

Summary of the Working Group on Al for Nuclear Fusion

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State of the Art



Data drives fusion experiments, from design to analysis and optimization



Noisy

ML models used for performance optimization of **MFE** experimental operations



ML/AI tools used to improve quantitative data extraction in IFE



Trained networks to extract quantities based on simulations/ model





ML/AI applied fusion research highlights existing bottlenecks:



Data accessibility

Big databases needed for ML/AI model training/validation





Workforce development

Hybrid curricula needed to fully leverage ML/AI tools **Community integration**

Common frameworks needed to create ONE community of practice focused on AI/ML for fusion (of all flavors)





Next Steps

More computing power allows for AI/ML to advance capabilities in IFE/MFE - need to explore extrapolation to unseen domains



Al can enhance analysis of instrumentation (especially imaging) and potentially extract inferred quantities using models or surrogates

Predicted image and its error map



Predicted image Error map (pixel value) (pixel-wise variance)

Adapted from John Kline, IAEA WG-AI4NF, 10/26/2021

Al predictive modeling can be used for plasma performance optimization, event and anomaly detection and plant operations



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Common bottlenecks preventing AI from accelerating IFE and MFE find analogies in other TM working groups themes/challenges

Hesitancy to share data

Ownership

Security Intellectual property

Poor communication

Stovepiped communities Fear of being scooped

Gatekeeping by community members

Low understanding of new methods Reluctance to change from familiar methods Lack of large scale coordinated effort

Low numbers of trained personnel

Fusion subject matter experts need re-skilling for AI/ML





Accelerating Progress - IAEA's Role

We identified a multitude of enabling activities for IAEA to accelerate progress in MFE and IFE





workshops: basics of ML, advanced ML (multimodal), huge datasets, fusion data for fusion researchers, ...

host

data



pper ECP-ARA Schema and Materian Specific in Proving Disconnect Fundamental Methods for Alarra, Waterian and Materian Properties in Proving Disconnect 19 - 32 April 2018, Microsov - Tradis, National Science - Statella, National Science - Science



be a data steward: manage data embargos, establish common formats



act as a broker/ambassador between the data and user community

> educate and engage workforce



create new ways to engage the community (Kaggle-like competitions, hack-a-thons)

For Business Problems Bring in the world's largest community of

Learn more

data scientists to consult on your toughest business problems.



IAEA's role: host data beyond national barriers

- Fusion databases hosted at different facilities in different formats, behind firewalls and not connected (worldwide)
- Need dedicated resources for db curation and management
 - IAEA internships, fellowships, consultancy opportunities could be leveraged







IAEA's role: host data beyond national barriers - open data access



IAEA can help build open MFE/IFE databases by:

- > surveying existing efforts in fusion communities and move past their shortcomings
- developing a centralized repository of experimental/simulation/plant operation data;
- coordinating efforts to expedite willingness to share from agencies/facilities through official call for data;
- building AI/ML projects that can consume the data;
- opening participation to fusion communities worldwide;
- offering committee/staff to adhere to OPEN/FAIR principles.



IAEA's role: leverage open access to engage broader interest in AI for Fusion



IAEA can act as a broker/ambassador between the data and user community.

Open data access crucial for attracting talents by engaging AI/ML community and expand Fusion workforce. Open databases used for:

- > outreach activities, workshops/schools, ...
- Kaggle-like competitions, hack-a-thons, ...

thus finding creative ways to engage, energize, and integrate many communities, e.g., ML experts, students, science researchers, other relevant cross-field researchers, private entities, ...



IAEA's role: educate a diverse workforce in Fusion



IAEA can establish a network to bring AI/ML/Fusion scientists together and help educate a diverse workforce.

Added value: bringing small AI activities scattered across many institutions together, and thereby gaining critical mass, avoiding duplication, and exploiting synergies

Thanks to IAEA supporting role, the Fusion workforce will see:

- increasing participation,
 diversification to AI/ML curricula
 - potential reference academic curricula definition
 - increasing outreach to non-physics communities.





Expected Outcomes

IAEA networks and Coordinated Research Activities will accelerate progress in AI for Fusion



IAEA will have a transformative role enabling key Fusion research opportunities

A roadmap for AI deployment in Fusion

Machine Learning for Real-time MFE System Behavior Prediction, Identification & Optimization Improve IFE Physics Understanding through Simulation, Theory and Experiment Using ML/AI Methods

- → Data stewardship
 - Democratization of resources, data standards, ...
- → Community integration
 - Partnering with industries, common frameworks, licensing qualification ...
- → Workforce development
 - Outreach, education, training, ...



Thanks to all the Nuclear Fusion Working Group members for their incredible effort!



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Thank you!

