

Proteomics: Large-Scale Identification of Proteins

Ulf Leser

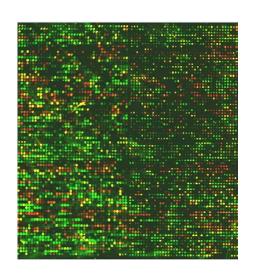
This Lecture

- Proteomics
- Separation
- Identification: Mass Spectrometry

Proteomics

- Genomics =
 Determining the genome of a species
- Transcriptomics =
 Determining the mRNA of a cell / tissue / state
- Proteomics =
 Determining the proteins in a cell / tissue / state
- Proteomics and transcriptomics have mostly identical goals
 - Understanding the processes happening in a cell
 - Differentiate between states, tissues, developmental state, ...
 - Biomarker: Finding protein/mRNA/... (forms, concentrations) that are characteristic for a certain phenotype (e.g., a disease)
- Metabolomics, epigenomics, bibliomics, ...

Proteomics versus Transcriptomics

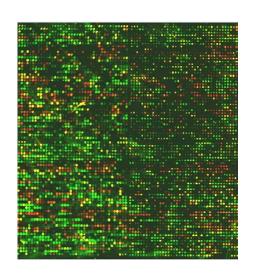




Advantages

- Proteins make you live, not mRNA
- mRNA is only indirect evidence with little correlation with proteome
 - Regulation by miRNA, post-translation modifications, decay, ...
- Protein survive (some time), mRNA is (mostly) transient
- Proteins are favorite drug targets

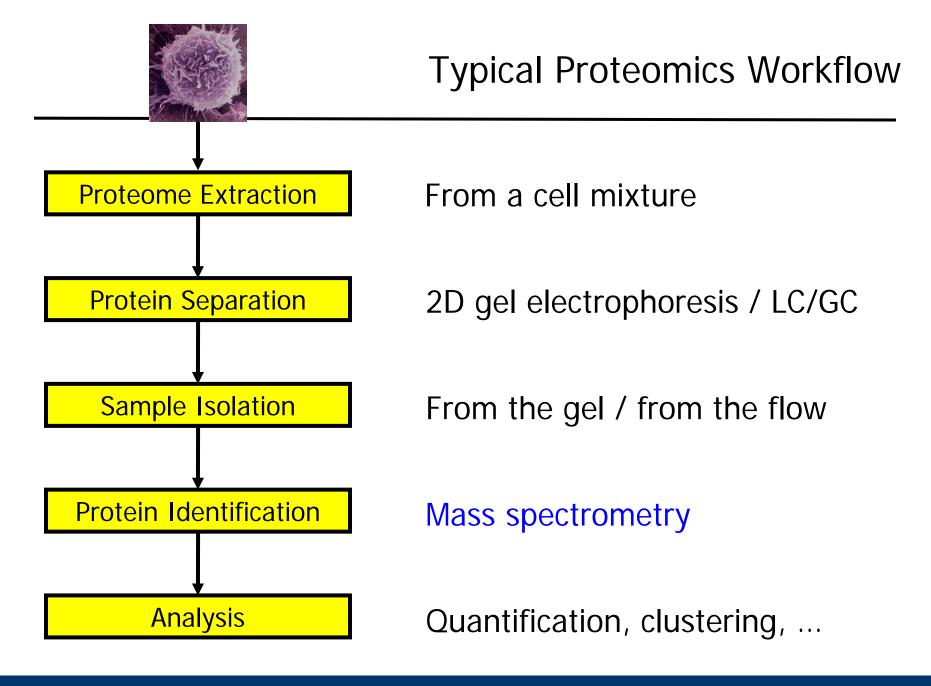
Proteomics versus Transcriptomics





Disadvantages

- Scale: ~20K genes, ~300K proteins, ~1M protein forms
- Handling: No PCR, no hybridization, no simple synthesis, no sequencing, no long-term "storage" as clones, high reactivity, ...
- Behavior highly context-dependent: Temperature, solution, pH, ...

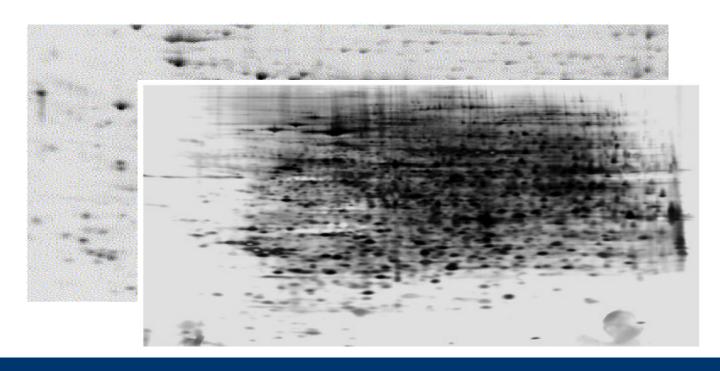


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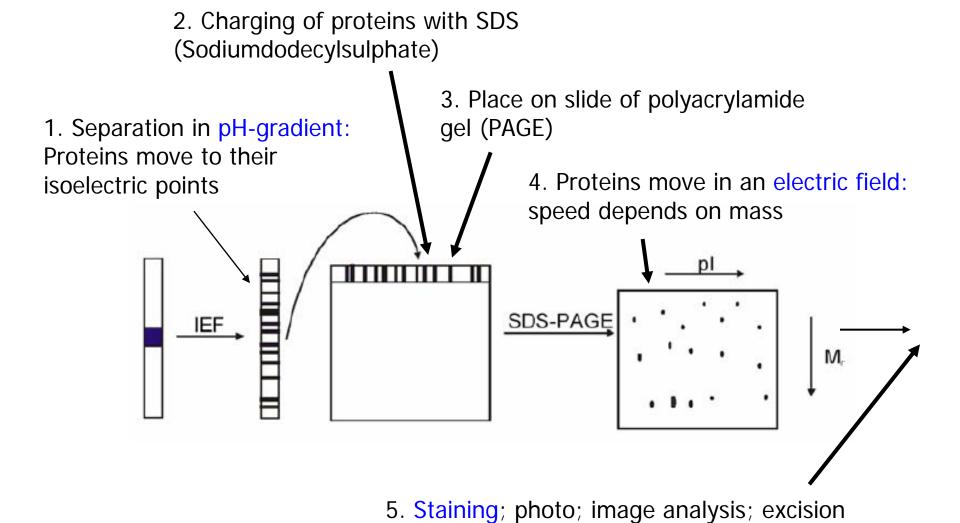
- Proteomics
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2D Gel Elektrophoresis

- Separation of proteins in two dimensions
 - Mass
 - Charge
- Every "spot" one protein (hopefully)



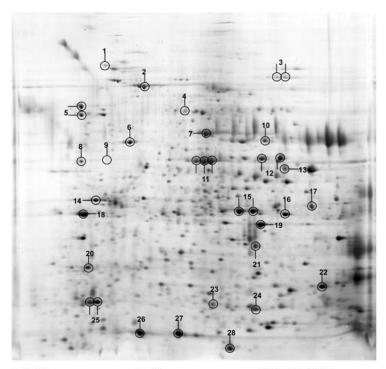
Method



Ulf Leser: Introduction to Bioinformatics

Analysis

- 2D-Page may separate up to 10.000 proteins
- Under identical conditions, the position of a particular protein is fairly stable
- Software for identification of proteins by position
 - After photo and image analysis
 - Align image to reference
- Various databases of 2D-Gels



- ATP:Guanidino Kinase
- Adenylate Dehydrogenase Calreticulin Actin
- Enolase
- Tropomyosin
- Serpin-like 10 Phosphoglycerate kinase
- 12 Aldolase
- 13 GAPDH
- 14 14-3-3 e
- 15 GST28 16 Triose Phosphate Isomerase
- 17 Elongation Factor Ia 18 14-3-3 homolog I
- 19 GST26 20 Calpain

- 21 Myosin Light Chain
 - 22 Cycophilin
 - 23 Superoxide Dismutase 24 Fatty Acid Binding Protein (Sm14)
 - 25 SME16 26 Thioredoxin
 - 27 Dynein Light Chain
 - 28 Ubiquitin 29 Adenylate Kinase

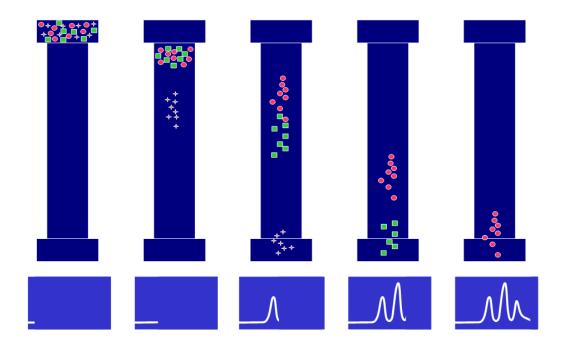
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Pro / Contra

- Comparably simple and cheap
- Disadvantages
 - No high-throughput much manual work
 - No robust quantification (spot intensity depends on staining)
 - Similar proteins (e.g. protein forms) build overlapping spots
 - Many restrictions
 - No proteins with <20KD or >200KD
 - No highly charged proteins
 - No detection of low concentrations
 - No membrane proteins (depending on method)
 - ...
 - No de-novo protein identification
 - Limited accuracy in comparative identification

Liquide / Gas Chromatography

- Other option: GC/LC
 - Chamber contains two phases (liquid / liquid, liquid/gas)
 - Different speeds depending on mass/charge ratio
 - Separation by retention times

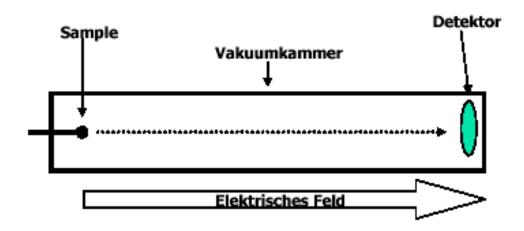


This Lecture

- Proteomics
- Separation
- Identification: Mass Spectrometry
 - Method
 - Algorithms: Naïve, probabilistic

Mass Spectrometry

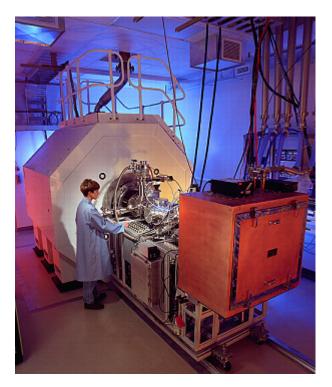
- Accelerate particles (must be charged) in an electric field
- Detector measures hits at back wall
- Time of flight (ToF) proportional to mass
 - Other techniques exist (magnetic drift, ...)
- Spectrum of mass peaks is used to identify particle



Mass Spectrometry



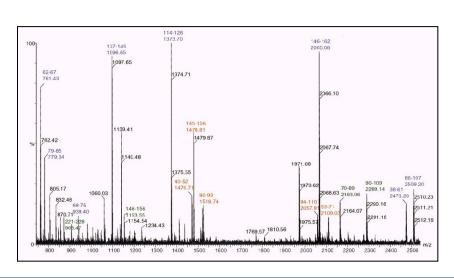
Source: http://imr.osu.edu



Source: http://www.sysbio.org

MS for Protein Identification

- Problem: Proteins are fragile and break during acceleration
- Solution
 - Break proteins into peptides before acceleration (digestion)
 - Measure peptides ToF (each peptide one signal)
 - Identify protein based on spectrum of peptide signals
- In theory, every protein has an almost unique spectrum
 - Using modern MS/MS, even different forms of the same protein are separable



Digestion

Trypsin:

Cleaves after Arginine und Lysine if next AA is not Proline

N-Asp-Ala-Gly-Arg

Ser-Glu-Asn-Leu-Ile-Arg

His-Cys-Lys-Pro Lys

Thr-Tyr-C

Chymotrypsin:

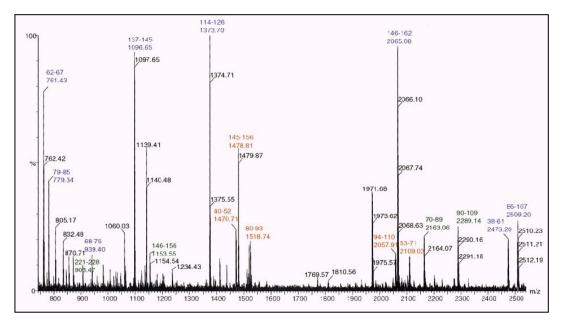
After Tyr, Trp, Phe, Met

Ionization

- Problem: Peptides often are uncharged no acceleration
- Solution
 - MALDI Matrix Assisted Laser Desorption / Ionization
 - Peptide are embedded in a "matrix"
 - Crystallization with charged, light-sensitive molecules
 - Fire on crystal with laser
 - Light-sensitive molecules vaporize and carry peptides with them
 - Accelerate
- Other techniques known
 - E.g. ESI: electrospray ionization

From Spectra to Peaks

- Detecting peaks and assigning them to peptides is difficult
 - Technical biasin runs / machines
 - Inaccuracies of measurement
 - Inhomogeneous sample preparation
 - Matrix etc.
 - Different quantities of peptides



- Creating a spectrum: Signal processing (not covered here)
 - Peak detection, peak disambiguation, noise filtering, ...

This Lecture

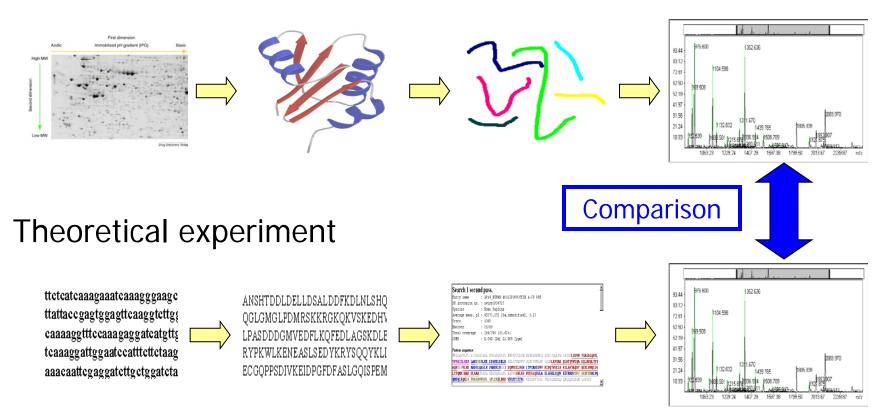
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 - Algorithms: Naïve, probabilistic

Algorithms for Protein Identification from Spectra

- We focus on database-based identification
- Idea
 - We have a database D of protein sequences d₁, d₂, ...
 - Each d_i is subjected to electronic digestion set of peptides
 - · For each peptide, we know its theoretical ToF
 - Compute a theoretical spectrum s_i for each d_i
 - Measure real spectrum s of unknown protein k
 - Compare empirical spectrum s with all theoretical spectra s_i
- We can only find what we already know

Illustration

Real experiment



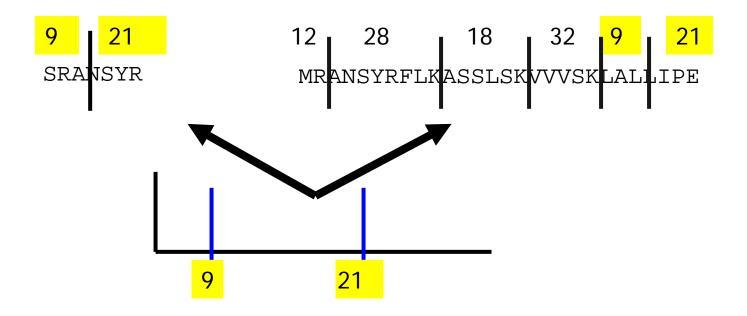
Naive Algorithm: Hitcount

- Compare measured spectrum s with all s_i in DB
- Protein d_i which has the most peaks in common wins
 - Input: $s = \{p_1, ..., p_m\}$, database D with many $s_i = \{q_{i1}, ..., q_{ij}\}$
 - For each s_i : Compute $|s \cap s_i|$
 - Protein d_i where s_i has maximal overlap wins
- Complexity?
 - Keep peak lists s and s_i sorted
 - We need to compare |s| hits with |D| proteins in DB
 - Let q be the average number of peaks in a database spectrum
 - Together: $\sim (|s|+q)^*|D|$ comparisons
 - Can be sped-up further (indexing)

Why "Naïve"?

- Peptide masses are not really equal (e.g. isotopes)
 - Small deviation nearest peak; match might not be unique
- Some (short) peptides are more frequent than others
 - Some peptides appear in almost all proteins little signal
 - Smaller peptides are much more frequent but much less specific
 - And peptide length is stochastic
 - Frequent peptides should have a lower impact
- Proteins have different lengths
 - Longer proteins have a higher a-priori chance for more peak matches

Example



Which one would you prefer?

More Problems

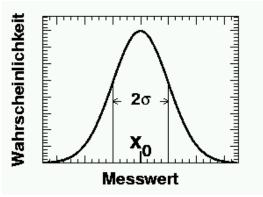
- Many sources of error
 - Enzymes don't work 100%
 - Theoretical spectra don't match
 - Protein sequences in DB contain errors
 - Especially when directly translated from genome
 - Leads to theoretical spectra not existing in nature
 - Posttranslational modifications modify real spectra
 - MS is not perfect spurious, shifted, missing peaks
 - Lead to false positive and false negative peak matches
- Closed-world assumption
 - What if real sequence is not in the database?
 - Some protein always has the highest count high enough?
 - No confidence scores

Practically Relevant Algorithms ??? SEQUAN?

- Heuristic: MOWSE (outdated)
 - Considers total protein mass and peptide frequencies
 - Generates a score
- Probabilistic algorithm: Profound
 - Copes with measurement errors, deviation in protein mass, and different peptide frequencies
 - Generates a probability of match for each protein (~ confidence)
- Many more (and newer) algorithms
 - MASCOT, PeptIdent, ProteinProspector, SEQAN, ...

Example of a Probabilistic Method: ProFound [zcoo]

- Given: Measured spectrum D and a protein k
 - D: Previously s; k: previously s_i
- ProFound computes prob. p(k|D) that D was produced by k
- The formula is complex; its derivation is even more complex and skipped
- Basic assumption: Measured peptide masses are normally distributed around the "canonical" value
 - Most probable isotope composition



ProFound Approach

- First step: Assign peaks from k to closest peaks from D
 - A-priori assignment is a strong first filter; errors are propagated
- Then compute probabilities using

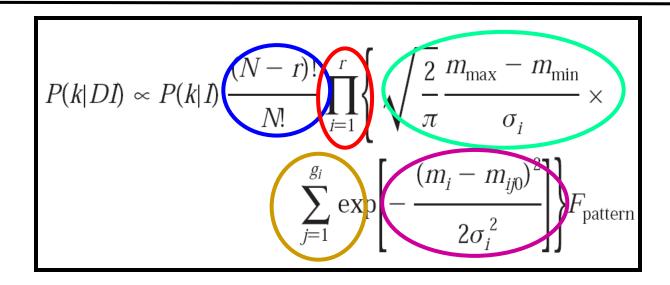
$$P(k|DI) \propto P(k|I) \frac{(N-r)!}{N!} \prod_{i=1}^{r} \left\{ \sqrt{\frac{2}{\pi}} \frac{m_{\text{max}} - m_{\text{min}}}{\sigma_i} \times \sum_{j=1}^{g_i} \exp\left[-\frac{(m_i - m_{ij0})^2}{2\sigma_i^2}\right] \right\} F_{\text{pattern}}$$

Legend

$$P(k|DI) \propto P(k|I) \frac{(N-r)!}{N!} \prod_{i=1}^{r} \left\{ \sqrt{\frac{2}{\pi}} \frac{m_{\text{max}} - m_{\text{min}}}{\sigma_i} \times \sum_{j=1}^{g_i} \exp \left[-\frac{(m_i - m_{ij0})^2}{2\sigma_i^2} \right] \right\} F_{\text{pattern}}$$

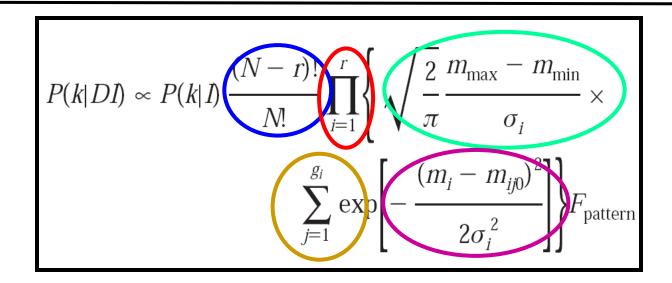
- p(k|D,I) = prob. that protein k was observed by spectrum D given the background information I
- p(k|I): A-priori probability of k in the given species / cell / tissue
- N: Predicted number of peptides of database protein k
- r: Number of hits between D and k (results from initial assignment)
- m_{max}, m_{min} range of observed masses for current peak (background)
- σ_i standard deviation of current peak (background)
- g_i: How often is the i'th peptide contained in k?
- m_i: Mean mass of the DB peak (background)
- m_{ii0}: Measured mass of j'th occurrence of this peptide
- F_{pattern}: Heuristic factor dealing with "overlapping peaks"

ProFound Explanation



- How many of the expected peptides of k did we observe?
- Multiply probabilities of all hits
- "Freedom" of measurements of hits for this peptide
- Many predicted peaks may create only one measured peak
- Probability of the difference between the expected mass and the measured mass (assuming normal distribution)

ProFound Intuition



- Many hits (r ~ N) score goes down (outweighs influence of more factors in the red product)
- Hits with a small stddev or a broad range score goes up
- Many observed peaks match the predicted peaks score goes up
- Observed peaks close to canonical peaks score goes up
- Theoretical peak as high stddev scores go down (also green)

Critique

- Score assumes that protein is in the database
 - Better: formulate "null" hypothesis, compute prob. of the spectrum given the null hypothesis, and report the log-odds ratio as score
 - But this is not as simple done as said
- Assumes that every peak comes from "the" protein
 - But measurements might be contaminated with peptides from other proteins
- Assumes that observed peaks can be assigned clearly to predicted peaks
 - This problem is tried to be covered by F_{pattern}

Further Reading

- Basics on proteomics: Every Bioinformatics book
- Zhang, W. and Chait, B. T. (2000). "ProFound: an expert system for protein identification using mass spectrometric peptide mapping information." *Anal Chem 72(11): 2482-9.*
- Pappin, D. J. C., Hojrup, P. and Bleasby, A. J. (1993).
 "Rapid identification of proteins by peptide-mass fingerprinting." *Current Biology* 3(327-332).
- Survey: Colinge J, Bennett KL (2007) Introduction to Computational Proteomics. PLoS Comput Biol 3(7): e114