

2nd European Photon & Neutron EOSC Symposium

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Machine Learning-based Spectra Classification

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PaNOSC and ExPaNDS projects have received funding from the European Union's Horizon 2020 research and innovation programme under grant agreements 823852 and 857641, respectively.

Background

Background (Big data):

• Experiments in photon sciences at synchrotrons or XFELs (e.g. SCS, FXE and HED experiments) always generate a large volume of data.

□ ML algorithms and its requirements:

• A great amount of data which are clearly annotated and complete for covering the requirements of scientific reuse.

□ ML, Big Data requirements on Ontologies, and Application Definition:

- The research outputs should align with the 'FAIR' principles, meaning that data, software, models, and other outputs should be *Findable*, *Accessible*, *Interoperable*, and *Reusable*;
- For data re-use, the metadata shall follow a scientific experiment data model; Appropriate NeXus application definitions shall be defined.





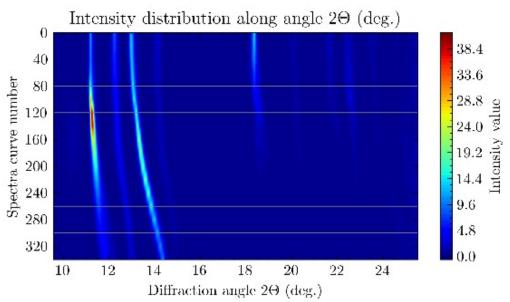


Example of Data Reuse

HED Diffraction Spectra data

[Mg0.2,Fe0.8] 50s ramp, 100 ms exposure n = 0n = 20n = 40n = 60n = 80n = 100Intensity n = 120n = 140n = 160n = 180n = 200n = 220n = 240n = 260n = 280n = 300n = 320n = 34010 12 16 22 24 26 14 18 20 Diffraction angle 20 (deg.)

Data set: 349 samples with each of 4023 features **Two phases**: Low and High pressure **Training data**: 4 (original)+40 (simulated) = 44



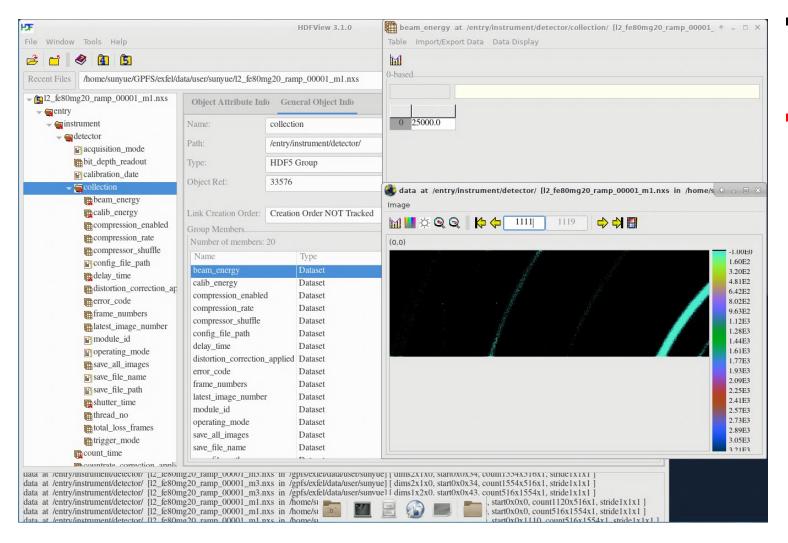
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Measured at PETRA III beamline P02.2 using a 25.6 keV beam.

The major spectral changes we aim to capture are

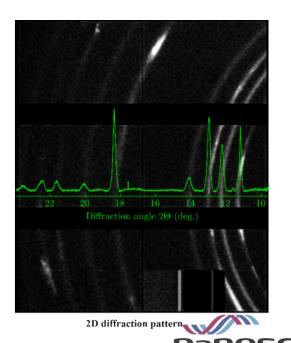
- the change of intensity distribution (e.g. drop or appearance) of peaks at certain locations, or
- the shift of those in the spectrum.

NeXus Raw data file example



One of the NeXus Raw data file displayed in HDFview software

- NeXus base classes are used, but no application definitions (only free key-values under NXcollection).
- No real experimental data was registered (e.g. beam_energy).

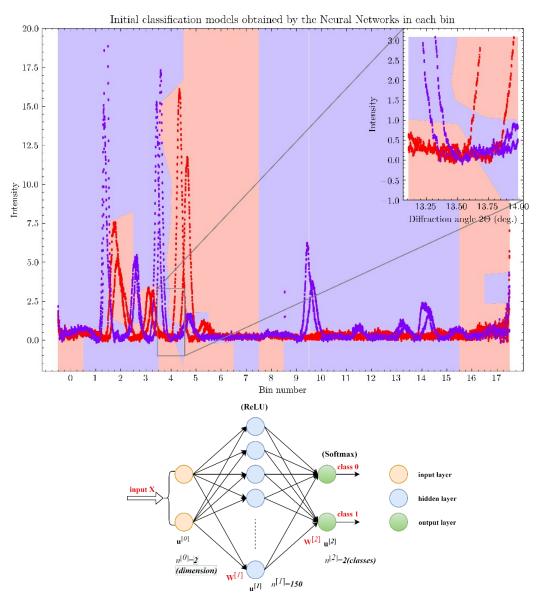


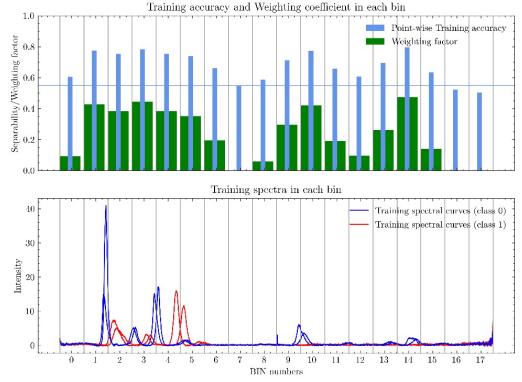


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Two-layer NN with Weighting technique for spectra classification

In this work, the spectra classification task is regarded as a general two-dimensional (2D) space segmentation problem.





The distribution of point-wise training accuracy (separability), and the believability weighting factor for each bin.

Believability weighting factor:

 $w_b = \max(0, 1.4337A_b^2 - 0.4337)$

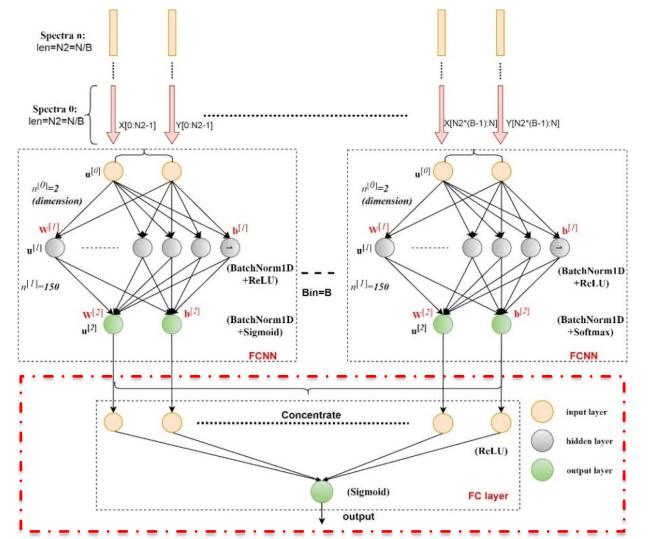


Ambigious regions corresponding to different number of bins. Relationship between confidence and number of bins BIN=1 Confidence vs Bnum -150-140BIN = 5-160Intensity Intensity sity -50BIN = 11-1600.8 Inten -170-100-180-180confidence 0.6 -200-150-19020 25sification 10 15 2510 1520 2510 15 20Diffraction angle 2Θ (deg.) Diffraction angle 2Θ (deg.) Diffraction angle 2Θ (deg.) -170Final cla 0.4 -175-180-180-180Intensity -180 Intensity Intensity 0.2-185BIN = 18-190BIN = 24BIN = 29190 -200-195-2000.0 -20015 20 2530 0 5 10 10 15 20 2510 15 20 2515 202510Number of bins Diffraction angle 2Θ (deg.) Diffraction angle 2Θ (deg.) Diffraction angle 2Θ (deg.) N_B $L_{curve} = \operatorname*{argmax}_{k \in \{0, \cdots, C-1\}} \sum_{h=1}^{\infty} w_b \sum_{i=1}^{\tilde{\nu}} \hat{C}_k^{(i)}$ Performance metrics: $P_{conf} = 1 - \frac{N_f}{N_f}$

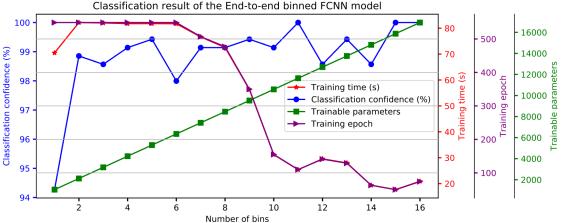
- Submitted a paper 'Machine Learning Applied for Spectra Classification in XFEL Sciences' to Data Science Journal;
- <u>https://github.com/European-XFEL-examples/panosc-ml-spectra-classification</u>.



End-to-End FCNN with automatically capturing weighting factors model



Classification results under different number of bins:



As the number of bins increases: Classification confidence increases; Trainable parameters increase linearly; Training epochs decreases; Training time decreases.



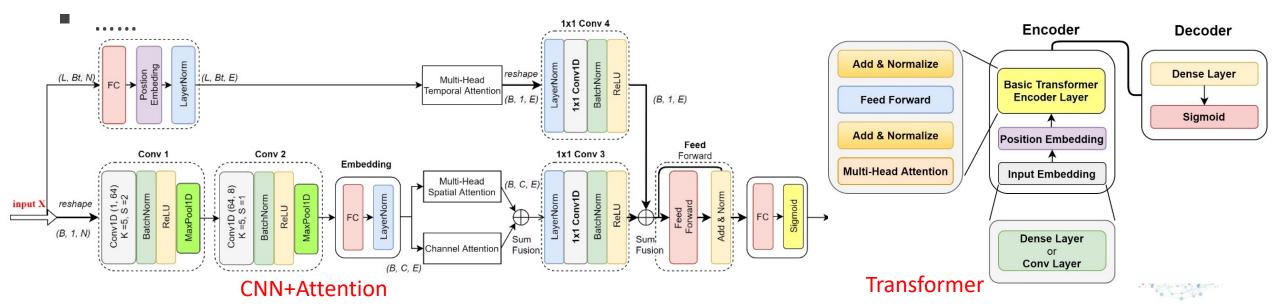


End-to-end FCNN with automatically capturing weighting factors model

Other classification models

Under the setting of 1D spectral time series classification:

- 1D Fully Connected NN (FCNN)
- Convolutional Neural Network (CNN) solution, ResNets solution
- LSTM-based solution
- Transformer-based solution
- CNN+Multi-head Attention



Conclusion and Perspectives

- The process of selecting/creating the training set is still limited, because the data is not properly annotated and some key information (such as the pressure value for each diffractogram) is missing.
- Appropriate NeXus Application Definitions needs to be developed.
- PaN portal is enabling 'FAIR' principle:

• Data search;

- Download / Data Analysis and Visualization cloud environment;
- Automatic data interpretation via NeXus ontologies for
 - interoperability, and
 - reusability





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