

Summary of the Working Group on AI for Nuclear Data

WG Coordinators

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#AI4Atoms Virtual Event

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Relevant tasks in Nuclear Data



- Efficient information extraction from experimental measurements and associated publications
- UQ for all input of nuclear data evaluations including outlier detection, uncertainty estimate of model parameters & experimental data, etc.
- Validation of nuclear data libraries
- Generation/Improved accessibility of databases of sensitivities of integral responses to nuclear data and experimental data
- Trainings on how to apply ML methods in the nuclear data pipeline
- Consideration of physics constraints, e.g., through probability distributions, on nuclear theory model parameters (method developments)

Technical requirements

- Good databases
- Efficient use of computer simulations
- Use of emulators where necessary
- Documentation of APIs

State of the art

- Nuclear and atomic data evaluation:
 - Generalized Least Squares method still prevalent
 - Gaussian processes applied and show promise
 - Monte Carlo approaches (e.g., BMC) continue to improve
 - ML methods applied to determine parameters of physical models and fitting functions (e.g., R-Matrix)
 - First successful applications of neural networks (e.g., mass evaluations)

State of the art

- Validation:
 - ML approaches can identify incorrect nuclear data, problematic features of experiments and potential biases
 - Bayesian experimental design process (ML-driven approach to find optimal designs)

Limitations

- Common Gaussian distribution assumption is often not justified (e.g., outliers)
- ML approaches require **high-quality data**, which is an obstacle at present
- Information in **databases very difficult to parse** automatically
- **Unbalanced UQ and quality control efforts** in different domains of data (weakest link determines overall quality)

Possible next steps

- Continue **identifying important open questions/issues** that could benefit from the application of ML (our WG already made significant progress in this respect)
- **Enrich experimental databases** (e.g., EXFOR, ALADDIN) with well-documented API and rich meta data to facilitate the development and testing of ML approaches
- **Define diverse and broad test sets** with domain-specific data to validate ML models (relevant physics quantities) for comparison studies
- **Educational activities**, e.g., workshops, to teach ML methods, nuclear physics knowledge relevant for the application of ML
- **Crowd-sourcing campaigns** / open-science challenges to improve databases, models, etc. (e.g., hackathons, Kaggle as example)

Accelerating Progress – IAEA's role



- IAEA hosts prominent databases, e.g., EXFOR, RIPL, etc. so bring database maintainers, ML developers, and nuclear data users together to discuss how to **make databases more accessible** for ML studies.
- Draft a document laying out a **standard approach for data handling** (e.g., FAIR) and collaborative open-source practices with a focus on EXFOR and potentially others, considering input from nuclear data users, ML developers and database maintainers
- Organize meetings for **comparison exercises and educational workshops**
- Host repositories with **reference datasets** for domain-specific ML studies

Expected Outcomes

- **Reference test sets** to validate ML models
- **Clear licensing** of data in databases and repositories
- **Roadmap** to future measurements and theory needs
- **Detection of issues** in data and models impeding scientific progress
- Significant speed-up of **evaluation and validation** work through the application of ML

Thank you!

