

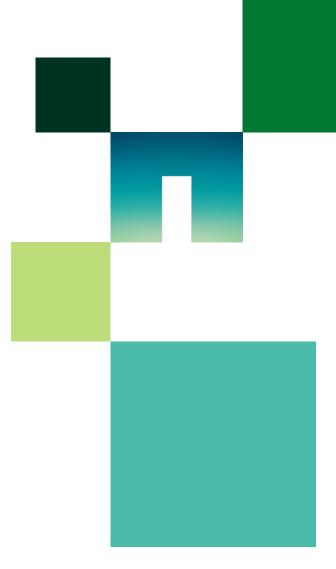
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# Unlocking Business Transformation with an AI Center of Excellence

NetApp, in partnership with NVIDIA

Santosh Rao, NetApp; Tony Paikeday, NVIDIA;

April 2, 2020



## Agenda

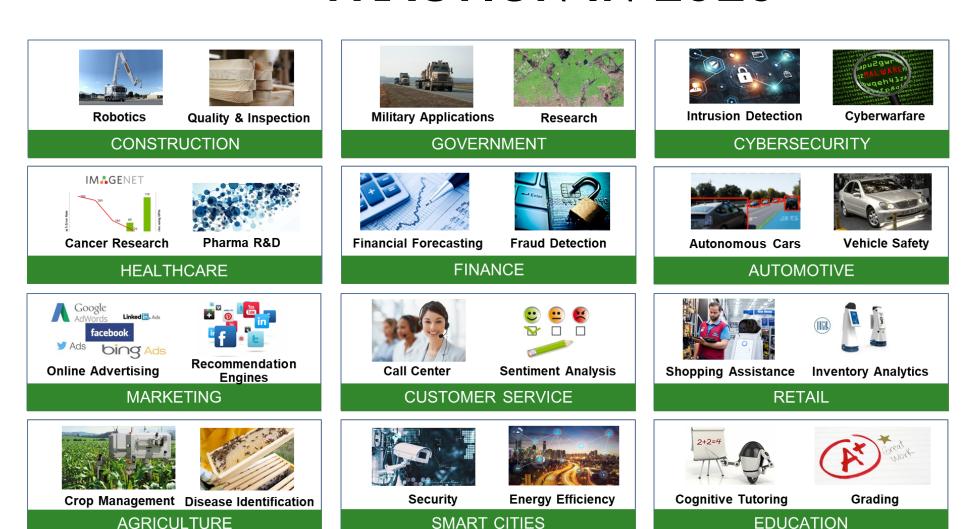
- 1) Al Use Cases in 2020
- 2) Why Al Infrastructure Matters
- 3) Enabling AI with an Integrated Data Pipeline
- 4) How to Get Started





## Al Use Cases in 2020

# MORE AI INDUSTRY APPLICATIONS GAIN TRACTION IN 2020



# CREATING SMARTER, SAFER CITIES

AnyVision builds safer cities with high-speed, real-time recognition from surveillance video streams and the ability to detect 115M individuals in 0.2 seconds.

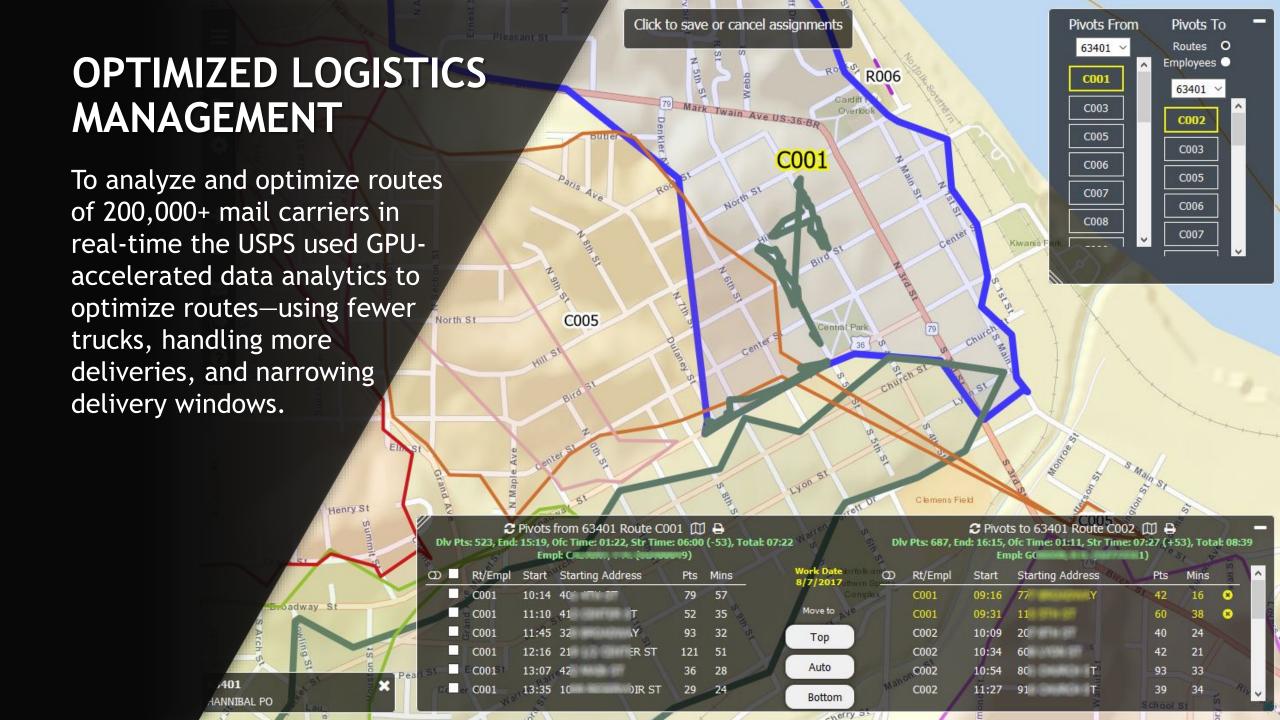




Avitas Systems predicts its solution will decrease annual inspection costs by 25%, reduce maintenance downtime by 15%, and increase inspection turnaround time by 25%.

The robots can handle the heat and use infrared cameras and chemical and other sensing technologies to inspect assets under dangerous conditions and keep production running.







# Why AI Infrastructure Matters

## DIFFERENT ROLES, SAME GOALS

Everyone wants the best AI tools—nobody wants to design/build it

**Data Scientists** 



**Data Engineers** 



IT Management



#### PLAN / CODE / BUILD / TEST / DEPLOY / OPERATE / MONITOR

What if you could iterate on models much faster than today?

What if you could automate reproducibility?

What if your DL/ML projects could get deployed 3 months faster?

# THE CASE FOR AN (IT-LED) AI INFRASTRUCTURE

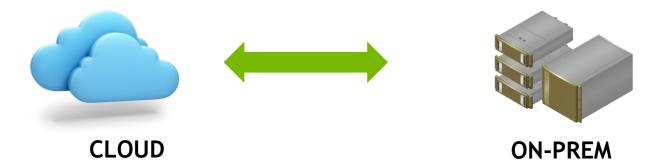
#### The Benefits of AI Centers of Excellence (CoE)

- 1. Build communities of practice
- 2. Centralize AI talent pipeline
- 3. Consolidate experience in going from (AI ideas)  $\rightarrow$  (PoCs)  $\rightarrow$  (Production)
- 4. Reduce CapEx and accelerate ROI through infrastructure centralization
- 5. Democratize AI across the company, accessible to every product, service and supply chain



## DO I BUILD IT ON PREM OR IN THE CLOUD?

Understanding how each supports development workflow



- Early exploration
- Limited access to capital/budget
- Modest datasets already native/local to cloud provider
- Fewer experiments / slower pace of experimentation

- "Deep learning enterprise"
- Requires "GPU-ready" data center
- Large datasets local to onpremises
- Frequent experiments (often in parallel), rapid pace

# FACTORS TO WEIGH

Data Gravity, sovereignty and security

Maintaining lowest cost per training run

Ensuring ability to fail fast, learn faster



## THE 3RD OPTION YOU NEED TO CONSIDER

Bringing AI training closer to the nexus of clouds and data

#### IF YOU:

- Don't have an Already data center
- Are budgetchallenged in updating yours
- Need an affordable
   OpEx model for Al optimized facilities

## Colocation services for Al infrastructure



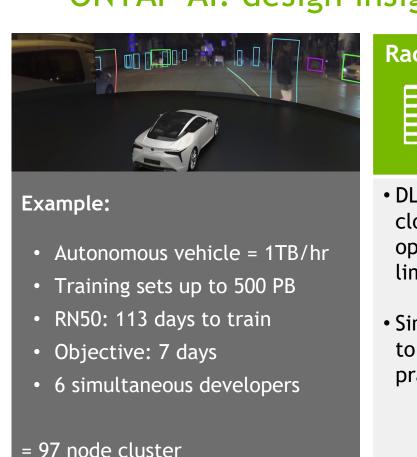
#### **BENEFITS:**

- Already optimized for Alinfrastructure
- Many already have their data sets residing at colo
- Faster deployment, less CapEx
- Low-latency, high BW, direct-connect to major clouds
- Cloud-like ease, performance of on-prem



## "A-HA" MOMENTS IN AI INFRASTRUCTURE

ONTAP AI: design insights gained from deep learning data centers

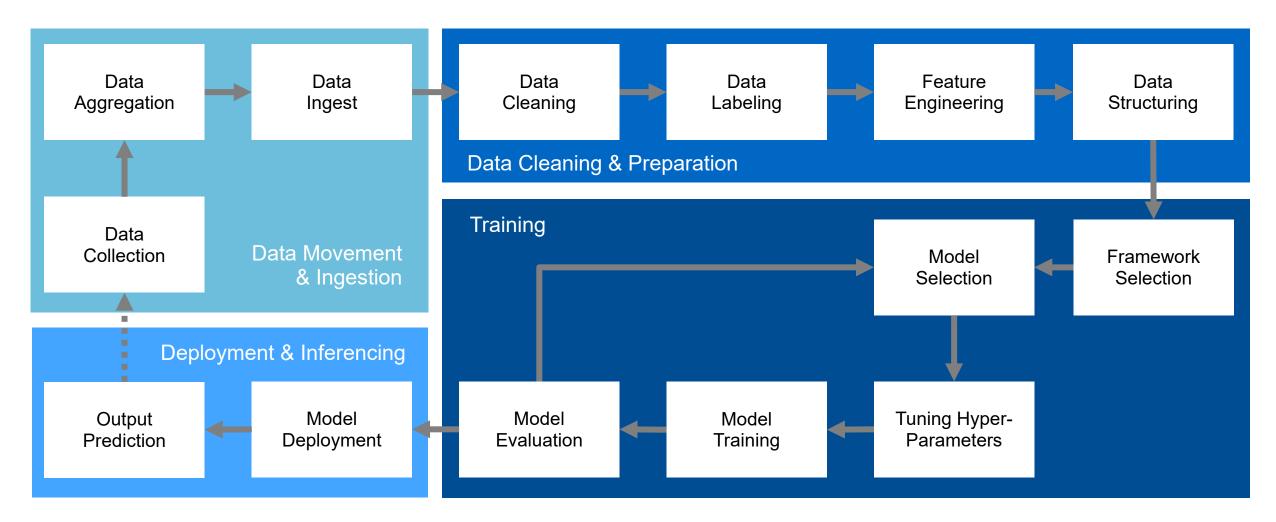


		-		
Rack Design	Networking	Storage	Facilities	Software
<ul> <li>DL drives         close to         operational         limits</li> <li>Similarities         to HPC best         practices</li> </ul>	<ul> <li>•100G EDR or</li> <li>100GbE</li> <li>preferred</li> <li>• High-</li> <li>bandwidth,</li> <li>ultra-low</li> <li>latency</li> </ul>	<ul> <li>Datasets range from 10k's to millions objects</li> <li>Terabyte levels of storage and up</li> </ul>	<ul> <li>Assume         higher watts         per-rack</li> <li>Higher         FLOPS/watt         = DC less         floorspace         required</li> </ul>	• Scale requires "cluster- aware" software



# What Is an Al Pipeline?

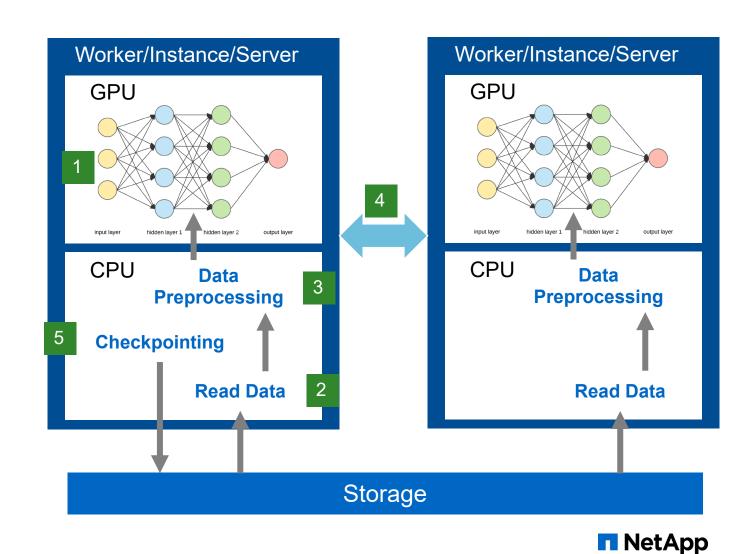
### Four phases for AI and ML pipelines



#### Training is the dominant phase in the pipeline

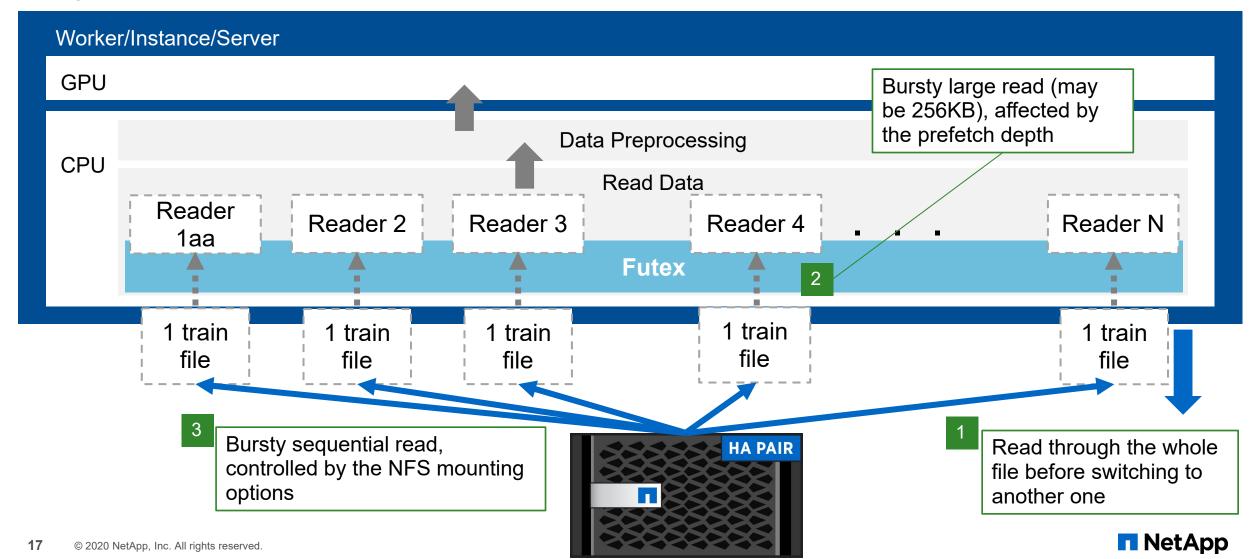
#### A closer look at the training phase

- Select a model and a combination of Hyper-parameters
- 2. Read a batch of examples from the storage
- 3. Process the example before feeding into GPU
- 4. Sync the learned model after processing a batch of examples
- 5. Save the latest learned model for a certain period of time
- 6. Repeat from step 2 until accurate

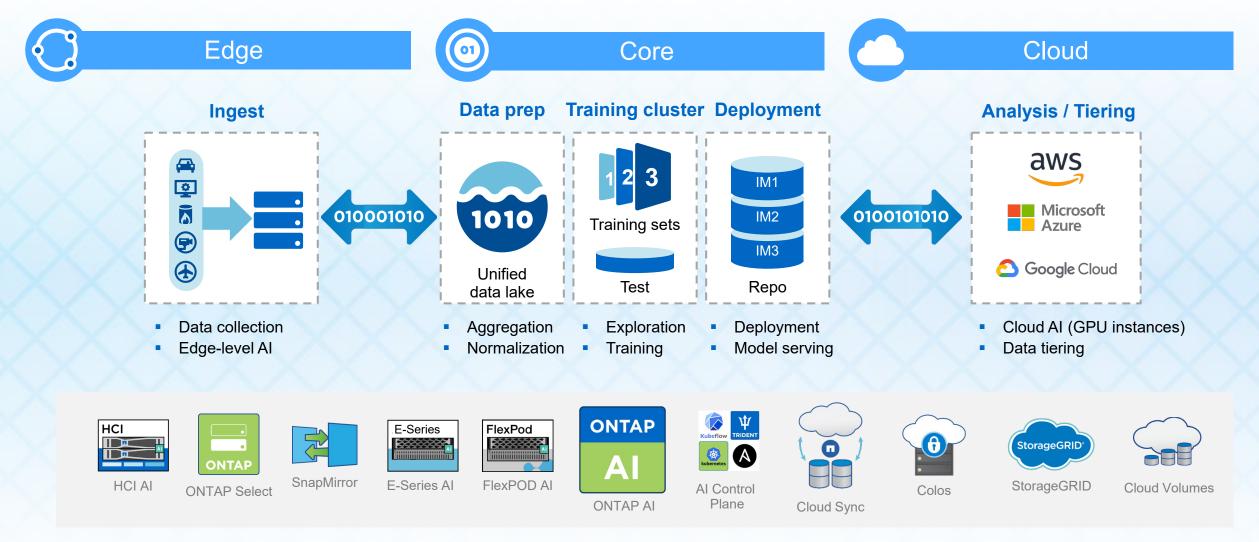


### Data reading from the storage

Using Tensorflow as an example



## Edge to core to cloud data pipeline for Al







# Automate Your Al Data Pipeline

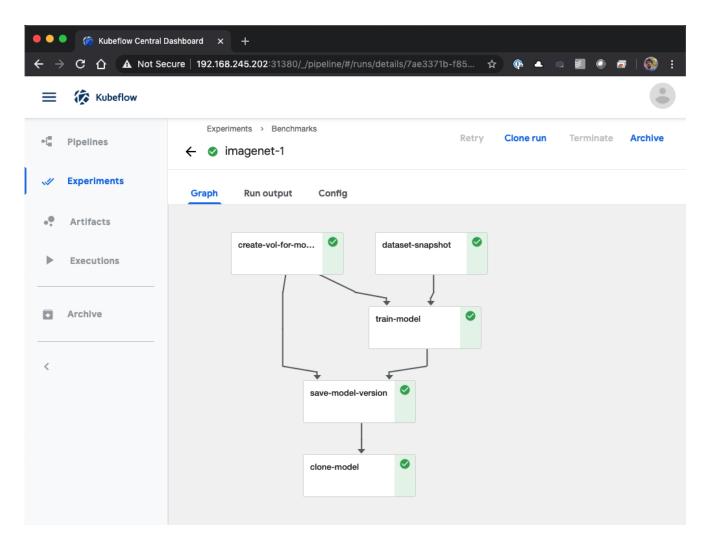
**Control Plane Solutions** 



## Automated data prep, training, Dev/Test, and deployment

#### Automate your AI data pipeline

- Kubeflow Pipelines: Platform/standard for defining and deploying portable and scalable AI/ML workflows
  - Python SDK–familiar and comfortable for data scientists
- Example pipeline steps
  - 1. Create new volume for storage of model
  - Create NetApp® Snapshot™ copy of existing dataset volume (for traceability)
  - 3. Execute containerized AI/ML training job
  - Create NetApp® Snapshot™ copy of model volume (versioned model baseline)
  - Create clone of model volume for testing





#### Example pipelines

Plug and play workflow automation

- Al training run with automatic traceability and versioning
  - Plug in data prep, training, and validation commands, then press play!
- Create an exact copy of production data for a development workspace
  - Create near-instantaneous, space-efficient copy of production dataset(s)
  - Experiment without fear of "messing up" production

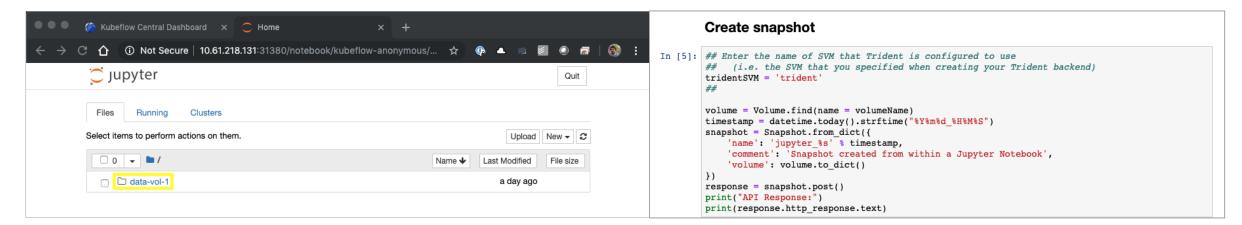
Pipelines		
Filter	pipelines	
	Pipeline name 个	
	ai-training-run	
	create-data-scientist-workspace	



#### On-demand data scientist/developer workspaces

Instant data accessibility; facilitating rapid experimentation

- Jupyter Notebooks
  - Wiki-like documents that contain live code and descriptive text
  - Widely used in AI/ML community as means of document, storing, and sharing projects
- NetApp AI Control Plane implementation
  - Petabytes of data accessible from within familiar interface
  - Protect production data while still making it accessible to Data Scientists for experimentation
  - Trigger NetApp® Snapshot™ copy creation from within notebook for dataset/model versioning/baselining

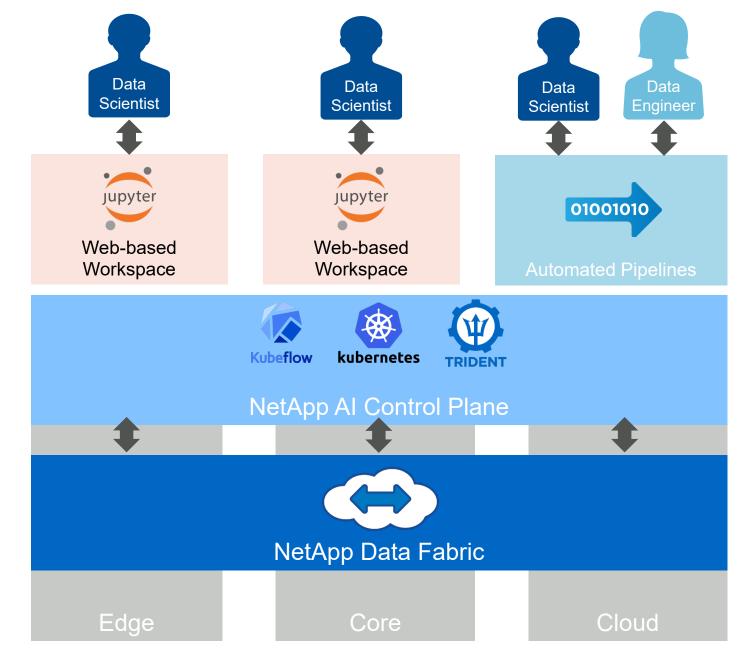




#### The bigger picture

Full-stack AI data and workload management

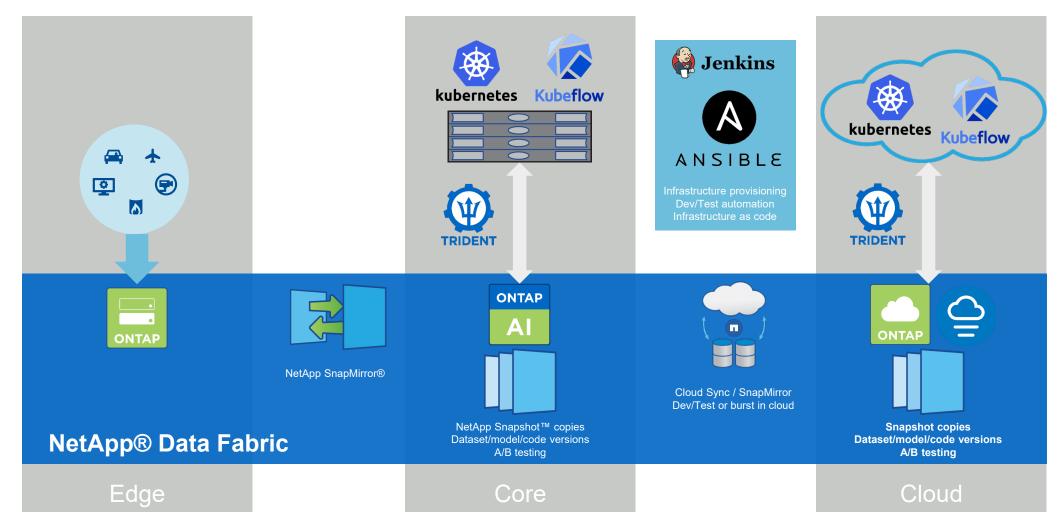
- On-demand Jupyter Workspaces
  - Full access to production datasets
- Automate data prep, training, and deployment workflows
- Workspaces and workloads can span edge, core, and cloud
  - Choice of any compute and/or cloud
  - Cross-site Data Scientist collaboration
- Built-in versioning
  - Full dataset to model traceability
  - Seamlessly switch between model versions for dev/test, A/B testing, etc.





#### Flexible architecture

Data and workloads are available whenever and wherever they are needed

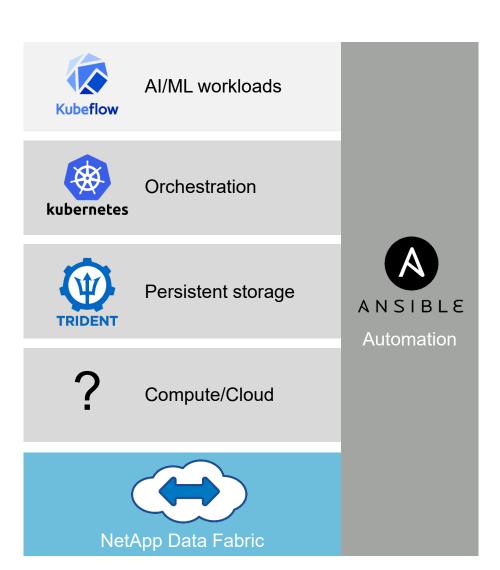




### Fully open stack

Enabling portability, scalability, automation, and simplicity

- Kubeflow = an AI/ML toolkit for k8s
  - Standard open-source platform for deploying AI/ML workloads
- Kubernetes (k8s) = container orchestration
  - Industry-standard, open-source container platform
- NetApp® Trident = storage provisioner for k8s
  - Enterprise-class storage presented in Kubernetes-native format
- Choice of compute and/or cloud
- NetApp Data Fabric = data portability and protection
  - Edge to core to cloud data movement
- Ansible = deployment automation, infrastructure as code







## **Data Mover Solutions**

## Customer requirements and challenges



Data in data lake



Data lake data into Al



Data sync between HDFS/MapRFS and NFS

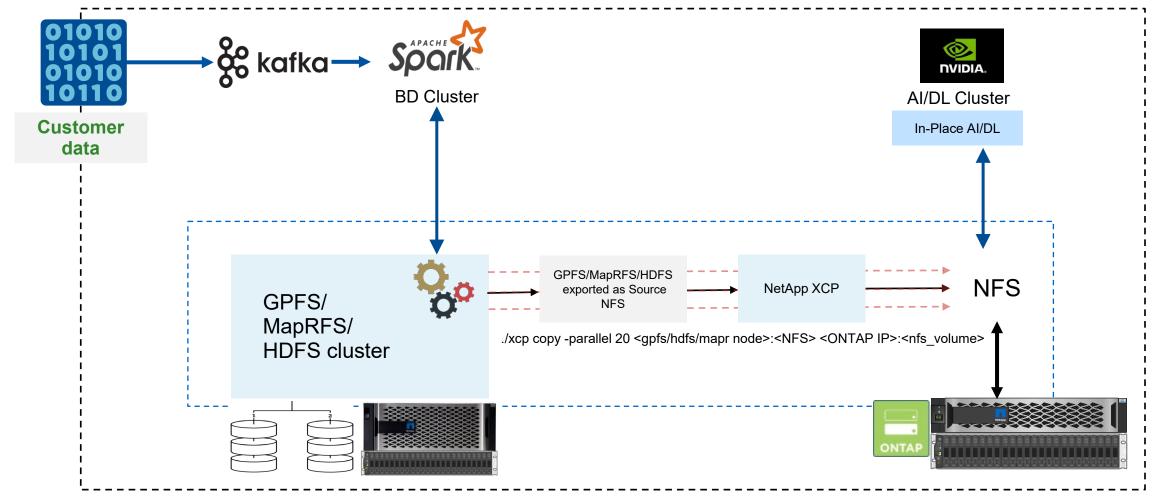


HPC (GPFS) data into Al

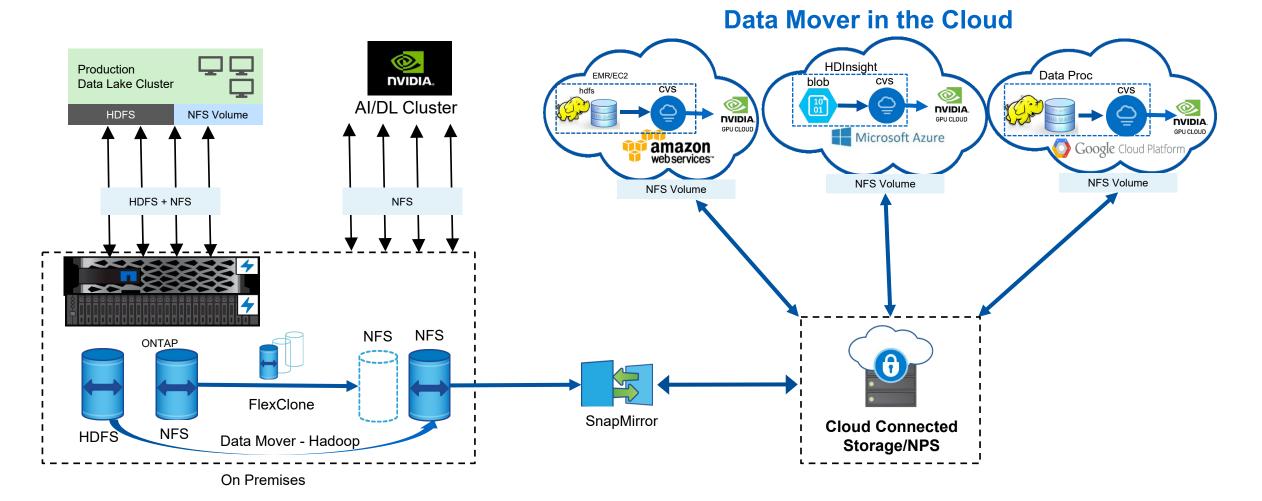


#### NetApp Data Mover Solution for AI – XCP

Extends from edge to core to cloud; federates data sources and GPUs for AI processing



#### **Data Mover Solution**





# **Getting Started**

Chart your path to success in 2020

























Cloud Volumes



#### NetApp ONTAP AI

Simplify, accelerate, and integrate your data pipeline for deep learning with NetApp, NVIDIA and Mellanox

#### Proven architecture for DL

 Powered by NVIDIA DGX-1 and DGX-2 systems and NetApp cloud-connected all-flash storage with Mellanox fabric

#### Simple to deploy

- Eliminate design complexity and guesswork
- Speed innovation and experimentation

#### Deliver performance and scalability

- Accelerate results
- Start small and grow non-disruptively

#### Build an integrated data pipeline

- Intelligently manage your data from edge to core to cloud
- Backed by AI expertise and single point of contact support





# Don't let infrastructure delay your ROI for AI

#### For organizations who:

- Want to deploy Al infrastructure built on ONTAP Al
- Lack the CapEx budget to build a data center for GPUcomputing
- Need access to Al-ready data center facilities - NOW
- Need an affordable OpEx model for hosting their ONTAP AI investment



## **ONTAP AI-Ready Data Center**

A suite of AI infrastructure solutions from partners

- Get a world-class Al-ready data center now
- De-risk deployment with a Test Drive
- Leading-edge infrastructure without CapEx

No data center? No problem!

ONTAP AI Hosting



Co-location services for customer-owned ONTAP AI infrastructure

Try it now!
Buy it if you love it!

ONTAP AI Test-Drive



Kick-the-tires using coloprovided ONTAP AI infrastructure Cloud-like ease + on-prem performance

ONTAP AI as-a-Service



Rent ONTAP Al infrastructure from colo provider

Learn more, contact: <u>testdriveprogram@nvidia.com</u>



## Get started on your Al journey

- Schedule a time to talk to our Al experts
- Work with us to identify and prioritize use cases
- Participate in a tailored workshop
- Visit our Briefing Centers
- Sign up for a Customer Proof of Concept (CPoC)
- Learn more at <u>www.netapp.com/ai</u>



