



# Spectra Processing, Compound Annotation, Functional Insight and Causal Analysis using **MetaboAnalyst 6.0**

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# Schedule

## Part I: 2:15 pm – 4:15 pm

- **2:15 – 2:30:** General introduction
- **2:30 - 3:00:** Untargeted metabolomics
  - ✓ LC-MS & MS/MS spectral processing
  - ✓ From peaks to functions
- **3:05 – 3:25:** Live demo
- **3:25 – 4:15:** Hands on practice

## Part II: 4:30 p.m. – 6:30 p.m.

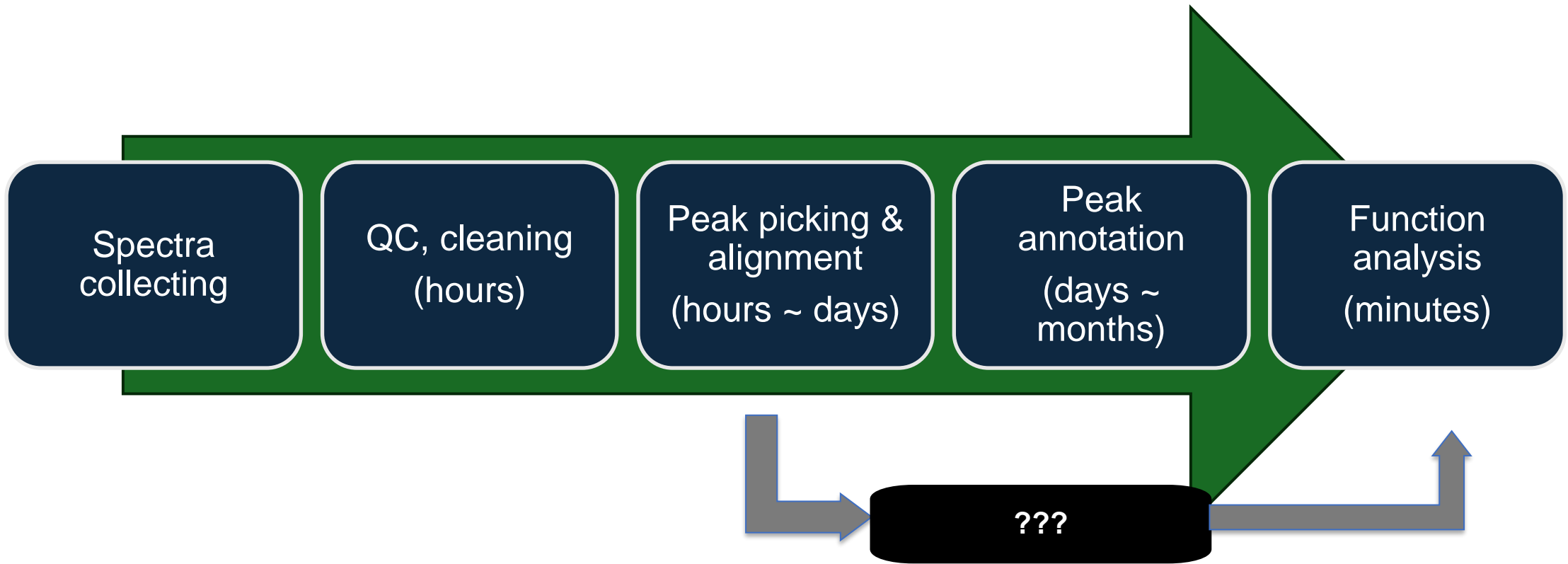
- **4:30 – 5:15:** Background
  - ✓ Statistical analysis
  - ✓ Causal analysis
- **5:15 – 5:35:** Live demo
- **5:40 – 6:20:** Hands on practice
- **6:20 – 6:30:** Summary & discussion

# Github Repository

- [https://github.com/xia-lab/Metabolomics\\_2024](https://github.com/xia-lab/Metabolomics_2024)
- Slides (in PDF format);
- Example data;
- Reference literatures;
- Contact information.

# Functional Analysis

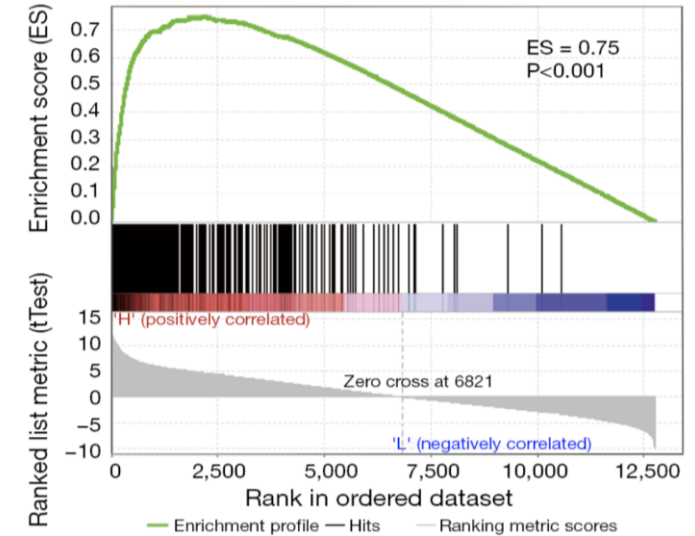
# Conventional approaches is time consuming



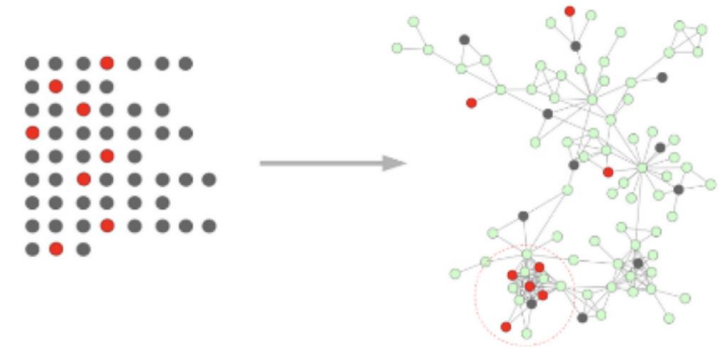
Can we perform enrichment analysis direct from peaks?

# Key concept

- Biological systems showing **coordinated changes or group behaviors**
- Leveraging this collective power inherent in biological systems can tolerate the random errors/inaccuracies based on individual genes/metabolites
  - Gene set enrichment analysis (GSEA)
- Can we apply this concept on MS peaks?
  - → The ***mummichog*** approach



A Subramanian *et al.*, *PNAS* (2005).



S. Li *et al.*, *Plos Computational Biology* (2013).

# Critical: Input preparation

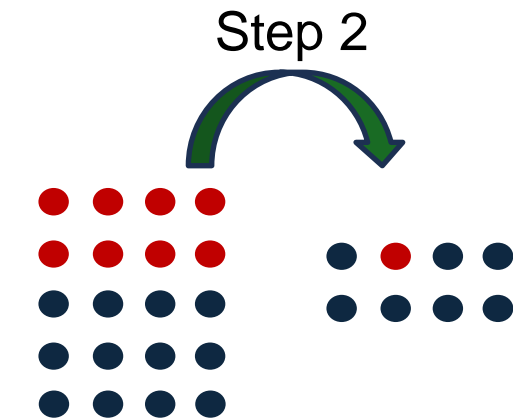
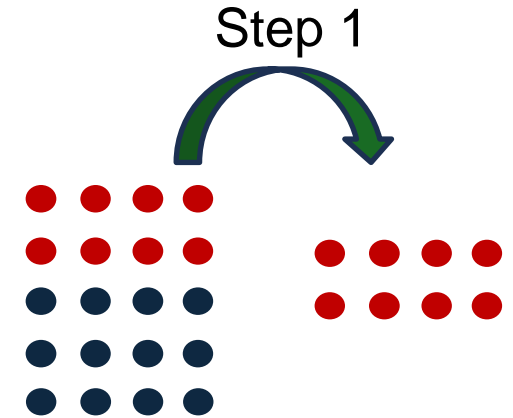
- LC - **high-resolution** MS (LCHR-MS)
  - Orbitrap, Q-TOF
  - Reason: putative annotation needs to be approximately correct (better guess leads to more accurate functional analysis)
- Needs to be **complete** peak list or peak intensity table
  - Not just significant peaks
  - Reason: mummichog using permutation to estimate the null/background distribution
- In general, the algorithm works well for **> 3000** peaks (assuming human plasma samples).

# How does mumichog work?

**Step 1:** Match the peaks to tentative metabolites. Looked up all the significant metabolites in each pathway and calculate the p-value using Fisher's exact test

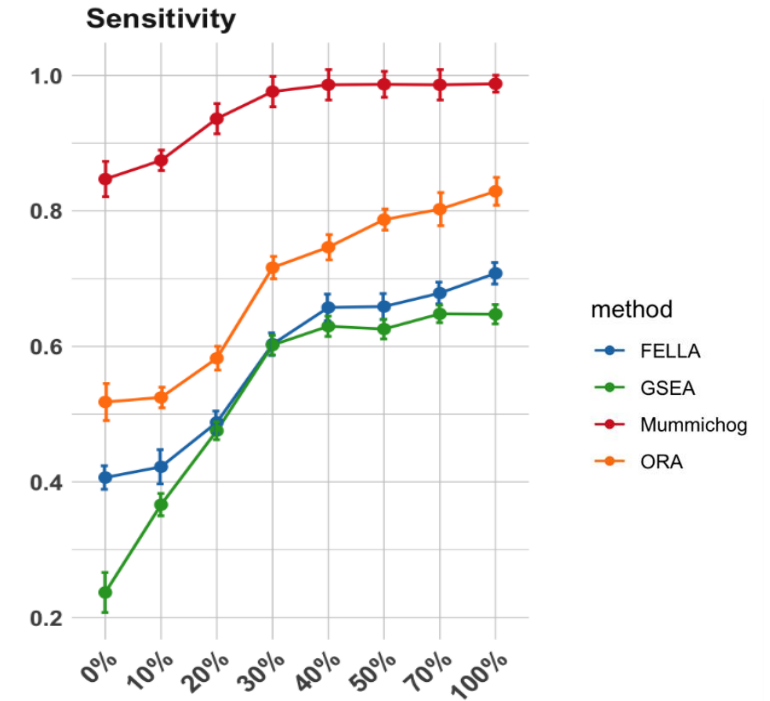
**Step 2:** Randomly pull features with the same length as the significant ones, and repeat Step 1 for 100 ~ 1000 times

**Step 3:** Test if certain pathways are enriched in the significant peaks as compared to null models (Gamma distribution)



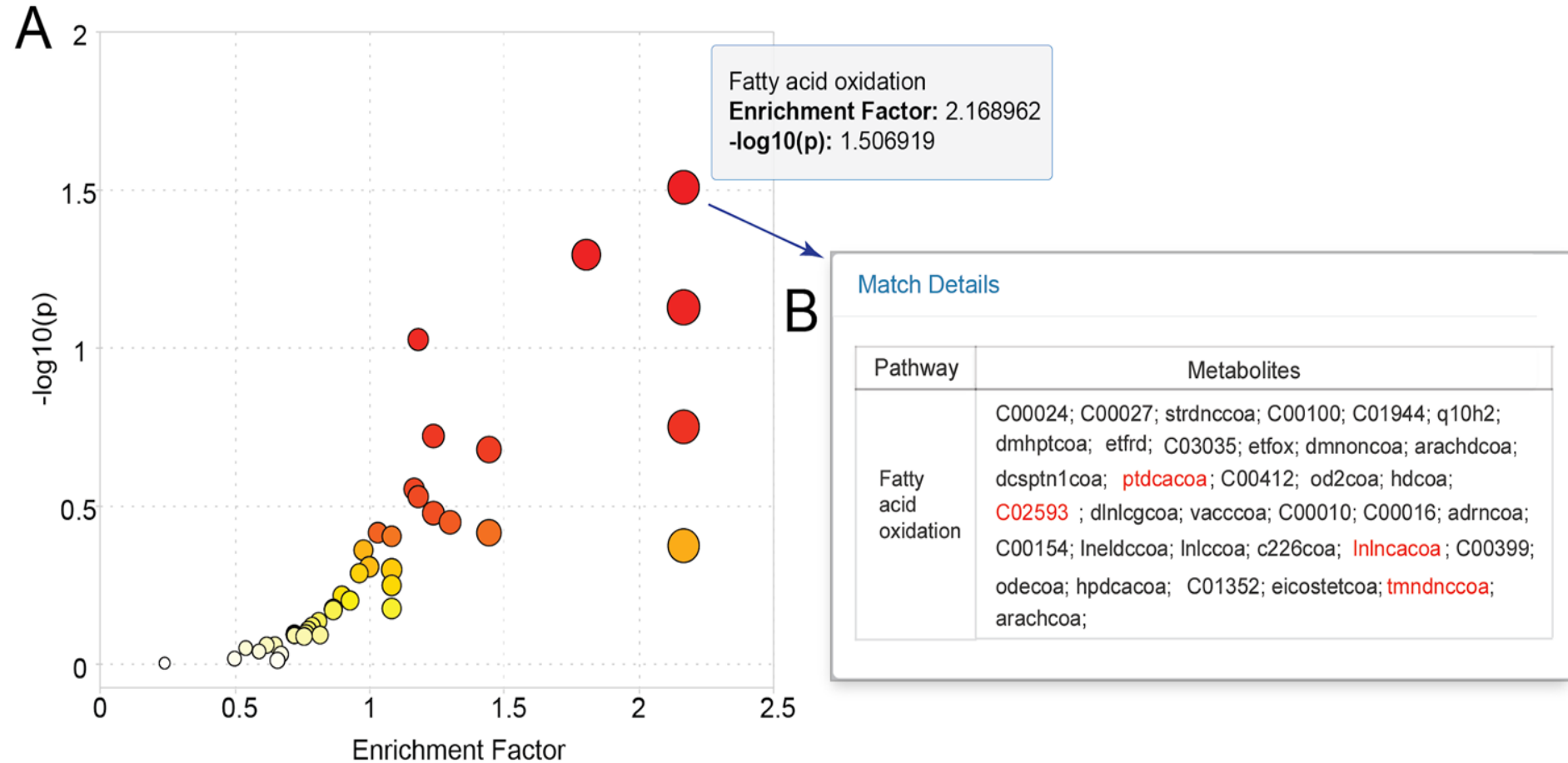


# Mummichog is more sensitive, robust compared to GSEA



Y. Lu et al., *Briefings in Bioinformatics* (2023).

# From ranked peak lists to functions

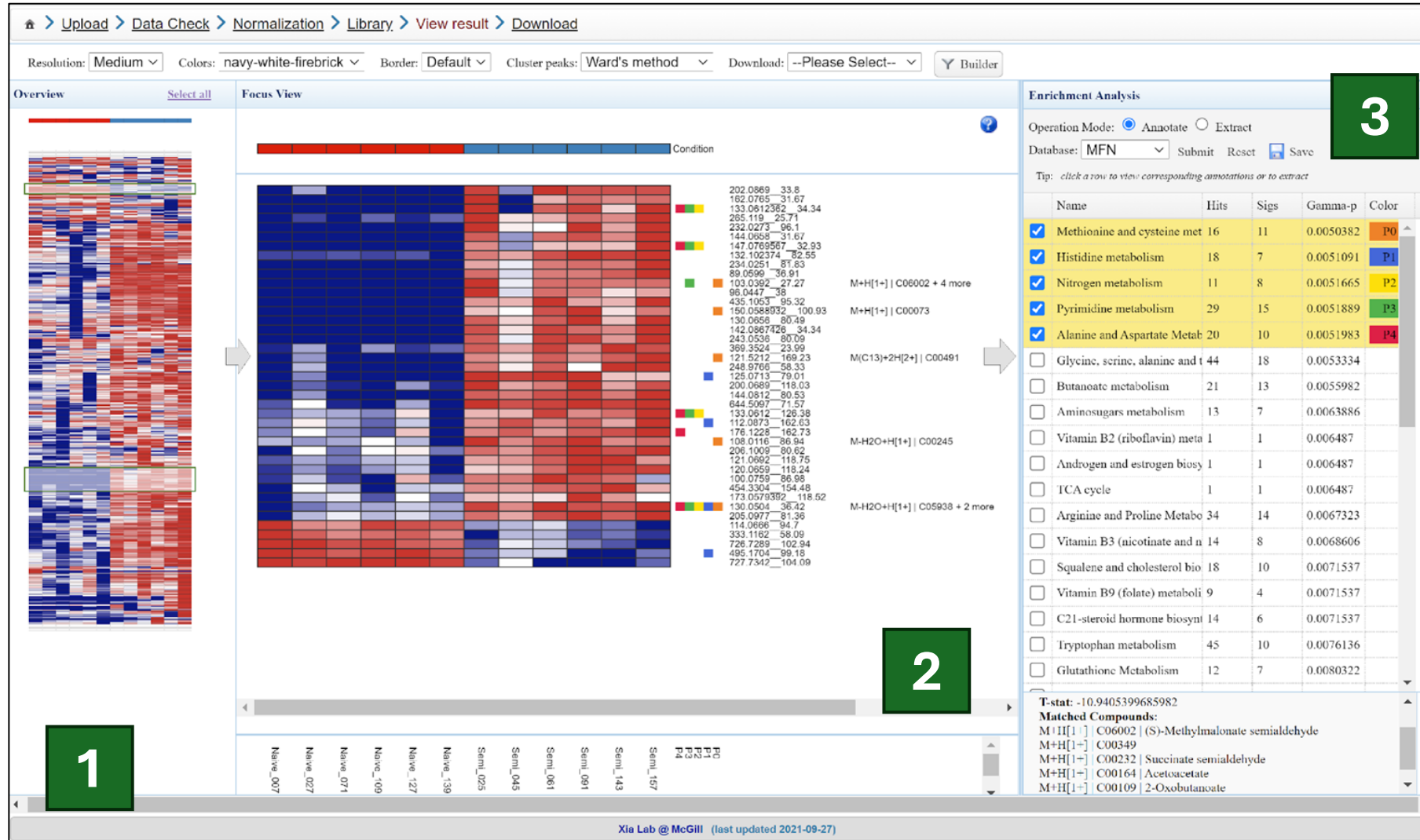


# How to interpret result table (key parameters)

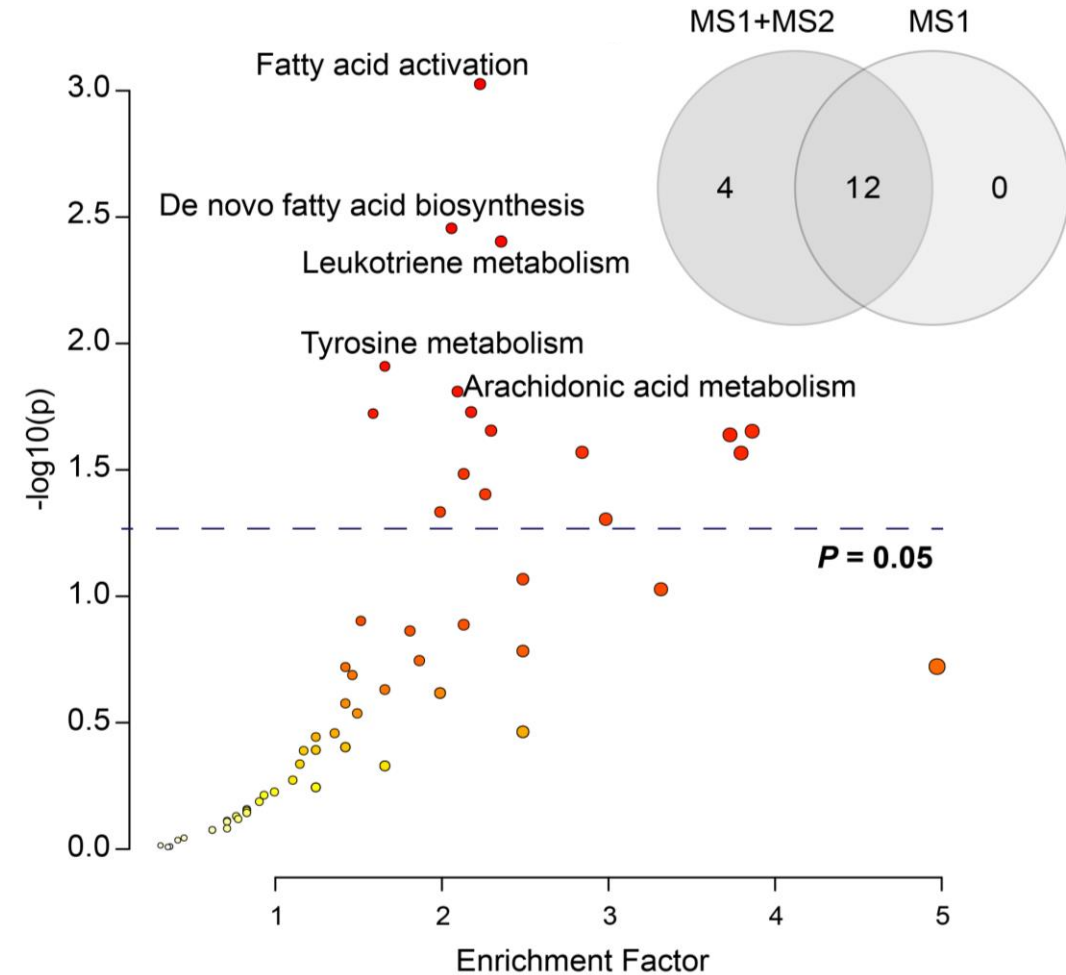
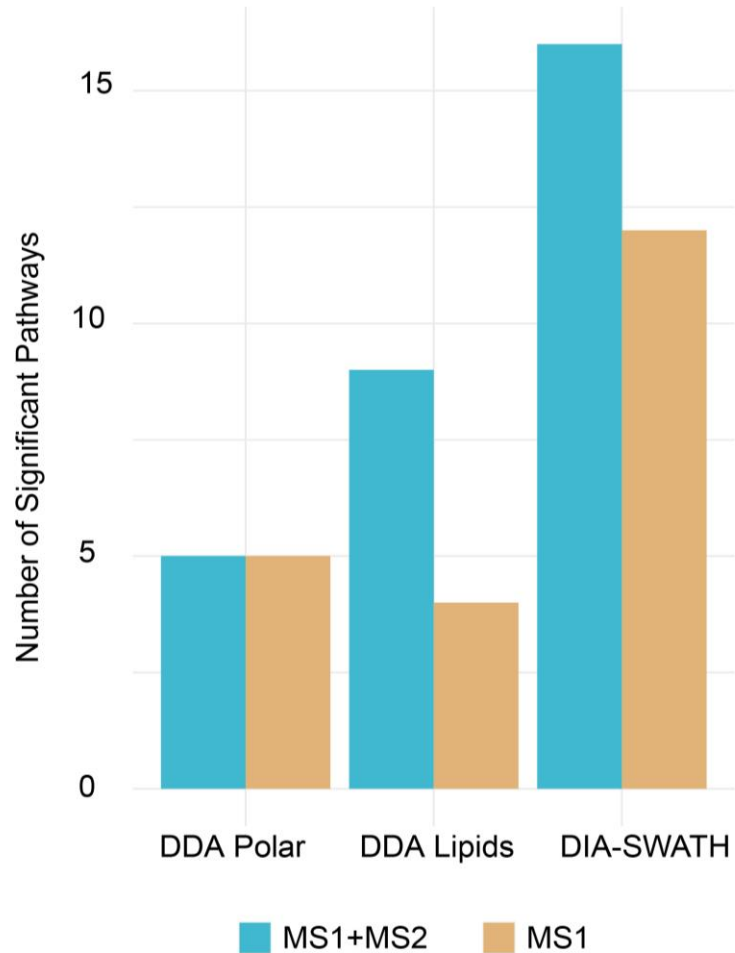
Pathway Name	Total ↑↓	Hits (all) ↑↓	Hits (sig.) ↑↓	Expected ↑↓	P(Fisher) ↑↓	P(Gamma) ↑↓	Details
Vitamin E metabolism	54	38	15	5.0563	0.030024	0.025523	<a href="#">View</a>
Carnitine shuttle	72	25	10	6.7418	0.06554	0.028334	<a href="#">View</a>

- **Total:** the total number of the given pathway
- **Hits (all):** all the peaks mapped to the pathway
- **Hits (sig):** all the significant peaks mapped to the pathway
- **Expected:** The expected number of metabolite hits in the pathway.
- **P(Fischer):** The Fisher's exact p-value for the pathway
- **P(Gamma):** P-values derived from Gamma distribution based on permutation tests for the pathway.

# Not just significant peaks



# Integrating LC-MS and MS/MS improves results



# Recap: a unified workflow for global metabolomics

