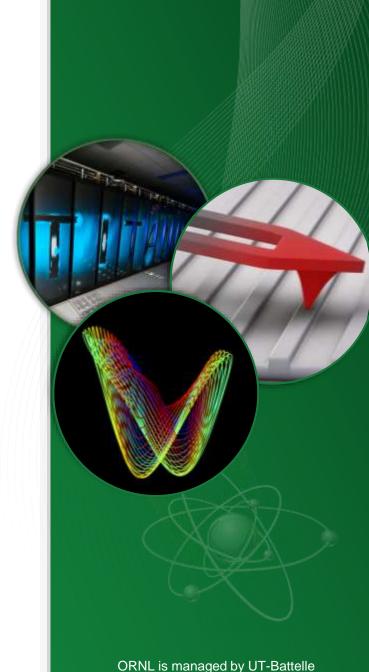
Imaging in the Information Dimension

- Suhas Somnath
- Chris R. Smith
- Stephen Jesse







ORNL is managed by UT-Battelle for the US Department of Energy

Multitude of Instruments



Micro Raman Microscope



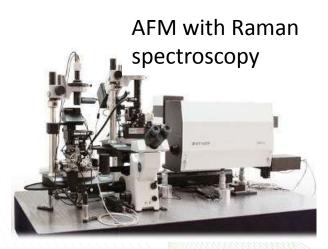
Atomic Force Microscope (AFM)



<u>AFM</u> with <u>Infrared</u> spectroscopy (AFM-IR)







What we wanted





Instrument Tier



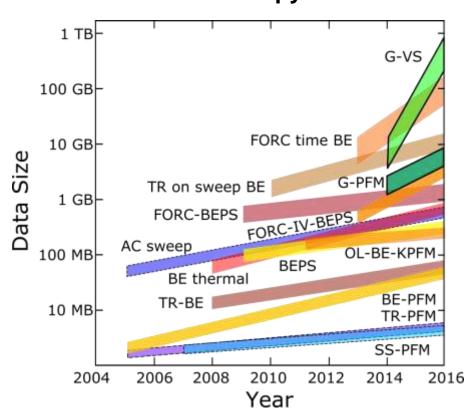




Interactive visualization, analysis, storage on supercomputers

Growing Data Sizes and Dimensionality

Evolution of Scanning Probe Microscopy Data



 Data sizes have grown from ~ 10 MB to ~ 1 TB in 10 years!

 Dimensionality ranges from 1D spectra to 7D hyperspectral datasets

 Cannot use laptops to analyze data

Instrumentation Software Inadequate for Analysis



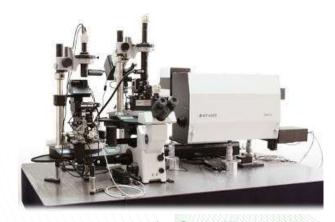




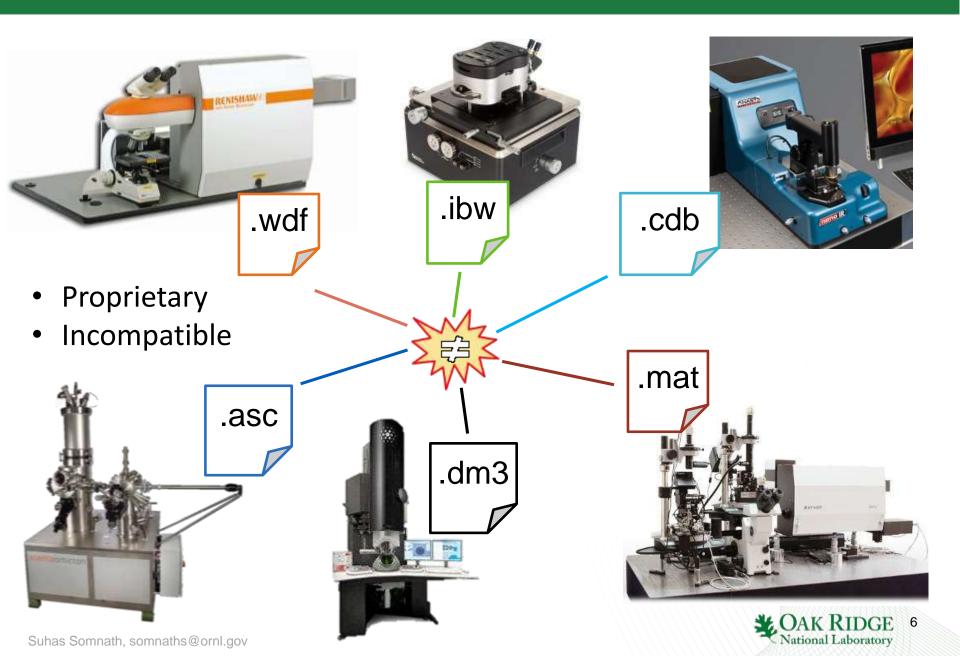
- Software provided for controlling instruments typically only comes with basic data analysis capabilities
- Integrating user-developed functionality often impossible





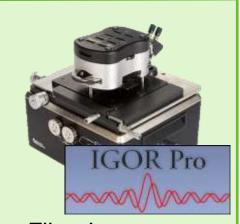


Multitude of File Formats



Disjoint & Unorganized Communities





- Filter Image
- Register Image ...



SVD Filtering ...

- **FFT Filtering**
- SVD Filtering ...



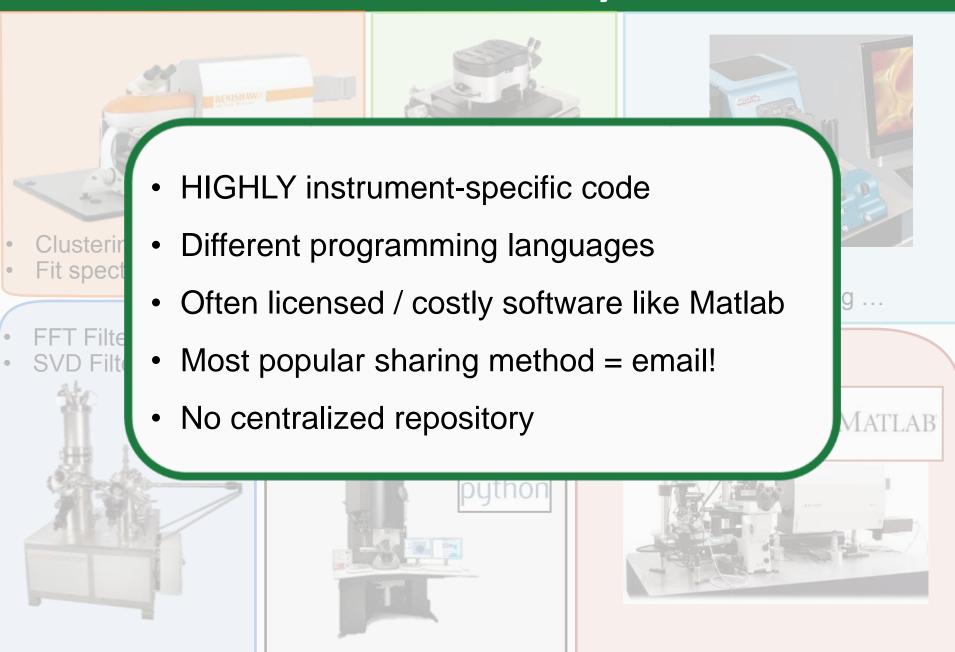
- FFT Filtering
 - Classify Images ...



Register Images



Cannot Share Code Efficiently



Problems Opportunities in Imaging

- 1. Closed science
 - a. No traceability for data analysis
 - b. Results not (readily) reproducible
- 2. Multiple, incompatible, proprietary data formats
- 3. Disorganized and unorganized communities
- 4. No proper analysis software
- 5. Growing data volumes, variety, and dimensionality

The Solution





Instrument Tier



Automated, standardized, modularized data acquisition



Instrument-agnostic, self-describing, model in HPC-friendly file format







Centralized repository for data processing, analysis





Interactive visualization + analysis + storage on supercomputers

Expectation of Data Model

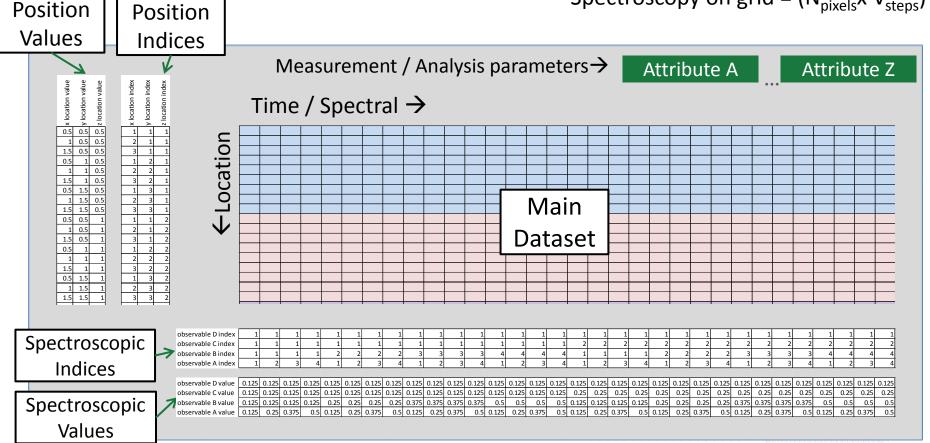
- Accommodate data of different shapes, dimensionalities, precision and sizes.
- Accommodate data without N-dimensional form
 - Compressed sensing / sparse sampling
 - Not all combinations of spectroscopic variables
 - Incomplete experimental data

Universal Imaging and Spectroscopic Data (USID)

- Data stored as 2D matrix of (position x spectral values) regardless of dimensionality
- Ancillary datasets explain the data

Example data types:

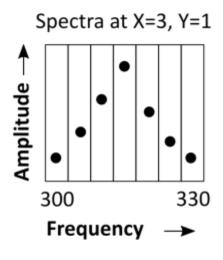
- 2D images = (N_{pixels} x 1)
- Single spectra = (1 x Z_{steps})
- Spectroscopy on grid = $(N_{pixels}x V_{steps})$



USID – 1D spectra

Original N-dimensional form

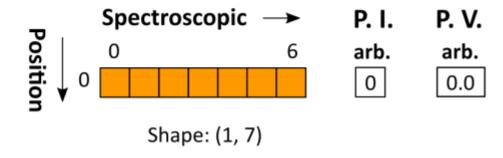
USID 2-dimensional form



Shape: (7,)

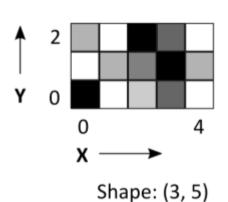
Quantity: Amplitude

Units: V



- **S. I.** Frequency 0 1 2 3 4 5 6
- **S. V. Frequency** 300 305 310 315 320 325 330

USID – 2D Image



Original N-D

<u>form</u>

Quantity: Intensity Units: arb. units

Spectroscopic P. I.

XY

P. V.

Х	Υ
-250	0
-125	0
0	0
125	0
250	0
-250	3.5
-125	3.5
0	3.5
125	3.5
250	3.5
-250	7
-125	7
0	7
125	7
250	7

USID 2D form

S. I. arb. 0

Position

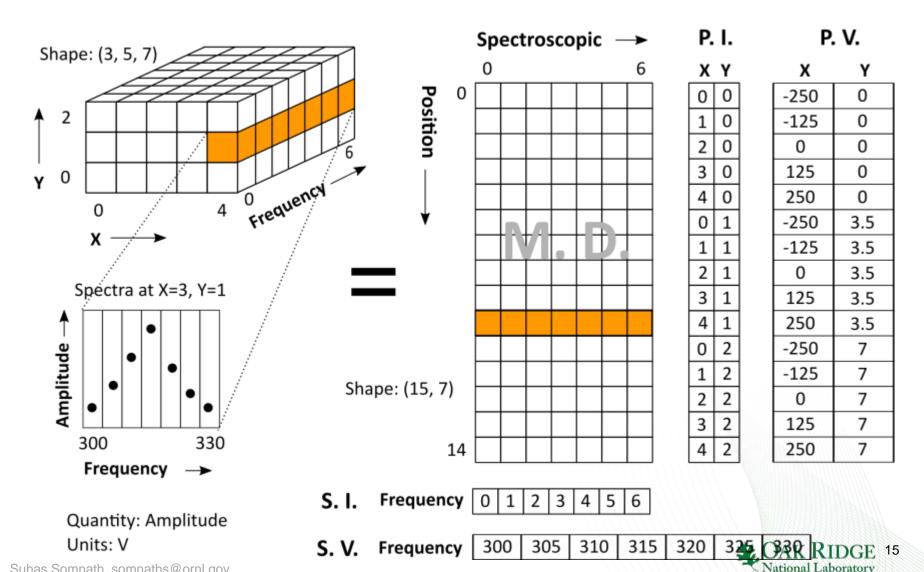
S. V. arb. 0.0

Shape: (15, 1)

USID – Spectra on Grid (3D)

Original N-dimensional form

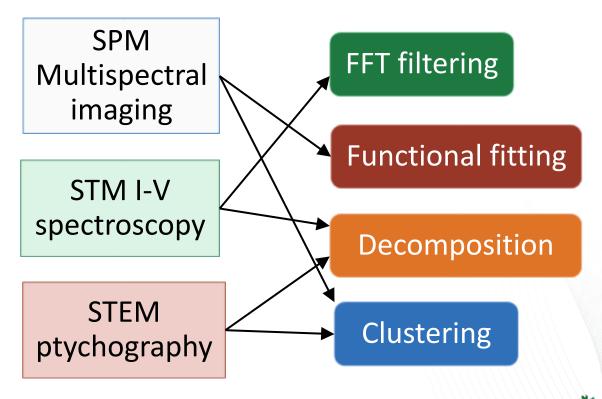
USID 2-dimensional Form



Suhas Somnath, somnaths@ornl.gov

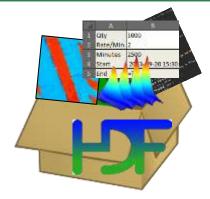
USID - Instrument Agnostic Code

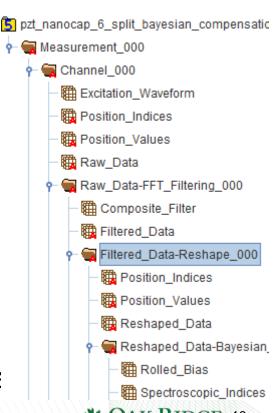
- Instrument-agnostic data allows instrument-agnostic code
- Single version of analysis and processing routine
- Brings multiple scientific communities together



Hierarchical Data Format (HDF5)

- A HDF5 file is a smart container
 - Capable of storing multidimensional datasets, Images, text, measurement parameters, etc.
 - Contents organized like traditional folders and files
 - **Groups** Analogous to file folders
 - Dataset 1 to N dimensional data
 - Integer, floating point, complex numbers etc
 - Attributes {Key: value} pairs useful for describing data and experimental parameters, etc.
- Easily accessible C, C++, python, Java....
- Tree structure + nomenclature +attributes are records of workflow applied to dataset
- Parallel read / write, HPC & cloud compatible



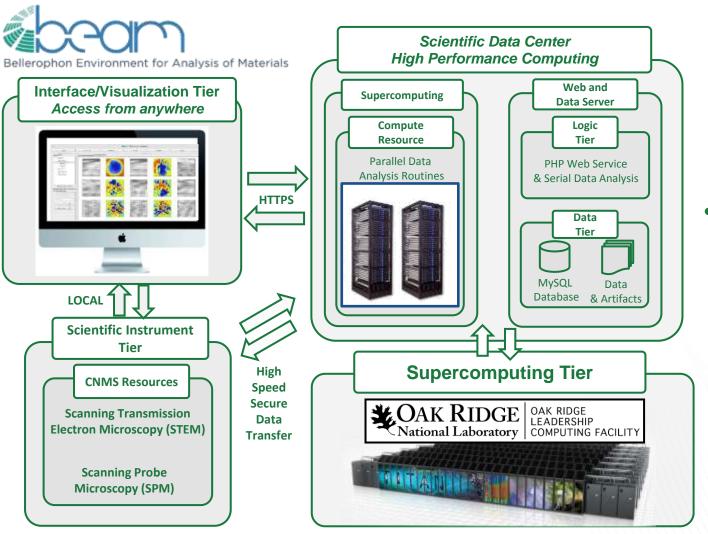


Expectation from Software

- Easy to learn and understand
- Strong support-base
- Established community standard
- Straightforward to implement and maintain
- Optimized libraries for scientific and numeric algorithms
- Access to existing imaging related packages
- Free
- Scalable to multiple CPU cores + distributed computing

(Purely) Programmer-Driven Solution

Software connecting scientific instruments to supercomputers



Successes:

- Easy to use –Point-click
- Fast on super-computers

Shortcomings:

- Very long development cycle
- Very expensive
- Brittle (points of failure)
- Scientists had no control!!



Python for Scientific Research

Very easy to learn + code

Numerous, **powerful** libraries for science









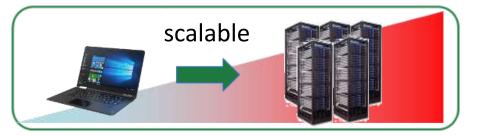
- Facilitates innovation
- More robust code
- Improved adoption of new methods / standards
- Accelerates scientific progress

Crossplatform









Established standard for:

- Microscopy
- Data science
- Microbiology
- Neutron science
- Deep learning
- More!

Strong user community



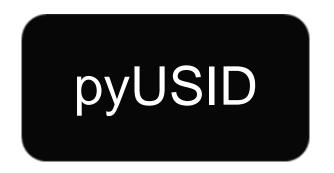
All for a princely sum of **\$0!**

See Jake Vanderplas' Pycon 2017 Keynote talk: http://www.youtube.com/watch?v=ZyjCqQEUa8o&t=19m20s



(2) Software Packages

- Written in Python
- Open source & free
- Written by scientists
- Data centric
- Instrument-independent data model in HDF5
- Instrument-independent analysis algorithms
 - Reusable across scientific domains





Software Organization

pycroscopy

1/0

 Data translators (proprietary formats to HDF5)

Visualization

- Plotting utilities
- Jupyter widgets

Simulation

AFM Force-distance ...

Analysis

- Physical model specific
- Fitting to model, etc.
- Physics based regression

Processing

- Physical model agnostic
- Image filtering, registration,
- Multivariate analysis

pyUSID

- HDF5 file i/o operations
- Base data processing,
 - visualization...



Software Organization

Science / data analytics applications



File / data tools

Well documented

Beginner topics

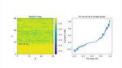
To learn how to use pyUSID, Please go through the following documents in the recommended order:



01. Primer to HDF5 and h5py



02. The USIDataset



03. Translation and the NumpyTranslator



04. Plotting utilities



05. Utilities for reading h5USID files

Intermediate topics

To learn how to write to h5USID files, write data processing classes, or adding functionality to pyUSID, go through these additional documents in the recommended order: Those interested in contributing to pyUSID are encouraged to read our guidelines for contributing code



06 Utilities for handling

07 Speed up

OR Utilities that assist

reshape_to_Ndims(h5_main, h5_por=None, h5_spec=None, get_labels=False, verbose=False, sort_clims=False| | jumpos|

Reshape the input 2D matrix to be N-dimensions based on the position and spectroscopic datasets.

To view the filti

('ifft2'), Remem

necessary to us

transform. Also

the inverse tran

symmetric abou

result in the inv

times smaller ti

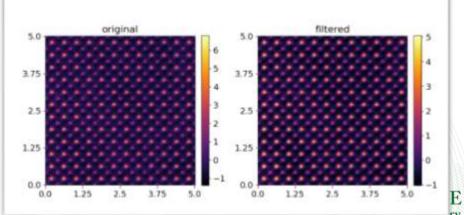
kept.

- Parameters: h5_main (HDF5 Datased 2D data to be reshaped
 - . h5 pos (HDF5 Dataset, optional Position indices corresponding to rows in
 - . h5_spec (HDFS Dataset, optional) Spectroscopic indices corresponding to columns in h5 main
 - . get_labels (bool optional Whether or not to return the dimension labels.
 - · verbose (book optional Whether or not to print debugging statements
 - . sort_dims (bool If True, the data is sorted so that the dimensions are in order from fastest to slowest If False, the data is kept in the original order If get, labels is also True, the labels are sorted as well.

Returns:

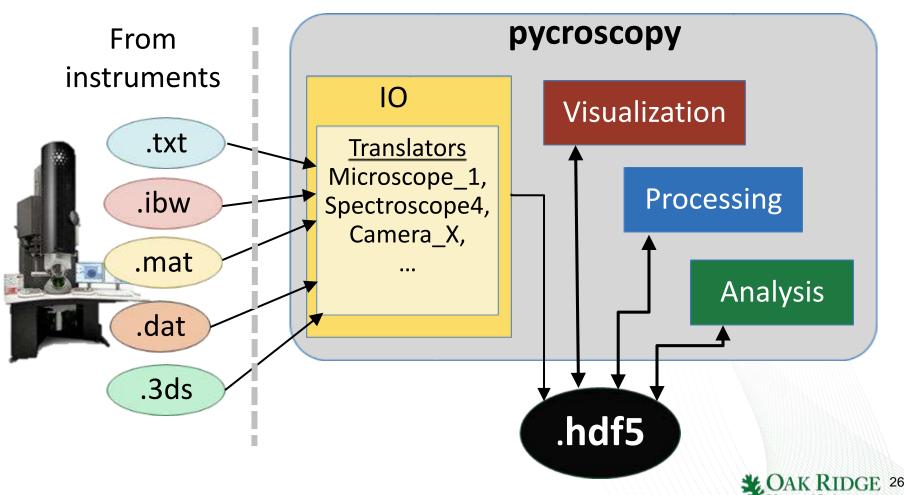
. ds_Nd (N-D numpy array) - N dimensional numpy array arranged as image_filtere [positions slowest to fastest, spectroscopic slowest to fastest] inage filtere

```
fig. axes = plt.subplots(ncols=2, figstre=(10, 5))
for axis, ing, title in zip(axes, [image_raw, image_filtered], ['original', 'filtered']):
    - px.plot_utils.plot_map(axis, img, cmap=plt.cm.inferno.
                              x_size=x_edge_length, y_size=y_edge_length, num_ticks=5)
    awis.set_title(title)
fig.tight_layout()
```



Entering the USID Ecosystem

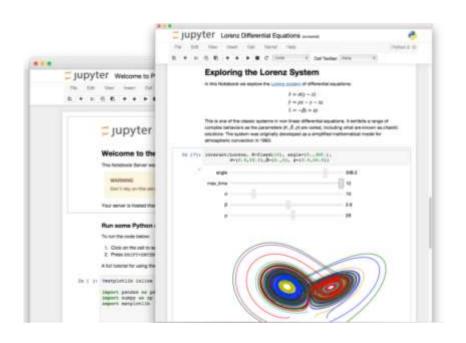
- hdf5 file is the hub for all operations
- Analysis, processing, visualization available after translation to .hdf5



Jupyter Notebooks



Jupyter Notebook



- Interactive documents
- Exploratory programming
- Code
- Text
- Images
- Interactive slice through data, pan, move, rotate ...

Truly Achieving Open Science, Reproducibility

Aim – <u>ALL</u> scientific journal papers accompanied with:

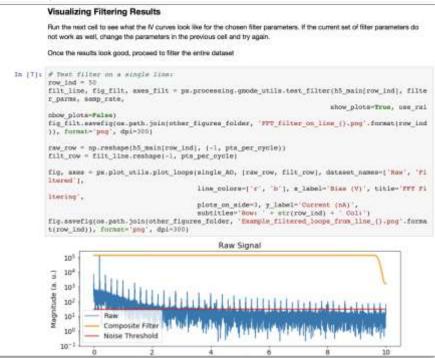
- Jupyter notebook that shows all analysis (raw data → figures).
- Data with DOI number



yielding >100,000 I-V curves in <20 min. This allows detection of switching currents in the

nanoscale capacitors, as well as determination of the dielectric constant. These experiments

show the potential for the use of full information capture and Bayesian inference toward extracting physics from rapid I-V measurements, and can be used for transport measureJupyter notebook associated with paper

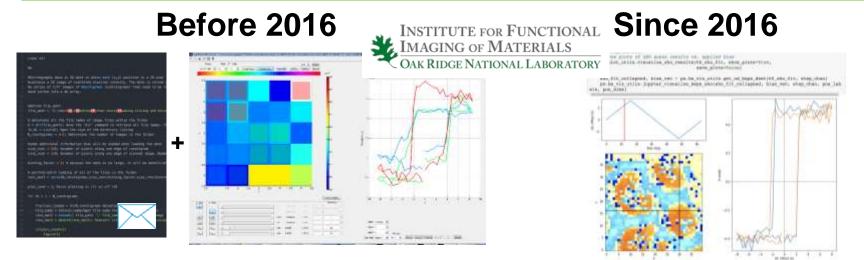


DOI associated with data (raw → paper figures)



ments in both atomic force and scanning tunneling microscopy.

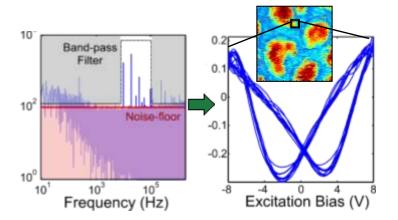
Pycroscopy - Supporting User Research



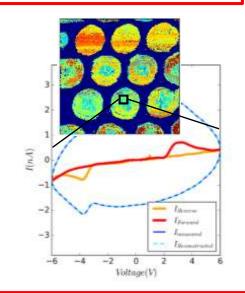
Scripts + complicated, Matlab GUI	Set of simple Jupyter notebooks
Witten by dedicated software engineer	Written by material scientists
Not customizable	Completely customizable.
2-3 hours of training before use	Notebooks include instructions. NO training required!
Deployed only on two offline workstations due to licensing restrictions = queue	Each user gets VMs with jupyter notebook server
Will remain on off-line desktops Suhas Somnath, somnaths@ornl.gov	In the process of switching to computations on clusters OAK RIDGE 29

Pycroscopy - Scientific Advancements

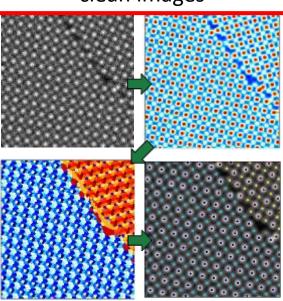
3,500x faster imaging via adaptive signal filtering, linear unmixing of signals



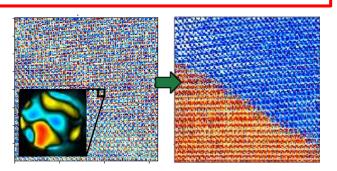
200x faster spectroscopy via Bayesian inference



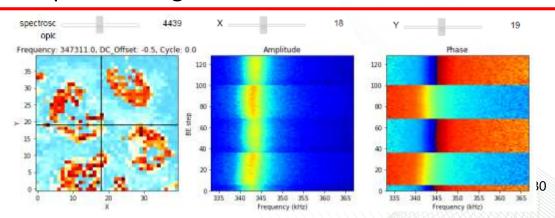
Separating uncorrelated data from correlated data to clean images



Identifying invisible patterns using multivariate analysis



Simplified navigation multidimensional data - users



Software Progression

Scaling up Computing:









2016 Single core 2017 Multi-core Single CPU 2018 Multi-core Multi CPU

2019 JupyterHub On HPC

Emphasis always on ease-of-development instead of raw performance

New Scientific Domains:

Atomic Force Microscopy

2016

+ Scanning
Transmission
Electron
Microscopy

Mass
Spectrometry
+ BioChemistry

+

2017

2018

2019

Thank you

Questions?



File / data tools related- https://groups.google.com/forum/#!forum/pyusid

Science, data analysis, reading proprietary instrument data - https://groups.google.com/forum/#!forum/pycroscopy