



How to build a GPU cluster from scratch for your ML team

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Lambda

Who am I?

I'm the co-founder and CEO of **Lambda**. We've built large scale GPU clusters for the Fortune 500, the world's leading academic research institutions, and the DOD.

I'm also the lead architect of the **Lambda Echelon**, a turn-key GPU cluster. This talk is based on the Lambda Echelon reference design whitepaper and the experience we've gained deploying these large clusters.

Overview

- Introduction
- Cluster use case overview
- Cluster design
 - Compute
 - Storage
 - Cluster Networking
 - Data center floor planning & power distribution checklist
 - Software
- Rack design
 - Rack elevations
 - Power basics
 - Rack power distribution
- Node design
 - GPU selection
 - Node PCIe topology / NVLink topology / real life examples
 - GPUDirect RDMA
- Putting it all together
- Citations

5 Stages of GPU Cloud Grief

It all starts with the **shock** of an expensive cloud bill.

Greetings from Public Cloud Provider,

This e-mail confirms that your latest billing statement, for the account ending in 1234, is available on the Cloud web site. Your account will be charged the following:

Total: \$18,445.92

You can see a complete break down of all charges on the Billing & Cost Management page.

Thank you for using the public cloud.

Sincerely,
Public Cloud Provider

Stage 1 - Denial

“This won’t happen again next month.”

Stage 2 - Anger

“The bill doubled again!”

Stage 3 - Bargaining with your account manager.

Hi Team,

Are you available in the next few days for a 15 min call to discuss lowering your public cloud spend using reserved instances?

Based on your January usage, purchasing a 1-year reserved instance would save you up to 20% on your public cloud bill!

Thank you for being a loyal public cloud customer.

Cheers,
Public Cloud Provider

Stage 4 - Depression

“Spot instances and reserved instances aren’t enough, this is hopeless.”

Stage 5 - Acceptance

“GPU cloud services are expensive. Managing hardware is scary.”

Solution

Learn how to build a GPU cluster.

Good reasons to consider building an on-prem cluster



Big data sets &
expensive egress



Data sovereignty
& security



More compute for
less money

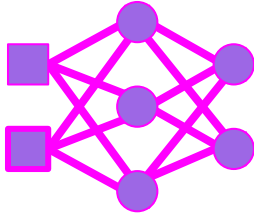
Rule Number 1: Know your ML team.

Your use case should inform the cluster design

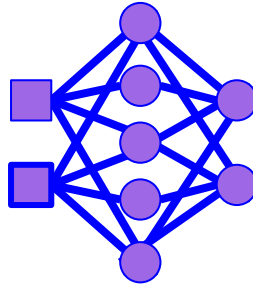
- **Hyperparameter search**
 - “Finding the best model.”
- **Large scale distributed training**
 - “Quickly training a model.”
- **Production inference**
 - “Deploying the model at scale to production.”

Hyperparameter search

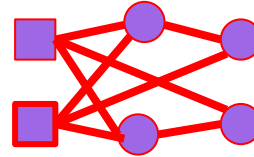
Which neural network performs best (highest accuracy)?



A



B

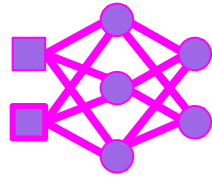


C

Hyperparameter Search

Answer: _____ ? _____

Hyperparameter search scoring

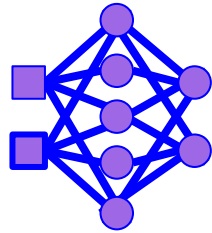


=>



=>

37% Accuracy

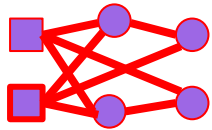


=>



=>

91% Accuracy



=>



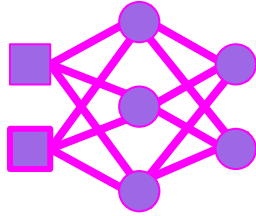
=>

30% Accuracy

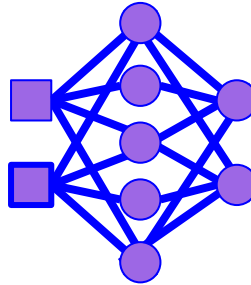
Hyperparameter Search

Hyperparameter search

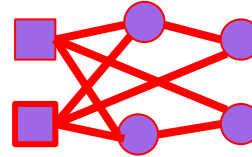
Give a score to each, select the highest score.



A
Score: 37%



B
Score: 91%

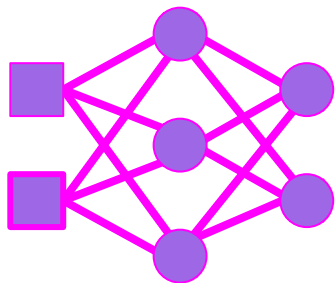


C
Score: 30%

Answer: ____ **B** ____

Hyperparameter Search

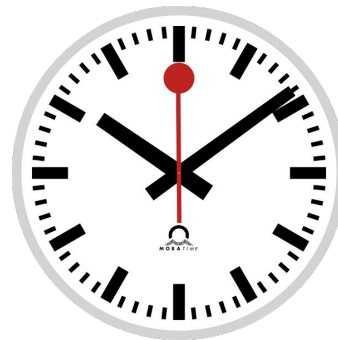
Each training run can take 1 to 48+ hours!



=>



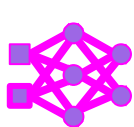
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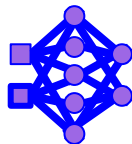
Hyperparameter Search

Most people start small

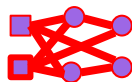
Which neural network performs best?



A



B



C

=>

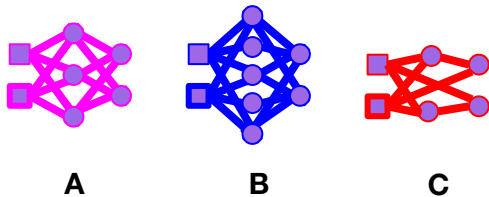
Answer: ____ ? ____



Hyperparameter Search

Then scale to faster GPUs

Which neural network performs best?



=>

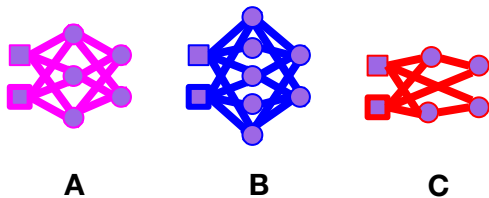
Answer: ____ ? ____



Hyperparameter Search

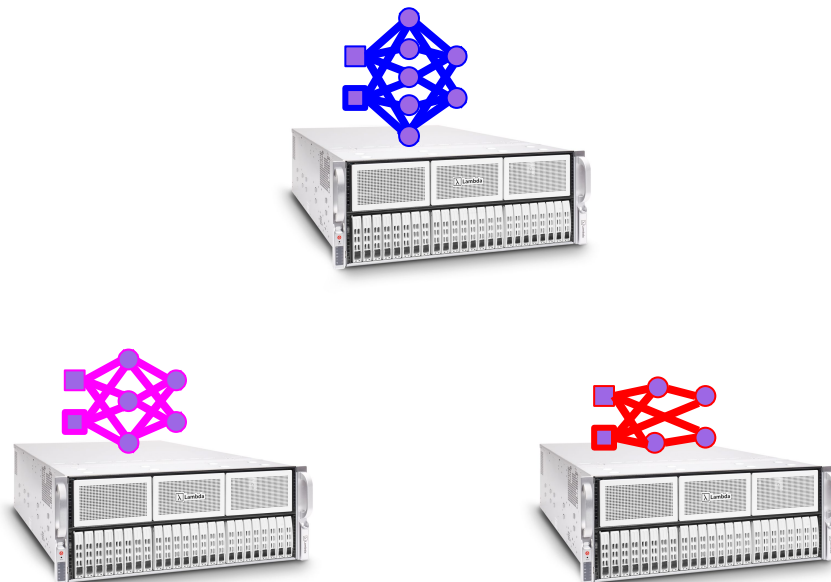
Then scale to multiple servers to run the jobs simultaneously

Which neural network performs best?



Answer: ____ ? ____

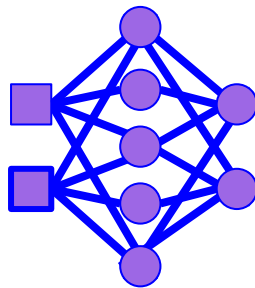
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Hyperparameter Search

Large scale distributed training

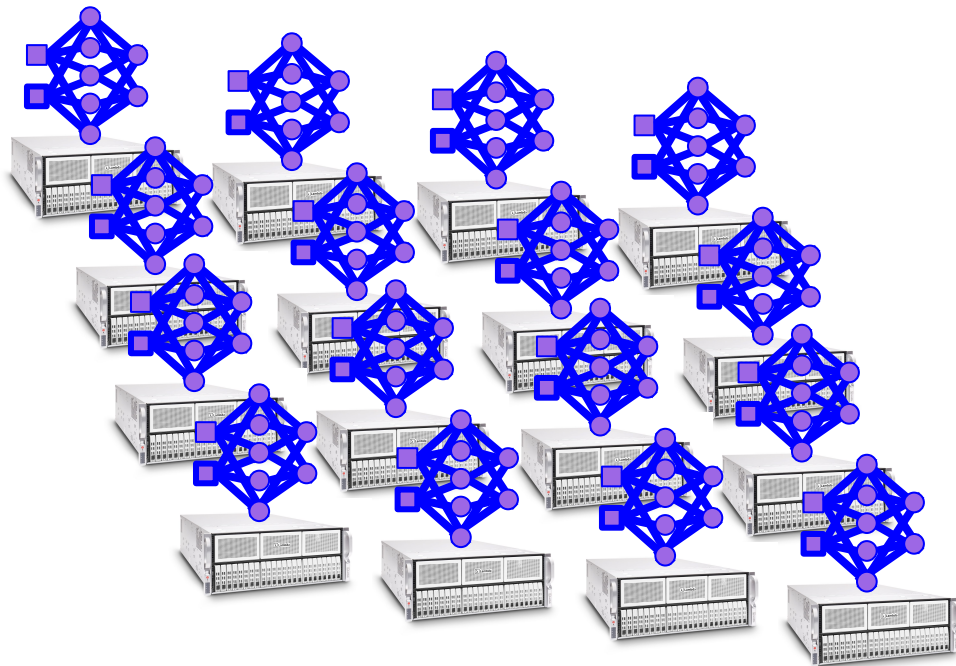
Train this model as fast as possible



B

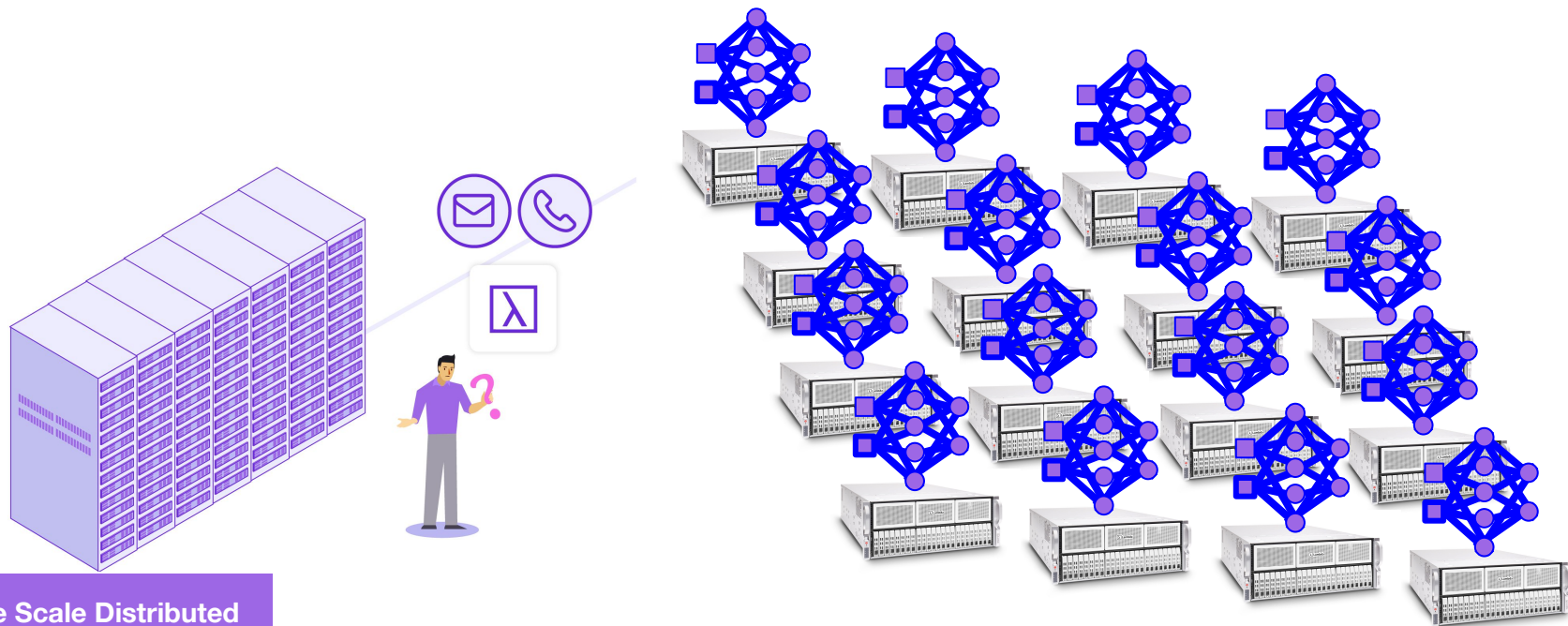
Large Scale Distributed
Training

Have multiple servers help with the training



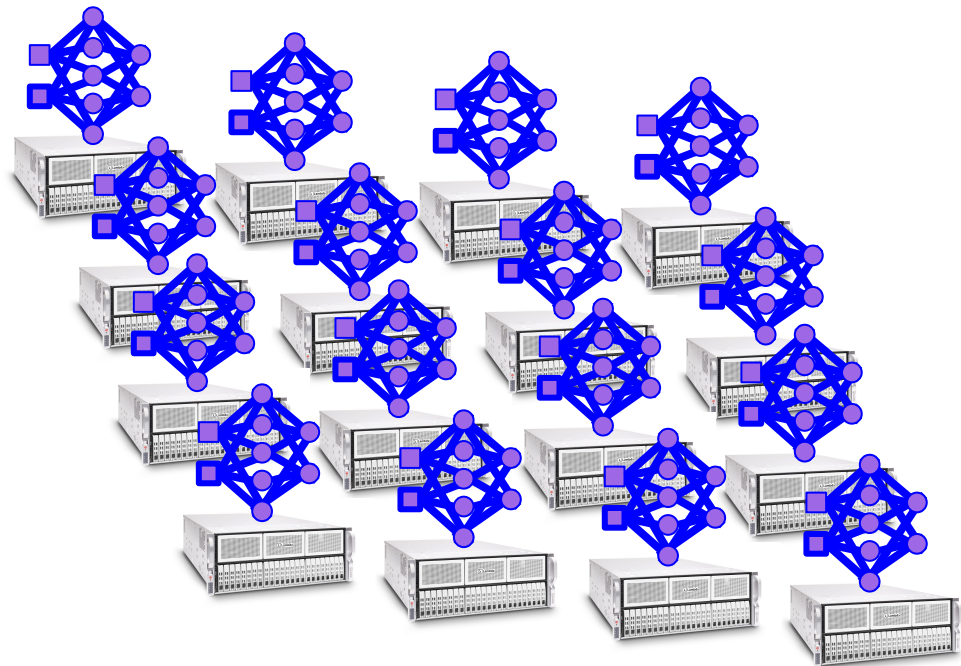
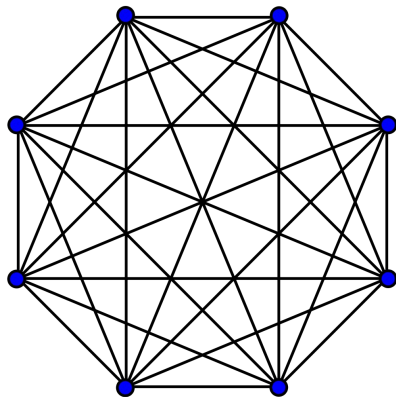
Large Scale Distributed
Training

Coordinating the servers is hard



Large Scale Distributed
Training

Lots of server-to-server communication = high speed network required

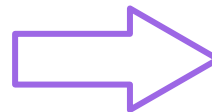
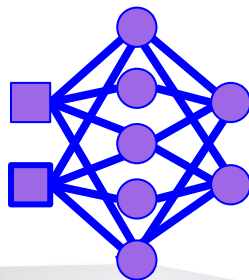
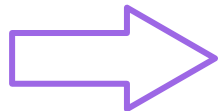


Large Scale Distributed
Training

Production inference



Hey Lambda, translate
"Hello" to Chinese

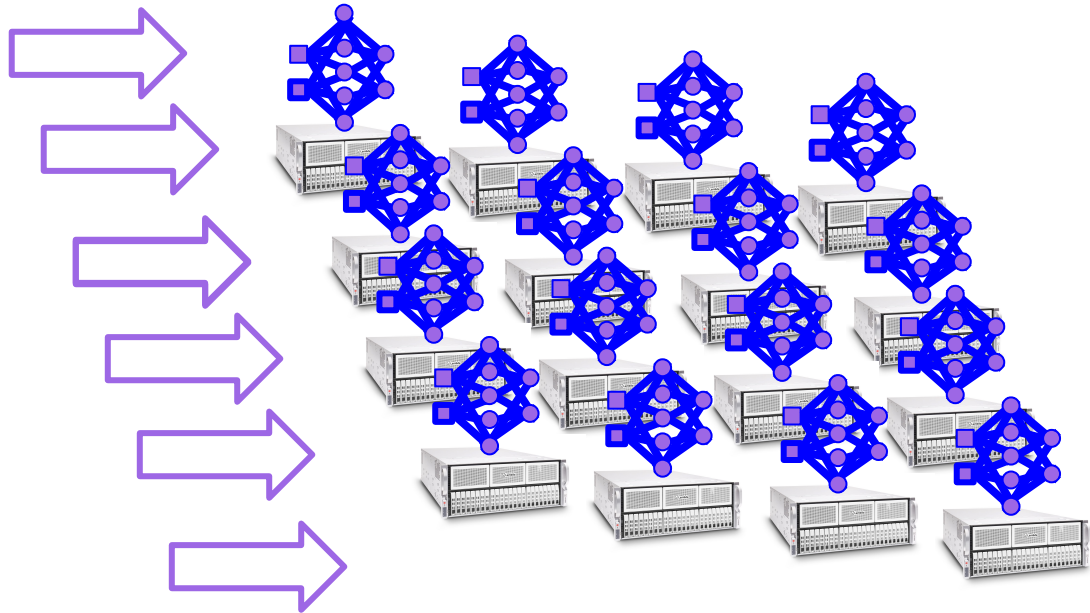
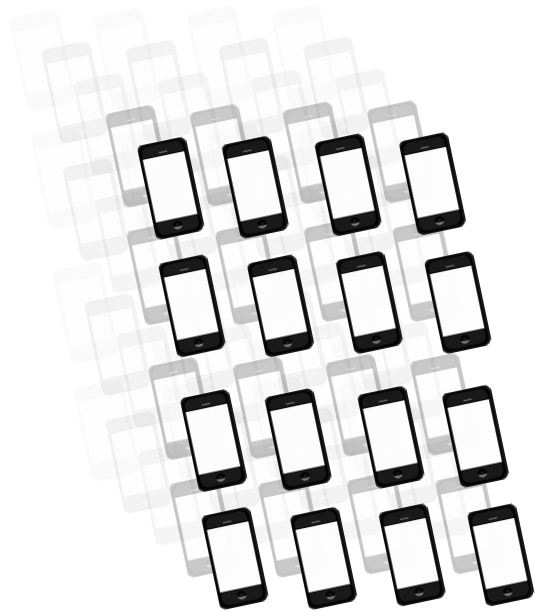


你好 - nǐ hǎo



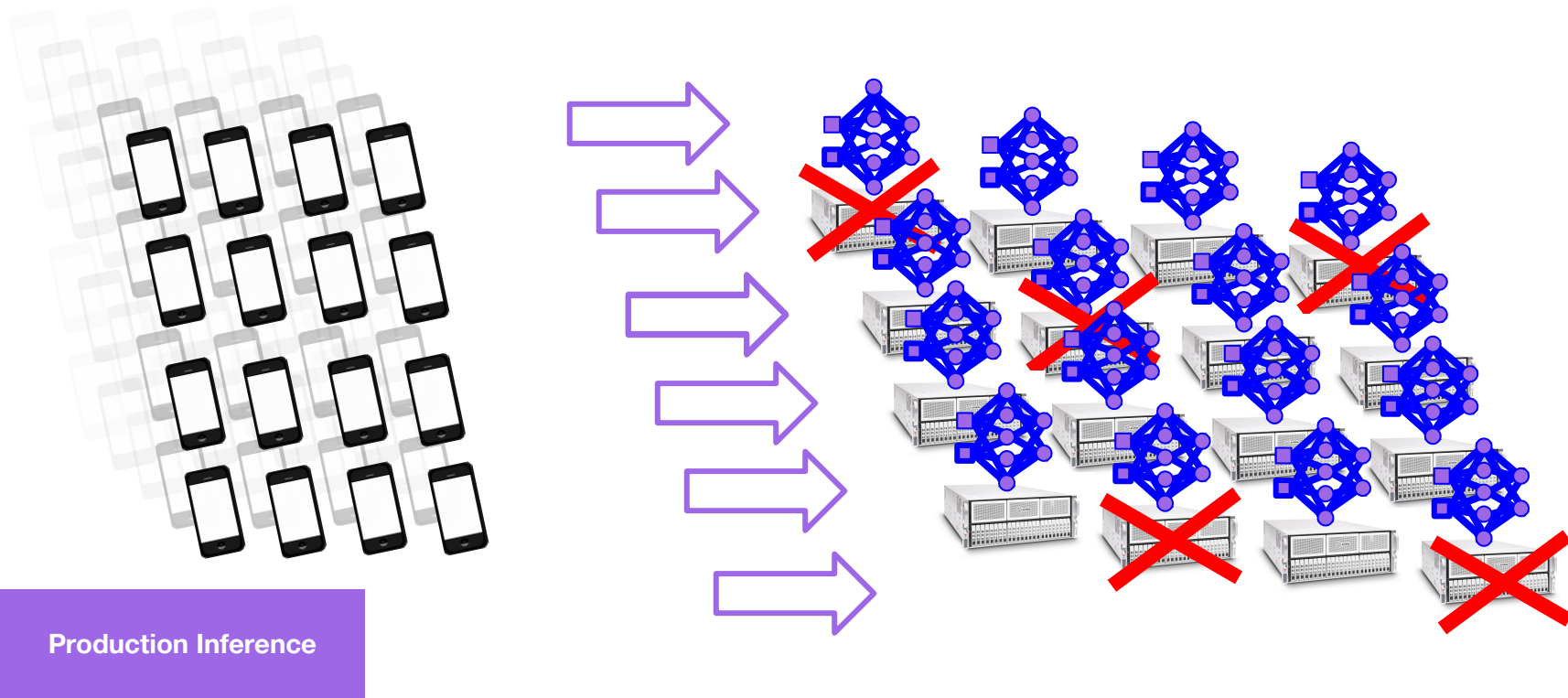
Production Inference

Inference clusters handle thousands of simultaneous requests



Production Inference

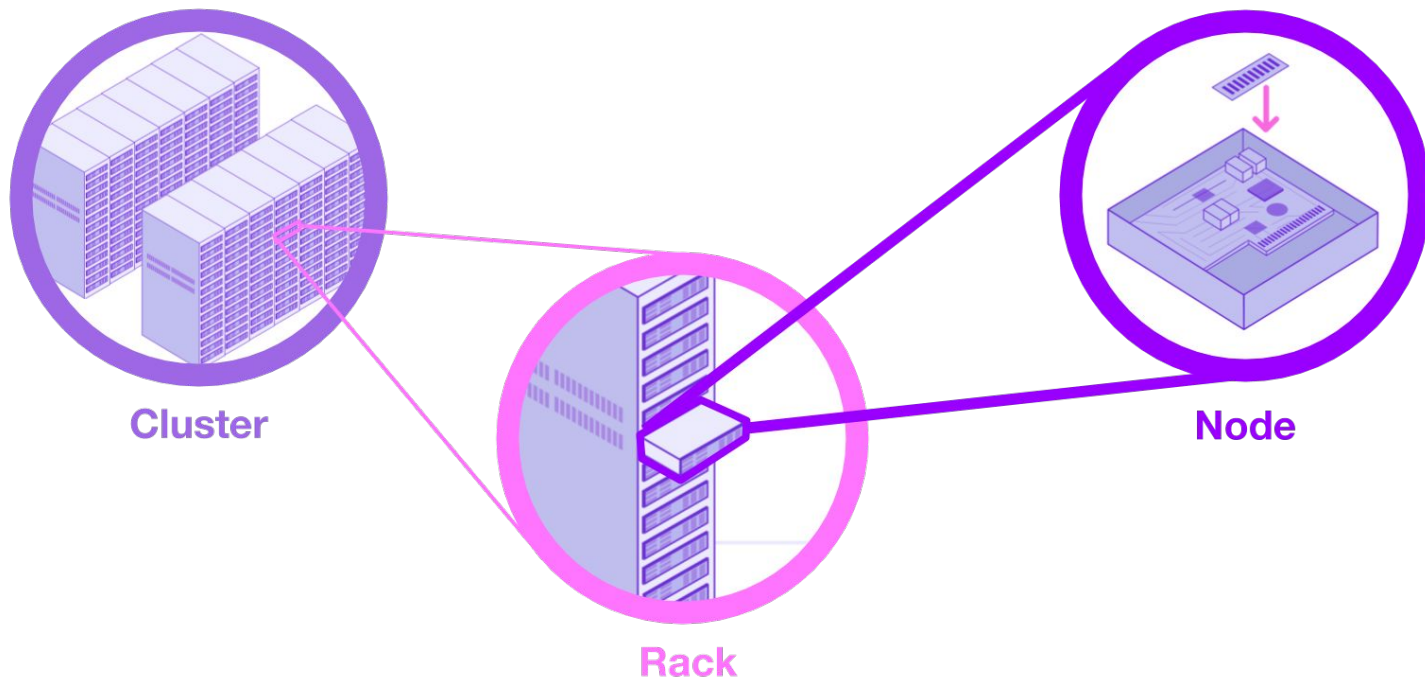
Inference clusters must be robust to outages



What's the right cluster for each use case?

	Hyperparameter search	Large scale distributed training
Node to Node Bandwidth	Low	High
Key Metric	Training throughput / \$	Time to train a single large model
Price	\$\$	\$\$\$\$
1+1 Redundancy	Optional	Optional
Operating Mode	Usually offline, job-queue based	Usually offline, job-queue based
	Production inference	
Node to Node Bandwidth	Low	
Key Metric	High availability & throughput	
Price	\$\$\$\$\$	
1+1 Redundancy	Critical	
Operating Mode	Often online, real time results	

Three levels of abstraction



The three levels of cluster design

1. **Cluster design:** the highest level of abstraction. Entire racks are just dots on the screen. Data center floor plans, capacity planning (power, network, storage, compute), and network topologies are the main output products at this level.
2. **Rack design:** Rack elevations describe the layout and exact position of individual nodes in a particular rack. Cable and port counts are important details at this layer.
3. **Node design:** Individual node bills of materials and cluster design goals drive component selection for each node.

Software: it's important to remember that there is a two way design dependency between the cluster hardware and the software that runs on it. Selecting the right software depends on the use case and scale of your planned deployment.

Work products for each layer

Abstraction Layer	Work Products	Goals
Cluster	• Cluster BOM (Bill of Materials) • Data center floor plan & checklist • Capacity plan • Network topology • Installation plan	• Working within constraints of data center floor space and power availability • Connecting all of the racks together • Determining the overall storage/compute/network capacity
Rack	• Rack BOM • Rack elevation	• Properly distributing the capacities planned for in the cluster stage • Rack level power density & capacity • Node placement within rack
Node	• Node BOM	• Hitting the performance requirements defined by the compute capacity of the cluster
Software	• Software architecture • Software installation plan	• Tying all of the hardware together so that it's easy to use and administer

Cluster design

Cluster architecture

Cluster architectures have five components:

1. **Compute:** to do the actual work. GPU & CPU nodes.
2. **Storage:** to serve data sets and store trained models / checkpoints.
3. **Networking:** multiple networks for compute, storage, in-band management, out-of-band management.
4. **Data center power distribution & floor planning:** understanding the electrical and physical layout of the deployment data center informs the rack elevations, cable lengths, and networking.
5. **Software:** cluster orchestration, job scheduling, resource allocation, container orchestration, and node level software stacks.

Compute

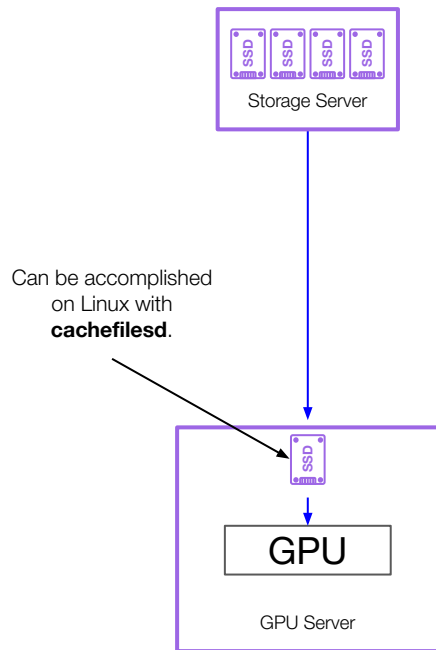
You need to answer three questions:

1. How much compute do you need?
2. How fast do you want the compute to talk to each other?
3. How fast do you need the compute to be able to talk to the storage?

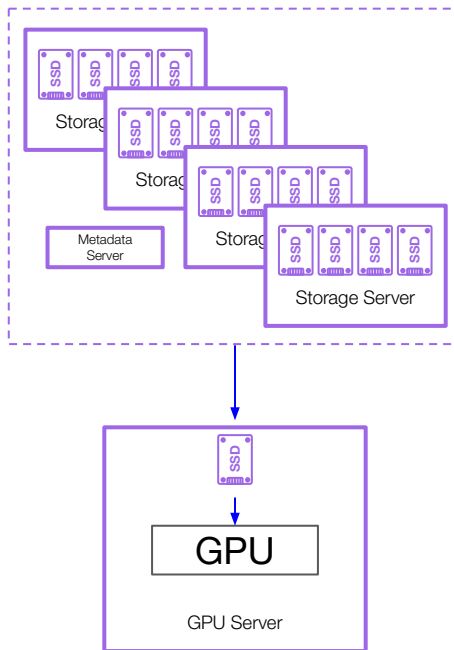
Storage

- Often, storage clusters will become the bottleneck in a highly optimized cluster. It's important to remove this bottleneck because it reduces the utilization of the most expensive part of the cluster, the compute.
- There are a two ways to design a storage cluster:
 - Work with a storage partner
 - Roll your own

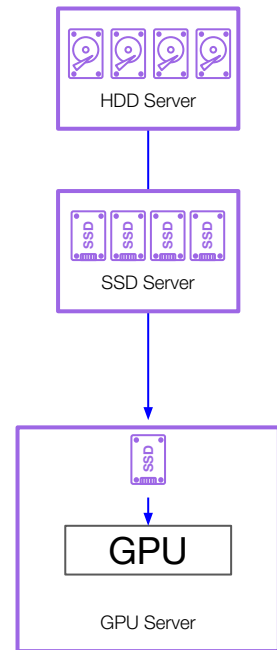
Storage architecture diagrams



**Single Tier NFS
+ local cache**

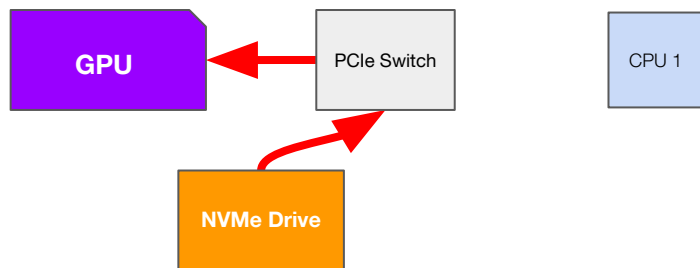


**Parallel cluster file system
+ local cache**

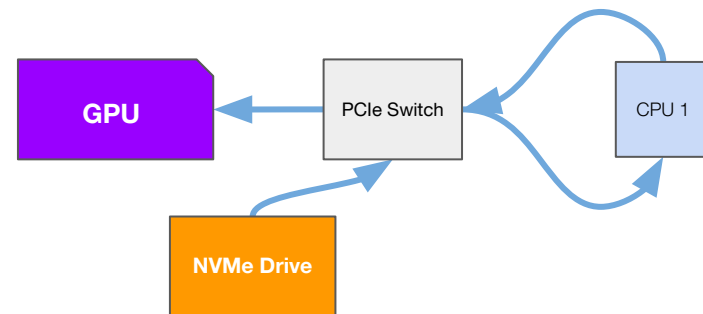


**Tiered storage system
+ local cache**

GPUDirect Storage



Data pathway with GPUDirect Storage
(Directly to the GPU via PCIe switch)



Data pathway without GPUDirect Storage
(Additional copy to CPU memory)

Storage choices

Open source storage options



Proprietary storage options



Cluster networking

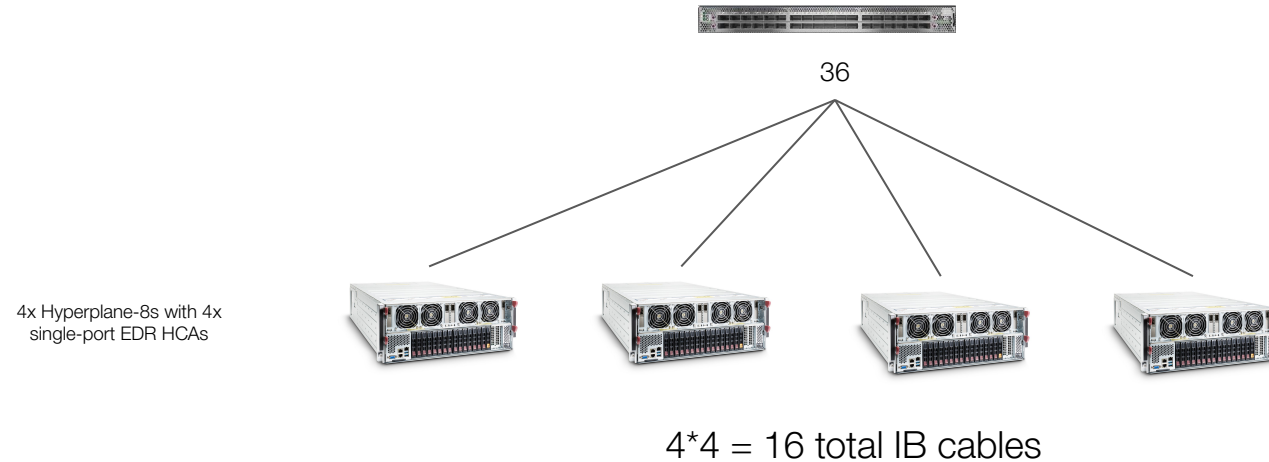
Networking basics: switches

MSB7800 - 36 port EDR (100Gb/s) IB switch - 7.2Tb/s aggregate switch throughput
 $7.2 \text{ Tb/s} / 36 \text{ ports} = 7200 \text{ Gb/s} / 36 \text{ ports} = \mathbf{200 \text{ Gb/s / port} \Rightarrow (100 \text{ Gb/s EDR IB)}$



Networking basics: single switch topology

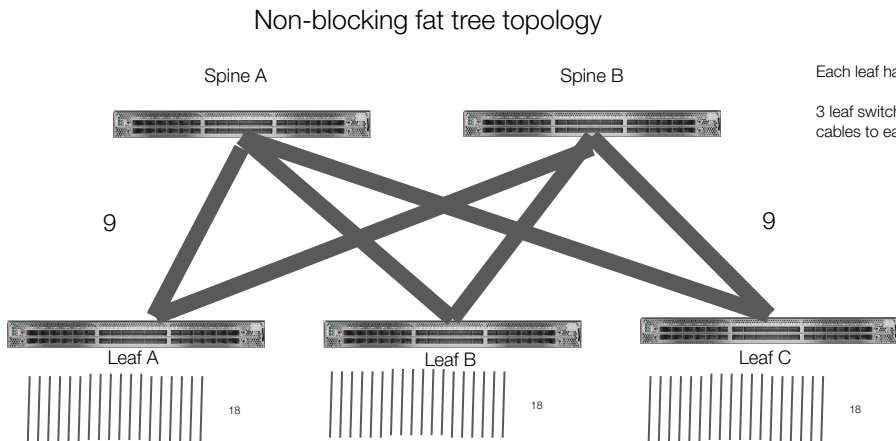
Non-blocking 100% port to port speed



Networking basics: spine and leaf topology

To guarantee non blocking, ensure equal number of “southbound” and “northbound” ports on the leaf switches

$36/2 = 18$
southbound and 18
northbound



Each leaf has 9 cables going northbound, 3 leaf switches

3 leaf switches * 9 northbound cables = 27 northbound leaf switch cables to each spine switch.

48 needed cables / 18
southbound = 2.666, round up
to 3 => 3 leaf switches to
provide full 48 ports.

54 total ports - we're using
48/54.

12x Hyperplane-8s with 4x
single-port EDR HCAs



$4 * 12 = 48$ total IB cables

Total southbound
bandwidth of
5.4Tb/s available.

100% port-to-port bandwidth spine & leaf topology

Legend



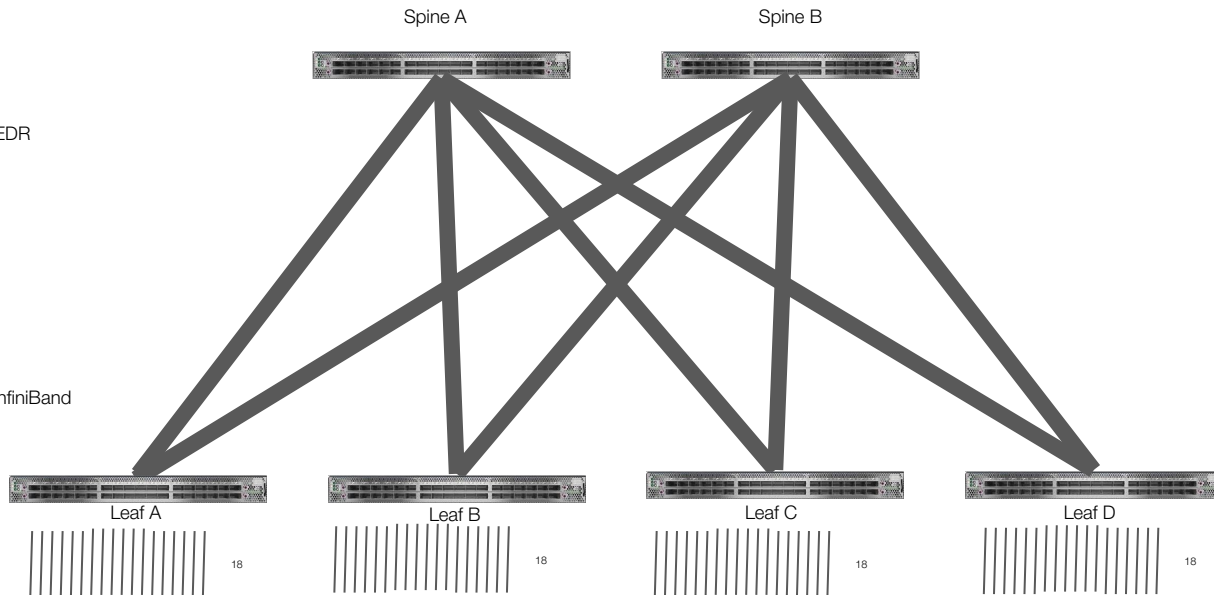
A 36-port MSB7800 100Gb/s IB EDR Switch



Each thick grey line represents 9 InfiniBand cables.



Each thin black line represents 1 InfiniBand cable.



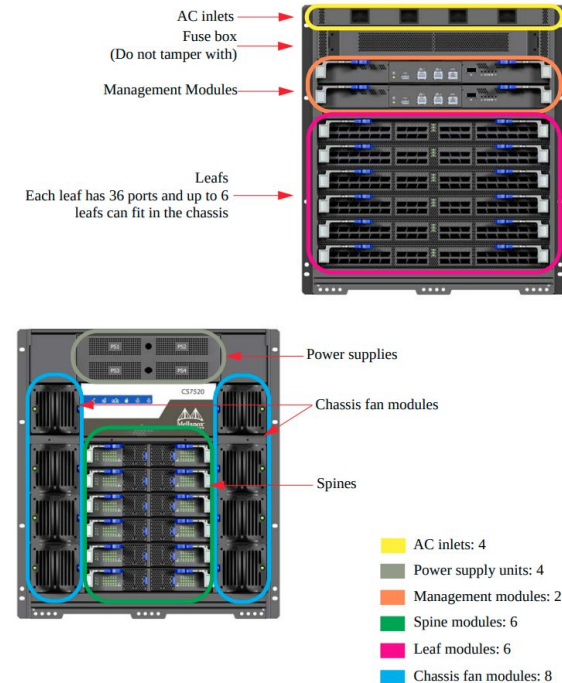
Networking basics: director switches

- Use a backplane instead of cables to build out the internal (spine <=> leaf) network topology.
- To get 100% non-blocking bandwidth in the CS7520 you NEED to have all 6 spines installed.
- CS7520 has 216 southbound ports (and thus 216 northbound ports).
- Assuming each spine has only 36 ports, how many spines do you need to support 216 ports?
 - 36 ports / spine, 216 ports needed
 - 216 ports / 36 (ports / spine) = 6 spine switches

For more info see:

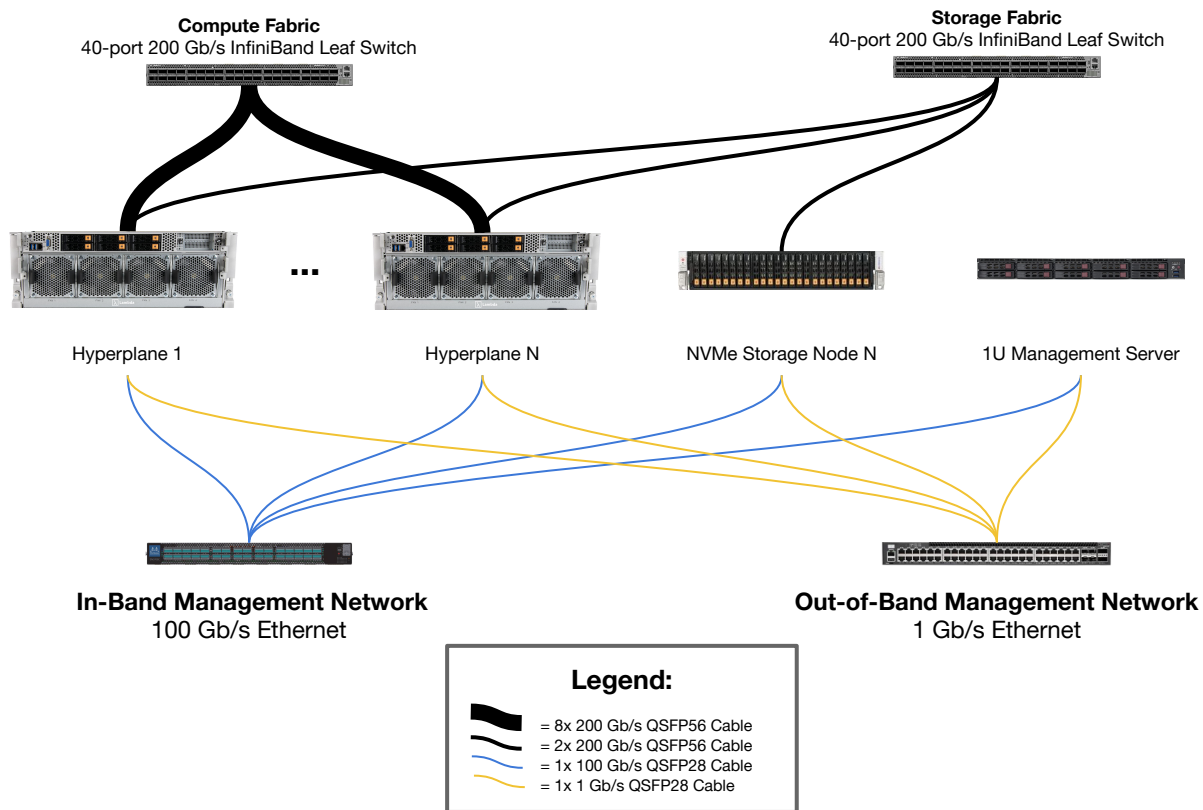
https://www.mellanox.com/related-docs/prod_ib_switch_systems/CS7520_Dismantling_Guide.pdf

Figure 1: Front and Rear View of the CS7520



Lambda Echelon Network Topology

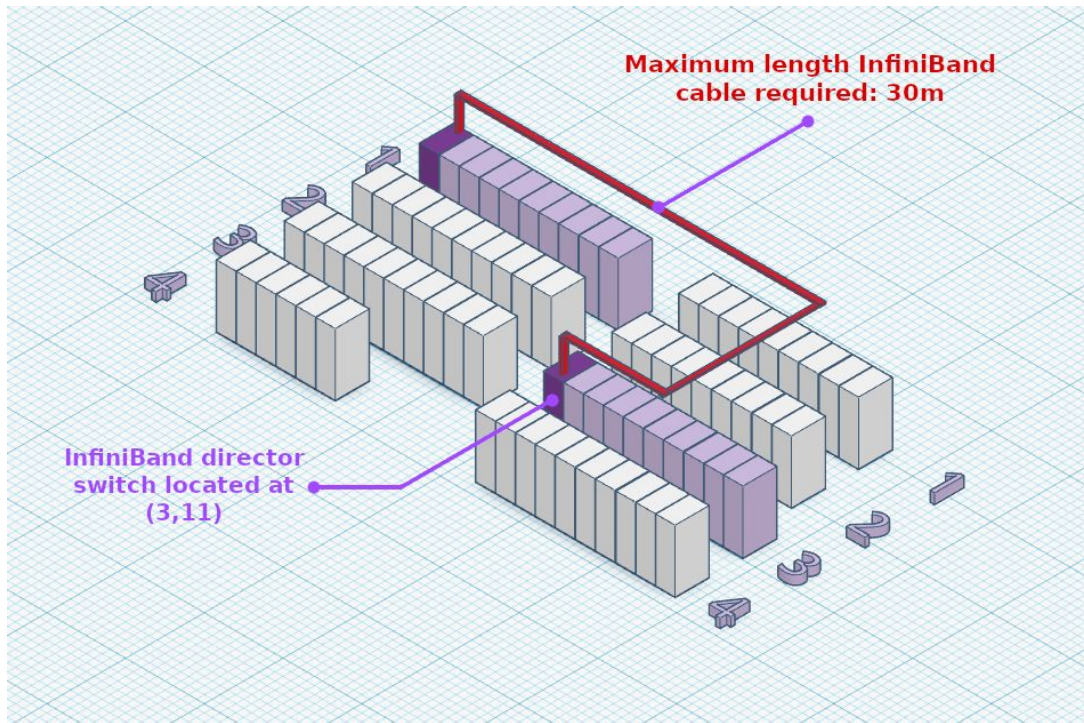
(Single Rack Configuration)



Data center floor planning & power distribution

Checklist:

- ❑ What rack does your DC provide?
- ❑ Max power (kW) per rack?
- ❑ What PDUs are provided, if any?
- ❑ What's the floor plan of the DC?
- ❑ Pull floor plan into DCIM/CAD to calculate cable lengths.



Software

There's a lot of software you'll want to run on your cluster. Some of this software is installed on the compute nodes, some of it is installed on the management nodes. Depending on your use case, these are some common packages you'll encounter.

- **Experiment management software** (Weights & Biases, Determined.ai, Spell.ml, etc.) - manage experiments and version control
- **Notebooks** - Jupyter Lab or iPython notebooks
- **Containers** - Docker mostly but also Singularity
- **Kubernetes** - manage instances and execute containers on them
- **Job scheduler** - (SLURM, Kubeflow)
- **Logging** - (Prometheus, Grafana, etc.)
- **Production inference** - (TensorFlow Serving, TensorRT)
- **Lambda Stack** - drivers, CUDA, TensorFlow, upgrade management, Dockerfile templates, etc.

GPU Accelerated Docker Containers:
Open Source Dockerfiles & NGC Container Support

System Wide Installation:
Works directly in your terminal or python notebook

Frameworks:

 TensorFlow

 PYTORCH

 Caffe

 Keras

 Caffe2

Drivers:
NVIDIA Drivers, CUDA,
InfiniBand Drivers

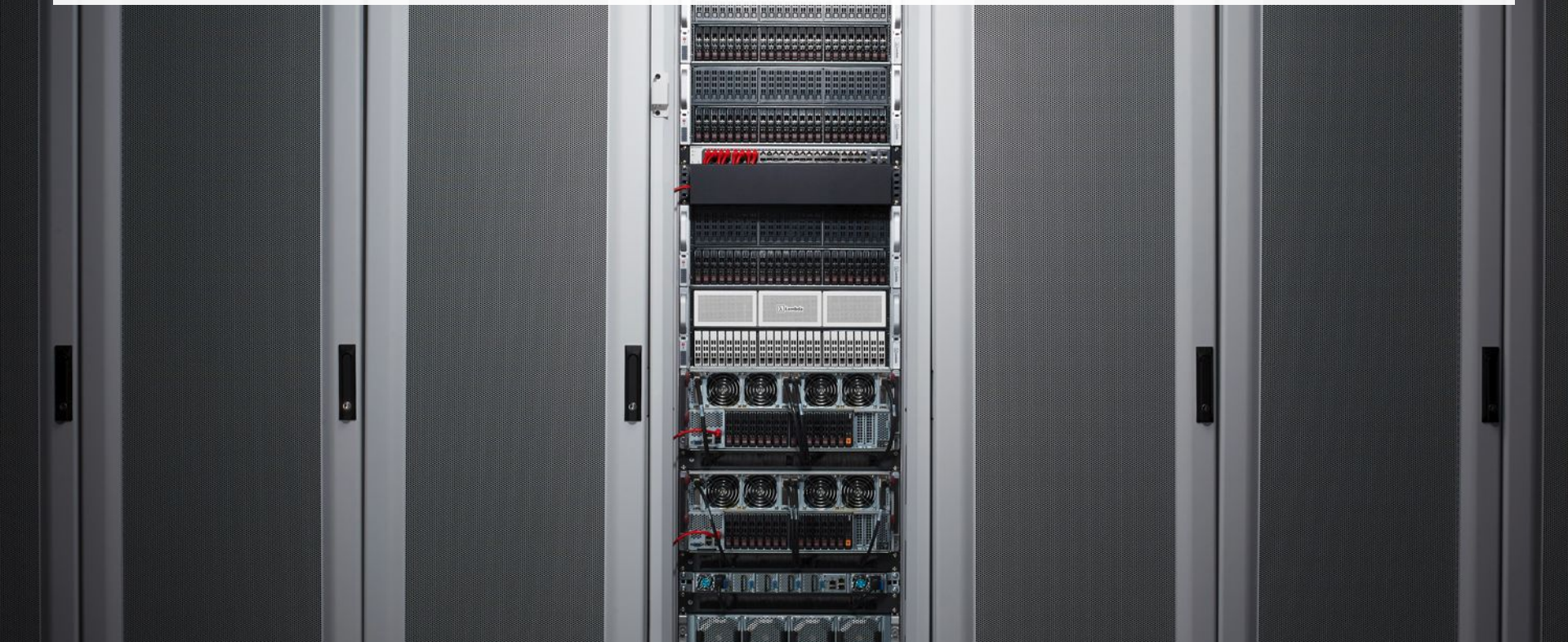
System Libraries:
protobuf, cuDNN, NCCL

Developer Tools:
git, vim, curl, emacs, tmux,
valgrind, ipython

Operating System:
Ubuntu 20.04 LTS, Ubuntu 18.04 LTS or Ubuntu 16.04 LTS
(Long Term Support)

For more info on Lambda Stack, see: <https://lambdalabs.com/lambda-stack-deep-learning-software>

Rack design



Lambda Echelon Single Rack Elevation

Mellanox SN3700C

32-port 100 Gb/s Ethernet
In-Band Management Switch

Mellanox QM8700

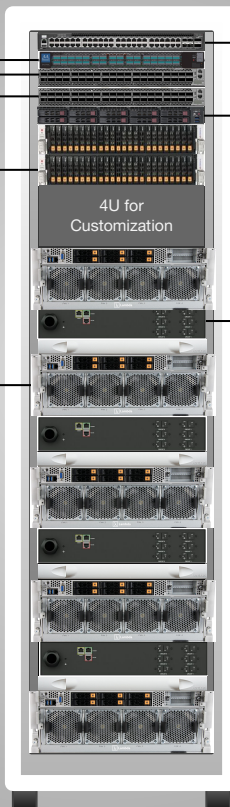
40-port Non-blocking Managed HDR 200Gb/s
Compute Fabric Switch (1)
Storage Fabric Switch (1)

Lambda Storage

Up to 384 TB of Flash NVMe
Storage
Storage Node (2)

Lambda Hyperplane

8x Tesla A100 GPUs with NVLink
Compute Node



Edgecore AS4610-54T

48-port 1Gb/s
Out-of-Band Management Switch

1U Management Server

Remote Access,
Cluster Control, Parameter Server

Raritan PDU

34.5kW 3-Phase Switched
PDU (415V / 60A)

APC NetShelter SX

45U
1200mm Depth Rack

	Single Rack Cluster
	Compute, Storage, Networking Rack
1	Edgecore AS4610-54T - 1G IPMI Switch
2	Mellanox SN3700C - In Band Switch
3	Mellanox QM8700 - Compute Switch MGR
4	Mellanox QM8700 - Storage Switch MGR
5	Lambda - Management Node
6	Lambda Matrix - Flash Storage
7	Up to 384TB NVMe
8	Lambda Matrix - Flash Storage
9	Up to 384TB NVMe
10	4U
11	Customizable
12	Space
13	
14	Lambda Hyperplane
15	8x A100 Server
16	
17	
18	Raritan
19	PX3-5008I2R-Q1 34.5kW PDU
20	1U Blanking Panel
21	
22	Lambda Hyperplane
23	8x A100 Server
24	
25	Raritan
26	PX3-5008I2R-Q1 34.5kW PDU
27	1U Blanking Panel
28	
29	Lambda Hyperplane
30	8x A100 Server
31	
32	Raritan
33	PX3-5008I2R-Q1 34.5kW PDU
34	1U Blanking Panel
35	
36	Lambda Hyperplane
37	8x A100 Server
38	
39	Raritan
40	PX3-5008I2R-Q1 34.5kW PDU
41	1U Blanking Panel
42	
43	Lambda Hyperplane
44	8x A100 Server
45	

Lambda Echelon - Single Rack Cluster - Bill of Materials

Description	Qty	TDP Draw (kW)	Ext. TDP (kW)
Lambda Hyperplane - 8x A100 Server	5	6.5	32.5
1U Management Node	1	1	1
2U Flash Storage Server	2	0.75	1.5
Mellanox SN3700C - 32 port - 100 Gb/s Ethernet - In-Band Switch	1	0.25	0.25
Mellanox QM8700 - 40 port - 200 Gb/s IB- Compute Fabric Switch	1	0.8	0.8
Mellanox QM8700 - 40 port - 200 Gb/s IB - Storage Fabric Switch	1	0.8	0.8
Edgecore AS4610-54T - 48 port - 1 Gb/s Ethernet - IPMI Switch	1	0.1	0.1
Mellanox MFS1S00-H003E - 3m active 200 Gb/s IB cables	49	0	0
Mellanox MFA1A00-C003 - 3m active 100 Gb/s Ethernet cables	8	0	0
Cat 5e Ethernet Cables - 1m/2m/3m	16	0	0
Line A - Raritan PX3-5008I2R-Q1 - 34.5kW PDU 415V 60A Input IEC 60309 3P+N+E 6h 60A (4P5W)	2	0	0
(Optional Line B) Raritan PX3-5008I2R-Q1 - 34.5kW PDU 415V 60A Input IEC 60309 3P+N+E 6h 60A (4P5W)	2	0	0
1U Blanking Panel	4	0	0
APC Netshelter SX 3305W - 45U - 600mm x 1200mm Rack	1	0	0
C14 to C13 cables (2 ft)	14	0	0
C13 to C20 cables (2 ft)	20	0	0
Rack TDP Draw (kW):			36.95

Lambda Echelon - Single Rack Cluster - Bill of Materials

Description	Qty	TDP Draw (kW)	Ext. TDP (kW)
Lambda Hyperplane - 3x A100 Server	5	6.5	32.5
1U Management Node	1	1	1
2U Flash Storage Server	2	0.75	1.5
Mellanox SN3700C - 32 port - 100 Gb/s Ethernet - In-Band Switch	1	0.25	0.25
Mellanox QM8700 - 40 port - 200 Gb/s IB- Compute Fabric Switch	1	0.8	0.8
Mellanox QM8700 - 40 port - 200 Gb/s IB - Storage Fabric Switch	1	0.8	0.8
Edgecore AS4610-54T - 48 port - 10 Gb/s Ethernet - In-Band Switch	1	0.1	0.1
Mellanox MFS1S00-H003E - 3m active 200 Gb/s IB cables	49	0	0
Mellanox MFA1A00-C003 - 3m active 100 Gb/s Ethernet cables	8	0	0
Cat 5e Ethernet Cables - 1m/2m/3m	16	0	0
Line A - Raritan PX3-5008I2R-Q1 - 34.5kW PDU 415V 60A Input IEC 60309 3P+N+E 6h 60A (4P5W)	2	0	0
(Optional Line B) Raritan PX3-5008I2R-Q1 - 34.5kW PDU 415V 60A Input IEC 60309 3P+N+E 6h 60A (4P5W)	2	0	0
1U Blanking Panel	1	0	0
APC Netshelter SX 3305W - 45U - 600mm x 1200mm Rack	1	0	0
C14 to C13 cables (2 ft)	14	0	0
C13 to C20 cables (2 ft)	20	0	0
Rack TDP Draw (kW):			36.95

The most important
number on this page:
Total rack TDP.

Back to basics

Single phase systems

$$P \text{ (watts)} = V \text{ (volts)} * I \text{ (amps)}$$

We use *I* from the French, *intensité du courant*.
Because that's what André-Marie Ampère used.

3-phase systems

$$P = 3 * V / \sqrt{3} * I$$

This simplifies to $P = \sqrt{3} * V * I$
Because $3 / \sqrt{3} = \sqrt{3}$

Real life 3-phase systems

$$P = \sqrt{3} * V * I * 0.8$$

It's very common to see an 80% regulatory derating factor applied to PDUs.

How do PDU manufacturers calculate power capacity?

From the APC8966 Data Sheet:



Example PDU:
APC 8966

Input frequency

50/60 Hz

Number of Power Cords

1

Load Capacity

17300VA

Maximum Input Current

60A

Maximum Line Current

48A

Regulatory Derated Input Current (North America)

48A

Nominal Output Voltage

208V

Nominal Input Voltage

208V 3PH

Input Connections

IEC 60309 60A 3P + PE



$$P = \sqrt{3} * V * I * 0.8$$

Plug in the numbers from the data sheet:

$$P = \sqrt{3} * 208 * 60 * 0.8$$

$$P = 17292.7953$$

$$P = \underline{17.3\text{kVA}}$$

See how they derate the maximum input current of 60A to the "Regulatory Derated Input Current (North America)" 48A? That's $48A = 60A * 0.8$.

That's where the 0.8 came from in the previous slide.

Some common PDU input plugs



IEC60309 - 60A 3-phase plug - 208V

Blue means the system is between 200 and 250V.



IEC60309 - 60A 3-phase plug - 415V

Red means the system is above 400V.



NEMA L15-30P 30A 3-phase plug - 208V

Receptacles & Plugs

Plug Photo	Name	Plugs into	Receptacle Photo	Max Amps
	IEC C13 Plug	IEC C14 Receptacle (on server)		15A (Max power ~2.5kW)
	IEC C14 Plug	IEC C13 Receptacle (on PDU)		15A (Max power ~2.5kW)
	IEC C19 Plug	IEC C20 Receptacle (on server)		20A (Max power ~3.2kW)
	IEC C20 Plug	IEC C19 Receptacle (on PDU)		20A (Max power ~3.2kW)

IEC stands for International Electrotechnical Commission, an international standards organization headquartered in Switzerland.

Lambda Echelon - Single Rack Cluster - Bill of Materials

Description	Qty	TDP Draw (kW)	Ext. TDP (kW)
Lambda Hyperplane - 8x A100 Server	5	6.5	32.5
1U Management Node	1	1	1
2U Flash Storage Server	2	0.75	1.5
Mellanox SN3700C - 32 port - 100 Gb/s Ethernet - In-Band Switch	1	0.25	0.25
Mellanox QM8700 - 40 port - 200 Gb/s IB- Compute Fabric Switch	1	0.8	0.8
Mellanox QM8700 - 40 port - 200 Gb/s IB - Storage Fabric Switch	1	0.8	0.8
Edgecore AS4610-54T - 48 port - 1 Gb/s Ethernet - IPMI Switch	1	0.1	0.1
Mellanox MFS1S00-H003E - 3m active 200 Gb/s IB cables	49	0	0
Mellanox MFA1A00-C003 - 3m active 100 Gb/s Ethernet cables	2	0	0
Cat 5e Ethernet Cables - 1m/2m/3m	16	0	0
Line A - Raritan PX3-5008I2R-Q1 - 34.5kW PDU 415V 60A Input IEC 60309 3P+N+E 6h 60A (4P5W)	2	0	0
(Optional Line B) Raritan PX3-5008I2F-Q1 - 13.5kW PDU 415V 60A Input IEC 60309 3P+N+E 6h 60A (4P5W)	2	0	0
1U Blanking Panel	1	0	0
APC Netshelter SX 3305W - 45U - 600mm x 1200mm Rack	1	0	0
C14 to C13 cables (2 ft)	14	0	0
C13 to C20 cables (2 ft)	20	0	0
Rack TDP Draw (kW):			36.95

Now you understand
how much work goes
into this bill of materials.

Lambda Echelon rack elevations

	Single Rack Cluster
	Compute, Storage, Networking Rack
1	Edgecore AS4610-54T - 1G IPMI Switch
2	Mellanox SN3700C - In Band Switch
3	Mellanox QM8700 - Compute Switch MGR
4	Mellanox QM8700 - Storage Switch MGR
5	Lambda - Management Node
6	Lambda Matrix - Flash Storage
7	Up to 384TB NVMe
8	Lambda Matrix - Flash Storage
9	Up to 384TB NVMe
10	4U
11	Customizable
12	Space
13	
14	Lambda Hyperplane
15	8x A100 Server
16	
17	
18	Raritan
19	PX3-5008I2R-Q1 34.5kW PDU
20	1U Blanking Panel
21	
22	Lambda Hyperplane
23	8x A100 Server
24	
25	Raritan
26	PX3-5008I2R-Q1 34.5kW PDU
27	1U Blanking Panel
28	
29	Lambda Hyperplane
30	8x A100 Server
31	
32	Raritan
33	PX3-5008I2R-Q1 34.5kW PDU
34	1U Blanking Panel
35	
36	Lambda Hyperplane
37	8x A100 Server
38	
39	Raritan
40	PX3-5008I2R-Q1 34.5kW PDU
41	1U Blanking Panel
42	
43	Lambda Hyperplane
44	8x A100 Server
45	

40 GPUs

Training time for Mask R-CNN on MSCOCO: 25 mins

Source: <https://mlperf.org/training-results-0-7/>

Four Rack Cluster

(1) Compute + Network + Storage Rack

1	Edgecore AS4610-54T - 1G IPMI Switch
2	Mellanox SN3700C - In-Band Switch
3	Mellanox QM8700 - Storage Switch
4	Lambda - Management Node
5	Lambda Matrix - Flash Storage Up to 384TB NVMe
6	Lambda Matrix - Flash Storage Up to 384TB NVMe
7	
8	
9	
10	5U Customizable Space
11	
12	
13	
14	Lambda Hyperplane 8x A100 Server
15	
16	Raritan PX3-5008I2R-Q1 34.5kW PDU
17	
18	1U Blanking Panel
19	
20	Lambda Hyperplane 8x A100 Server
21	
22	
23	
24	
25	Raritan PX3-5008I2R-Q1 34.5kW PDU
26	
27	1U Blanking Panel
28	
29	Lambda Hyperplane 8x A100 Server
30	
31	
32	
33	Raritan PX3-5008I2R-Q1 34.5kW PDU
34	
35	1U Blanking Panel
36	
37	Lambda Hyperplane 8x A100 Server
38	
39	
40	Raritan PX3-5008I2R-Q1 34.5kW PDU
41	
42	1U Blanking Panel
43	
44	Lambda Hyperplane 8x A100 Server
45	

(1) Compute + Networking Rack

1	Edgecore AS4610-54T - 1G IPMI Switch
2	Mellanox QM8700 - Spine Switch MGR
3	Mellanox QM8790 - Spine Switch
4	Mellanox QM8790 - Spine Switch
5	Mellanox QM8790 - Spine Switch
6	Mellanox QM8790 - Leaf Switch
7	Mellanox QM8790 - Leaf Switch
8	Mellanox QM8790 - Leaf Switch
9	Mellanox QM8790 - Leaf Switch
10	Mellanox QM8790 - Leaf Switch
11	Mellanox QM8790 - Leaf Switch
12	Mellanox QM8790 - Leaf Switch
13	Mellanox QM8790 - Leaf Switch
14	
15	Lambda Hyperplane 8x A100 Server
16	
17	
18	Raritan PX3-5008I2R-Q1 34.5kW PDU
19	
20	1U Blanking Panel
21	
22	Lambda Hyperplane 8x A100 Server
23	
24	
25	
26	Raritan PX3-5008I2R-Q1 34.5kW PDU
27	
28	1U Blanking Panel
29	
30	Lambda Hyperplane 8x A100 Server
31	
32	
33	Raritan PX3-5008I2R-Q1 34.5kW PDU
34	
35	1U Blanking Panel
36	
37	Lambda Hyperplane 8x A100 Server
38	
39	
40	Raritan PX3-5008I2R-Q1 34.5kW PDU
41	
42	1U Blanking Panel
43	
44	Lambda Hyperplane 8x A100 Server
45	

(1) Compute + Storage Rack

1	Edgecore AS4610-54T - 1G IPMI Switch
2	Lambda Matrix - Flash Storage Up to 384TB NVMe
3	
4	Lambda Matrix - Flash Storage Up to 384TB NVMe
5	
6	
7	
8	
9	
10	8U Customizable Space
11	
12	
13	
14	
15	Lambda Hyperplane 8x A100 Server
16	
17	
18	Raritan PX3-5008I2R-Q1 34.5kW PDU
19	
20	1U Blanking Panel
21	
22	Lambda Hyperplane 8x A100 Server
23	
24	
25	
26	Raritan PX3-5008I2R-Q1 34.5kW PDU
27	
28	1U Blanking Panel
29	
30	Lambda Hyperplane 8x A100 Server
31	
32	
33	Raritan PX3-5008I2R-Q1 34.5kW PDU
34	
35	1U Blanking Panel
36	
37	Lambda Hyperplane 8x A100 Server
38	
39	
40	Raritan PX3-5008I2R-Q1 34.5kW PDU
41	
42	1U Blanking Panel
43	
44	Lambda Hyperplane 8x A100 Server
45	

(1) Compute + Storage Rack

1	Edgecore AS4610-54T - 1G IPMI Switch
2	Lambda Matrix - Flash Storage Up to 384TB NVMe
3	
4	Lambda Matrix - Flash Storage Up to 384TB NVMe
5	
6	
7	
8	
9	
10	8U Customizable Space
11	
12	
13	
14	
15	Lambda Hyperplane 8x A100 Server
16	
17	
18	Raritan PX3-5008I2R-Q1 34.5kW PDU
19	
20	1U Blanking Panel
21	
22	Lambda Hyperplane 8x A100 Server
23	
24	
25	
26	Raritan PX3-5008I2R-Q1 34.5kW PDU
27	
28	1U Blanking Panel
29	
30	Lambda Hyperplane 8x A100 Server
31	
32	
33	Raritan PX3-5008I2R-Q1 34.5kW PDU
34	
35	1U Blanking Panel
36	
37	Lambda Hyperplane 8x A100 Server
38	
39	
40	Raritan PX3-5008I2R-Q1 34.5kW PDU
41	
42	1U Blanking Panel
43	
44	Lambda Hyperplane 8x A100 Server
45	

160 GPUs

Training time for
Transformer WMT E-G
Big: ~1.02 mins

Source:

<https://mlperf.org/training-re-sults-0-7/>

21 Rack Cluster		
(15) Compute + Storage Rack		
1	Edgecore AS4610-54T - 1G IPMI Switch	
2	1U Blanking Panel	
3	1U Blanking Panel	
4	Lambda Matrix - Flash Storage	
5	Up to 384TB NVMe	
6	Lambda Matrix - Flash Storage	
7	Up to 384TB NVMe	
8	6U Customizable Space	
9		
10		
11		
12		
13		
14	Lambda Hyperplane	
15		8x A100 Server
16		
17	Raritan	
18		PX3-5008I2R-Q1 34.5kW PDU
19		1U Blanking Panel
20	Lambda Hyperplane	
21		8x A100 Server
22		
23	Raritan	
24		PX3-5008I2R-Q1 34.5kW PDU
25		1U Blanking Panel
26	Lambda Hyperplane	
27		8x A100 Server
28		
29	Raritan	
30		PX3-5008I2R-Q1 34.5kW PDU
31		1U Blanking Panel
32	Lambda Hyperplane	
33		8x A100 Server
34		
35	Raritan	
36		PX3-5008I2R-Q1 34.5kW PDU
37		1U Blanking Panel
38	Lambda Hyperplane	
39		8x A100 Server
40		
41	Raritan	
42		PX3-5008I2R-Q1 34.5kW PDU
43		1U Blanking Panel
44	Lambda Hyperplane	
45		8x A100 Server
46		

800 GPUs

Training time for ResNet 50 on ImageNet:
1.06 mins

Source:

<https://mlperf.org/training-results-0-7/>
<https://github.com/NVIDIA/DeepLearningExamples/blob/2de29455a696b5608e24daa95f579f689b2d59e1/PyTorch/Translation/Transformer/README.md>

Node design



Choosing the right GPU depends on your use case



Quadro RTX 8000

Hyperparameter
Search



NVIDIA A100

Large scale
distributed training



NVIDIA T4

Production
inference

GPU benchmarking is hard!



<https://lambdalabs.com/blog/>



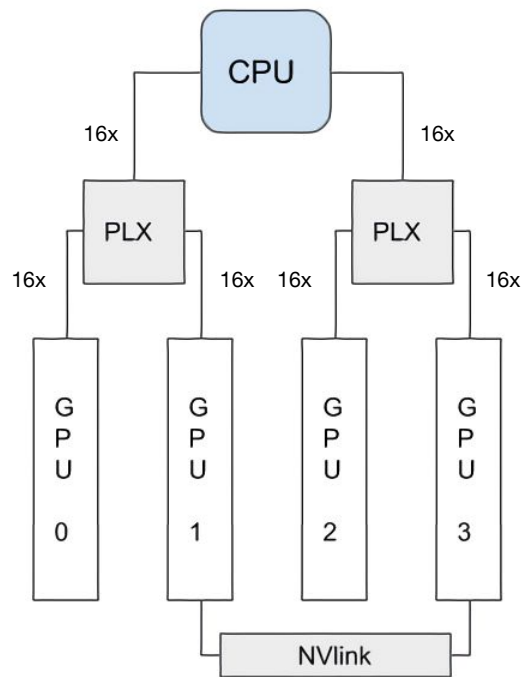
<https://mlperf.org>

What to look for in a node / compute system

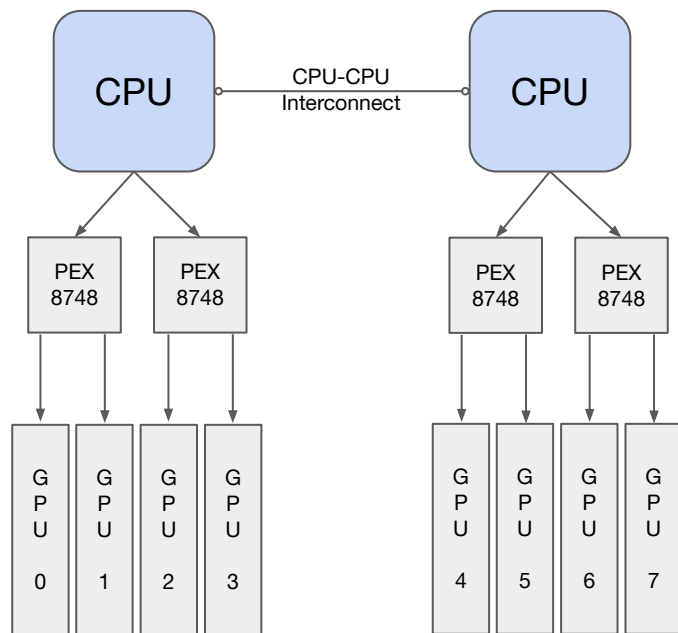
1. Number of PCIe lanes. (Affects total bandwidth.)
2. PCIe topology. (Tradeoff between CPU<>GPU and GPU<>GPU transfer.)
3. PCIe generation. (Gen 4 is 2x bandwidth of Gen 3.)
4. NUMA node topology. (Affects GPU peering & virtualization.)
5. FLOPS / \$.

GPU peering & PCIe topology

Example PCIe topology

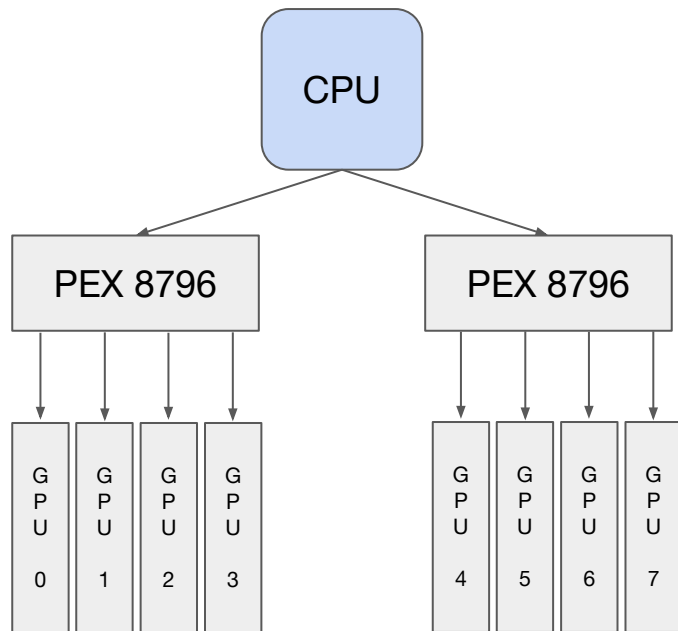


Dual root PCIe topology



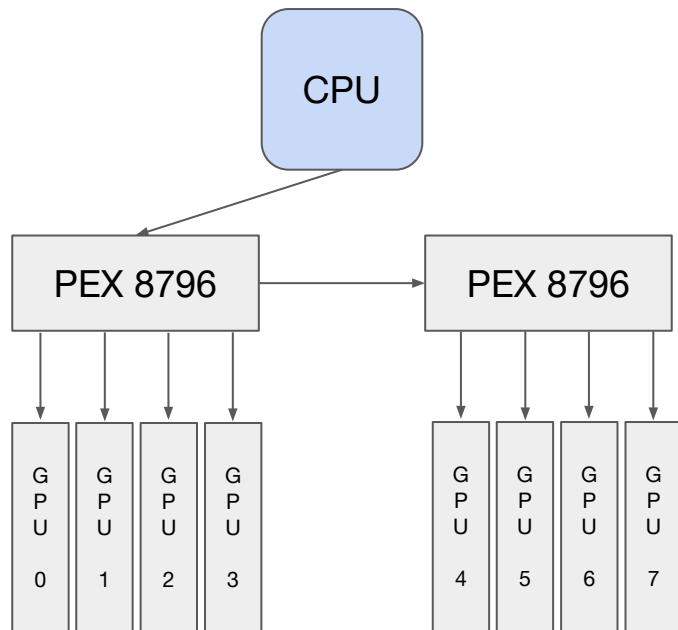
→
Arrow is 16x PCIe
Connection

Single root PCIe topology



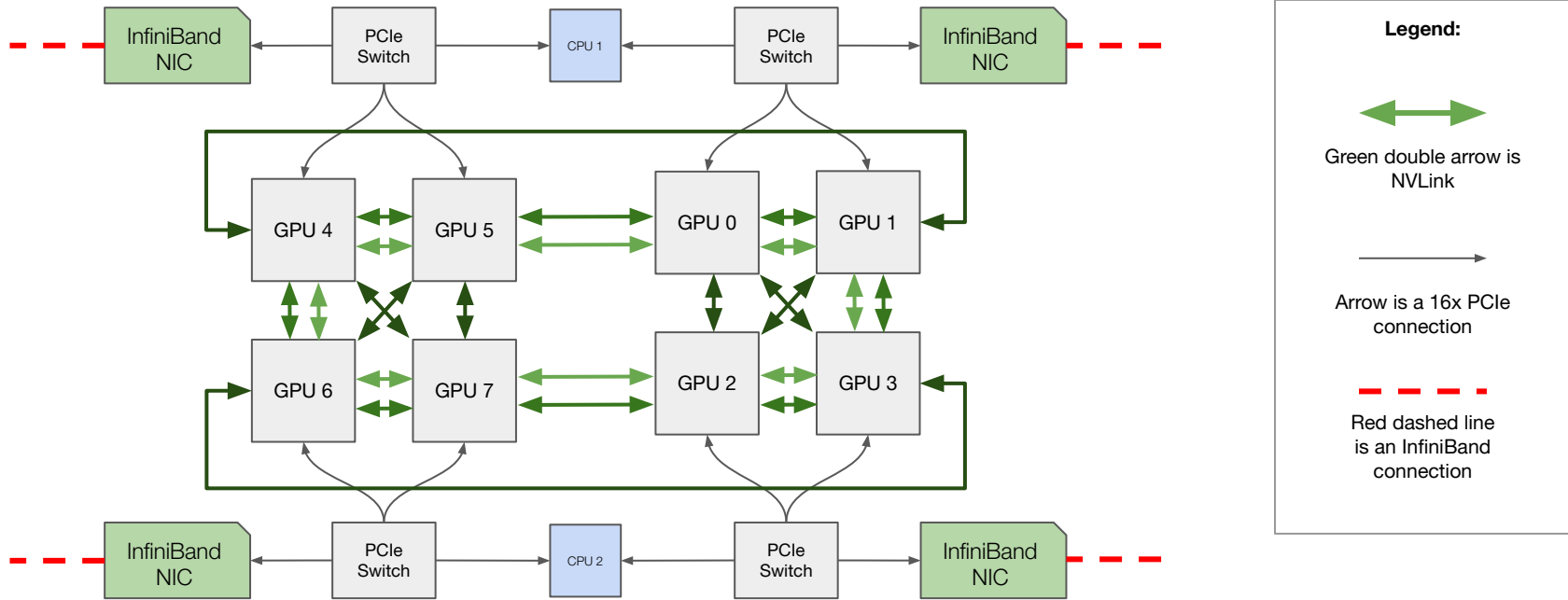
→
Arrow is 16x PCIe
Connection

Cascaded PCIe topology

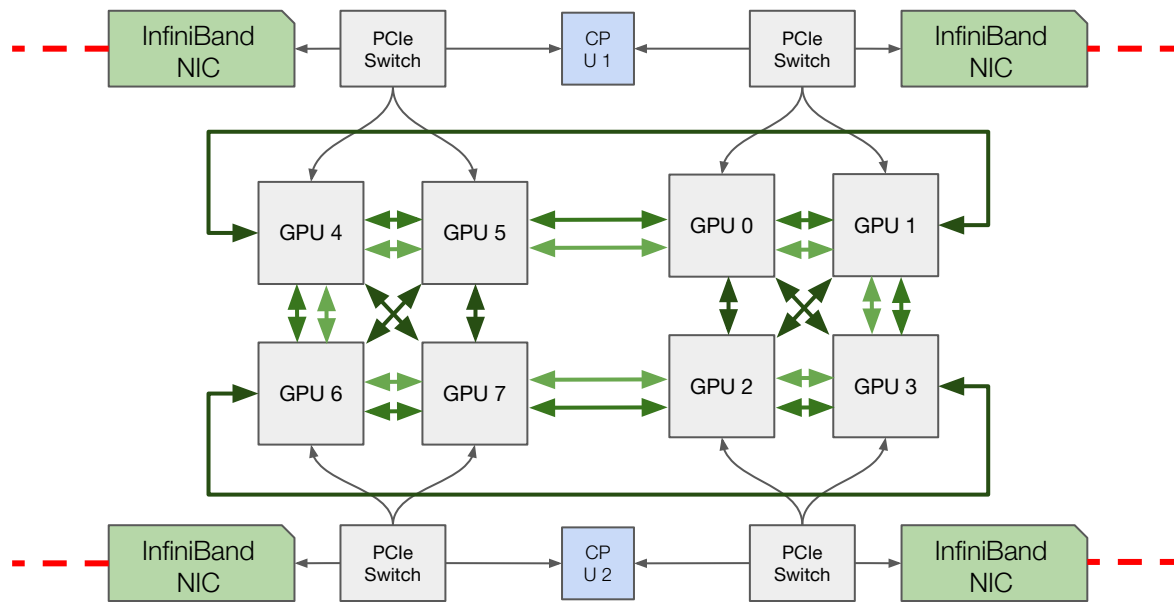


→
Arrow is 16x PCIe
Connection

NVLink hybrid cube-mesh topology

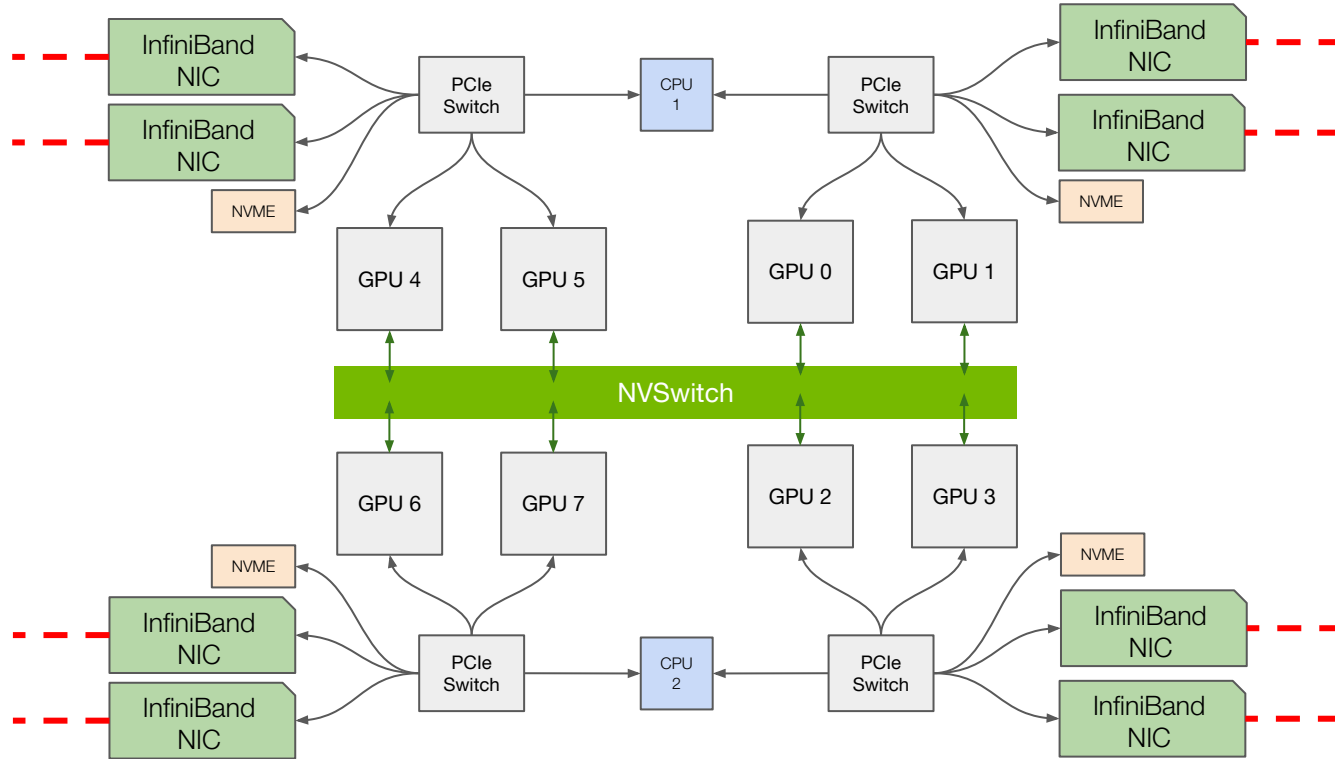


Four non-overlapping pathways

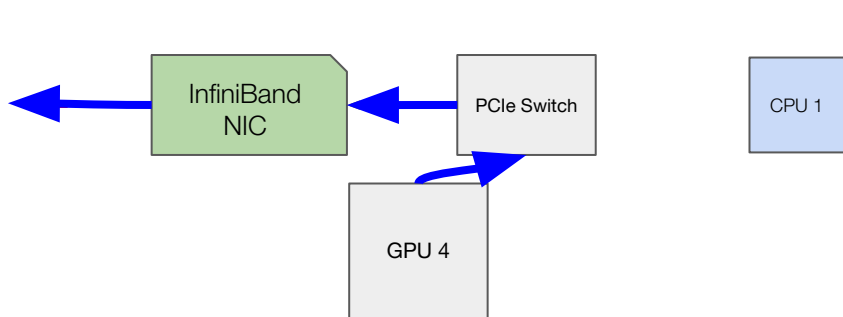


Each GPU has six NVLink connections which allows for four non-overlapping pathways to be drawn through all eight GPUs on the system and out an InfiniBand card, optimizing both GPU to GPU and node to node communication during distributed all-reduce. (See Sylvain Jeaugey's "NCCL 2.0" presentation for more information.)

NVSwitch Topology



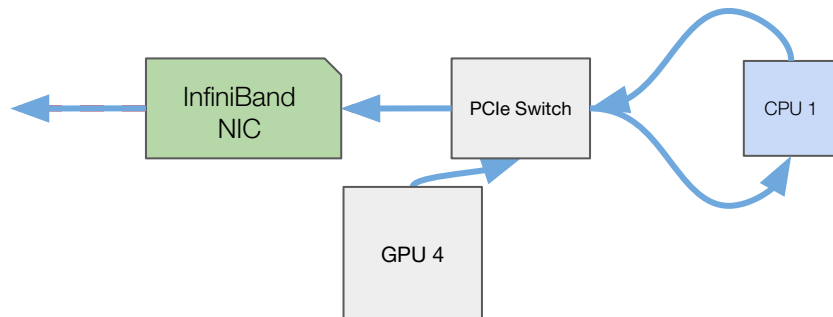
GPUDirect RDMA over InfiniBand



Data pathway with RDMA

(Directly out the door via PCIe switch)

Approximately doubles the peak node-to-node bandwidth vs. double copy.

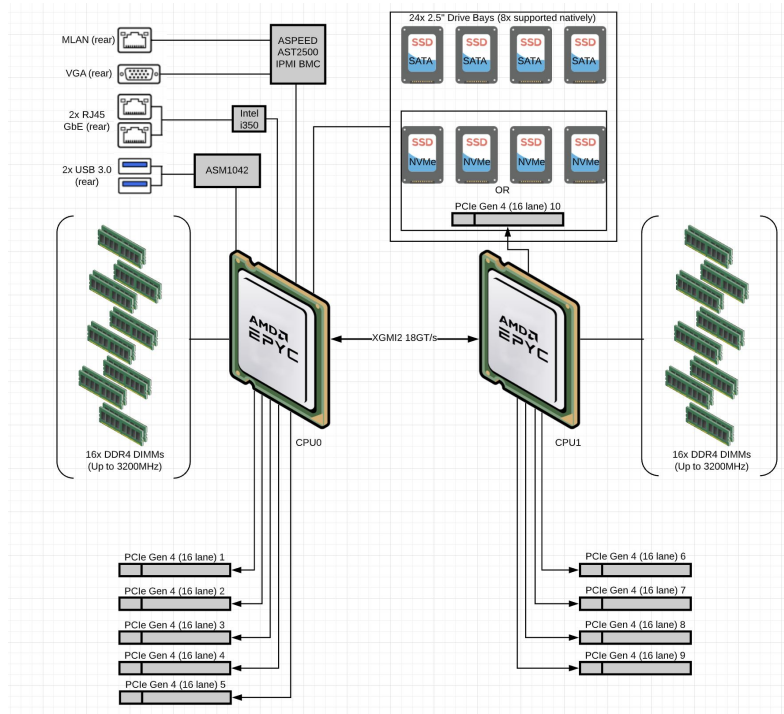


Data pathway without RDMA

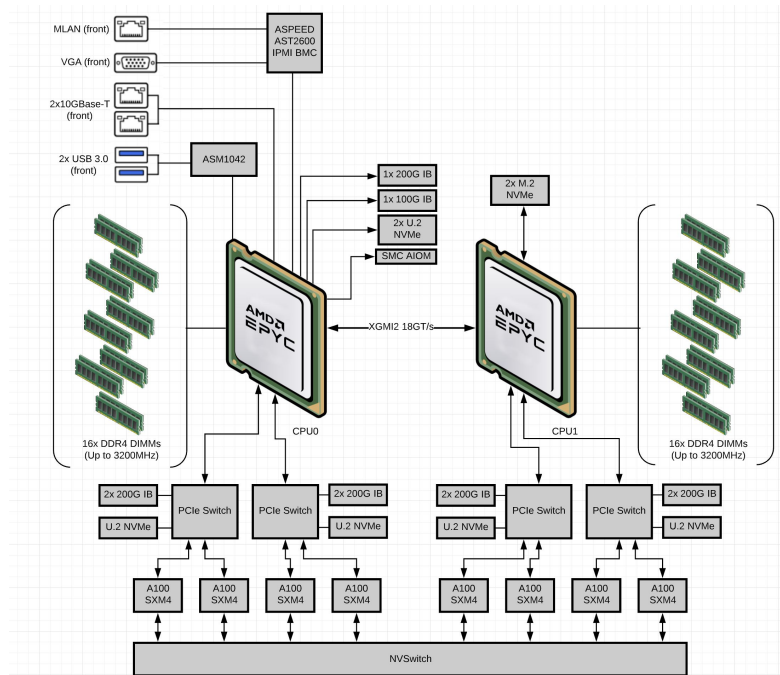
(Additional copy to CPU memory)

Real life examples

Lambda Scalar GPU server PCIe topology



Lambda Hyperplane-8 A100 GPU server PCIe topology



Putting it all together

- ❑ Write your Cluster, Rack, and Node BOMs.
- ❑ Price your BOMs.
- ❑ Negotiate a colocation contract.
- ❑ Order all of the parts on your BOMs.
- ❑ Assemble the servers.
- ❑ Ship the racks / servers to the colo / DC.
- ❑ Rack, stack, label and cable the servers.
- ❑ Install your software.
- ❑ Turn it on.
- ❑ Maintain it.

Or, work with Lambda to build a customized Lambda Echelon GPU cluster for your team.



You're done!



Citations

- Balaban, Sohmers, et al. **Lambda Echelon Deep Learning GPU Cluster - Reference Design Whitepaper**. <https://lambdalabs.com>
- Sergeev, Alexander and Mike Del Balso. **Horovod: fast and easy distributed deep learning in TensorFlow**. <https://arxiv.org/pdf/1802.05799.pdf>
- Pitch Patarasuk and Xin Yuan. **Bandwidth optimal all-reduce algorithms for clusters of workstations**. J. Parallel Distrib. Comput., 69:117–124, 2009. <https://www.cs.fsu.edu/~xyuan/paper/09jpdcc.pdf>
- Jeaugey, Sylvain. **NCCL 2.0**. (2017). <http://on-demand.gputechconf.com/gtc/2017/presentation/s7155-jeaugey-nccl.pdf>
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- Mellanox Cable Management Guidelines. https://www.mellanox.com/sites/default/files/products/interconnect/pdf/Mellanox_Cable_Management_Guidelines_and_FAQs_Application_Note.pdf
- Berkeley Lab's report on ASHRAE Thermal Guidelines. https://datacenters.lbl.gov/sites/all/files/ASHRAE%20Thermal%20Guidelines_%20SVLG%202015.pdf
- Mellanox CS7520 Dismantling Guide: https://www.mellanox.com/related-docs/prod_ib_switch_systems/CS7520_Dismantling_Guide.pdf

Additional thanks to Chuan Li, Steve Clarkson, and Thomas Sohmers for assistance with this document.

About me



- CEO of Lambda.
- Started using CNNs for face recognition in 2012.
- First employee at Perceptio. We developed image recognition CNNs that ran locally on the iPhone. Acquired by Apple in 2015.
- Research published in SPIE and NeurIPS.

About **Lambda**



Lambda provides AI infrastructure to the Fortune 500, major research institutions, and the DOD.

Our products include the Lambda Quad workstation, Lambda Hyperplane server, **Echelon GPU cluster**, and the Lambda GPU cloud.

www.lambdalabs.com