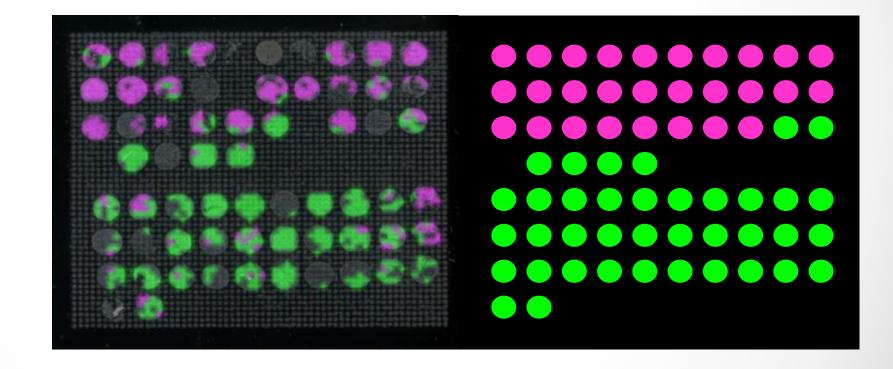
#### Practical Considerations for Clinical Mass Spectrometry Imaging Studies

Erin H Seeley, PhD



## The Three Keys to Success

## Reproducibility

# Reproducibility

# Reproducibility

## **Clinical MS Studies**

- May take weeks to years to complete
- May involve tens to thousands of samples
- Generally detecting intensity differences between disease states, rarely all-ornothing
- Reproducibility is of utmost importance to obtain accurate results
- Automation and robotics help to ensure consistent results

### How Many Samples Do I Need?

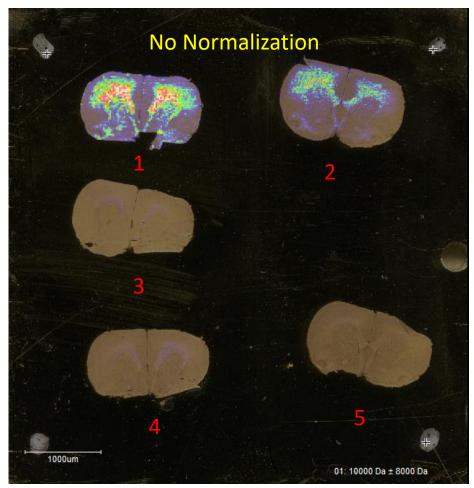
#### • Multiple sets needed

- Training creation of a classification algorithm
- Validation evaluation of classification algorithm accuracy
- Testing evaluation of algorithm performance in a clinical setting
- Power calculations have shown that typically a minimum of 25 human samples per cohort are needed for statistical significance
- More than that are preferred for a good representation of the expected patient population
- Are there multiple phenotypes/subtypes I need to account for?
  - Distribution of phenotypes/subtypes should be balanced
- What is my end goal?
  - Publication may need 50-100 samples
  - Clinical Assay may need 300-1000+ samples

### Do I Need to Care about Ambient Conditions?

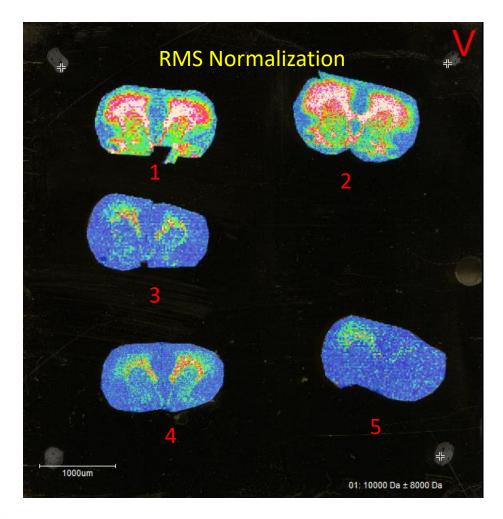
- YES!!!
- Changes in temperature and humidity may affect digestion efficiency or evaporation time
- How to standardize
  - Temperature and humidity control for the lab
  - Well regulated and calibrated ovens
  - Principle of deliquescence for digestion\*
  - Heated tray in HTX Sprayer for constant evaporation
  - $\circ$  etc.

## What About Instrument Performance?

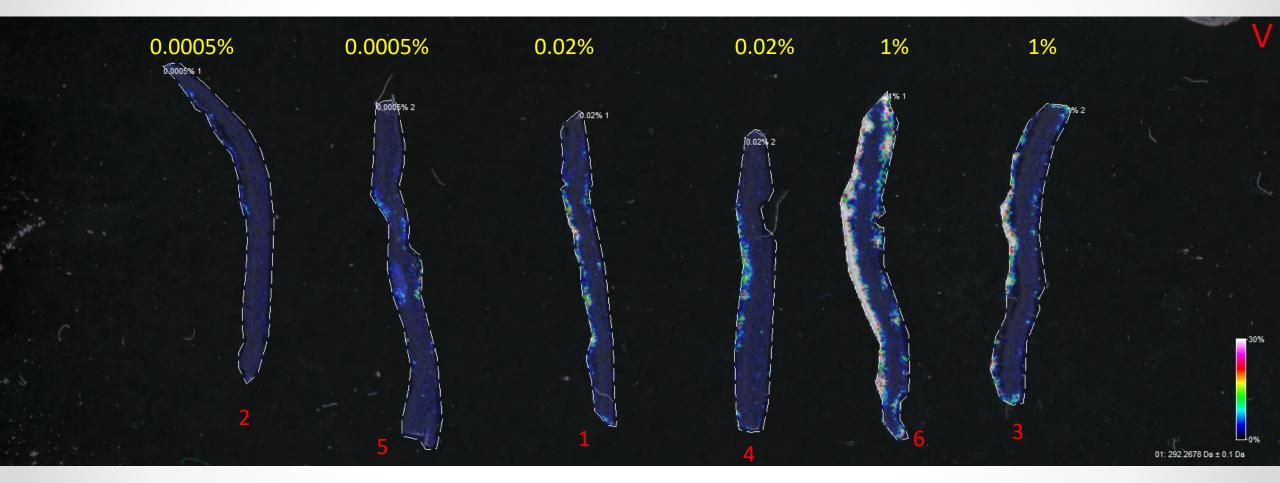


~22,000 pixels, 100 shots per pixel = 2.20 million shots

### But I Can Fix That With Normalization, Right?

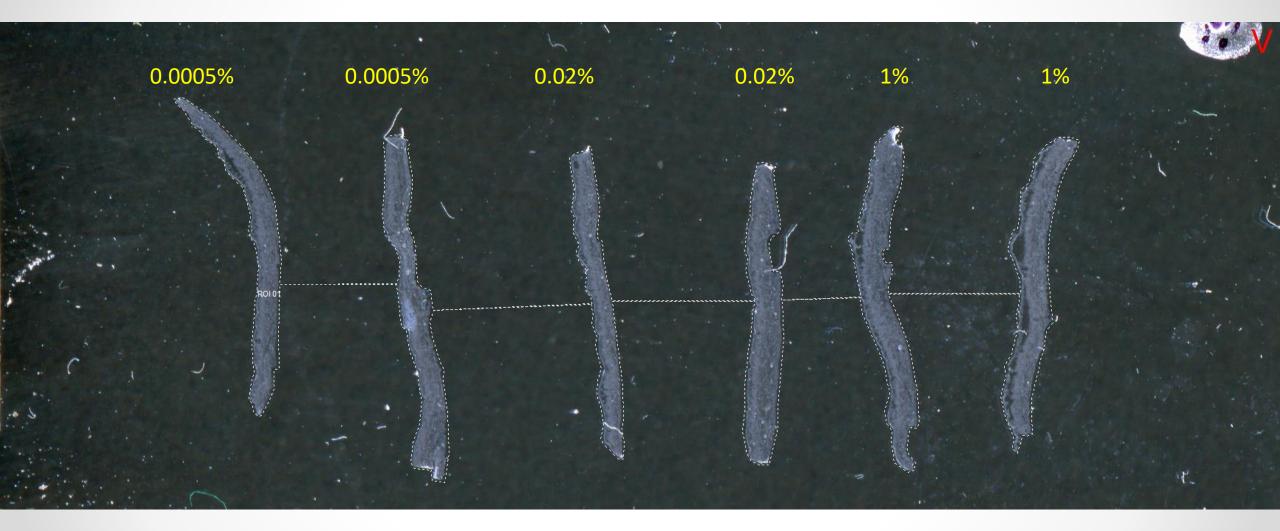


#### **Randomize Collection Order**



Skin treated with varying concentrations of an antifungal compound

### **Collect All Samples as One Region**



## **Minimizing Sampling Bias**

- Verify instrument performance each day with a known standard
  - Particularly important after instrument maintenance is performed
  - Different service engineers may tune differently
- Include a "standard sample" with each analysis
- Ensure that each batch of samples analyzed contains a mix of diagnostic states
- Mix data collection so that types to be compared are not collected sequentially
- Perform regular source cleaning/maintenance so that signal is not degraded over time
- Minimize data volume
  - Lower spatial resolution
  - Limited regions for data collection
  - Histology-guided profiling
  - Use of Tissue MicroArrays (TMAs)
- Carry out data analysis in a single batch

### **Histology-Guided** Mass Spectrometry Profiling

- A targeted approach in which only discrete areas within a tissue section are analyzed
- Histological staining is used to guide the acquisition of spectra
- Each analyzed spot is enriched for a single cell type
  Ensure high quality/purity of annotations
- Conducive to statistical analysis and classification algorithm generation
- Can provide biological insight not attainable by standard histology (disease outcome, improved diagnostics, treatment response, etc.)

## **Melanocytic Skin Lesions**

 $\bullet$   $\bullet$   $\bullet$ 

Histology-Guided Mass Spectrometry Profiling – FFPE Tissue In Collaboration with Dr. Rossitza Lazova, California Skin Institute

### Melanoma Diagnosis Discordance

- 4+ M biopsies performed per year in the US to rule out melanoma.<sup>+</sup> Of these, over 10% cannot be definitively classified using routine histopathology.<sup>‡</sup>
- There can be considerable disagreement among dermatopathologists in the diagnosis of melanocytic lesions
- In a 1996 study\*, 8 pathologists reviewed the same 37 slides
  - 13 cases had complete agreement
  - 10 cases had one discordance
  - 14 cases had 2 or more discordances, 8 of these had 3 or more
- More objective diagnosis is needed to help with difficult cases

<sup>+</sup> Br J Dermatol, **2017**, 176, 949-954 <sup>‡</sup>J Am Acad Dermatol, **2012**, 67, 727-735 \*Human Pathol. **1996**, 27, 528-531

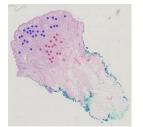
## **Study Parameters**

- Data collected over a period of 4 years
- 756 specimens, mixed between biopsies and tissue microarrays (TMAs)
- Age of specimens 5-40+ years
- Two different sprayer platforms [SunChrom SunCollect and HTX M5(2)]
- Two different mass spectrometers [Bruker ultrafleXtreme and rapifleX(2)]
- Three different personnel collected data

#### **Histology-Guided MS Profiling Workflow**



Tissue section a MALDI target



Stained serial section annotated



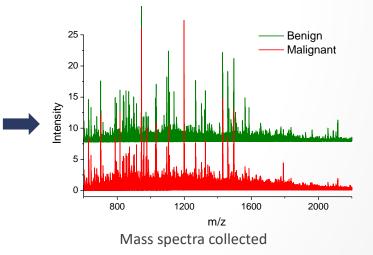
Many tissues in one experiment



Images of stained and unstained sections merged



Trypsin and matrix applied

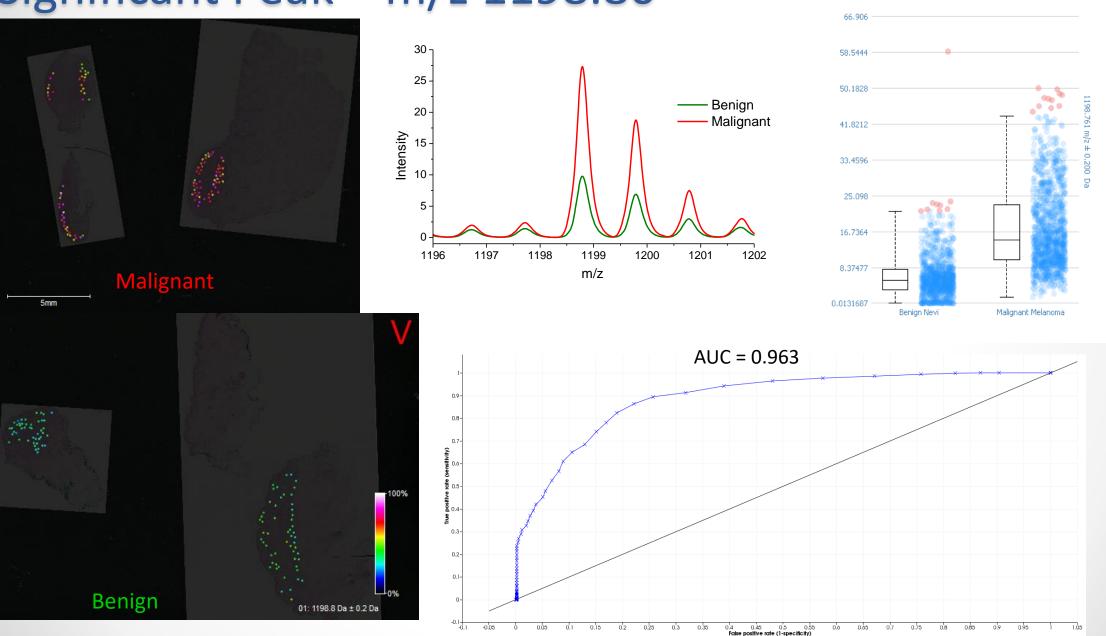


### **Statistical Analysis Methods**

- Spectra loaded into SCiLS Lab software and baseline corrected, normalized, and peaks picked
- Samples sorted into Training and Validation/Testing sets
- Hypothesis testing carried out on training set
- Linear discriminant analysis (LDA) classification algorithm generated and validated

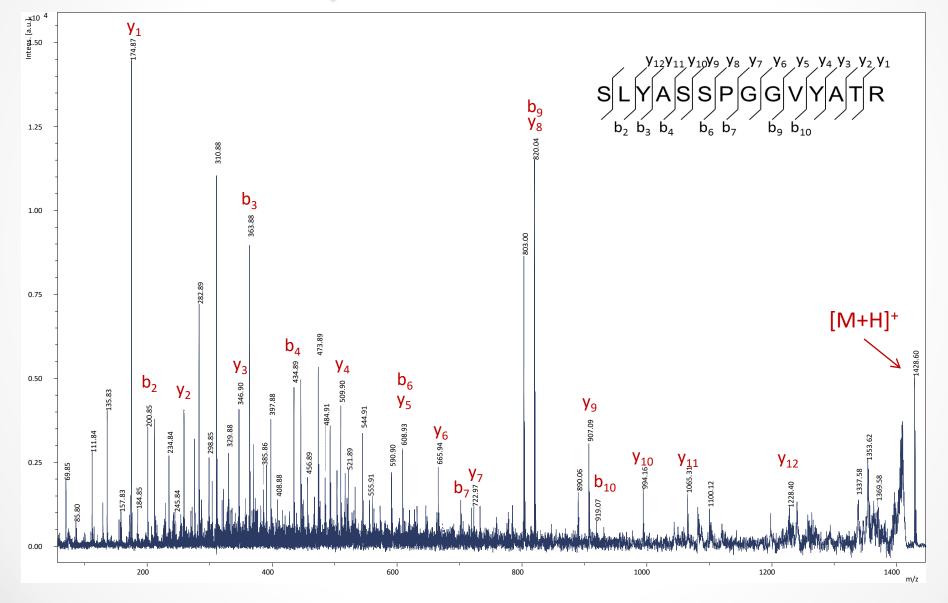
327/1075 peaks were significant after Bonferroni correction

#### Significant Peak – m/z 1198.80



75.2676

#### **Identification of a Peptide from Vimentin**



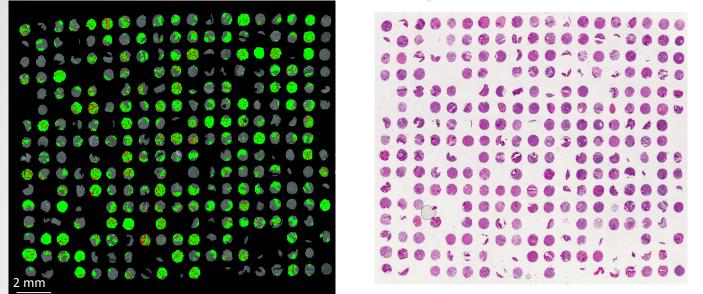
#### **Classification Algorithm Generation**

- Melanoma subtypes included: acral, desmoplastic, lentigo maligna, nevoid, nodular, Spitzoid, and superficial spreading. Nevi subtypes included: acral, conventional, and Spitz
- Linear discriminant analysis (LDA) classification algorithm built using leave-10%-out repeated random sub-sampling cross validation

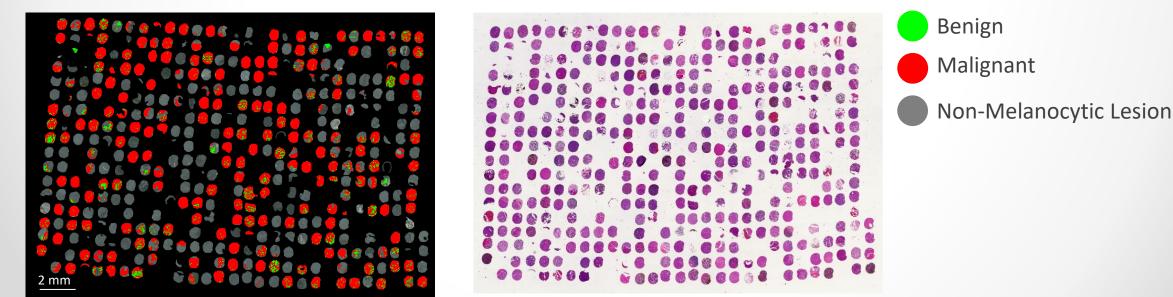
|                    | Number of Patients | Internal Cross Validation<br>Spectral Accuracy |
|--------------------|--------------------|--|
| Malignant Melanoma | 100                | 94.1%  |
| Benign Nevi        | 111                | 94.1%  |

- Optimized classifier applied to independent set of unambiguous malignant melanoma and benign nevi specimens traditional biopsies and TMAs
  - Malignant Melanoma 288 patients
  - Benign Nevi 257 patients
  - o 93% Sensitivity, 98% Specificity\*

#### **Analysis of Melanocytic Lesion TMAs**

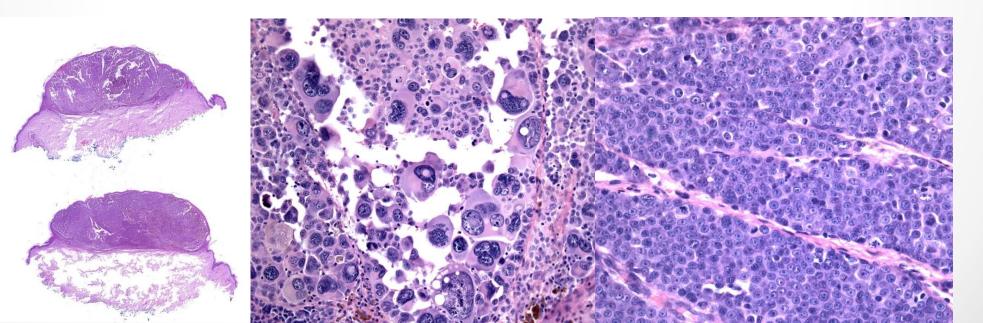


- 380 samples originated from 6 total TMAs (172 malignant, 208 benign)
- Only melanocytic component evaluated
- Core with no data inappropriate for testing



### **Case Study**

- 37 year old pregnant woman presents with lesion on upper arm
- Excisional biopsy performed and determined to be malignant
- Insufficient margins taken for size of lesion
- No further treatment during pregnancy



#### **Case Study**

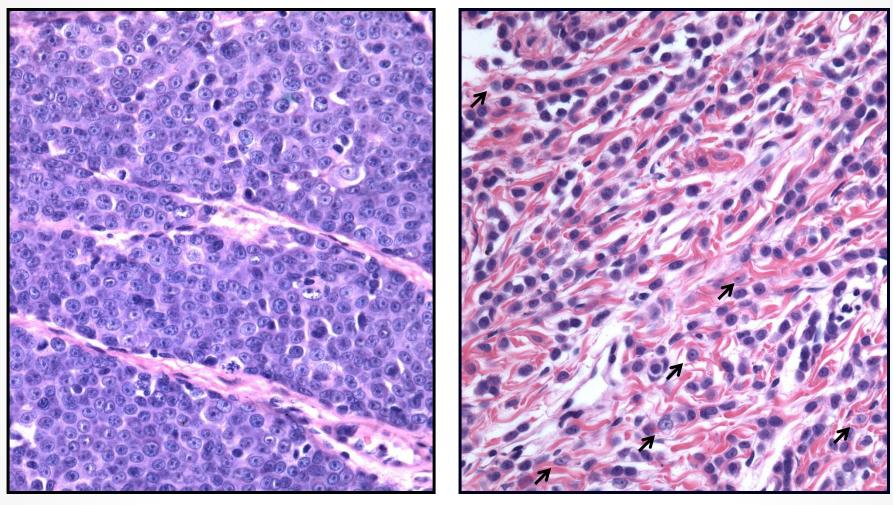
• Two months later, male baby born with melanocytic lesions



Alomari A, Glusac EJ, Choi J, Hui P, Seeley EH, Caprioli RM, Watsky KL, Urban J, Lazova R. J Cutan Pathol. 2015, 42, 757-64

#### Mother





Metastases or Congenital Nevi?

#### **Mass Spectrometry Analysis**

| Mother                 |  |  |
|------------------------|--|--|
| Histological Diagnosis | Mass Spectrometry<br>(number of spectra) |  |
| Malignant Melanoma     | Malignant Melanoma – 29/29               |  |

| Baby Skin Lesions        |  |  |
|--------------------------|--|--|
| Histological Diagnosis   | Mass Spectrometry<br>(number of spectra) |  |
| Biopsy A – Indeterminant | Benign Nevus – 23/23                     |  |
| Biopsy B – Indeterminant | Benign Nevus – 9/9                       |  |

Cells within lesions on baby contained y chromosome

#### Summary

- Mass spectrometry imaging is a powerful technology for clinical applications
- The use of machine learning allows for development of classification algorithms for diagnostic and prognostic assays
- Provides biological insight not available by other techniques
- Helps to provide solutions to clinically challenging problems
- Reproducibility is key to successful clinical studies

## Acknowledgements

- Rossitza Lazova
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