



STREAMLINE

# Real-time pre-processing for serial crystallography

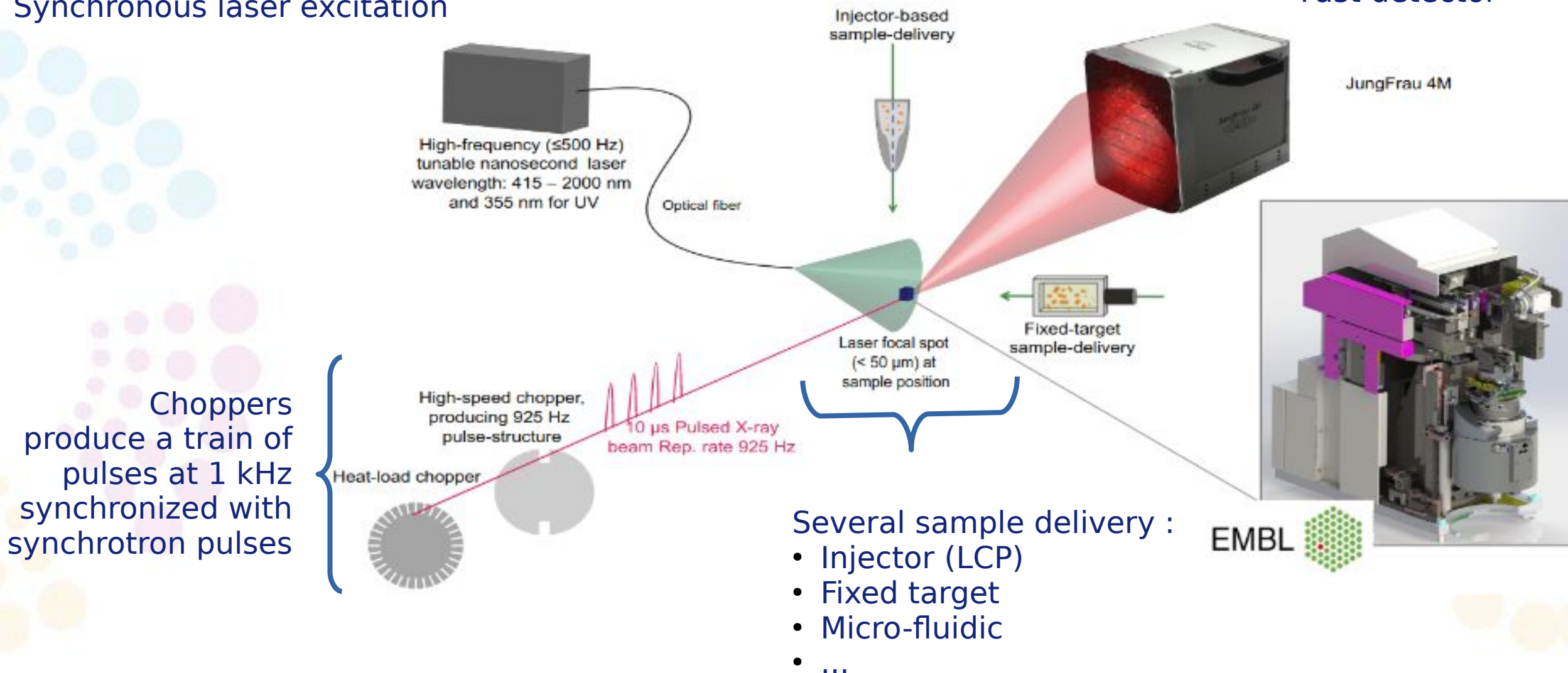
**Jérôme Kieffer**<sup>1</sup>, Nicolas Coquelle<sup>1</sup>, Gianluca Santoni<sup>1</sup>,  
Shibom Basu<sup>2</sup>, Samuel Debionne<sup>1</sup>, Alejandro Homs<sup>1</sup>,  
Andy Götz<sup>1</sup>, Daniele De Sanctis<sup>1</sup>.

<sup>1</sup>ESRF - Grenoble (France), <sup>2</sup>EMBL - Grenoble (France)

- Serial crystallography at the ESRF ID29 beamline
- Image analysis for single crystal frames
- Lossy data compression
- Peak-finding
- Conclusions

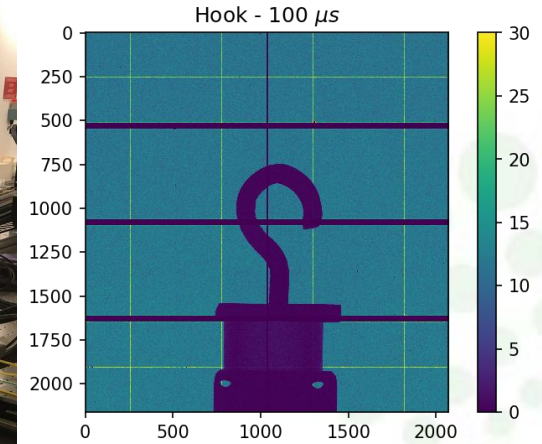
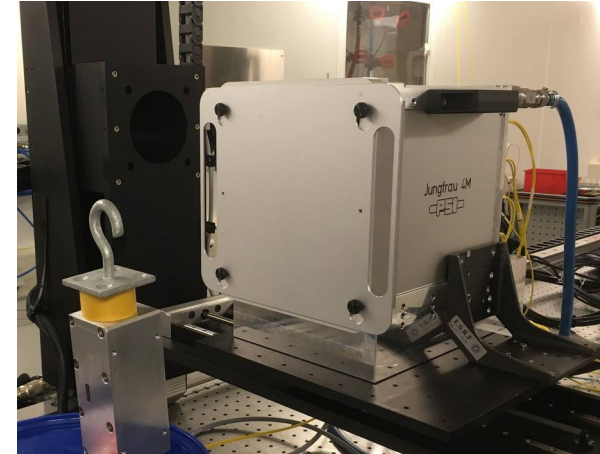
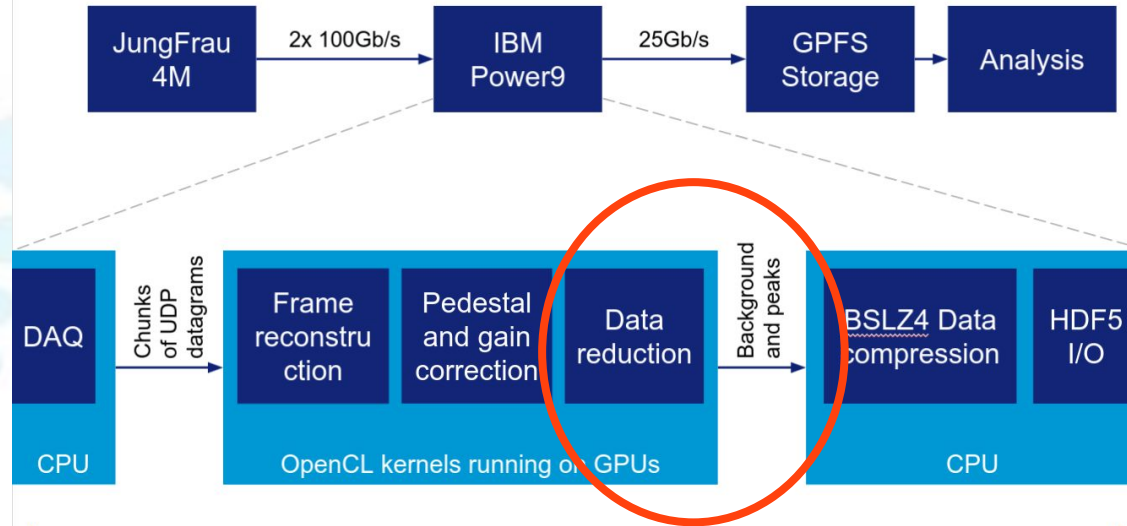
## Synchronous laser excitation

## Fast detector



Credit: Julien Orlans

# Lima2 controls the Jungfrau 4M detector

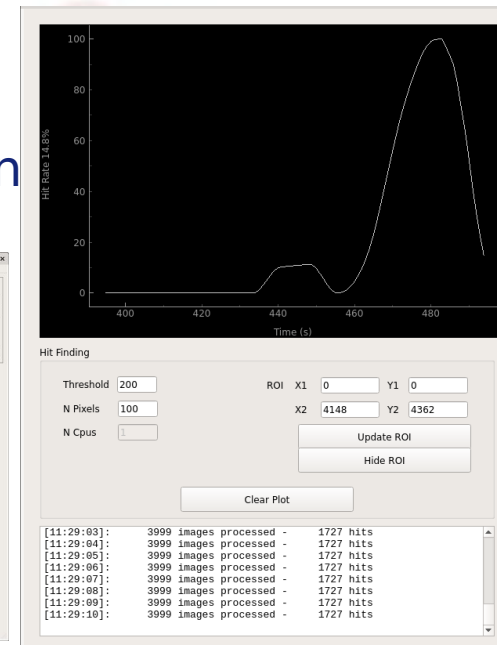
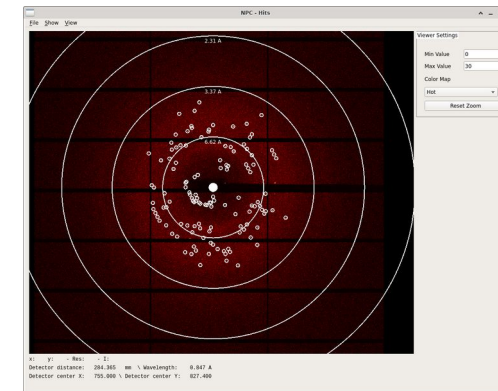


LIMA2 Receivers  
multiple instances  
collaborating through MPI

Pipeline  
running  
@2KHz



## NanoPeakCell: Live feedback of peak position during acquisition



Debionne, S., Homs, A., Claustre, L., Kieffer, J., De Sanctis, D., Santoni, G., Goetz, A. & Meyer, J. (2022). In Proceedings of the 14 th international conference on Synchrotron Radiation Instrumentation (SRI2021). <https://indico.desy.de/event/27430/abstracts/>



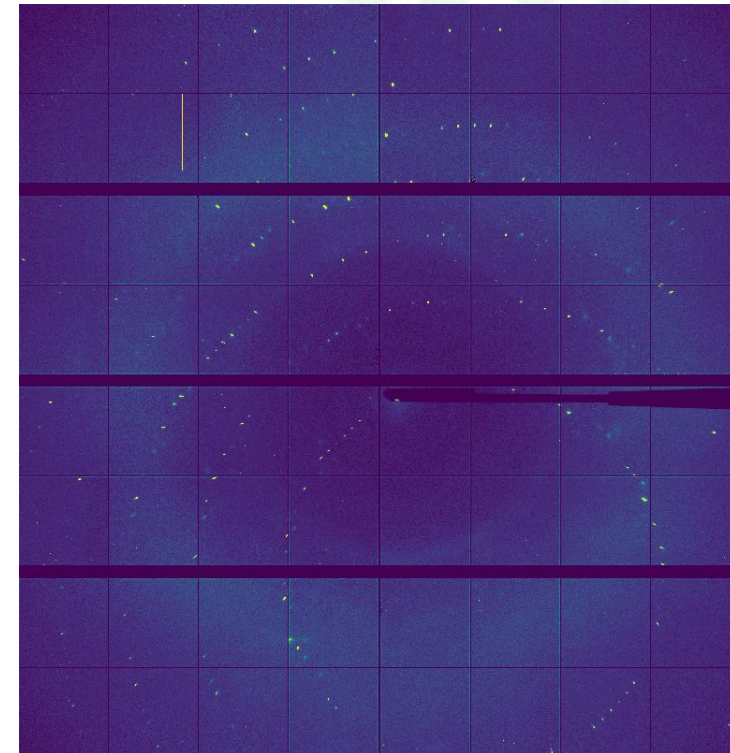
- Integration
- Indexing ← Most difficult
- Peak-finding
- Intense pixel saving } This contribution
- Veto algorithm
  - Leonarsky & al. Struct.Dyn. 7, 014305 (2020)
- Image reconstruction
  - Debionne & al., SRI2021
- Dump data to disk



Holton J. M., see  
[https://bl831.als.lbl.gov/~jamesh/lossy\\_compression/](https://bl831.als.lbl.gov/~jamesh/lossy_compression/)

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First diffraction image  
obtained at the ID29



- Pixel intensity needs to be corrected:

$$I_{cor} = \frac{I_{raw} - I_{dark}}{F \cdot \Omega \cdot P \cdot A \cdot I_0} = \frac{signal}{normalization}$$

- Intensity average per ring:

- Pixel splitting:  $c_{i,r}$  is the fraction of pixel  $i$  in the ring  $r$
- Normalization issue due to polarization, ...

→ this is a weighted average:  
implemented in pyFAI

$$\bar{I}_r = \frac{\sum_{i \in bin_r} c_{i,r} \cdot signal_i}{\sum_{i \in bin_r} c_{i,r} \cdot normalization_i} = \frac{V_{bin_r}}{\Omega_{bin_r}}$$

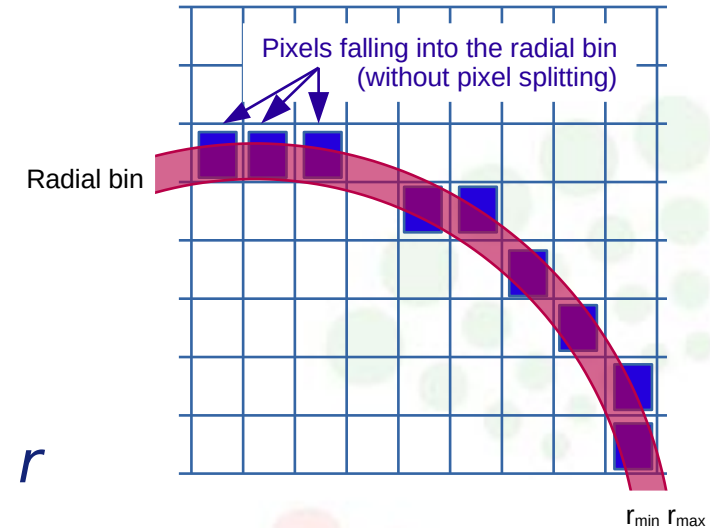
- Use of accumulators:

- Simplifies notation
- Suitable for parallel reduction

$$V = \sum \omega \cdot v = \sum c \cdot signal$$

$$\Omega = \sum \omega = \sum c \cdot normalization$$

$$\Omega \Omega = \sum \omega^2 = \sum c^2 \cdot normalization^2$$



# Uncertainties in azimuthal integration (1)

- Uncertainties on the average value
  - Called *sem* and reported by pyFAI
  - Not of interest for background evaluation

$$\sigma(\bar{I}_r) = \frac{\sqrt{\sum_{i \in \text{bin}_r} c_i^2 \cdot \text{variance}_i}}{\sum_{i \in \text{bin}_r} c_i \cdot \text{normalization}_i} = \frac{\sqrt{VW_r}}{\Omega_r}$$

- Uncertainties on pixel value
  - Called *std* and larger than *sem* by a factor  $\sqrt{N}$

$$\sigma(I_r) = \sqrt{\frac{\sum_{i \in \text{bin}_r} c_i^2 \cdot \text{variance}_i}{\sum_{i \in \text{bin}_r} c_i^2 \cdot \text{normalization}_i^2}} = \sqrt{\frac{VW_r}{\Omega \Omega_r}}$$

- Poisson error model:
  - For all pixels belonging to a common distribution:  
 $\text{variance} = \langle \text{signal} \rangle$
  - Usually simplified in:

$$\begin{cases} \text{variance}_i = \text{signal}_i \\ VW = \sum c^2 \cdot \text{signal} \end{cases}$$

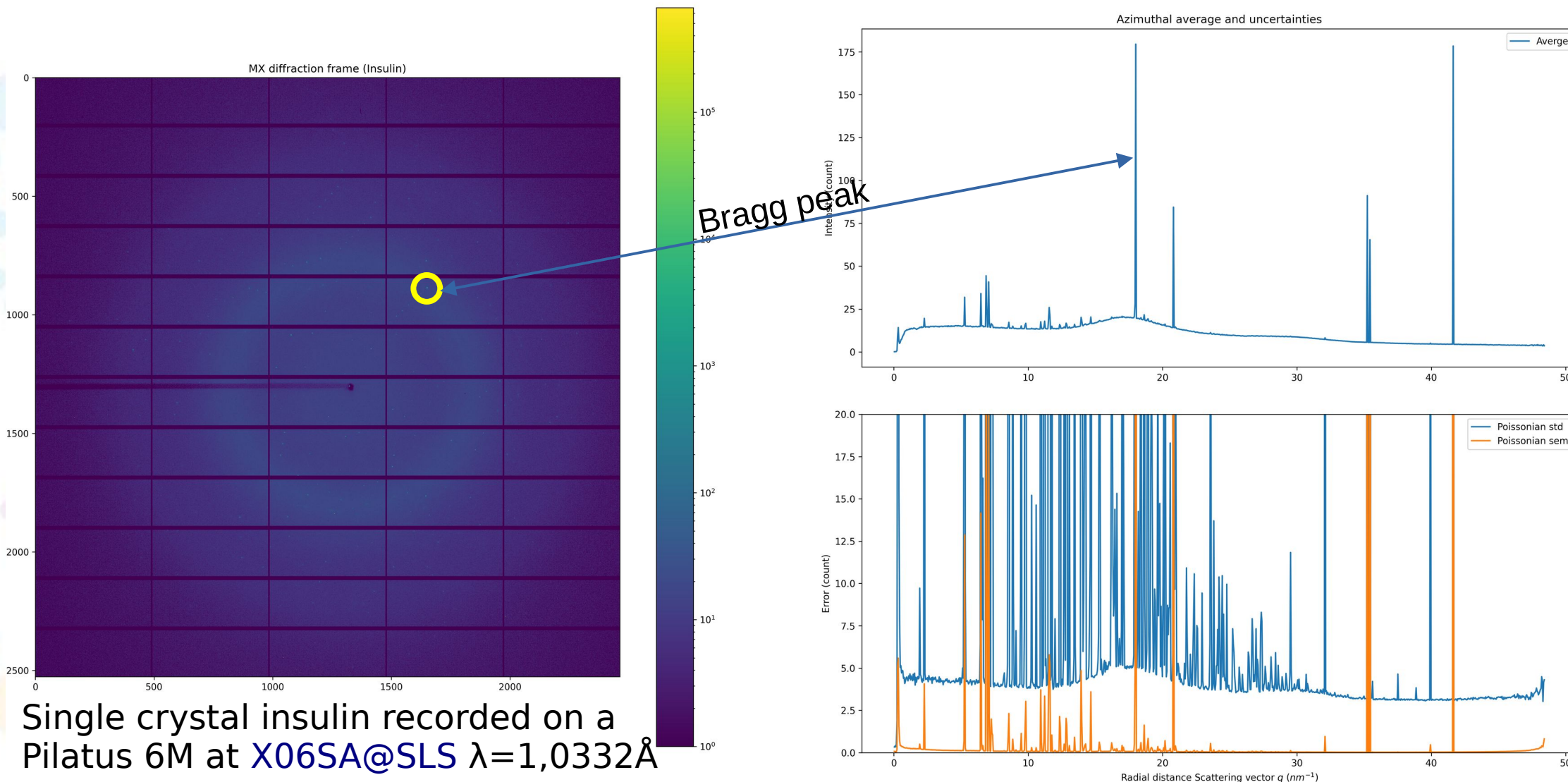
$$V = \sum \omega \cdot v = \sum c \cdot \text{signal}$$

$$\Omega = \sum \omega = \sum c \cdot \text{normalization}$$

$$\Omega \Omega = \sum \omega^2 = \sum c^2 \cdot \text{normalization}^2$$



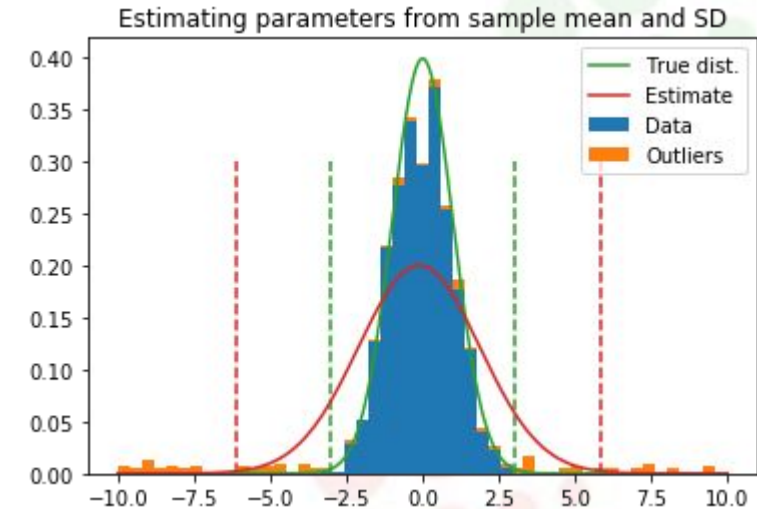
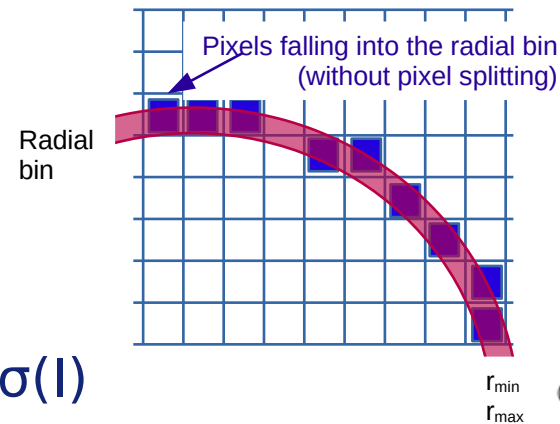
# Example on an insulin diffraction frame:



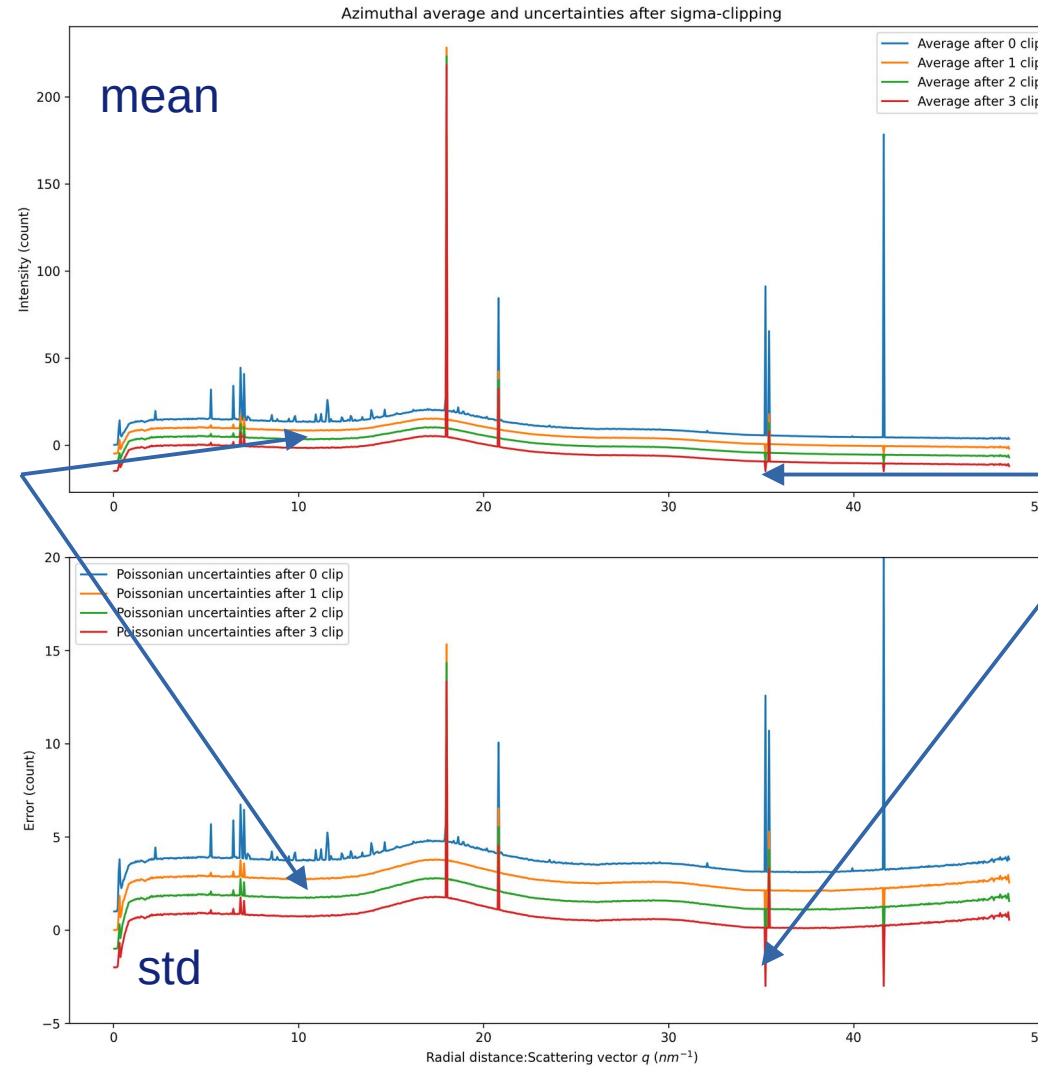
Single crystal insulin recorded on a Pilatus 6M at X06SA@SLS  $\lambda = 1.0332 \text{ \AA}$

# Sigma-clipping

- Iterative algorithm:
  - Integrate to calculate  $\bar{I}$  and  $\sigma(I)$
  - Mask out any pixel with:  $|I - \bar{I}| > n \cdot \sigma(I)$
- Removes both tails from the distribution:
- Good approximation of the background
- Number of iterations:
  - 3 to 5 are common
- Cut-off parameter (SNR)
  - Default value provided by Chauvenet:
  - Discard at worse 1 pixel per ring per cycle on a normal distribution
  - Depends on the size, thus on the number of bins:  $SNR_{clip} = 2.7 \sim 3.5$



$$SNR_{chauvenet} = \sqrt{2 \log \left( \frac{N}{\sqrt{2 \pi}} \right)}$$



Remove most peaks with few cycles

Empty bins results with mean=0 & std=0

Jeopardizes subsequent analysis

- Limits of the Poisson error model:
  - Requires all pixels in a ring to be from the **same** distribution
  - Thus incompatible with Bragg-peaks!
  - Consider for example a distribution of 2 pixels of value 1 and 99:
    - Mean: 50, std: 10, both pixels are at  $5\sigma$  -> empty ensemble

- Azimuthal error model:
 
$$\left\{ \begin{array}{l} \text{variance}_i = \omega_i^2 \cdot (v_i - \bar{v}_r)^2 \\ \text{VV} = \sum \omega^2 \cdot \left( \frac{\text{signal}}{\text{normalization}} - \frac{V}{\Omega} \right)^2 \end{array} \right.$$

- Single-pass implemented with:

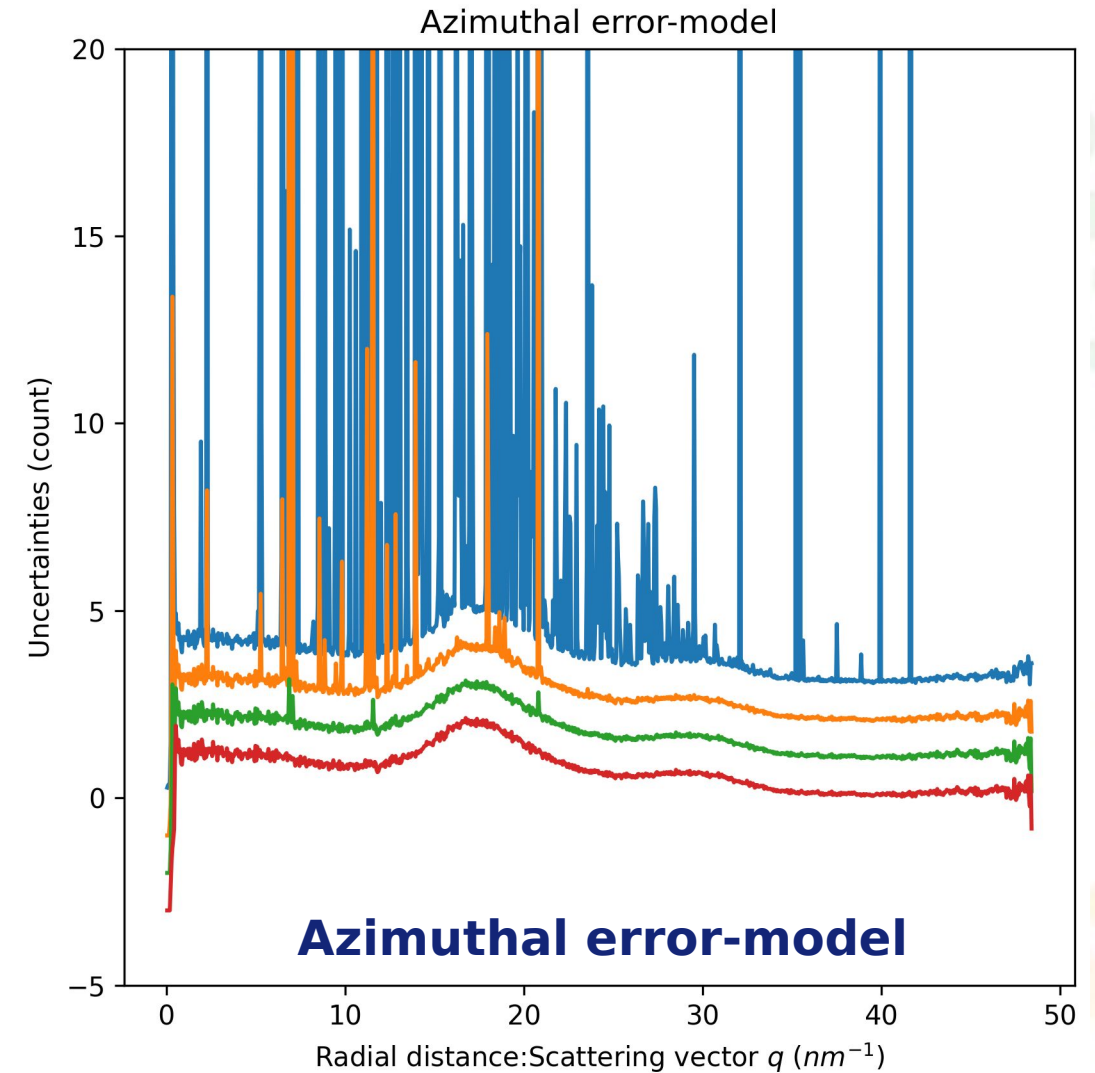
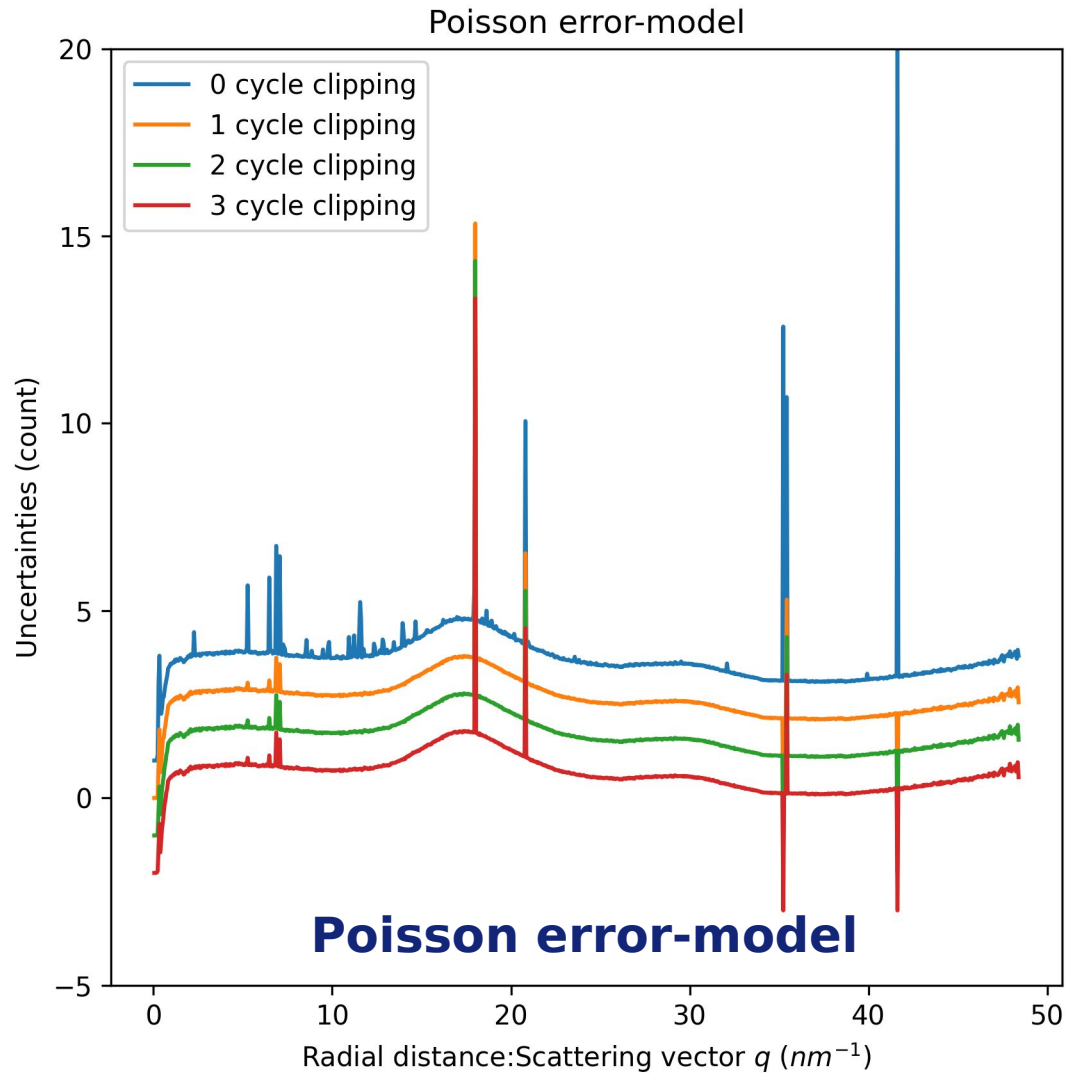
$$\text{VV}_{A \cup b} = \text{VV}_A + \omega_b^2 \left( v_b - \frac{V_A}{\Omega_A} \right) \left( v_b - \frac{V_{A \cup b}}{\Omega_{A \cup b}} \right)$$

$$V_{A \cup b} = \sum \omega \cdot v = V_A + \omega_b \cdot v_b$$

$$\Omega_{A \cup b} = \sum \omega = \Omega_A + \omega_B$$

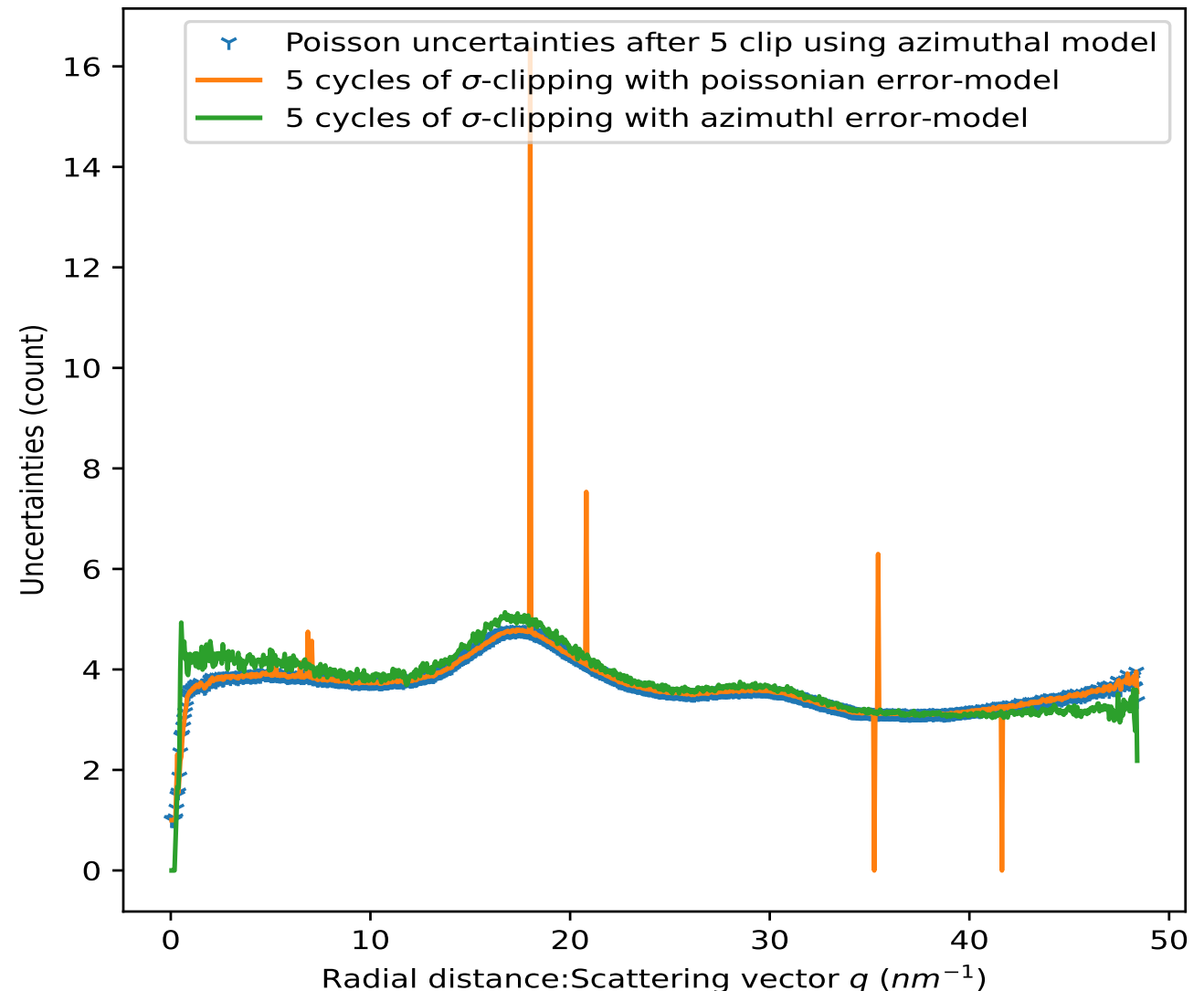
$$\Omega \Omega_{A \cup b} = \sum \omega^2 = \Omega \Omega_A + \omega_B^2$$





- Use azimuthal model for  $\sigma$ -clipping
  - Robust to Bragg-peaks
- Use Poisson model for subsequent analysis
  - Less noisy
  - Limits of Poisson when count  $\rightarrow 0$

Uncertainties from different error-models after  $\sigma$ -clipping

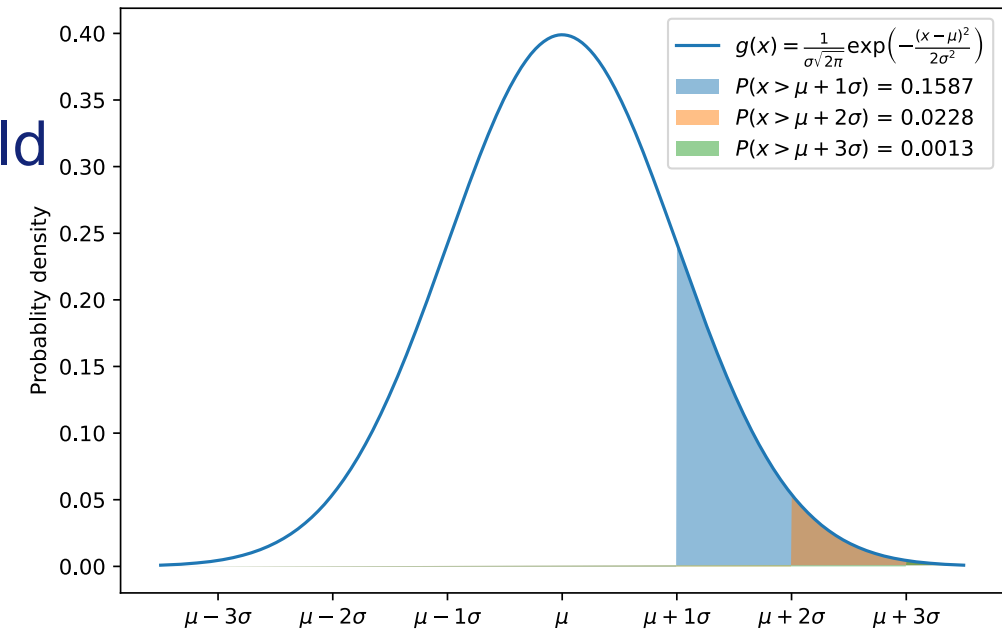


# Save only intensity of pixel of interest

- Serial crystallography at the ESRF ID29 beamline
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- Sparsification:

- Store positive outlier with  $\text{SNR} > \text{threshold}$
- Record also its position
- Record background avg ( $\mu$ ) & std ( $\sigma$ )
- Compression-rate can be estimated assuming a normal distribution
- Implemented using OpenCL in pyFAI



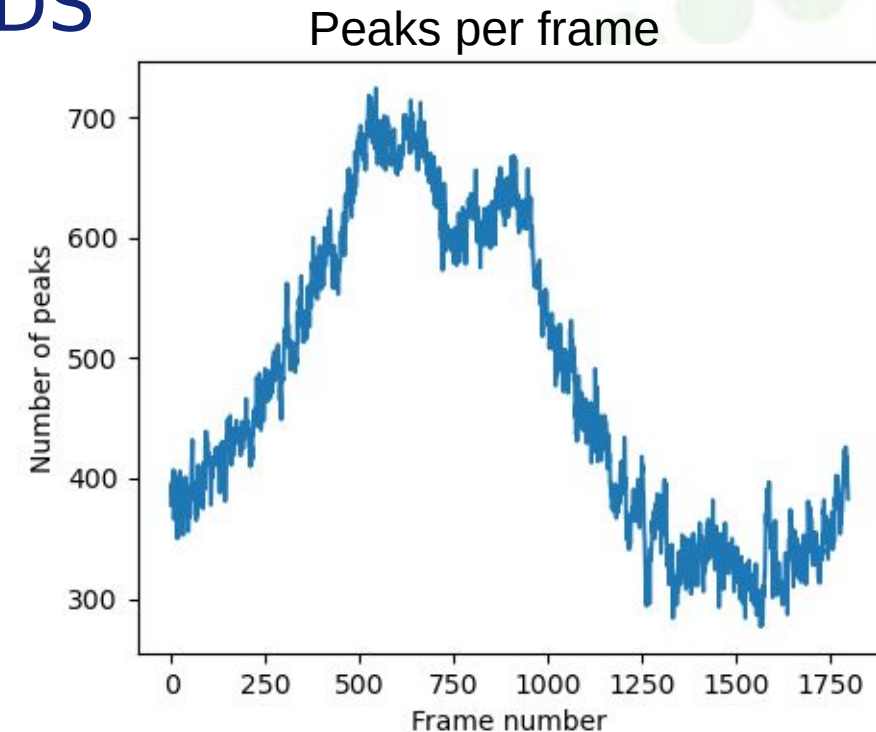
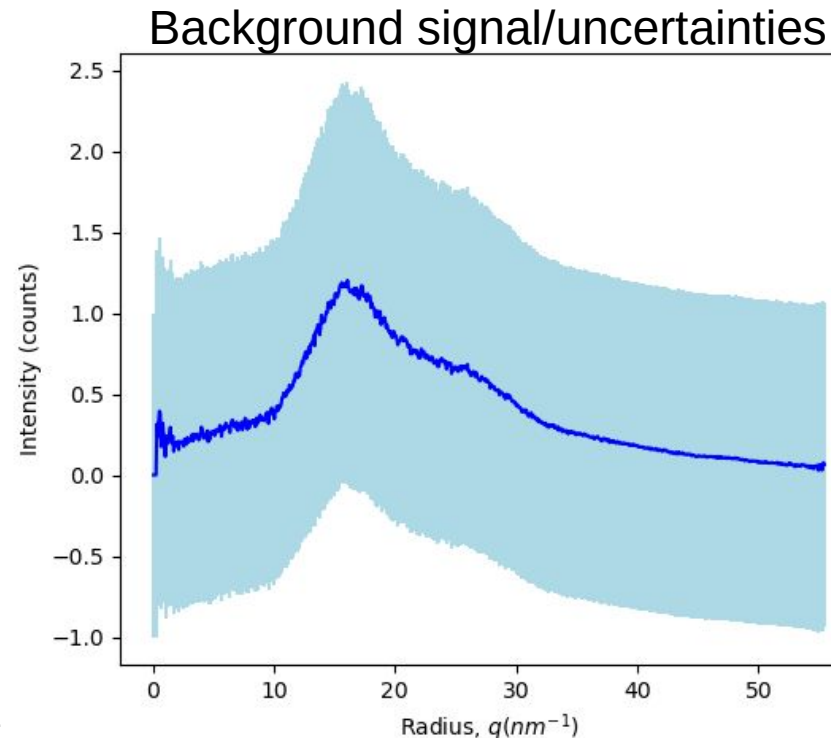
- Densification:

- Available as part of FabIO
- Restores frames with (or without) background noise
- Implemented in C (GIL-free) + multi-threading



# Validation of sparsified dataset:

- Raw dataset: Insulin acquired at SLS with an Eiger4M
- Comparison of quality indicator from XDS
- Sparse data compressed with:
  - Poissonian error-model
  - $\text{SNR}_{\text{clip}}$ : automatic
  - $\text{SNR}_{\text{pick}}$ :  $1\sigma$
  - $\text{SNR}_{\text{peak}}$ :  $5\sigma$
  - Cycles: 5
  - Bins: 800



Signal null at large  $q$

- std tend to 1
- pixels  $\geq 2$  get recorded

# Performances & quality:

- Compression of a factor: **5x** when cut-of at  $1\sigma$
- Compression speed: **250 fps** (GPU)
- Decompression speed: **200 fps** (CPU)
- Limits of the Poisson model at low count rate :  $\mu=0 \rightarrow \sigma=1$

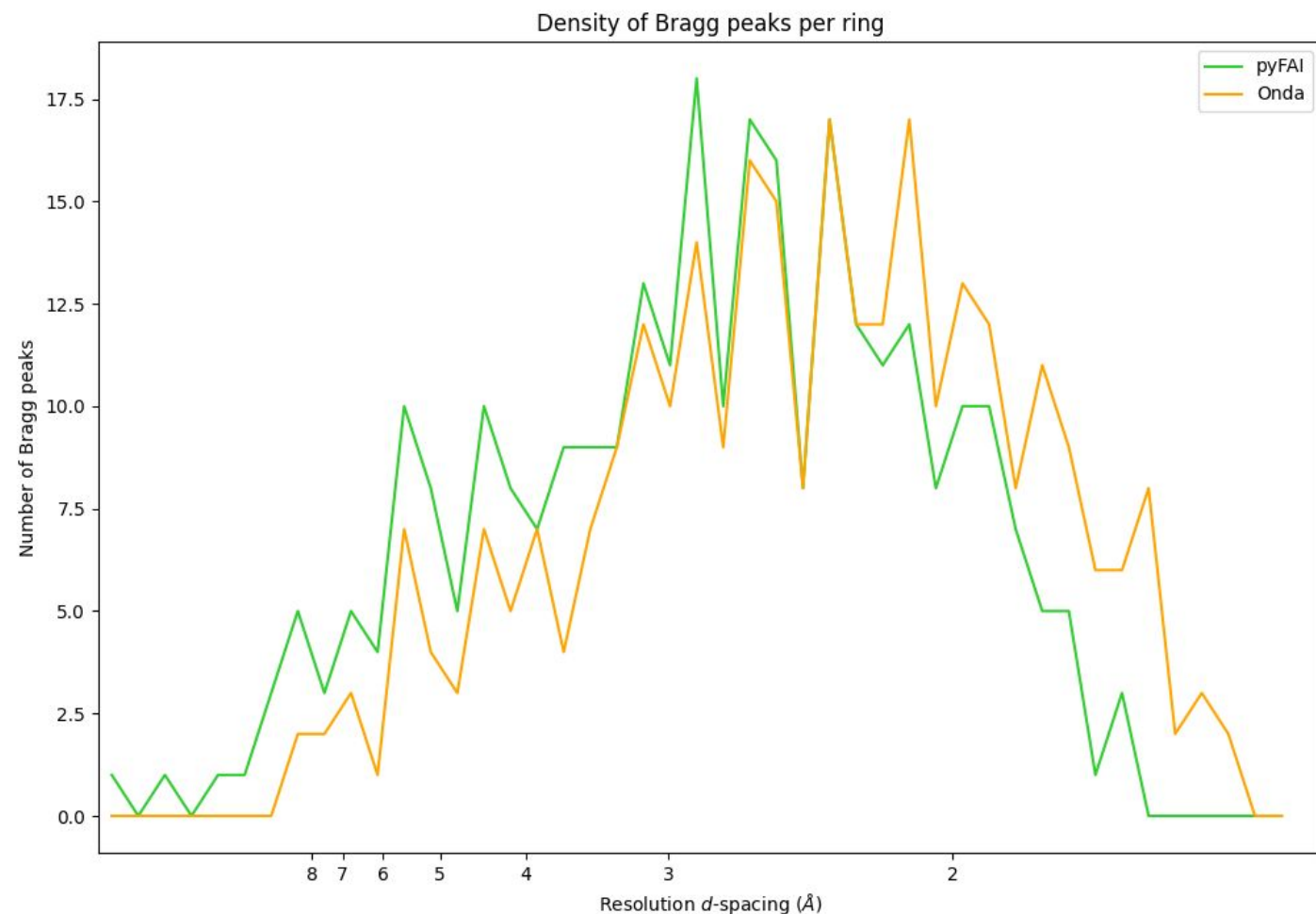
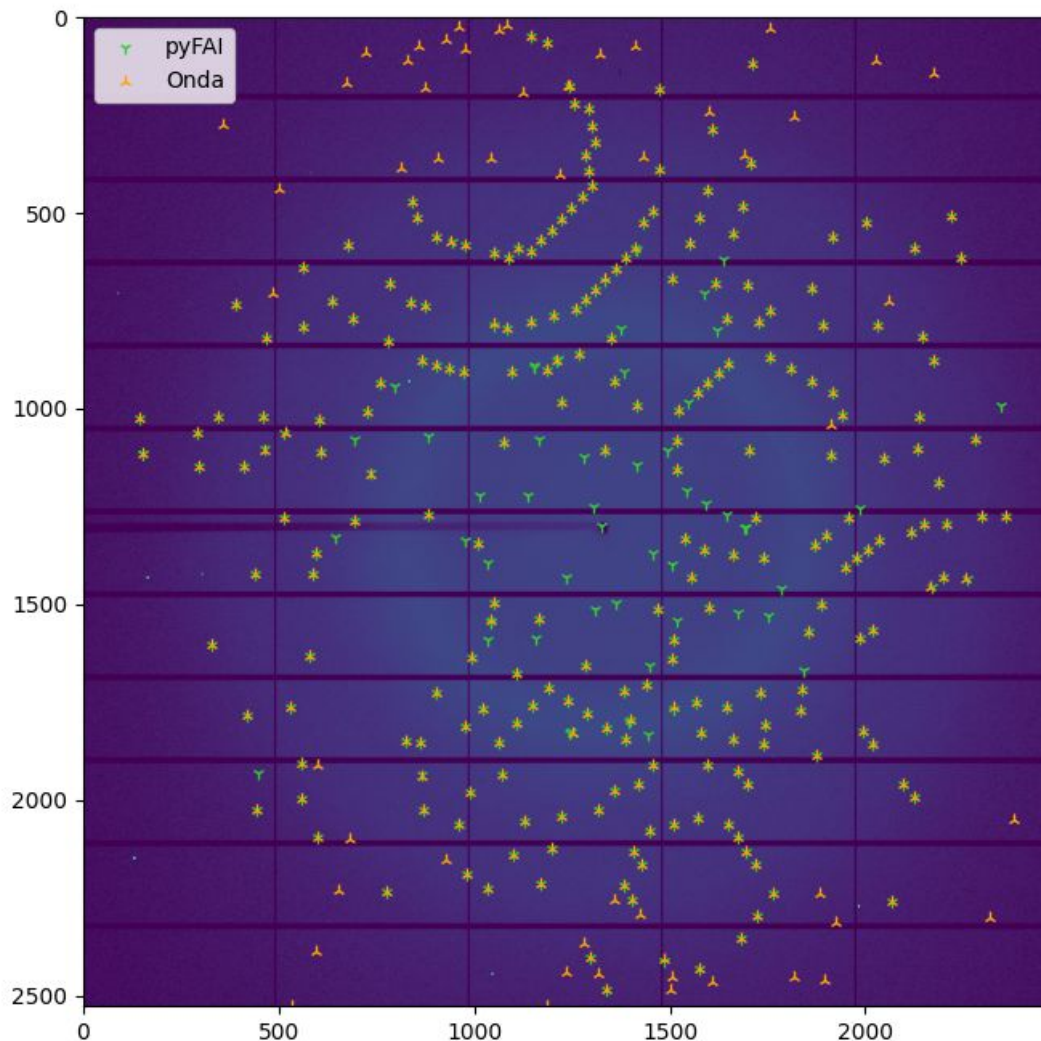
Indicator	Raw data			Spasified ( $1\sigma$ , poisson) + densified (noise)		
Size	2357 MB			439 MB		
Shell	2.91 Å	2.06 Å	total	2.91 Å	2.06 Å	total
Completeness	99.8 %	93.7 %	92.9%	99.8 %	94.1 %	93.2 %
$R_{\text{obs}}$	9.8 %	56.9 %	12.4%	8.9 %	67.8 %	11.0 %
$R_{\text{exp}}$	8.7 %	73.7 %	14.7%	8.0 %	85.6 %	12.0 %
$R_{\text{meas}}$	10.3 %	60.8 %	13.1%	9.3 %	72.6 %	11.6 %
$CC_{1/2}$	99.7	94.6	99.7	99.7	94.4	99.8
$I/\sigma$	25.86	5.38	10.54	26.85	3.70	10.14

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# Layout of the peak-picking algorithm:

- Subtract background intensity (from  $\sigma$ -clipping)
  - Clip to 0 negative values. Those are all discarded.
- Pixel is a peak if:
  - Maximum within the local neighborhood (3x3 or 5x5)
  - Subtracted signal is greater than a picking threshold ( $\text{SNR}_{\text{pick}}$ )
  - At least 2 or 3 other pixels in the neighborhood meet the  $\text{SNR}_{\text{pick}}$  criteria
- Then:
  - Sum subtracted intensities on the neighborhood (+ uncertainties propagation)
  - Calculate the center of mass of the peak
- Implemented on GPU using OpenCL
  - Same execution time as sparsification





OnDA : Mariani, V et al. J. Appl. Cryst. 49, 1073-1080 (2016).

Cheetah: Barty, A. et al., J. Appl. Crystallogr. 47, 1118–1131 (2014).

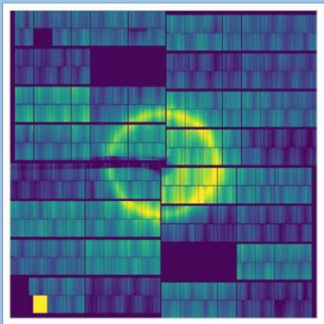
Indexer :	XGANDALF		XGANDALF-Fast	
Peak-picker	Indexation rate	Run-time	Indexation rate	Run-time
Zaef	10 %	2178 s	10 %	430 s
PeakFinder8	49.5 %	10397 s	48.5 %	1757 s
PeakFinder9	44.2 %	8328 s	43.5 %	1436 s
Robust PeakFinder	22.4 %	6314 s	21.2 %	1628 s
PyFAI peakfinder	50.2 %	9325 s	50.0 %	1595 s

1000 micro-crystal from HEWL Lysozyme collected on an Eiger 4M at ESRF-ID30a3

- Separation of Bragg-peaks from amorphous background using  $\sigma$ -clipping
  - Several error-models: Poisson, azimuthal and hybrid
  - Performance critical section for all algorithms ( $\sim 3$ -4 ms for 4 Mpix)
- Sparse & lossy data compression for single crystal diffraction
  - Compression rate 5-100x (tuneable thanks to  $\text{SNR}_{\text{pick}}$ )
  - Compression speed: 250 fps, single GPU stream
  - Decompression on CPU with background reconstruction
  - Data quality validated with XDS reduction software
- Peak-finder
  - Similar in many point to the PeakFiner8 from Cheetah (Barty, 2014)
  - Implemented on GPU @ 250 fps
  - Peak-position validated by indexing with CrystFEL

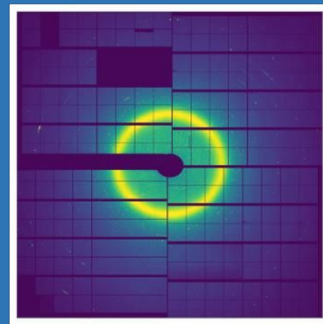
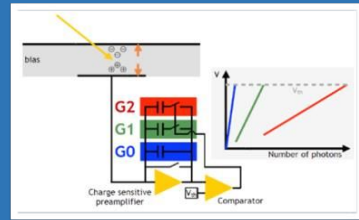
- Modify CrystFEL to be able to read sparse-frames
- Implement it online at 1kHz within LImA2 (needs 4 GPUs in //)

## ► Image Reconstruction



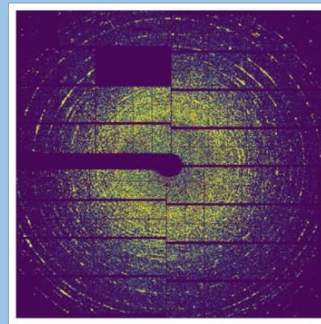
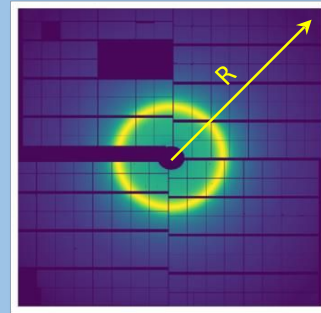
UDP packets data from detector are geometrically assembled (CPU)

## ► Background and Gain Corrections



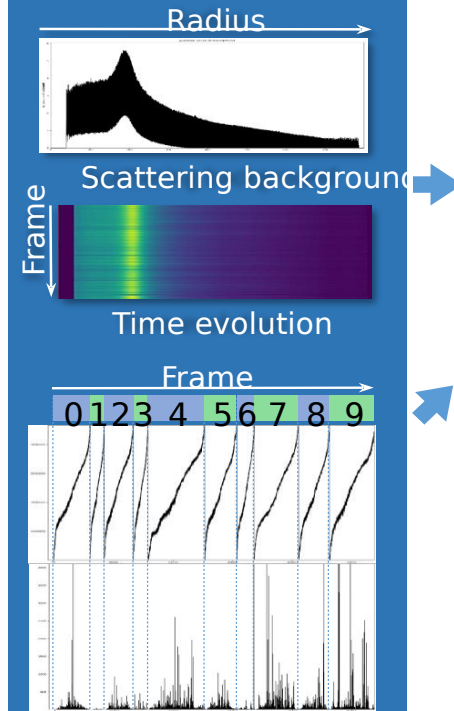
Per pixel & per frame gain selection: 3 pedestal + 3 gain maps

## ► pyFAI-based Data Reduction



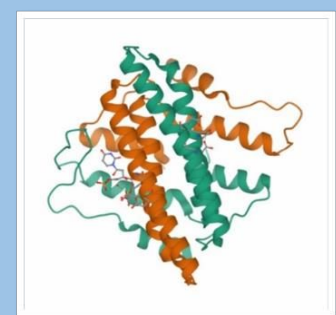
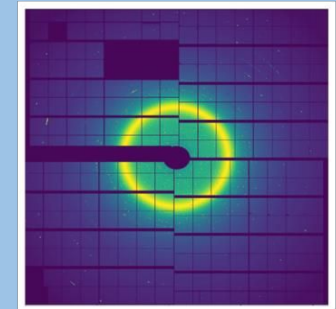
Peak finding and background extraction using Sigma Clipping

## ► Sparse Data Saving



Peak pixel position & intensity per frame (CSR)

## ► Offline re-densification



Performed to feed SMX data processing software