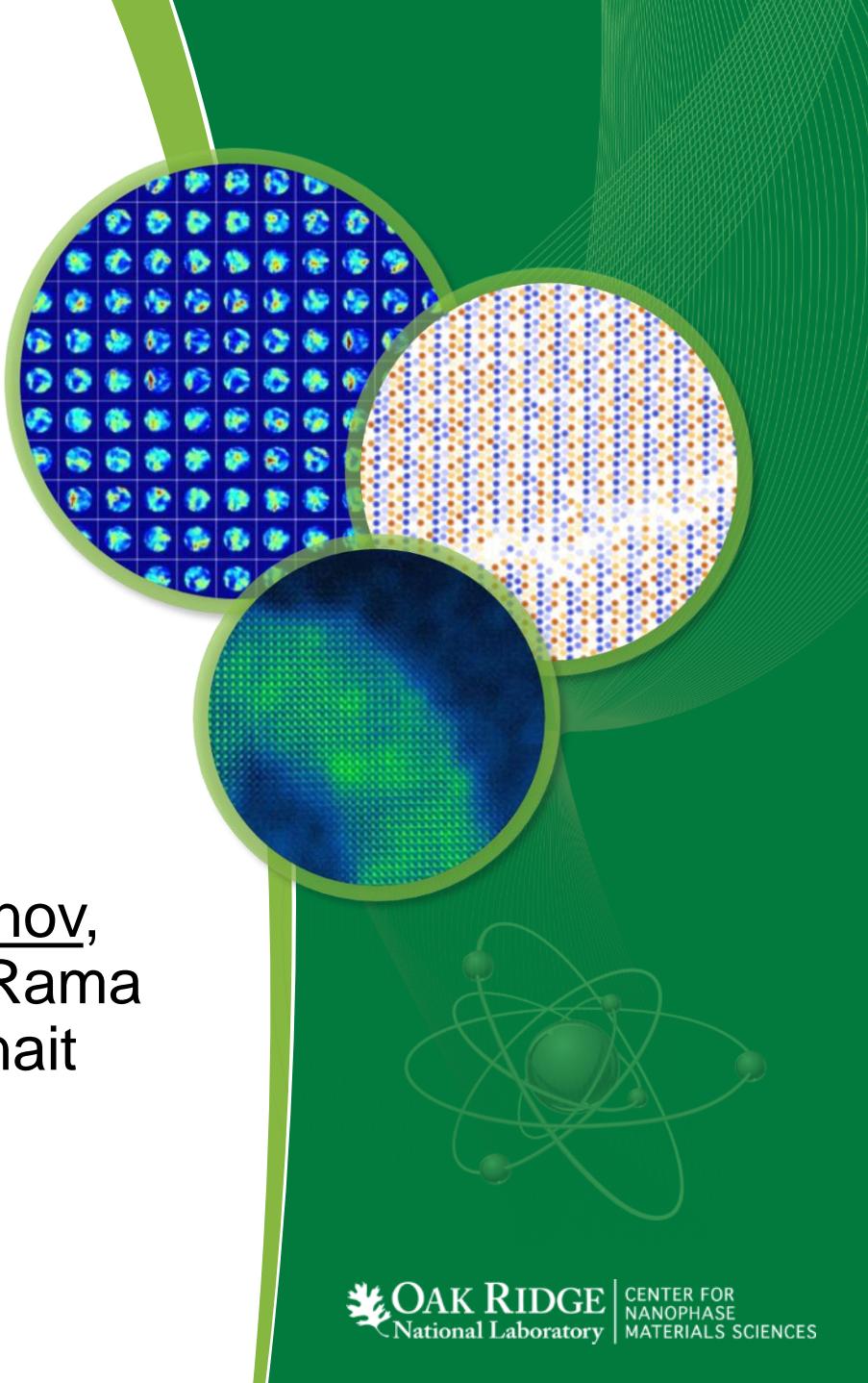


Processing of Atomic Resolution Images and Multispectral Data

Stephen Jesse, Alex Belianninov,
Suhas Somnath, Chris Smith, Rama
Vasudevan, Noumanne Laanait



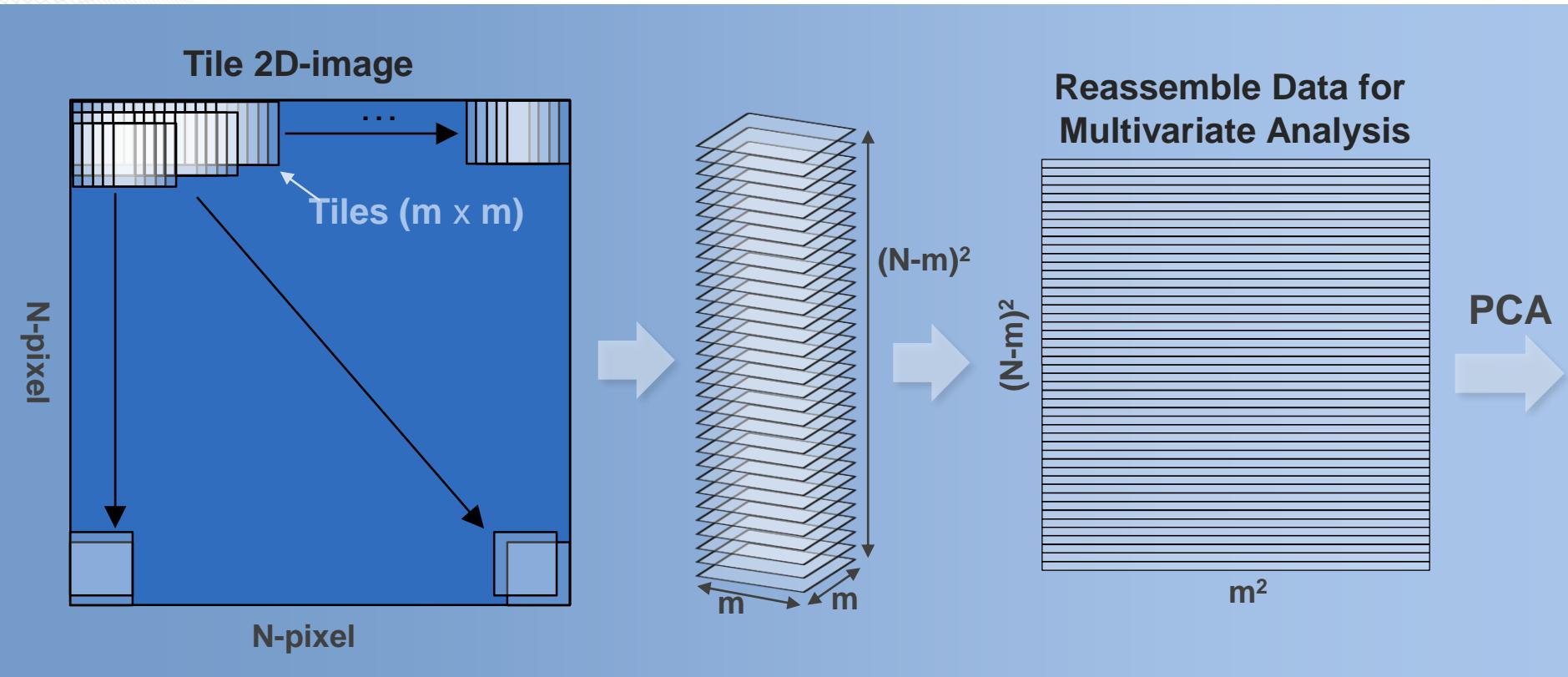
Outline

- Setting up images for multivariate analysis
- Multivariate analysis on imaging data
 - De-noising
 - PCA
 - K-means
 - Atom finding (pattern matching)
 - Analysis on found atoms
- 4D-STEM with multivariate analysis

Image Processing using SVD

*How can you express 2D images in a way that
you can use Multivariate Techniques?*

Image Cleaning: Sliding Window PCA



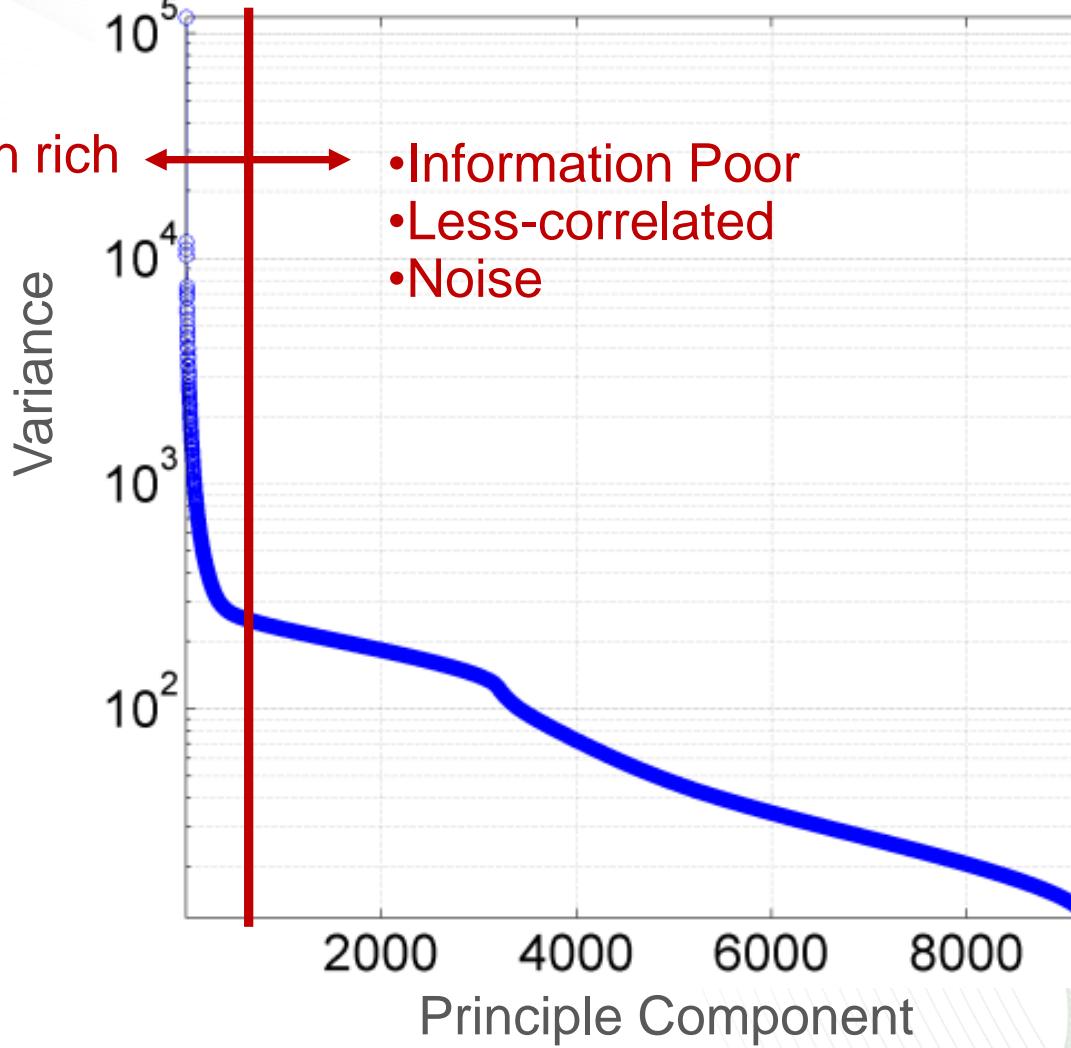
Principal Component Analysis (PCA)

$$D \rightarrow USV^*$$

Scree Plot -ranked significance of PC's

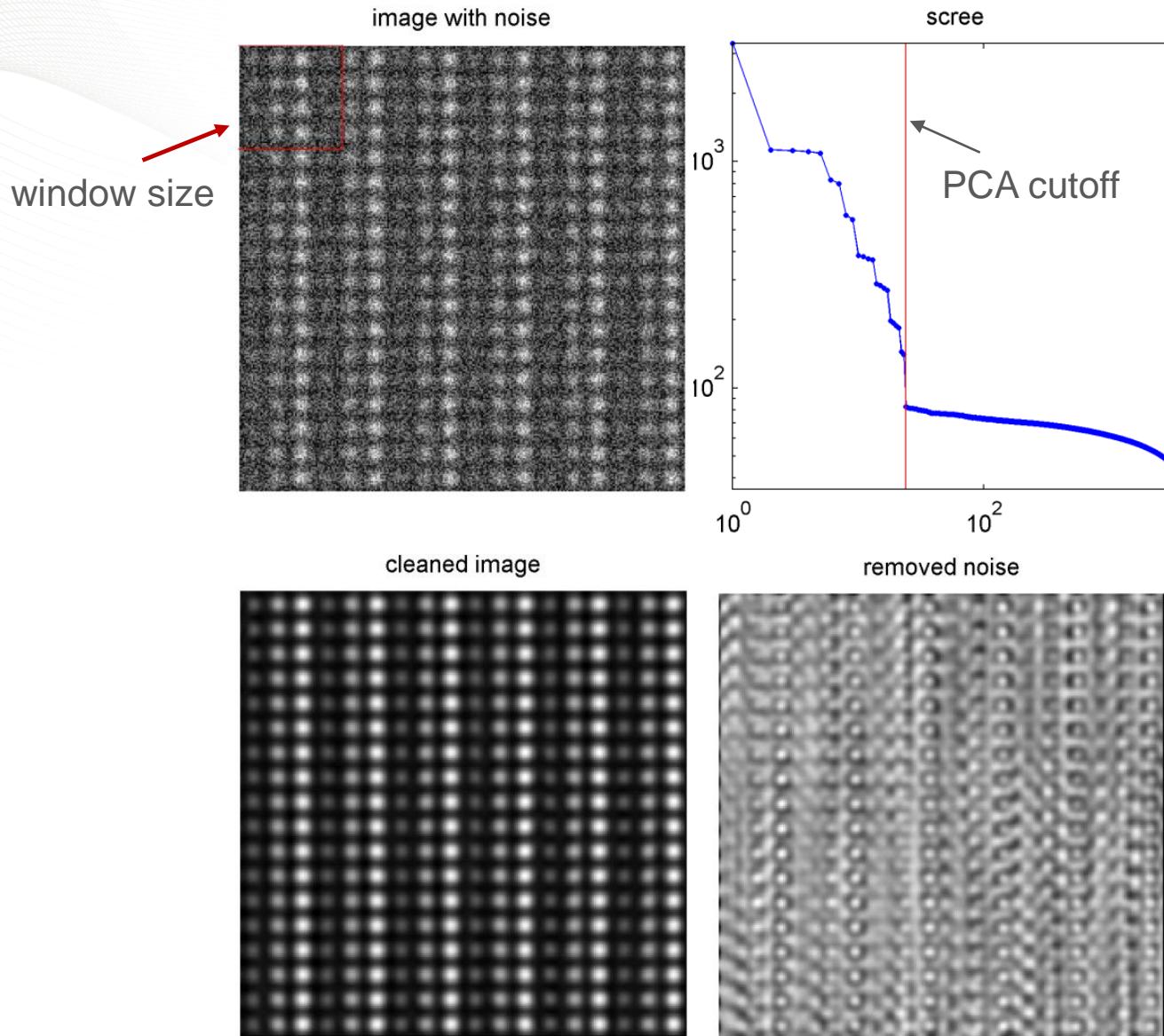
Information rich

- Information Poor
- Less-correlated
- Noise



Sliding Window PCA: De-noising

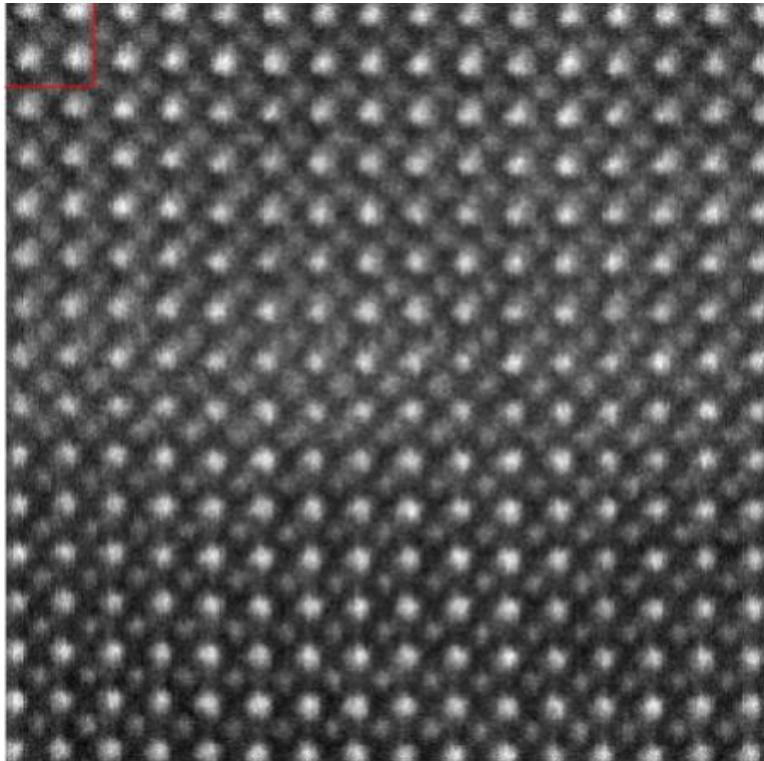
Model Image



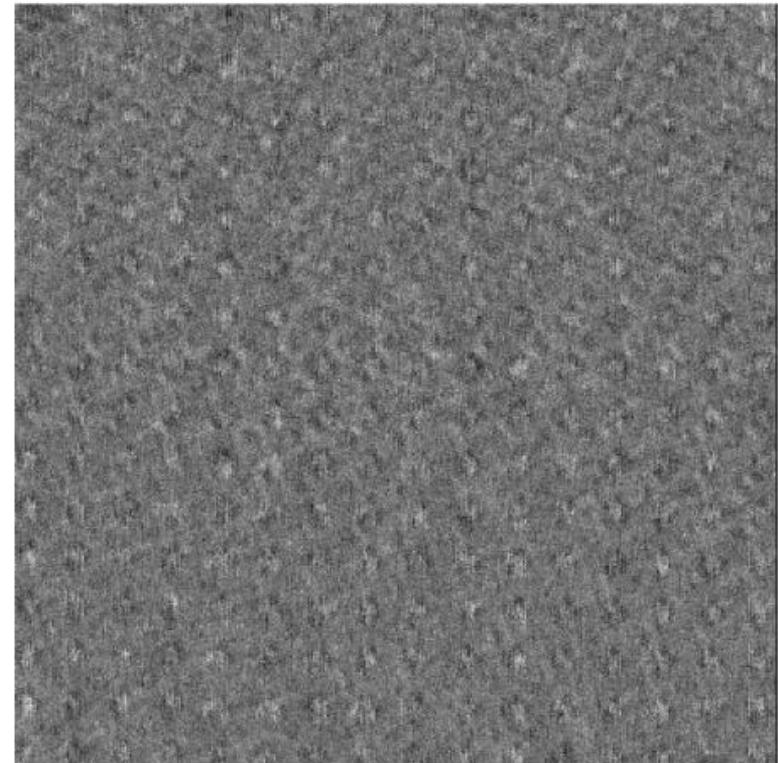
Sliding Window PCA: De-noising

Experimental Image

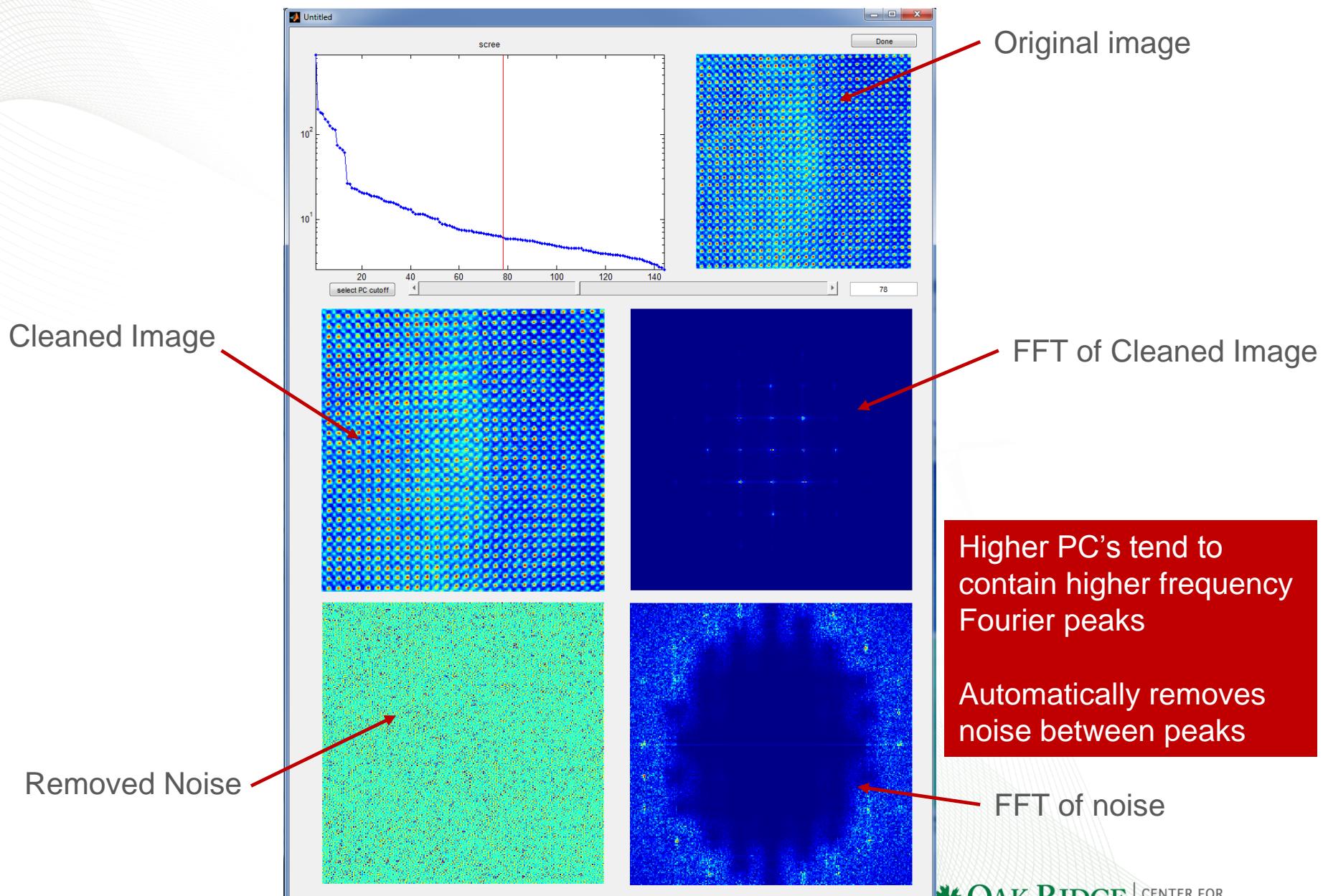
Original Image



Removed Noise

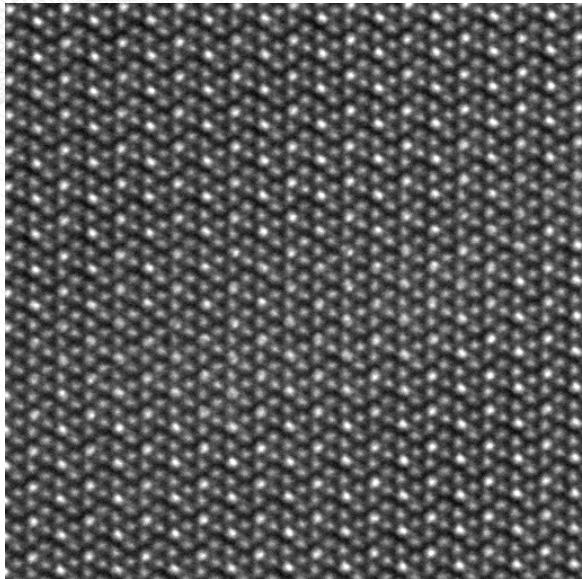


Effects of Number of Retained PC's

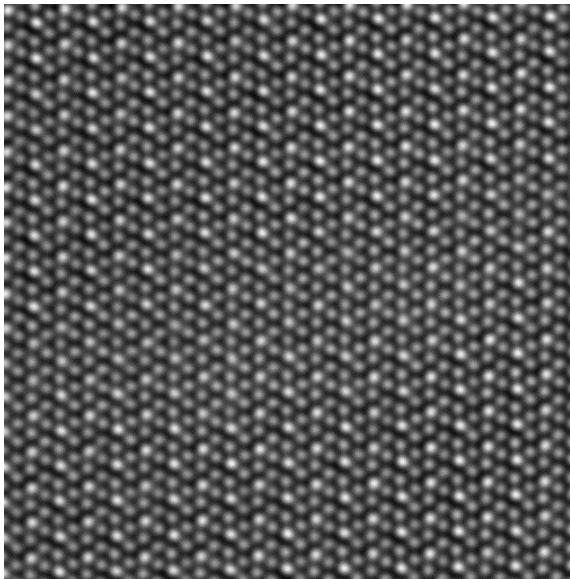


Sliding Window PCA-based Pattern Finding

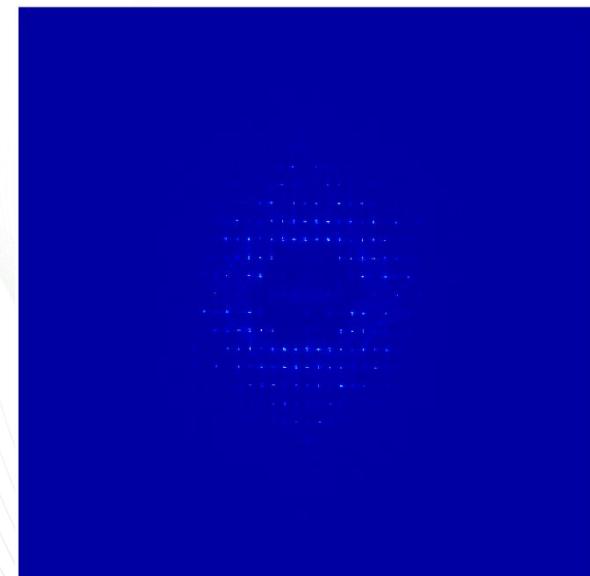
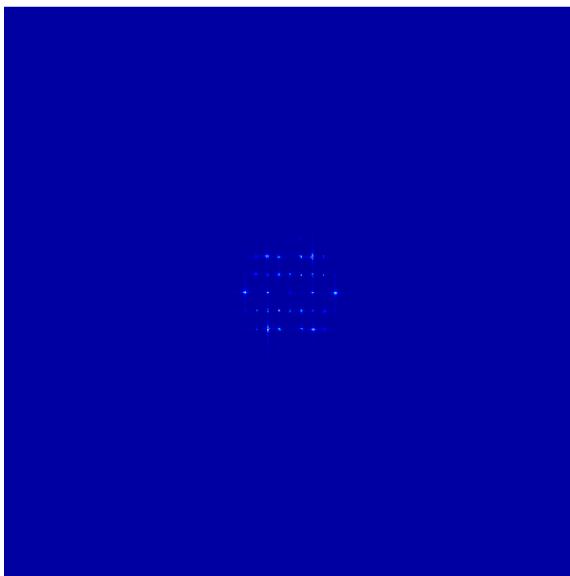
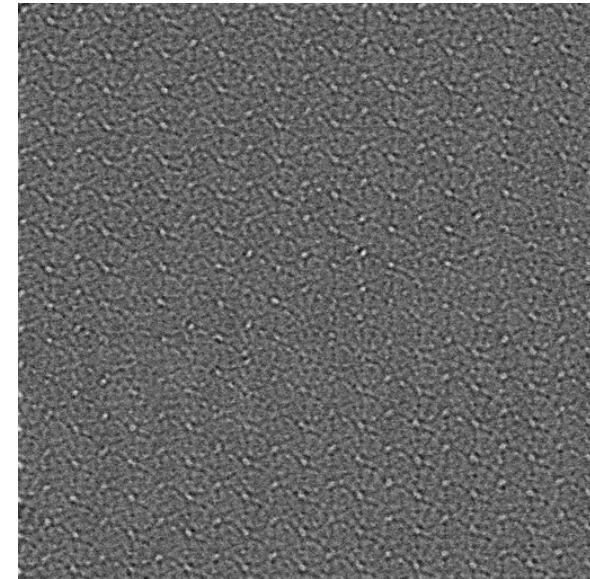
original



cleaned

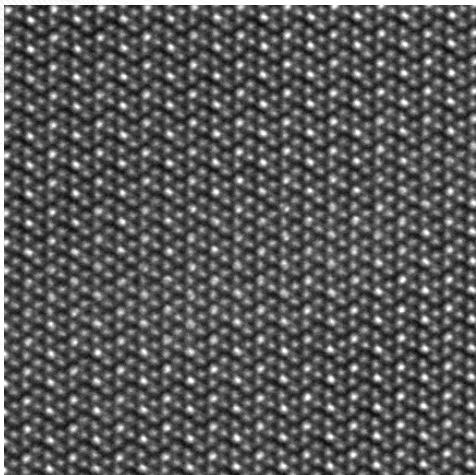


removed noise

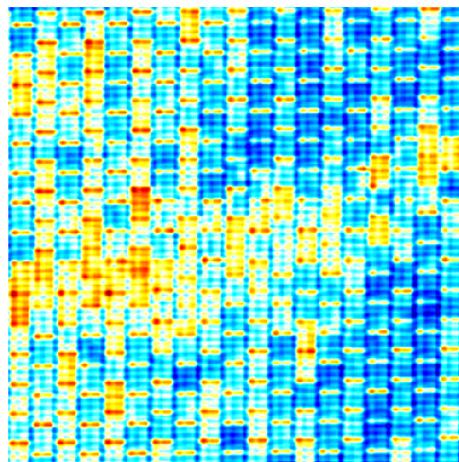


Sliding Window PCA-based Pattern Finding

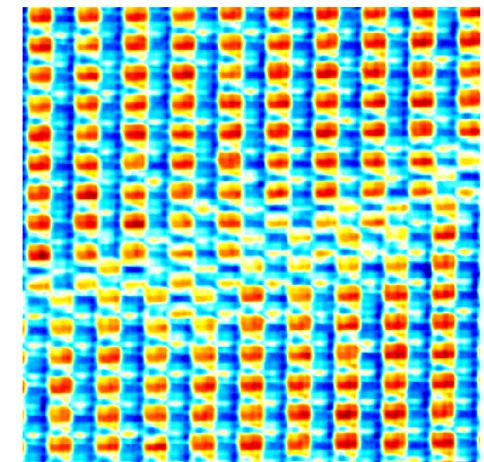
Original Image



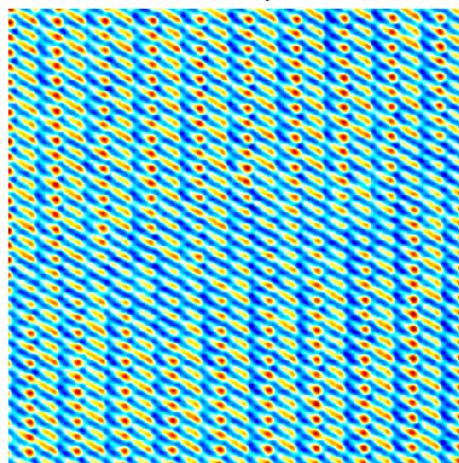
PC map 1



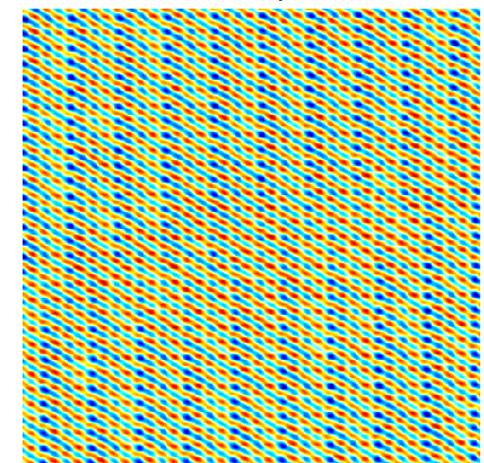
PC map 2



PC map 3



PC map 4



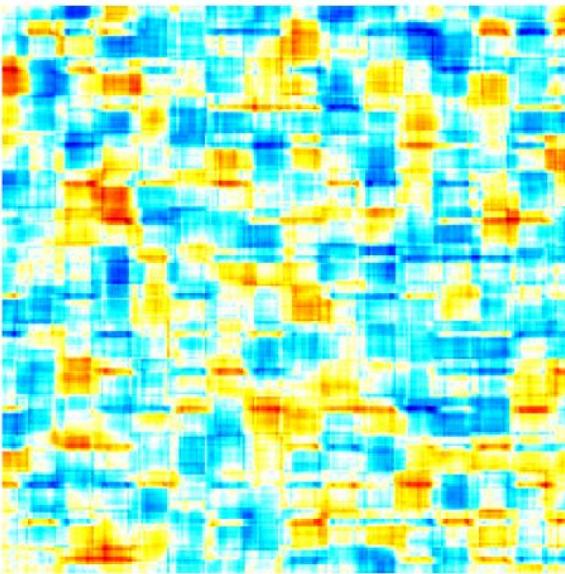
Sliding Window PCA can reveal patterns in images on its own without the need to identify features

Sliding Window PCA-based Pattern Finding 😊😊

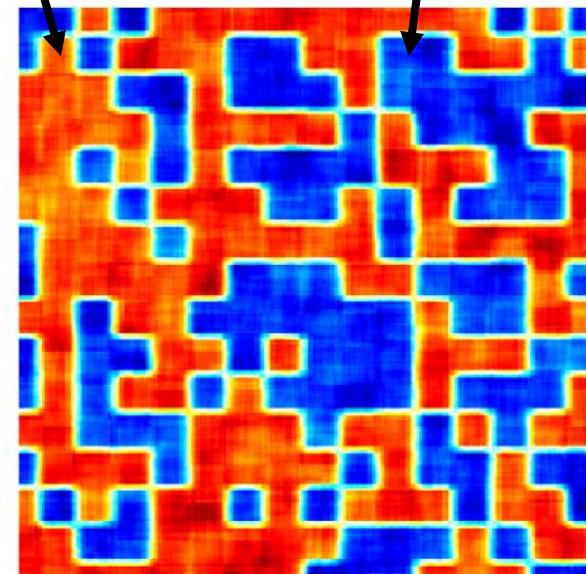
Start with a random grid
of smiley and frowny faces



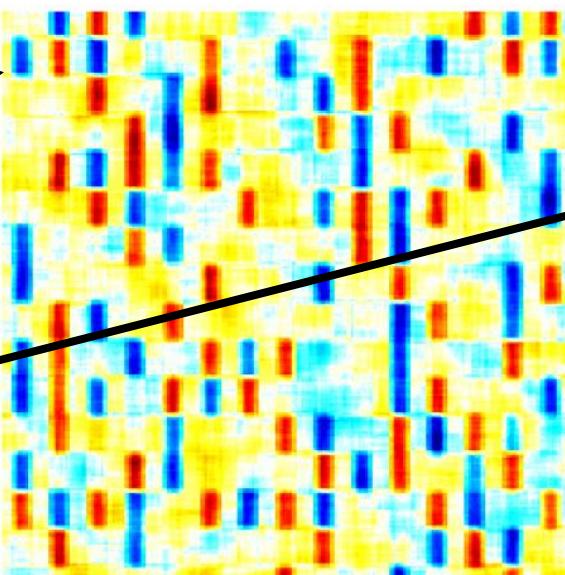
Sliding Window PCA-based Pattern Finding 😊😊



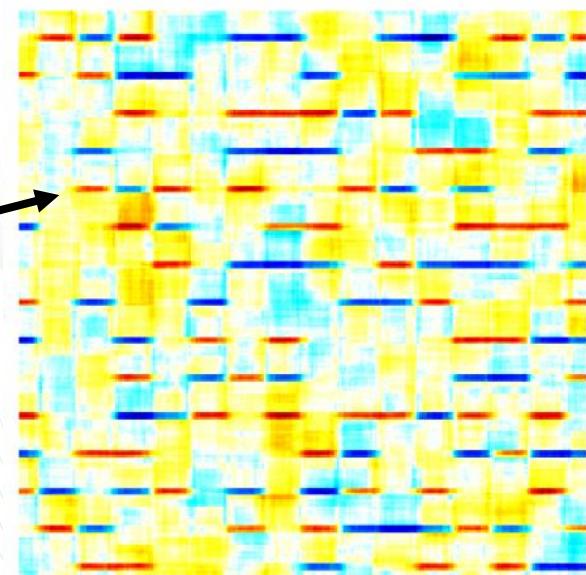
Happy face (red) Sad face (blue)



Horizontal Transitions
(changes in emotion)

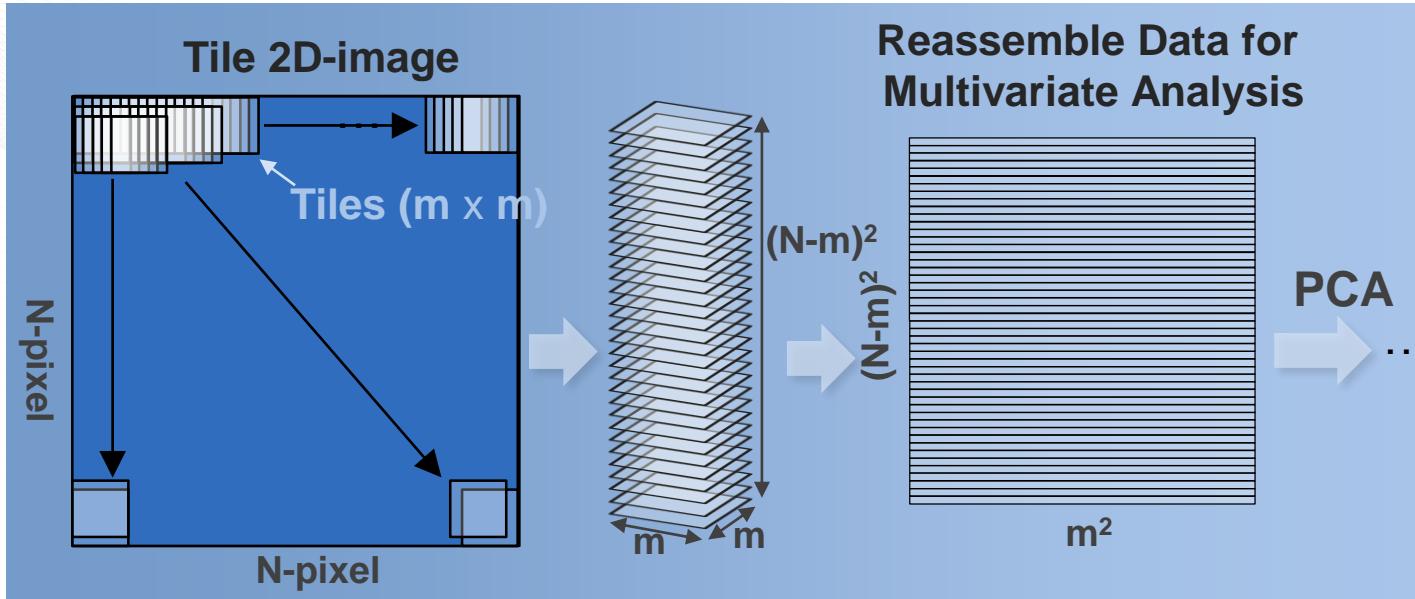


Vertical Transitions

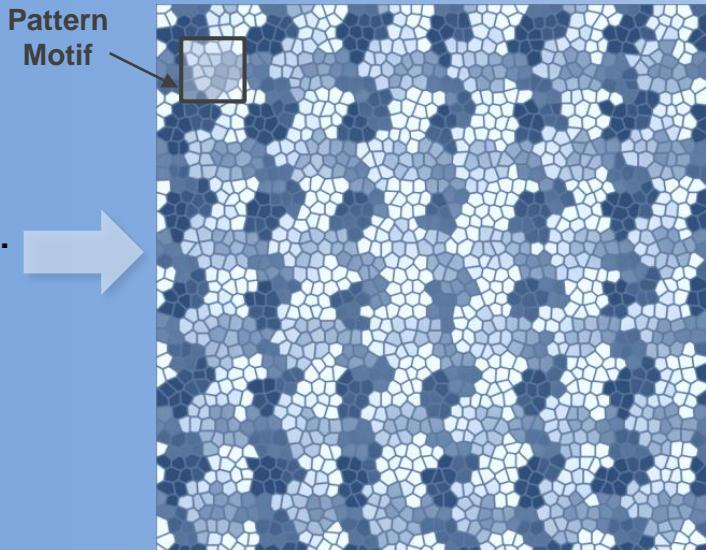


Clustering and Pattern Matching for Atom Finding

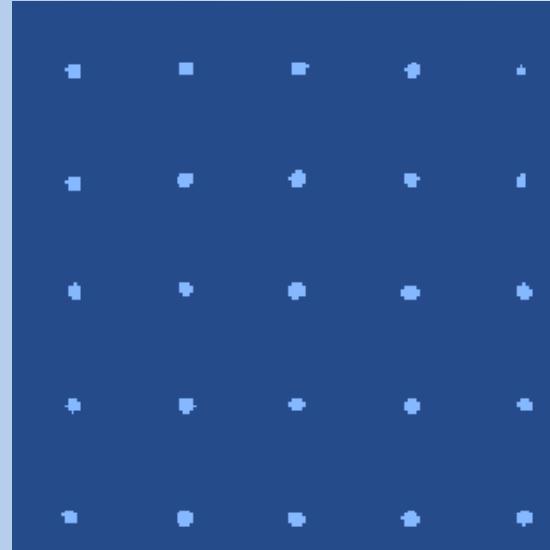
Sliding Window PCA + Clustering



Cluster on PCA components

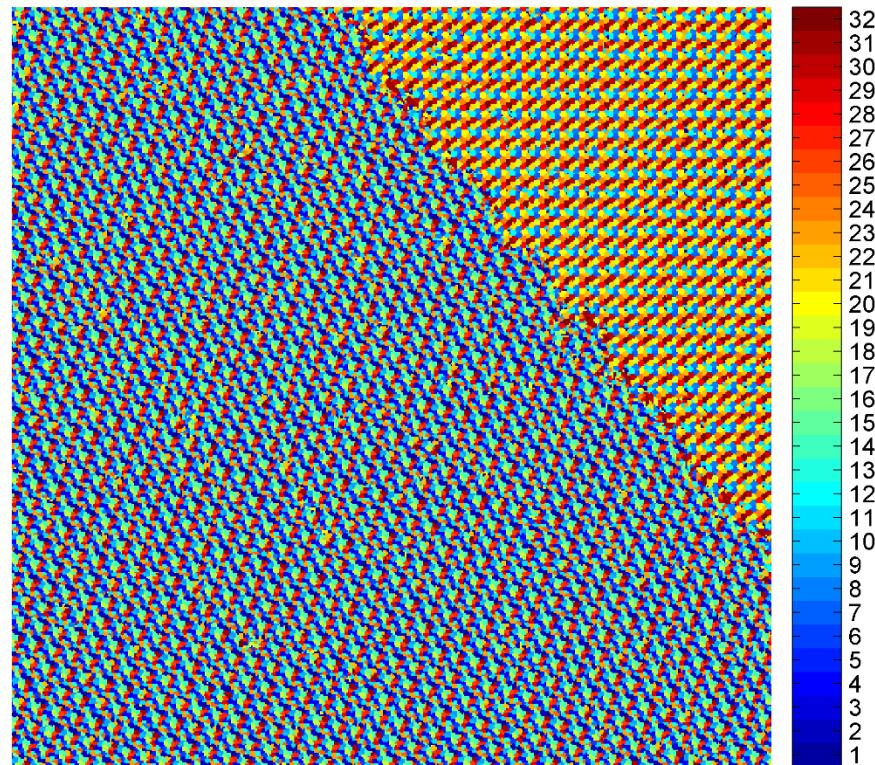
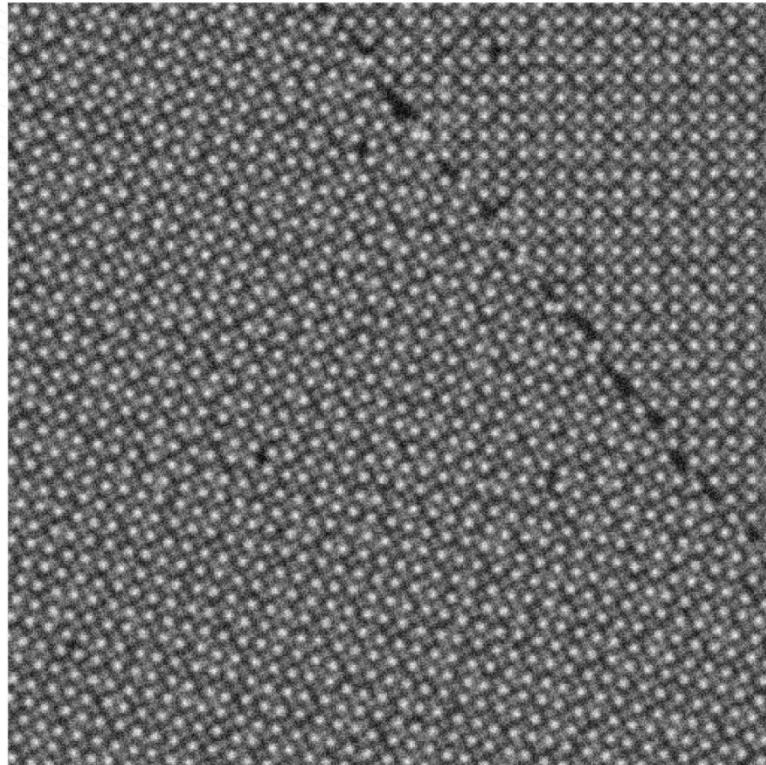


Motif Match on Cluster Map

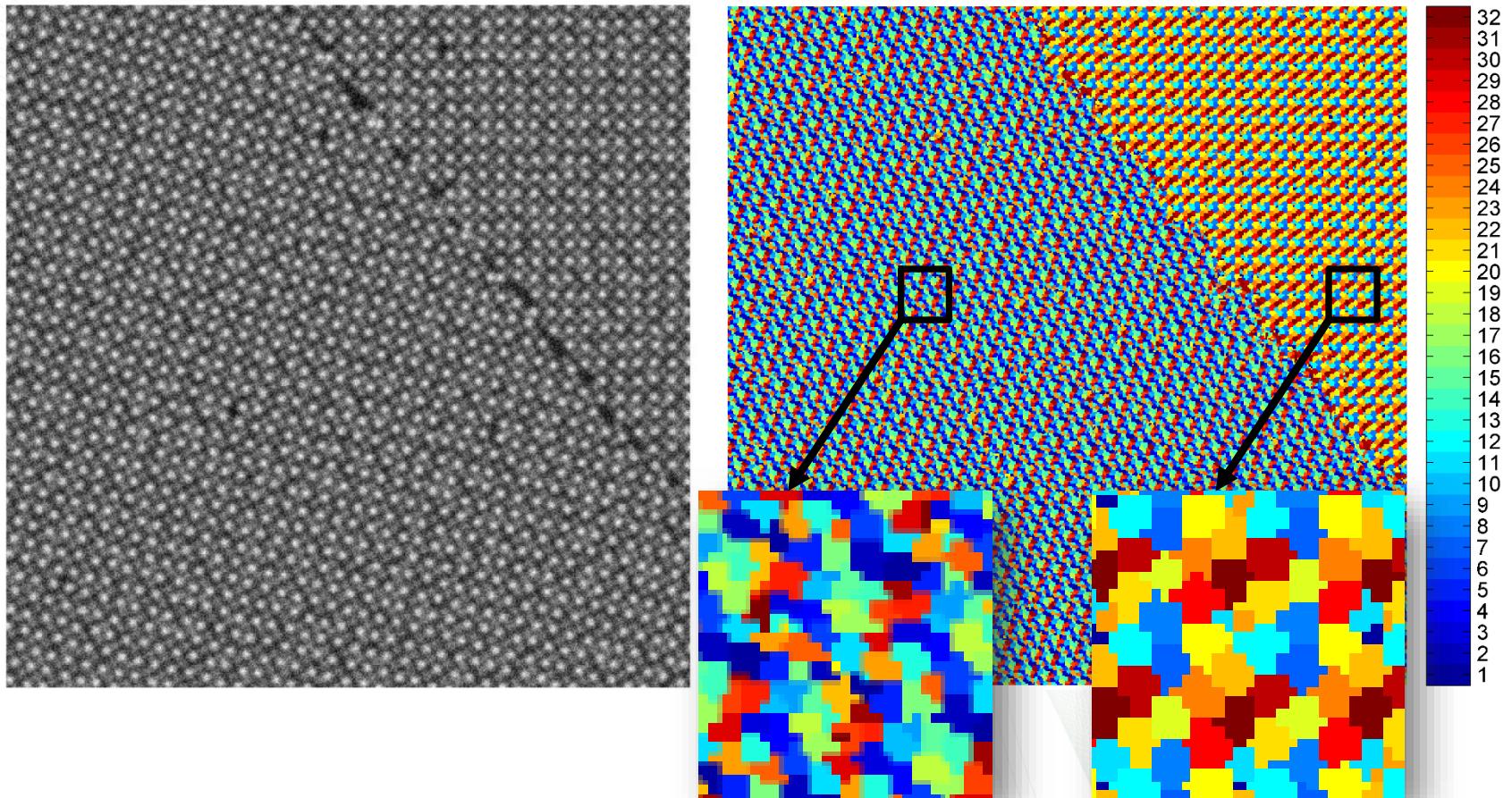


Cluster analysis

Perform clustering on PCA weighting vector from every pixel

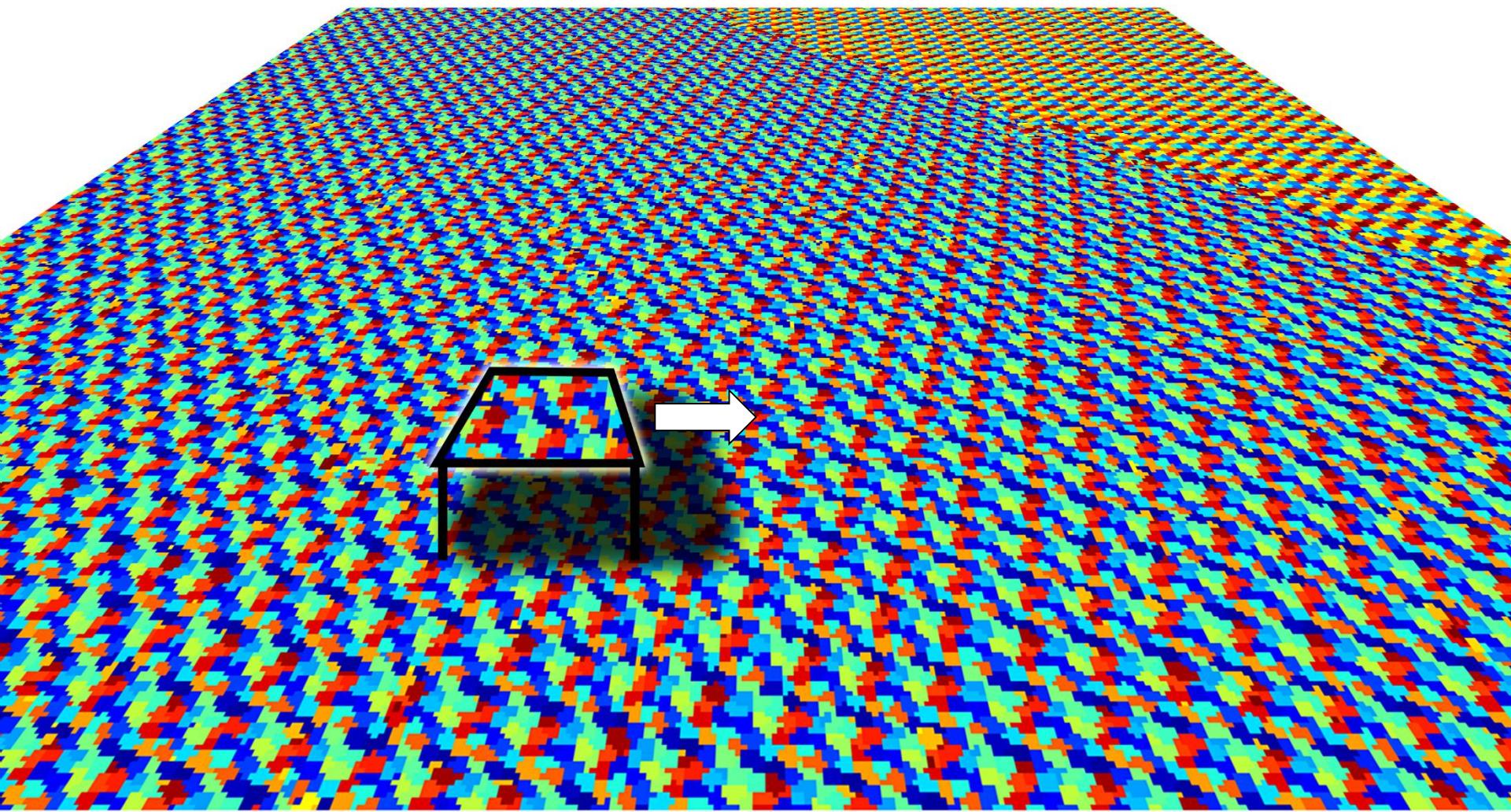


Cluster analysis

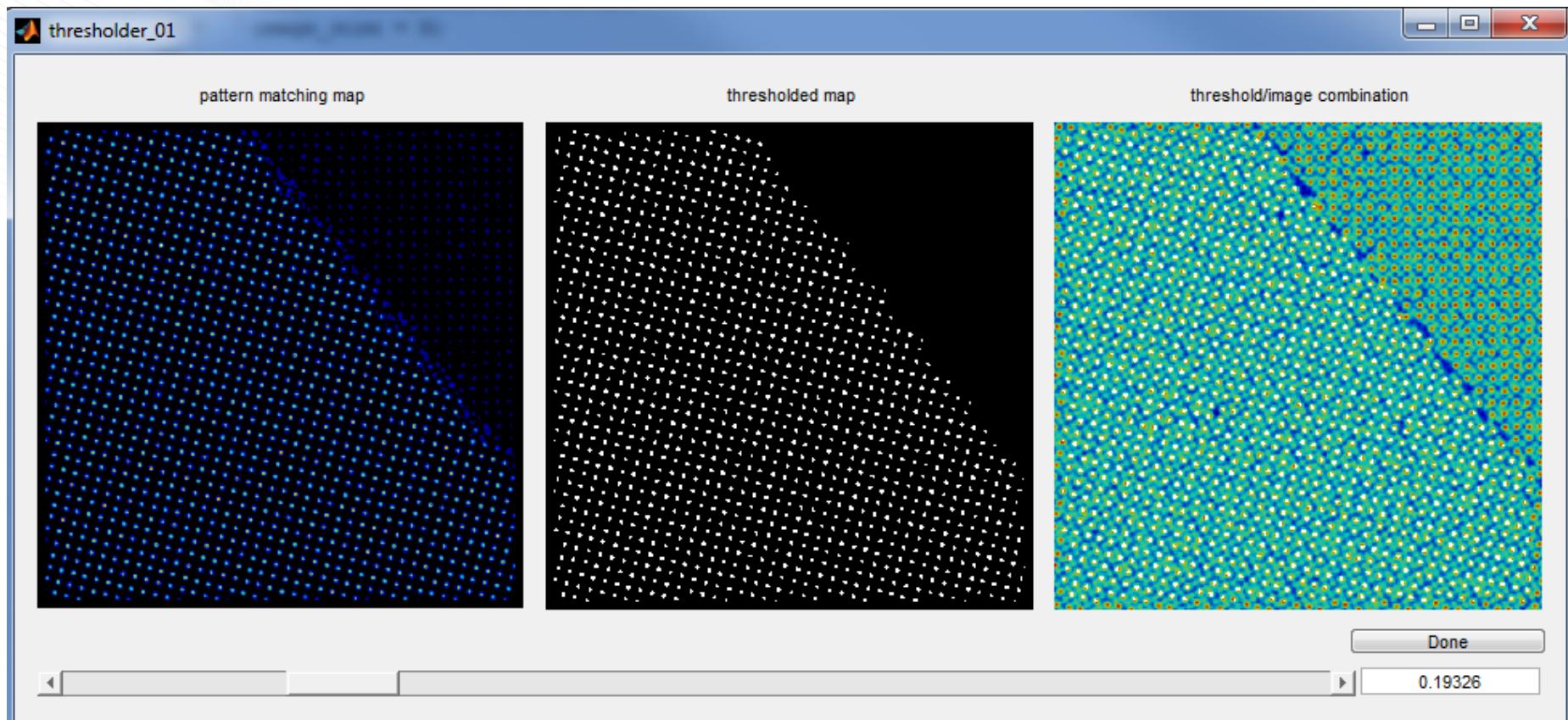


We can use local motifs to locate similar atomic arrangements across an image

Motif Matching

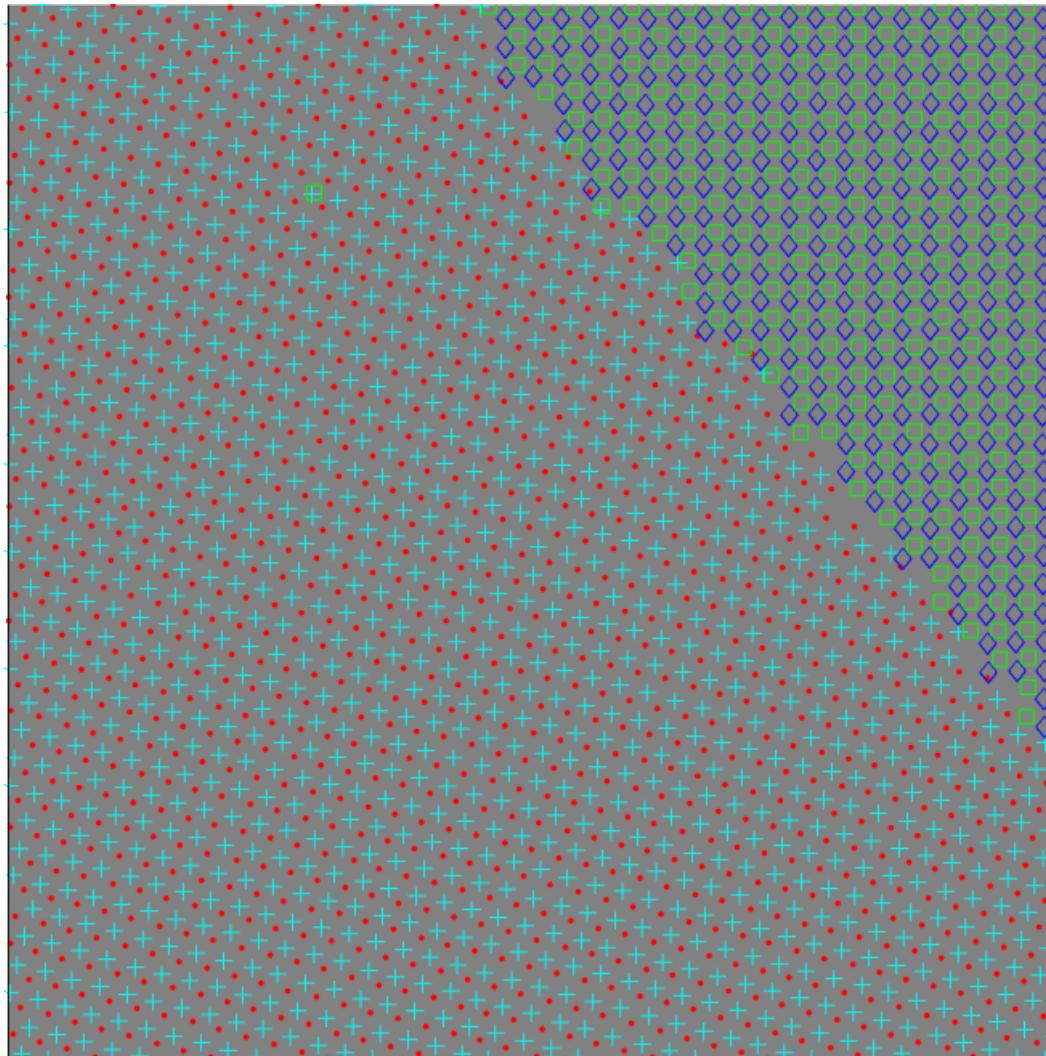


Motif Matching



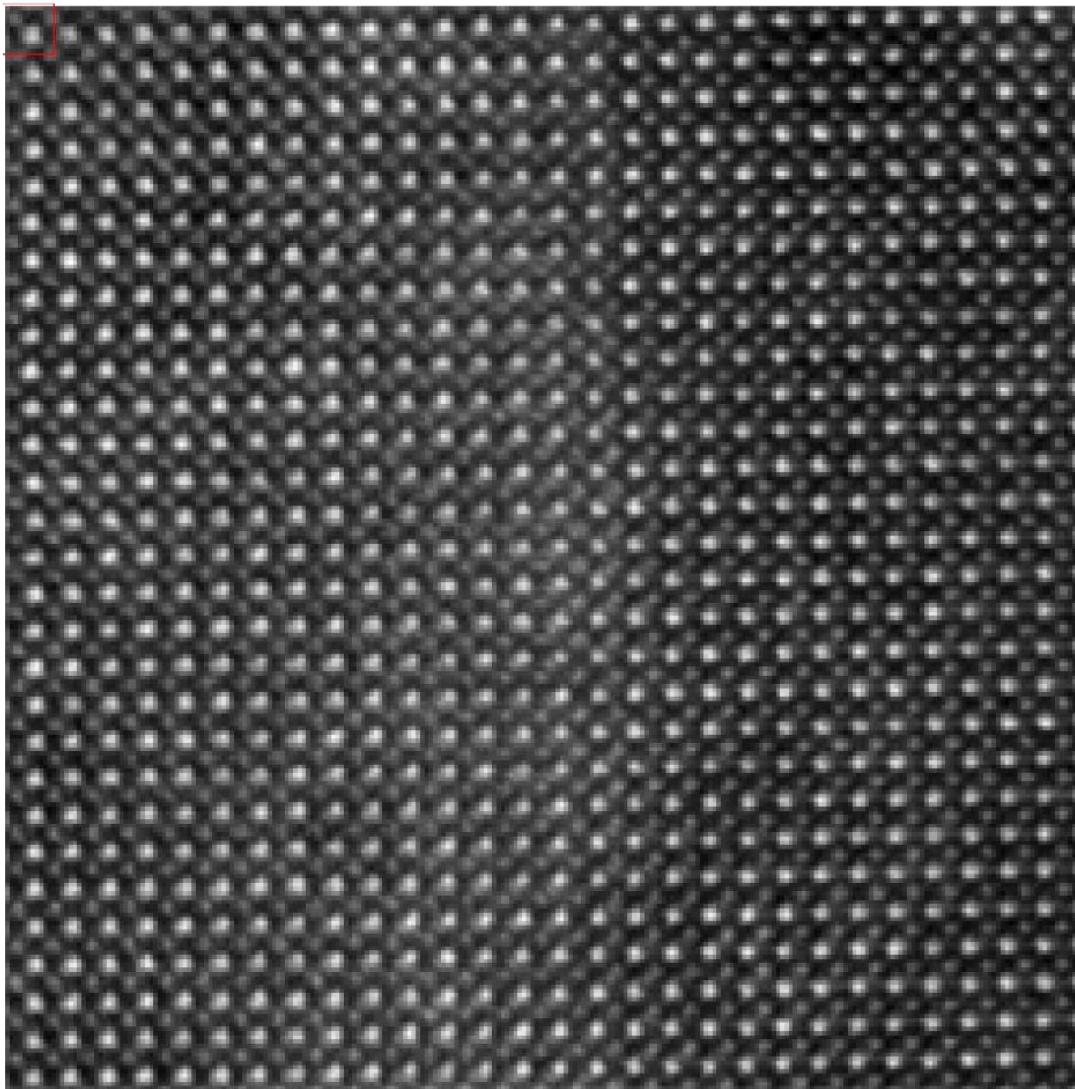
Sliding Window PCA + Clustering

After motif matching...



Sliding Window PCA + Clustering

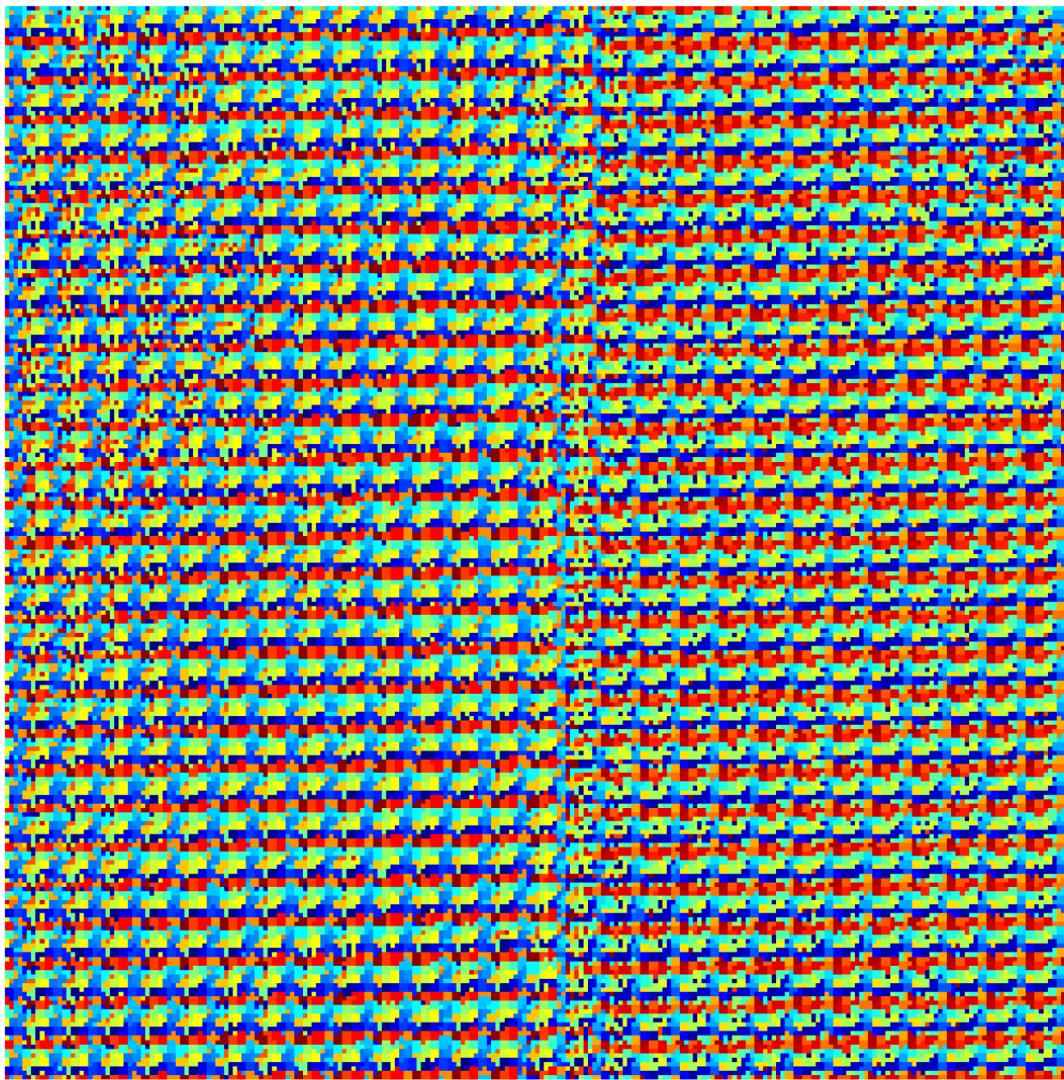
original image



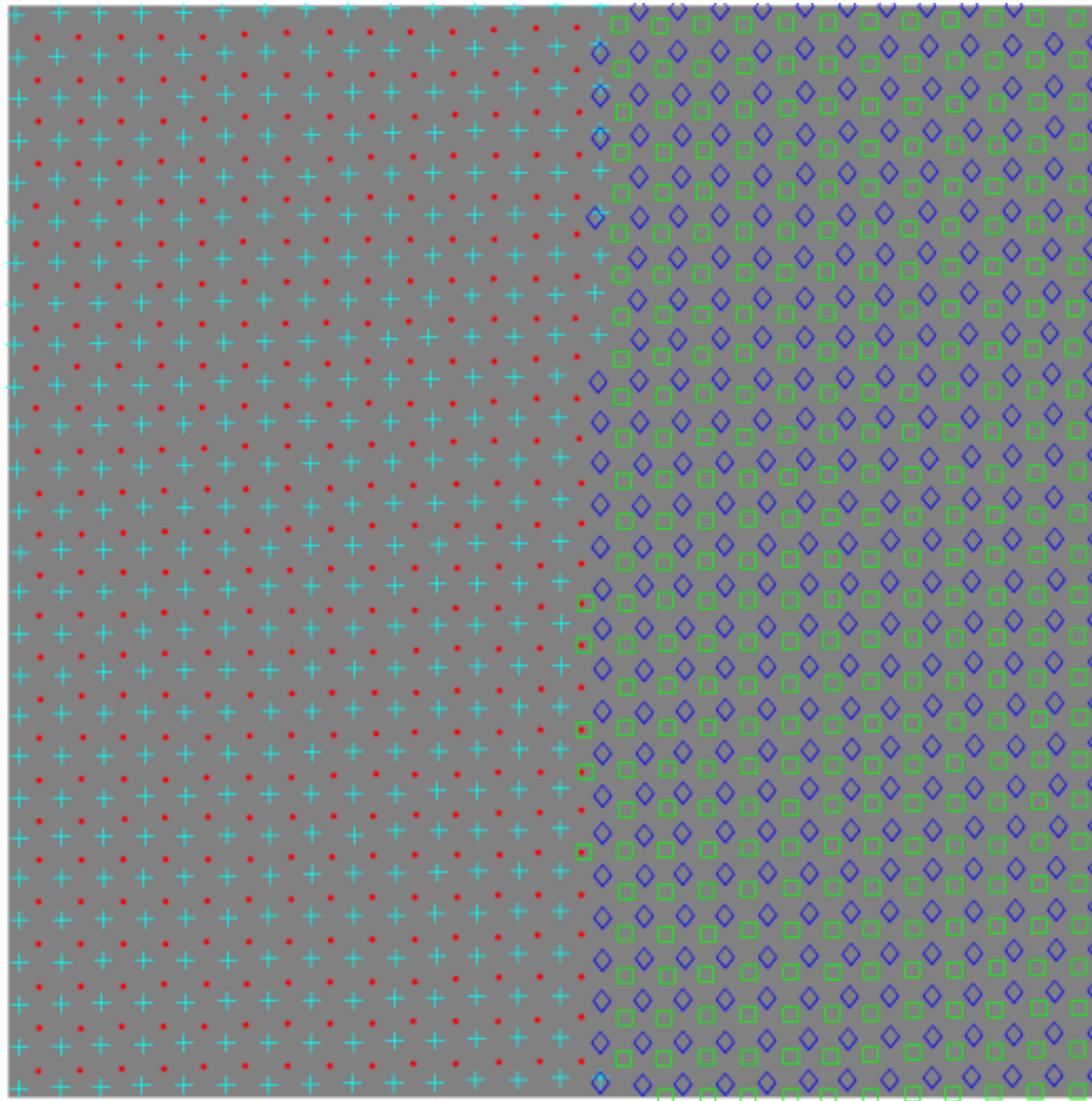
Sliding Window PCA + Clustering

Clustering of the first 32 PC's

36 Clusters

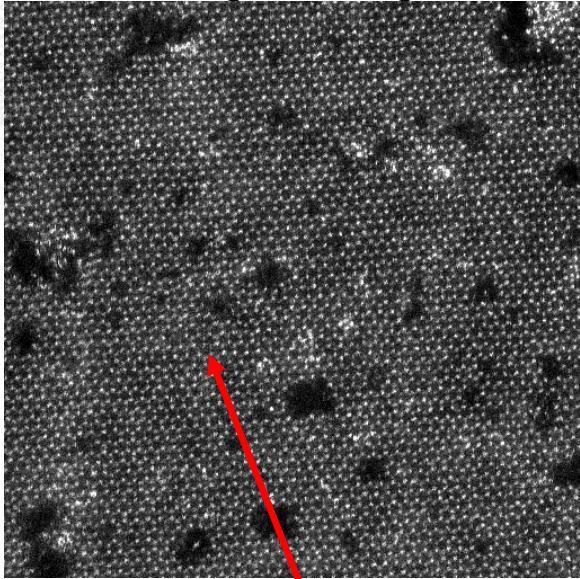


Sliding Window PCA + Clustering

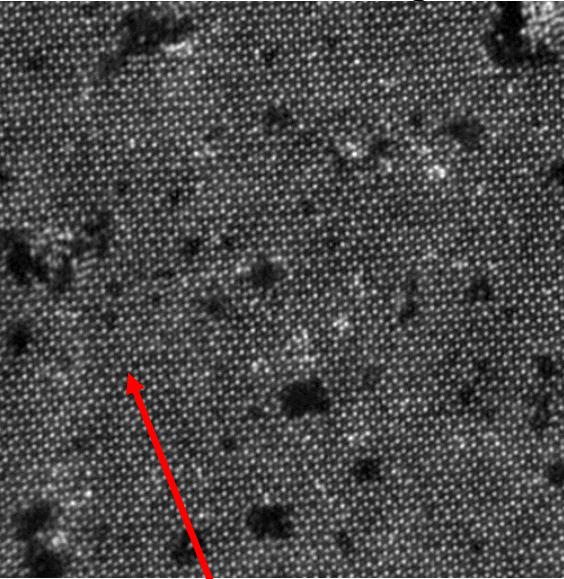


Sliding Window PCA + Clustering

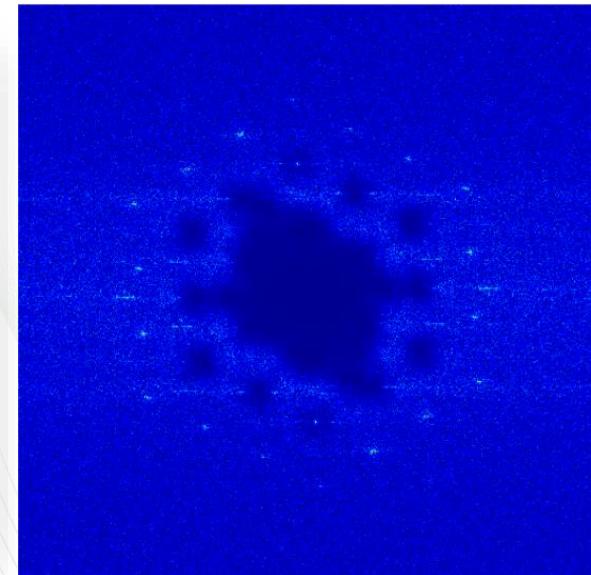
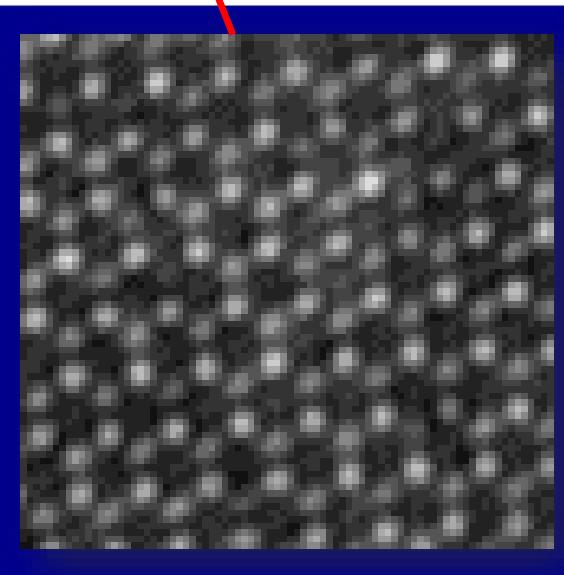
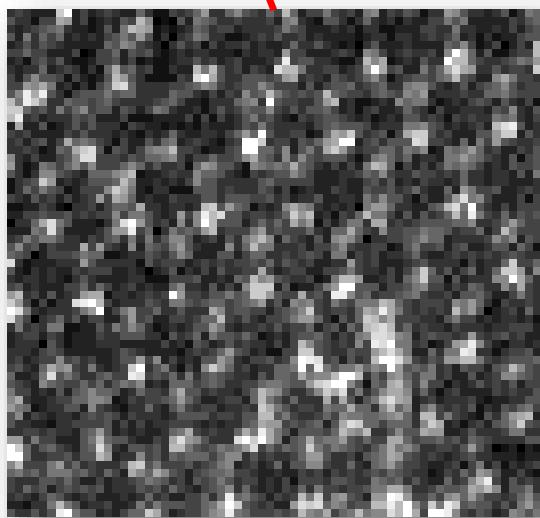
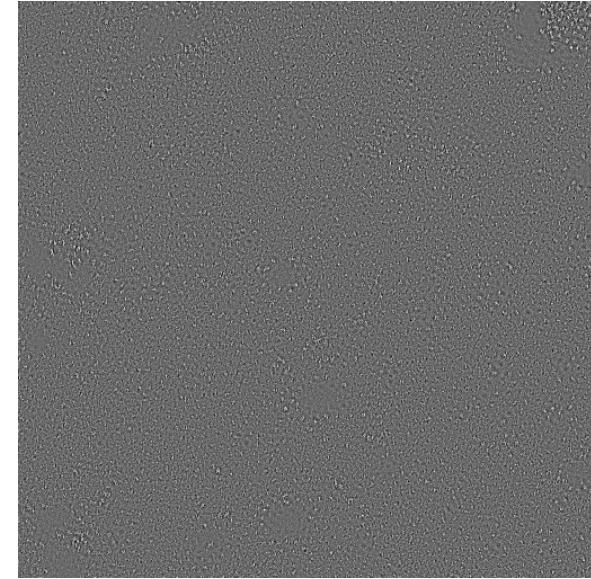
Original image



Denoised image

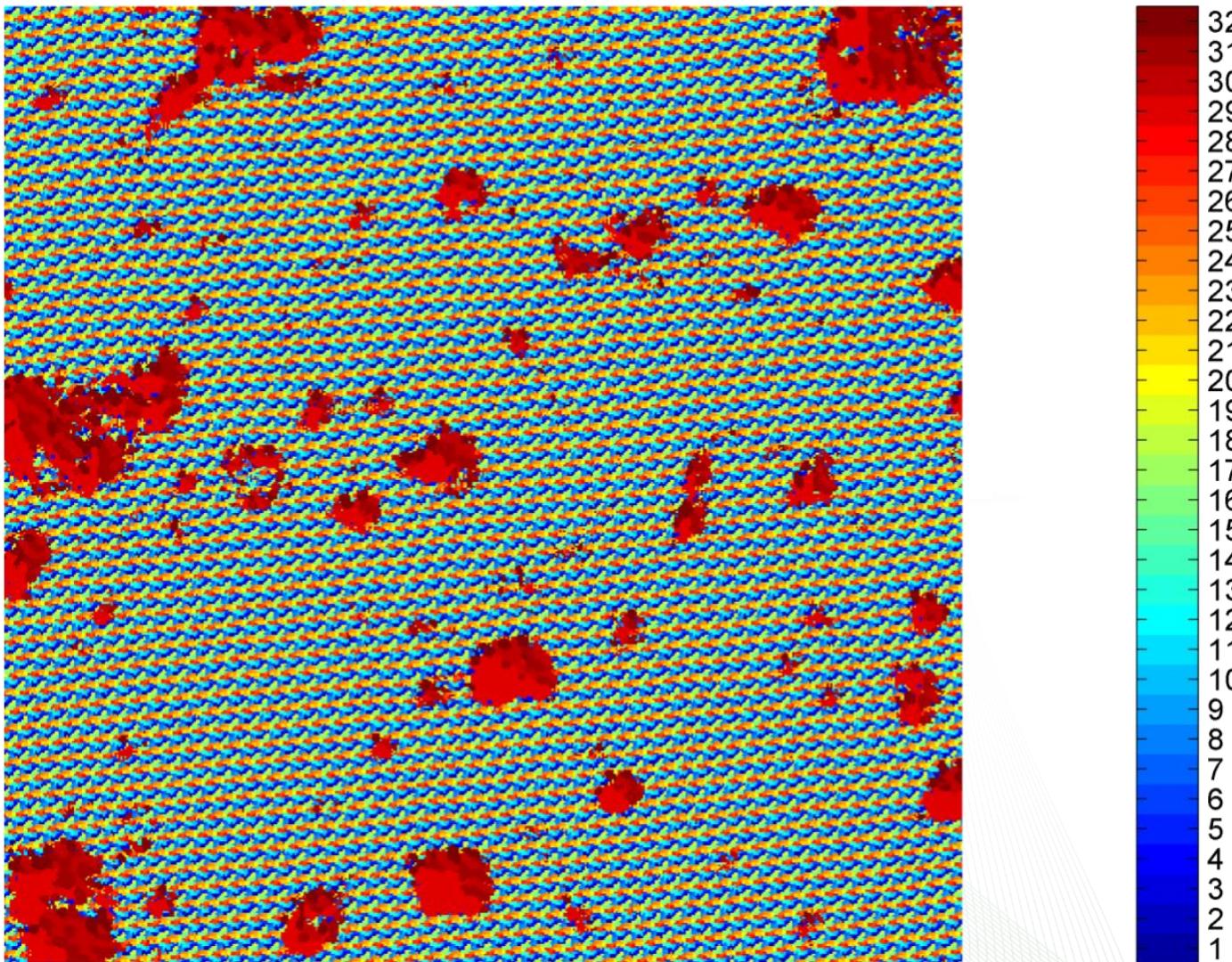


Removed noise



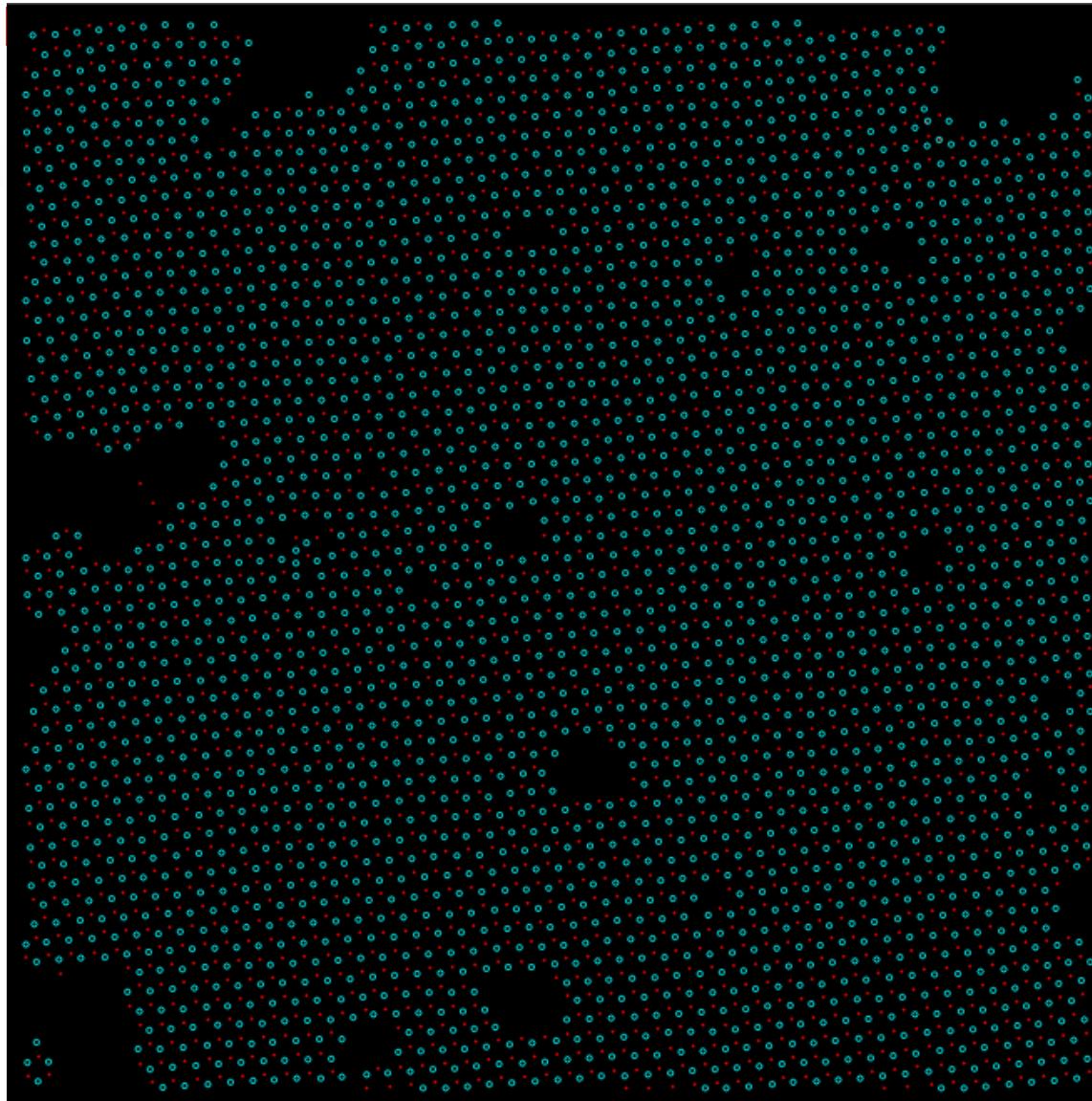
Sliding Window PCA + Clustering

32 Clusters



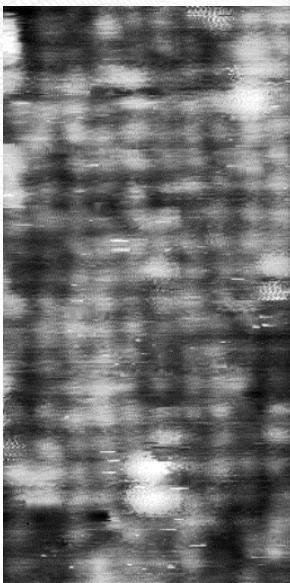
Sliding Window PCA + Clustering

Found atoms

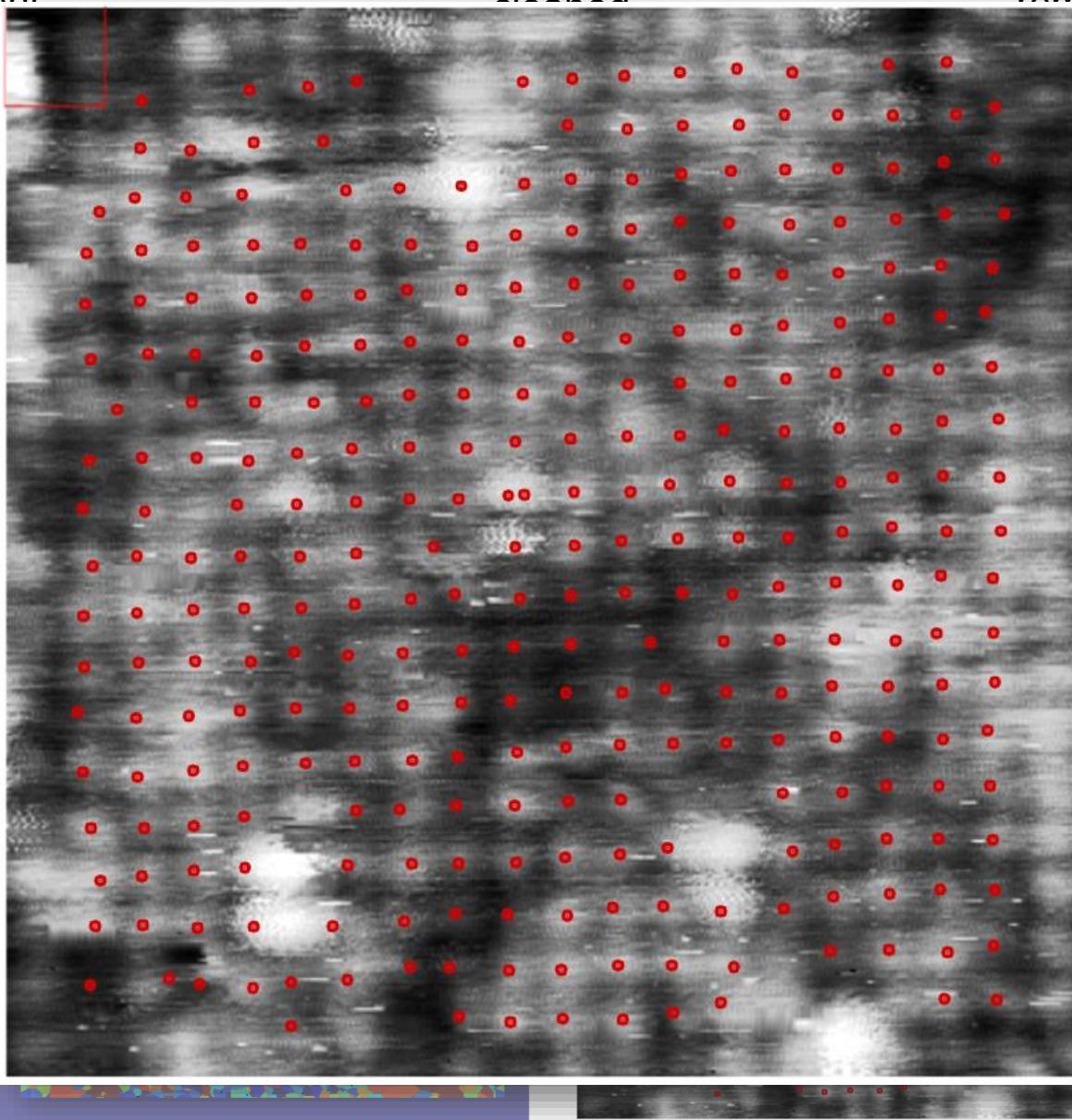


Sliding Window PCA + Clustering

original



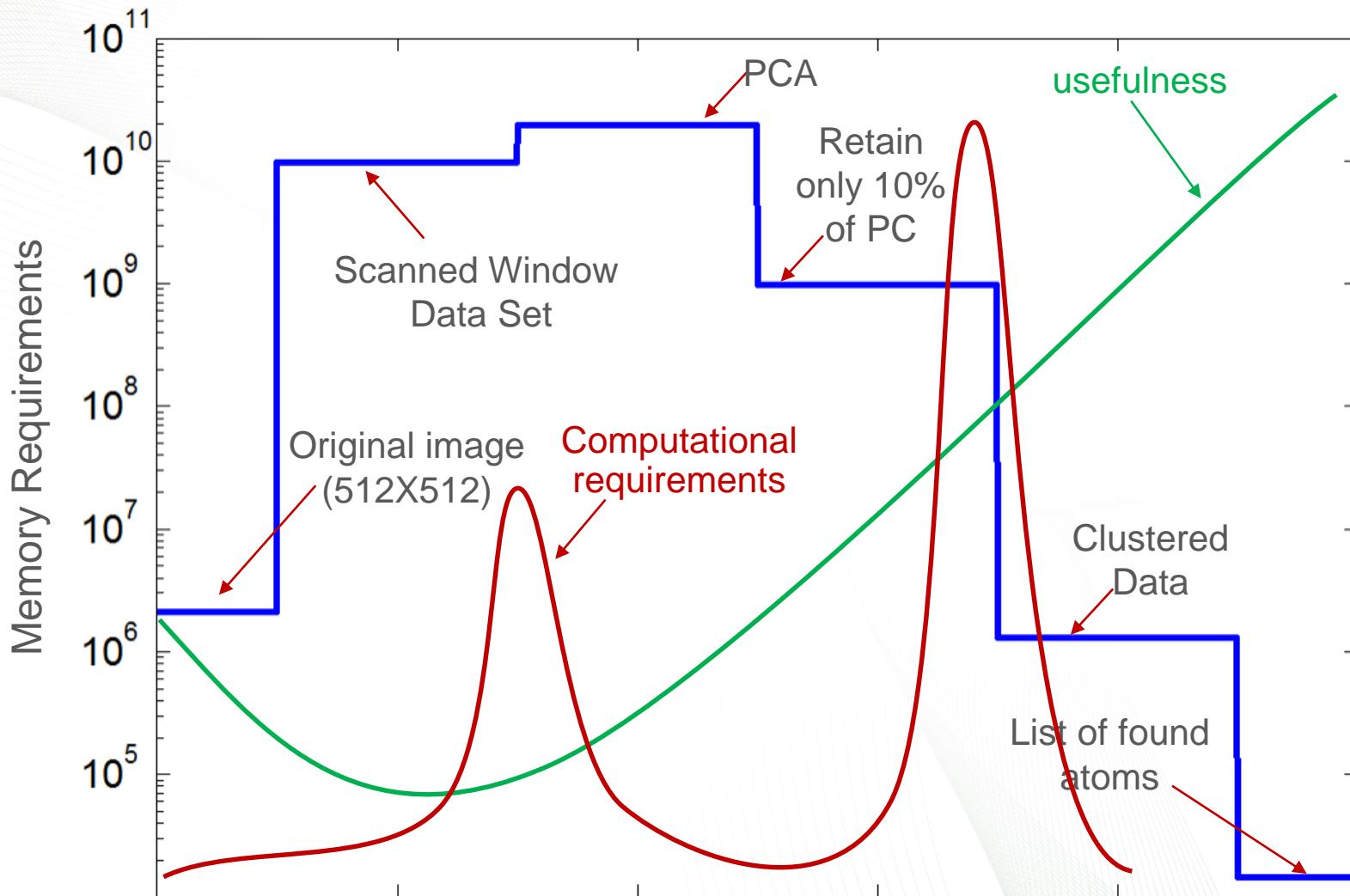
cleaned



removed noise



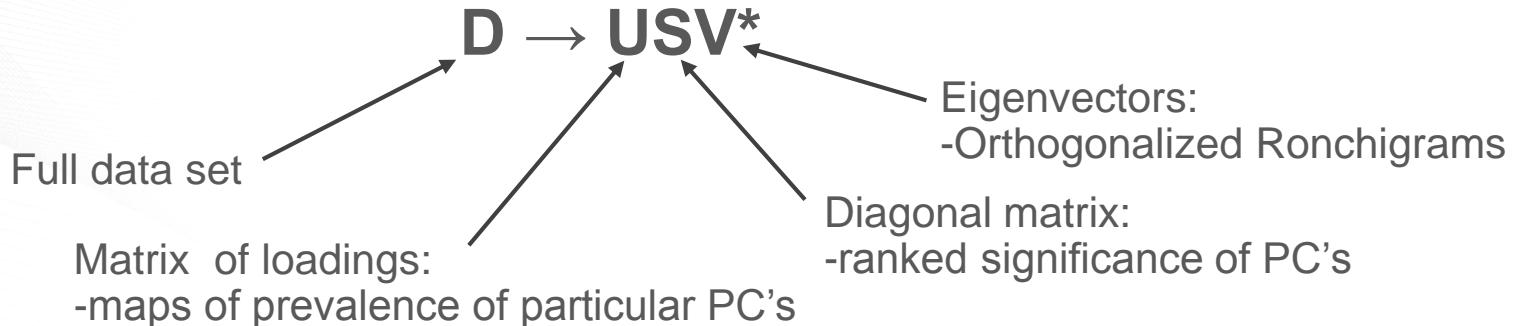
Sliding Window PCA + Clustering as Big Data



Alternate atom finding K-means on found atom positions

Multivariate Analysis of 4D STEM Data

Principal Component Analysis (PCA)



- Compares every individual observable with every other observable
- Find the most common behavior
- Find the next most information rich (orthogonal) variation on the above behavior
- Find the next most information rich (orthogonal) variation on the above behavior
- ... loop

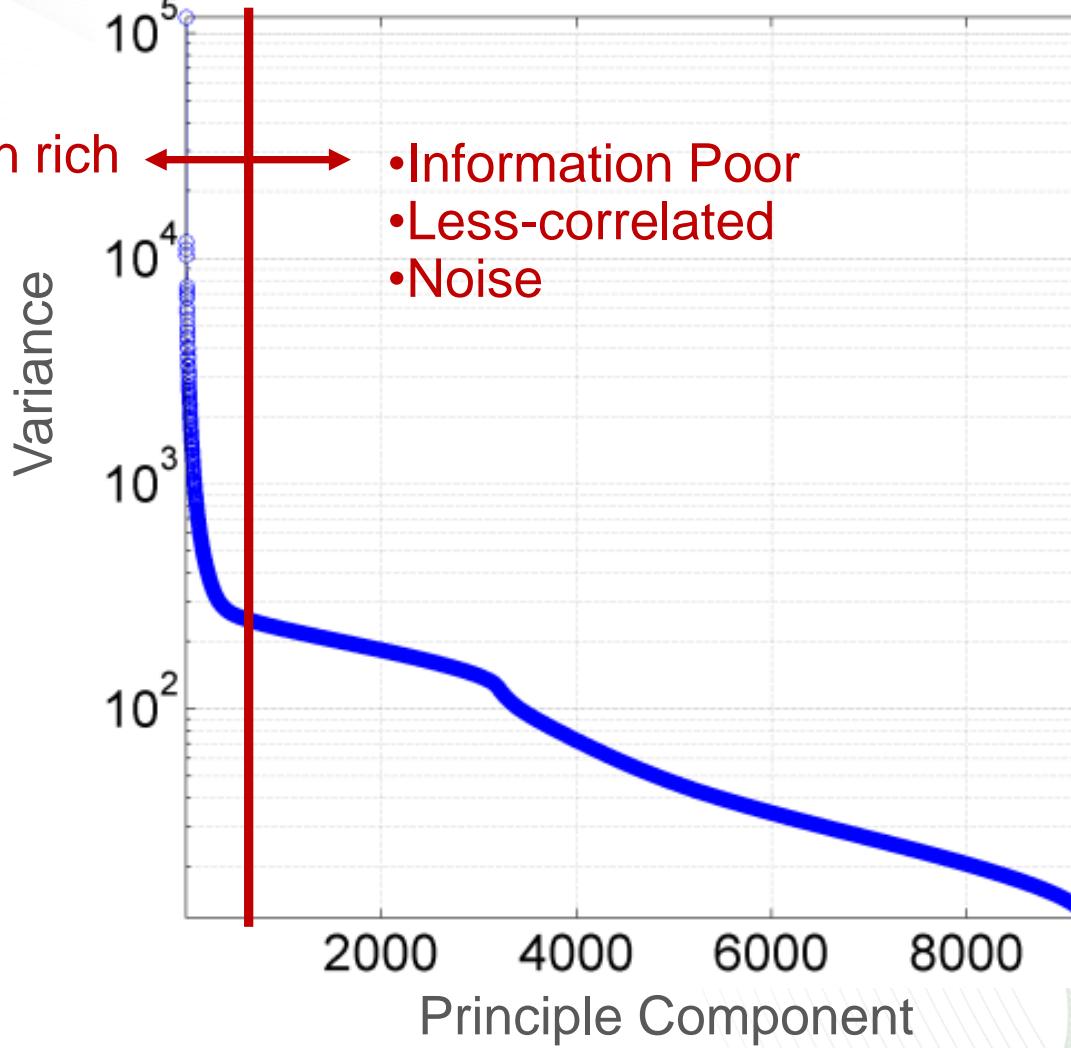
Principal Component Analysis (PCA)

$$D \rightarrow USV^*$$

Scree Plot -ranked significance of PC's

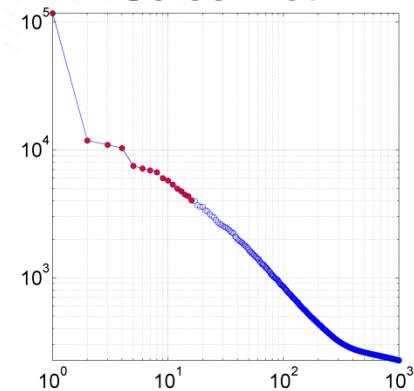
Information rich

- Information Poor
- Less-correlated
- Noise

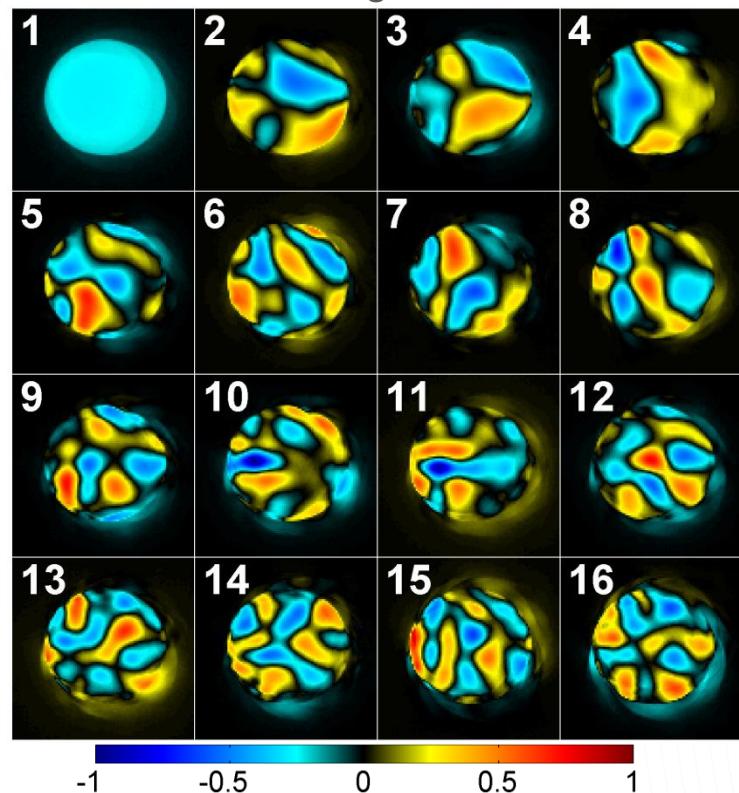


Principal Component Analysis (PCA)

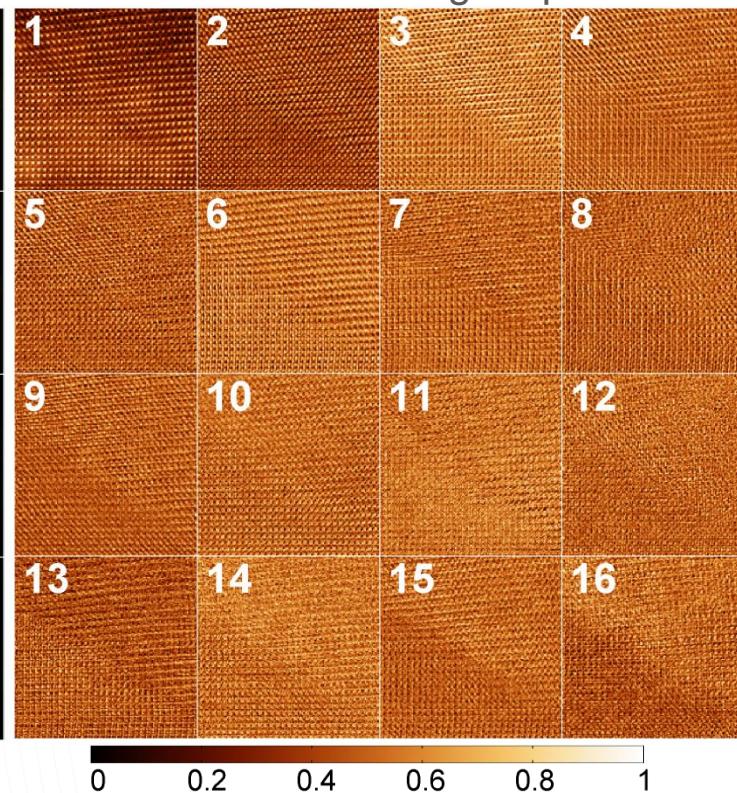
Scree Plot



First 16 eigenvectors

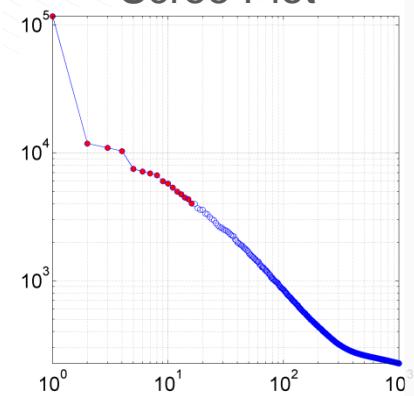


First 16 loading maps

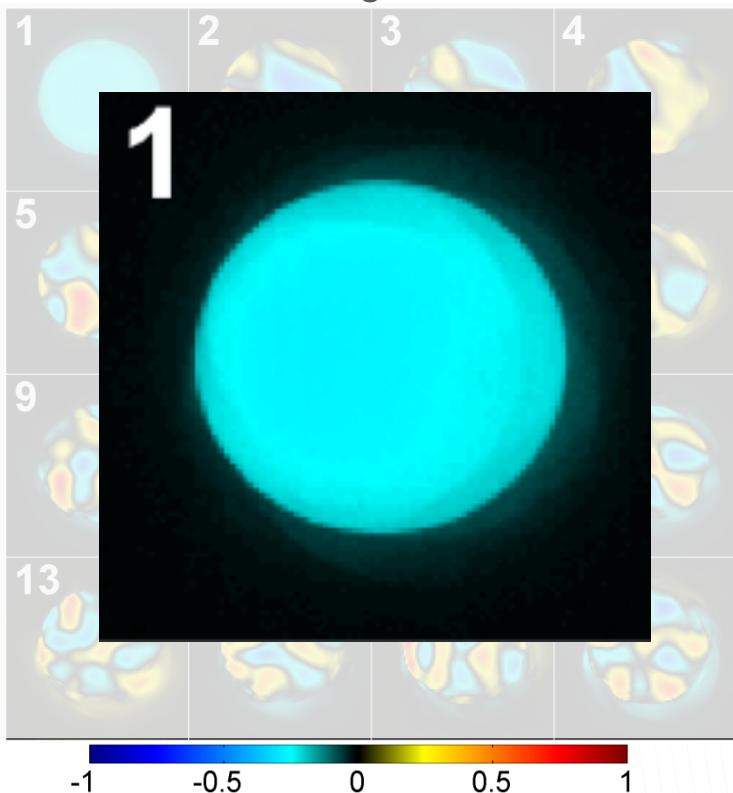


Principal Component Analysis (PCA)

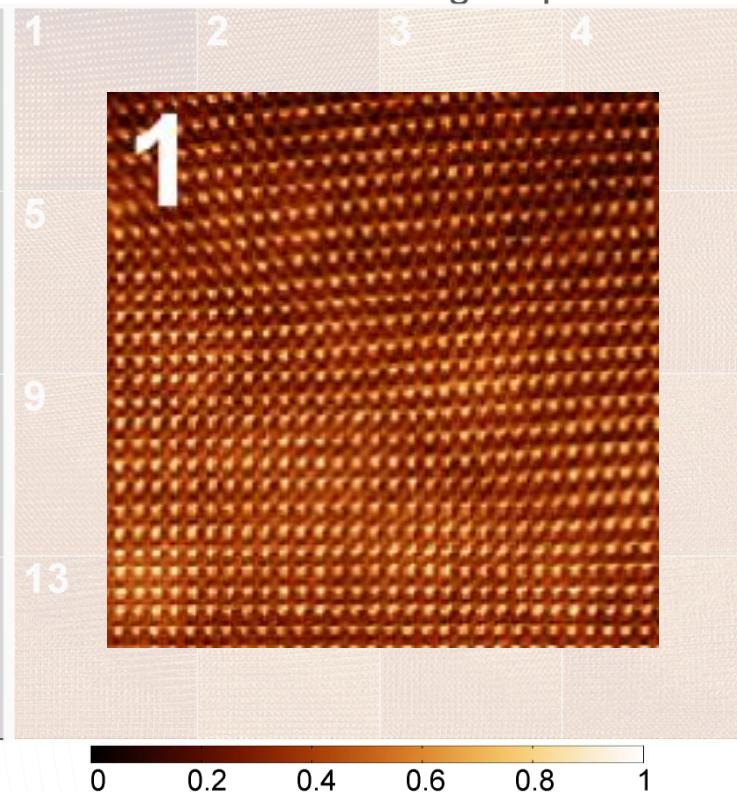
Scree Plot



First 16 eigenvectors

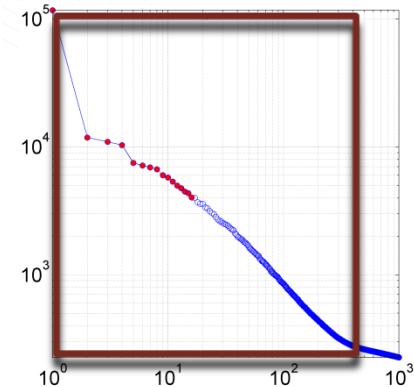


First 16 loading maps

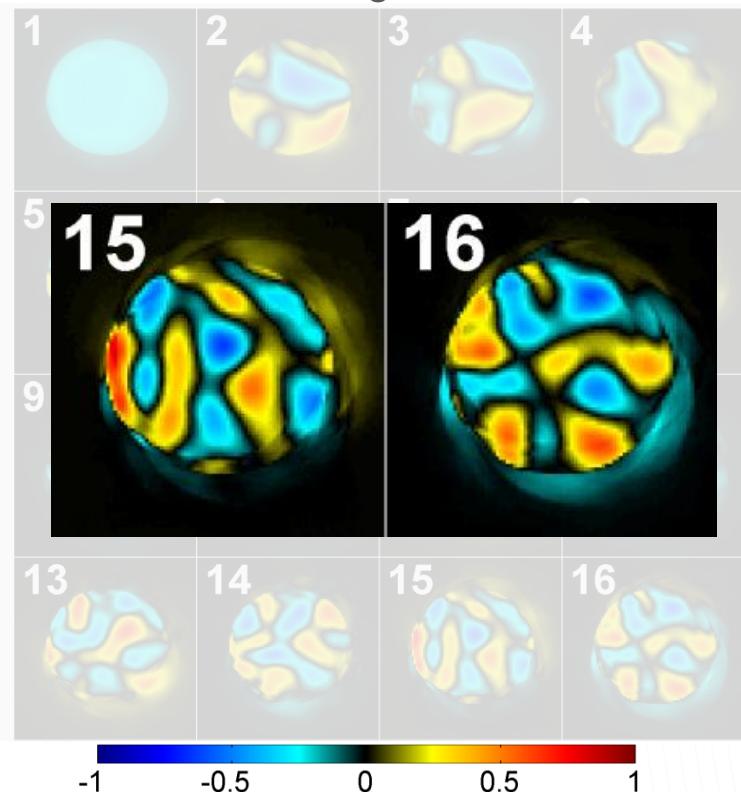


Principal Component Analysis (PCA)

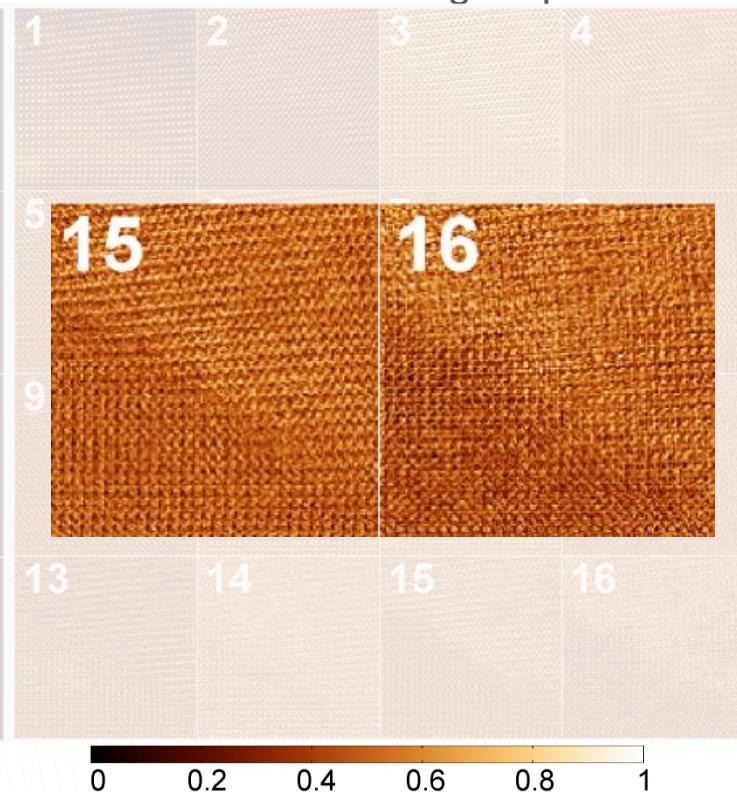
Scree Plot



First 16 eigenvectors



First 16 loading maps



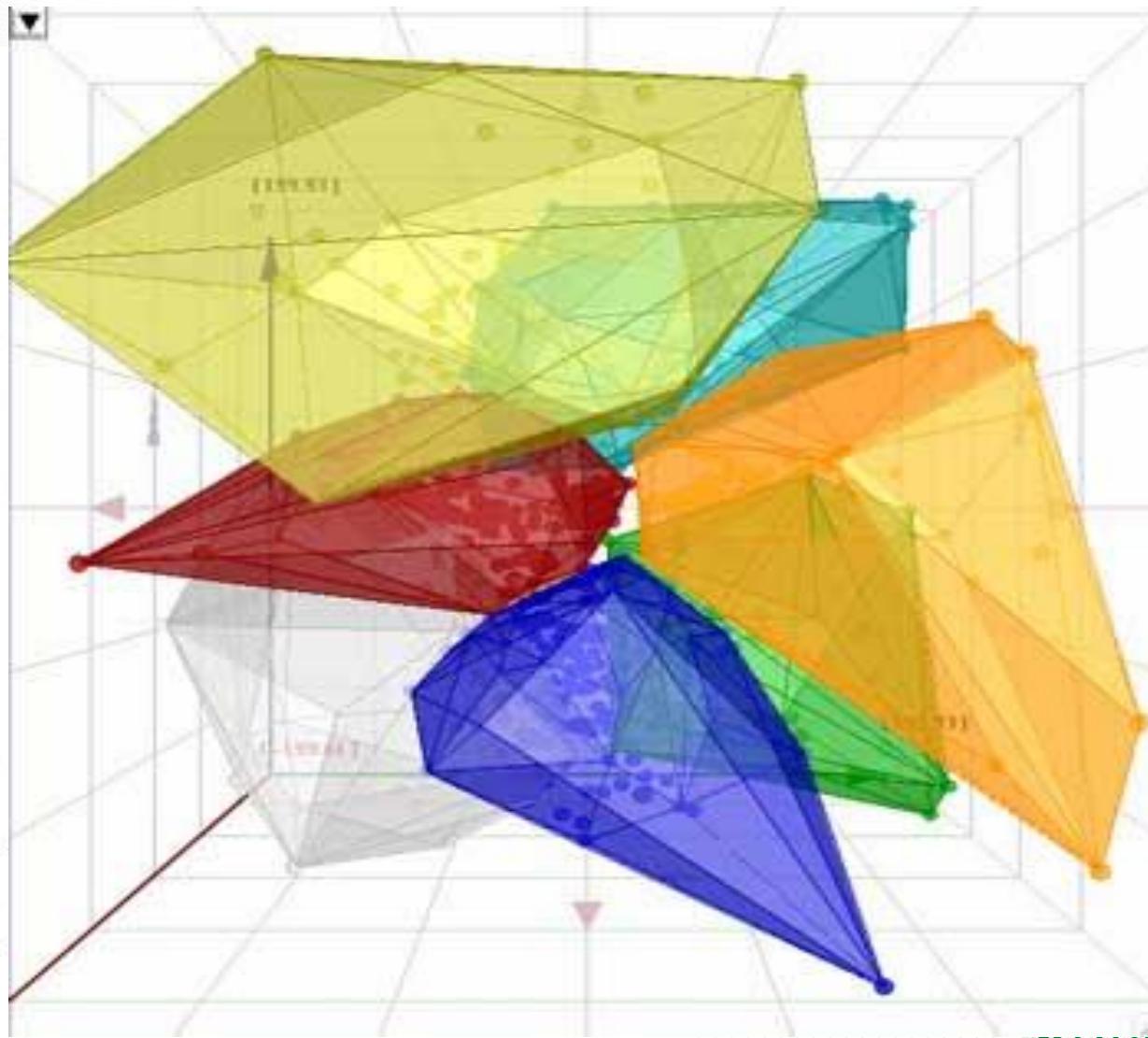
PCA typically fails at: elucidating underlying physics

PCA typically succeeds at:

- being fast and efficient
- de-noising
- data compression

Cluster analysis

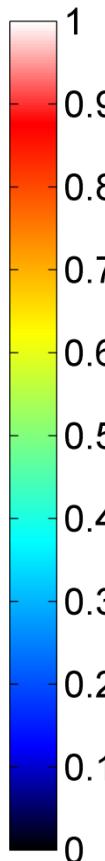
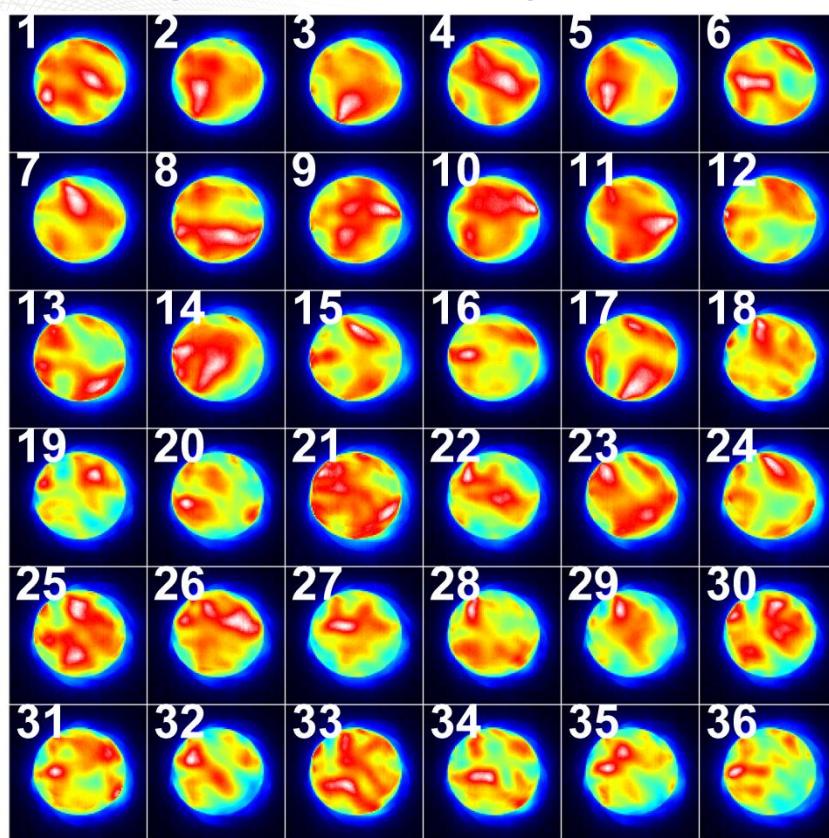
Group observables by “distance” to one another



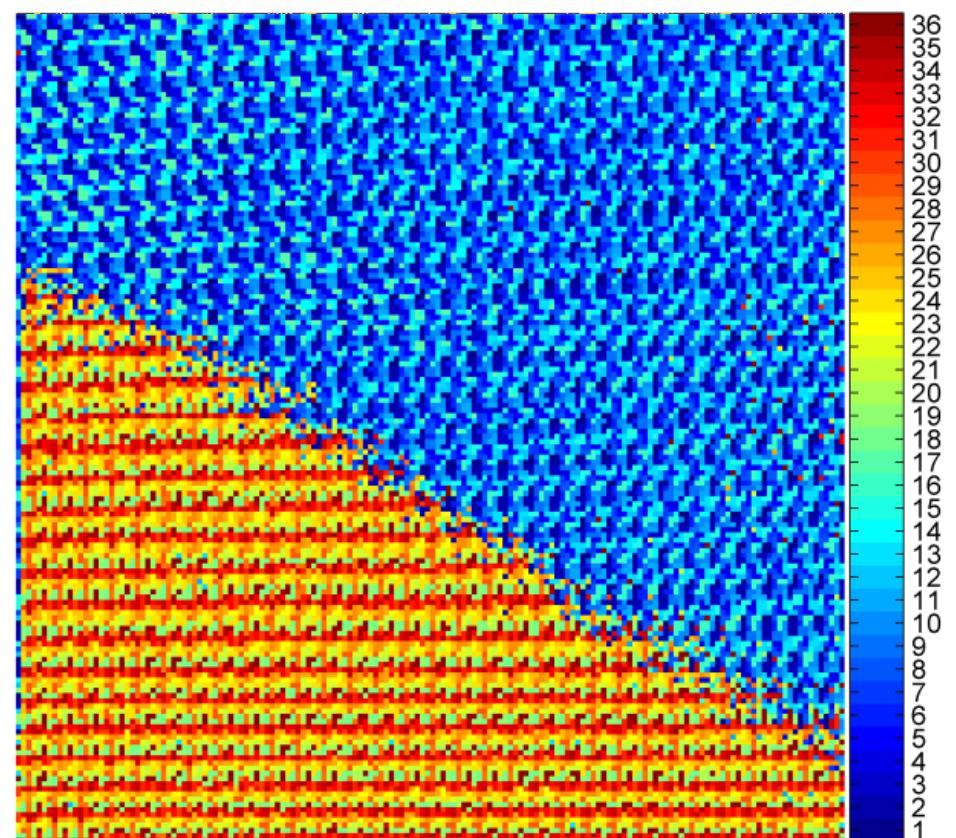
Cluster analysis

K-means clustering: 36 Clusters

Ronchigram corresponding to each cluster



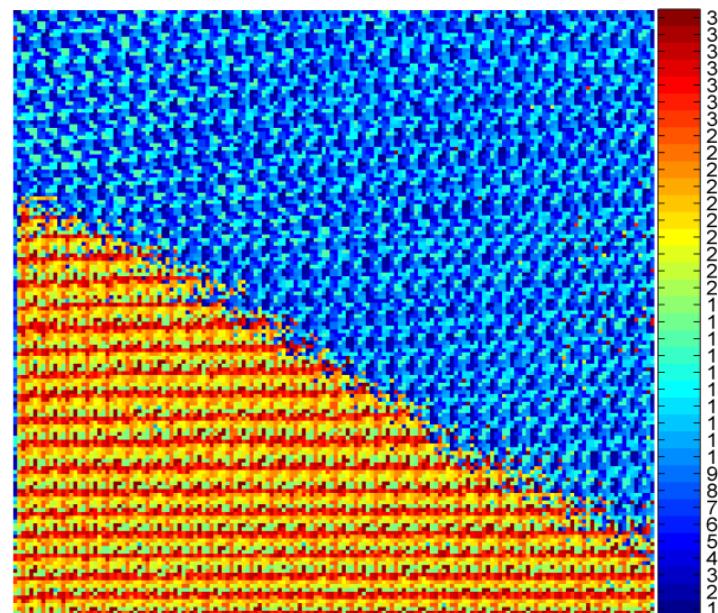
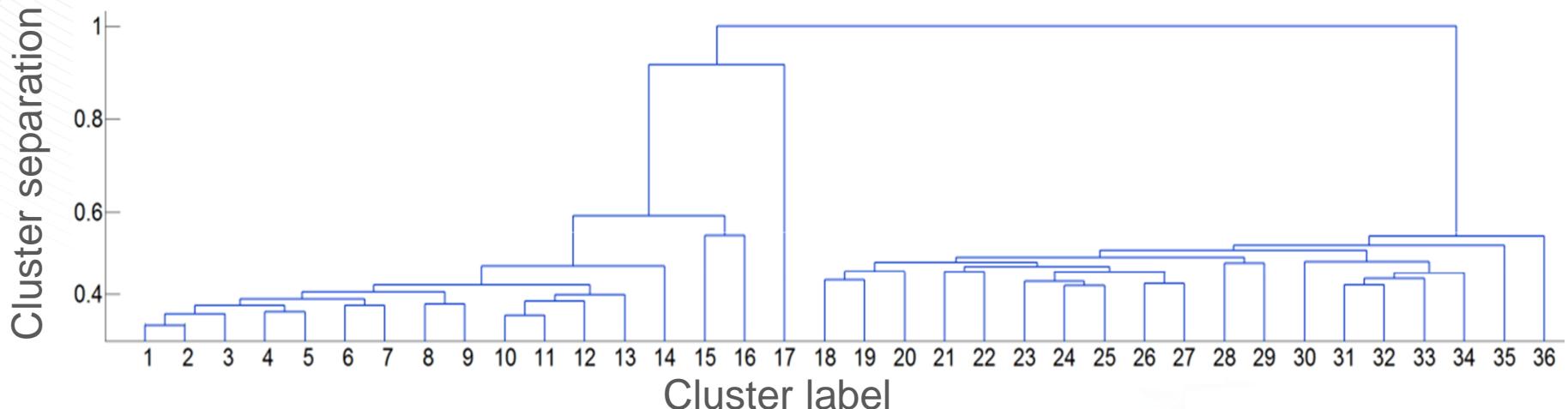
Cluster Locations



Cluster analysis

How many clusters is the right number of clusters?

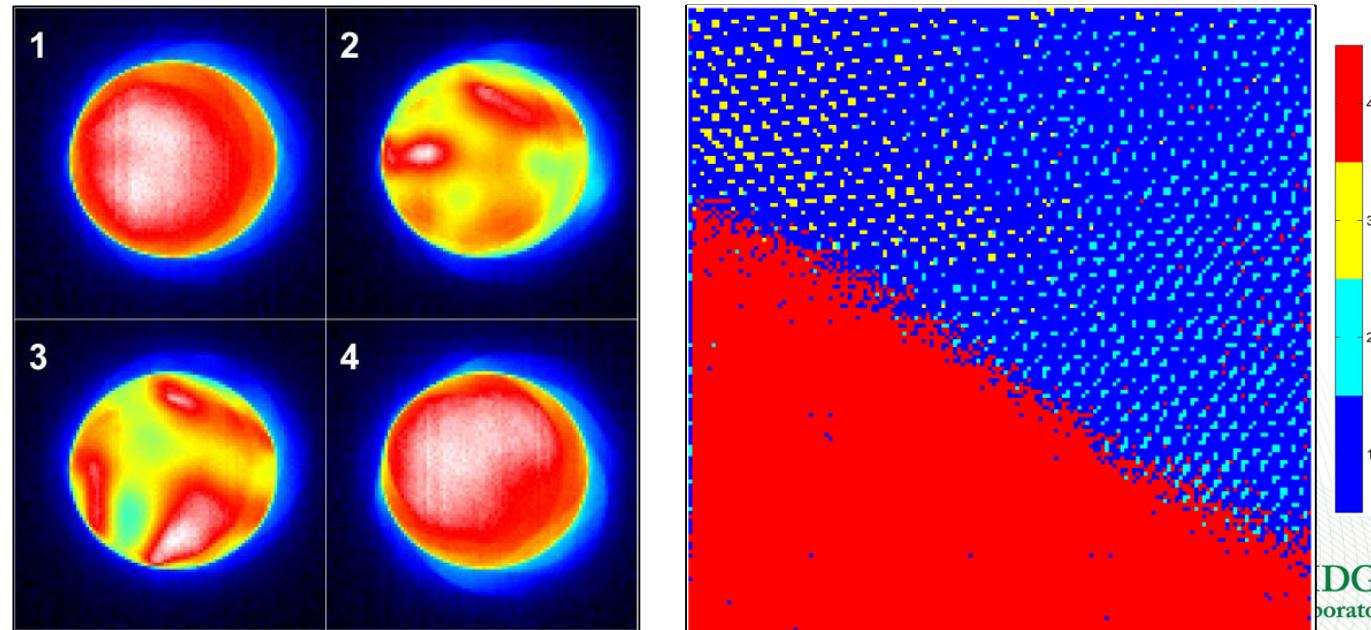
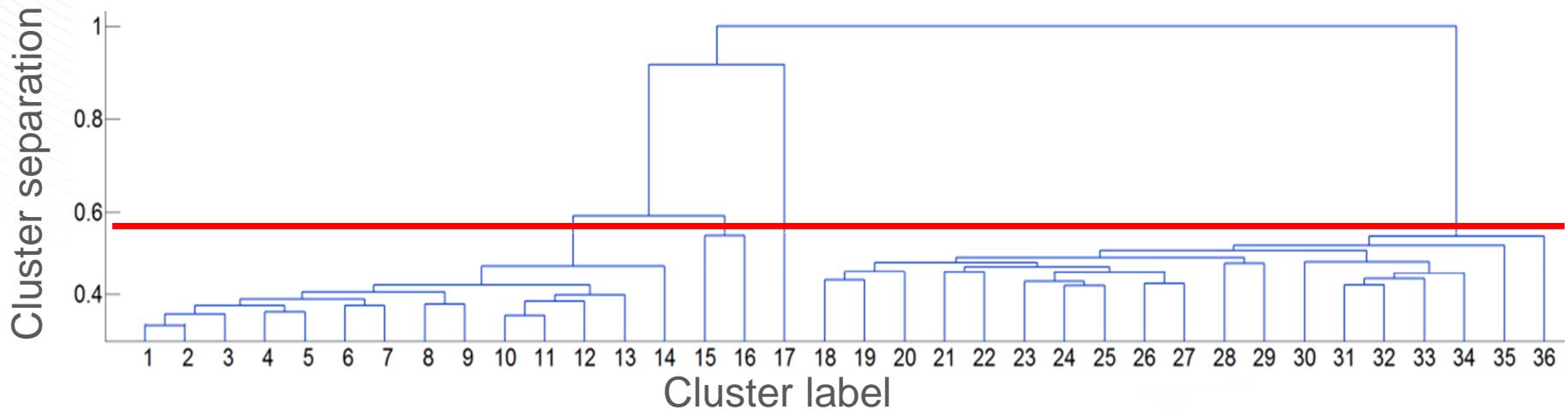
Dendrogram



Cluster analysis

How many clusters is the right number of clusters?

Dendrogram



- Scientific user facility
- 2 Page Proposal
- sjesse@ornl.gov

