

## Preprocessing, Management, and Analysis of Mass Spectrometry Proteomics Data

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## Outline

- Mass Spectrometry-based Proteomics
  - MS data
- Preprocessing of MS Data
  - Noise Reduction
  - Binning
  - Peaks Alignment
- MS-Analyzer, a platform for spectra analysis
  - Ontology+Workflow
- Conclusions and future work

## **Mass Spec**trometry-based Proteomics



### • A spectrum is a large sequence of value pairs (*I*, *m*/*Z*):

- I (intensity) depends on the quantity of the detected biomolecules,
- m/Z (mass to charge ratio) depends on the molecular mass of detected biomolecules

### **MALDI-TOF** spectrum



# **Preprocessing of MS Data**

- Each spectrum data is the result of two measurements (*I*, *m*/*Z*) that are corrupted by noise
- Preprocessing aims to correct *intensity* and *m*/*Z* values in order to:
  - (i) reduce noise,
  - (ii) reduce amount of data, and
  - (iii) make spectra comparable (Data calibration or Peaks Alignment).

## **Noise Reduction**

- Each mass spectrum exhibits a base intensity level (a baseline) which varies from fraction to fraction and consequently needs to be identified and subtracted.
  - Base line subtraction flattens the base profile of a spectrum
  - **Smoothing** reduces the noise level in the whole spectrum.



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## Normalization

- Normalization aims to make intensity comparable across different spectra. It is usually applied to each spectrum.
- Different normalization scheme

- Direct 
$$I_{j\_norm} = \frac{I_j - I_{min}}{I_{max} - I_{min}}$$

- Inverse  $I_{j\_norm} = 1 - \frac{I_j - I_{min}}{I_{max} - I_{min}}$ 

- Canonical 
$$I_{j\_norm} = \frac{I_j}{\sum I_j}$$

- Logarithmic
  - This method performs a logarithmic transformation on a set of K spectra. The intensity measure I<sub>hj</sub>for sample h, (h = 1,...,K), at m/Z value j, is transformed because of its skewed distribution.

# **Data Reduction: Binning**

- Binning performs data dimensionality reduction by grouping measured data into *bins*.
- It calculates, for each interval of m/Z values (bin),
  - an aggregate intensity (e.g. the sum of the intensities in the bin)
  - and a representative m/Z value (e.g. the median or the one with maximum intensity)
- and elects this couple as representative of the bin.



## **Data Reduction: Binning**



### **Raw Spectrum**



### **Binned Spectrum:**

- Window Size: 100
- Type: fixed

## **Peaks Alignment**

- Without alignment, the same peak (e.g. the same peptide) may have different m/Z values across samples
- This method finds a common set of peak locations (i.e. m/Z values) in a set of spectra, so that all spectra have common m/Z values for the same biological entities.

– Applied to an entire dataset.

- Window in which m/Z can be shifted is defined as window of potential shift (intrinsically related to instrument)
  - A precise definition of this value has to consider instrument data sheet or a set of measurement (a sort of calibration).
  - For MALDI-TOF MS instrument this values can be set as 0.5% of current m/Z values (variable window shift).

# **MS Data Mining Steps**

- 1. **Loading** of the raw spectra produced by mass spectrometer,
- 2. Preprocessing of the raw spectra data,
- 3. **Preparation** of the data mining input file (i.e. the Weka ARFF file),
- 4. Data Mining **analysis** (i.e. classification) of mass spectra,
- 5. Knowledge Models visualization (e.g. decision tree)
  - represented in a standard language as PMML

M. Cannataro, C. Comito, A. Congiusta, P. Veltri, PROTEUS: a Bioinformatics Problem Solving Environment on Grids, *Parallel Processing Letters*, World Scientific, 2004.

# MS-Analyzer: a software platform supporting MS Analysis Workflows

- MS-Analyzer is a Grid-based Problem Solving Environment for the preprocessing, management and data mining analysis of MS data. It uses
  - domain ontologies to model
    - software tools (e.g. preprocessing and mining) and their relationships
    - data sources (e.g. MALDI-TOF, LC-MS/MS spectra datasets)
    - constraints (e.g. binning cannot be applied twice)
  - and workflow techniques to design complex in silico experiments.
- MS-Analyzer is able to:
  - interfacing mass spectrometers (i.e. spectra data sources)
  - acquiring, preprocessing and managing MS data on the Grid
  - offering composable filtering, preprocessing, and DM services
  - sharing experiments (data, workflows and knowledge)
- enabling a Virtual Proteomics Laboratory





## **Loading Spectra in different Formats**

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Spectra List	Home Modelling Viewer WfD		
	Default settings		
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	[03/10/2005 - 17:03.54] - MS-Anal [03/10/2005 - 17:04.10] - Setting V	yzer v1.0 - Welcome! Vindow opened.	

## **Loading Domain Ontologies**

🥩 MS-Analyzer - Magna File Help	Graecia, Catanzaro University	
Spectrum Tree	Worksheets	
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MS_EXP28	[03/10/2005 - 17:04.10] - Setting Window opened. [03/10/2005 - 17:07.22] - Reading of file(s) done.	
MS_EXP29	[03/10/2005 - 17:07.44] - Reading of file(s) done.	

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# **Ontology-based Workflow Design**

File Help  Spectrum Tree  Morksheets  Modelling Viewer WID  MS_EXP1  MS_EXP2  MS_EXP1  MS_EXP4  MS_EXP5  MS_EXP6  MS_EXP6  MS_EXP7  MS_EXP7  MS_EXP8  MS_EXP10  MS_EXP10  MS_EXP11  MS_EXP12  MS_EXP11  MS_EXP12  MS_EXP11  MS_EXP12  MS_EXP13  MS_EXP13  MS_EXP14  MS_EXP15  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP10  MS_EXP12  MS_EXP11  MS_EXP12  MS_EXP11  MS_EXP12  MS_EXP13  MS_EXP12  MS_EXP13  MS_EXP14  MS_EXP12  MS_EXP15  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP12  MS_EXP13  MS_EXP12  MS_EXP13  MS_EXP13  MS_EXP13  MS_EXP15  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP17  MS_EXP12  MS_EXP13  MS_EXP13  MS_EXP13  MS_EXP13  MS_EXP15  MS_EXP17  MS_EXP
Spectrum Tree Spectrum Tree Spectr
MS_EXP18       Fork         MS_EXP19       Merge         MS_EXP20       Merge         MS_EXP21       Node         MS_EXP23       End         MS_EXP24       End         MS_EXP26       Sequencial-Patterr

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# Performance evaluation of preprocessing techniques

- We considered three different mass spectra datasets publicly available on Internet:
  - (i) Pancreatic Cancer dataset
    - 142 spectra each one with 6,772 (m/Z, intensity) measurements,
    - two classes: healthy and diseased patients;
  - (ii) Prostate Cancer dataset
    - 322 spectra each one with 15,154 measurements,
    - four classes: no disease, benign cancer, pc410, pcg10 patients;
  - (iii) Ovarian Cancer dataset
    - 49 spectra each one with 59,386 measurements,
    - two classes, 25 control and 24 disease.
- We measured: execution times, memory occupancy, classification quality

### **Spectra size [Kbytes]**



### **Execution time vs Preprocessing**



# **Quality of classification**

### Table 7. Classification indexes with none preprocessing

No Prep.	TP Rate	Precision	Recall
Pancreatic Cl1	0.513	0.5	0.513
Pancreatic Cl2	0.488	0.5	0.488
Prostate Cl1	0.774	0.786	0.774
Prostate Cl2	0.794	0.704	0.794
Prostate C13	0.269	0.269	0.269
Prostate Cl4	0.395	0.447	0.395
Ovarian Cl1	0.84	0.913	0.84
Ovarian Cl2	0.917	0.846	0.917

### Table 10. Classification indexes with binning and alignment

Bin.+Alig.	TP Rate	Precision	Recall
Pancreatic Cl1	0.613	0.563	0.613
Pancreatic Cl2	0.525	0.575	0.525
Prostate Cl1	0,937	0,922	0,937
Prostate Cl2	0,937	0,967	0,937
Prostate Cl3	0,577	0,556	0,577
Prostate Cl4	0,698	0,732	0,698
Ovarian Cl1	0.96	0.96	0.96
Ovarian Cl2	0,958	0,958	0,958

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### Table 9. Classification indexes with binning

Binning	TP Rate	Precision	Recall
Pancreatic Cl1	0.536	0.578	0.536
Pancreatic Cl2	0.614	0.573	0.614
Prostate Cl1	0.847	0.885	0.847
Prostate Cl2	0.825	0.813	0.825
Prostate C13	0.577	0.405	0.577
Prostate Cl4	0.558	0.615	0.558
Ovarian Cl1	0,849	0,913	0,845
Ovarian Cl2	0,934	0,944	0,917



## **Conclusions and future work**

- We surveyed some preprocessing techniques and presented a first prototype of MS-Analyzer
  - ontology-based workflow design simplifies workflow building and helps enforcing constraints
  - selective use of preprocessing techniques may improve execution times, memory occupancy and quality of data mining
- Future work will regard
  - the implementation of MS-Analyzer functions as Grid services and
  - the integration of a spectra relational database to allow "inside database" preprocessing

## **Questions?**

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- MS-Analyzer (standalone version) can be downloaded at
  - www.icar.cnr.it/proteus