

From Data to Structure in Neutron Total Scattering at the CSNS Multi-Physics Instrument (CSNS-MPI)

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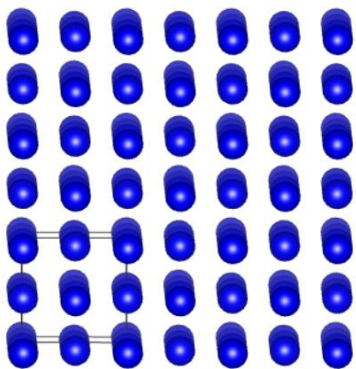
China Spallation Neutron Source Center

ICANS XXV March 2026 Malmö

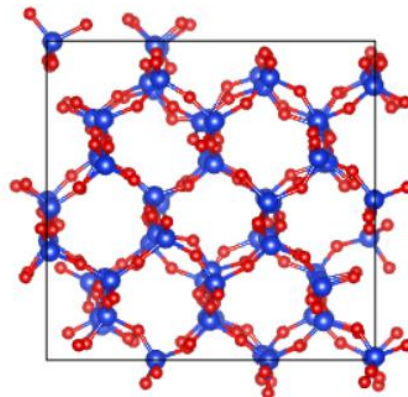
- **Neutron Total Scattering and PDF**
Physical Foundations for Probing Multi-range Structure
- **Data Processing and Modeling**
From Raw Scattering Data to Reliable Structural Models
- **Intelligent Enhancement**
Advanced Algorithms and AI Empowering Structure Analysis
- **Summary and Outlook**

Multis-order structures

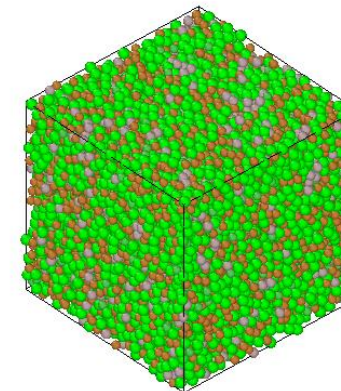
Long range ordered crystal



Local ordered/disordered structure

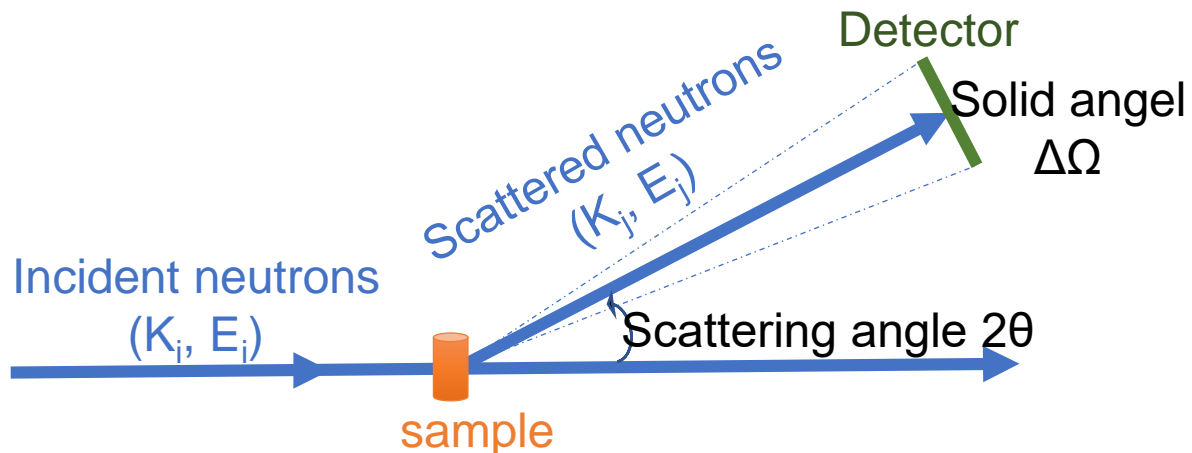


Disordered material



Can analysis matter structure in different order, from atomic scale to nanoscale.

Elastic interaction of neutrons with matter



Elastic scattering section:

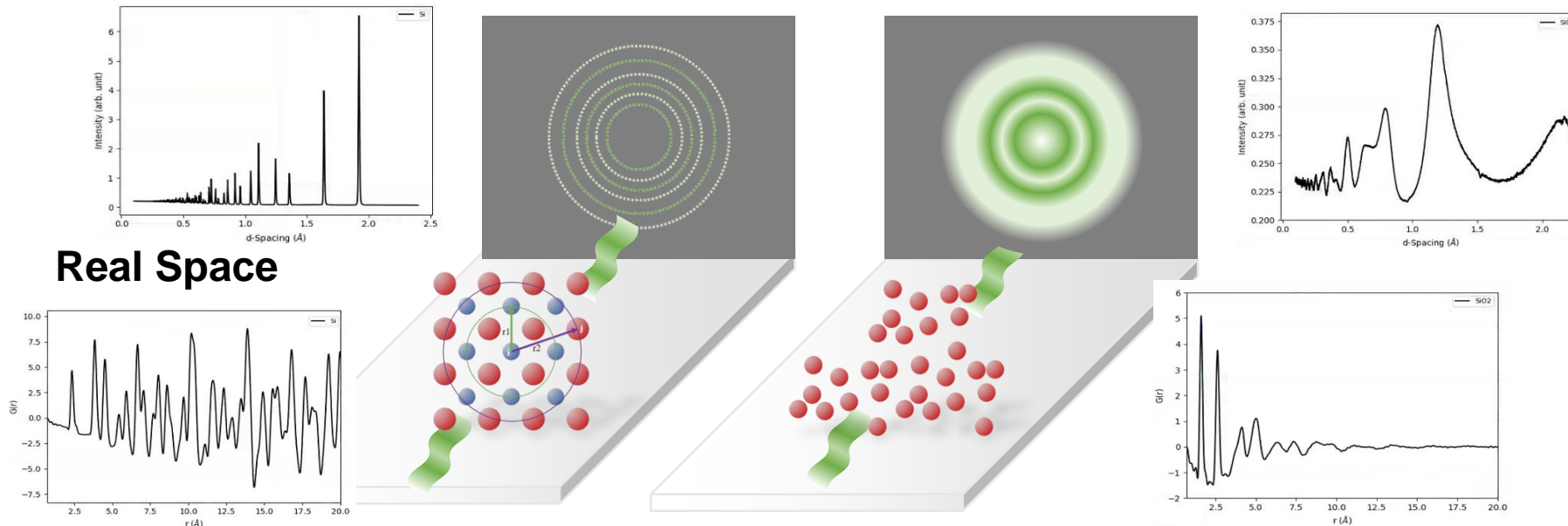
$$\left(\frac{d\sigma}{d\Omega}\right)_{c,el} = \frac{(2\pi)^3}{v_0} \sum_{\tau} |F(\tau)|^2 \delta(\mathbf{Q} - \tau),$$

$$\left(\frac{d\sigma}{d\Omega}\right)_{c,el} = \frac{1}{N_c} \left| \sum_{nj} [c_{jk} + P_{nj}] \langle \bar{b}_k \rangle e^{i\mathbf{Q} \cdot (\mathbf{l}_n + \mathbf{d}_j)} \left[1 + i\mathbf{Q} \cdot \sum_{n'j'k'} \Delta P_{n'j'k'} \mathbf{u}_{nj}^{n'j'k'} \right] \right|^2$$

From pattern to atomic structure

$$s(Q) = \frac{1}{N} \sum_{j,k} \overline{b_j b_k} \sin(Qr_{ij}) / Qr_{ij} = \frac{1}{N} \left(\sum_j \overline{b_j}^2 + \sum_{j \neq k} \overline{b_j b_k} \sin(Qr_{ij}) / Qr_{ij} \right) \quad S_{dist}(Q) = \rho_0 \int_0^\infty 4\pi r^2 G(r) \frac{\sin Qr}{Qr} dr$$

Reciprocal Space



Real Space

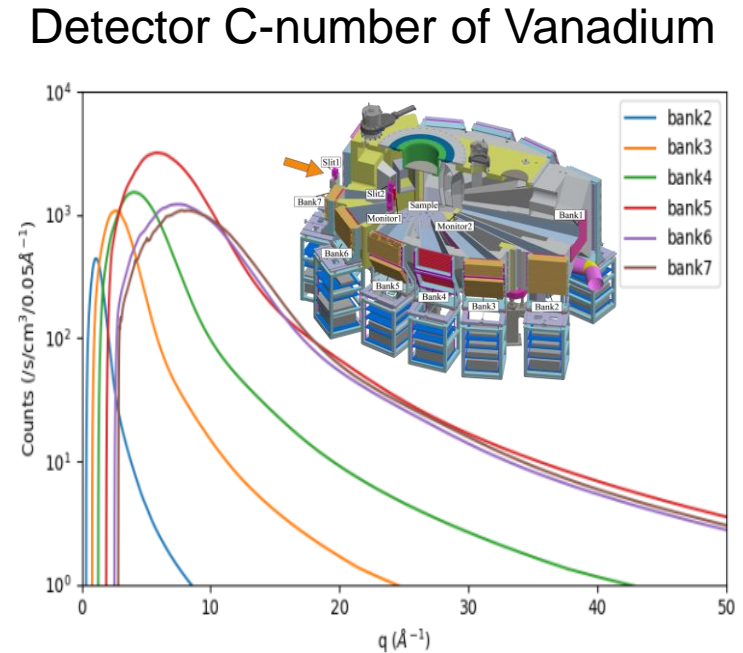
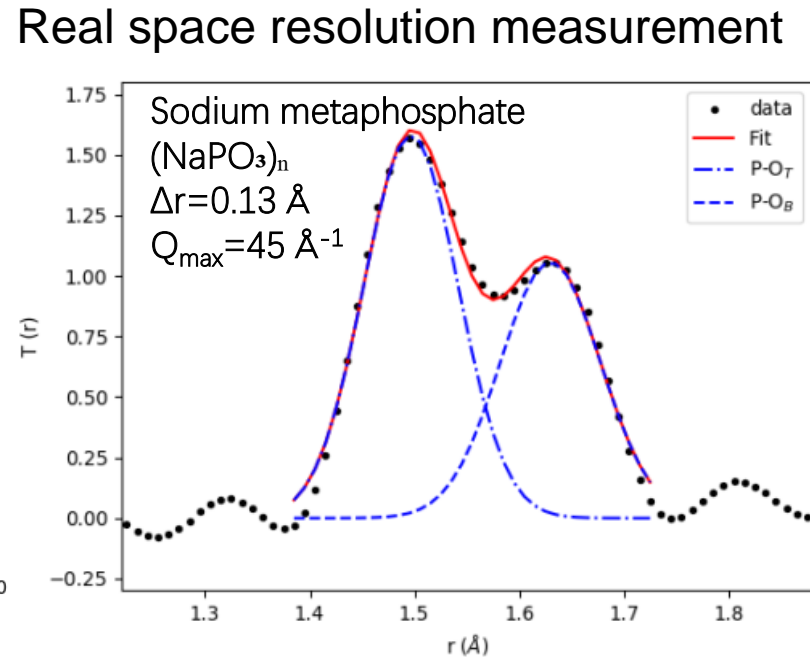
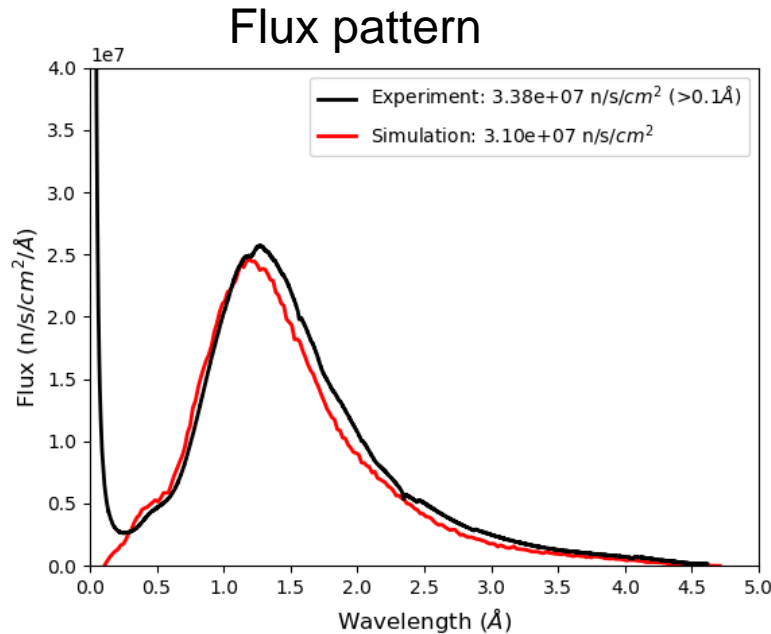
$$G(r) = \sum_{m,n} c_m c_n \overline{b_m} \overline{b_n} (g_{mn}(r) - 1)$$

$$G(r) = \frac{1}{(2\pi)^3 \rho_0} \int_0^\infty 4\pi Q^2 S(Q) \frac{\sin Qr}{Qr} dQ$$

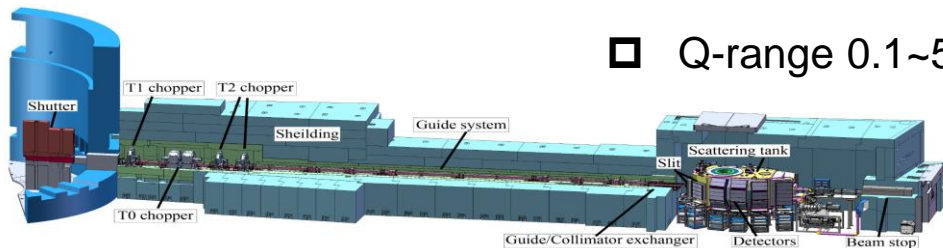
Peak positions in the PDF spectrum yield bond lengths, while peak areas yield coordination numbers.

Neutron Total Scattering Multi-Physics Instrument (MPI)

CSNS-MPI a diffractometer has a wide Q range to get a high resolution in real space



- ❑ Flux at sample position: 4.9×10^7 n/s/cm²@160kW (High Neutron Flux)
- ❑ Q-range 0.1~50 Å⁻¹, best resolution dQ/Q<0.4%, 0.1Å (High Real Space Resolution)



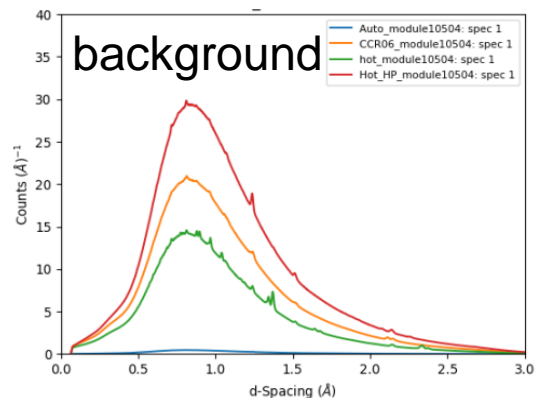
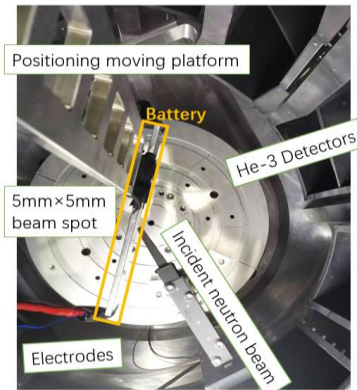
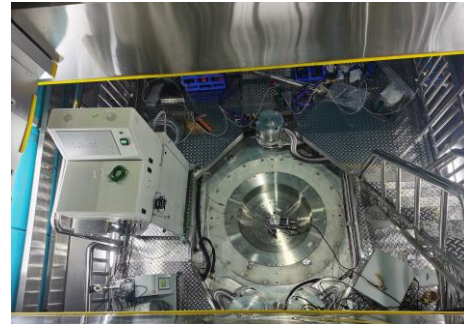
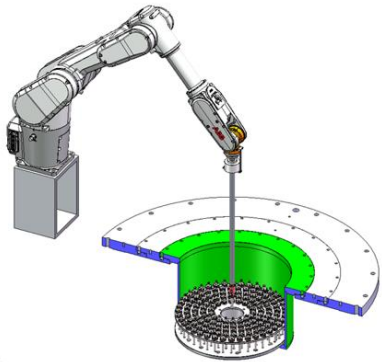
Neutron Total Scattering Multi-Physics Instrument (MPI)

➤ **Challenges** Background, Multi-scattering effect, Absorption

Multi-field coupled sample environment

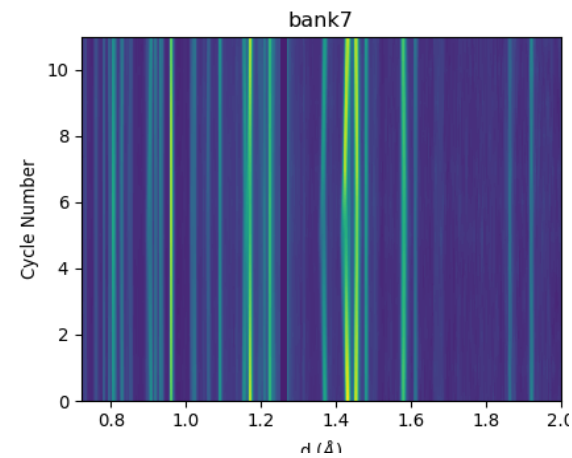
Robot Sample change

Gas absorption +CCR

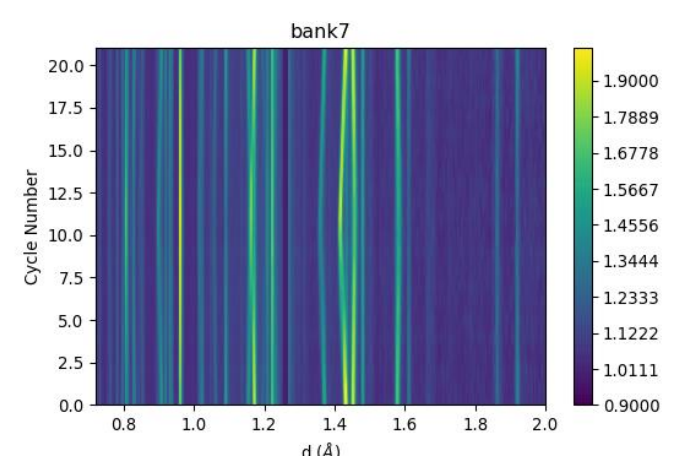


Fast-charging structural evolution

10C



5C

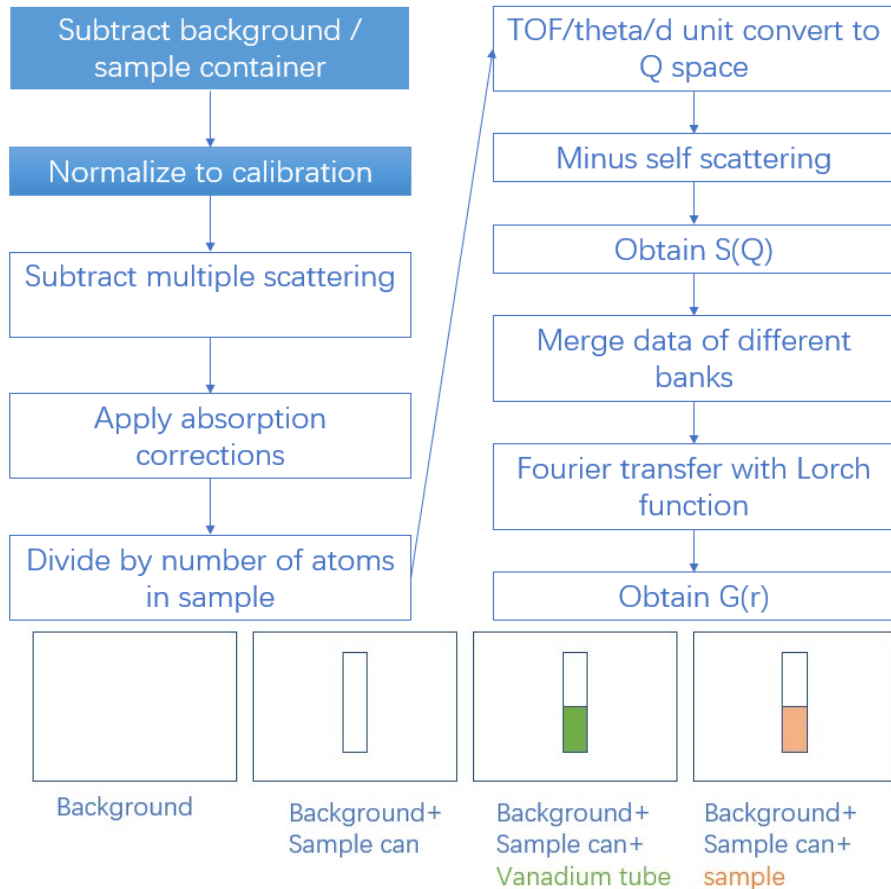


Through a data processing approach a fast time resolution of sub-minute level was achieved at MPI.

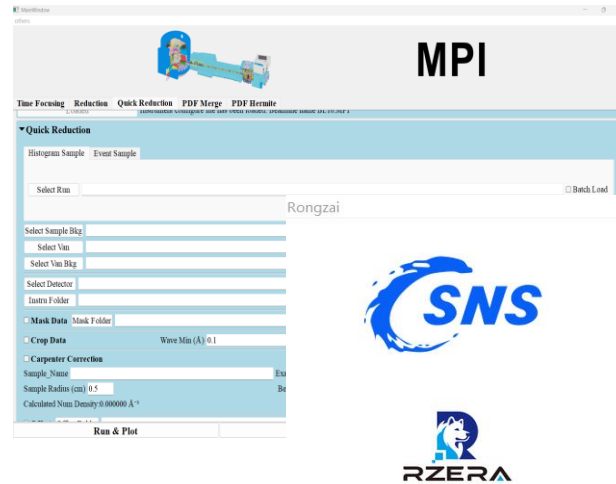
A reasonable data strategy not only yields reliable spectra, but also expands the functional boundaries of the diffractometer.

From Raw Counts to Reproducible S(Q)

Logical flow diagram of data reduce

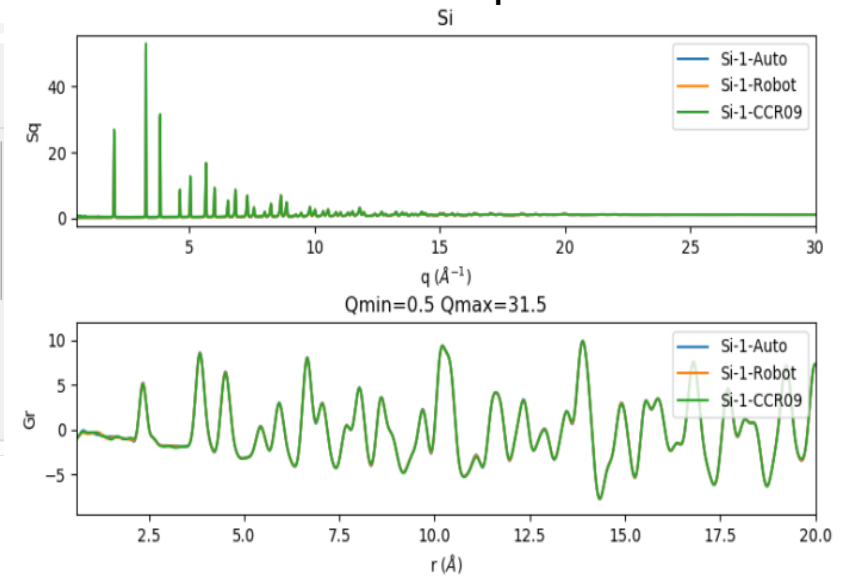


Reduce Software



$$I = \frac{(S - S_B) - \alpha(C - C_B)}{V - V_B}$$

consistent spectra

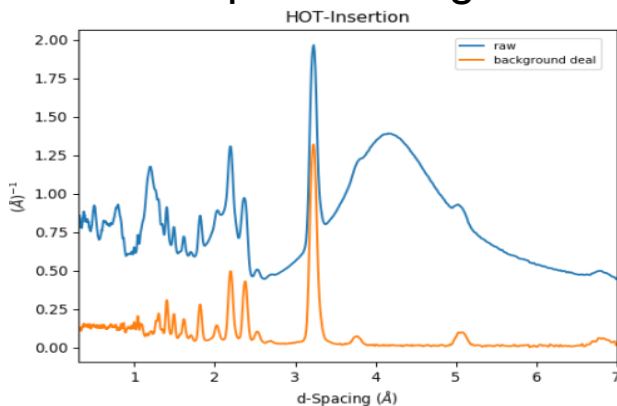


dBOT - Classify experiment types and establish a robust data correction method tuning to automatic reduction.

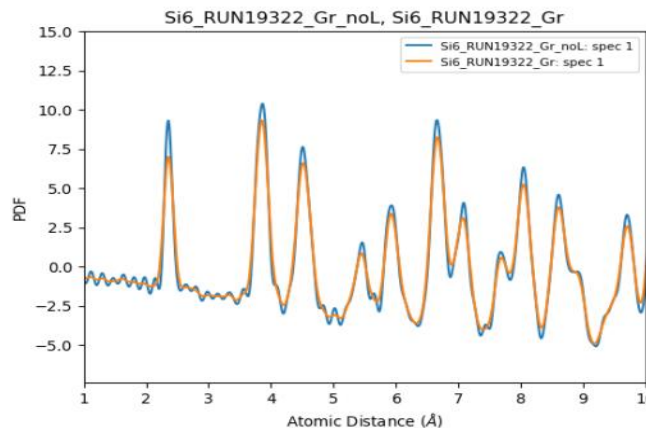
Standardized parameter processing ensures consistency across different sample environments and repeated measurements.

Quality Control Ensuring Fidelity

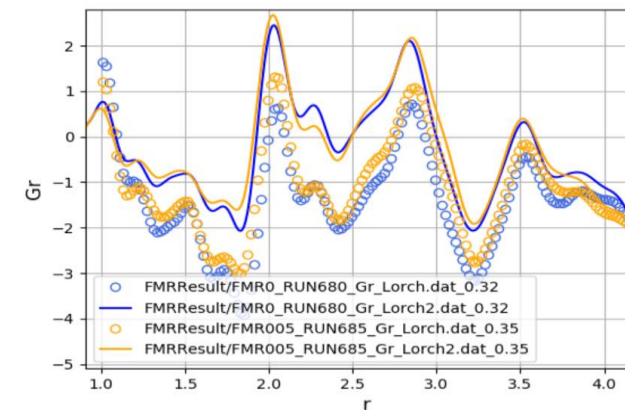
Decomposition Signals



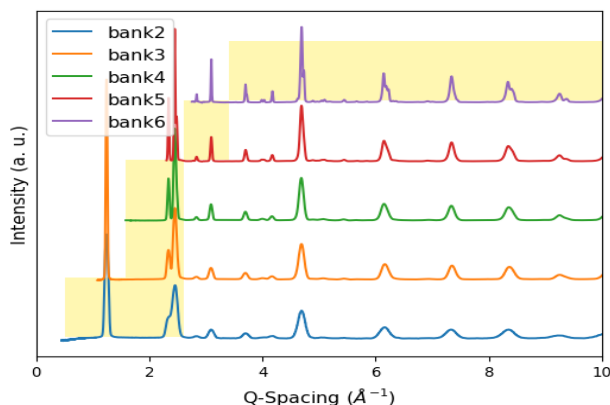
Lorch Modification



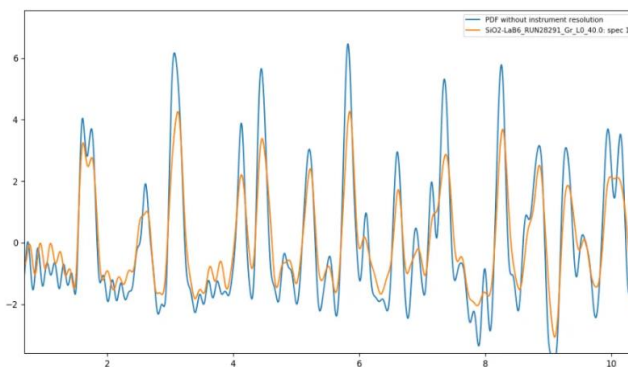
Small r correction



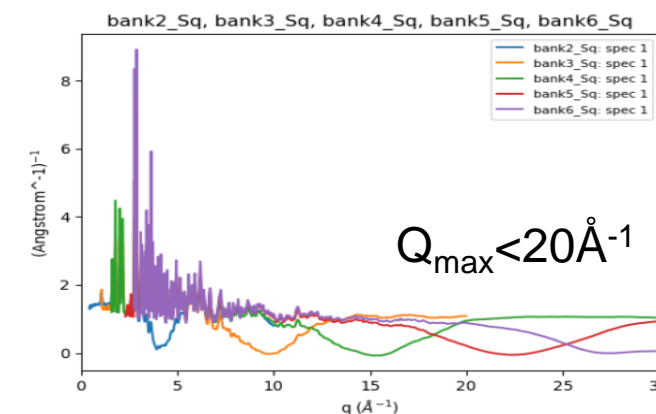
Merge data of different banks



Resolution improve

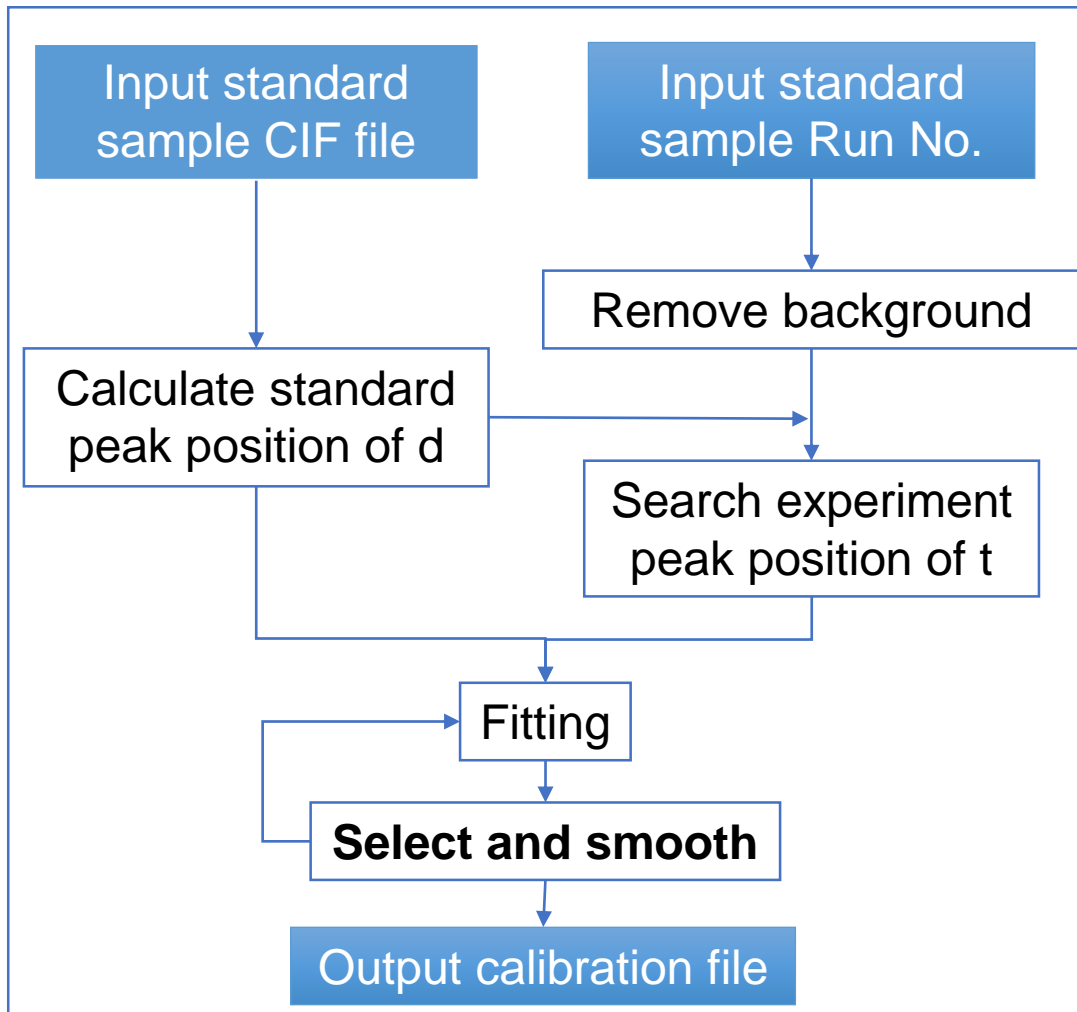


Strong absorption



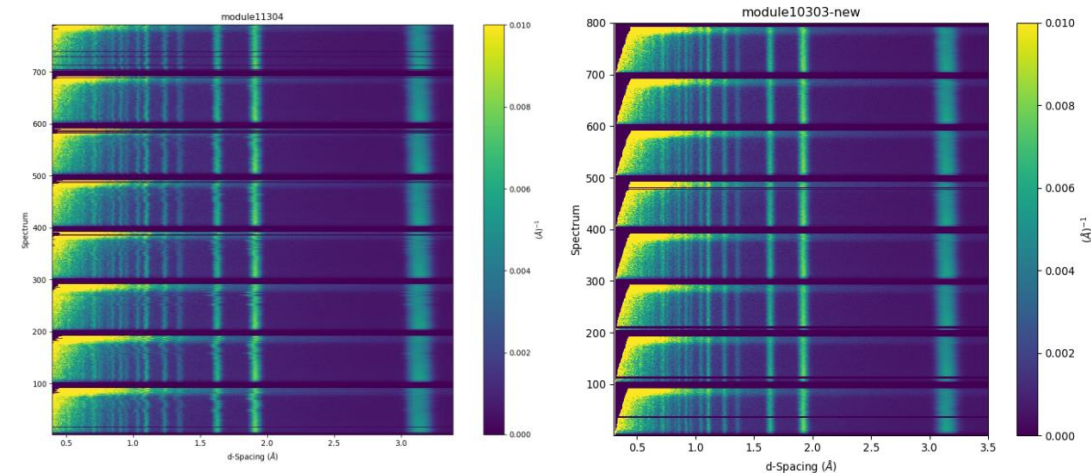
Structure analysis starts with disciplined correction of systematic effects.

Neutron flight time focus



Wide-Q obtained through the automatic time focusing of combined multiple standard samples

Combined calibrated by the Si, LaB6, MOF...



Once set, it runs. The automated calibration converts instrument stability into guaranteed data quality.

Robust, Rapid, Repeatability

Refinement Philosophy

NPD	Background	Sample	Diffuse Scattering	Local structure, amorphous ratio, lattice dynamics, magnetic disorder scattering	
			Inelastic Scattering		
		Scattering from Sample Environment, Air, etc.			
	Peaks	Position	Unit Cell & Magnetic	Crystal lattice parameters & space group, magnetic unit cell parameters & space symmetry group, qualitative phase analysis, macroscopic strain	
				Intensity	Structural Parameters
		Peak Shape	Instrumental Parameters		
			Sample Broadening	Microstrain, crystallite size, magnetic domain size	

Support popular software

Diffraction Refinement

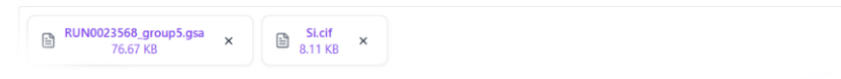
Minimize $R_{wp} = \sqrt{\frac{\sum_i w_i [I_{obs}(Q_i) - I_{calc}(Q_i)]^2}{\sum_i w_i I_{obs}^2(Q_i)}}$

PDF Refinement

Minimize $R_{pdf} = \frac{\sum |G_{obs}(r) - G_{calc}(r)|}{\sum |G_{obs}(r)|}$

RongZai_Agent

蓉仔——你的粉末中子衍射精修助手 by中国散裂中子源数据分析组

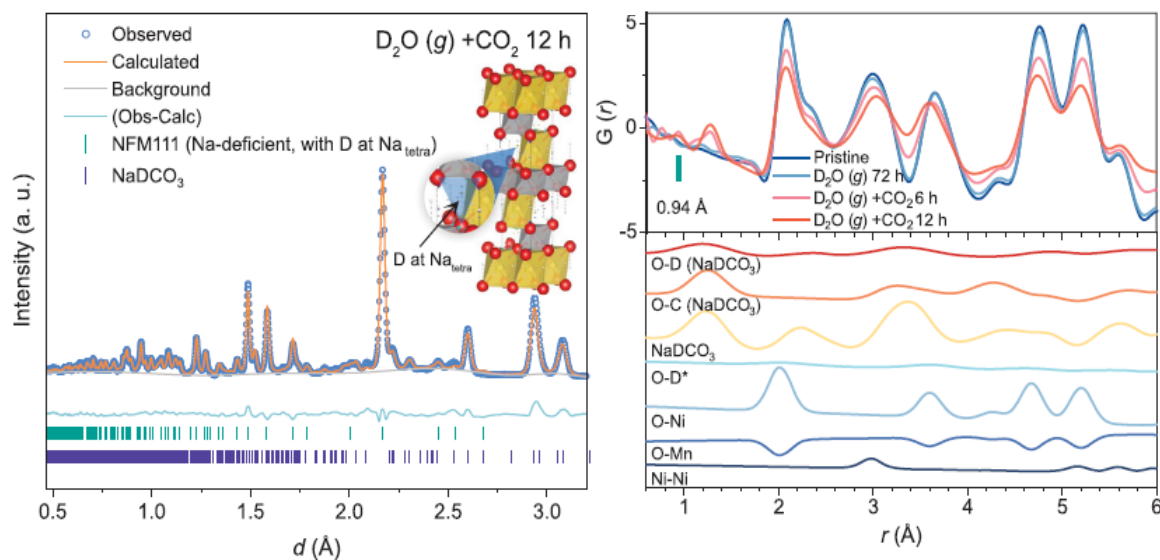


Upload 2 files to do the refinement, use the instrument file of the group5 of the BL16

Forward modeling seeks the best structural model whose calculated NPD/PDF matches the experimental data by Least Squares Method.

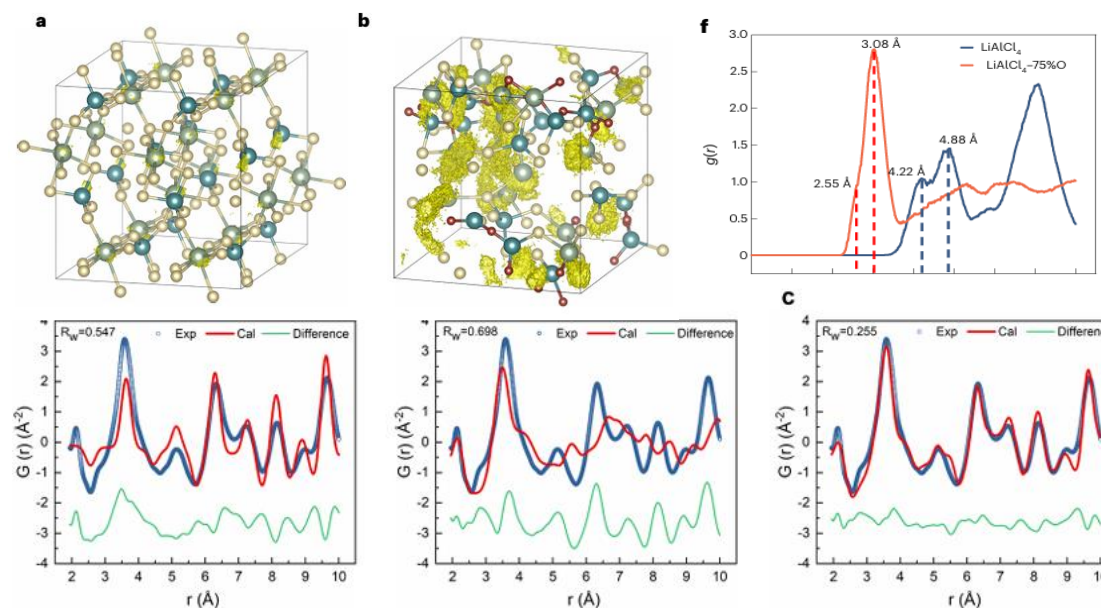
Local structure in energy field

Air sensitivity mechanism of Na-layered oxides



Under the interaction of water and carbon dioxide, the exchange of H^+ (D) from water with Na^+ in the Na layer of NLO leads to acid hydrolysis.

Pressureless inorganic solid-state batteries



Pressure-free Li/Na-based solid-state batteries have been achieved in a class of viscoelastic inorganic glasses by substituting oxygen for chlorine in tetrachloroaluminate.

Intelligent Enhancement Can AI help analysis?



➤ Limitations

The traditional “forward modeling” method has trouble with complex, disordered materials. It needs a good starting guess and takes a lot of computing time.

Structure model-dependent

Heavily relies on empirical experience

Strong parameter correlations

Prone to local minimum traps

Incapable of batch processing

Computationally expensive

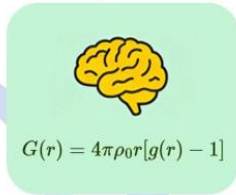
➤ Opportunities

Intelligent MPI Workflow

Automated experiment



Intelligent data



“Intelligent” is based on real physical model

Proposal
Collect

Before
Project
Design

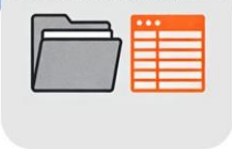
Virtual
Assistant

Underway
Data
Acquire

Structure
Analysis

After
Data
Reduce

Information administration



AI

Classify space groups

- Convolutional Neural Networks

Model atomic graphs

- Graph Neural Networks

Accelerate refinement

and so on...

End-to-end mapping:

$S(Q)$ or $G(r) \rightarrow$ structural parameters

physical constrained neural networks

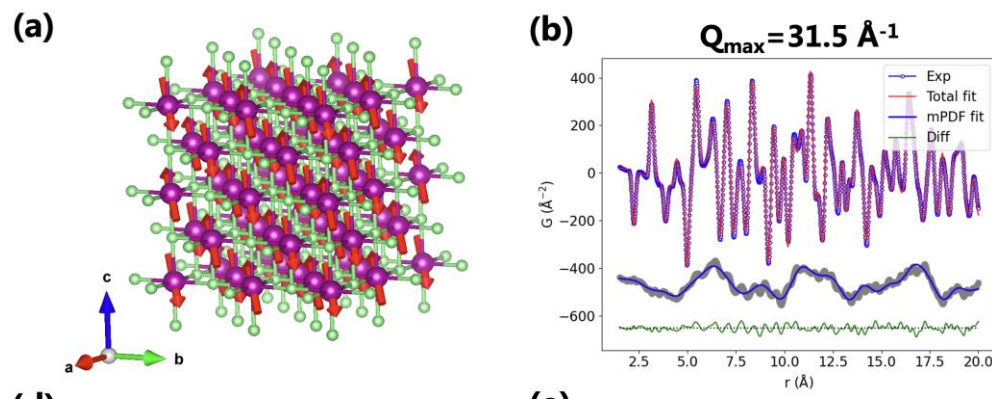
Real-time, unattended structural determination

Data are no longer just output—they drive the discovery of new structures.

AI helps speed up data analysis—but it's new physics-based algorithms that really decide what we can discover.

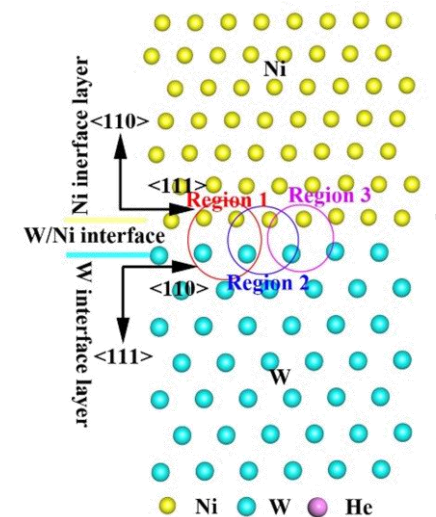
New algorithm ideas keep pushing the MPI field forward.

Magnetic pair distribution function method : mPDF@MPI



Short-range magnetic correlations at the subnanometer scale are revealed in real space.

Interface analysis method : iPDF@MPI

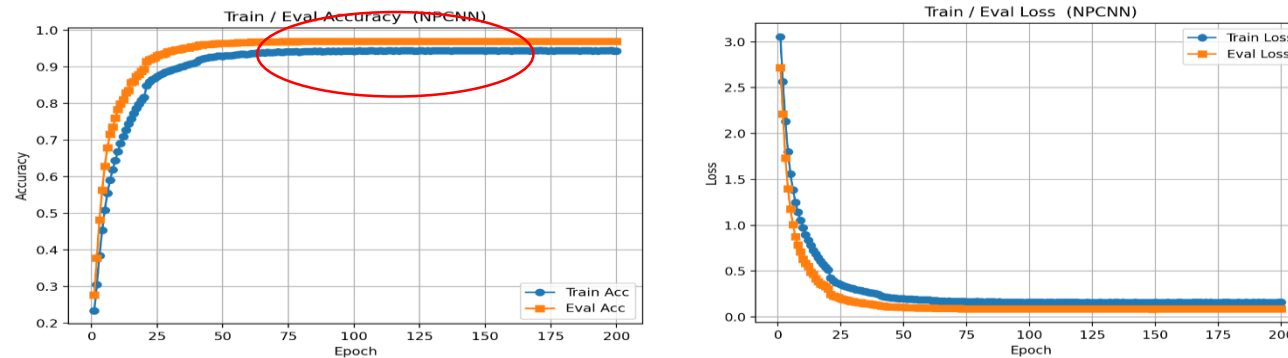


MMCs exhibit a narrow near-interface region and weak scattering signals.

Fundamental algorithmic innovation continues to propel advances in the field of MPI.

Rapid data analysis for battery materials

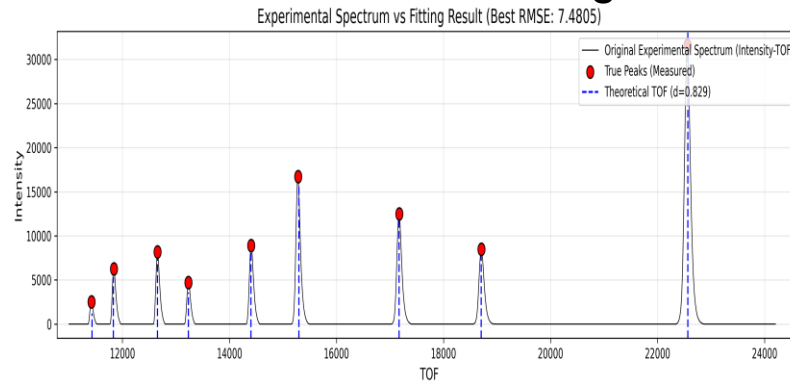
Space group retrieval based on Convolutional Neural Networks (CNN)



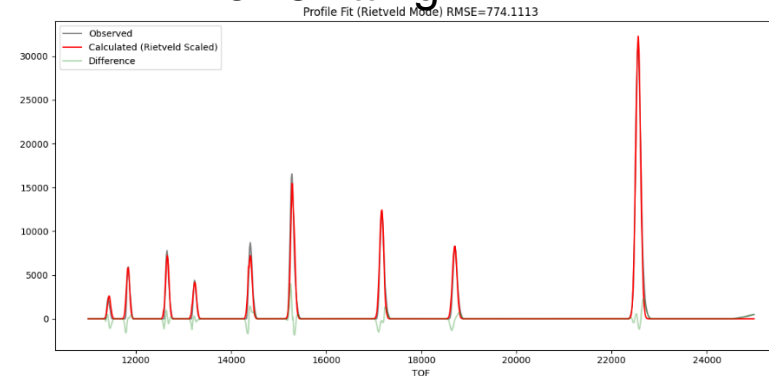
Loss and accuracy curves for the training and validation of space group classification

Crystal structure fitting based on Reinforcement Learning (RL)

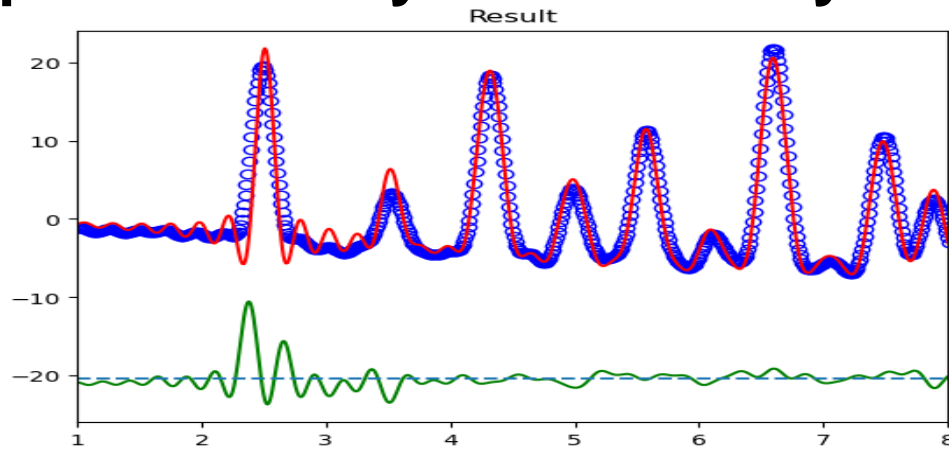
Peak fitting



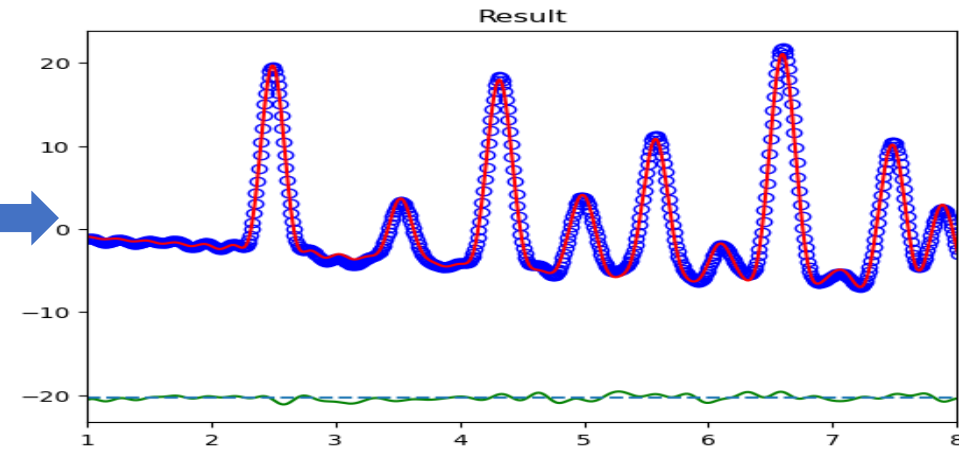
Profile fitting



Rapid data analysis for battery materials



Least squares method:
 $Rw = 0.16268$

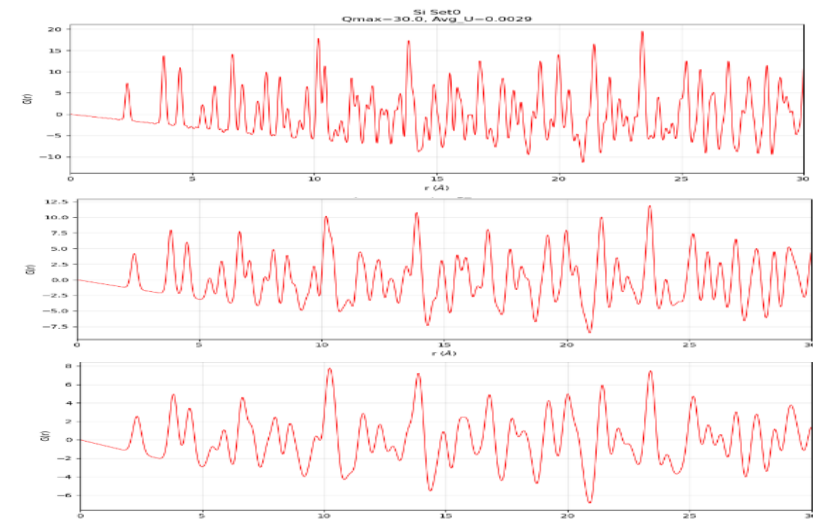


Reinforcement learning based on the PPO algorithm:
 $Rw = 0.05181$

Local optimum trap:

Global optimization methods (genetic algorithm, simulated annealing) are required to avoid convergence to wrong solutions.

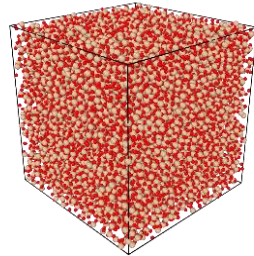
Compared with the least squares method, it achieves a better goodness of fit and avoids getting stuck in local minima.



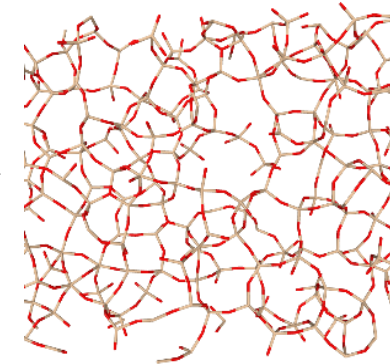
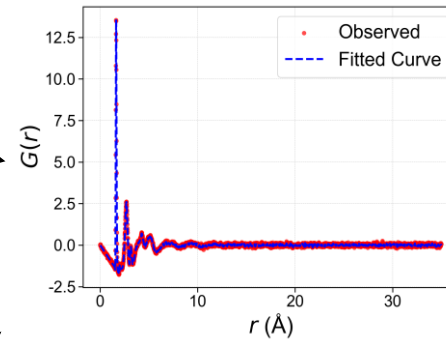
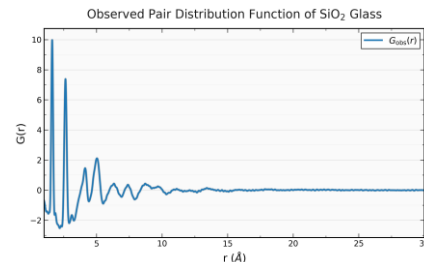
dataset augmentation

CNN-based PDF study of glass system

create physically reasonable initial structural using molecular dynamics



Experimental PDF data collection



Obtain the material structure

PDF → CNN → Structure info
Fast and Simple

For measured data of the PDF data of glass, the proportions obtained by CNN are in good agreement with the actual values.

- 1 Rigorous data engineering — from automated calibration and standardized processing to critical corrections
- 2 Innovative intelligent algorithms — advanced agents based on traditional analysis, physics-driven new methods, or AI tools including CNN and ML, expanding perceptual boundaries and analytical efficiency.

Outlook

- 1 Multimodal data fusion — combining neutron, X-ray data to construct a more complete structural picture
- 2 Collaborating with users and professional team— promoting the development of open-source toolchains.



Thank you!

