#### Today in HST.480/6.092

#### **Gil Alterovitz**



#### Announcements

Homework 2/3 due this Fri 5p
Projects: In progress

Today
 Intro to Proteomics II, Mass spec



#### **Organization: Levels of Abstraction**

Part I: Sequence

Part II: Expression

Part III: Proteomics

Part IV: Systems/Misc.

### Computer Science/Algorithms Perspective

- 6.096 Algorithms for Computational Biology (Spring)- Prof. Manolis Kellis
- This new course covers the algorithmic foundations of computational biology, combining theory with practice. We study the principles of algorithm design for biological datasets, analyze influential algorithms, and apply these to real datasets.

#### Topics include:

- Strings: biological sequence analysis, gene finding, motif discovery, RNA folding, global and local sequence alignment
- Genomes: genome assembly, comparative genomics, genome duplication, genome rearrangements, evolutionary theory
- Networks: gene expression, clustering algorithms, scale-free networks, machine learning applications to genomics



### **Evolution Perspective**

- 6.891 Computational Evolutionary Biology (Fall)- Prof. Robert C. Berwick
- Course Description
- Evolution from a computational, modeling, and engineering perspective. Why has it been easier to develop a vaccine to eliminate polio than to control influenza or AIDS? Has there been natural selection for a 'language gene'? Why are there no animals with wheels? When does 'maximizing fitness' lead to evolutionary extinction? How are sex and parasites related? Why don't snakes eat grass? Why don't we have eyes in the back of our heads? How does modern genomics illustrate and challenge the field? Extensive hands-on laboratory exercises in modelbuilding and analyzing evolutionary data.



### **HST-Perspective**

#### HST.512/HST.513 Genomic Medicine

Subject studies the use of industrialized methods of data acquisition and analysis to improve medical care. Questions addressed are: What new benefits of genomics can be anticipated in the near future in terms of new drugs and treatments? How can diagnosis and the diagnostic process be changed today? How do our prognostic abilities change? How does one manage the deluge of clinically relevant genomic data? What constitutes a genomic clinical trial? What are the useful features of alternative genomic technologies today and for the near future? What are the different kinds of genomic informational resources and databases? Are they useful and how? What are the ethical individual and corporate challenges ahead? What are the key limitations we face? Enrollment limited. I. Kohane, A. Butte, J. Drazen, T. Golub, S. Greenberg, J. Hirschorn, S. Lory, P. Park, M. Ramoni, A. Riva, Z. Szallasi, S. Weiss



### Mass Spec Lab Techniques

Harvard Chemistry 165. Experimental Physical Chemistry Catalog Number: 0667 Frank N. Keutsch Half course (spring term). Lectures: F., 1–2:30; laboratories M., or Tu., 1–5. EXAM GROŬP: 6, 7 Introduction to methods and techniques used in physical chemistry/chemical physics research laboratories. Nine of eleven laboratory assignments involve experiments conducted in current CCB Research Groups: molecular beams; mass spectrometry; Fourier transform infrared and NMR spectroscopies; laser ablation; laser spectroscopy; cavity ring-down spectroscopy; scanning tunneling and atomic force microscopy; kinetics. Computer-based methods of data acquisition and analysis are used throughout. Note: Recommended as an efficient preparation for research in experimental physical chemistry/chemical physics and related sciences. Prerequisite: Chemistry 160 or Physics 143a.

#### Proteomics/Mass Spec

 Harvard: BCMP 301 (formerly \*Genetics 327). High Throughput Functional Proteomics
 Catalog Number: 1535
 Edward E. Harlow (Medical School) 2863
 BCMP = Biological Chemistry and Molecular Pharmacology

 Harvard: Cell Biology 332. Mass Spectrometry and Proteomics Catalog Number: 1568 Steven P. Gygi (Medical School) 3939

#### **Proteomics and Mass Spec**



#### Gil Alterovitz HST.480/6.092



### Paradigm Shifts in Bioinformatics

- Sequencing (1980's to early 1990's)
  - DNA/RNA/Protein Sequence Analysis/sequence storage
- 3-D Protein Structure Prediction (Mid-1980's-late 1990's)
  - Databases of Protein structures
- DNA/RNA Microarray Expression Experiments (Mid-1990's to 2000's)
  - Databases of expression data
- Protein interaction experiments (Early 2000's to Present)
  - Databases with pairwise interactions
- Mass Spec proteomic pattern experiments (Early 2000's to Present)
  - Databases with mass spec, protein identifications, proteomic patterns
- Integration of multiple modalities (Ongoing)



### New Flexibility with SELDI-TOF



Division of Health Science & Technology

#### Fractionation



Figure by MIT OCW.

Alterovitz, G., et al., *Analysis and Robot Pipelined Automation for SELDI-TOF Mass Spectrometry.* Proceedings of the International Conference of IEEE Engineering in Medicine and Biology, San Francisco, CA, USA, 2004.



# Steps in SELDI-TOF

# **Washings** Non-binding proteins are washed \$ \$ 6°+ 1001 away, eliminating sample "noise". **Add Matrix**

#### SELDI-TOF Mass Spec





# Computational Proteomics $\cong$ Bioinformatics for Genomics



MDKSELVQKAKLAEQAERYDDMAAMKAVTEQGHELSNEERNLLSVAYFNYGWRR....

Figure by MIT OCW.



#### **SELDI-TOF Mass Spec Schematic**



$$\frac{m/z}{U} = a(t-t_0)^2 + b$$

#### Where: t = time of flight ( $\mu$ s) m = mass (Da) z = charge U = 20,000 Volts a = 0.272, b = 0, t<sub>0</sub> = 0.0038 are constants

#### Figure by MIT OCW.



#### Data Axes





#### Proteomic Pattern Clustering in N-Space



Figure by MIT OCW.



#### The Challenges: SELDI Issues

- Different Operator (reproducibility), repeated measurements by same operator (repeatability) =>Hardware/software automation
- 2. Not one:one mapping.
  - Many Peaks → 1 Protein (e.g. variability in machine measurement, different charges will appear on different parts of m/z axis). =>Hardware/software automation fractionation, Biological Validation
  - One peak → Many Proteins (e.g. too many proteins with similar mass). => Biological Validation
- 3. Current models are typically 'black boxes': Proteomic profiles rather than protein identifications. Proteins are typically not identified.





#### Ambiguity in SELDI: A Linguistic Analogy Language Representation: Spelling (& Pronunciation)

		= Homograph (Homophone) words spelled (& pronounced) alike (i.e. masses within +/- machine variability)	≠ Heterograph (Heterophone) words spelled (& pronounced) differently
'Meaning'	= Synonym	Same word	Synonym
	≠	Homonym	Different words/meanings
<u>(ey</u> and	: nuage Representation = SELDI '	Mass ' m/z	Which is same protein?

Language Representation = SELDI 'Mass,' m/z Word meaning='protein'

m/z

#### Analyzing States and Control in Proteomics

#### Chameleon version:

Key:

- 1. Researcher wants to test hearing in chameleons.
- 2. But, how to get chameleon to respond?
- 3. Researcher remembers reading that 'A chameleon darker color can be a sign of distress or anger.'
- 4. Researcher's experiment: Test hearing of chameleon by provoking it (loud noises) -> color change to signal that animal can hear\*.
- 5. Researcher add chemical that specifically kills chromatophore *cells*\*
- 6. Result: Chameleon does not respond to loud noises. Extrapolating to humans, researcher writes paper: 'Human Audition Potentially Mediated by Chromatophore Cells'
- 7. Our conclusion- we need to look at more than just analysis differential. We need to look at flow of control
- \* Skin color changes are initiated by moving small, black granules (melanosomes) from
   In *chromatophore cells*.



Photograph courtesy of Alastair de Wet and stock.xchng

Light Green © <-> Brown Ø Normal <-> 'Disease State'



Photograph courtesy of Christian Burger and stock.xchng

#### States (e.g. cancer vs. normal)= color vs no color change. Control = hearing controls hormones->chromatophore cell receptors->release granules (melanosomes)

#### Quantifying Automation Reduction in Peak Intensity Variation



Figure by MIT OCW.

- One hundred (20 manual, 80 automated) biological samples done with replicates (2x). Coefficient of Variation (CV) is 27.8% for automation vs. 45.1% for manual.
- Statistically significant with P < 0.001</p>

 $CV = \sigma I \mu$  (standard deviation/mean)

Alterovitz, G., et al., *Analysis and Robot Pipelined Automation for SELDI-TOF Mass Spectrometry.* Proceedings of the International Conference of IEEE Engineering in Medicine and Biology, San Francisco, CA, USA, 2004.

### Hardware and Analysis Automation Components



#### **Robotic Automation**



#### Analysis Pipeline



### **Robot Sample Preparation**





SELDI MS-TOF: Surface-enhanced laser desorption/ionization time-offlight mass spectrometry





# Raspap: Robot Automated Sample Preparation and Analysis Pipeline for Proteomics



HST

### BAP: Bioinformatics Automated Pipeline





### **Object-Oriented Tree Structure of** BAP



Harvard-MIT

**Division of Health** Science & Technology

### **Machine Learning Results**



Figure by MIT OCW.

Alterovitz, G., et al., *Analysis and Robot Pipelined Automation for SELDI-TOF Mass Spectrometry.* Proceedings of the International Conference of IEEE Engineering in Medicine and Biology, San Francisco, CA, USA, 2004.





#### Maximum range: 35000

Does 18,012 have a single charge? What about 15,740?



### Biological Protein Peak Identification

Image removed due to copyright considerations

Gel image



### Tandem Mass Spec

- Take advantage of high sensitivity at low peptide resolution (without a matrix that can add irreproducible 'noise' in that region).
- Use this to sequence small cut bits of proteins (puzzle pieces)
- Compare cleaved proteins sequences with database to identify the protein in the sample (complete puzzle).
  - Via cross-correlation of spectra with hypothesized spectra of database entries
- Yields: protein identification and abundance (via peak area/intensity.



### Challenges

Protein may not be in database
 Cleaved protein may match several database entries



Amino Acid	Symbol	Average molecular weight (Da)
Alanine	А	71.0788
Arginine	R	156.1876
Asparagine	N	114.1039
Aspartic Acid	D	115.0886
Cysteine	C	103.1448
Glutamine	Q	128.1308
Glutamic Acid	E	129.1155
Glycine	G	57.0520
Histidine	Н	137.1412
Isoleucine	1	113.1595
Leucine	L L	113.1595
Lysine	K	128.1742
Methionine	M	131.1986
Phenylalanine	F	147.1766
Proline	P	97.1167
Serine	S	87.0782
Threonine	Т	101.1051
Tryptophan	W	186.2133
Tyrosine	Y	163.1760
Valine	V	99.1326

### Tandem MS/MS with HPLC

 In this approach, the proteins in a sample are first digested (cleaved into smaller peptides) using a protease such as trypsin.

 Trypsin cuts proteins on the carboxyl side of positively charged amino acid residues (e.g. lysine and arginine). Image removed due to copyright considerations

Trysin 3-D Structure



### High Performance Liquid Chromatography

- The chromatography involves a separation based on attributes such as:
  - Hydrophobicity: lacking attraction to water
  - Strong cation exchange: net positive charge
  - Strong anion exchange: net negative charge
  - Size separation: size/molecular weight
  - Special affinity: interaction with particular functional groups



Figure by MIT OCW.



#### Schematic of Tandem QqTOF (quadrupole-timeof-flight) Mass Spectrometer Pass-through (ion guide)



Highly charged ions formed (>> 3 seen in SELDI) Figure by MIT OCW. (analyte solution pushed through needle into electric field->ionized droplets



The window generated by Q1 is  $\sim$  3 amu wide and a spectrum is generated of the selected peak.

Figure by MIT OCW.

Source: Samuel Lunenfeld Research Center



### Tandem Mass Spec



Figure by MIT OCW.





Fragmentation between the C and the amino Ns are often used for sequencing. The N terminal portions of these fragmentations are referred to as the b-series, the C-terminal portions are the y-series.

Figure by MIT OCW.





If we focus on the b-series ion peaks, we can see how each pair of peaks is separated by the mass of 1 particular amino acid.

Figure by MIT OCW.





Figure by MIT OCW.

