

# From wet lab to the Cloud

Alexandre BUREL









### Presentation of the LSMBO



PHC

Institut Pluridisciplina Hubert CU

- Laboratoire de Spectrométrie de Masse BioOrganique
  - Part of Institut Pluridisciplinaire Hubert Curien (UMR7178)
  - Under joint supervision of CNRS and University of Strasbourg
  - 12 Mass spectrometers



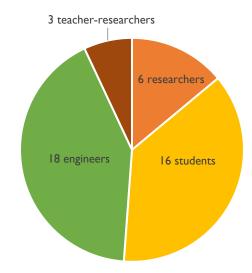
de Strasbourg

Université

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- Who we are
  - Around 40 people
  - Half are permanent staff
  - A majority of chemists
  - But also biologists, pharmacists, bioinformaticians, etc.







Université					
	de Strasbourg				

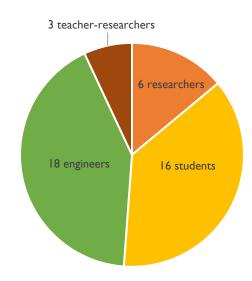
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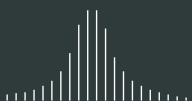
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  - 12 Mass spectrometers
- Who we are
  - Around 40 people
  - Half are permanent staff
  - A majority of chemists
  - But also biologists, pharmacists, bioinformaticians, etc.
- What we do
  - Specialized in proteomics
  - Identification and quantification of proteins
  - Search for biomarkers
  - Characterization of therapeutic proteins

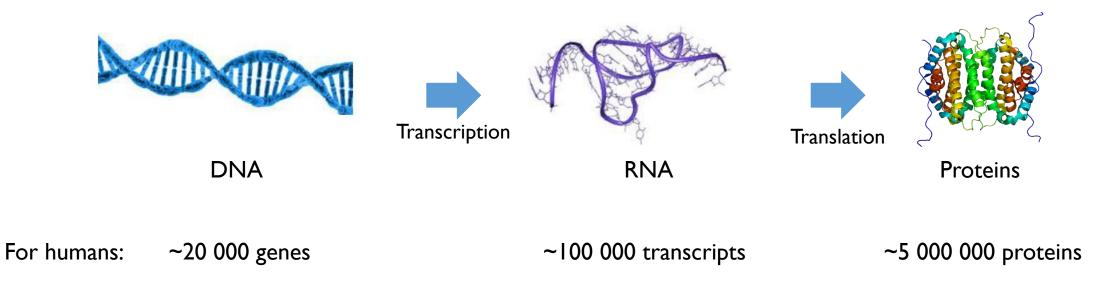


### A few definitions



### • Proteomics

- Study of all the proteins in a cell, tissue or organism
- Proteins are essential molecules for most functions in all living organisms
- Genes are the source code for proteins

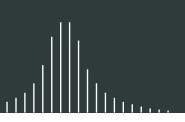


### A few definitions

- Mass spectrometry
  - Technique used to measure the mass of molecules
  - Proteins are cut into small pieces called peptides
  - Each peptide is charged
  - Then accelerated through an flight tube
  - The flight time to reach the detector is measured

Bruker timsTOF Ultra





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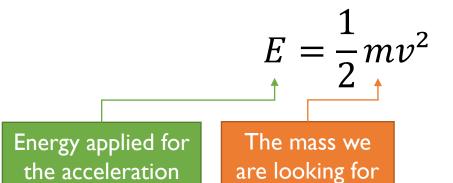
$$E = \frac{1}{2}mv^2$$
The mass we are looking for



Bruker timsTOF Ultra



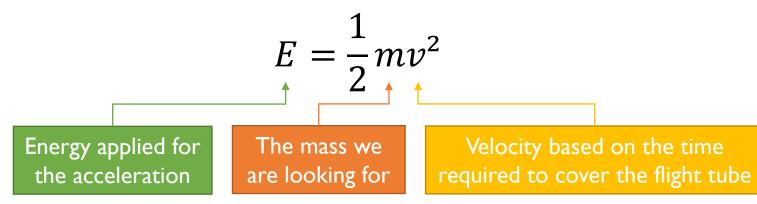
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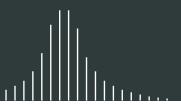


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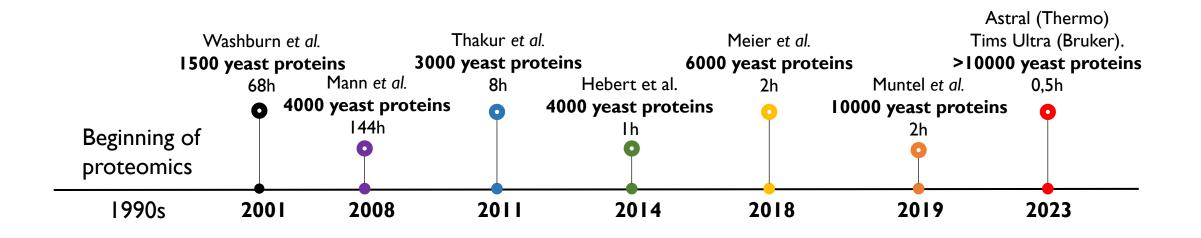
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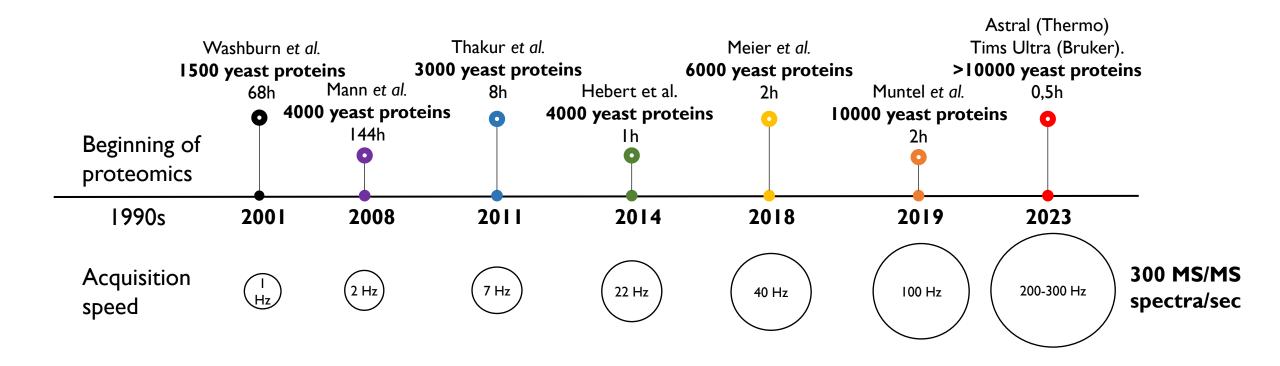




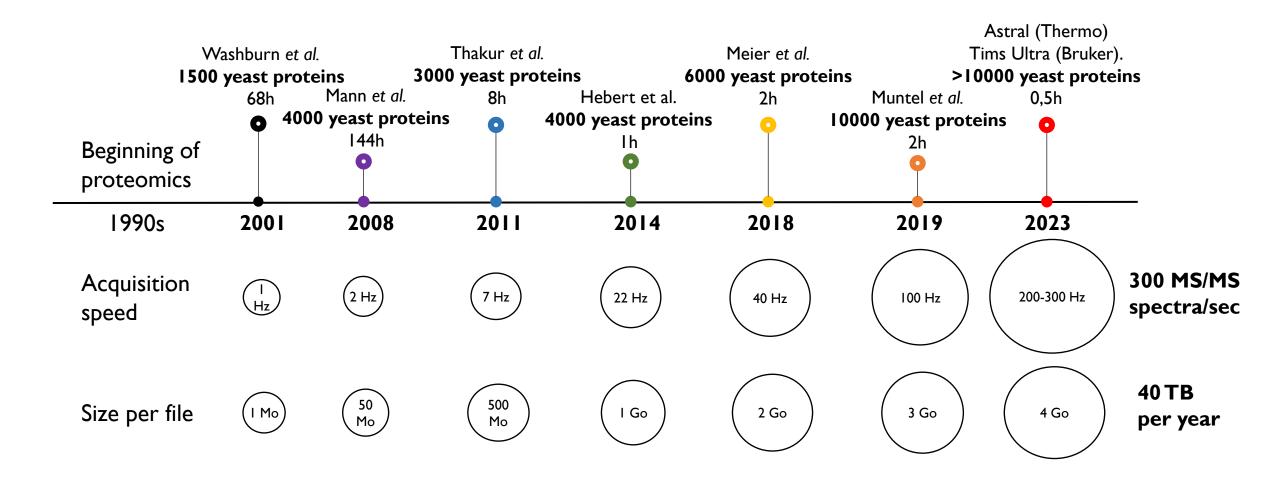
### Improvements in mass spectrometry



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- Many software involved for different purposes
  - Identification & quantification (Proline, Maxquant, Dia-NN, etc)
  - Functional analysis (Kegg, GeneOntology, Reactome, etc.)
  - Statistical analysis (Prostar, MSqRob, R)
  - FAIR data practices (PRIDE, ProteomeXchange)



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- The solution is in the Cloud!







#### In a few words

- SCIGNE a platform offering compute and storage services hosted by IPHC
- Includes support to help researchers to manage and analyse large amounts of data in several scientific fields (physics, chemistry, biology and ecology)
- Labelised by IN2P3 and by the University of Strasbourg (CORTECS)
- Involved in several national and international scientific projects
- Managed by a team of 8 highly-skilled engineers

### **Expertise & Services**

- Processing and analysis of large amounts of scientific data
- Computation reproducibility studies
- Data and software management plans, making the data FAIR
- Software development (including GPU & IA), source code opening
- IT Security & Green computing
- High-throughput computing, Cloud computing and scientific data management services

#### https://scigne.fr

#### • Advantages

- Plenty of resources available
- Stable and powerful environment
- The SCIGNE engineers are responsive and competent

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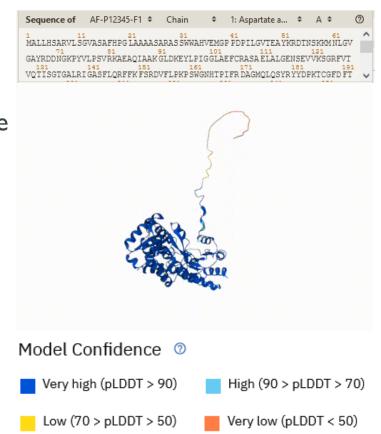
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  - Best fit for Linux command line software
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- Software
  - Occasionally: Proline, Brownotate, NetMHCpan, etc.
  - On a daily basis: Alphafold, ionbot, Dia-NN, Galaxy

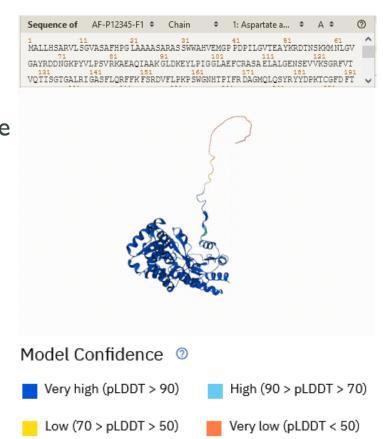
AlphaFold

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  - Predicts the 3D structure of a protein from its amino acid sequence
- Why is it important?
  - Characterization of proteins and protein complexes
  - Better understanding of the interactions with other molecules
  - Accelerate drug discovery
  - Understanding diseases linked to misfolded proteins



Jumper, J et al. Highly accurate protein structure prediction with AlphaFold. Nature (2021)

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- Why using the Cloud?
  - Study of proteins not predicted yet
  - Not depend on the public server
  - Our VM: 8 CPU, 64GB RAM, 4TB

Jumper, J et al. Highly accurate protein structure prediction with AlphaFold. Nature (2021)

1 21 31 41 MALLHSARVL SGVASAFHPG LAAAASARAS SWWAHVEMGP PDPIL 
 71
 81
 91
 101
 111
 121

 GAYRDDNGKPYVLPSVKAEAQIAAK
 GOLKEYLPIGGLAEFCRASA ELALGENSEVVKSGRFVT
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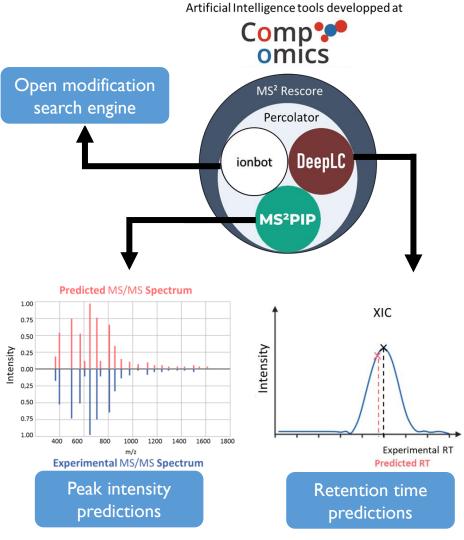
 VQTISGTGALRIGASFLQRFFK FSRDVFLPKP SWGNHTPIFR DAGMQLQSYR YYDPKTCGFD FT
 151
 161
 171
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 191
 Model Confidence 0 Very high (pLDDT > 90) High (90 > pLDDT > 70) Low (70 > pLDDT > 50)Very low (pLDDT < 50)</p>

1: Aspartate a...

8

Sequence of AF-P12345-F1 Chain

### ionbot

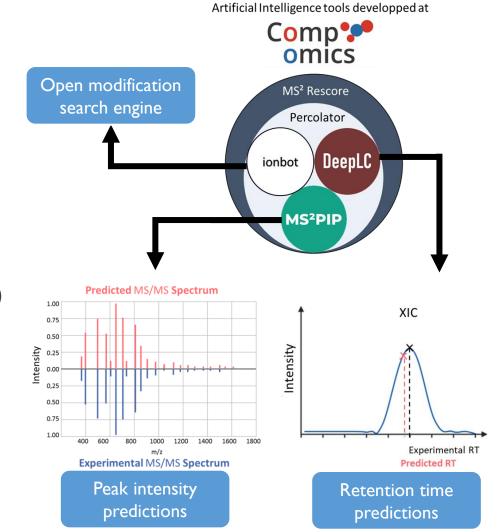


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- Developed by Compomics (VIB, Ghent University, BE)
- Open modification search engine
- Using Deep Learning to model the complex behaviour of peptide molecules in a mass spectrometer

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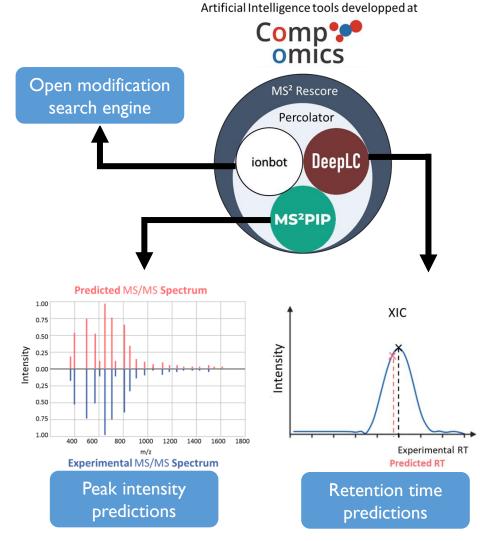
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- Why is it important?
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- Why using the Cloud?
  - Need for important resources
  - Searches can take hours in some cases
  - Our VM: I6CPU, 64GB RAM, ITB



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### Dia-NN

- What is it?
  - Developed by Vadim Demichev (Charité Universitätsmedezin Berlin, DE)
  - Quantify large-scale experiments using neural networks

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Experiment name: 10/22/2024 14:16:03	
Input	Output
Raw .d (DIA) Clear list Convert to .dia	Use existing .quant files when available
A	Main output D:\Projets\DiaNN\1.9\report.tsv
	Temp/.dia dir
	☐ Generate spectral library
	Output library
	Generate Prosit input from FASTA or spectral library
	Precursor FDR (%) 1.0 🗭 Threads 4 🖨
	PDF XICs Log level 1
~	Additional options
Spectral library	
Add FASTA	
Clear list	
Clear list Reannotate	
Reannotate	
Reannotate	Run Not started Stop
Reannotate Contaminants	Run Not started Stop
Reannotate       Contaminants       DIA-NN exe       diann.exe	
Reannotate       Contaminants       DIA-NN exe       diann.exe   Precursor ion generation	Algorithm
Reannotate         Contaminants         DIA-NN exe         diann.exe         Precursor ion generation         FASTA digest for library-free search / library generation         Deep learning-based spectra, RTs and IMs prediction	Algorithm Mass accuracy 0.0 🜩 🗌 Unrelated runs
Reannotate         Contaminants         DIA-NN exe         diann.exe         Precursor ion generation         FASTA digest for library-free search / library generation         Deep learning-based spectra, RTs and IMs prediction	Algorithm Mass accuracy 0.0 文 🗋 Unrelated runs MS1 accuracy 0.0 文 Peptidoforms
Reannotate         Contaminants         DIA-NN exe         diann.exe         Precursor ion generation         FASTA digest for library-free search / library generation         Deep learning-based spectra, RTs and IMs prediction         Protease       Trypsin/P ~         Missed cleavages       1	Algorithm Mass accuracy 0.0
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Experiment name: 10/22/2024 14:16:03								
Input Raw .d (DIA) Clear list Convert to .dia	Output           Output           Use existing .quant files when available							
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☑ N tem M excision ☑ C carbamidomethylation	Protein inference Genes ~							
□ Ox(M) □ Ac(N+em) □ Phospho □ K-GG	Neural network classifier Single-pass mode $\scriptstyle \lor$							
Peptide length range 7 - 30 -	Quantification strategy $$$ Quant UMS (high precision) $$ $$ $$ $$ $$ $$ $$ $$ $$ $$ $$ $$ $$$							
Precursor charge range 1 - 4 -	Cross-run normalisation $RT$ -dependent $\lor$							
Precursor m/z range 300 - 1800 -	Library generation $$\rm IDs,RT$ & IM profiling $$\sim$$							
Fragment ion m/z range 200 🚖 - 1800 🜲	Speed and RAM usage Optimal results $\checkmark$							

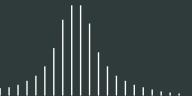
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  - One of the most widely used software and one of the most demanding
- Why using the Cloud?
  - Computation can take weeks on our local servers
  - 2VMs with 64 CPU and 128GB RAM
  - More VMs are about to be created with a in-house software solution to dispatch the jobs automatically

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- Why using the Cloud?
  - Convenient to keep these tools in one place
  - Easier maintenance
  - Access from home
  - Our VM: I6CPU, 32Go RAM, ITo

## Galaxy





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- Local servers are not able to process such data
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- Perspectives
  - Development of Cumulus to dispatch the jobs on different VM on the Cloud
  - RSync client to automate data transfer
  - Will be used for Dia-NN, other software will follow
  - <u>https://github.com/stars/AlexandreBurel/lists/cumulus</u>

### Acknowledgments



Directors : Christine Carapito, Sarah Cianferani





