Interactive Visualization and Automated Analysis of Neutron Scattering Experiments

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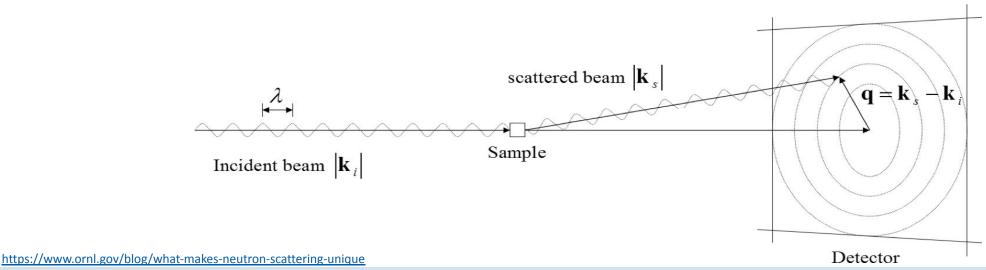
adiasoft



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Neutron Scattering probes within Materials

- Why do we shoot particles at things?
 - X-ray scattering probes low-Z materials and are scattered by electrons
 - Electron scattering is great for low-Z materials and is strongly interacting
- Neutrons are weakly scattering!
 - Neutrons will either pass straight through most materials or only single scatter off the atomic nuclei in the material
 - Neutrons can penetrate materials that stop X-rays or electrons
 - Beneficial for studying materials near absolute zero

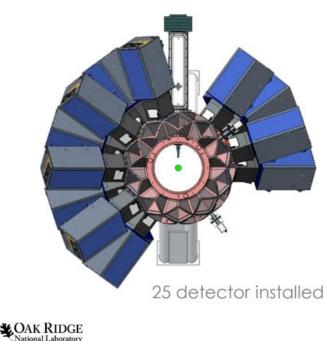


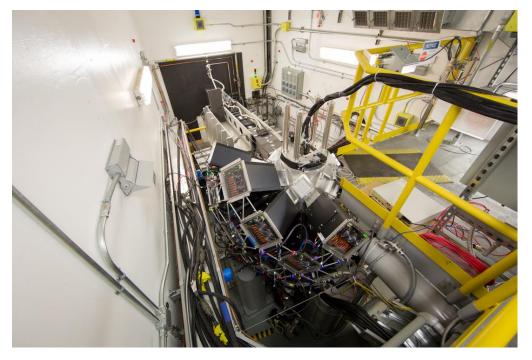


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TOPAZ Single-crystal Diffractometer

- Located at ORNL and receives neutrons from the Spallation Neutron Source
- Samples can be measured with high precision for volumetric sampling in reciprocal space (momentum measurements)
 - Samples are rotated to measure all aspects of the lattice
 - Temperature control from 5 K 450 K
 - Broad Q coverage

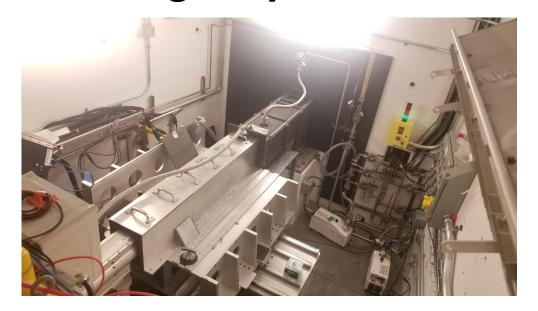


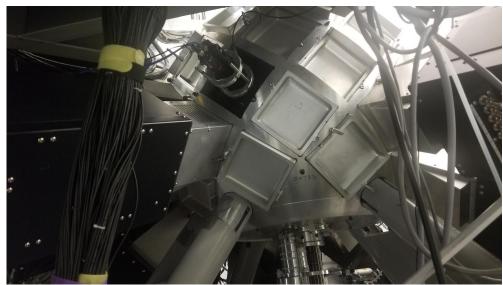


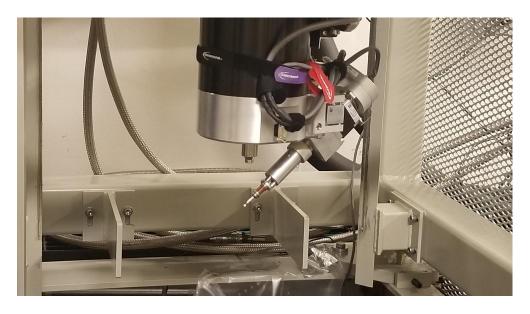
https://neutrons.ornl.gov/topaz



TOPAZ Single-crystal Diffractometer









Analysis methods for TOPAZ data

- Mantid
 - Open source community developed application
 - Algorithms use raw data processed within 2D slices of the 3D data
- Some challenges faced by the community
 - Large datasets can be up to 100 Gb in size and current tools limit interactivity
 - Displaying data requires a high level of user interaction
 - 2D slices may miss key features of the diffraction data
 - May be slow to run due to optimization shortfalls
 - Timescale: I-2 hours
- Working collaboratively to leverage Mantid
 - Machine learning DBSCAN, U-Nets
 - Automated Bragg peak identification
 - Automated background subtraction
- Also improving the interactivity of the data viz



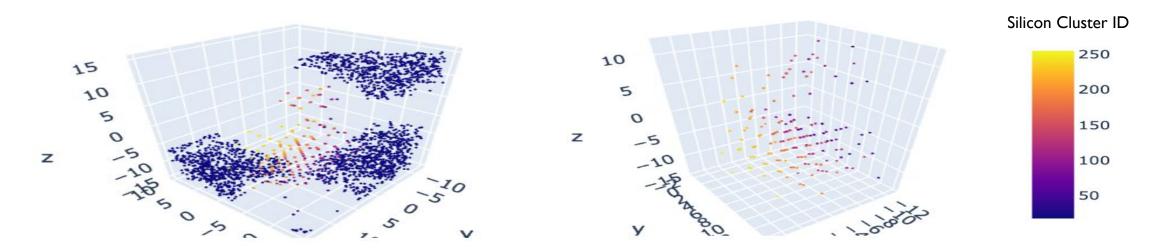


https://www.mantidproject.org/

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Analysis of Scattering Data for a Single Silicon Crystal

- Preprocessing the data
 - Measurements of the particles in momentum space
 - Looking for pockets of constructive interference
 - Measurements projected onto a 3D mesh
- DBSCAN is a density based clustering algorithm
 - Can identify oblong clusters and attribute data to noise if it doesn't match anything
 - Sparse background can be removed easily
 - Provides real time identification of Bragg peaks
 - Agrees with Mantid reconstruction of Bragg peaks



https://scikit-learn.org/stable/modules/generated/sklearn.cluster.DBSCAN.html#sklearn.cluster.DBSCAN



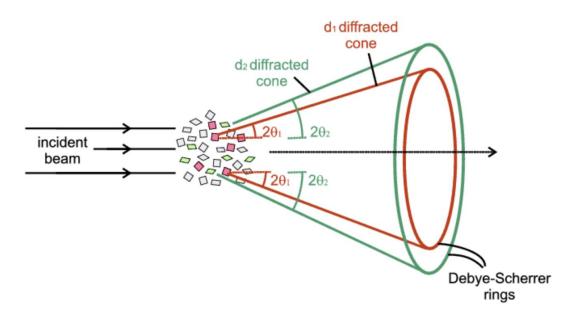
Background Removal Processes

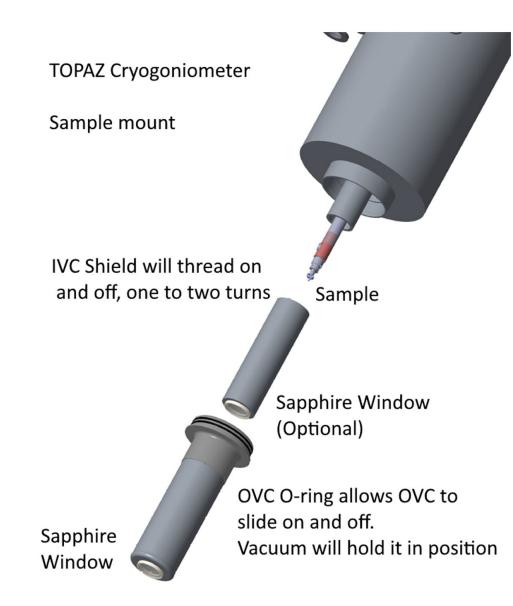
Sample mounting

- Crystal sample can be tested at low temperatures
- Aluminum cylinders have a randomly oriented crystal structure
- These rings will scatter neutrons into a sphere or ring

Scattered neutrons

- Each plane orientation will create rings within reciprocal space
- Translates into a sphere in 3D representation





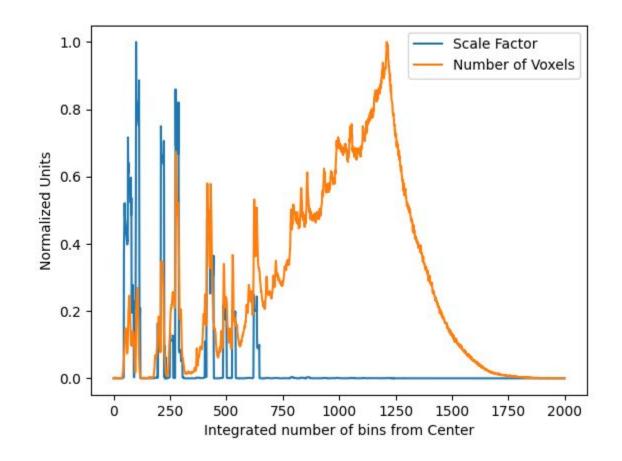


Identify Accurate Ground Truths

- Properties of Aluminum Rings
 - Monochromatic in neutron intensity
 - Concentrated at a fixed radius
 - Varies from sample to sample
- Define a scale factor
 - Simple clustering
 - Tune DBSCAN parameters to get rings
 - Gradient based definitial for clustering
 - Normalized by the mean number of voxels

$$SF_i = max(\sum_{i}^{N+i} \nabla w_i)$$

How do these two methods compare?

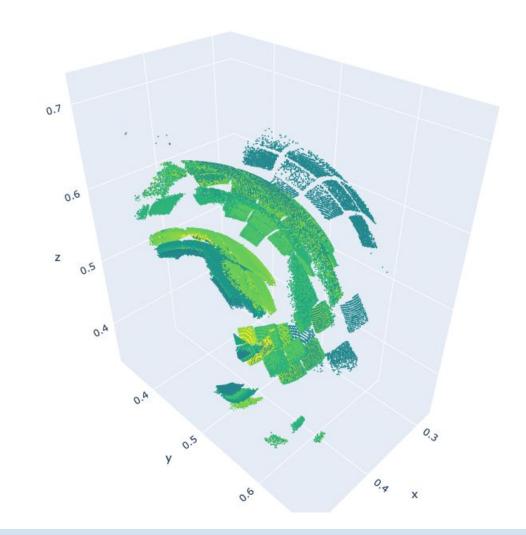


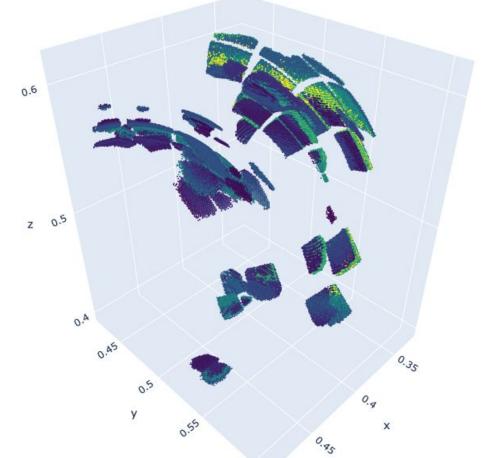
Aluminum Labels

DBSCAN

0.6

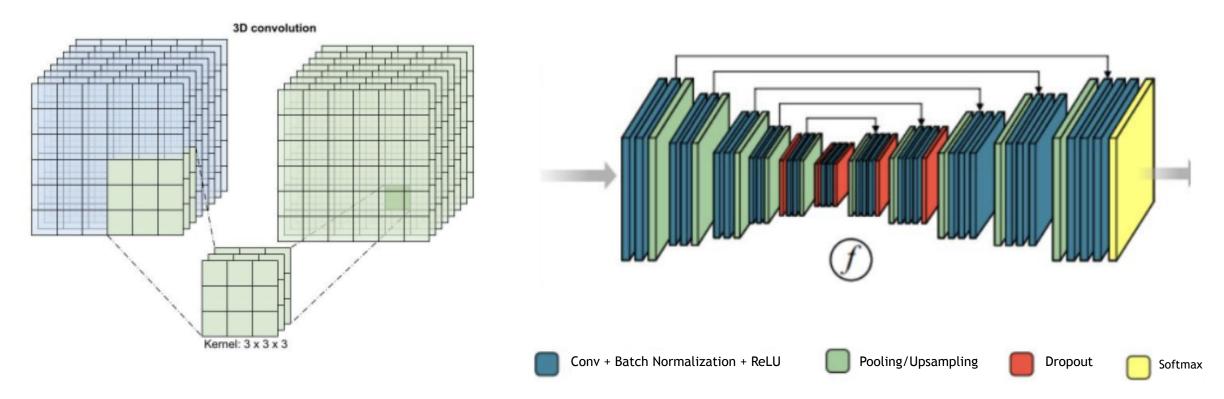






Identification of Aluminum Rings

- Building a neural network
 - Identify a 3D CNN to train and predict aluminum rings
 - General Unet mainly used for image segmentation
 - Fully connected
 - Contraction and expansion paths
 - 3D Unet used for learning dense volumetric segmentation from sparse annotations





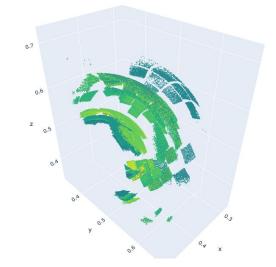
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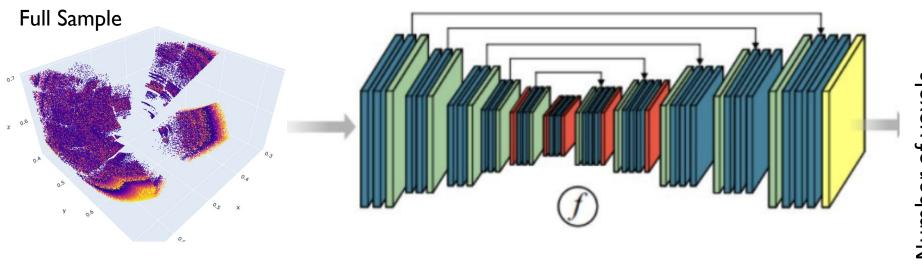
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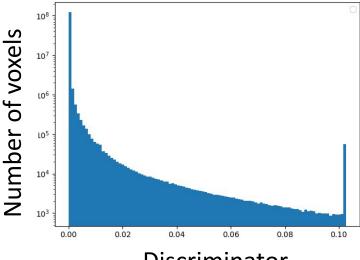
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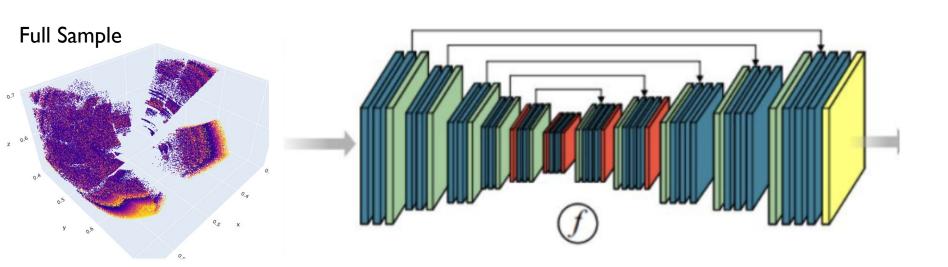




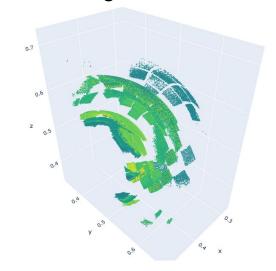
Discriminator

Identification of Aluminum Rings

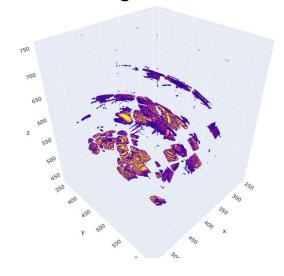
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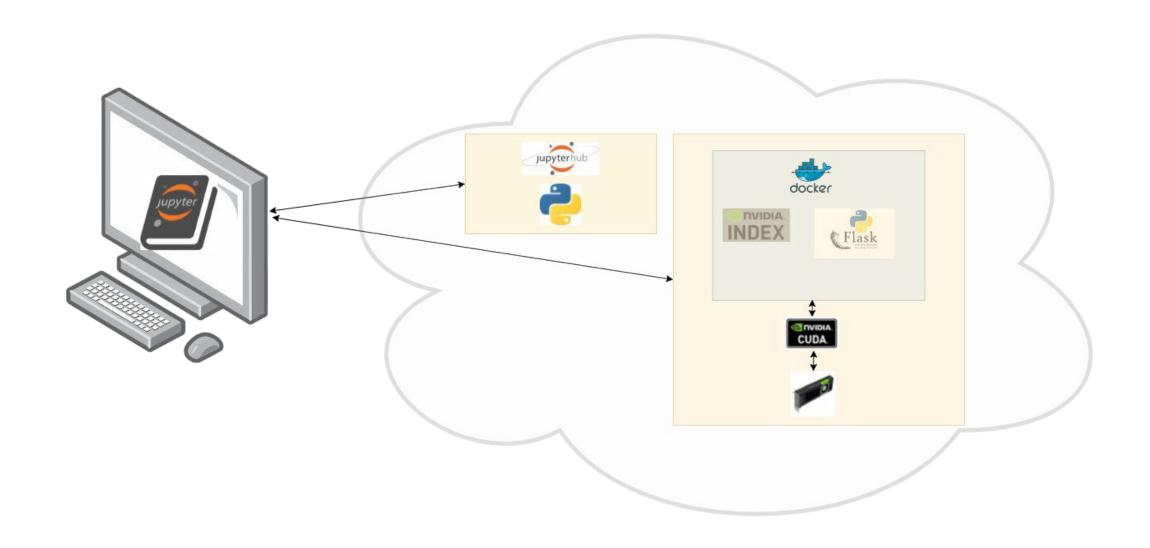
Labelled Rings



Predicted Rings

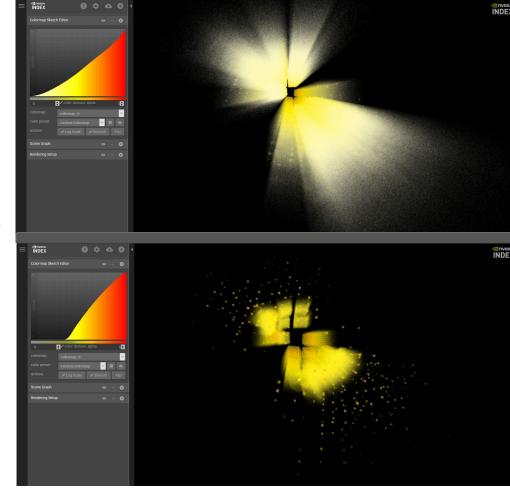


System Architecture, with Jupyter Hub Interface

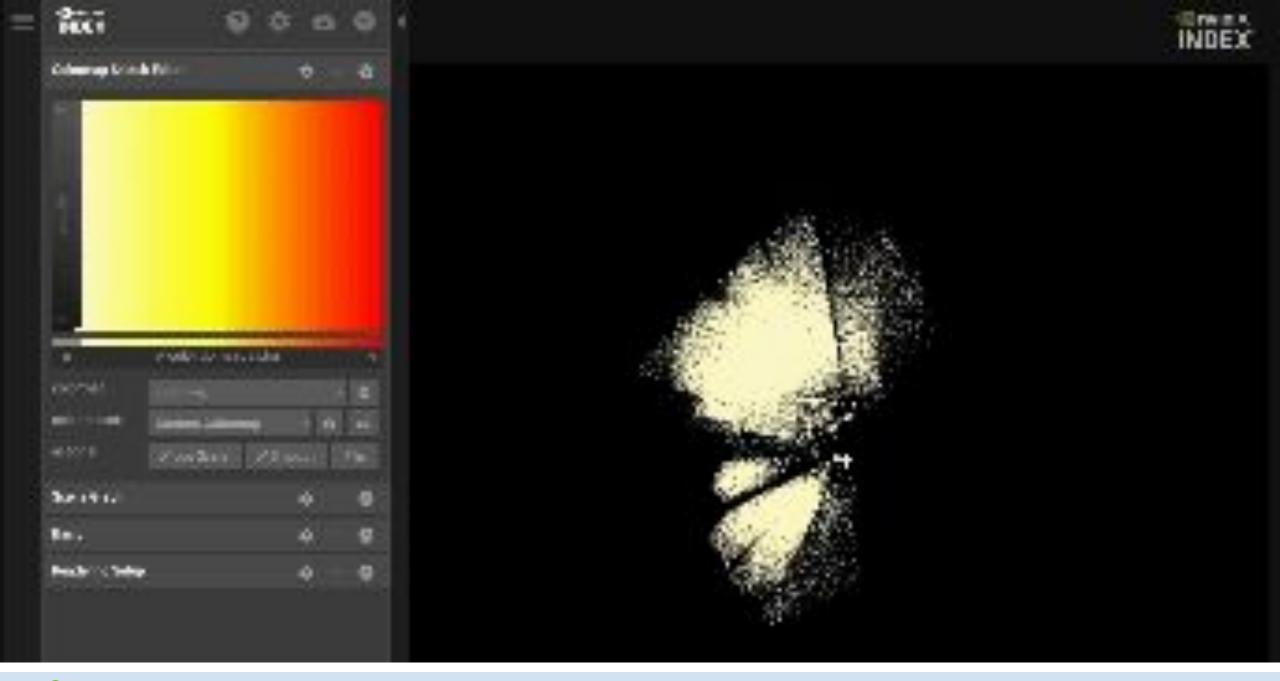


NVIDIA's IndeX: Browser-based 3D visualization

- IndeX: interactive 3D volumetric visualization framework
 - rotate, pan, zoom into large 3D volumes
 - executes remotely on a GPU server
 - o streams video to the user's browser
- TOPAZ event data is converted into a voxel database
 - voxels are a regular 3D grid
 - each is colored and/or made transparent based on the number of neutron events within it
 - colors & transparency can be dynamically edited
- The voxel database can be dense or sparse
 - large fraction of voxels are empty, so the sparse data format is very efficient
- Interactive response is robust to WiFi quality
 - even with a poor network connection, the typical delay after a mouse click is < 0.25 s









Summary

- Take the user out of the loop
 - Automated sample identification
 - Real time sample visualization and analysis
 - Information at your fingertips
- 3D visualization
 - Integrated on a single GPU
 - Interactive analysis
- Open-source collaboration
 - Applications to many real time 3D datasets
 - o Broad applications to samples within a beamline

THANK YOU!

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