

# Metabolomics

## From data to insights

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Stockholm - SciLifeLab - 18<sup>th</sup> February 2020

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Science for Life Laboratory - KTH - Stockholm, Sweden



THE HUMAN PROTEIN ATLAS

SciLifeLab

# Learning Objectives

*You will learn about*

- Metabolomics / Targeted and Untargeted
- Analytical technologies
- Metabolomics applications & Limits
- Main processing pipelines
- Tips, pitfalls and traps

# Outline

**What is metabolomics?**

**Experimental Design**

**Data Generation / Analytical technologies**

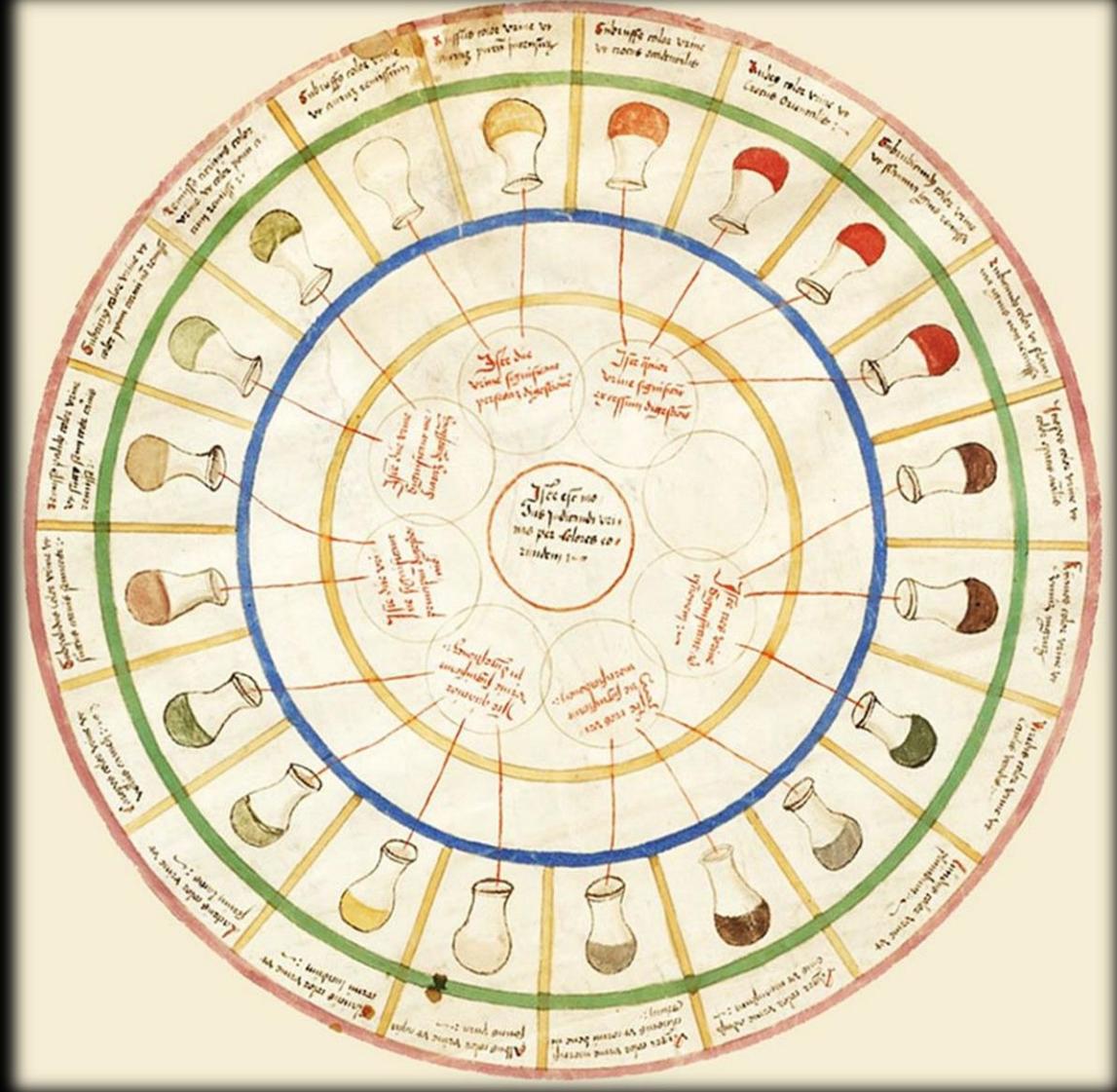
**Applications and Limits**

**Preprocessing**

**Data cleaning**

**Key steps and best practice**

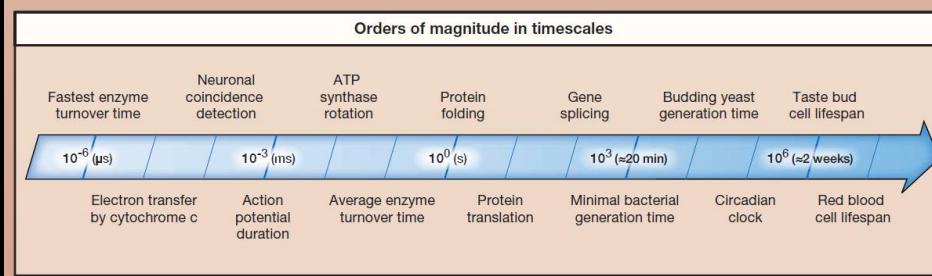
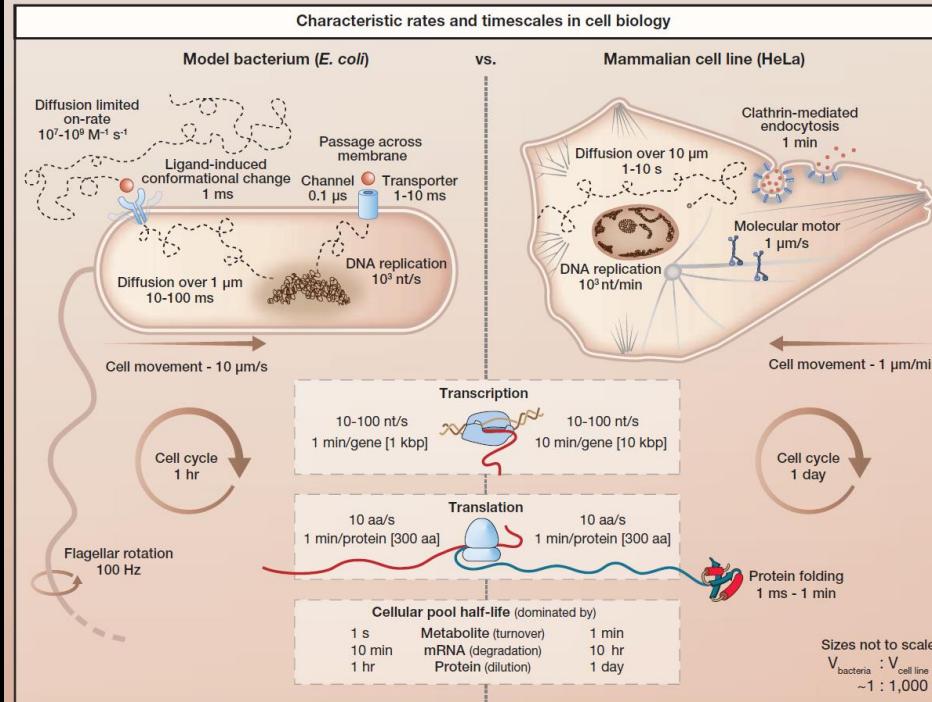
## *The urine wheel*



# Time scales in biology

## SnapShot: Timescales in Cell Biology

Maya Shamir,<sup>1</sup> Yinon Bar-On,<sup>1</sup> Rob Phillips,<sup>2</sup> and Ron Milo<sup>1</sup>  
<sup>1</sup>Weizmann Institute of Science, Rehovot 7610001, Israel  
<sup>2</sup>California Institute of Technology, Pasadena, CA 91125, USA



# Metabolomics

*What can happen*

Genomics

DNA

*What appears to be happening*

Transcriptomics

RNA

*What makes it happen*

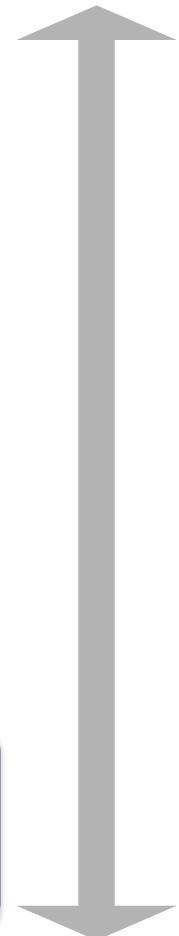
Proteomics

Proteins

*What has happened and is happening*

Metabolomics

Metabolites  
Amino acids, sugars, nucleotides  
lipids (Lipidome)



# What is a metabolite?

## Metabolites:

**Small molecules (<1500 Da)**

**Ultimate support of the biological information**

**Includes human & microbial products**

**Endogenous metabolites:** produced by the host organism

**Exogenous metabolites:** not produced by the host organism

## Metabolome:

**refers to the complete set of metabolites in a biological sample**

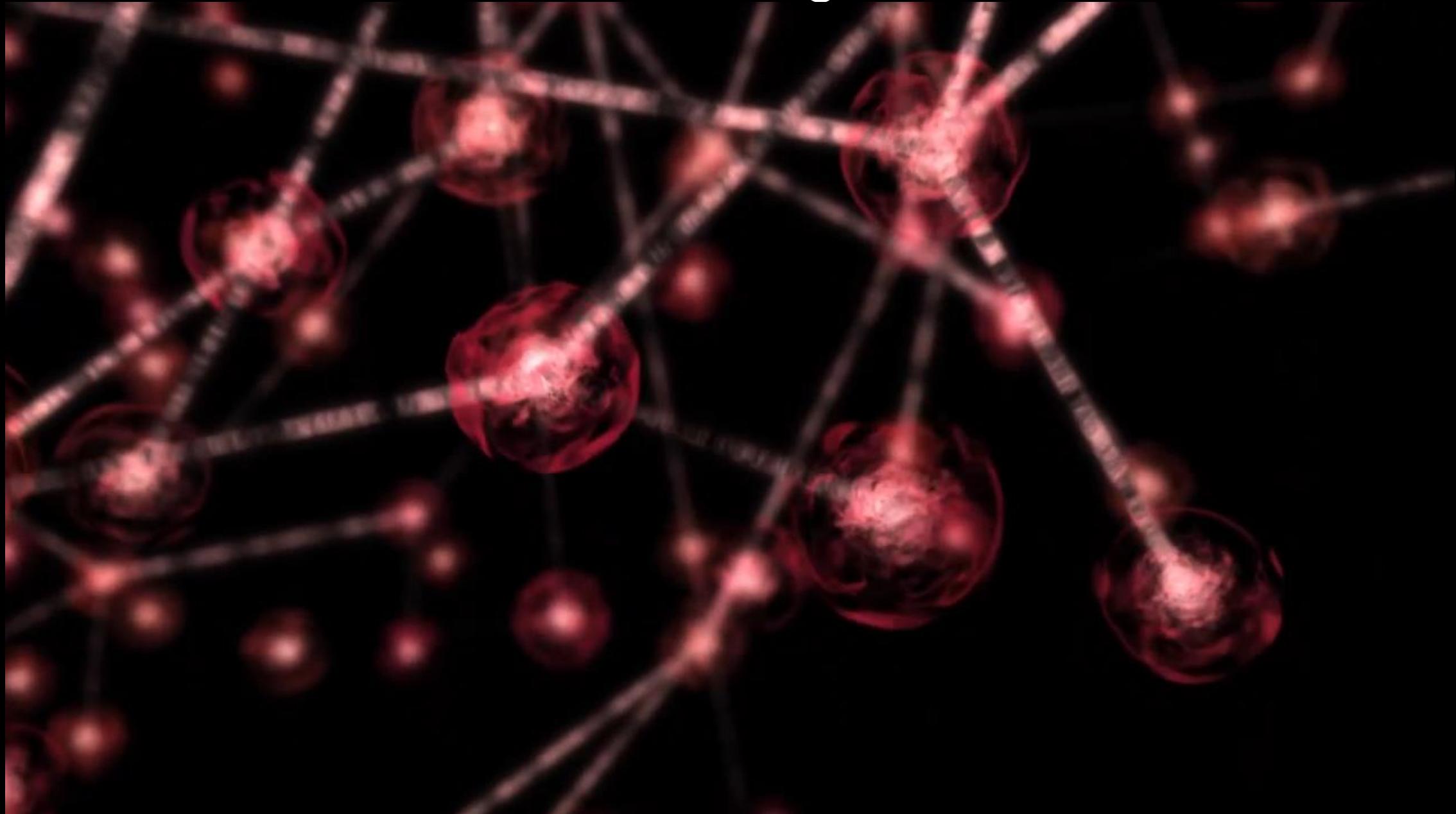
## MetaboLomics / MetaboNomics:

**→ Metabolic Profiling / Metabotyping**

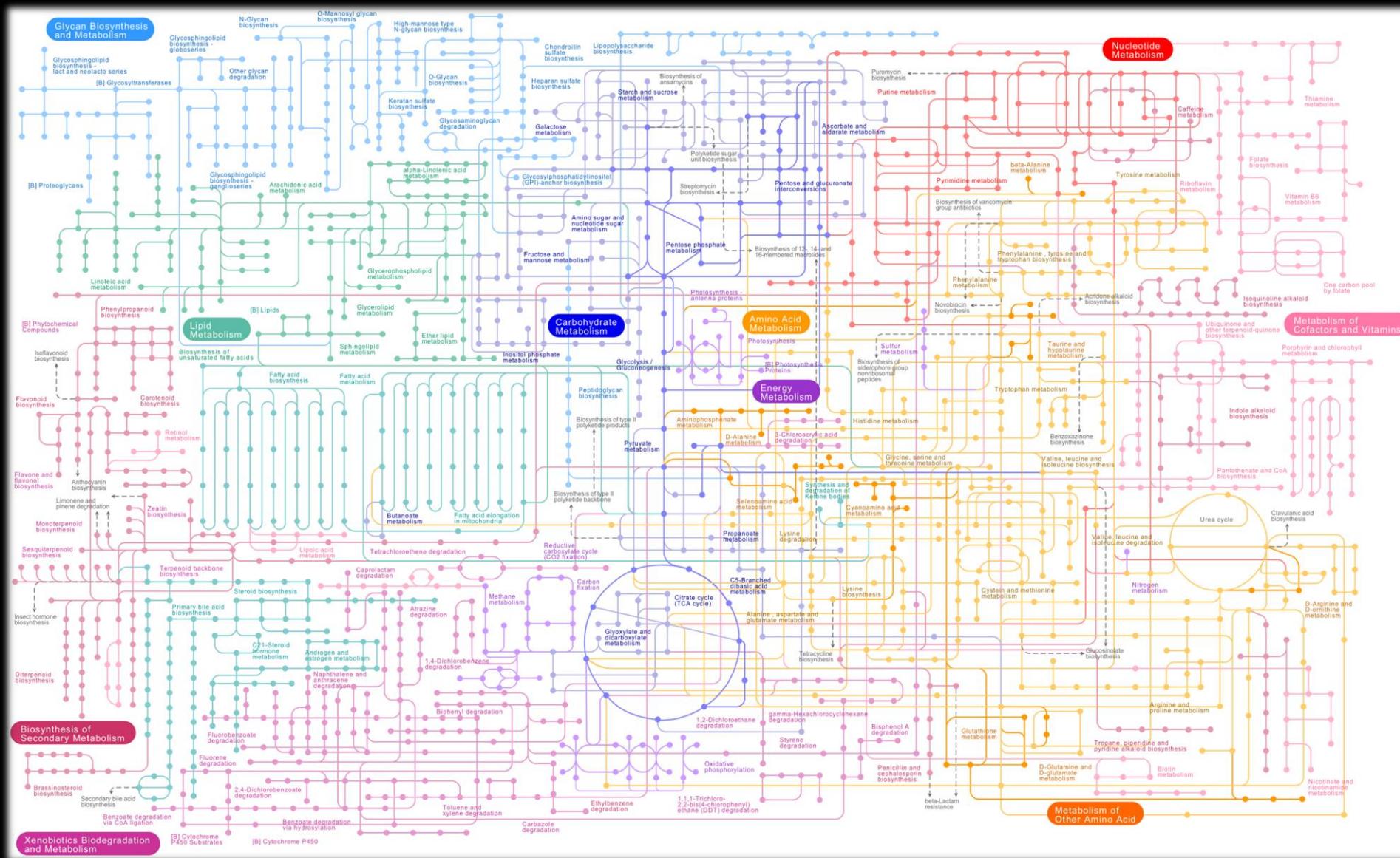
“the quantitative measurement of the metabolic **responses** of **complex systems** to a pathophysiological **stimulus** or genetic modification”.

(Nicholson, J. K., et al 1999, Xenobiotica, 29, 1181-89.)

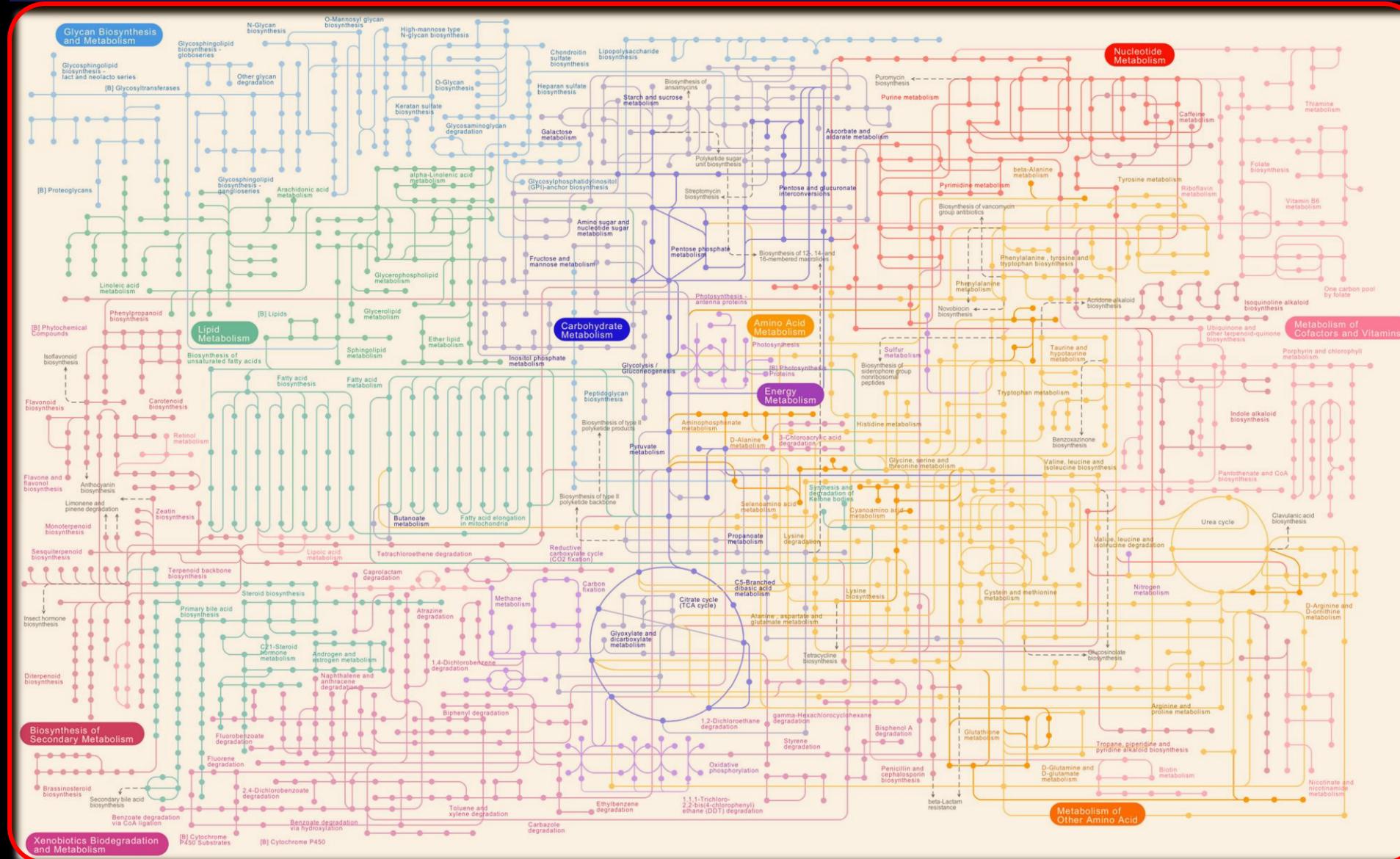
# Metabolism is an integrated network



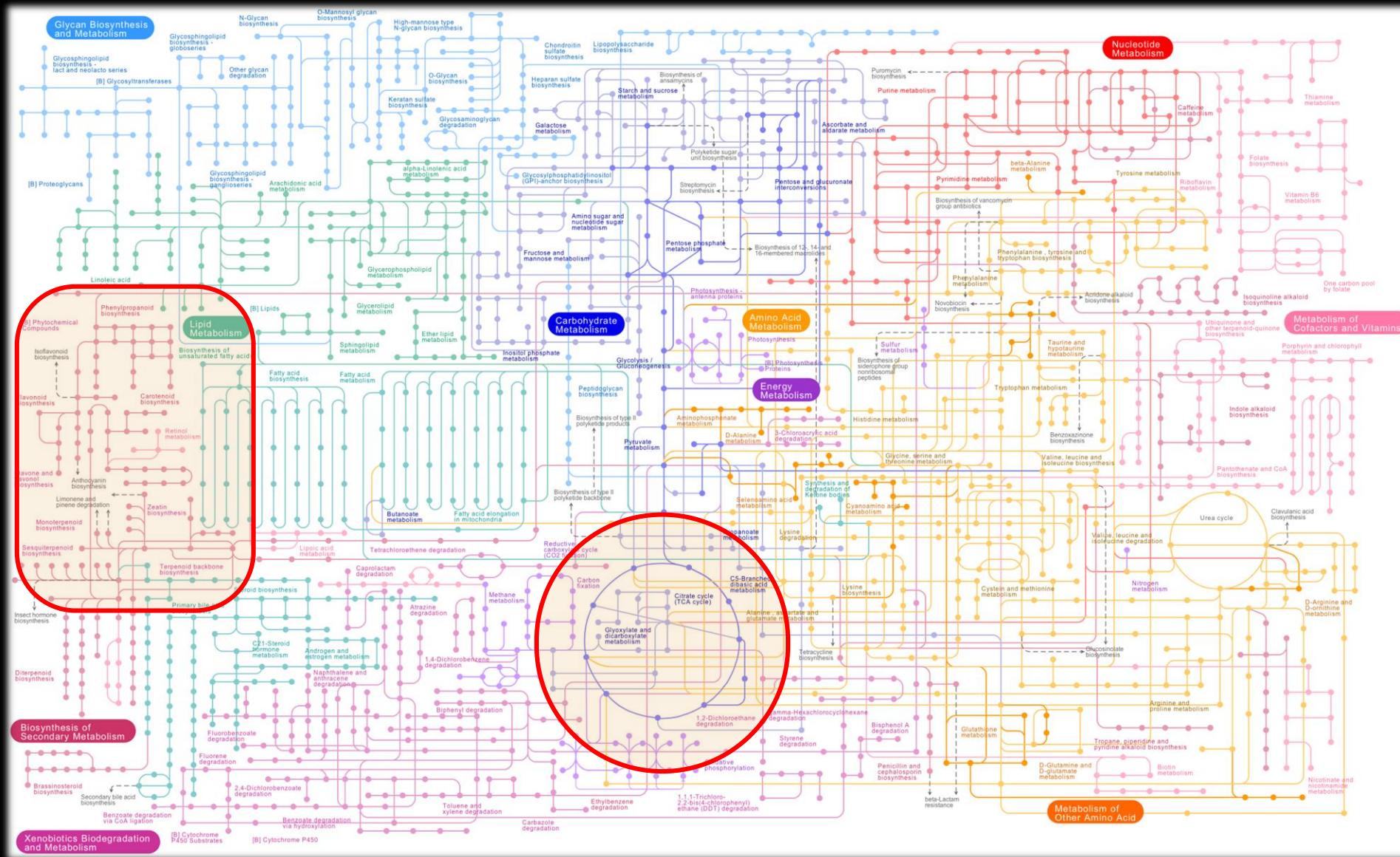
# Metabolomics and Metabolism



# Metabolomics (Untargeted)



# Metabolomics (Targeted)



# Biological Information Generation



# Experimental Design

**“To consult the statistician after an experiment is finished is often merely to ask him to conduct a post-mortem examination. He can perhaps say what the experiment died of.”**

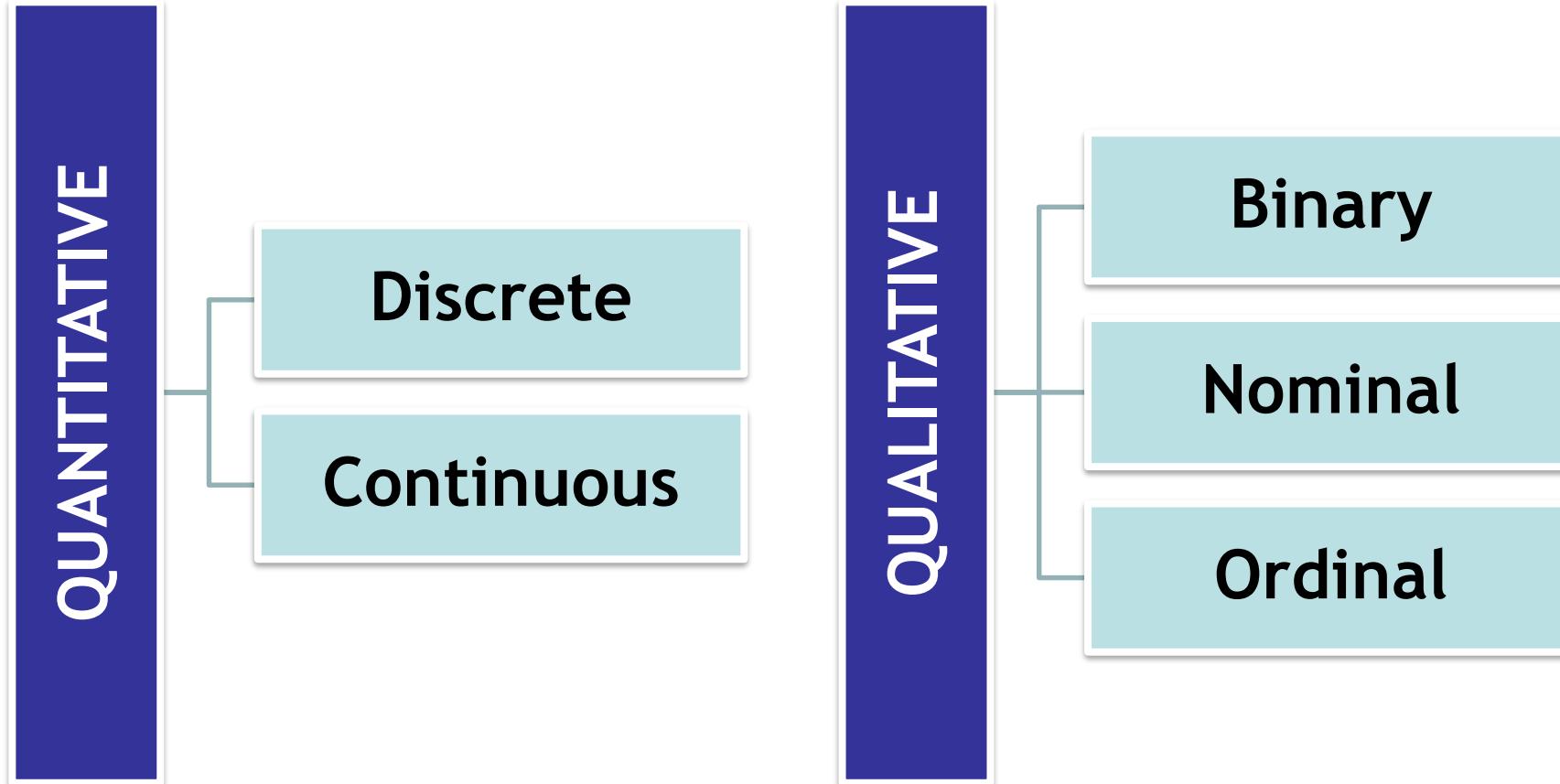
***Sir Ronald Fisher (1938)***



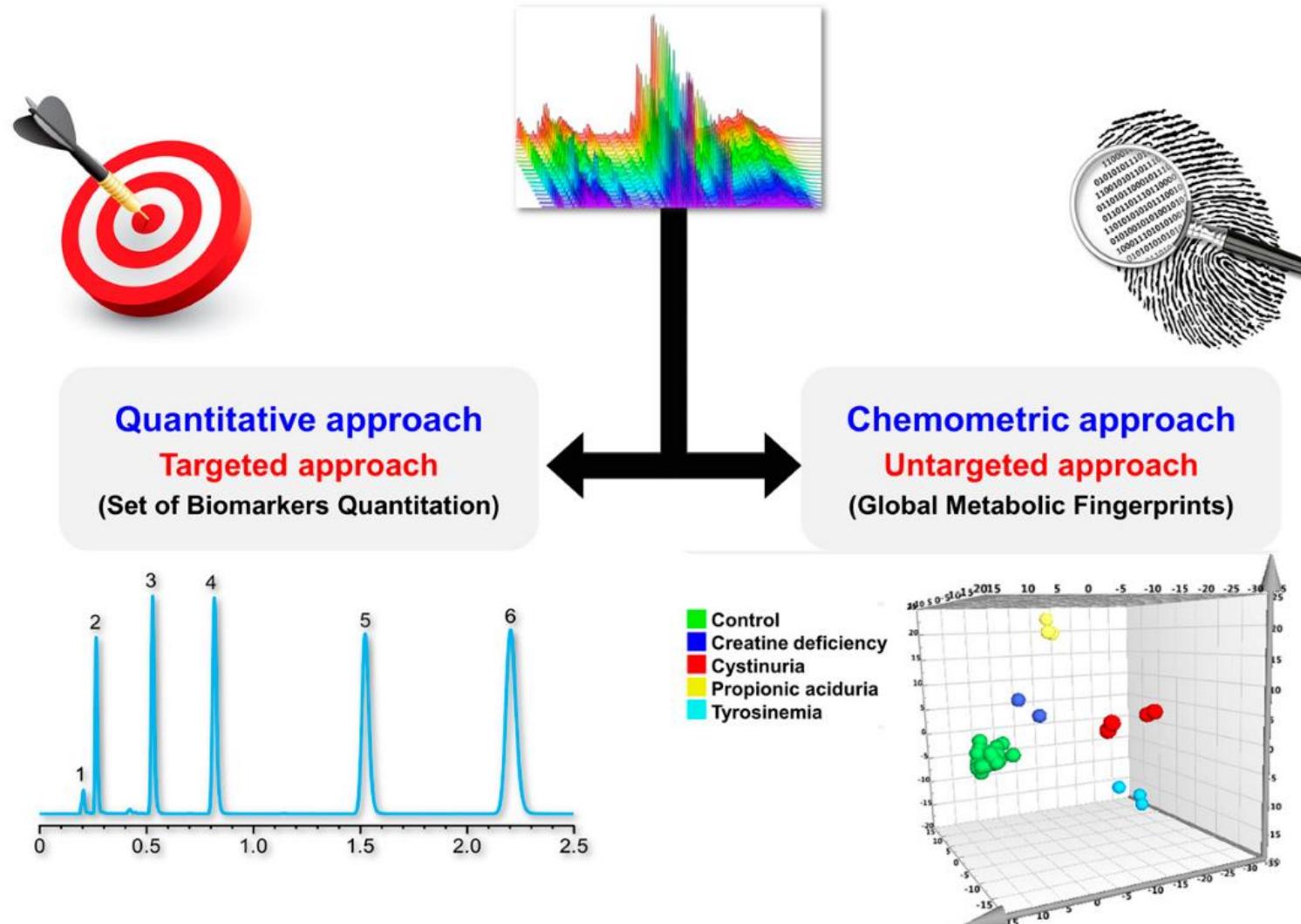
# Experimental Design **may help**

1. Clear and precise study objective
2. Sample type and size
3. Sampling and sample preparation strategy
4. Number of samples / Biological - Analytical replicates
5. Analytical technology(ies)
6. Collection of meta-data (Categorical, continuous, ordinal...)
7. Confounding factors
8. Randomization
9. Data analysis strategies (univariate vs. multivariate)
10. Biological Interpretation and insights
11. Validation (Biological/Analytical)
12. ...**Name it**

# Know your data



# Metabolomics



Tebani et al. IJMS. 2016

# Metabolomics Workflows Overview



Tebani et al. JIMD. 2017

# Metabolomics worflow

1

2

3

Tebani, A. *IJMS*. 2016, 17, 1167

# Biological Information Extraction



NMR spectroscopy



Mass spectrometry

# Biological Information Extraction

## NMR spectroscopy

Tissues, biofluids and extracts

Interaction of spin active nuclei ( $^1\text{H}$ ,  $^{13}\text{C}$ ,  $^{31}\text{P}$ ) with electromagnetic fields gives molecular information

Non-destructive  
Cross-instrument robustness

## Mass spectrometry

Tissues, biofluids and extracts

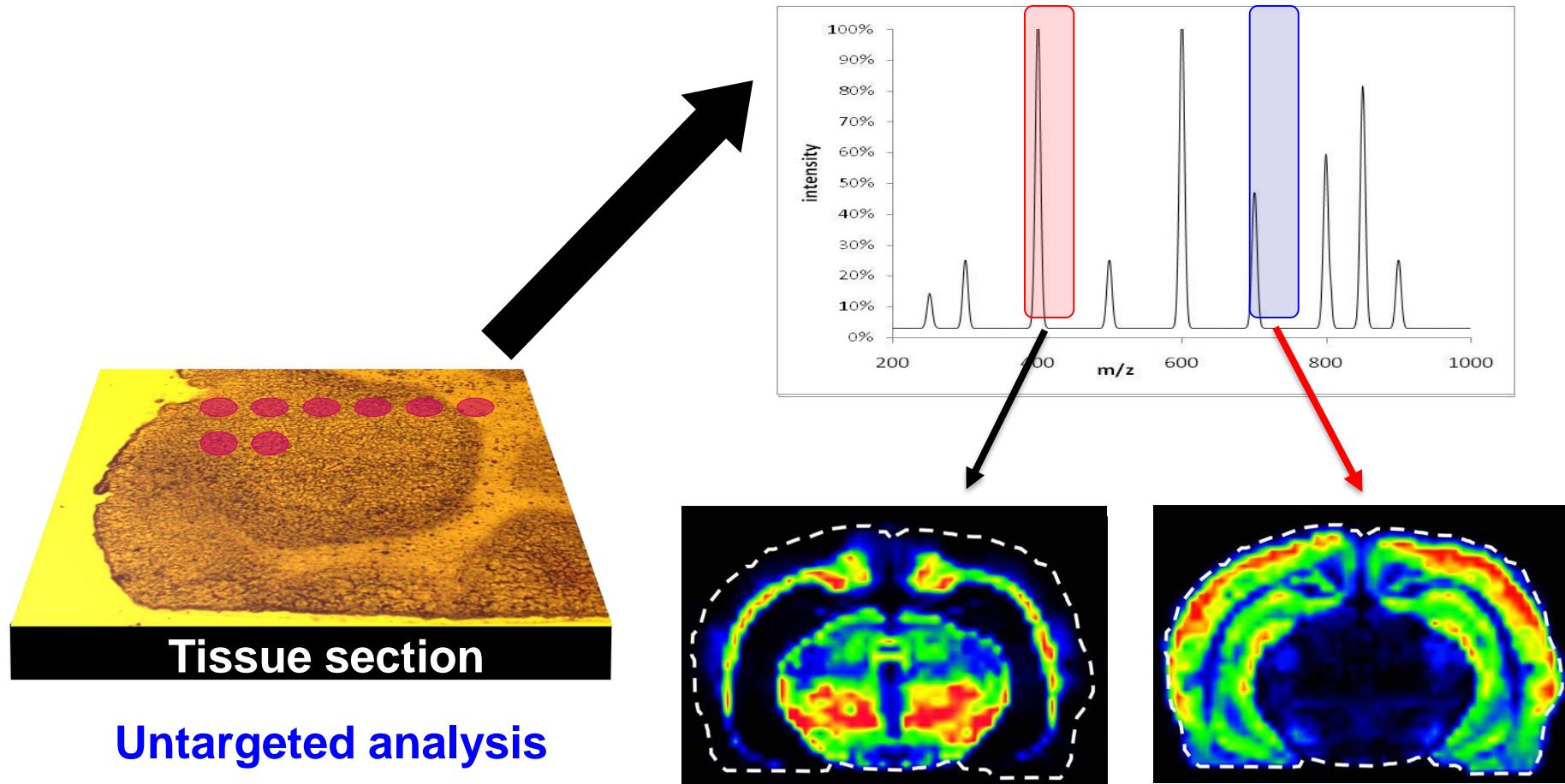
Mass to charge ratio ( $m/z$ )

Sensitivity  
Higher metabolome coverage

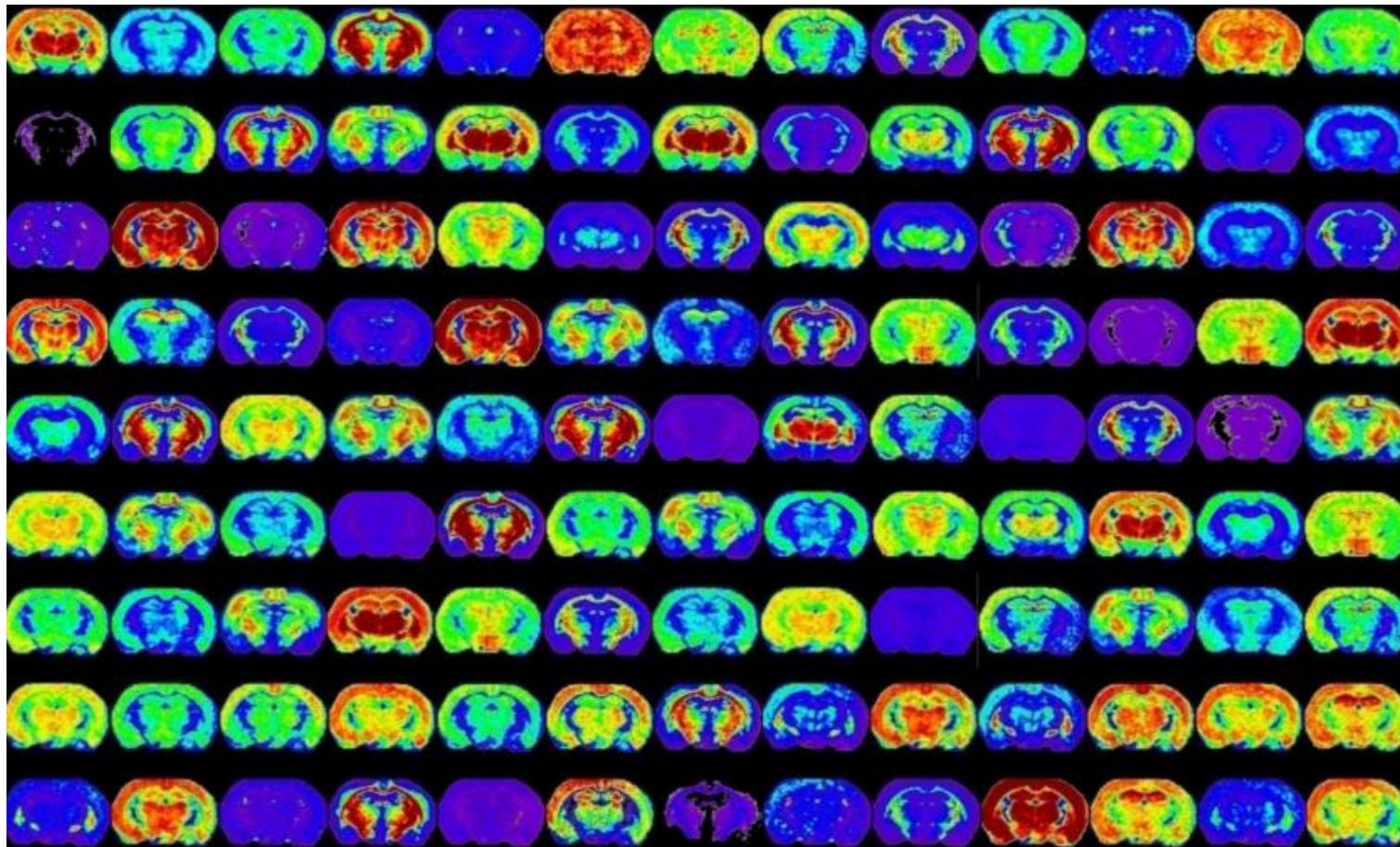
*Tebani et al. JIMD. 2016*

# Metabolomics-based imaging

## Metabolomic Imaging



# Metabolomics-based imaging



**Journal of Mass Spectrometry**

Volume 46, Issue 2, pages 209-222, 24 JAN 2011 DOI: 10.1002/jms.1876

<http://onlinelibrary.wiley.com/doi/10.1002/jms.1876/full#fig1>

# Applications

## Metabolomics

### ENABLING TECHNOLOGIES

#### Bioinformatics / Machine Learning

Predictive Analytics  
Actionable visualization technologies  
Data integration - Network Analysis

#### Advanced Analytical Strategies

Nuclear Magnetic Resonance  
Mass spectrometry

### Next-Generation Diagnostics

Clinical Chemistry - Pathology - Precision Surgery - Microbiology

Bekri S. *Expert Review of Precision Medicine and Drug Development* 1.6 (2016): 517-532  
<https://doi.org/10.1080/23808993.2016.1273067>

## Metabolomics paths towards Precision Medicine



### Populational Profiling

Epidemiological stratification  
Disease-risk biomarker discovery  
Large-scale association studies  
Public health prevention



### Individual Profiling

Patient stratification  
Personalized therapies  
Pharmacometabonomics  
Nutritional assessment



### Drug Discovery

Proof of mechanism  
Proof of action  
Pharmacokinetics  
Pharmacodynamics

# Applications

Article



molecular  
systems  
biology

The gut microbiota modulates host amino acid and glutathione metabolism in mice

Adil Mardinoglu<sup>1,2,\*†</sup>, Saeed Shoaei<sup>1,†</sup>, Mattias Bergenthal<sup>3,4</sup>, Pouyan Ghaaffari<sup>1</sup>, Cheng Zhang<sup>2</sup>, Erik Larsson<sup>3,4</sup>, Fredrik Bäckhed<sup>3,4</sup> & Jens Nielsen<sup>1,2</sup>

## Host-Microbiota interactions

*Int. J. Mol. Sci.* 2016, 17(7), 1167; doi:10.3390/ijms17071167

Open Access

Review

**Clinical Metabolomics: The New Metabolic Window for Inborn Errors of Metabolism Investigations in the Post-Genomic Era**

Abdellah Tebani <sup>1,2,3</sup> , Lenaig Abily-Donval <sup>2,4</sup> , Carlos Afonso <sup>3</sup> , Stéphane Marret <sup>2,4</sup> and Soumeya Bekri <sup>1,2,\*</sup>

## Inherited Metabolic Diseases

RESEARCH ARTICLE

CANCER DIAGNOSTICS

**Intraoperative Tissue Identification Using Rapid Evaporative Ionization Mass Spectrometry**

Júlia Balog,<sup>1,\*</sup> László Sasi-Szabó,<sup>2,\*</sup> James Kinross,<sup>3,4</sup> Matthew R. Lewis,<sup>3</sup> Laura J. Muirhead,<sup>3,4</sup> Kirill Veselkov,<sup>3</sup> Reza Mirnezami,<sup>4</sup> Balázs Dezső,<sup>5</sup> László Damjanovich,<sup>2</sup> Ara Darzi,<sup>4</sup> Jeremy K. Nicholson,<sup>3†</sup> Zoltán Takáts<sup>3†</sup>

## Pathology and Cancer

## Pharmacometabolic Investigation of Dynamic Metabolic Phenotypes Associated with Variability in Response to Galactosamine Hepatotoxicity

Muireann Coen,<sup>\*,†</sup> Françoise Goldfain-Blanc,<sup>‡</sup> Gaëlle Rolland-Valognes,<sup>§</sup> Bernard Walther,<sup>||</sup> Donald G. Robertson,<sup>†</sup> Elaine Holmes,<sup>†</sup> John C. Lindon,<sup>†</sup> and Jeremy K. Nicholson<sup>\*,†</sup>

## Responder Non-responder Prediction

An Integrative Approach for Identifying a Metabolic Phenotype Predictive of Individualized Pharmacokinetics of Tacrolimus

PB Phapale<sup>1,2</sup>, S-D Kim<sup>2</sup>, HW Lee<sup>1,2</sup>, M Lim<sup>1,2</sup>, DD Kale<sup>1,2</sup>, Y-L Kim<sup>2,3</sup>, J-H Cho<sup>4</sup>, D Hwang<sup>4</sup> and Y-R Yoon<sup>1,2</sup>

## Human Drug Pharmacokinetics



ARTICLE

<https://doi.org/10.1038/s41467-019-10893-6>

OPEN

Assessing the causal association of glycine with risk of cardio-metabolic diseases

Laura B.L. Wittemans<sup>1</sup>, Luca A. Lotta<sup>1</sup>, Clare Oliver-Williams<sup>2,3</sup>, Isobel D. Stewart<sup>1</sup>, Praveen Surendran<sup>2</sup>, Savita Karthikeyan<sup>2</sup>, Felix R. Day<sup>1</sup>, Albert Koulman<sup>1,4</sup>, Fumiaki Imaamura<sup>1</sup>, Lingyao Zeng<sup>5,6</sup>, Jeanette Erdmann<sup>7,8,9</sup>, Heribert Schunkert<sup>5,6</sup>, Kay-Tee Khaw<sup>10</sup>, Julian L. Griffin<sup>11</sup>, Nita G. Forouhi<sup>1</sup>, Robert A. Scott<sup>1</sup>, Angela M. Wood<sup>2</sup>, Stephen Burgess<sup>1,2,12</sup>, Joanna M.M. Howson<sup>1,2</sup>, John Danesh<sup>2,13</sup>, Nicholas J. Wareham<sup>1</sup>, Adam S. Butterworth<sup>1,2</sup> & Claudia Langenberg<sup>1</sup>

## Large-scale epidemiological profiling

# Metabolomics-based imaging

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**'Intelligent Knife' Tells Surgeon Which Tissue Is Cancerous**

July 17, 2013 — Scientists have developed an "intelligent knife" that can tell surgeons immediately whether the tissue they are cutting is cancerous or not.

**Related Topics**  
Health & Medicine   Article

**RESEARCH ARTICLE**

**CANCER DIAGNOSTICS**

**Intraoperative Tissue Identification Using Rapid Evaporative Ionization Mass Spectrometry**

Júlia Balog,<sup>1\*</sup> László Sasi-Szabó,<sup>2\*</sup> James Kinross,<sup>3,4</sup> Matthew R. Lewis,<sup>3</sup> Laura J. Muirhead,<sup>3,4</sup> Kirill Veselkov,<sup>3</sup> Reza Mirnezami,<sup>4</sup> Balázs Dezső,<sup>5</sup> László Damjanovich,<sup>2</sup> Ara Darzi,<sup>4</sup> Jeremy K. Nicholson,<sup>3†</sup> Zoltán Takáts<sup>3†</sup>

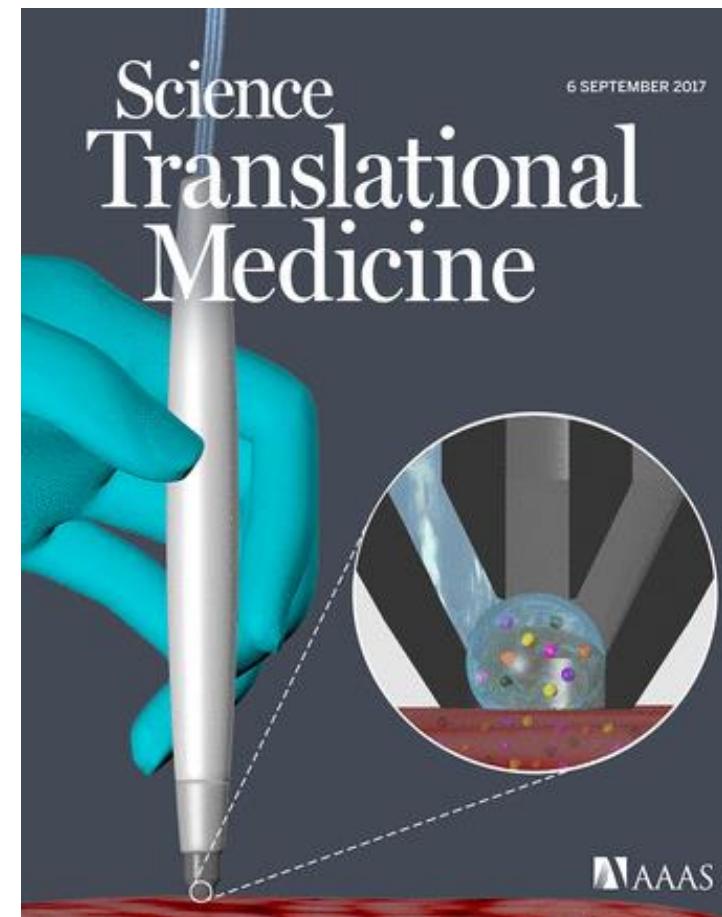
**SCIENCE TRANSLATIONAL MEDICINE | RESEARCH ARTICLE**

**CANCER DIAGNOSTICS**

**Nondestructive tissue analysis for ex vivo and in vivo cancer diagnosis using a handheld mass spectrometry system**

Jialing Zhang,<sup>1</sup> John Rector,<sup>1,2</sup> John Q. Lin,<sup>1</sup> Jonathan H. Young,<sup>1</sup> Marta Sans,<sup>1</sup> Nitesh Katta,<sup>2</sup> Noah Giese,<sup>1</sup> Wendong Yu,<sup>3</sup> Chandandeep Nagi,<sup>3</sup> James Suliburk,<sup>4</sup> Jinsong Liu,<sup>5</sup> Alena Bensussan,<sup>1</sup> Rachel J. DeHoog,<sup>1</sup> Kyana Y. Garza,<sup>1</sup> Benjamin Ludolph,<sup>1</sup> Anna G. Sorace,<sup>6</sup> Anum Syed,<sup>2</sup> Aydin Zahedivash,<sup>2</sup> Thomas E. Milner,<sup>2</sup> Livia S. Eberlin<sup>1\*</sup>

## Real time Metabolomics: Precision Surgery



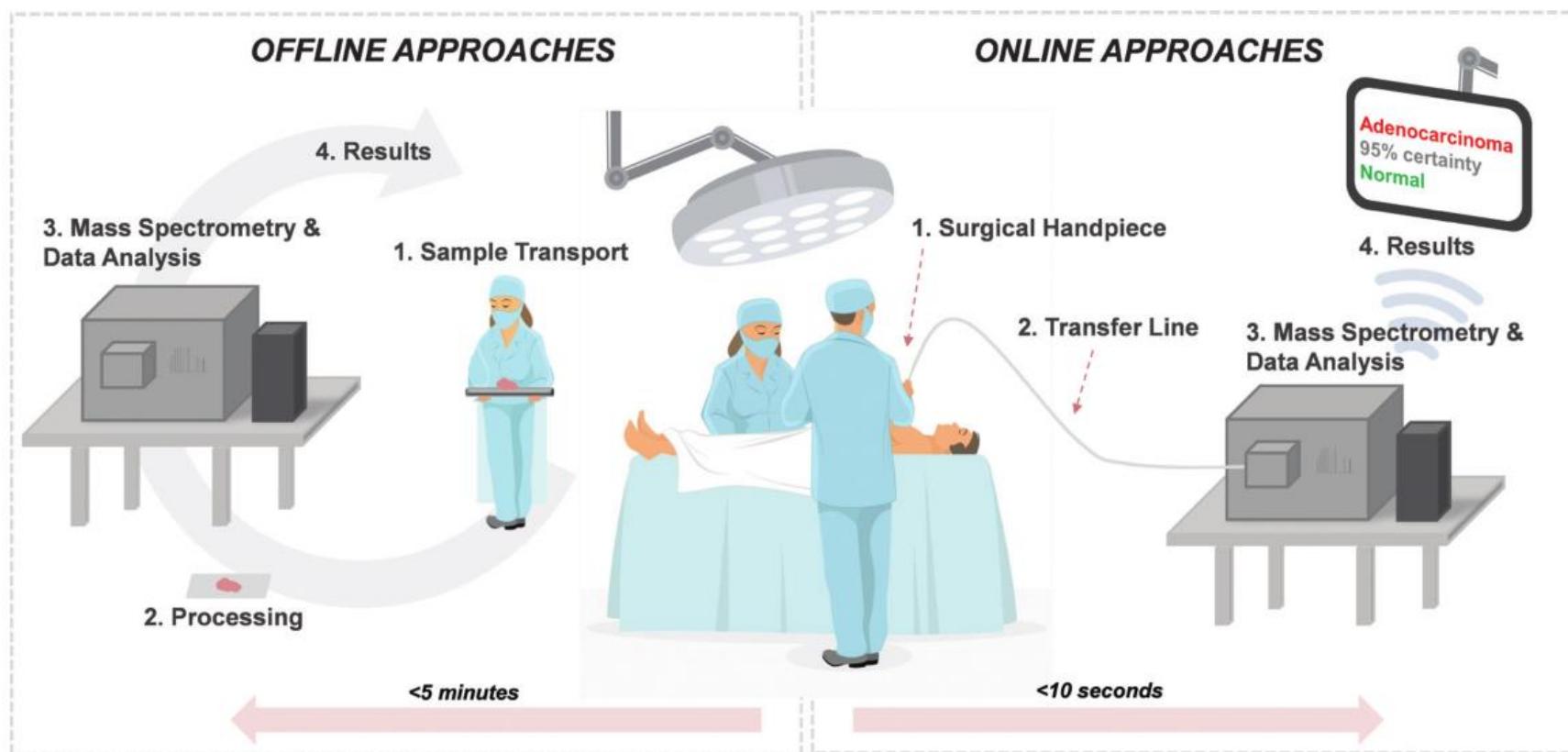
# Metabolomics-based imaging



Sci Transl Med 17 July 2013: Vol. 5, Issue 194, p. 194ra93

# Metabolomics-based imaging

## Real time Metabolomics: Precision Surgery



Demian R. Ifa et al. Clin Chem 62:1 (2016)

Sci Transl Med 17 July 2013: Vol. 5, Issue 194, p. 194ra93

# Limits

**Metabolite Identification** are the main bottlenecks of metabolomics for large adoption in both translational and clinical context.

Lack of standardized **annotation** of the metabolome is important for functional analysis and integration with other omics through GEMs

More **absolute quantification** of metabolites is needed (targeted and untargeted) to achieve reliability and robustness

**Standardization and Harmonization** is a prerequisite for large adoption

**Miniaturization** will enhance high-throughput

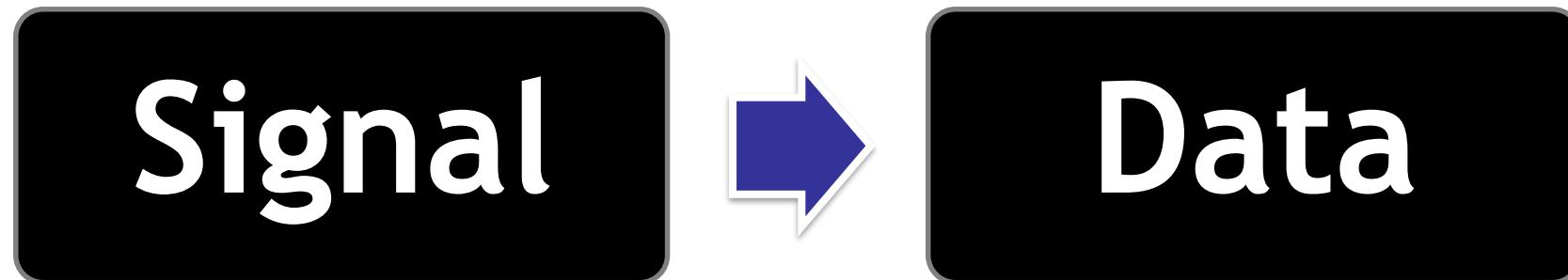
**Automation, Data Visualization and Clinical Actionability** at different stages, instrument-, pre- and post-analytic levels including data processing, integration and interpretation are very important issues for large clinical adoption of any diagnostic innovation

# **Data Analysis**

## **From signals to numbers**

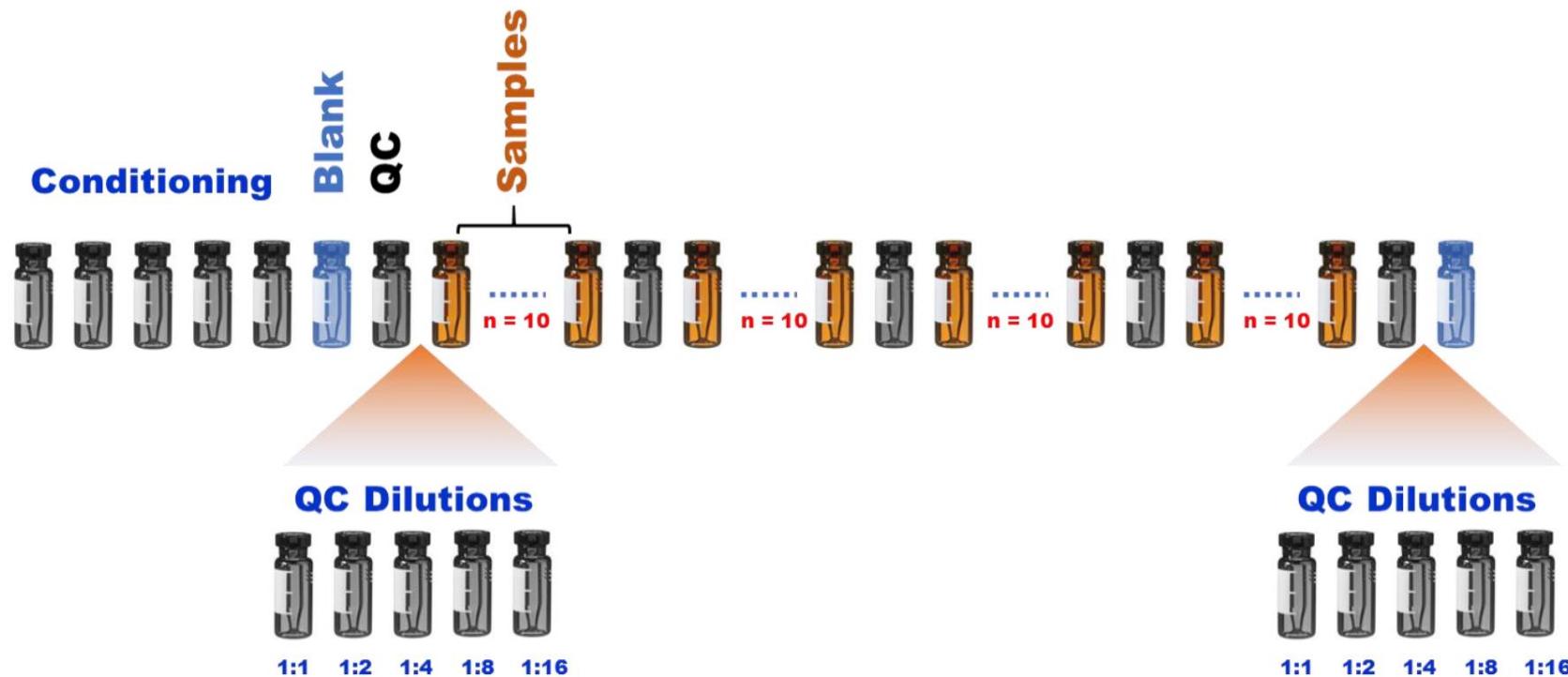
# **PREPROCESSING**

# Data processing

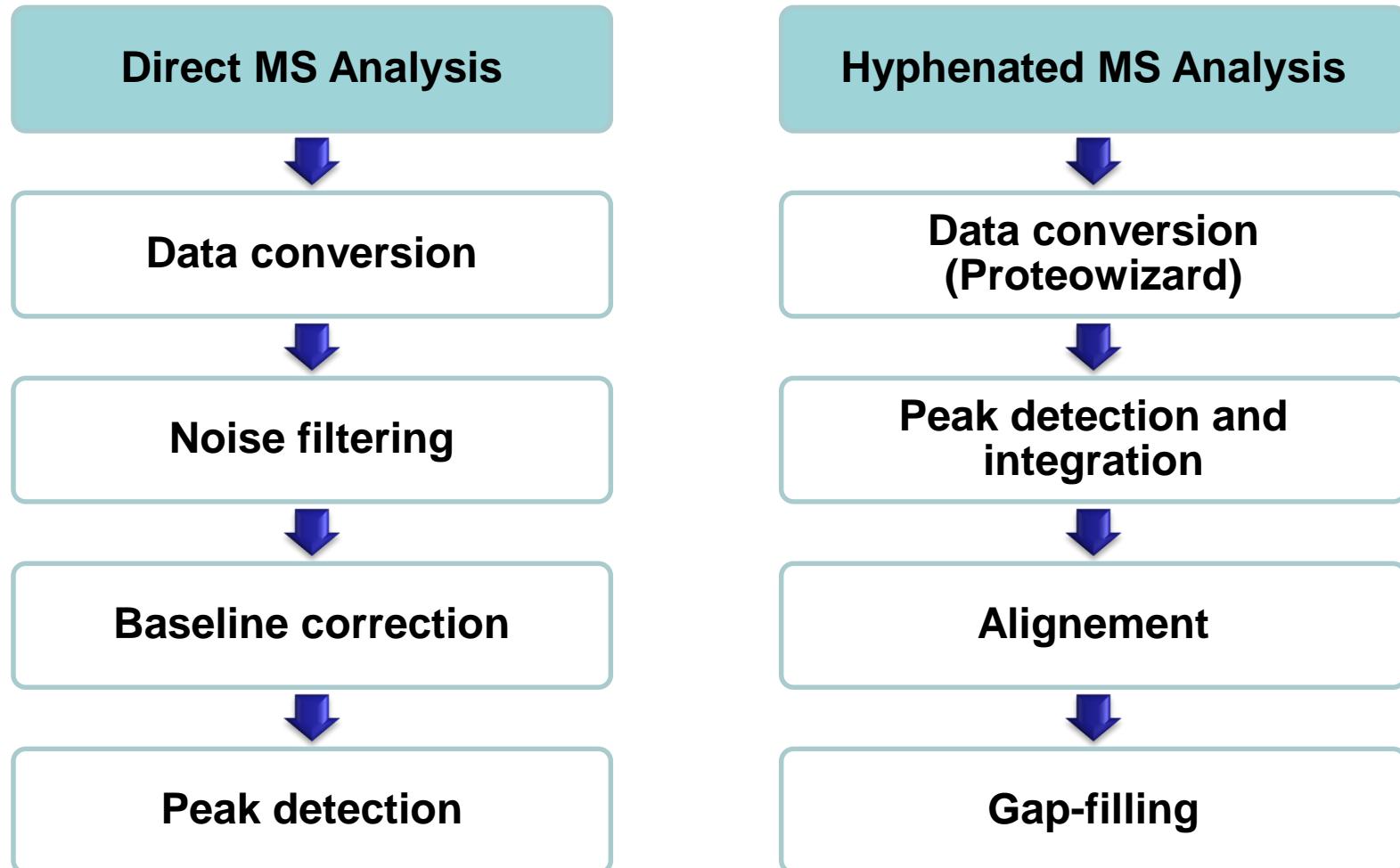


Variables									
Samples	1	2	3	4	5	6	7	8	9
6	0	4	8	0	2	2	7	2	2
0	1	8	0	8	0	0	3	4	9
4	4	0	4	0	1	4	0	1	8
8	0	2	8	2	4	8	4	0	
0	4	6	0	5	2	0	8	0	2
2	8	9	5	5	5	2	0	4	6
4	9	5	6	5	4	5	2	8	9
1	8	0	8	5	0	1	8	0	8
4	0	4	0	0	4	2	0	4	0
0	2	8	2	5	8	0	2	8	2

# Data processing



# Data processing



# Data processing softwares

**Table 1** Software tools commonly used for the preprocessing of metabolomics data

Tool	Instrument data type	Software type	Website	References
XCMS	LC-MS, GC-MS	R Package	<a href="http://bioconductor.org/packages/release/bioc/html/xcms.html">http://bioconductor.org/packages/release/bioc/html/xcms.html</a>	Smith et al. (2006)
OpenMS—FeatureFinderMetabo	LC-MS	GUI	<a href="http://ftp.mi.fu-berlin.de/pub/OpenMS/release-documentation/html/TOPP_FeatureFinderMetabo.html">http://ftp.mi.fu-berlin.de/pub/OpenMS/release-documentation/html/TOPP_FeatureFinderMetabo.html</a>	Bertsch et al. (2010)
MetAlign	LC-MS	Windows GUI	<a href="http://www.wageningenur.nl/en/show/MetAlign-1.htm">http://www.wageningenur.nl/en/show/MetAlign-1.htm</a>	Lommen & Kools (2012)
MS-DIAL	LC-MS	Windows GUI	<a href="http://prime.psc.riken.jp/Metabolomics_Software/MS-DIAL/index.html">http://prime.psc.riken.jp/Metabolomics_Software/MS-DIAL/index.html</a>	Tsugawa et al. (2015)
mzMatch	LC-MS	R Package	<a href="http://mzmatch.sourceforge.net/index.php">http://mzmatch.sourceforge.net/index.php</a>	Scheltema et al. (2011)
IDEOM	LC-MS	Excel Template	<a href="http://mzmatch.sourceforge.net/ideom.php">http://mzmatch.sourceforge.net/ideom.php</a>	Creek et al. (2012)
AMDIS	GC-MS	Windows GUI	<a href="http://chemdata.nist.gov/dokuwiki/doku.php?id=chemdata:amdis">http://chemdata.nist.gov/dokuwiki/doku.php?id=chemdata:amdis</a>	Meyer et al. (2010)
MetaboliteDetector	GC-MS	CLI, GUI	<a href="http://md.tu-bs.de">http://md.tu-bs.de</a>	Hiller et al. (2009)
MET-IDEA	GC-MS	Windows CLI	<a href="http://bioinfo.noble.org/download">http://bioinfo.noble.org/download</a>	Broeckling et al. (2006)
MeltDB	LC-MS, GC-MS	Web App	<a href="https://meltdb.cebitec.uni-bielefeld.de/cgi-bin/login.cgi">https://meltdb.cebitec.uni-bielefeld.de/cgi-bin/login.cgi</a>	Kessler et al. (2013)
metaMS	GC-MS	R Package	<a href="http://bioconductor.org/packages/release/bioc/html/metaMS.html">http://bioconductor.org/packages/release/bioc/html/metaMS.html</a>	Wehrens et al. (2014)
MSeasy	GC-MS	R Package	<a href="https://cran.r-project.org/web/packages/MSeasy/index.html">https://cran.r-project.org/web/packages/MSeasy/index.html</a>	Nicolè et al. (2012)
SpectConnect	GC-MS	Web App	<a href="http://spectconnect.mit.edu">http://spectconnect.mit.edu</a>	Styczynski et al. (2007)
rNMR	NMR	R Package	<a href="http://rnmr.nmrfam.wisc.edu">http://rnmr.nmrfam.wisc.edu</a>	Lewis et al. (2009)

*CLI* command line interface, *GUI* graphical user interface

# Data processing softwares

*Electrophoresis* 2019, 40, 227–246

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Biswapriya B. Misra<sup>1</sup>   
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Received October 11, 2018

Revised November 9, 2018

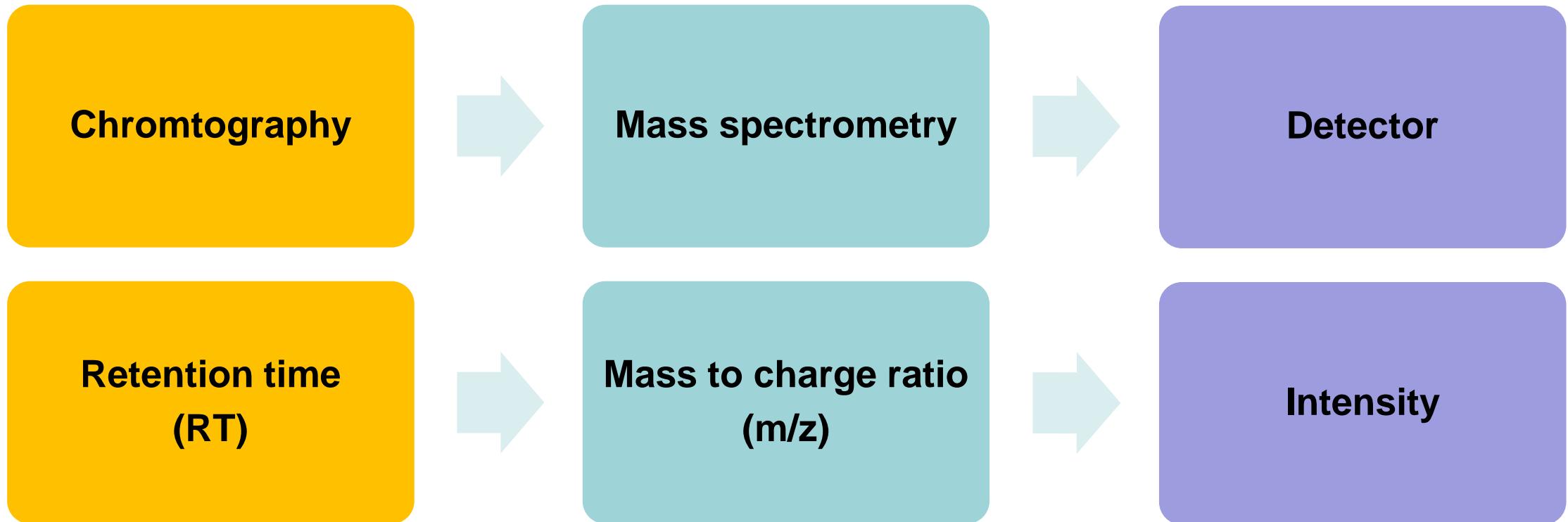
Accepted November 9, 2018

## Review

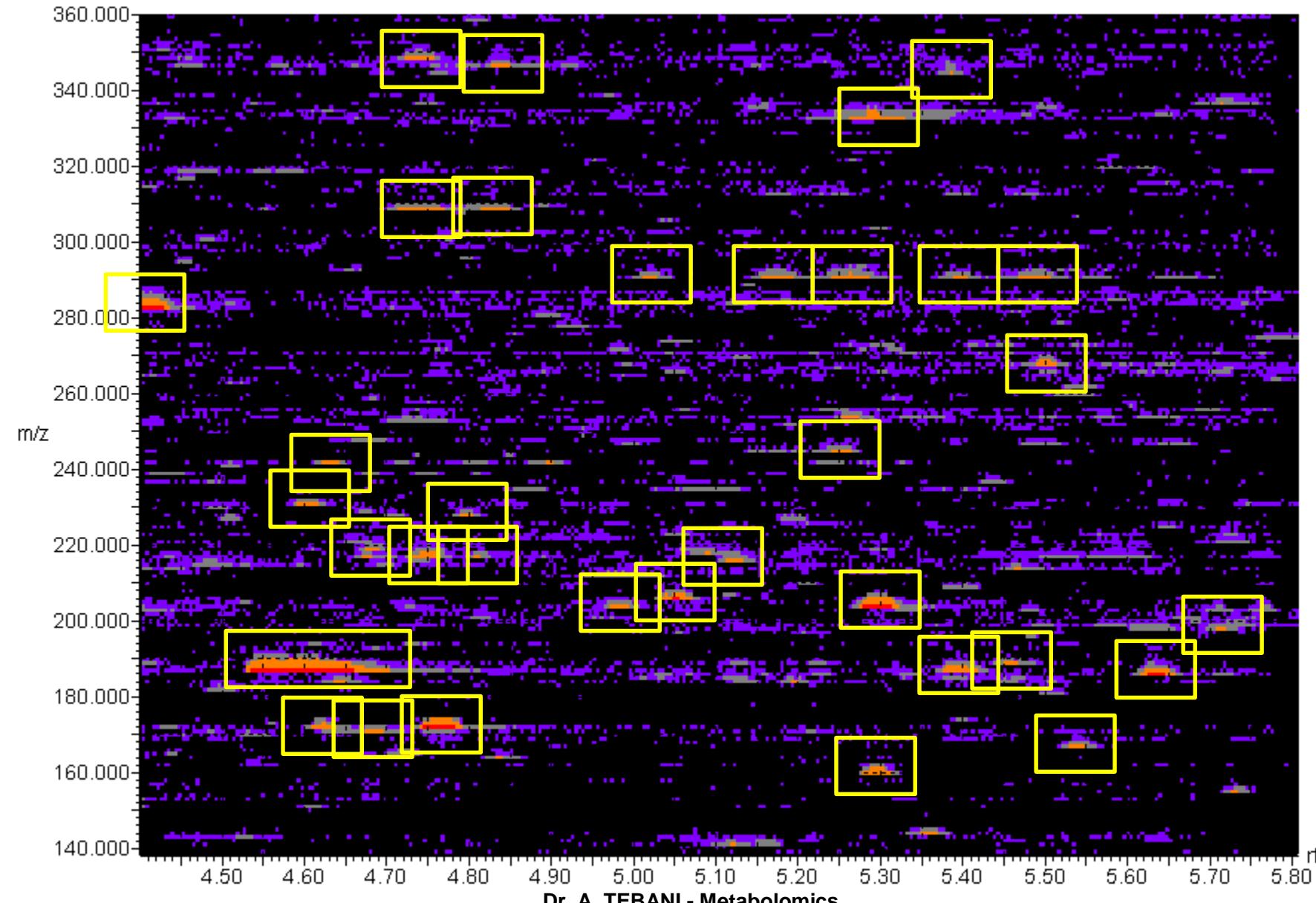
### Tools and resources for metabolomics research community: A 2017–2018 update

The scale at which MS- and NMR-based platforms generate metabolomics datasets for both research, core, and clinical facilities to address challenges in the various sciences—ranging from biomedical to agricultural—is underappreciated. Thus, metabolomics efforts spanning microbe, environment, plant, animal, and human systems have led to continual and concomitant growth of *in silico* resources for analysis and interpretation of these datasets. These software tools, resources, and databases drive the field forward to help keep pace with the amount of data being generated and the sophisticated and diverse analytical platforms that are being used to generate these metabolomics datasets. To address challenges in data preprocessing, metabolite annotation, statistical interrogation, visualization, interpretation, and integration, the metabolomics and informatics research community comes up with hundreds of tools every year. The purpose of the present review is to provide a brief and useful summary of more than 95 metabolomics tools, software, and databases that were either developed or significantly improved during 2017–2018. We hope to see this review help readers, developers, and researchers to obtain informed access to these thorough lists of resources for further improvisation, implementation, and application in due course of time.

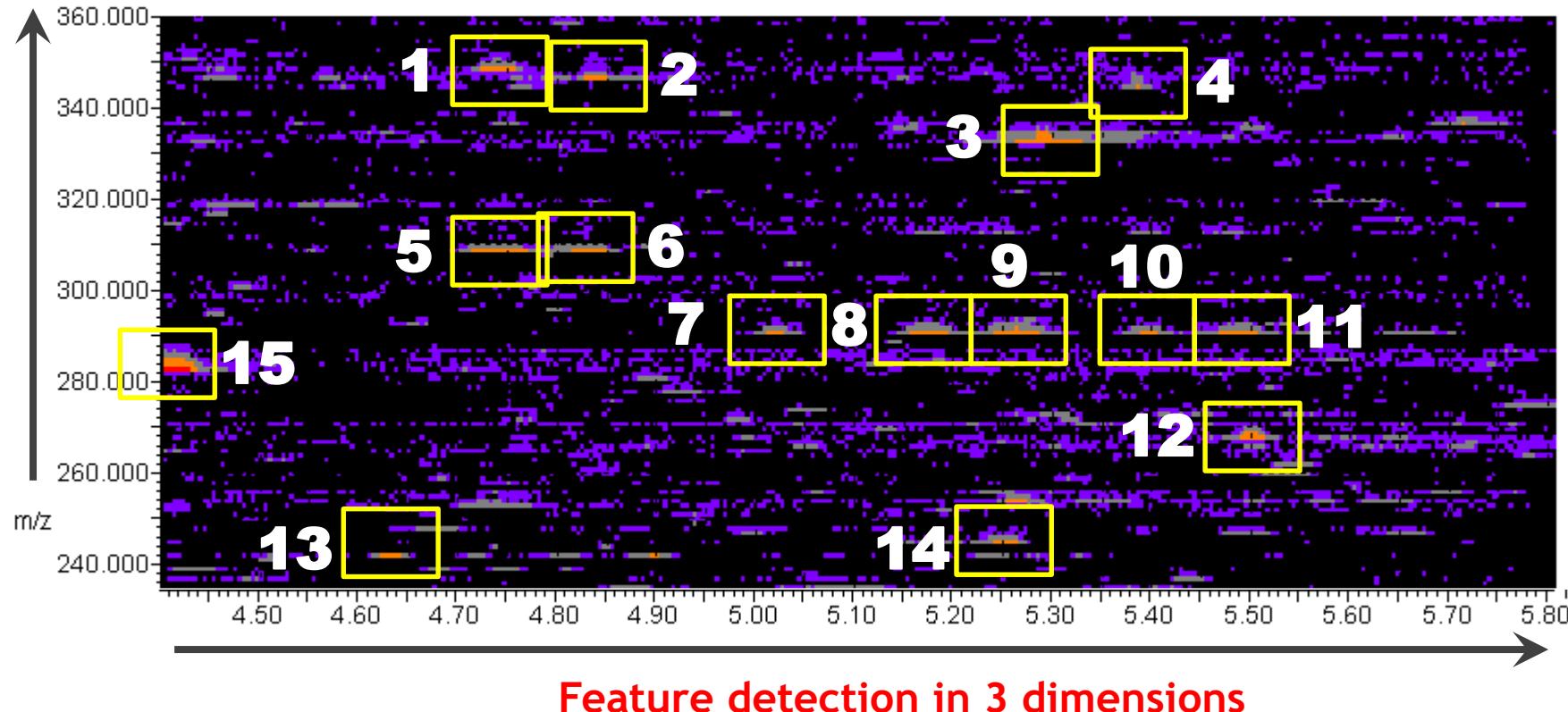
# Output data structure



# Feature detection in 3-dimensional data (LC-MS)



# Feature detection in 3-dimensional data (LC-MS)



Concatenated to  
single term  
representing each  
feature

**Feature = RT\_mz**

1. Mass (m/z)
2. Chromatographic retention time (RT)
3. Intensity (“counts”)

# Output data structure

Feature Identifier

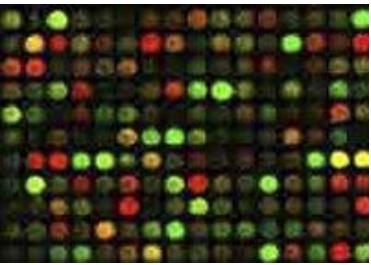


Sample identifier

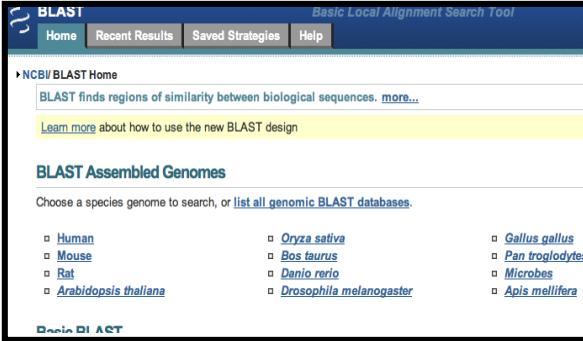
Feature.name Label	mz	rt	QC.nor.rsd	R2	X180501_1805235_T_NEG T	X180501_1805236_F_NEG F	X180501_1805237_T_NEG T	X180501_1805238_F_NEG F	X180501_1805239_QC_NEG QC
	mz	rt	QC.nor.rsd	R3					
13.22_855.6905m/z	855,6905	13,22	2,78	0,90	1,595756608	1,127168446	0,90987049	1,502157626	1,5242749
12.78_1175.8173m/z	1175,8173	12,78	3,75	0,89	1,012068895	1,924264871	0,941052344	1,24659619	0,700269515
13.11_832.6373m/z	832,6373	13,11	2,58	0,90	1,895778753	0,878647757	0,668815029	3,531989245	1,739529775
13.84_857.7065m/z	857,7065	13,84	2,50	0,91	1,446563144	1,459917821	0,95356938	1,72439594	1,702321205
5.97_339.2102m/z	339,2102	5,97	2,61	0,96	0,462711866	0,319314738	0,528005243	0,457088002	0,340008931
13.20_829.6733m/z	829,6733	13,20	3,62	0,86	1,694567338	1,470567656	0,934753971	1,746620672	1,462169722
12.41_899.6548n	898,6100	12,41	5,13	0,93	1,651790124	0,927875874	0,937857223	1,247611732	1,462013507
9.87_745.5764m/z	745,5764	9,87	4,30	0,93	1,218828015	1,276430502	1,070145633	1,561077878	1,428686119
11.62_828.6054m/z	828,6054	11,62	2,83	0,94	1,503987724	0,941322885	0,941501372	1,483197405	1,458004305
11.49_807.5658n	852,6062	11,49	5,46	0,91	1,031303648	0,798662614	0,683650659	2,074864895	1,016034483
10.74_747.5927m/z	747,5927	10,74	1,41	0,91	1,426032378	0,958717603	1,165625673	0,822211366	1,333198095
10.54_824.5737m/z	824,5737	10,54	2,97	0,92	1,02955304	0,68418959	0,920950584	1,168736087	2,084530777
12.31_854.6227m/z	854,6227	12,31	1,90	0,91	0,92279377	0,893270646	0,801607214	2,58524926	0,979421296
13.84_925.6952m/z	925,6952	13,84	2,97	0,91	1,541006208	1,549950309	0,964287935	1,703201186	1,722283028
11.21_826.5906m/z	826,5906	11,21	2,25	0,91	0,883019997	1,045390523	0,81326039	1,939078305	1,314143816
11.30_802.5890m/z	802,5890	11,30	1,88	0,93	1,3456936	1,026355676	1,006204223	0,940871989	1,513826994
11.30_870.5778m/z	870,5778	11,30	3,85	0,91	1,311794457	0,989558753	1,212741308	0,797155748	1,214987377
13.02_803.6564m/z	803,6564	13,02	5,33	0,91	1,430486792	1,911681286	1,338784094	1,130707907	1,540004947
13.12_900.6262m/z	900,6262	13,12	5,84	0,86	2,015477313	0,820250179	0,604150649	3,738009603	1,8536893
7.16_303.2418m/z	303,2418	7,16	4,66	0,91	1,379045696	0,893287758	1,514480954	1,358690786	2,126265644
11.64_934.6489n	915,6310	11,64	5,34	0,75	2,603452294	1,936584452	2,406736471	1,606282757	1,385685385
9.47_915.6317m/z	915,6317	9,47	7,79	0,88	2,289866624	1,380027238	1,588154586	0,952194138	1,786313204
13.86_993.6850m/z	993,6850	13,86	5,55	0,77	1,390865724	1,187803446	0,929504794	1,249722929	1,466552135
12.06_804.6054m/z	804,6055	12,06	2,78	0,92	1,709644369	0,933713733	0,696277652	1,561432257	1,782679331
6.03_566.3656m/z	566,3656	6,03	2,90	0,86	1,165464794	0,949195488	0,790945442	1,089173428	1,673271311
12.72_856.6375m/z	856,6375	12,72	3,32	0,92	1,420020977	0,930475089	0,884645361	2,959682093	0,998135122

Feature intensity

# The biggest challenge is Annotation

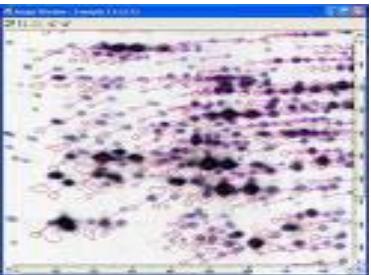


DNA/RNA

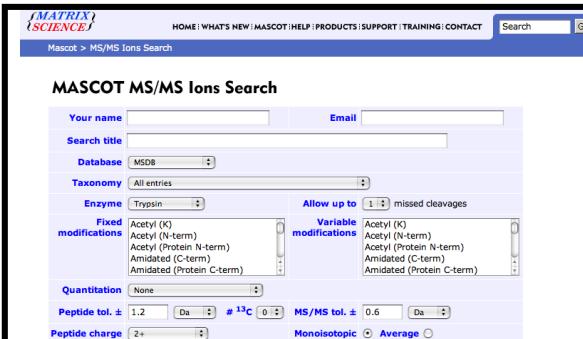


BLAST  
Home Recent Results Saved Strategies Help  
NCBI/BLAST Home  
BLAST finds regions of similarity between biological sequences. [more...](#)  
[Learn more](#) about how to use the new BLAST design  
BLAST Assembled Genomes  
Choose a species genome to search, or [list all genomic BLAST databases](#).  
 Human       *Oryza sativa*       *Gallus gallus*  
 Mouse       *Bos taurus*       *Pan troglodytes*  
 Rat       *Danio rerio*       *Microbes*  
 *Arabidopsis thaliana*       *Drosophila melanogaster*       *Apis mellifera*  
Basic DI ACT

Gene IDs +  
Transcript  
Abundance

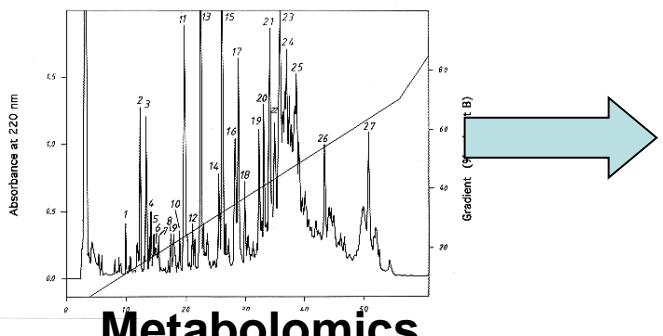


Proteomics

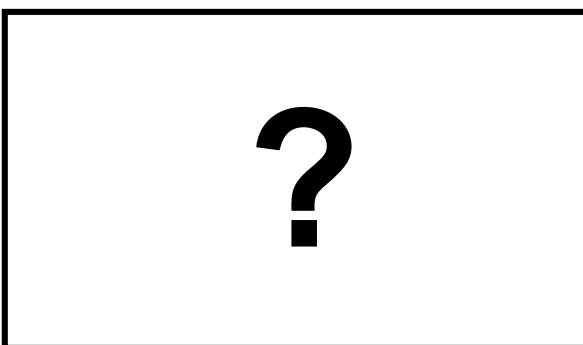


MATRIX / SCIENCE HOME: WHAT'S NEW | MASCOT | HELP | PRODUCTS | SUPPORT | TRAINING | CONTACT Search Go  
Mascot > MS/MS Ions Search  
MASCOT MS/MS Ions Search  
Your name \_\_\_\_\_ Email \_\_\_\_\_  
Search title \_\_\_\_\_  
Database: **MSDB** Taxonomy: All entries  
Enzyme: Trypsin Allow up to 1 missed cleavages  
Fixed modifications: Acetyl (K) Acetyl (N-term)  
Variable modifications: Acetyl (K) Acetyl (N-term)  
Acetyl (Protein N-term)  
Amidated (C-term)  
Amidated (Protein C-term)  
Quantitation: None  
Peptide tol.  $\pm$  1.2 Da #  $^{13}\text{C}$  0 MS/MS tol.  $\pm$  0.6 Da  
Peptide charge: 2+ Monoisotopic Average

Protein IDs +  
Concentrations



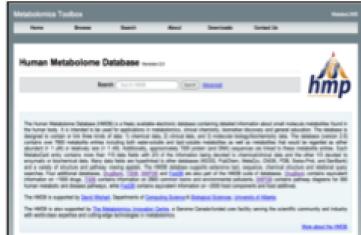
Metabolomics



Metabolite IDs +  
Concentrations

# Metabolome Databases

<http://metabolomicssociety.org/resources/metabolomics-databases>



[www.hmdb.ca](http://www.hmdb.ca)



[www.drugbank.ca](http://www.drugbank.ca)



[www.ymdb.ca](http://www.ymdb.ca)



[www.phenol-explorer.eu](http://www.phenol-explorer.eu)



[www.ecmdb.ca](http://www.ecmdb.ca)



[www.foodb.ca](http://www.foodb.ca)



[www.cowmetdb.ca](http://www.cowmetdb.ca)



[www.t3db.ca](http://www.t3db.ca)



[www.smpdb.ca](http://www.smpdb.ca)



[www.csfmetabolome.ca](http://www.csfmetabolome.ca)



[www.serummetabolome.ca](http://www.serummetabolome.ca)



[www.urinemetabolome.ca](http://www.urinemetabolome.ca)

## Levels of Metabolite Identification in MS

### 1. Positively identified compounds

Confirmed by match to known standard

### 2. Putatively identified compounds

Match to MS + RT or MS/MS + RT

### 3. Compounds putatively identified in a compound class

### 4. Unknown compounds

## Commercial tools

**Agilent MassHunter Profinder**

**Bruker's ProfileAnalysis**

**Thermo SIEVE™**

**Waters' Progenesis QI**

**SkyLine**

## Free options

**XCMS Online**

**MZmine**

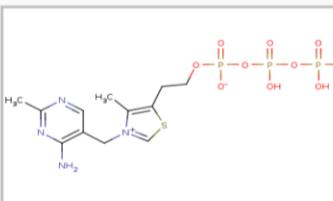
## XCMS

- The first open source tool for spectra processing
- Does peak picking, peak matching and retention time alignment
- Available as a program and a server
- Accepts multiple formats: **mzXML, mzData, .cdf (NetCDF), .d folders (Agilent; Bruker), .wiff files (AB SCIEX)**
- Metabolite identification is not the focus in XCMS (linked to Metlin)

Home\*  isoMETLIN Simple Search Advanced Search Batch Search Fragment Similarity Search Neutral Loss Search MS/MS Spectrum Match Search MRM▼  Logout [ tebanidz ]

## Simple Search

Show 10 entries Search:

	METLIN ID	Mass	$\Delta$ PPM	Name	KEGG	MS/MS	Structure
Mass	505,988						
Tolerance	30 PPM						
Charge	Neutral Positive Negative						
Adducts	M+H M+NH4 <b>M+Na</b> M+H-2H2O M+H-H2O M+K M+ACN+H M+ACN+Na M+2Na-H M+2H M+3H M+H+Na M+2H+Na M+2Na M+2Na+H M+Li M+CH3OH+H	[M+H] <sup>+</sup> <u>m/z</u> <b>505.0107</b> M 504.0035	21	Thiamin triphosphate <i>Formula: C12H19N4O10P3S</i> <i>CAS: 3475-65-8</i>	C03028	NO	
METLIN ID	Mass	$\Delta$ PPM	Name	KEGG	MS/MS	Structure	

Showing 1 to 1 of 1 entries

Previous 1 Next

Peptides Remove Peptides from Search Activate Windows

# Annotation Conversion

CTS - The Chemical Translation Service    [Simple Conversion](#)    [Batch Conversion](#)    [Services](#)

## Simple Conversion

To convert a single identifier, enter it in the box below, select source and target types, and hit the Convert button.

Chemical Name    Enter ID for conversion    Convert →    InChIKey

Issues? Let us know on [BitBucket](#).

Finley and King Labs, Harvard Medical School  
FLUKA  
ForeChem  
Fragmenta  
Georganics  
GlaxoSmithKline (GSK)  
GLIDA, GPCR-Ligand Database  
GNF / Scripps Winzeler lab  
Golm Metabolome Database (GMD), Max Planck Institute of Molecular Plant Physiology  
Hangzhou APIChem Technology  
Hangzhou Trylead Chemical Technology  
HDH Pharma  
**Human Metabolome Database**  
HUMGENEX  
IBCH RAS  
IBM  
ICCB-Longwood/NSRB Screening Facility, Harvard Medical School  
Immunology Lab, Department of Biotechnology, Calicut University  
InChI Code

# **GETTING DATA READY**

# Data Analysis

## Input

A matrix containing numerical values

Concentrations (Targeted)

Peak intensities (Untargeted)

Meta-data

Class labels, experimental factors

## Output

Discriminant features

Clustering patterns

Biological Inference

Biomarkers

Predictive models

# Data Analysis

Variables										
Samples	Metabolites									
6	0	4	8	0	2	2	7	2	2	2
0	1	8	0	8	0	0	3	4	9	
4	4	0	4	0	1	4	0	1	8	
8	0	2	8	2	4	8	4	0	0	
0	4	6	0	5	2	0	8	0	2	
2	8	9	5	5	5	2	0	4	6	
4	9	5	6	5	4	5	2	8	9	
1	8	0	8	5	0	1	8	0	8	
4	0	4	0	0	4	2	0	4	0	
0	2	8	2	5	8	0	2	8	2	

## Variables

Metabolites

Samples

Information  
+  
Noise

**Remove as much as possible noise**

**Extract as much as possible information**

# Data Analysis / Data cleaning



## Data cleaning

- Missing values imputation
- Filtering (Min, IQR, RSD, CV, R<sup>2</sup> ...)
- Normalization

# Data Analysis / Data cleaning

Cut-off



Feature.name	mz	rt	QC.nor.rsd	R2	X180501_1805235_T_NEG	X180501_1805236_F_NEG	X180501_1805237_T_NEG	X180501_1805238_F_NEG	X180501_1805239_T_NEG
Label	mz	rt	QC.nor.rsd	R3	T	F	T	F	T
13.22_855.6905m/z	855,6905	13,22	2,78	0,90	1,595756608	1,127168446	0,90987049	1,502157626	1,5242749
12.78_1175.8173m/z	1175,8173	12,78	3,75	0,89	1,012068895	1,924264871	0,941052344	1,24659619	0,700269515
13.11_832.6373m/z	832,6373	13,11	2,58	0,90	1,895778753	0,878647757	0,668815029	3,531989245	1,739529775
13.84_857.7065m/z	857,7065	13,84	2,50	0,91	1,446563144	1,459917821	0,95356938	1,72439594	1,702321205
5.97_339.2102m/z	339,2102	5,97	2,61	0,96	0,462711866	0,319314738	0,528005243	0,457088002	0,340008931
13.20_829.6733m/z	829,6733	13,20	3,62	0,86	1,694567338	1,470567656	0,934753971	1,746620672	1,462169722
12.41_899.6548n	898,6100	12,41	5,13	0,93	1,651790124	0,927875874	0,937857223	1,247611732	1,462013507
9.87_745.5764m/z	745,5764	9,87	4,30	0,93	1,218828015	1,276430502	1,070145633	1,561077878	1,428686119
11.62_828.6054m/z	828,6054	11,62	2,83	0,94	1,503987724	0,941322885	0,941501372	1,483197405	1,458004305
11.49_807.5658n	852,6062	11,49	5,46	0,91	1,031303648	0,798662614	0,683650659	2,074864895	1,016034483
10.74_747.5927m/z	747,5927	10,74	1,41	0,91	1,426032378	0,958717603	1,165625673	0,822211366	1,333198095
10.54_824.5737m/z	824,5737	10,54	2,97	0,92	1,02955304	0,68418959	0,920950584	1,168736087	2,084530777
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11.30_802.5890m/z	802,5890	11,30	1,88	0,93	1,3456936	1,026355676	1,006204223	0,940871989	1,513826994
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7.16_303.2418m/z	303,2418	7,16	4,66	0,91	1,379045696	0,893287758	1,514480954	1,358690786	2,126265644
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13.86_993.6850m/z	993,6850	13,86	5,55	0,77	1,390865724	1,187803446	0,929504794	1,249722929	1,466552135
12.06_804.6054m/z	804,6055	12,06	2,78	0,92	1,709644369	0,933713733	0,696277652	1,561432257	1,782679331
6.03_566.3656m/z	566,3656	6,03	2,90	0,86	1,165464794	0,949195488	0,790945442	1,089173428	1,673271311
12.72_856.6375m/z	856,6375	12,72	3,32	0,92	1,42020977	0,930475089	0,884645361	2,959682093	0,998135122

# Data Analysis / Normalization

## Sample normalization (row-wise)

To remove systematic variation between experimental conditions unrelated to the biological differences (i.e. dilutions, mass)

Total signal, sum of signals

Reference compound: Internal standards, endogenous metabolites

Reference sample: QC's, Controls...

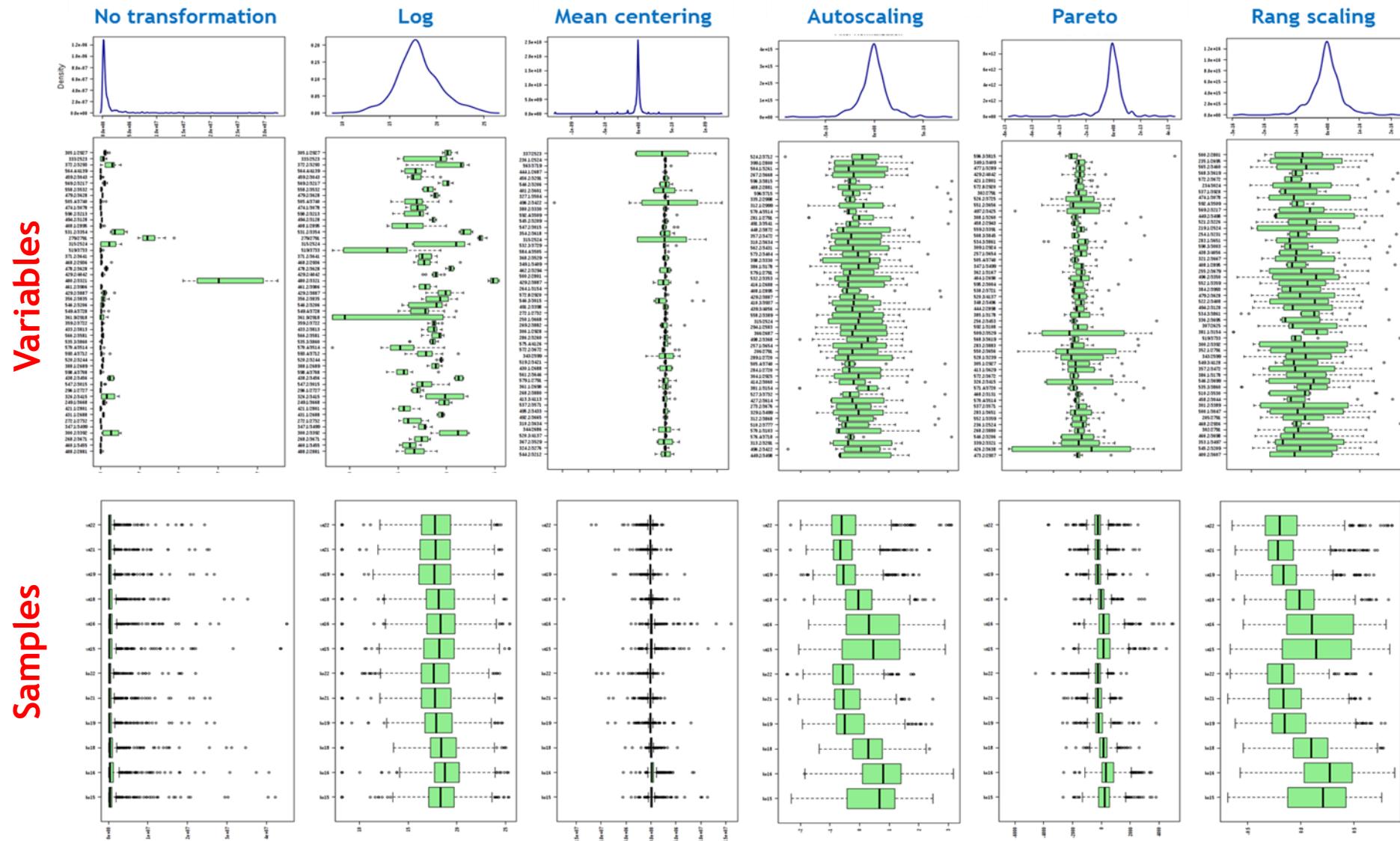
## Feature normalization (column-wise)

To bring variances of all features close to equal

Log transformation

Scaling

# Data Analysis / Normalization



# Data Analysis / Normalization

W162–W170 *Nucleic Acids Research*, 2017, Vol. 45, Web Server issue  
doi: 10.1093/nar/gkx449

Published online 19 May 2017

## NOREVA: normalization and evaluation of MS-based metabolomics data

Bo Li<sup>1,†</sup>, Jing Tang<sup>1,†</sup>, Qingxia Yang<sup>1,2,†</sup>, Shuang Li<sup>1</sup>, Xuejiao Cui<sup>1</sup>, Yinghong Li<sup>1</sup>, Yuzong Chen<sup>3</sup>, Weiwei Xue<sup>1</sup>, Xiaofeng Li<sup>1</sup> and Feng Zhu<sup>1,2,\*</sup>

Preprints (www.preprints.org) | NOT PEER-REVIEWED | Posted: 3 July 2018

[doi:10.20944/preprints201807.0059.v1](https://doi.org/10.20944/preprints201807.0059.v1)

Peer-reviewed version available at *Metabolites* 2018, 8, 47; doi:10.3390/metabo8030047

Review

## Data Normalization in NMR-based Metabolomics

Helena U. Zacharias<sup>1</sup>, Michael Altenbuchinger<sup>2</sup> and Wolfram Gronwald<sup>3\*</sup>

# Multivariate Data Analysis

Tow main objectives

## **Descriptive data analysis (Unsupervised learning)**

Mining massive datasets to discover hidden  
*data structures*  
*hidden relationships*  
*patterns, trends and clusters*  
*outliers*

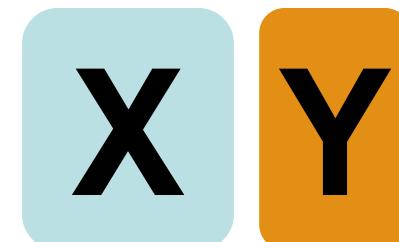
Dimension reduction



## **Predictive data analysis (Supervised learning)**

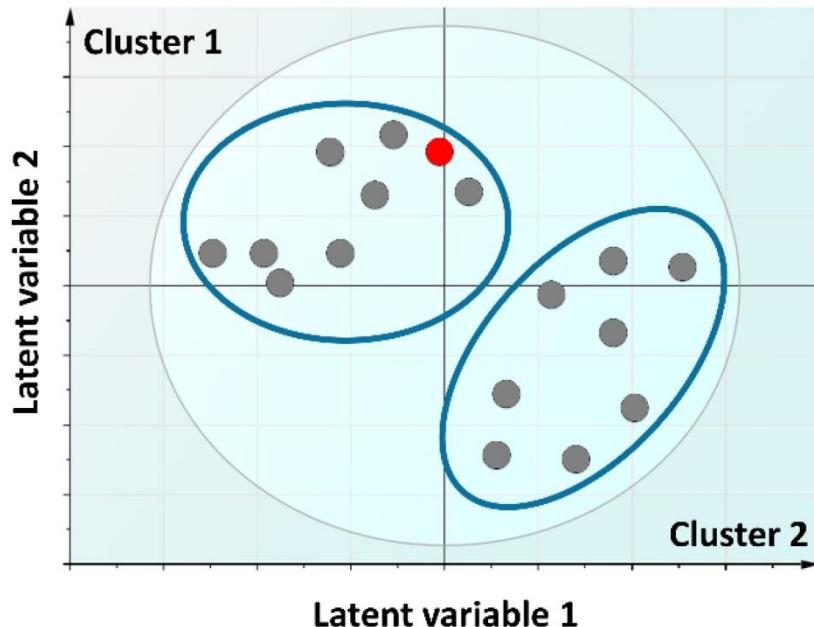
Building models for specific tasks using training datasets  
*regression*  
*classification,*  
*pattern recognition*  
*machine learning tasks*

Assessing the predictive accuracy of the models using new datasets

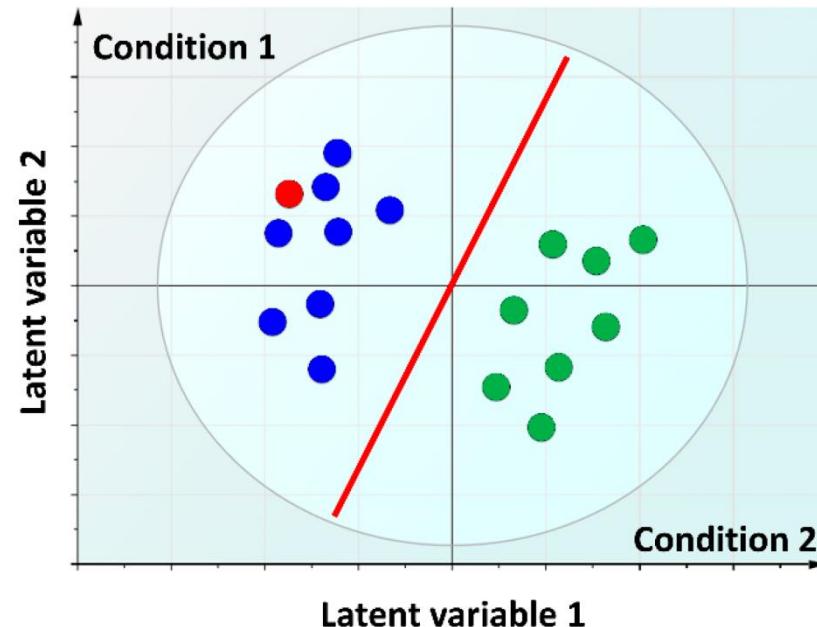


# Multivariate Data Analysis

## Unsupervised learning



## Supervised learning



Tebani A et.al. Int J Mol Sci. 2016 Sep; 17(9): 1.

# Data Analysis / Unsupervised learning

## Clustering

Organize the 1000s of variables into blocks

Variables in each block are more homogenous

Key parameter: similarities (Distance, Spearman, Pearson ...)

Similarity between samples - Similarity between clusters

Visualization using Heatmaps

- *K-means*
- *Hierarchical Methods*

## Dimension reduction

Reduce the high-dimensional data

1000s into low-dimensions (Latent variables)

- *Principal component analysis (PCA)*

**Linear Discriminant Analysis**

**Partial Least Squares**

**k-Nearest Neighbors**

**Random Forest**

**Support Vector Machines**

**Bayesian networks**

**Neural Networks**

**... name it**

<https://jawbone.com/blog/classifying-steps-machine-learning/>

## Model validation

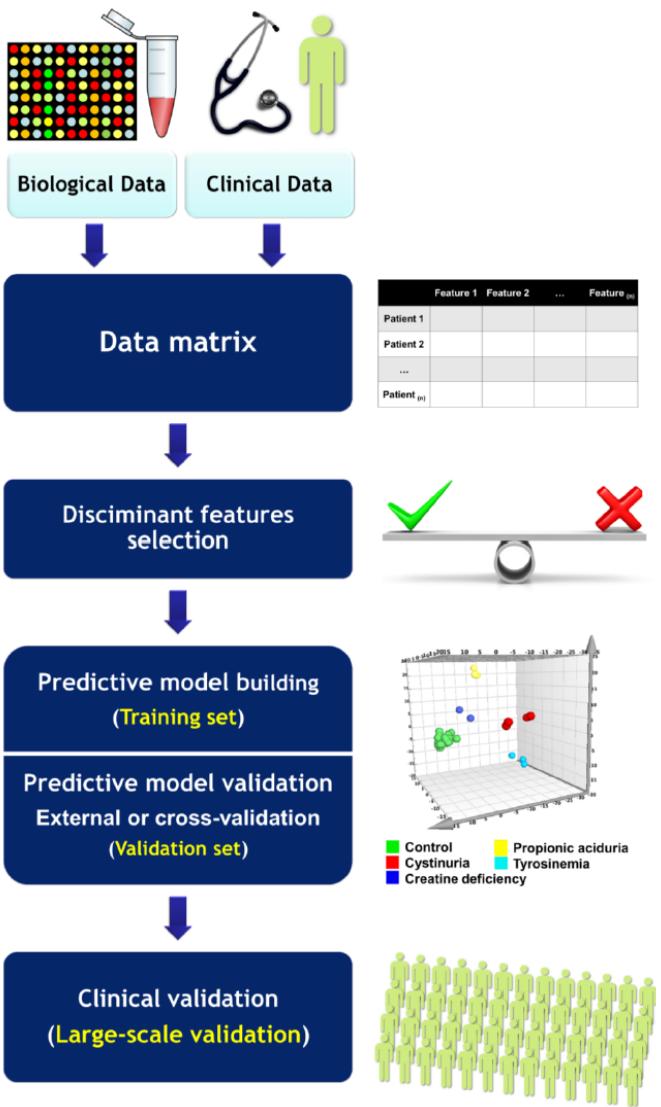
A *statistically valid* model

- Has **good fit** to the data
- Is **predictive** of new data

Validation methods

- Training set / Testing set
- Cross-validation
- Permutation test

# Data Analysis / Supervised learning



**Training dataset:** a set of examples used to build the predictive model

**Validation dataset:** a set of examples used to refine the model parameters and estimate the error.

**Test dataset:** used only to evaluate the predictive performance of the model. They are never used during the learning or testing process.

Tebani et al. Int J Mol Sci. 2016 Sep; 17(9): 1.

# Take home message

**Experimental design ++++++**

**Know your data**

**Main technologies are MS and NMR**

**Annotation is challenging in untargeted metabolomics**