# ESRF User Meeting 2025 PyFAI tutorial

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Edgar Gutiérrez Fernández 10/02/2025



### **PyFAI** tutorial - overview

- 10/02/2025 / 14:00 17:00
- Each user should use its own computer (Windows, Linuc, MacOS)
- 1<sup>st</sup> half: concepts of pyFAI
  - Motivations
  - Applications
  - Working philosophy
- Coffee break (~15 minutes)
- 2<sup>nd</sup> half: hands-on
  - Installation of python: venv/conda/visa
  - Installation of pyFAI
  - Calibration GUI
  - AzimuthalIntegratrion
  - Other pyFAI applications: integrate, diffmap-view, worker...





# PyFAI = Python Fast Azimuthal Integrator



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### X-ray scattering techniques

• Scattering is the deflection of photons upon interaction with matter.





### X-ray scattering techniques: diffraction / scattering

• If the material is **crystalline**, the scattered photons create constructive interferences, like water waves.



• Constructive interference between scattered X-ray takes place if Bragg relation is fulfilled:

• If the material is **disordered**, the scattered photons create broad distributions of intensity.



https://biosaxs.com/technique.html

• More ambiguous and hard to analyze. Usually requires complementary techniques.



### X-ray scattering techniques: diffraction / scattering

- The study of highly **crystalline** materials (metals, ceramics, oxides...) is named '**diffraction**''.
- Powder Diffraction: isotropic
  - Phase identification
  - Crystallinity
  - Lattice parameters
  - Thermal expansion
  - Phase transition
  - Strain/crystallite size

- The study of largely/inherently **disordered** materials (polymers, proteins, colloids...) is named '**scattering**".
- Wide-Angle X-ray Scattering (**WAXS**): analog to diffraction:
  - Phase identification
  - Crystalllinity/orientation
- Small-Angle X-ray Scattering (SAXS): micro/nano scale prove:
  - Particle shape/surface
  - Proteins domains
  - Protein folding
  - Colloid parameters
  - Fiber orientation



### X-ray scattering techniques: diffraction / scattering

- The study of highly crystalline materials (metals, ceramics, oxides...) is named 'diffraction''.
- Powder Diffraction: isotropic

- The study of largely/inherently **disordered** materials (polymers, proteins, colloids...) is named '**scattering**".
- **WAXS**: analog to diffraction.
- **SAXS**: micro/nano scale prove.
- Both rely on the same transformation: **2D image to azimuthal average.**



• PyFAI is the first tool to be used after data acquisition.



# pyFAI: pythonic tool to reduce 2D patterns



- Python as the widespread prog science.
  - the most accessible programming language
    - and in



- Data acquisition (BLISS)
- Data visualization (sil) / A
- Data analysis (PyMCA

PyFAI combines python API with fast algorithms written in Cython and OpenCL
 Uthon OpenCL



### Alternatives to pyFAI

### Fit2D

<sup>~</sup> MIT licensed from ESRF, written in Fortran, now discontinued

### XRDUA

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GPL licensed from U. Antwerp, written in IDL, focuses on diffraction mapping

### DAWN

EPL licensed from Diamond Light Source, written in Java

### DataSqueeze

Freeware from U. Pennsylvania

### Foxtrot

Commercial, from XENOCS & SOLEIL synchrotron, written in Java

### MAUD

Freeware from U. Trento, written in Java

### GSAS-II

- Freeware tool from U. Chicago & APS, written in Python
- Scikit-beam
  - BSD licensed from NSLS-II, written in Python

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• Import detector from detector\_factory (e.g. Pilatus1M from Dectris)

### 1) Detector instance: use pyFAI.detector\_factory

Import the module

3]: from pyFAI import detector\_factory

We just need to know the name of the detector

- [38]: my\_detector = detector\_factory(name="Eiger2\_9M")
- [39]: print(f"My detector is {my\_detector.name}\nIts shape is {my\_detector.shape} pixels")

My detector is Eiger2 9M Its shape is (3262, 3108) pixels

Set the binning

- [76]: my\_detector.set\_binning((2,2))
- [77]: print(f"My detector is {my\_detector.name}\nIts shape is {my\_detector.shape} pixels")

My detector is Eiger2 9M Its shape is (1631, 1554) pixels



• Import detector from **detector**\_factory (e.g. Pilatus1M from Dectris)

	1) Detector instance: use pyFAI.detector_factory
	Import the module
[3]:	<pre>from pyFAI import detector_factory</pre>
	We just need to know the name of the detector
[38]:	<pre>my_detector = detector_factory(name="Eiger2_9M")</pre>
[39]:	<pre>print(f"My detector is {my_detector.name}\nIts shape is {my_detector.shape} pixels")</pre>
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### Detector information: • Pixel size 1 • Pixel size 2 • Binning: default (1,1)• Max\_shape • Orientation $(1\rightarrow 4)$



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# Detector information: Pixel size 1 Pixel size 2 Binning: default (1,1) Max\_shape Orientation (1→4)



### **Detector instance**

• Import detector from **detector**\_factory (e.g. Pilatus1M from Dectris)





### Geometry instance

- A geometry is fully defined by:
  - A detector instance
  - Sample to detector **distance** (in meters)
  - Wavelength of the beam (in meters)
  - Three rotations of the detector
  - Coordinates of the point of normal incidence (**PONI**), from the sample to the detector plane.

Normally, the calibration of the geometry is fully done through the graphical interface.





# Calibration of geometry: pyFAI-calib2

- Calibration is made after measuring a standard sample:
  - LaB6, Cr2O3, AgBh...
- Choosing the correct detector (+ orientation, binning, mask...)
- Selecting the Debye-Scherrer rings associated to the standard
- Fitting the rings
- Refinement
- Validation
- Saving of .poni file



.poni file





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.poni file













# PyFAI: creating an integrator

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Integrator		
<pre>home &gt; edgar &gt; E eiger9m_bin.poni     # Nota: C-Order, 1 refers to the Y axis, 2 to the X axis     # Calibration done on Sat Feb 8 15:34:34 2025     poni_version: 2.1     Detector: Eiger2_9M     Detector_config: {"orientation": 3}     Distance: 0.1850492110887503     Pon11: 0.13007134675074164     Pon22: 0.1124624632248358     Rot12: 0.012061870930842628     Rot2: -0.012061870930842628     Rot2: 0.012061870930842628     Rot3: 0.0     Wavelength: 3.75709692221819e-11     # Calibrant: Ce02     # Image: Eiger2_bin2_Ce02_33keV.h5 </pre>	[130]: [131]:	After using pyFAI-calib2, we load the poni file <pre>from pyFAI import load file_poni = "eiger9m_bin.poni" ai = load(filename=file_poni) ai.detector = my_detector  print(f"Information from the AzimuthalIntegrator:\n\n {ai}") Information from the AzimuthalIntegrator: Detector Eiger2 9M PixelSize= 150µm, 150µm BottomRight (3) Wavelength= 3.757097e-11 m SampleDetDist= 1.850492e-01 m PONI= 1.300713e-01, 1.124625e-01 m rot1=0.014603 rot2=-0.012062 rot3=0.000000 rad DirectBeamDist= 185.082 mm Center: x=731.734, y=852.260 pix Tilt= 1.085° tiltPlanRotation= -140.440° λ= 0.376Å</pre>



### PyFAI: creating an integrator



### After using pyFAI-calib2, we load the poni file

[130]: from pyFAI import load file poni = "eiger9m bin.poni" ai = load(filename=file poni) ai.detector = my detector

### [131]: print(f"Information from the AzimuthalIntegrator:\n\n {ai}")

Information from the AzimuthalIntegrator:

Detector Eiger2 9M PixelSize= 150µm, 150µm BottomRight (3) Wavelength= 3.757097e-11 m SampleDetDist= 1.850492e-01 m PONI= 1.300713e-01, 1.124625e-01 m rot1=0.014603 rot2=-0.012062 rot3=0.000000 rad DirectBeamDist= 185.082 mm Center: x=731.734, y=852.260 pix Tilt= 1.085° tiltPlanRotation= -140.440° λ= 0.376Å

### Novelty in pyFAI 2025.01: FiberIntegrator

[132]: fiber integrator = load(filename=file poni, type ="pyFAI.integrator.fiber.FiberIntegrator",

[133]: print(f"Information from the FiberIntegrator:\n\n {fiber integrator}")

Information from the FiberIntegrator:

Detector Eiger2 9M PixelSize= 75µm, 75µm BottomRight (3) Wavelength= 3.757097e-11 m SampleDetDist= 1.850492e-01 m PONI= 1.300713e-01, 1.124625e-01 m rot1=0.014603 rot2=-0.012062 rot3=0.000000 rad DirectBeamDist= 185.082 mm Center: x=1463.468, y=1704.519 pix Tilt= 1.085° tiltPlanRotation= -140.440° λ= 0.376Å Incident angle: 0.00° Tilt angle 0.00° Sample orientation 1

\*FiberIntegrator is targeted to Grazing Incidence experiments









# PyFAI: loading data

- Importing data is made through other python packages:
  - FabIO
  - Silx
  - h5py
- Common file formats:
  - .edf
  - .tiff
  - .h5
- Visualizing is made through:
  - matplotlib
  - seaborn
  - silx



# PyFAI: loading data



# PyFAI: loading data



### What happens during an integration

- 1) Get the coordinates of every corner of every pixel of the detector (in meters).
- <sup>2</sup> 3 coordinates per corner, 4 corners per pixel
- Detector of  $1000 \times 1000 = 10^6$  pixels = 1Mpix \* 4 (corners) \* 3 (coordinates) \* 4 (bytes) = 48 Mbytes
- 2) Offset the detector's origin to the PONI and rotate around the sample.
- 3) Calculate the radial (2theta) and azimuthal (chi) positions of each corner.
- 4) Calculate normalization matrix: polarization, solid-angle, flat-field...
- 5) Assign each pixel to one or multiple **bins**.
- 6) Histogram bin position with associated intensities
- 7) Histogram bin position with **associated normalizations**
- 8) Return bin position and the ratio of **sum of intensities / sum of norm.**



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### Integration algorithms

### Histogram

- <sup>~</sup> Pixel by pixel procedure.
- <sup>2</sup> Each pixel is split over the bins it covers.
- Corner coordinates have to be calculated (4x slower initialization).
- The slow down is function of the oversampling factor, for every image.
- Serial read  $\rightarrow$  Random write



### Sparse Matrix Multiplication

- Bin by bin procedure.
- Creates and stores a sparse matrix with all the integration information.
- The sparse matrix can be huge: longer initialization related to the oversampling factor.
- No performance penalty on the integration itself.
- Serial write ← Random read





### **Pixel splitting**

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No splitting: one pixel to one single bin upon in which bin the center of the pixel is falling.





The intensity of each bin is the sum of the intensity of the pixels whose center falls into the radius bin



### **Pixel splitting**

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**Splitting**: each pixel intensity is shared between consecutive bins.







Python

- Use numpy methods
- No initialization
- . Slow, no cache
- Non-recommended (not popular)

### Cython

- . Use cython methods
- No initialization
- Faster than python, no cache
  - Recommended to integrate 10s frames

**P**ython

### OpenCL

Cache

- Use parallelization through CPU/GPU
- Initialization ~1-3 s



- Potentially fast, depending on the GPU
- Best option for large data

[15]: # Method = ("split-pixel scheme, algorithm, implementation")
method\_default = ("bbox", "csr", "cython")
method\_fast = ("bbox", "csr", "opencl")



# pyFAI interfaces

- Applications:
  - GUI applications: pyFAI-calib2, pyFAI-integrate, pyFAI-diffmap-view
  - Batch mode: worker
  - Scriptable applications: pyFAI-average, pyFAI-saxs, pyFAI-waxs, diff\_tomo
- Python interface:
  - High level, direct API: scripts, Jupyter notebook
  - Mid level: manually creation of detectors, integrators, units...
  - Low level: manually setting the integrator engines

It is up to the user to choose the way he/she uses pyFAI



### Latest news from pyFAI

- PyFAI-2025.1.0: release on 31/01/2024
- Median filter (cython, OpenCL)
- Unified WorkerConfiguration
- New API for Grazing Incidence experiments



### Project management: Silx & pyFAI

- Compatible with Windows, MacOS, Linux
- MIT licensed: compatible with both science and business
- **PyFAI** is embedded in the silx-kit project: <u>https://github.com/silx-kit/</u>
- **Silx-kit** project is python-based, developed at the ESRF and includes:
- Open to collaborations:
- About 20 direct contributors from ESRF, from other synchrotrons, XFELs (Soleil, NSLS-II, Petra-III, Eu-XFEL) and companies (Xenocs)
- Used by ~90 other projects from many other X-ray sources in the world (SLAC, ALS, APS, ALBA, NSLS-II, Petra-III, Soleil, Diamond, SLS, MaxIV...)









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### **PyFAI** is used in most European and American synchrotrons/FELs



User support is provided via the mailing list: pyFAI@esrf.fr (185 subscribers)

- Bugs are discussed via Github issue tracker
  - https://github.com/silx-kit/pyFAI/issues



### Reasons to choose pyFAI

### Faster than others

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- <sup>~</sup> First tool using sparse matrix multiplication to perform integration
- <sup>~</sup> All computation intensive kernels are ported to C, C++ or OpenCL
- <sup>~</sup> PyFAI is the only azimuthal integration tool benefiting from GPU

# Versatile (increasing with every version)

- Wide space to vary the integration protocol
- <sup>~</sup> Generic geometry
- Most detectors already defined (+ custom detector through Nexus file)
- <sup>~</sup> Graphical user interface alternatives (thanks to Valentin Valls)



### Acknowledgements

- Main author: Jerome Kieffer
- **Contributors: 43**

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- Former data analysis unit colleagues:
- ~ Valentin Valls
- <sup>~</sup> Loic Huder
- <sup>~</sup> Thomas Vincent
- <sup>~</sup> Claudio Ferrero

### **ESRF Beamlines:**

 <sup>~</sup> BM01, BM02, ID02, ID11, ID13, ID15a, ID15b, ID21, ID22, ID23, BM26, ID27, BM28, ID28, BM29, ID29, ID30, ID31...

### Trainees:

- Aurore Deschildre
- Frederic Sulzmann
- Guillaume Bonamis
- Other synchrotron/labs
  - Soleil: Fred Picca
  - Clemens Prescher (Dioptas)
  - Sesame: Philipp Hans
  - NSLS-II, ALS, APS...
- International Grants:
  - LinkSCEEM-2 grant:
    - . Dimitris Karkoulis
    - . Giannis Ashiotis
    - · Zubair Nawaz



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