

MULTI-OMICS INTEGRATION

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MDAW workshop: multi-domain data integration

21st April 2021



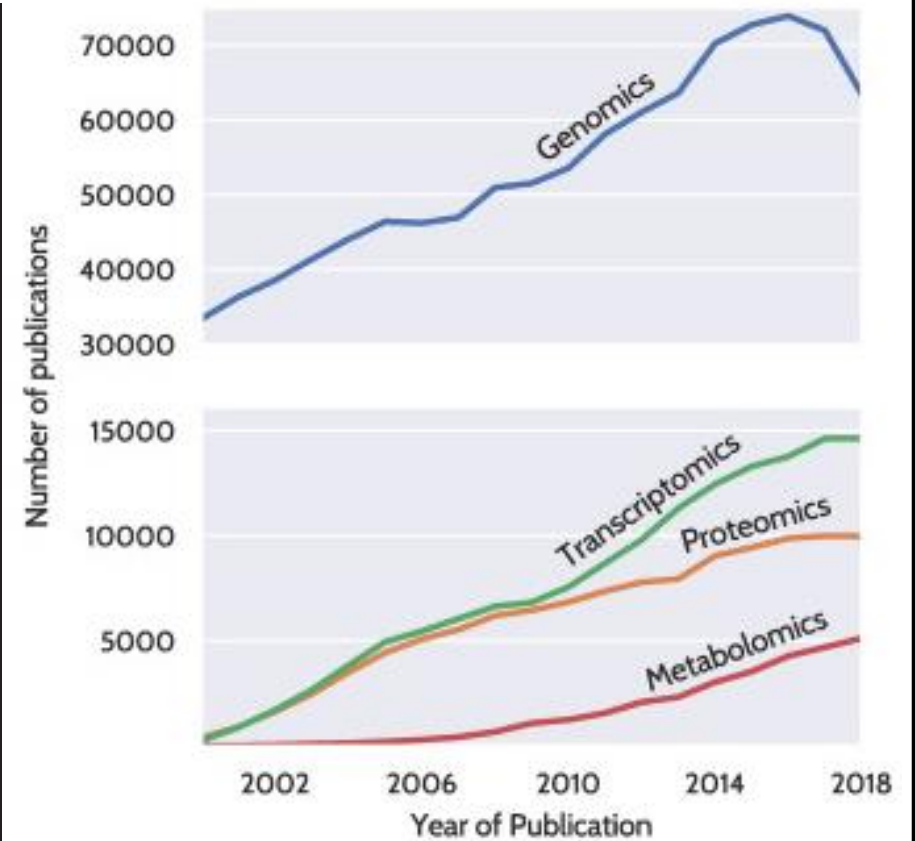
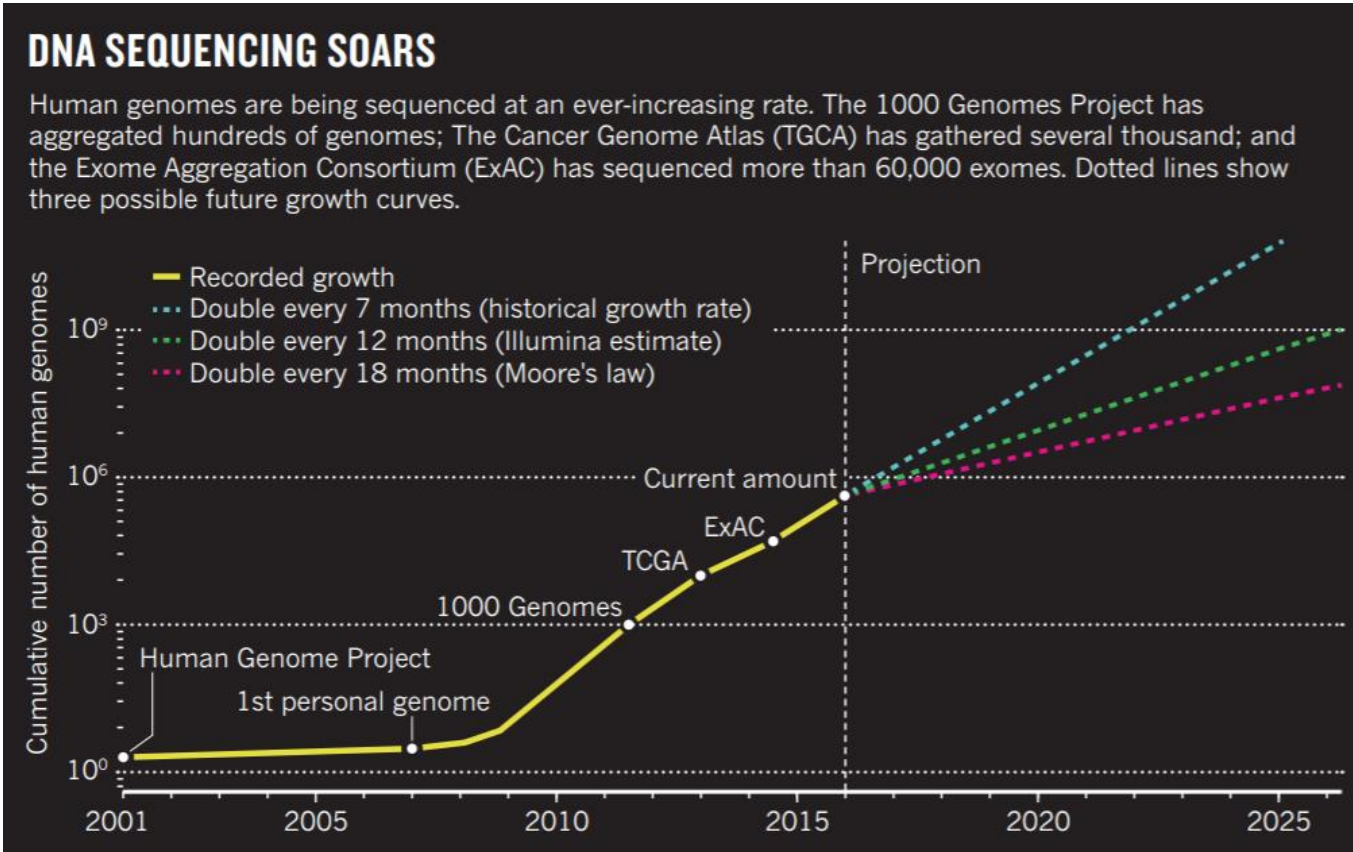
Outline

- **The multi-omics era**
- Methods
- Example 1
- Example 2
- The microbiome and the multi-omics

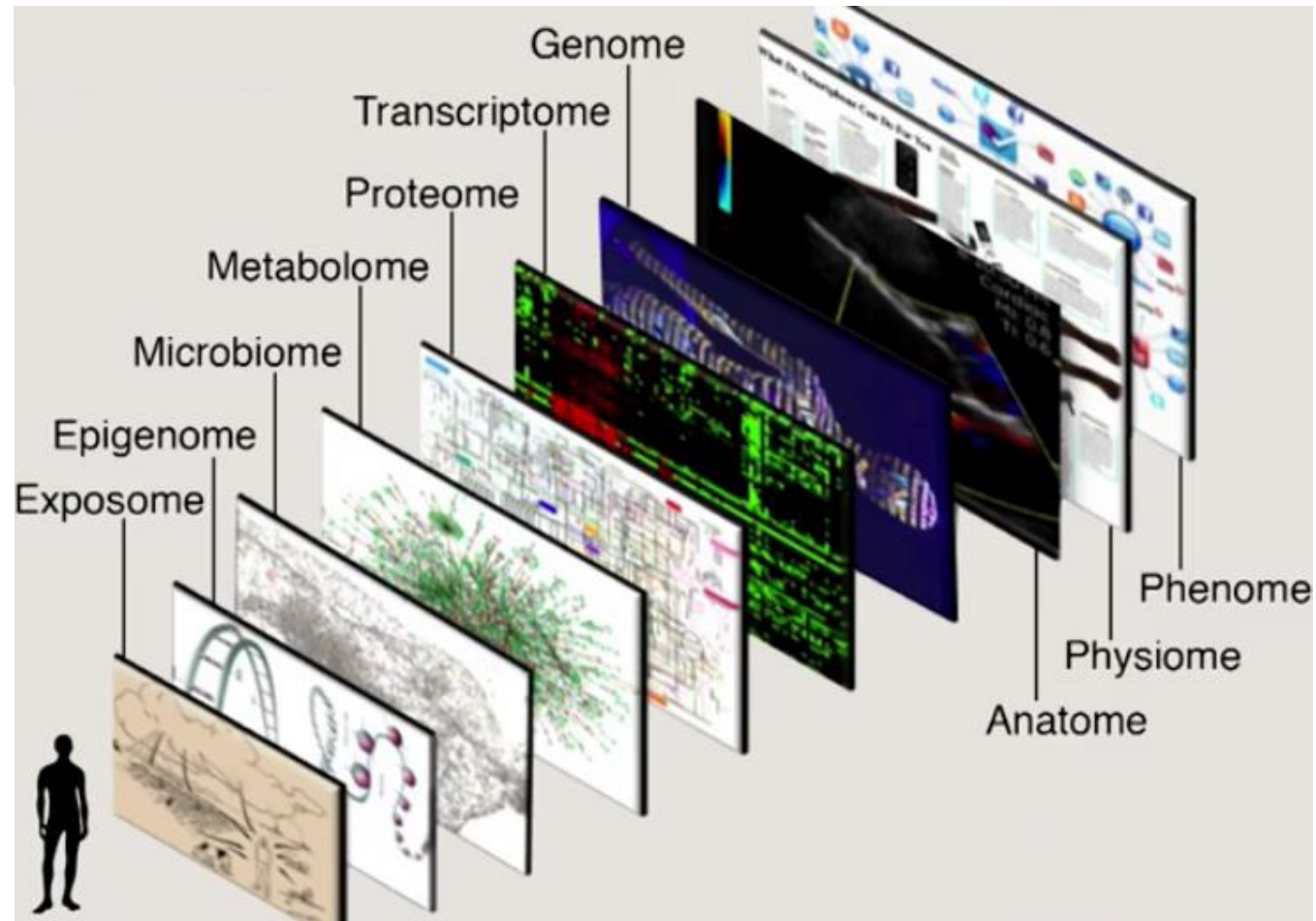




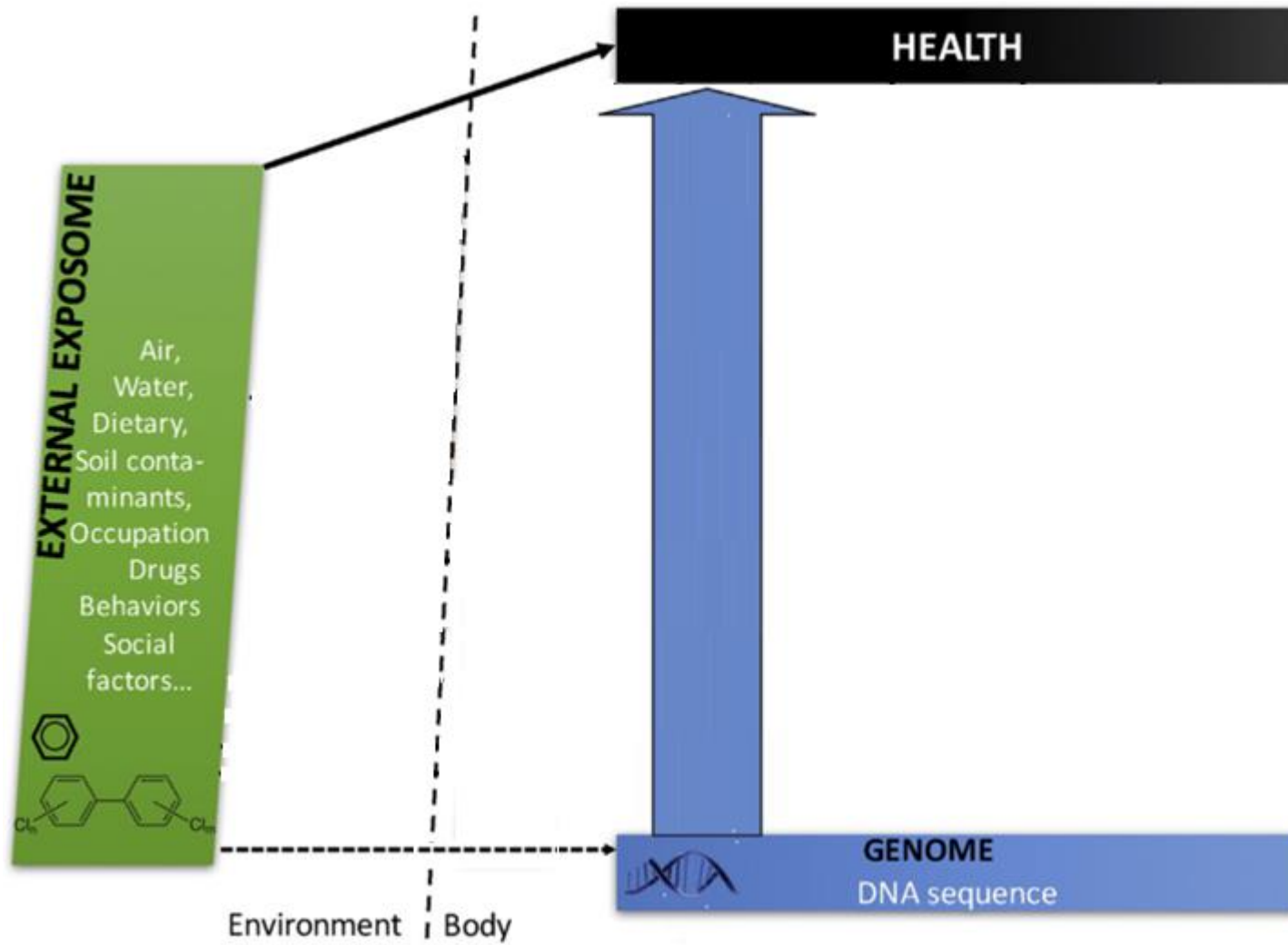
Big data



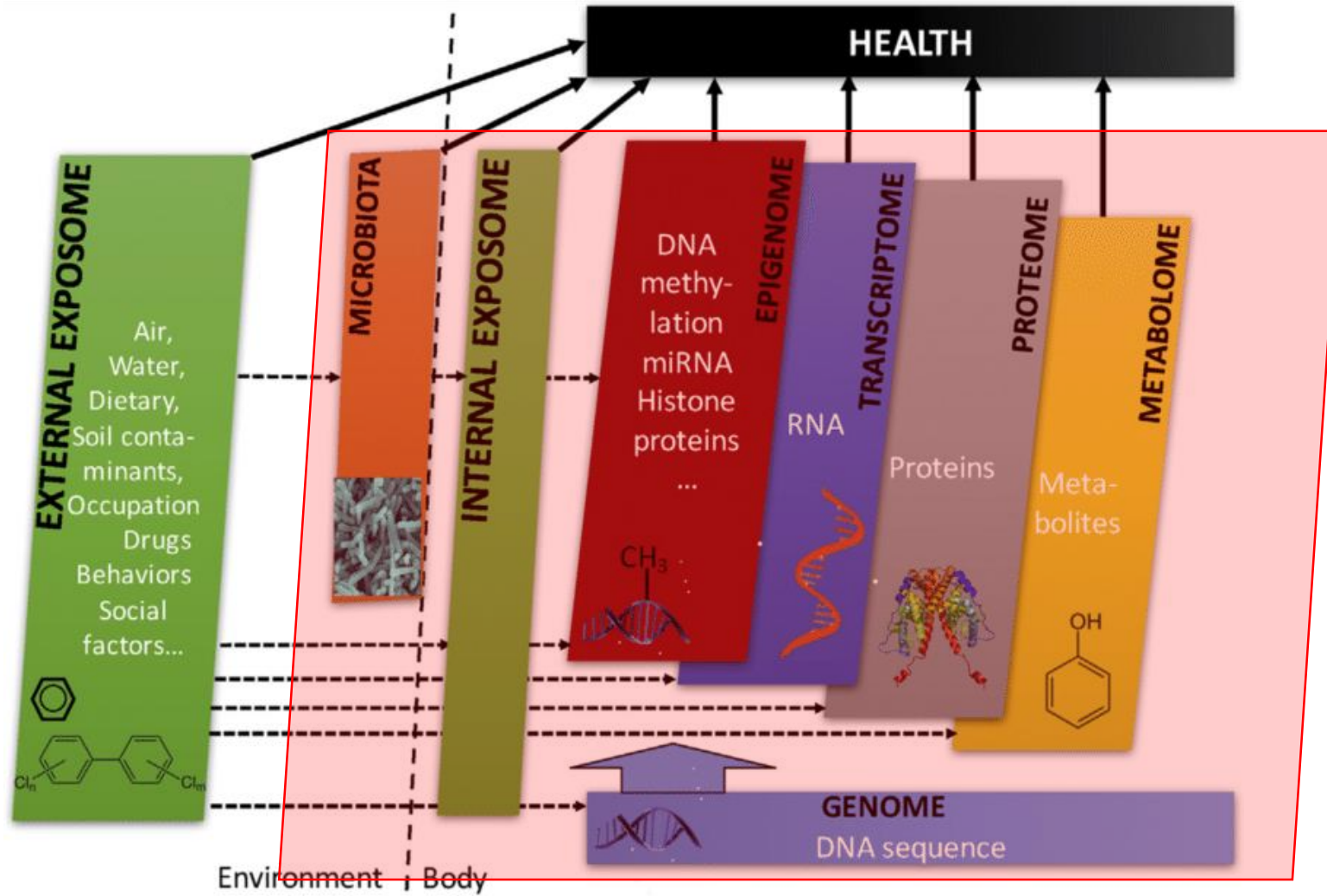
The human GIS



Omic



Omic



OMICS



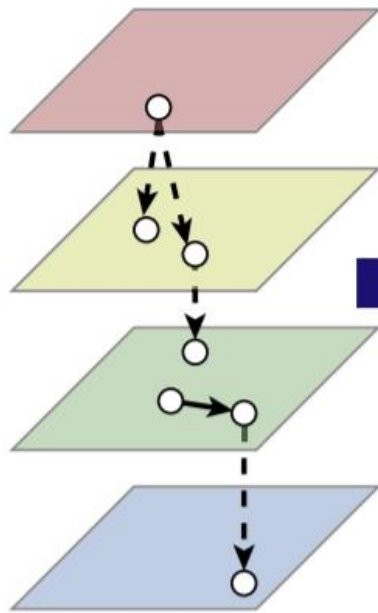
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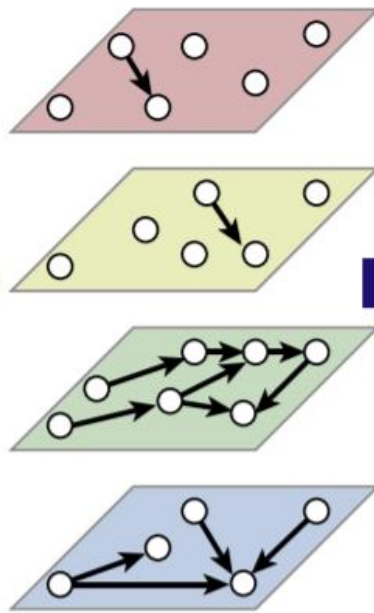
Conventional biology vs multi-omics

Conventional molecular biology



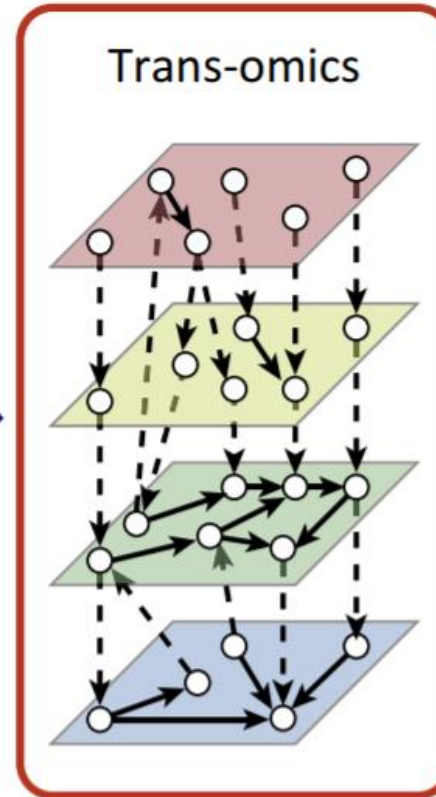
reductionist

Single omics



hypothesis free

Trans-omics



holistic

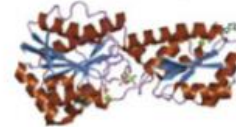
Genome



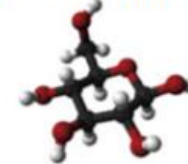
Transcriptome



Proteome

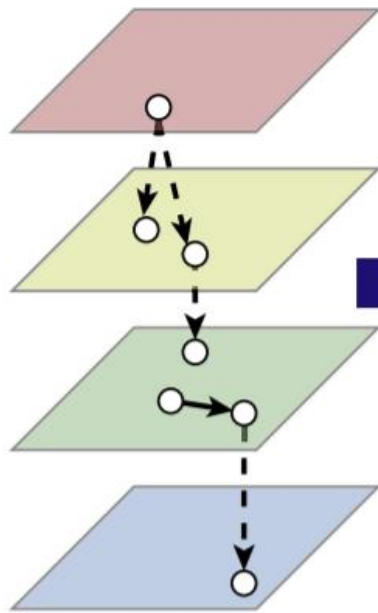


Metabolome



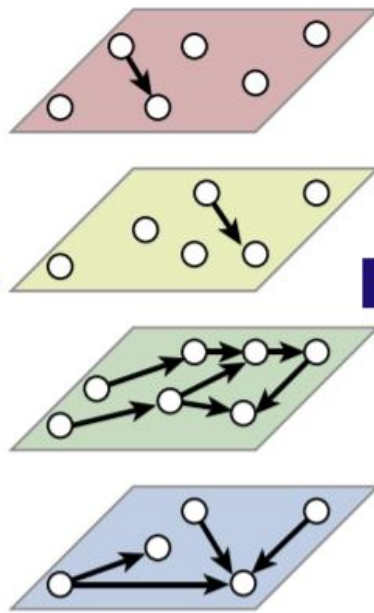
Conventional biology vs multi-omics

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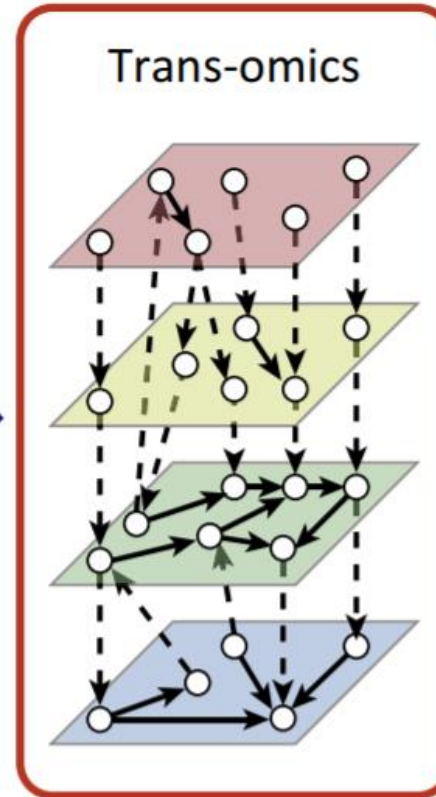
reductionist

Single omics



hypothesis free

Trans-omics



holistic

complementary approach

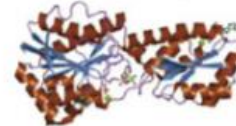
Genome



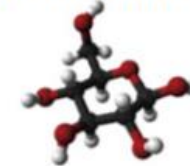
Transcriptome



Proteome





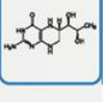


Metabolome



Omics integration challenges

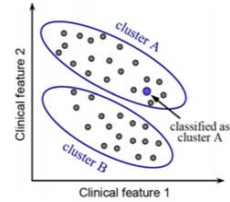
- Data collected from different sources and **techniques**
- **Biological** heterogeneity
- High data **dimensionality**:
 - Small sample size ($N < p$)
 - Data dimensionality different by omic type
- **Missing values**
- **Methods** for integrative analyses

	Supporting Structure	Platforms (log ₁₀ order of magnitude)	Features
	DNA	Microarrays (6) Sequencing (9)	Categorical data Distance-driven correlation Extremely stable over time
	DNA methylation Histone modifications Non-coding RNA	Microarrays (5) Bisulfite sequencing (1)	Continuous data Affected by time and exposures (with reduced plasticity)
	mRNA	Microarrays (5) RNA sequencing (9)	Continuous data Affected by time and exposures Strong measurement noise
	Proteins	Microarrays (5) Mass spectrometry (5)	Continuous data Affected by time and exposures
	Small molecules	Mass spectrometry (5) NMR spectroscopy (4)	Continuous data Structured correlation Strongly affected by exposures

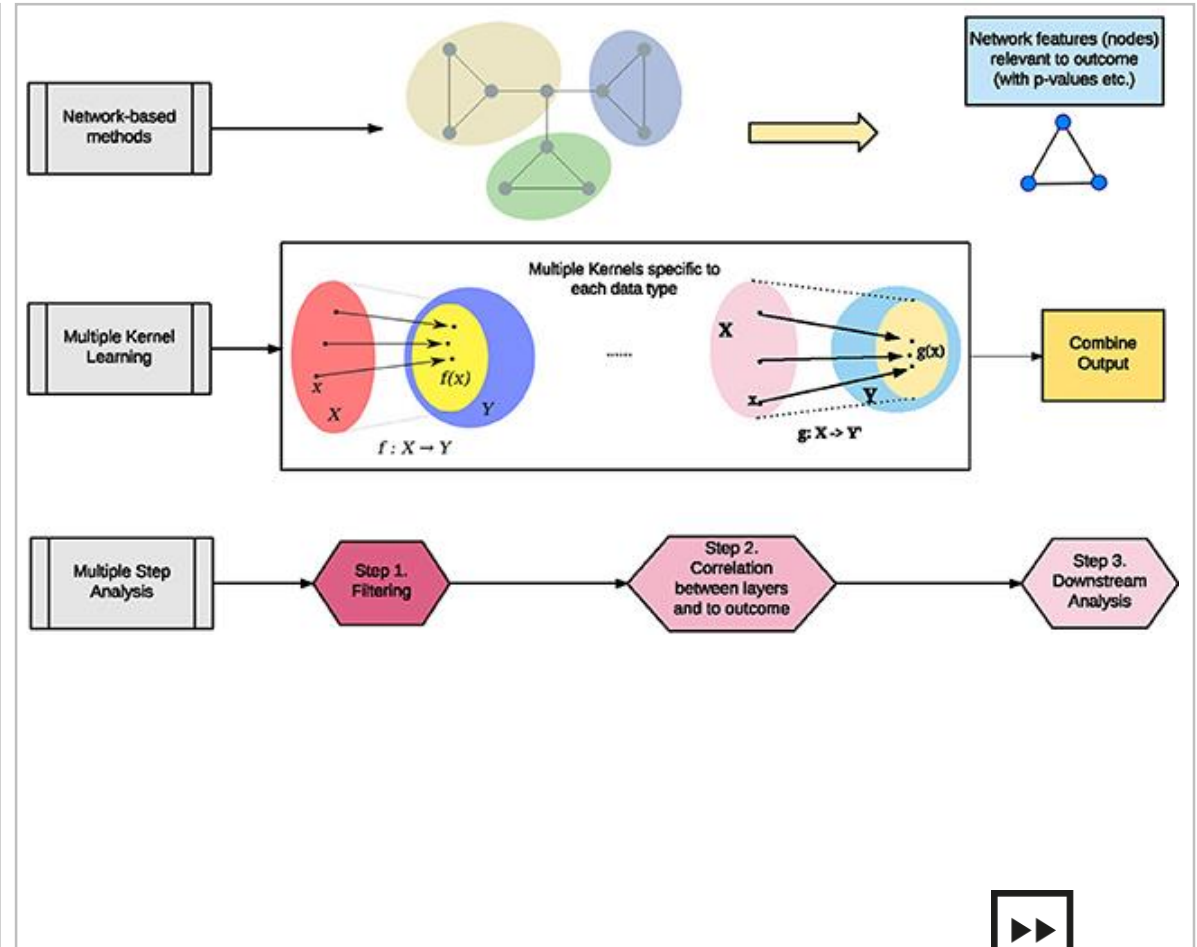
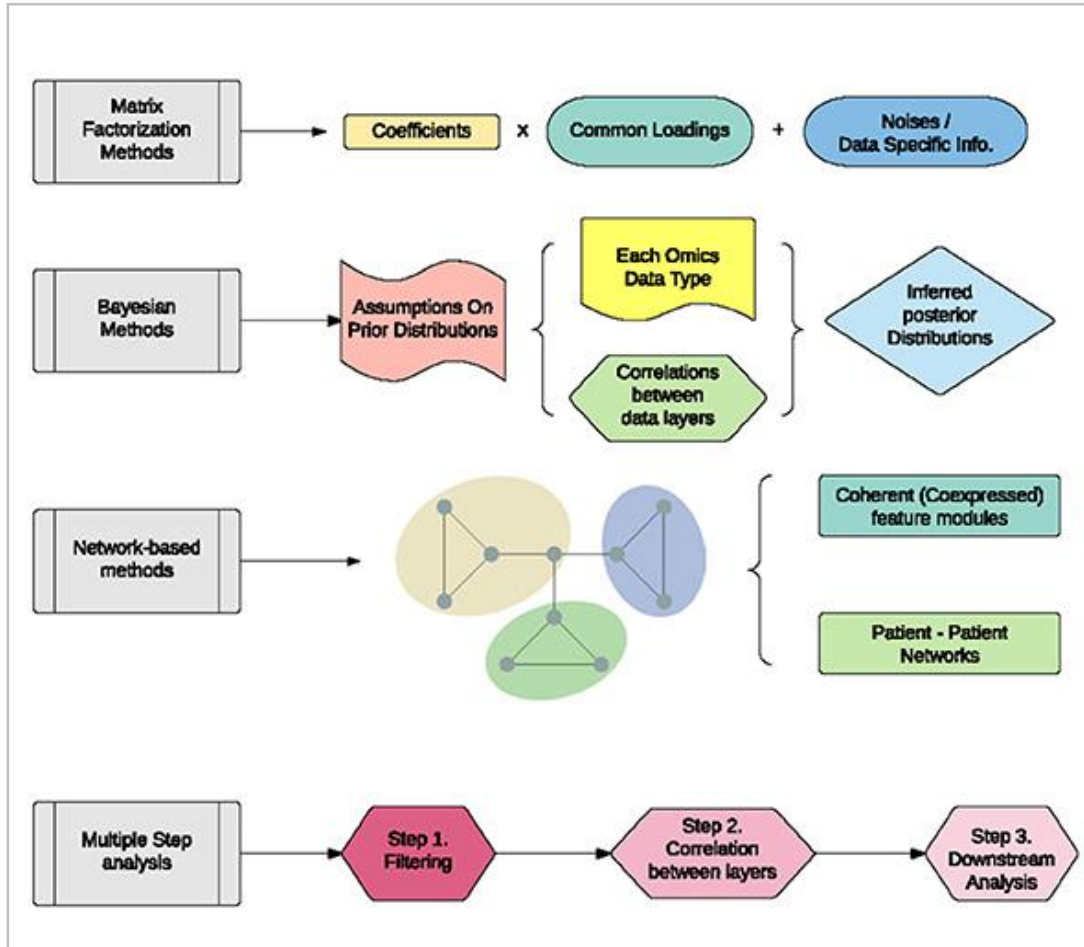
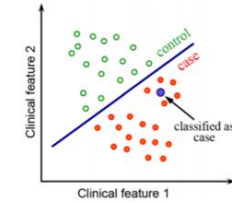


Methods

Unsupervised

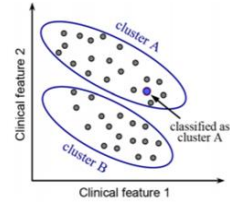


Supervised

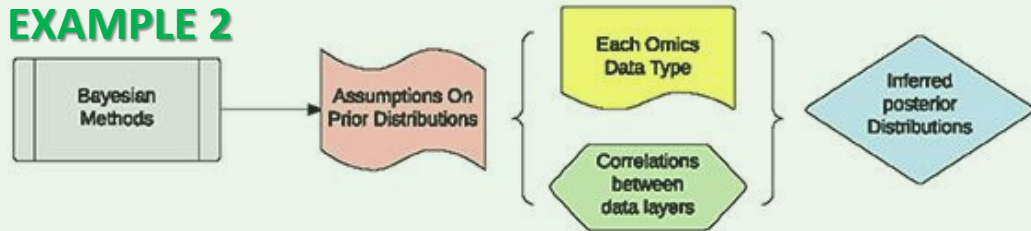


Methods

Unsupervised

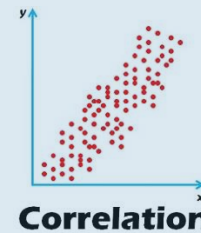


EXAMPLE 2



MOFA
Multi-Omics Factor Analysis V2 (MOFA+)

EXAMPLE 1



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Science and Technology in
childhood Obesity Policy

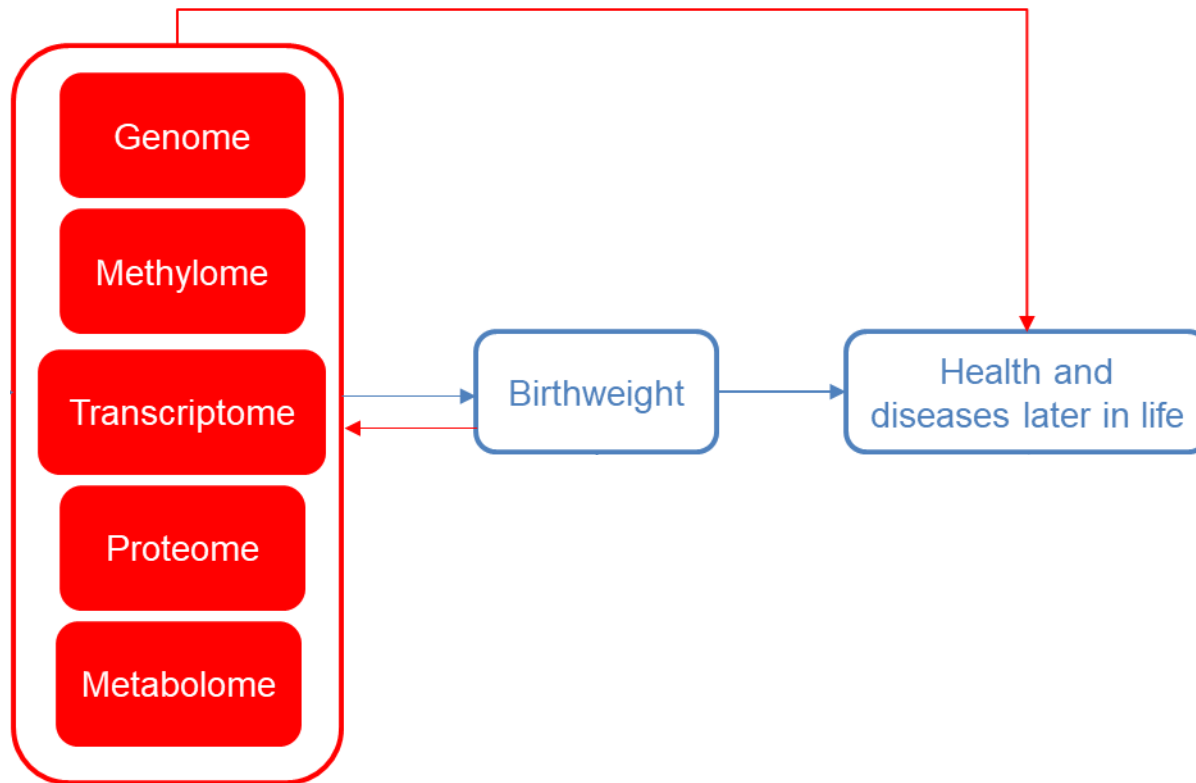
MULTI-OMICS ANALYSIS OF BIRTHWEIGHT



UHASSELT

KNOWLEDGE IN ACTION

Birthweight



Population and samples



4 EXPOsOMICS birth cohorts (n=483)



Cord blood

Methylome

485,577 CpGs

Transcriptome

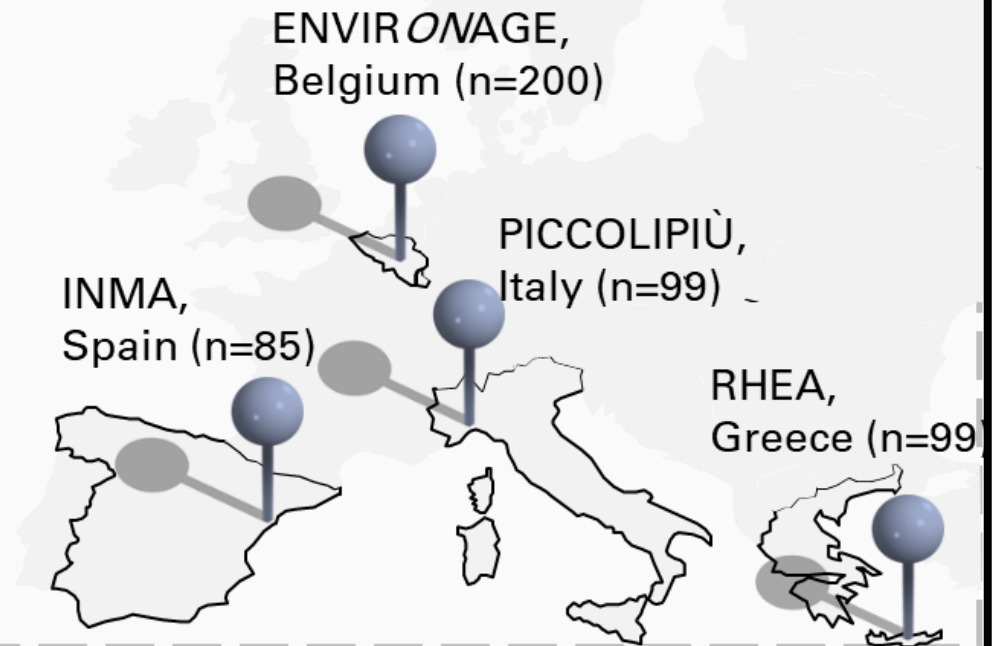
29,164 Transcripts
(ENVIRONAGE only)

Inflammatory proteins

16 Inflammatory proteins

Metabolome

4,712 Metabolites

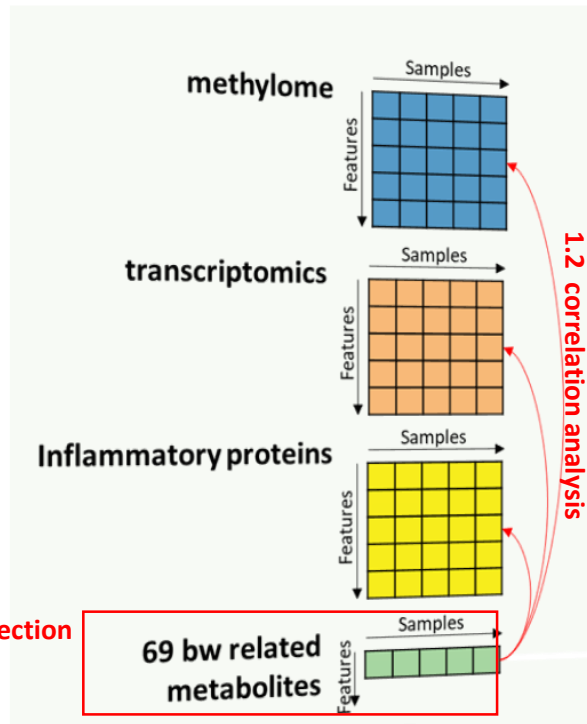


Multi-step method

Model overview

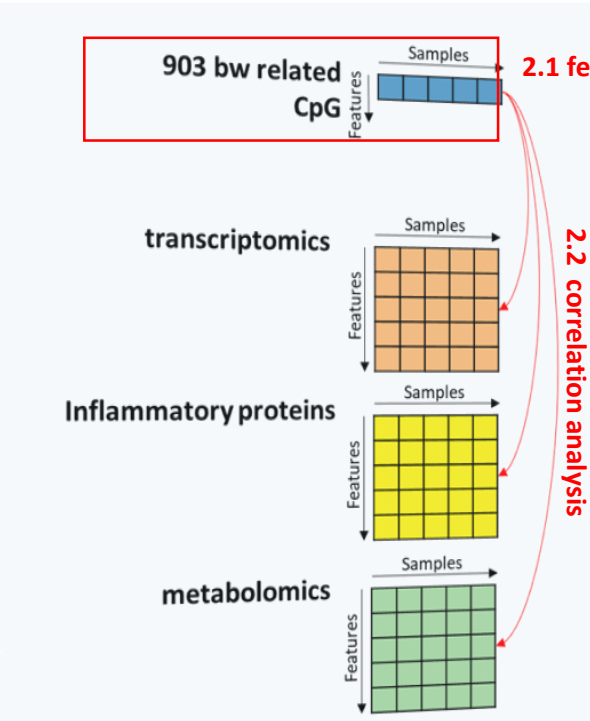
STEP 1

Metabolite-driven approach



STEP 2

Methylation-driven approach



1.1 feature selection

2.1 feature selection

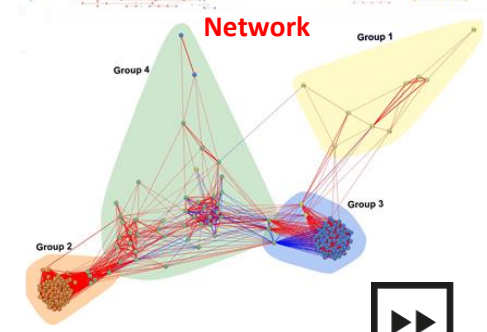
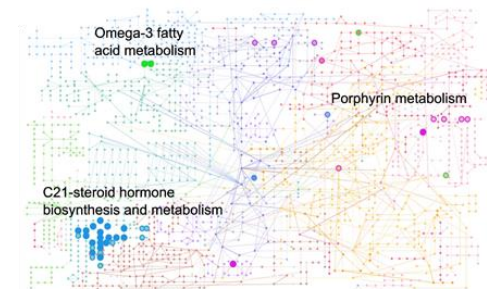
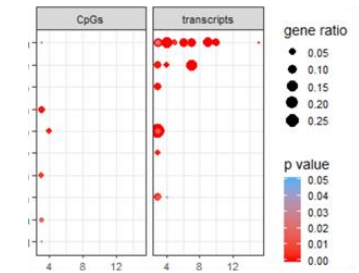
2.2 correlation analysis

STEP 3

Common signals identification

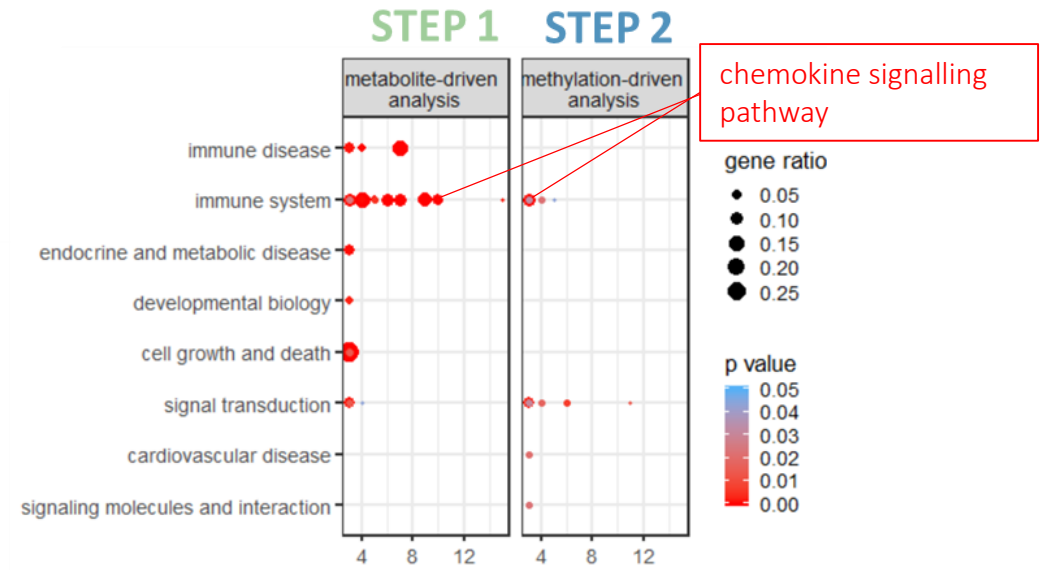
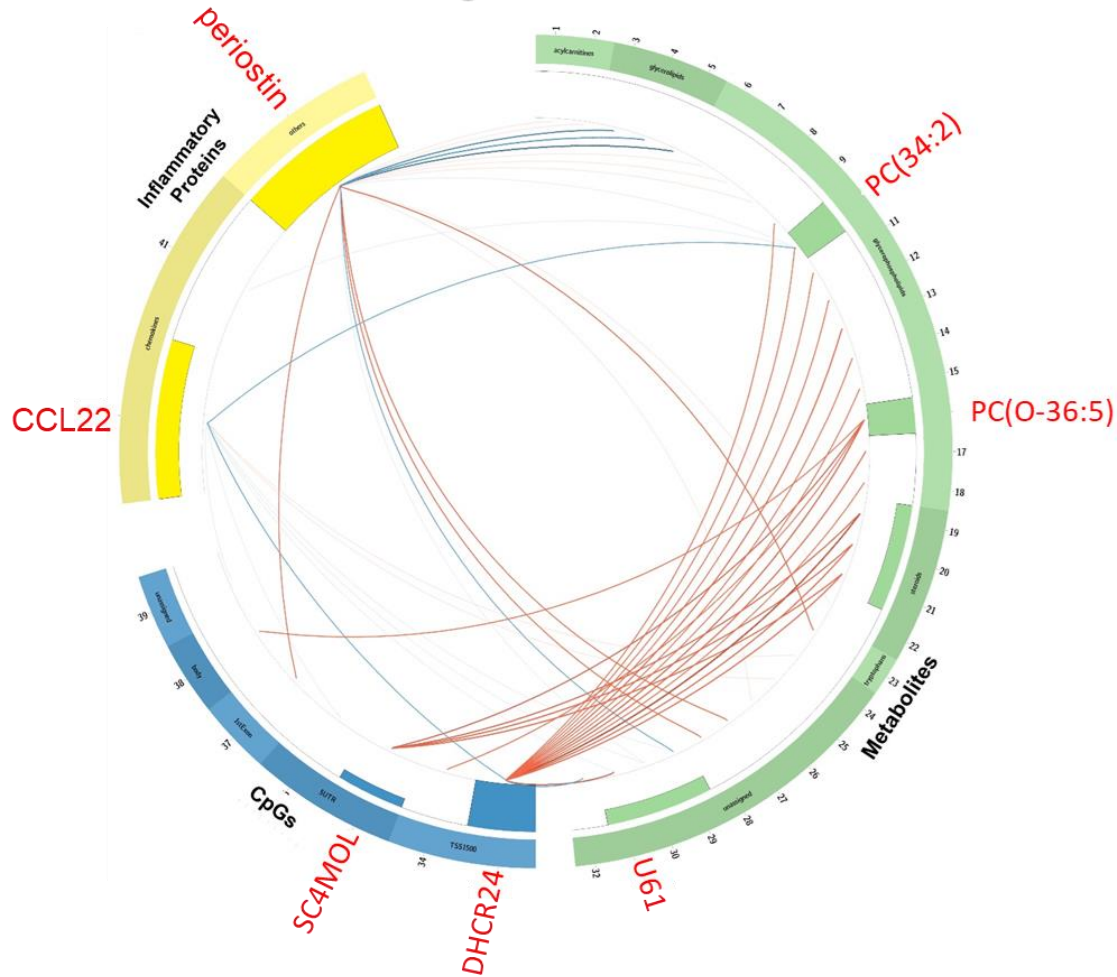
Downstream analyses

Pathways

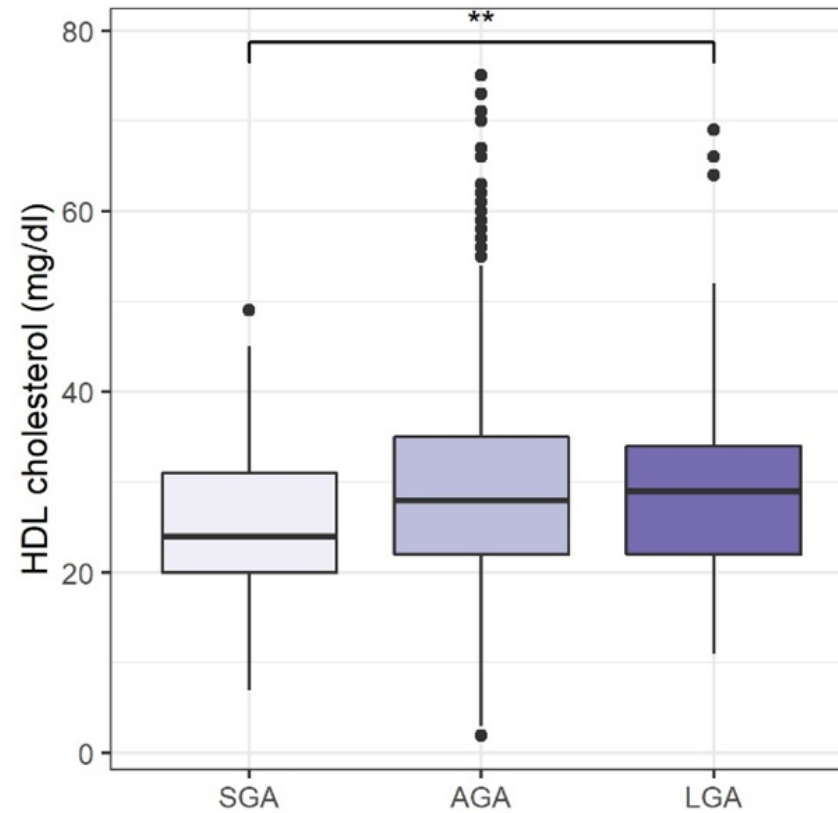


Results

STEP 3 Common signals identification



Results



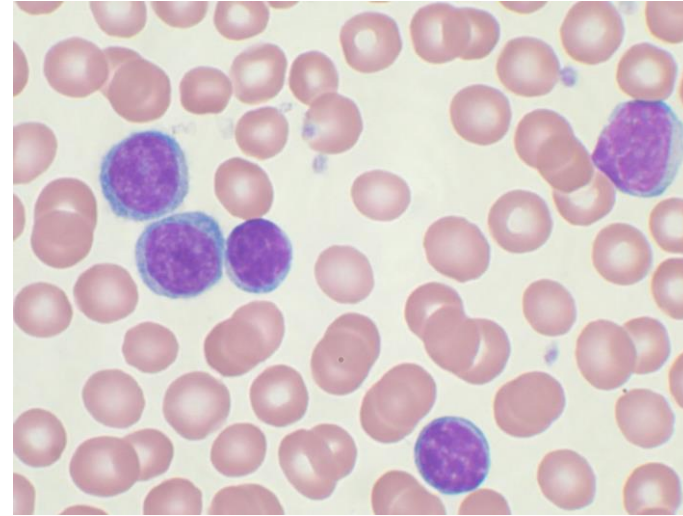
N=1097 newborns



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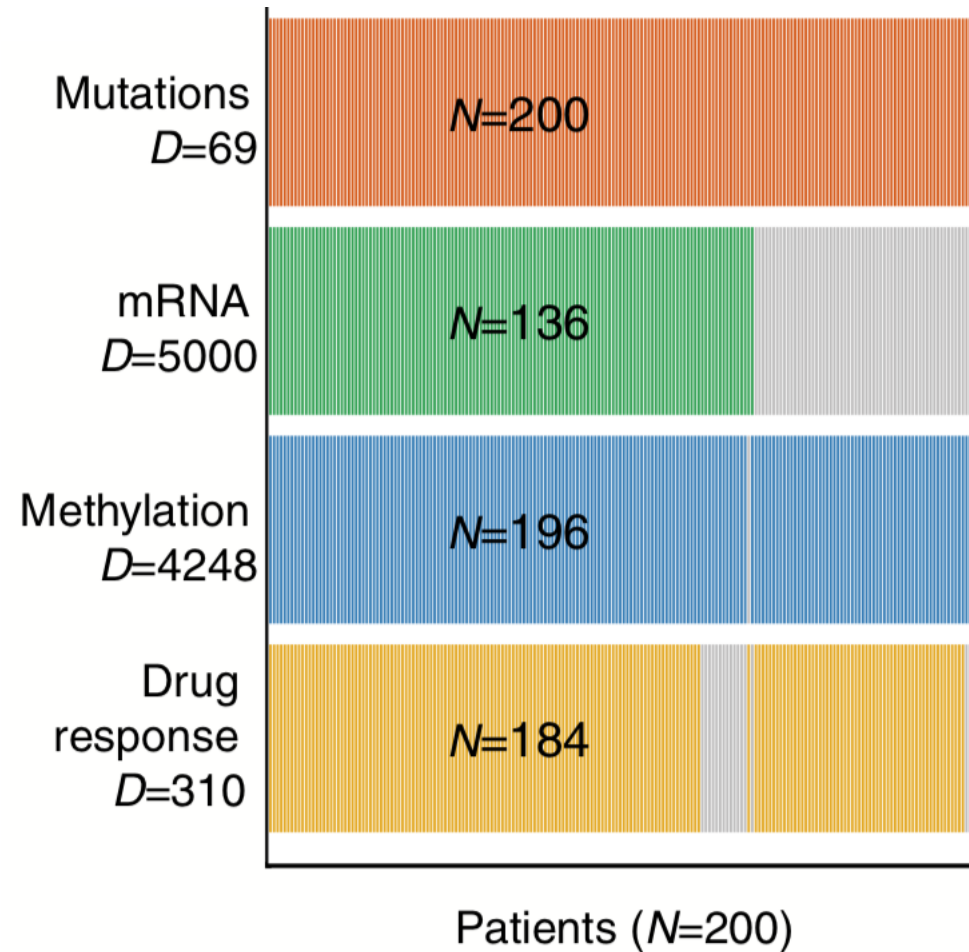
Multi-Omics Factor Analysis (MOFA) of chronic lymphocytic leukemia

Chronic lymphocytic leukemia study

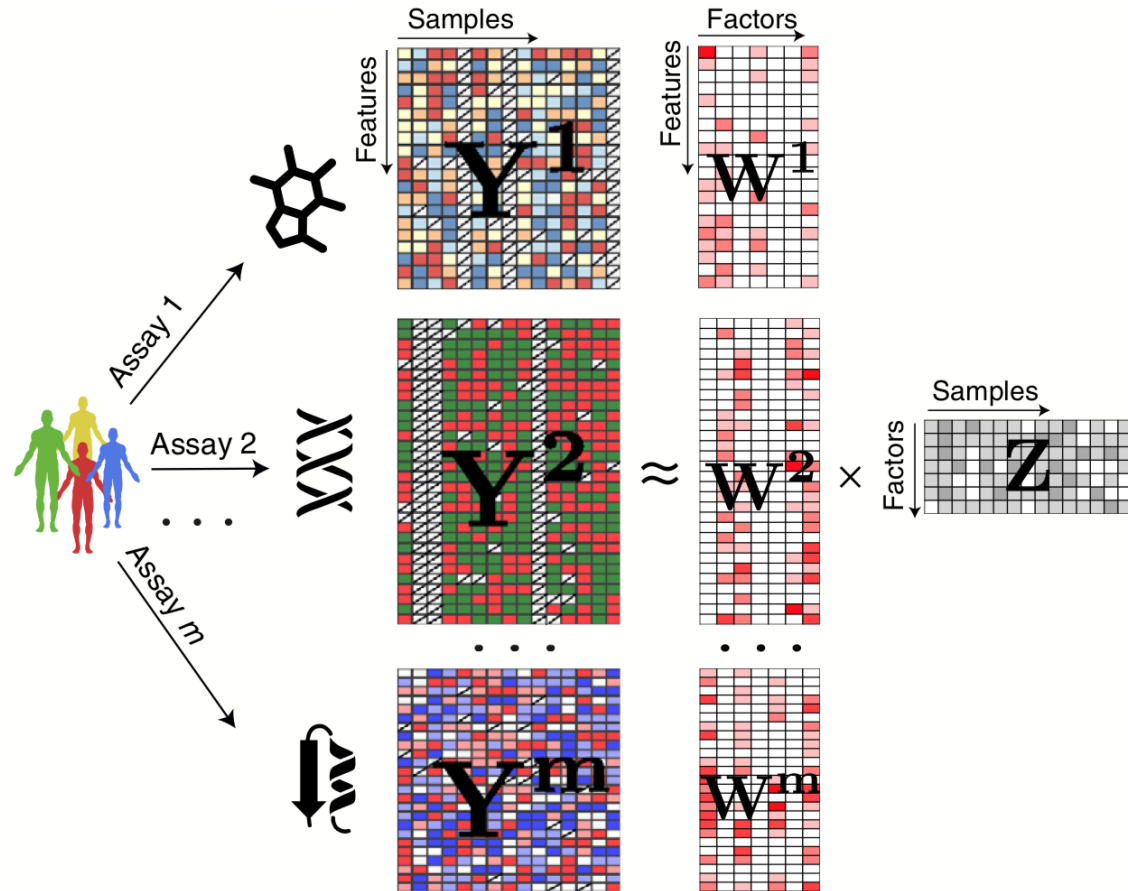
Study of chronic lymphocytic leukaemia (CLL), which combined:

- somatic mutation status
- transcriptome profiling
- DNA methylation assays
- ex vivo* drug response measurements

Nearly 40% of the 200 samples were profiled with some but not all omics types

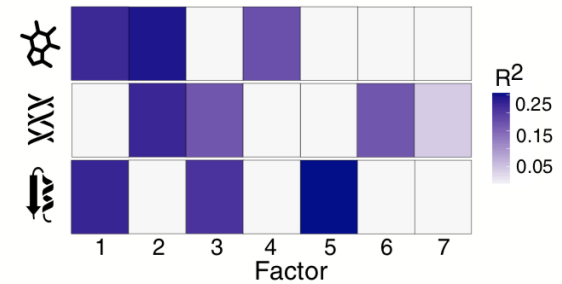


MOFA method Model overview

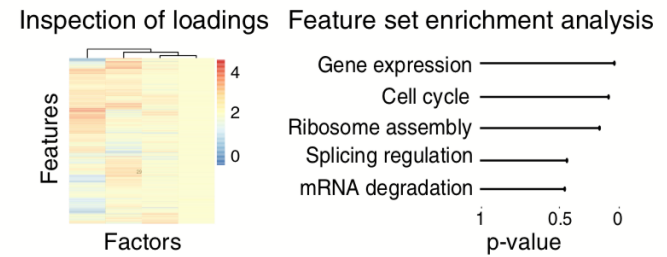


Downstream analyses

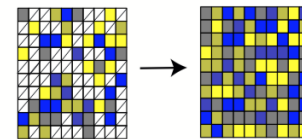
Variance decomposition by factor



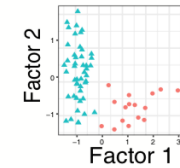
Annotation of factors



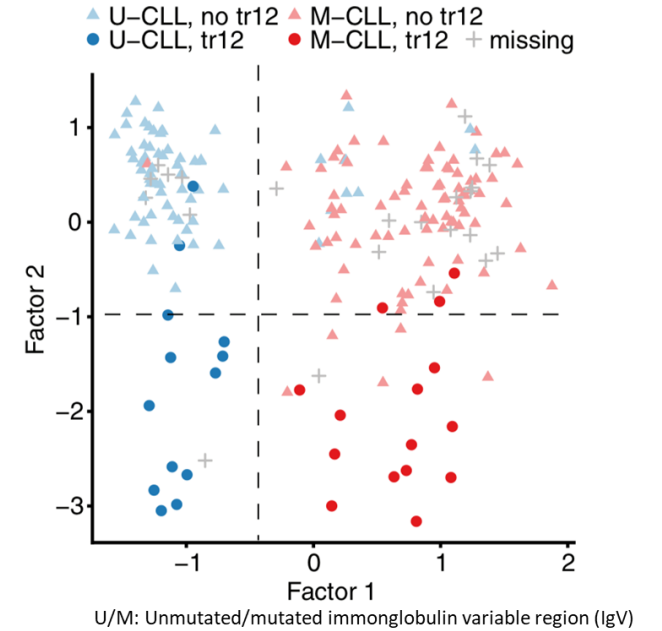
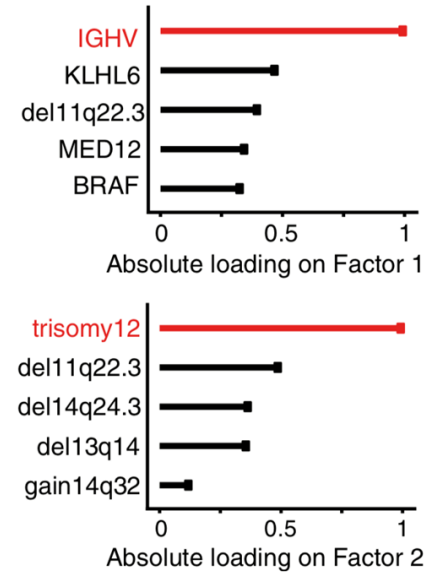
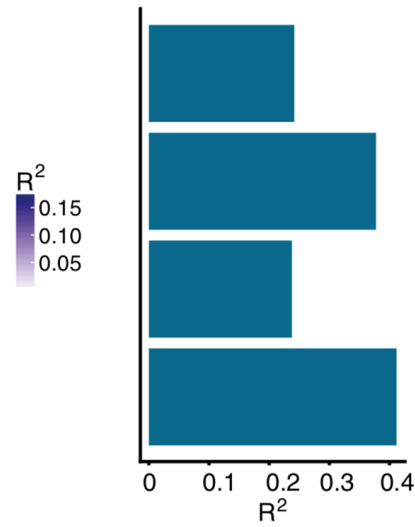
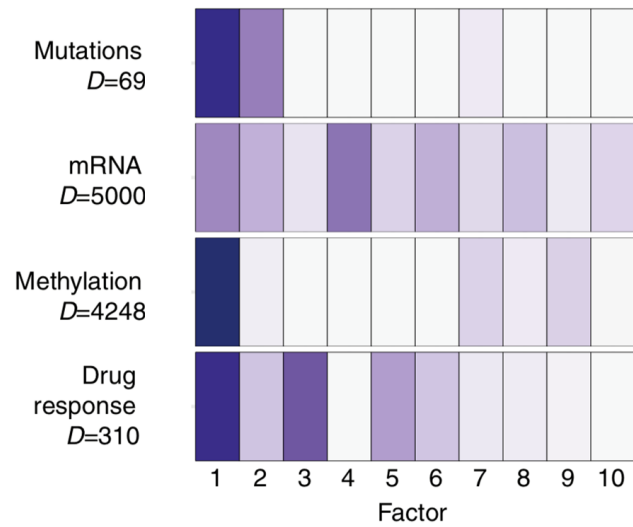
Imputation of missing values



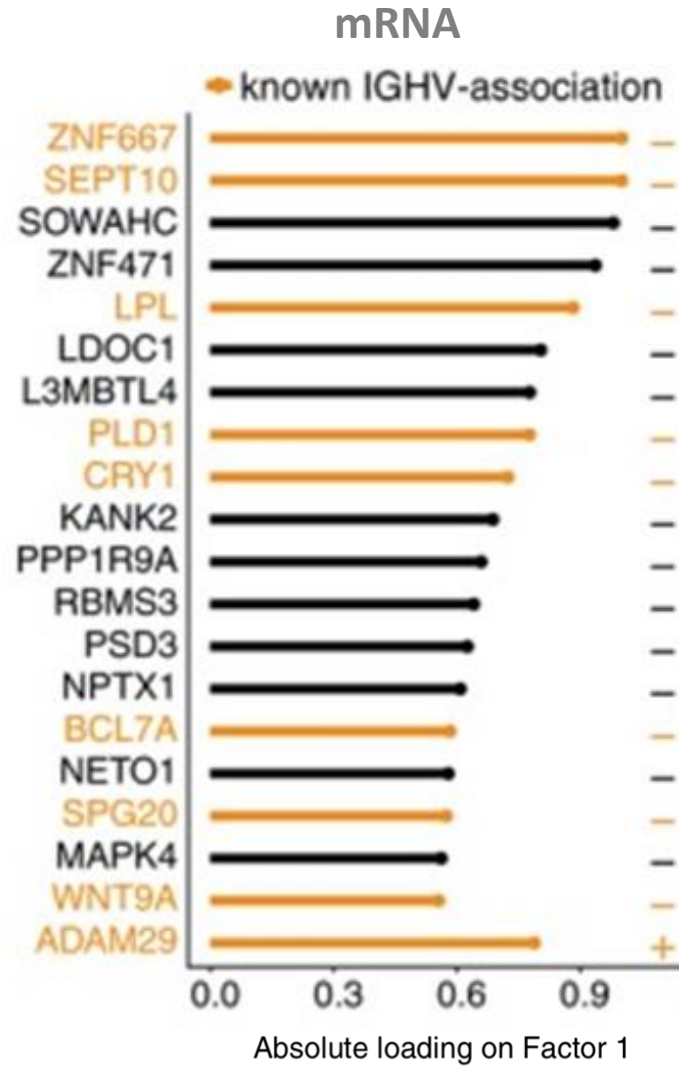
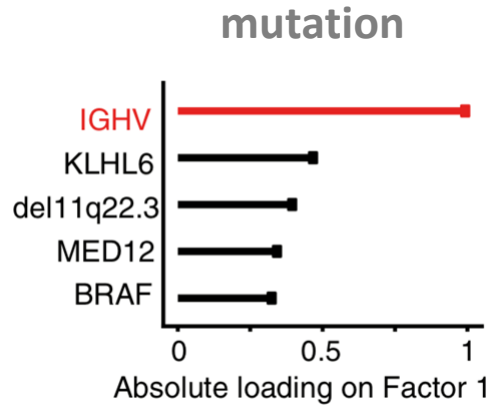
Inspection of factors



Results



Results

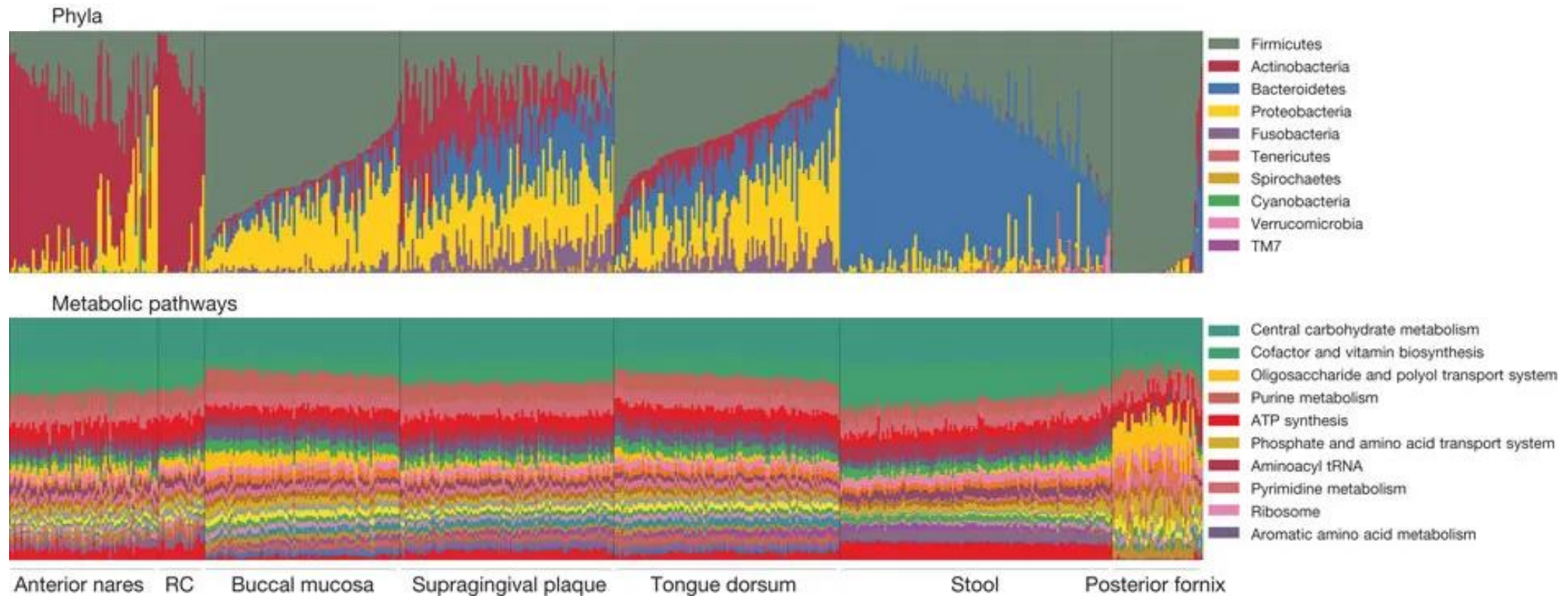


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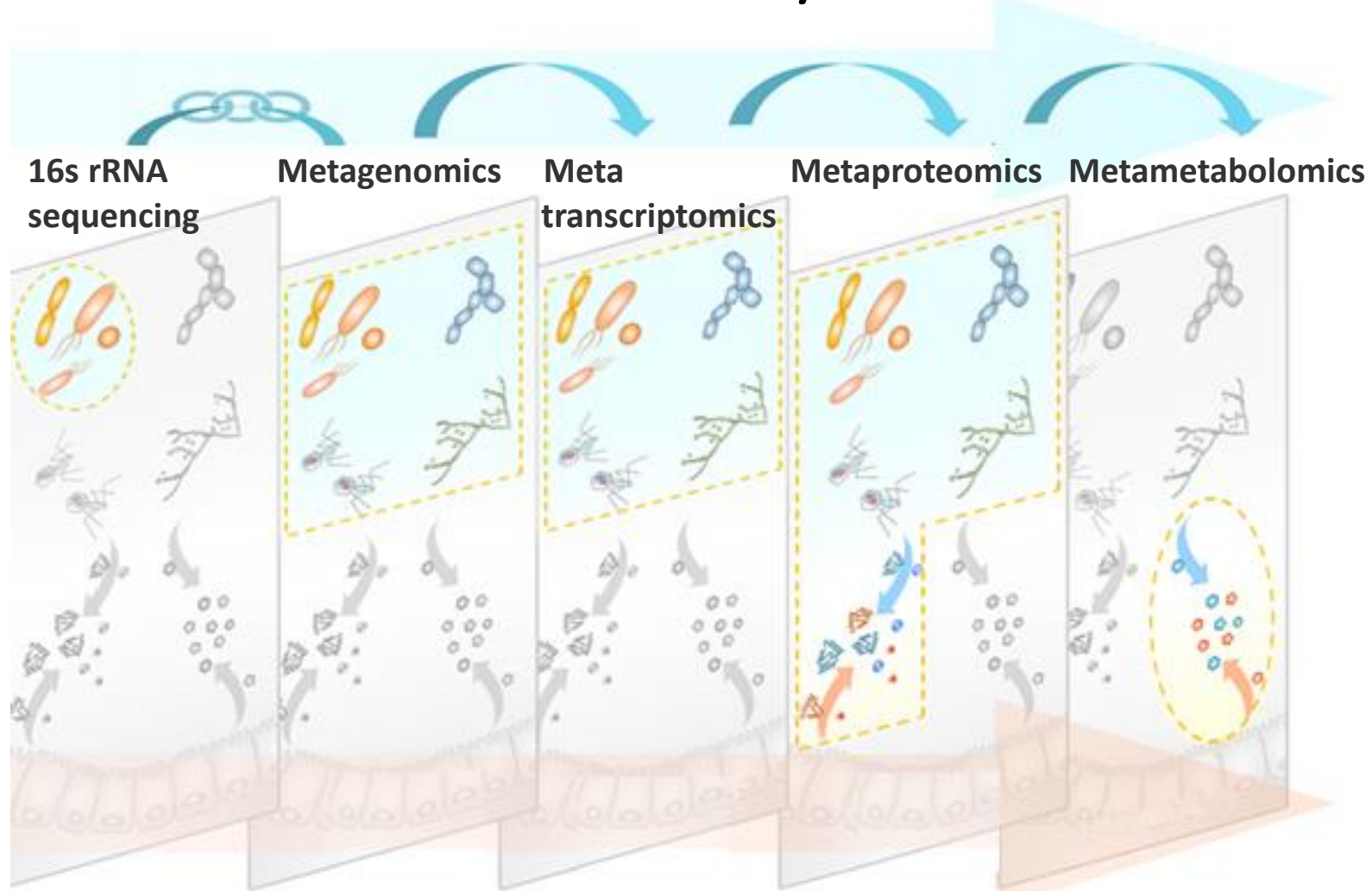
Why 16s rRNA is not enough?



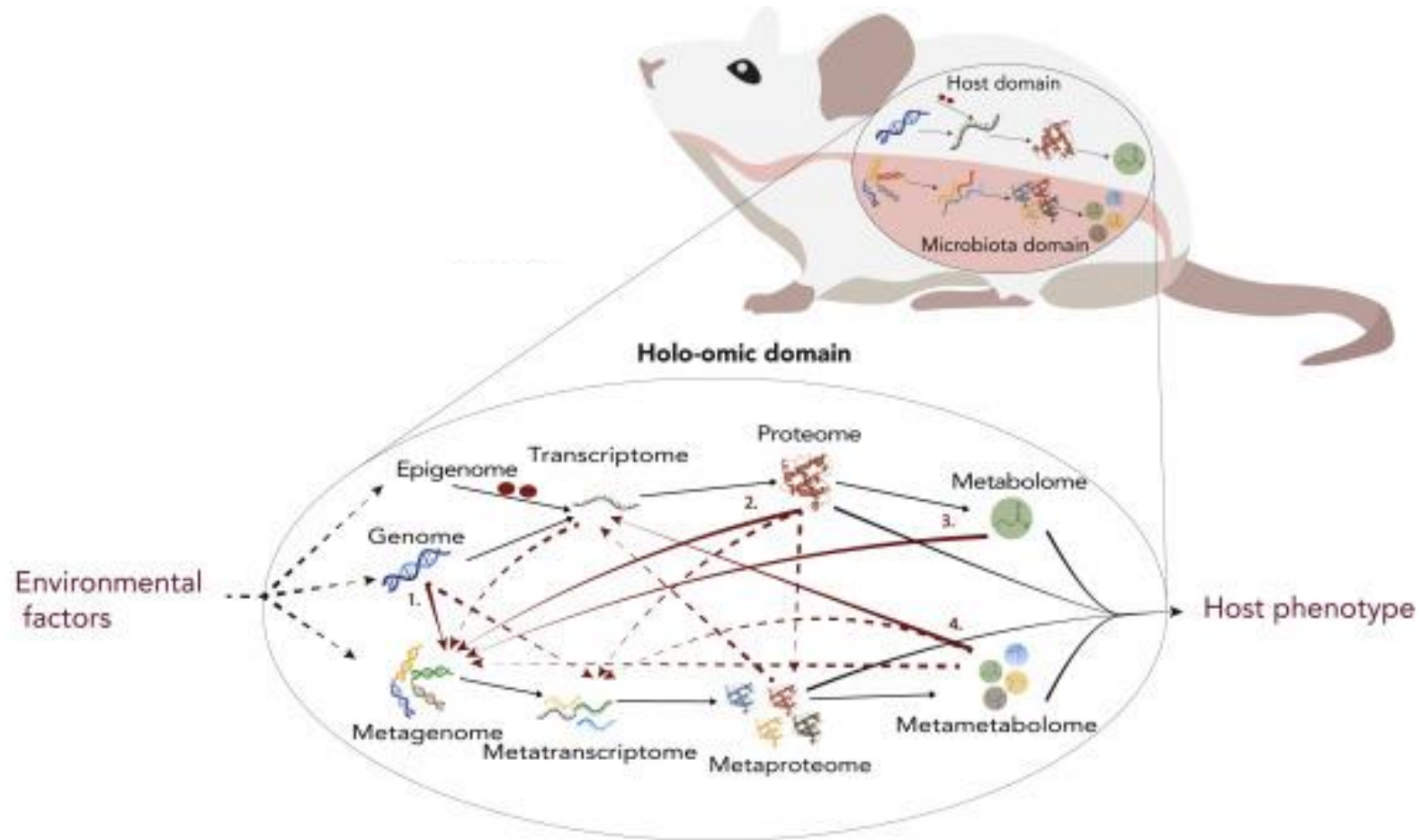
Carriage of microbial taxa varies while metabolic pathways remain stable within a healthy population



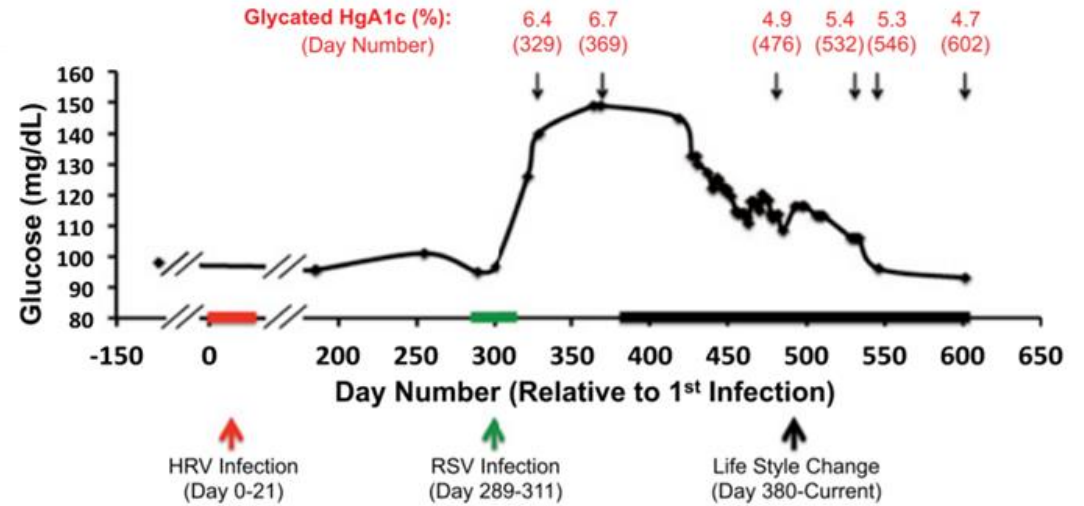
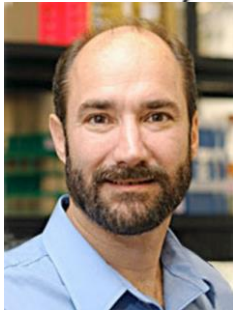
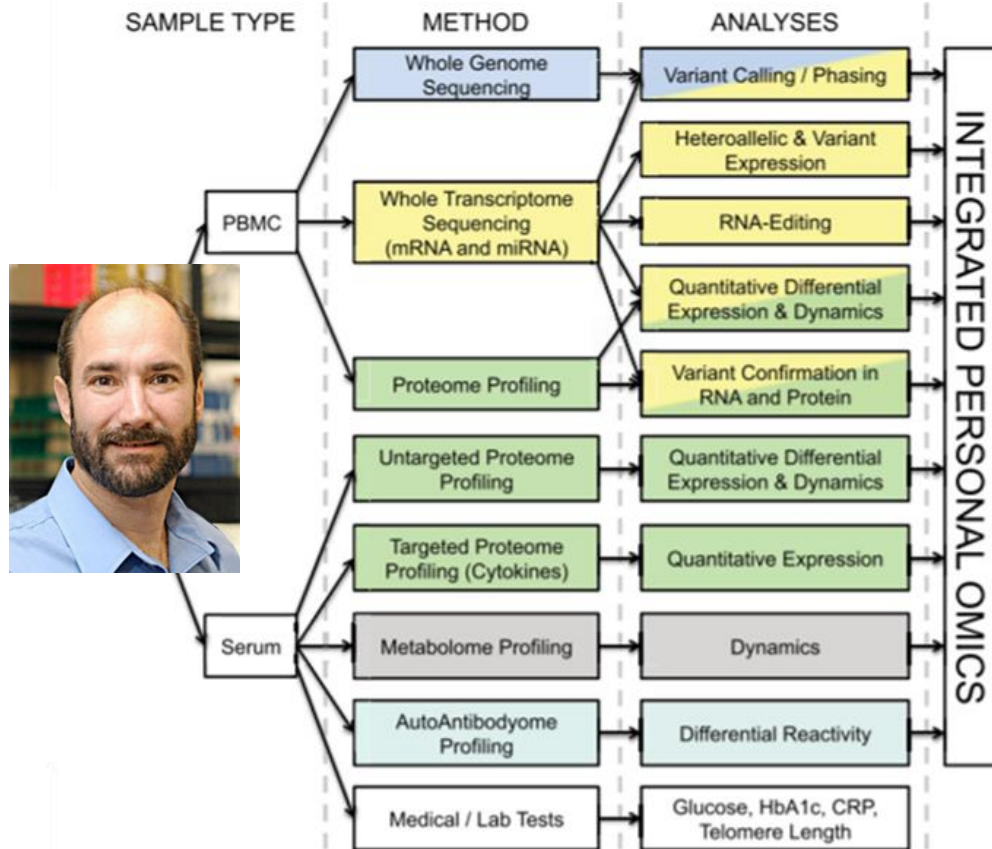
Each meta-omic reveals different layers of information



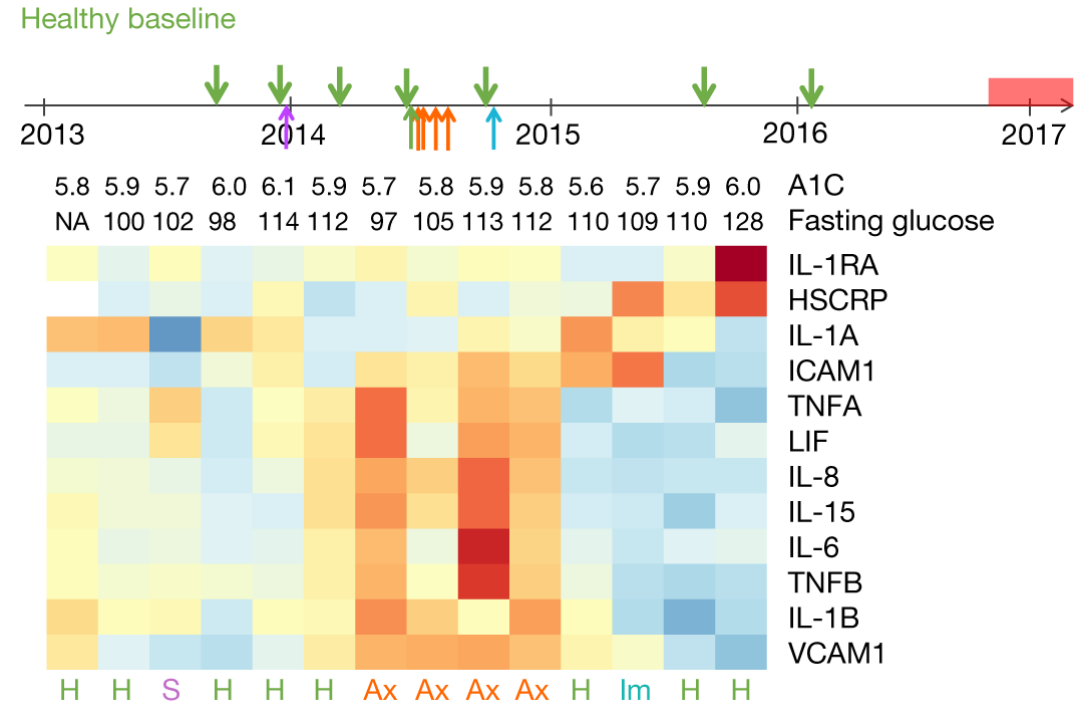
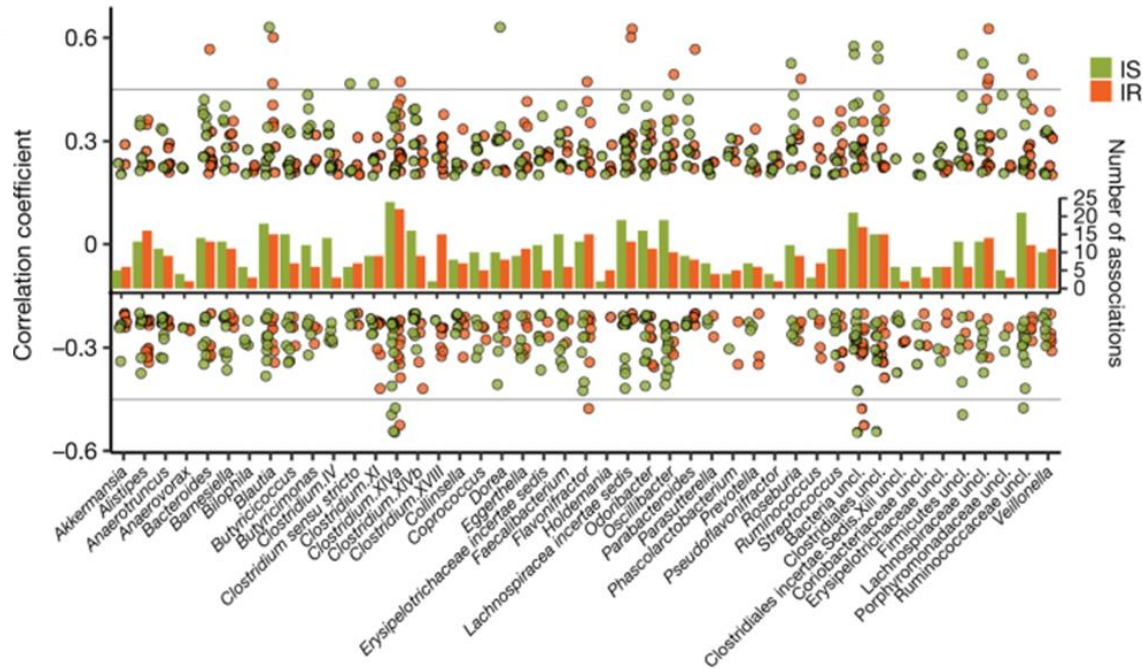
Holo-omic domain

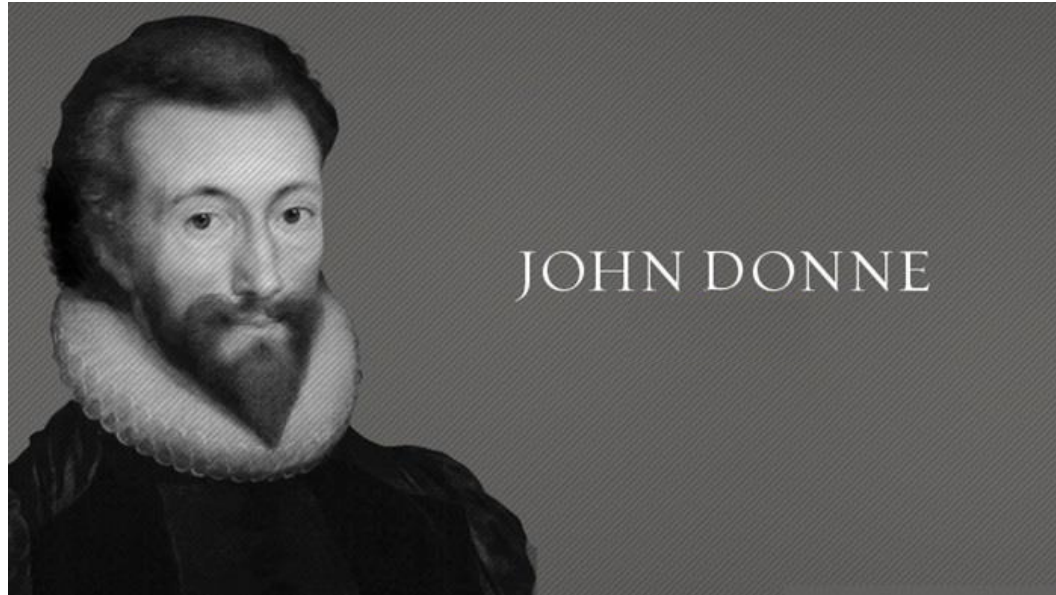


Integrated Personal Omics Profiling (iPOP)



Host–microbe multi-omics dynamics in prediabetes





NO MICROBIOME IS AN ISLAND ENTIRE OF ITSELF

Adapted from Zhang et al, Mol Cell Proteomics 2019

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Prof. Paolo Vineis



Dr Oliver Robinson







Imperial College
London



Hands-on session 4

🕒 03:30 PM - 05:00 PM | 📍 Workshop room

Hands-on session # 4: Visualisation of multi-omics data

-  [Leo Lahti](#)
-  [Rosella Alfano](#)
-  [Brigitte Reimann](#)
-  [Congrong Wang](#)

