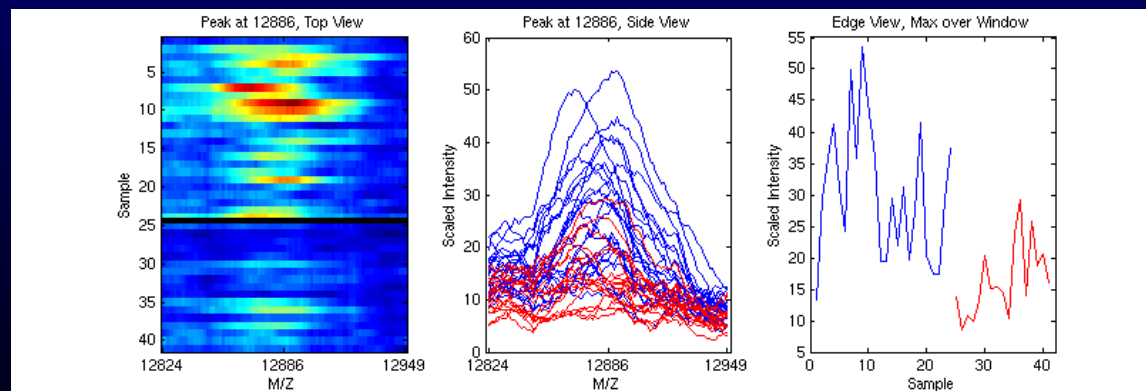


# The Analysis of Proteomic Spectra from Serum Samples

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# What Are Proteomic Spectra?

DNA makes RNA makes Protein

Microarrays allow us to measure the mRNA complement of a set of cells

Mass spectrometry allows us to measure the protein complement (or subset thereof) of a set of cells

Proteomic spectra are mass spectrometry traces of biological specimens

## Why Are We Excited?

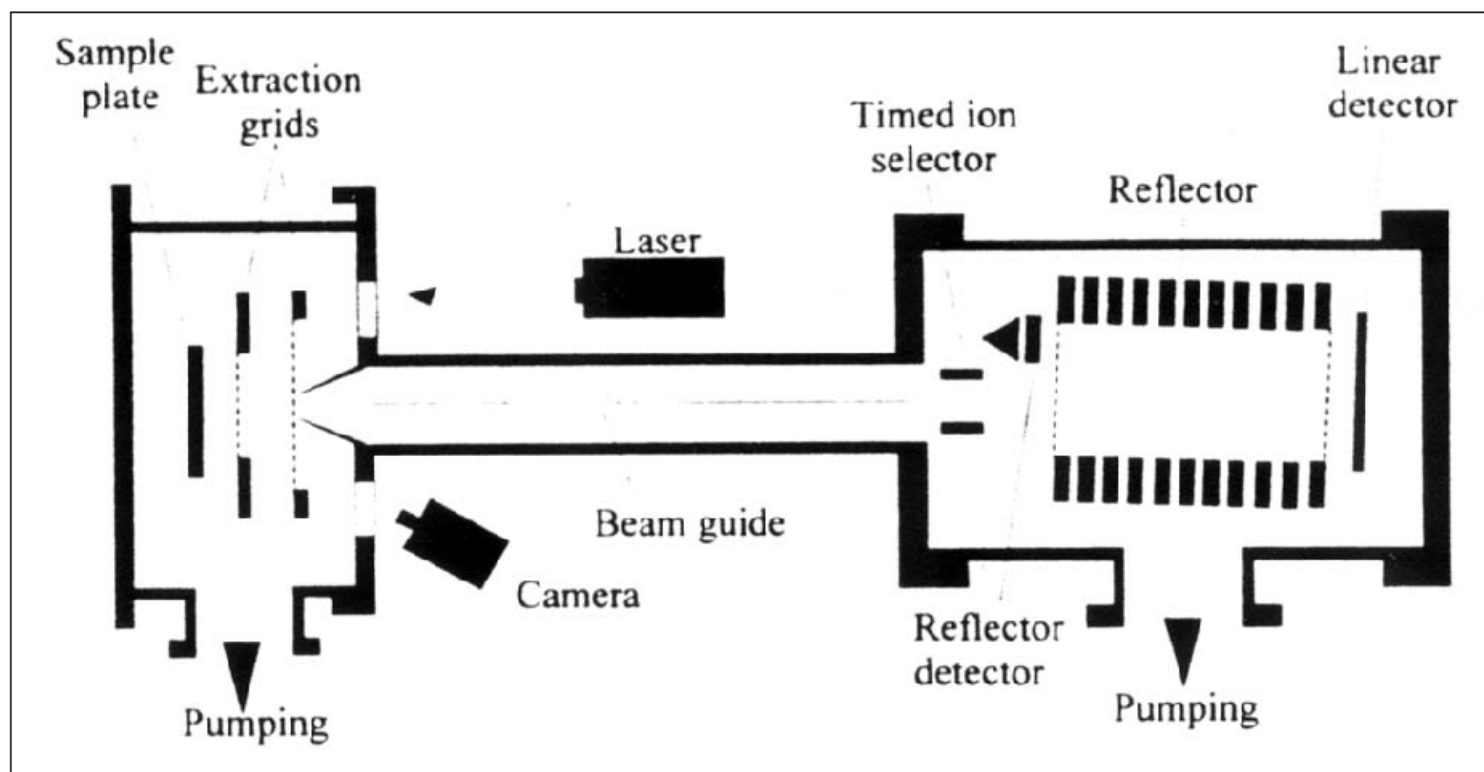
Profiles at this point are being assessed using serum and urine, not tissue biopsies

Spectra are cheaper to run on a per unit basis than microarrays

Can run samples on large numbers of patients

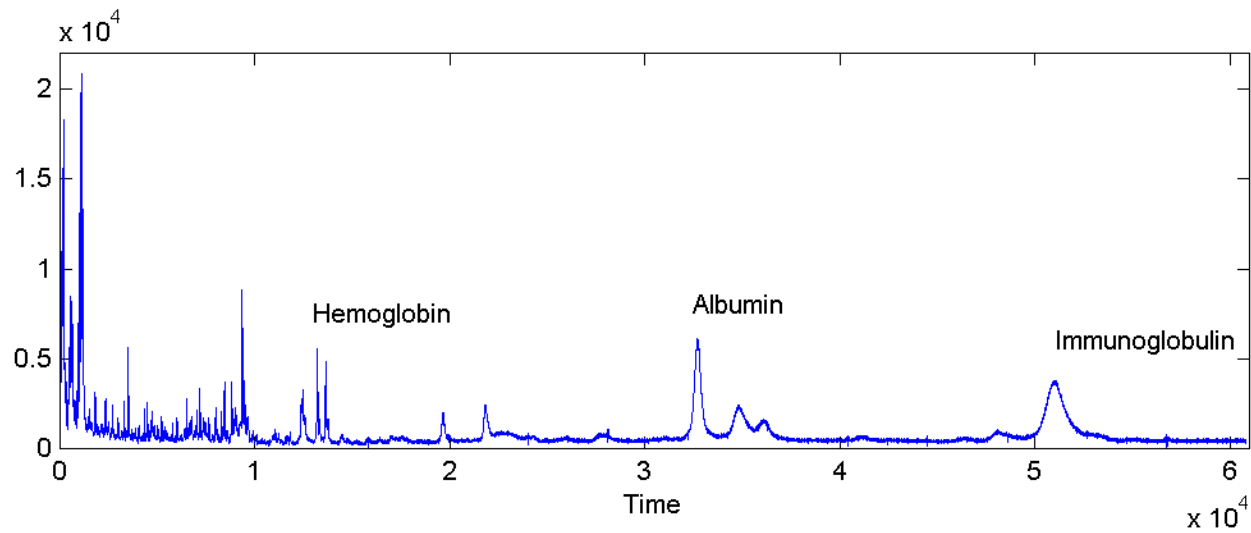
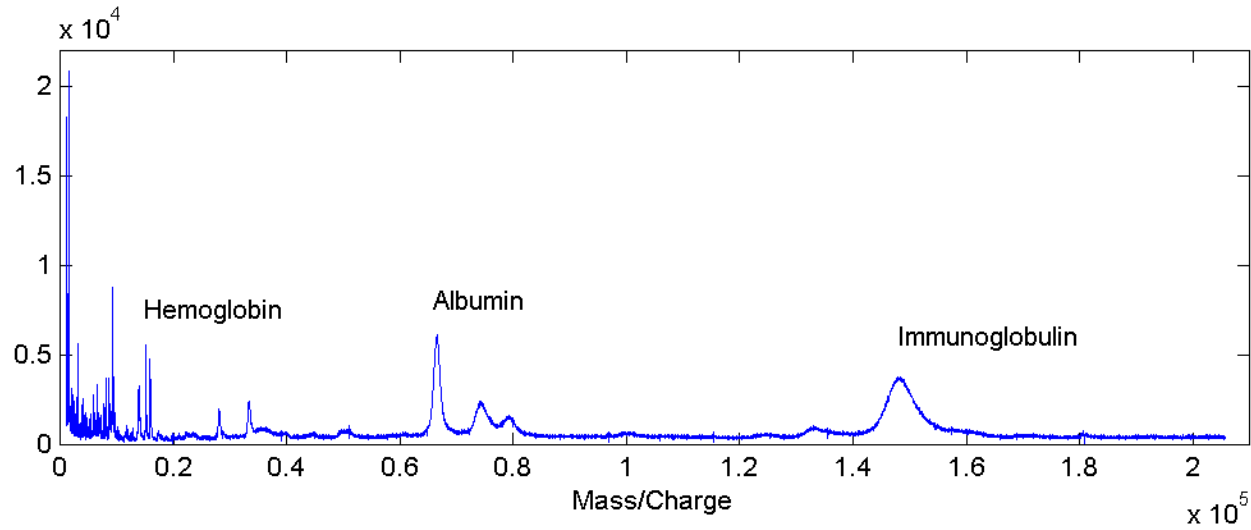
# How Does Mass Spec Work?

## Block Diagram of a MALDI-TOF

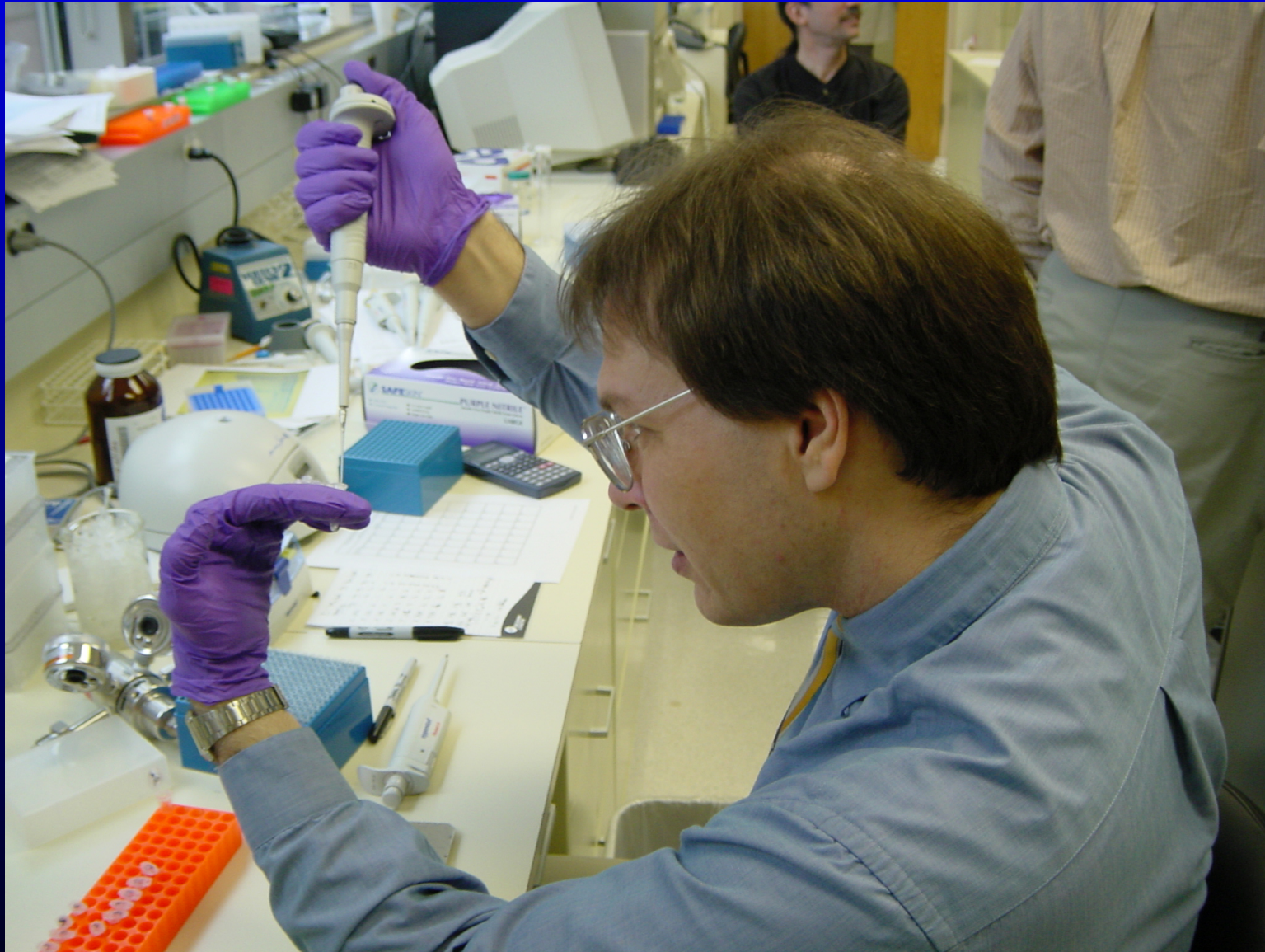


Vestal and Juhasz. *J. Am. Soc. Mass Spectrom.* 1998, 9, 892.

# What Do the Data Look Like?

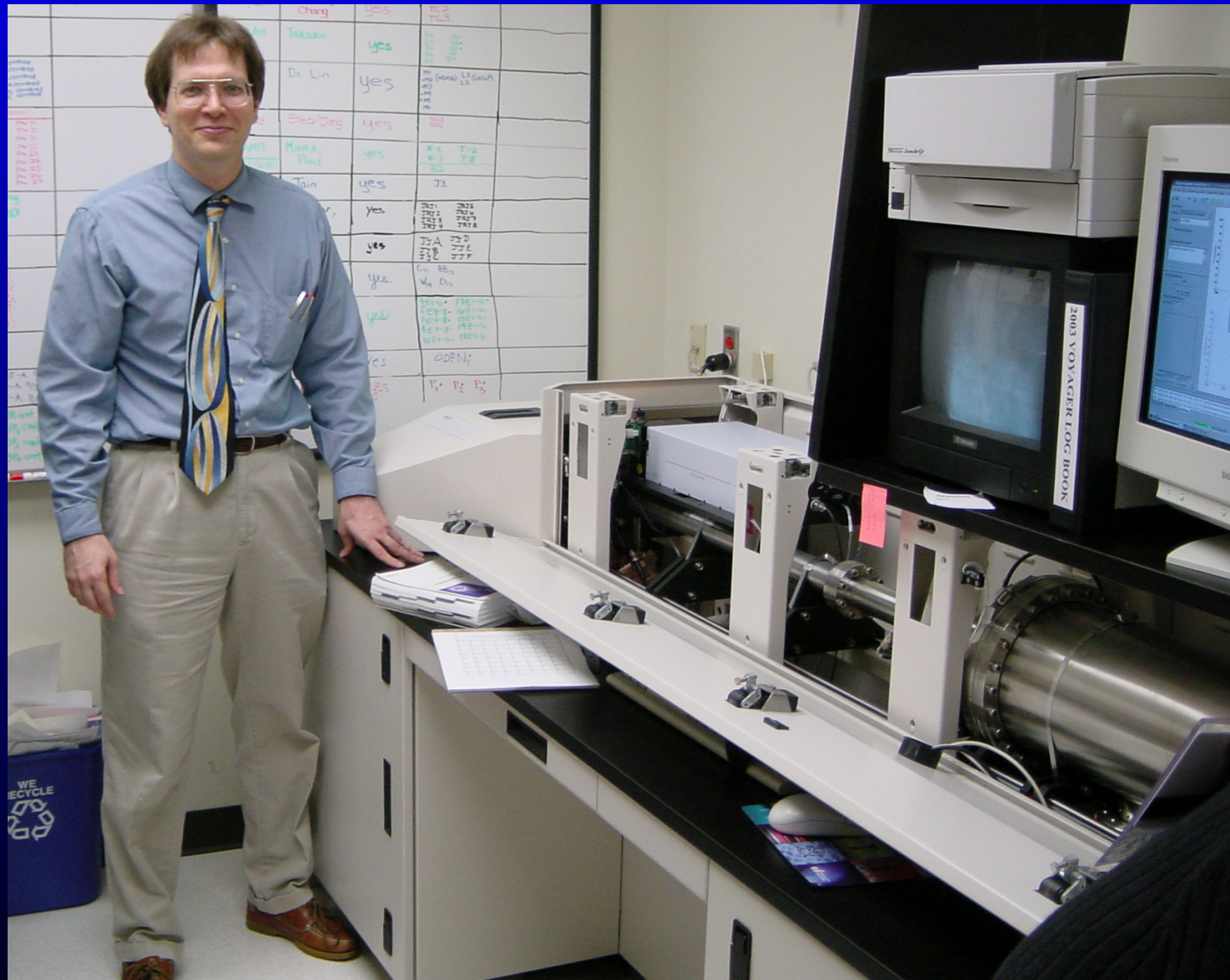


# Learning: Spotting the Samples



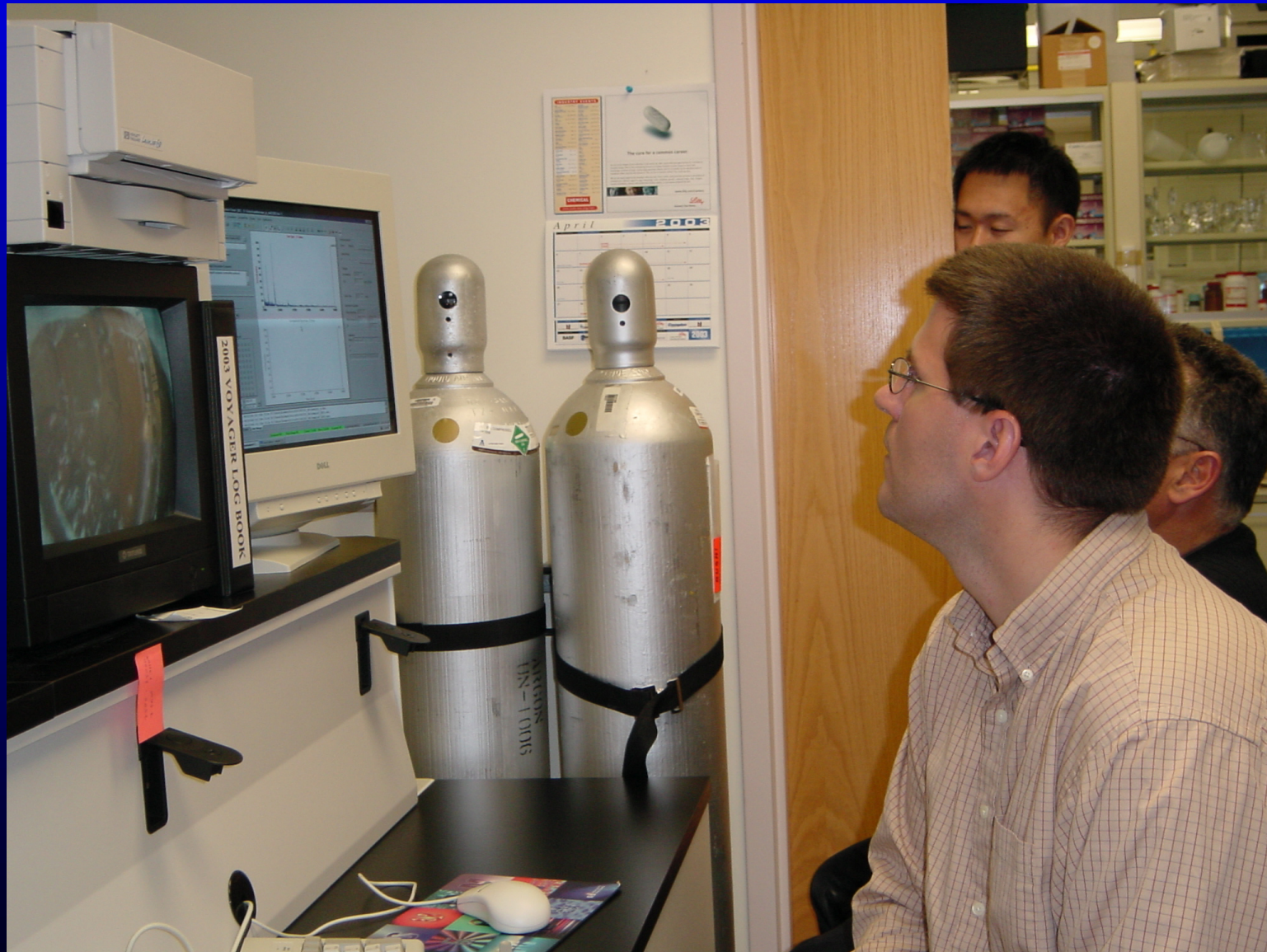


# What the Guts Look Like





# Taking Data





## Some Other Common Steps

Fractionating the Samples

Changing the Laser Intensity

Working with Different Matrix Substrates

## SELDI: A Special Case

[www.ciphergen.com](http://www.ciphergen.com)

Precoated surface performs some preselection of the proteins for you.

Machines are nominally easier to use.



## **A Tale of Two Examples**

Example 1 – Learning from the literature (SELDI)

Example 2 – Testing out our understanding (MALDI)

A story in pictures

# A SELDI Example: Feb 16 '02 Lancet

MECHANISMS OF DISEASE

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## Mechanisms of disease

### 🕒 Use of proteomic patterns in serum to identify ovarian cancer

*Emanuel F Petricoin III, Ali M Ardekani, Ben A Hitt, Peter J Levine, Vincent A Fusaro, Seth M Steinberg, Gordon B Mills, Charles Simone, David A Fishman, Elise C Kohn, Lance A Liotta*

---

- 100 ovarian cancer patients
- 100 normal controls
- 16 patients with “benign disease”

Use 50 cancer and 50 normal spectra to train a classification method; test the algorithm on the remaining samples.



## Their Results

- Correctly classified 50/50 of the ovarian cancer cases.
- Correctly classified 46/50 of the normal cases.
- Correctly classified 16/16 of the benign disease as “other”.

Data at <http://www.ncifdaproteomics.com> (used to be at <http://clinicalproteomics.steem.com>)

Large sample sizes, using serum

## The Data Sets

3 data sets on ovarian cancer

**Data Set 1** – The initial experiment. 216 samples, baseline subtracted, H4 chip

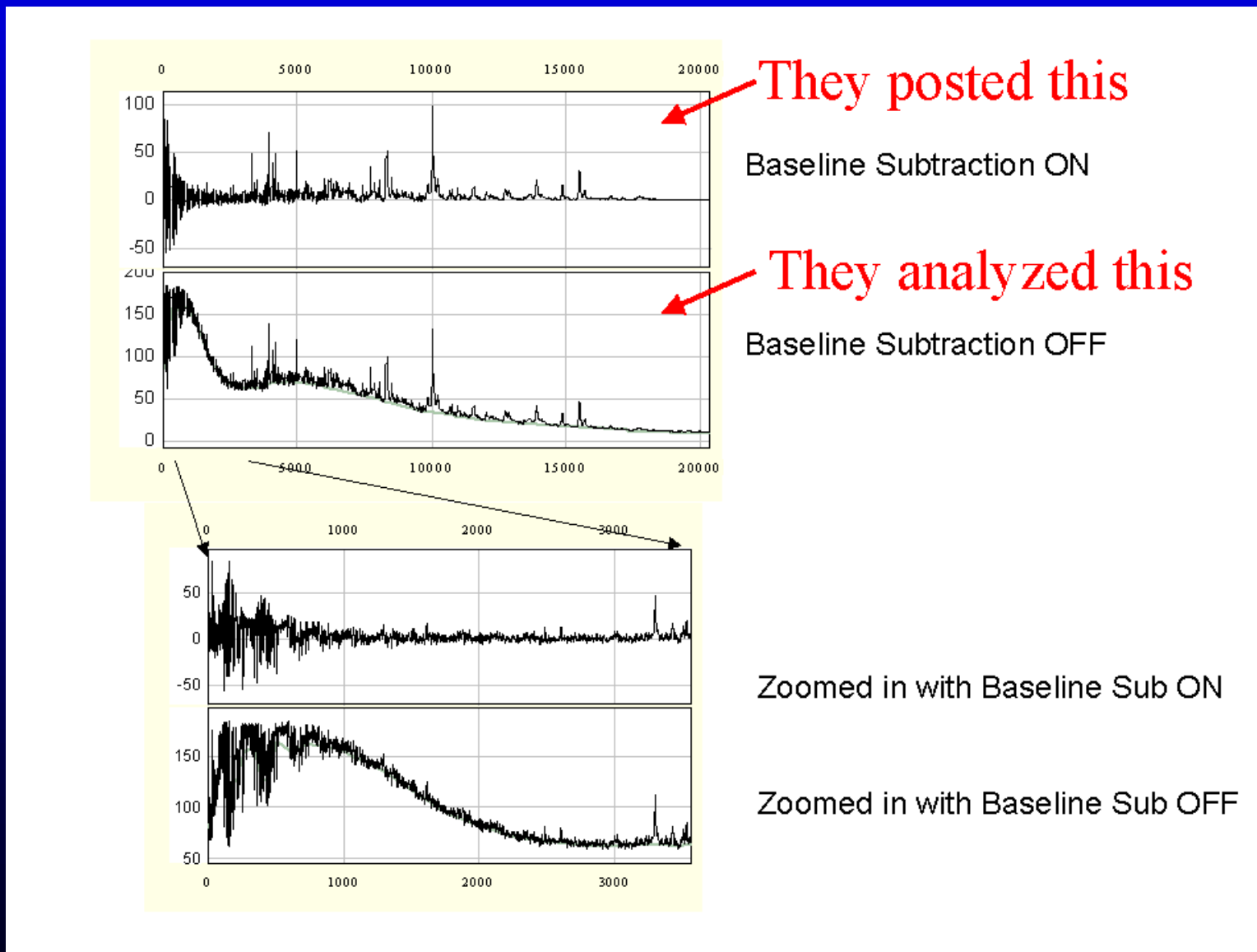
**Data Set 2** – Followup: the same 216 samples, baseline subtracted, WCX2 chip

**Data Set 3** – New experiment: 162 cancers, 91 normals, baseline NOT subtracted, WCX2 chip

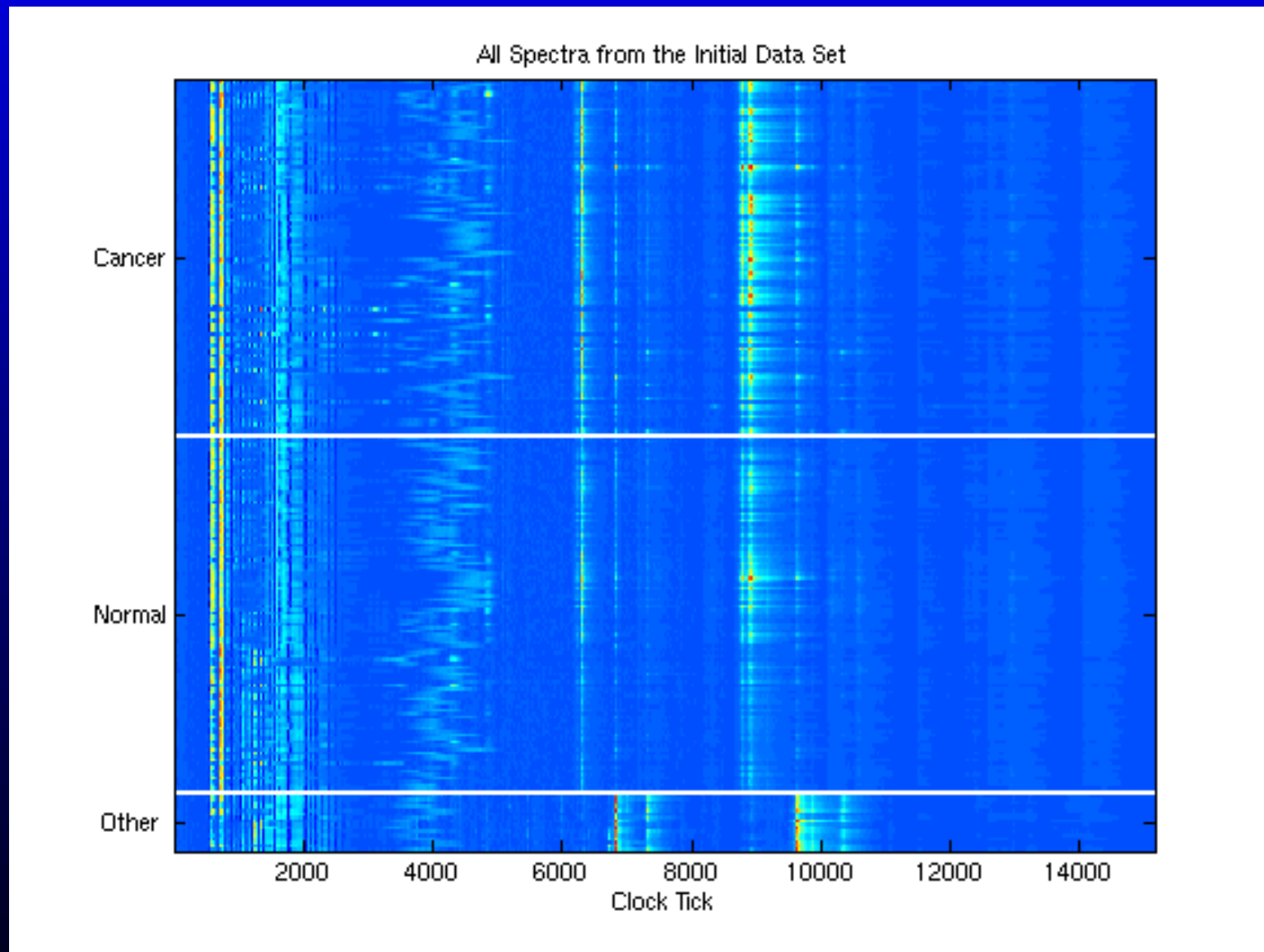
A set of 5-7 separating peaks is supplied for each data set.

We tried to (a) replicate their results, and (b) check consistency of the proteins found

# We Can't Replicate their Results (DS1 & DS2)

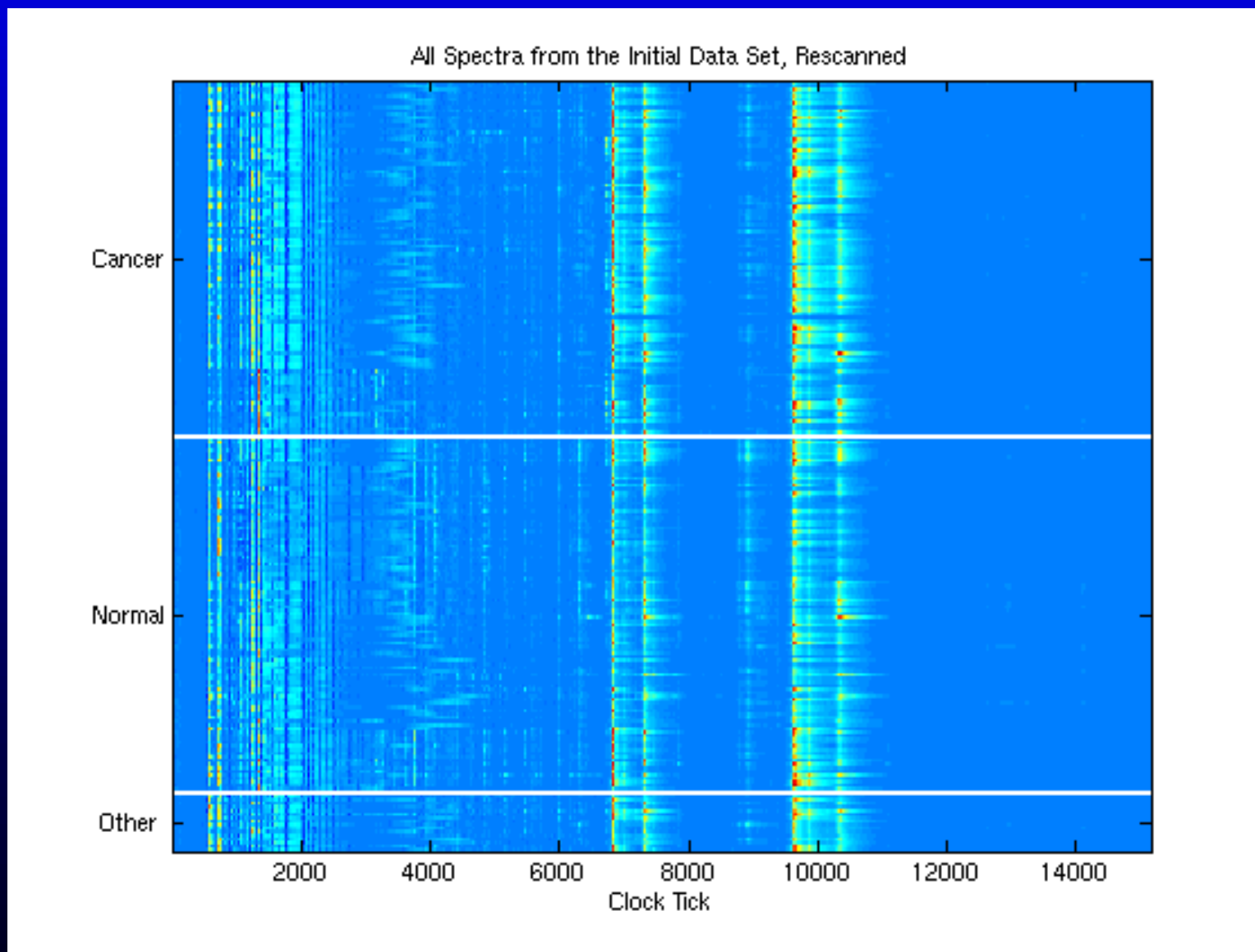


# Some Structure is Visible in DS1

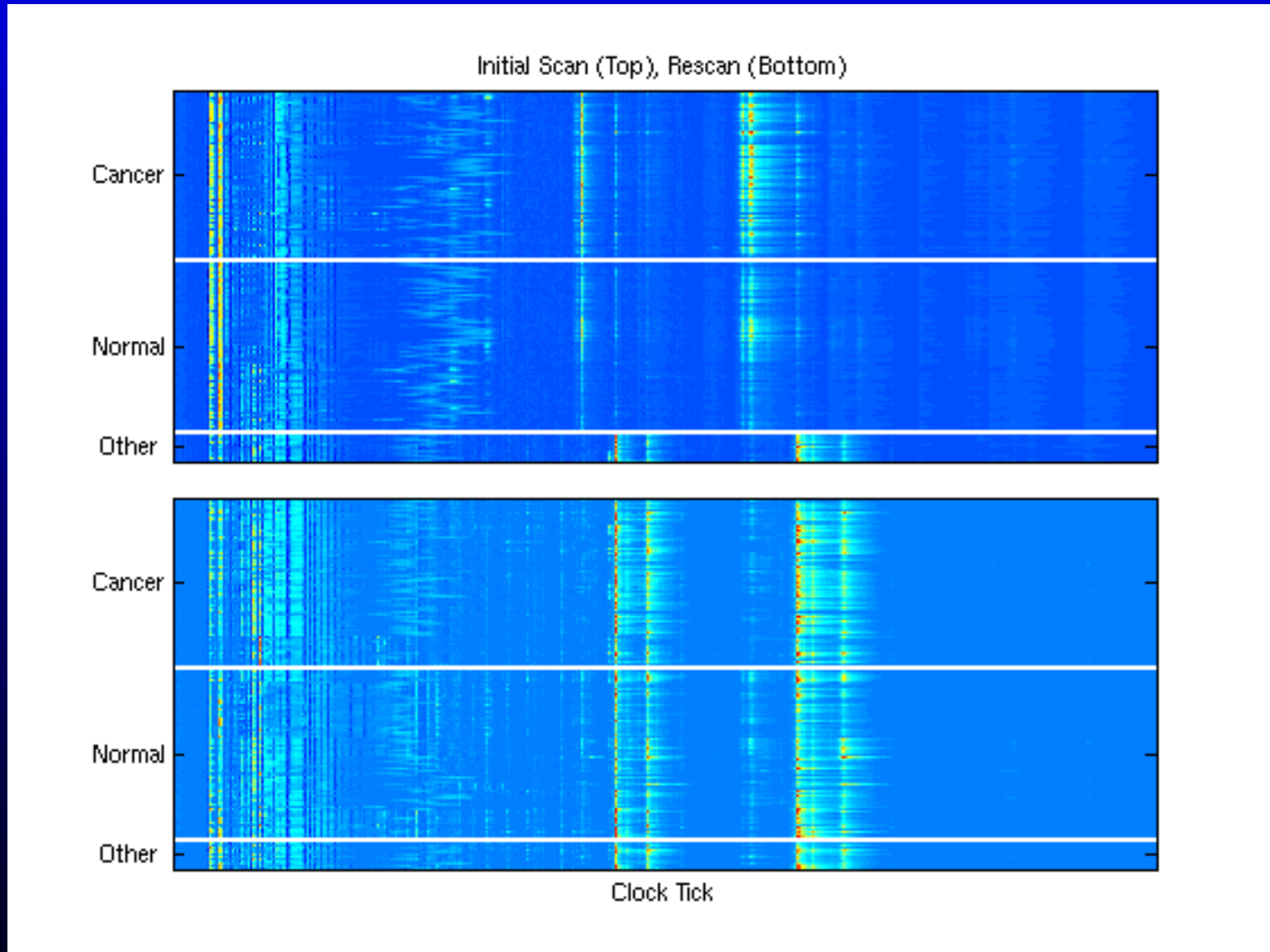




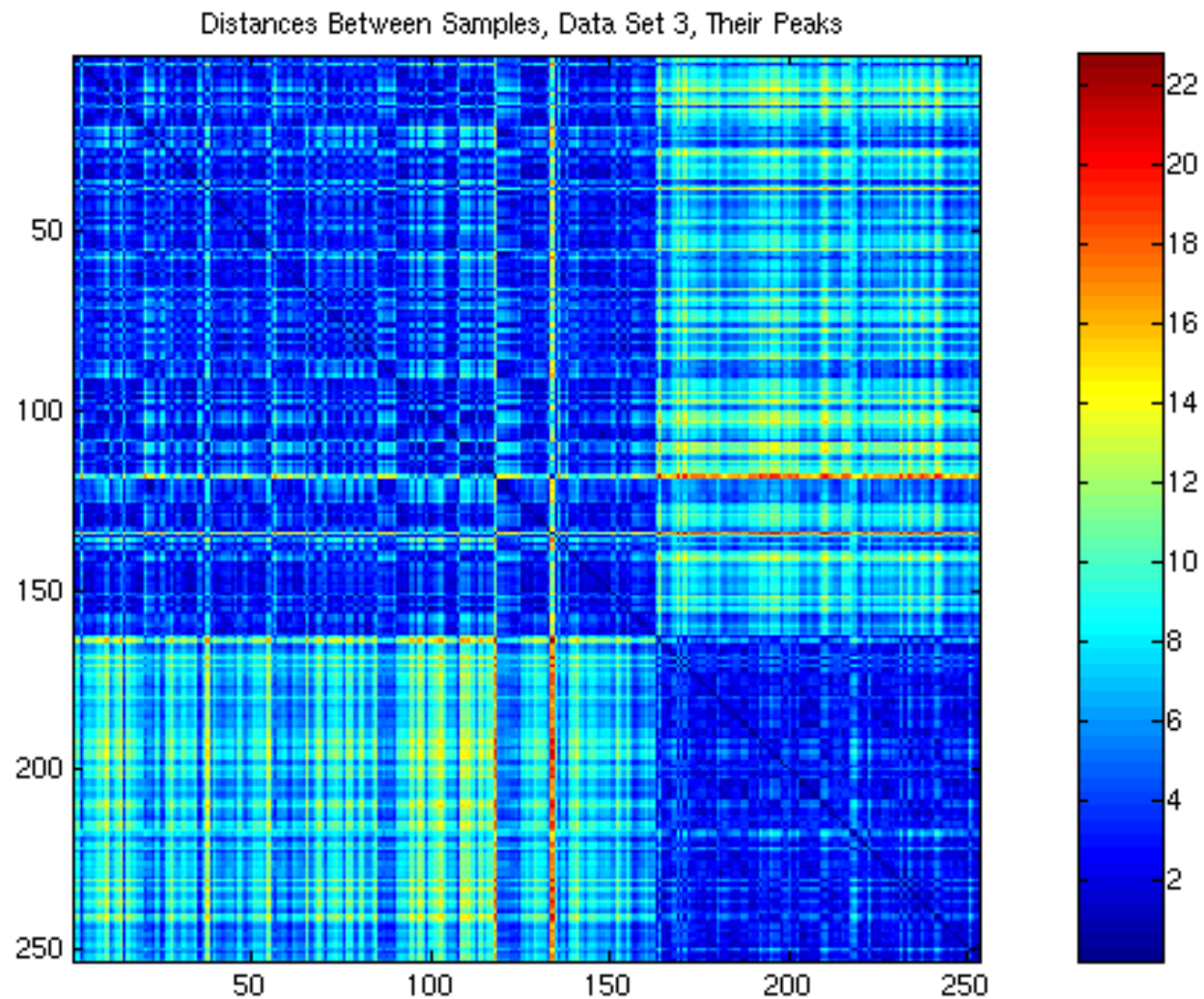
# Or is it? Not in DS2



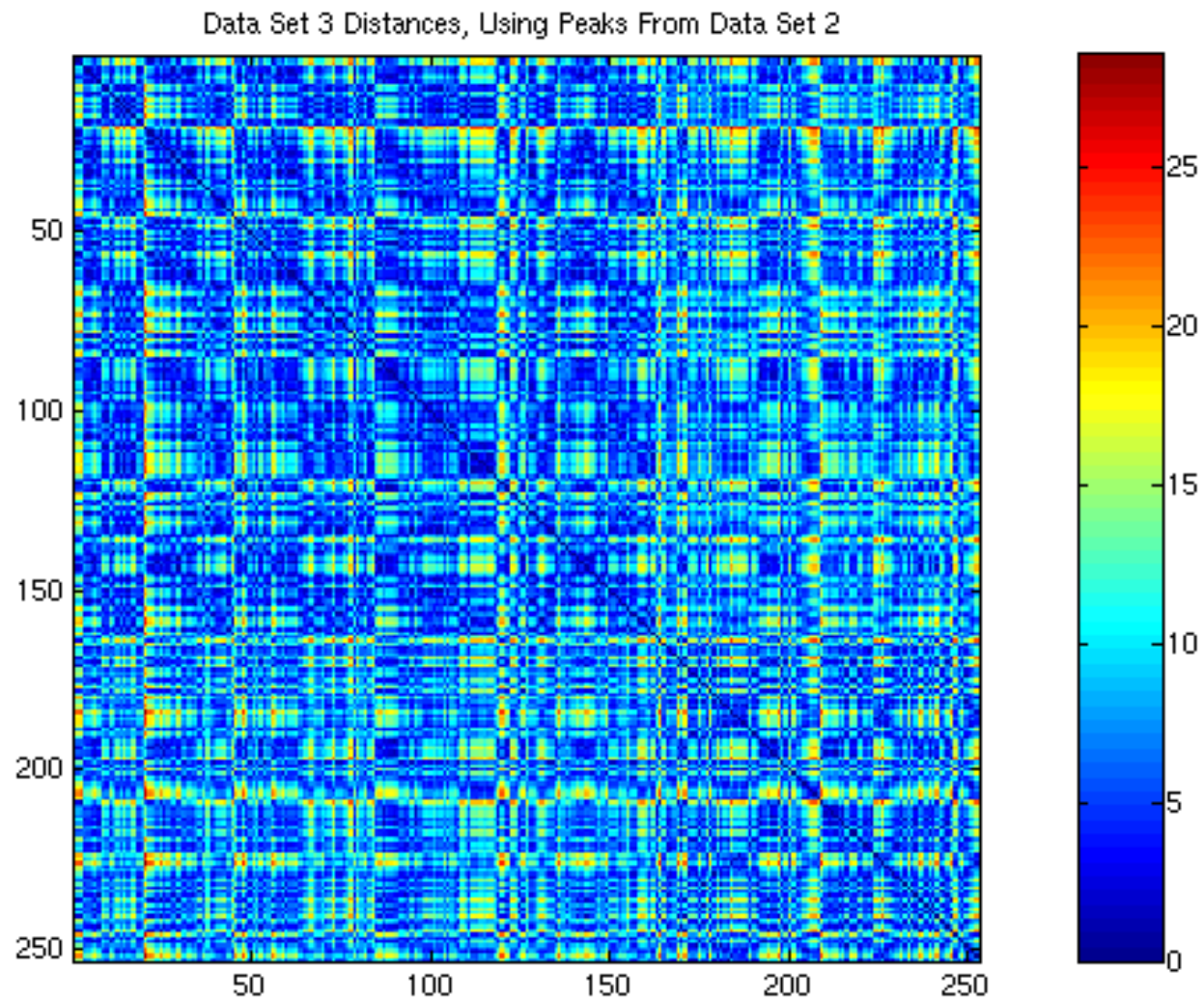
# Processing Can Trump Biology (DS1 & DS2)



# We Can Analyze Data Set 3!

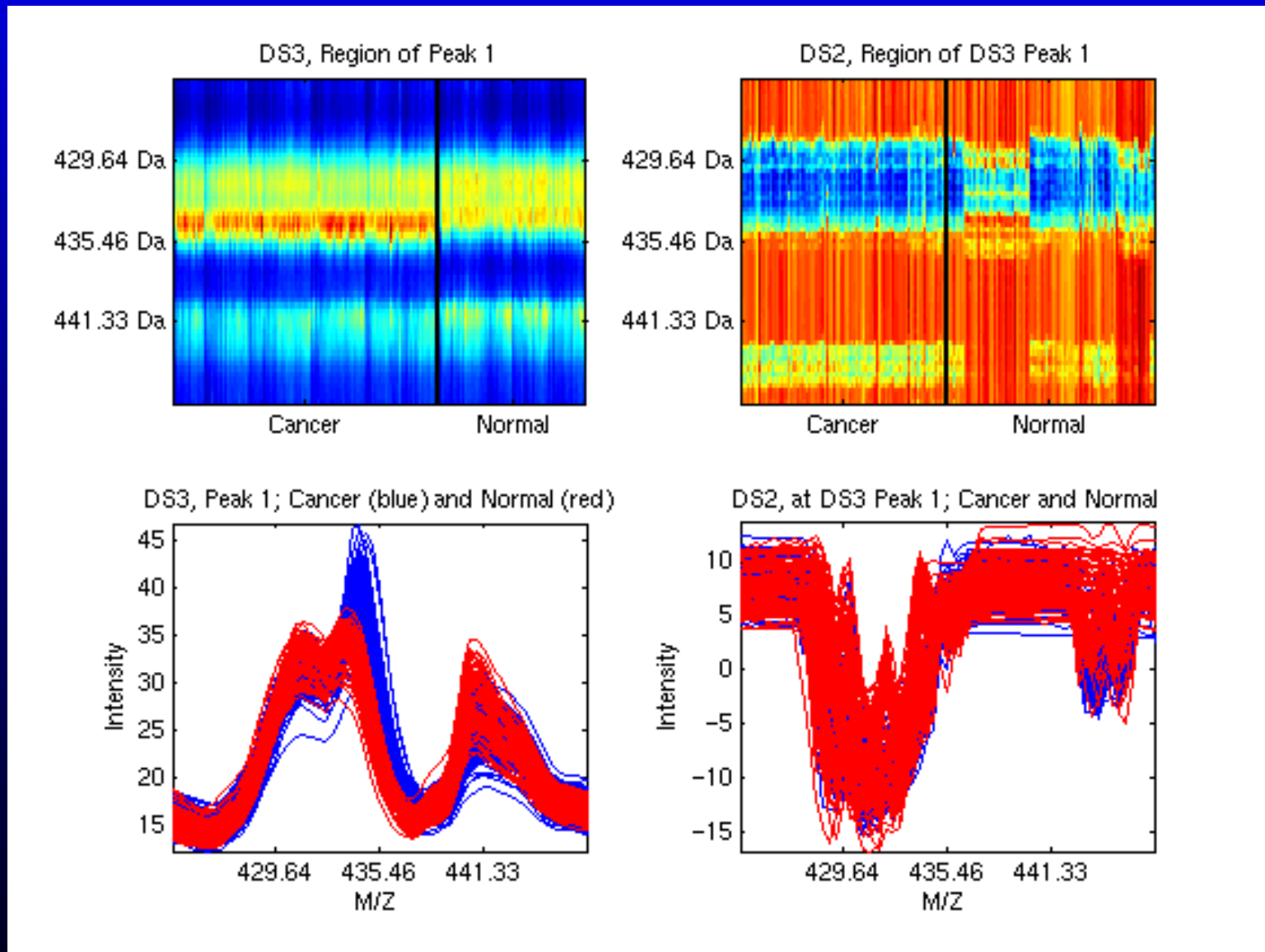


# Do the DS2 Peaks Work for DS3?

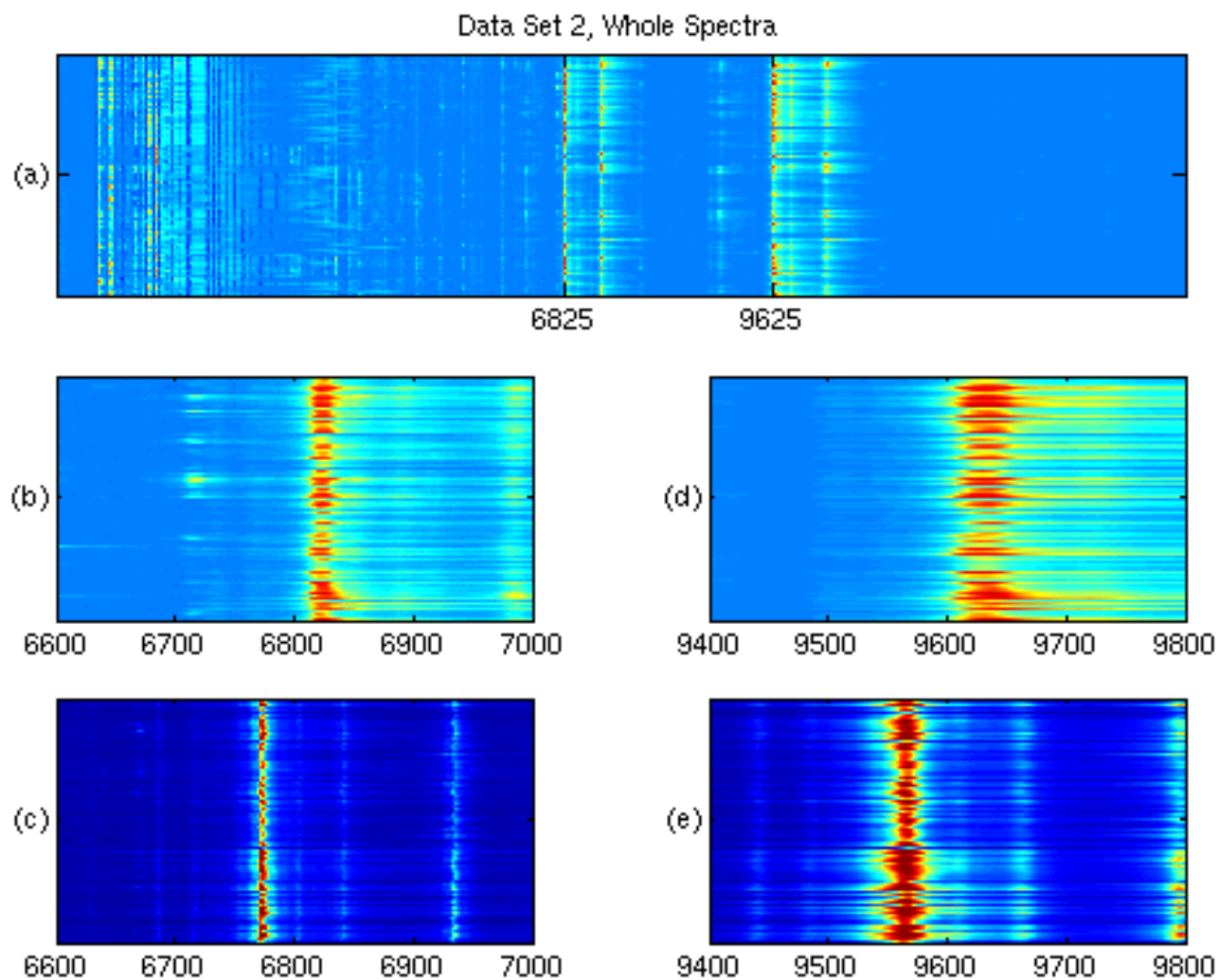




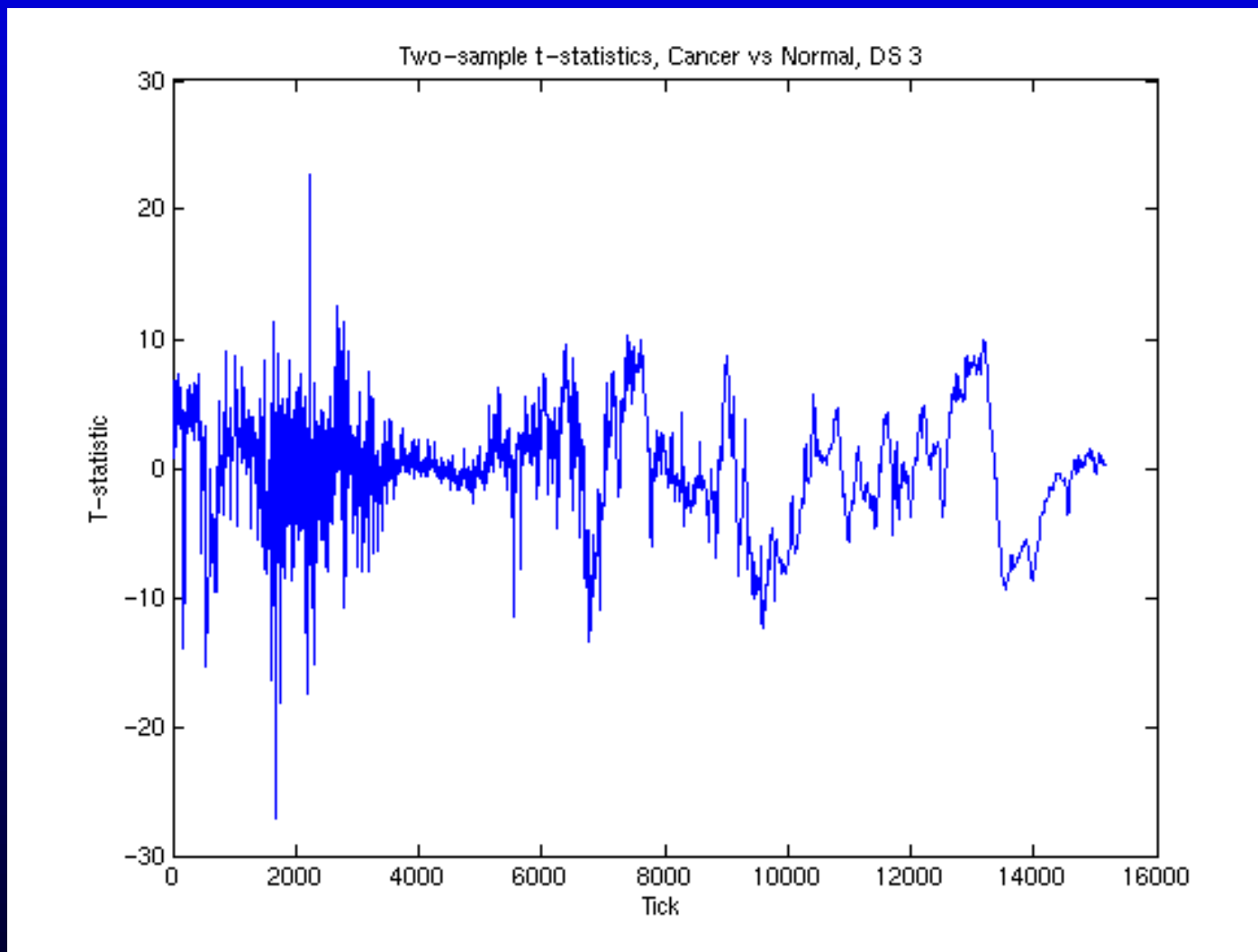
# Do the DS3 Peaks Work for DS2?



# Peaks are Offset

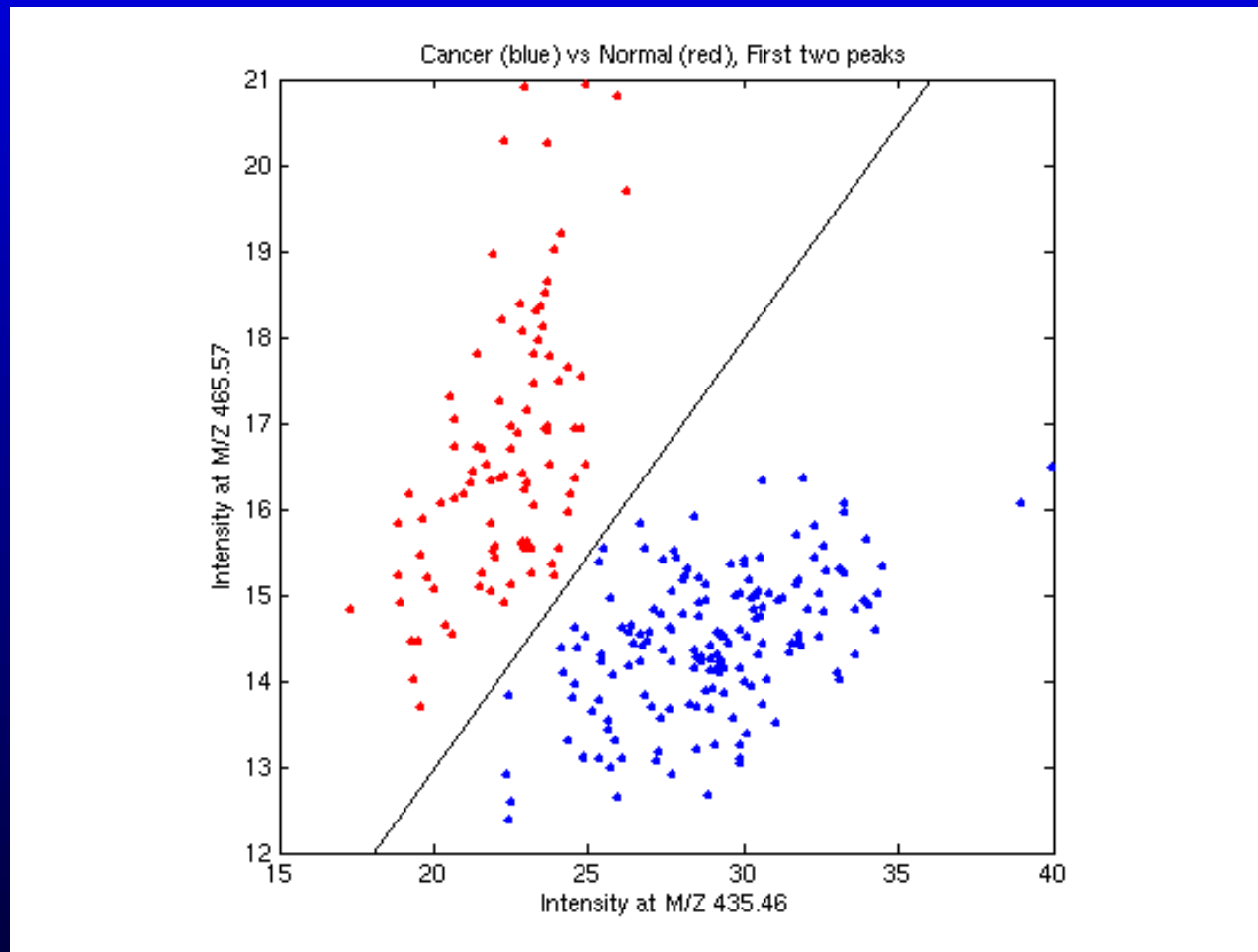


## Which Peaks are Best? T-statistics



Note the magnitudes: t-values in excess of 20 (absolute value)!

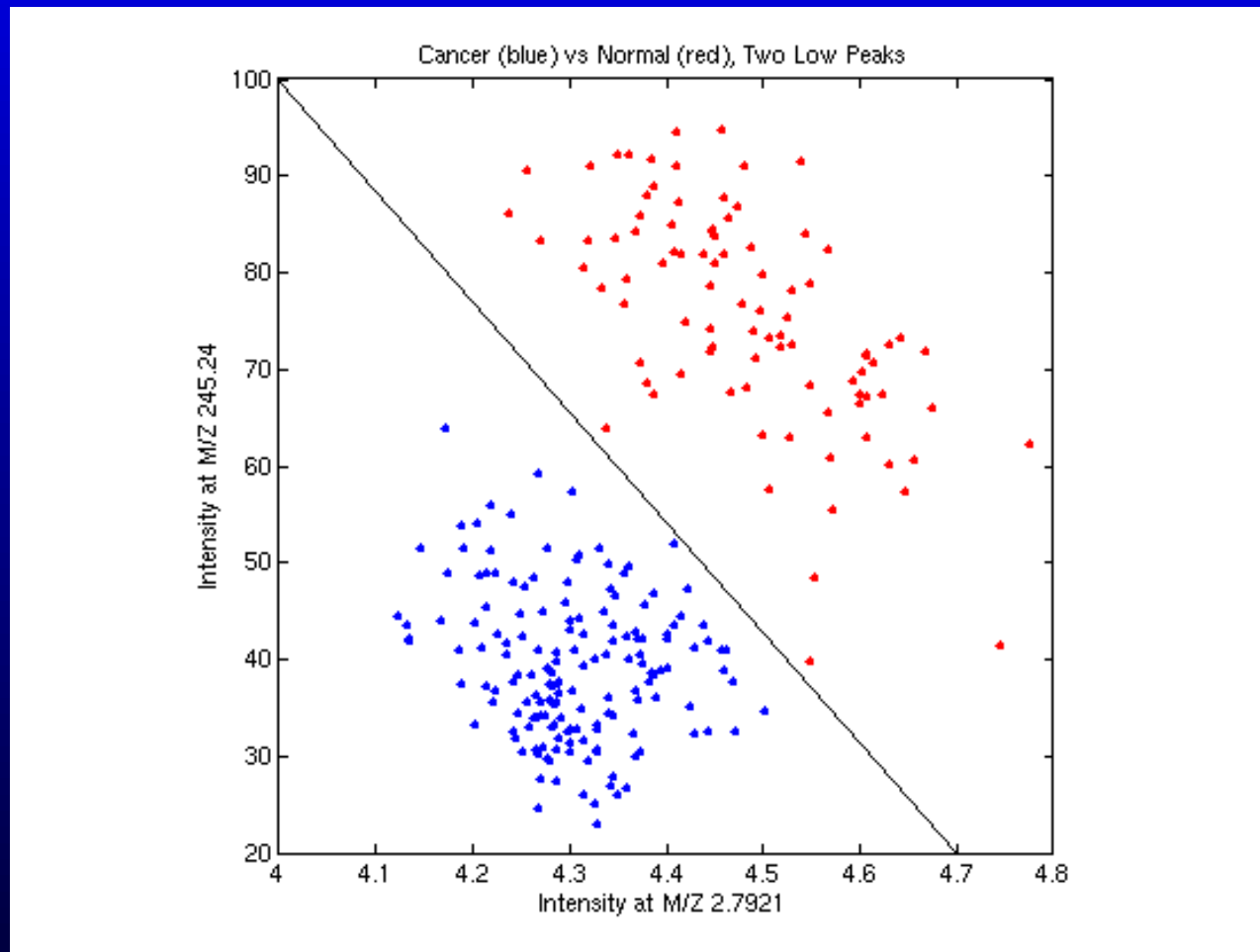
# One Bivariate Plot: $M/Z = (435.46, 465.57)$



Perfect Separation. These are the first 2 peaks in their list, and ones we checked against DS2.



## Another Bivariate Plot: $M/Z = (2.79, 245.2)$



Perfect Separation, using a completely different pair. Further, look at the masses: this is the noise region.

## Perfect Classification with Noise?

This is a problem, in that it suggests a qualitative difference in how the samples were processed, not just a difference in the biology.

This type of separation reminds us of what we saw with benign disease.

(Sorace and Zhan, BMC Bioinformatics, 2003)

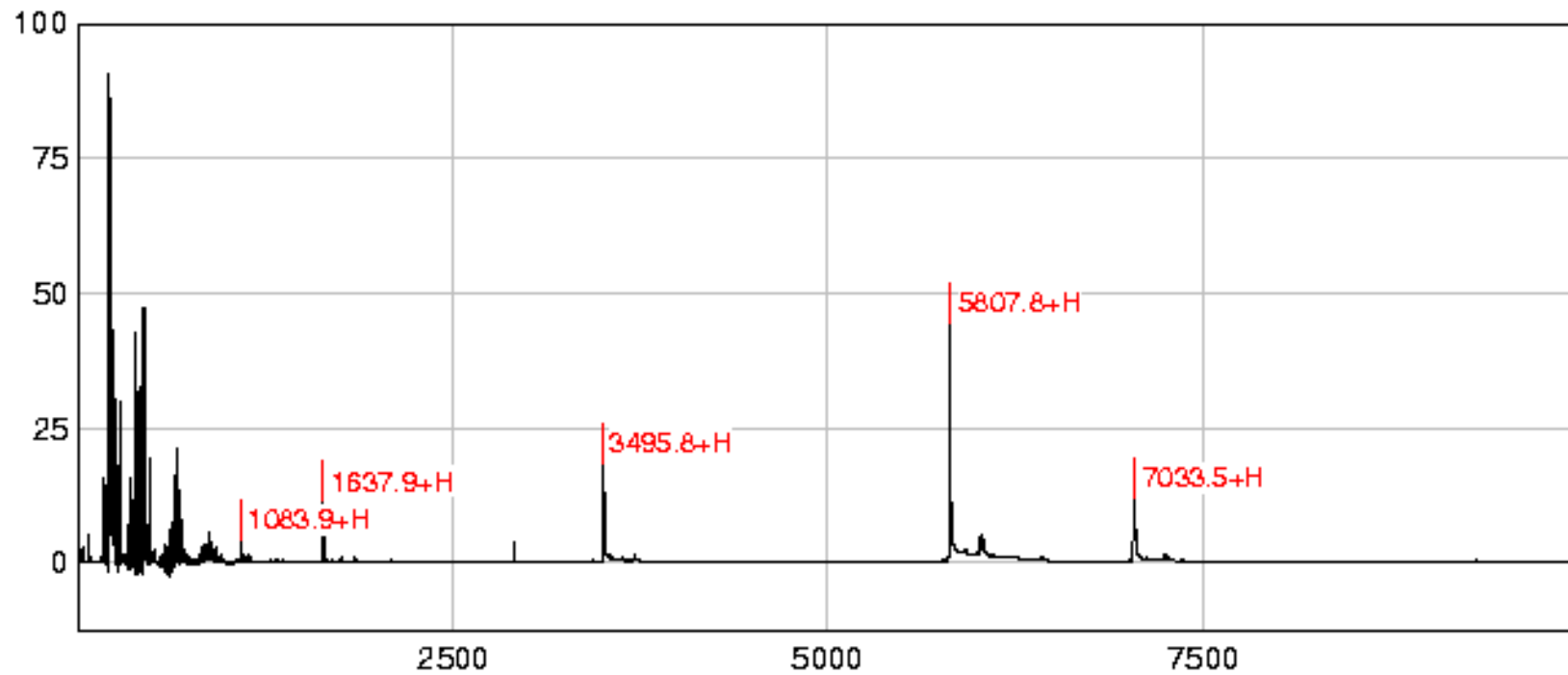
## Mass Accuracy is Poor?

A tale of 5 masses...

Feb '02 DS1	Apr '02 DS2	Jun '02 DS3
-7.86E-05	-7.86E-05	-7.86E-05
2.18E-07	2.18E-07	2.18E-07
9.60E-05	9.60E-05	9.60E-05
0.000366014	0.000366014	0.000366014
0.000810195	0.000810195	0.000810195

# How are masses determined?

Calibrating known proteins



## Calibration is the Same?

M/Z vectors the same for all three data sets.

Machine calibration the same for 4+ months?



# What is the Calibration Equation?

The CIPHERGEN equation

$$\frac{m/z}{U} = a(t - t_0)^2 + b, \quad U = 20K, t = (0, 1, \dots) * 0.004$$

Fitting it here

$$a = 0.2721697 * 10^{-3}, \quad b = 0, \quad t_0 = 0.0038$$

## What is the Calibration Equation?

The CIPHERGEN equation

$$\frac{m/z}{U} = a(t - t_0)^2 + b, \quad U = 20K, t = (0, 1, \dots) * 0.004$$

Fitting it here

$$a = 0.2721697 * 10^{-3}, \quad b = 0, \quad t_0 = 0.0038$$

These are the default settings that ship with the software!

## Other issues

Prostate Cancer

Q-star data different

clinical trials?

## A MALDI Example: Proteomics Data Mining

41 samples, 24 with lung cancer\*, 17 controls.

20 fractions per sample.

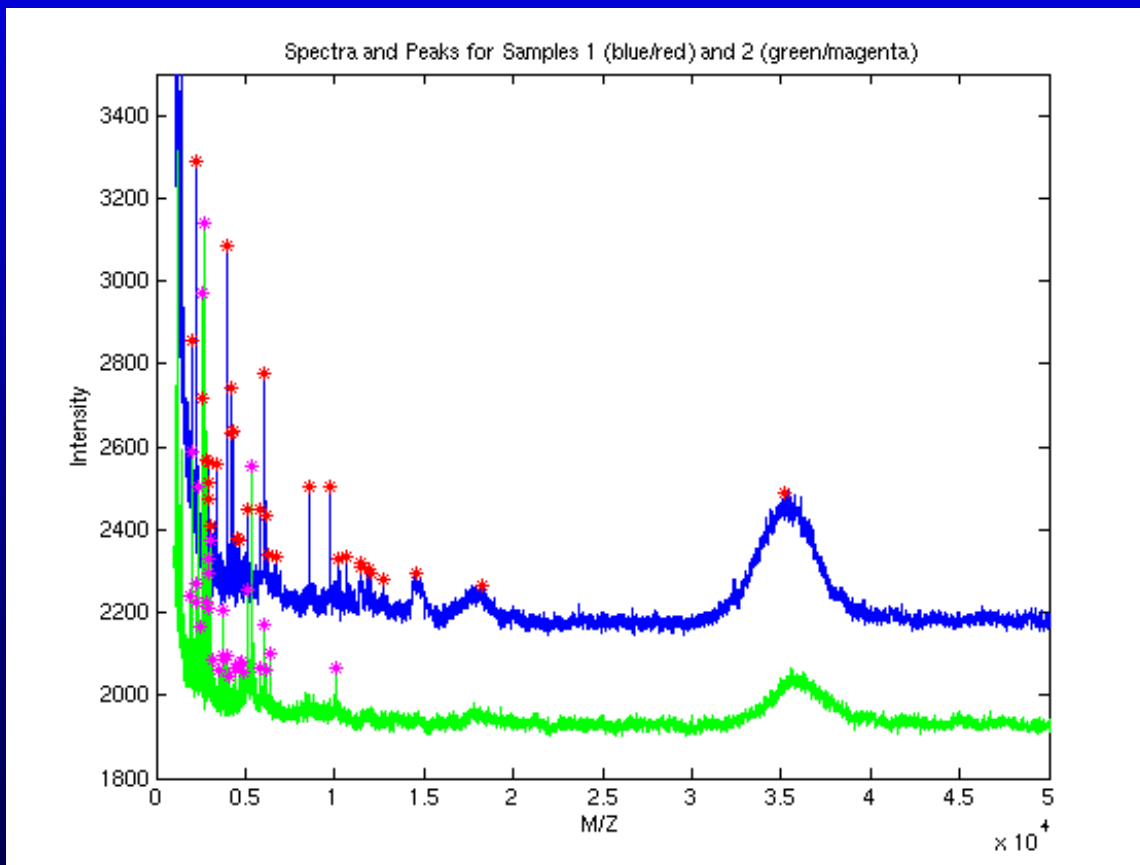
Goal: distinguish the two groups; we know this can be done due to the “zip effect”.

Data used to be at

<http://www.radweb.mc.duke.edu/cme/proteomics/explain.htm>

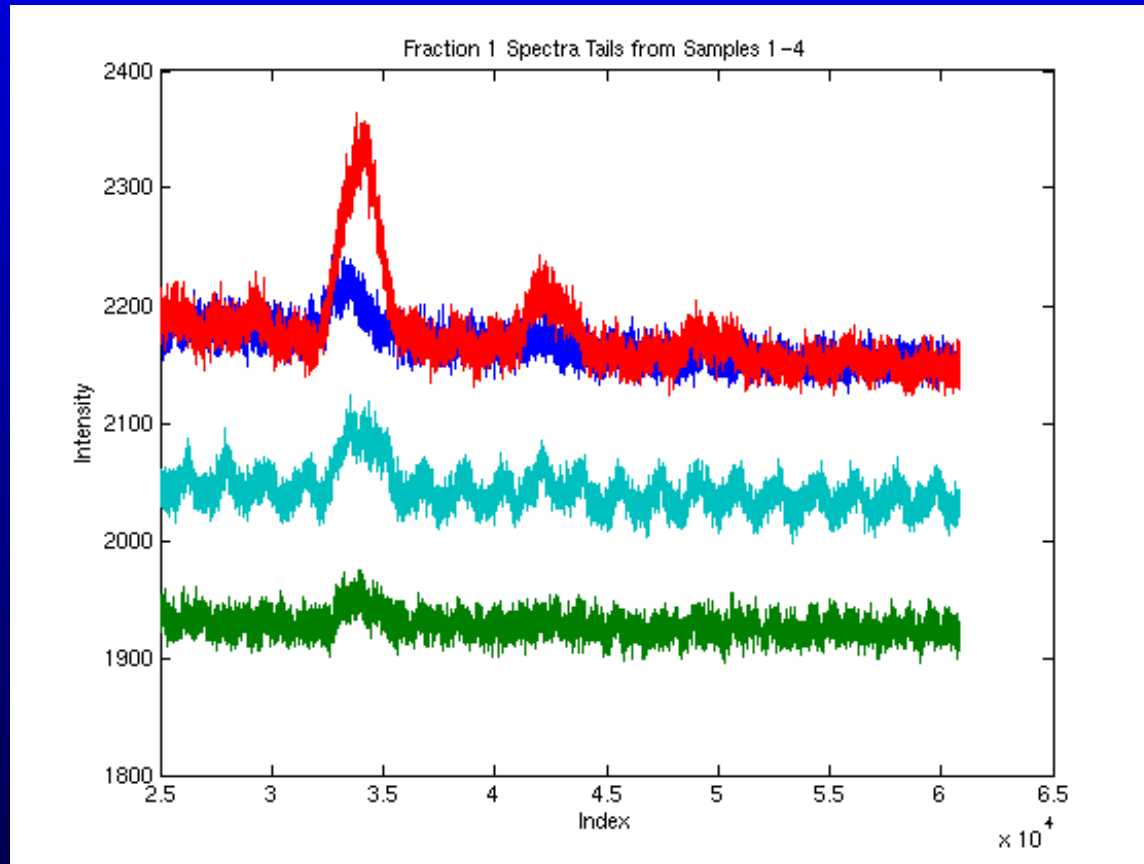
but the site has been retired. Send email to Ned Patz or Mike Campa at Duke if interested (Campa002@mc.duke.edu, patz0002@mc.duke.edu).

# Raw Spectra Have Different Baselines



Note the need for baseline correction and normalization.

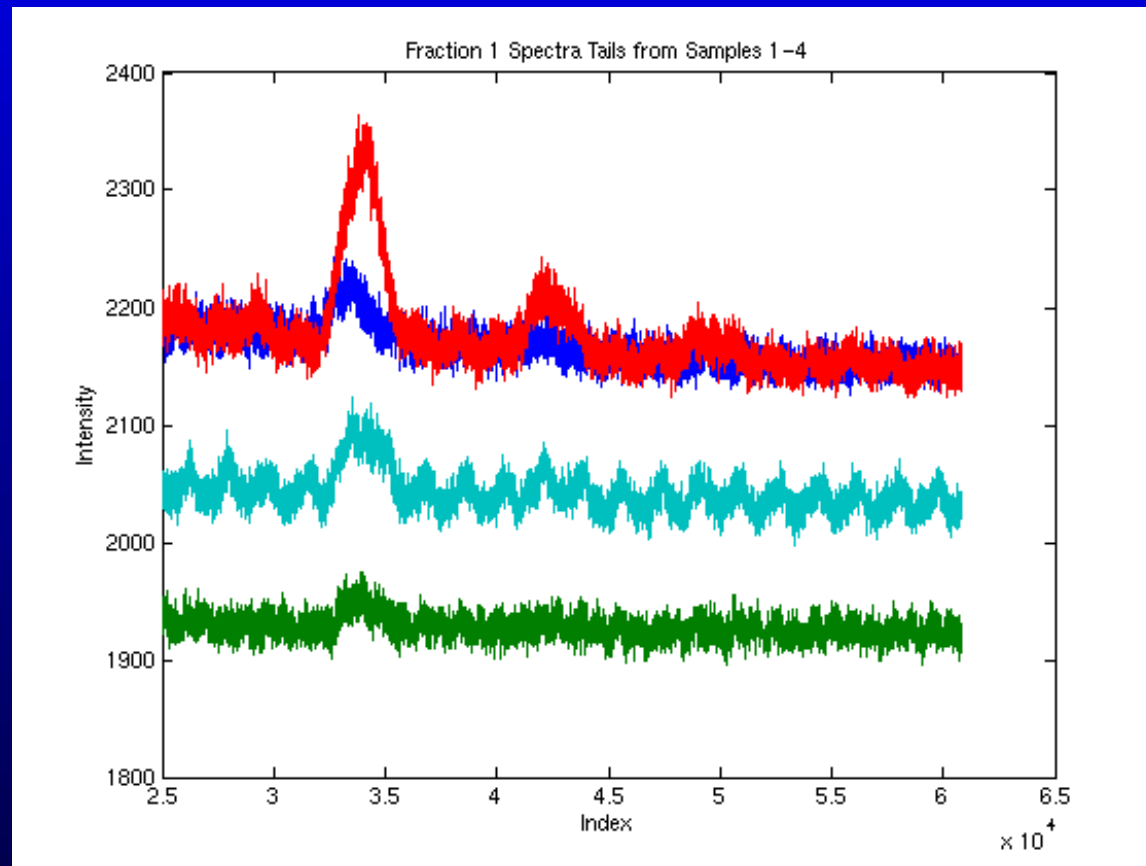
# Oscillatory Behavior..



Roughly half the spectra have sinusoidal noise.

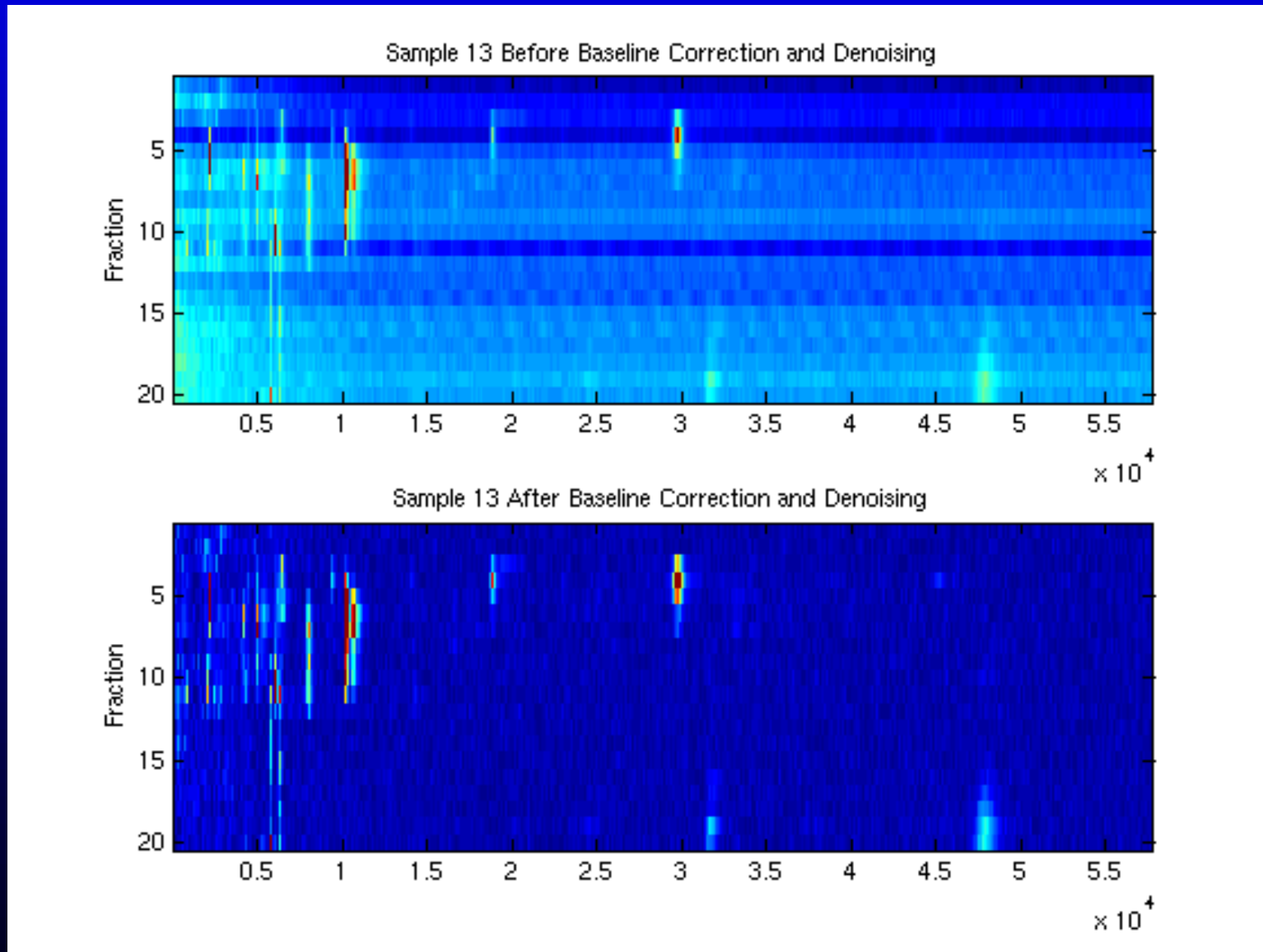


# Oscillatory Behavior..

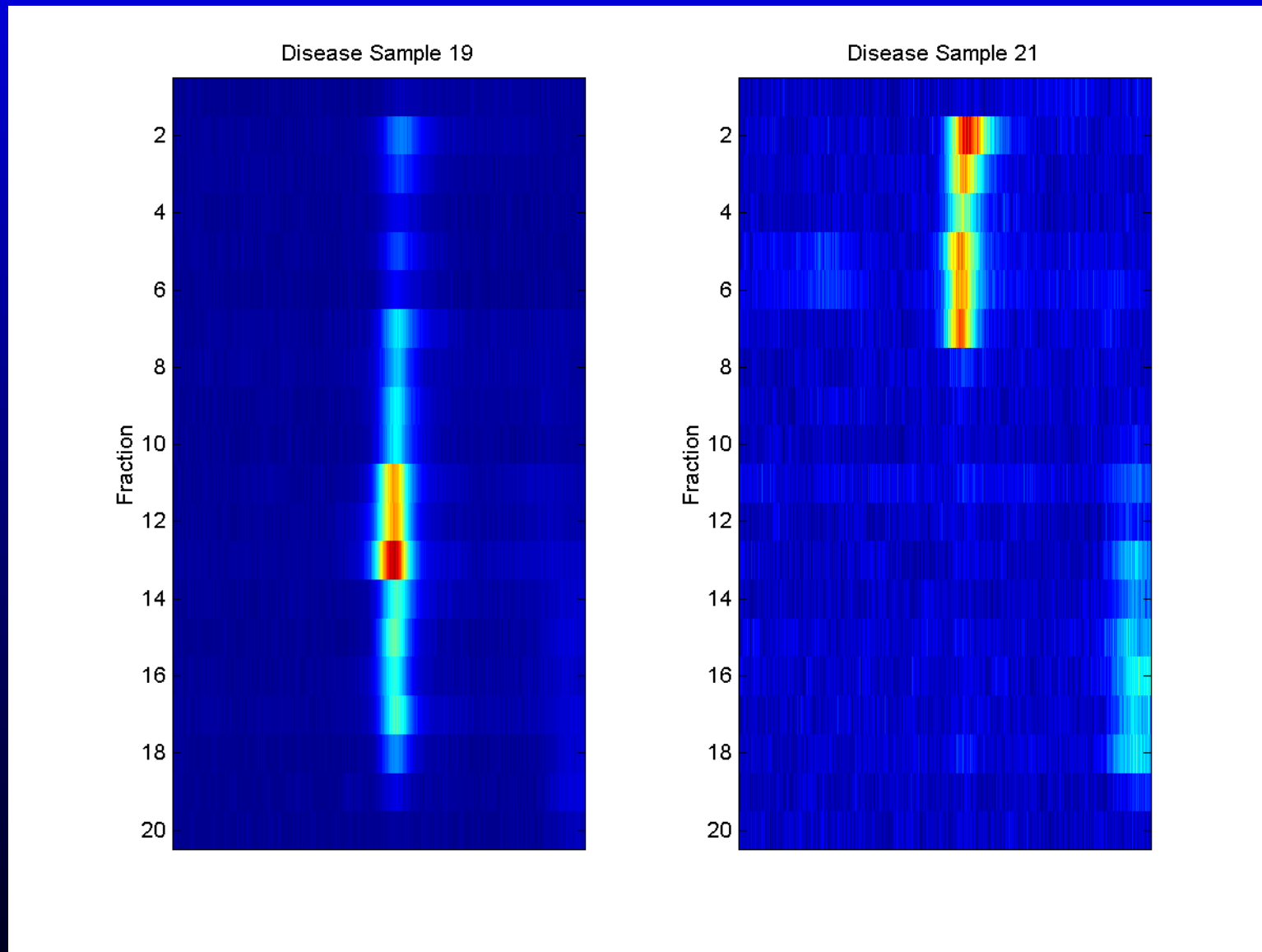


Roughly half the spectra have sinusoidal noise. We're seeing the A/C power cord.

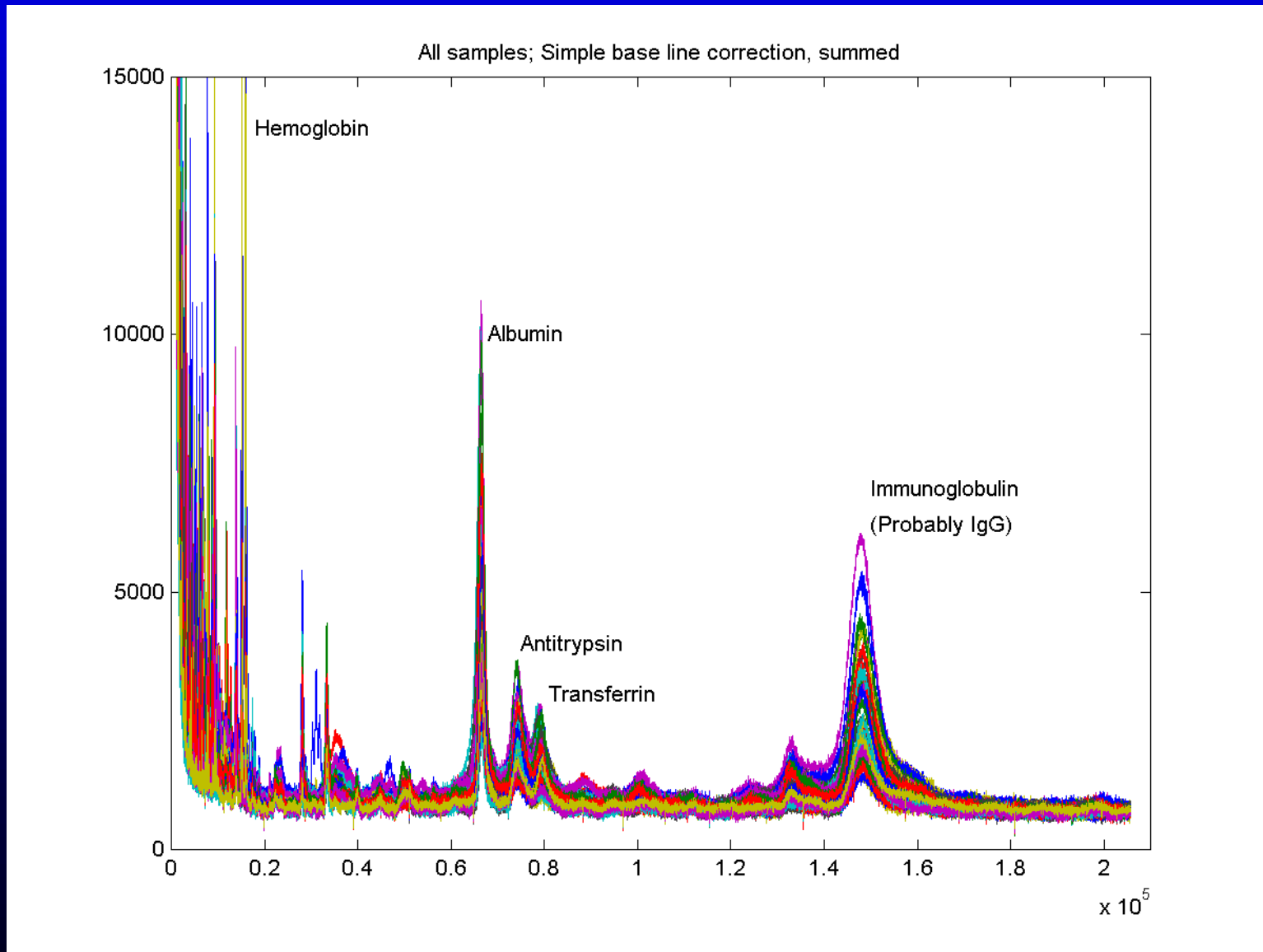
# Baseline Adj: Fraction Agreement, Before & After



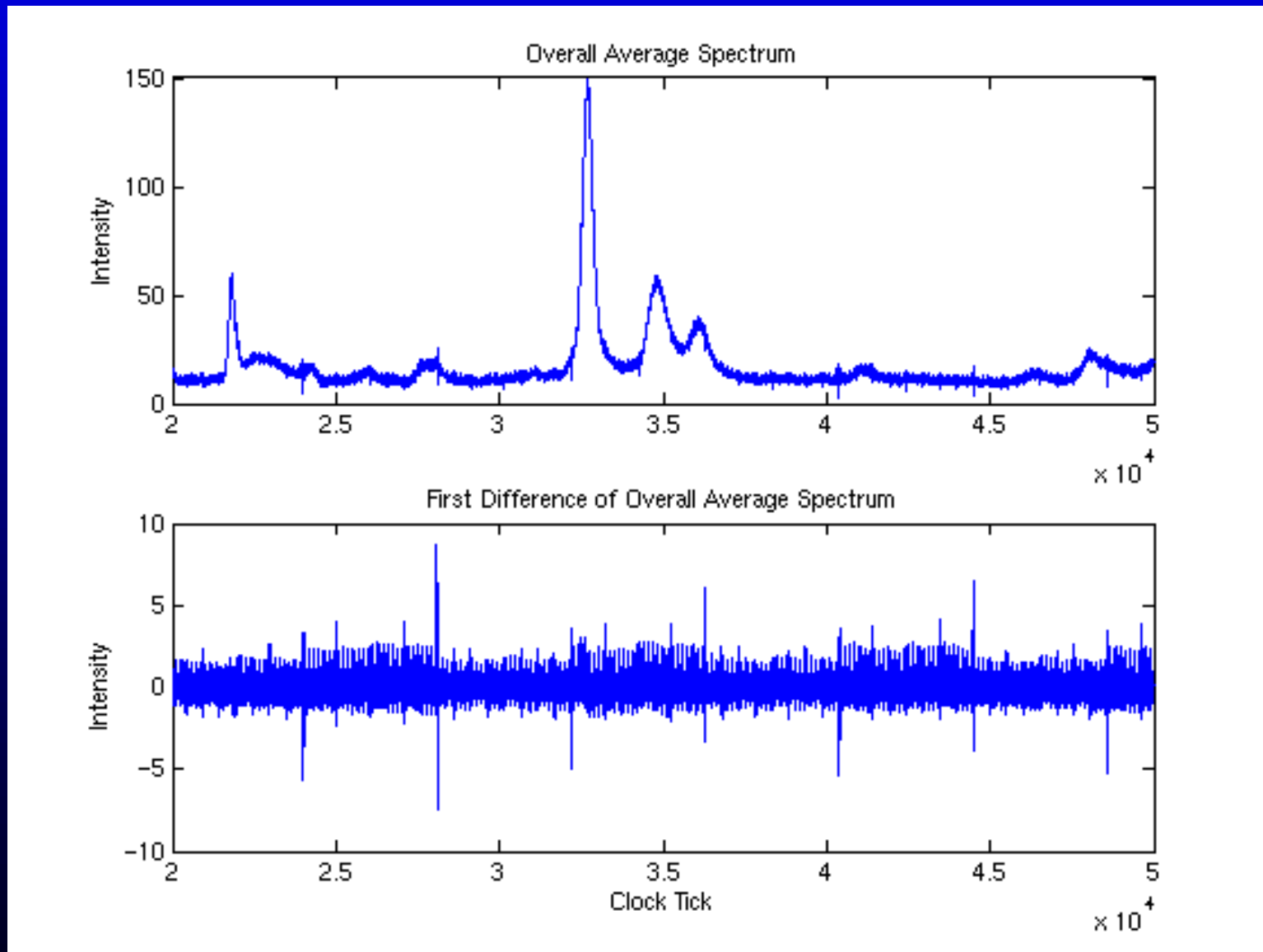
# Fractionation is Unstable



# Unfractionating the Data

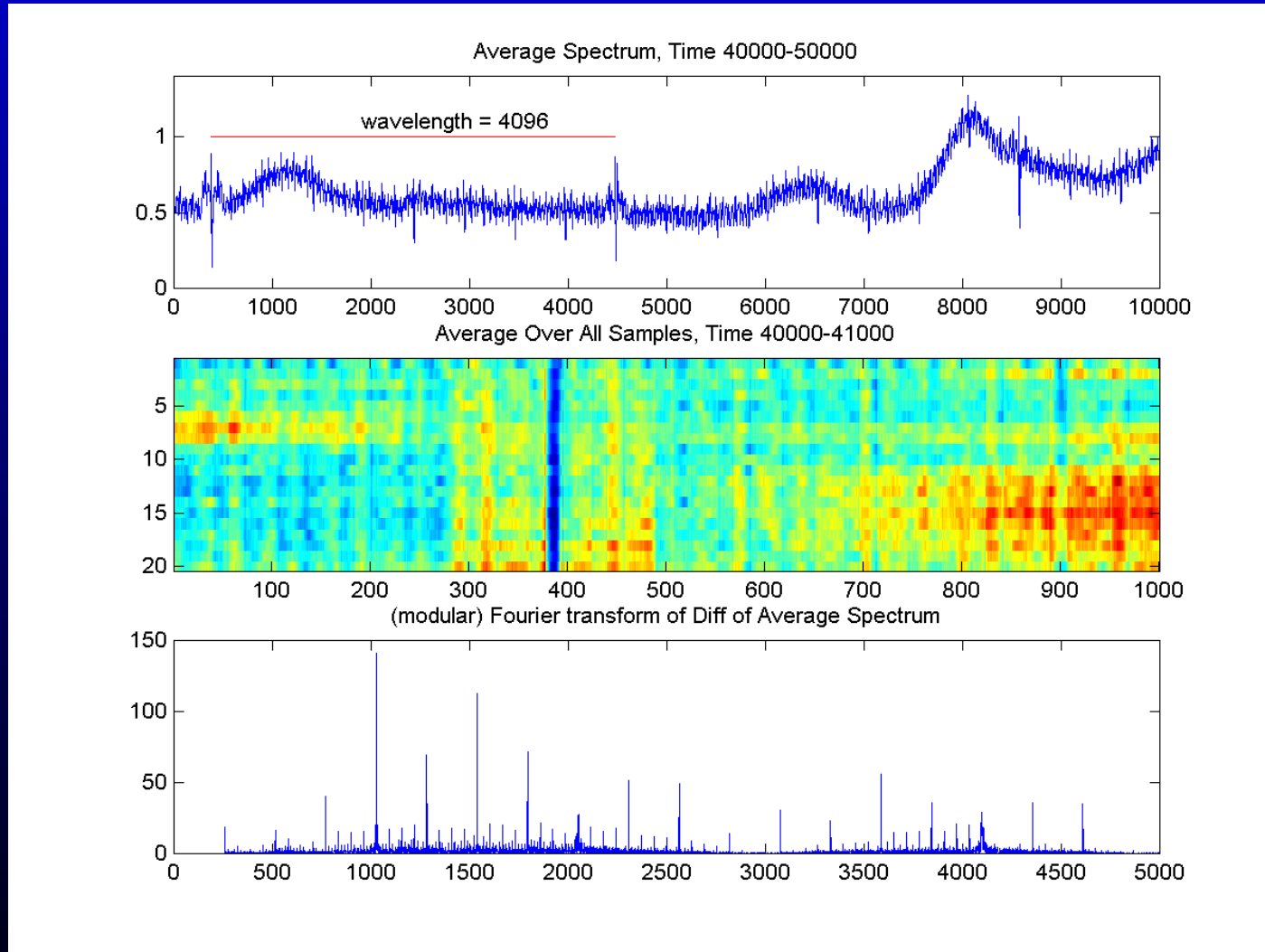


# The Overall Average Shows Spikes. Difference It.



# Computer Buffer?

Spike spacing has a wavelength of  $4096 = 2^{12}$ .

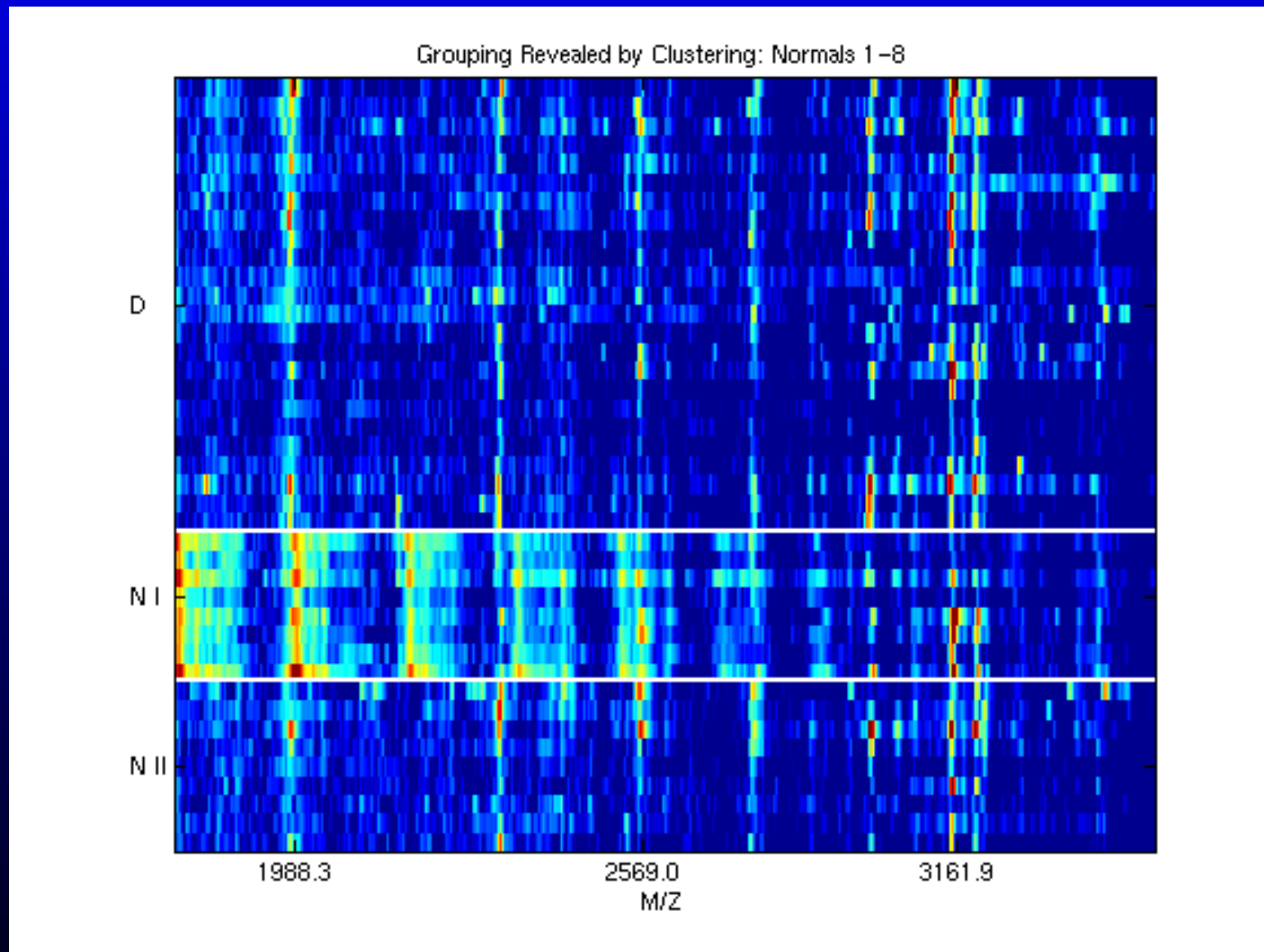




## Are We Done Cleaning Yet?

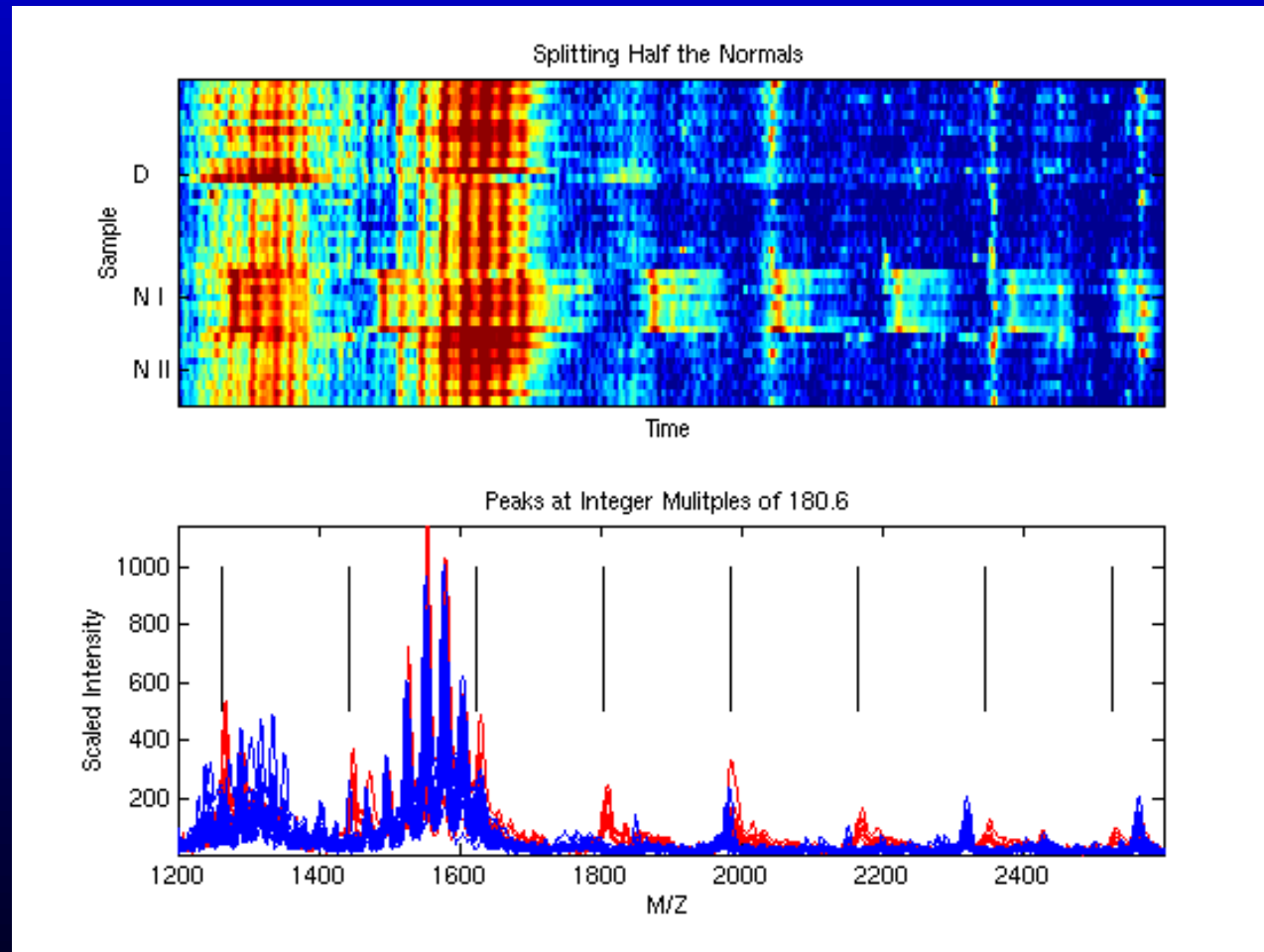
Give the problem a chance to be easy, try some simple clustering.

# PCA Splits off Half the Normals



# Peaks at Integer Multiples of M/Z 180.6!

This suggests a polymer. No Amino Acid dimers fit.



# Cleaning Redux

- Baseline Correction and Normalization
- Inconsistent Fractionation
- Computer Buffers
- Polymers in some Normal Spectra
- Peak Finding (Use Theirs)

Data reduced to 1 spectrum/patient, with 506 peaks per spectrum.

## Find the Best Separators

Peaks	MD	P-Value	Wrong	LOOCV
12886	2.547	$\leq 0.005$	11	11
8840, 12886	5.679	$\leq 0.01$	5	6
3077, 12886 74263	9.019	$\leq 0.01$	3	4
5863, 8143 8840, 12886	12.585	$\leq 0.01$	3	3
4125, 7000 9010, 12886 74263	23.108	$\leq 0.01$	1	1

There are 9 values that recur frequently, at masses of 3077, 4069, 5825, 6955, 8840, 12886, 17318, 61000, and 74263.

P-values are not from table lookups!

## Testing Reality (Significance)

Generate a bunch of “random noise” data matrices, each  $41 \times 506$  in size.

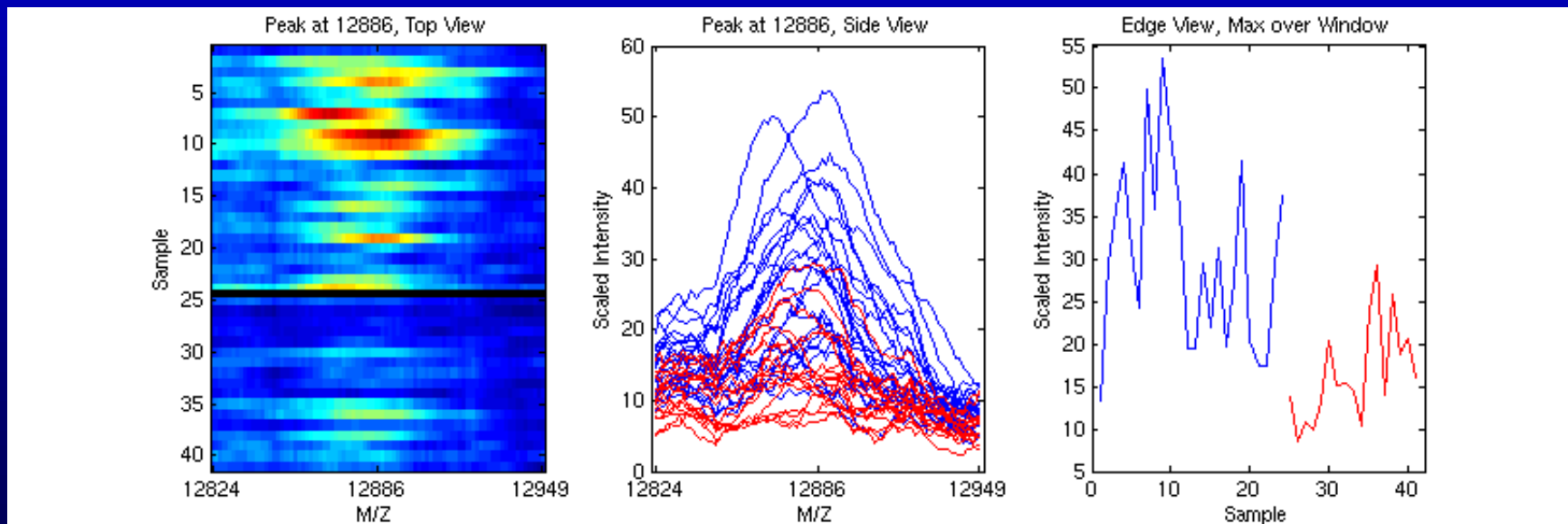
For each matrix, split the 41 noise “samples” into groups of 24 and 17.

Repeat our search procedure on the random data, and see how well we can separate things.



# The Eyeball Test

We applied one last filtering step and actually *looked* at the regions identified. All 9 peaks listed above passed the eye test.



Blue lines = Cancers

Red lines = Controls

## Other Stuff

We were the only ones to notice the sinusoidal noise.

## Other Stuff

We were the only ones to notice the sinusoidal noise.  
and the clock tick.

## Other Stuff

We were the only ones to notice the sinusoidal noise.

and the clock tick.

and they were looking at power cables and other stuff.

## Other Stuff

We were the only ones to notice the sinusoidal noise.

and the clock tick.

and they were looking at power cables and other stuff.

and they gave us a nice shiny plaque!

## The Deluge at MDA

Brain Cancer

Bladder Cancer

Leukemia

Pancreatic Cancer

Breast Cancer

Several show real structure, several show processing effects.

“If you’re not working on a proteomics project, you will be soon!”

Kevin Coombes to Bioinf section at MDA

# The Punchlines

- There is no magic bullet here. (Sorry.)
- Data preprocessing is extremely important with this type of data, and there is still much room for improvement.
- Use Simple Tests and Pictures
- Insist on Good Experimental Design
- There is structure in this data and it can be found!



## Our Own Reports

On the *Lancet* data:

Baggerly, Morris and Coombes (2004), *Bioinformatics*, **20(5)**:777-785.

On the Proteomics Data Mining Conference Data:

Baggerly, Morris, Wang, Gold, Xiao and Coombes (2003), *Proteomics*, **3(9)**:1677-1682.

More methodology:

Coombes et al (2003), *Clinical Chemistry*, **49(10)**:1615-1623.

pdf preprints are available.

## Partners in Crime

**Kevin Coombes, Jeff Morris**

---

Jing Wang, David Gold, Lian-Chun Xiao

---

Ryuji Kobayashi, David Hawke, John Koomen

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