



**Connect**

# GenAI mit Parasol AI Studio

Hands-On Day Darmstadt 18.11.2025

**Jochen Cordes**

Solution Architect  
EMEA TelCo CoE  
Red Hat

**Alessandro Arrichiello**

Solution Architect  
EMEA TelCo CoE  
Red Hat





# Jochen Cordes

Solution Architect  
EMEA TelCo CoE  
Red Hat  
<https://www.linkedin.com/in/jochencordes/>



# Alessandro Arrichiello

Solution Architect  
EMEA TelCo CoE  
Red Hat



Any Model  
Any Accelerator  
Any Cloud

# The Myth of Universal Solutions

Model choice depends on use-case, precision, performance and resource constraints

Use-case Families	Gen Models	Nongen AI	Optimization	Simulation	Rules/Heuristics	Graphs
Prediction/Forecasting	Low	High	Low	High	Medium	Low
Planning	Low	Medium	High	Medium	High	Medium
Decision Intelligence	Low	Medium	High	Medium	High	Low
Autonomous Systems	Low	Medium	High	High	Medium	Low
Segmentation/Classification	Medium	High	Low	Low	Medium	High
Recommendation Systems	Medium	High	Medium	Low	Low	High
Perception	Medium	Low	Low	High	Low	Medium
Intelligent Automation	Medium	High	Medium	High	Medium	Medium
Anomaly Detection/Monitoring	Medium	High	Low	Medium	Low	High
Content Generation	High	Low	Low	Low	Medium	Low
Conversational Interfaces	High	Medium	Low	High	Medium	Low
Knowledge Discovery	High	Medium	Low	Low	Medium	High

Color Code for Recommendation Level: L (Low): Low M (Medium): Medium H (High): High

## Avoid Hype-Driven Adoption

- Using GenAI for unsuitable cases can lead to high failure rates. Evaluate feasibility and appropriateness for each use case.

## Focus on Alternative AI Techniques

- Established techniques like ML, optimization, simulation, and rule-based systems may be more suitable and reliable.

## Combine AI Techniques for Robust Solutions

- Combining GenAI with other AI techniques can mitigate limitations like inaccuracies. Use GenAI for interfaces and rule-based systems for decision-making.

## GenAI's Limitations in Specific Use Cases

- GenAI isn't ideal for prediction, planning, decision intelligence, or autonomous systems. It's better for content generation, conversational interfaces, and knowledge discovery.

## Manage GenAI-Specific Risks

- Consider risks like output unreliability, data privacy, IP issues, cybersecurity, and regulatory compliance. Evaluate these risks for each use case.

# The Advantages of Having a Choice

Matching hardware to model needs enables faster AI operationalization and cost savings



## Performance

Different AI models perform better on specific accelerators



## Cost Efficiency

Selecting the most suitable accelerator avoids over-provisioning or under-utilizing resources leading to significant cost savings



## Innovation

As the AI landscape is evolving rapidly the ability to choose accelerators allows for adapting new hardware innovation

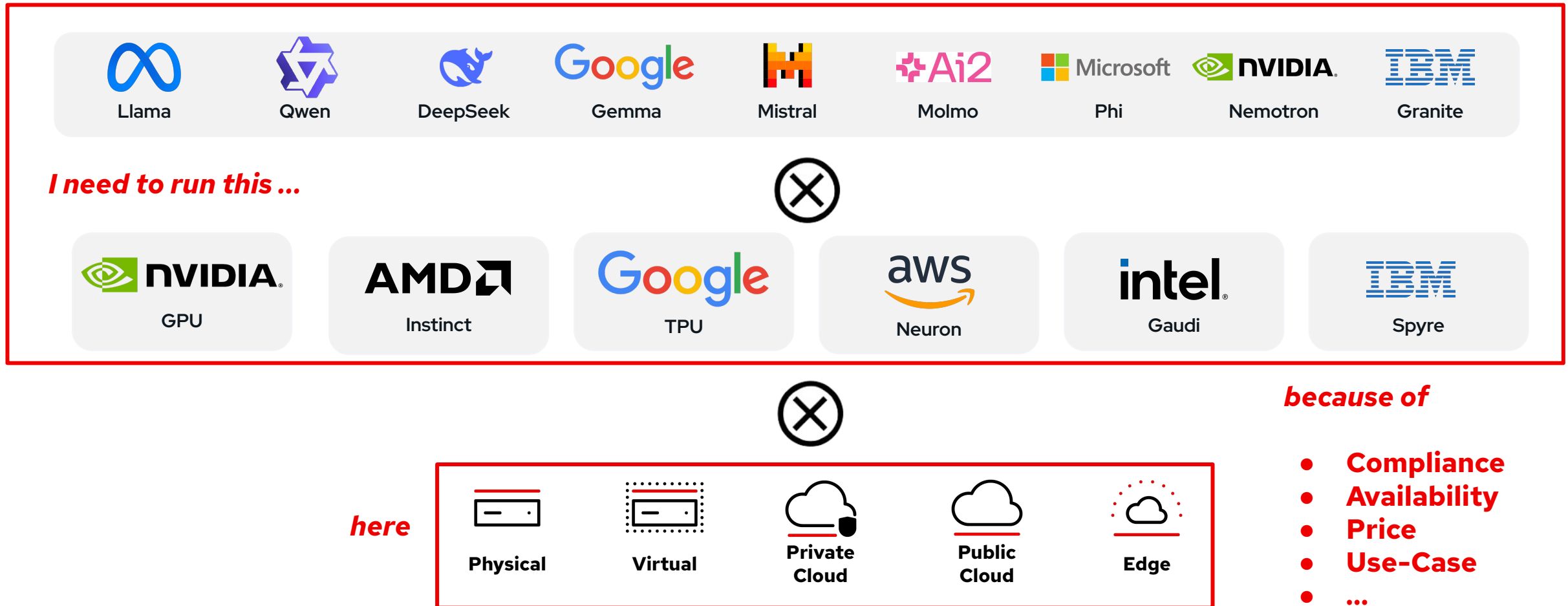


## Operationalize

Operationalize AI faster by matching the hardware to the specific needs of models

# Flexibility Meets Hybrid Cloud

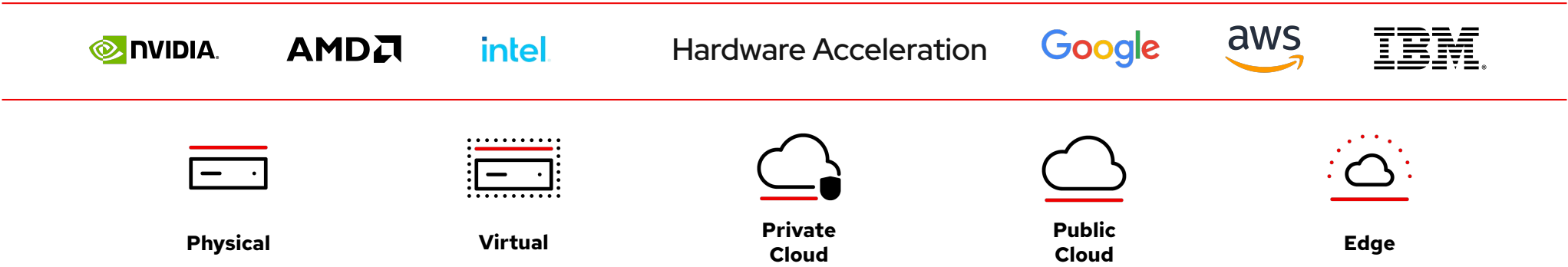
Aligning models, accelerators and cloud to your needs



- **Compliance**
- **Availability**
- **Price**
- **Use-Case**
- ...



Trusted, Consistent and Comprehensive foundation



\* NVIDIA, AMD, Intel, Google TPU supported in Red Hat AI. AWS Inferentia/Neuron IBM AIU are on our roadmap



## Red Hat AI Inference Server

### Gen AI model inference

- ▶ Packaging: Linux container
- ▶ Red Hat vLLM inference server
- ▶ Validated & optimized model repository
- ▶ LLM Compressor tool
- ▶ Certified: RHEL/RHEL AI and OpenShift/OpenShift AI
- ▶ 3rd Party Support Policy: Non-Red Hat Linux & Kubernetes platforms

I need Gen AI model Inference  
on RHEL/Linux or  
OpenShift/Kube



## Red Hat Enterprise Linux AI

### AI model inference & training

- ▶ Packaging: Linux server appliance
- ▶ Granite family models
- ▶ InstructLab model alignment
- ▶ Optimized RHEL image with integrated accelerators
- ▶ **Includes Red Hat AI Inference Server**

I need an integrated AI Linux  
server appliance for inference &  
training



## Red Hat OpenShift AI

### AI model inference, training & LLMOps

- ▶ Packaging: Kubernetes distributed cluster
- ▶ Supports Gen AI & Predictive AI
- ▶ Distributed Training, Tuning & Inference in OpenShift Kubernetes
- ▶ LLMOps & MLOps / Day 2 Mgt
- ▶ **Includes RHEL AI**
- ▶ **Includes Red Hat AI Inference Server**

I need a complete distributed AI  
platform for inference, training and  
LLMOps



## Flexible and Efficient Inference

- ▶ GA distributed inference (llm-d)
- ▶ New validated and optimized models
- ▶ vLLM enhancements
- ▶ LLM Compressor GA

## Connecting Models to Data

- ▶ Modular and extensible approach for: data ingestion, synthetic data generation, tuning, evaluations.
- ▶ RAG enhancements & partner integrations
- ▶ Continual Post Training Algorithm
- ▶ Feature Store GA



## Agentic AI

- ▶ AI experiences: AI hub and gen AI studio
- ▶ Model Context Protocol support & MCP Server access in gen AI studio
- ▶ Llama Stack API integration

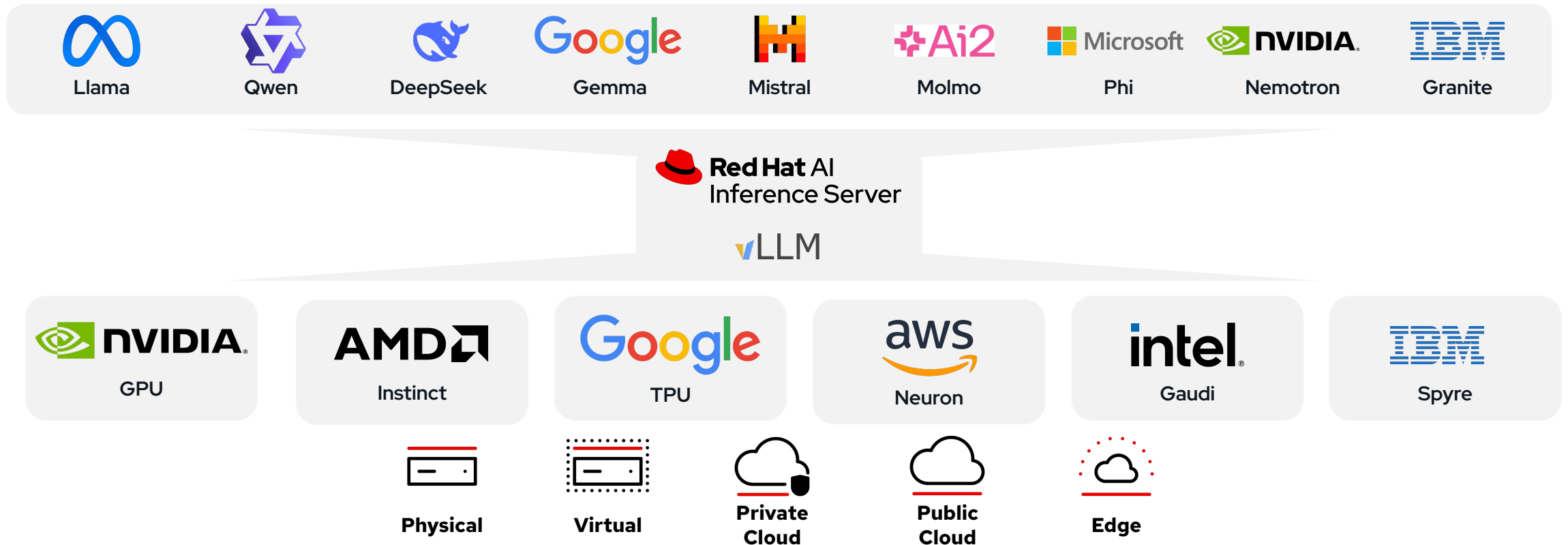
## AI Platform

- ▶ Model catalog and registry GA
- ▶ Model as a Service provider enhancements and API Mgt integration
- ▶ GPU as a Service enhancements

**Single platform to run any model, on any accelerator, on any cloud**

# Optimized Inference

vLLM connects model creators to accelerated hardware providers



Single platform to run any model, on any accelerator, on any cloud

# Red Hat AI repository on Hugging Face

A collection of third-party validated and optimized large language models

## Broad Collection of models



Llama



Qwen



Gemma



Mistral



DeepSeek



Microsoft

Phi



Molmo

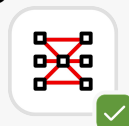


Granite



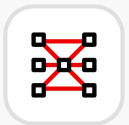
Nemotron

## Validated models



- ▶ Tested using realistic scenarios
- ▶ Assessed for performance across a range of hardware
- ▶ Done using GuideLLM benchmarking and LM Eval Harness

## Optimized models



- ▶ Compressed for speed and efficiency
- ▶ Designed to run faster, use fewer resources, maintain accuracy
- ▶ Done using LLM Compressor with latest algorithms

# Red Hat AI tooling for model optimization

Optimize and validate your choice of model



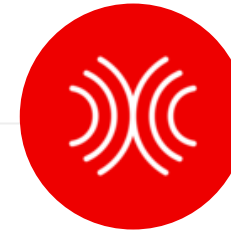
## Inference benchmarks with GuideLLM

Tool for evaluating LLM performance to guarantee efficient, scalable, and affordable inference serving.



## Accuracy evaluation with LM-eval-harness

A unified framework for evaluating the accuracy of LLMs across a variety of tasks and benchmarks.

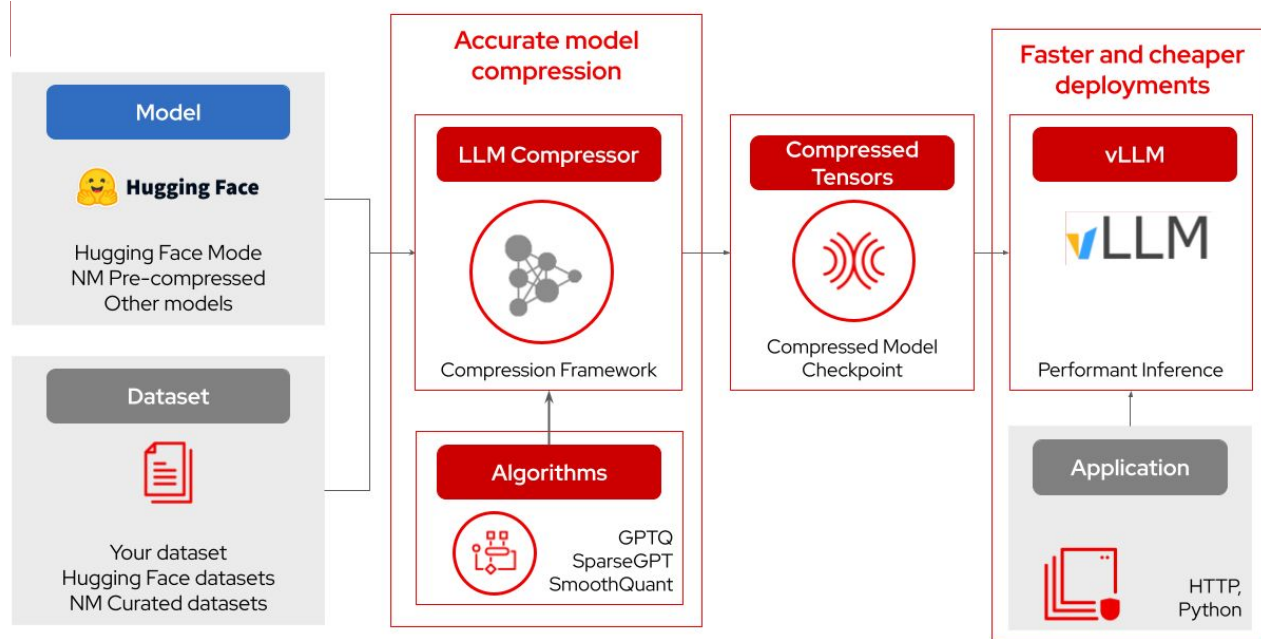


## LLM Compression tools

Framework for reducing the size and computational requirements of a LLMs while preserving accuracy

**Receive tailored capacity planning guidance from our experts**

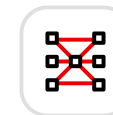
# Optimization Support with vLLM-Compressor framework



- ▶ Comprehensive set of algorithms in unified interface
  - GPTQ, AWQ, SmoothQuant
  - FP8, INT8, INT4, MxFp4 and W4A8
- ▶ Seamless integration w/ HF AutoModel
- ▶ Safetensors-based checkpoint format compatible with vLLM
- ▶ Large model support via HF accelerate

Optimize fine-tune models for inference

# Why Use Red Hat AI Compressed Models?



Delivering near-baseline accuracy and reliability through rigorous engineering and evaluation

## Exceptional Quality and Accuracy

- ▶ Achieve **near-perfect (~99%) accuracy** recovery compared to the original, uncompressed baseline.
- ▶ Derived from **intensive hyperparameter tuning**, not a simple quantization run.

## Rigorously Evaluated and Reliable

- ▶ **Evaluated on diverse, rigorous benchmarks** (Arena-Hard, etc.) to ensure baseline performance.
- ▶ **Extensive testing** provides a trustworthy and reliable model for end-users.

## Competitive Differentiation

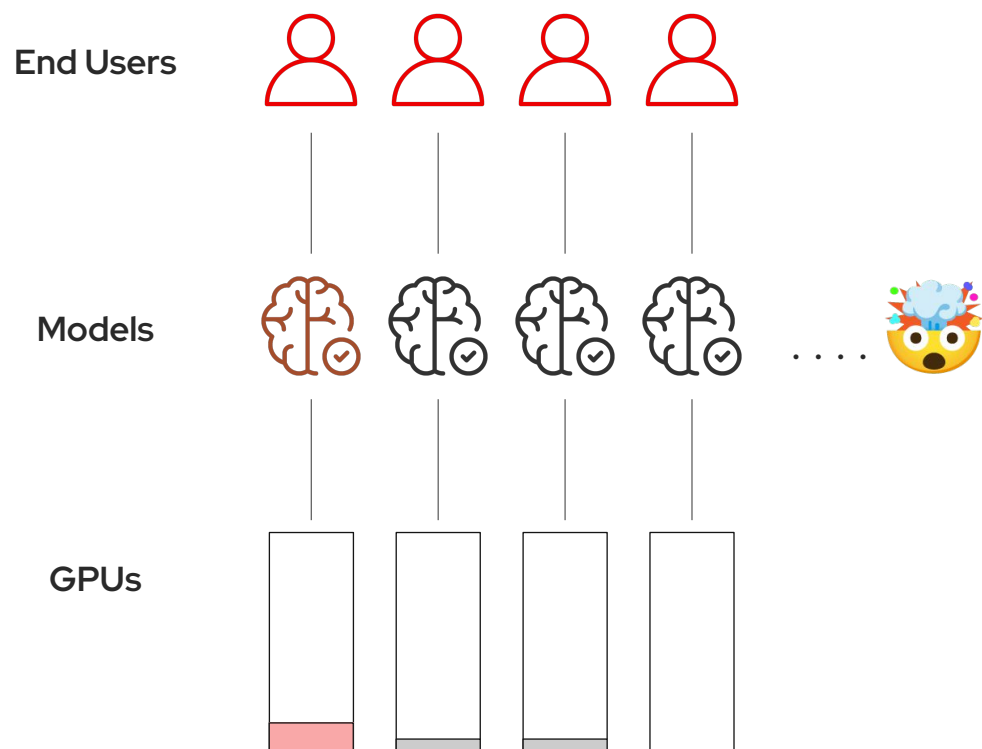
- ▶ Our **comprehensive compression tuning** is resource-intensive, requiring multiple runs for proper recovery.
- ▶ This **commitment to quality at scale** provides is key to the unique value Red Hat provides.

# Model-as-a-Service

Resources Optimization on Shared  
Inference Servers



# Infrastructure as a Service<sup>CPU & especially GPU</sup> can be costly



Self-Service is good for plentiful resources & small teams

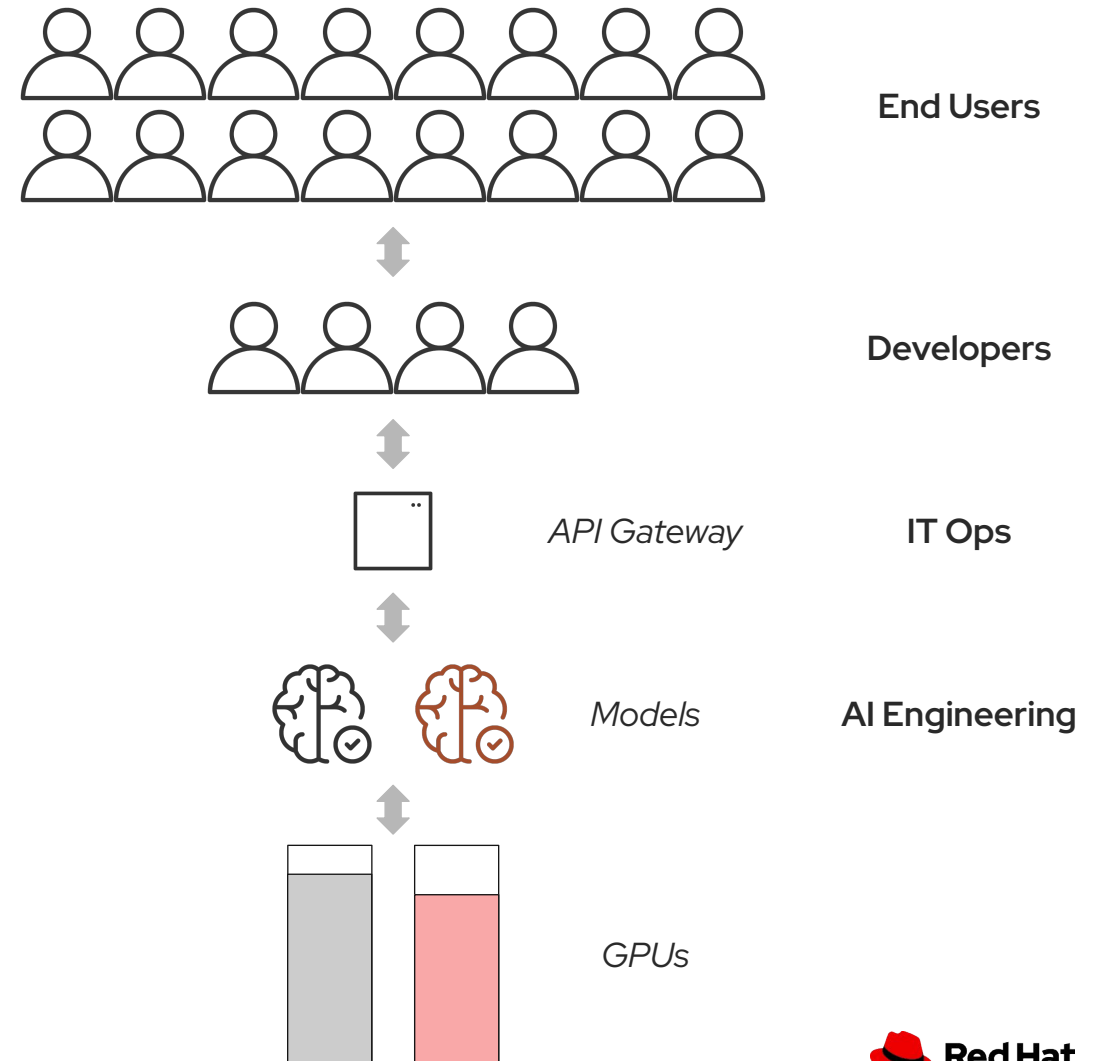
- Throwing GPUs at the problem is risky
- Few people know how to use them correctly
- Leads to duplication and underutilization
- Leads to high costs
- Most people want an LLM endpoint, not a GPU



# Models as a Service

Offering AI **models as *the* service** to a larger audience

- IT serves common models centrally
  - Generative AI focus, applicable to any model
  - Centralized pool of hardware
  - Platform Engineering for AI
  - AI management (versioning, regression testing, etc)
- Models available through API Gateway
- Developers consume models, build AI applications
  - For end users (private assistants, etc)
  - To improve products or services through AI
- Shared Resources business model keeps costs down



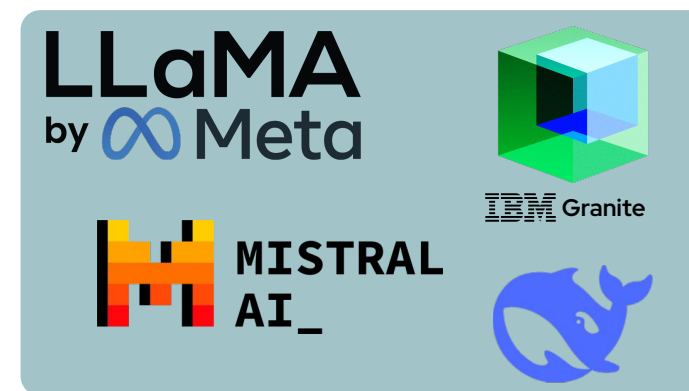
Hosted AI services are **not** the only option



**Risks & Challenges:**

- Costs at scale
- Data privacy and security policies
- IP leakage

Become the **Private AI Provider**



**Models-as-a-Service Benefits:**

- Cost effective & optimize performance
- Easy to use
- Consistent with data & security requirements

## AI Applications

### Anything LLM

- ▶ Granite 3.3 8B Instruct LLM
- ▶ Weaviate Vector Database
- ▶ Chatbot
- ▶ Document embeddings

### Continue

- ▶ VS Code with continue.dev plugin
- ▶ Granite 3.1 8B Code Instruct LLM



### Docling

- ▶ Transformation of
  - Spreadsheet
  - PDFto Markdown



### Stable Diffusion

- ▶ Stable Diffusion LLM
- ▶ Generation of images

## Models-as-a-Service



## AI Platform



# Connection Details

Wifi:

**Red Hat Summit: Connect 2025**

Password:

**redhat\_2025**



# red.ht/3JtNcSx

Replace **studentX** with your actual assigned user!





# Jetzt Session bewerten!

Einfach QR-Code scannen,  
Session aus der Liste wählen  
und bewerten. **Vielen Dank!**

[red.ht/rhsc-darmstadt-feedback](https://red.ht/rhsc-darmstadt-feedback)



# Open Tech Quest

Solve technical challenges around  
Ansible, OpenShift & RHEL as a team

```
... blueprint, p... SET_BLUE...
...um Object.assign({}, blueprint, { past: [] pr...
...ges: [] }, future: [] }); return blueprint; } } case...
...ent.id === action... blueprint... Object.assign({,
...payload.comment... BLUEPRINT_USE...
...on.payload.blueprin... int, { present: Object.as...
... blueprint.p... zation... yload.users }) } } return bluep...
...atum [ ...s... nt => { if... eprint.p... nt.id === action... blueprint...
...int.past... => { retu... Object.as... 'f', pastB... action.p...
...({}), past/... ons, { use... action.p... user }) } }
...rsion: ac... version, customiza... present
...mization... rint.future.map(...sign({}, futu...
...rint.version... sign({}, futu... T_HOST
...); } return blue... T_HOST
...eprint.id) } r... sign({},... resent: t...
...sent.custo... ne: acti... ostnam...
...D: return... int => { if...
...ueprint.p... rint => { re...
...bject.assign({}, pastBlueprint.custom/...
...blueprint.present, { version: action.r...
...action.payload.blueprint.custo...
...eprint, { v... n.payload...
...eprint.cl... hostname... sent.id...
...int => { if... sent.id... load.de...
...description... ueprint => { re...
...rint => { re...
...}), }, }, pres...
...p), future...
...cription... { compo...
...modules... action.payloa...
...if (bluep... nt.id === acti...
...vload.bl... localPer...
```

## Nicht vergessen!

Schließt euch zu Teams von  
rund 5 Personen zusammen,  
legt eure jeweiligen  
Benutzeraccounts an und  
startet dann ab 14:45 Uhr  
gemeinsam durch!

[red.ht/otq](https://red.ht/otq)





Connect

# Thank you



[linkedin.com/company/red-hat](https://linkedin.com/company/red-hat)



[facebook.com/redhatinc](https://facebook.com/redhatinc)



[youtube.com/user/RedHatVideos](https://youtube.com/user/RedHatVideos)



[twitter.com/RedHat](https://twitter.com/RedHat)

