

# Running Deep Learning Applications on HPC System

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Introduction

ML Perf

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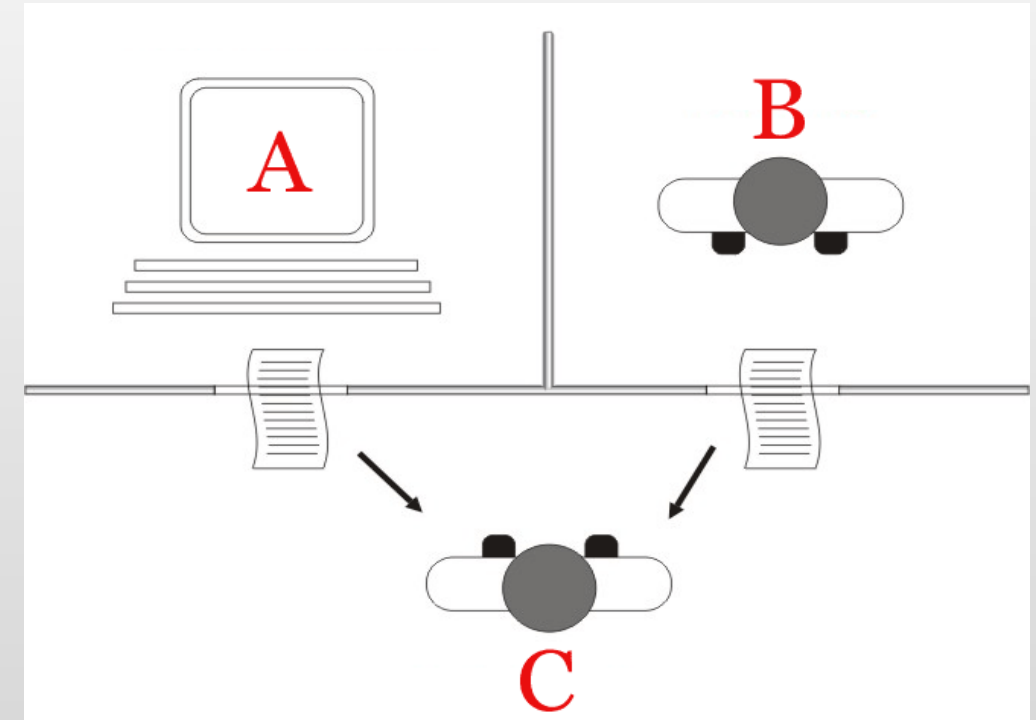
Running DL  
Models

# History of AI

## Pre-1950s: Foundations of AI

*The formal study of AI began with mathematicians and philosophers like Alan Turing.*

*In 1950, Alan Turing proposed the famous "Turing Test," a measure of a machine's ability to exhibit human-like intelligence. This concept laid the groundwork for future AI research.*



The "standard interpretation" of the Turing test, in which player C, the interrogator, is given the task of trying to determine which player – A or B – is a computer and which is a human. The interrogator is limited to using the responses to written questions to make the determination.

# History of AI

## 1950s and 1960s: Early Research and Symbolic AI

*The term "artificial intelligence" was coined in 1956 at the Dartmouth Conference, where the field of AI research was officially established.*

*During this period, researchers focused on symbolic AI, using rules and logic to mimic human problem-solving processes.*



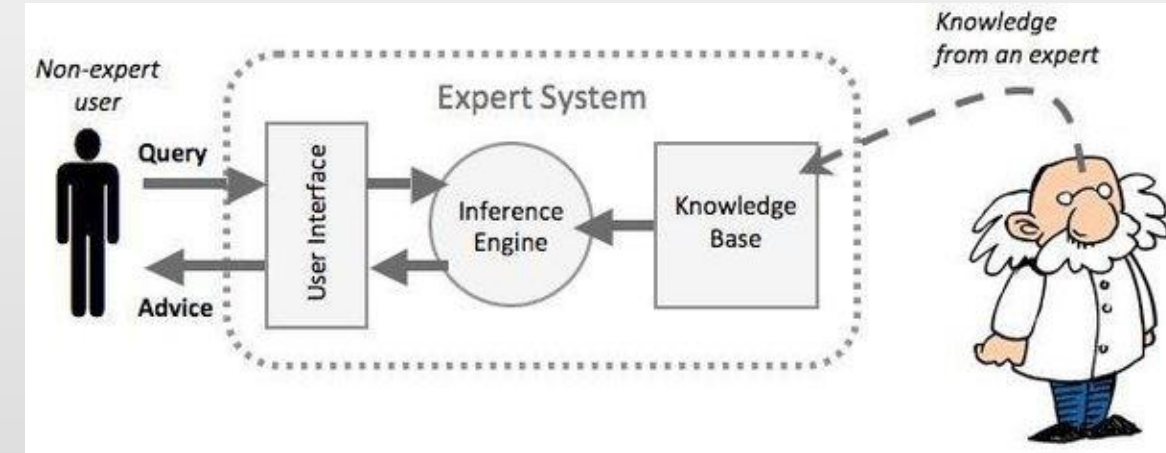
# History of AI

## 1970s and 1980s: Knowledge-Based Systems AI

*Research saw a shift towards knowledge-based systems and expert systems.*

*These systems utilized large knowledge bases to reason and draw conclusions about specific domains.*

*Though promising, early expert systems faced limitations due to the complexity of representing all human knowledge in a formalized manner.*

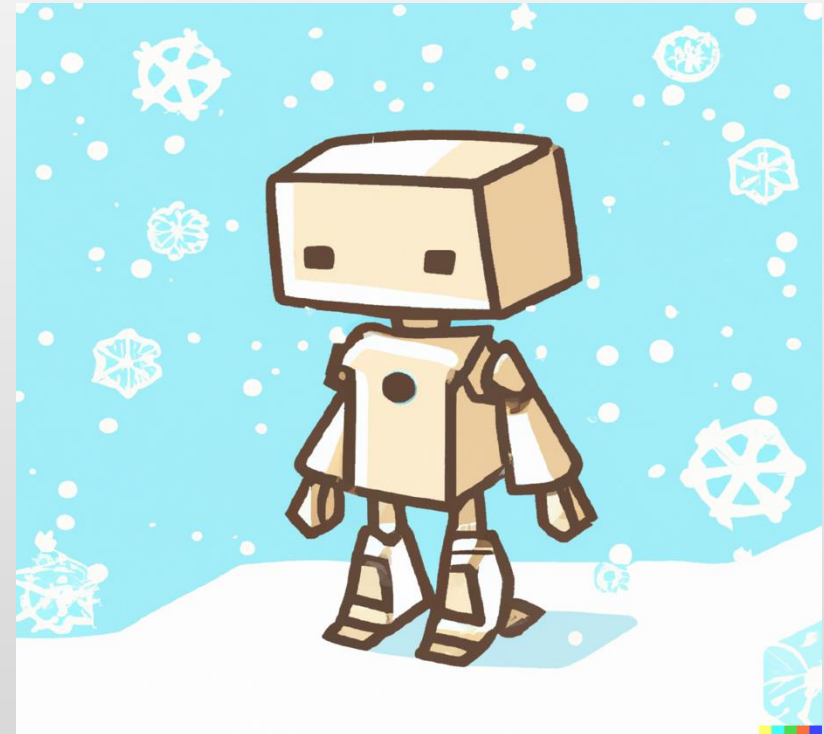


# History of AI

## 1980s and 1990s: AI Winter

*During this period, AI faced a decline in interest and funding due to unmet expectations and the inability to deliver on grand promises.*

*Progress in AI did not match initial optimism, leading to a phase known as the "AI Winter."*



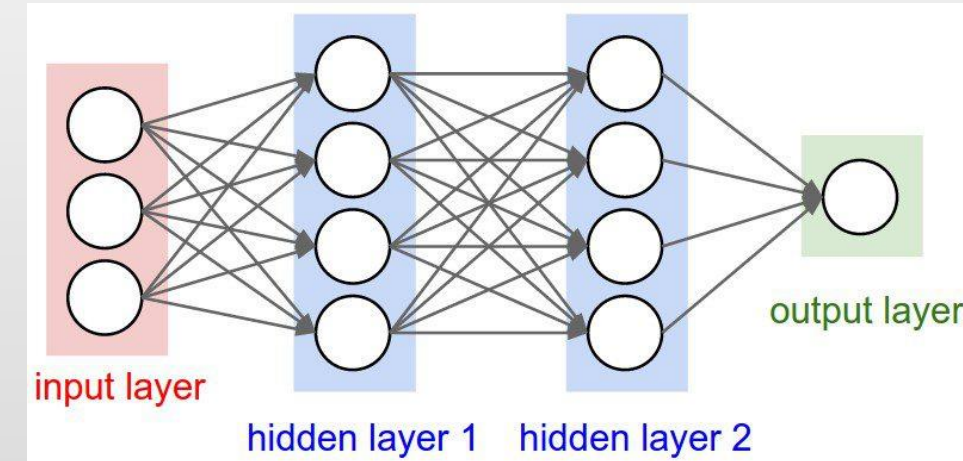
# History of AI

## Late 1990s and 2000s: Machine Learning and Neural Networks

*The resurgence of AI came with the advent of machine learning techniques, particularly neural networks.*

*Improved computational power and access to vast amounts of data allowed neural networks to excel in tasks like image recognition and language processing.*

*Support vector machines and other machine learning algorithms also gained popularity.*

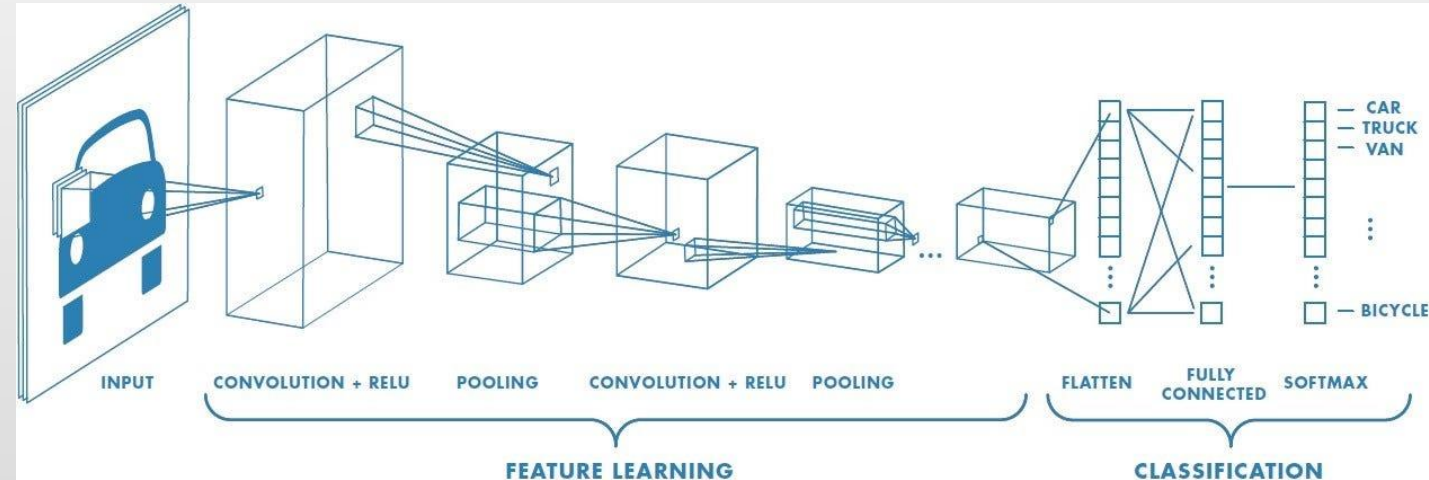


# History of AI

## 2010s: Deep Learning and AI Breakthroughs

*Deep learning, a subset of machine learning using artificial neural networks with multiple layers, became the driving force behind many AI breakthroughs.*

*Applications of AI proliferated in various fields, including natural language processing, computer vision, robotics, and autonomous vehicles.*





# History of AI

## Present and Beyond

*AI continues to evolve rapidly, with ongoing research in areas like reinforcement learning, explainable AI, Large Language Model and AI ethics.*

*Integration of AI in everyday life through virtual assistants, smart devices, and recommendation systems is becoming increasingly common.*

*However, concerns about AI's societal impact, privacy, and ethical implications persist.*



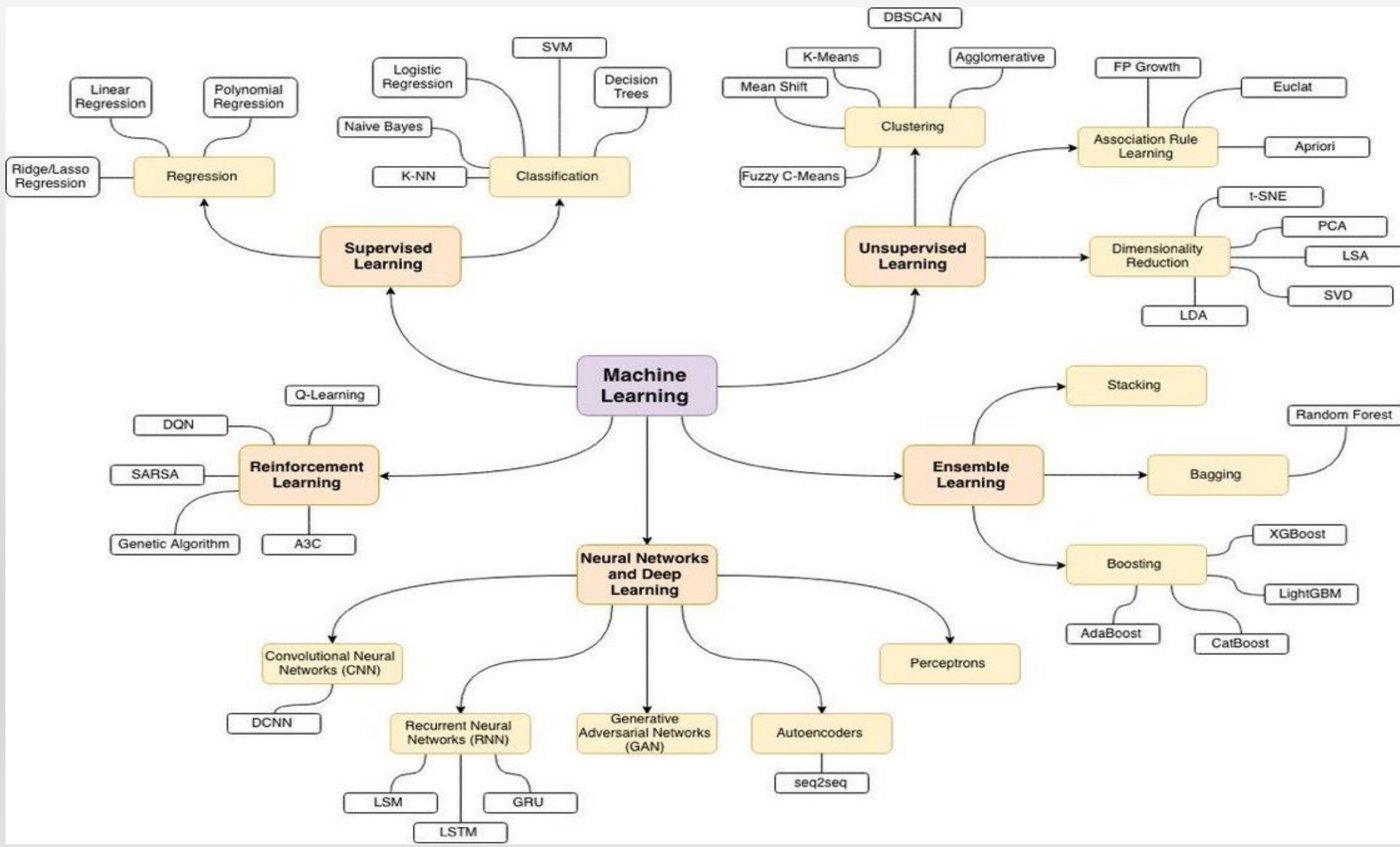
# Introduction of Machine Learning

Introduction

ML Perf

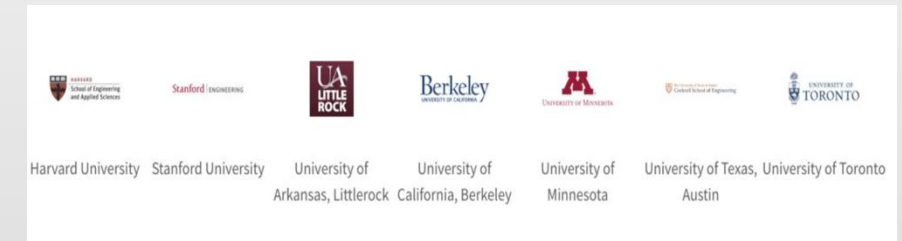
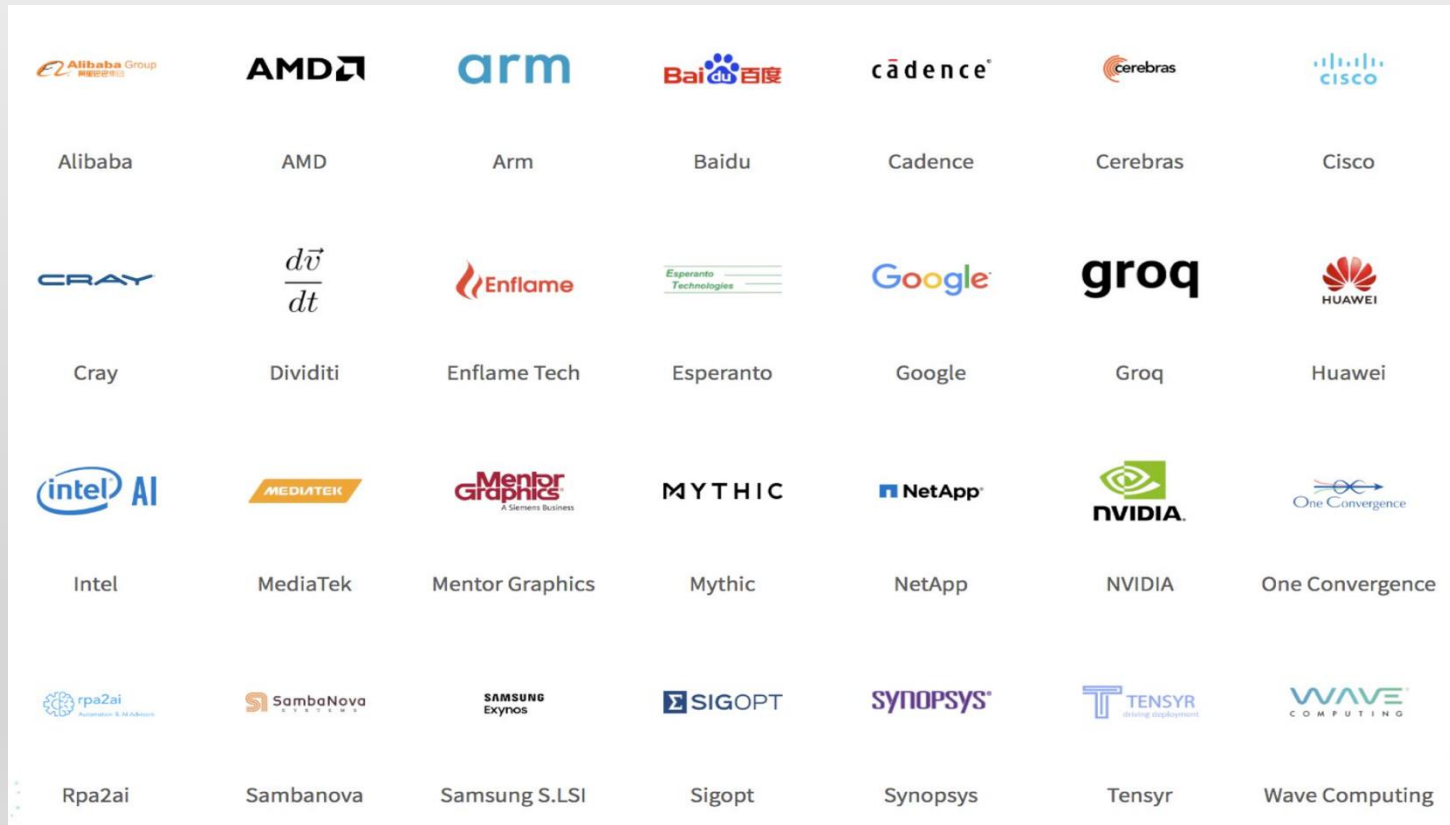
PyTorch

Running  
DL Models



# ML Perf Benchmark

ML Perf: A broad ML benchmark suite for measuring the performance of ML software frameworks, ML hardware accelerators, and ML cloud and edge platforms



# ML Perf Benchmark

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Area	Benchmark	Dataset	Quality Target	Reference Implementation Model	Latest Version Available
Vision	Image classification	ImageNet	75.90% classification	ResNet-50 v1.5	v3.1
Vision	Image segmentation (medical)	KITS19	0.908 Mean DICE score	3D U-Net	v3.1
Vision	Object detection (light weight)	Open Images	34.0% mAP	RetinaNet	v3.1
Vision	Object detection (heavy weight)	COCO	0.377 Box min AP and 0.339 Mask min AP	Mask R-CNN	v3.1
Language	Speech recognition	LibriSpeech	0.058 Word Error Rate	RNN-T	v3.1
Language	NLP	Wikipedia 2020/01/01	0.72 Mask-LM accuracy	BERT-large	v3.1
Language	LLM	C4	2.69 log perplexity	GPT3	v3.1
Commerce	Recommendation	Criteo 4TB multi-hot	0.8032 AUC	DLRM-dcnv2	v3.1
Marketing, Art, Gaming	Image Generation	LAION-400M-filtered	FID<=90 and CLIP>=0.15	Stable Diffusionv2	v3.1
Commerce	Recommendation	1TB Click Logs	0.8025 AUC	DLRM	v2.1
Research	Reinforcement learning	Go	50% win rate vs. checkpoint	Mini Go (based on Alpha Go paper)	v2.1
Vision	Object detection (light weight)	COCO	23.0% mAP	SSD	v1.1
Language	Translation (recurrent)	WMT English-German	24.0 Sacre BLEU	NMT	v0.7
Language	Translation (non-recurrent)	WMT English-German	25.00 BLEU	Transformer	v0.7

# ML Perf Benchmark

## ML Perf Training

*The MLPerf Training benchmark measures how fast systems can train models to a target quality metric.*

- MLPerf Training
- MLPerf Training: HPC



Edge: Lenovo SE450 Edge Server

## ML Perf Inference

*The MLPerf inference benchmark suite measures how fast systems can process inputs and produce results using a trained model*

- MLPerf Inference: Datacenter
- MLPerf Inference: Edge
- MLPerf Inference: Mobile
- MLPerf Inference: Tiny



Mobile: Xiaomi note12 turbo

## ML Perf Storage

*The MLPerf Storage benchmark suite measures how fast storage systems can supply training data when a model is being trained.*

- MLPerf Storage



Tiny: Cora Z7 for ARM/FPGA SoC Development

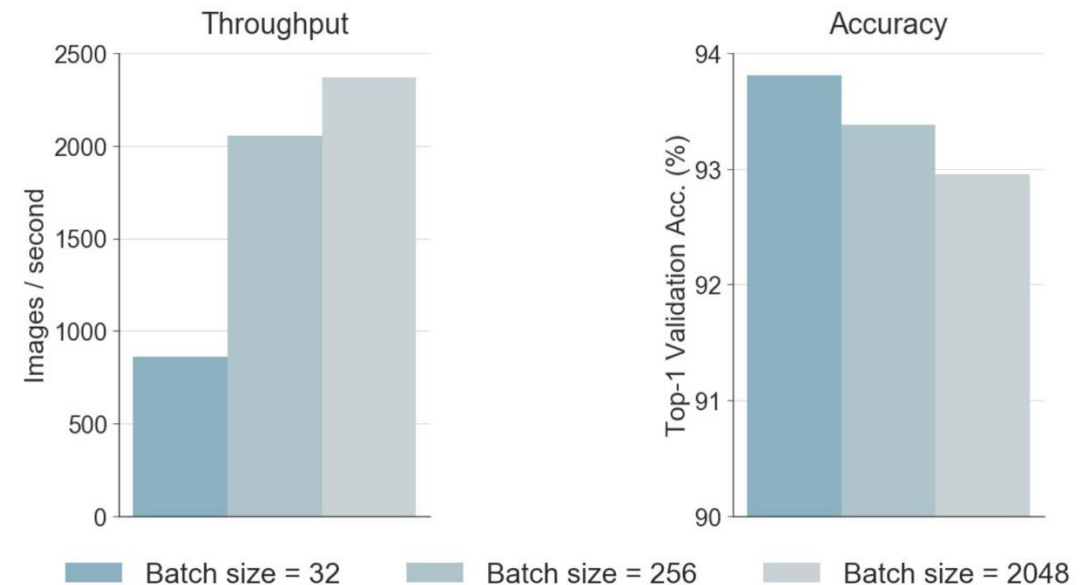
# ML Perf Training Benchmark

Performance

How fast is a model for training, inference?

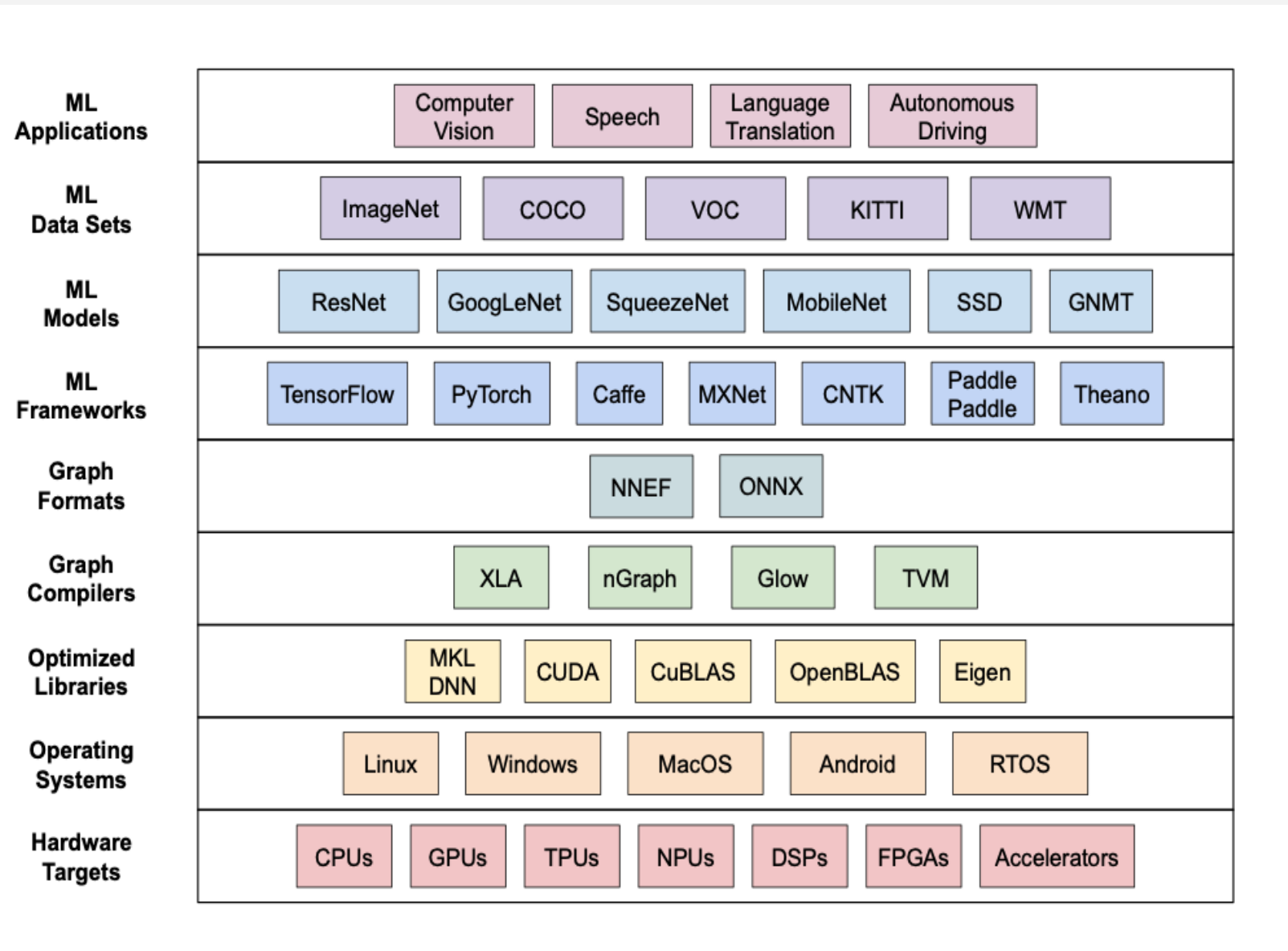
Quality

How good are a model's predictions?



- End to end training of a Resnet56 CIFAR10 model
- Nvidia P100 with 512GB of memory and 28 CPU cores
- TensorFlow 1.2 compiled from source with CUDA 8.0 and CuDNN 5.1

# ML Perf Benchmark





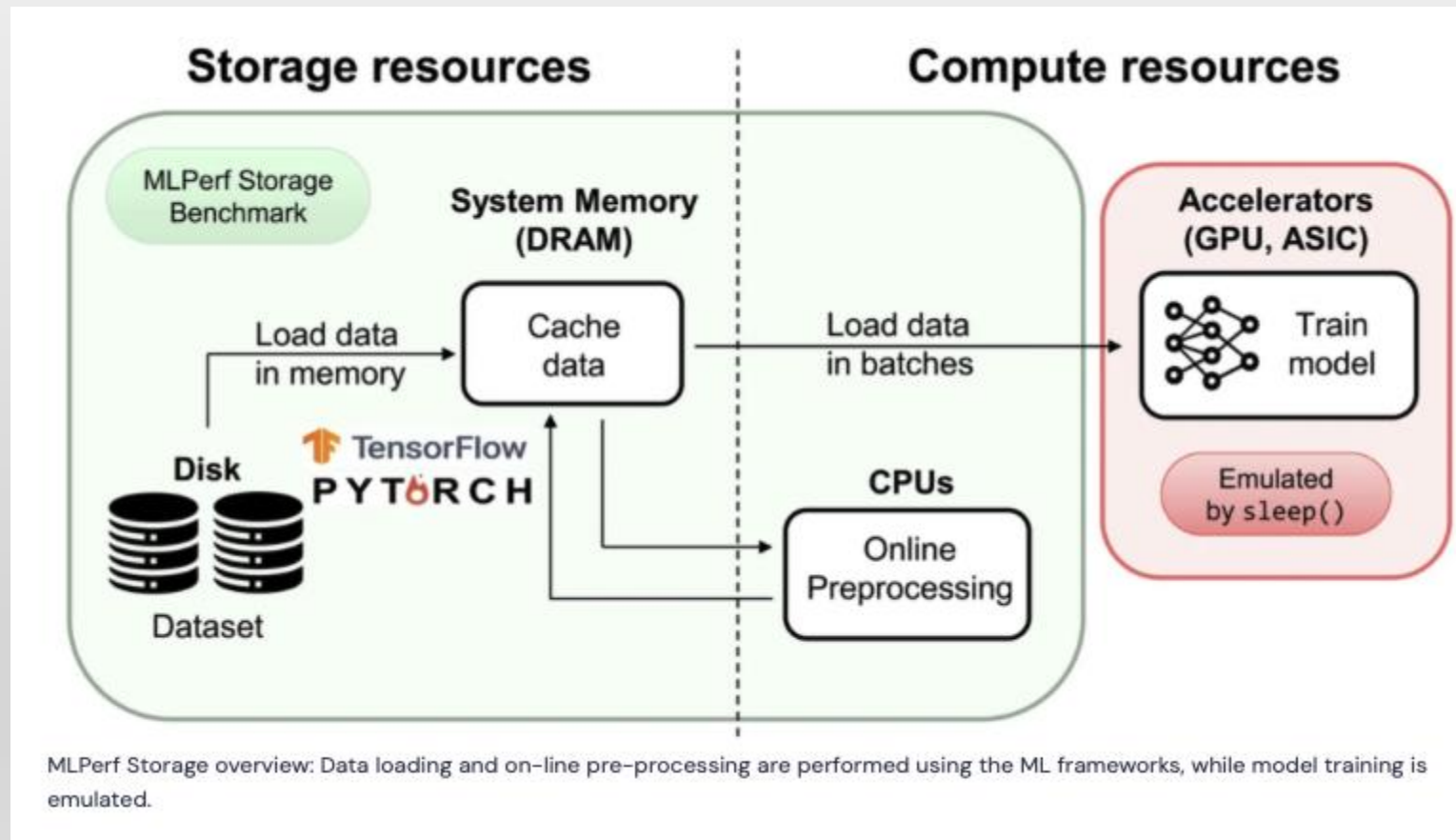
# ML Perf Benchmark

Training v1.1																	
Closed		Open															
ID	Submitter	System	Processor	#	Accelerator	#	Software	Benchmark results (minutes)								Details	Code
								Image classification	Image segmentation (medical)	Object detection, light-weight	Object detection, heavy-weight	Speech recognition	NLP	Recom-mendation	Reinforce-ment Learning		
								ImageNet	KITS19	COCO	COCO	LibriSpeech	Wikipedia	1TB Clickthrough	Go		
								ResNet	3D U-Net	SSD	Mask R-CNN	RNN-T	BERT [1]	DLRM	Minigo		
Available cloud																	
1.1-2000	Azure	nd96amsr_a100_v4_ngc21.09_merlin_hugectr	AMD EPYC 7V12	2	NVIDIA A100-SXM4-80GB (400W)	8	Merlin HugeCTR with NVIDIA DL Frameworks Release 21.09							1.875	<a href="#">details</a>	<a href="#">code</a>	
1.1-2001	Azure	nd96amsr_a100_v4_ngc21.09_mxnet	AMD EPYC 7V12	2	NVIDIA A100-SXM4-80GB (400W)	8	MXNet NVIDIA Release 21.09	29.720	25.400	8.309					<a href="#">details</a>	<a href="#">code</a>	
1.1-2002	Azure	nd96amsr_a100_v4_ngc21.09_pytorch	AMD EPYC 7V12	2	NVIDIA A100-SXM4-80GB (400W)	8	PyTorch NVIDIA Release 21.09				47.064	37.550	21.213		<a href="#">details</a>	<a href="#">code</a>	
1.1-2003	Azure	nd96amsr_a100_v4_ngc21.09_tensorflow	AMD EPYC 7V12	2	NVIDIA A100-SXM4-80GB (400W)	8	TensorFlow NVIDIA Release 21.09							319.410	<a href="#">details</a>	<a href="#">code</a>	
1.1-2004	Azure	nd96amsr_a100_v4_n4_ngc21.09_pytorch	AMD EPYC 7V12	8	NVIDIA A100-SXM4-80GB (400W)	32	PyTorch NVIDIA Release 21.09				14.912				<a href="#">details</a>	<a href="#">code</a>	
1.1-2005	Azure	nd96amsr_a100_v4_n8_ngc21.09_mxnet	AMD EPYC 7V12	16	NVIDIA A100-SXM4-80GB (400W)	64	MXNet NVIDIA Release 21.09	4.587		1.517					<a href="#">details</a>	<a href="#">code</a>	
1.1-2006	Azure	nd96amsr_a100_v4_n8_ngc21.09_pytorch	AMD EPYC 7V12	16	NVIDIA A100-SXM4-80GB (400W)	64	PyTorch NVIDIA Release 21.09						3.111		<a href="#">details</a>	<a href="#">code</a>	
1.1-2007	Azure	nd96amsr_a100_v4_n9_ngc21.09_mxnet	AMD EPYC 7V12	18	NVIDIA A100-SXM4-80GB (400W)	72	MXNet NVIDIA Release 21.09		3.800						<a href="#">details</a>	<a href="#">code</a>	
1.1-2008	Azure	nd96amsr_a100_v4_n16_ngc21.09_pytorch	AMD EPYC 7V12	32	NVIDIA A100-SXM4-80GB (400W)	128	PyTorch NVIDIA Release 21.09					4.533			<a href="#">details</a>	<a href="#">code</a>	
1.1-2009	Azure	nd96amsr_a100_v4_n32_ngc21.09_tensorflow	AMD EPYC 7V12	64	NVIDIA A100-SXM4-80GB (400W)	256	TensorFlow NVIDIA Release 21.09							30.714	<a href="#">details</a>	<a href="#">code</a>	
1.1-2010	Azure	nd96amsr_a100_v4_n34_ngc21.09_pytorch	AMD EPYC 7V12	68	NVIDIA A100-SXM4-80GB (400W)	272	PyTorch NVIDIA Release 21.09				3.908				<a href="#">details</a>	<a href="#">code</a>	
1.1-2011	Azure	nd96amsr_a100_v4_n48_ngc21.09_tensorflow	AMD EPYC 7V12	96	NVIDIA A100-SXM4-80GB (400W)	384	TensorFlow NVIDIA Release 21.09							24.802	<a href="#">details</a>	<a href="#">code</a>	
1.1-2012	Azure	nd96amsr_a100_v4_n96_ngc21.09_mxnet	AMD EPYC 7V12	192	NVIDIA A100-SXM4-80GB (400W)	768	MXNet NVIDIA Release 21.09		1.262						<a href="#">details</a>	<a href="#">code</a>	
1.1-2013	Azure	nd96amsr_a100_v4_n128_ngc21.09_mxnet	AMD EPYC 7V12	256	NVIDIA A100-SXM4-80GB (400W)	1024	MXNet NVIDIA Release 21.09	0.583		0.455					<a href="#">details</a>	<a href="#">code</a>	
1.1-2014	Azure	nd96amsr_a100_v4_n128_ngc21.09_pytorch	AMD EPYC 7V12	256	NVIDIA A100-SXM4-80GB (400W)	1024	PyTorch NVIDIA Release 21.09						0.656		<a href="#">details</a>	<a href="#">code</a>	
1.1-2015	Azure	nd96amsr_a100_v4_n192_ngc21.09_pytorch	AMD EPYC 7V12	384	NVIDIA A100-SXM4-80GB (400W)	1536	PyTorch NVIDIA Release 21.09					3.205			<a href="#">details</a>	<a href="#">code</a>	
1.1-2016	Azure	nd96amsr_a100_v4_n224_ngc21.09_tensorflow	AMD EPYC 7V12	448	NVIDIA A100-SXM4-80GB (400W)	1792	TensorFlow NVIDIA Release 21.09							17.439	<a href="#">details</a>	<a href="#">code</a>	
1.1-2017	Azure	nd96amsr_a100_v4_n256_ngc21.09_mxnet	AMD EPYC 7V12	512	NVIDIA A100-SXM4-80GB (400W)	2048	MXNet NVIDIA Release 21.09	0.438							<a href="#">details</a>	<a href="#">code</a>	
1.1-2018	Azure	nd96amsr_a100_v4_n256_ngc21.09_pytorch	AMD EPYC 7V12	512	NVIDIA A100-SXM4-80GB (400W)	2048	PyTorch NVIDIA Release 21.09					0.422			<a href="#">details</a>	<a href="#">code</a>	
Available on-premise																	
1.1-2019	Baidu	1_node_8_A100_NGC21.05_MXNet	Intel(R) Xeon(R) Platinum 8350C	2	NVIDIA A100-SXM4-80GB (400W)	8	MXNet NVIDIA Release 21.05	28.605							<a href="#">details</a>	<a href="#">code</a>	
1.1-2020	Baidu	1_node_8_A100_PaddlePaddle	Intel(R) Xeon(R) Platinum 8350C	2	NVIDIA A100-SXM4-80GB (400W)	8	PaddlePaddle (branch: develop, commitID: 605e7f0)	28.613							<a href="#">details</a>	<a href="#">code</a>	
1.1-2021	Dell	DSS8440x8A100-PCIE-40GB	Intel(R) Xeon(R) Gold 6248R CPU @ 3.00GHz	2	NVIDIA A100-PCIE-40GB (250W)	8	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	38.871		11.193	61.505		66.631		393.431	<a href="#">details</a>	<a href="#">code</a>
1.1-2022	Dell	DSS8440x8A100-PCIE-40GB-NVBridge	Intel(R) Xeon(R) Gold 6248R CPU @ 3.00GHz	2	NVIDIA A100-PCIE-40GB (250W)	8	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	37.083		10.899	58.571				405.424	<a href="#">details</a>	<a href="#">code</a>
1.1-2023	Dell	R750xax4A100-PCIE-80GB	Intel(R) Xeon(R) Gold 6338 CPU @ 2.00GHz	2	NVIDIA A100-PCIE-80GB (300W)	4	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	62.949	60.586	18.321	93.134	84.025	56.260		590.009	<a href="#">details</a>	<a href="#">code</a>
1.1-2024	Dell	R750xax4A100-PCIE-80GB-8368	Intel(R) Xeon(R) Platinum 8368 CPU @ 2.40GHz	2	NVIDIA A100-PCIE-80GB (300W)	4	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	64.131		18.390	91.562		56.131		<a href="#">details</a>	<a href="#">code</a>	
1.1-2025	Dell	2xR750xax4A100-PCIE-80GB	Intel(R) Xeon(R) Gold 6338 CPU @ 2.00GHz	4	NVIDIA A100-PCIE-80GB (300W)	8	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	32.087							<a href="#">details</a>	<a href="#">code</a>	
1.1-2026	Dell	DSS8440x8A100-PCIE-80GB	Intel(R) Xeon(R) Gold 6248R CPU @ 3.00GHz	2	NVIDIA A100-PCIE-80GB (300W)	8	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	33.553	28.543	9.835	55.033	60.763	41.269		<a href="#">details</a>	<a href="#">code</a>	
1.1-2027	Dell	DSS8440x10A100-PCIE-80GB	Intel(R) Xeon(R) Gold 6248R CPU @ 3.00GHz	2	NVIDIA A100-PCIE-80GB (300W)	10	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	28.187		8.223	46.948		36.859		<a href="#">details</a>	<a href="#">code</a>	
1.1-2028	Dell	4xR750xax4A100-PCIE-80GB	Intel(R) Xeon(R) Gold 6338 CPU @ 2.00GHz	8	NVIDIA A100-PCIE-80GB (300W)	16	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	17.336							<a href="#">details</a>	<a href="#">code</a>	
1.1-2029	Dell	8xR750xax4A100-PCIE-80GB	Intel(R) Xeon(R) Gold 6338 CPU @ 2.00GHz	16	NVIDIA A100-PCIE-80GB (300W)	32	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	10.586		3.477					<a href="#">details</a>	<a href="#">code</a>	
1.1-2030	Dell	XE8545x4A100-SXM-40GB	AMD EPYC 7763 64-Core Processor	2	NVIDIA A100-SXM4-40GB (400W)	4	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	61.820		16.998	95.157	79.563			<a href="#">details</a>	<a href="#">code</a>	
1.1-2031	Dell	XE8545x4A100-SXM-80GB	AMD EPYC 7713 64-Core Processor	2	NVIDIA A100-SXM4-80GB (500W)	4	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	56.326	55.999	16.244	83.774	106.542	38.855	9.522	451.293	<a href="#">details</a>	<a href="#">code</a>
1.1-2032	Dell	2xE8545x4A100-SXM-80GB	AMD EPYC 7713 64-Core Processor	4	NVIDIA A100-SXM4-80GB (500W)	8	NGC MXNet 21.09 , NGC PyTorch 21.09 , NGC TensorFlow 21.09-If1	30.123		8.735	48.788	35.068	26.547		<a href="#">details</a>	<a href="#">code</a>	
1.1-2033	Fujitsu	PRIMERGY-GX2460M1-mxnet	AMD EPYC 7502 32-Core Processor	2	NVIDIA A100-PCIE-40GB (250W)	4	MXNet NGC21.09	70.294	49.946	20.916					<a href="#">details</a>	<a href="#">code</a>	
1.1-2034	Fujitsu	PRIMERGY-GX2460M1-pytorch	AMD EPYC 7502 32-Core Processor	2	NVIDIA A100-PCIE-40GB (250W)	4	Pytorch NGC21.09					109.216	127.843		<a href="#">details</a>	<a href="#">code</a>	



# ML Perf Storage Benchmark

The MLPerf Storage benchmark suite measures how fast storage systems can supply training data when a model is being trained.

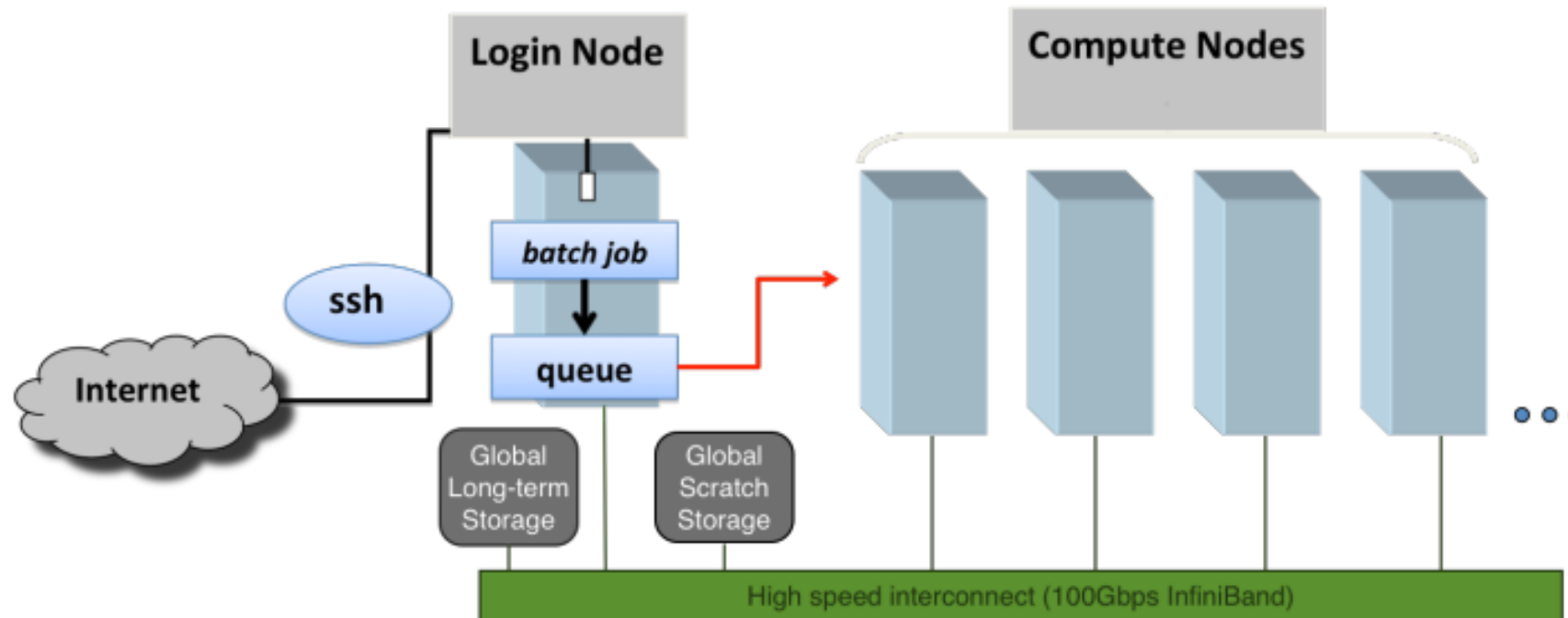


## An example of Training

The problem we're going to solve today is to train a model to classify MNIST datasets. The MNIST database of handwritten digits has a training set of 60,000 examples, and a test set of 10,000 examples.



# CARC Cluster



## Slurm script for job submission

```
git clone https://github.com/uschpc/Running-DL-Applications.git
```

## Common types of GPU

Partition	CPU model	CPU frequency	CPUs/node	GPU model	GPUs/node	Memory/node	Nodes
gpu	xeon-6130	2.10 GHz	32	V100	2	184 GB	29
gpu	xeon-2640v4	2.40 GHz	20	P100	2	123 GB	38
gpu	epyc-7282	2.80 GHz	32	A40	2	248 GB	12
gpu	epyc-7513	2.60 GHz	64	A100	2	248 GB	12

# Common types of GPU

## GPU specifications in the GPU partition

There are four kinds of GPUs in the GPU partition: A100, A40, V100, P100. The following is a summary table for the GPU specifications:

GPU model	Architecture	Memory	Memory Bandwidth	Base Clock Speed	Cuda Cores	Tensor Cores	Single Precision Performance (FP32)	Double Precision Performance (FP64)
A100	ampere	40GB	1.6TB/s	765MHz	6912	432	19.5TFLOPS	9.7TFLOPS
A40	ampere	48GB	696GB/s	1305MHz	10752	336	37.4TFLOPS	584.6GFLOPS
V100	volta	32GB	900GB/s	1230MHz	5120	640	14TFLOPS	7TFLOPS
P100	pascal	16GB	732GB/s	1189MHz	3584	n/a	9.3TFLOPS	4.7TFLOPS