

Metabolomics

From data to insights

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Stockholm - SciLifeLab - 18th February 2020

Division of Systems Biology - Department of Protein Science
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

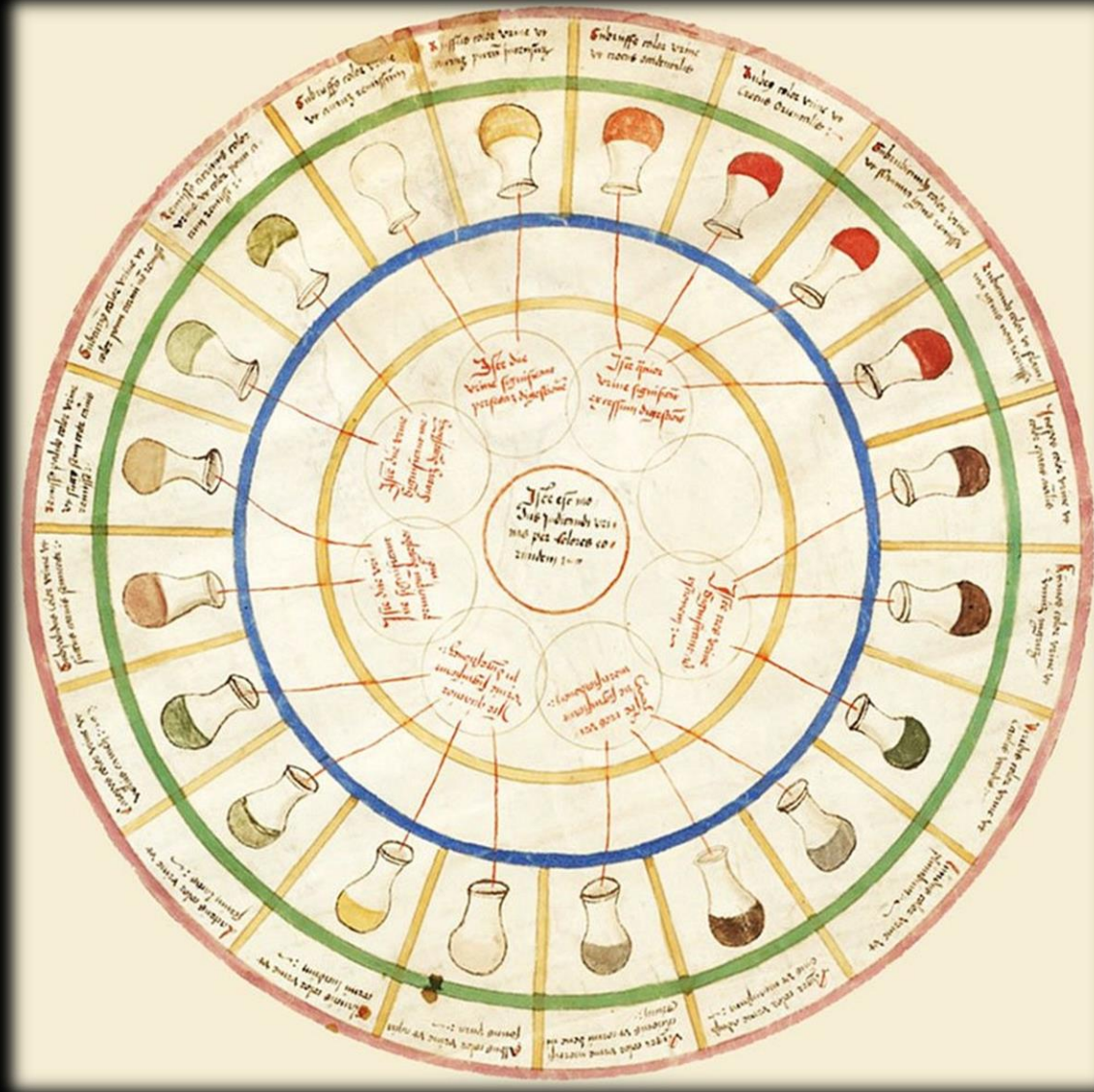
Applications and Limits

Preprocessing

Data cleaning

Key steps and best practice

The urine wheel



Time scales in biology

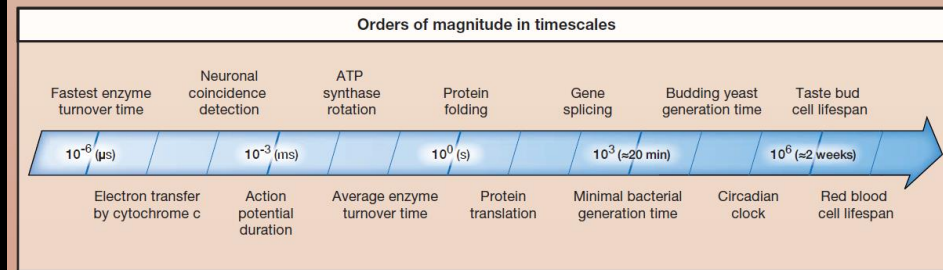
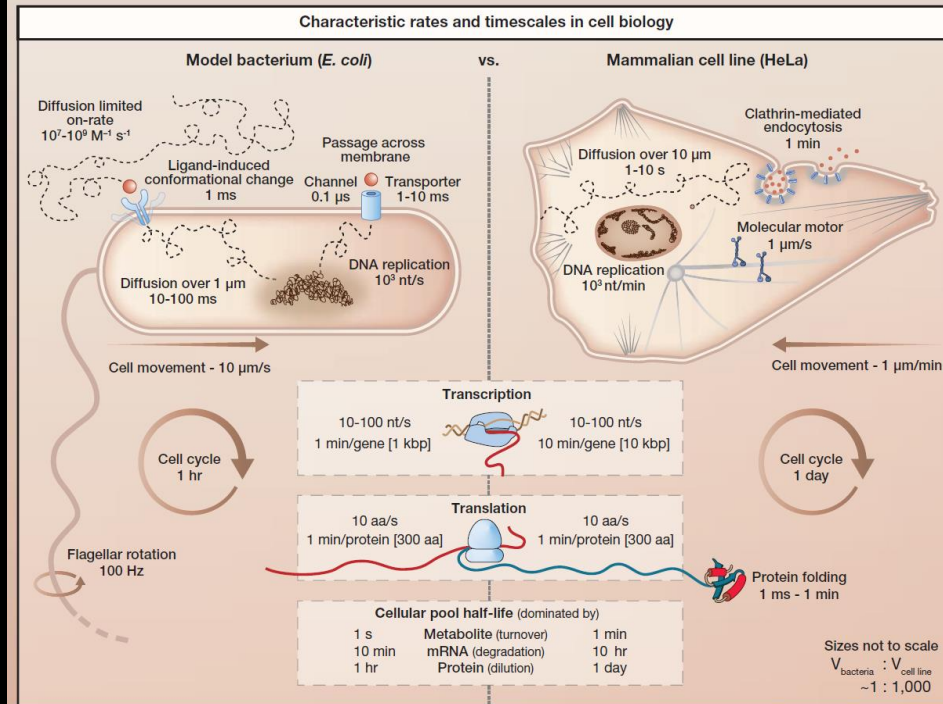
SnapShot: Timescales in Cell Biology

Cell

Maya Shamir,¹ Yinon Bar-On,¹ Rob Phillips,² and Ron Milo¹

¹Weizmann Institute of Science, Rehovot 7610001, Israel

²California 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



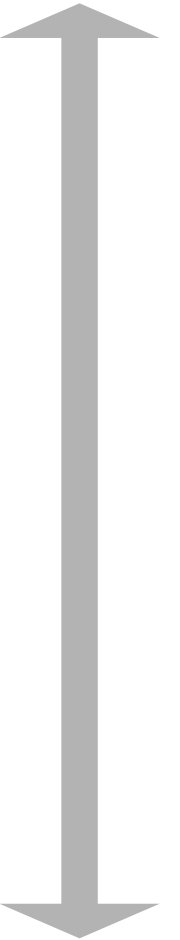
Proteins

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

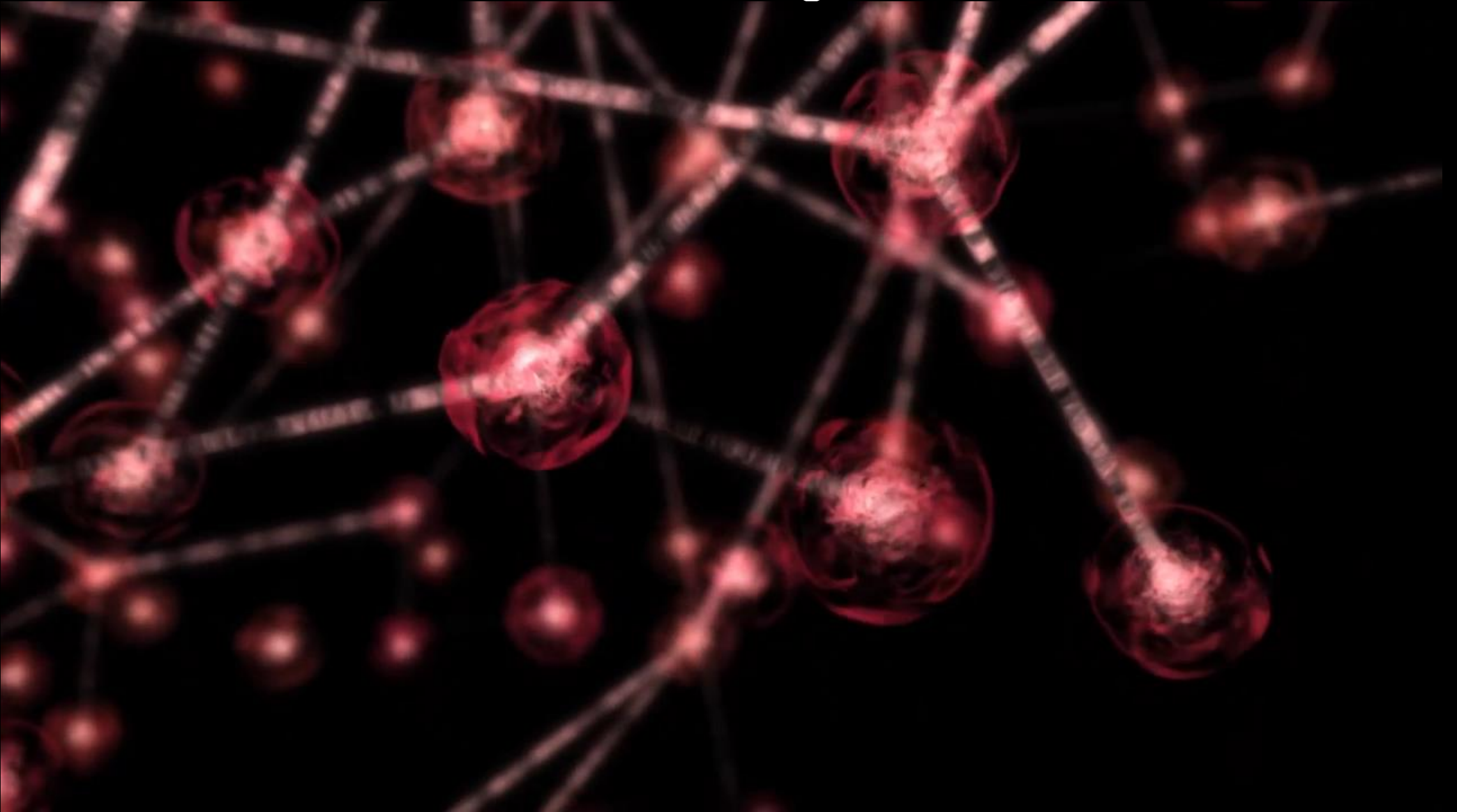
MetaboLomics / MetaboNomics:

- ➔ Metabolic Profiling / Metabotyping

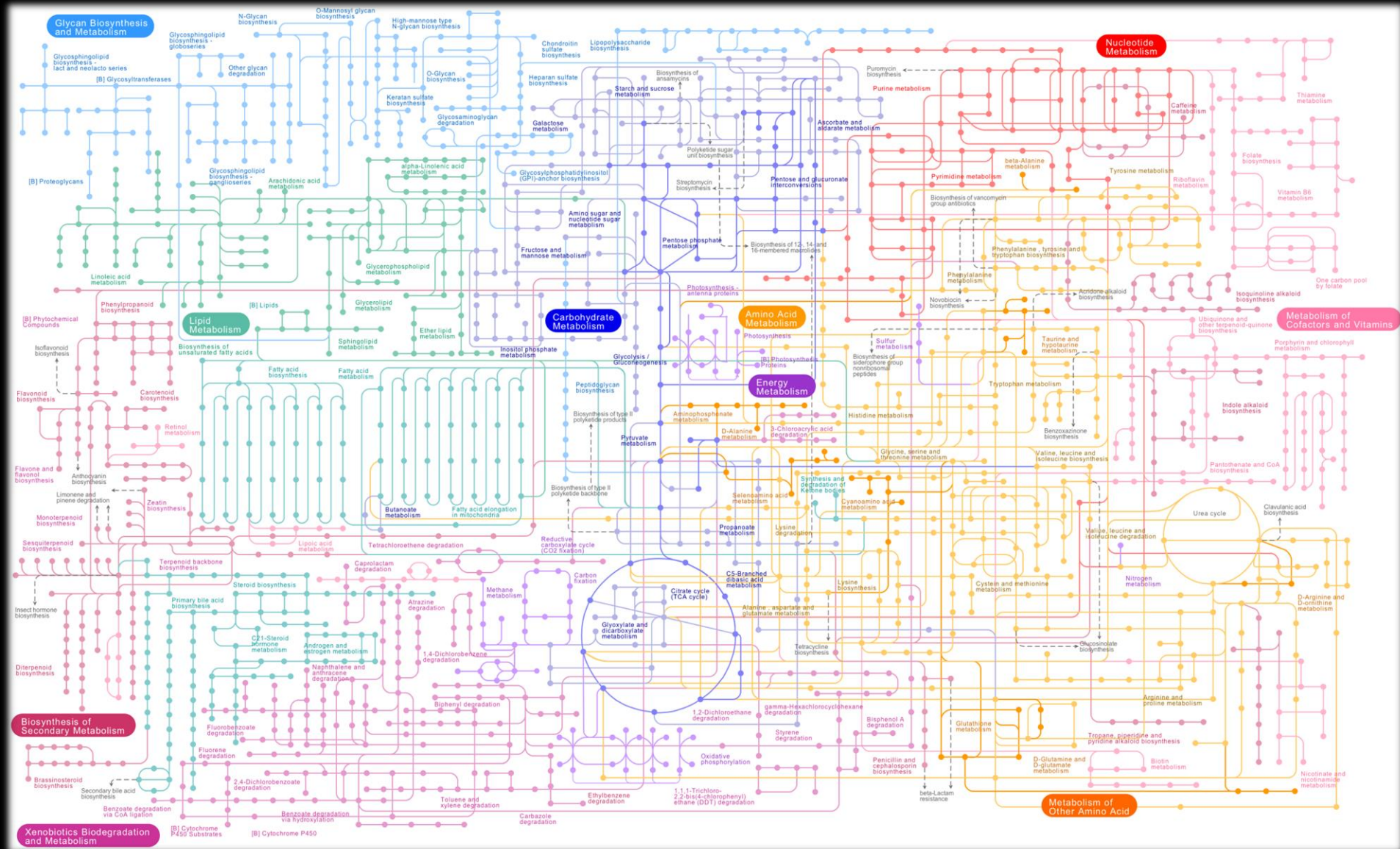
“the quantitative measurement of the metabolic *responses* of *complex systems* to a pathophysiological *stimulus* or genetic modification”.

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

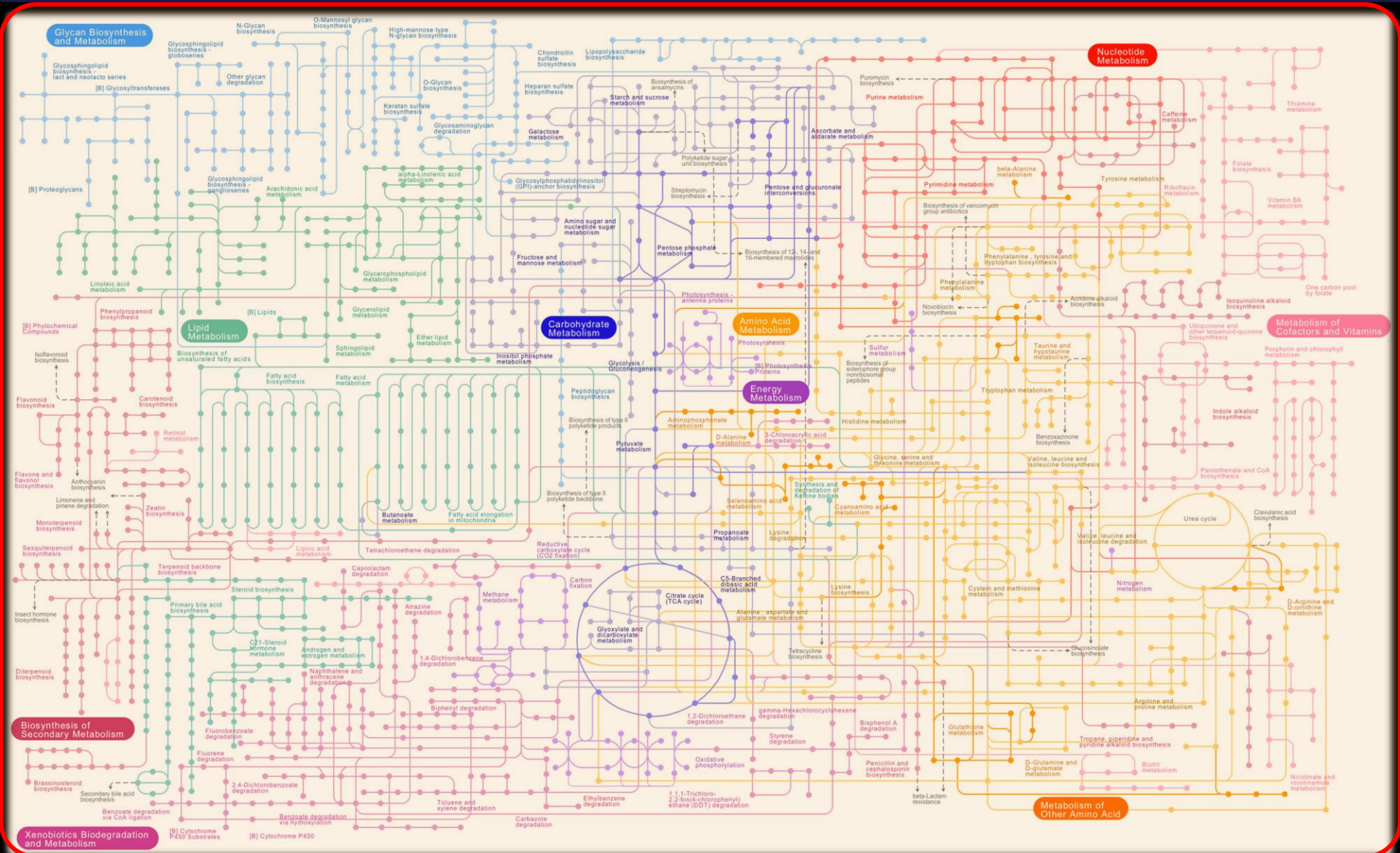
Metabolisme is an integrated network



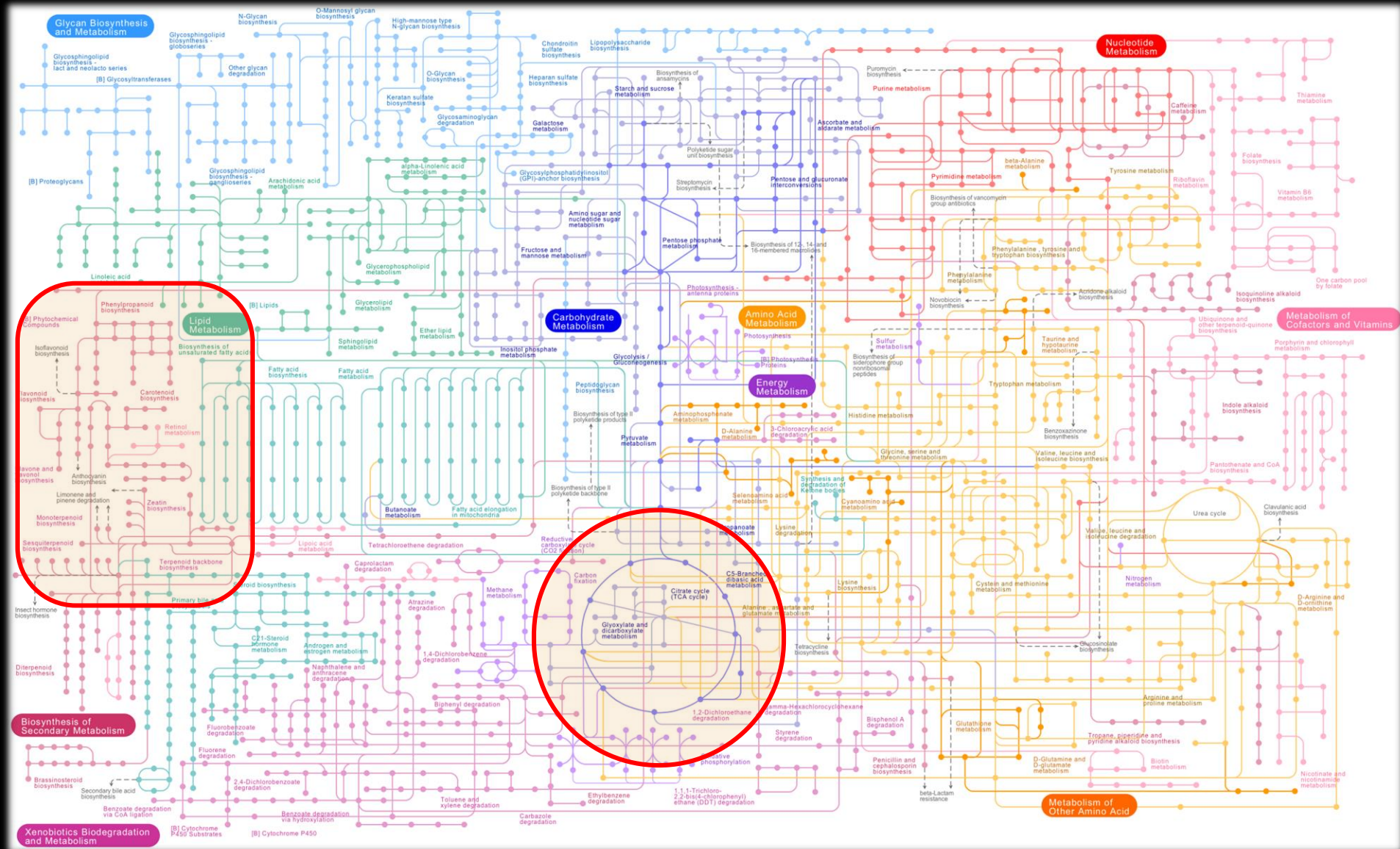
Metabolomics and Metabolism



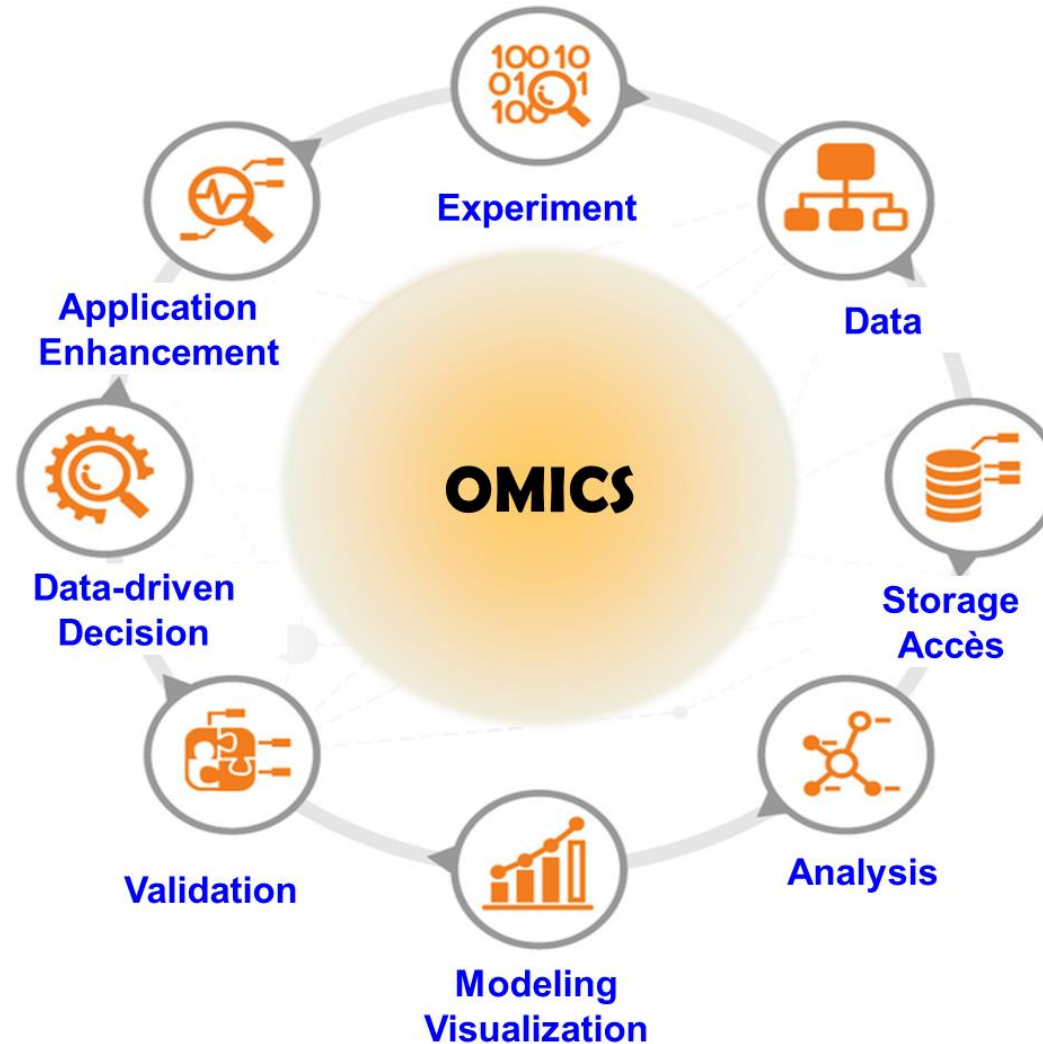
Metabolomics (Untargeted)



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.”

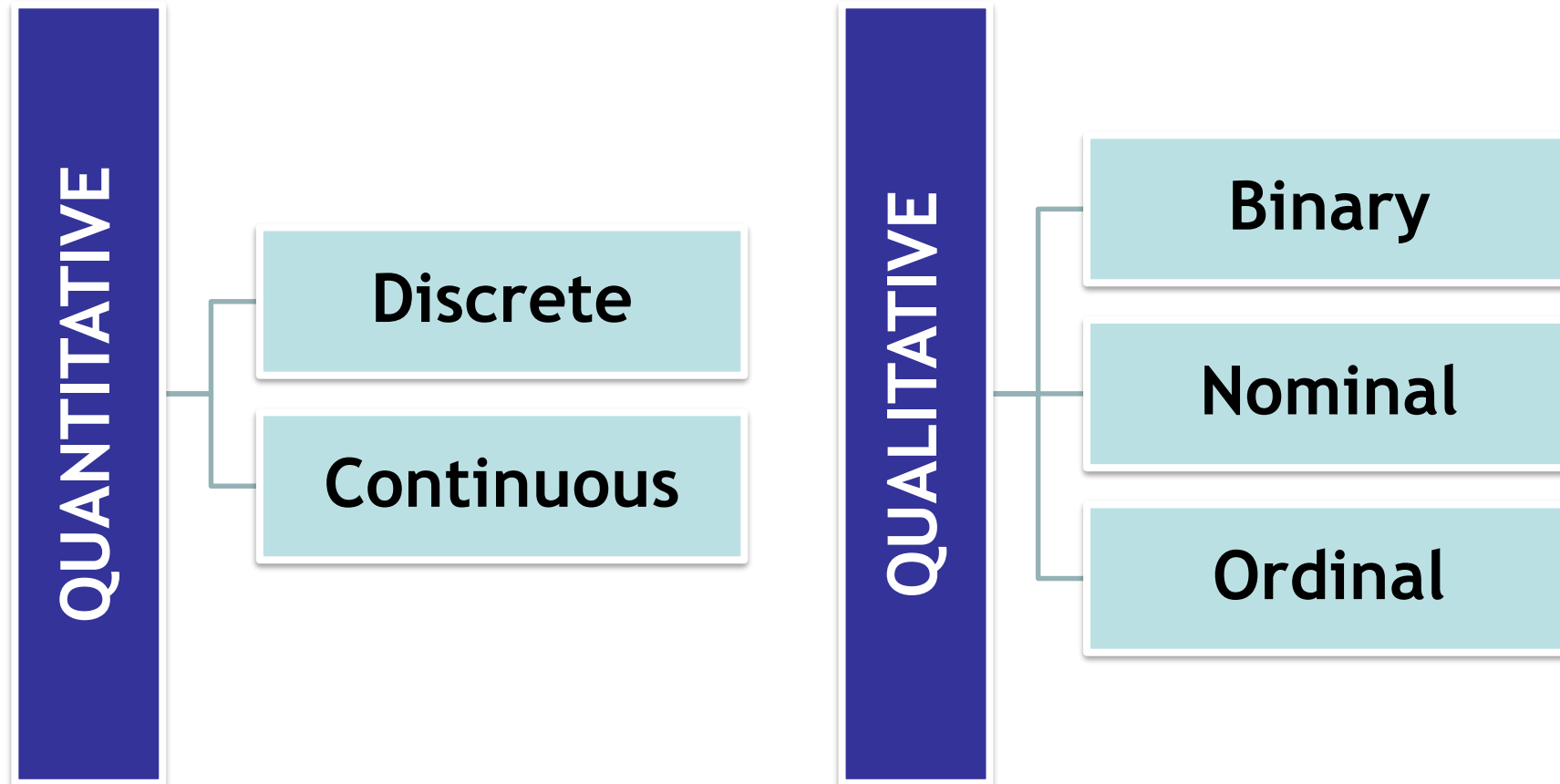
Sir Ronald Fisher (1938)



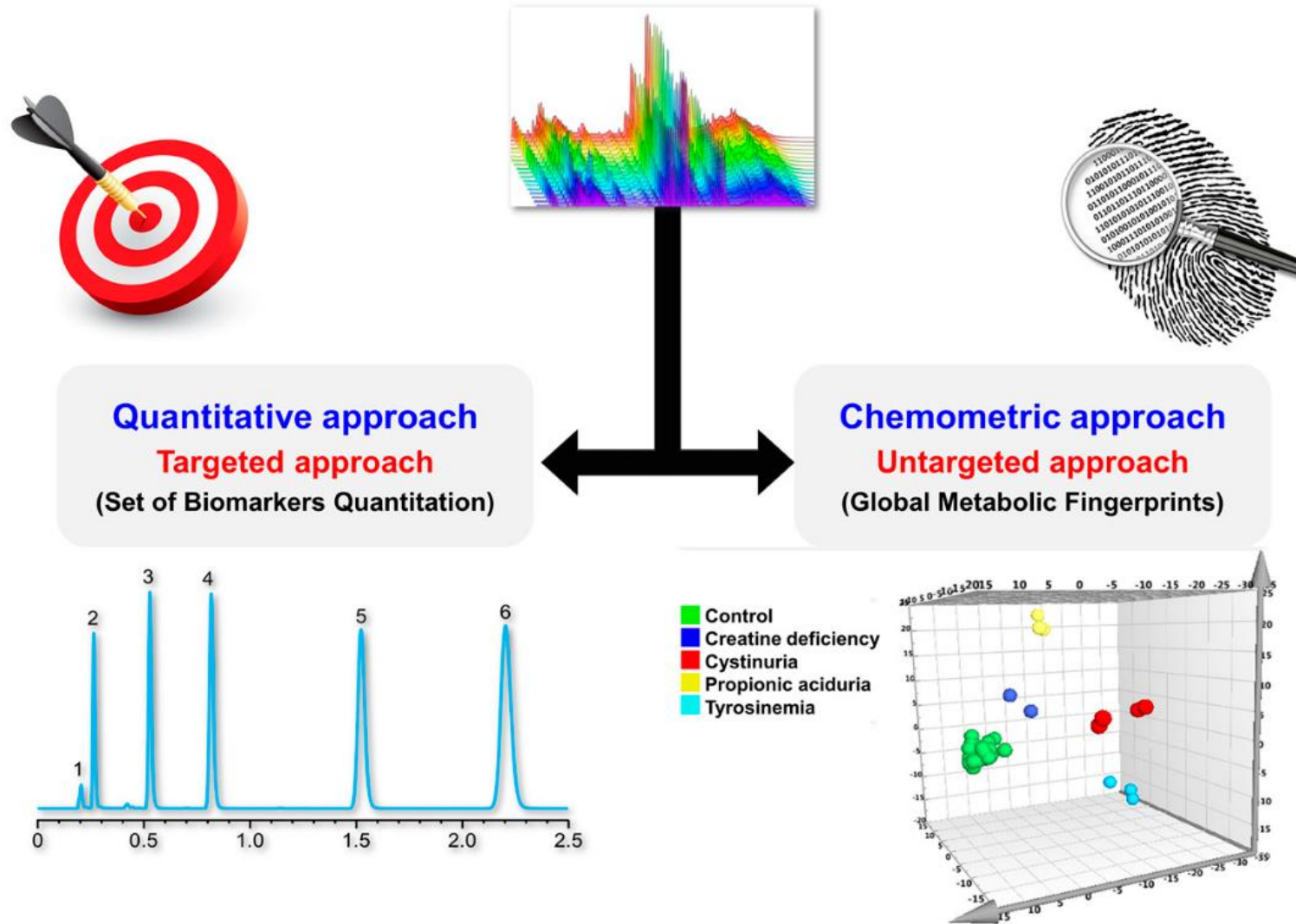
Experimental Design **may help**

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

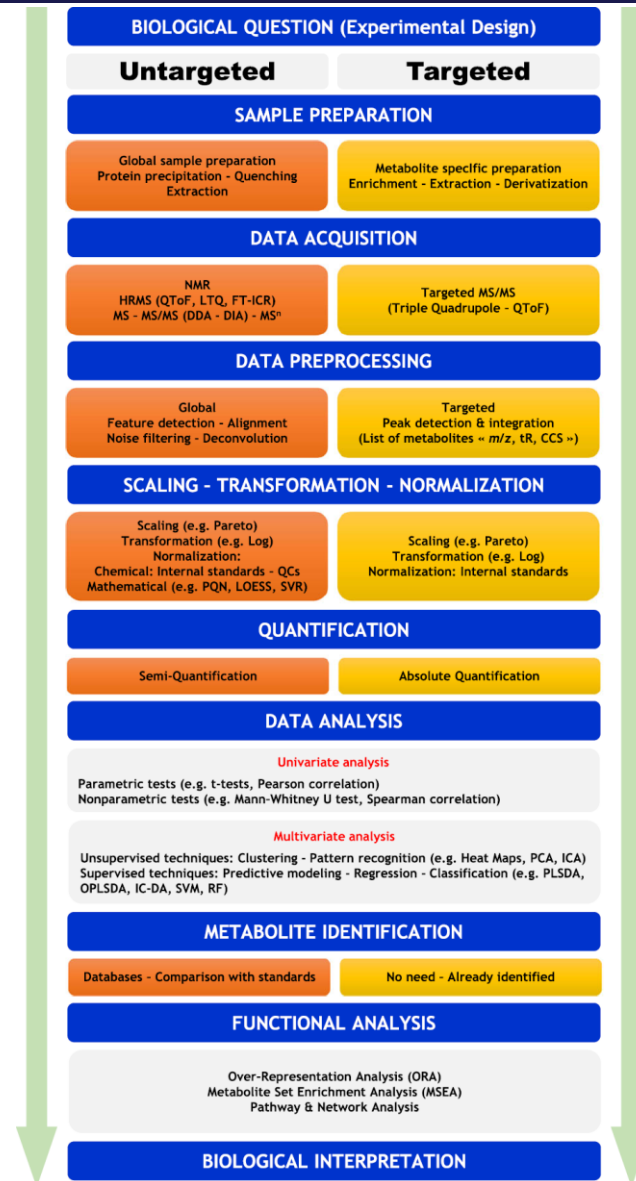


Metabolomics



Tebani et al. IJMS. 2016

Metabolomics Workflows Overview



Tebani et al. JIMD. 2017

Metabolomics workflow

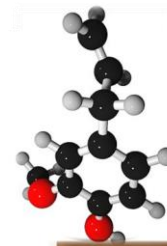
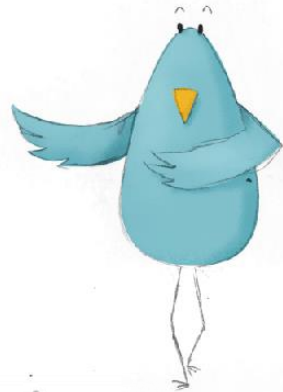
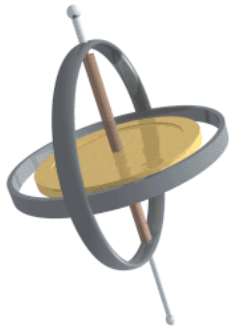
1

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 (^1H , ^{13}C , ^{31}P) with electromagnetic fields gives molecular information

Non-destructive
Cross-instrument robustness

Mass spectrometry

Tissues, biofluids and extracts

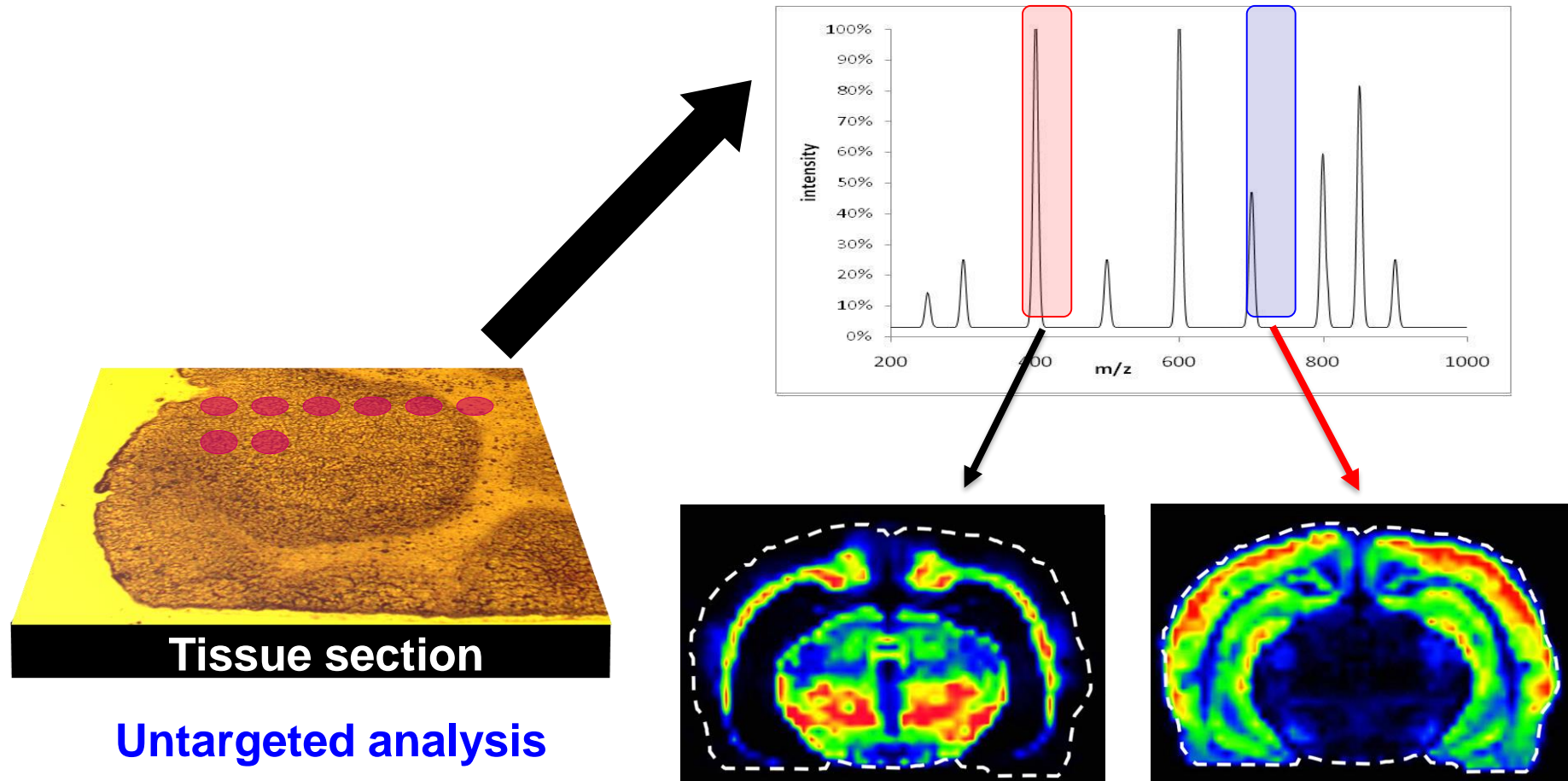
Mass to charge ratio (m/z)

Sensitivity
Higher metabolome coverage

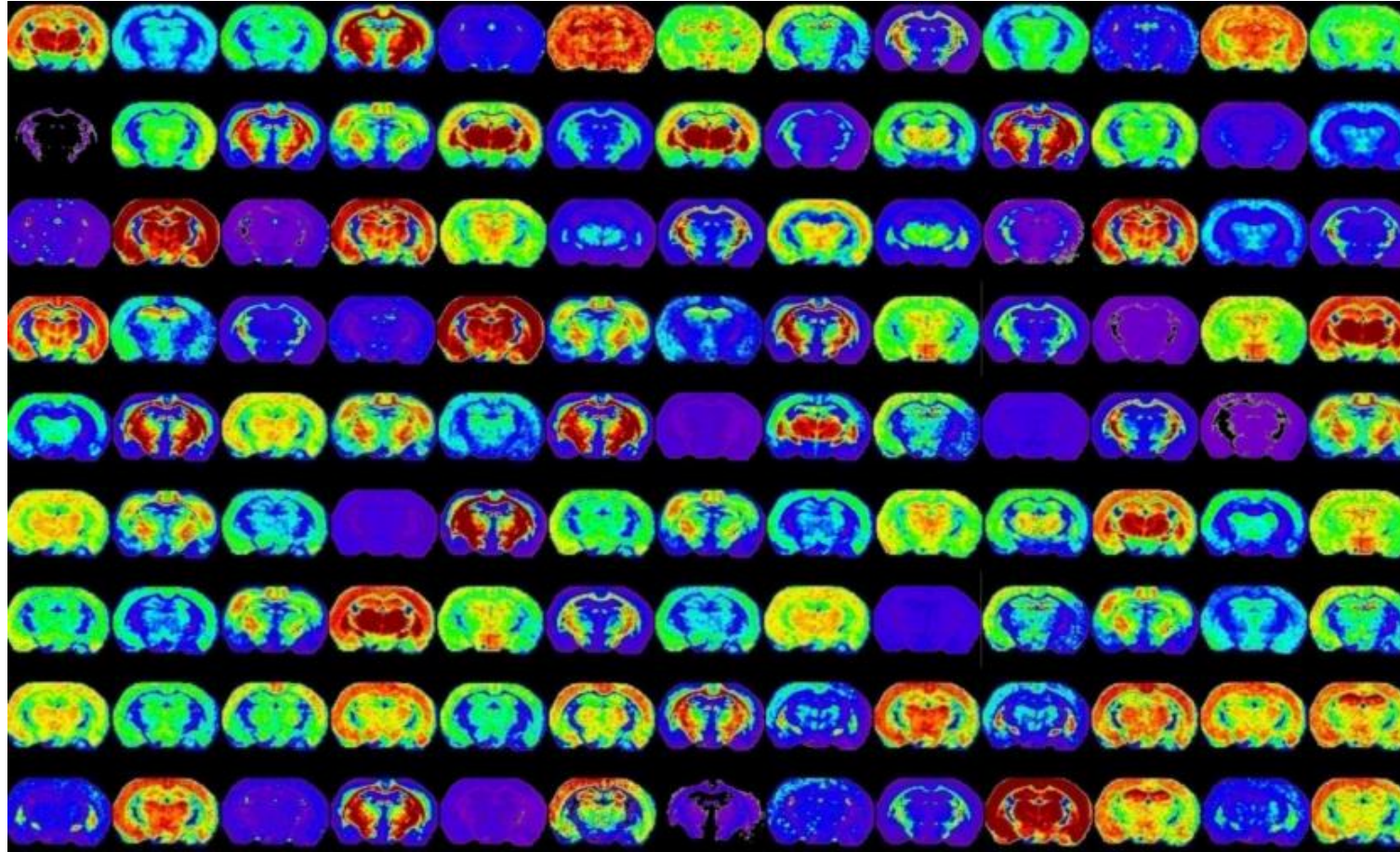
Tebani et al. JIMD. 2016

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>

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

Metabolomics paths towards Precision Medicine

```
graph TD; A[Metabolomics paths towards Precision Medicine] --> B[Populational Profiling]; A --> C[Individual Profiling]; A --> D[Drug Discovery];
```

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^{1,2,†,*}, Saeed Shoaie^{1,†}, Mattias Bergental^{3,4}, Pouyan Ghaffari¹, Cheng Zhang², Erik Larsson^{3,4}, Fredrik Bäckhed^{3,4} & Jens Nielsen^{1,2}

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^{1,2,3} ✉, Lenaig Abily-Donval^{2,4} ✉, Carlos Afonso³ ✉, Stéphane Marret^{2,4} ✉ and Soumeia Bekri^{1,2,*} ✉

Inherited Metabolic Diseases

RESEARCH ARTICLE

CANCER DIAGNOSTICS

Intraoperative Tissue Identification Using Rapid Evaporative Ionization Mass Spectrometry

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

Pathology and Cancer

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

Muireann Coen,^{*,†} Françoise Goldfain-Blanc,[‡] Gaëlle Rolland-Valognes,[§] Bernard Walther,^{||} Donald G. Robertson,[‡] Elaine Holmes,[†] John C. Lindon,[†] and Jeremy K. Nicholson^{*,†}

Responder Non-responder Prediction

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

PB Phapale^{1,2}, S-D Kim², HW Lee^{1,2}, M Lim^{1,2}, DD Kale^{1,2}, Y-L Kim^{2,3}, J-H Cho⁴, D Hwang⁴ and Y-R Yoon^{1,2}

Human Drug Pharmacokinetics



ARTICLE

<https://doi.org/10.1038/s41467-019-08936-1>

OPEN

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

Laura B.L. Wittemans¹, Luca A. Lotta¹, Clare Oliver-Williams^{2,3}, Isobel D. Stewart¹, Praveen Surendran², Savita Karthikeyan², Felix R. Day¹, Albert Koulman^{1,4}, Fumiaki Imamura¹, Lingyao Zeng^{5,6}, Jeanette Erdmann^{7,8,9}, Heribert Schunkert^{5,6}, Kay-Tee Khaw¹⁰, Julian L. Griffin¹¹, Nita G. Forouhi¹, Robert A. Scott¹, Angela M. Wood², Stephen Burgess¹², Joanna M.M. Howson¹³, John Danesh^{2,13}, Nicholas J. Wareham¹, Adam S. Butterworth² & Claudia Langenberg¹

Large-scale epidemiological profiling

Metabolomics-based imaging

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Science News ... from universities, journals, and other research institutions

'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,^{1*} László Sasi-Szabó,^{2*} James Kinross,^{3,4} Matthew R. Lewis,³ Laura J. Muirhead,^{3,4} Kirill Veselkov,³ Reza Mirnezami,⁴ Balázs Dezső,⁵ László Damjanovich,² Ara Darzi,⁴ Jeremy K. Nicholson,^{3†} Zoltán Takáts^{3†}

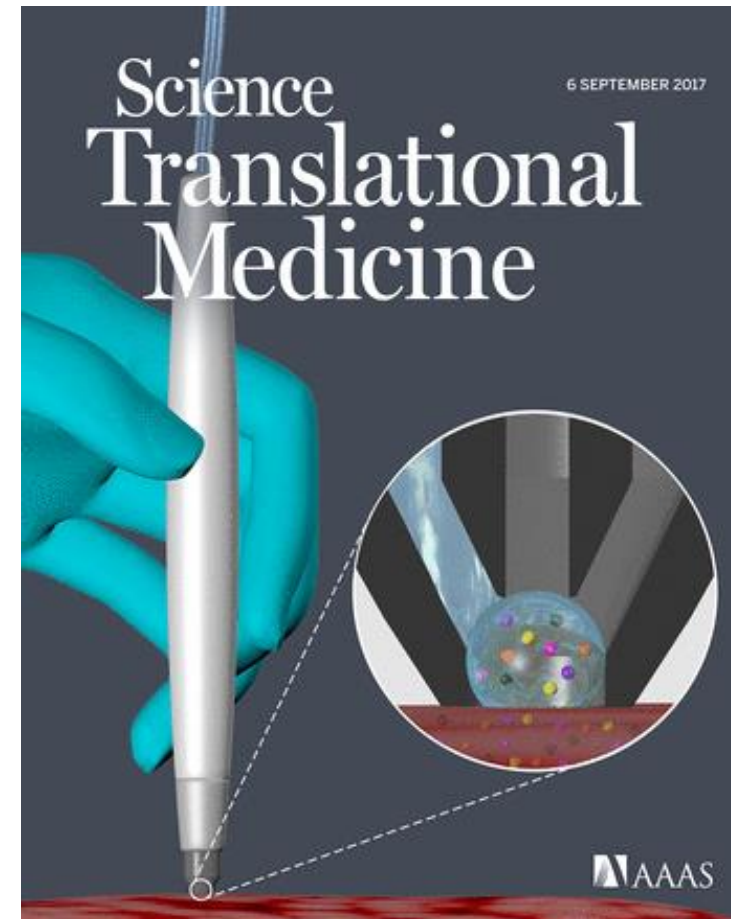
SCIENCE TRANSLATIONAL MEDICINE | RESEARCH ARTICLE

CANCER DIAGNOSTICS

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

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

Real time Metabolomics: Precision Surgery



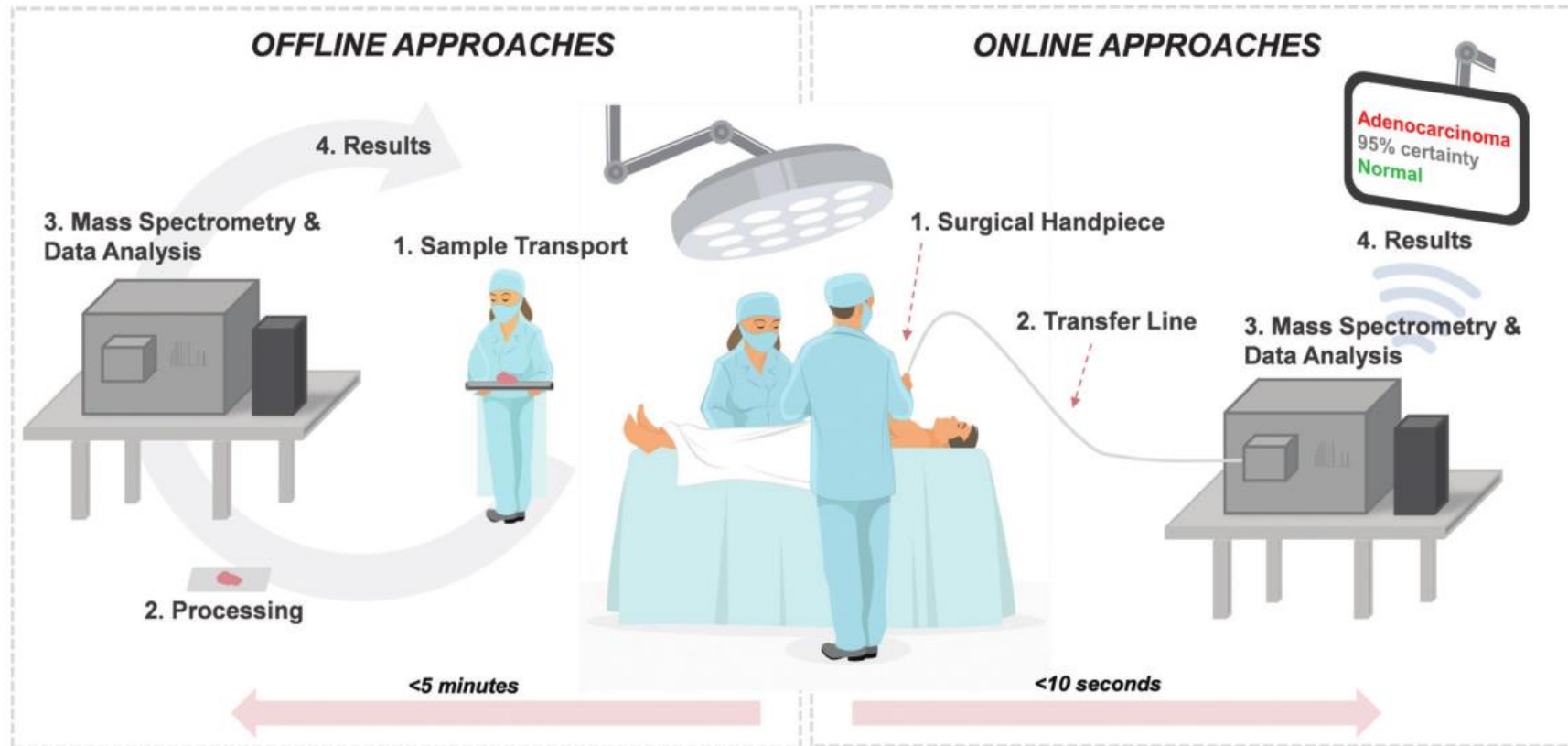
Metabolomics-based imaging



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

Signal



Data

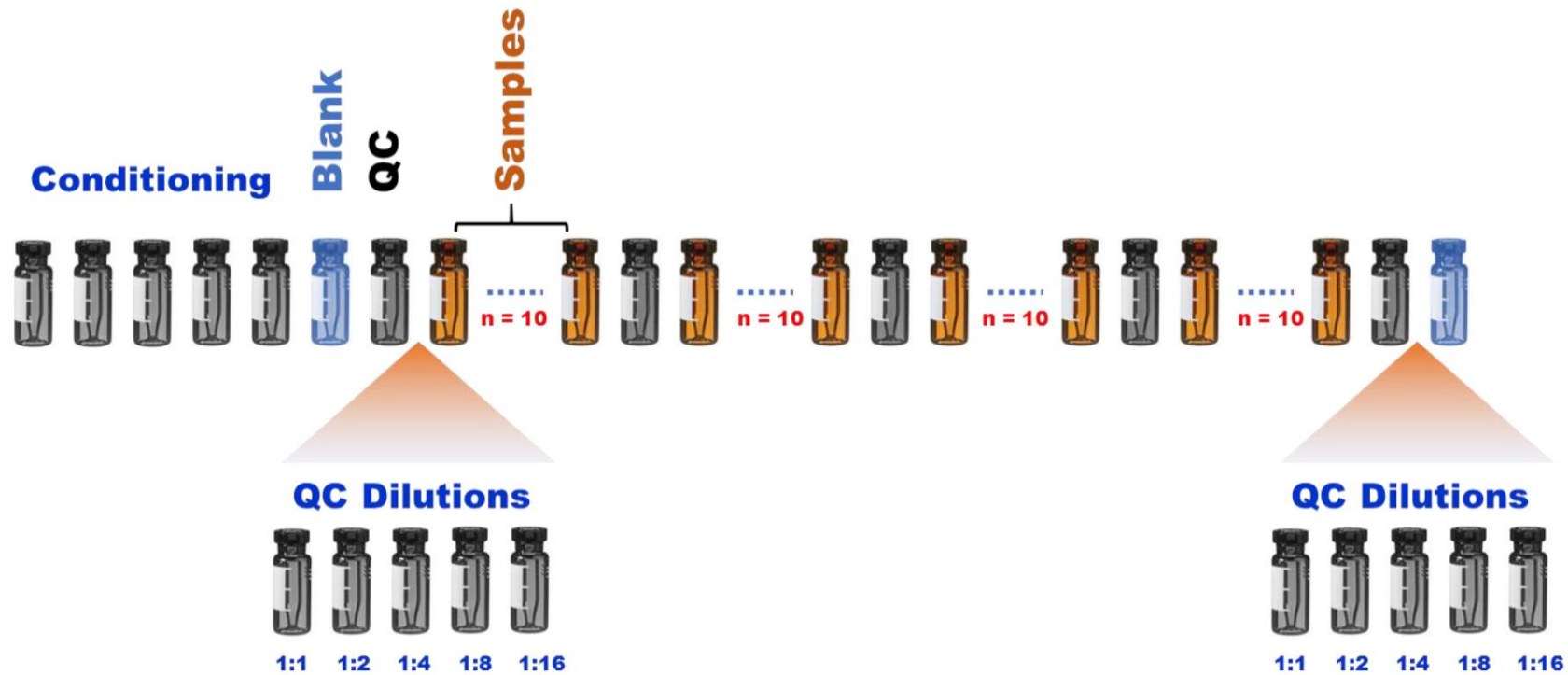


Variables

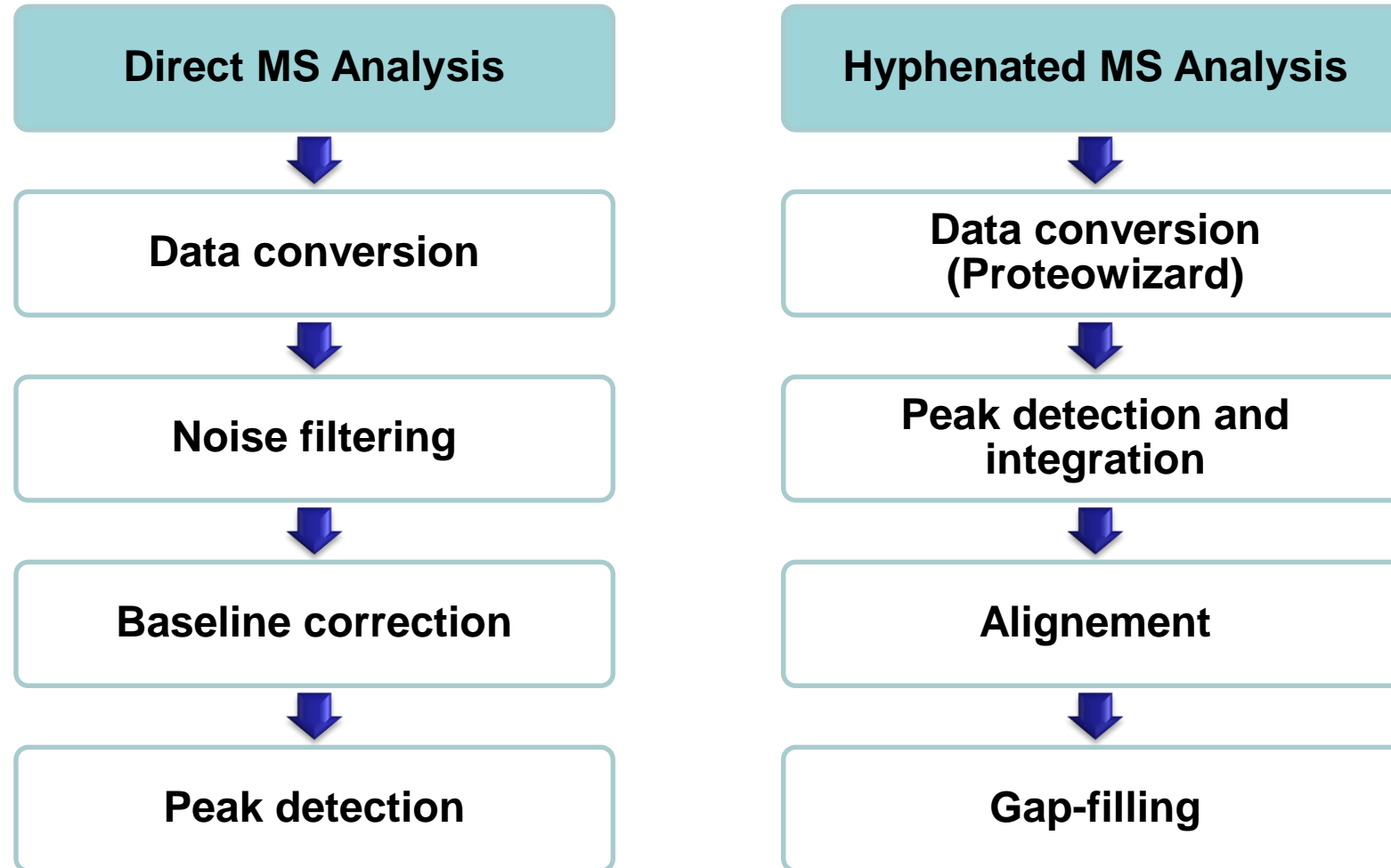
Samples

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

CLI command line interface, *GUI* graphical user interface

Spicer, R., Salek, R.M., Moreno, P. et al. *Metabolomics* (2017) 13: 106

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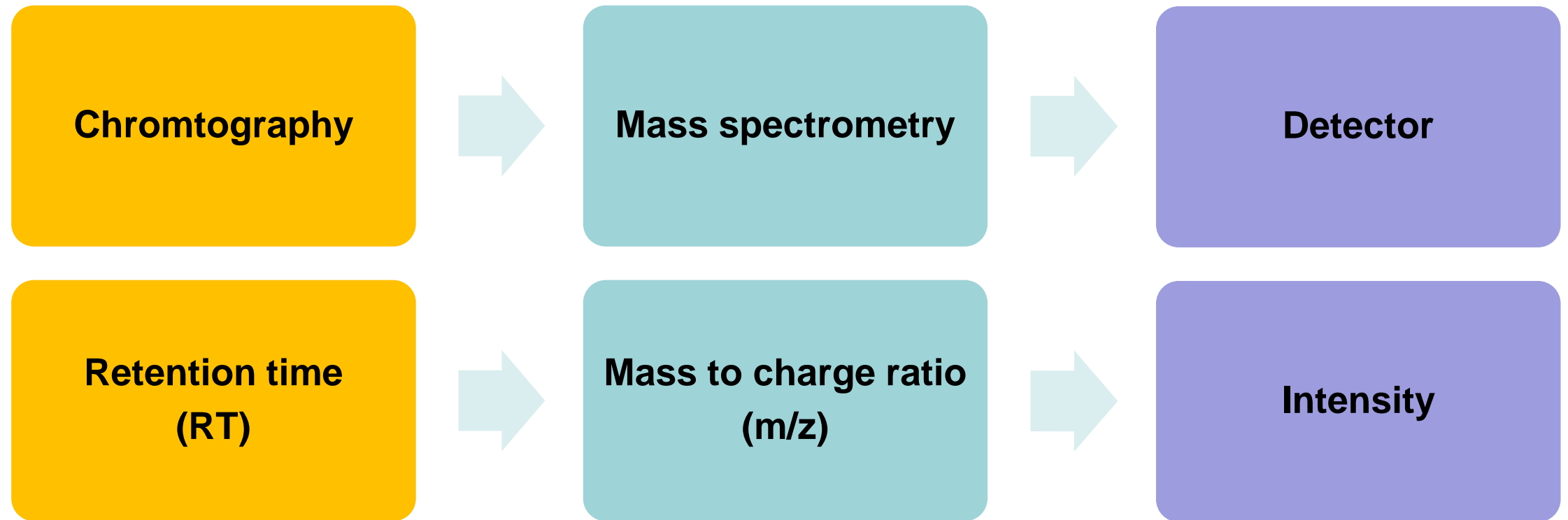
Accepted November 9, 2018

Review

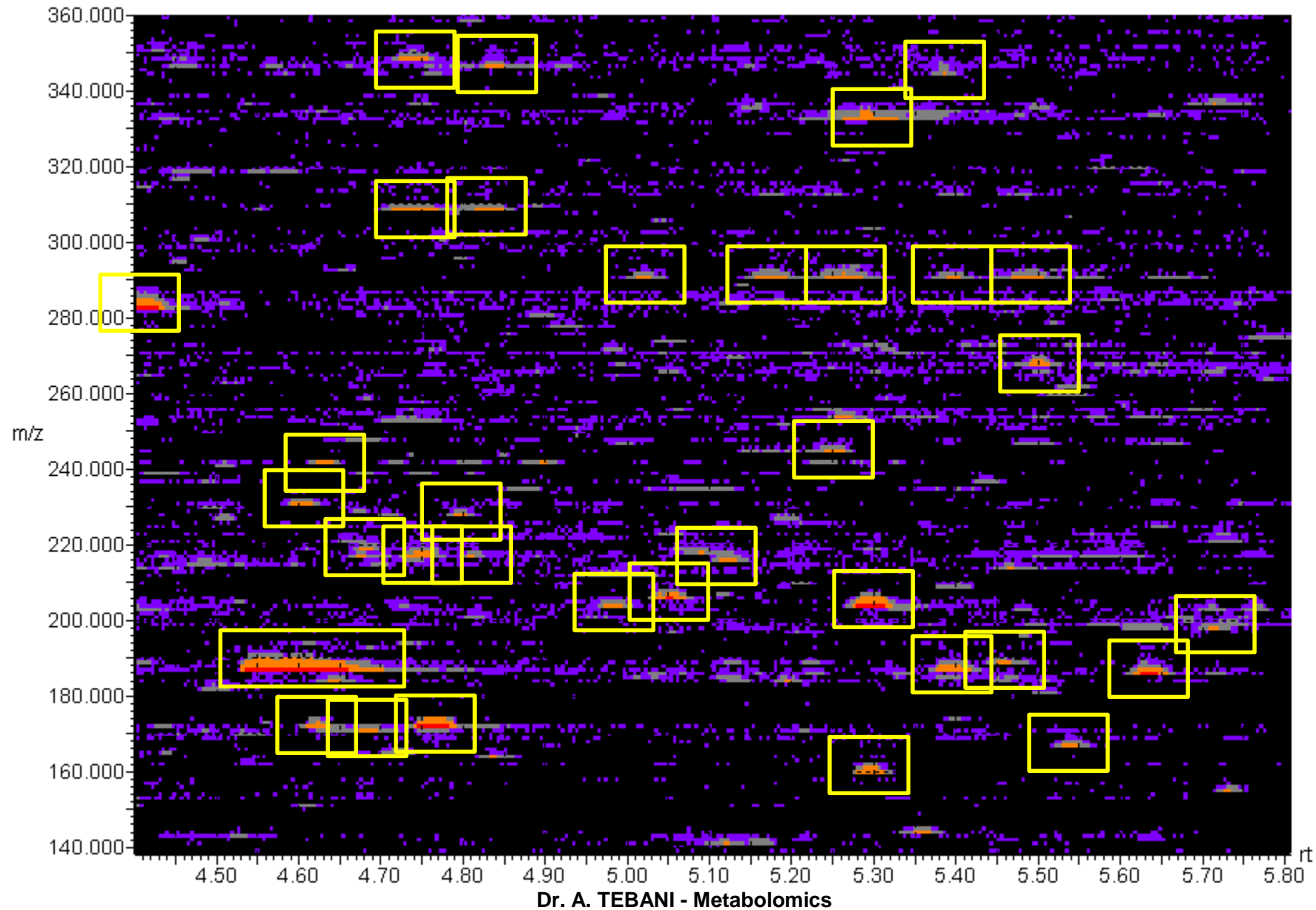
Tools and resources for metabolomics research community: A 2017–2018 update

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.

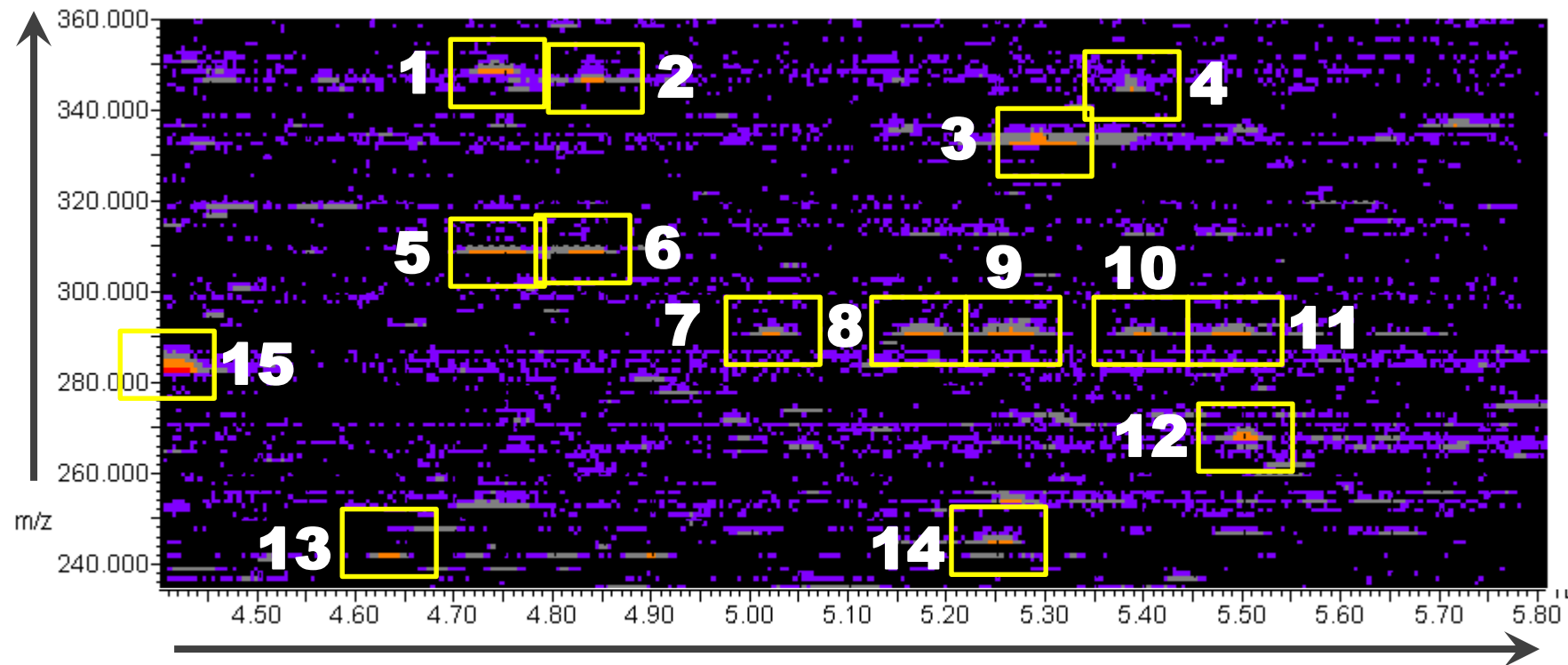
Output data structure



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



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



Feature detection in 3 dimensions

Concatenated to
single term
representing each
feature

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

Feature = RT_mz

Output data structure

Feature Identifier

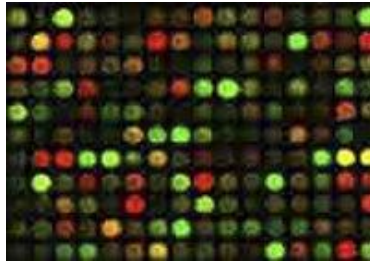


Sample identifier

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_QC_NEG
Label	mz	rt	QC.nor.rsd	R3	T	F	T	F	QC
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	0,922779377	0,893270646	0,801607214	2,58524926	0,979421296
13.84_925.6952m/z	925,6952	13,84	2,97	0,91	1,541006208	1,549550309	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

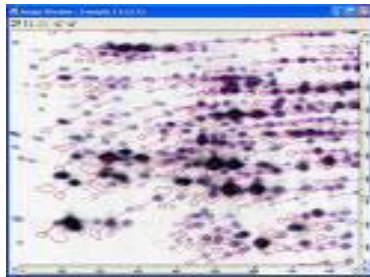
Feature intensity

The biggest challenge is Annotation



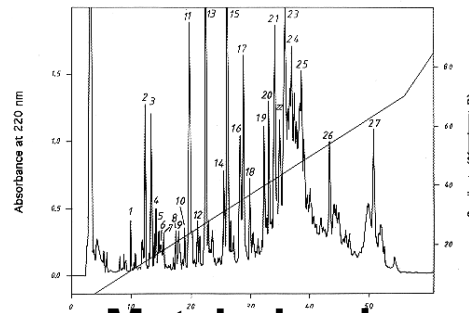
DNA/RNA

Gene IDs +
Transcript
Abundance

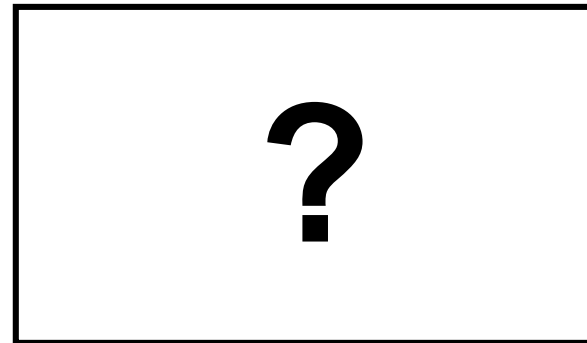


Proteomics

Protein IDs +
Concentrations



Metabolomics



Metabolite IDs +
Concentrations

Metabolome Databases

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



www.hmdb.ca



www.drugbank.ca



www.ymdb.ca



www.phenol-explorer.eu



www.ecmdb.ca



www.foodb.ca



www.cowmetdb.ca



www.t3db.ca



www.smpdb.ca



www.csfmetabolome.ca



www.serummetabolome.ca



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)

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

Mass:

Tolerance: PPM

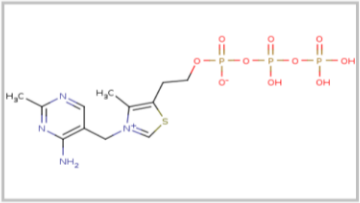
Charge: Neutral **Positive** Negative

Adducts: **M+H** M+NH4 **M+Na** 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

Peptides: Remove Peptides from Search

Show entries

Search:

METLIN ID	Mass	Δ PPM	Name	KEGG	MS/MS	Structure
3586	[M+H] ⁺ <u>m/z</u> 505.0107 M 504.0035	21	Thiamin triphosphate <i>Formula: C₁₂H₁₉N₄O₁₀P₃S</i> <i>CAS: 3475-65-8</i>	C03028	NO	

METLIN ID	Mass	Δ PPM	Name	KEGG	MS/MS	Structure
-----------	------	--------------	------	------	-------	-----------

Showing 1 to 1 of 1 entries

Previous **1** Next

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

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

Convert →

InChIKey

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

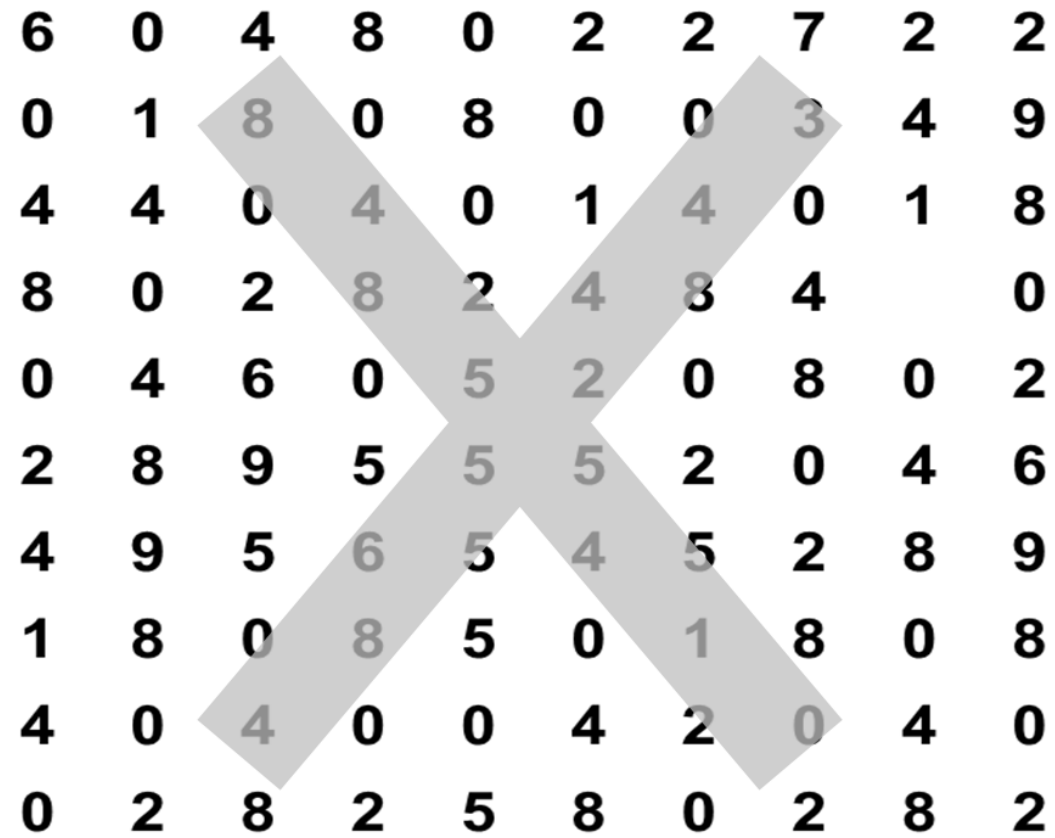
Predictive models

Data Analysis

Variables

Metabolites

Samples



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 Analysis

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, R2 ...)
- 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
12.31_854.6227m/z	854,6227	12,31	1,90	0,91	0,922779377	0,893270646	0,801607214	2,58524926	0,979421296
13.84_925.6952m/z	925,6952	13,84	2,97	0,91	1,541006208	1,549550309	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

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: QCs, Controls...

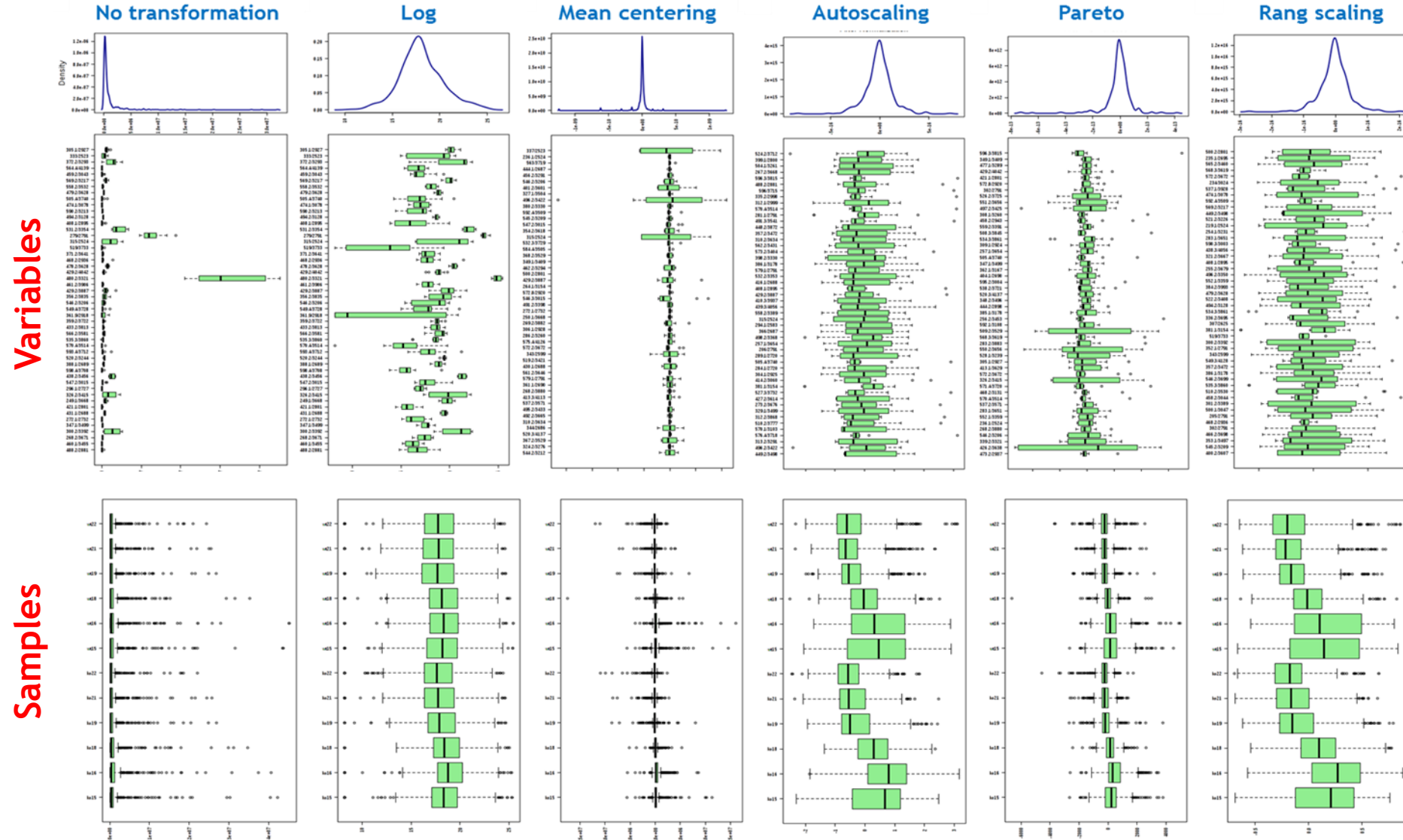
Feature normalization (column-wise)

To bring variances of all features close to equal

Log transformation

Scaling

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^{1,†}, Jing Tang^{1,†}, Qingxia Yang^{1,2,†}, Shuang Li¹, Xuejiao Cui¹, Yinghong Li¹,
Yuzong Chen³, Weiwei Xue¹, Xiaofeng Li¹ and Feng Zhu^{1,2,*}

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

doi: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. Zacharias¹, Michael Altenbuchinger² and Wolfram Gronwald^{3*}

Multivariate Data Analysis

Two main objectives

Descriptive data analysis (Unsupervised learning)

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



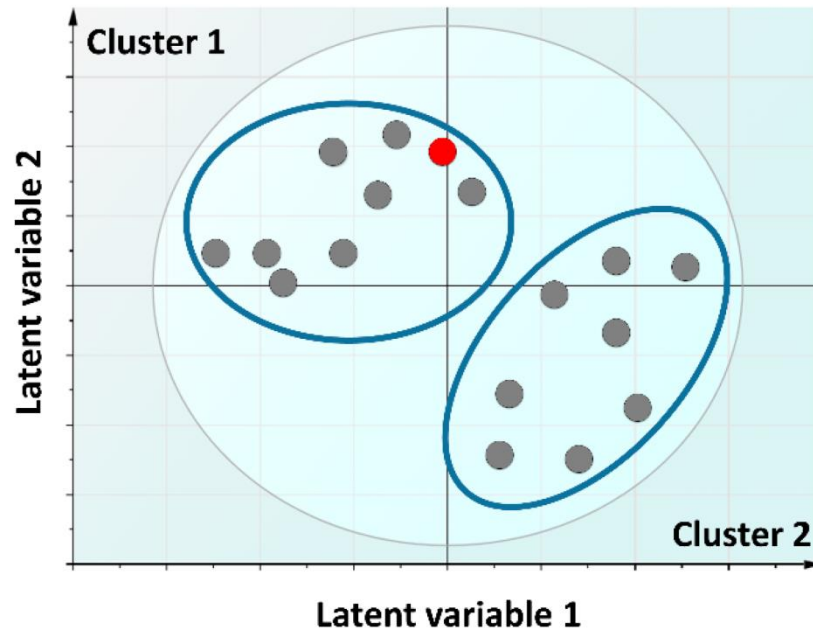
Predictive data analysis (Supervised learning)

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

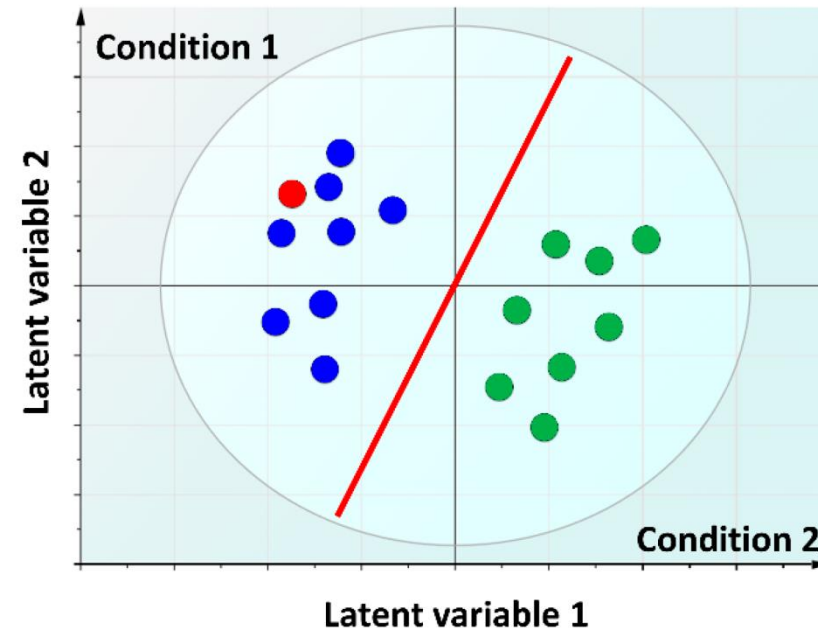


Multivariate Data Analysis

Unsupervised learning



Supervised learning



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

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

Model validation

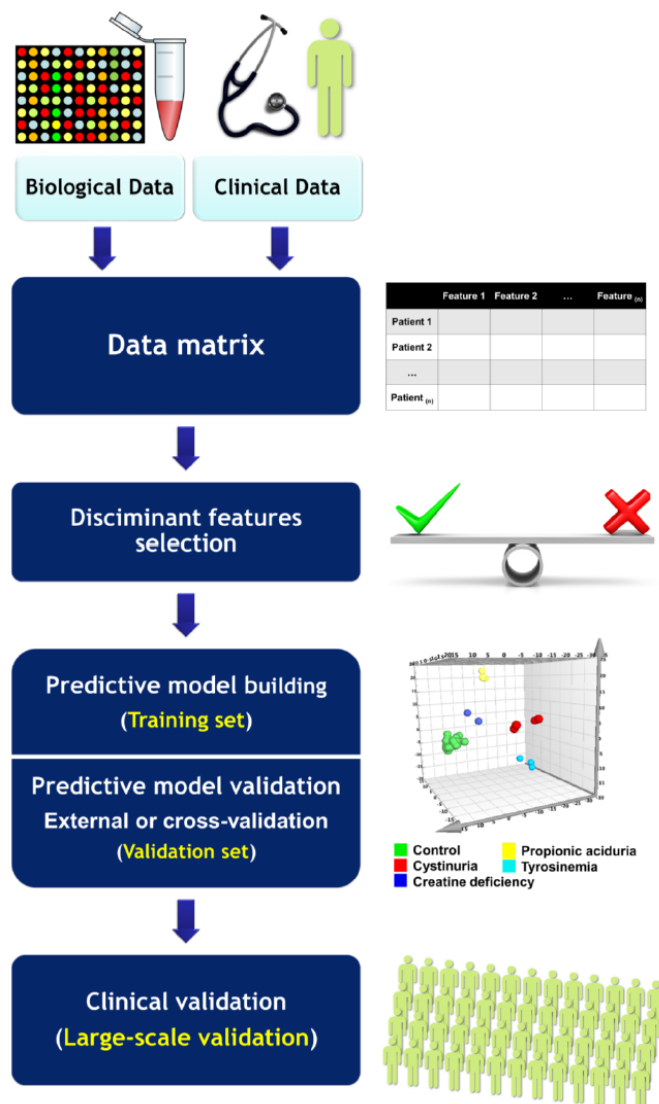
A statistically valid model

- 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 you data

Main technologies are MS and NMR

Annotation is challenging in untargeted metabolomics