KFS – Digitalization priorities

ErUM DATA Meeting

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Komitee Forschung m

Synchrotronstrahlung

Outreach activities to and input from the community

- 2/2018 KFS user survey on needs for Data management
- 3/2018 KFS+KFN workshop #1 "Digital transformation" at DESY
- 9/2018 KFS+KFN symposium at SNI Munich
- 6/2019 KFS+KFN workshop #2 "Digital transformation" at DESY
- 2018-2020 Participation in the BMBF ErUM Data process
- 2019/2020 DAPHNE4NFDI proposal





Research data management

DAPHNE4NFDI

- Universities
- Research institutions
- KFS-Committee for Synchrotron Research
- KFN-Committee for Neutron Research
- Large-scale photon and neutron research
 facilities
- Wider community





Challenge #1 - variety of disciplines, X-ray techniques and samples, set-ups and data



Degree of automatization and standardization in experiments and data varies across the techniques

Protein crystallography – experiment and data highly standardized



Most others: data, metadata highly user specific



Challenge #2: increasing data rate



| | Per year (2019) | BESSY II | PETRA III | FLASH | EuXFEL | ELBE | MLZ | Total | ESRF EBS |
|---|--|----------------|----------------|--------------|--------------|--------------|---------------------------|-----------------|----------|
| 500 fps, 12 MPix 12MB/image 21 TB/hr (6 GB/s) 2PB free space 'full in a few days' | Beamlines (BL) Measuring stations (MS) | 27 BL 37 MS | 22 BL 42 MS | 2 BL 7 MS | 3 BL 6 MS | 1 BL 7 MS | 26 BL 26 MS (+7 BL) | 81 BL 125 MS | 44 |
| | Experiments/ year | 800 | 1400 | 38 | 25 | 70 | 620 | approx. 2950 | 2000 |
| | Individual Users | 1500 | 3000 | 350 | 499 | 100 | 600 | approx. 6050 | 9000 |
| | User visits | 3000 | 5000 | 500 | | | >1000 | >9500 | |
| | Publications | 500 | 500 | 30 | 94 | 40 | 345 | approx. 1500 | 2000 |
| | Data generated | 1-2 PB | 5 PB | 1 PB | 20 PB | 0.5 PB | 6TB | 28.5 PB | > 20 PB |
| | Expected 2025 | 8 PB | 20 PB | 4 PB | 100 PB | 2 PB | 0.5 PB | 134.5 PB | 60 PB |



New (ML/AI) software and algorithms for coping with data avalanche

Five possible objectives - more welcome

1. Faster and better classification / analysis of X-ray data and feedback loop to simulation/models/theory.

Example: Ptychography, ML training

- on a small subset of the data/sample
- on 180k artifical samples

Allows to improve speed and relax on sampling conditions

Neural network reconstruction 300 x faster than Phase retrieval- outlook for real time imaging



1. Faster and better classification / analysis of X-ray data and feedback loop to simulation/models/theory.

X-ray techniques

- X-ray reflectivity
- GISAXS
- SAXS
- WAXS
- Ptychography
- Tomography
- XANES, EXAFS
- X-ray spectroscopy
- Resonant magnetic scattering
- XPCS
- Serial crystallography
- FEL diffraction pattern classification

ML techniques

- Pattern/image recognition and classification
- Machine learning for inversion problems
- Automated segmentation in tomography
- Superresolution

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2. Towards autonomous experiments



Noack et al. Scientific reports (2019) 9:11809

Coffee ring with nanoparticles



Sample

Grid scannning





(a) (b)

2. Towards autonomous experiments

Noack et al. Scientific reports (2019) 9:11809 Credit animation: Talk S. Jethian (Berkely + CAMERA), ESRF-ILL workshop 2019

Grid

Add: beamline control accelerator

SMART





A. Chen et al. InfoMat. 2020;2:553–576.



3. X-ray data mining ML based cross analysis using repositories etc.

Ren et al., Sci. Adv. 2018;4:eaaq1566

Example: Predicting material properties



ML predicted glass forming ability (GFA) of tenary metallic glasses based on X-ray data



Requires data research management

- complete metadata
- electronic logbooks
- curated and documented analysis software
- proper repositories
- ontologies
- descriptors for material properties. (PRL 114, 105503 (2015)

4. Benchmarking, development, use and understanding of ML algorithms





5. Parallelization, GPU + FPGA computing

Example: X-ray photon correlation spectroscopy.

Pixel by pixel correlation

$$g_2(\vec{Q},t) = \frac{\langle I(\vec{Q},t')I(\vec{Q},t'+t)\rangle}{\langle I(\vec{Q},t')^2 \rangle}$$

Benefits from parallel computing





Priorities for KFS

- We are striving for science- and user-driven software solutions for our community.
- Close collaboration with the national facilities including European XFEL and ESRF
- Evaluate cross community efforts e.g. together with KFN, KFB, KFSI
- Include ML/AI stakeholders and other ErUM communities
- Understanding ErUM Data as a long-term project for the KFS community aiming for establishing new algorithms, new data management, federated infrastructure and data science education.
- Today and tomorrow: KFS encourages you to discuss and plan possible collaborations for ErUM Data.