

# DeepHyper: Automated Machine Learning at Scale

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# **AI/ML** for DOE applications



#### Diverse data types



# **Degrees of Freedom in Neural Networks Design**





~10<sup>10</sup> possible designs



Manual design: trial and error and time consuming (aka graduate student descent)



## **Degrees of Freedom in Neural Networks Design**

#### **Algorithm Hyperparameters**



Optimizer: SGD, RMSprop, Adam... Learning rate Minibatch size Learning rate scheduler Adaptative batch size

. . .

#### **Architecture Variables**



Number of layers Type of the layer: Fully Connected, Convolution, Recursive... Activation function Dropout rate Skip connection

. . .



# **DeepHyper: Scalable AutoML**



DeepHyper documentation: http://deephyper.readthedocs.io



# **Asynchronous Bayesian Optimization**

Algorithm: Asynchronous BO /\* Initialization \*/ 1 optimizer  $\leftarrow$  optimizer() 2 for  $i \leftarrow 1$  to W do  $configs.h_m \leftarrow random_point(H_m)$ 3 submit\_evaluation(configs) // Nonblocking 4 5 end /\* Main loop \*/ 6 while not done do // Query results  $results \leftarrow get_finished_evaluations()$ 7 if |results| > 0 then 8 optimizer.tell(results.h<sub>m</sub>, results.valid\_accuracy) 9  $next \leftarrow optimizer.ask(|results|)$ 10 submit\_evaluation(next) // Nonblocking 11 end 12 13 end



Unevaluated parameter

#### *Surrogate Model Fitted to Sampled Performance (iterative refinement improves the learning model)*



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# **Configuring Neural Architecture Search**

### How do we define a space of neural networks?

A neural network search space can be represented as a directed acyclic graph with nodes and edges.

- Nodes represent possible operations, for example:
- 1. Add an identity layer
- 2. Add a layer with 40 neurons
- 3. Add a layer with 60 neurons
- 4. Add a dropout operation
- 5. Add a skip connection to another node
- Nodes can be constant (i.e., predefined and immutable during the search)
- Nodes can be variable (i.e., the search can tweak these to get better performance)
- Each variable node has an upper bound on the number of operations (which may be expressed as a categorical variable). Edges define the flow of the tensor in the graph





## Skip Connections



#### Loss surface: with and without skip connections Physics-informed neural networks

Li, Hao, Zheng Xu, Gavin Taylor, Christoph Studer, and Tom Goldstein. "Visualizing the loss landscape of neural nets." Advances in neural information processing systems 31 (2018). A. Krishnapriyan, A. Gholami, S. Zhe, R. Kirby, and M. W. Mahoney. "Characterizing possible failure modes in physics-informed neural networks." Advances in Neural Information Processing Systems 34 (2021).



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## DeepHyper NAS-API

search\_space = create\_search\_space(num\_layers=5)
ops = [random() for \_ in range(search\_space.num\_nodes)]
search\_space.set\_ops(ops)
model = search\_space.create\_model()
model.summary()
plot\_model(model, to\_file='sampled\_neural\_network.png', show\_shapes=True)
print("The sampled\_neural\_network.png file has been generated.")

#### Fun to generate random architectures!

172,424

parameters







#### 344,424 Parameters (more skips/layers)



Istm\_30: LSTM input: (None, 8, 16) output: (None, 8, 5)

## **Exploring Search Space**

Regularized ageing evolution to explore the search space of possible architectures



Real, Esteban, et al. "Regularized evolution for image classifier architecture search." Proceedings of the AAAI conference on artificial intelligence. Vol. 33. 2019.



## Searching for a Surrogate LSTM: Sea Surface Temperature Forecasting



R. Maulik, R. Egele, B. Lusch, and P. Balaprakash. Recurrent Neural Network Architecture Search for Geophysical Emulation. In SC '20: IEEE/ACM International Conference on High Performance Computing, Networking, Storage and Analysis, 2020.

## Searching for a Surrogate LSTM: Sea Surface Temperature Forecasting



ALCF/Theta

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# Cancer Drug Response: predicting the activity of a drug treatment against a cancer cell

#### Method–GPU(s)/Eval–Nodes



#### ALCF/ThetaGPU

R. Egele, P. Balaprakash, I. Guyon, V. Vishwanath, F. Xia, R. Stevens, Z. Liu. AgEBO-tabular: joint neural architecture and hyperparameter search with autotuned data-parallel training for tabular data. In SC '21: IEEE/ACM International Conference on High Performance Computing, Networking, Storage and Analysis, 2021.

## **Active Research Topics**

# • AutoML at (exa-)scale

- Deep ensembles and uncertainty quantification
- Multiobjective optimization (accuracy, inference time, #params)
- Designing insect-brain inspired learning algorithms for neuromorphic computing
- Software/hardware co-design, compiler optimization, data services management
- Continual/transfer learning across similar tasks

# **The DeepHyper Community**



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