

The Power of AEM and Data Analytics: Examples from Materials Science (with a little geology too)

Paul G. Kotula

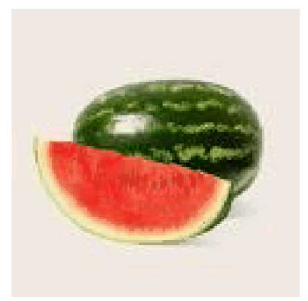
Sandia National Laboratories, Albuquerque, NM, USA

Thanks to Michael Keenan, Mark Van Benthem, and Hongkyu Yoon

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Why are we called Sandia?



?

- Largest National Laboratory in the US with over 13,000 employees
- Originally a division of Los Alamos Scientific Laboratory
- Named after the Sandia Mountains near Albuquerque
- Those mountains were named by the early Spanish settlers in ~1540
 - From a distance the Sandia's look like a watermelon on its side
 - The Spanish settlers mistook gourds being grown by the natives for watermelons



Sandia Mountains, Albuquerque, NM, USA, Courtesy USDA-NRCS

Materials Characterization Tetrahedron

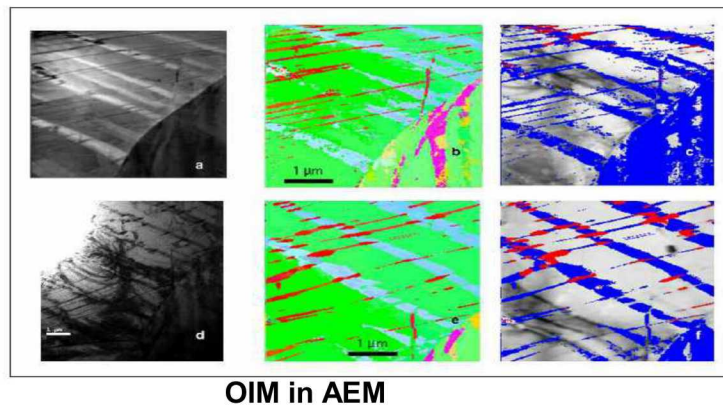
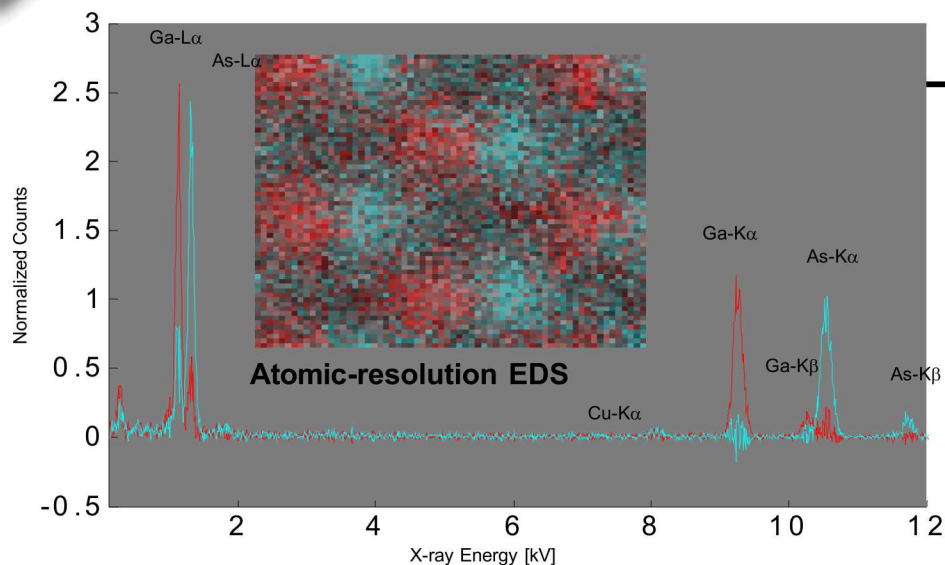
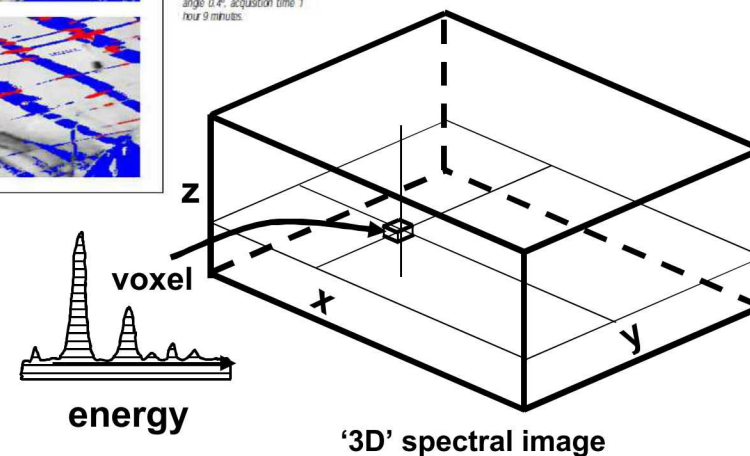
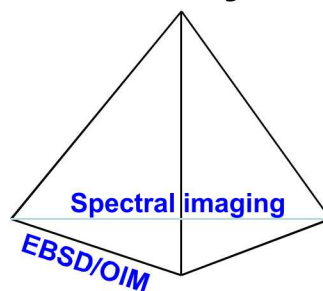


Figure 7: Austenitic stainless TRIP steel containing three distinct phases: γ -Fe matrix, high ϵ -bands and low ϵ -martensite at band intersections. (a) Index map coupled to (b) orientation (b) and phase (d) maps acquired without precession. (c) Brightfield image coupled to orientation (e) and phase (f) maps acquired with precession. Stop size 22 nm, precession angle 0.4°, acquisition time 1 hour 9 minutes.

Data Analysis

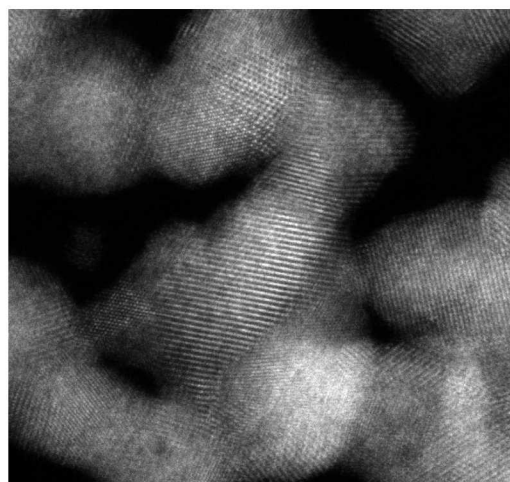
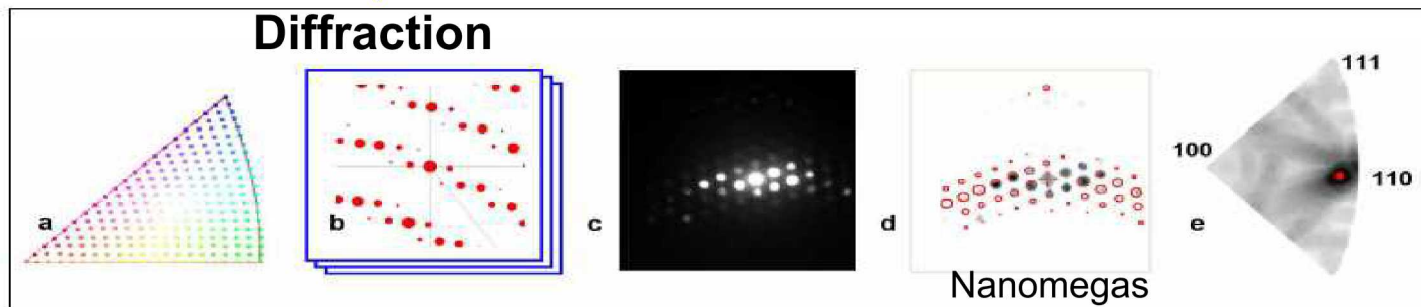


Imaging



Microanalysis

Diffraction



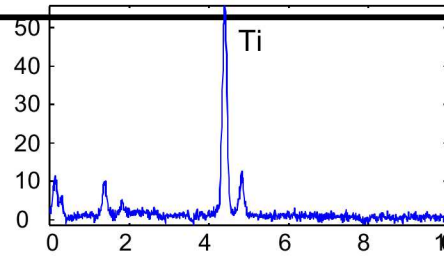
Atomic-resolution STEM image

AEM Spectral Imaging

Focused Electron Probe

Thin foil specimen

Counts



**kV
EDXS**

Data Analysis

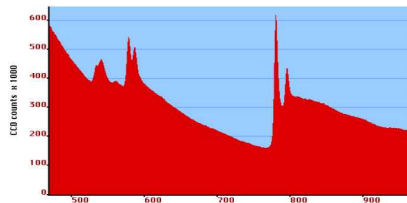
Imaging

Spectral imaging

Microanalysis

Diffraction

Counts

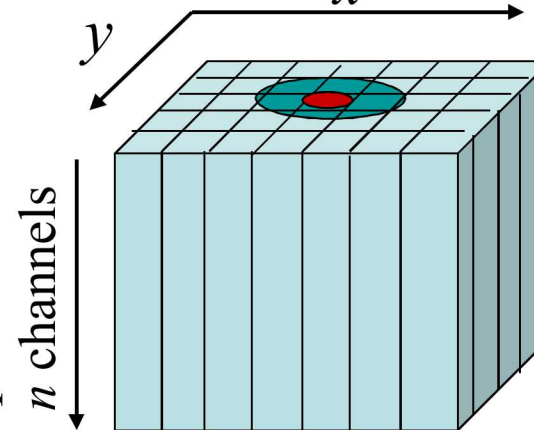


Energy loss

EELS

spatial coordinate

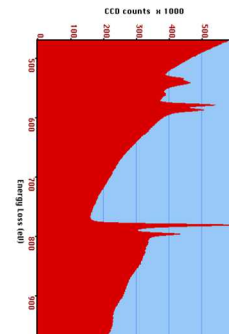
**spectra with
 n channels**



**Spectral image
data cube**

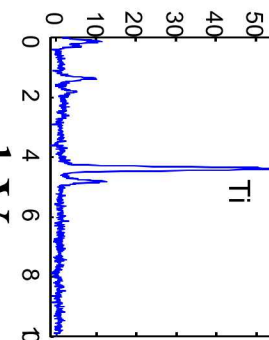
Energy loss

Counts

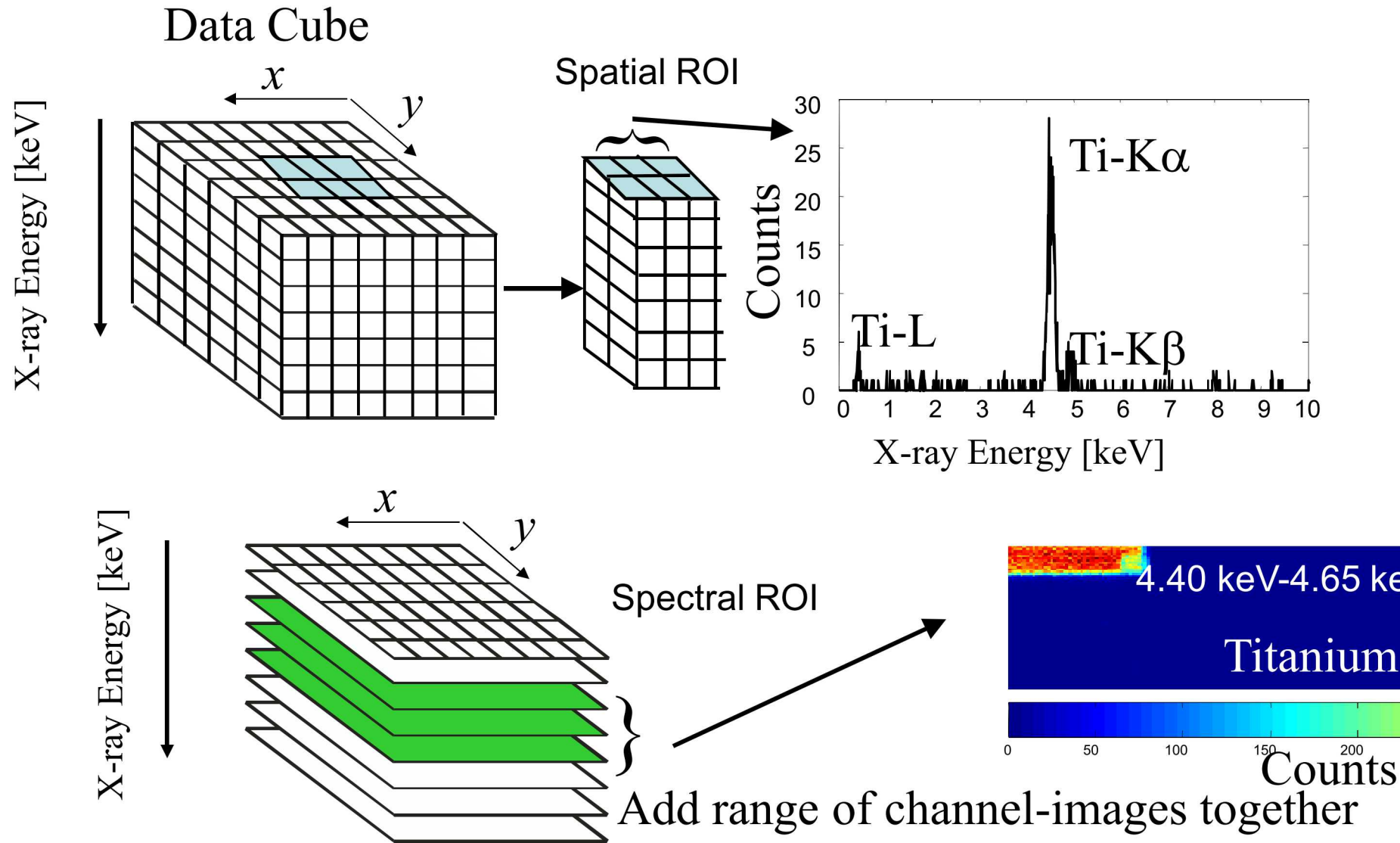


kV

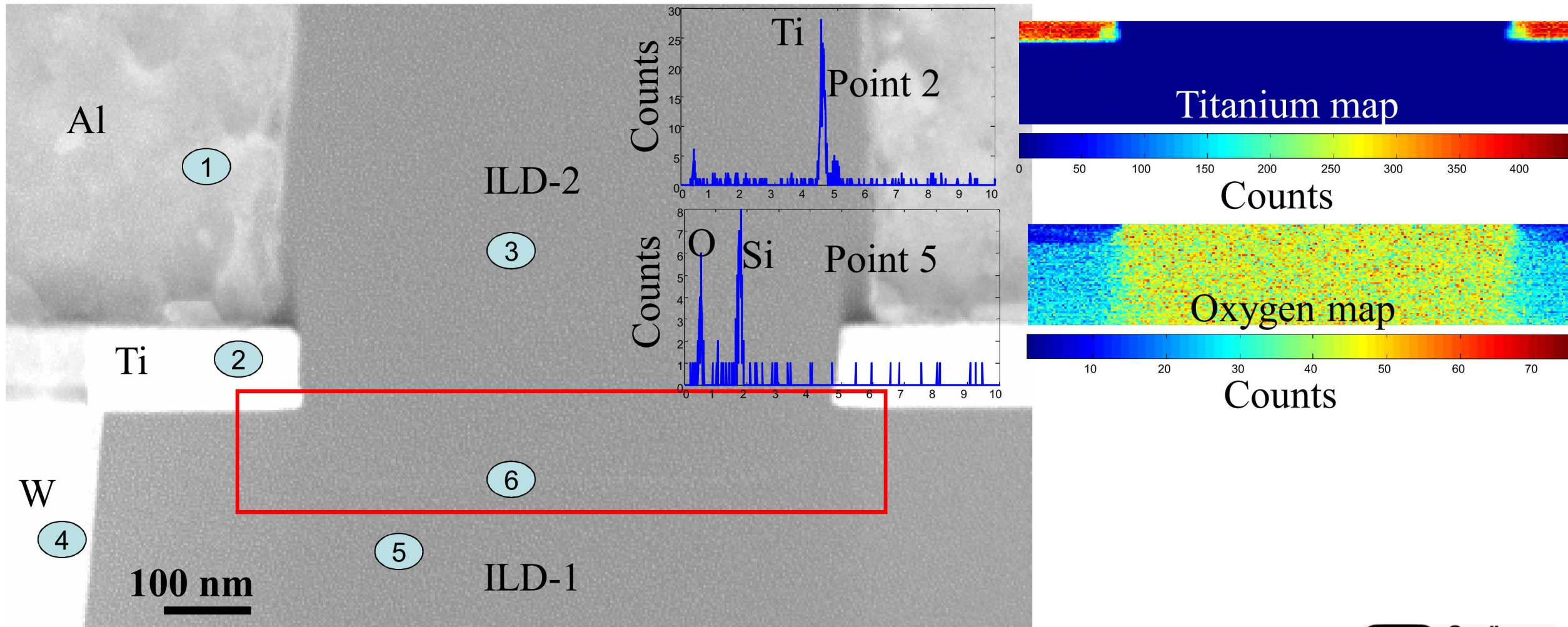
Counts



Conventional Data Analysis

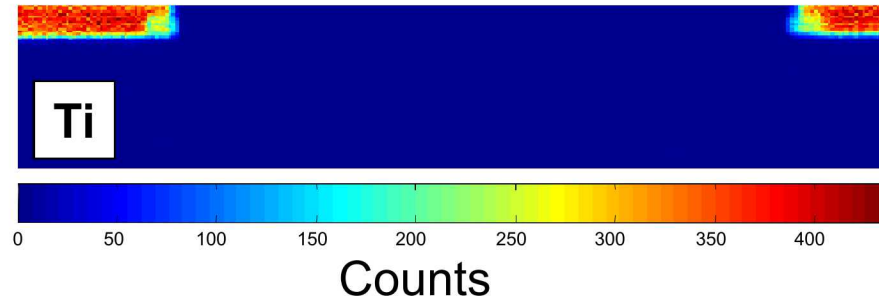
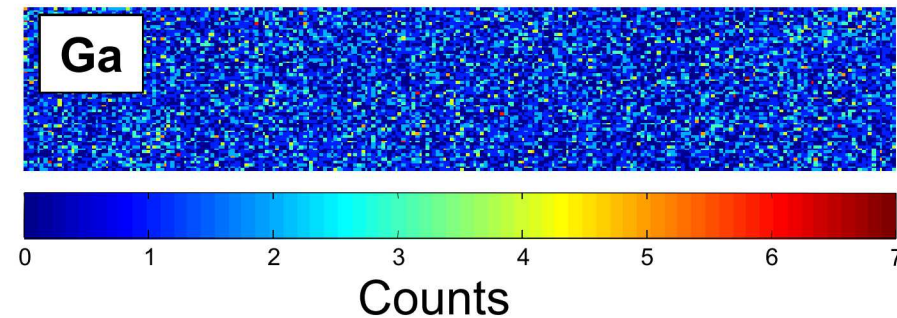
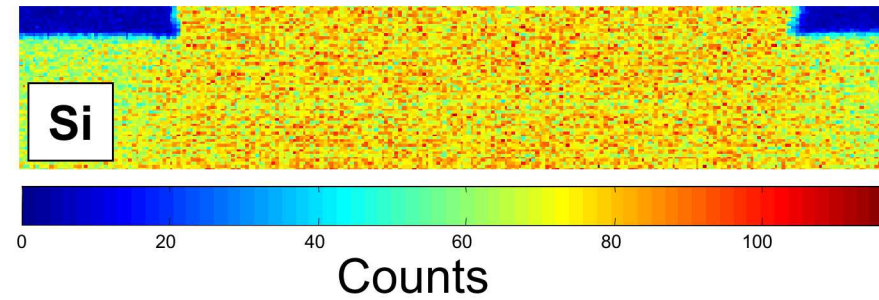
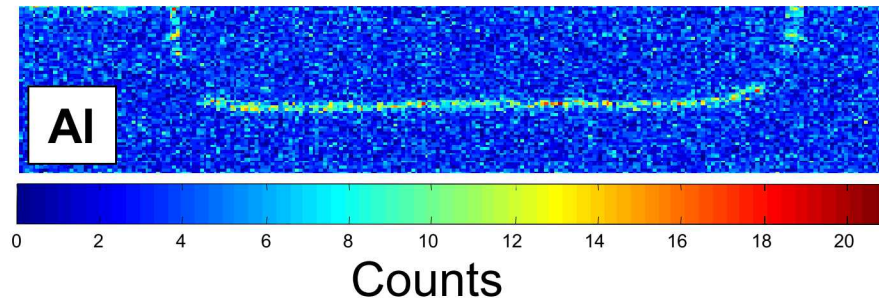
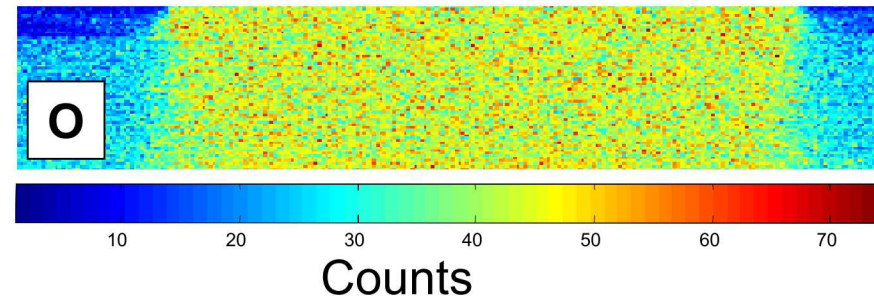
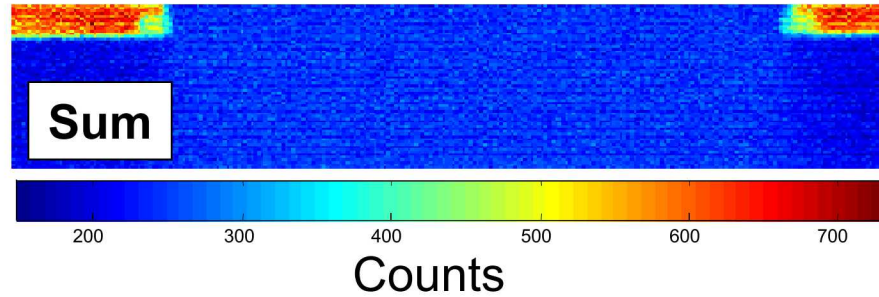


Example of Conventional Analysis



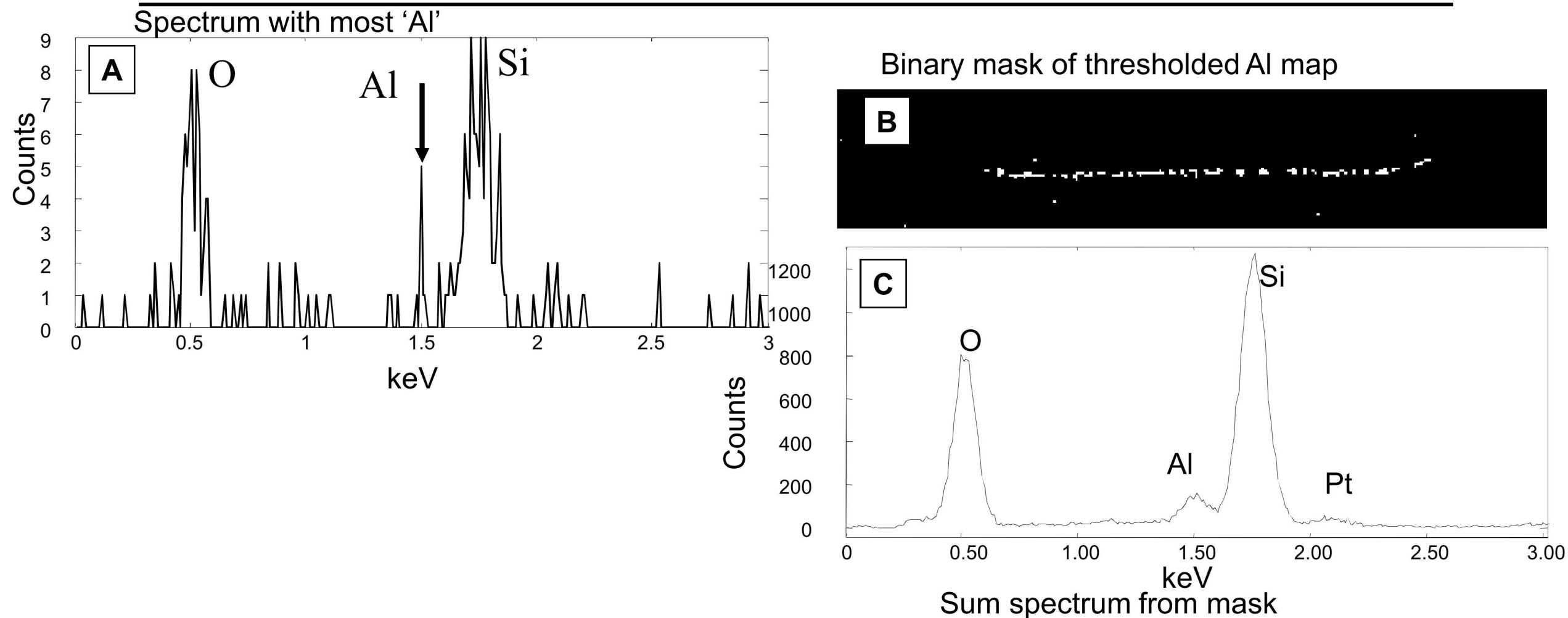
P.G. Kotula, et al. *Microsc. Microanal.* **12** [6] 538-544.

Example of Conventional Analysis

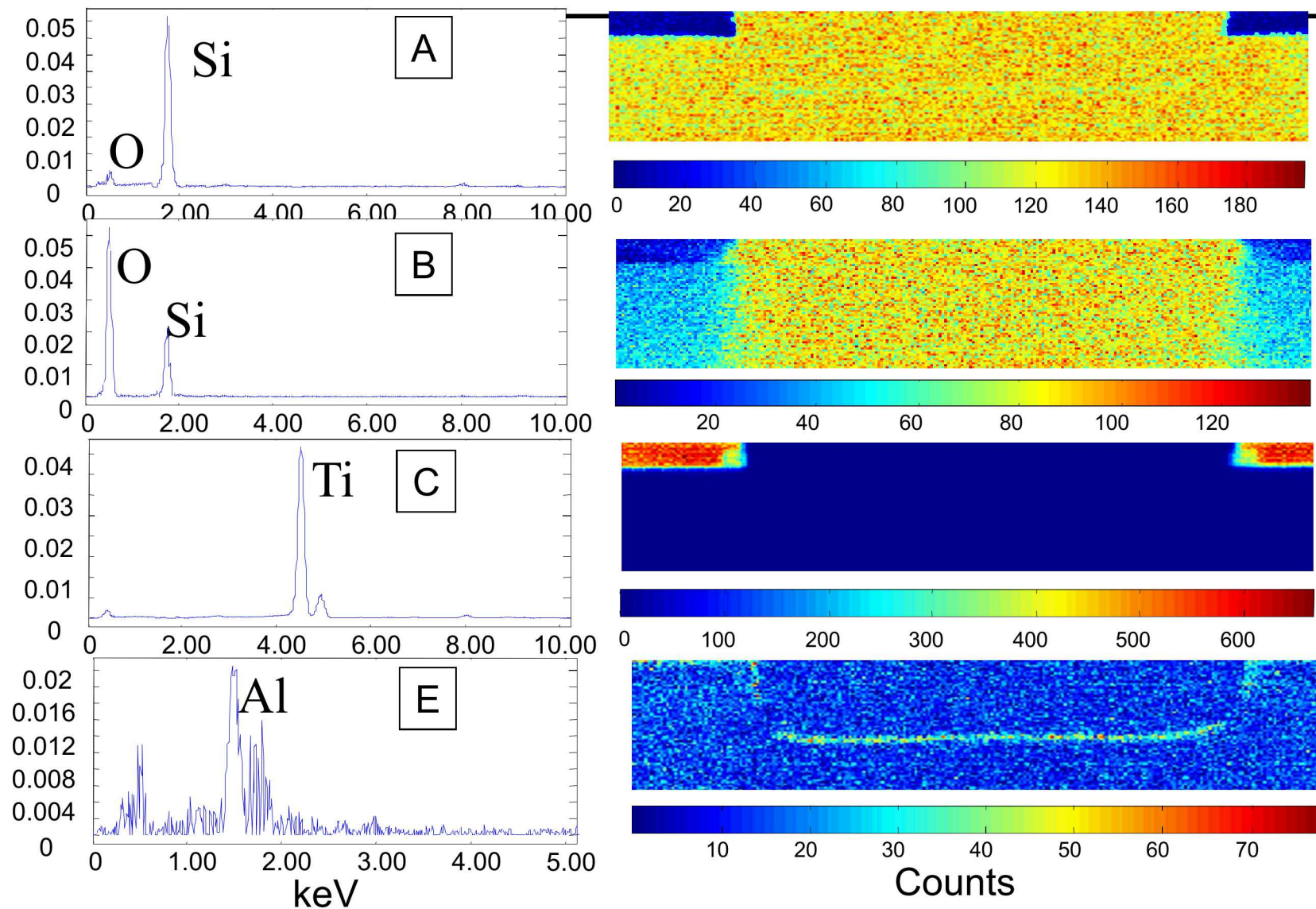


P.G. Kotula, et al. *Microsc. Microanal.* **12** [6] 538-544.

Example of Conventional Analysis



Multivariate Statistical Analysis of the same data



P.G. Kotula, et al. *Microsc. Microanal.*
12 [6] 538-544.

Why is MSA important?

- Spectral images are large and getting larger
- Impossible to manually interrogate
- MSA can reduce the dimension of the data
- Separate chemical information from redundancy and noise

Keenan, M.R., *Multivariate analysis of spectral images composed of count data*, in *Techniques and applications of hyperspectral image analysis*, H. Grahn and P. Geladi, Editors. 2007, John Wiley & Sons: Chinchester.

Marcellus shale

4 μm

RGB image

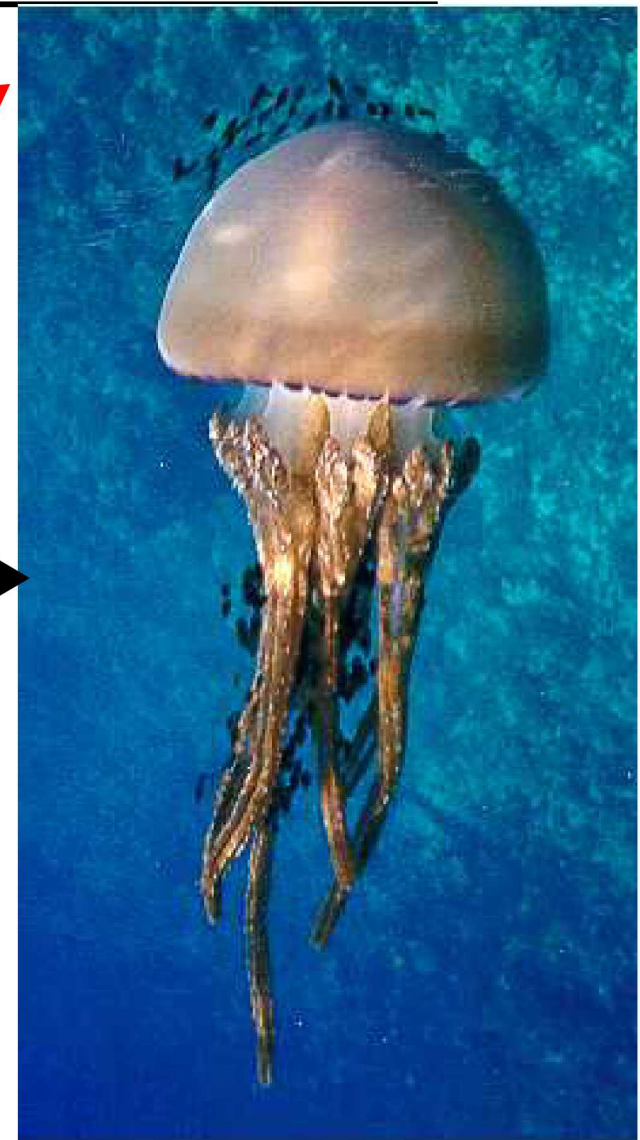
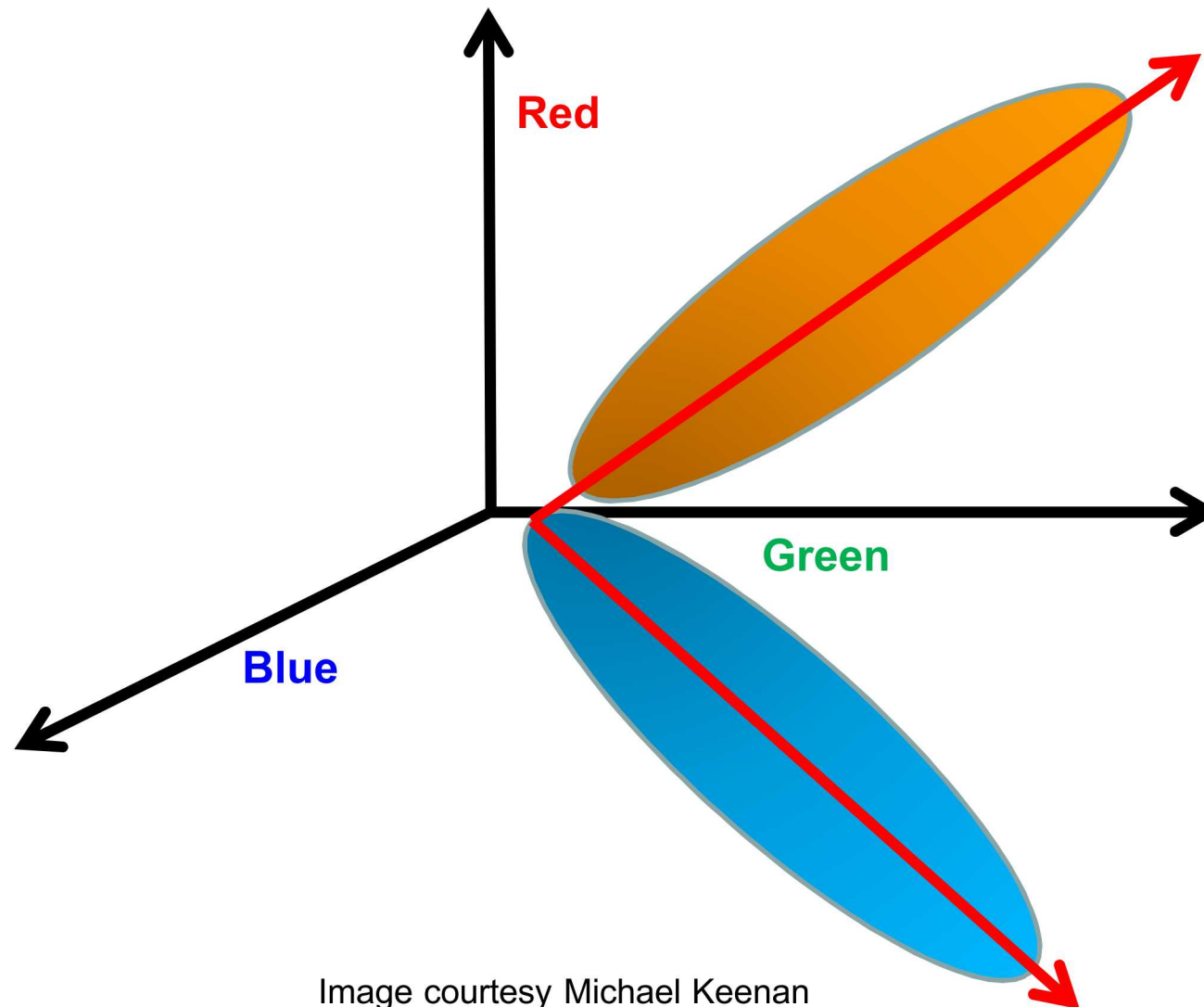
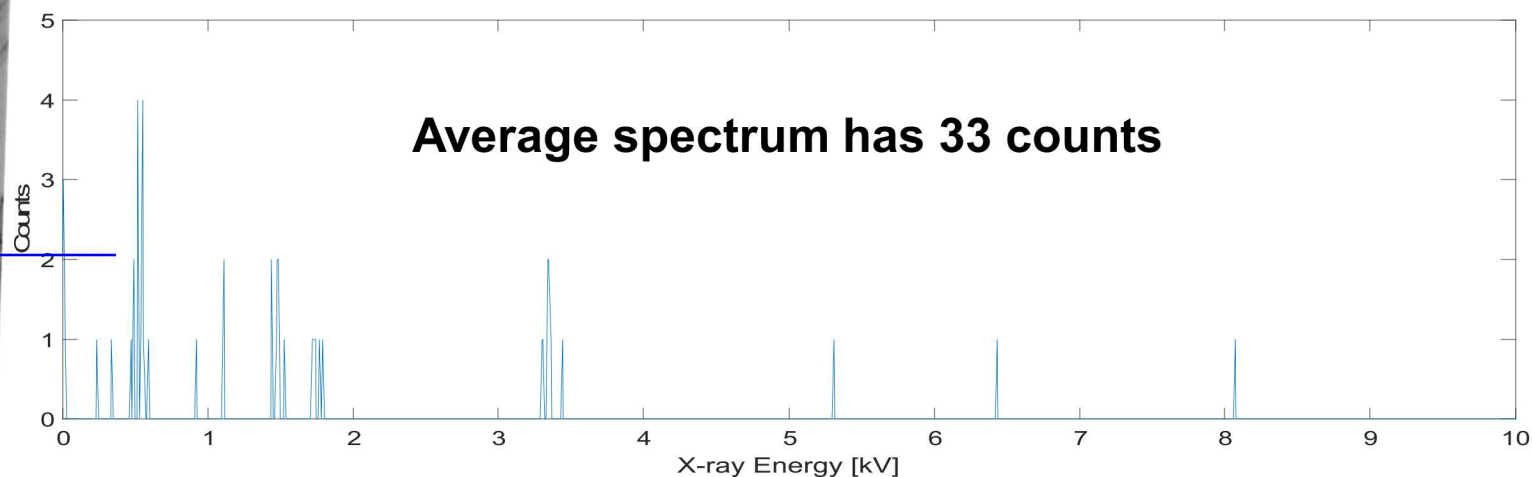
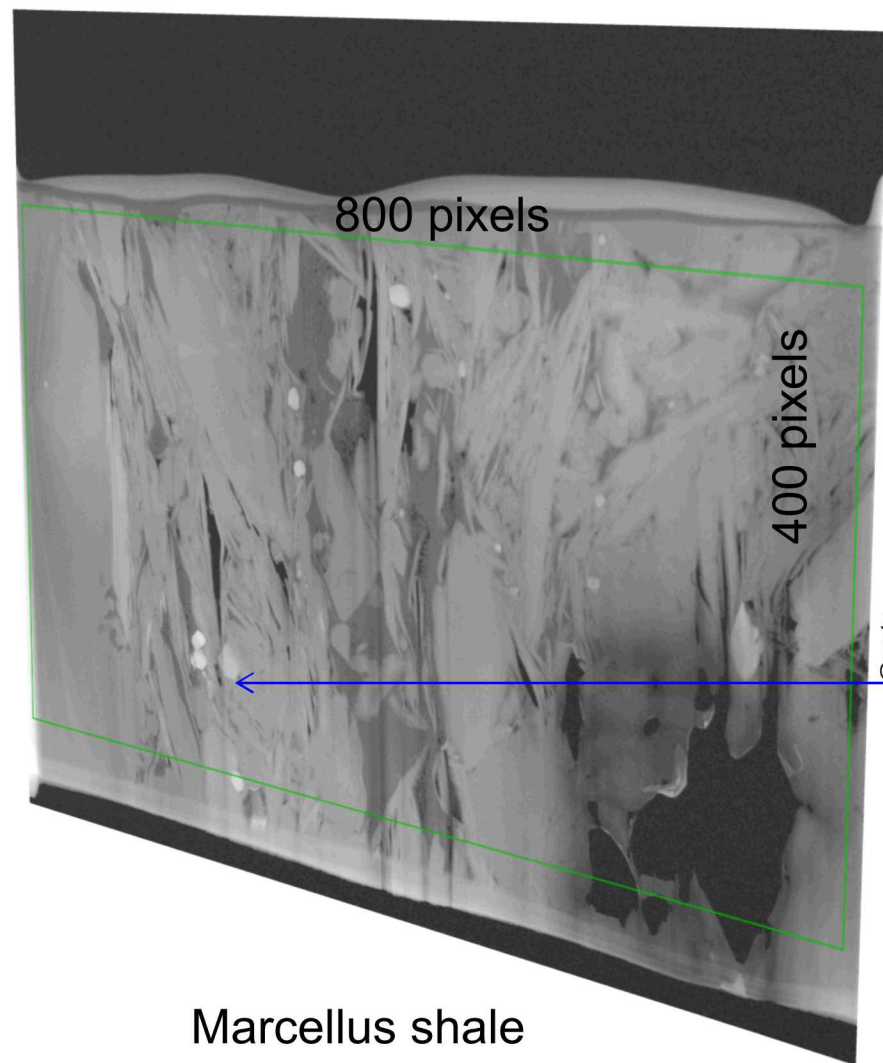
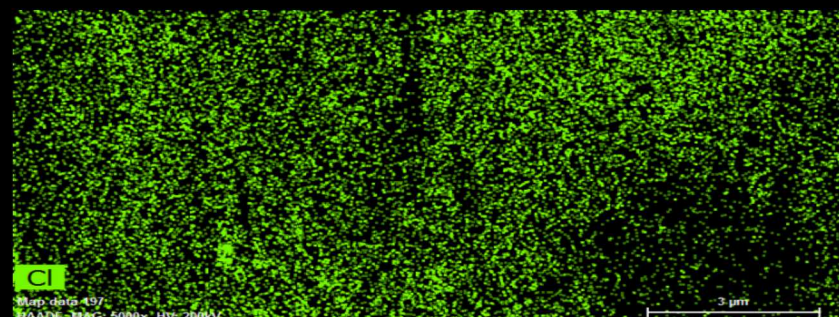
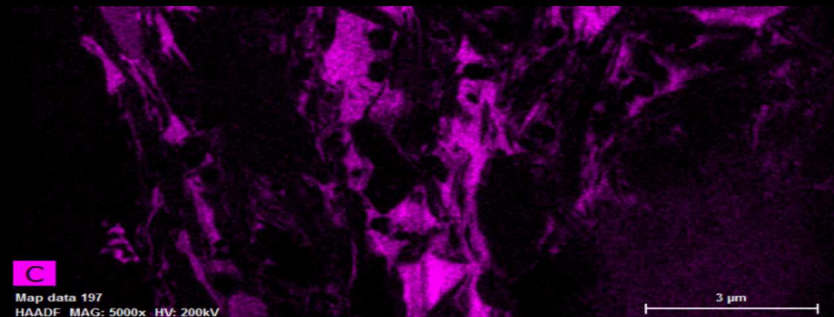
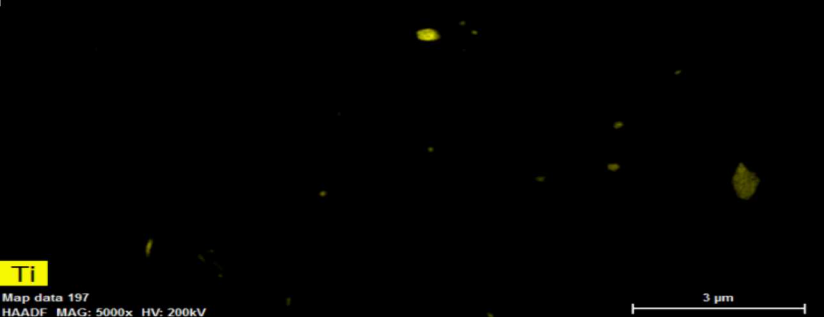
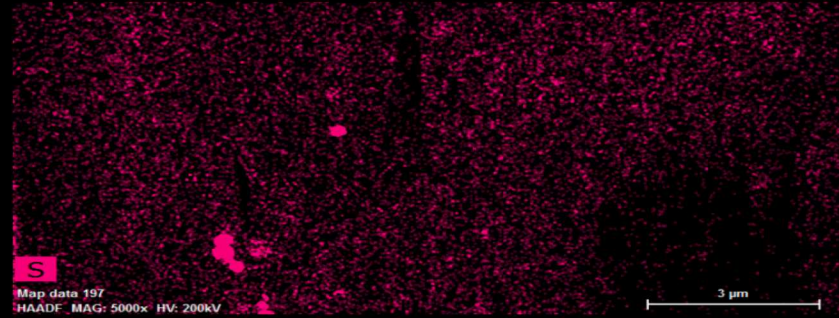
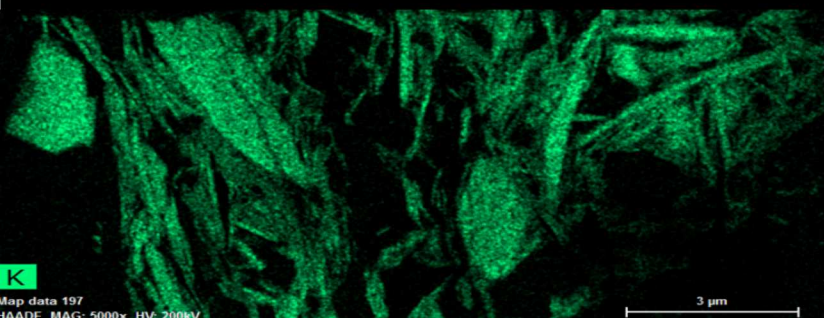
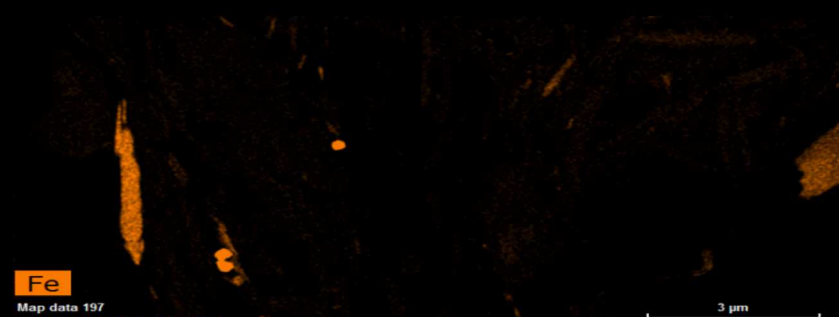
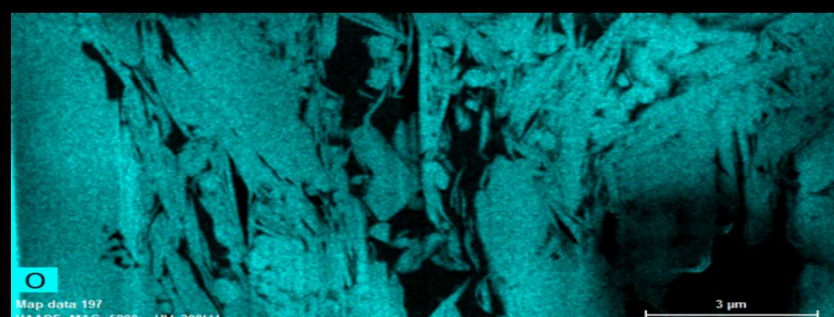
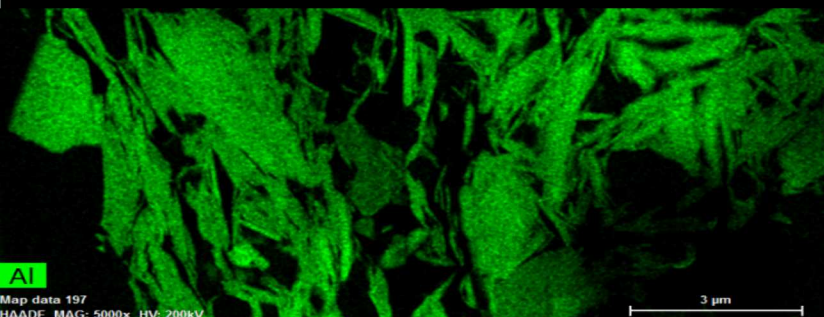
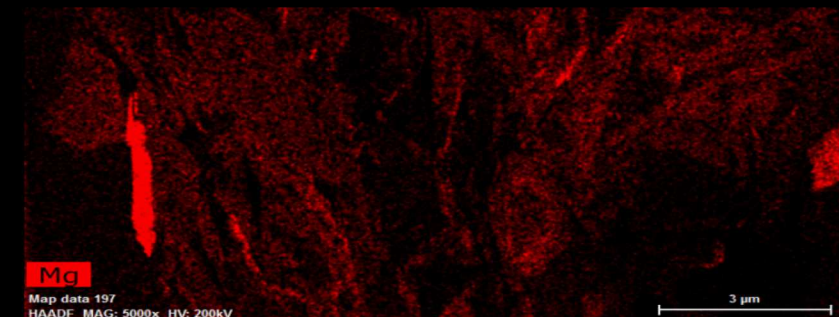


Image courtesy Michael Keenan

The Data Set

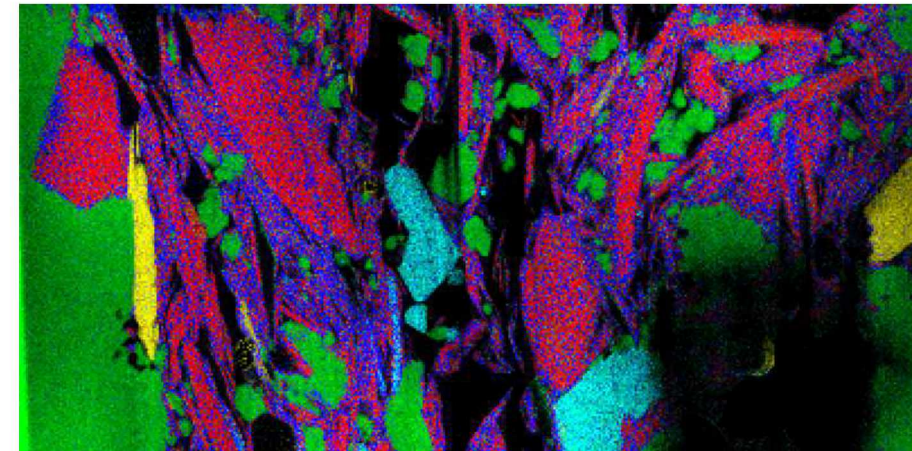
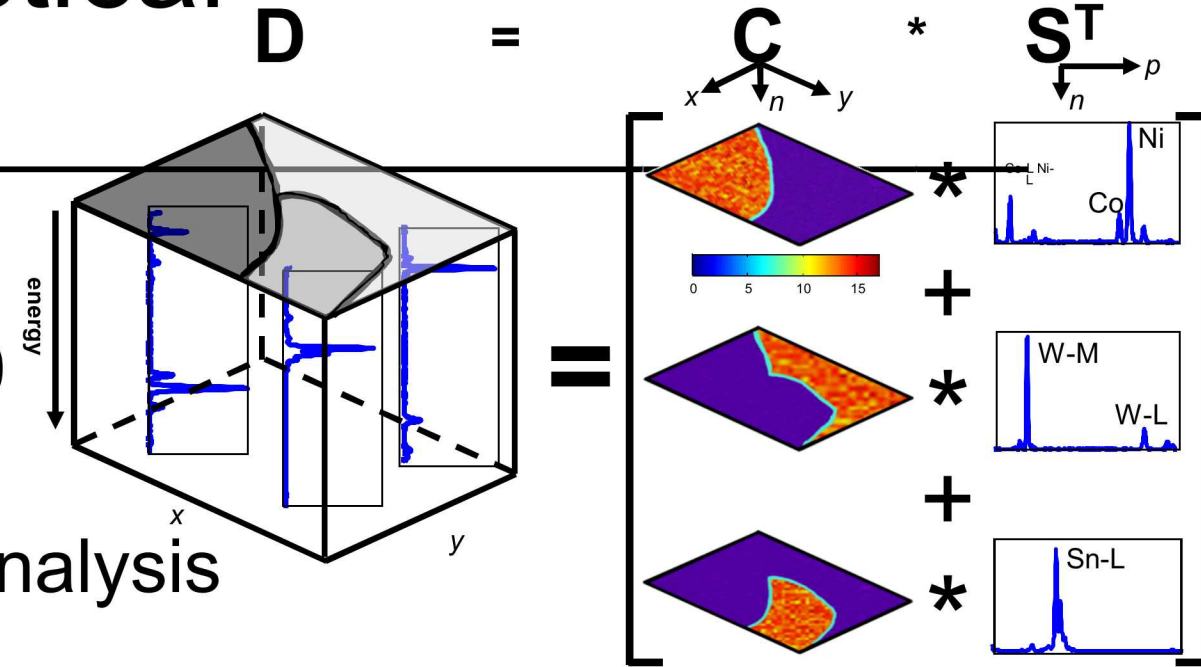
- A little over 1 hour of AEM data acquisition
- 320,000 spectra each with 2048 channels
- Spectra are noisy
- Can MSA make anything out of this?



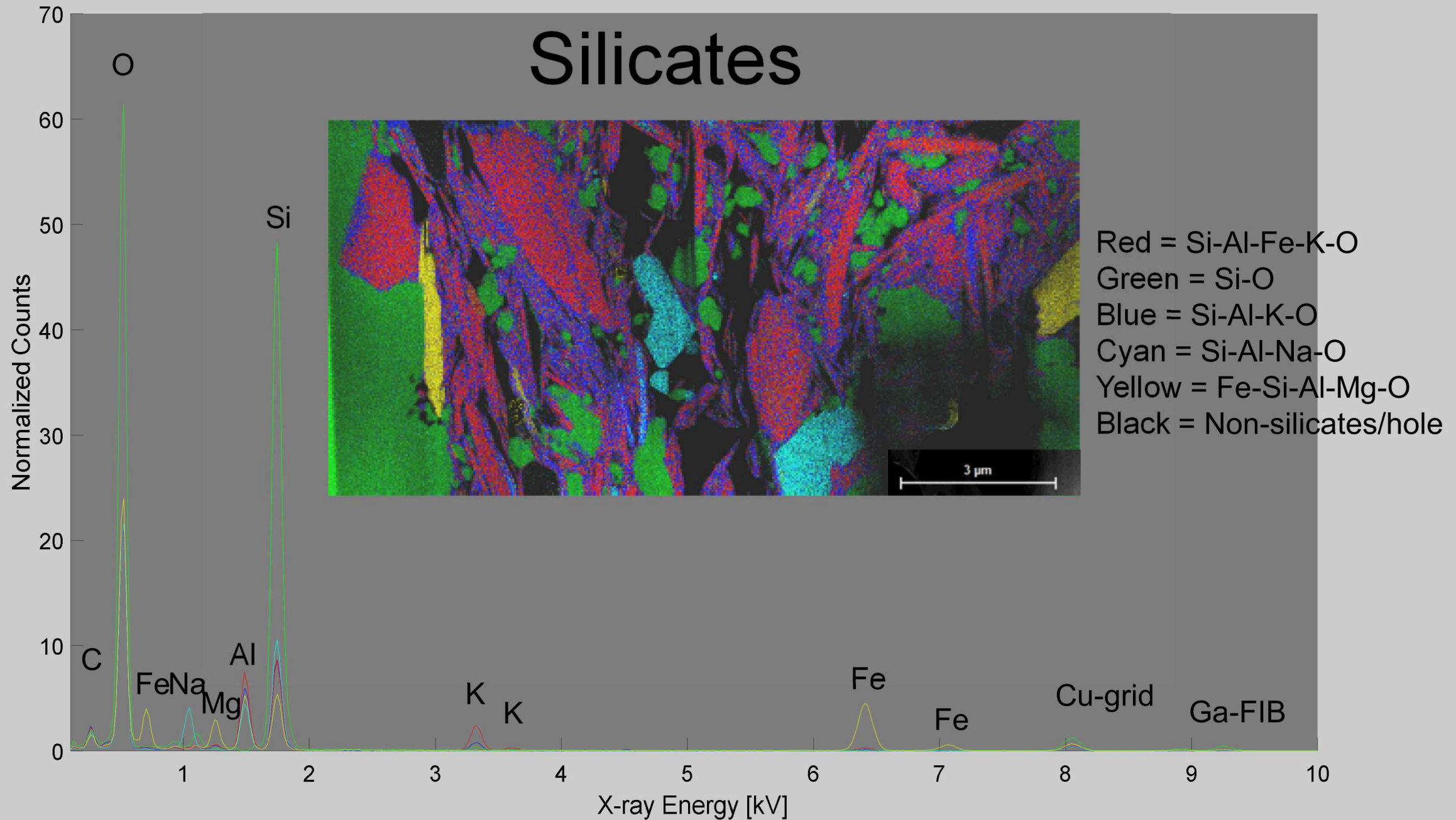


Multivariate Statistical Analysis

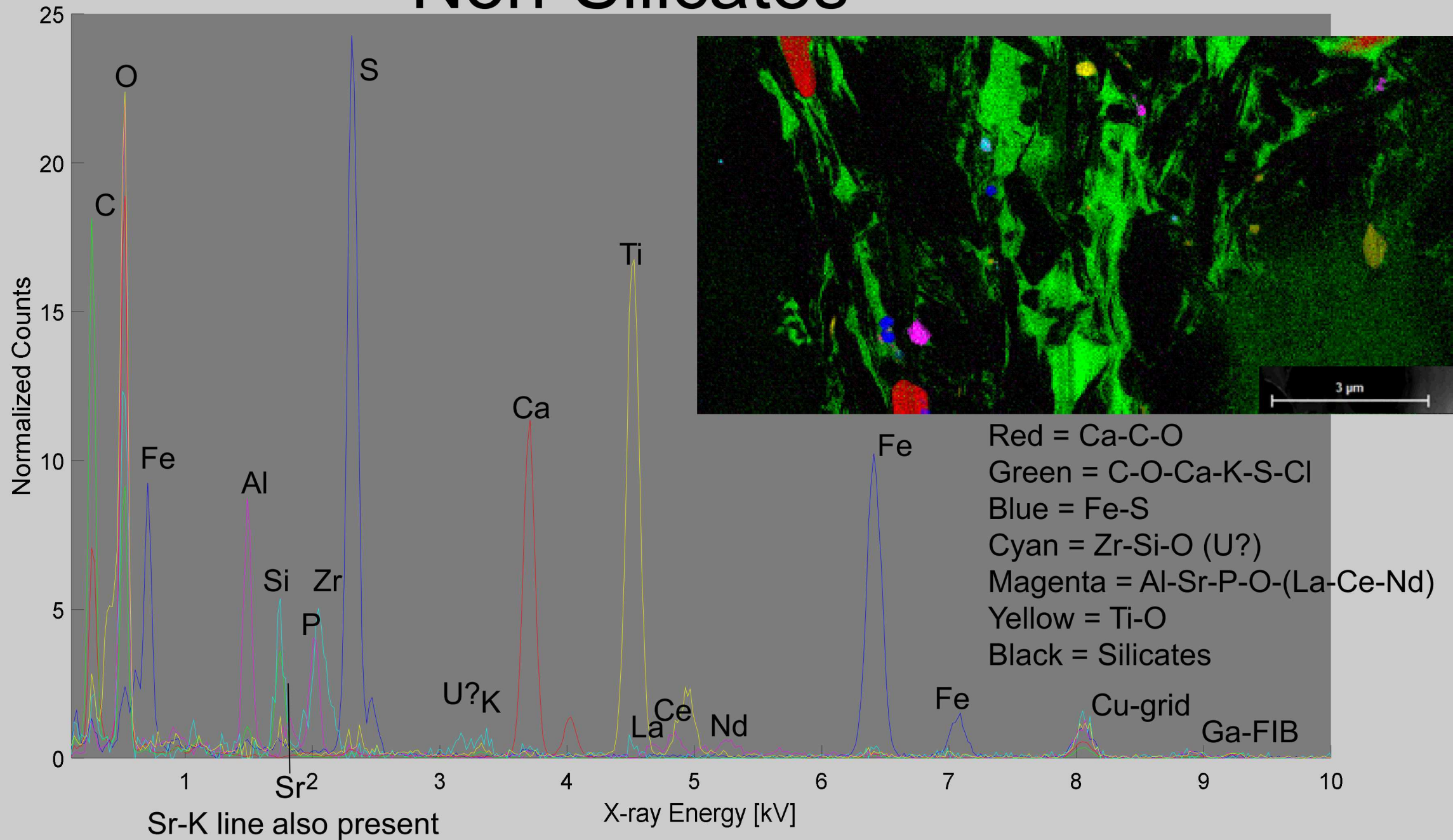
- $D = CS^T$
- **Compress** data to get more than 100 counts/new pixel
- **Normalize** for Poisson noise (Eigenanalysis for rank determination...13)
- Compute **Principal Components**
- **Rotate** to simple **spatial** or spectral structure
- Apply **non-negativity** constraint
- **Inverse normalization** for noise
- **25 seconds later...ready for interpretation**



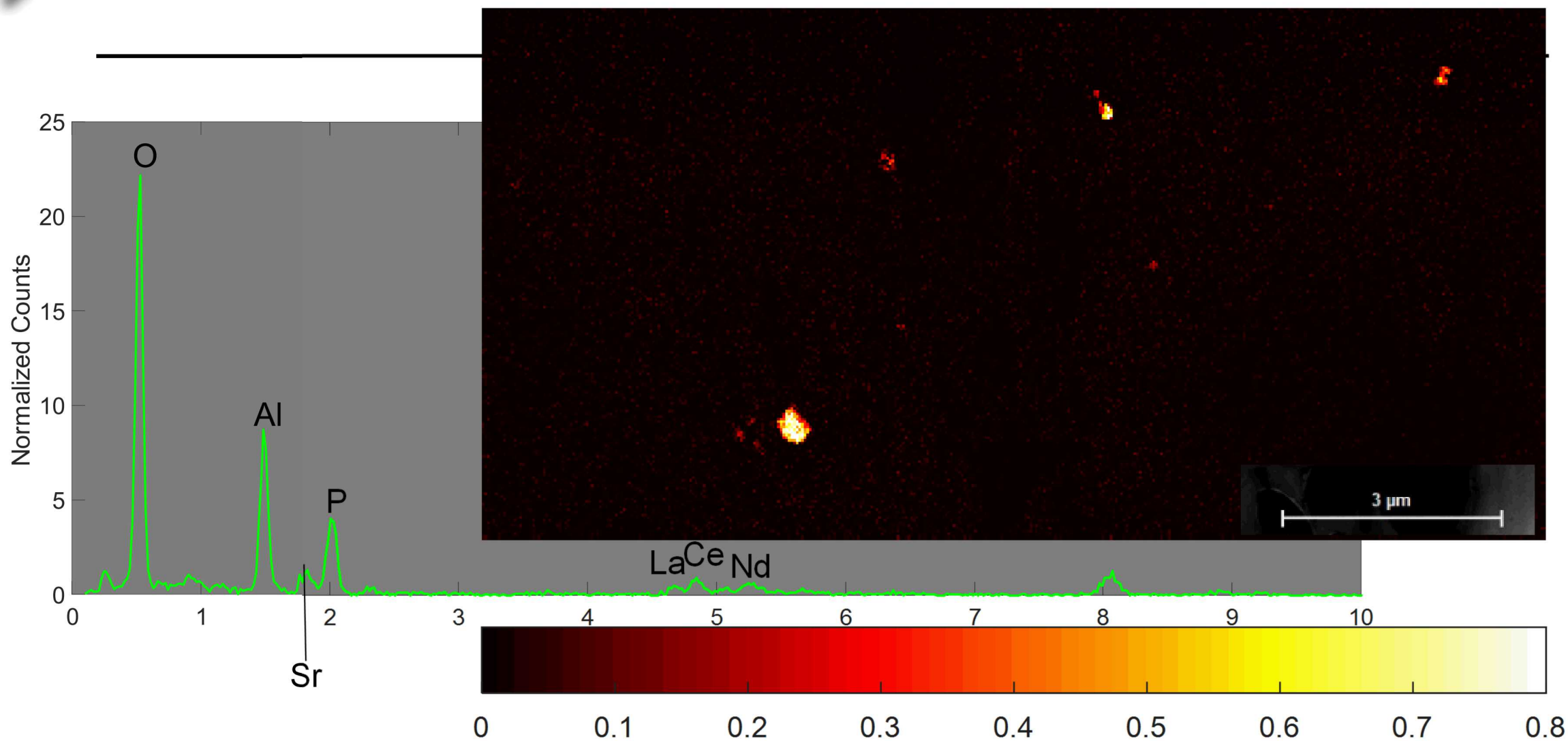
Silicates



Non-Silicates

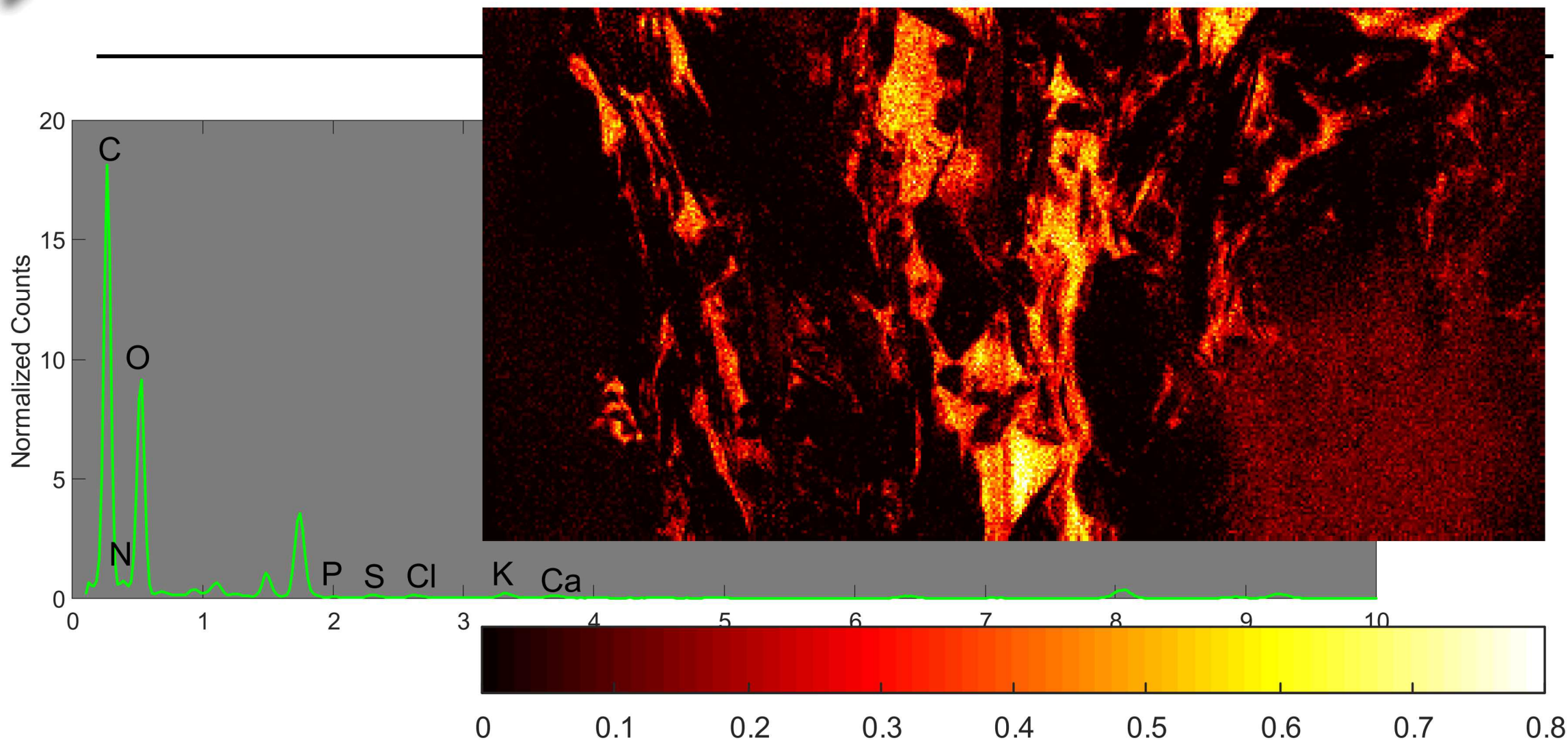


Interesting RE-containing phase



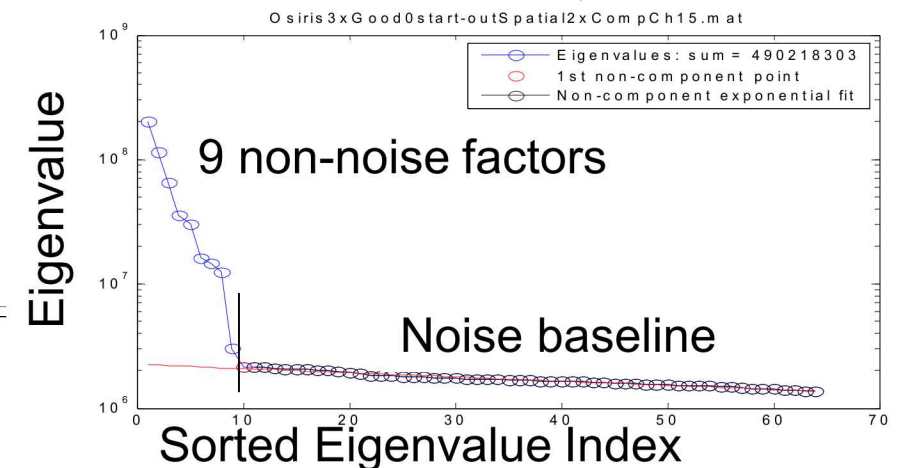
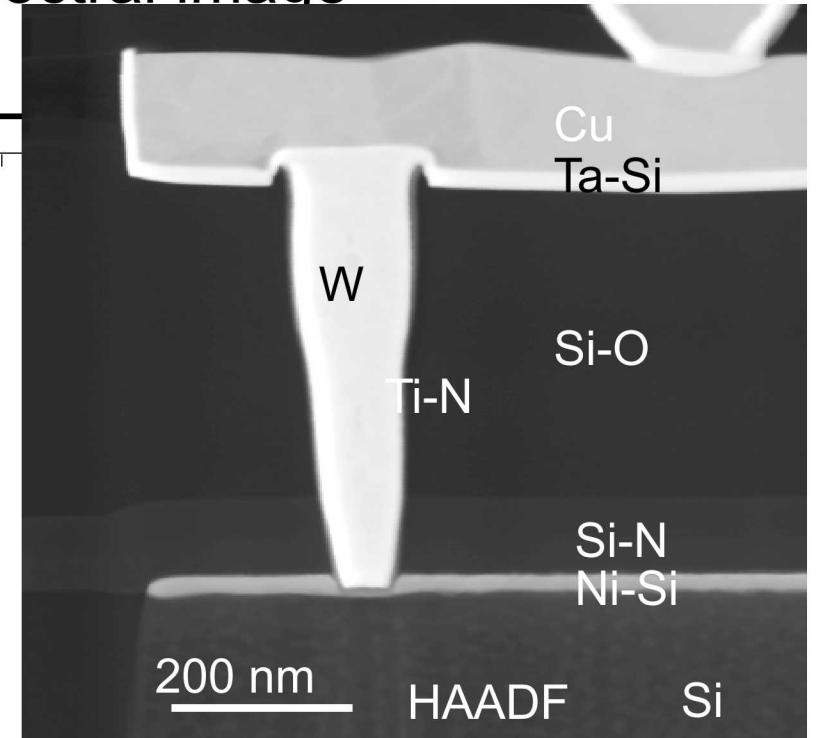
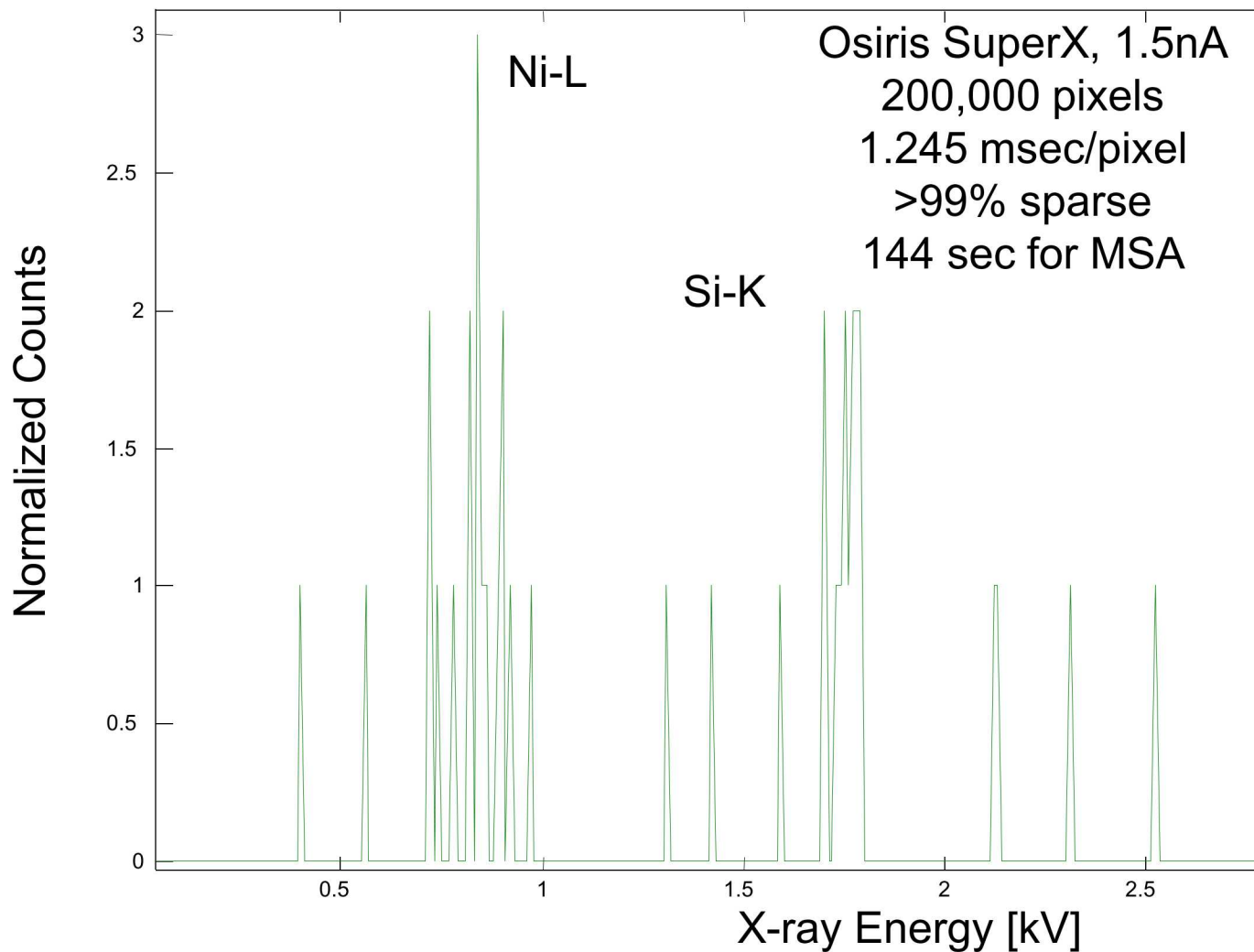
The power of statistics!

Carbon-containing material

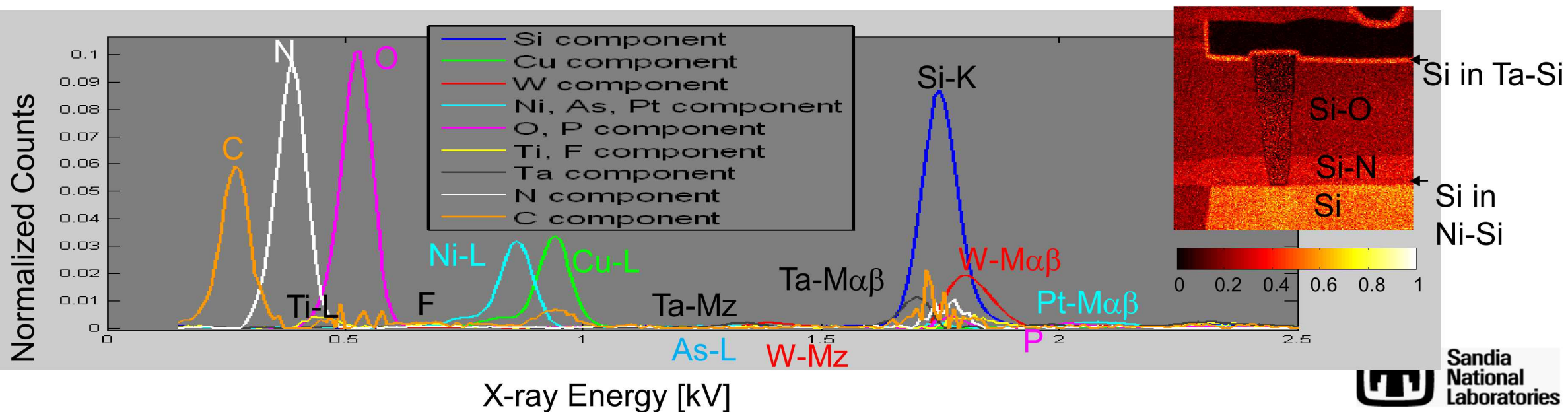
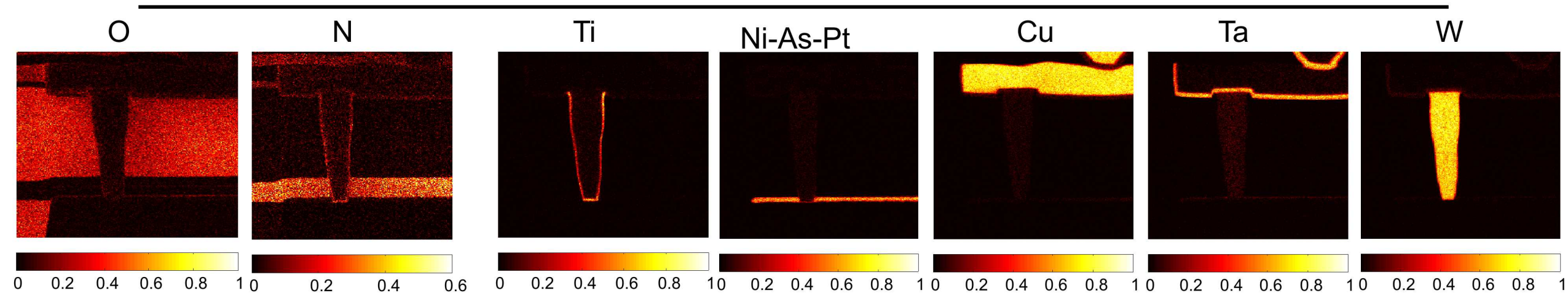


Low end spatially, high end for sensitivity

Raw spectrum from the CMOS spectral image

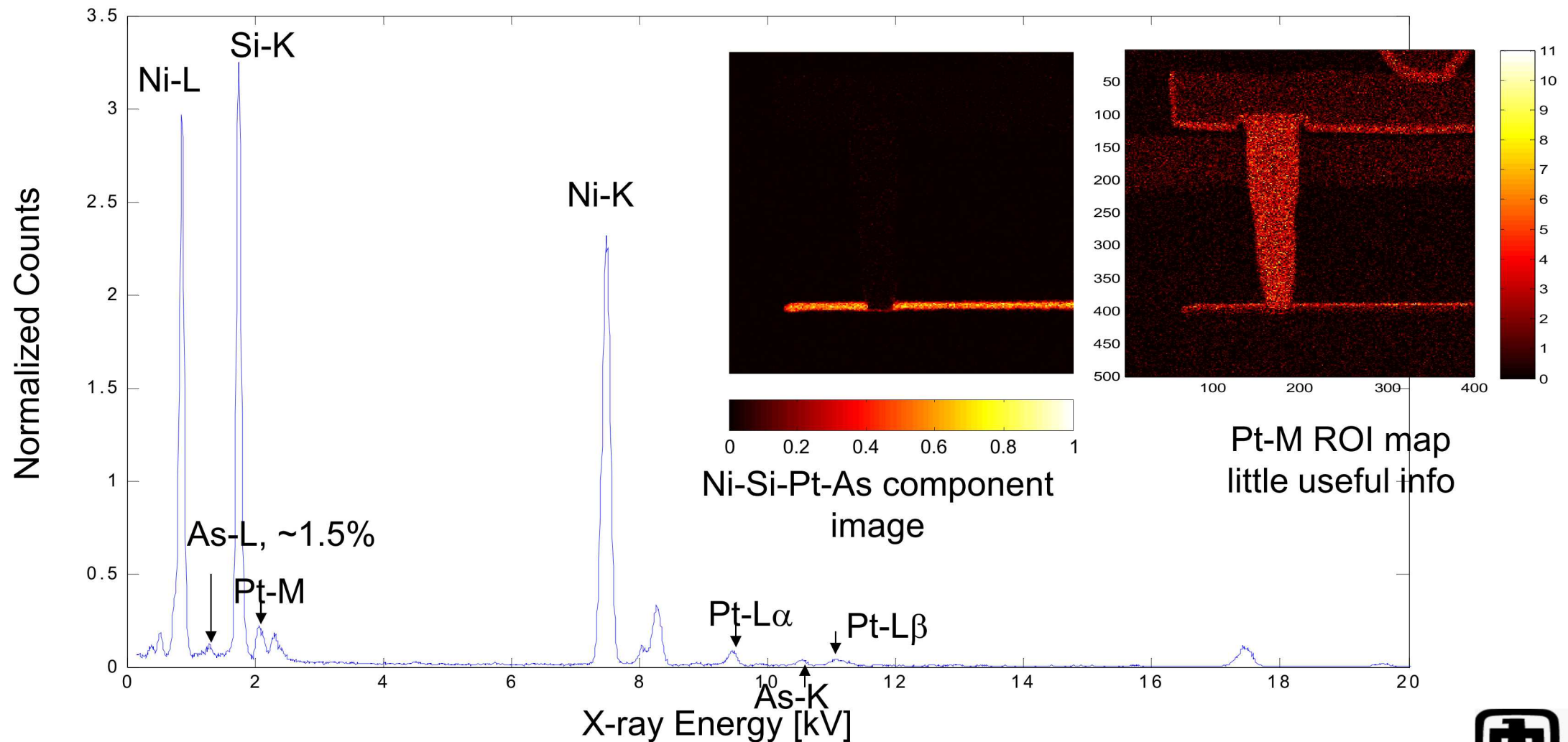


Low end, Spectral-Domain Simplicity Best Spectral or Elemental 'Contrast'



Low end, Spatial-Domain Simplicity

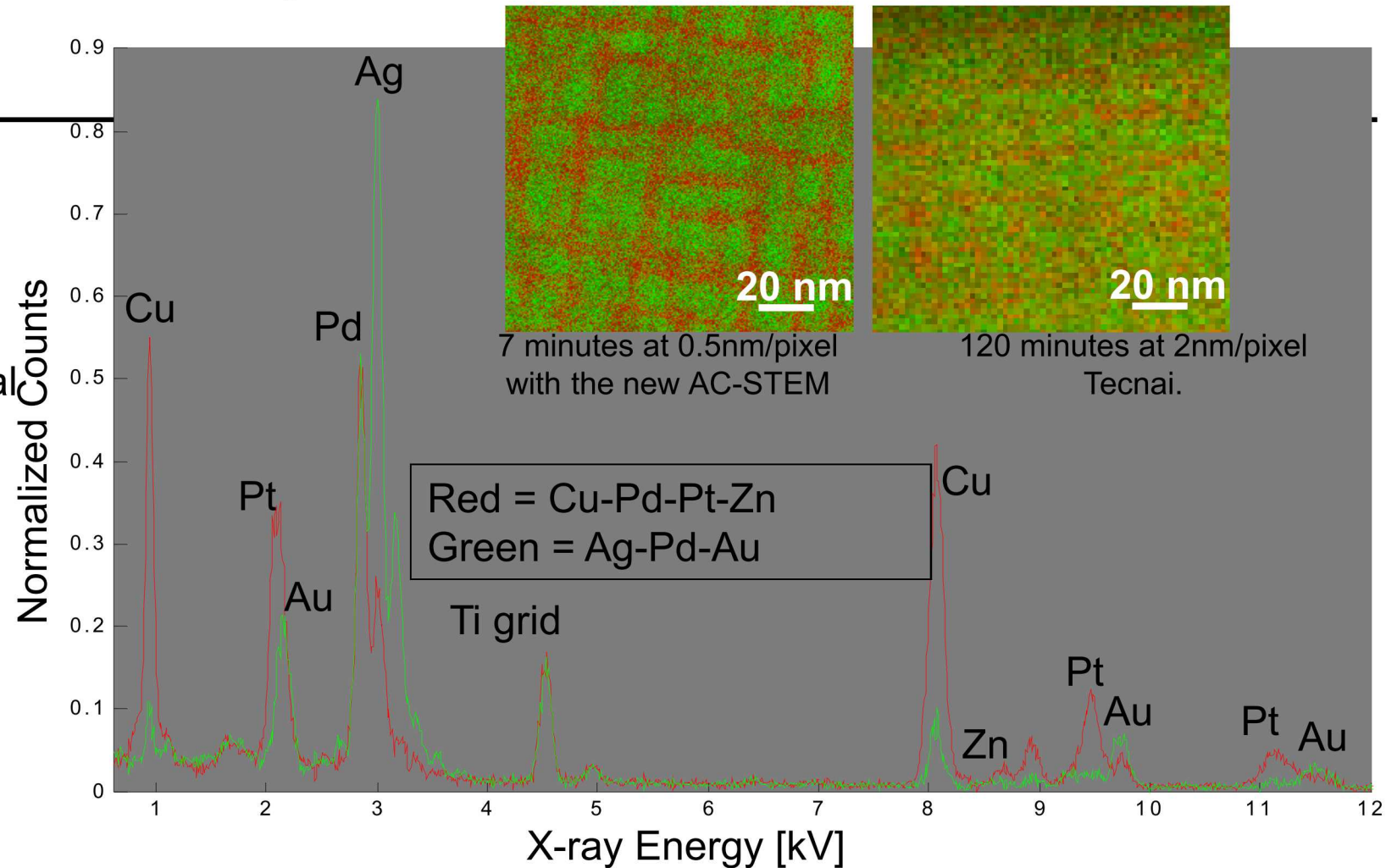
Ni-silicide contact, MSA shows minor elements



Medium end analysis

Sub-nm microanalysis of electrical contact materials

Paliney 7, electrical contact material nanometer-scale spinodal decomposition.

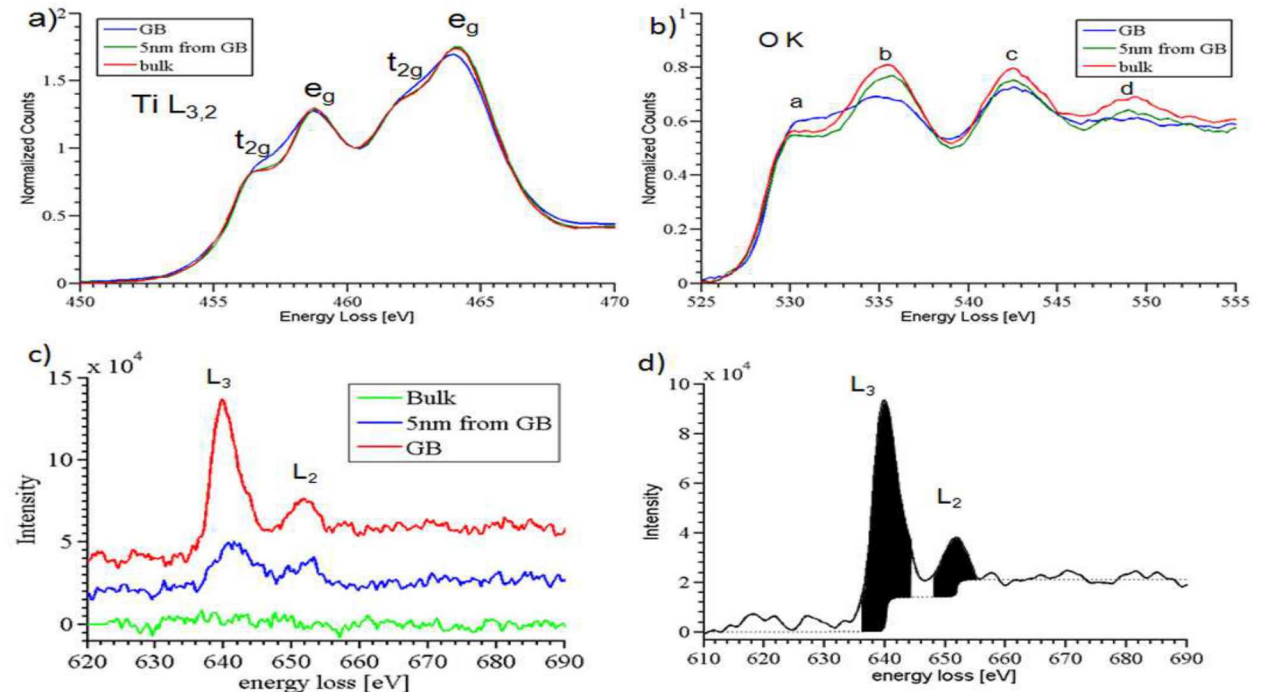
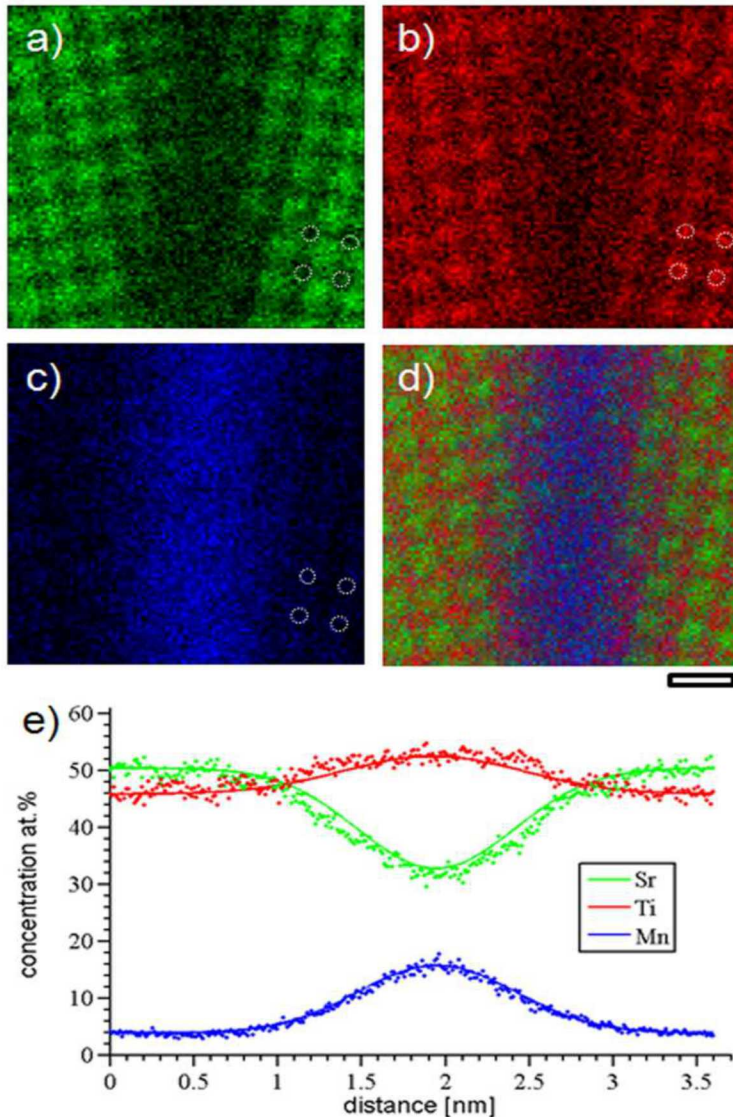


The analytical power of the AC-STEM is *at least 70x* better than the older analytical microscope at Sandia.

Analysis of Mn-doped STO $\Sigma=13$ Boundaries

$\Sigma 13$ (510)/[001]

Quantitative EDS and EELS



Mn²⁺ at boundary
Mn⁴⁺ in bulk near boundary
(substitutional with Ti)

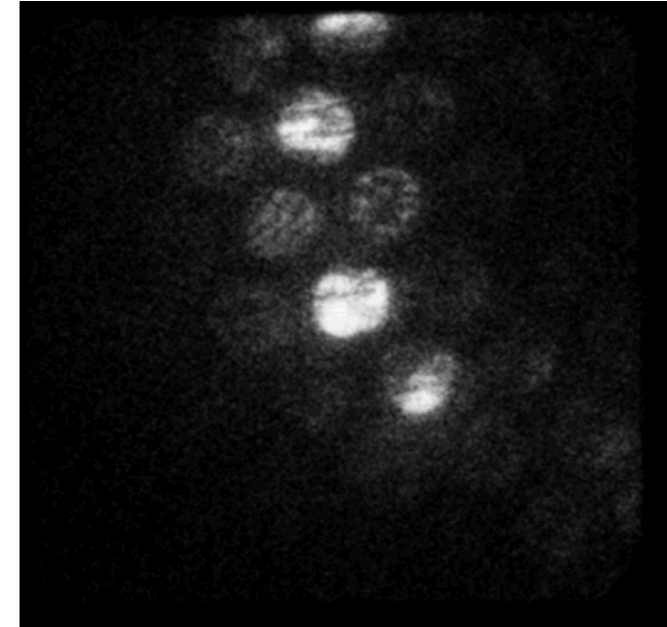
FWTM enrichment at the
boundary less than 2nm

Hao Yang, Paul G. Kotula, Yukio Sato, Yuichi Ikuhara, Nigel D. Browning. "Segregation of Mn²⁺ Dopants as Interstitials in SrTiO₃ Grain Boundaries," Submitted (2013).

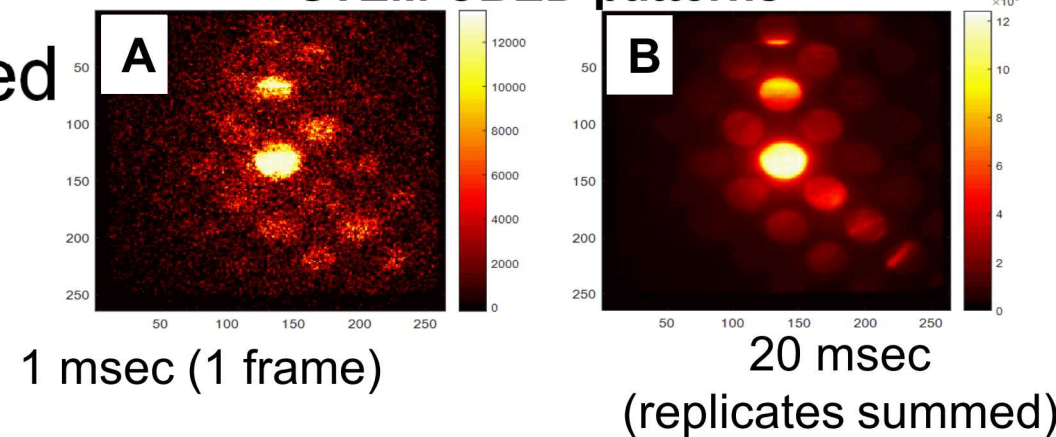
4D-STEM

STEM-CBED and MSA...A Tale of Big Data

- Titan G2 80-200 STEM
- pnCCD mounted below the projection chamber
- 4.8 mrad convergence angle in probe with 30 pA
- 256x256 real-space pixels
- 264x264 reciprocal-space pixels
- 4×10^9 data elements as summed
- No “small” dimension

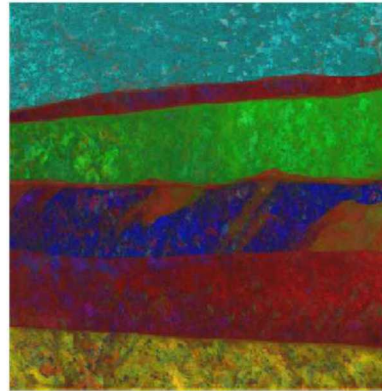
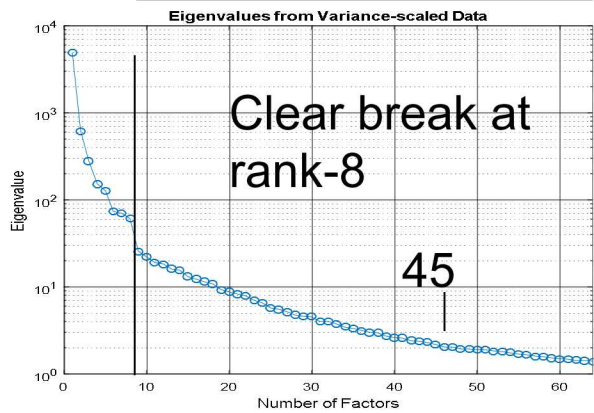


STEM CBED patterns

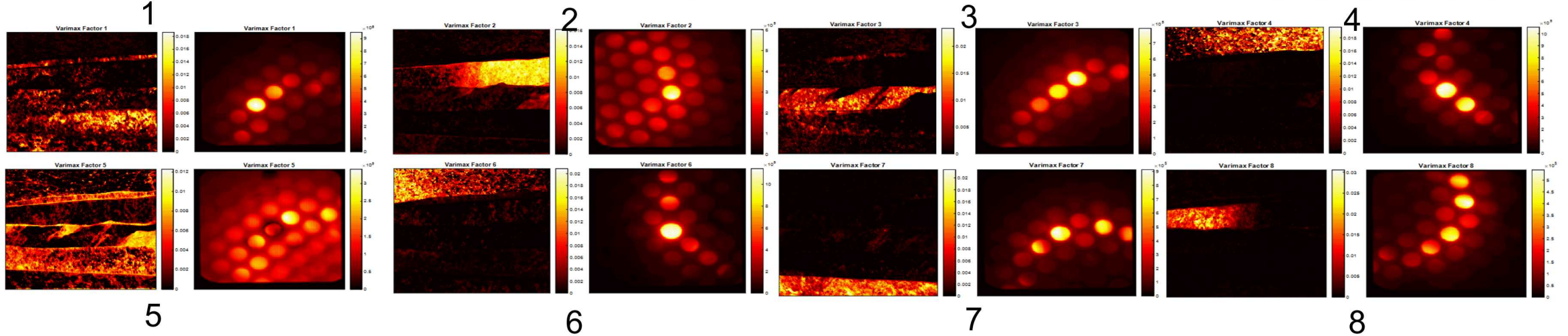
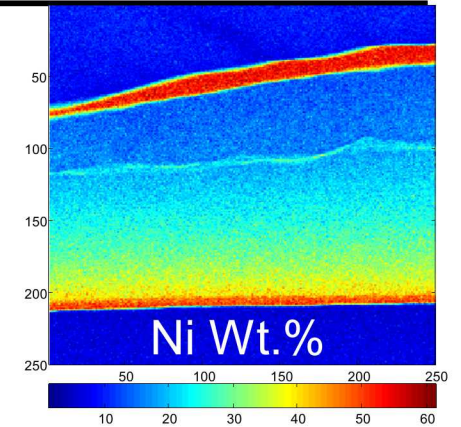


STEM-CBED and MSA...Real-space simplicity

38 Gbyte in, 9 Mbyte after MSA, Compression factor of ~21000



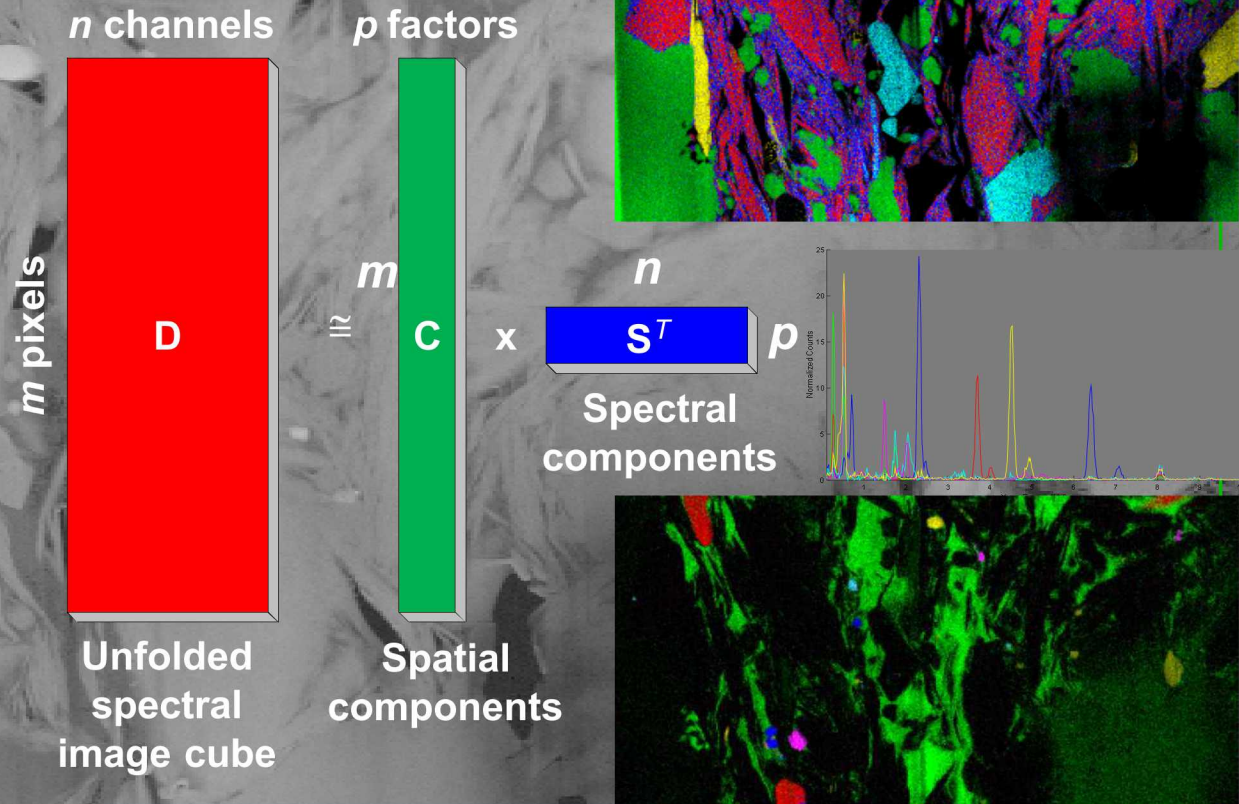
Red = 1+5 (fcc)
Green = 2+8 (fcc)
Blue = 3 (fcc)
Cyan = 4+6 (bcc)
Yellow = 7 (bcc)



First-pass MSA breaks out largely by crystallographic phase/orientation

Conclusions

- MSA methods are powerful and highly relevant today
- General phase analysis
- Needle in the haystack
- Additional data modalities
- Commercially and non-commercially available
- Paul.Kotula@sandia.gov



Analysis goal: Obtain an easily interpretable representation of the data