

Machine Learning for Nuclear Astrophysics

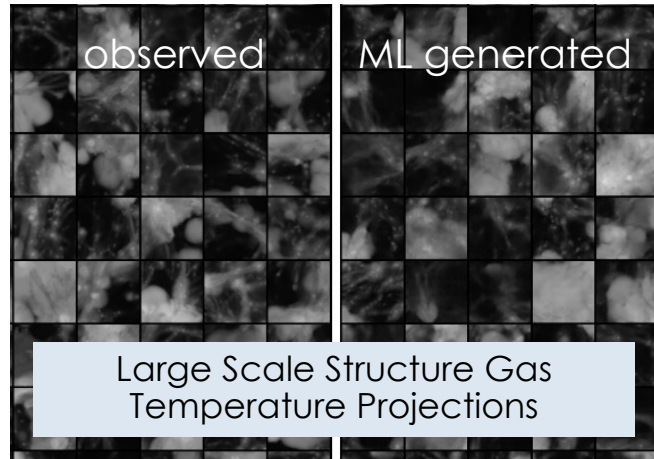
Michael Smith
Physics Division
ORNL

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REVIEWS OF MODERN PHYSICS

Colloquium: Machine learning in nuclear physics

Amber Boehnlein, Markus Diefenthaler, Nobuo Sato, Malachi Schram, Veronique Ziegler, Cristiano Fanelli, Morten Hjorth-Jensen, Tanja Horn, Michelle P. Kuchera, Dean Lee, Witold Nazarewicz, Peter Ostroumov, Kostas Orginos, Alan Poon, Xin-Nian Wang, Alexander Scheinker, Michael S. Smith, and Long-Gang Pang
Rev. Mod. Phys. **94**, 031003 – Published 8 September 2022

- Machine learning (ML) widely used in Astrophysics, beginning to be used in nuclear physics
- tremendous untapped potential for *nuclear astrophysics*
- recommend that
 - the potential of ML to advance nuclear astrophysics be discussed in NSAC LRP
 - ML be a part of future relevant Funding Opportunity Announcements
 - ML toolkits for non-experts be developed for nuclear physics & nuclear astrophysics

Machine Learning in Nuclear Astrophysics - Possibilities

- *Experimental Planning, Design, Execution*
 - **speed** sensitivity studies with emulators, feature selection (DT, RF, SVM)...
 - **design** detectors with BO, GP ...
 - **tune** separators and systems with BO, GP ...
 - **choose** energies with NN, DT, RF, SVM ...
- *Data Analysis*
 - **identify / classify** events with KM, KNN ...
 - **removing hidden fitting variables** with DT, FA, RF ...
- *Computational / Theoretical*
 - **identify outliers** in inputs (e.g., rates) & in outputs (e.g., abundances) of nucleosynthesis simulations using RF, SVM, LR, KM, IF ...
 - **find patterns** in nucleosynthesis flows with NN, CNN, SVM, KM ...
 - **interpret** nucleosynthesis tracer particle results with KM, KNN, RF, PCA ...
 - **determine uncertainties** in models with BNN, emulators ...
 - **speed up** full hydro + nuclear simulations with emulators
 - **find functional forms** using NN + SR
 - **reduce dimensionality** (remove non-observable params) with PCA, VAE, GDA..

GLOSSARY

DT	Decision Tree
RF	Random Forests
SVM	Supp. Vector Mach.
BO	Bayesian Optimization
GP	Gaussian Process
NN	Neural Network
KM	K means
KNN	K Nearest Neighbors
FA	Factor Analysis
RF	Random Forests
LR	Logistic Regression
IF	Isolation Forest
CNN	Convolutional NN
BNN	Bayesian NN
SR	Symbolic Regression
PCA	Principal Comp/ Analy.
VAE	Var. AutoEncoders
GDA	Gen. Discrim. Analy.