

Fair Universe

Unbiased Data Benchmark Ecosystem for Physics HiggsML Uncertainty Challenge

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https://fair-universe.lbl.gov/













Fair Universe: Unbiased Data Benchmark Ecosystem for Physics

Project aims to:

- Provide a large-compute-scale Al ecosystem for hosting challenges and benchmarks.
- Organize a challenge series initially focused on measuring and minimizing the effects of systematic uncertainties in HEP (particle physics and cosmology).



Large-compute-scale AI ecosystem for hosting challenges and benchmarks

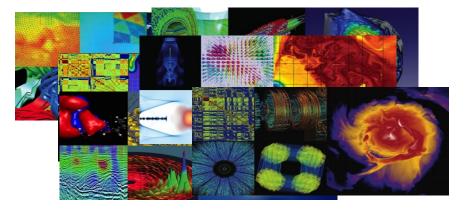
NERSC: Mission HPC for the Dept. of Energy Office of Science



Large compute systems

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- E.g. Perlmutter: ~7k A100 GPUs Also high-capacity/ fast filesystems, 1 Tbit/s WAN and flexible services
 - E.g. SPIN: Rancher/K8s platform for user-defined services



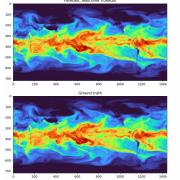
Broad science user base

- > 10,000 users.
- > 1000 projects
- Across all DoE Science e.g. HEP; NP; Climate; Fusion Chemistry; Materials; Genomics: etc ...



NERSC-AI Ecosystem:

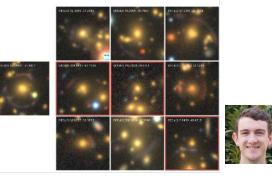
- Deploy optimized hardware and software (working with vendors)
 Improve performance, e.g through benchmarking (MLPerf HPC)
 - **Apply** cutting edge AI for science: e.g. "NESAP" program with postdocs
- *Empower* through e.g. over 20 DL@Scale tutorials, 1000s of total participants: (<u>SC23</u>)
- Many Al for science highlights not covered here. e.g.





Collab with Nvidia, Caltech

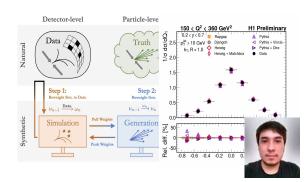
First DL model with skill of numerical weather prediction (NWP) Train up to 1000s of GPUs Forecasts 1000s times faster than NWP



Self-supervised sky

SUIVEYS Stein et. al. (2021)

Foundation-like model trained at scale - used for downstream tasks E.g. uncovered thousands of undiscovered strong-lenses



Unfolding particle physics

H1 Collaboration ([...] <u>Mikuni et. al.</u>): Reanalyzing previous experiment data to unfold many quantities Utilized Perlmutter for 1000s of bootstrapping and UQ runs 5

Codabench/"Fair Universe" Platform

- Codabench open source platform for Al benchmarks and challenges
 - Originally (CodaLab) Microsoft/Stanford now a Paris-Saclay/<u>LISN</u> led community
 - > 500 challenges since 2013
 - Allows code submission as well as results
 e.g. for evaluation timing or reproducibility
 - Also data-centric AI "inverted competitions"
 - Organizers can define scoring functions
 - Queues for evaluation can run on diverse compute resources
 - Platform itself can be deployed on different compute resources

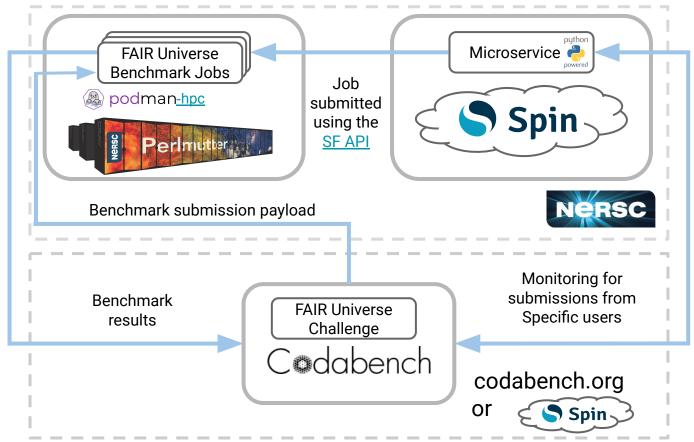
"Fair Universe" brings Codabench to HPC at NERSC!



Based on https://www.codabench.org/



Fair Universe Platform: Codabench/NERSC integration

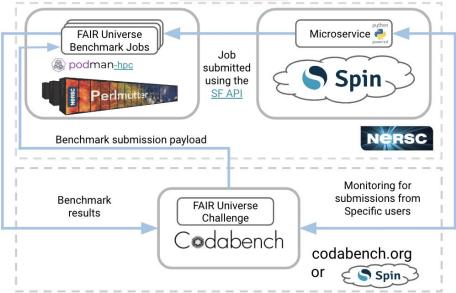




Fair Universe Platform: Codabench/NERSC integration

- Benchmark submissions pulled to workers running on Perlmutter:
 - Use <u>podman(-hpc</u>) container runtime: secure and scalable
 - Enable parallelism/scale for
 - Intensive methods use multiple
 A100 GPUs for training or evaluation
 - Many participants through running many parallel workers
 - Many evaluations e.g for Uncertainty Quantification
- Workers submitted as needed by microservice on SPIN service platform
 - NERSC's "<u>SF API</u>" for job submission
 - Monitor/filter submissions

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- Also deploy instances of Codabench platform itself within SPIN
 - Customization and future OIDC integration with NERSC authorization

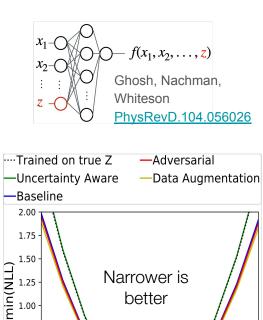
Organize a challenge series focused on measuring and minimizing the effects of systematic uncertainties in HEP

Bias and uncertainty in ML in HEP

- ML trained on simulation with estimated systematics ("Z")
 - Then applied in data with different state Z'
- Commonly dealt with by shifting Z and measuring impact Non-optimal so many other techniques proposed - e.g.:
 - Decorrelation augmentation; adversarial training
 - "Uncertainty-aware" Ghosh, Nachman, Whiteson <u>PhysRevD.104.056026</u>
 Darameterize classifier using Z
 - Parameterize classifier using Z
 - Other novel approaches e.g. (not comprehensive)
 - Inferno: <u>arxiv:1806.04743;</u>

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- Full profile-likelihood: e.g. arxiv:2203.13079
- Mainly demonstrated on single systematic with limited data
 - Can be hard to scale, e.g. retraining and profiling expensive
- Need for larger datasets and novel metrics to compare uncertainty quantification for ambitious approaches



1.0

μ

0.8

1.2

[•] 0.75 **TI** 0.50

0.25

.00 1.25

0.75

0.6



Fair Universe: HiggsML Uncertainty Challenge

Search Competitions		Ť *	Benchmarks 🝷 📑 Resou	rces 🛛 🖵 Queue Managem	ent 🛆 wbhimji 👻	
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	CURE PARTCEPARTS Submission's Dumps regare ORGANIZED BY: Ihsan Ullah (fair-universe@lbl.gov) CURRENT PHASE ENDS: March 14, 2024 At 5:00 PM PDT CURRENT SERVER TIME: February 29, 2024 At 11:37 AM PST Docker image.chl/fair_universeć26/7393 I Secret url: https://www.codabench.org/competitions/2044/?secret_key=300b4901-767c-4399-8e77-40a558357/c5 I					
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Get Started	Phases	My Submissions	Results	Forum	?	
Overview	Overniew					
Evaluation	Overview					

- Extension of <u>HiggsML Kaggle challenge</u> $(H \rightarrow \tau \tau)$ from 2014
- HiggsML Uncertainty Challenge Pilot launched today
 - Evaluate methods and metrics and gain feedback
- Full HiggsML Uncertainty Challenge ~May-Oct 2024

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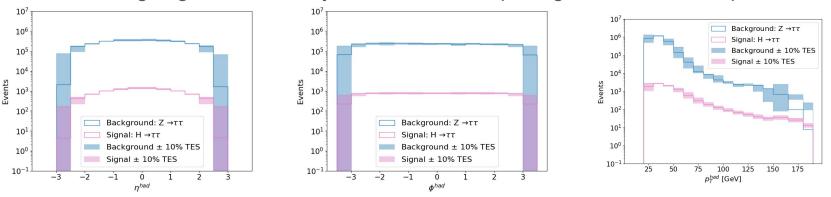
Submitted as <u>NeurIPS competition</u> - results presented at NeurIPS in December

http://go.lbl.gov/fair-u niverse-higgsml-spri ng24

HiggsML Uncertainty Challenge - Datasets

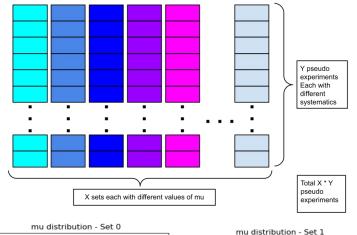
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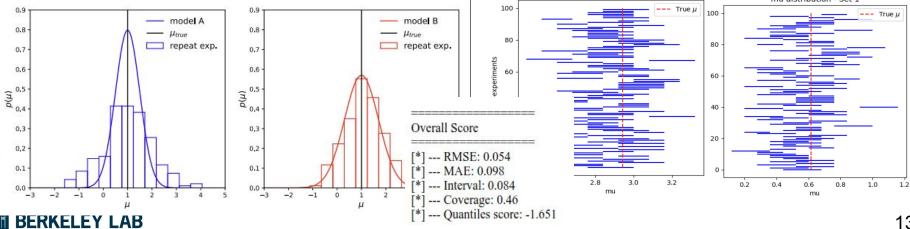
- Much larger datasets than original HiggsML challenge (12*10⁶ training events)
- <u>Delphes</u> detector simulation (see "Data" tab on competition page for details)
- Apply parameterized systematics (Nuisance Parameters) :
 - In current pilot: Tau Energy Scale shift hadronic Tau Pt (and correlated MET)
 - Further planned for full competition: Jet Energy Scale (and correlated MET impact); Additional randomised Soft MET; Background normalisation; W-boson background normalisation
 - Also ongoing work to add systematics into Delphes generation for comparison



HiggsML Uncertainty Challenge - Evaluation

- Form multiple pseudo-experiment test sets: different signal strengths (μ) and systematics Current pilot - 4μ and 50 pseudo-experiments
- Task: predict uncertainty interval $[\mu_{16}, \mu_{84}]$
 - E.g. 68% quantile of likelihood or assume 1σ 0
- Score: balance accuracy and precision with matching the true uncertainty and accuracy





Uncertainty Quantification Metric

- Interval width (w) averaged over N test sets
- **Coverage (c)**: fraction of time μ is contained
- Combined using a **coverage function f(x)**:

$$x \ge 0.68 - 2\sigma_{68}$$
 and $x \le 0.68 + 2\sigma_{68}$: 1. $x < 0.68 - 2\sigma_{68}$: $1 + |rac{x - (0.68 - 2\sigma_{68})}{\sigma_{68}}|^4$ $x > 0.68 + 2\sigma_{68}$: $1 + |rac{x - (0.68 + 2\sigma_{68})}{\sigma_{68}}|^3$ with $\sigma_{68} = rac{\sqrt{(1 - 0.68)0.68N)}}{N}$

- N dependance for equivalent ideal coverage
- Penalizes undercoverage more

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• Final score (s) designed to avoid large values or gaming

$$w = \frac{1}{N} \sum_{i=0}^{N} |\mu_{84,i} - \mu_{16,i}|$$

= $\frac{1}{N} \sum_{i=0}^{N} 1 \operatorname{if}(\mu_{true,i} \in [\mu_{84,i} - \mu_{16,i}])$
 $\underbrace{\mathbb{E}_{10^{0}}}_{10^{0}} \int_{0.5}^{10^{1}} \int_{0.6}^{0.6} \int_{0.7}^{0.8} \int_{0.8}^{0.9} \int_{0.9}^{0.8} \int_{0.9}^{0.9} \int_{0.9}^{1.0}$

$$s=-\ln\left((w+\epsilon)f(c)
ight)$$

х

See also <u>Sascha Diefenbacher's AISSAI Workshop presentation</u>

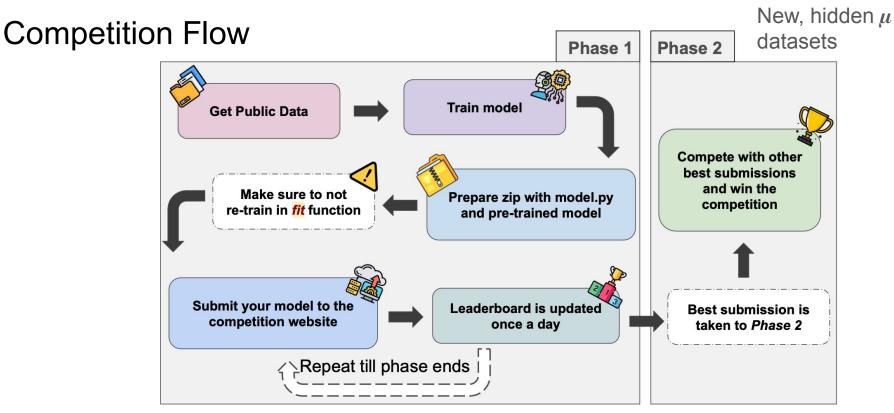
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Enter the HiggsML Uncertainty Challenge Pilot!

See more detailed: walkthrough slides; and "starting kit" \bullet

http://go.lbl.gov/ fair-universe-hig asml-spring24

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	0	FAIR UNIVERSE - HIGGSML UNCERTAINTY CHALLENGE	
		ORGANIZED BY:FAIR Universe CURRENT PHASE ENDS: 31 March 2024 At 05:00 GMT+ CURRENT SERVER TIME:5 March 2024 At 15:05 GMT+ Decker Image: newsc/fak_universe:1980bal ₪	6. Check out the starting kit
	Get Started	Apr 2024 Apr 2024 Phases My Submissions Results	FAIR UNIVERSE - HIGGSML UNCERTAINTY CHALLENGE
	Evaluation Data Starting Kit	Overview	ORGANIZED BY:FAIR Universe CURRENT PHASE ENDS:31 March 2024 At 05:00 GMT+5 CURRENT SERVER TIME:5 March 2024 At 15:05 GMT+5 Dockeringer news/far_universe12900at ₽
	Example Estimators Terms Files	In 2012, the Nobel-prize-winning discovery of the Higgs Boson by the ATI Collider (LHC) at CERN In Geneva, Switzehard was a major millestone in validation it provided of the Standard Model of particle physics (SM), the SM does not answer. One promissing approach to uncover some of thes detail, as the rate of Higgs Boson production and its decay properties mill other phenomena not explained by the SM.	Mar 2004 Apr 2004 Get Started Phases My Submissions Results Forum
		The LHC collides protons together at high energy and at a high rate. Each	Cverview Evaluation Starting Kit and Sample Submission
			Starting Kit Use are providing a starting kit a Google Colab notebook to demonstrate the problem and a solution for it which can be submitted as a submission the comprise. You can copy the Colab notebook and make changes as you want. Terms
			Files Dummy Sample Submission Dummy sample submission is provided to make you understand what is expected as a submission. You can modify the sample submission the way you want but make sure the format is the same as instructed in the sample submission
BERKEL	EY LAB		Dummy Sample Submission



- Submissions are run, and the leaderboard updated, at least once a day
- Participants code is run on the platform, in parallel on many test sets, so there can be differences to when tested locally contact us on <u>slack</u> with any issues

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Conclusions

- We have built a flexible platform for hosting challenges and benchmarks extending Codabench, backed by HPC at NERSC
- Launching a series of challenges for uncertainty aware methods for HEP
 - You can enter the HiggsML Uncertainty Challenge Pilot now!
 - <u>http://go.lbl.gov/fair-universe-higgsml-spring24</u>
- Open to feedback on metrics/datasets to encourage advanced approaches and ensure challenge is interesting

Help and feedback: <u>#higgsml-uncertainty-challenge-spring-24</u> channel on the <u>Fair Universe Slack workspace</u>

Ongoing information Google Group: Fair-Universe-Announcements

Collaborations, questions, comments: <u>fair-universe@lbl.gov</u>

General collaboration on NERSC-AI: <u>wbhimji@lbl.gov</u>

