



SCIENTIFIC
NETWORK MIMS

2022-2023

Coordination

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INRAE units involved

StatSC

BIA

QuaPA

SPO

LBE

ITAP

MaIAGE

CSGA Centre des Sciences du Goût et de
l'Alimentation

UNH Unité Nutrition Humaine

PhAN

LABERCA

Micalis

Prose

BioForA

LBLGC

AGAP Institut

SELMET

Partnerships

Faculté des Sciences, Paris

INRIA

University of Genève

University of Toulouse

ANSES

CNAM

University of Paris-Saclay

University of Montpellier

ADLIN

French Wine and vine Institut

Cross Methodological Insights for Multi-source Data Integration

Context and challenges

In biology, as in other scientific fields, the integration of multi-source data is more relevant than ever. Indeed, the data collected are increasingly complex and their volume is growing, due to the development of analytical platforms, imaging techniques, the rise of omics data, etc.

This context has stimulated the search for new methods allowing the joint analysis of several data sets (structured data, multi-block, multi-channel) in many fields, such as:

- Machine Learning, where several approaches are considered for the processing of multi-source data (matrix factorisation, probabilistic approach).
- Chemometrics, where different methods are proposed to establish a chemical mapping of samples using several analytical techniques (generalisations of canonical analysis, NIPALS algorithm and tensor decompositions)
- Bioinformatics, where integrative methodological approaches allow the most complete picture possible of the dynamics of molecular systems to be drawn.

In order to contribute to meeting the challenge of analysing and exploiting these multi-source data from an exploratory, but also predictive perspective, it is essential to bring together different viewpoints, practices and paradigms in order to reconcile these different approaches. It is also necessary to encourage collaboration between "method generators" and "data generators" in the various application fields.

This is the challenge that the MIMS consortium proposes to take up, by bringing together an interdisciplinary community working on approaches to the analysis and integration of multi-source data.



Logo MIMS



Goals

MIMS is a multidisciplinary consortium gathering more than 60 researchers, whose objective is to examine the analysis and exploitation of multi-source data, both in an exploratory and predictive perspective.

This consortium brings together multidisciplinary skills: information processing, biological sciences and analytics. The implementation of this multi-disciplinarity and its management will be based on the sharing of data, practices and methods between the partners, with the aim of formalising a scientific project to meet a common challenge: the optimal analysis of multi-source data for exploratory and predictive purposes.

Research units involved and partnerships

INRAE scientific division	INRAE research units	Expertises
<u>Sciences for food, bioproducts and waste engineering</u>	<u>USC StatSC</u>	Sensometry, Chemometrics, Statistics, Multispectral imaging
	<u>BIA</u>	Chemometrics, computer science
	<u>QuaPA</u>	Volatolomics, MRI Chemometrics, Data Analysis, Image Analysis, System & Data Management
	<u>SPO</u>	Chemometrics
<u>Mathematics, computer and data sciences, digital technologies</u>	<u>LBE</u>	Biostatistics, machine learning
	<u>ITAP</u>	Chemometrics
	<u>MAIAGE</u>	Mathematical statistics, applied statistics, bioinformatics
<u>Human nutrition and food safety</u>	<u>CSGA</u> Centre des Sciences du Goût et de l'Alimentation	Chemometrics
	<u>UNH</u> Unité Nutrition Humaine	Bioinformatics, metabolomics, chemometrics
	<u>PhAN</u>	Perinatal nutrition and metabolic diseases, Bioinformatics, Data analysis, metagenomics and metabolomics
<u>Microbiology and the food chain</u>	<u>LABERCA</u>	Metabolomics, Chemometrics, Expology, Epidemiology
	<u>Micalis</u>	Biologist, Microbiota, Data Analysis
<u>Ecology and biodiversity of forest, grassland and freshwater environments</u>	<u>Prose</u>	
	<u>BioForA</u>	Quantitative Genetics, Modelling
<u>Plant biology and breeding</u>	<u>LBLGC</u>	Physiology
	<u>AGAP</u> Institut	Quantitative genetics, Genomics, Biochemistry, Evolutionary genetics, Selection, Ecophysiology, Biostatistics, Bioinformatics



Animal physiology and livestock systems		
External partners		Expertises
Faculté des Sciences, Paris	SELMET	Biometrics, Chemometrics, Machine Learning, Agronomy
INRIA	Centre Boreli	Unsupervised learning, Statistics, Graph networks, Bioinformatics
University of Genève	Équipe projet LORIA	Knowledge Discovery, Life Sciences
University of Toulouse	Sciences Analytiques	Metabolomics, Chemometrics
ANSES	Institut de mathématique de Toulouse	Statistics, Multi-omics data analysis and integration
CNAM	Laboratoire de Ploufragan-Plouzané	Statistics, multi-block methods Epidemiology
University of Paris-Saclay	EPN6 - Mathématiques et Statistique	Analysis of complex heterogeneous data, Clusterwise methods, High dimensional classification
University of Montpellier	Signaux et Statistique	Multi-block data analysis, tensor analysis (high dimensional), Structural equation models
ADLIN	Institut Montpellierain Alexander Grothendieck	Supervised component models, classification
French Wine and vine Institut	ADLIN	Finance, Strategy, Multi-omics, Bioinformatics, Transcriptomics, Visualisation
	IFV	Chemometrics, Analytical Chemistry

References

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