

Imperial College
London

LC-MS Applications in Metabonomics: from Toxicology to Disease Biomarkers

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Biomolecular Medicine
Imperial College, London



Overview

Challenges in Metabonomics

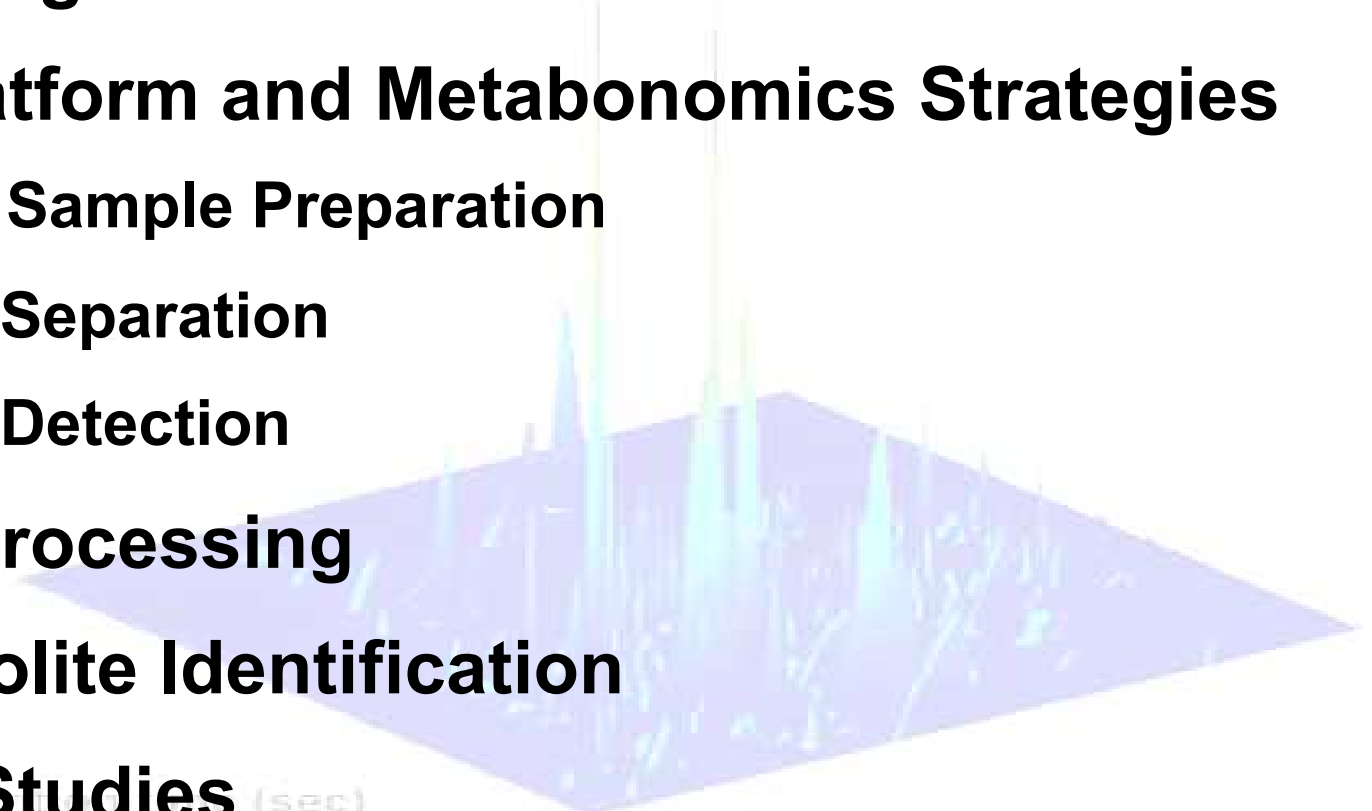
MS platform and Metabonomics Strategies

- Sample Preparation
- Separation
- Detection

Data Processing

Metabolite Identification

Case Studies (sec)



Challenges in Metabonomics

Untargeted analysis

Hundreds of molecules in complex matrices

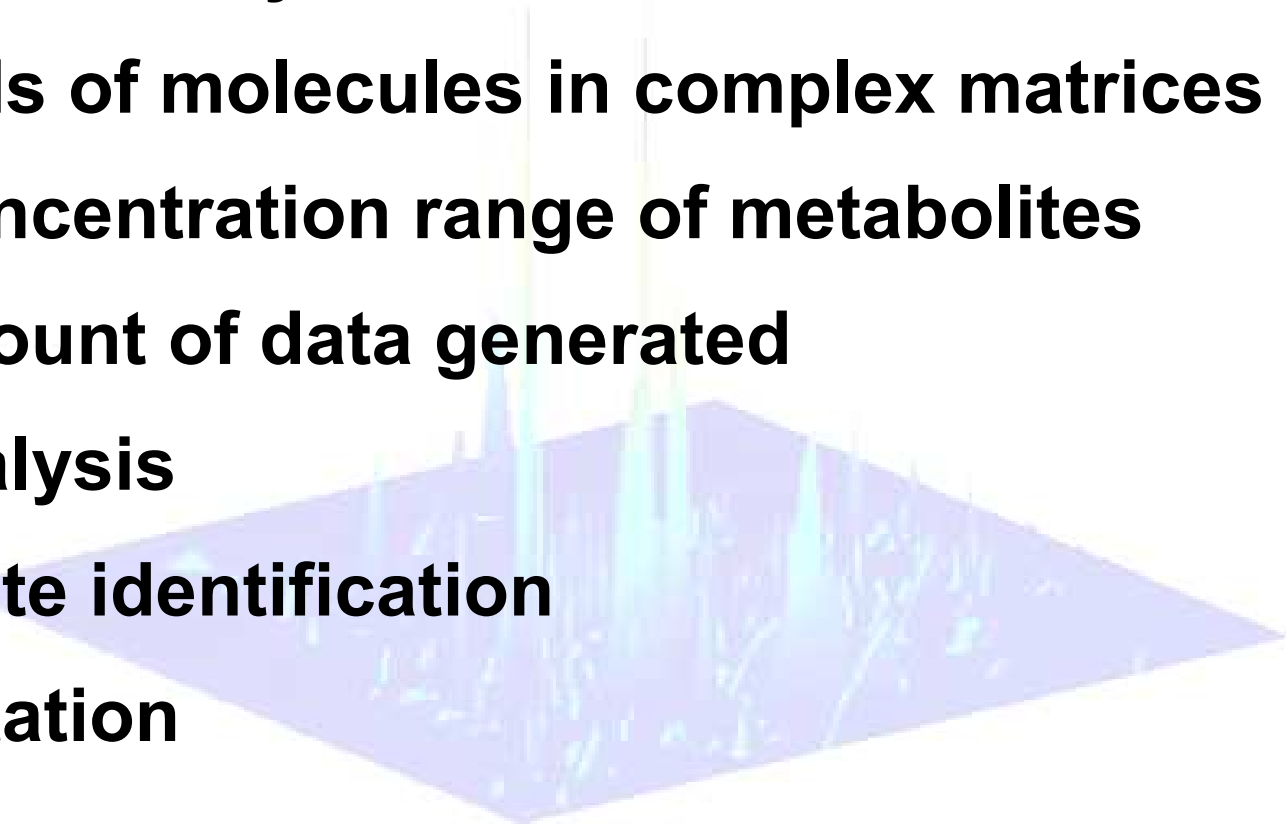
Wide concentration range of metabolites

Vast amount of data generated

Data analysis

Metabolite identification

Interpretation



Why use MS?

Complementary to NMR
Reproducibility
Sensitivity
Dynamic range
Sample throughput
Structural information
Quantitation
Data analysis automation



MS in Metabonomics

Cancer

- kidney & ovarian cancer, colorectal cancer, brain tumours

Toxicology

- COMET and COMET2

Disease biomarkers

- cardiovascular disease, diabetes

Nutrition

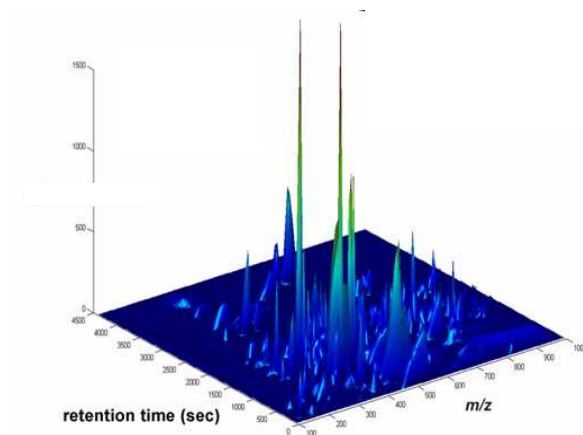
- Effects of green tea, flavonoids

Plants

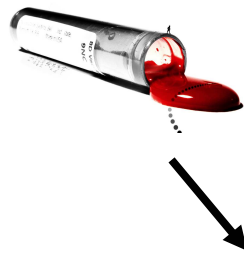
- plant-host interactions, growth rate

Other organisms

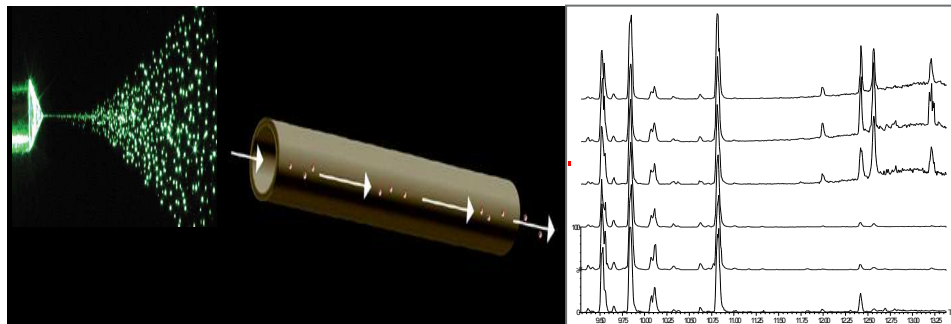
- yeast, fungi



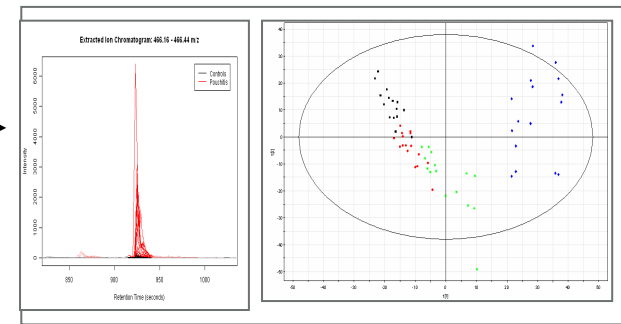
Metabonomics Platform



Sample
preparation



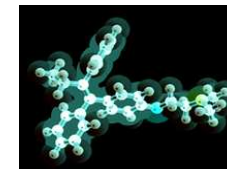
LC/GC-MS Analysis



Data Analysis



Metabolite
Identification

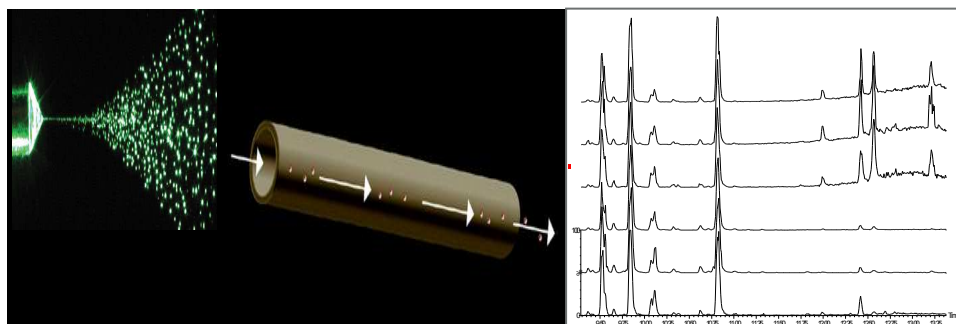


Metabonomics Platform

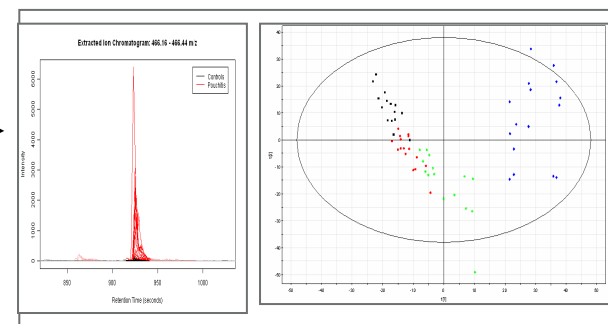


Sample
preparation

Considerations



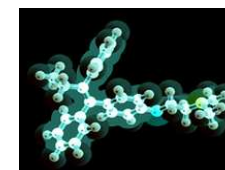
LC/GC-MS Analysis



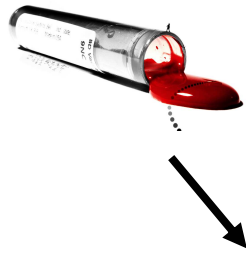
Data Analysis



Metabolite
Identification

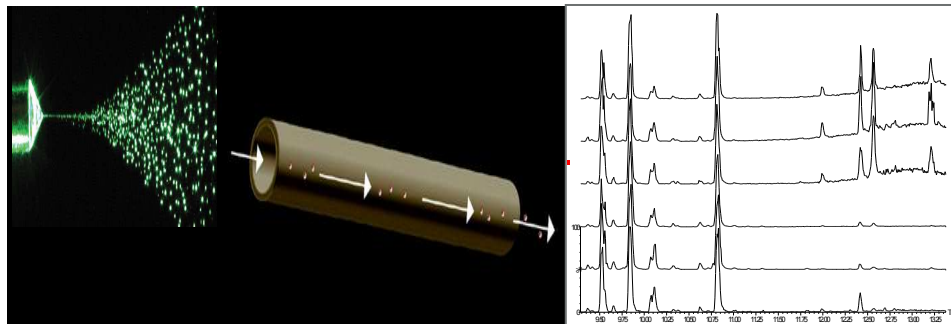


Metabonomics Platform



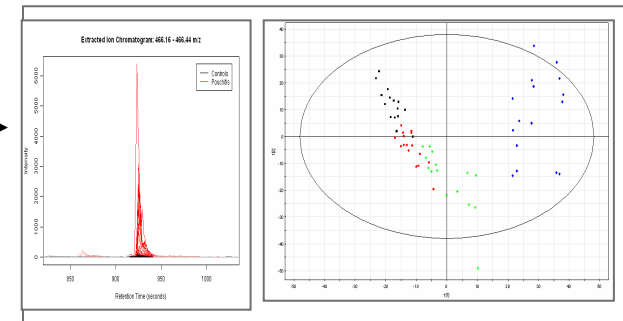
**Sample
type**

Considerations



**Chromatography
Mass Spectrometer**

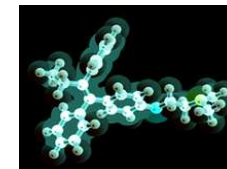
**Prior knowledge
e.g. NMR data**



Data Processing



**Metabolite
Identification**



Sample Preparation

Dependent on goal

- Untargeted analysis: minimal sample pre-treatment to prevent loss of metabolites
- Sample pre-concentration techniques i.e. SPE for low level metabolites

Serum/plasma

- Methanol/acetonitrile protein precipitation

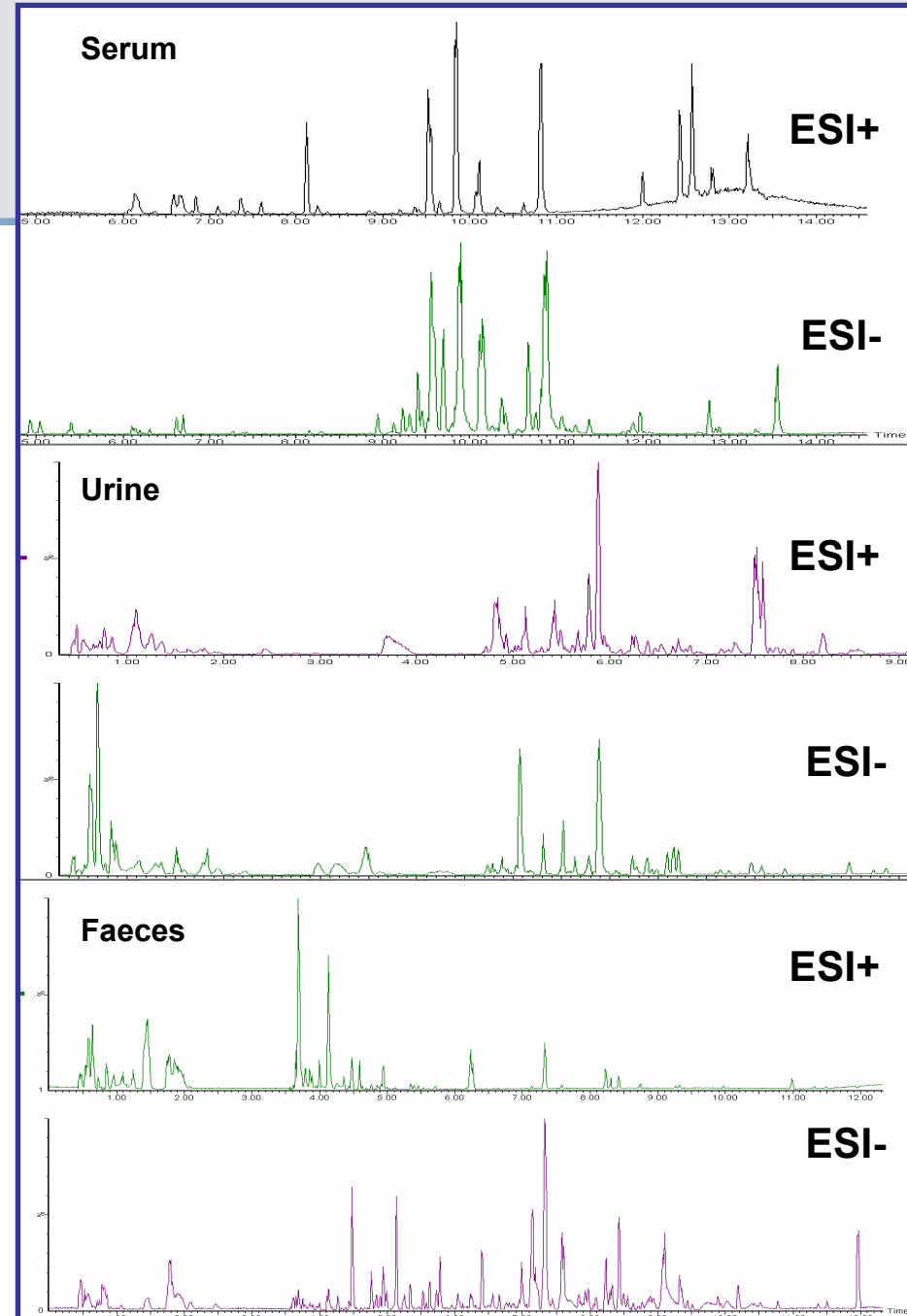
Urine

- Centrifugation & dilution

Faeces

- Aqueous/organic extracts

UPLC-MS Base Peak Intensity Chromatograms



Sample Preparation

Bile

Dilution 1:4 with water

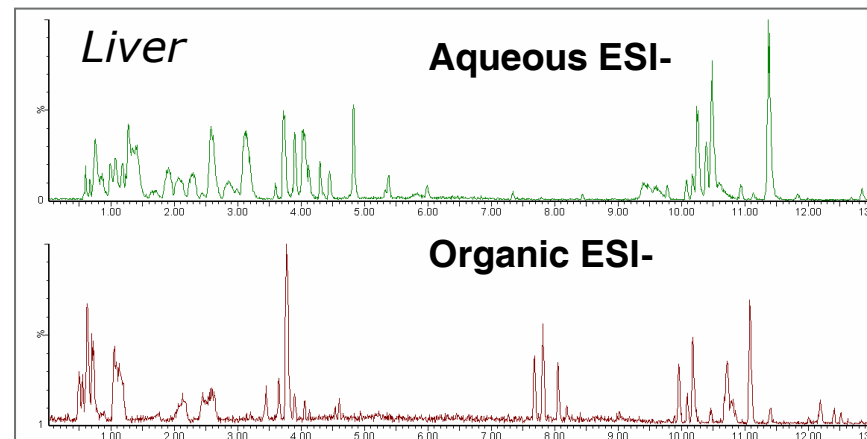
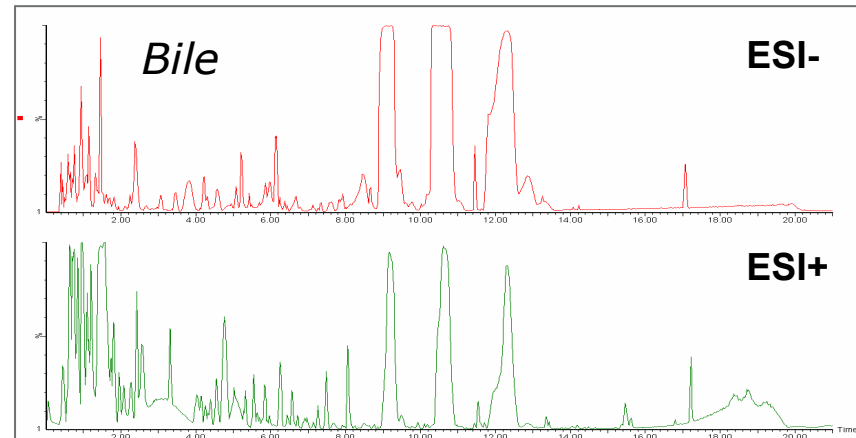
Centrifugation 13000rpm 10mins

Cerebrospinal fluid

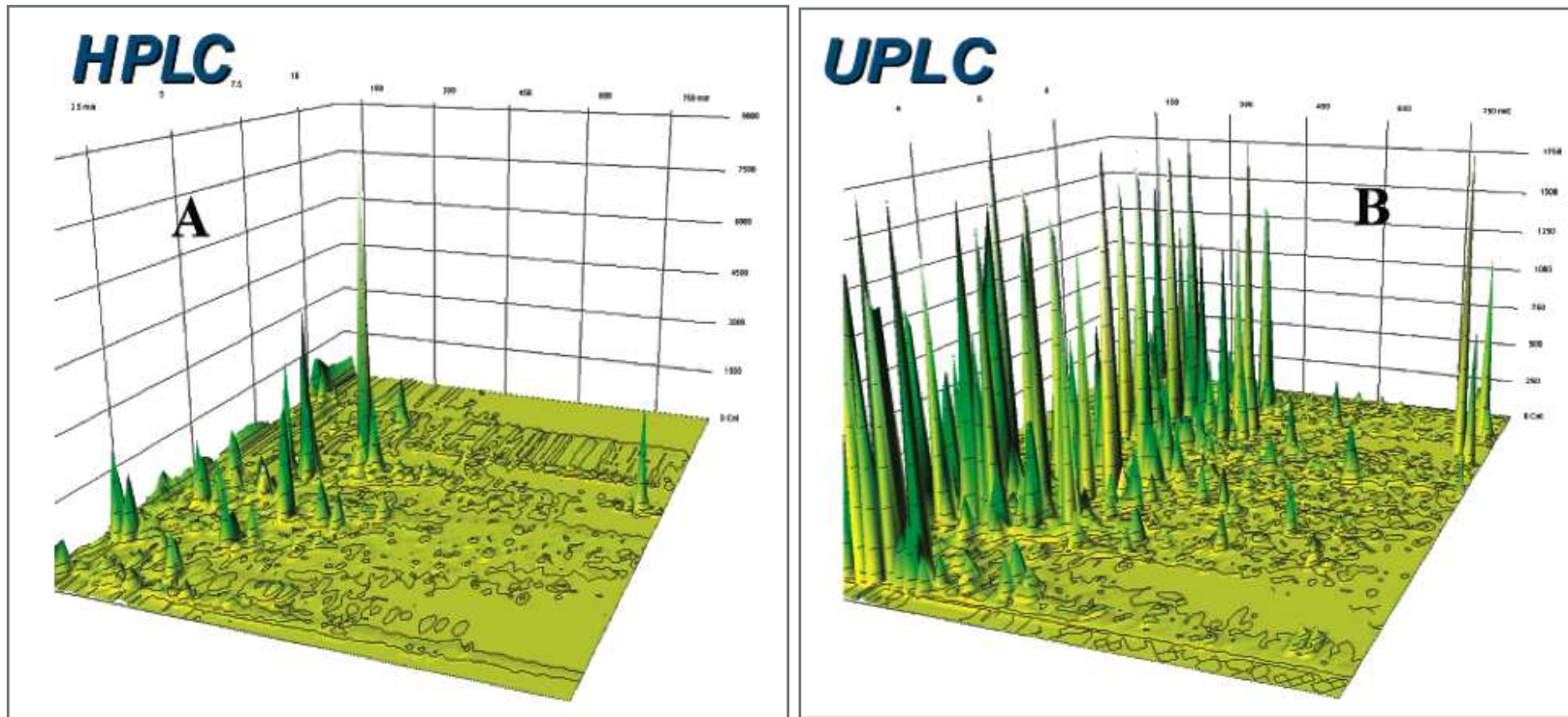
Methanol protein precipitation
(as serum/plasma)

Tissues e.g. liver

Chloroform: methanol extraction:
aqueous and organic extracts



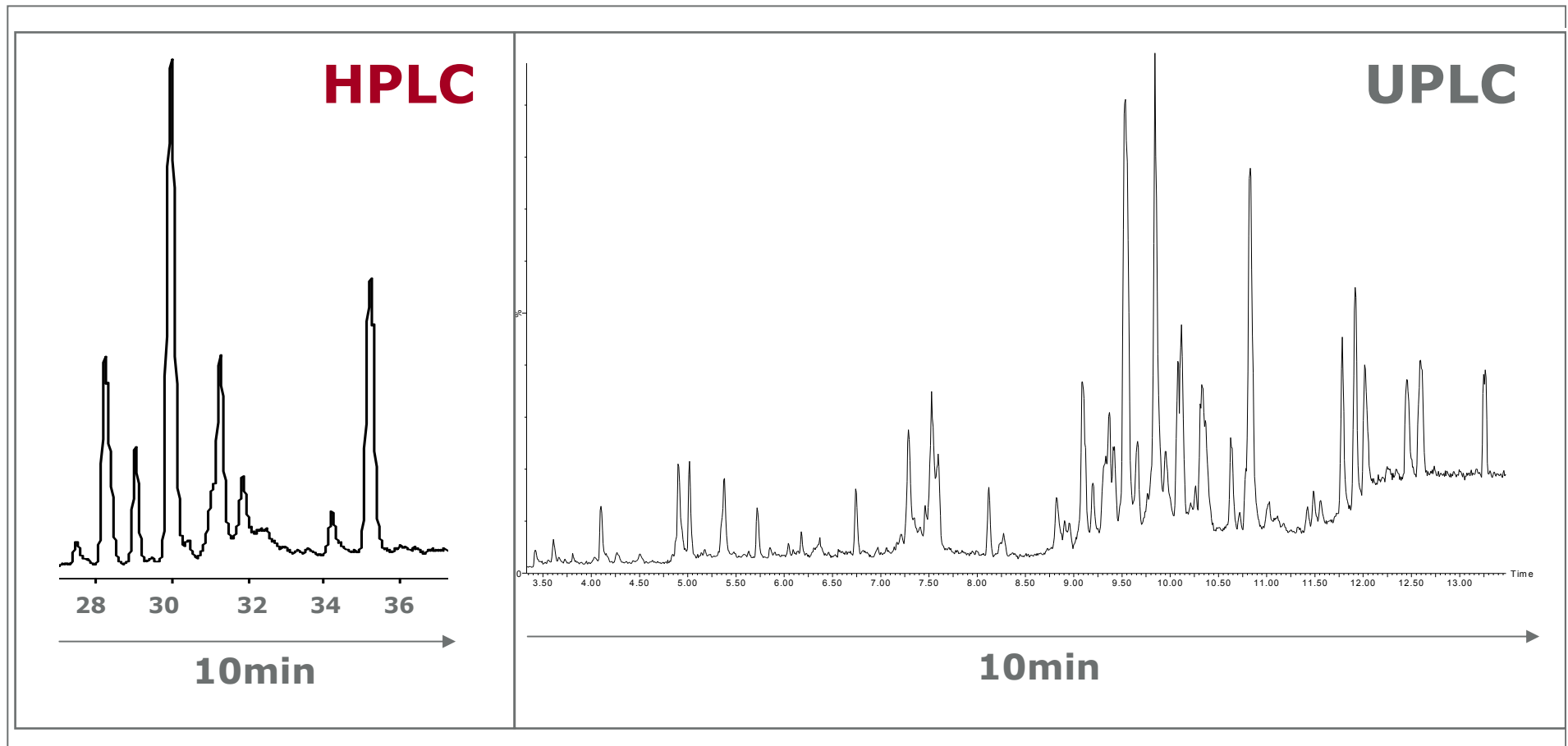
Liquid chromatography



Wilson ID *et al.*, J Proteome Res. 2005. 4(2):591-8.

UPLC *vs* HPLC: Serum sample (methanol extract)

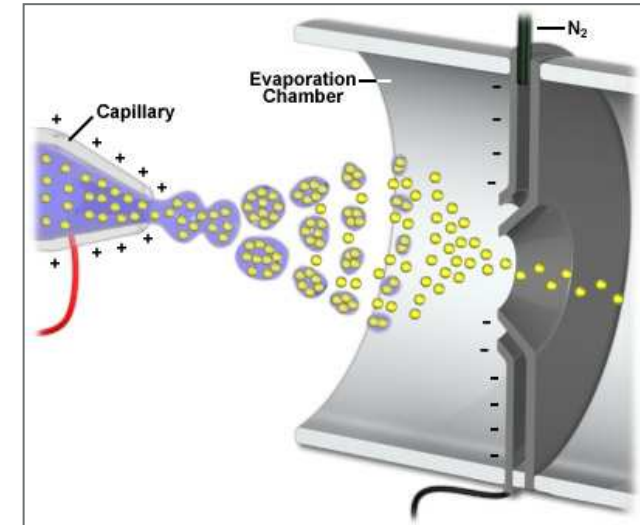
1.7 μ m particle size columns can withstand higher backpressures and flow rates



MS: Ionisation and Detection

Electrospray ionisation (ESI)

- Can be interfaced to liquid chromatography
- Readily amenable to MS analysis
- No matrix necessary

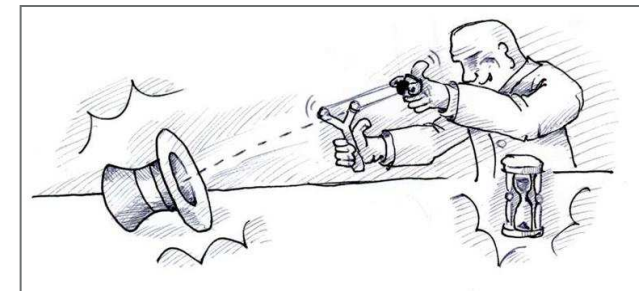


Time of flight (ToF)

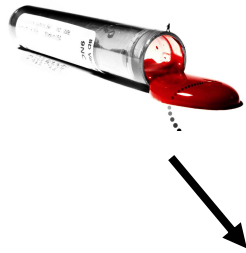
- good mass accuracy

Quadrupole time of flight (Q-ToF)

- good mass accuracy, MS/MS

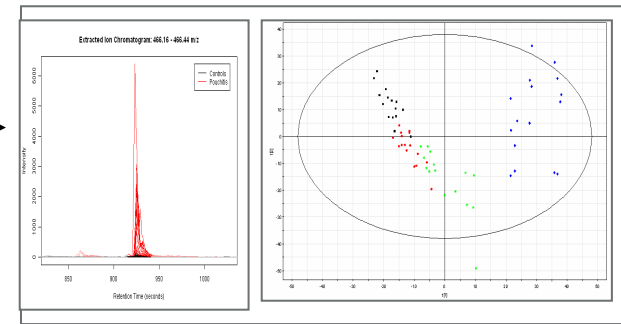
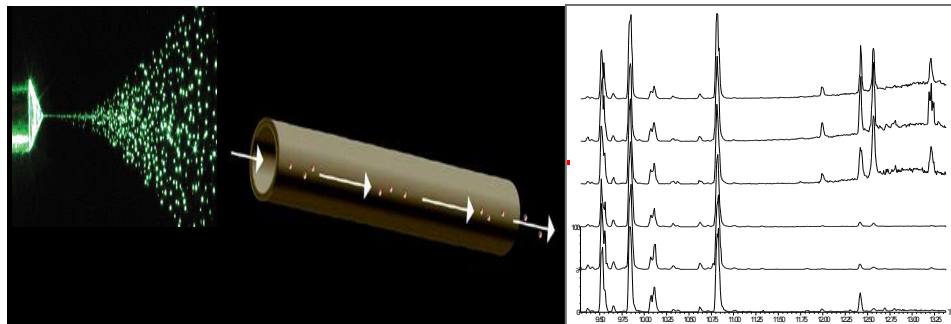


Metabonomics Platform



**Sample
type**

Considerations

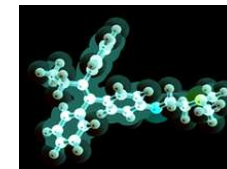


**Chromatography
Mass Spectrometer**

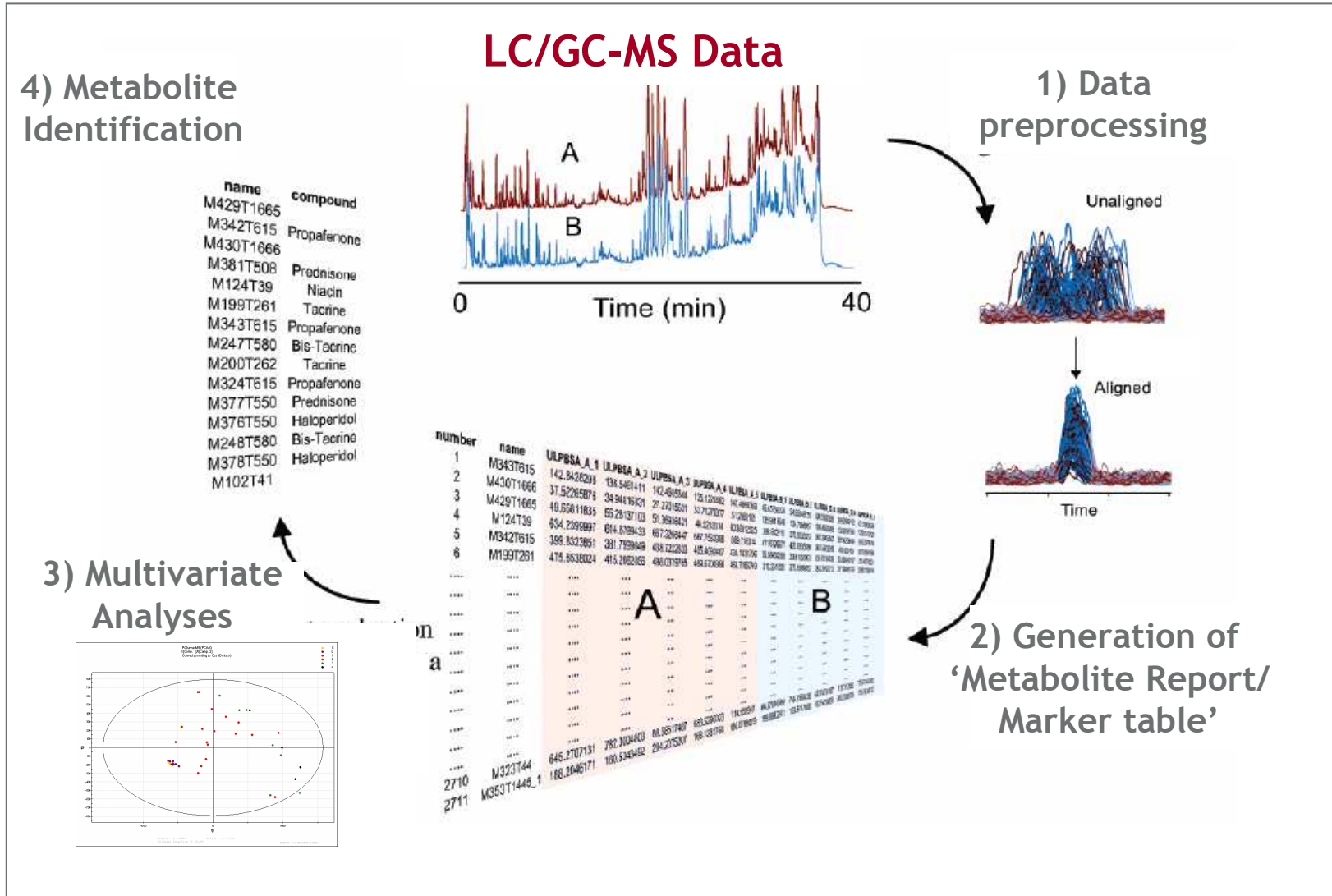
Data Processing

**Prior knowledge
e.g. NMR data**

**Metabolite
Identification**



Data Processing Workflow



Data Preprocessing

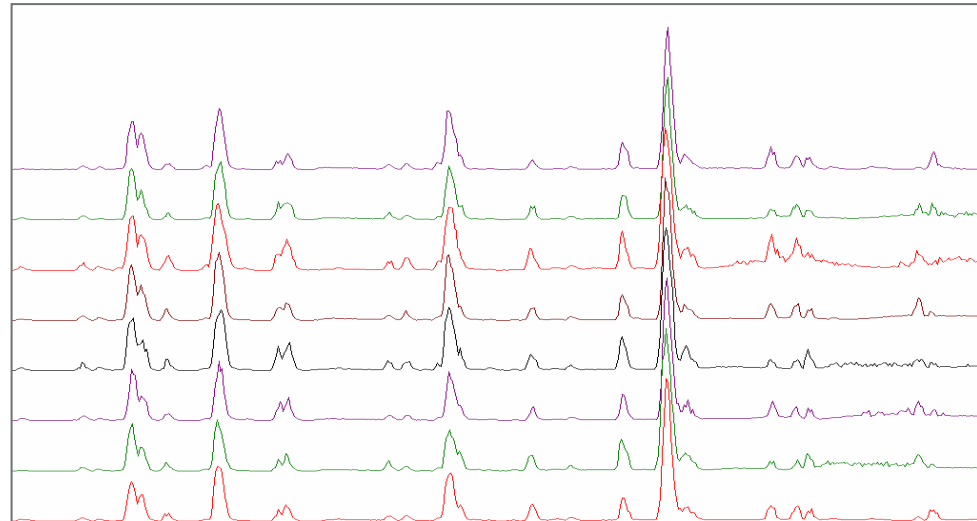
Objectives

Peak picking and alignment

‘Matching’ peaks across samples

Determination of differences between samples

Normalisation



Data Preprocessing Challenges

1) Dataset Complexity

Thousands of peaks

Isotopes, adducts, dimers, fragments

Noise

Positive & negative mode data

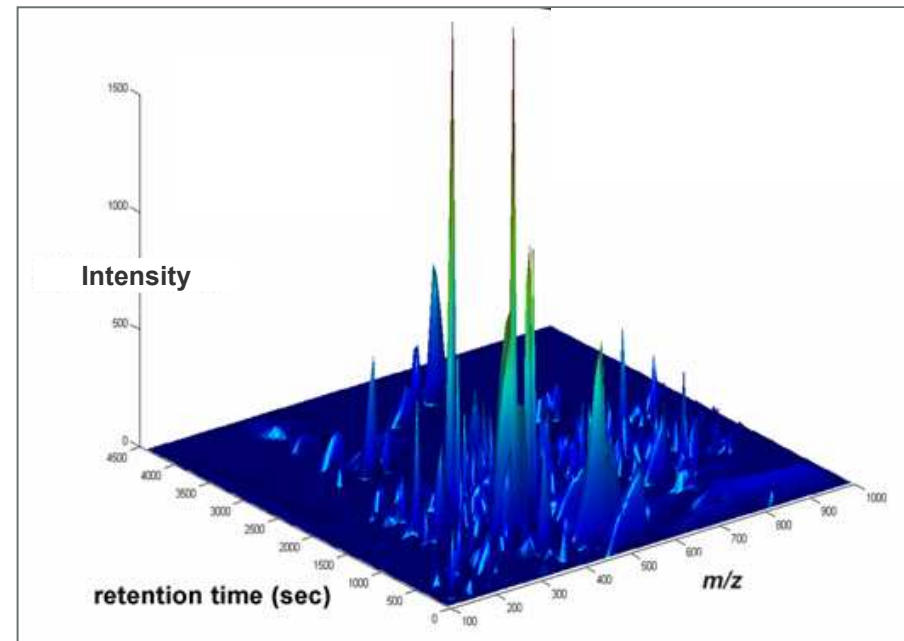
2) Peak Shifts

Temperature

Mobile phase changes

Stationary phase changes

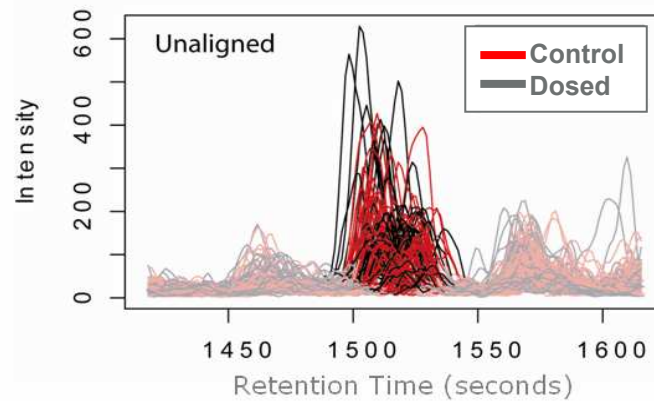
Sample composition



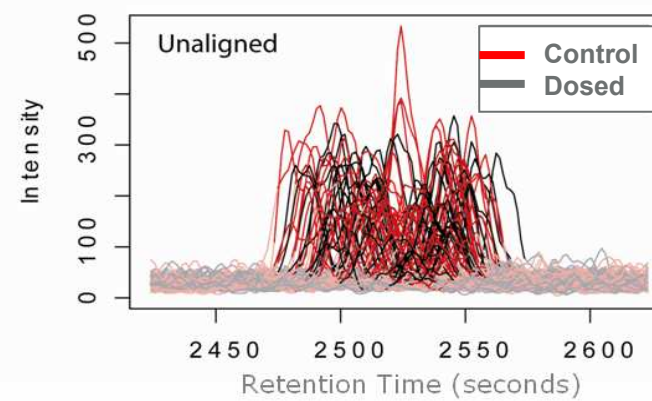
Importance of Alignment

Unaligned

Extracted Ion Chromatogram: 550.2 - 550.5 m/z

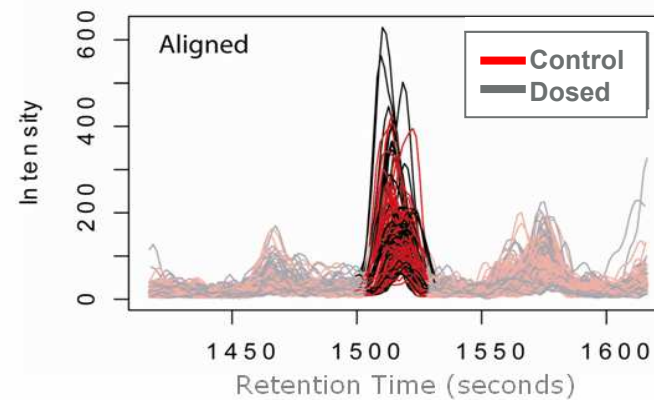


Extracted Ion Chromatogram: 320.1 - 320.4 m/z

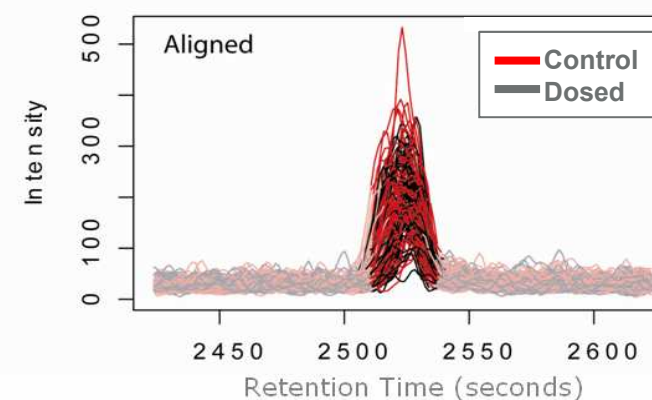


Aligned

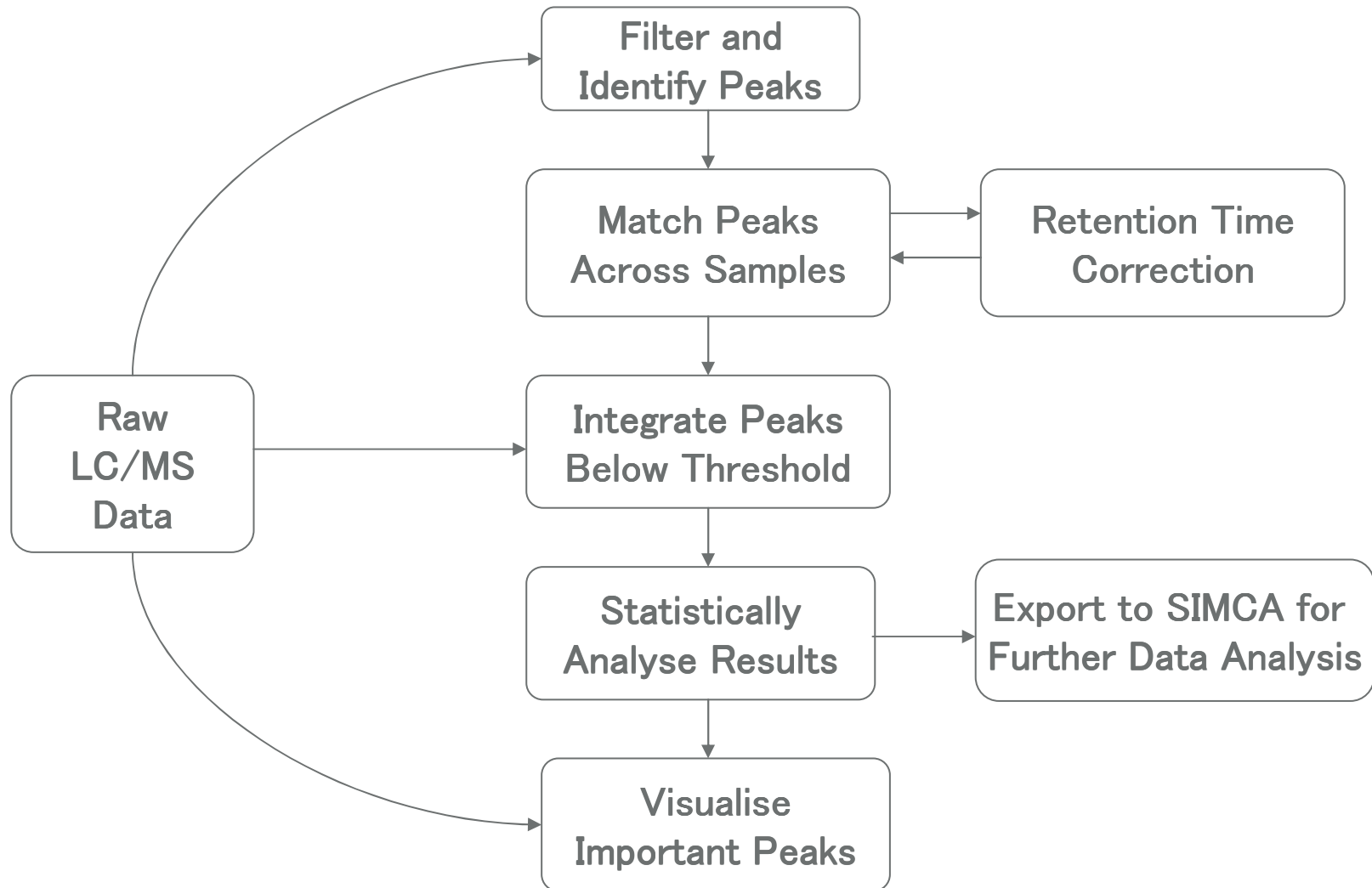
Extracted Ion Chromatogram: 550.2 - 550.5 m/z



Extracted Ion Chromatogram: 320.1 - 320.4 m/z



General Preprocessing Approach



Software Options

Commercial Software

Markerlynx



MassHunter



MarkerView



MassFrontier



Metabolic Profiler



Platform Independent Freeware

MZMINE

Katajamaa M, Oresic M. BMC Bioinformatics. 2005.18;6:179.

Katajamaa M, Miettinen J, Oresic M. Bioinformatics. 2006. 22(5):634-6.

MSFACTS

Duran AL, et al., Bioinformatics. 2003. 19(17):2283-93.

METALIGN

De Vos RC, et al., Nat Protoc. 2007;2(4):778-91.

Tikunov Y, et al., Plant Physiol. 2005 Nov;139(3):1125-37.

MET-IDEA

Broeckling CD, et al., Anal Chem. 2006 Jul 1;78(13):4334-41.

MATHDAMP

Baran R et al., BMC Bioinformatics. 2006. 7:530

XCMS

Smith CA, et al., Anal Chem. 2006. 78(3):779-87.

Metabolite Identification

Query databases

Isolate metabolite of interest (prep LC)

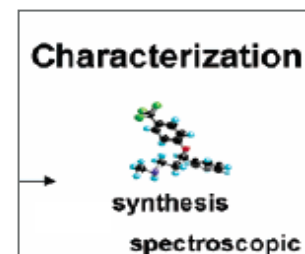
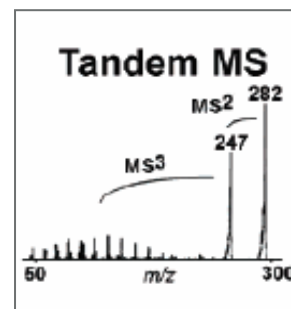
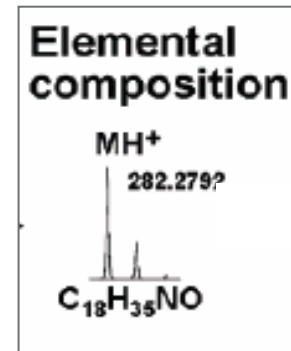
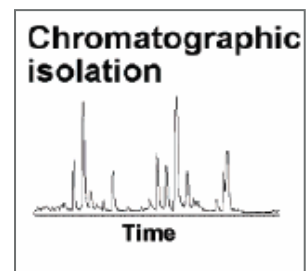
Obtain high accuracy mass data (FTMS)

Fragmentation data (MS^E , MS/MS)

Other spectroscopic techniques for further characterisation (NMR)

Purchase/synthesise standard

- Compare retention time
- Accurate mass
- Fragmentation



Metabolite Identification



LipidBank

Suite of databases and associated software



Tomato Metabolite Database



18.5 million unique chemical compounds



>6500 metabolites



Plant metabolites and lipids



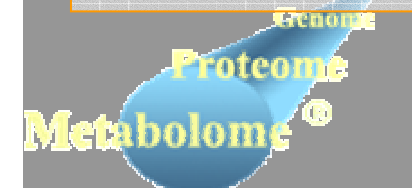
> 23,000 metabolites

Must purchase a license

MDL® Metabolite Database

<http://www.mdl.com/products/predictive/metabolite/>

Tumor metabolome



Metabolic Database

GMD@CSB.DB
- The Golm Metabolome Database -
Hosted at Max Planck Institute of Molecular Plant Physiology

Custom MS libraries & metabolite profiling experiments



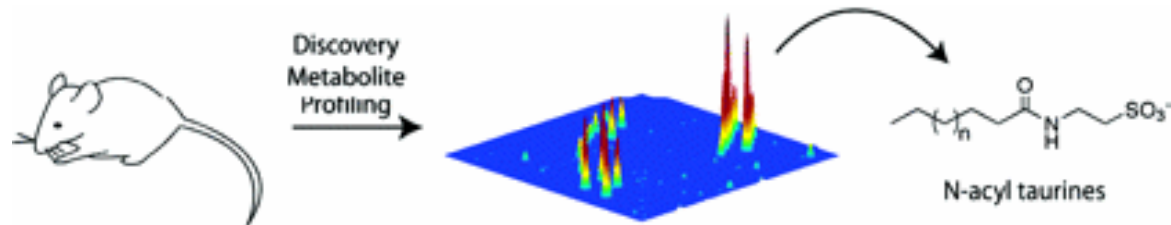
BioCYC Database Collection

Includes MetaCyc



NIST: contains over 100,000 compounds (<http://www.nist.gov/srd/nist1a.htm>)

Identification of Novel Brain Lipids



Identification of enzyme substrates by untargeted LC-MS analysis of WT and KO mouse brain

Brain lipids regulated by FAAH *in vivo*:

- known signalling molecules - anandamide
- novel family of taurine-conjugated fatty acids

14332

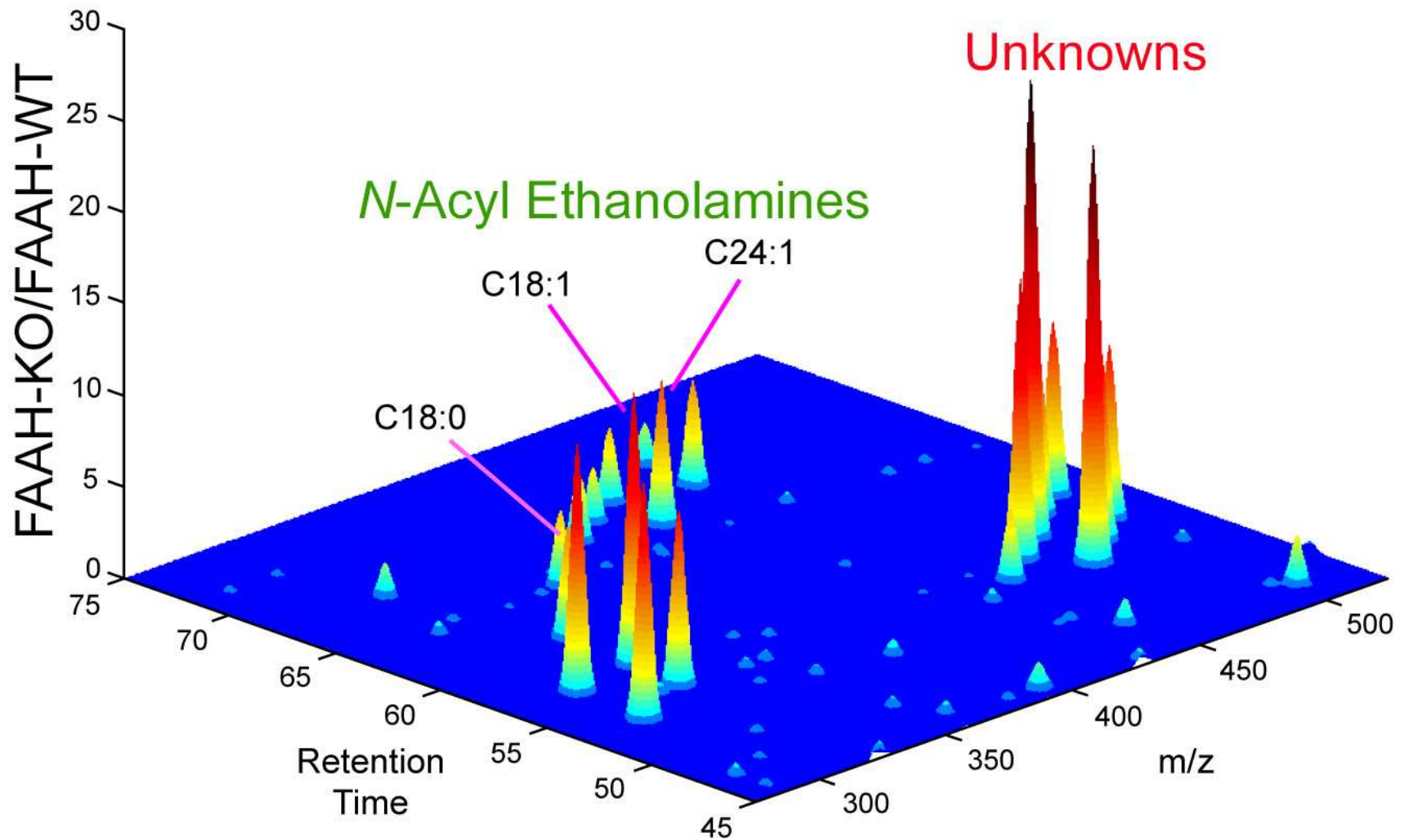
Biochemistry 2004, 43, 14332–14339

Assignment of Endogenous Substrates to Enzymes by Global Metabolite Profiling[†]

Alan Saghatelian, Sunia A. Trauger, Elizabeth J. Want, Edward G. Hawkins, Gary Siuzdak, and Benjamin F. Cravatt*

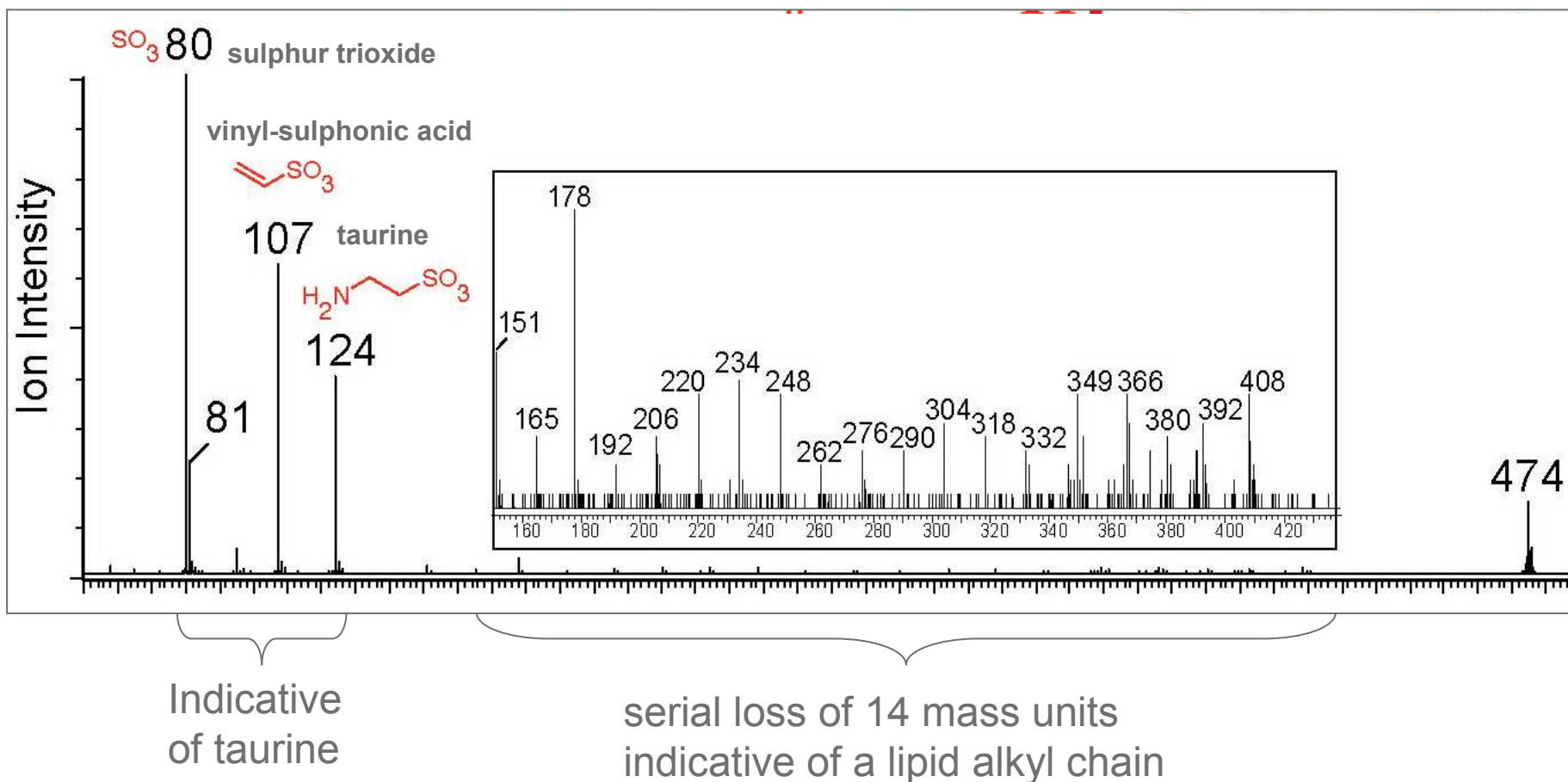
Unknown Brain Lipids

Regulated by Fatty Acid Amide Hydrolase (FAAH)



MS/MS: Structural Information

Highly related fragmentation patterns for the unknown metabolites

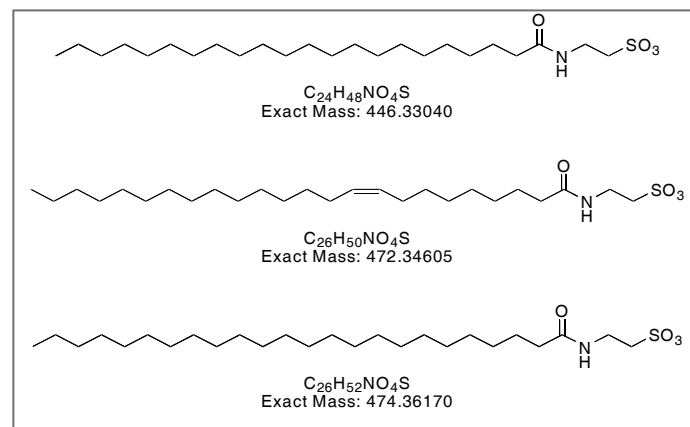


FTMS: Accurate Mass

446.3310

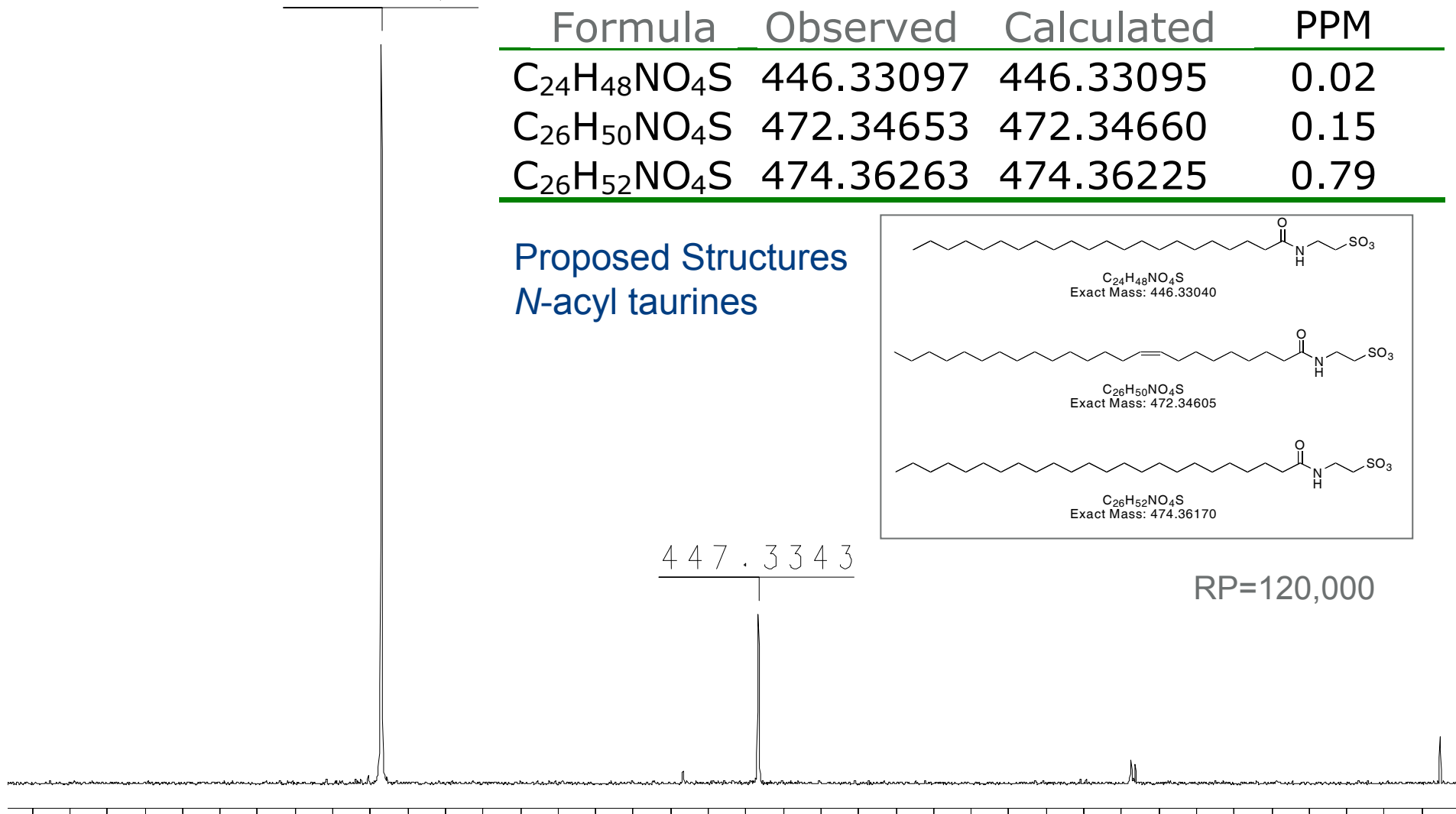
Molecular Formula	m/z Observed	m/z Calculated	PPM
C ₂₄ H ₄₈ NO ₄ S	446.33097	446.33095	0.02
C ₂₆ H ₅₀ NO ₄ S	472.34653	472.34660	0.15
C ₂₆ H ₅₂ NO ₄ S	474.36263	474.36225	0.79

Proposed Structures
N-acyl taurines



447.3343

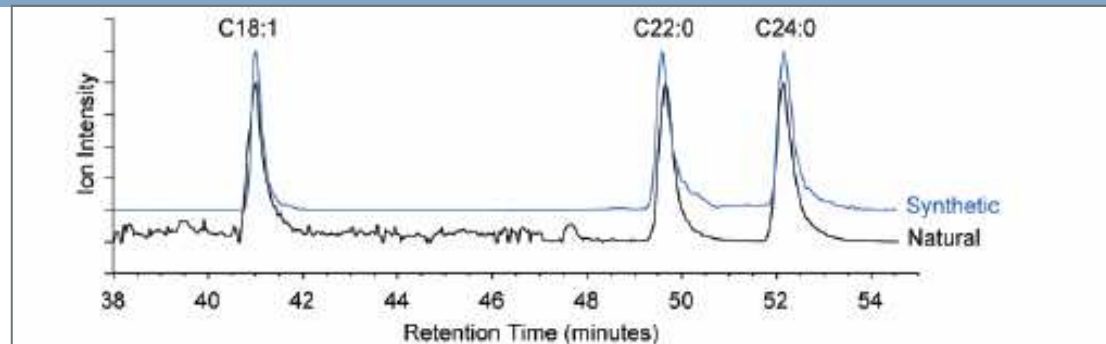
RP=120,000



Confirmation of Identification

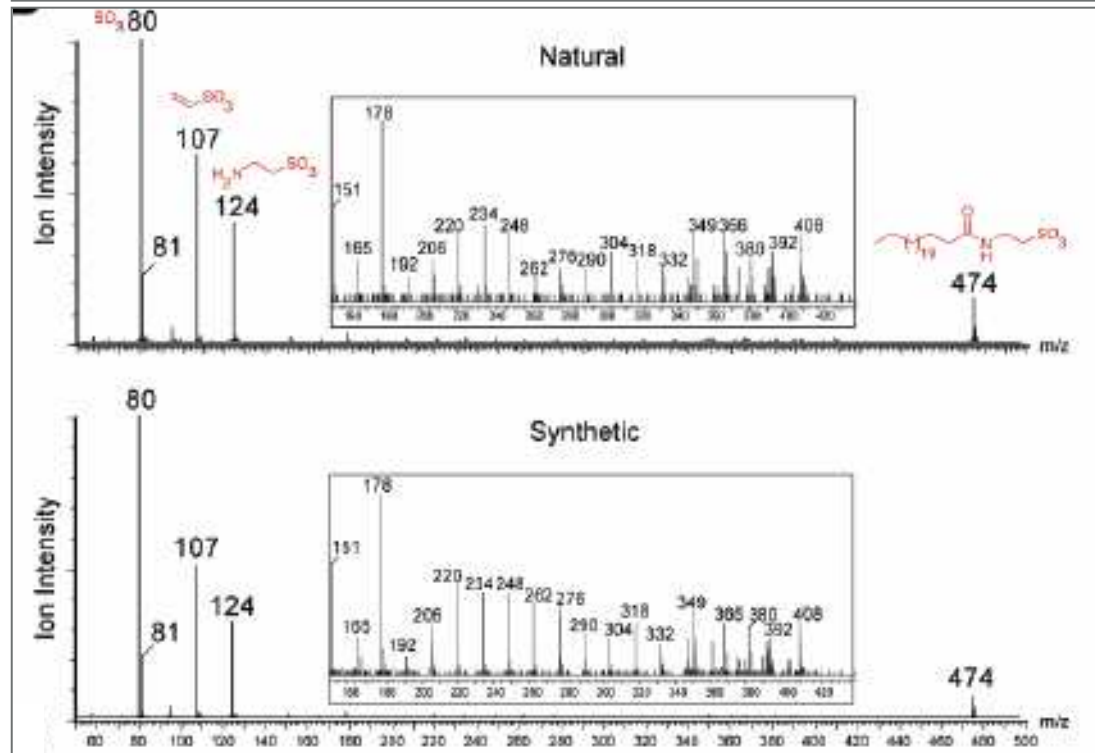
LC-MS

Co-migration of natural and synthetic NATs

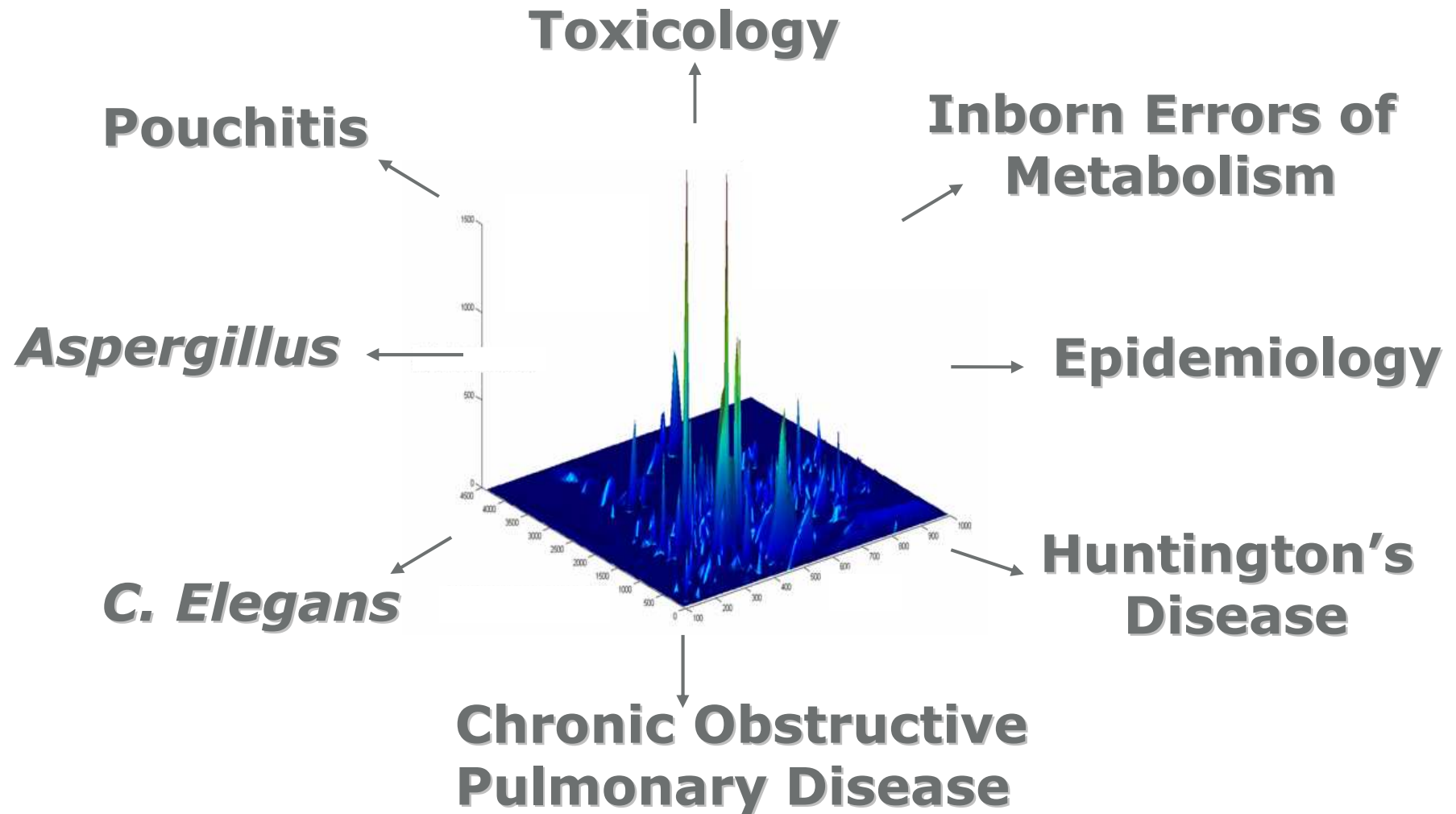


MS/MS

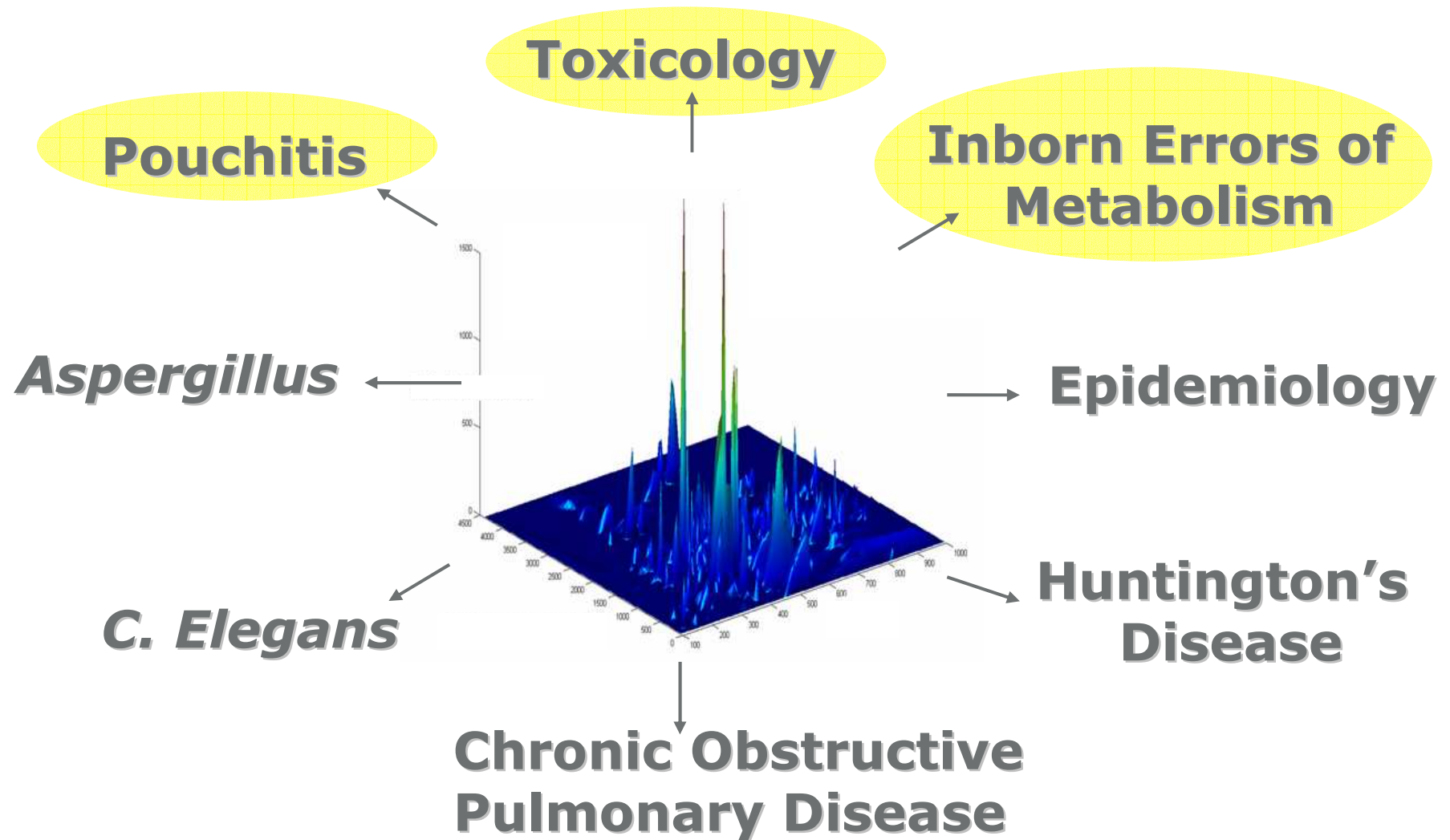
Spectrum of endogenous metabolite matched the C24:0 NAT standard



MS Studies in Biomolecular Medicine



MS Studies in Biomolecular Medicine



1) Toxicology: COMET 2 Project

Pfizer Global R&D

Bristol-Myers Squibb

Servier

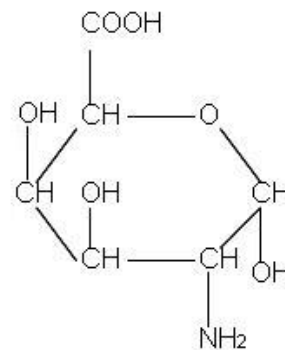
Sanofi-Aventis

Waters

NMR and UPLC-MS Analyses

Galactosamine (galN)

Model hepatotoxin



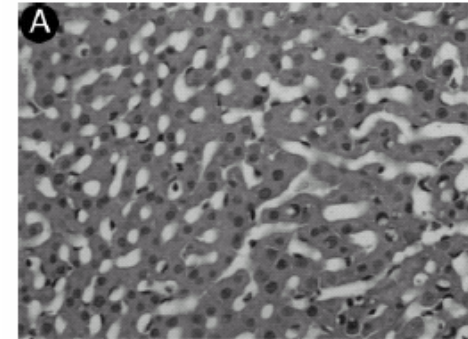
Bromoethanamine (BEA)

Model nephrotoxin

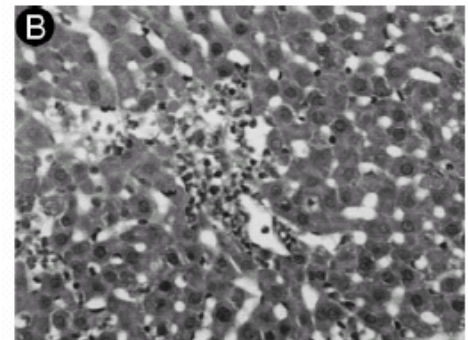


GaIN-induced hepatotoxicity

- **Selective hepatotoxin: produces dose-dependent, reversible liver damage**
- **Morphologically and biochemically similar to human hepatitis**
- **Severity of the response to gaIN is often quite variable**
- **Mechanism not yet fully resolved:**
 - **depletion of uridine nucleotide levels which inhibits RNA and protein synthesis (uridine or precursors protect)**
 - **alters gut permeability and increases bacterial translocation leading to endotoxemia (co-administration of LPS increases toxic response)**
- **Glycine protects against liver damage**



Control



24 hr after GaIN

GaIN-induced hepatotoxicity

- Selective hepatotoxin: produces dose-dependent, reversible liver damage
- Morphologically and biochemically similar to human hepatitis

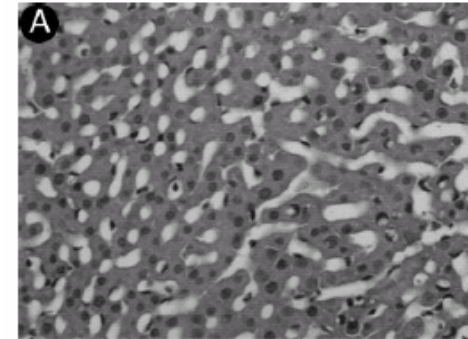
• Severity of the response to gaIN is often quite variable

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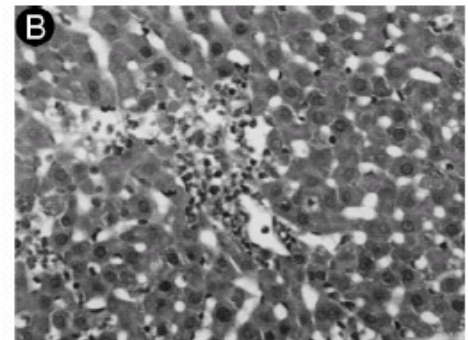
- depletion of uridine nucleotide levels which inhibits RNA and protein synthesis (uridine or precursors protect)
- alters gut permeability and increases bacterial translocation leading to endotoxemia (co-administration of LPS

increases toxic response)

• Glycine protects against liver damage



Control



24 hr after GaIN

UPLC-MS Strategy

Serum
methanol extraction



UPLC: HSS column (2.1 x 100mm, 1.7 μ m) gradient:
water + ACN 0.1% FA

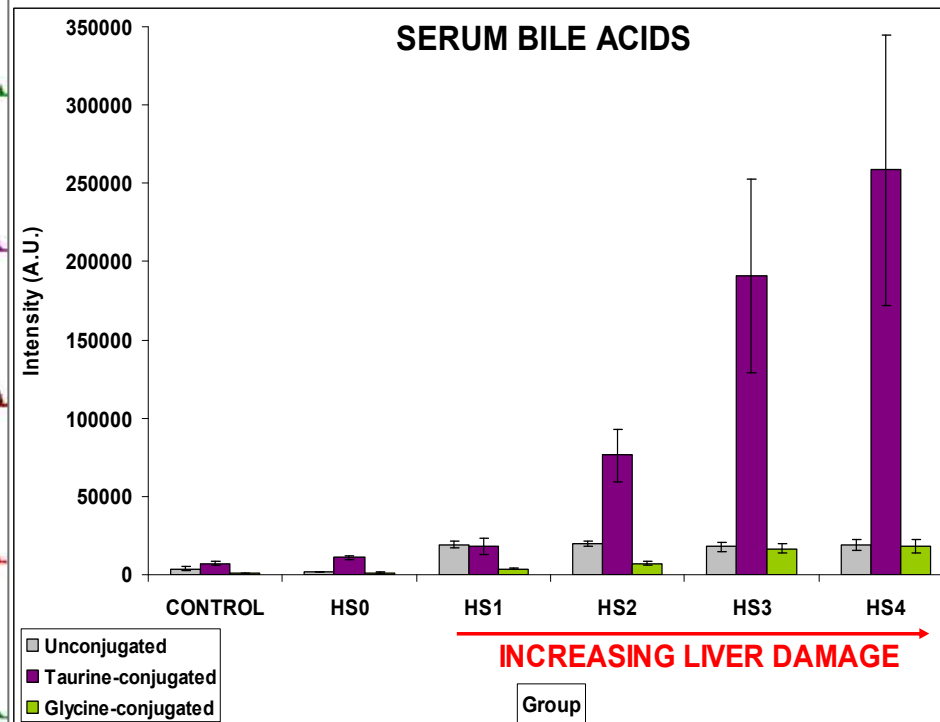
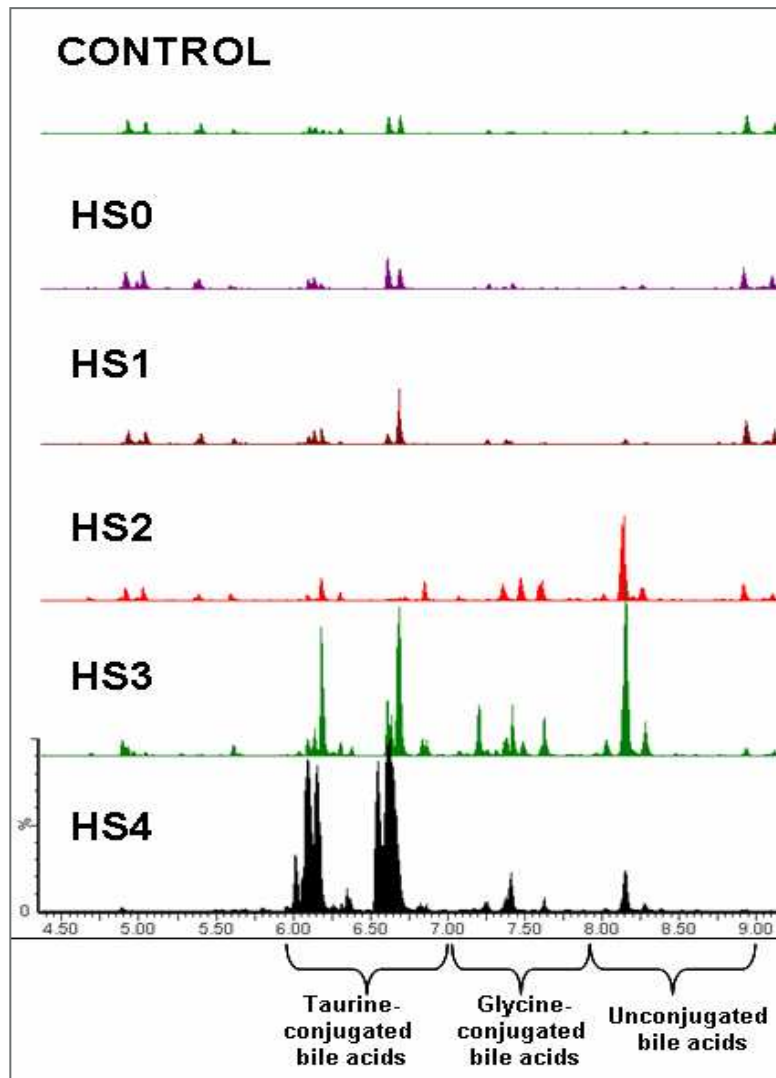


MS: Q-ToF Premier, Negative mode
ESI
50-1000 m/z , MS(E)



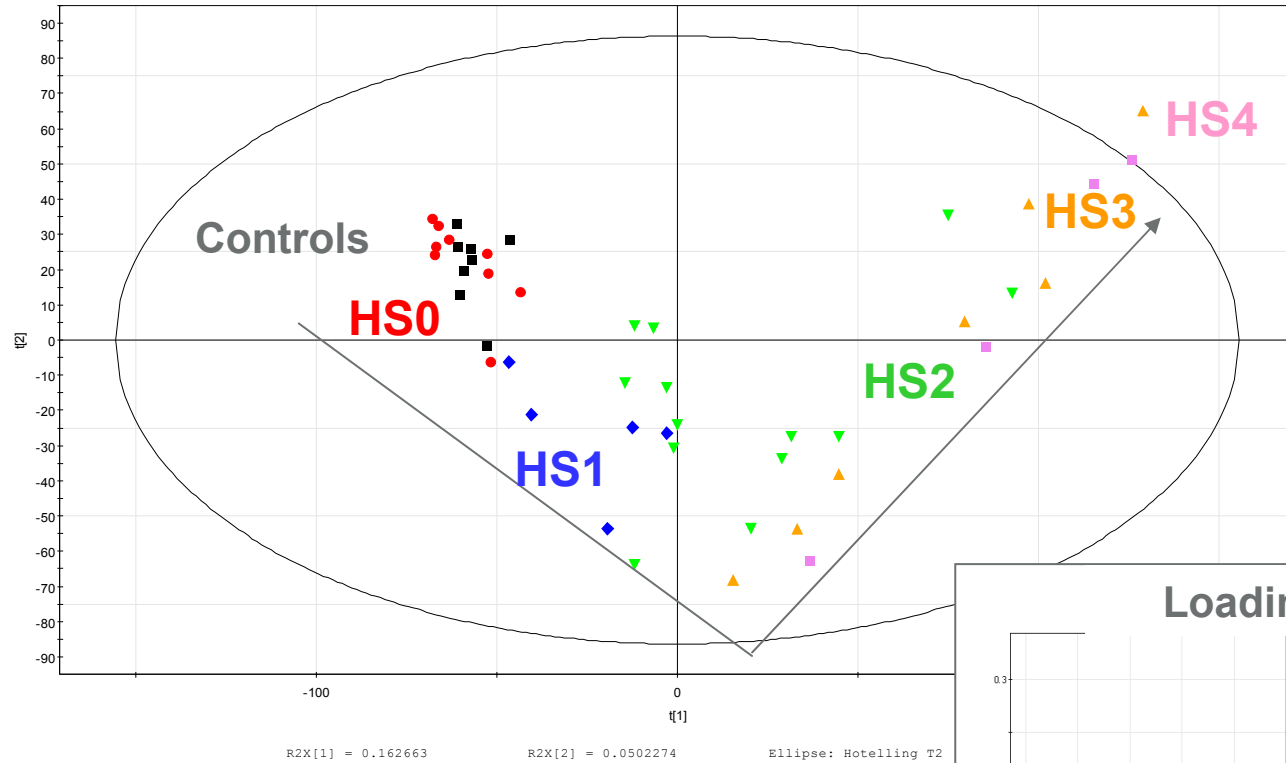
Data analysis
Peak picking and alignment
generation of intensity plots
generation of marker table
SIMCA-P – PCA

Intra-animal Variability

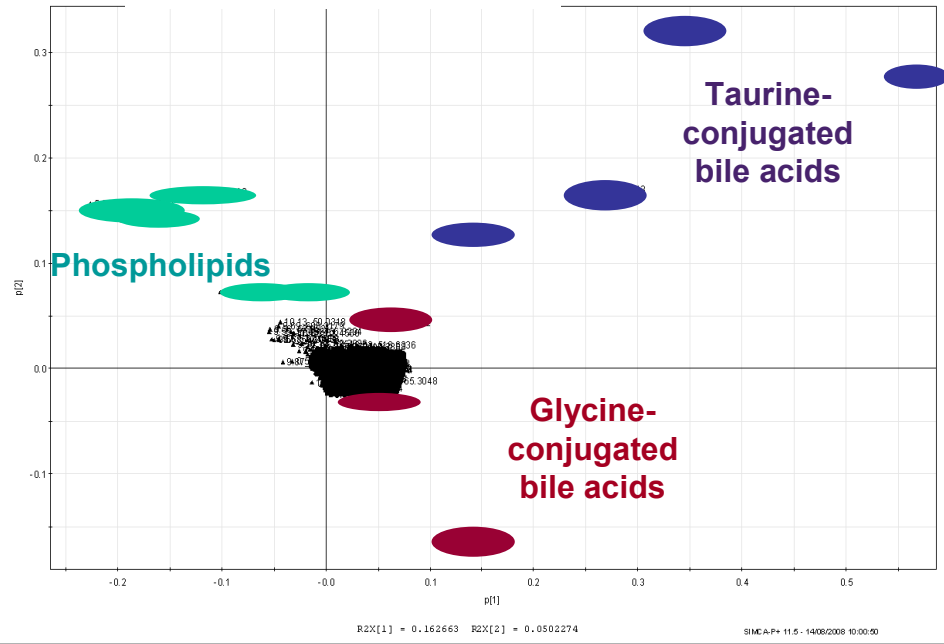


PCA Scores plot

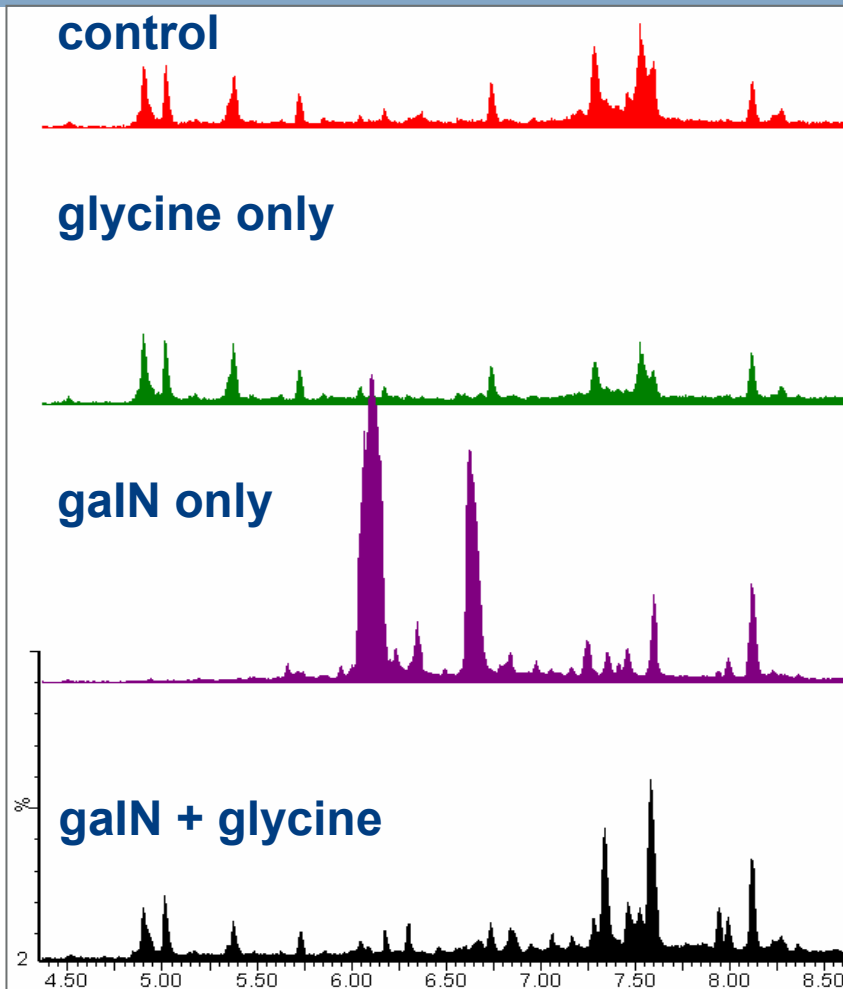
- Control
- HS0
- HS1
- HS2
- HS3
- HS4



Loadings plot



Protective Effect of Glycine

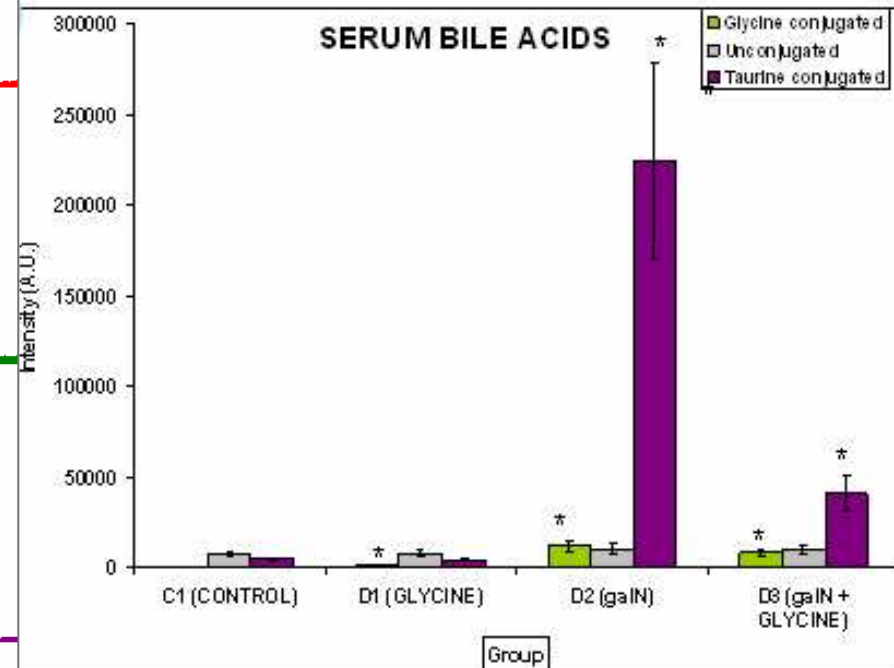


RT (mins)

Taurine-
conjugated
bile acids

Glycine-
conjugated
bile acids

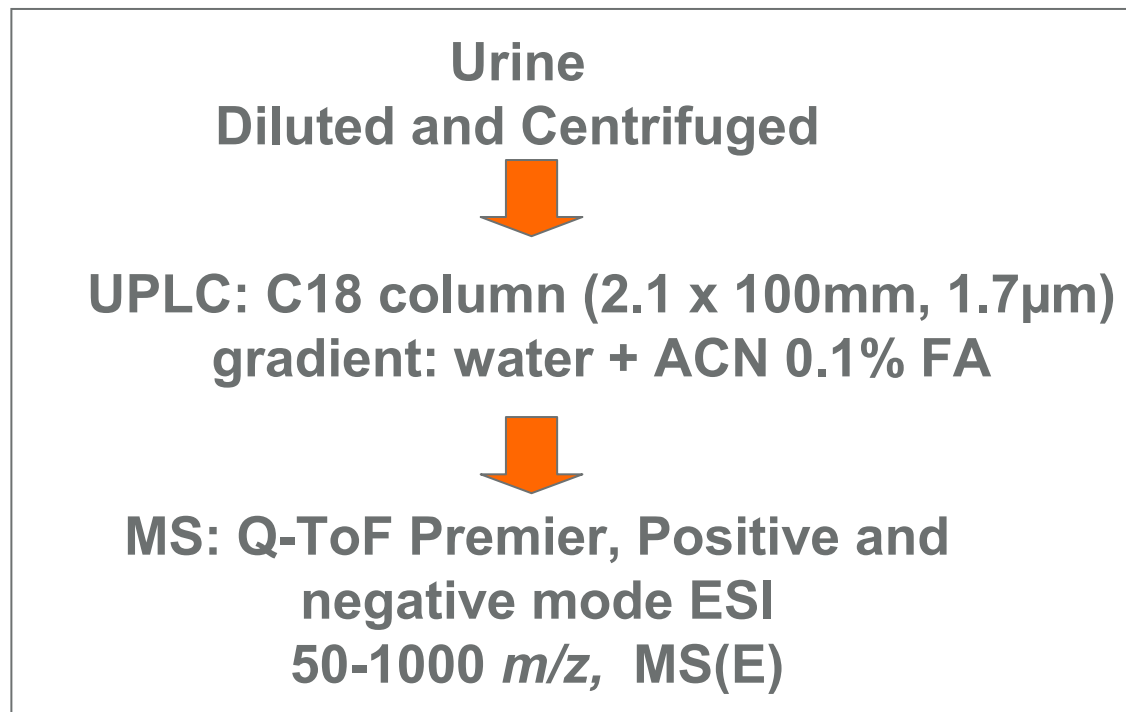
Unconjugated
bile acids



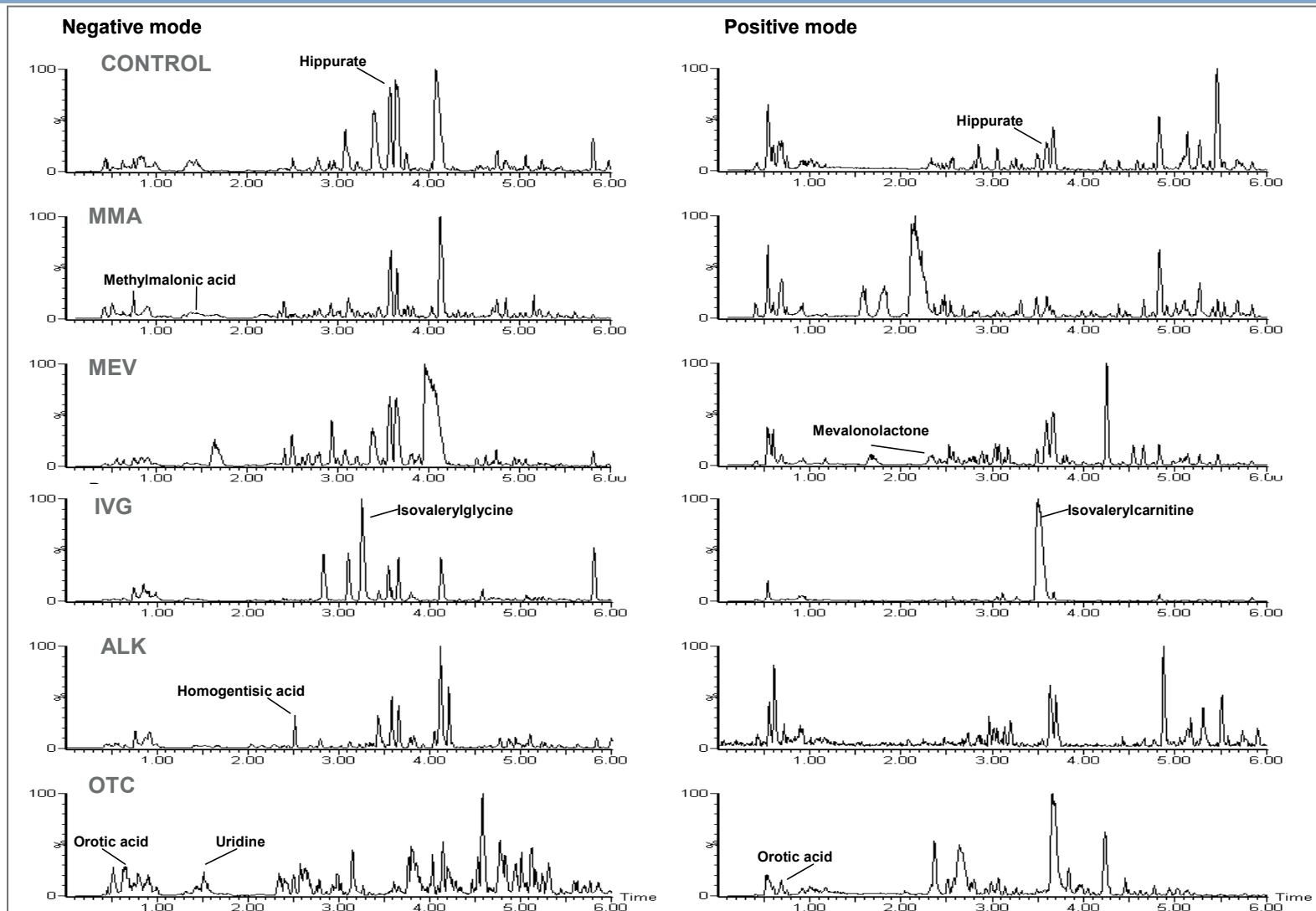
2) Organic Acidurias

AIM: to evaluate the applicability of UPLC-MS for the identification of organic acidurias

Urine samples were screened from patients with five different organic acidurias



Urinary Metabolite Profiles



3) Pouchitis

Ileo-anal pouches created in the management of some patients with ulcerative colitis

Proportion of these patients develop pouchitis

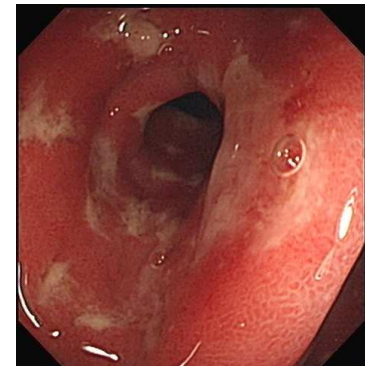
Inflammation of pouch lining

AIM: to determine cause of pouchitis

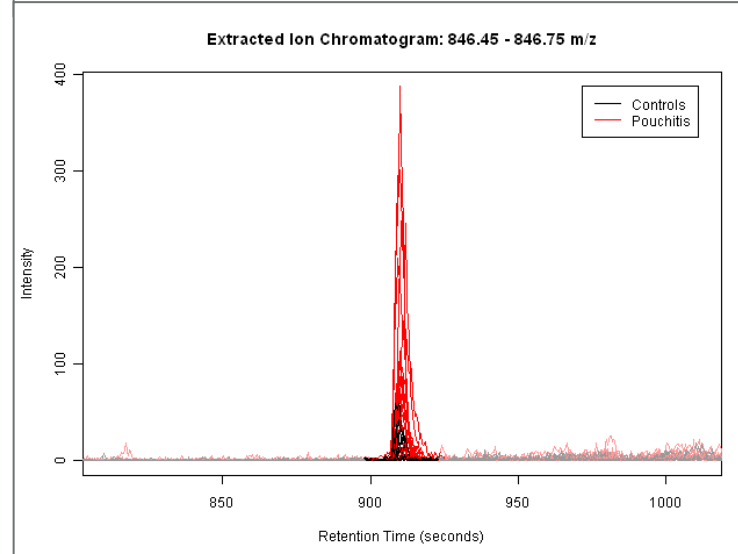
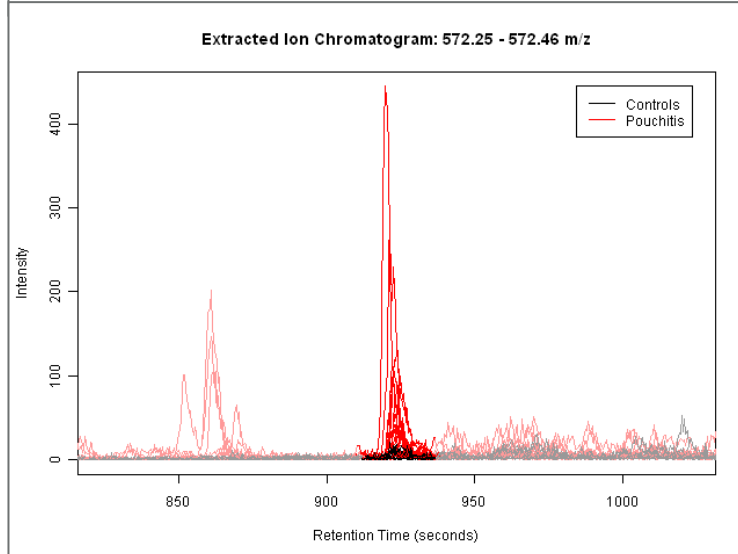
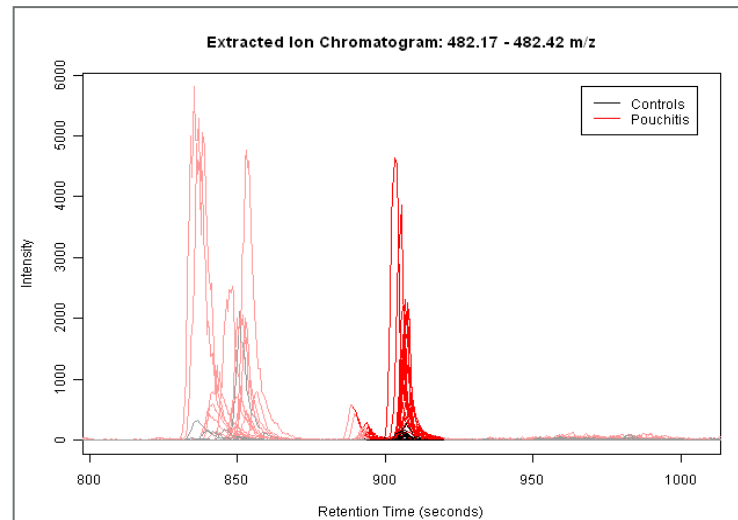
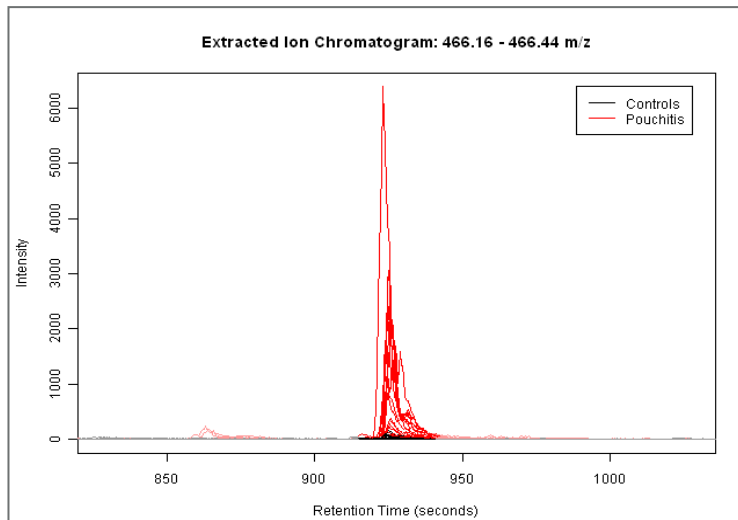
Metabonomic analysis of plasma, urine & faeces

NMR and UPLC-MS

UPLC-MS data processed using XCMS



Elevated Faecal Lipids in Pouchitis



Summary

Mass spectrometry-based metabonomics studies can offer

- Sensitivity
- Reproducibility
- Sample throughput
- Complementary information to NMR

Strategy varies depending on question being asked

Need to consider

- Sample preparation
- Separation approaches
- Mass spectrometer
- Data analysis

Challenges still remain in data analysis

- Software & databases

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