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## Machine Learning for Nuclear Data

## REVIEWS OF MODERN PHYSICS

## Colloquium: Machine learning in nuclear physics

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- I. Introduction
- II. Machine Learning for Nuclear Physics in Broad Strokes
- III. Nuclear Theory
- IV. Experimental Methods
- V. Accelerator Science and Operations

### VI. Nuclear Data

- A. Overhauling the nuclear data pipeline
- B. Improving compilations and evaluations
- C. Building emulators and surrogate models
- VII. Summary and Perspectives



# Transforming the Nuclear Data Pipeline

#### Measurements

- **design** experiments / detectors with BO, emulators, GPs, NNs ...

#### Theory

- speed up large-scale calculations with HPC resources and emulators
- determine uncertainties in theory model using BNN

### Compilation

- choose keywords in NSR with NLP
- extract tables / plots / text with CNNs

#### Evaluation

- update ENDF using BN and GPs
- extract GDR energies & widths using NNs and MTL
- identify outliers using RF, and using experimental info w/ SVM, LR ...
- **combine data** differential / integral / theory using RFs

#### Processing / Validation and Verification

- reaction evaluation processing / V&V using BN and GP
- **design** validation experiments with BO

ML projects are already improving the pipeline

they have the potential to transform it ...

#### **GLOSSARY**

**BO** Bayesian Optimization

**GP** Gaussian Process

**NN** Neural Network

**BNN** Bayesian NN

**NLP** Natural Lang. Proc.

**CNN** Convolutional NN

**BN** Bayesian Network

MTL Multi-Task Learning

**SVM** Supp. Vector Mach.

LR Logistic Regression

**RF** Random Forests

### Applications

- identify criticality factors using RF
- Uncertainty Quantification (UQ) using emulators

# Transforming the Nuclear Data Pipeline

future trends...

Reaction pipeline formulated as a BN

UQ and radiation transport with emulators

Physics-driven loss terms in NNs

Validation & Verification using BO

- Recommend that the potential of ML to transform Nuclear Data Pipeline be discussed in NSAC LRP and NSAC-Nuclear Data Subcommittee Report
- Recommend that ML be a part of future Funding Opportunity Announcements from NDIAWG / DOE NP
- Recommend that USNDP establish a "Machine Learning Task Force" to share approaches / algorithms / strategies
- Recommend development of toolkits for non-experts on ML for nuclear data