

Connect

I servizi di inferenza per l'Al con HPE e Red Hat Openshift

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Agenda

HPE STRATEGY

HPE MACHINE LEARNING INFERENCE SOFTWARE

HPE | RED HAT OCP INFRASTRUCTURE BLUEPRINT

HPE | RED HAT OCP GPUS CONCURRENCY MODEL

Q&A

2



HPE Strategy – The key Pillars of Digital Enterprise

Edge Connect your edge

Control and harness data to innovate at the edge Data **Turn data into intelligence** make smart decisions

Cloud Create your hybrid cloud

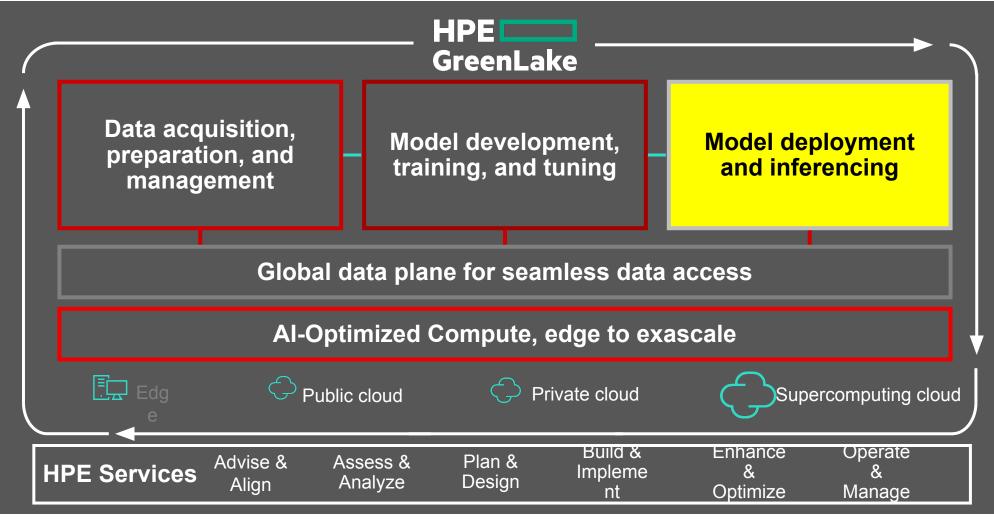
Achieve the cloud experience everywhere Securit Secure your data

Secure your data from edge to cloud

Sustainability as a catalyst



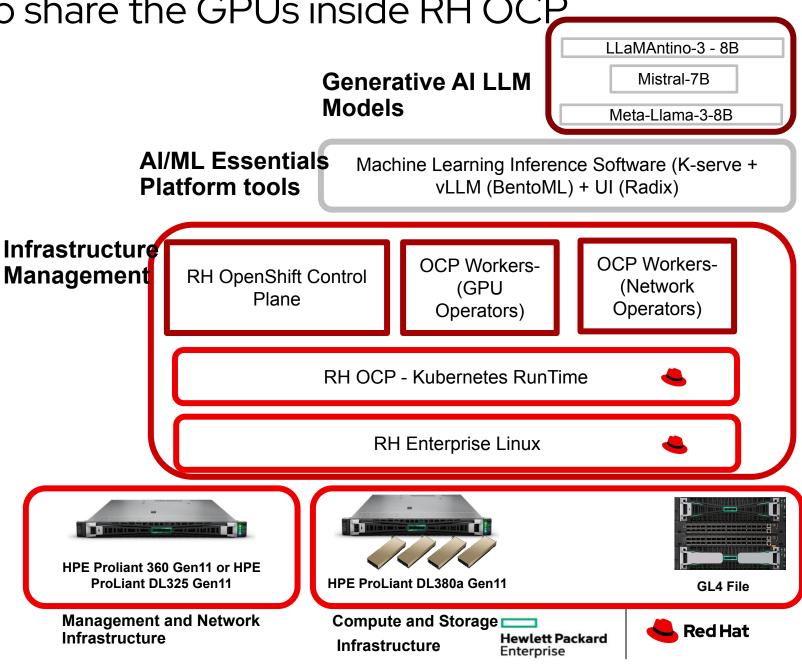
Unlock your competitive advantage with responsible AI at any scale



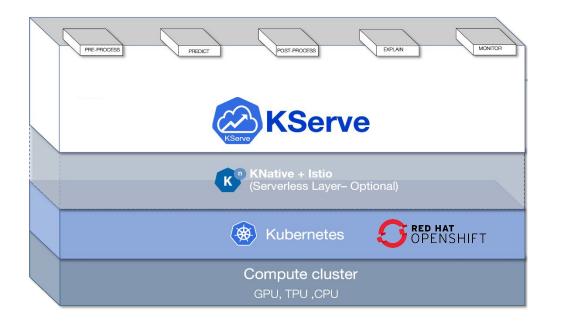


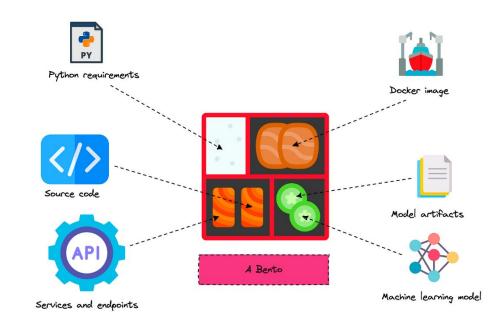
NVIDIA MIG to share the GPUs inside RH OCP

- HW Compute stack based on HPE • Proliant Gen11 architecture
- HW Network stack switch at 100GB/s based on Aruba 8325 Switch for data and 6300 Switch for Mgmt at 1GB/s
- **HW Storage** based on latest GL4File NFS/S3 standard density rack
- **Gen Al models** choice either import • from custom model or foundation models trained
- AI/ML Platform sw tools: choice of HPE MLIS (Machine Learning Inference Software)
- Services: HPE Deployment • Inference Startup Service (DIY)



HPE MLIS Open source Components





KServe

- Kubernetes-based platform for deploying models at scale
- Autoscaling, canary rollouts, and batch inferencing capabilities

BentoML

- SDK for standardizing model packaging for services
- Serving standards for REST interfaces, logging, metrics
- OpenLLM Support for optimized LLM deployments
- **vLLM** tackles the bottleneck of slow LLM inference, optimizing performance



HPE Developed software Components

• UI/UX

• Interface for managing and monitoring models, services, deployments, access tokens.

Security and authentication

- User management
- Auth integration and access token management

Deployment APIs

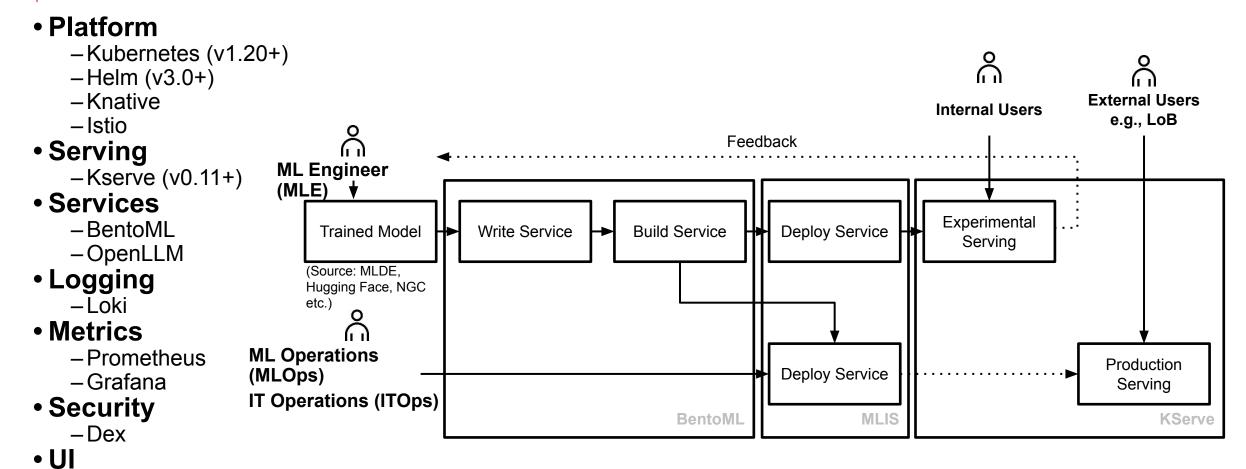
- Reduce Kubernetes deployment friction
- CLI and Python-native calls

Inferencing databases

- (Optionally) Capture data predictions
- Integrated logging
- Metrics and Operations
- LLM deployment and support



HPE MLIS Stack



-Radix



Inference Service Deployment for an LLM in 5 steps: 1) connect to a registry Connect to existing registries: **Connect to your model registry** Add new registry

Select model

Configure Resources

Configure Scaling

Interact with your model

– NVIDIA NIM (NGC)

- OpenLLM (Hugging Face)
- AWS S3 Bucket
- Minio Registry

ridd new registry		
A registry is required in ord deployments. Learn how to		
Name		
hugging-face		
Туре		
OpenLLM		~
Endpoint		
https://huggingface.co		
HuggingFace token		
huggingface token		
	Cancel	Create registry



Inference Service Deployment for an LLM in 5 steps: 2) select model from a registry

Connect to your model registry

Select model

Configure Resources

Configure Scaling

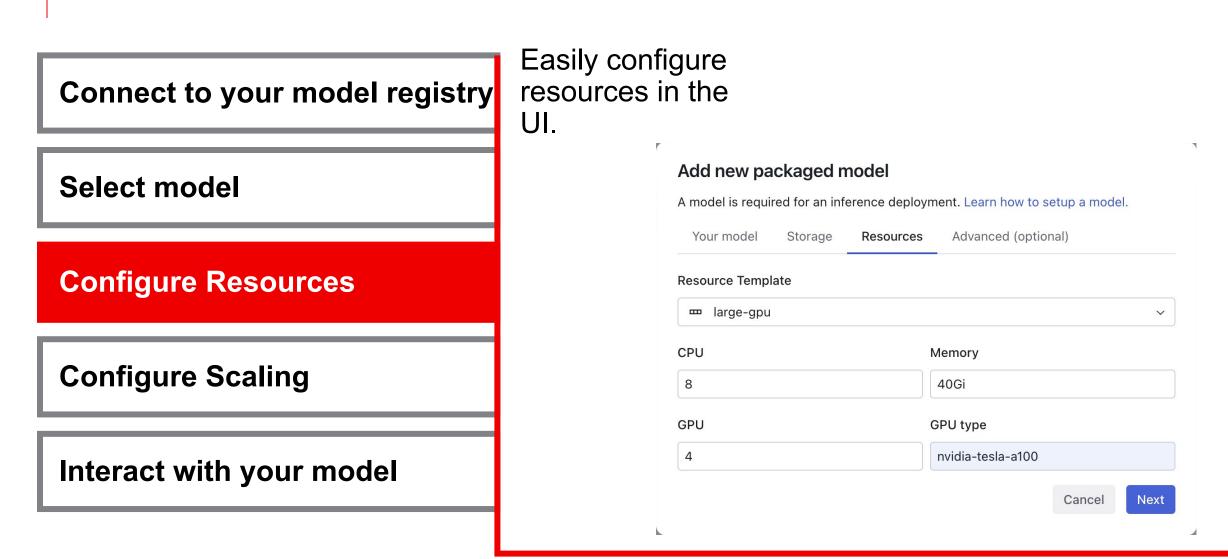
Interact with your model

Select a model from the registry.

NousResearch/llama-2-7b-hf	
mistral	
HuggingFaceH4/zephyr-7b-alpha	
HuggingFaceH4/zephyr-7b-beta	
mistralai/Mistral-7B-Instruct-v0.2	
mistralai/Mistral-7B-Instruct-v0.1	
mistralai/Mistral-7B-v0.1	
mixtral	
mistralai/Mixtral-8x7B-Instruct-v0.1	
mistralai/Mixtral-8x7B-v0.1	
mpt	
mosaicml/mpt-7b	
mosaicml/mpt-7b-instruct	
mosaicml/mpt-7b-chat	
select a model	~
	Cancel



Inference Service Deployment for an LLM in 5 steps: 3) Configure model's resources





Inference Service Deployment for an LLM in 5 steps: 4) Set deployment scaling

Connect to your model registry

Select model

Configure Resources

Configure Scaling

Interact with your model

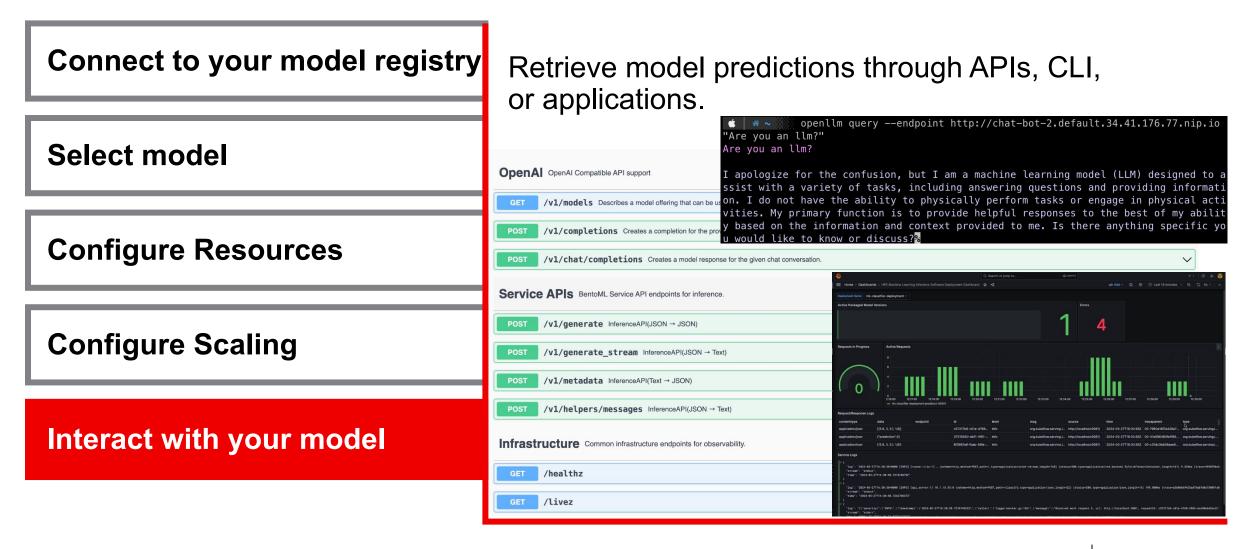
Set your
deployment to scale
according to load

A deployment is a running instance of a packaged model. Learn how to setup a deployment.

Deployment	Packaged Model	Infrastructure	Scaling	Advanced (o	ptional)
Auto scaling targ	jets template				
© scale-0-to	-8-rps-20				~
Minimum instanc	ce	Maxin	num instanc	es	
0		8			
Auto scaling targ	jet				
rps ~ 20					
				Cancel B	ack Next

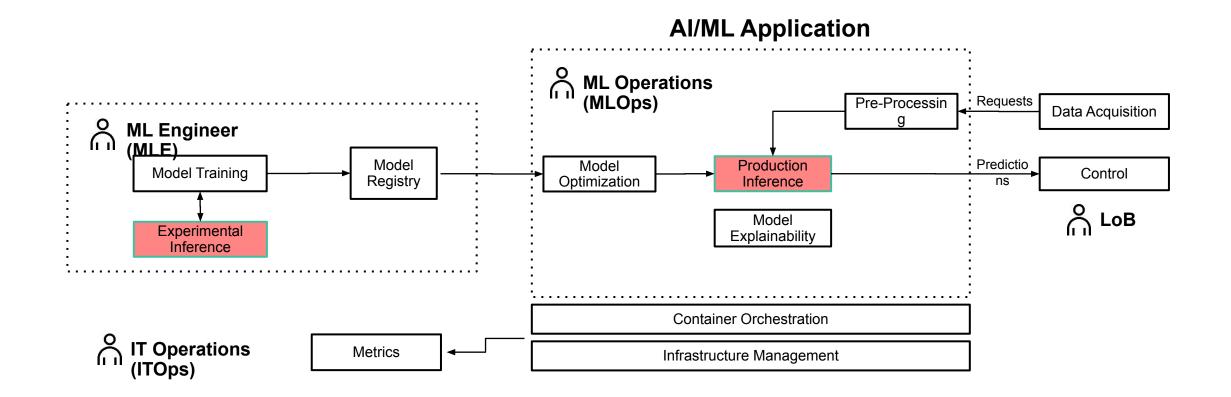


Inference Service Deployment for an LLM in 5 steps: 5) Retrieve model predictions



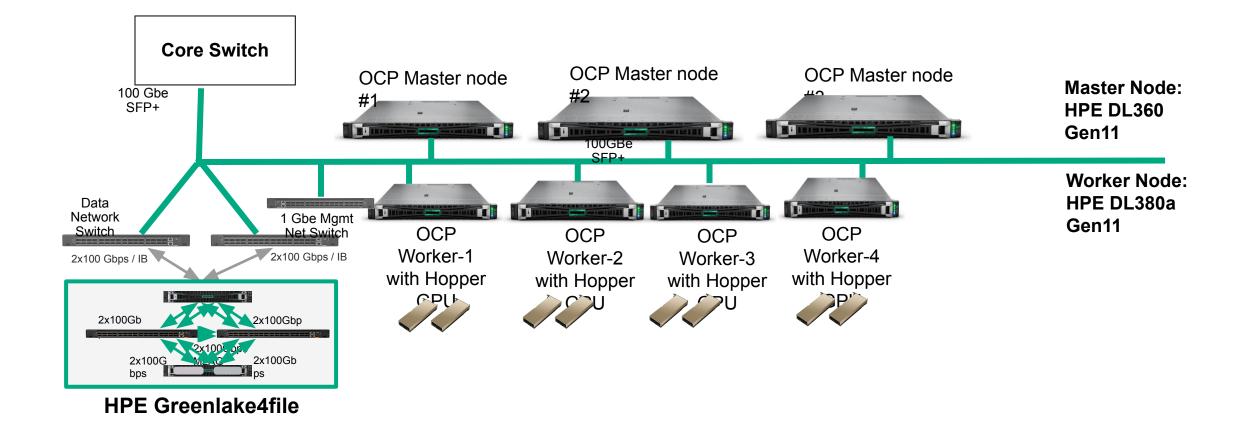


Deploying AI/ML Models into Production





Example of physical Architecture with RH Openshift

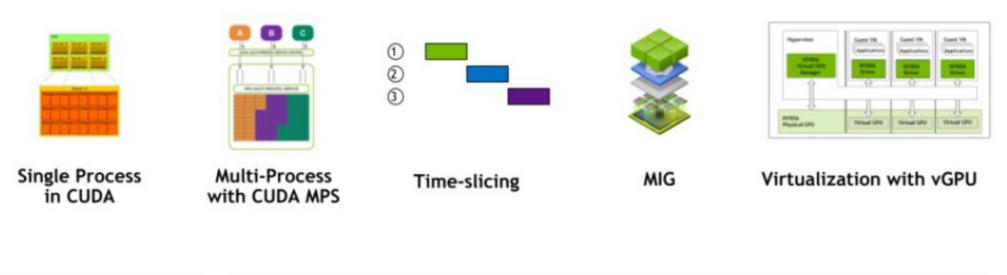




NVIDIA GPUs Concurrency choices

GPU "CONCURRENCY"

Choices







NVIDIA GPUs Concurrency choices

	Streams	MPS	Time-Slicing	MIG	vGPU
Partition Type	Single process	Logical	Temporal (Single process)	Physical	Temporal & Physical – VMs
Max Partitions	Unlimited	48	Unlimited	7	Variable
SM Performance Isolation	No	Yes (by percentage, not partitioning)	Yes	Yes	Yes
Memory Protection	No	Yes	Yes	Yes	Yes
Memory Bandwidth QoS	No	No	No	Yes	Yes
Error Isolation	No	No	Yes	Yes	Yes
Cross-Partition Interop	Always	IPC	Limited IPC	Limited IPC	No
Reconfigure	Dynamic	At process launch	N/A	When idle	N/A
GPU Management (telemetry)	N/A	Limited GPU metrics	N/A	Yes – GPU metrics, support for containers	Yes – live migration and other industry virtualization tools
Target use cases (and when to use each)	Optimize for concurrency within a single application	Run multiple applications in parallel but can deal with limited resiliency	Run multiple applications that are not latency- sensitive or can tolerate jitter	Run multiple applications in parallel but need resiliency and QoS	Support multi- tenancy on the GPU through virtualization and need VM management benefits





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Thank you



