

September 19 – 22, 2022, PSI, Villigen, Switzerland

A SEDCNN Machine Learning Model for Textured SAXS/WAXD Image Denoising

Zhongzheng Zhou^{1,2}, Chun Li^{1,3}, Xiaoxue Bi¹, Chenglong Zhang¹, Yingke Huang¹, Jian Zhuang^{1,3}, Wenqiang Hua⁴, Zheng Dong^{1,3}, Lina Zhao^{1,2}, Yi Zhang^{1,2}, and Yuhui Dong^{1,2}

1 Beijing Synchrotron Radiation Facility, Institute of High Energy Physics, Chinese Academy of Sciences, Beijing, 100049, China

2 University of Chinese Academy of Sciences, Beijing, 100049, China

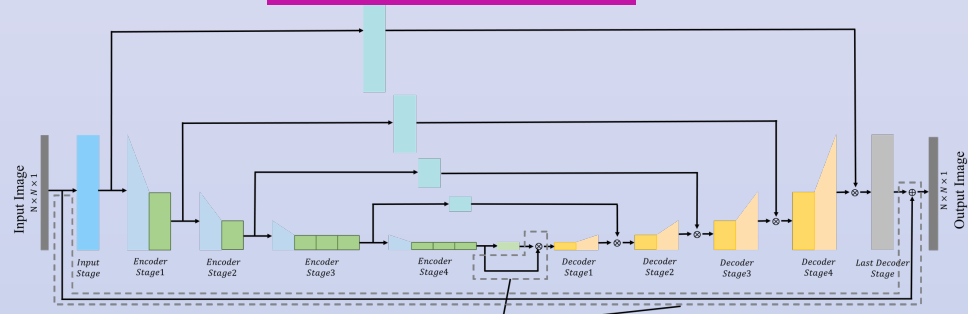
3 Chinese Spallation Neutron Source Science Centre, Dongguan, Guangdong, 523808, China

4 Shanghai Synchrotron Radiation Facility, Shanghai Advanced Research Institute, Chinese Academy of Sciences, Shanghai, 201204, China

Introduction

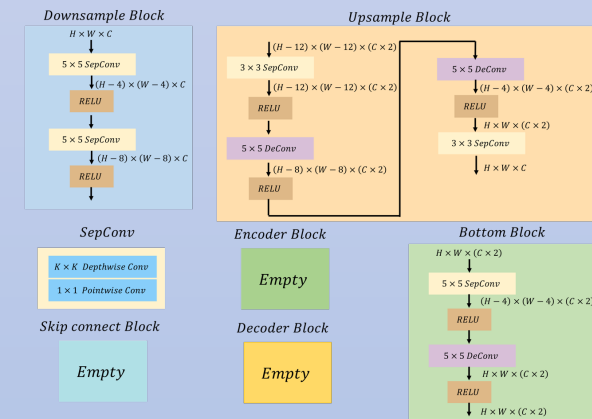
- Experiments on next-generation beamlines are evolving into multi-modal, multi-scale, time-resolved, in-situ characterization.
- To maintain high signal-to-noise ratio (SNR), prolonged exposure and excessive dose occurring in SAXS/WAXD experiments remains a serious concern.
- This work algorithmically improves the SNR of SAXS/WAXD images, allowing more redundancy in exposure time and dose reduction during experiments.

Network Design

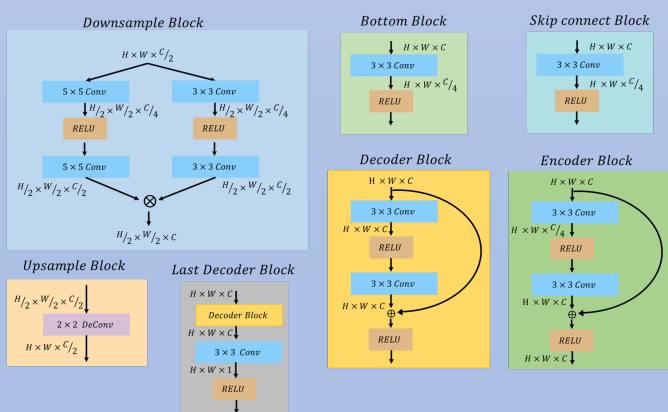


SAXS Size	$N^2 \times 48$	$(N-8)^2 \times 48$	$(N-16)^2 \times 48$	$(N-24)^2 \times 48$	$(N-32)^2 \times 48$	$(N-32)^2 \times 48$	$(N-24)^2 \times 48$	$(N-16)^2 \times 48$	$(N-8)^2 \times 48$	$N^2 \times 1$
WAXD Size	$N^2 \times 16$	$(N/2)^2 \times 32$	$(N/4)^2 \times 64$	$(N/8)^2 \times 128$	$(N/16)^2 \times 256$	$(N/16)^2 \times 64$	$(N/8)^2 \times 64$	$(N/4)^2 \times 32$	$(N/2)^2 \times 32$	$(N/2)^2 \times 16$

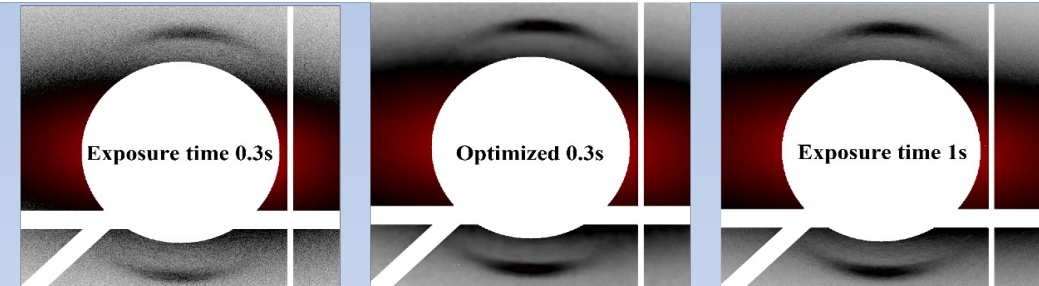
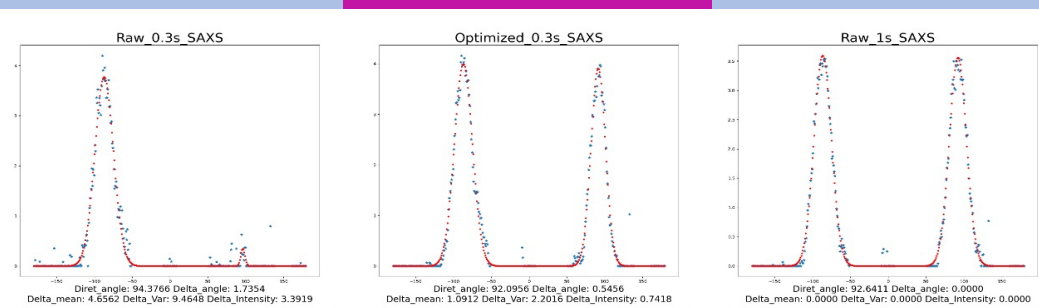
SAXS Part



WAXD Part



SAXS



Methods

Preprocessing: masking and resizing

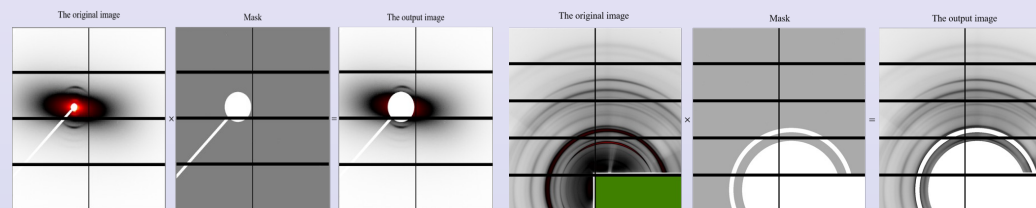


Fig. 1: preprocessing procedures, including masking and resizing

Customized metrics reflecting physical attributes

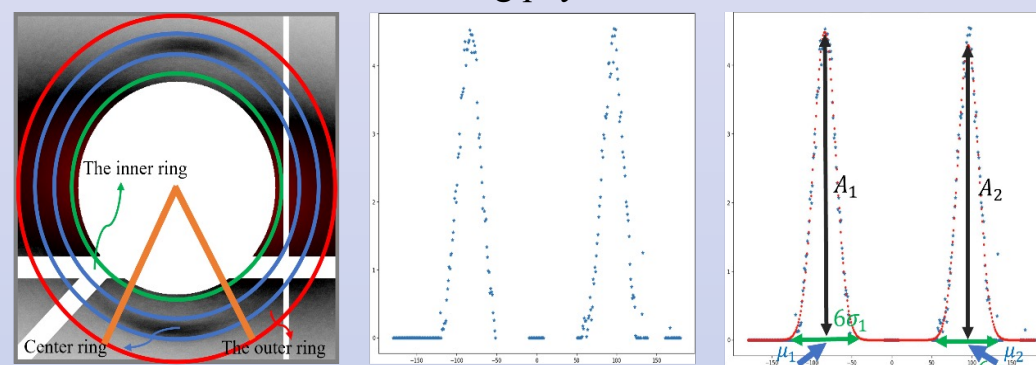
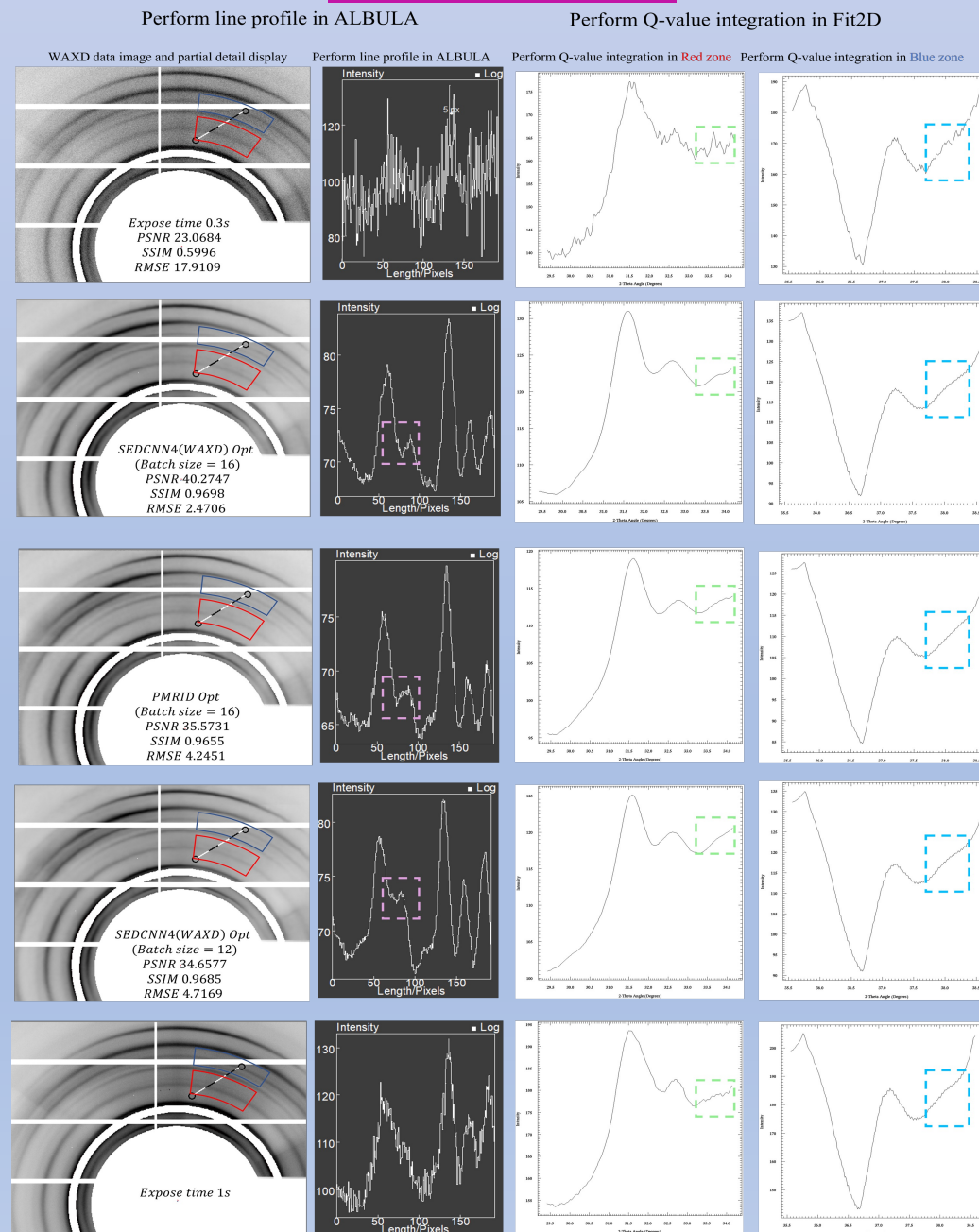


Fig. 2: radial integration, upon which various customized metrics are built

- Block-based training scheme empowers small network models.
- Fine-tuning hyperparameters to reach optimal performance.

WAXD



Conclusions

- The proposed model provides bespoke denoising solution for SAXS/WAXD images, from data preprocessing to network design and to final performance evaluation metrics.
- The model demonstrates superior denoising performance on highly textured SAXS/WAXD images, compared with classic natural image denoisers like U-Net, REDCNN, and PMRID.