## ICDD March meetings 20'io

### Employing μ-XRF for detection of Palladium defects

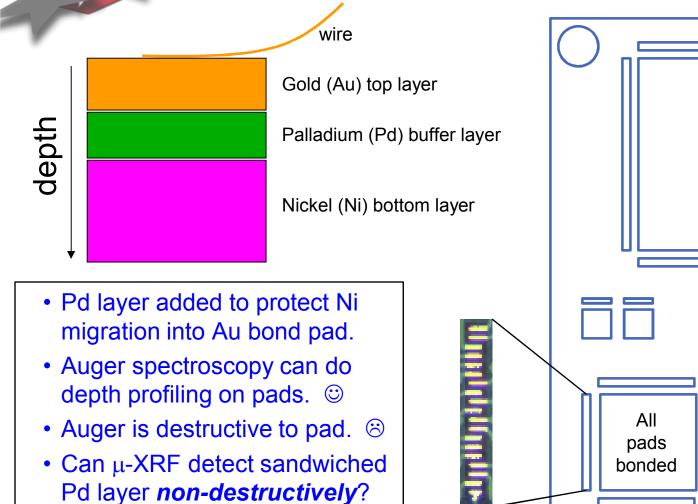
Mark A. Rodriguez, James J. M. Griego, Mark H. Van Benthem, Paul G. Kotula and Paul T. Vianco

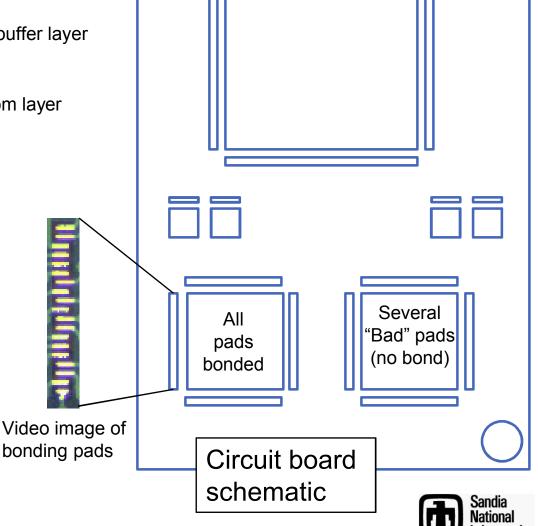
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### Problem: some wire-bonding pads will not bond.









- Use Bruker M4 Tornado μ-XRF to scan circuit boards and look for Pd interlayer on electrode pads.
- Apply Principal Component Analysis (PCA) with Multivariate Curve Resolution (MCR) algorithms to aid data extraction of weak signal embedded in data.



Courtesy Bruker -Nano

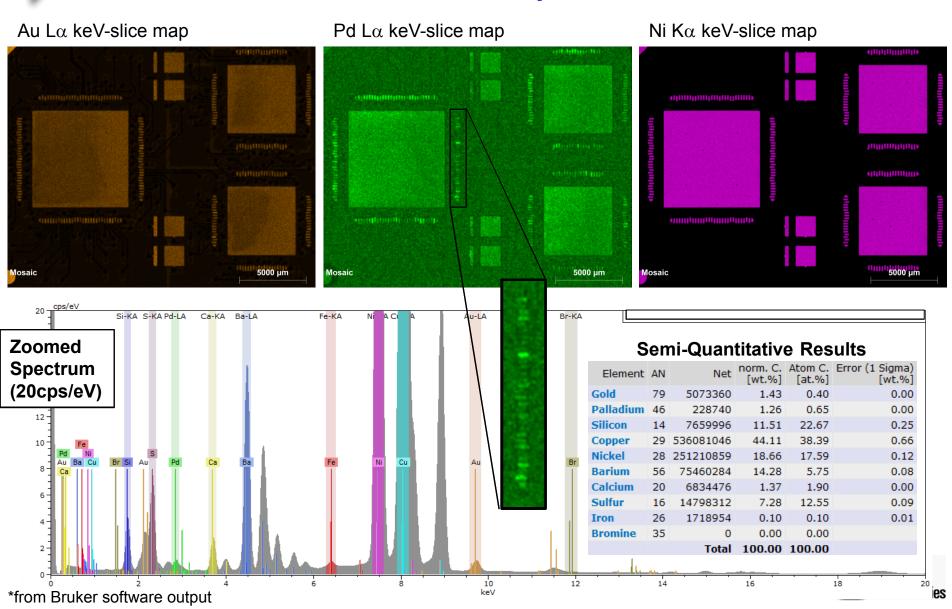
#### Bruker M4 Tornado μ-XRF

- Rh microfocus source 50 kV/ 600 μA
- High flux polycapillary optic
- ~30 μm spatial resolution
- XFlash® silicon drift detector

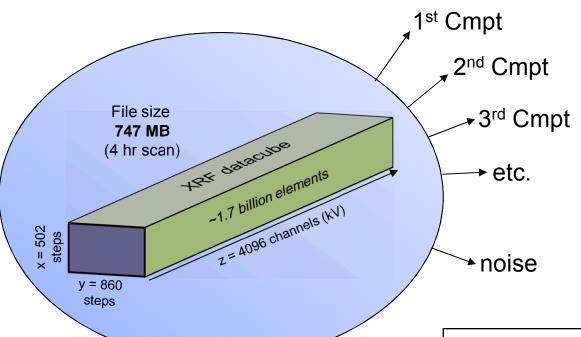


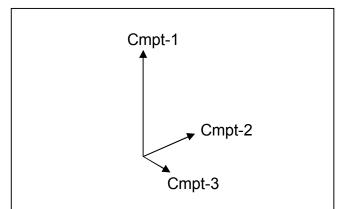


Although Cu, Ni, Si, Ba elements dominate the conventional XRF spectrum, we can *still* see spatial dependence of Pd and Au in **keV-slice maps\*** even at ~1 wt% levels.



## We employ Principal Component Analysis to reduce *massive* μ-XRF datasets





Each derived **component** is mathematically constrained to be orthogonal to others

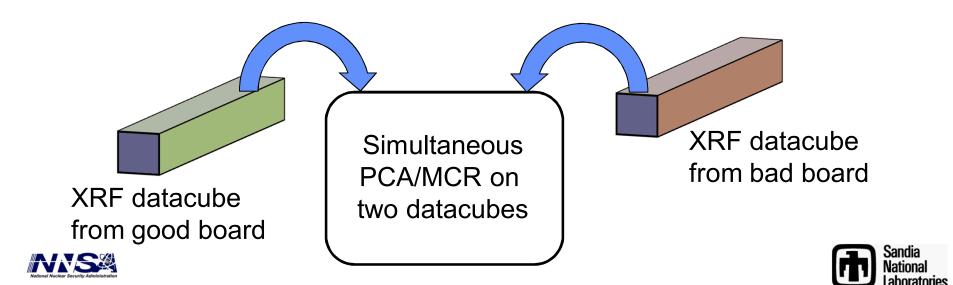
We assume that the entire XRF datacube can be represented by *linear combinations* of all species generating fluorescence signal

- PCA decomposes XRF spectra using the entire datacube to find energy features common to given spatial locations
- Multivariate Curve Resolution (MCR) helps enable more interpretable results

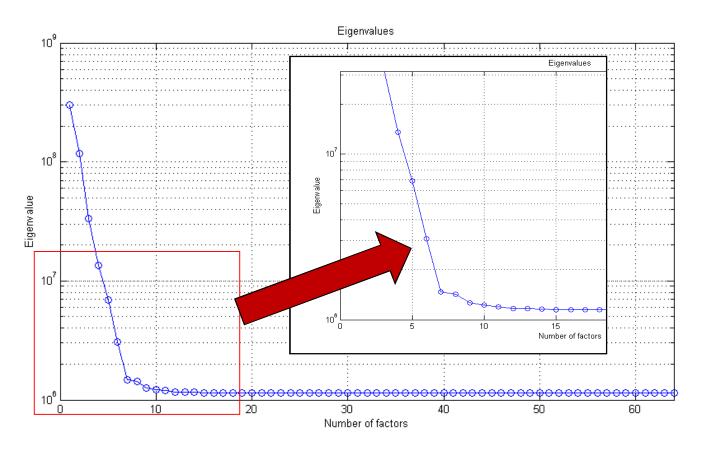


# Principal Component Analysis (PCA) did not easily extract Pd defects from individual datasets.

- Pd signal is highly spatially-correlated to Au and Ni signal
- Low Pd signal and significant Rh/Pd emission-line overlap
- The proportionality of Au/Pd/Ni ratios may prevent isolation of Pd as a component.
- Solution: Perform PCA/MCR with at least two datacubes
  - a good circuit board and one with problems in Pd pads.



# The Eigenvalue analysis suggests at least 9 factors present in the massive dataset for the combined "good-bad" circuit boards



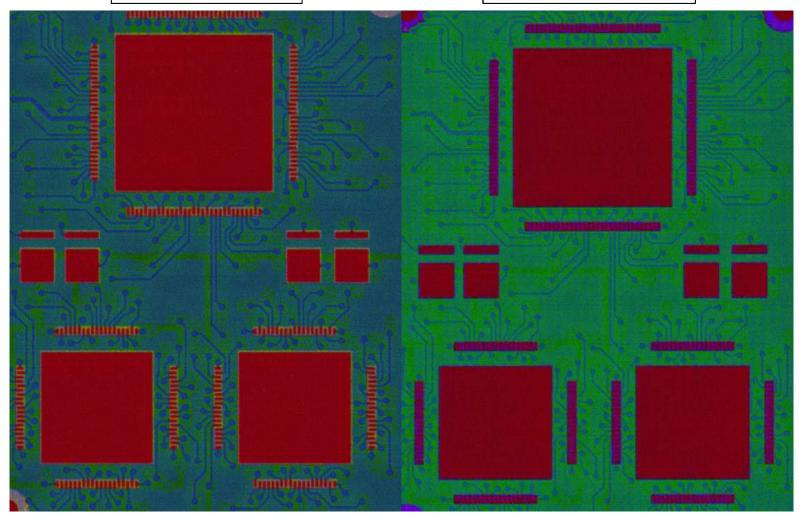




False color chemical maps can be generated from XRF spectrums of extracted components

"Bad" board

"Good" board

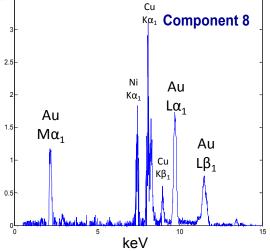


Red = Ni Green = Ba, S Blue = Cu Magenta = Al Yellow = Si

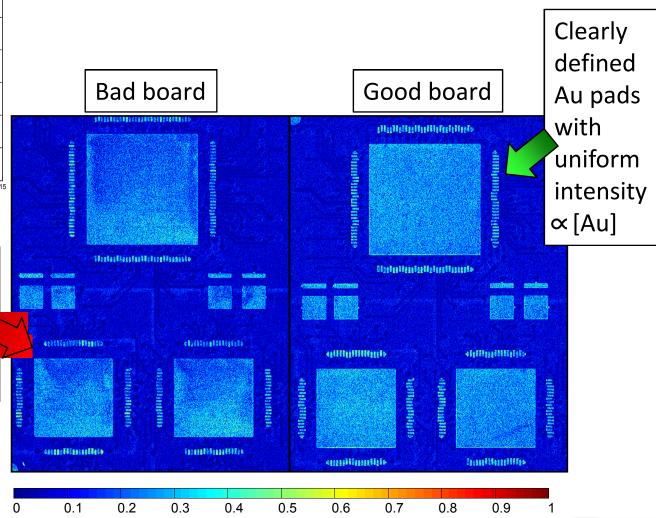




# PCA/MCR of Component 8 shows primarily **Gold** signal in spectrum with some mixing of Cu and Ni



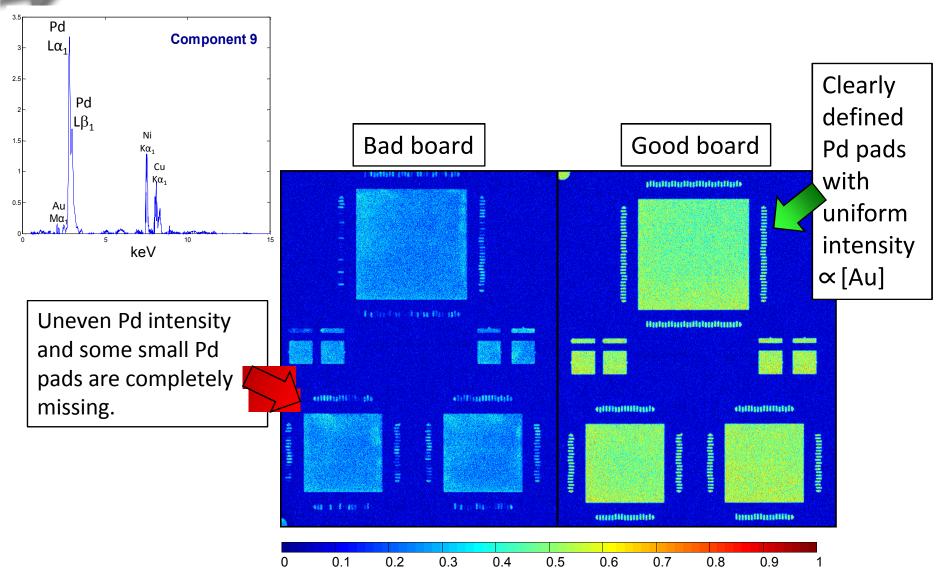
Some Au pads show uneven intensity and location of Au in pad is not always clearly defined.







# PCA/MCR for Component 9 shows primarily Palladium signal in spectrum with some mixing of Cu and Ni



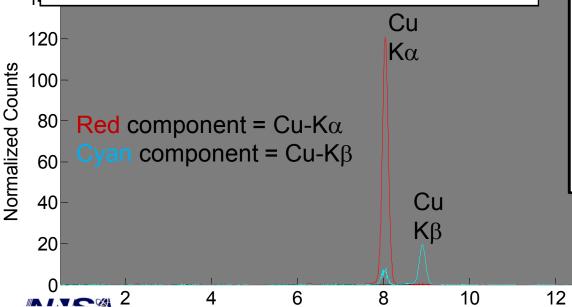




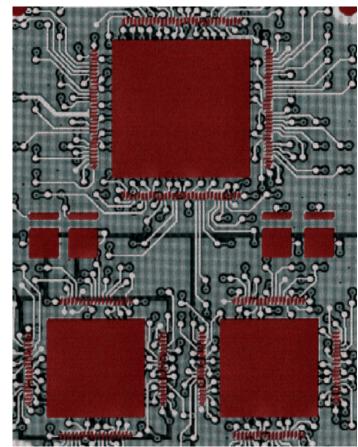
## PCA/MCR even sorts filtered and unfiltered radiation such as Ni-filtered Cu Kα wavelengths

Two *distinct* components were observed for Cu emission lines:

- $K\alpha$  emission only spectrum (red)
- another dominated by Kβ (cyan)
- The spatial map shows regions where the Cu  $K\alpha$  only is dominant (i.e. red blocks).
- Since red + cyan = white, the white circuit traces indicate a *combination* of the two components (i.e. unfiltered Cu spectrum).



X-ray Energy [kV]



The presence of Ni layer over the Cu circuitry renders these areas red.

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### **Summary**

- μ-XRF coupled with Principal Component Analysis can prove very helpful for detection of weak XRF signals.
- Simultaneous PCA on multiple circuit board datasets aided in detection of defects in sandwiched Pd metal layers.
- Filtering effects (e.g. Ni on Cu) can be diagnosed via separate components within PCA.

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