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LA METABOLOMIQUE ET SES APPLICATIONS EN BIOLOGIE: POURQUOI, COMMENT, BUT ET PERSPECTIVES



Le mot métabolisme provient du grec “μεταβολισμός” (metabo-lismo), qui signifie “changement.”



Concept du métabolisme par Ibn al-Nafis (1213–1288):

« Le corps et ses parties sont en état perpétuel de dissolution et de reconstitution, de sorte qu’ils subissent inévitablement un changement permanent »

Quelques définitions.....

Métabolite

Est une molécule organique <1500 Da détectable dans un organisme, incluant des molécules d'origine endogènes ou exogènes, issus des micro-organismes

Le Métabolome

Est un terme utilisé pour la première fois par Olivier et al en 1998 pour décrire un **ensemble de métabolites** synthétisés par un organisme

La Métabolomique

A été formulée par Oliver Fiehn et est définie par une **analyse complète non sélective** dans laquelle tous les métabolites d'un système biologique sont identifiés et quantifiés

La Métabonomique

Est définie par la **mesure quantitative de la réponse métabolique** multiparamétrique des systèmes vivants à un stimuli ou une altération génétique (Jeremy Nicholson)

Profilage des métabolites

Implique l'identification et la quantification d'un **ensemble ou d'une classe prédéfinis** de métabolites d'identités connus ou inconnus et appartenant à des voies métaboliques particulières.

Le métabolome représente l'ultime réponse d'un organisme à une altération génétique, une pathologie, une exposition à un toxique ou toute cause environnementale. Le métabolome d'un système peut ainsi à la fois permettre de lire la signature biologique d'une réponse adaptative ou pathologique, et également être le vecteur de cette réponse. Ce dernier aspect souligne l'implication directe du métabolome dans le déterminisme des phénotypes

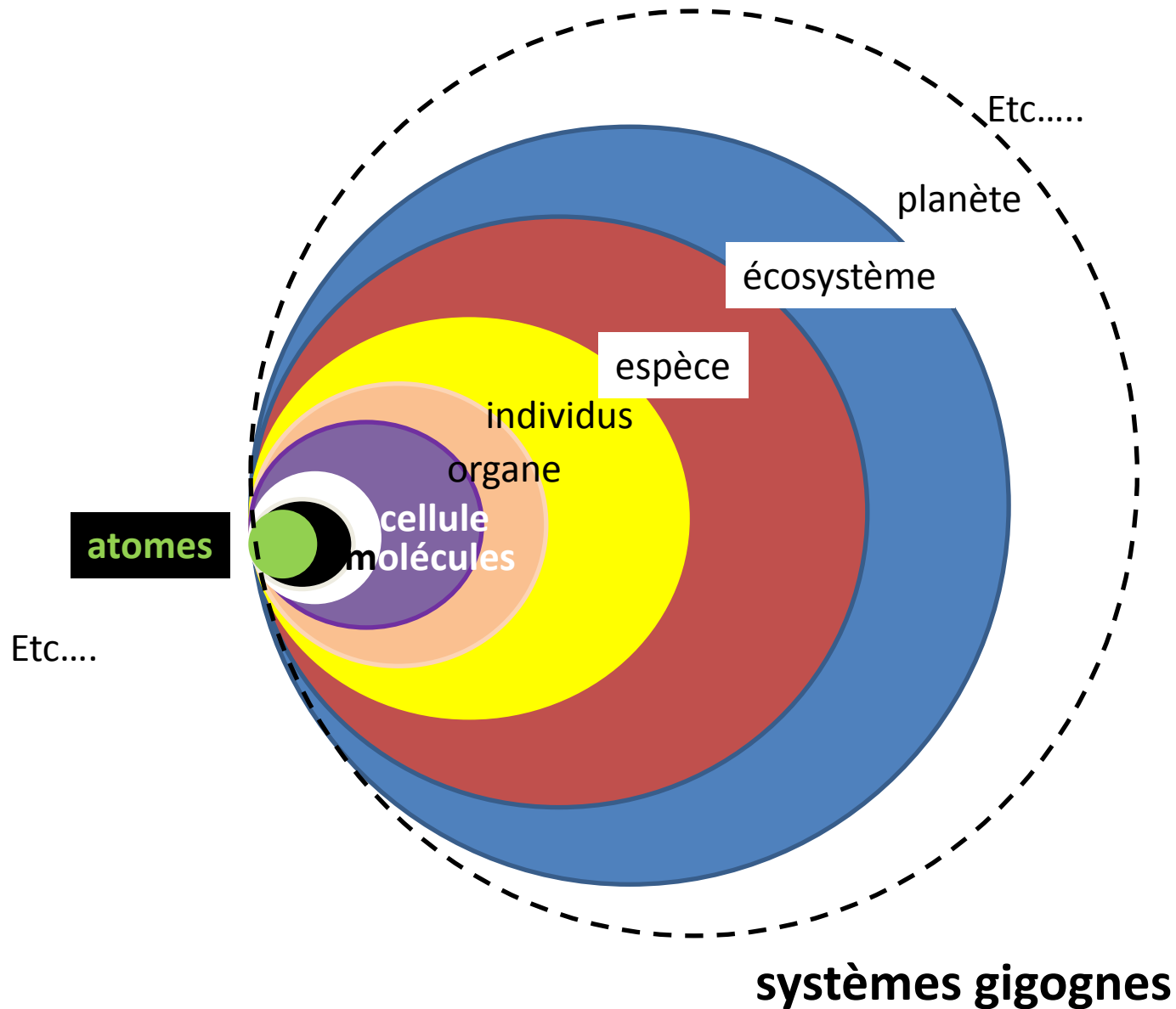
OPINION

2020 visions

For the first issue of the new decade, *Nature* asked a selection of leading researchers and policy-makers where their fields will be ten years from now. We invited them to identify the key questions their disciplines face, the major roadblocks and the pressing next steps. Visit go.nature.com/htW8uM to respond and to add your vision.

Microbiome Energy Synthetic
Perso **Metabolomics**
medicine Astronomy
Drug discovery Chemistry palaeontology
Lasers

Les petites molécules sont présentes dans tous les systèmes...



Le Metabolome est Connecté à tous les autres “Omes”

- Small molecules (i.e. AMP, CMP, GMP, TMP) are the primary constituents of the genome & transcriptome
- Small molecules (i.e. the 20 amino acids) are the primary constituents of the proteome
- Small molecules (i.e. lipids) give cells their shape, form, integrity and structure
- Small molecules (sugars, lipids, AAs, ATP) are the source of all cellular energy
- Small molecules serve as cofactors and signaling molecules for both the proteome and the genome
- *The genome & proteome largely evolved to catalyze the chemistry of small molecules*

Qing-zhao Wang · Chan-yuan Wu · Tao Chen ·
Xun Chen · Xue-ming Zhao

Integrating metabolomics into systems biology framework to exploit metabolic complexity: strategies and applications in microorganisms

Appl Microbiol Biotechnol (2006) 70: 151–161
DOI 10.1007/s00253-005-0277-2

Metabolomics tools for identifying biomarkers for neuropsychiatric diseases

Neurobiology of Disease 35 (2009) 165–176

Emerging high-throughput approaches to analyze bioremediation of sites contaminated with hazardous and/or recalcitrant wastes

Biotechnology Advances 26 (2008) 561–575

High-Throughput Technique for Comprehensive Analysis of Japanese Green Tea Quality Assessment Using Ultra-performance Liquid Chromatography with Time-of-Flight Mass Spectrometry (UPLC/TOF MS)

J. Agric. Food Chem. 2008, 56, 10706–10708

Metabolomics in monitoring kidney transplants

Wishart, David S

Current Opinion in Nephrology & Hypertension:

November 2006 - Volume 15 - Issue 6 - p 637-642

Metabonomics in pharmaceutical R & D

John C. Lindon, Elaine Holmes and Jeremy K. Nicholson

FEBS Journal 274 (2007) 1140–1151

Gut flora metabolism of phosphatidylcholine promotes cardiovascular disease

Zeneng Wang^{1,2}, Elizabeth Klipfell^{1,2}, Brian J. Bennett³, Robert Koeth¹, Bruce S. Levison^{1,2}, Brandon DuGar¹, Ariel E. Feldstein^{1,2}, Earl B. Britt^{1,2}, Xiaoming Fu^{1,2}, Yoon-Mi Chung^{1,2}, Yuping Wu⁴, Phil Schauer⁵, Jonathan D. Smith^{1,6}, Hooman Allayee⁷, W. H. Wilson Tang^{1,2,6}, Joseph A. DiDonato^{1,2}, Aldons J. Lusis³ & Stanley L. Hazen^{1,2,6}

7 APRIL 2011 | VOL 472 | NATURE | 57

Physiologia Plantarum 132: 162–175. 2008

Plant metabolomics and its potential application for human nutrition

Robert D. Hall^{a,b,*}, Inge D. Brouwer^c and Melissa A. Fitzgerald^d

Metabolomics in human nutrition: opportunities and challenges

Am J Clin Nutr 2005;82:497–503

Applications of Metabolomics in Agriculture

RICHARD A. DIXON,[†] DAVID R. GANG,[‡] ADRIAN J. CHARLTON,[§] OLIVER FIEHN,[#]

HARRY A. KUIPER,^{||} TRACEY L. REYNOLDS,[⊥] RONALD S. TJEERDEMA,[⊗]

ELIZABETH H. JEFFERY,[△] J. BRUCE GERMAN,[×] WILLIAM P. RIDLEY,^{*,⊥} AND

JAMES N. SEIBER[◇]

8984 *J. Agric. Food Chem.* 2006, 54, 8984–8994

Metabonomic modeling of drug toxicity

Hector C. Keun^{*}

Pharmacology & Therapeutics 109 (2006) 92 – 106

Regulation of floral scent production in petunia revealed by targeted metabolomics

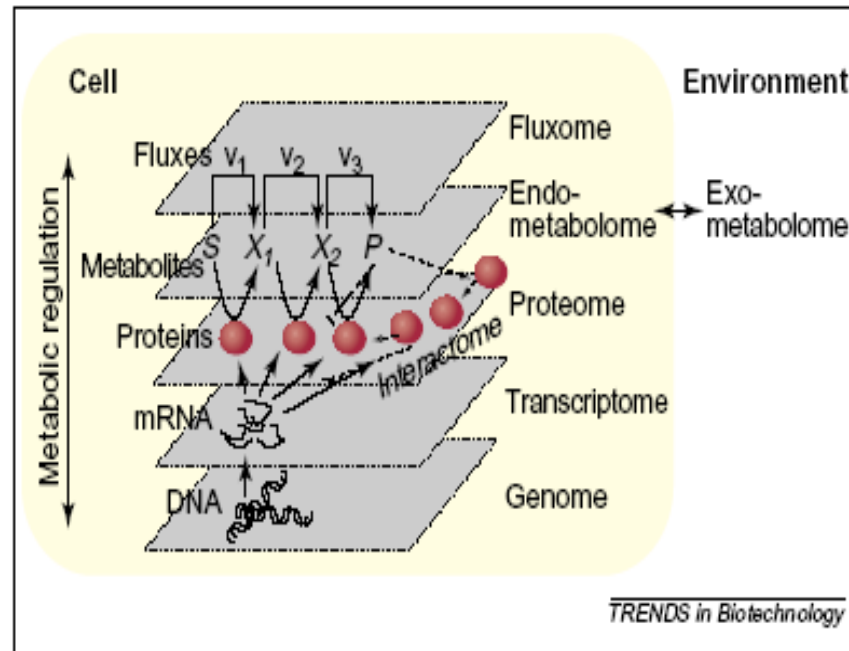
Phytochemistry 62 (2003) 997–1008

Metabolomic Profiling for Identification of Novel Potential Biomarkers in Cardiovascular Diseases

Journal of Biomedicine and Biotechnology

Volume 2011, Article ID 790132, 9 pages

Comment j'en suis arrivé à la métabolomique... et autres omiques



contexte pro-athérogène



Hamsters adultes

Régime CROQUETTES

+ 20% lipides (0,12% cholestérol)

12 semaines

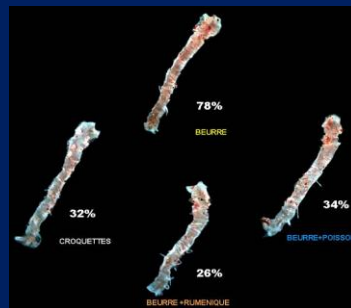
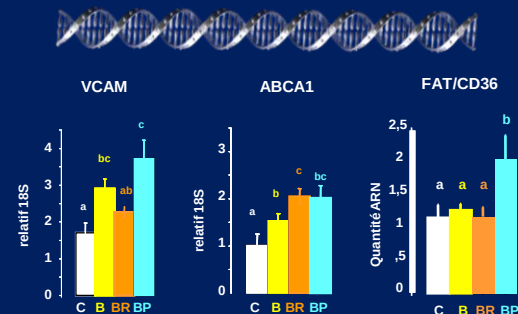
Croquettes

Beurre
(rum énique: 0,1%)

Beurre
+
rum énique
(qsp 1%)

Beurre
+
Huile de Poisson
(1%)

les stries lipidiques (dépôts d'esters de cholestérol) **marqueurs circulants** (lipoprotéines, paraoxonase, apo SAA) **expression gènes cibles** (VCAM-1, ABCA-1, COX2, cytokines, PPARs,...)



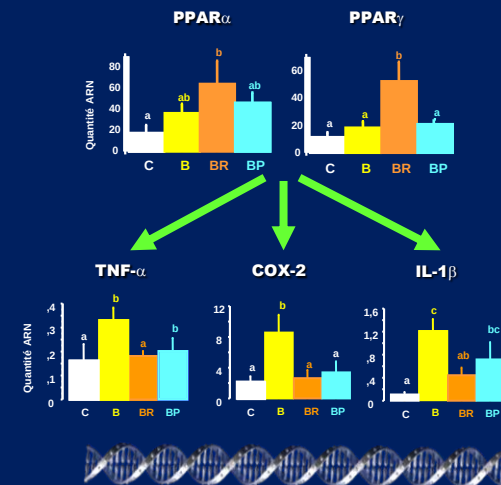
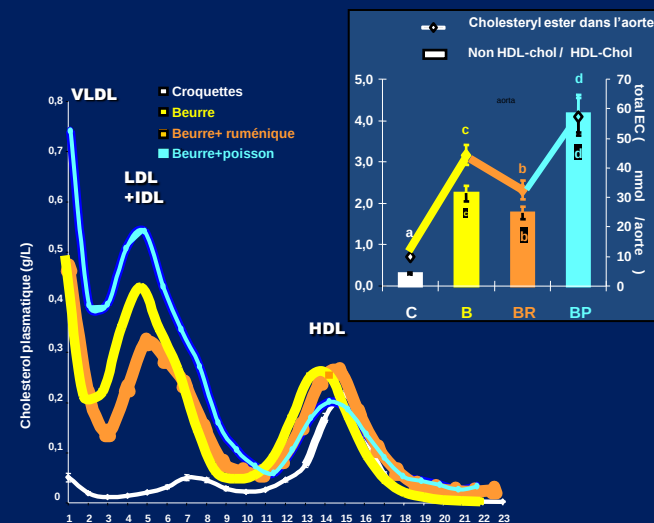
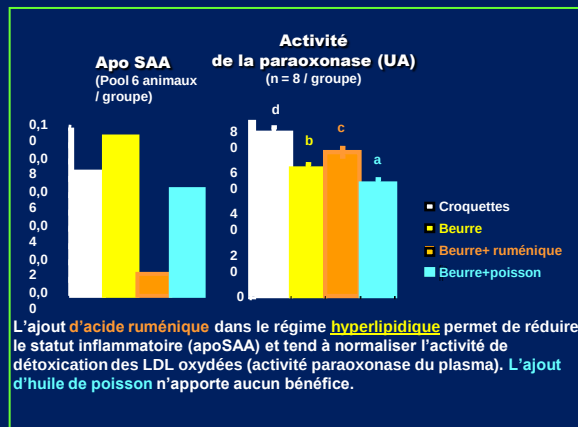
ATHEROSCLEROSE

RECRUTEMENT MONOCYTES
(VCAM)
CAPTURE LIPIDES
(FAT/CD36)

INFLAMMATION
LOCALE: $TNF\alpha$, COX-2, IL1 β
SYSTEMIQUE: apoSAA

EFFLUX CHOLESTEROL
(ABCA1)
DETOXICATION LDLox
(PARAOXONASE)

LDL
VLDL
TRANSPORT CHOLESTEROL
HDL



Conclusions

L'acide ruménique dans la matière grasse laitière est anti-athérogène (hamster hyperlipidémique)

↘ importance des dépôts lipidiques (EC)



Systémique: ↘ nonHDL-CH/HDL-CH

Local: ↗ Efflux chol (ABCA1)

↘ Recrutement des monocytes

↘ inflammation

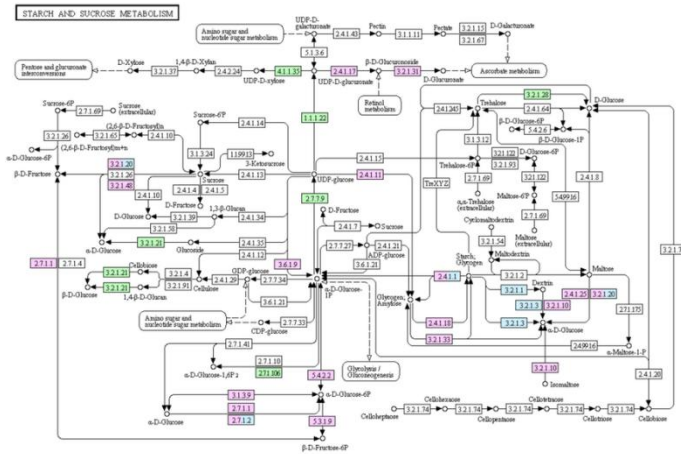
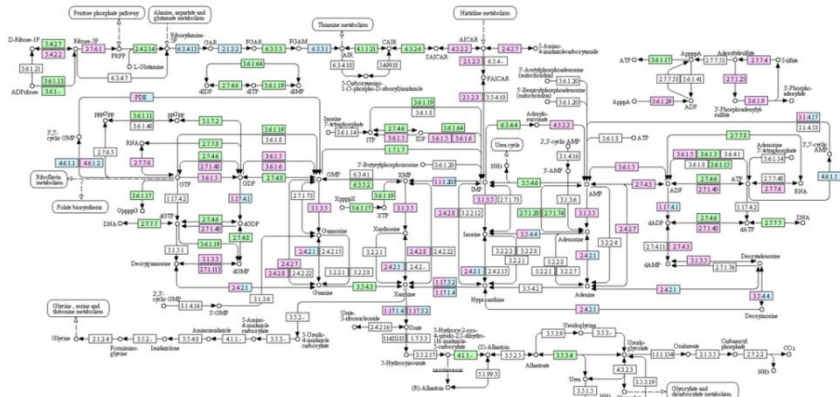
↗ Détoxification des oxLDL



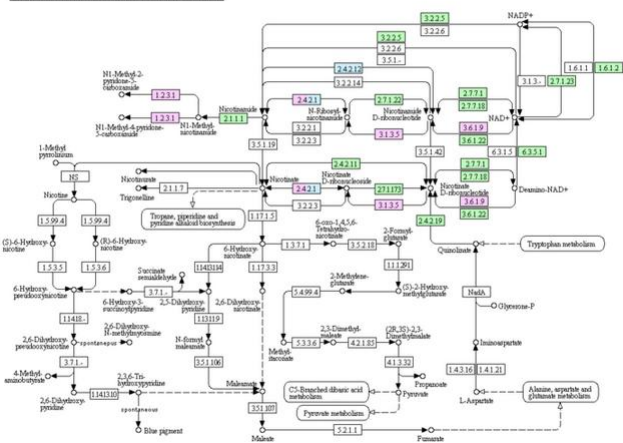
Est-ce suffisant?

Oui et non

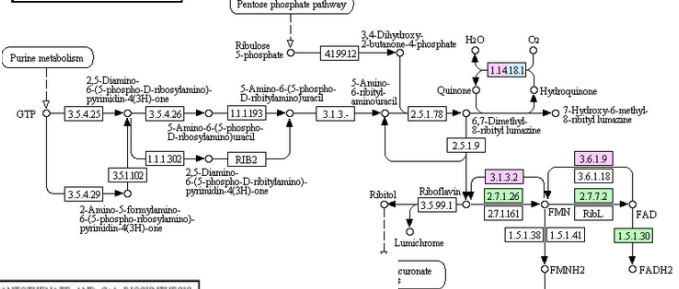
Mécanisme moléculaires de la calcification artérielle (KEGG [H01002](#))



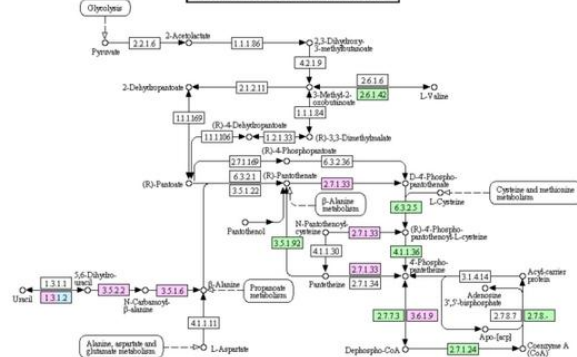
NICOTINATE AND NICOTINAMIDE METABOLISM



RIBOFLAVIN METABOLISM



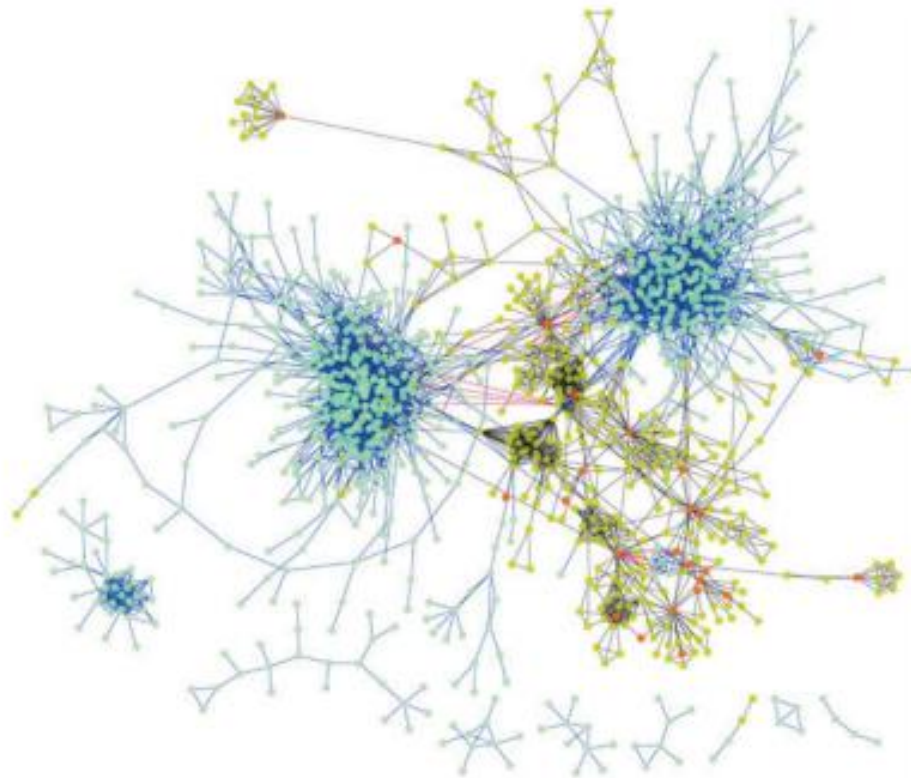
PANTOTHENATE AND CoA BIOSYNTHESIS



Porphyrin and chlorophyll metabolism

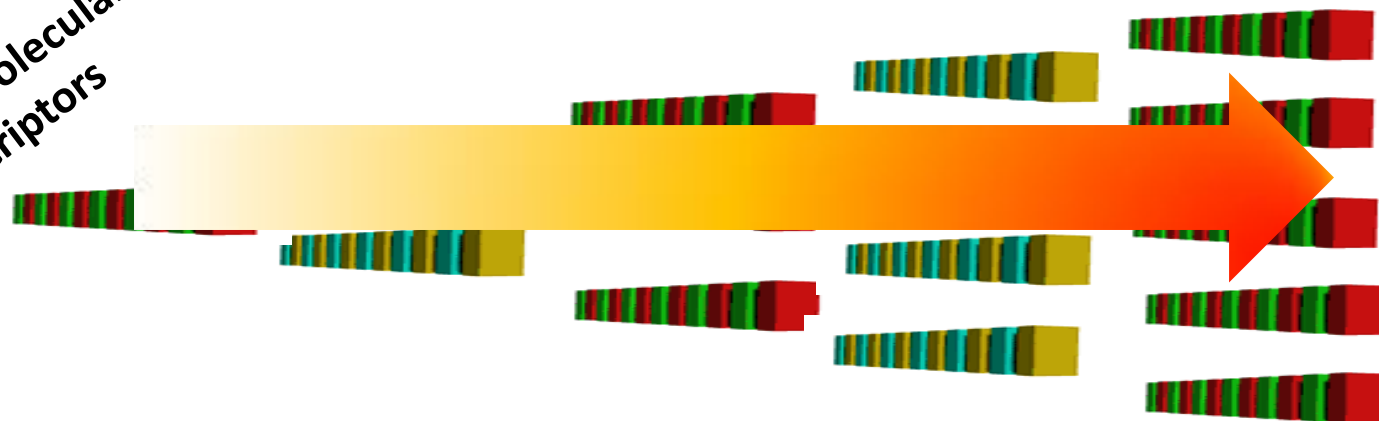


Réseau des gènes liés au sexe impliqués dans l'athérogénèse (composante sexuelle)
Reconstitution « in silico » partir des données de la littérature



Diez D, Wheelock AM, Goto S, et al. The use of network analyses for elucidating mechanisms in cardiovascular disease. Mol Biosyst 2010;6:289-304.

Limited molecular
descriptors



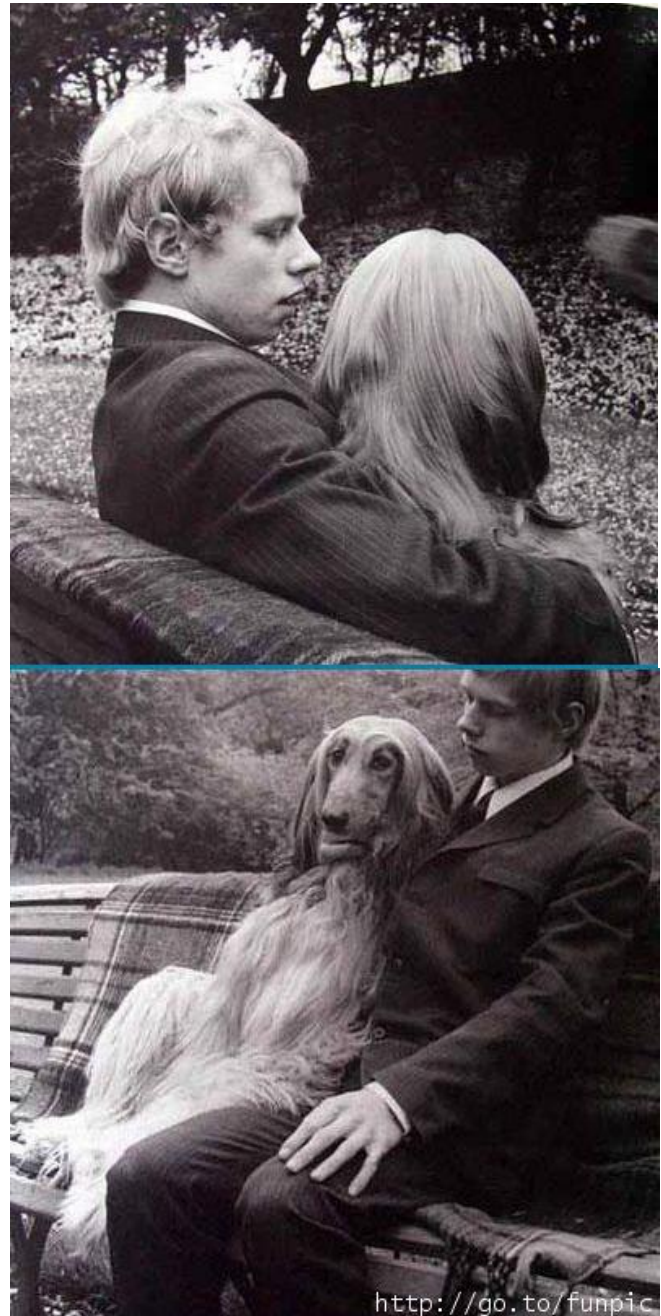
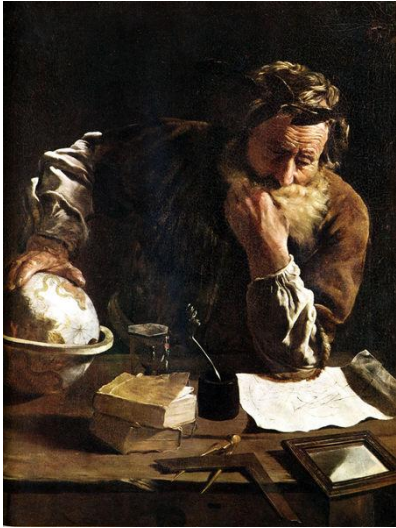
Complex
phenotypic traits

Blood lipids, CRP...

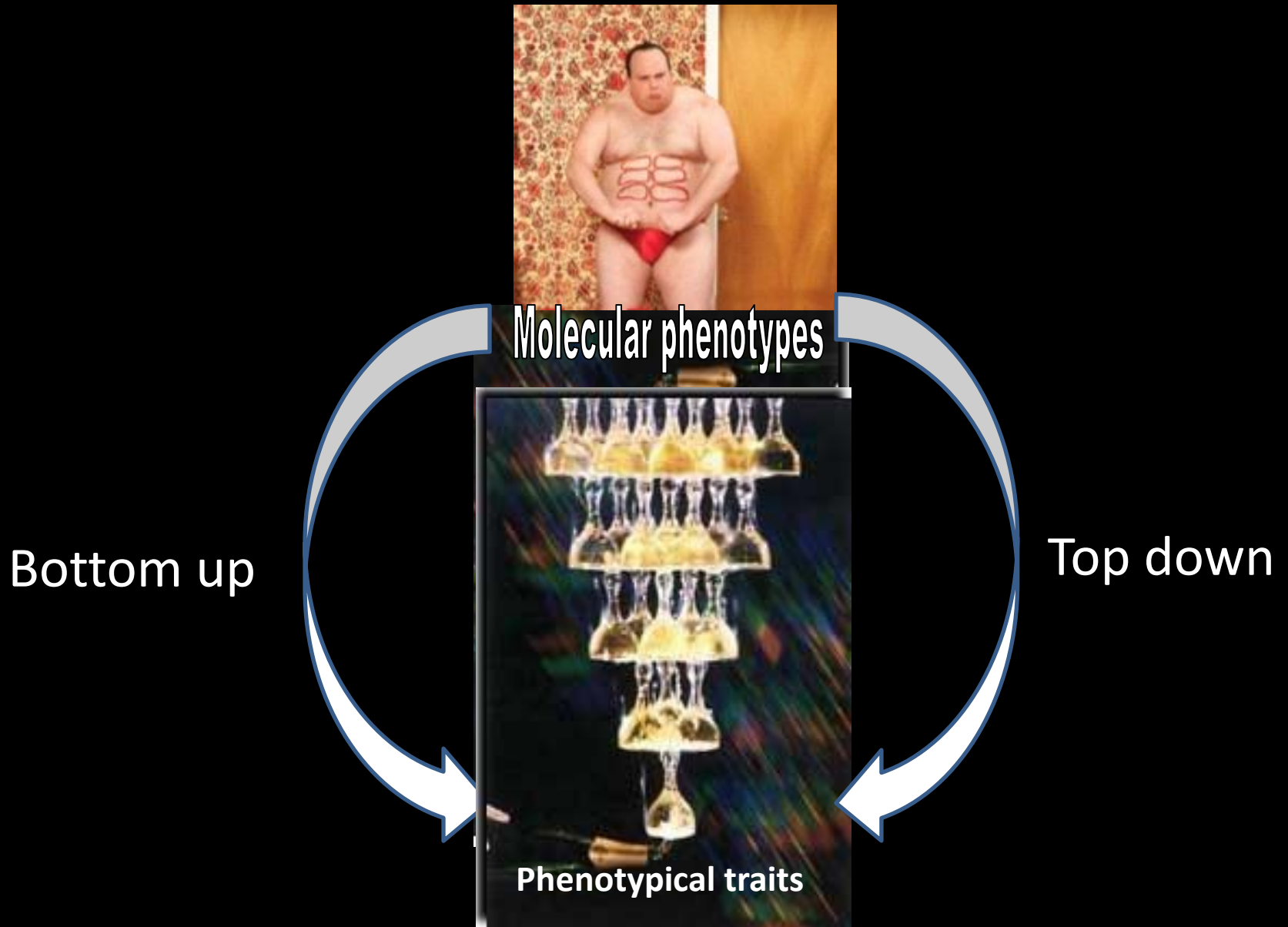
CVD, diabetes...

**« conventional » so-called bottom-up strategies:
explain complex phenotypes with only few targeted
molecular descriptors**

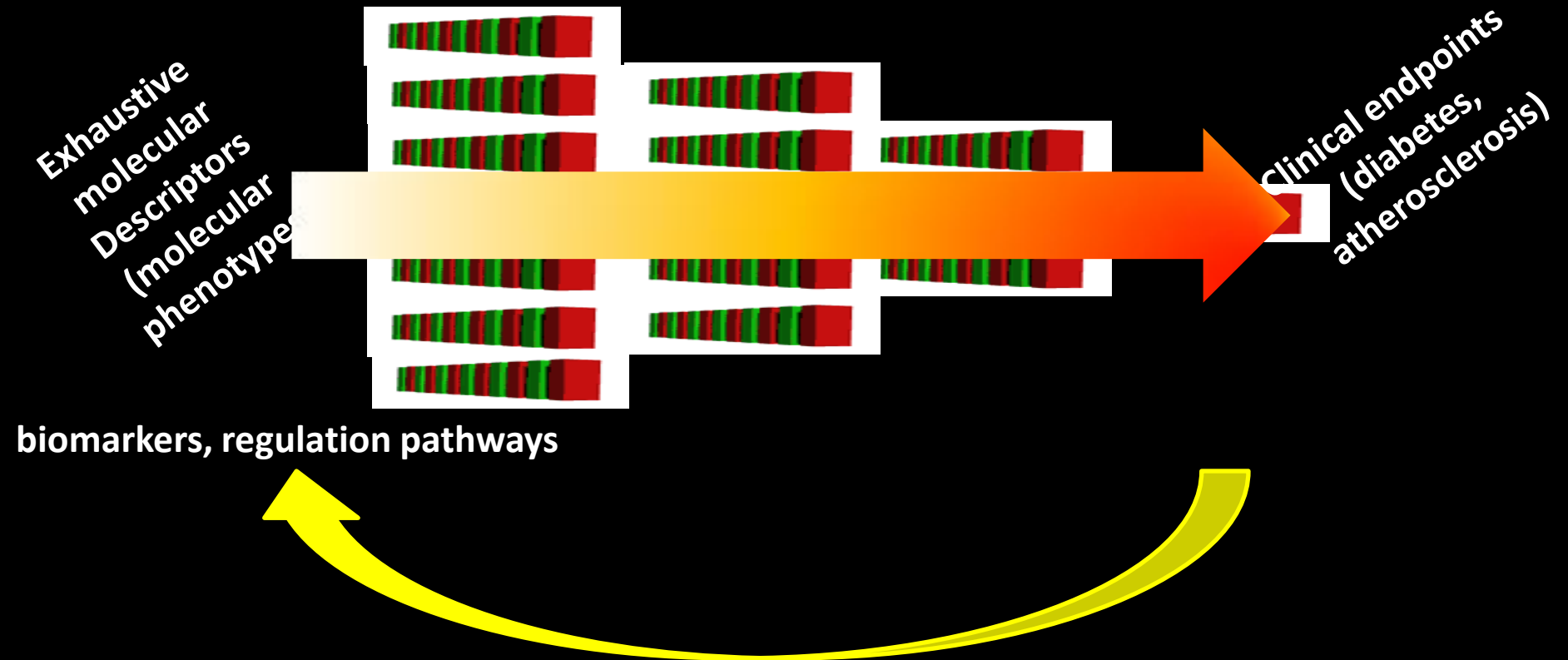
Reality is much complex
than anticipated.....

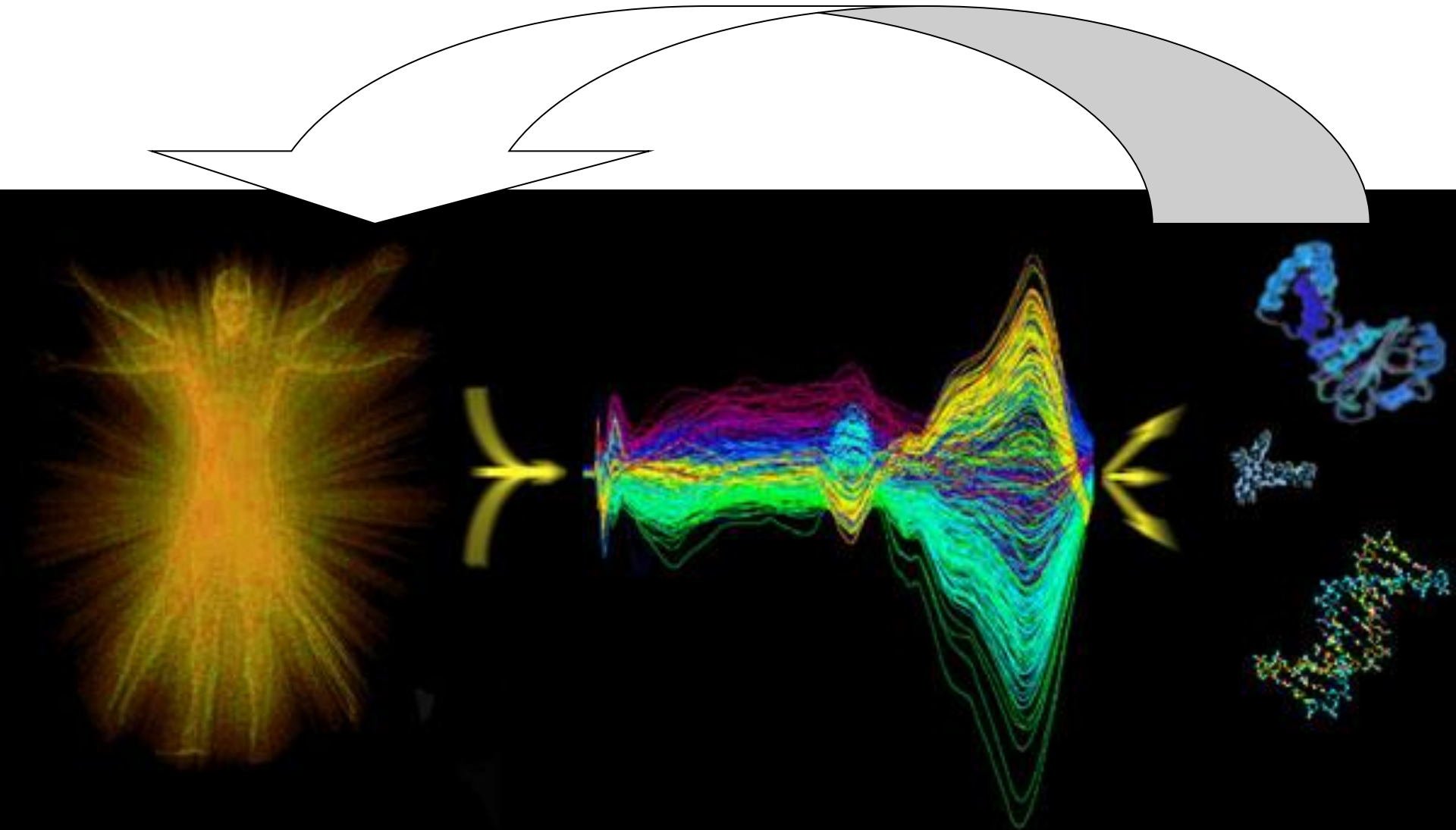


Need to address the molecular basis of phenotypes
the « top down » approach.....



The top-down approach



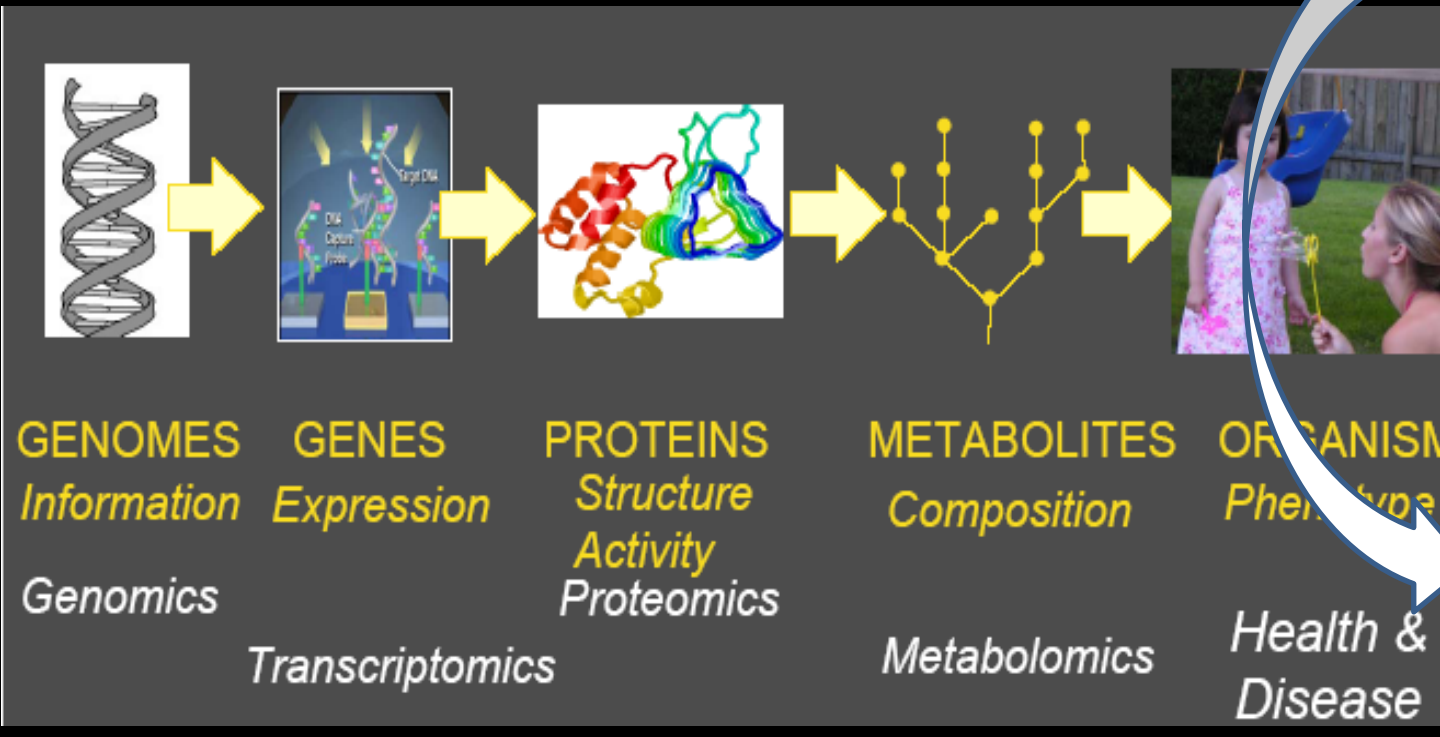


<http://masspec.scripps.edu/index.php>

Need to assess the molecular basis of phenotypes
the « omic» answer....



Molecular phenotypes



Contrairement au transcriptome ou au protéome, le métabolome représente l'ultime réponse d'un organisme à une altération génétique, une pathologie, une exposition à un toxique ou à tout autre facteur susceptible de perturber son fonctionnement.

La métabolomique (en santé) pour

la découverte de biomarqueurs et détecter l'insoupçonnable

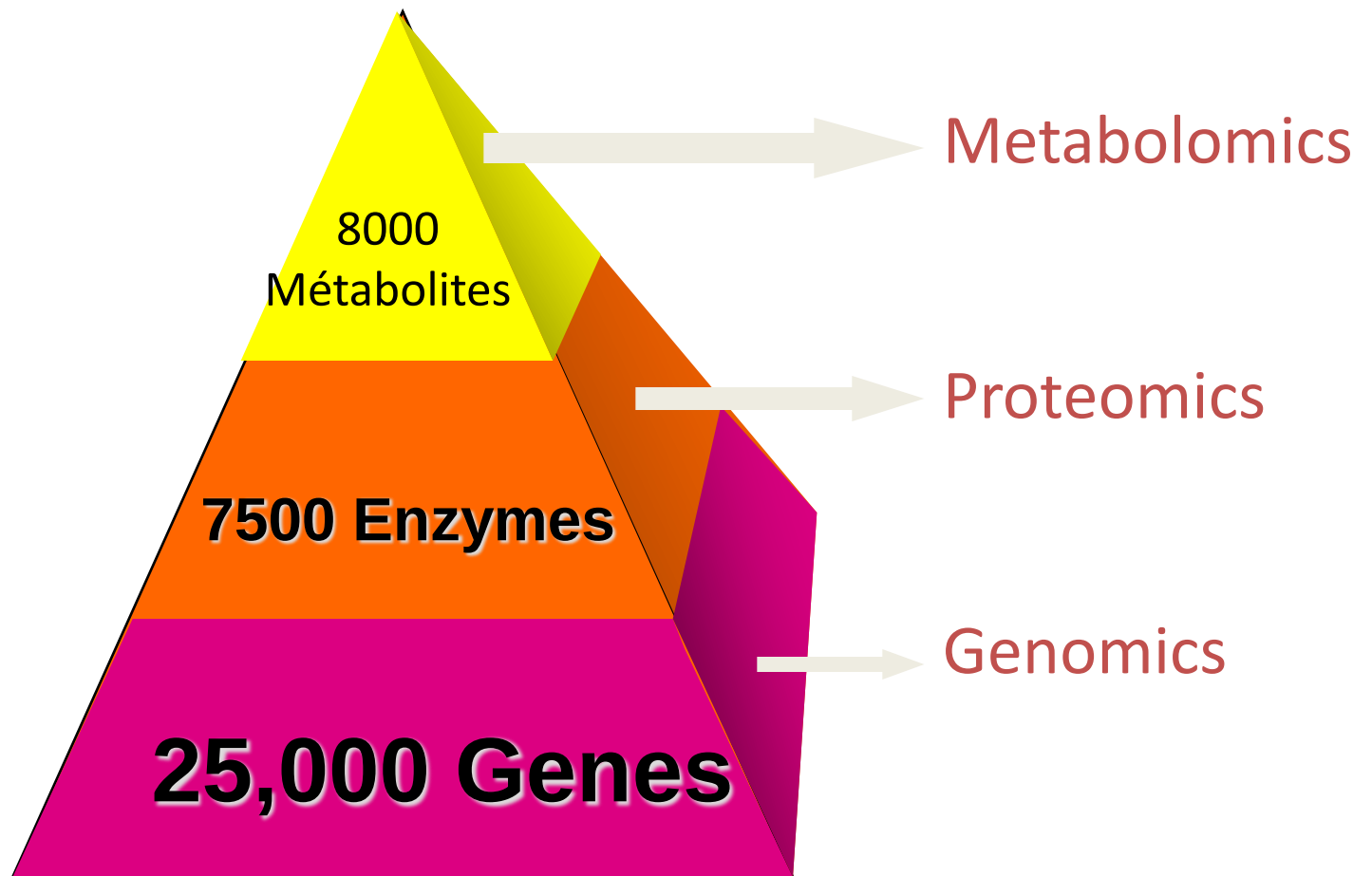


cartographier les voies métaboliques (mécanismes, flux)

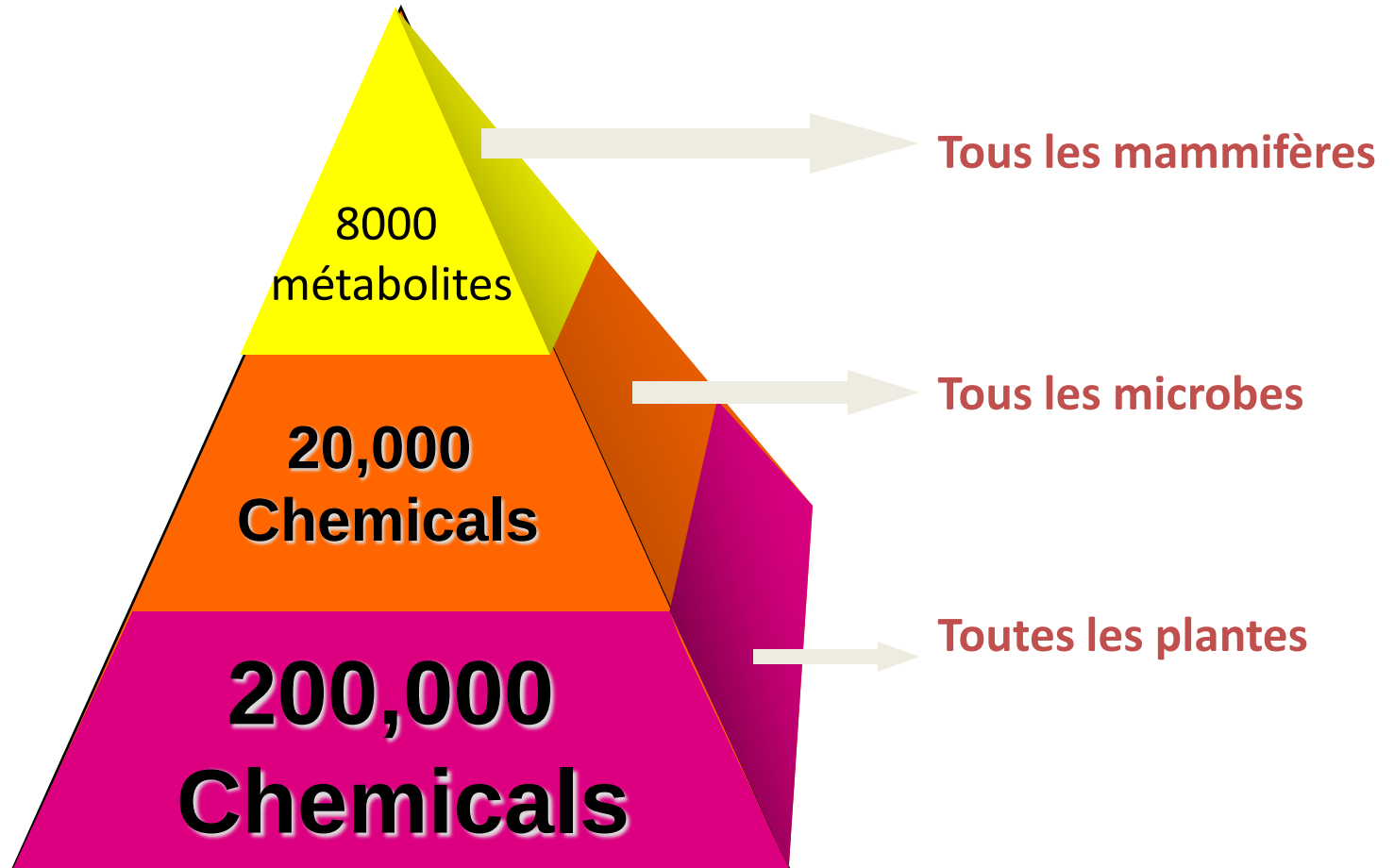
« marché des biomarqueurs sont évalués à 694 millions de \$ en 2008, montant qui pourrait atteindre 2,2 milliards de \$ en 2015, avec des perspectives de croissance annuelle de l'ordre de 12 à 13,5 %. »

History

- The first paper was titled, “Quantitative Analysis of Urine Vapor and Breath by Gas-Liquid Partition Chromatography”, by Robinson and Pauling in 1971.
- Many of the bioanalytical methods used for metabolomics have been adapted (or in some cases simply adopted) from existing biochemical techniques.
- Human Metabolome project – first draft of human metabolome in 2007



Differents Metabolomes



Pourquoi la métabolomique est difficile?

Ne peut être déduite du génome

Grande diversité chimique

gamme dynamique 7 log

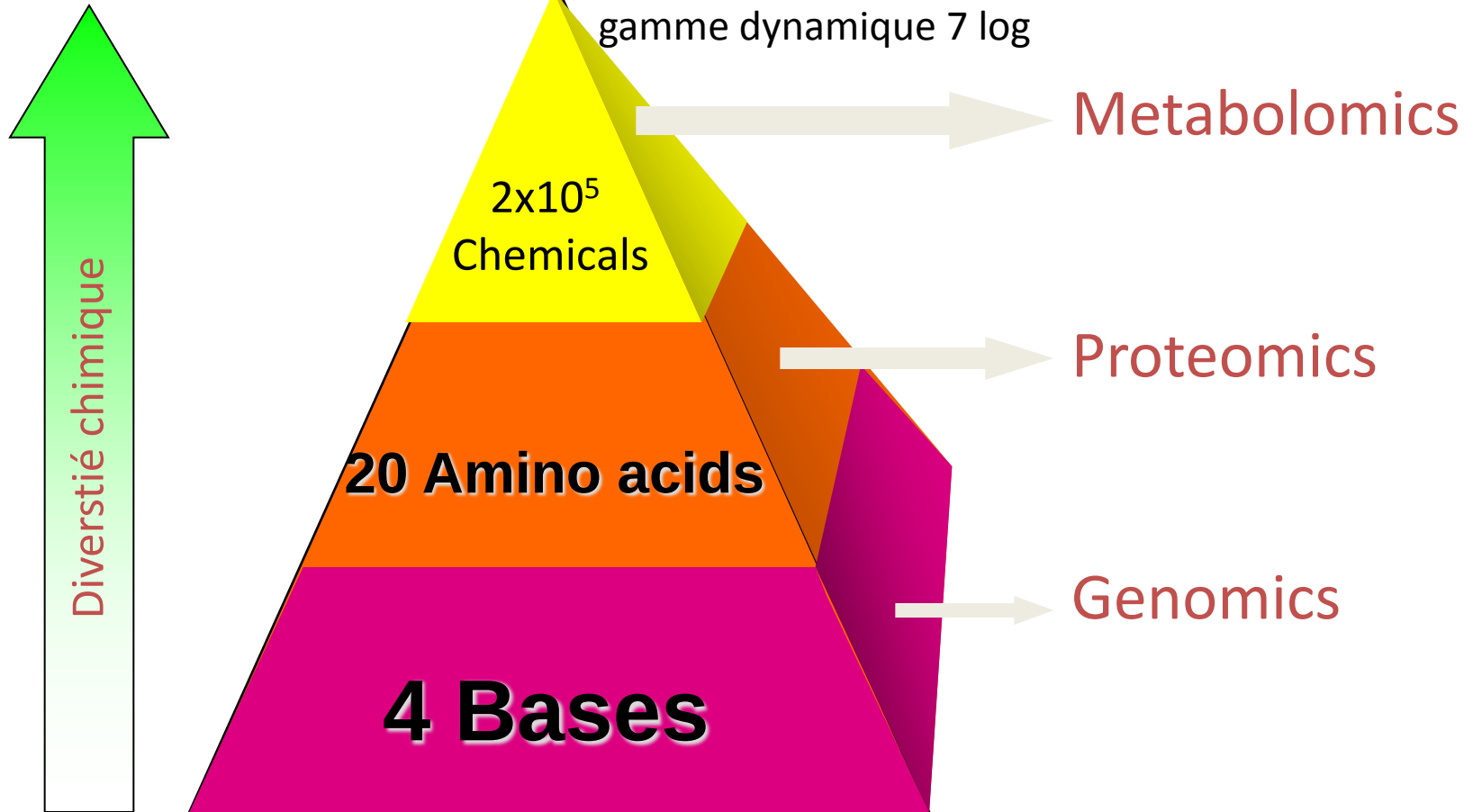


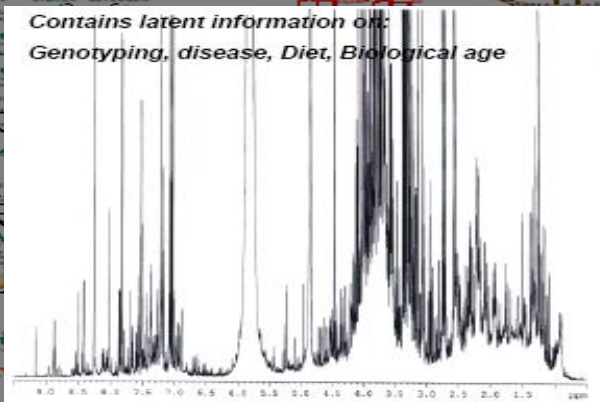


Plate-formes:
NMR
LC/MS
GC/MS



Sang, urines,
extrait tissulaire

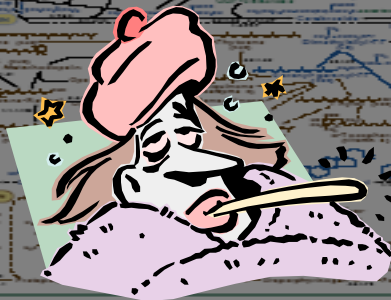
Contains latent information on:
Genotyping, disease, Diet, Biological age



Data acquisition

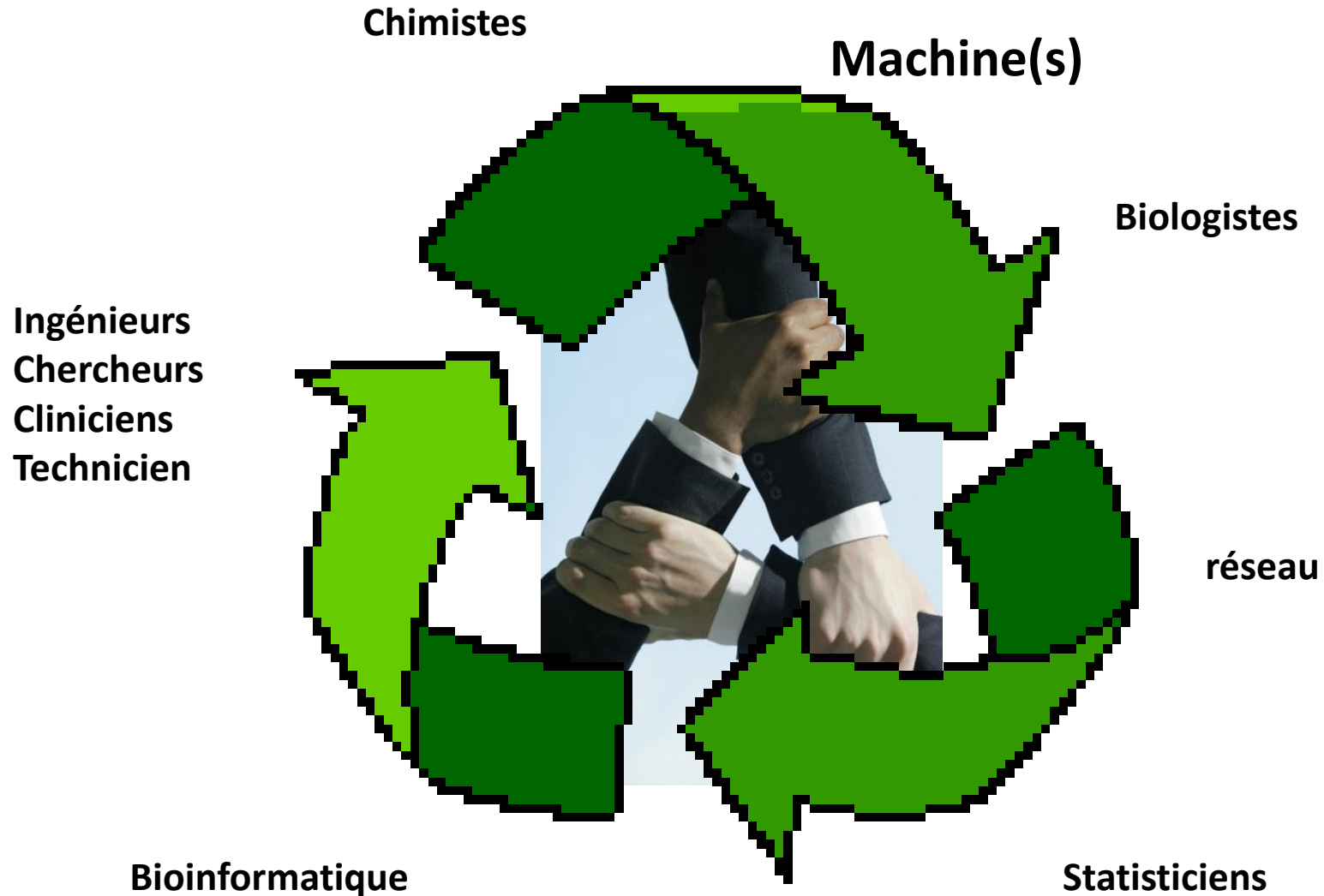


MVA
bioinformatic



Data mining

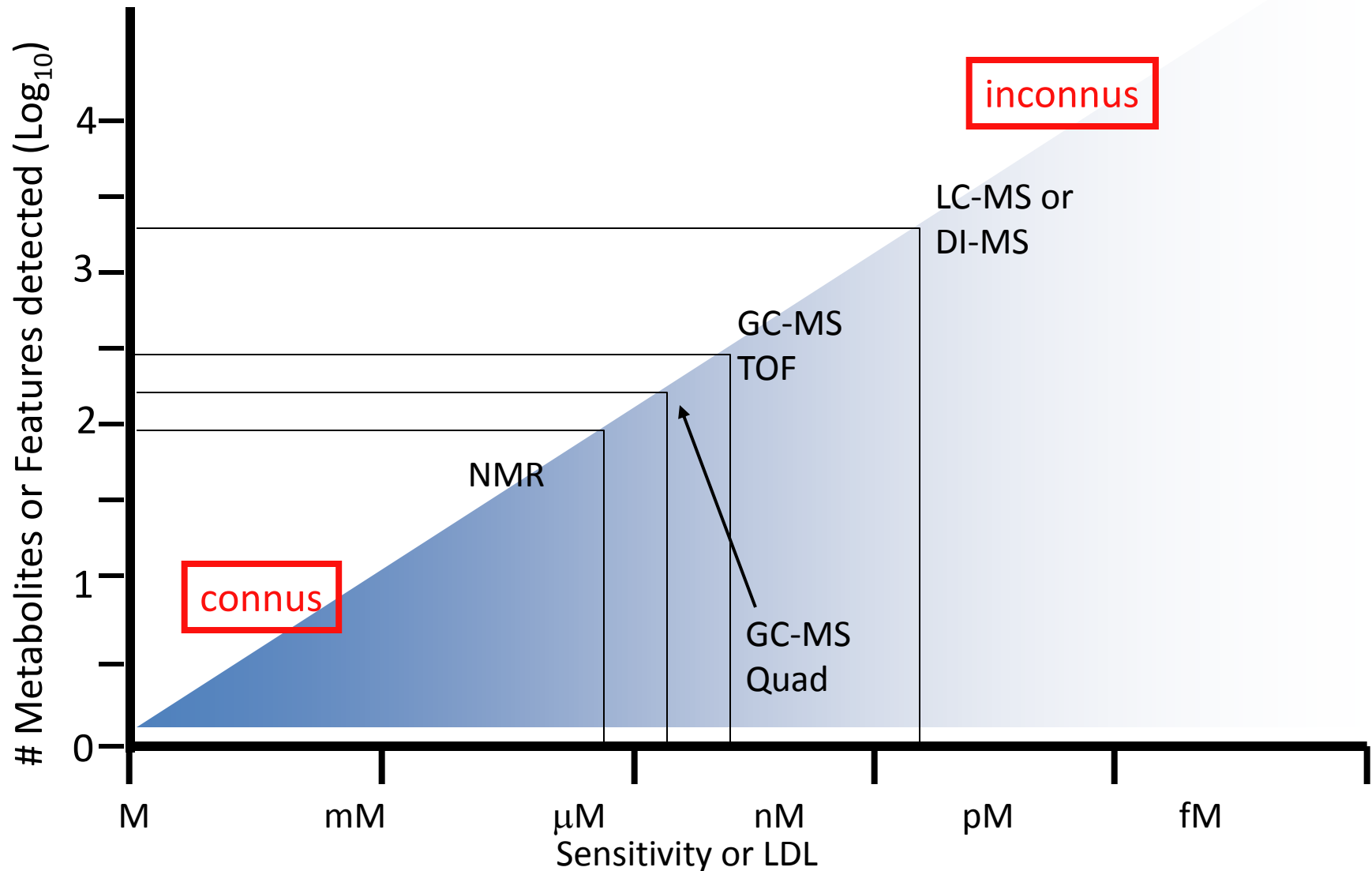
La métabolomique: quelle infrastructure.....



Quels sont les outils de la métabolomique?

outils	avantages	défauts
RMN	Robustesse et reproductibilité	Recouvrement sensibilité
GC-MS GC X GC TOF	sensibilité	dérivation
LC-MS	sensibilité	stabilité

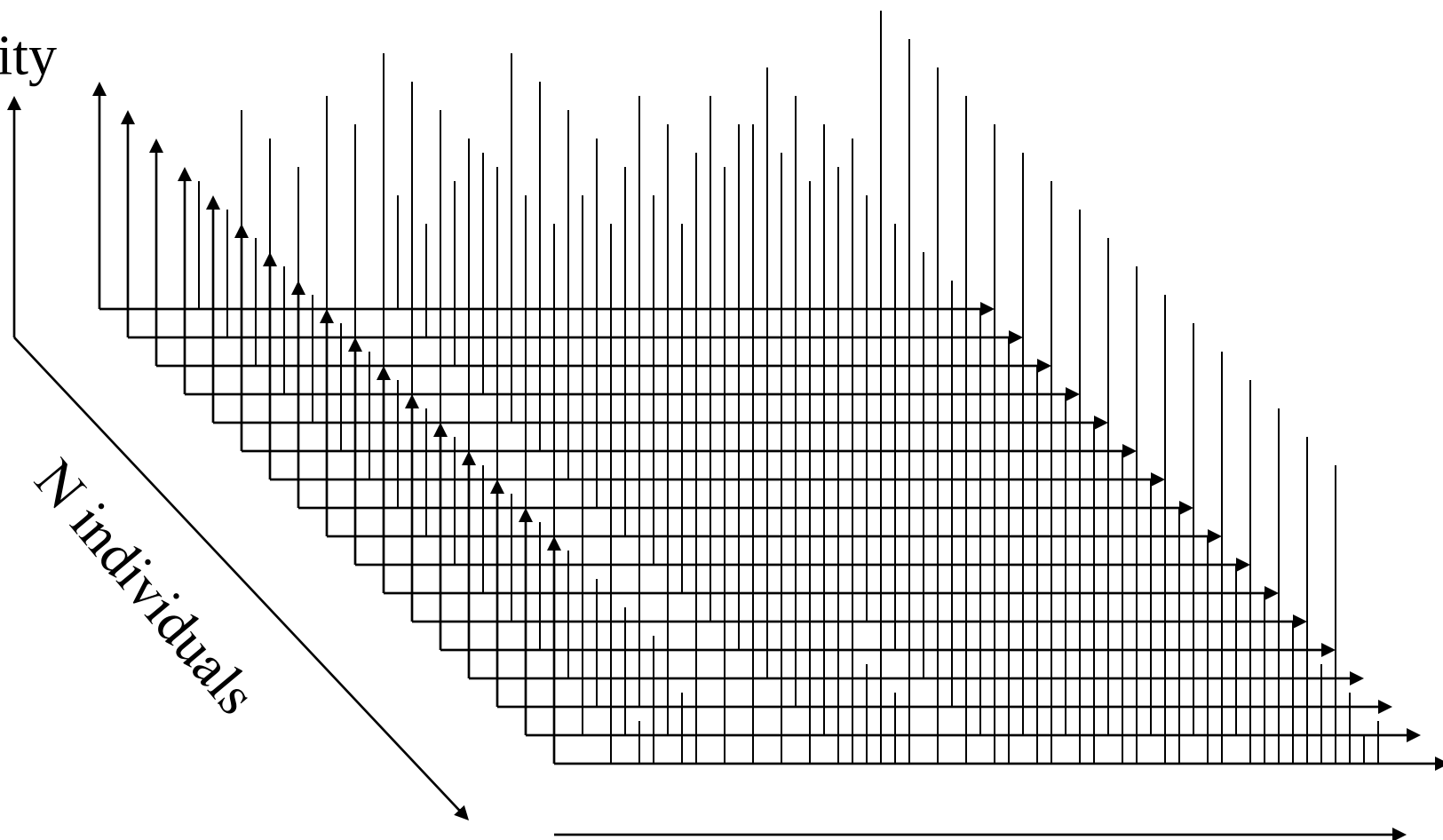
Technologie & Sensibilité



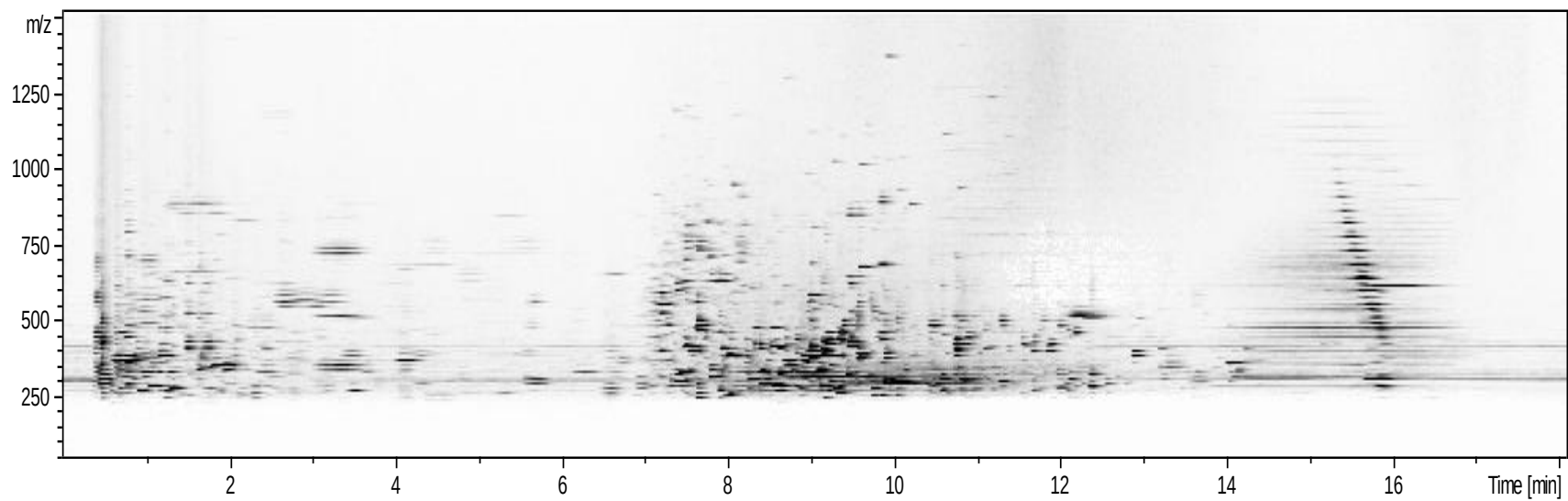
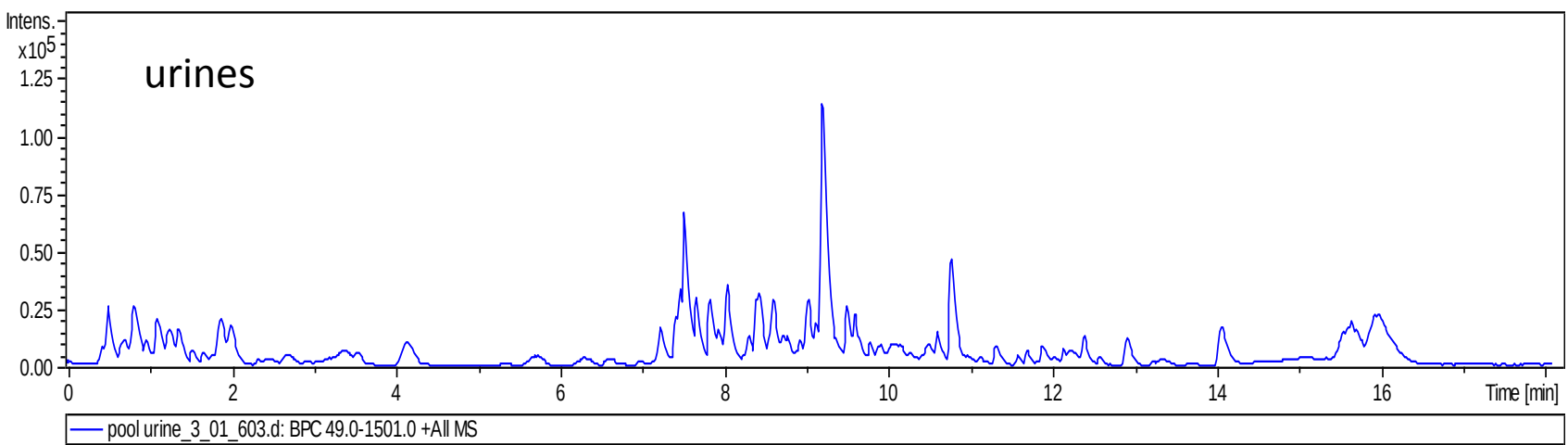
Que peut-on mesurer?

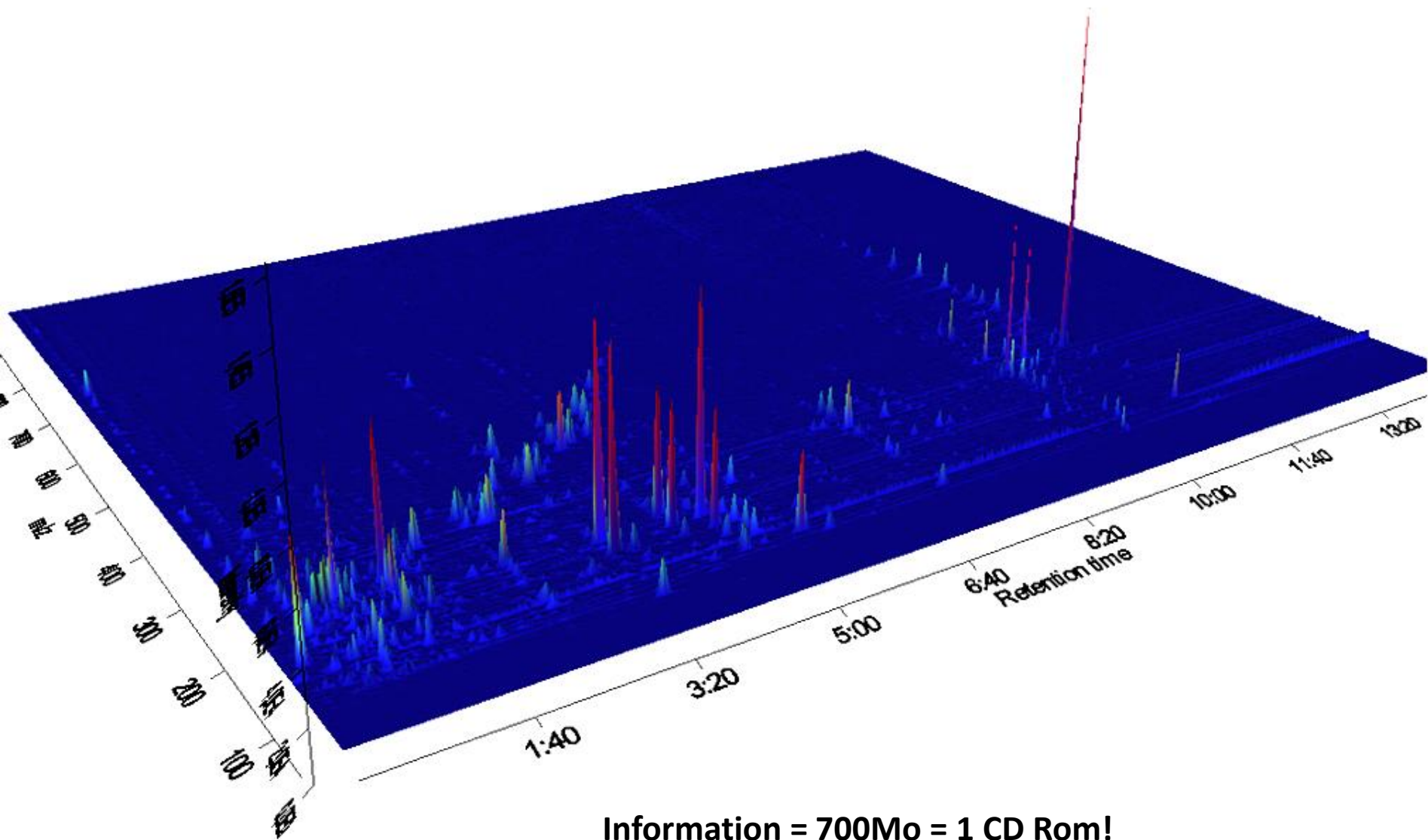
- NMR-based metabolomics (~50 metabolites identified/quantified, μM sensitivity)
- GC-MS based metabolomics (~70 metabolites identified/quantified, $<\mu\text{M}$ sensitivity)
- DI-MS based metabolomics (160 metabolites identified/quantified, nM sensitivity)
- LC-MS based metabolomics (300 metabolites identified/quantified, nM sensitivity)
- Lipidomics (3000 lipids identified and semi-quantified, nM sensitivity)
- Specialty phytochemical, nutrient, drug and pesticide analysis (mostly HPLC, nM sensitivity)

intensity



K manifest variables



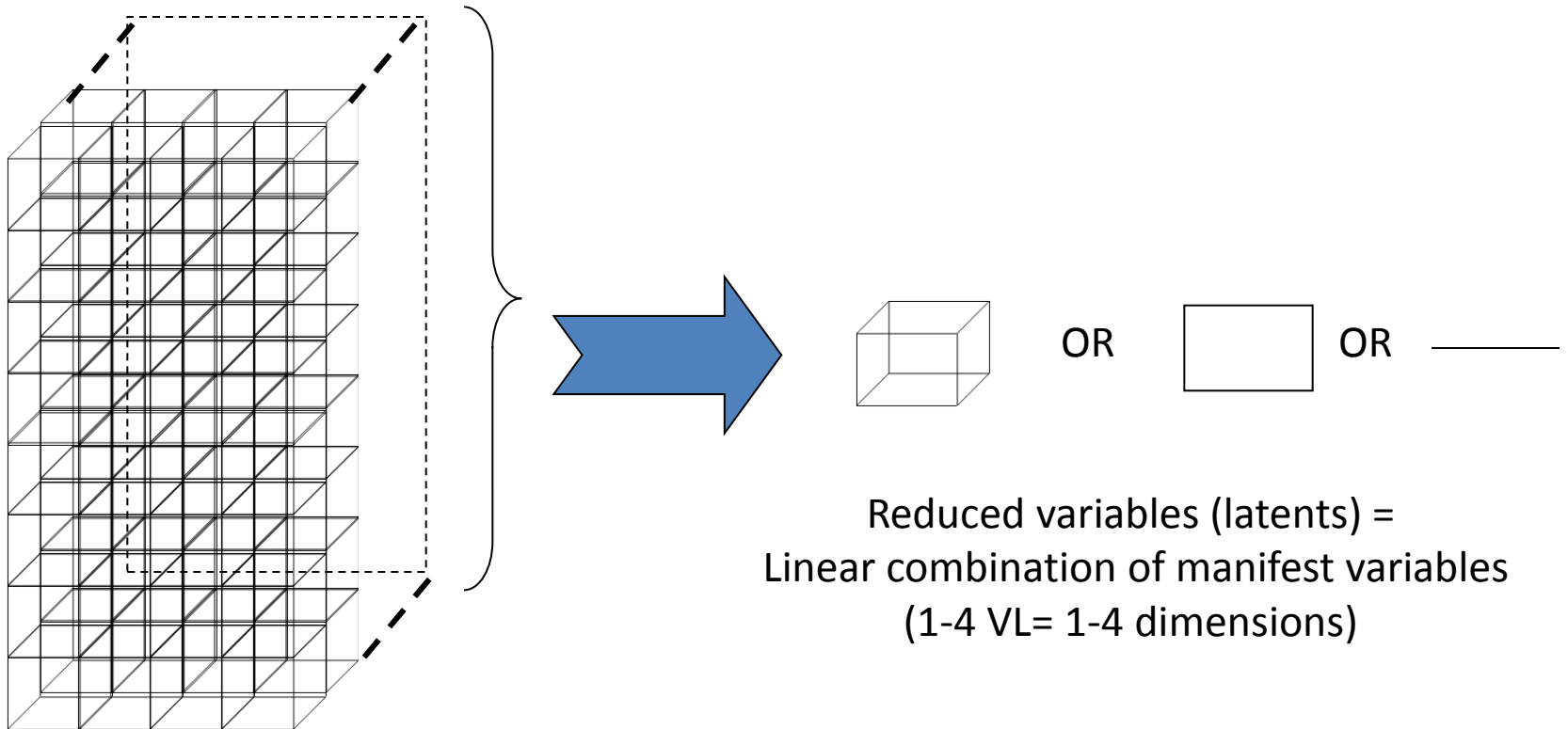


Information = 700Mo = 1 CD Rom!

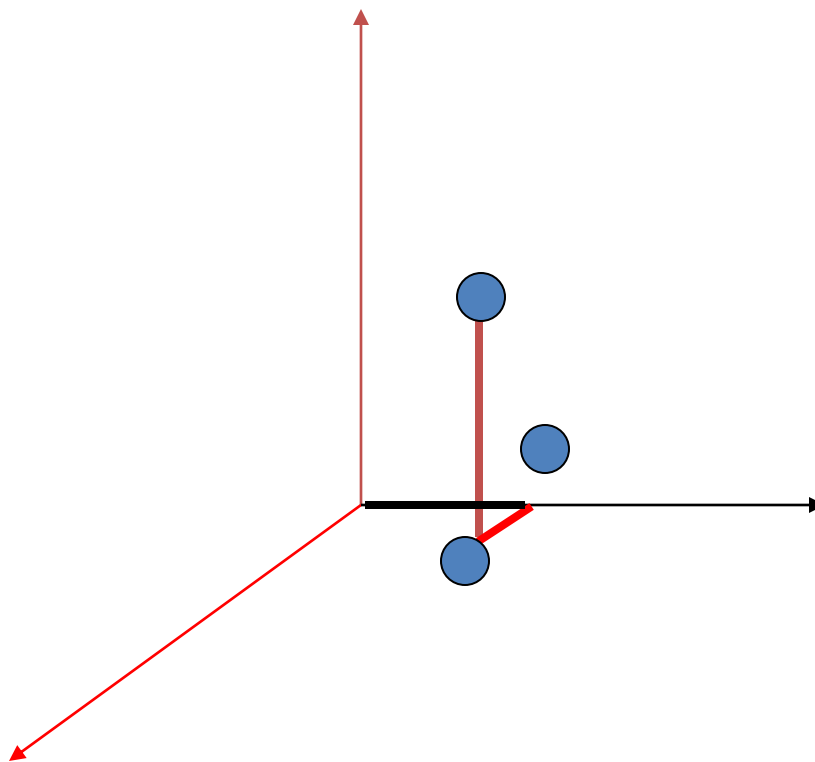
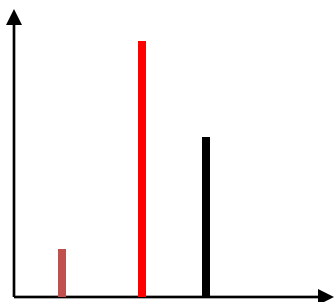
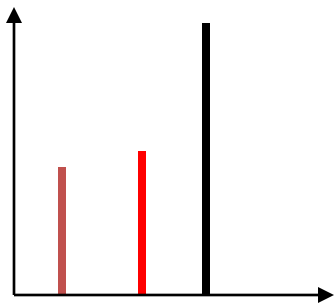
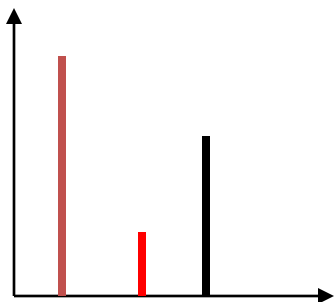
1 variable = 1 dimension

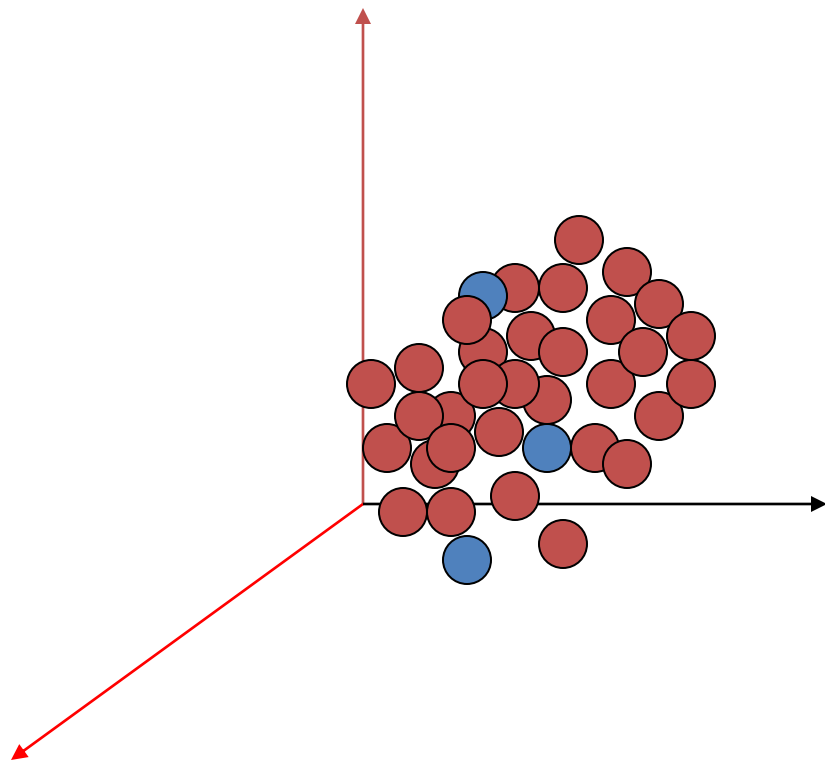
1000 variables = 1000 dimensions

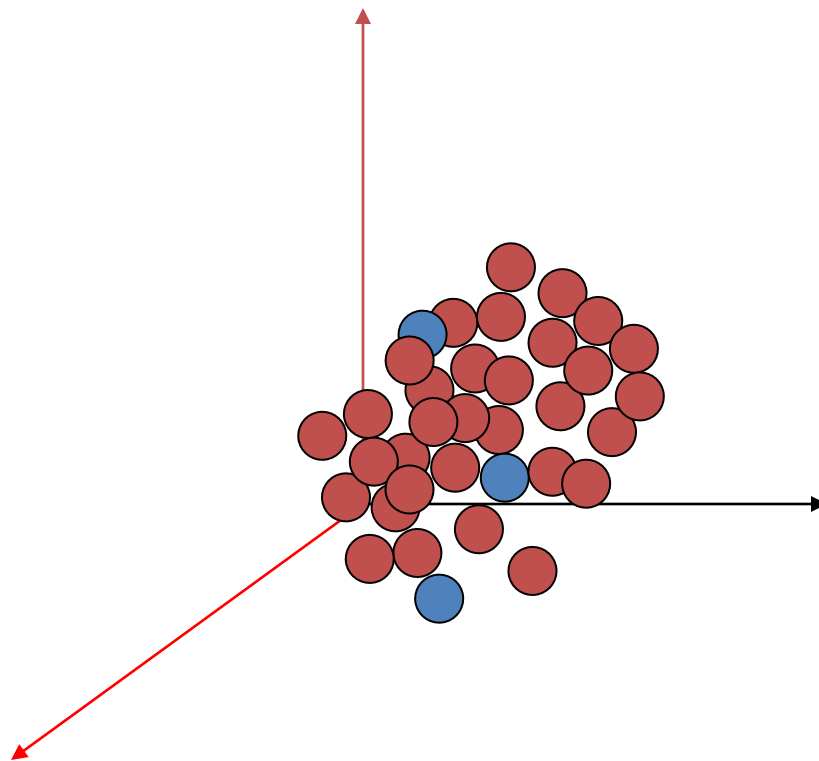
GOAL= REDUCTION WITHOUT LOSING INFORMATION



K manifest variables = k dimensional space
(Hyperspace)





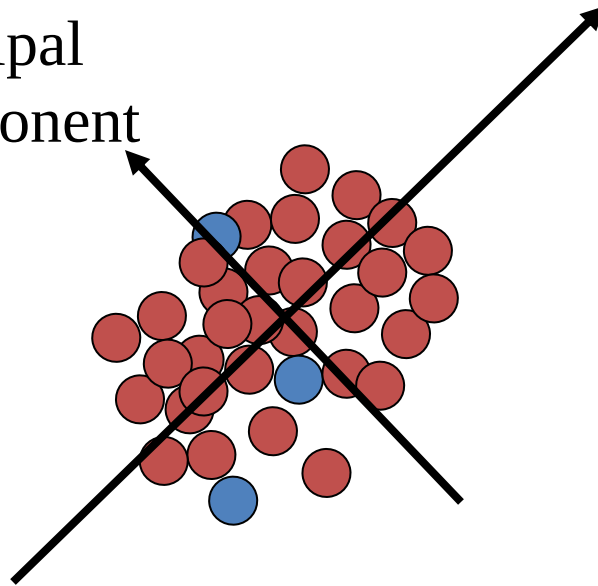
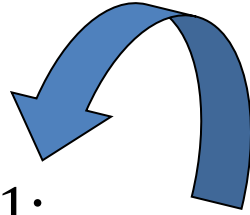


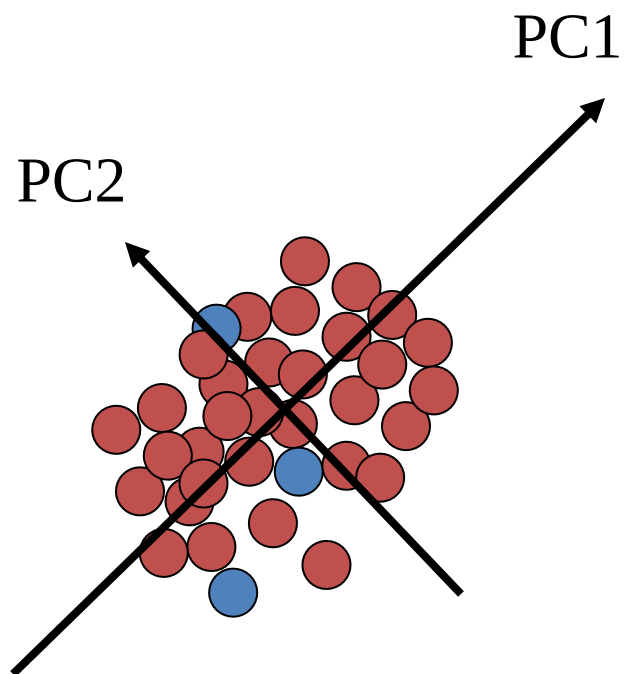
1^{ère} principal
component

Orthogonal to PC1:
Max of variance

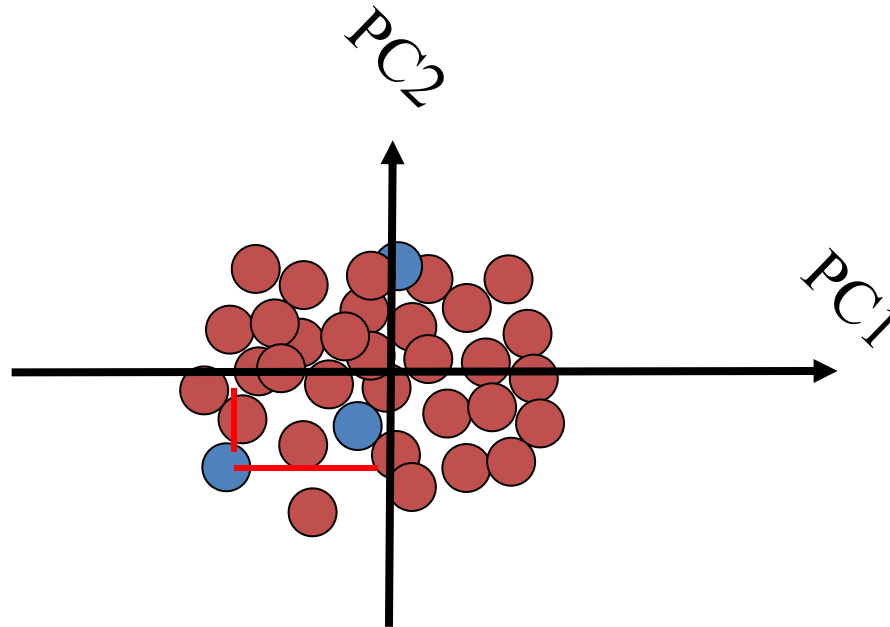
Second
principal
component

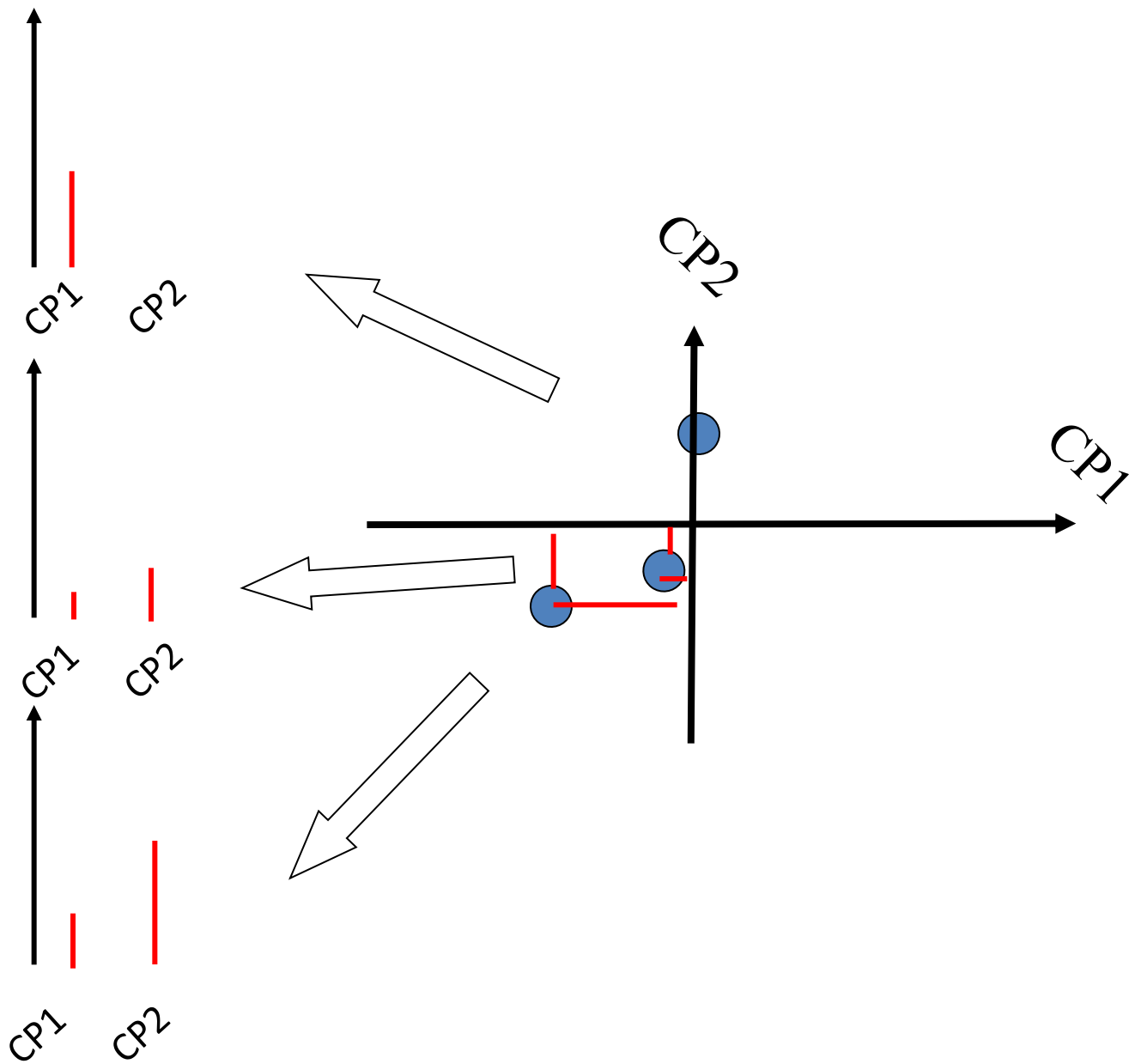
Axis of maximal deviation

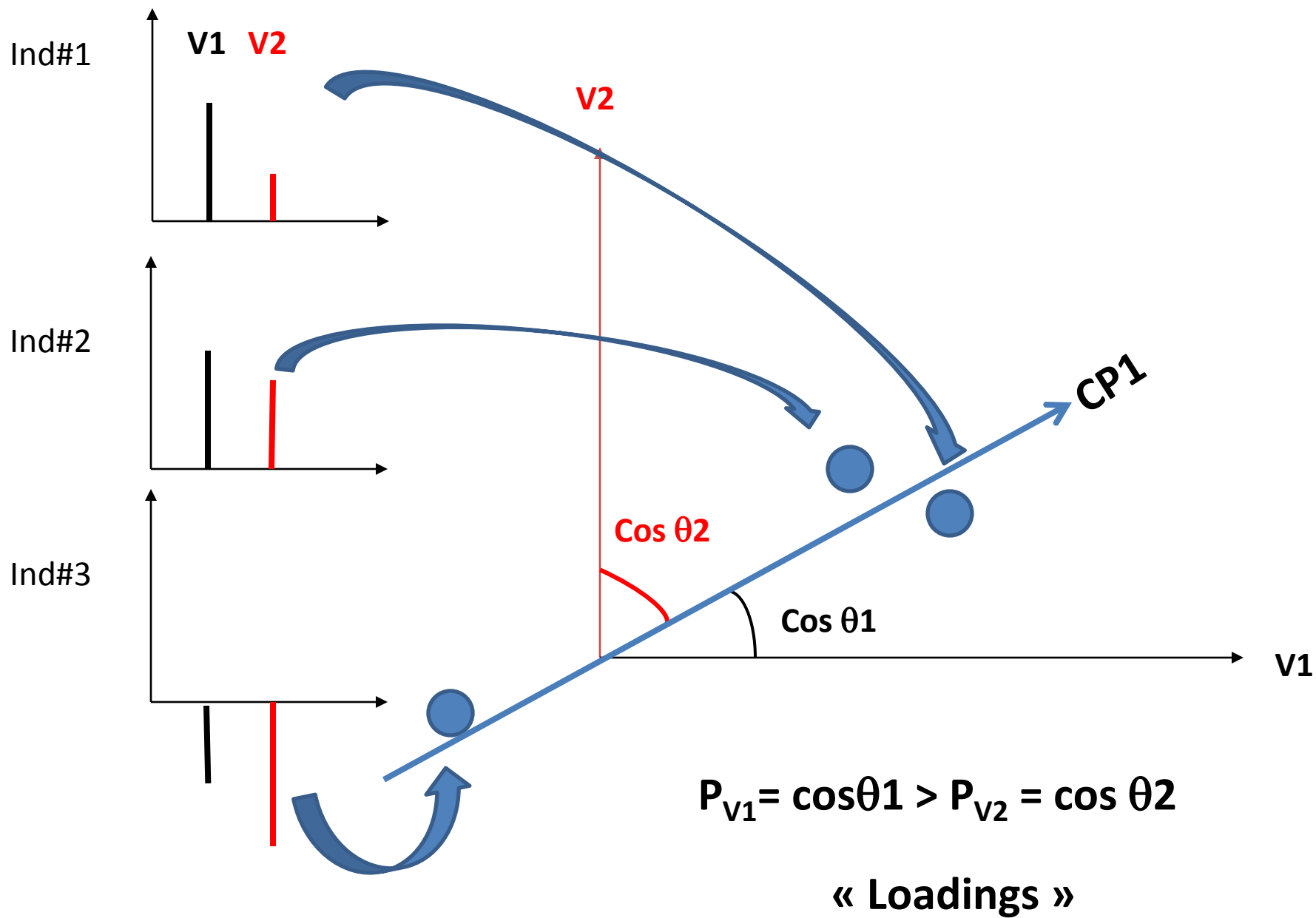




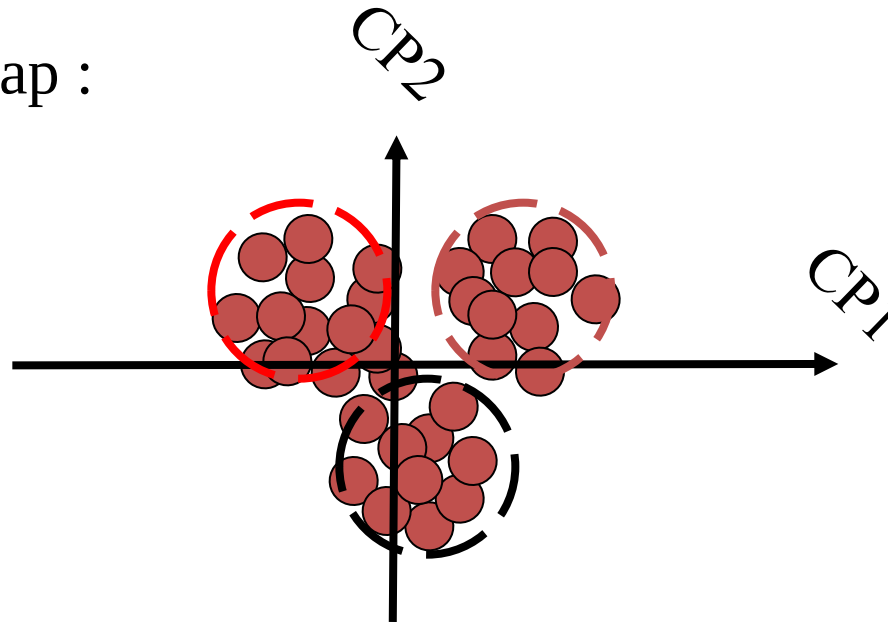
Observations map or « score plot »



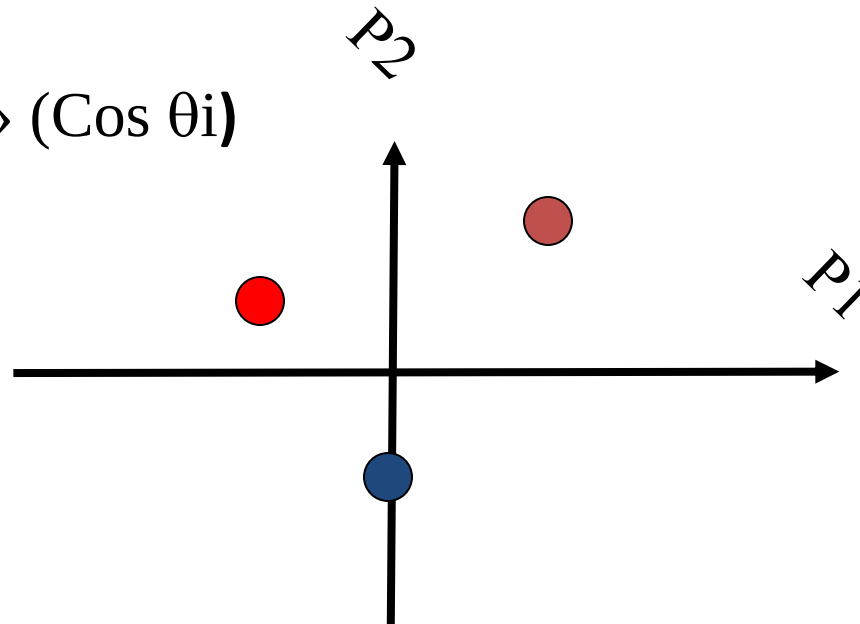




observations map :
« Score plot »



variables map :
« loading plot » ($\cos \theta_i$)



Pour aller plus loin....

Metabolomics (2014) 10:361–374

DOI 10.1007/s11306-013-0598-6

REVIEW ARTICLE

Reflections on univariate and multivariate analysis of metabolomics data

Edoardo Saccenti · Huub C. J. Hoefsloot ·
Age K. Smilde · Johan A. Westerhuis ·
Margriet M. W. B. Hendriks

Environmental and Molecular Mutagenesis 54:542–557 (2013)

Review Article

Deciphering the Complex: Methodological Overview of Statistical Models to Derive OMICS-Based Biomarkers

Marc Chadeau-Hyam,^{1*} Gianluca Campanella,¹ Thibaut Jombart,²
Leonardo Bottola,³ Lutzen Portengen,⁴ Paolo Vineis,^{1,5} Benoit Liquet,⁶ and
Roel C.H. Vermeulen^{4,7}

Gut flora metabolism of phosphatidylcholine promotes cardiovascular disease

Zeneng Wang^{1,2}, Elizabeth Klipfell^{1,2}, Brian J. Bennett³, Robert Koeth¹, Bruce S. Levison^{1,2}, Brandon DuGar¹, Ariel E. Feldstein^{1,2}, Earl B. Britt^{1,2}, Xiaoming Fu^{1,2}, Yoon-Mi Chung^{1,2}, Yuping Wu⁴, Phil Schauer⁵, Jonathan D. Smith^{1,6}, Hooman Allayee⁷, W. H. Wilson Tang^{1,2,6}, Joseph A. DiDonato^{1,2}, Aldons J. Lusi³ & Stanley L. Hazen^{1,2,6}

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50 cas avec 3 ans suivi (infarctus, AVC, mort)
50 témoins

Métabolomique LC MS plasma
= 2000 analytes mesurés



40 analytes fortement discriminants

Cohorte 1 = cohorte d'apprentissage

Cohorte 2 indépendante = cohorte de validation

**50 cas avec 3 ans suivi (infarctus, AVC, mort)
50 témoins**

**25 cas avec 3 ans suivi (infarctus, AVC, mort)
25 témoins**

Métabolomique LC MS plasma
= 2000 analytes mesurés

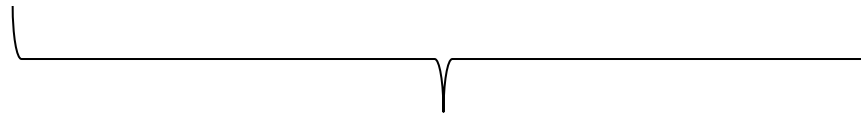


40 analytes fortement discriminants

Métabolomique LC MS plasma
= 2000 analytes mesurés



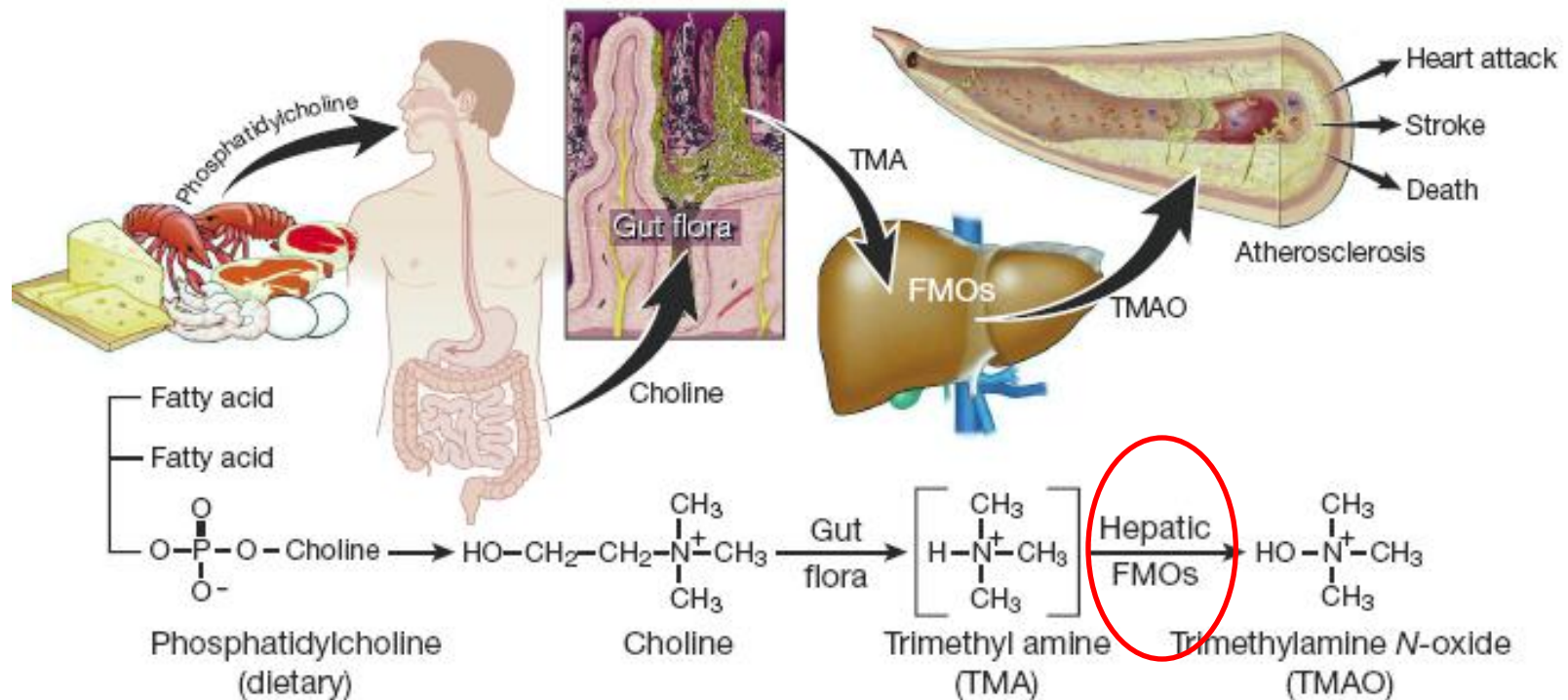
24 analytes fortement discriminants



18 analytes en commun

Top 3 = choline, betaine, triméthoxyde amine (TMAO)

choline, betaine, triméthoxyde amine (TMAO)= Métabolites de la phosphatidylcholine

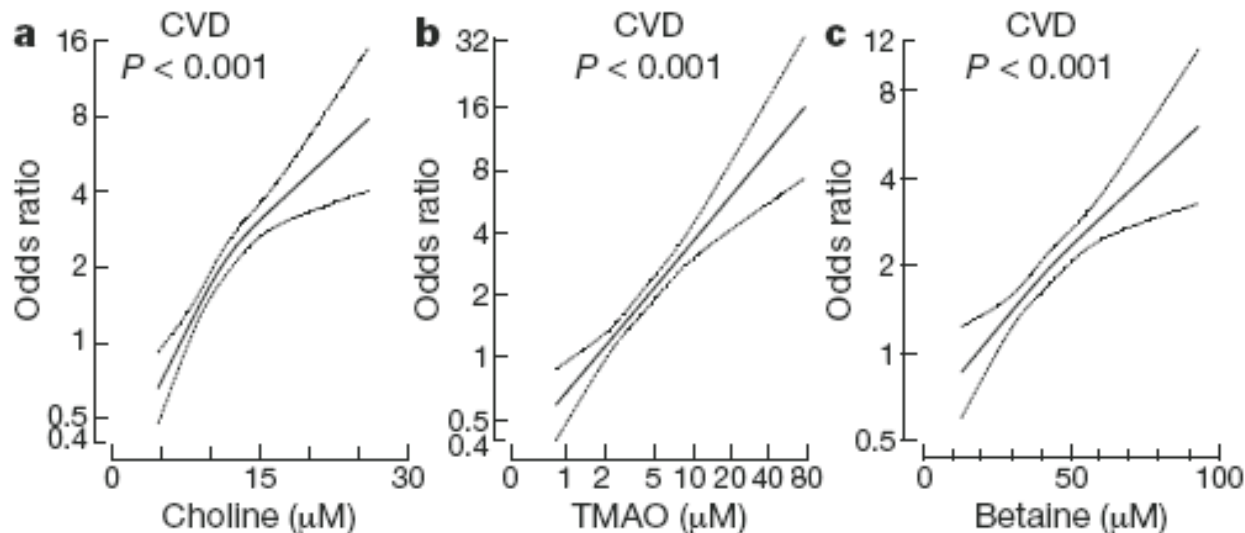


bétaine

Validation sur modèle souris athéromateux:

- nourris avec ou sans choline
- Nourris avec ou sans TMAO
- Traités aux antibiotiques
- Cohortes humaines 1876 sujets avec angiographies

Dosage des métabolites « athérogènes » dans une cohorte de 1876 sujets à pathologie cardiovasculaire variable (angiographie)



Implication = nouveaux marqueurs du risque cardiovasculaire, indépendant des marqueurs lipidiques classiques

Identifications de cibles thérapeutiques: flore intestinale, FMOs hépatique

Diagnostic development process

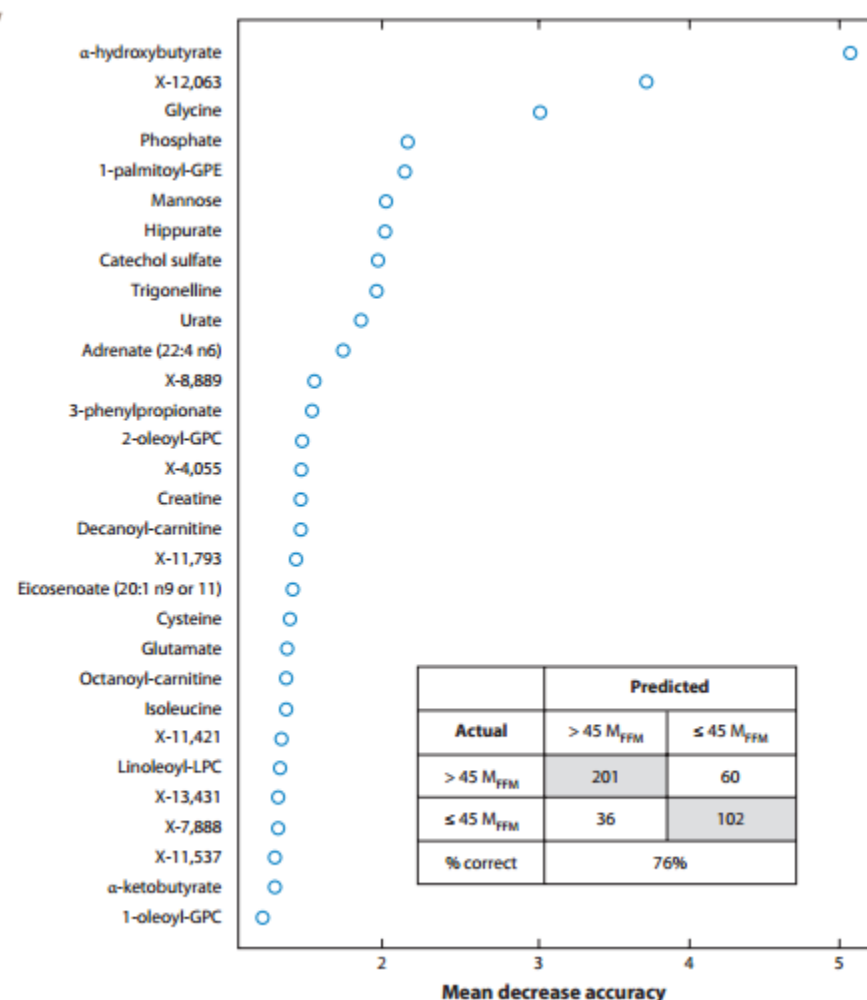
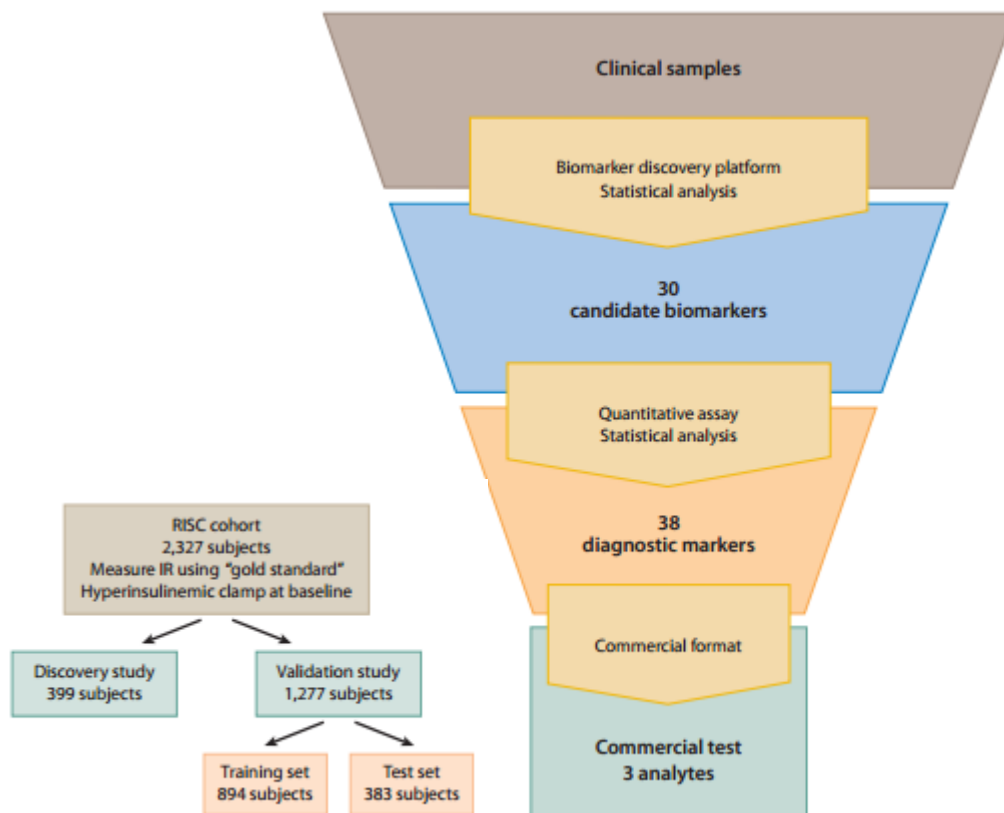
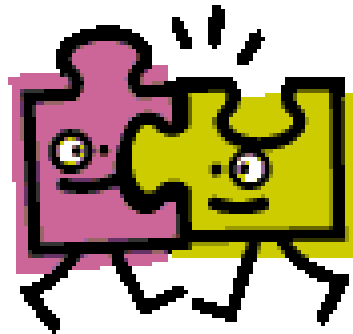


Table 1 Biomarker candidates

2-hydroxybutyrate	Linoleic acid	Stearate
3-hydroxybutyrate	Linolenic acid	Threonine
3-methyl-2-oxo-butyric acid	Margaric acid	Isovalerylcarnitine
3-phenylpropionate	Octanoyl carnitine	Linoleoyl-LPC
Catechol sulfate	Oleic acid	1,5-anhydroglucitol
Creatine	Oleoyl-LPC	Stearoyl-LPC
Decanoyl carnitine	Palmitate	1-palmitoyl-GPE
Docosatetraenoic acid	Palmitoleic acid	Octanoylcarnitine
Glutamic acid	Palmitoyl-LPC	alpha-ketobutyrate
Glycine	Serine	Cysteine

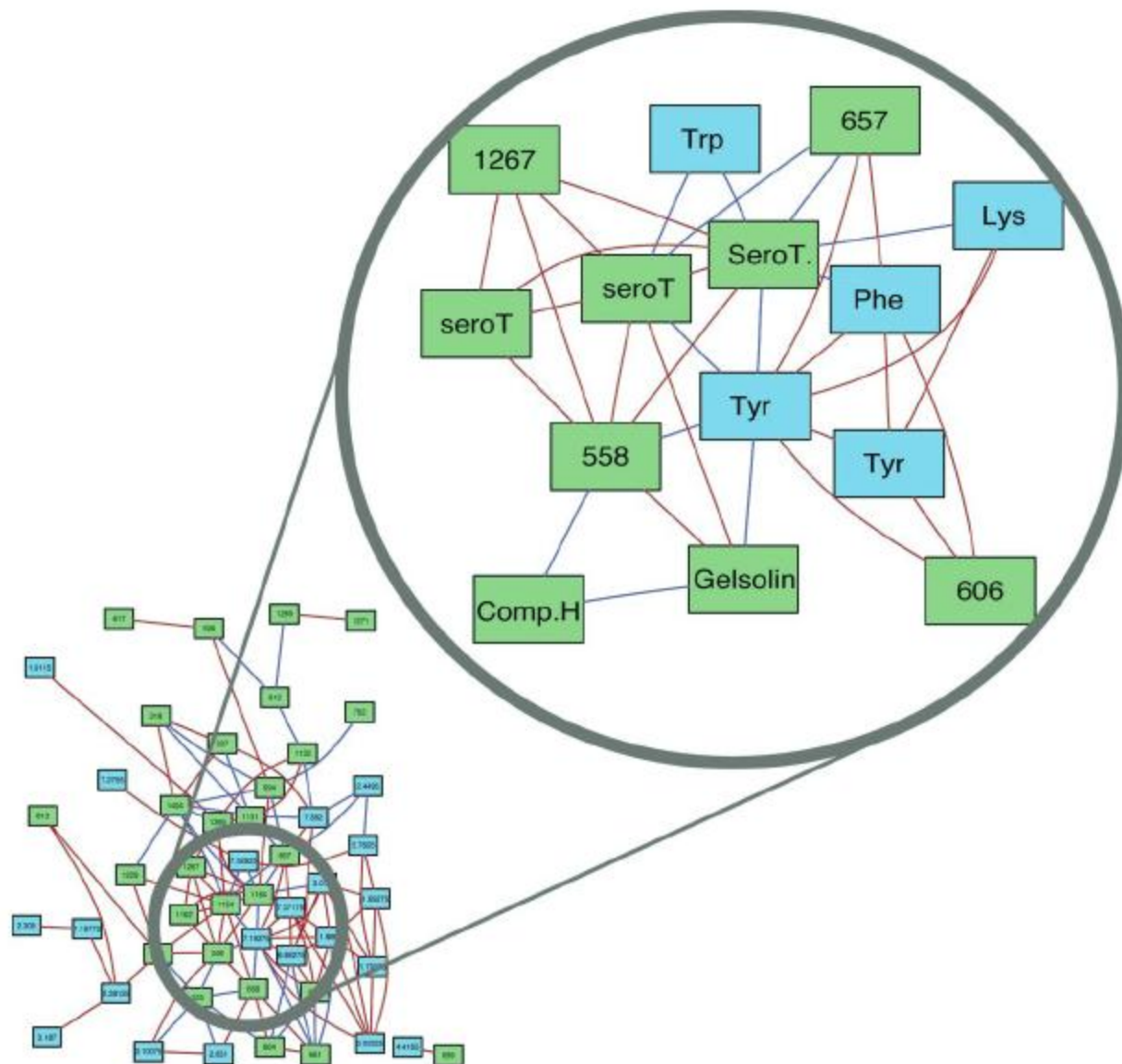
Couplage avec autres omiques



Statistically Integrated Metabonomic–Proteomic Studies on a Human Prostate Cancer Xenograft Model in Mice

Mattias Rantalainen,[†] Olivier Cloarec,[†] Olaf Beckonert,^{†,‡} I. D. Wilson,[‡] David Jackson,[§] Robert Tonge,[§] Rachel Rowlinson,[§] Steve Rayner,[§] Janice Nickson,[§] Robert W. Wilkinson,^{||} Jonathan D. Mills,^{||} Johan Trygg,^{*,⊥} Jeremy K. Nicholson,[†] and Elaine Holmes^{*,†}

Multiple correlations between metabolites and proteins were found, including associations between serotransferrin precursor and both tyrosine and 3-D-hydroxybutyrate. Additionally, a correlation between decreased concentration of tyrosine and increased presence of gelsolin was also observed. This approach can provide enhanced recovery of combination candidate biomarkers across multi-omic platforms, thus, enhancing understanding of in vivo model systems studied by multiple omic technologies



Proteasome Subunits mRNA Expressions Correlate With Male BMI: Implications for a Role in Obesity

Obesity (2009) **17**, 1044–1049.

Adipose Tissue IL-6 Content Correlates with Resistance to Insulin Activation of Glucose Uptake both *in Vivo* and *in Vitro*

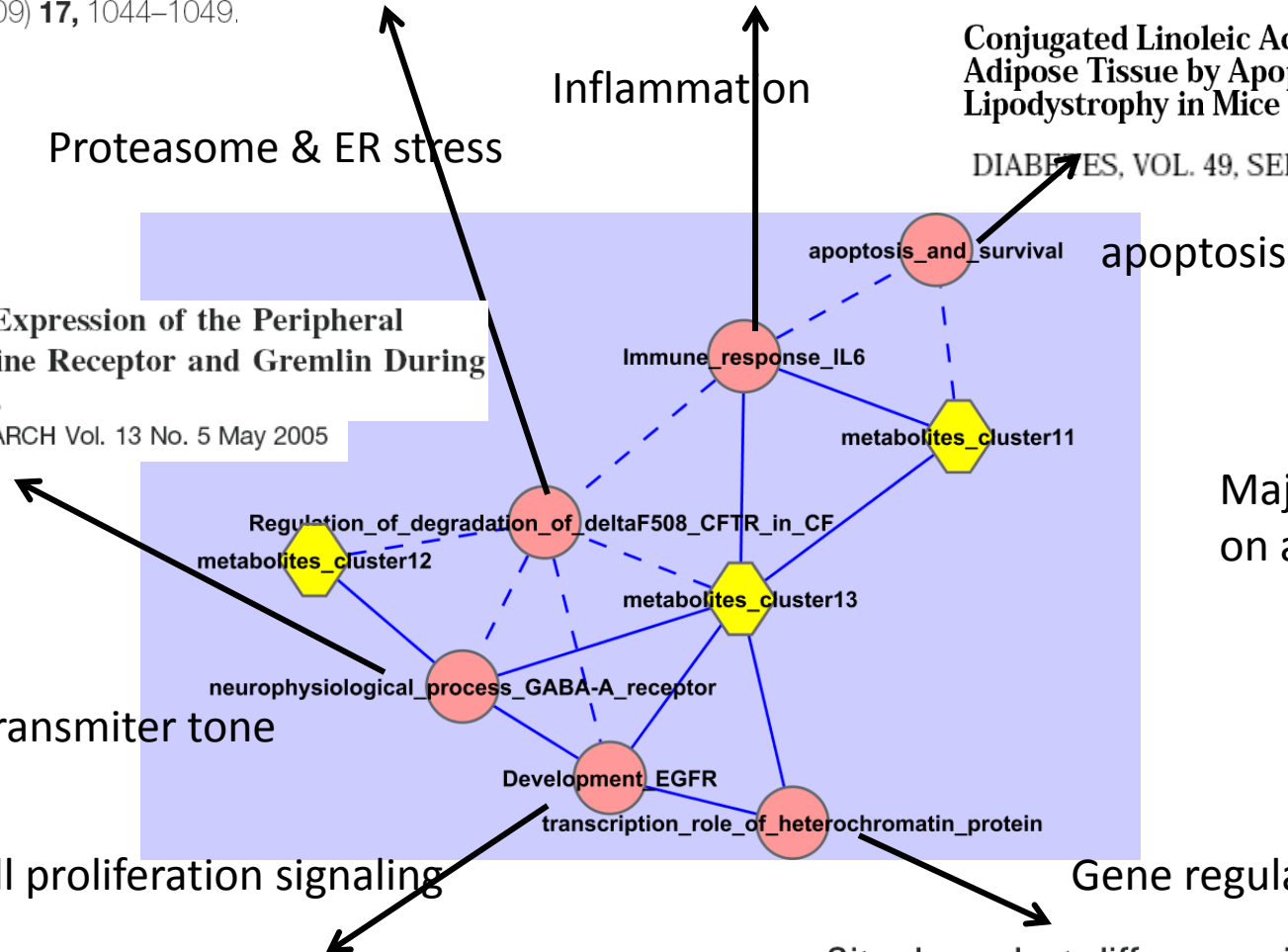
The Journal of Clinical Endocrinology & Metabolism 87(5):2084–2089

Conjugated Linoleic Acid Supplementation Reduces Adipose Tissue by Apoptosis and Develops Lipodystrophy in Mice

DIABETES, VOL. 49, SEPTEMBER 2000

Differential Expression of the Peripheral Benzodiazepine Receptor and Gremlin During Adipogenesis

OBESITY RESEARCH Vol. 13 No. 5 May 2005



Neurotransmitter tone

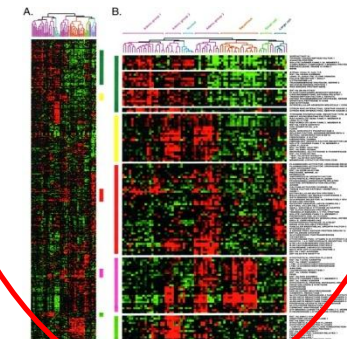
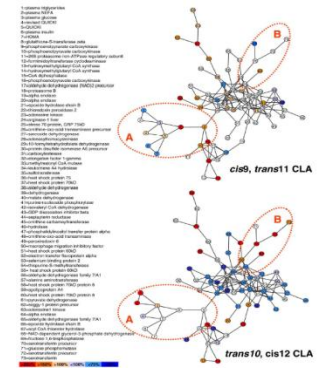
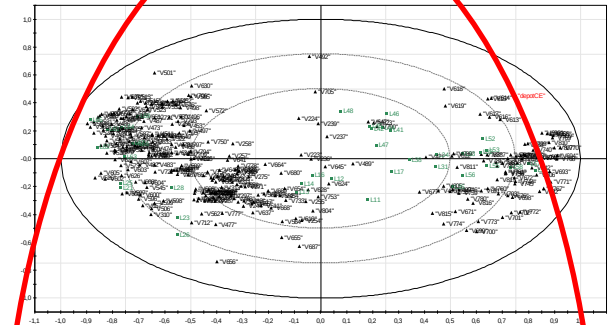
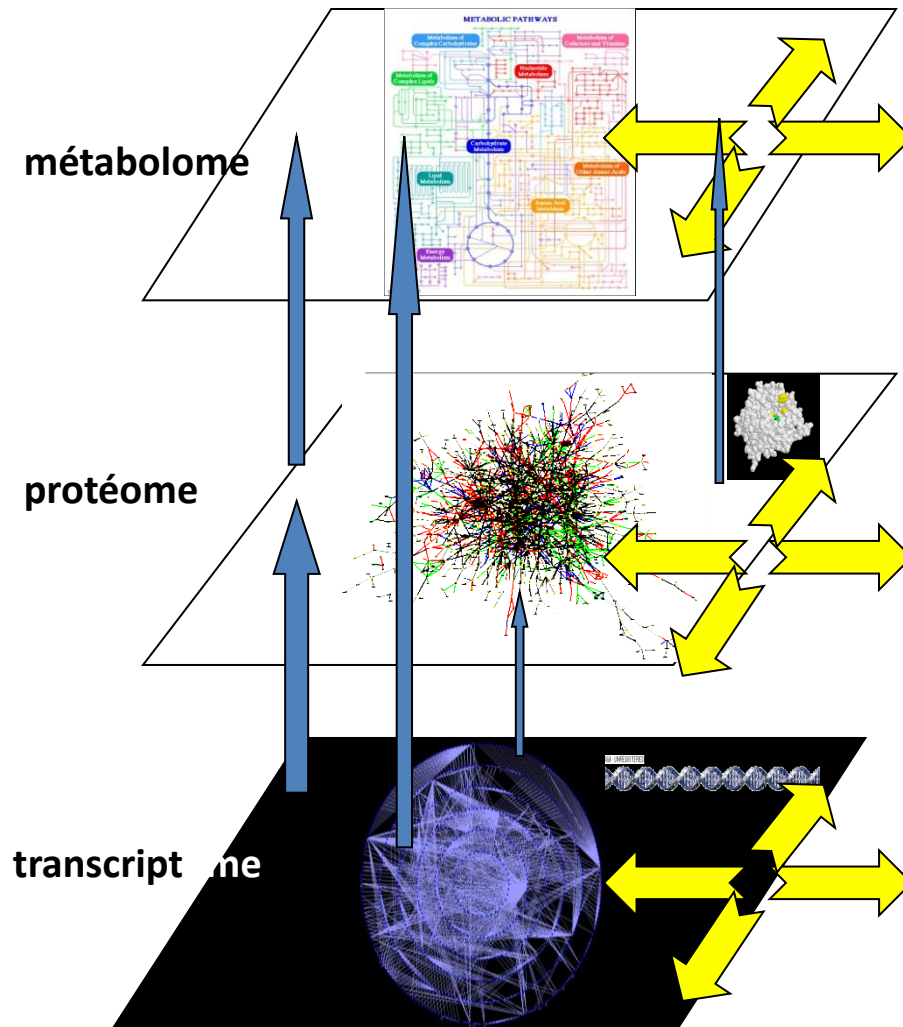
Cell proliferation signaling

EGFR Tyrosine Kinase Inhibitor (PD153035) Improves Glucose Tolerance and Insulin Action in High-Fat Diet-Fed Mice

DIABETES, VOL. 58, DECEMBER 2009

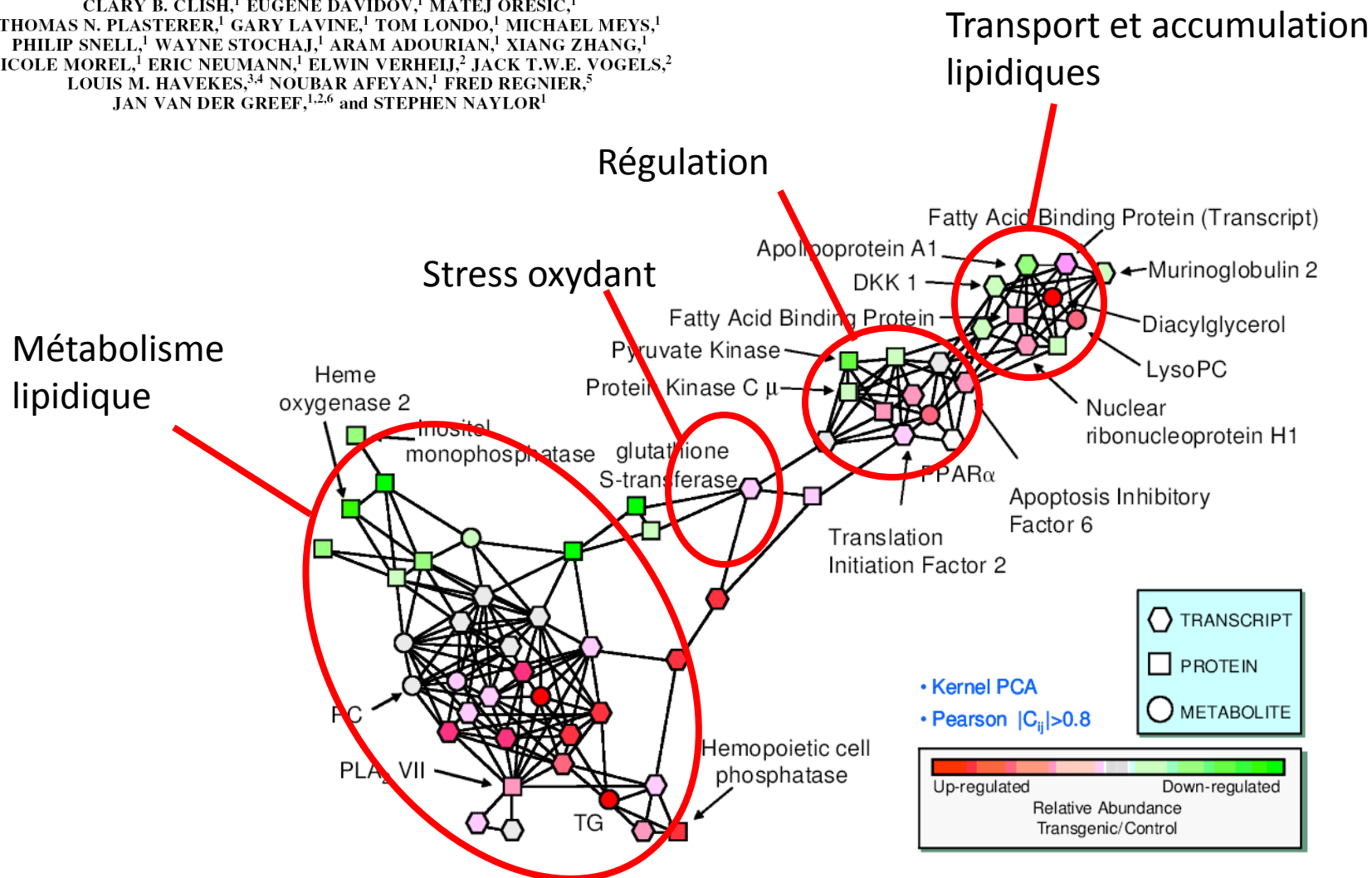
Site-dependent differences in both prelamins A and adipogenic genes in subcutaneous adipose tissue of patients with type 2 familial partial lipodystrophy

J Med Genet 2009;**46**:40–48.

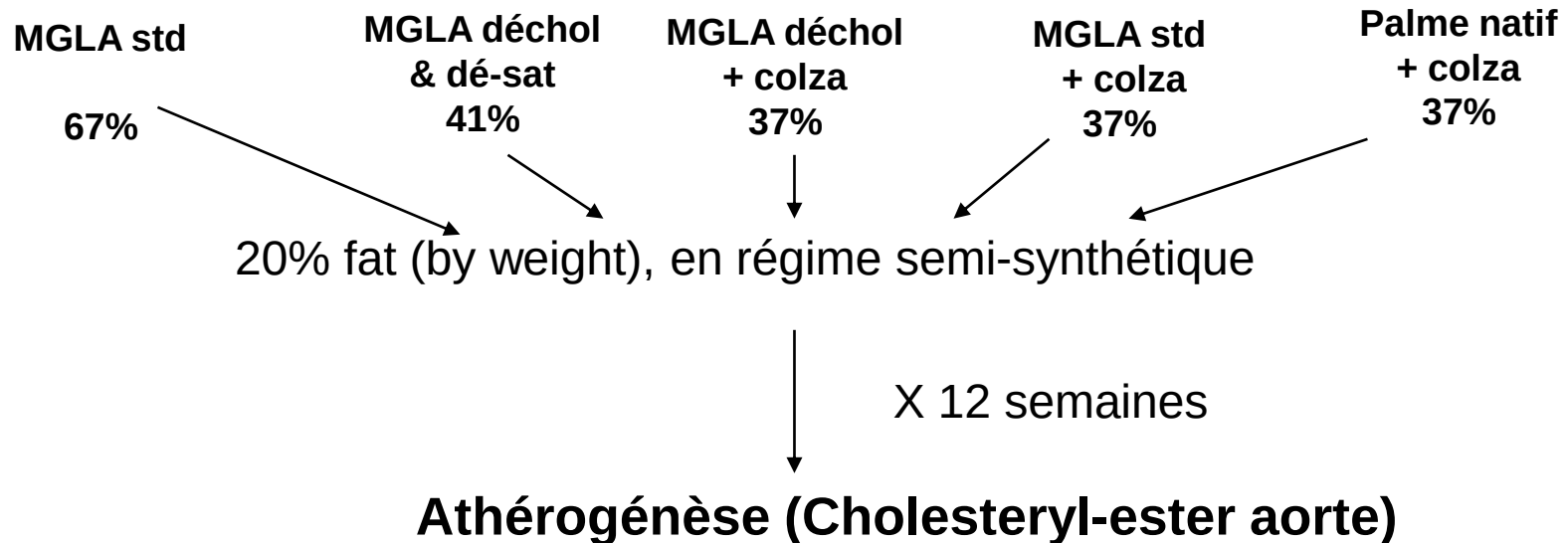


Integrative Biological Analysis of the APOE*3-Leiden Transgenic Mouse

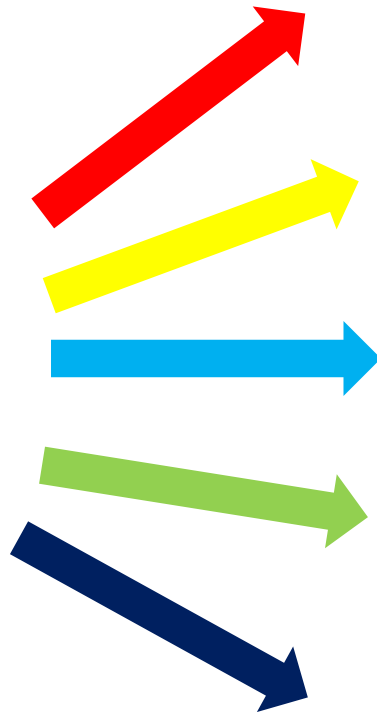
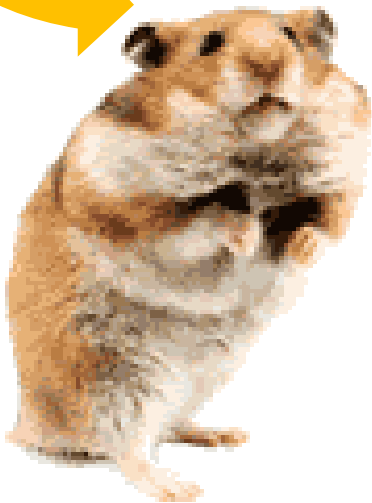
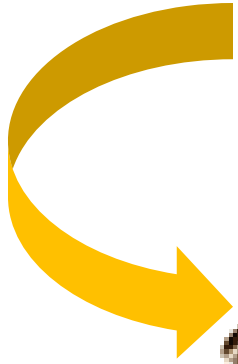
CLARY B. CLISH,¹ EUGENE DAVIDOV,¹ MATEJ ORESIC,¹
 THOMAS N. PLASTERER,¹ GARY LAVINE,¹ TOM LONDO,¹ MICHAEL MEYS,¹
 PHILIP SNELL,¹ WAYNE STOCHAJ,¹ ARAM ADOURIAN,¹ XIANG ZHANG,¹
 NICOLE MOREL,¹ ERIC NEUMANN,¹ ELWIN VERHELJ,² JACK T.W.E. VOGELS,²
 LOUIS M. HAVEKES,^{3,4} NOUBAR AFEYAN,¹ FRED REGNIER,⁵
 JAN VAN DER GREEF,^{1,2,6} and STEPHEN NAYLOR¹



Evaluer l'impact athérogénique de différentes matières grasses laitières anhydres +/-décholestérolisées et appauvries en AGS



Régimes tests



Metabolome plasmatique n=701 (LCMS)

Metabolome urines n=1224 (LCMS)

Chimie du sang =26 (biochimie: TG, Chol, HDLC, LDLC,...)

Acides gras foie & classes lipidiques du plasma n=124 (fast-GC)

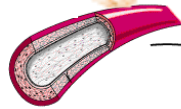
Expression gènes foie et sang n=44 (QPCR)

Total = 1666 variables biologiques/hamster + 1 (mesure indice athérogénicité)



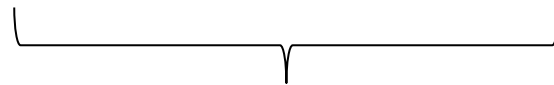
But

- identifier des biomarqueurs prédictifs de l'athérogénèse sensibles aux régimes;



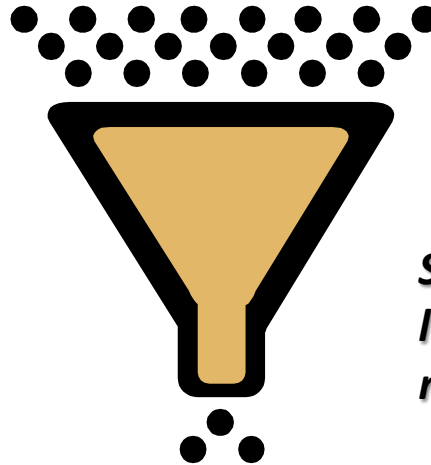
Lésions peu étendues

Lésions étendues



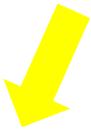
Marqueurs biologiques des lésions ($f(\text{régimes})?$)

1666 variables biologiques/hamster



*Sélection selon critère VIP de
l'analyse discriminante
multivariée PLS*

127 variables biologiques discriminantes



38 métabolome
urine



38 metabolome
plasma



23 acides gras
(PL, CE, TG,
Lipides totaux)



12 biochimie
sang



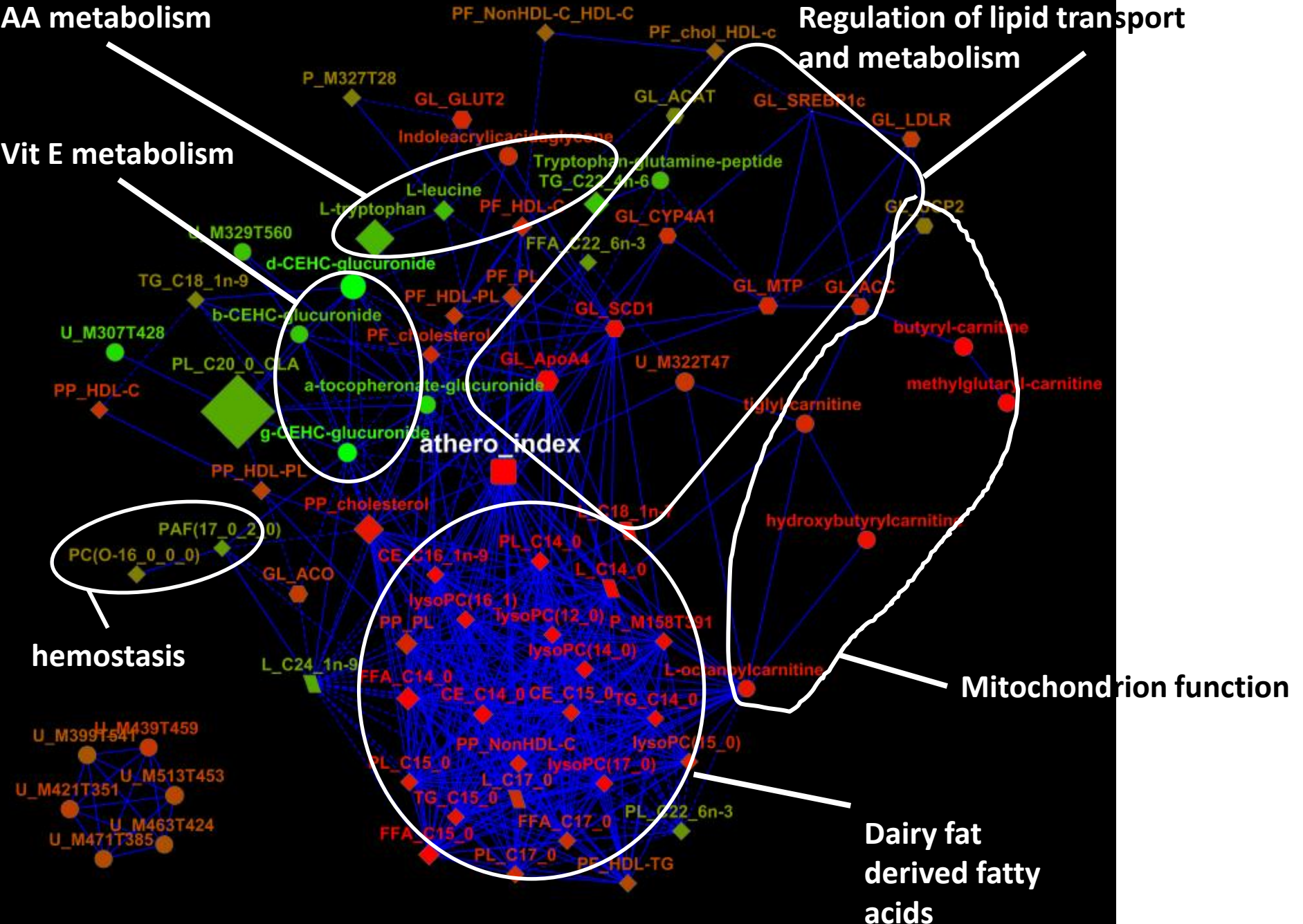
16 gènes

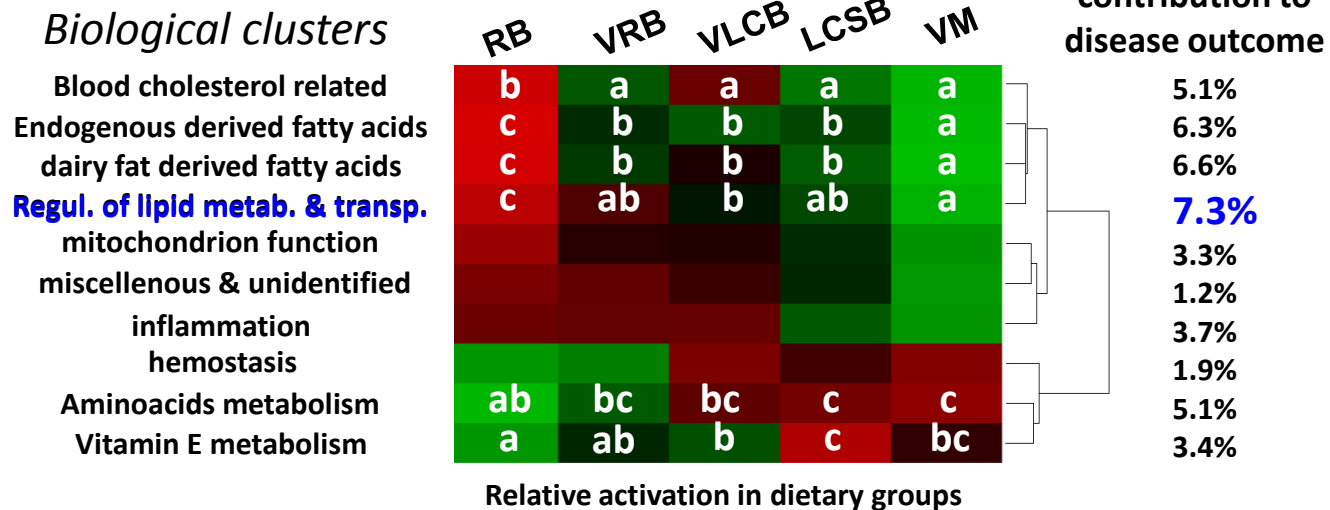
AA metabolism

Regulation of lipid transport and metabolism

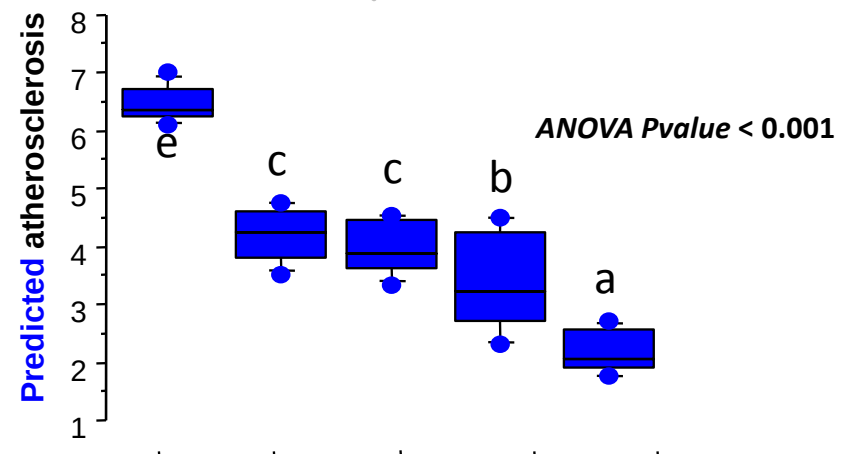
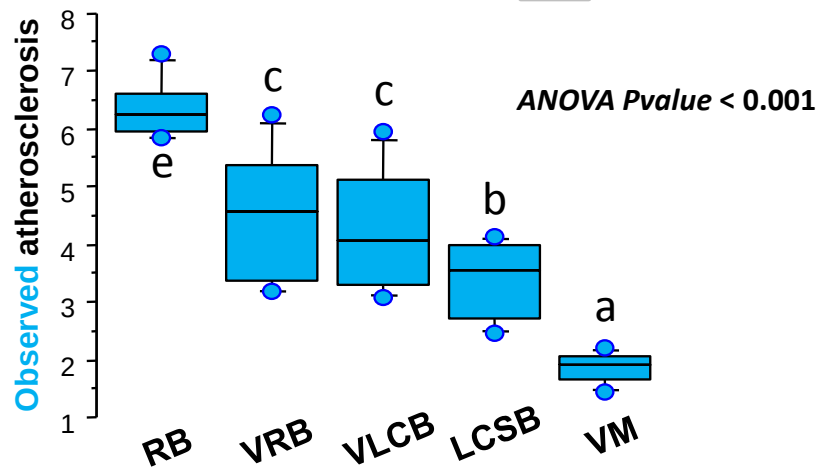
Vit E metabolism

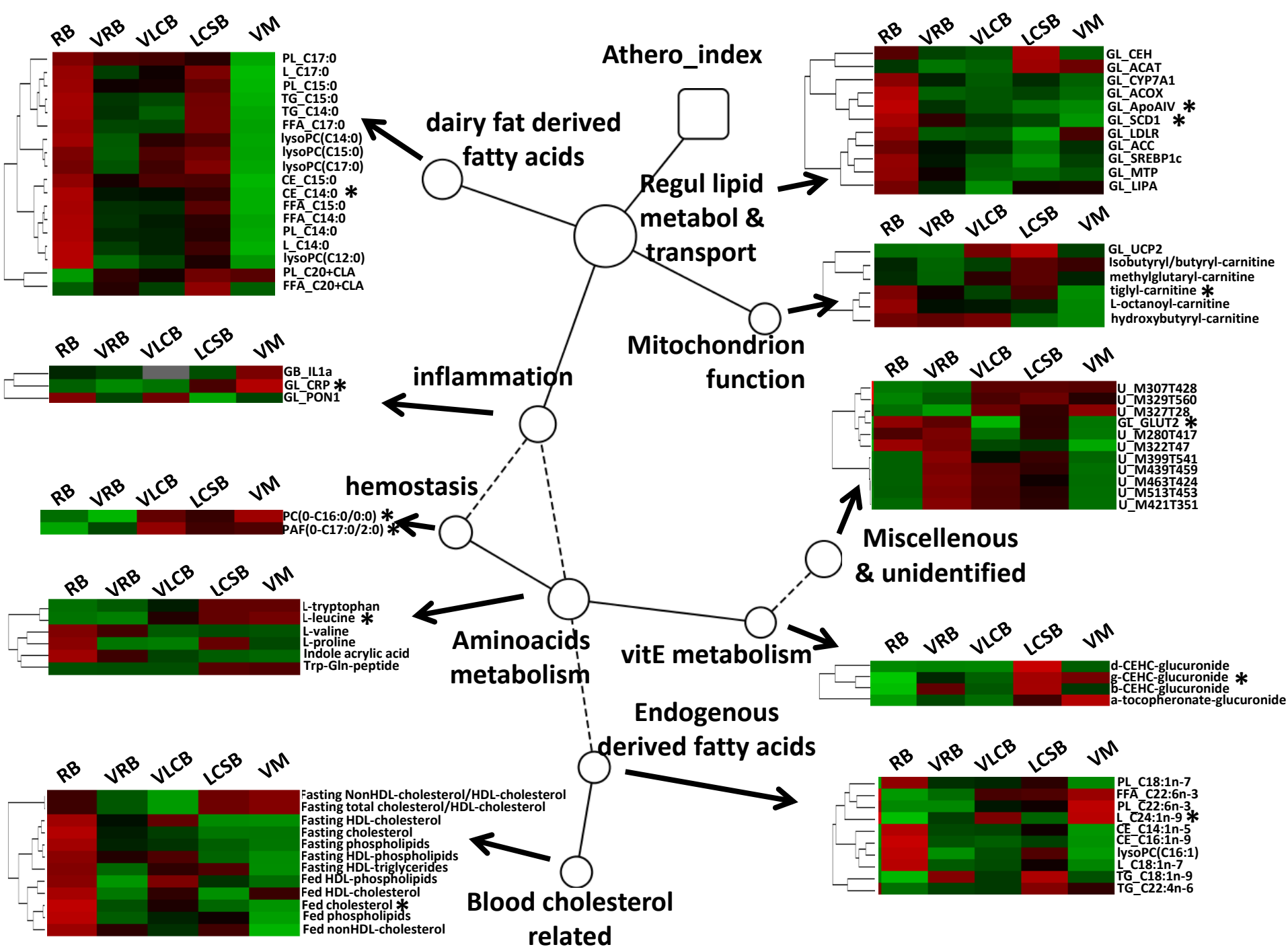
hemostasis





$$\begin{aligned}
 \text{Predicted Atherogenicity} = & 0.108091 * [\text{dairy fat derived fatty acids}] + 0.152669 * [\text{endogenous derived fatty acids}] - \\
 & 0.0340185 * [\text{mitochondrion function}] + 0.173429 * [\text{regul. Lipid metabol. Trans.}] - 0.0751566 * [\text{vitaminE metabolism}] \\
 & - 0.153142 * [\text{hemostasis}] - 0.110269 * [\text{aminoacids metabolism}] + 0.110269 * [\text{blood cholesterol related}] + \\
 & 0.231903 * [\text{inflammation}] - 0.00456344 * [\text{miscellenous \& unidentified}] + \mathbf{4.06731}.
 \end{aligned}$$





Quelques applications en sélection cultivars, origine produits, qualité

Table 1. A selection of metabolomic studies on common foods summarizing the purpose of these studies, the methodologies used and the major conclusions reached

Purpose	Plant/food	Sample preparation	Analysis	Data analysis	Outcome	Reference
Metabolite diversity/ breeding	Raspberries	Acetic acid/acetonitrile	DIMS-LC-MS	PCA	Variation attributed to anthocyanins	Stewart et al. [26]
Metabolite diversity/ breeding	Tomatoes	MeOH with H ₂ O/ chloroform partition	GC-MS	PCA	Identification of 889 metabolic loci, metabolites associated with yield trait	Schauer et al. [24]
Metabolite diversity/ breeding	Potatoes	MeOH with H ₂ O/ chloroform partition	GC-MS	PCA Correlation	Detected variation between cultivars and landraces	Dobson et al. [27, 28]
Geographic origin	Grapes/wine	95% EtOH	NMR	PCA/PLS-DA	Grapes from different regions can be distinguished	Son et al. [69]
Geographic origin	Coffee	MeOH/H ₂ O	GC-FID/LCMS	PCA	Samples from Asia, Africa, and S. America could be distinguished	Choi et al [35]
Geographic origin	Green tea	Hot H ₂ O	NMR	PCA/OPLS-DA	Growing region with high temperature, rainfall and sun increased the anine concentrations	Lee et al. [70]
Geographic origin	Olive oil	SPME	GC-MS	Linear discriminant analysis	Distinguished from different productions areas within a region	Cavaliere et al. [36]
Quality	Green tea	MeOH/H ₂ O/chloroform	GC-MS/LC-MS	PCA/OPLS-DA	Tea grown on greater shade had higher umami and less astringency and these attributes were related to sugars, amino acid and phenolics	Ku et al. [71]
Postharvest disorder	Apples	Lyophilisation/SPME/ extraction H ₂ O:MeOH	GC-MS	PCA/PLS (with validation)	Metabolites related to sun/shade treatments and mouth-feel parameters	Rudell et al. [38]
Sensory	Tomatoes		GC-MS and LC-MS			Thissen et al. [21]
Sensory	Wine/grapes	Wine directly added to NMR tubes	NMR	PCA/PLS-DA		Rochfort et al. [39]
GM assessment	Tomatoes	80% MeOH/H ₂ O	GC-MS, LC-MS, CE-MS	ANOVA/PCA	GM lines were within traditional cultivar variation	Kusano et al. [41]

Characterization and classification of Western Greek olive oils according to cultivar and geographical origin based on volatile compounds

Eirini Pouliarekou^a, Anastasia Badeka^a, Maria Tasioula-Margari^a, Stavros Kontakos^b,
Francesco Longobardi^c, Michael G. Kontominas^{a,*}

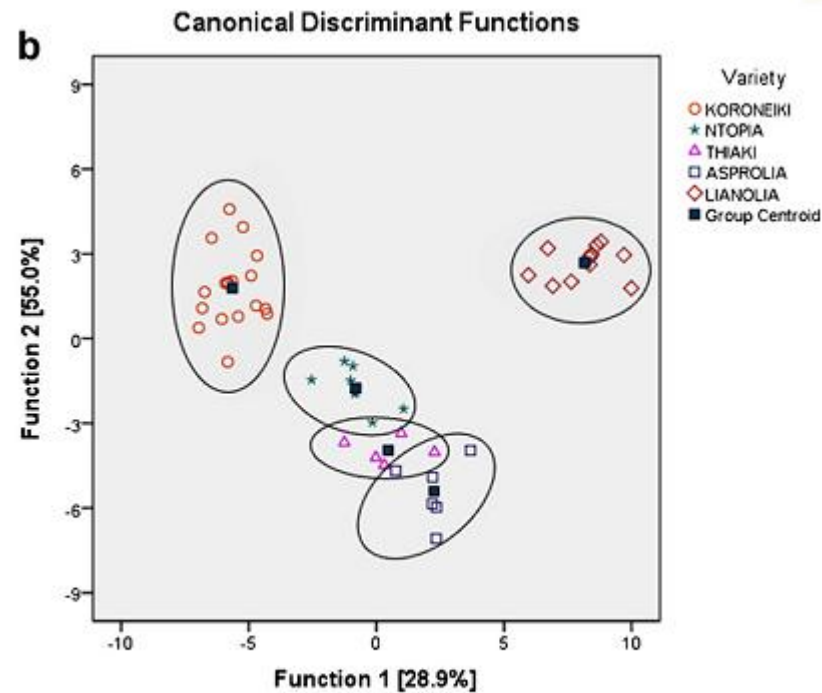
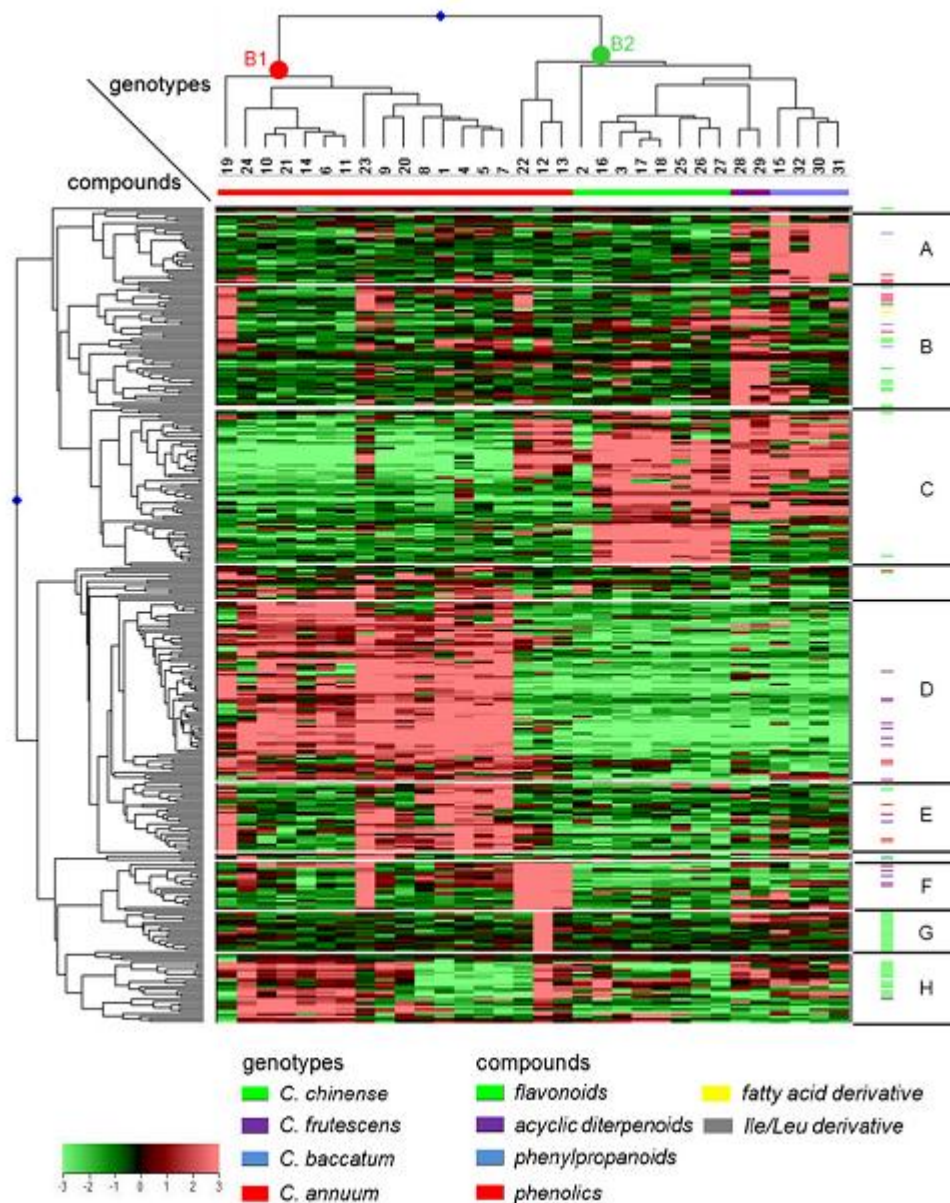


Fig. 3. Classification of olive oil samples from Western Greece according to cultivar using (a) 53 variables and (b) 19 variables.

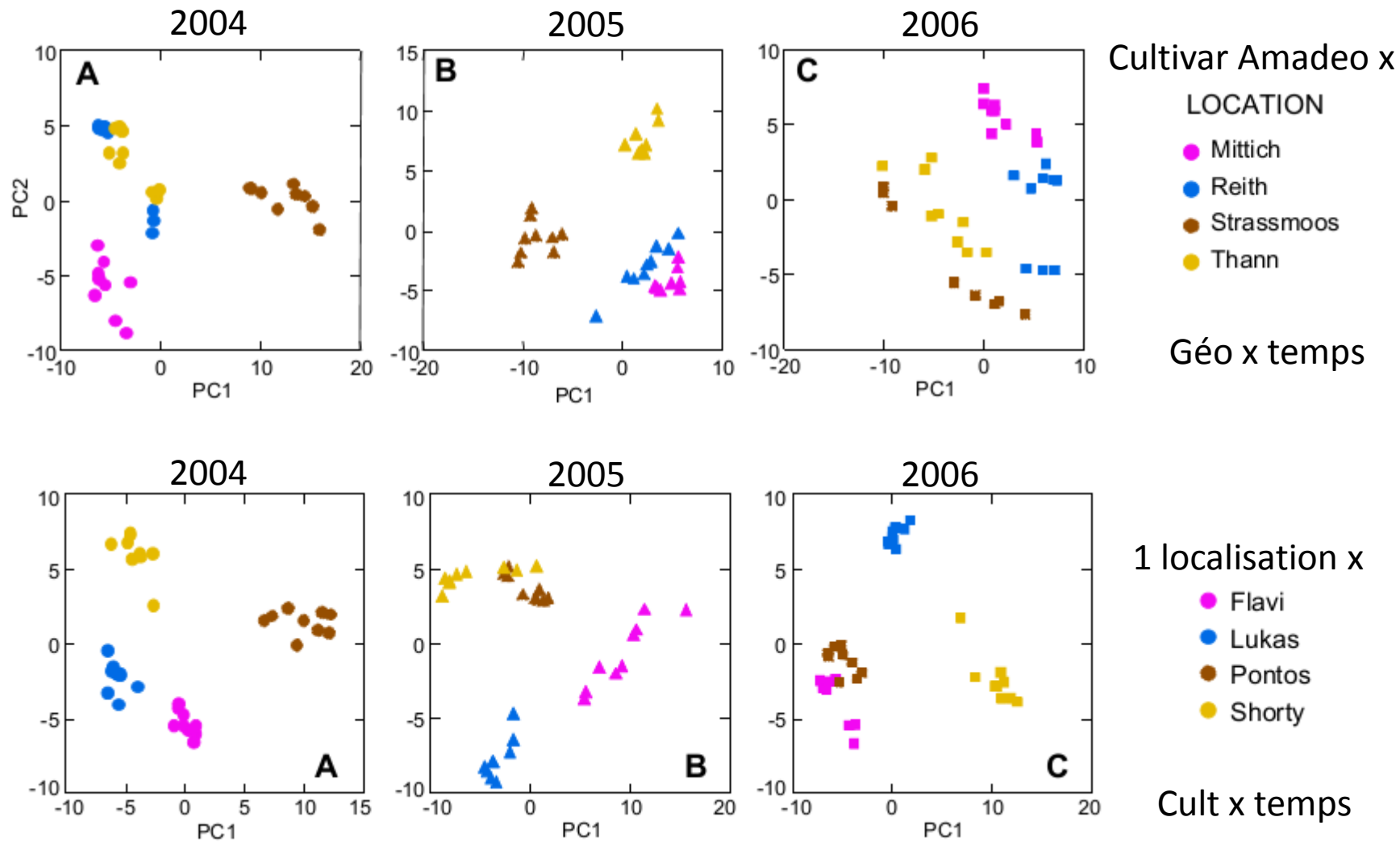
Metabolomics and molecular marker analysis to explore pepper (*Capsicum* sp.) biodiversity

Yuni Wahyuni · Ana-Rosa Ballester · Yury Tikunov ·
Ric C. H. de Vos · Koen T. B. Pelgrom · Awang Maharijaya ·
Enny Sudarmonowati · Raoul J. Bino · Arnaud G. Bovy
Metabolomics (2013) 9:130–144



Metabolite profiling of maize grain: differentiation due to genetics and environment

Richard M. Röhlig · Joachim Eder ·
Karl-Heinz Engel



Apple Peels, from Seven Cultivars, Have Lipase-Inhibitory Activity and Contain Numerous Ursenoic Acids As Identified by LC-ESI-QTOF-HRMS

Tony K. McGhie,* Sébastien Hudault, Rona C. M. Lunken, and John T. Christeller

microOTOF QII mass spectrometer

Table 3. Relative Amount of Each Compound Measured Normalized to Total Peak Area

no.	compound name	average	'Braeburn'	'Granny Smith'	'Sciros'	O.P. 'Red Field'	'Cripps Pink'	'Fuji'	'Scilate'
1	dihydroxy-urs-12-en-28-oic acid	0.13	0.14	0.00	0.22	0.24	0.07	0.12	0.07
2	unknown	1.64	1.37	0.13	0.53	1.24	6.39	0.73	0.59
3	unknown	0.76	0.44	1.21	0.48	2.14	0.53	0.26	0.26
4	unknown	1.76	1.04	1.60	1.77	4.65	1.26	0.73	1.01
5	trihydroxy-urs-12-en-28-oic acid	2.48	2.03	1.88	0.63	0.49	4.10	3.16	5.34
6	trihydroxy-urs-12-en-28-oic acid	0.11	0.28	0.01	0.26	0.24	0.00	0.00	0.00
7	unknown	2.50	1.00	1.91	3.63	6.64	0.75	0.77	2.49
8	unknown	3.45	2.75	2.56	3.07	5.14	4.18	3.34	2.56
9	tetrahydroxy-urs-12-en-28-oic acid	1.26	5.28	0.08	1.16	1.68	0.38	0.20	0.23
10	unknown	0.93	1.25	0.88	0.35	1.84	0.74	0.78	0.64
11	3-oxo-1,19 α -dihydroxy-urs-12-en-28-oic acid (annurcoic acid)	8.39	12.32	5.08	9.27	8.90	8.79	6.40	7.80
12	unknown	3.35	5.22	3.35	1.02	5.69	2.77	3.35	2.04
13	unknown	2.25	3.44	2.33	1.34	3.30	1.49	2.29	1.62
14	3-oxo-1 α -hydroxy-urs-12-en-28-oic acid	2.14	4.91	1.25	2.02	1.98	2.21	1.21	1.55
15	3 β -p-coumaroyloxy-dihydroxy-urs-12-en-28-oic acid	1.54	1.53	0.17	0.23	0.51	1.70	2.60	4.01
16	3 β ,19 α -dihydroxy-urs-12-en-28-oic acid (pomolic acid)	1.86	1.73	1.66	1.96	1.78	1.69	1.62	2.65
17	3-oxo-hydroxy-urs-12-en-28-oic acid	3.69	5.38	4.98	5.45	2.74	2.58	2.21	3.20



	'Braeburn'	'Granny Smith'	'Sciros'	O.P. 'Red Field'	'Cripps Pink'	'Fuji'	'Scilate'
total ursenoic acids peak area	3,630,859	3,110,778	4,243,395	4,381,954	4,207,580	4,753,673	3,507,671
% lipase inhibition	39	71	60	64	70	60	50
27	unknown	0.00	0.00	0.00	0.00	0.00	0.00
28	oxo-urs-12-en-28-oic acid	0.59	0.90	0.61	0.81	0.37	0.49
29	unknown (dihydroxy-urs-12-en-28-oic acid)	1.91	3.13	1.73	2.38	1.86	1.39
30	oxo-urs-12-en-28-oic acid	1.96	3.21	2.04	3.39	1.33	1.16
31	3-oxo-hydroxy-urs-12-en-28-oic acid	2.56	8.63	0.00	0.00	2.21	2.01
32	3-oxo-hydroxy-urs-12-en-28-oic acid	2.04	0.00	1.93	2.97	2.57	1.94
33	3 β -?-p-coumaroyloxy-hydroxy-urs-12-en-28-oic acid	2.00	1.08	1.94	2.83	1.21	1.49
34	3 β -trans-p-coumaroyloxy-2 α -hydroxy-urs-12-en-28-oic acid	6.34	3.67	4.94	8.94	4.18	5.29
35	3 β -cis-p-coumaroyloxy-2 α -hydroxy-urs-12-en-28-oic acid	4.35	1.33	7.18	5.76	2.32	3.55
36	oxo-urs-12-en-28-oic acid	2.83	3.39	4.29	4.20	2.25	1.63
37	3 β -hydroxy-lup-20(29)-en-28-oic acid (betulinic acid)	6.03	3.80	9.62	5.85	4.92	6.76
38	3 β -hydroxy-urs-12-en-28-oic acid (ursolic acid)	5.41	4.48	8.56	5.36	3.95	5.56
39	unknown	1.29	0.31	0.00	0.07	4.40	2.65
40	unknown	2.26	1.05	3.74	1.90	1.79	2.63
41	unknown	6.49	3.78	9.76	6.10	5.40	7.19
42	3 β -hydroxy-urs-12-en-28-ol (uvaol)	5.06	1.53	6.42	2.53	3.23	5.70
43	unknown	1.70	0.86	1.39	1.36	1.25	2.40
total ursenoic acids peak area	3,630,859	3,110,778	4,243,395	4,381,954	4,207,580	4,753,673	3,507,671
% lipase inhibition	39	71	60	64	70	60	50

Metabolomics-Driven Nutraceutical Evaluation of Diverse Green Tea Cultivars

Yoshinori Fujimura^{1,3}, Kana Kurihara^{2,3}, Megumi Ida², Reia Kosaka², Daisuke Miura¹, Hiroyuki Wariishi^{1,2,3}, Mari Maeda-Yamamoto⁴, Atsushi Nesumi⁴, Takeshi Saito⁵, Tomomasa Kanda⁵, Koji Yamada², Hirofumi Tachibana^{1,2,3*}

Analyse métabolomique de différents cultivars de thé vert

Test biologique (dysfonction vasculaire in vitro) des différents cultivars de thé vert

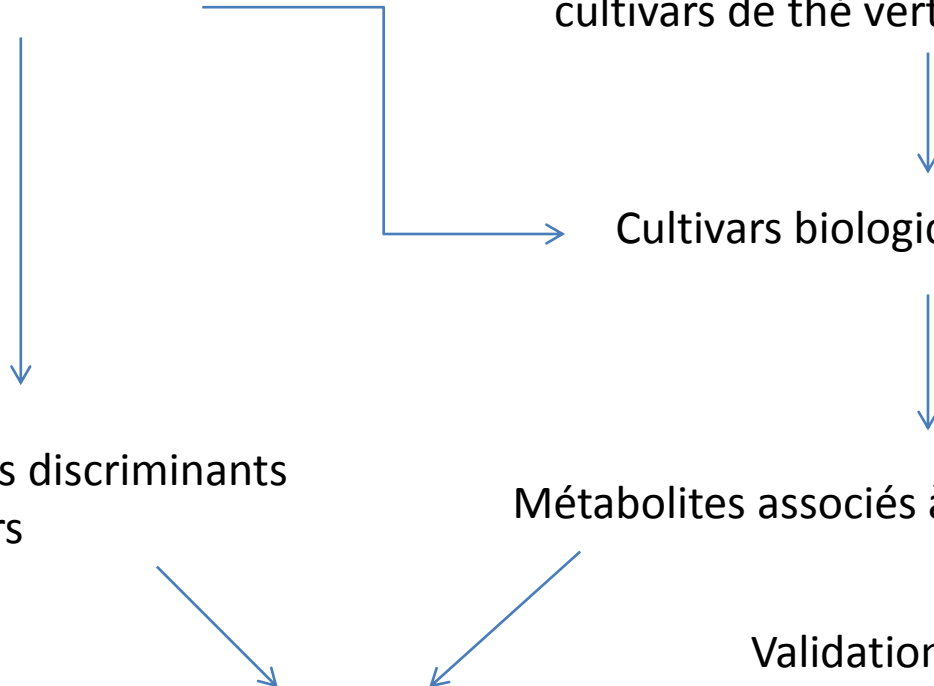
Cultivars biologiquement actifs/non actifs

Métabolites discriminants des cultivars

Métabolites associés à l'activité biologique

Cultivars possédant les métabolites actifs

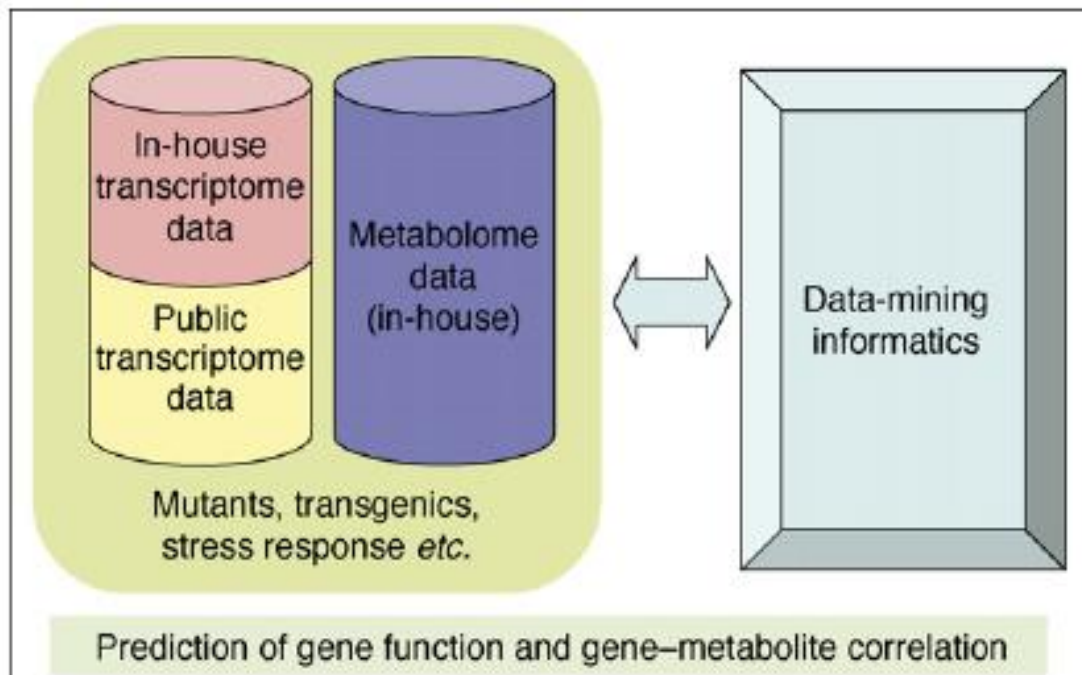
Validation = transfert de l'activité biologique par ajout des métabolites suspectés actifs à un cultivar initialement inactif (ex: importance bio THEANIN)

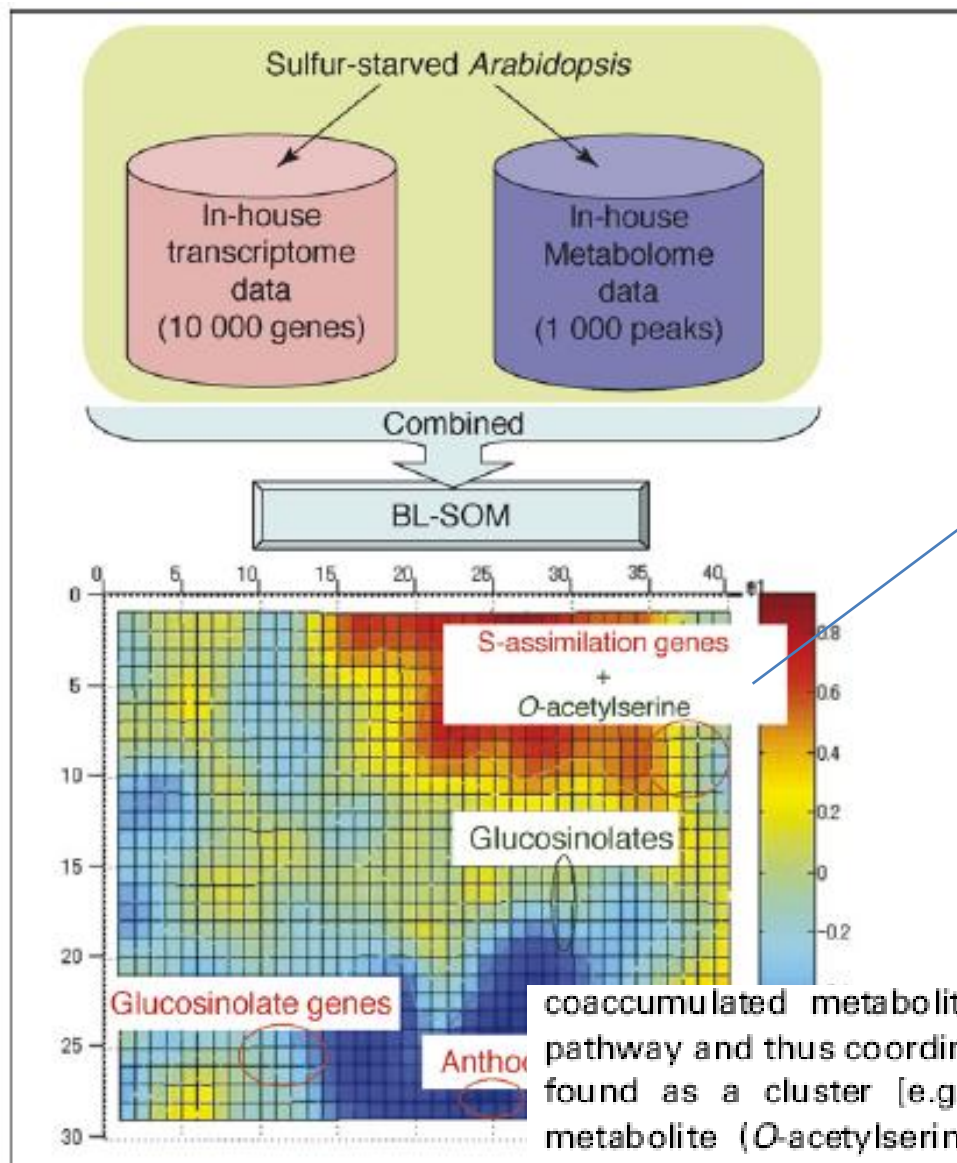


Decoding genes with coexpression networks and metabolomics – ‘majority report by precogs’

TRENDS in Plant Science Vol.13 No.1

Kazuki Saito^{1,2}, Masami Y. Hirai¹ and Keiko Yonekura-Sakakibara¹



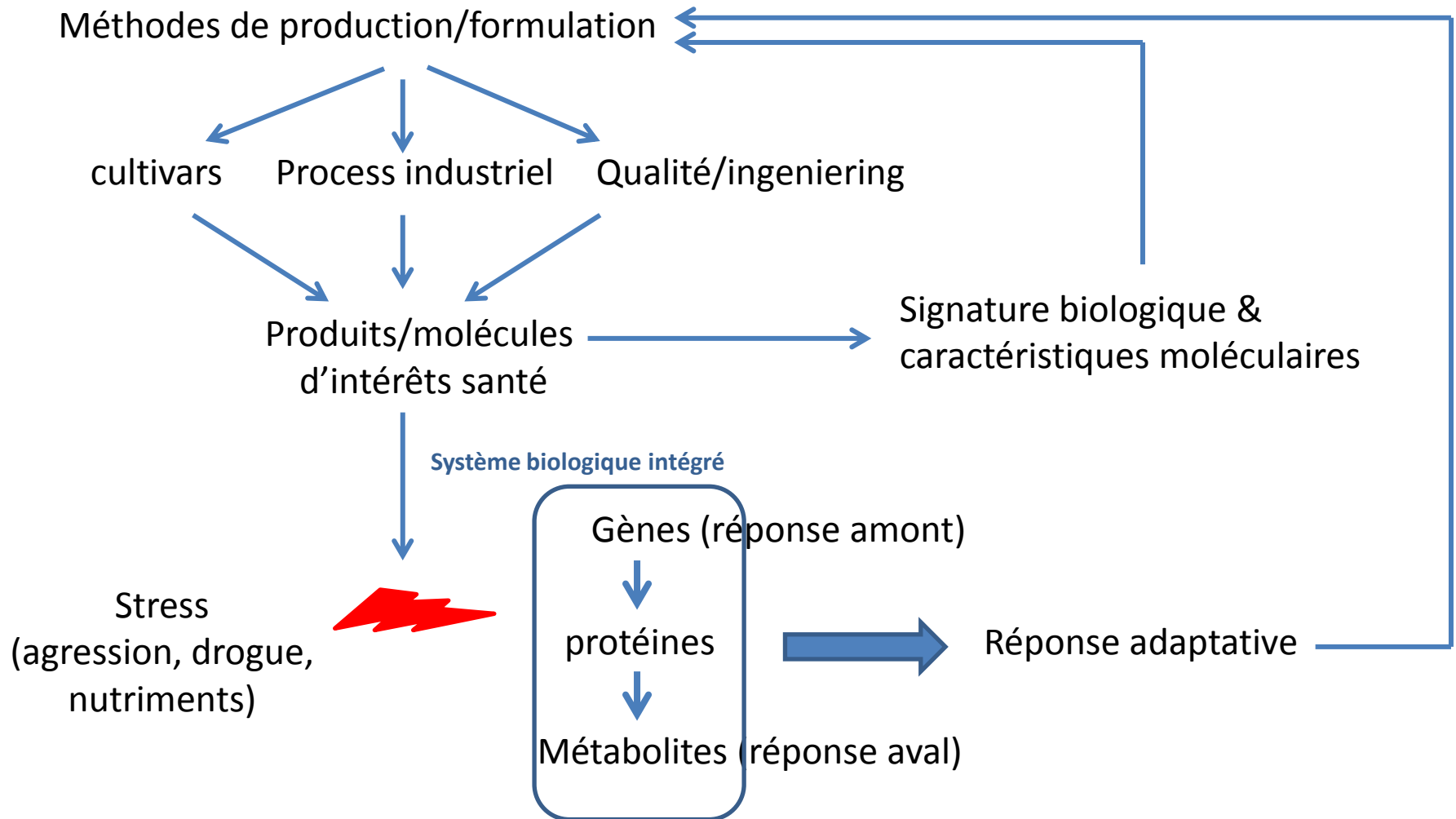


Co-expression de gènes associés à l'assimilation du soufre avec O-acétylsérine

Fonction de gènes inconnus

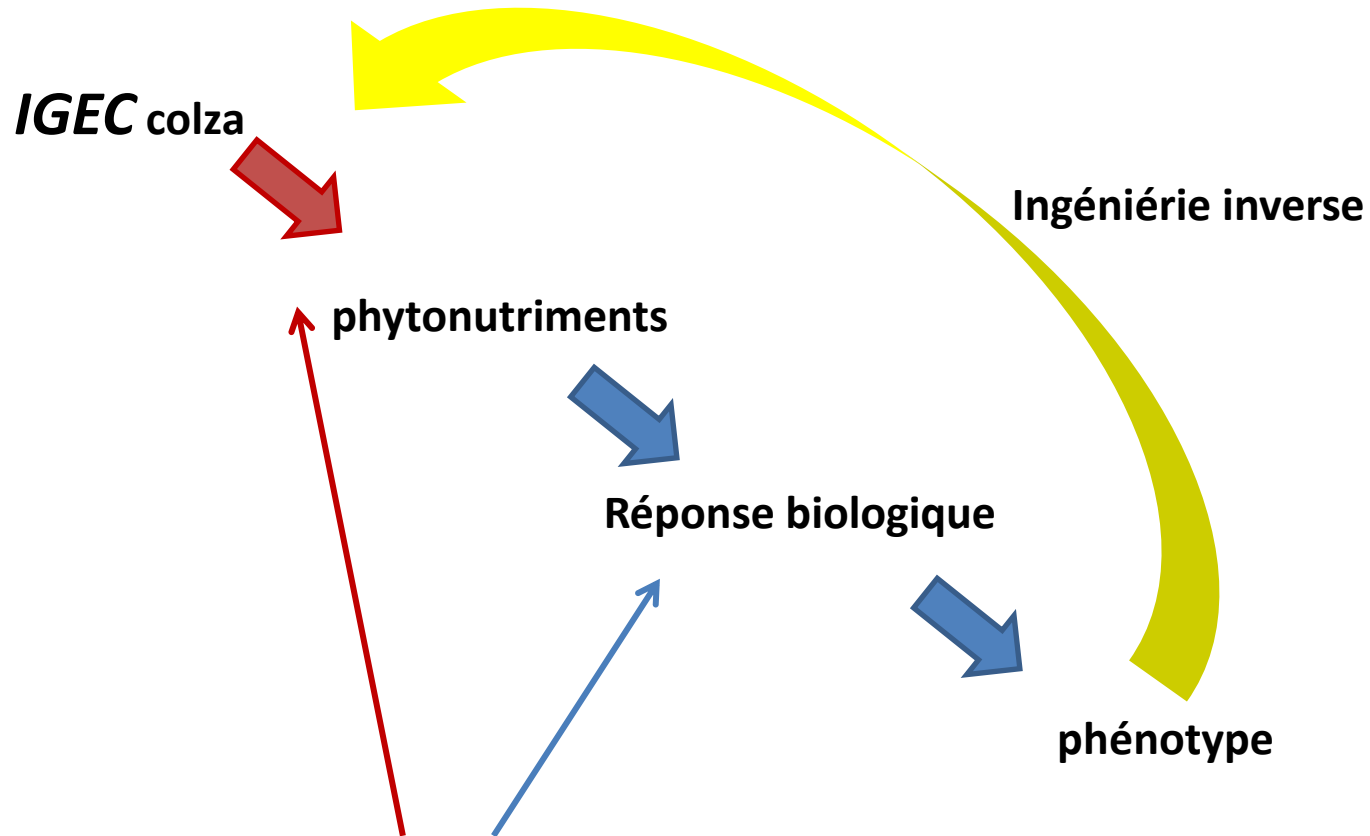
coaccumulated metabolites. A set of genes involved in the same metabolic pathway and thus coordinately regulated under the sulfur-starved condition were found as a cluster [e.g. genes of glucosinolate biosynthesis, genes and a metabolite (*O*-acetylserine) of the sulfur assimilation pathway and genes of anthocyanin biosynthesis]. These clusters contained novel genes for enzymes and transcription factors to be further characterized [17,20]. *O*-Acetylserine, an intermediary and regulatory metabolite of the sulfur assimilation pathway, was found in the same cluster as the genes of this pathway [17].

CRIBIOM, une plate-forme de criblage biologique



Projet « Polygone »

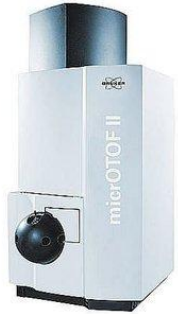
Onidol, PIVERT, U-Caen, AMU
GENESYS/SAS PIVERT



« omiques »

CRIBIOM

Métabolomique et lipidomique



LC QTOF



GCMS



LC orbitrap



fastGCFID

Transcriptomique



Plate forme microarray (IFR)

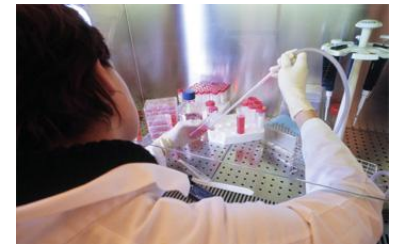


QPCR

Plate-formes expérimentales

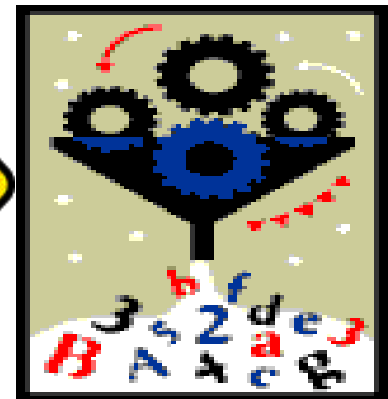


animalerie



Cultures cellulaires

Bioinformatique, statistiques et modélisation in silico



Quels sont les éléments qui fragilisent encore le développement de la métabolomique?



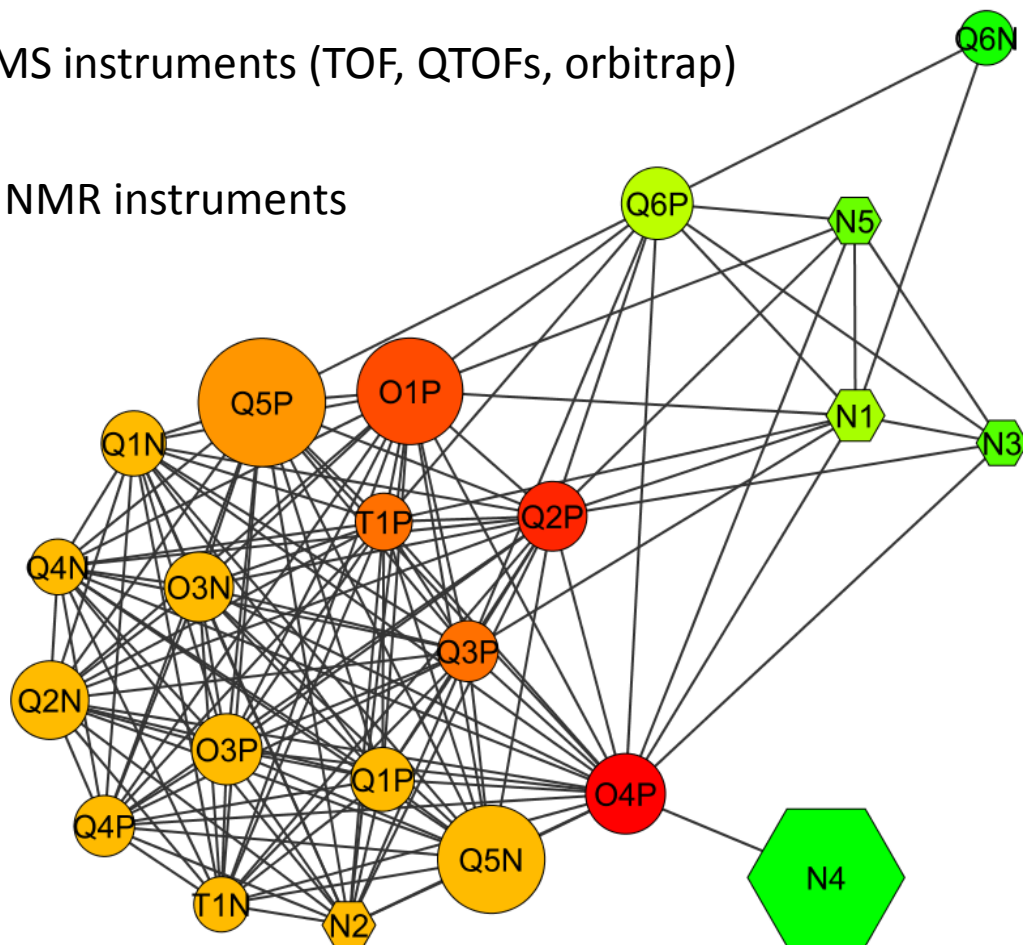


Reproductibilité inter-plateforme et standardisation



○ MS instruments (TOF, QTOFs, orbitrap)

⬡ NMR instruments



Metabo
Ring



Despite a large differences in the number of features among the instruments, the heterogeneity in the analytic conditions and data post-processing, the spectral information within (NMR and MS) and across methods (NMR vs MS) was highly converging (from 64% to 91% on average). No effect of the MS configuration (TOF, QTOF, Orbitrap) was noticed



Facteurs humains: scepticisme et conservatisme des biologistes, formation insuffisante



Les progrès à attendre

De l'identification manuelle à
l'identification automatique à
haut débit

De la découverte à la validation de nouveaux
biomarqueurs

Du relatif au quantitatif

Des biomarqueurs individuels aux
biomarqueurs multiplexes

Du statique au dynamique

Des petites cohortes à
l'épidémiologie

De centaines de
composés à des milliers

Du laboratoire R&D à la
clinique

De la preuve de principe
à la routine

Du groupe au
metabotype individuel

De l'instrument dépendant à
l'universel

Du descriptif au mécanistique

Du spectral à l'imagerie

De l'interrogation manuel à
l'automatisation

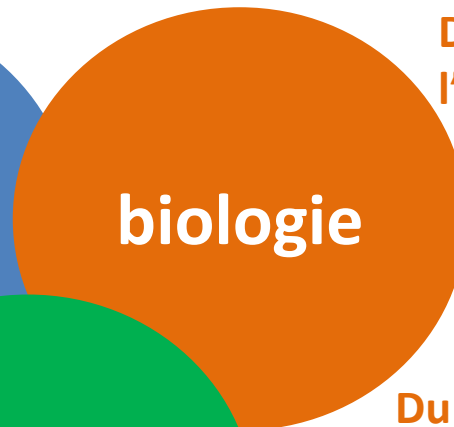
Des voies métaboliques aux fonctions
biologiques

Des outils maisons aux outils universels

D'un accès restreint à un accès universel
(visualisation)

Des statistiques à l'intégration au
text-mining et à l'interprétation

De la métabolomique à l'intégration omique





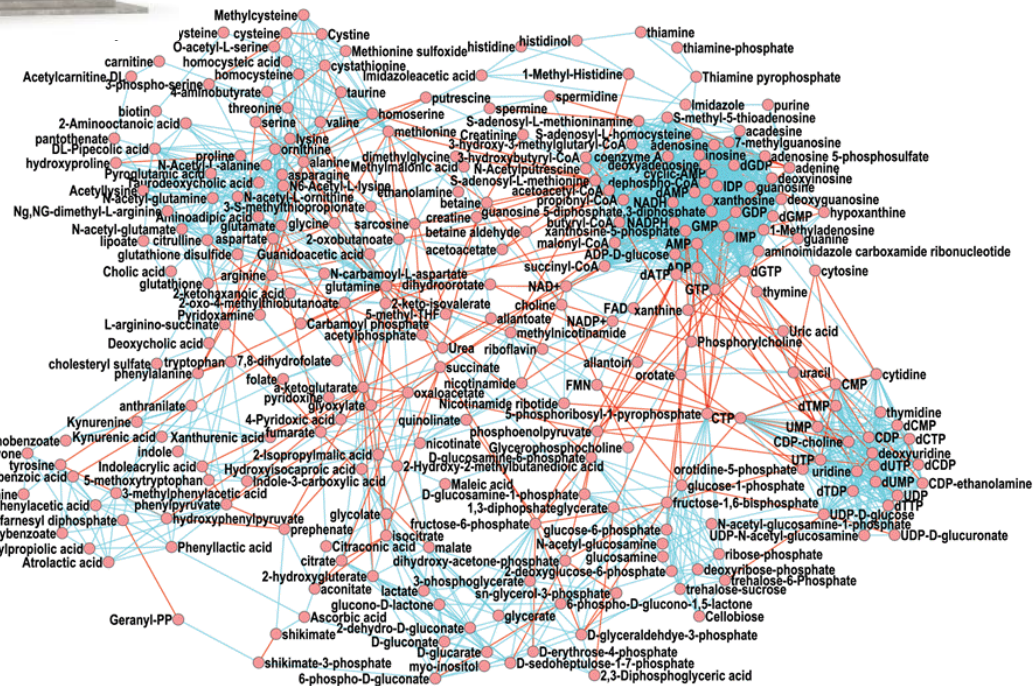
Spectro de masse



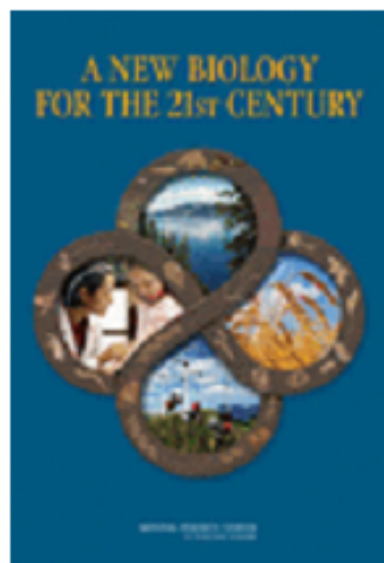
Banque de données



Identifier et annoter







A New Biology for the 21st Century

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Que pensez-vous de la métabolomique?

•A:

Très bien

•B:

Bien

•C:

Moyen

•D:

À supprimer