

Storage Developer Conference September 22-23, 2020

Use Cases for NVMe-oF for Deep Learning and HCI Pooling

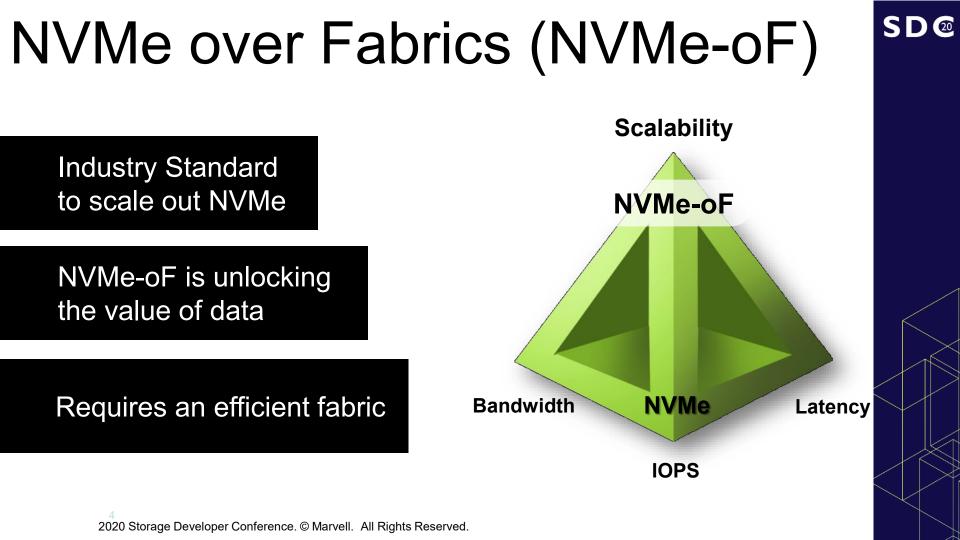
Nishant Lodha Director of Technologies, Marvell

Emerging Use Cases for NVMe-oF

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- Background and Motivation
- Use Cases by Fabric
- Accelerating Deep Learning
- Scaling Hyperconverged Infrastructure
- Key Takeaways

NVMe-oF Background



Scaling our NVMe Requires a (Real) Network

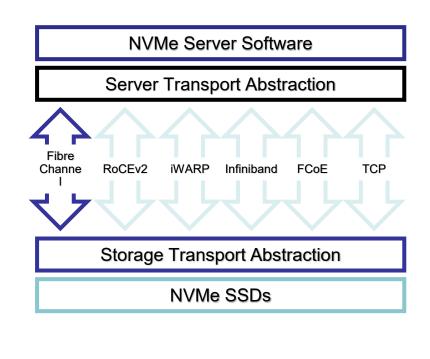
- Many options, plenty of confusion
- Fibre Channel is the transport for the vast majority of today's all flash arrays

FC-NVMe Standardized in Mid-2017

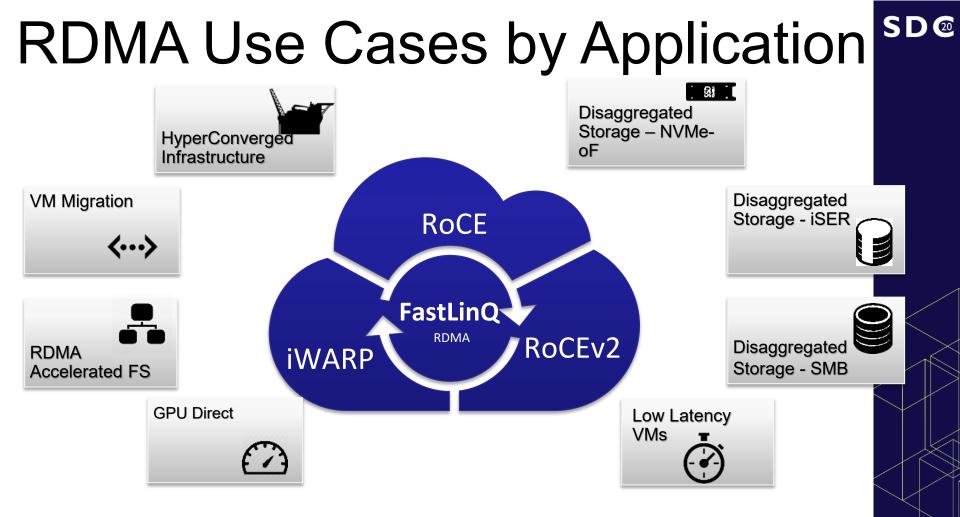
 RoCEv2, iWARP and InfiniBand are RDMA-based but not compatible with each other

NVMe-oF RDMA Standardized in 2016

 NVMe/TCP – is here! Standardized in NOV2018

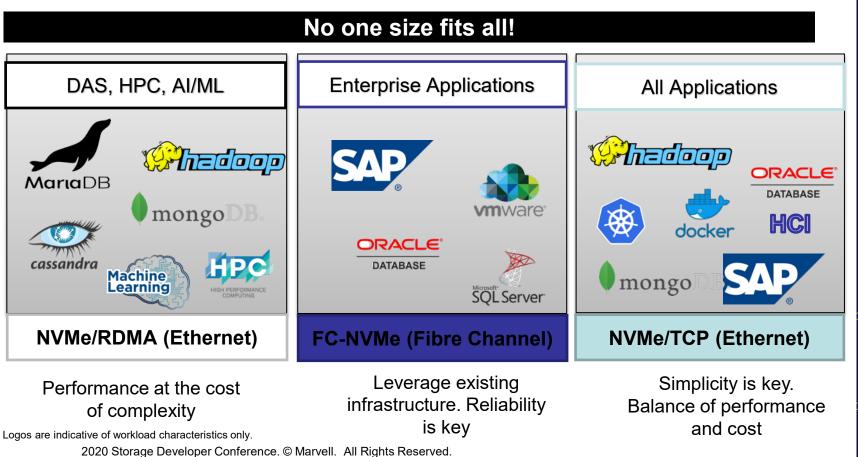


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SD@ NVMe-oF[™] RoCE – Limited Use Cases Small Scale, Contained and Well Managed Use Cases **RNIC** Upgrade Keeping the network Not Automatic Required 'lossless' **Not Precise** RDMA/OEFD Creates **Islands** expertise Not for everyone Congestion **Skillset Requirements Backward Compatibility** NEW SKILLS

Use Cases by Fabric



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NVMe-oF: NVMe/TCP

NVMe over TCP

- What: Defines a TCP Transport Binding layer for NVMe-oF
- Promoted by Facebook, Google, Intel, Marvell etc.
- Not RDMA-based, Standardized on 2018
- Why:
 - Enables adoption of NVMe-oF into existing datacenter IP network environments that are not RDMAenabled

Accelerated NVMe over TCP

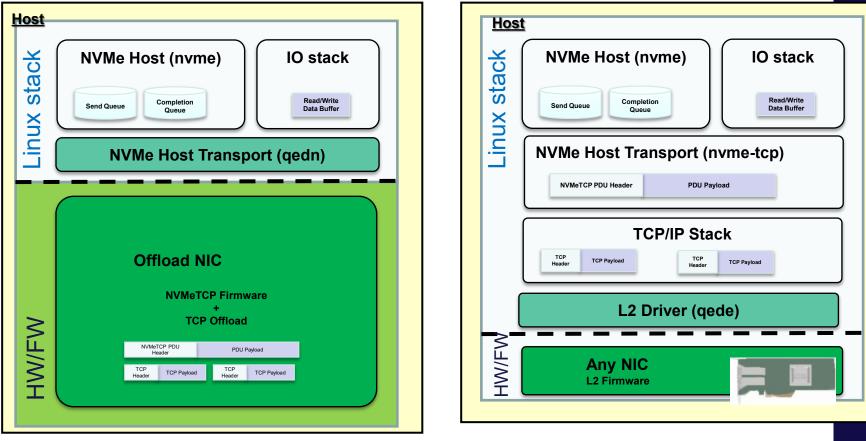
- Addresses Scalability and Congestion challenges with RDMA
- Enables adoption of NVMe-oF into existing datacenter IP network environments

KEY BENEFITS

- Ultimate Flexibility NVMe/TCP and NVMe/RDMA
- Exceptional Performance and full offload NVMe/TCP
 - Simplicity of TCP with RDMA like performance

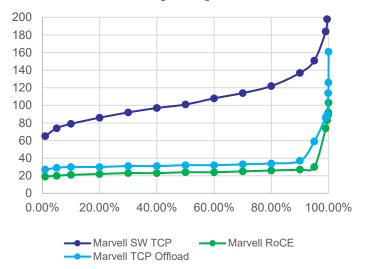
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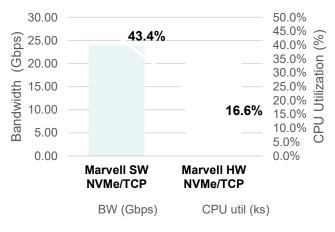


Reducing Latency, Freeing up CPU

4K Read IO - queued latency [usec]



NVMe/TCP Throughput Read 8KB, 16Jobs, 8QD SD @



fio --name=single-rd --rw=randread --bs=4k --time_based --refill_buffers --numjobs=16 --iodepth=16 --direct=1 --invalidate=1 --fsync_on_close=1 --randrepeat=0 --norandommap -group_reporting --ioengine=libaio --runtime=30 --filename=/dev/nvme0n1 --ramp_time=1 --rate_iops=12500

Use Cases for AI/ML

An Al Breaks the Writing Barrier

A new system called GPT-3 is shocking experts with its ability to use and understand language as well as human beings do

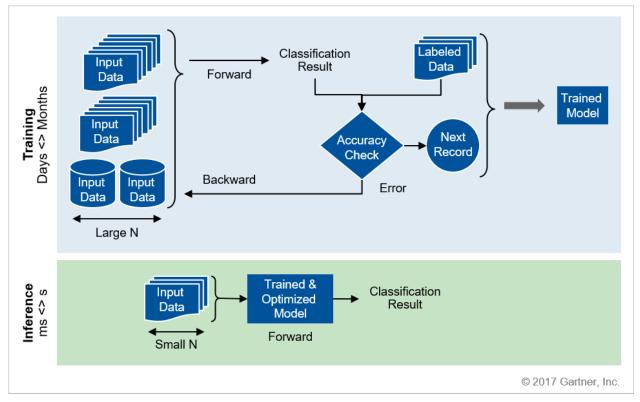
By David A. Price • Aug. 22, 2020 12:01 am ET

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Deep Learning

• Two step – Training (continuous) and Inference



Where and What

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Compute and Storage intensive
 In Datacenter

 On or off Prem

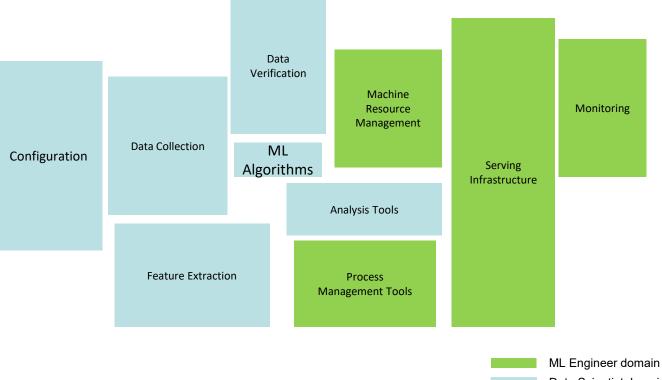
 Large Clusters of CPUs, GPUs and NVMe

 Latency sensitive
 In Datacenter
 Smaller set of CPUs/GPUs or mix
 At the edge
 CPUs, ASICs, FPGAs

Machine Learning IT Landscape **HYBRID/PUBLIC** low-latency inference inference training Compute Edge Device Edge Machine Deep Data Learnin Learning training inference low-latency inference PRIVATE

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Machine Learning Pipeline



Based on: Hidden Technical Debt in Machine Learning Systems, Scully, D., et al.

Data Scientist domain

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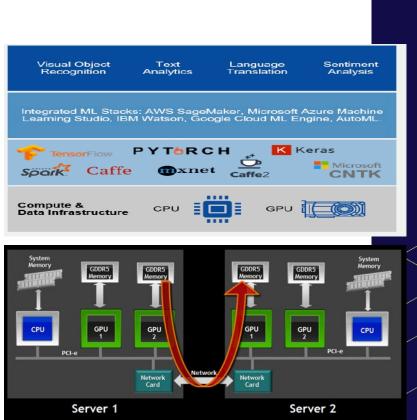
GPU Storage Bottleneck

ML training datasets typically far exceed GPU's local RAM capacity, creating an I/O chokepoint that analysts call the GPU storage bottleneck.

- Al and ML systems end up waiting, and waiting, to access storage resources their massive size impeding timely access and thus performance.
- Al and ML applications involve accessing a large number of small files from many GPU servers, deploying a parallel distributed file system as the storage infrastructure becomes a necessity
- NVMe-oF provides GPUs with direct access to an elastic pool of NVMe, so all resources can be accessed with local flash performance. It enables AI data scientists and HPC researchers to feed far more data to the applications so they can get to better results faster.

GPU to GPU Communications

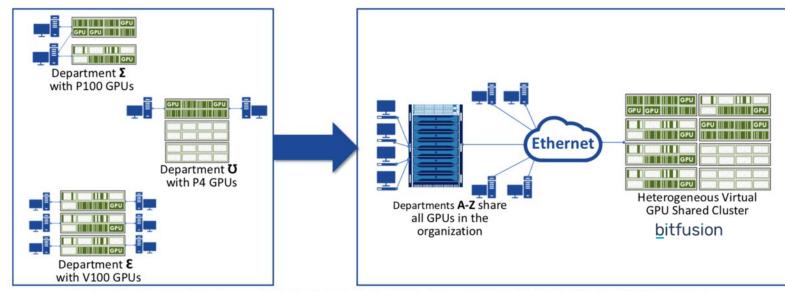
- 100GbE NIC on NVIDIA
 DGX-1 Compliant Servers
 - Supports GPUDirect RDMA
 - Lower latency
 - Higher throughput
 - Lower CPU utilization
 - Integrated with NCCL 2.0 (Inter-node Communication over RDMA)
 - Supports distributed TensorFlow / Horovod



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Scaling out GPUs over the Network

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From scattered, underutilized, non-optimized and uncoordinated GPUs deployment to Unified, Virtualized and Elastic GPU cluster

Delivers a high-performance data platform for deep

- learning, with performance on par with locally resident datasets
- GPU Direct technology enable direct GPU to GPU communication over RoCEv2
- Customer can reduce SKUs by disaggregating storage

Deep Learning Storage Optimization with NVMe-oF Visual Object Recognition Text Language

- **Problem: Captive \$torage in Deep Learning nodes**
- Solution: Storage Pool on EBOF vs. Captive per blade/R&T storage
 - A RDMA fabric provides excellent scalability for CNNs

acks: AWS SageMaker, Microsoft Azure IBM Watson, Google Cloud ML Engine Spork CNTK Compute & Data Infrastru NVMe-oF FBOF (DL datasets, models

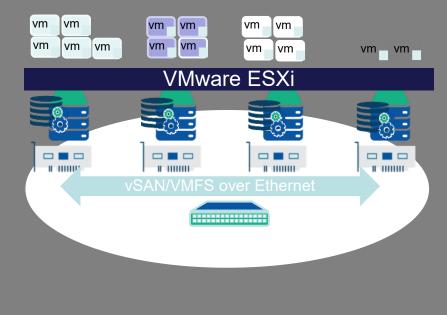
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Sentiment Analysis

Use Cases for HCI

VMware vSAN

- With ESXi 6.X/7.X running NVMe over Ethernet
- Key Benefit low CPU utilization, futureproof configuration
- I/O Capabilities require
 - 10GbE or 10/25GbE
 - SR-IOV, VXLAN, N-VDS(E)
 - Storage offload
 - NVMe/RoCE, NVMe/TCP



* Based on internal Marvell tests

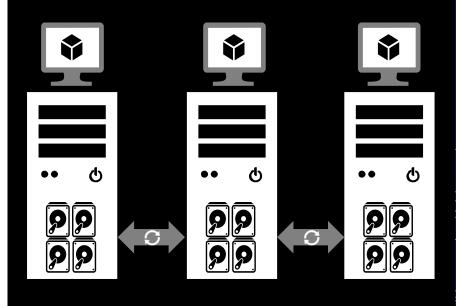


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Microsoft Storage Spaces Direct SD®

- What is Storage Spaces Direct (S2D)?
 - Software-defined storage / HCI
 - Highly available and scalable
 - Storage for Hyper-V and Private Cloud
 - Industry standard hardware servers, storage, networking
 - RDMA (RoCE or iWARP) network as storage fabric
 - 10/25GbE

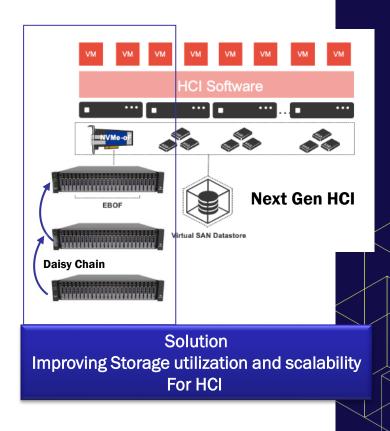
Storage Spaces Direct Cluster



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Scalable HCI

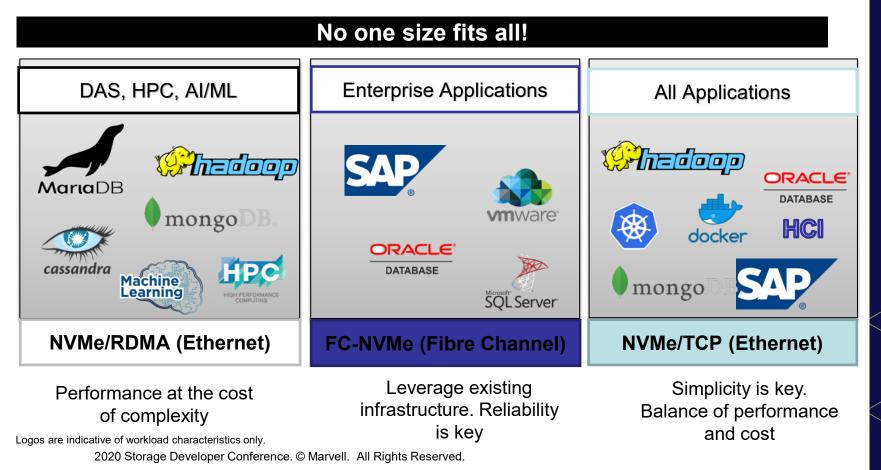
- Problem: Storage and Compute tied to the hip
- Solution: Shared compute-less storage
 - Marvell NVMe-oF NIC ⇔ EBOF's
 - Utilization: compute nodes are dedicated to run VMs/applications (lower overhead to manage in-node storage [DAS])
 - Scalability: greater networking and storage efficiency with reduced intra-node traffic in the cluster
 - Daisy chaining of EBOFs for scale up deployments
 - Next generation HCI fabrics being enabled to consume external EBOF
 - iWARP for easy deployment and lower OpEx



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Many Applications, Many Choices



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