A photograph of a server room with rows of server racks. The racks are filled with equipment, and there are many glowing lights, some orange and some blue, creating a bokeh effect in the background. The overall scene is dimly lit, with the primary light sources being the server components and their indicator lights.

Testing AI Datacenter Fabric with Emulated Collective Communication Library Workload

Spirent Communications, Akihiro NAKAMURA

Agenda

- 自己紹介
- ネットワークにAIを適用する展望
- AI/ML向けDCネットワークの現状と検証課題
- 検証例の紹介

自己紹介



- 2001年3月 大学卒業
- 2001年4月~2017年3月
測定器専門商社にて光通信・情報通信関連の製品・ソリューションを販売
 - 新製品の立ち上げや市場開拓、コミュニティ・学会にも積極的に参加
- 2017年4月~ 現職
- 好きなスポーツ：野球、ゴルフ
- MPLS Japan 2010ごろ？から参加中

Agenda

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AI's potential impact on **Telco business cases**

\$450B - \$680B
Value to the global telecommunications industry
- McKinsey | 2024 -



REDUCED OPEX

AI streamlines operations and cuts costs by proactively identifying and addressing inefficiencies, faults, and areas for improvement in real time, leading to more efficient resource utilization and reduced downtime.



REDUCED CAPEX

By enhancing operational efficiency and optimizing resource utilization, AI allows for more strategic and need-based management of equipment upgrades and energy costs, thus minimizing ongoing capital expenditures.



INCREASED CUSTOMER EXPERIENCE

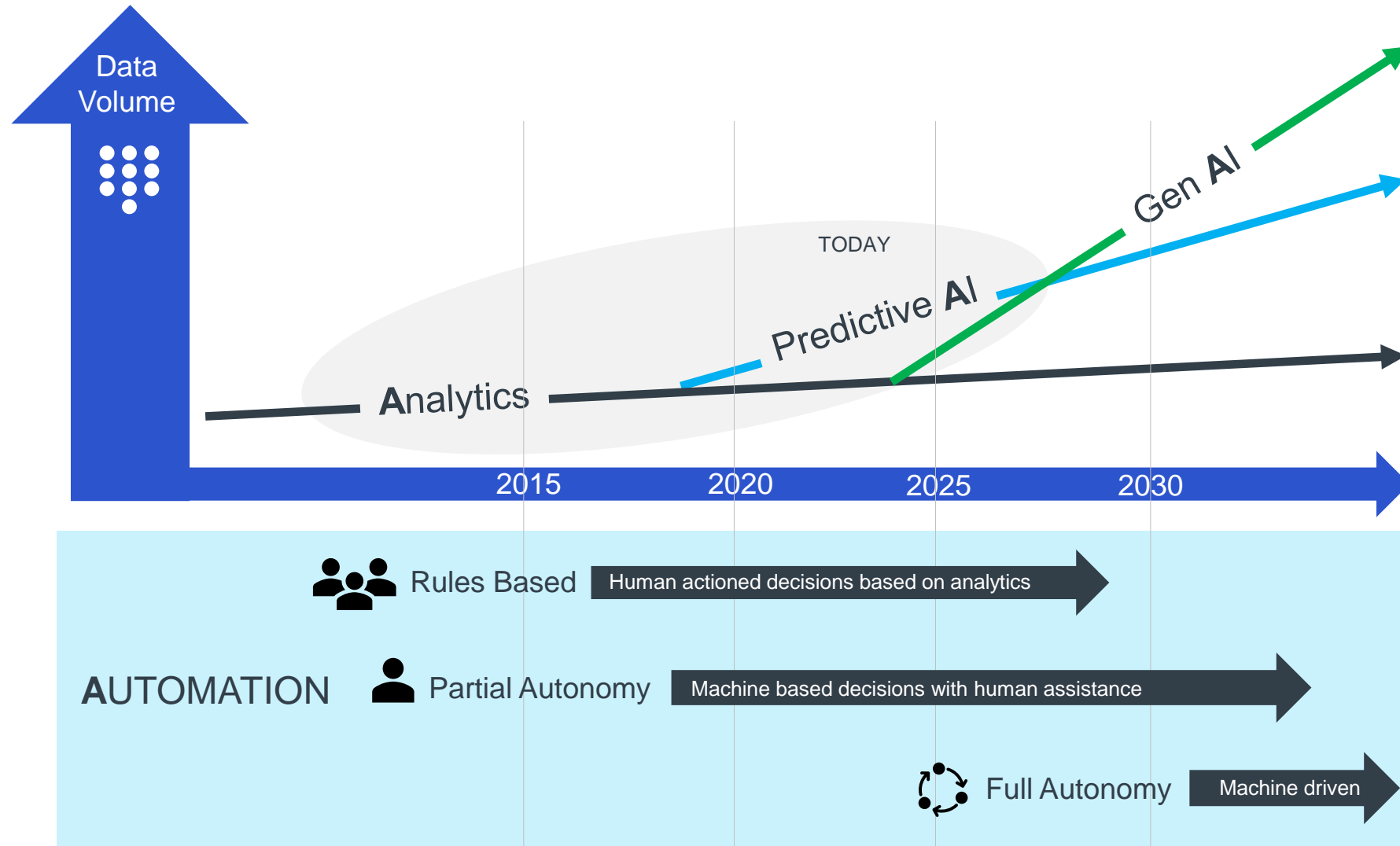
AI enhances customer satisfaction and service quality by analyzing engagement and sentiment, offering proactive recommendations, thereby reducing churn and boosting revenue.



INCREASED REVENUE

AI leverages customer engagement and behavioral trends to uncover new revenue opportunities, enhancing existing services and creating personalized offerings.




The power of 3 (Analytics, Automation & AI) over time





Use cases of AI on **Network & Infrastructure**

Seen across our global telecom operators

in NETWORK management

		
NETWORK DESIGN	OPERATIONS & MGT.	SECURITY
Traffic prediction	Network optimization	Threat prediction
Capacity planning	Predictive maintenance	Fraud detection
Radio map planning	Fault prediction	Resilience planning
	Anomaly detection	
	Root cause analysis	

in INFRASTRUCTURE

	
RAN	DEVICES (UE)
Energy optimization	GenAI on device
Spectral efficiency	AI/ML in NR Air Interface
Traffic steering	
Load balancing	
Mobility optimizations	

How GenAI will complement PredAI in Network Management

Today	
Analytics & Predictive AI/ML	
Top 5 network mgt. use cases (today):	
1	Network management
2	Predictive maintenance
3	Traffic prediction (network planning)
4	Anomaly detection
5	Performance Assurance



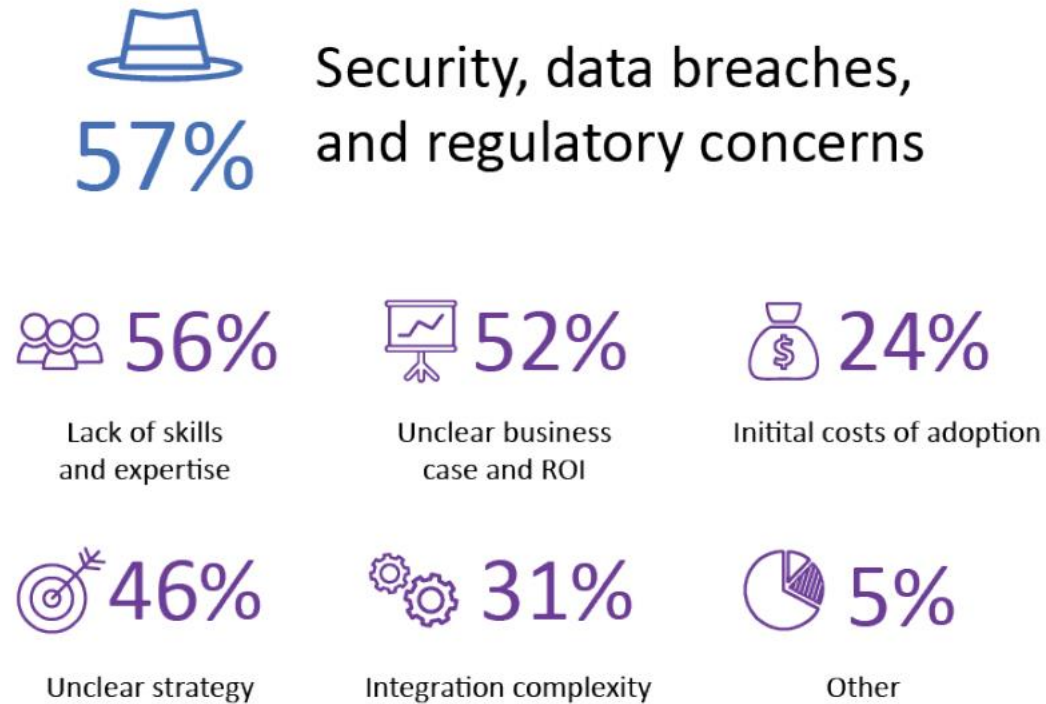
Evolution			
Generative AI			
For:	Content creation:	Querying (human-machine):	Supporting predictive models:
Use cases:	<ul style="list-style-type: none"> • Troubleshooting guides • Incident reports • Network topologies & coverage maps • Configuration scripts • Test traffic generation (for adversarial scenarios) 	<ul style="list-style-type: none"> • Document interrogation (field and support teams) • Consulting historical data (similar issues / resolutions) 	<ul style="list-style-type: none"> • Anomaly detection • Augmenting data sets • Preventing overfitting

Save \$30B - \$45B / Year

Incremental value to the global telecommunications industry

Barriers to adoption

Figure 3: Biggest barriers to widespread AI adoption in telecom

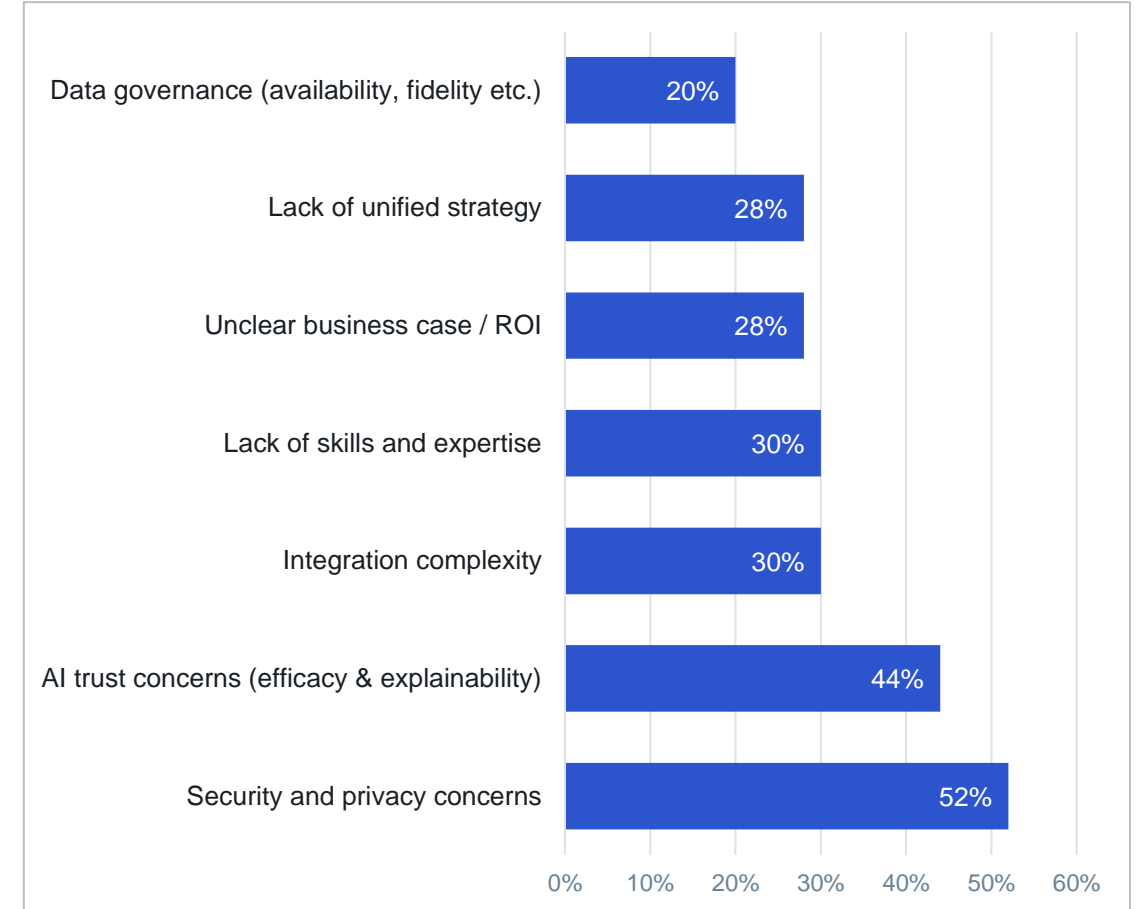


Note: Respondents could select multiple options.

Source: Omdia, Telecoms.com Annual Industry Survey 2023, n=87

© 2024 Omdia

What key challenges are slowing down the adoption of AI in your organization's network? (Select Top 3)



Spirent / RCR poll | N=50

Case study – Validating Security AI assistants

~99%

Firewall breaches are due to misconfigurations compounded by thousands of overlapping rules & policies

AI

dependency aware assistants

Can optimize & troubleshoot overlapping rules while adding new rules more efficiently = enhanced solution performance & trust.



How to validate AI assistant optimized rules & policies do not impact the security or performance of the network

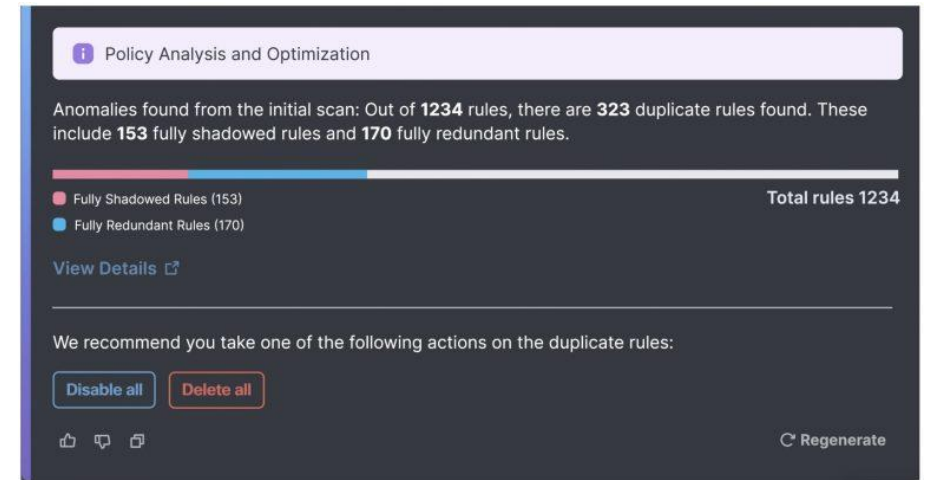


Automated & continuous Security testing

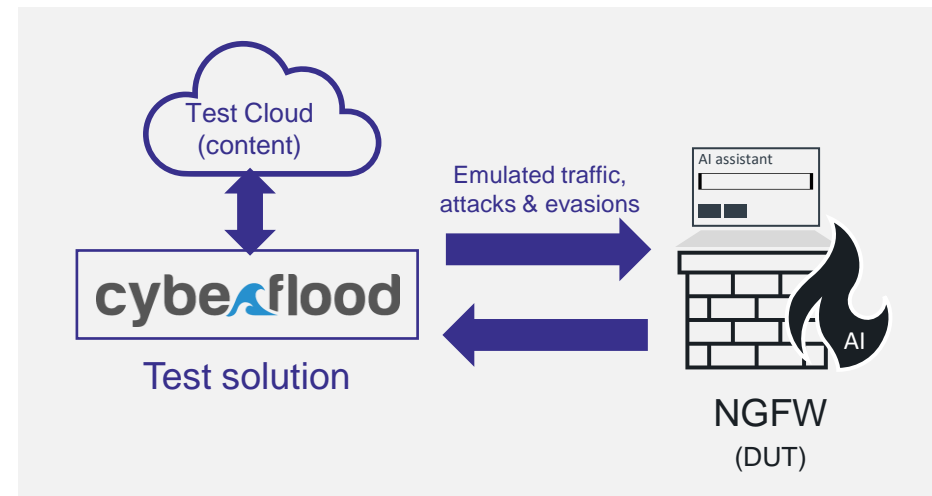
Realistic traffic & attack generation

Validates

- AI changes are implemented correctly.
- Efficacy of AI predictive blocking.
- Traffic performance impact
- Hallucinations or laziness



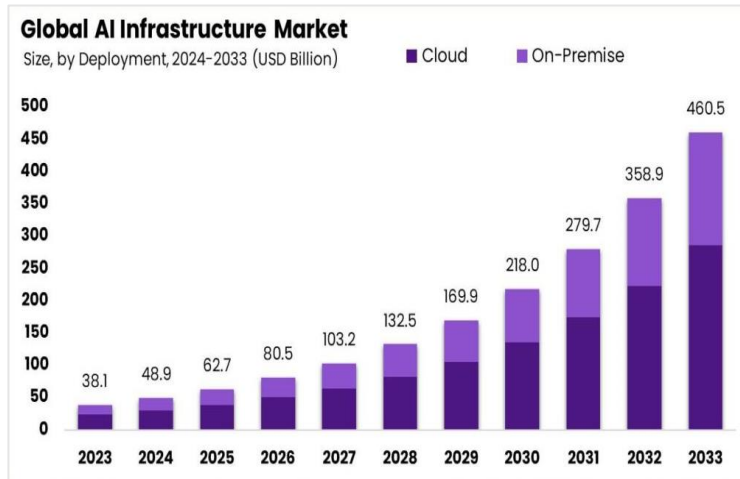
Example: Cisco Firewall AI Assistant



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AI Infra Market Trends



Market Scoop April '24

- Ten-year overall AI infrastructure market expected to grow from \$38B in '23 to \$460B in 2033.
- Ten-year CAGR of 28.3%

AI Adoption in Finance '24



- 46% are already using large language models (LLMs)
- NVIDIA 4/24 Survey

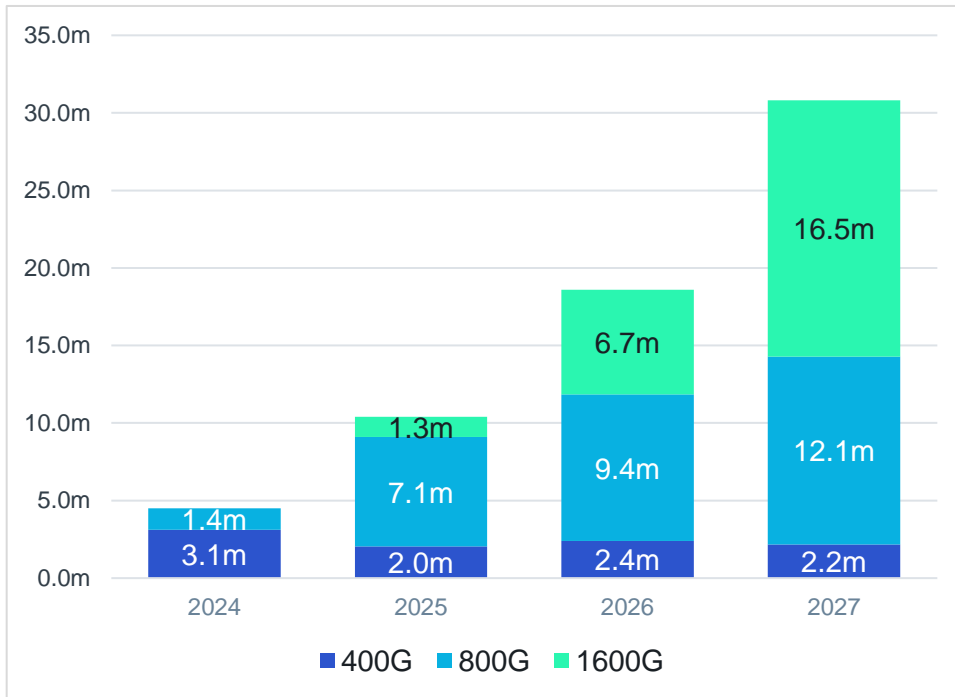
Ultra Ethernet Consortium



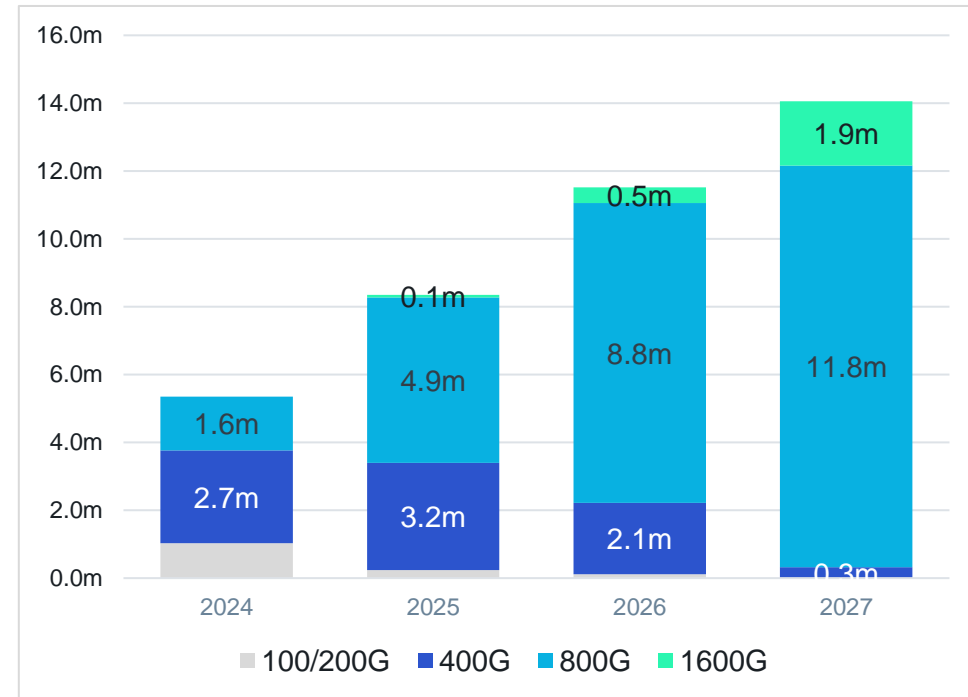
- Standards bodies defining next generation network specifications for evolving AI demands

Data Center Networking Link Speed update

AI Back-End Networks Ethernet speed growth (ports)



Front-End Networks Ethernet speed growth (ports)



Source: Dell'Oro Q3 2024

- Ethernet revenue share vs InfinBand will grow from 17% in 2023 to **51% in 2028**

Bottlenecks For Training Performance

Model name	Model Size (billion params)	Dataset size (billion tokens)	Training ZFLOPS (10 ²¹)	GPU	GPU FLOPS	GPU utilization	Training time (in weeks)	Number of GPUs
OPT	175	300	430	H100	3,000	0.5	1	474
<u>LLaMA</u>	65	1,400	600	H100	3,000	0.5	1	662
<u>LLaMA 2</u>	34	2,000	400	H100	3,000	0.5	1	441
LLaMA2	70	2,000	800	H100	3,000	0.5	1	882
GPT-3	175	300	420	H100	3,000	0.5	1	463
GPT-4* (est.)	1,500	2,600	31,200	H100	3,000	0.5	1	34,392



Training Performance Optimization

Memory Access

Network Performance

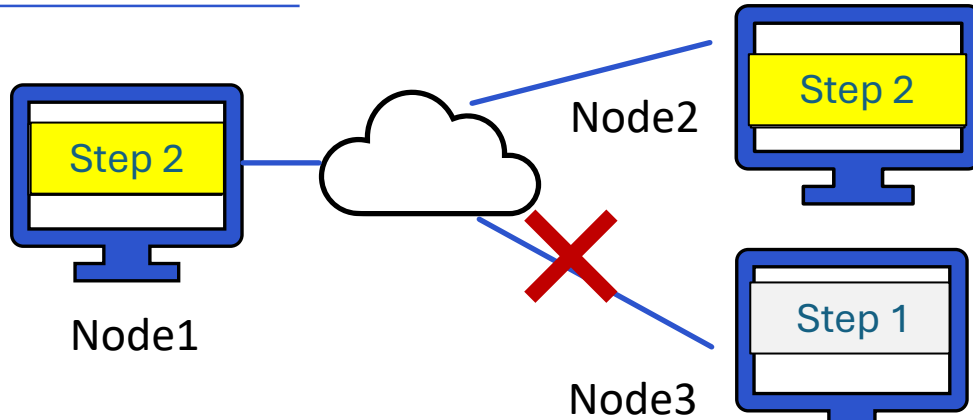
1% packet loss = 30%+ training performance down

Table 3 — GPU cluster size for training various LLM models with H100 GPUs, FP8 arithmetic, and 50% utilization.

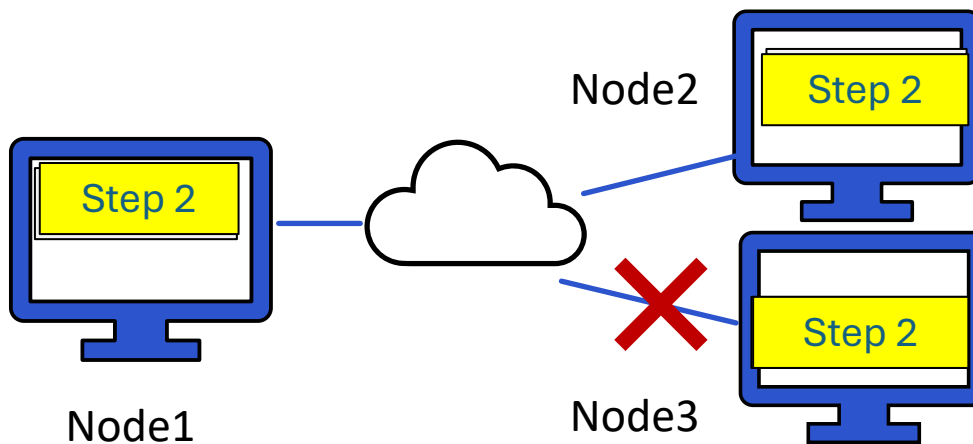
GPU fabrics for GenAI workloads By [Sharada Yeluri](#) on 8 Dec 2023

Collective Communication Amplifies The Network Bottleneck

Traditional P2P Communication



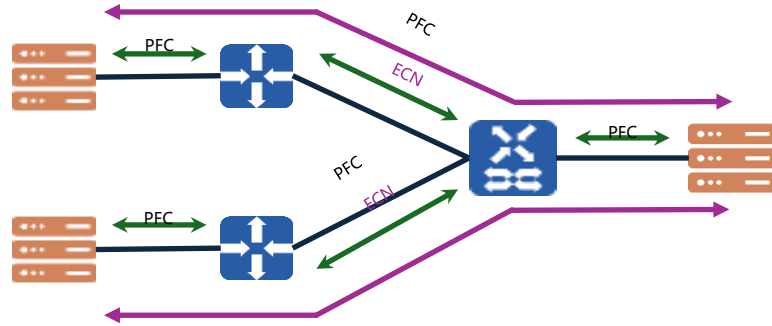
Collective Communication



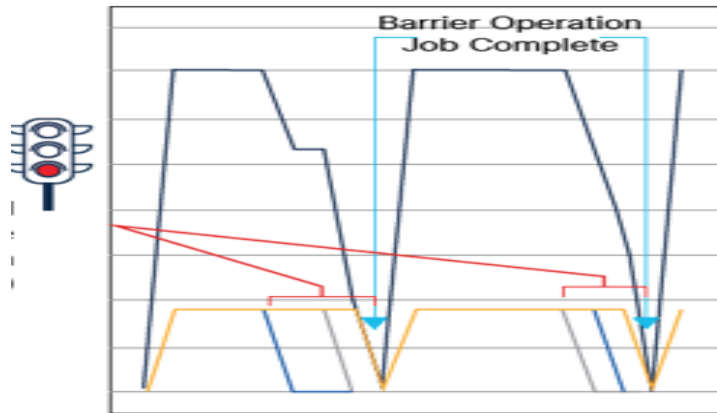
Wait for the recovery of Node 3

Challenges Of Network For LLM Training

Challenges – Network Congestions

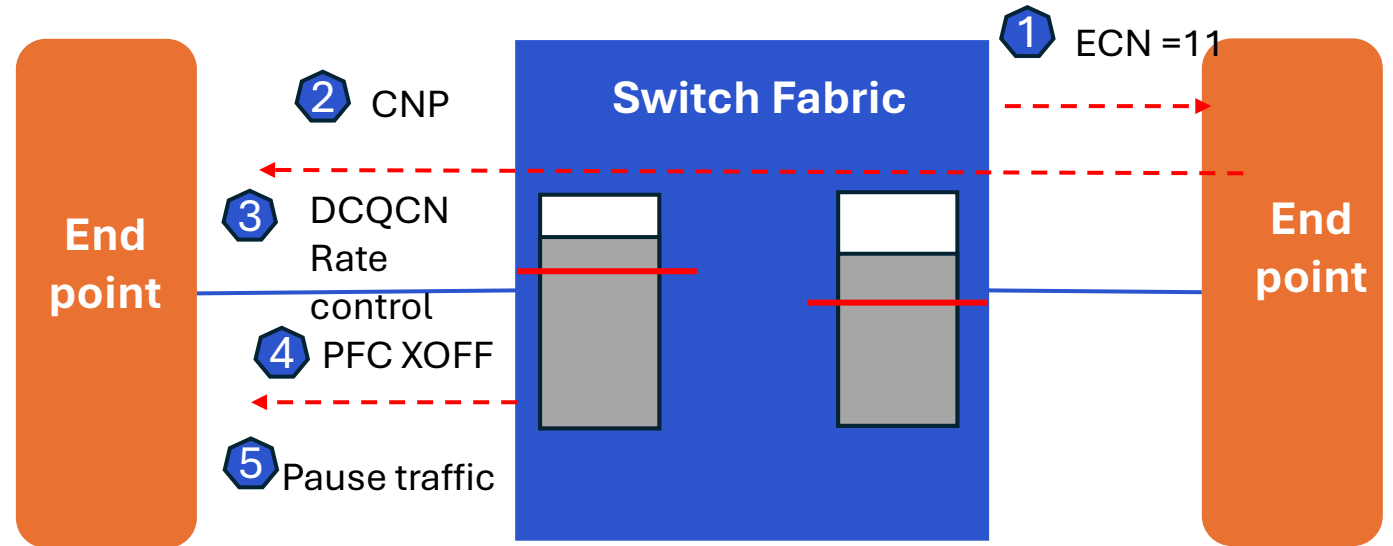


Frequent incast traffic



Bursty but elephant flows

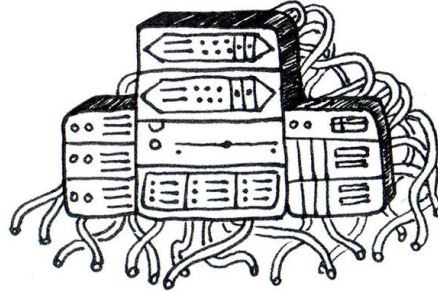
Congestion control – DCQCN w/ ECN & PFC



Testing AI Fabric with Real xPU-Based Systems



Takes a datacenter to test a datacenter



Burden is on the end customer at all levels of solution



Requires staff with specific knowledge of AI network landscape, traffic and protocols

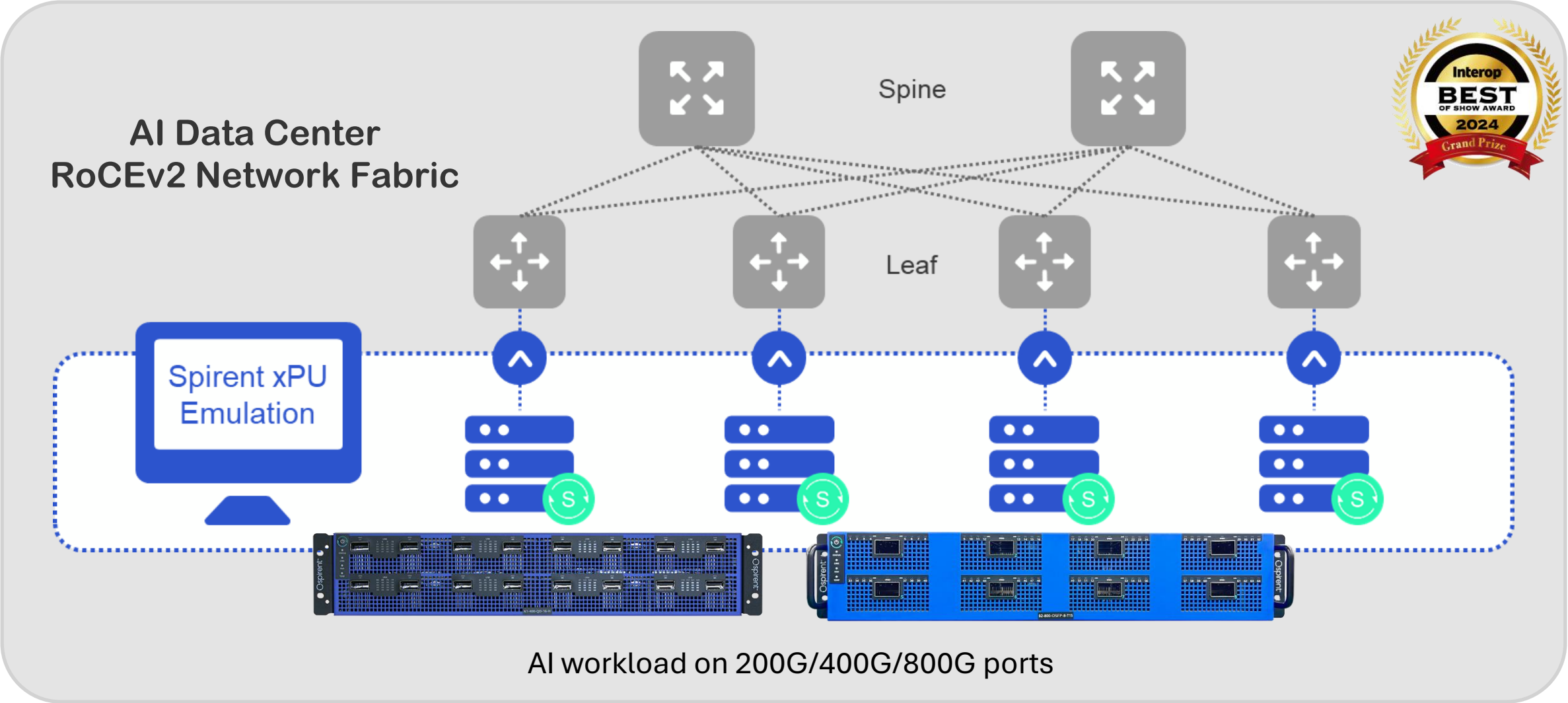
High Cost – Limited Repeatable Testing:

Procurement, Deployment, Management, Test Creation/Execution, Power Consumption, Results/Analytics and Ongoing Test Development

Agenda

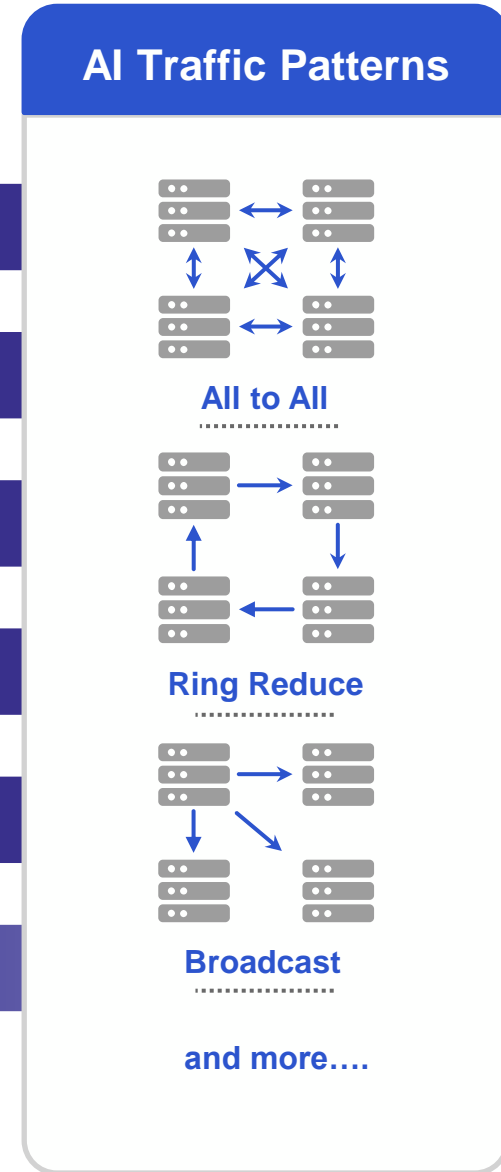
- 自己紹介
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Spirent A1: Hyper-Realistic AI Workload Emulation



Comprehensive Test Coverage

- ✓ Collective Communications Library (CCL) Traffic Patterns
- ✓ Congestion Scenarios
- ✓ Network Performance
- ✓ Service Performance
- ✓ Resiliency (Negative Traffic Testing)
- ✓ Multi-Speed Roadmap from 100G to 1.6Tb



Spirent vs “Do it Yourself”

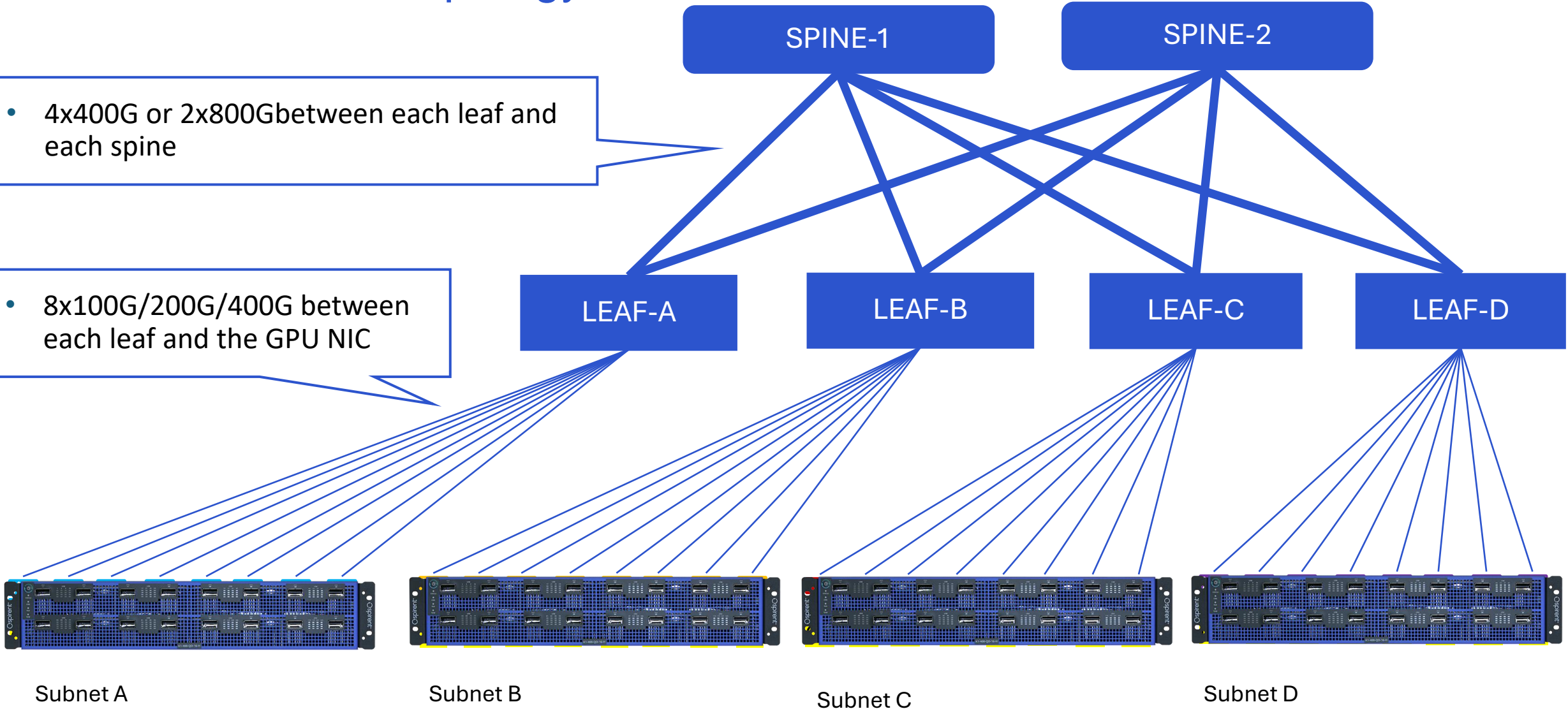


	Spirent	DIY
Equipment	One Spirent A1 with 16 x 400G	Two Eight-Card xPU Server with 400G NIC Four Eight-card xPU Server with 200G NIC
Material Cost	<ul style="list-style-type: none"> • Less than a real 16 x port xPU system from NVIDIA. • One time cost. 	<ul style="list-style-type: none"> • Varies per region; as high as \$500K+ /server in some regions. • Double the cost every time a new generation GPU releases
Power	4,500W	10,000W+ per server
Expertise/Skills	No additional cost. Current Spirent test experts can use our solution to test the AI fabric. Complexity of in building traffic patterns masked by the RocEv2 & Traffic Wizard.	Cost to train network engineer to have AI knowledge or hire new AI engineers. Complexity in creation of test scenarios.
Maintenance & Development Cost	Minimum maintenance cost.	Maintenance of the xPU test bed (incl. Servers).
Test Cases	Consistent and Repeatable.	Inconsistent across multiple runs.
Additional values	General test platform which can be used for AI tests and Routing/Switching tests greatly improves ROI.	Single purpose test bed ONLY for AI fabric testing.
Availability	Short lead times. Less than 4 weeks.	Long lead times due to xPU shortages.

A Test Practice – Topology

- 4x400G or 2x800G between each leaf and each spine

- 8x100G/200G/400G between each leaf and the GPU NIC



A Test Practice – Test Cases

CCL Pattern – 32-port

- 1 x 32-port Ring
- 8 x 4-port Ring
- 1 x 32 Alltoall with PXN
- 4 x 8 Alltoall w/o PXN

Scenarios

- Benchmarking
No congestion, ideal load balancing
- X 80% rate-limiting
- Imbalance
Break fibers

Parameters

- Packet size:
514B, 1582B, 8194B
- X Data size:
128MB-64GB
- DUT: per flow LB or
Packet Spray LB

Wizard To Configure CCL Testing

Select RoCEv2 Ports
Enable two or more ports to configure RoCEv2 servers. The RoCEv2 server will be emulated on the port.

PFC ECN DCQCN

Enable	Port Name	Enable PFC	Priority Queue	Enable ECN	ECN
<input checked="" type="checkbox"/>	Cisco //1/65	<input checked="" type="checkbox"/>	Priority 4	<input checked="" type="checkbox"/>	01:
<input checked="" type="checkbox"/>	Juniper //1/73	<input checked="" type="checkbox"/>	Priority 4	<input checked="" type="checkbox"/>	01:

Configuration CCL AllRingReduce.
Select Rocev2 Ports For CCL AllRingReduce.

Manual Select xPU Total: Server C

Filter: Apply Clear

Col Cor

Rovev2 Ports

- A1
- B1
- C1
- D1
- A2
- B2
- C2
- D2
- A3
- B3
- C3

Add ->
<- Del

AddAll =>
<= DelAll

Steps

- Select Wizard
- Configure RoCEv2 Ports Params
- Configure RoCEv2 Servers
- Select Queue Pair Traffic Endpoints
- Summary**

Summary
Summary description of configuration

- o MAC Address: 00:10:94:03:00:01
- o QP start value: 400
- o Number of QP Blocks: 1
- o Number of QP's per Block: 1
- Port: PortConfig5 //1/1 (offline) , PFC: False , ECN: True
 - o IPv4 Address: 192.168.104.2
 - o MAC Address: 00:10:94:04:00:01
 - o QP start value: 500
 - o Number of QP Blocks: 1
 - o Number of QP's per Block: 1
- Port: PortConfig6 //1/1 (offline) , PFC: False , ECN: True
 - o IPv4 Address: 192.168.105.2
 - o MAC Address: 00:10:94:05:00:01
 - o QP start value: 600
 - o Number of QP Blocks: 1
 - o Number of QP's per Block: 1

Selected QP Traffic

- Traffic Orientation: Bidirectional
- Number of Queue pairs selected: 2

===== **END OF SUMMARY** =====

Select CCL algorithm
Select CCL algorithm

CCL RingAllReduce Algorithm

CCL AllToAll With PXN Algorithm

CCL AllToAll Without PXN Algorithm

ation.

rs :

Number of QPs per Block	Frame Size	Total Data Size(Gb)	Reduce Time(ms)
	1498	400	200

Reset < Back Next > Preview Topology Finish Cancel

Results Analysis

Primary Collective KPIs

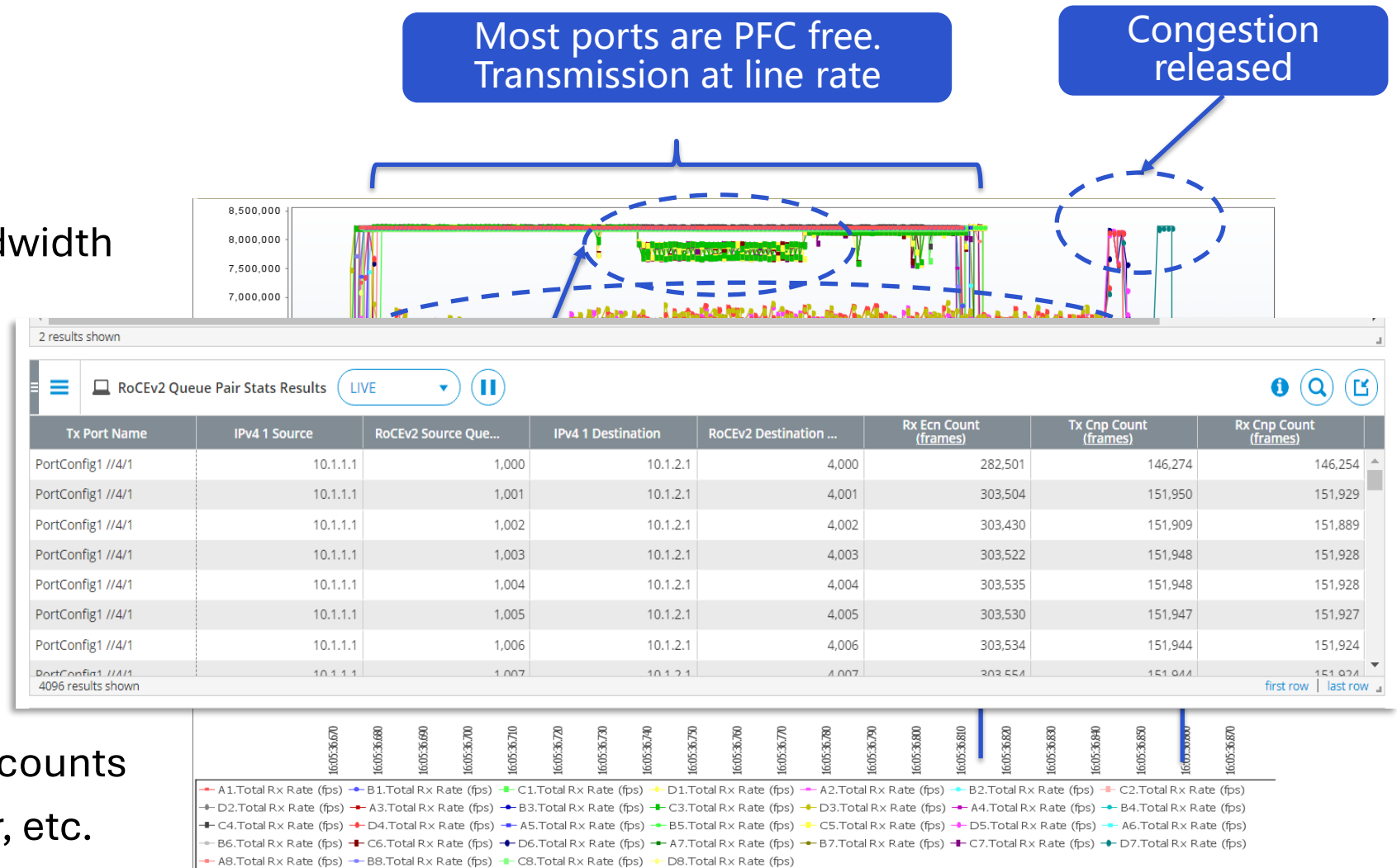
- Algorithm bandwidth, Bus bandwidth

Collective related Stats:

- Transmission Time
- Job Completion Time
- Tail Latency

Network Stats:

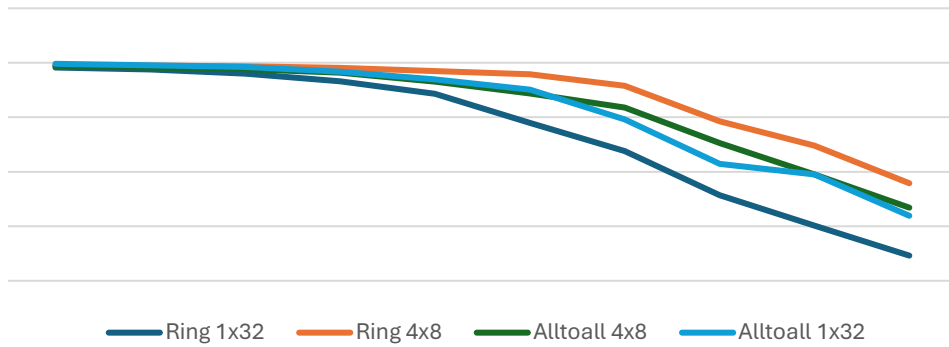
- PFC counts, ECN counts, CNP counts
- Packet loss, Latency, reorder, etc.



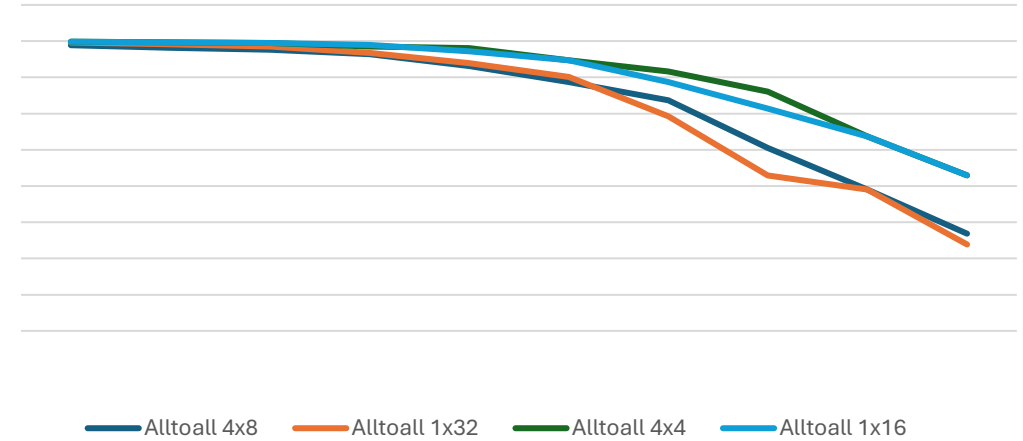
1ms high resolution sampling

A Test Practice – Sample Of Results

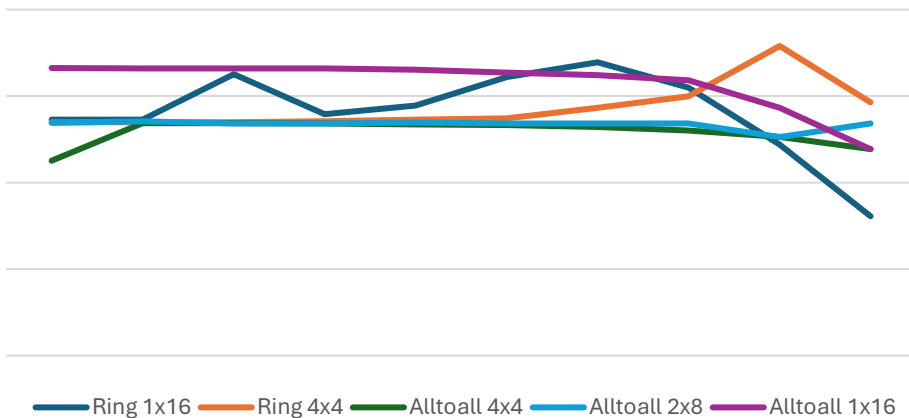
80% BW limitation



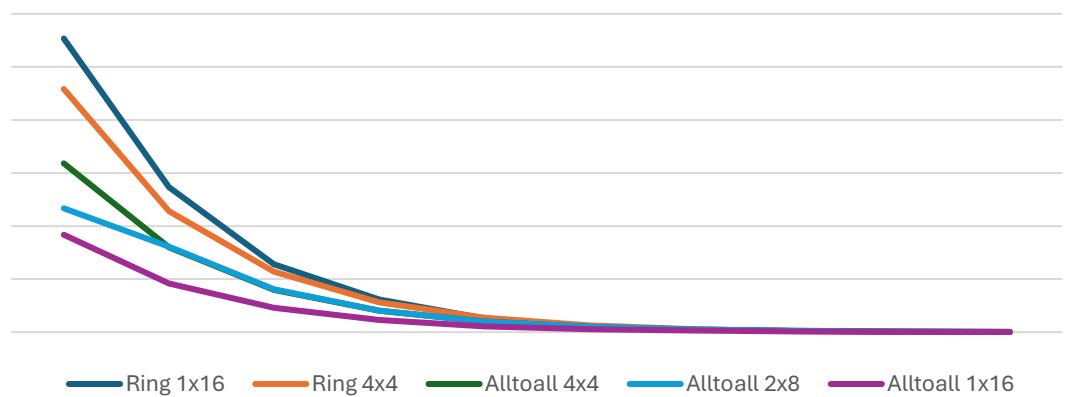
No congestion



BusBW Imbalance



Rx PFC number





Validate the readiness of new
Ethernet infrastructures for
next-gen AI applications



Thank you.

Akihiro.Nakamura@spirent.com

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