

Summary of the Working Group on AI for Nuclear Physics

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Germany

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Technical Meeting on Artificial Intelligence for Nuclear Technology and Applications

#AI4Atoms Virtual Event

25–29 October 2021

Davidson College, USA
Springer Nature,

IAEA, Austria

Participants

Rick Archibald (ORNL, USA)

Yassid Ayyad (USC, Spain)

Rebecca Auchettl (ANSTO, Australia)

Maria Liz Crespo (ICTP, Italy)

Auralee Edelen (SLAC, USA)

Cristiano Fanelli (MIT, USA)

Morten Hjorth-Jensen (University of Oslo, Norway and MSU, USA)

Tzany Kokalova Wheldon (University of Birmingham, UK)

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Karin Rathsman (ESS, Sweden)

Alexander Scheinker (LANL, USA)

Tiago Silva (University of São Paulo, Brazil)

Gianluca Valentino (University of Malta, Malta)



Nuclear Physics

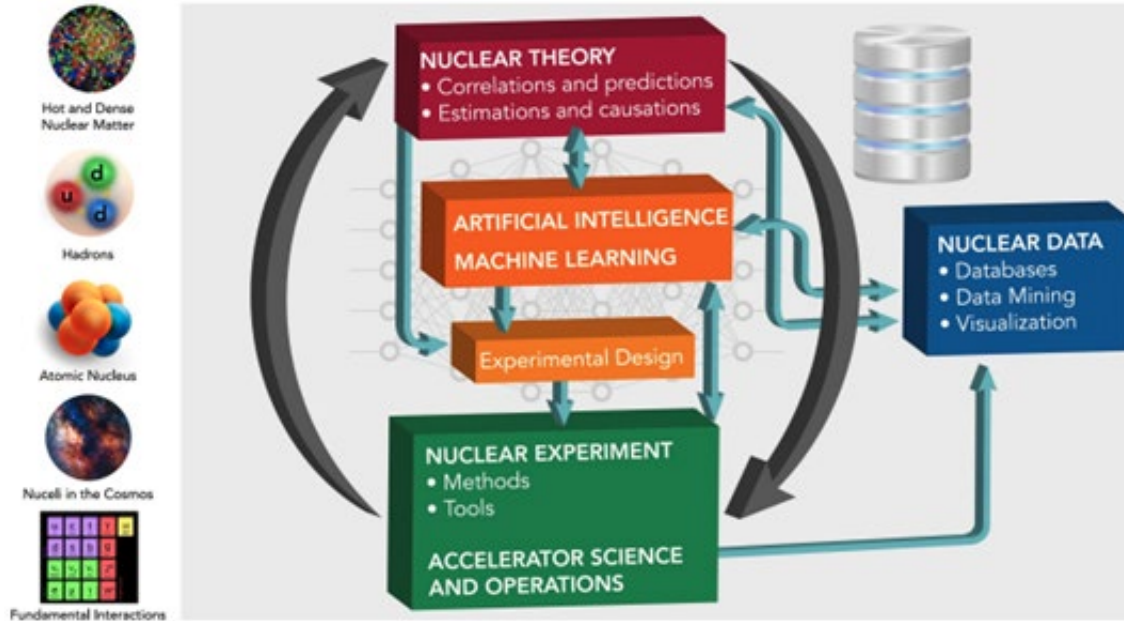


Image courtesy of W. Nazarewicz, FRIB, MSU, USA

Nuclear Physics

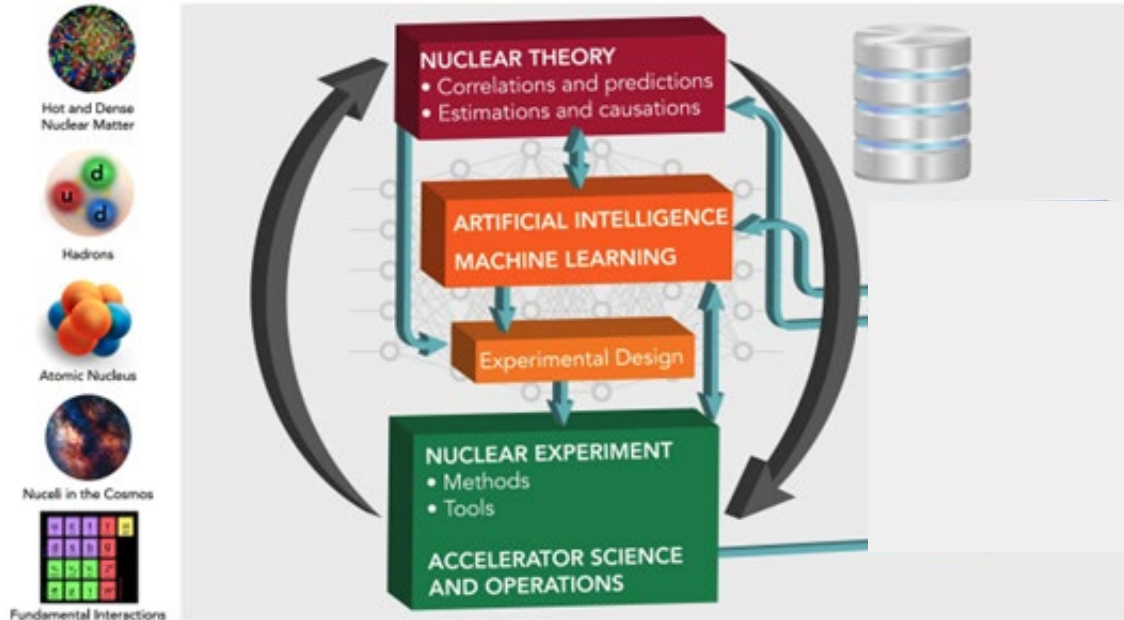
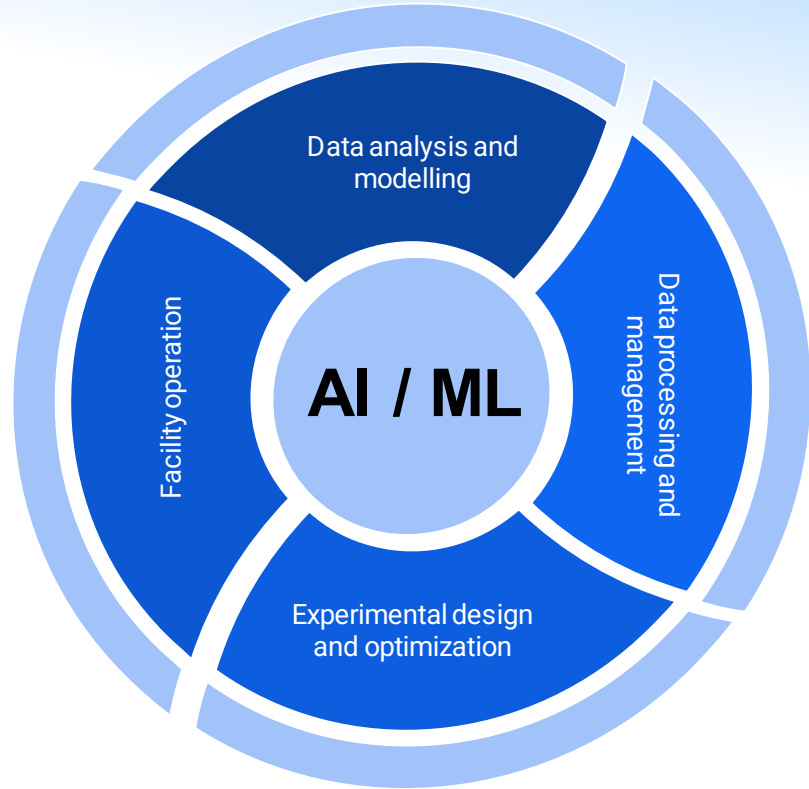


Image courtesy of W. Nazarewicz, FRIB, MSU, USA

AI in Nuclear Physics

- Analysis and Modeling
- Data Processing and Management
- Experimental Design and Optimization
- Facility Operation



State of the Art / Ongoing Efforts

Analysis & Modeling

ML in nuclear theory

Bayesian model averaging, see L. Neufcourt et al., Phys. Rev. Lett. 122, 062502 (2019)

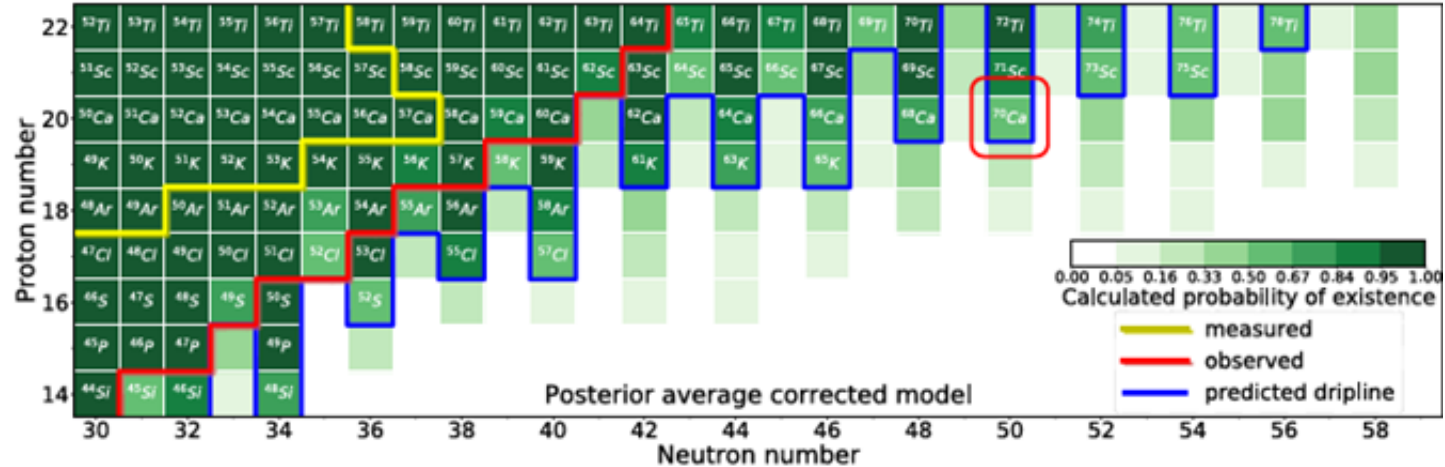


Image courtesy of W. Nazarewicz, FRIB, MSU, USA

Data Processing and Management

ML for event and particle identification

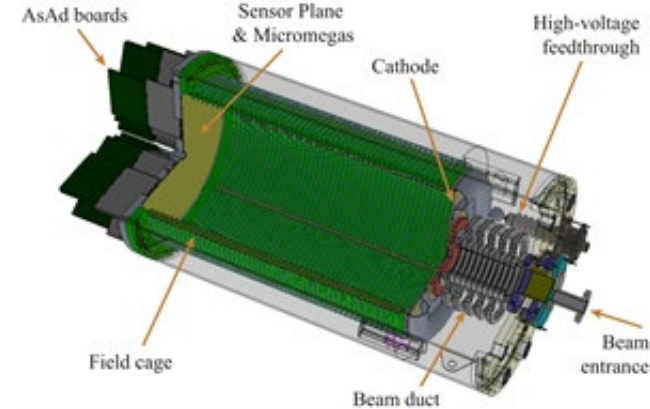
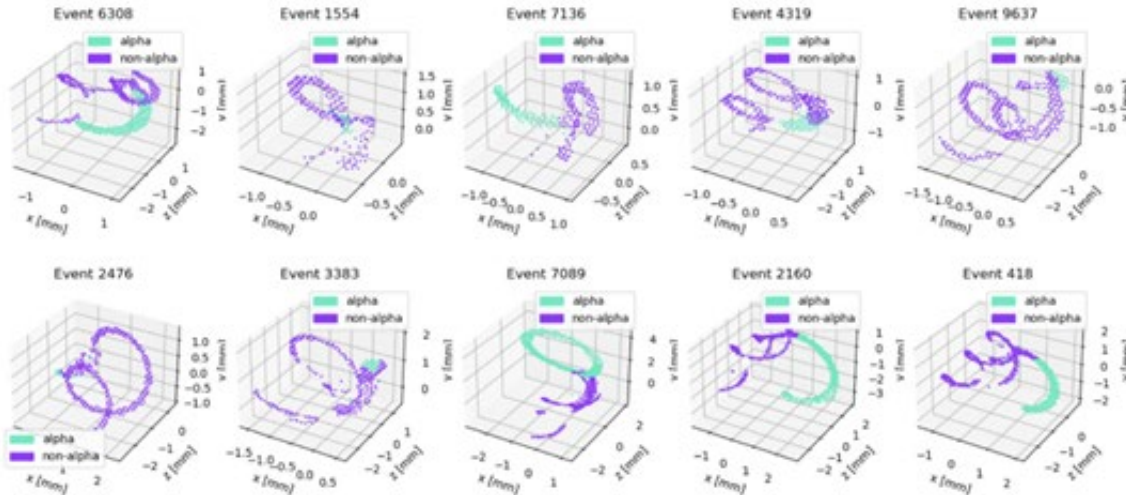
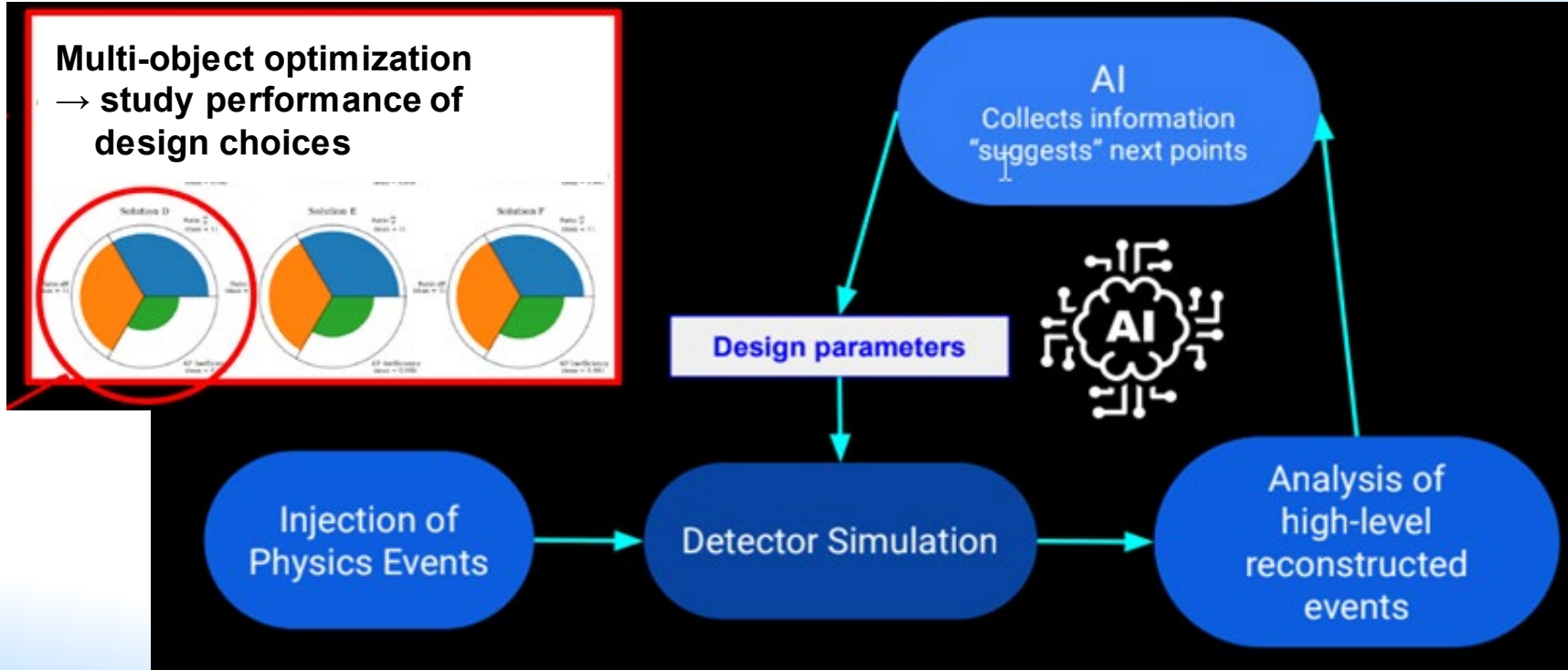


Image courtesy of Y. Ayyad, USC, Spain

Experimental Design

ML for experimental design

Image courtesy of C. Fanelli, MIT, USA



Facility Operation

ML in accelerator operation

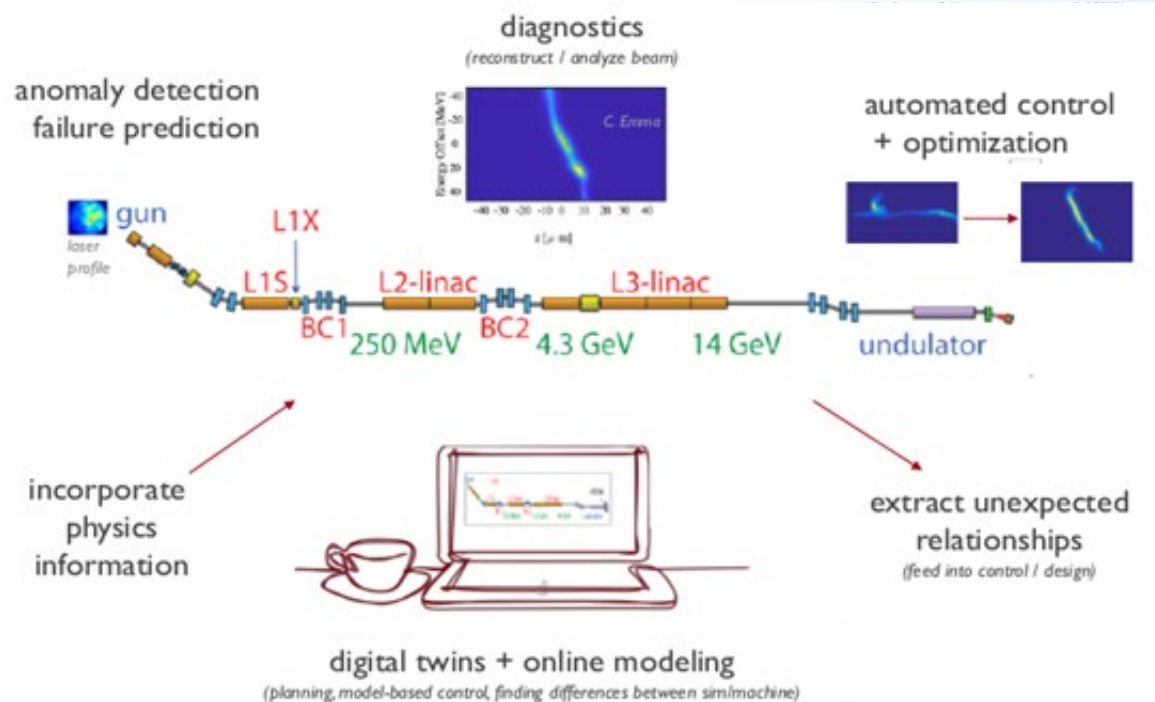


Image courtesy of A. Edelen, SLAC, USA

Facility Operation

ML in accelerator operation



Image courtesy of R. Auchetti, ANSTO, Australia



Machine-learning-based tuning in **seconds**

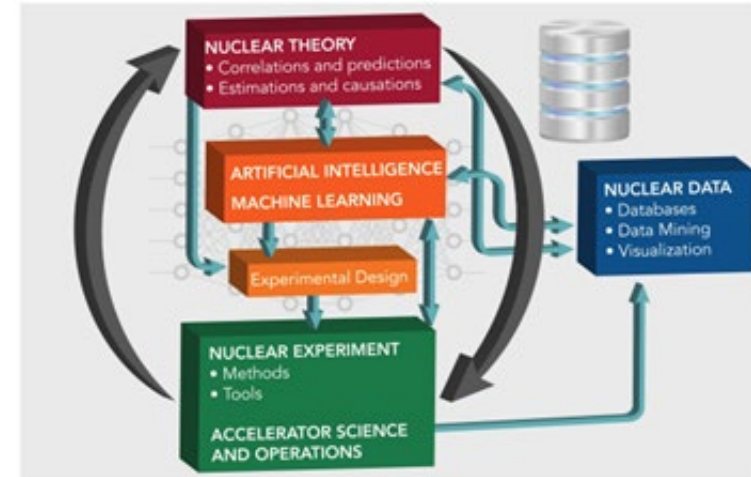
Hand tuning by expert in **10–20 min**

Image courtesy of A. Edelen, SLAC, USA

Cross-cutting Needs and Next Steps

Cross-cutting Needs

- Nuclear Science Drivers
 - Real-time systems
 - Uncertainty quantification in AI
- Education efforts
 - AI-focused summer schools and workshops
- Interdisciplinary funding
 - Positions
 - Research & Development
 - Production/deployment efforts



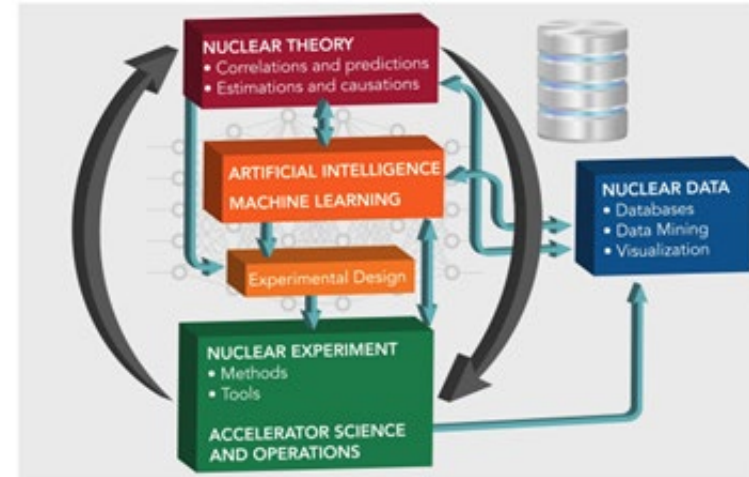
Cross-cutting Needs

Ongoing

- Nuclear Science Drivers
 - Real-time systems
 - Uncertainty quantification in AI

Next steps

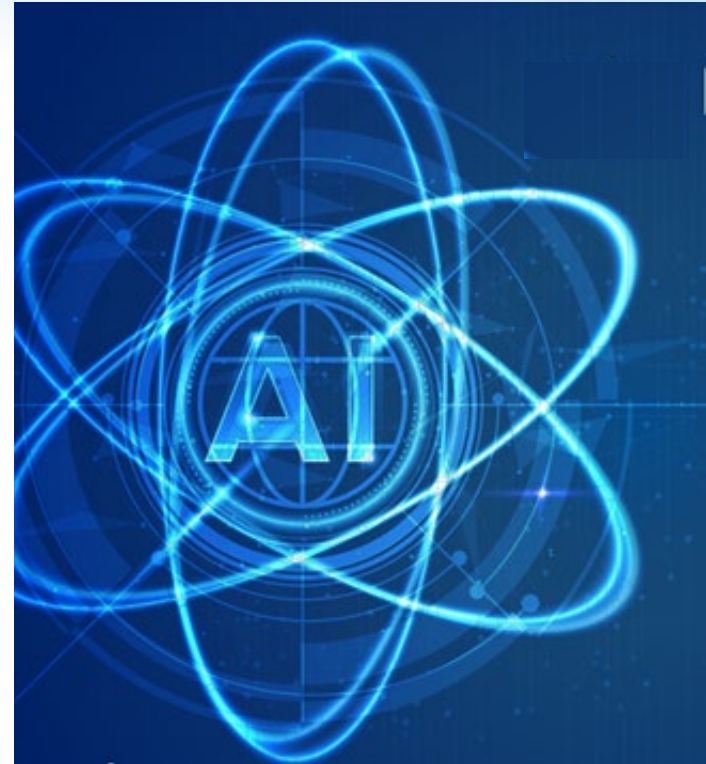
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Accelerating Progress—IAEA's Role



- Nuclear Science Drivers
 - Real-time systems
 - Uncertainty quantification in AI
- Sponsorship of community efforts
 - Data challenges / hackathons
 - Online resource center
 - Living review of AI in nuclear physics
 - Nuclear physics- based benchmarks
 - Datasets
 - Event listing
- Workforce development
 - Established summer schools
 - Internships and fellowships
- Interdisciplinary coordination
 - Networks
 - Coordinated research activities
 - Standards on non-expert controlled research



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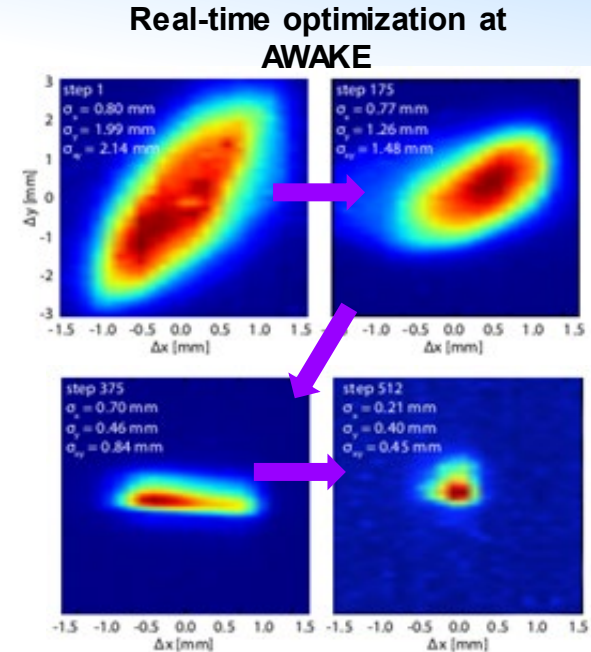
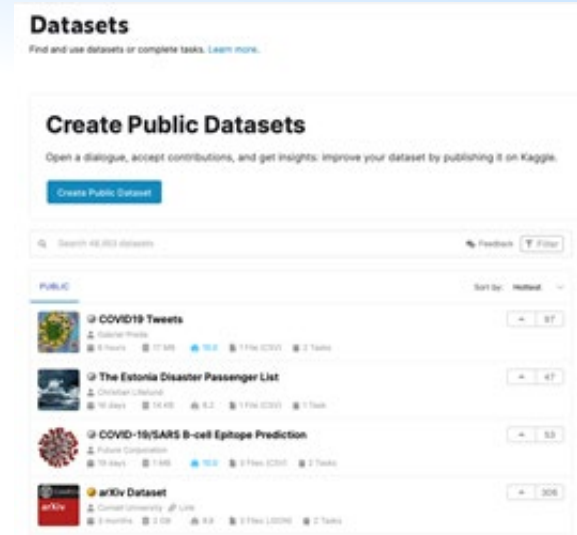


Image courtesy of A. Scheinker, LANL, USA

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HEPML-LivingReview

A Living Review of Machine Learning for Particle Physics

Modern machine learning techniques, including deep learning, is rapidly being applied, adapted, and developed for high energy physics. The goal of this document is to provide a nearly comprehensive list of citations for those developing and applying these approaches to experimental, phenomenological, or theoretical analyses. As a living document, it will be updated as often as possible to incorporate the latest developments. A list of proper (unchanging) reviews can be found within. Papers are grouped into a small set of topics to be as useful as possible. Suggestions are most welcome.

General Review

The purpose of this note is to collect references for modern machine learning as applied to particle physics. A minimal number of categories is chosen in order to be as useful as possible. Note that papers may be referenced in more than one category. The fact that a paper is listed in this document does not endorse or validate its content - that is for the community (and for peer-review) to decide. Furthermore, the classification here is a best attempt and may have flaws - please let us know if (a) we have missed a paper you think should be included, (b) a paper has been misclassified, or (c) a citation for a paper is not correct or if the journal information is now available. In order to be as useful as possible, this document will continue to evolve so please check back before you write your next paper. If you find this review helpful, please consider citing it using `[cite(hepmlivingreview)]` in HEPML bib.

• Reviews

• Modern reviews

- Jet Substructure at the Large Hadron Collider: A Review of Recent Advances in Theory and Machine Learning [DOI]
- Deep Learning and its Application to LHC Physics [DOI]
- Machine Learning in High Energy Physics Community White Paper [DOI]
- Machine learning at the energy and intensity frontiers of particle physics
- Machine learning and the physical sciences [DOI]
- Machine and Deep Learning Applications in Particle Physics [DOI]
- Modern Machine Learning and Particle Physics

• Specialized reviews

- The Machine Learning Landscape of Top Taggers [DOI]
- Dealing with Nuisance Parameters using Machine Learning in High Energy Physics: a Review
- Graph neural networks in particle physics [DOI]
- A Review on Machine Learning for Neutrino Experiments [DOI]

Accelerating Progress—IAEA's Role



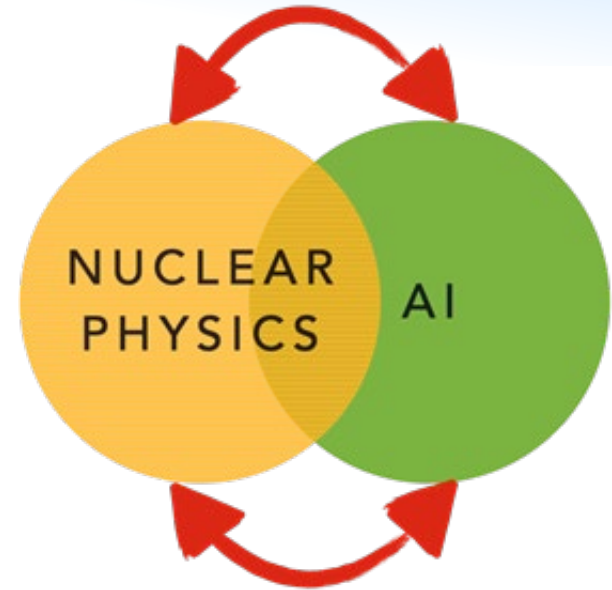
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Expected Outcomes



Acceleration of scientific discovery

Standards for non-export controlled research

Workforce development

More high-quality data

Data standards

Easier access to information

Improved beam quality

Common reporting standards

Engagement with adjacent fields

Increased beamtime / more experiments

Reproducibility

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Thank you!