichog

9. More data visualization – heatmap, boxplot

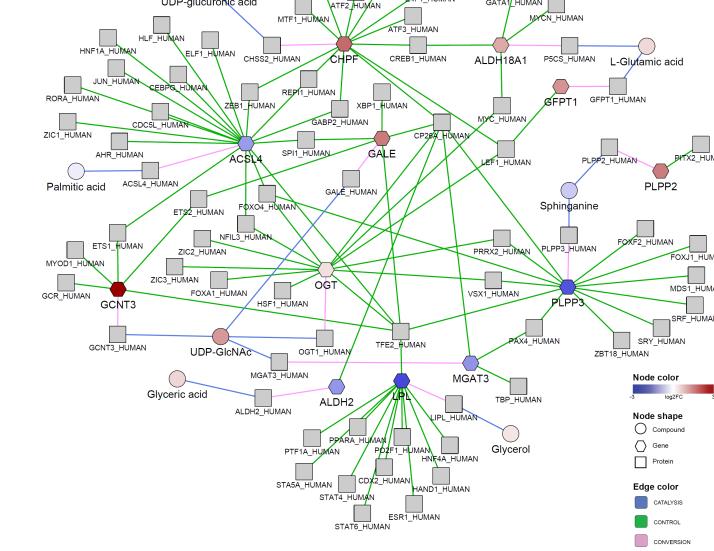
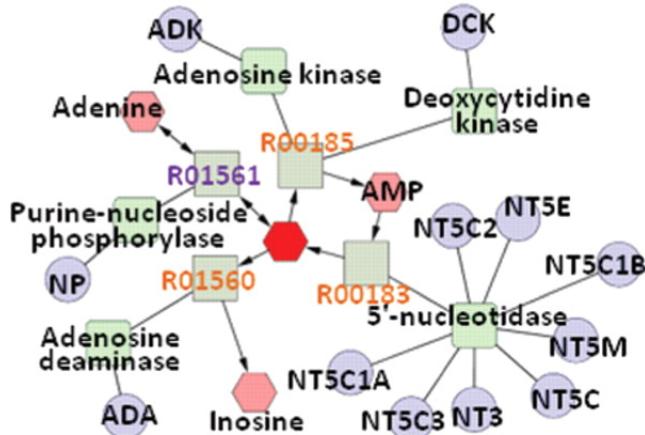


Example of manual layout  
of multiple data types  
Sabido et al, 2012. PMID:  
22361236

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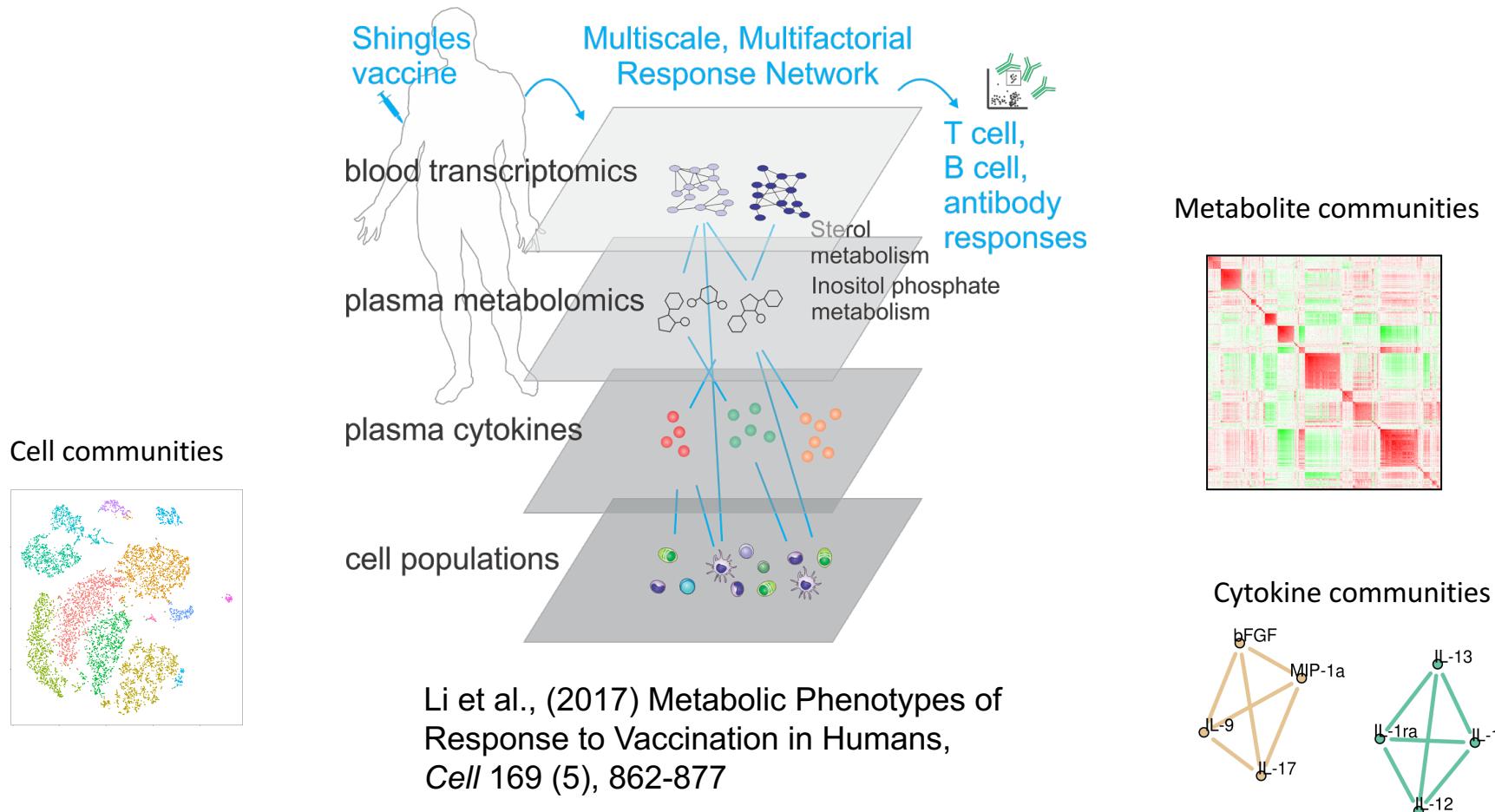
# Combining metabolite data with gene expression



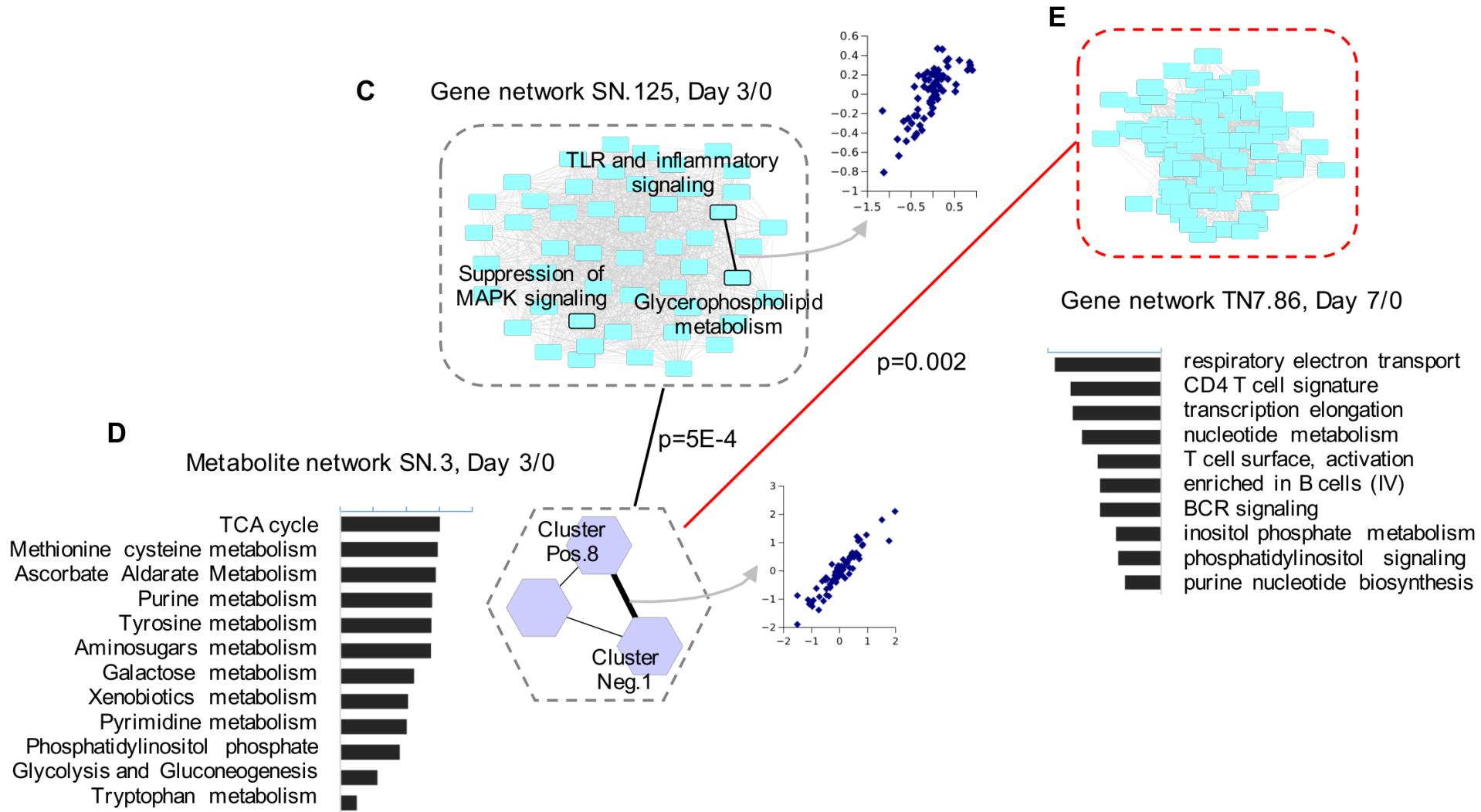
Metscape  
Karnovsky et al. (2012)  
*Bioinformatics*. 28:373

Metabox  
Wanichthanarak et al. (2017)  
*PloS ONE* 12(1):e0171046

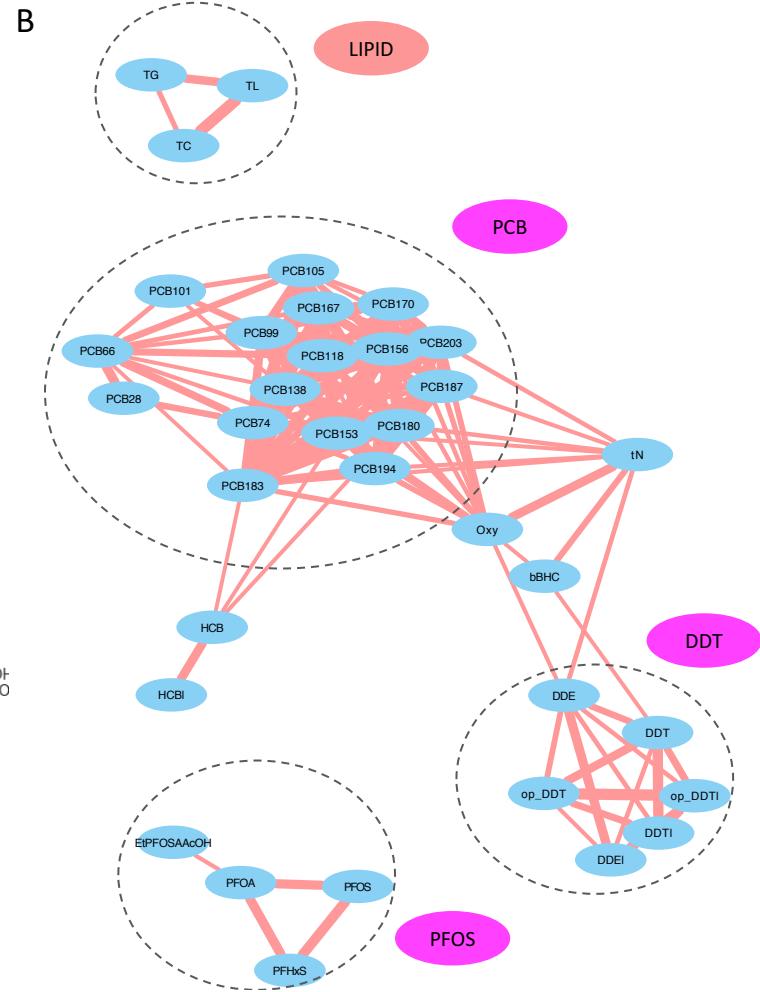
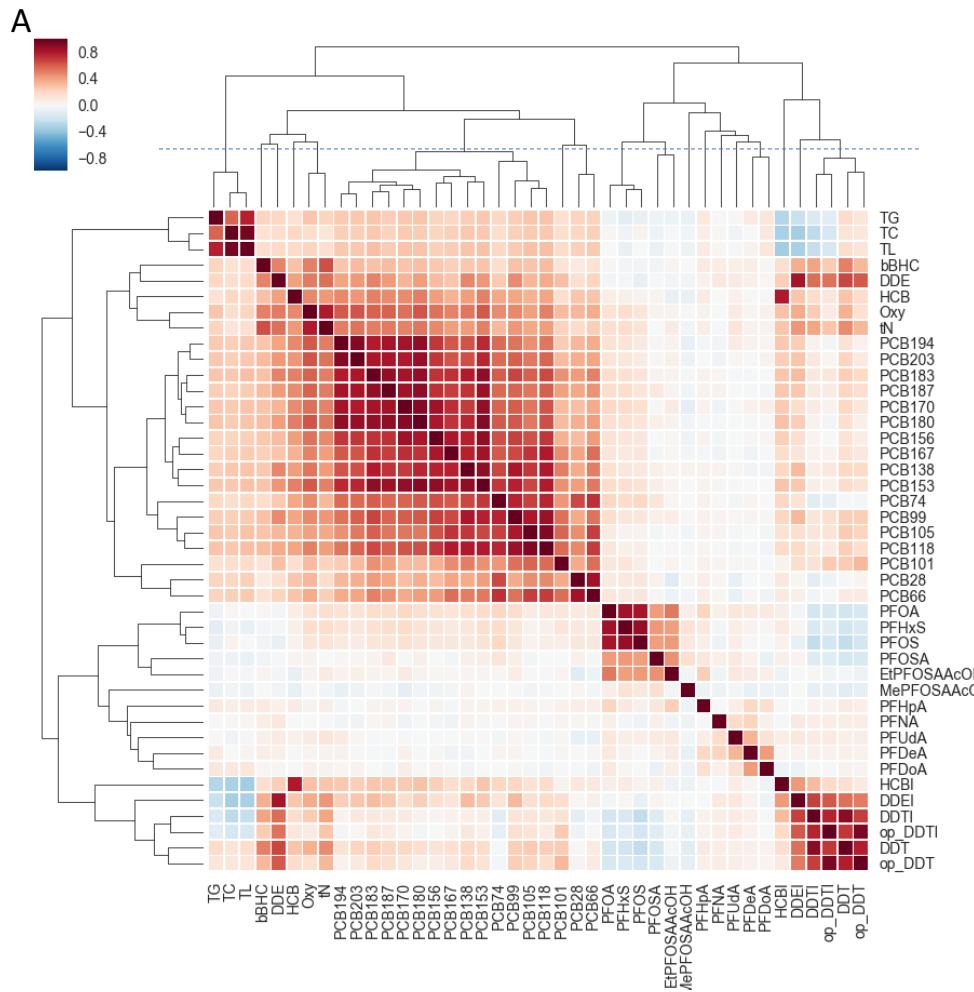
# Multi-omics in immune response



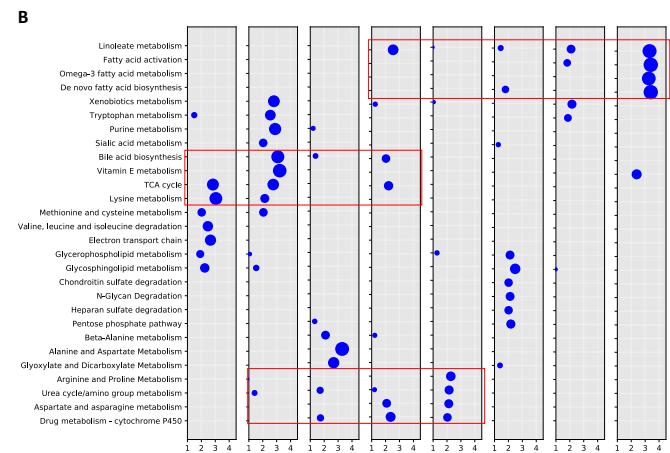
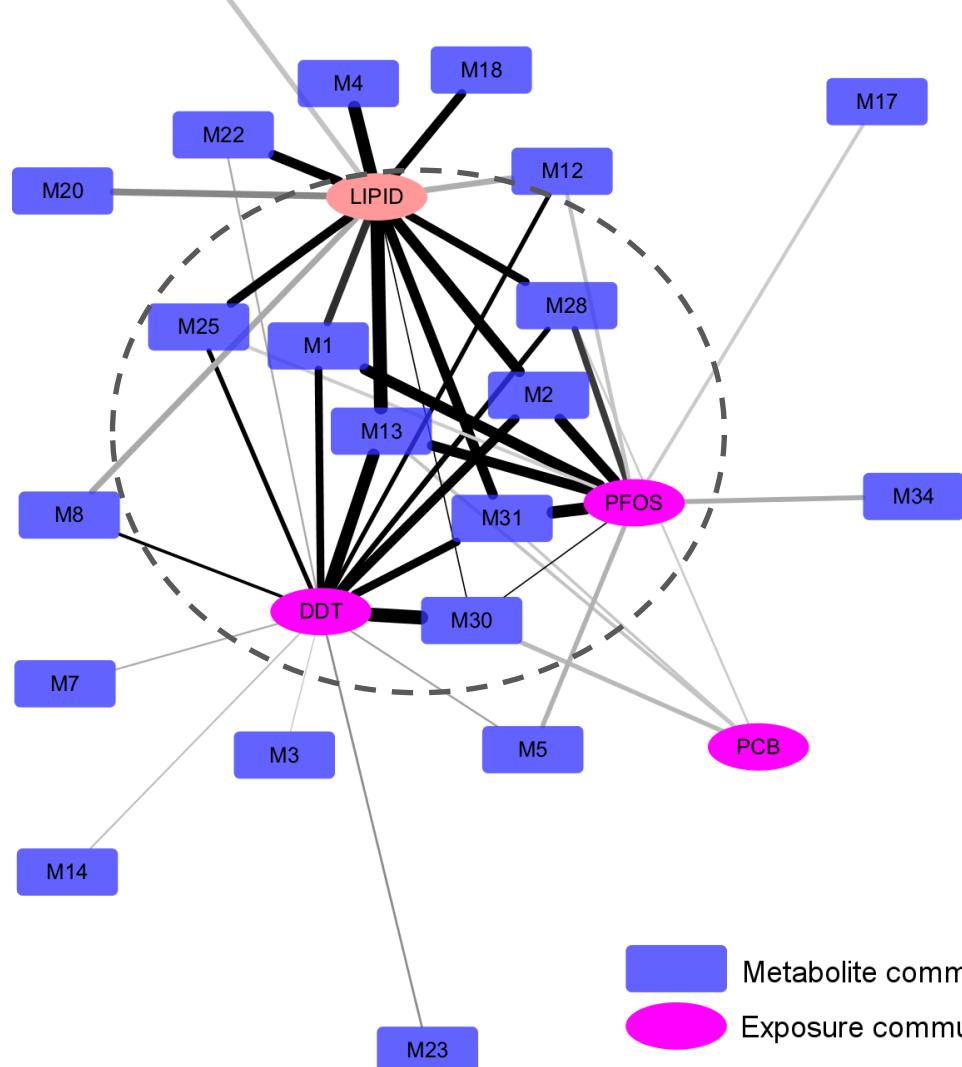
# Multi-omics in immune response



# Exposure communities in CHDS data



# Metabolomics × Exposures



# Outline

- What are pathways and metabolic models
- Pathway analysis for targeted data
- Untargeted data analysis and *mummichog*
- Application examples
  - Intracellular mechanisms
  - Population studies
  - Multi-omics integration
- Resources and issues

# Resources

## Metabolic pathways and models

KEGG: <http://www.genome.jp/kegg/kegg2.html>  
BioCyc: <https://biocyc.org>  
Recon: <https://www.vmh.life/#human/all>

## Metabolite ID conversion

<https://cts.fiehnlab.ucdavis.edu>  
<https://www.metaboanalyst.ca/faces/upload/ConvertView.xhtml>

## Metabolomics data analysis tools

MetaboAnalyst: <https://www.metaboanalyst.ca>  
MetScape: <http://metscape.ncbi.org/>  
MetExplore: <https://metexplore.toulouse.inra.fr/>  
Metabox/Met-DA: <http://metda.fiehnlab.ucdavis.edu>  
XCMS Online: <https://xcmsonline.scripps.edu>  
Mummichog: <http://mummichog.org>  
Commercial: MetaCore, Ingenuity Pathway Analysis

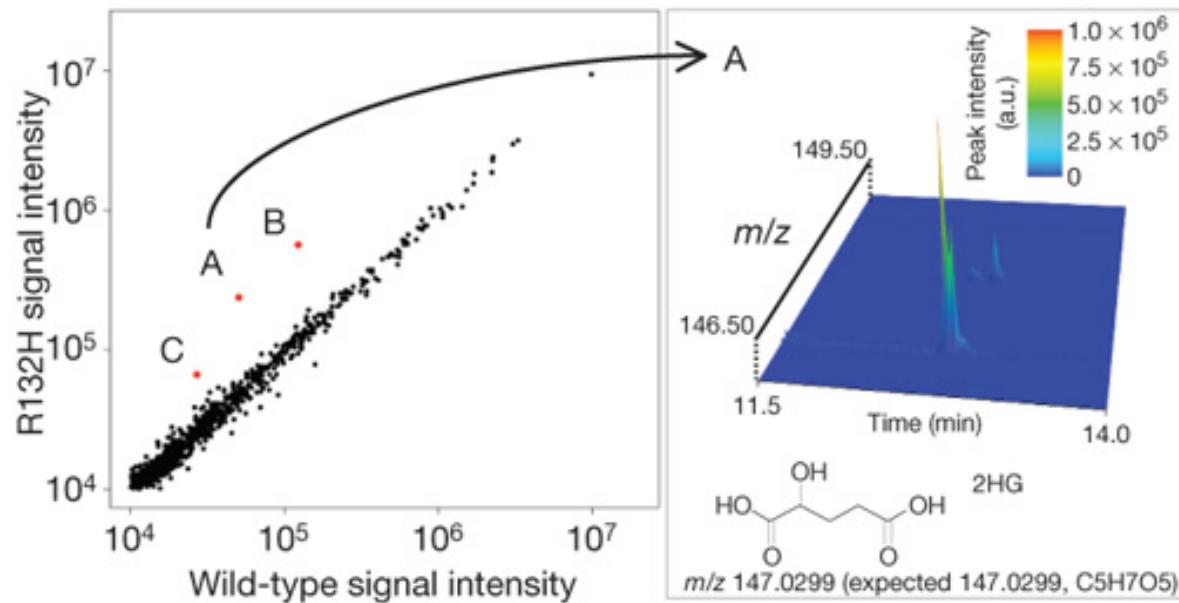
## Network visualization tools

Cytoscape: <http://cytoscape.org>  
Gephi: <https://gephi.org>

## General programming and Data analytics

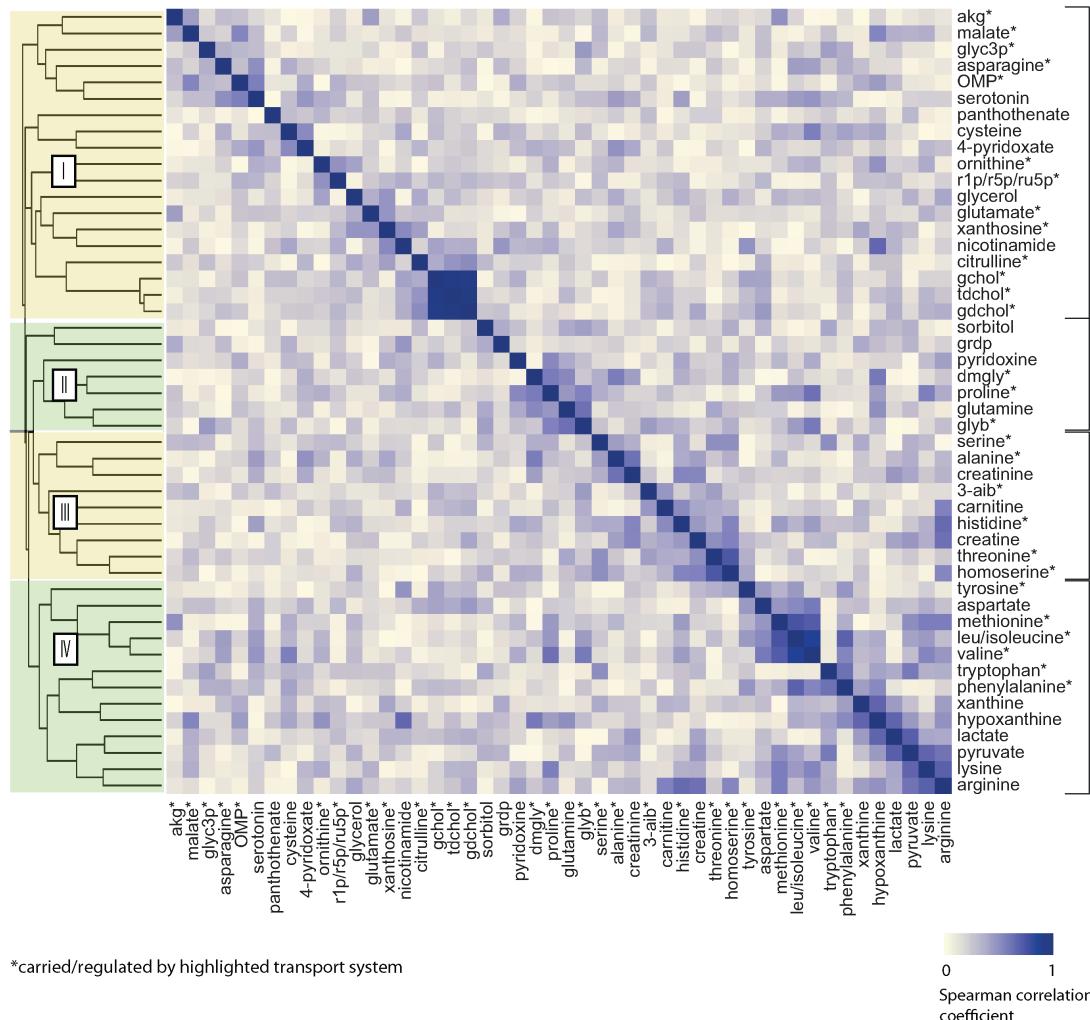
Python – Conda – Pandas – Seaborn  
R – Bioconductor – Rstudio  
Jupyter Notebooks

# Limits of Pathway analysis (I)



Dang *et al.* *Nature* **462**, 739 (2009) doi:10.1038/nature08617  
Cancer-associated IDH1 mutations produce 2-hydroxyglutarate

# Limits of Pathway analysis (II)



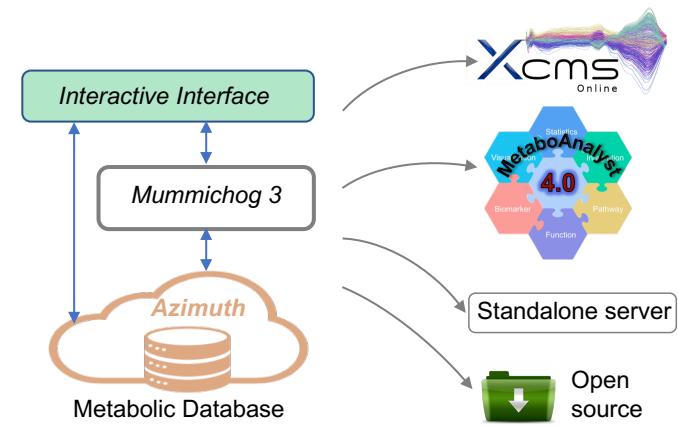
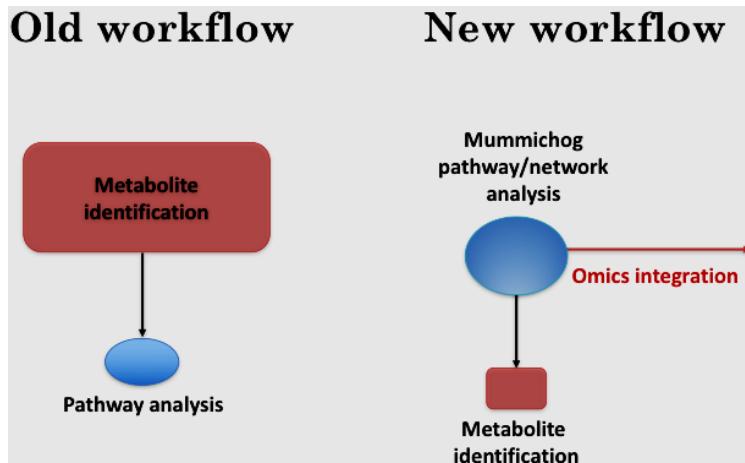
Deo et al. (2010) PLOS Computational Biology 6(2): e1000692.  
Glucose challenge induces many transporter reactions.

# Summary

- Metabolic pathways are part of metabolic models
- Pathway analysis for targeted metabolomics data typically involve over-representation test and mapping
- *Mummichog* identifies pathway patterns from untargeted metabolomics
- Statistical significance ≠ Biological significance
- Need validation of the hypotheses generated from bioinformatics tools
- Pay attention to metabolites and confirmation levels

# Mummichog & future development

- Active development of version 2 and server
- Separating metabolic models to a new database, updated independently from the software
- Send in your feature requests



- Coupling visual analytics
- Adding compatibility with upstream data processing
- Incorporating research in metabolic network reconstruction, especially using high-resolution mass spectrometry data
- Incorporating network alignment methods for data integration

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