

Advancing Scientific Research with AI: Collaborative Efforts in Training Large Foundation Models on the Frontier Supercomputer

Prasanna Balaprakash

Director of Al Programs

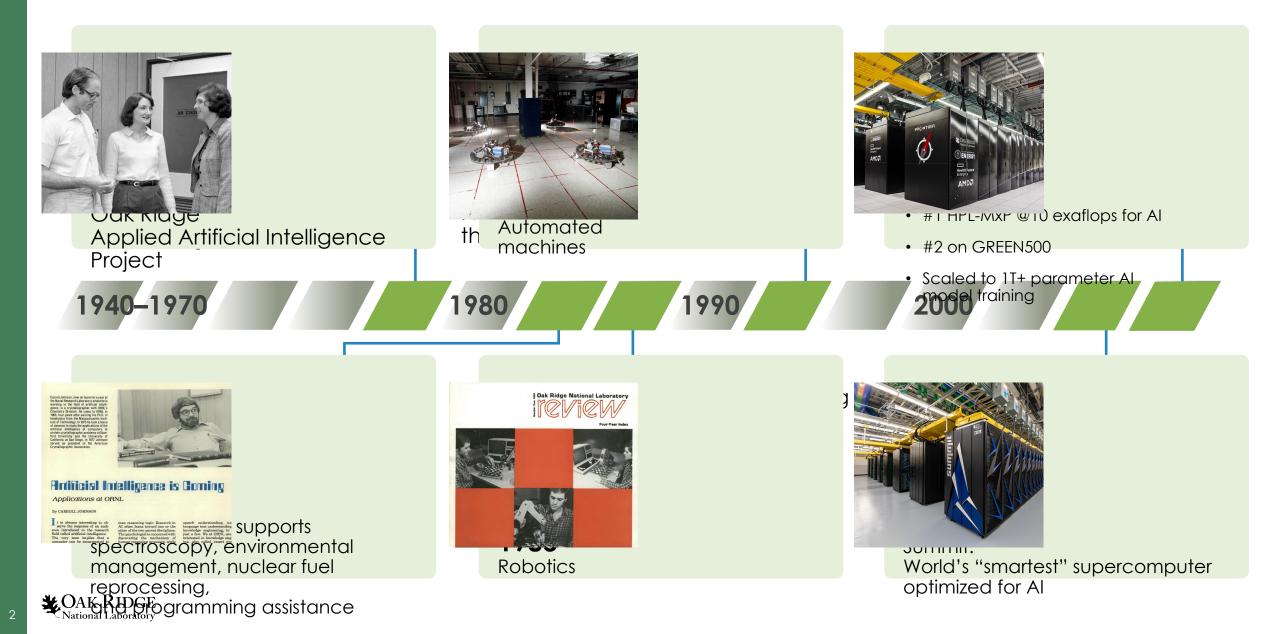
Oak Ridge National Laboratory

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Sajal Dash, Isaac Lyngus, Junqi Yin, Xiao Wang, Romain Egele, Austin Elis, Guojing Cong, Aristeidis Tsaris, Dan Lu, Feiyi Wang Massimiliano Lupo Pasini, Jong Choi, Xiao Wang, Pei Zhang, John Gounley, and Paul Laiu



ORNL has a rich history leveraging AI for science



Al transforming science and national security Accelerating scientific discovery, fortifying energy infrastructure, and enhancing national security





Grand challenges in AI for science nature

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EDITORIAL 27 September 2023

AI will transform science – now researchers must tame it

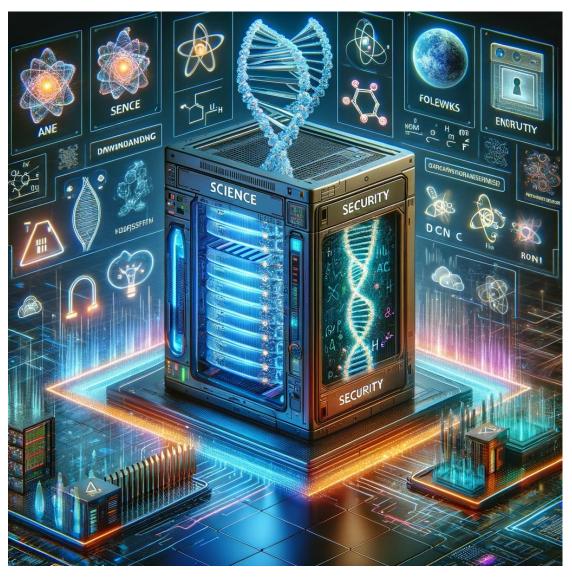
A new *Nature* series will explore the many ways in which artificial intelligence is changing science – for better and for worse.

OCTOBER 30, 2023

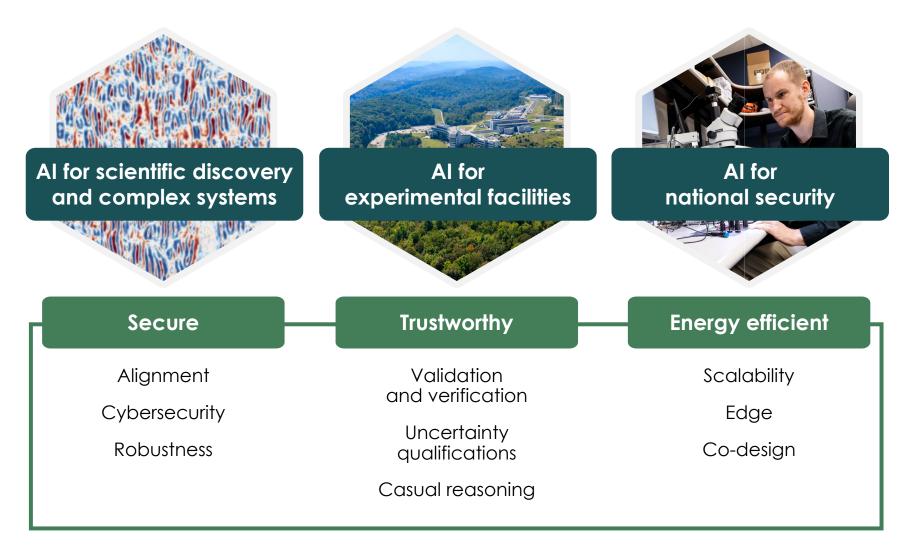
Executive Order on the Safe, Secure, and Trustworthy Development and Use of Artificial Intelligence

BRIEFING ROOM > PRESIDENTIAL ACTIONS

By the authority vested in me as President by the Constitution and the laws of the United States of America, it is hereby ordered as follows:



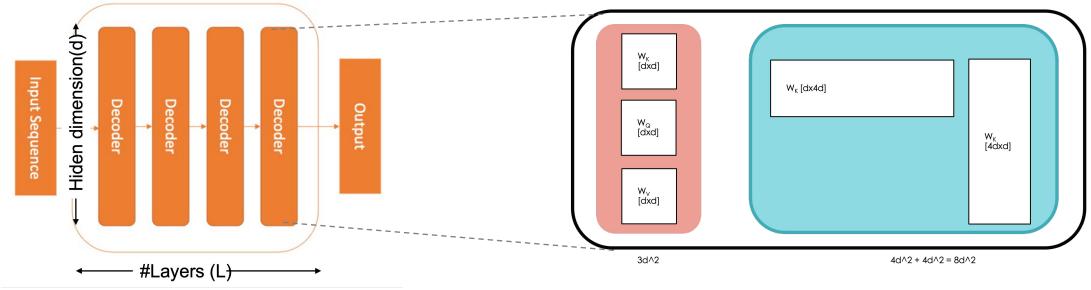
ORNL's Al initiative Secure, trustworthy, and energy-efficient Al





Resources needed to train LLM-like models

One layer has 11d² parameters



- GPT like models grows linearly with #Layers and quadratically with #Hidden dimensions
- Total parameters is roughly 12*L*d²

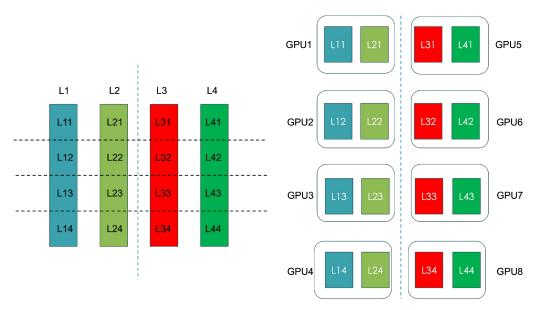
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- Training one parameter needs 20 Bytes of memory and 120 FLOPs
- Need a minimum of 8, 55, and 313 MI250X GCDs to fit models of size 22B, 175B, and 1T

	Memory Requirement		
Values	22B Model	175B Model	1T Model
Parameters (6x)	132 GB	1050 GB	6 TB
Gradients (4x)	88 GB	700 GB	4 TB
Optimizer States (8x)	176 GB	1.4 TB	8 TB
Total Memory (20x*)	440 GB	3.5 TB	20 TB

Optimizing distributed training strategies

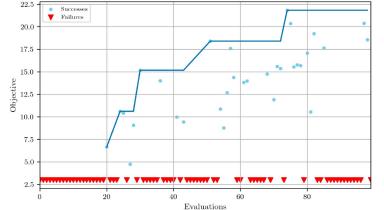
- Ported SOTA LLM training frameworks and libraries to Frontier (Megatron-DeepSpeed, FlashAttention 1&2)
- GPT model can be trained with a combination of Tensor (TP), Pipeline (PP), and Data parallelism (DP)
- TP slices model horizontally, PP vertically, and DP replicates
- TP should limit within node, PP should use large #micro-batches



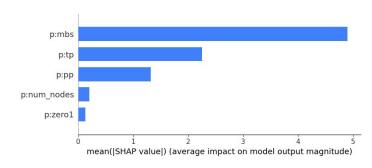


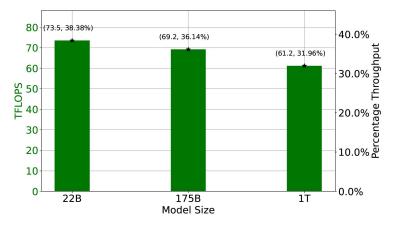
Training LLMs with 1 trillion parameters

- Automatically identify the best distribution strategies to achieve high GPU throughput and scaling efficiency
- Achieved 100% weak scaling efficiency and 87.05% strong scaling efficiency at 3072 GPUs.
- Trained the models for 10-100 iterations instead of training till completion.



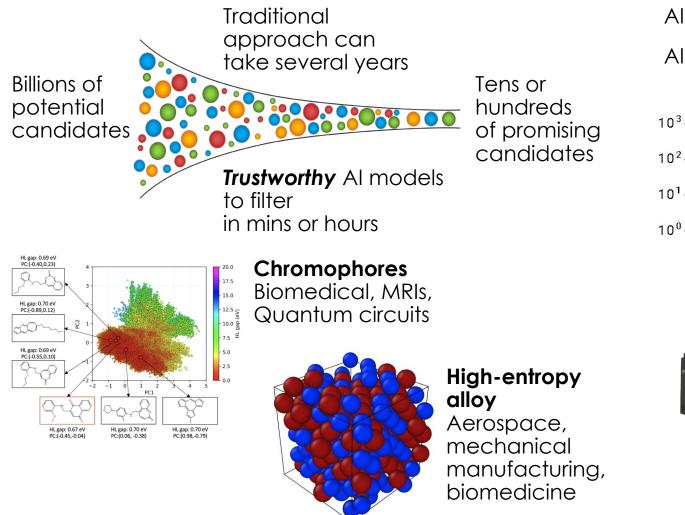
Hyperparameters	Value		
	175B Model	1T Model	
TP	4	8	
PP	16	64	
MBS	1	1	
GBS	640	1600	
ZeRO Stage	1	1	
Flash Attention	v2	v2	
Precision	fp16	fp16	
checkpoint-activations	True	True	







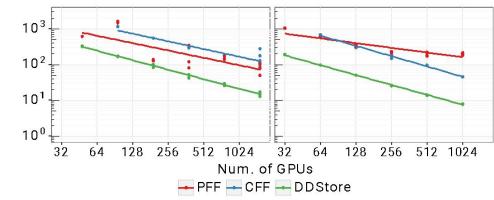
Accelerated materials discovery via trustworthy AI models on Frontier



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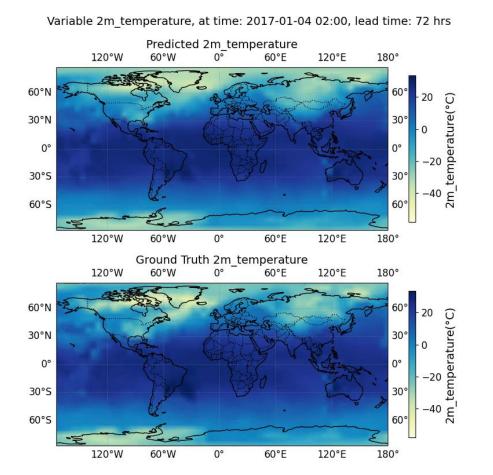
Al model training on 32 GPUs takes 100 hours

Al model training on 1,024 GPUs takes 10 hours

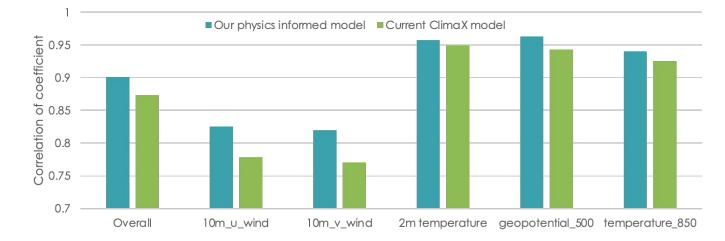




Foundation model for climate and weather forecasting

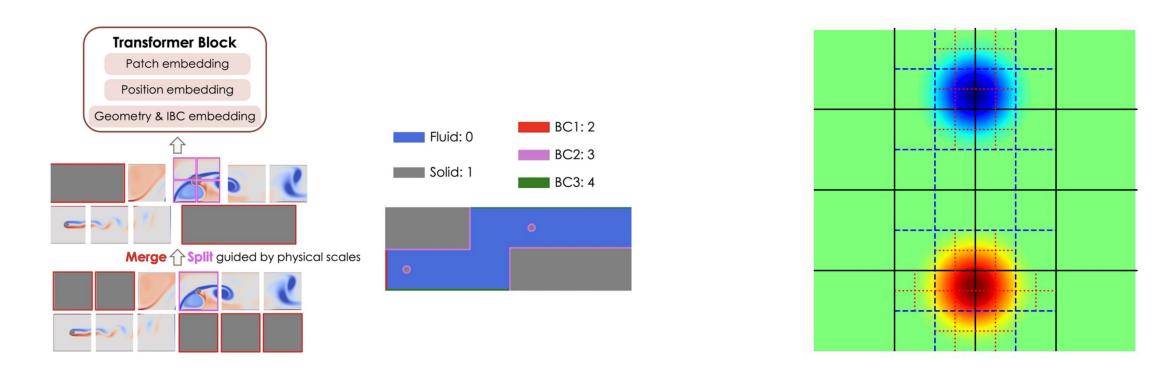


- Scaled our AI foundation model with 1B to 5B parameters on Frontier supercomputer;
- 87% scaling efficiency.
- Our Al foundation model accurately forecasts weather 72 hours ahead.





Foundation models for fluid dynamics



- - Adaptivity guided by physics, like Adaptive Mesh Refinement (AMR)
 - Hierarchical transformers: Transformers at various scales with a two-way coupling

Data credits:

Miniweather by Matt Norman (ORNL), https://github.com/mrnorman/miniWeather "CGL @ ETHZ - Data."

"CGL @ EIIZ - Data." https://cgl.ethz.ch/research/visualization/data.php (accessed Jun. 08, 2023). Open slide master to edit



Frontier first indicates path for energy efficient scaling



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251	Personal perspect	ives of a supercomputing enthusiast	
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Lockwood California,	Labels	January 13, 2024	
States 	ai applications	A CLOSER LOOK AT "TRAINING" A TRILLION-PAR. MODEL ON FRONTIER —	AMETER
		A paper titled "Optimizing Distributed Training on Frontier for Large Langu been making its rounds over the last few weeks with sensational taglines sa trained a trillion-parameter model using only a fraction of the Frontier supe superficiality of the discourse around this paper seemed suspicious to me, s interests of embracing my new job in AI systems design, I decided to sit dow manuscript and figure out exactly what the authors did myself.	ying the authors ercomputer. The so in the
computing		manuscript and ngure out exactly what the authors the myself.	

- First time in open science setting and on non-NVIDIA hardware
 - So far, large runs on only NVIDIA hardware
- No heroic effort in software engineering
 - Largely dependent on existing software like Megatron-DeepSpeed
- Train massive LLMs with good efficiency using an entirely NVIDIA-free supercomputer
- Don't need an army of researchers at a national lab to make
 productive use of AMD GPUs for training large LLMs

Frontier trained a ChatGPT-sized large language model with only 3,000 of its 37,888 Radeon GPUs — the world's fastest supercomputer blasts through one trillion parameter model with only 8 percent of its MI250X GPUs

ews By Matthew Connatser published January 07, 2024

Now you're playing with AI power!

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Researchers at Oak Ridge National Laboratory trained a large language model (LLM) the size of ChatGPT on the Frontier supercomputer and only needed 3,072 of its 37,888 GPUs to do it. The team published a research paper that details



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