

intel[®] ai
summit
英特爾 AI 科技論壇

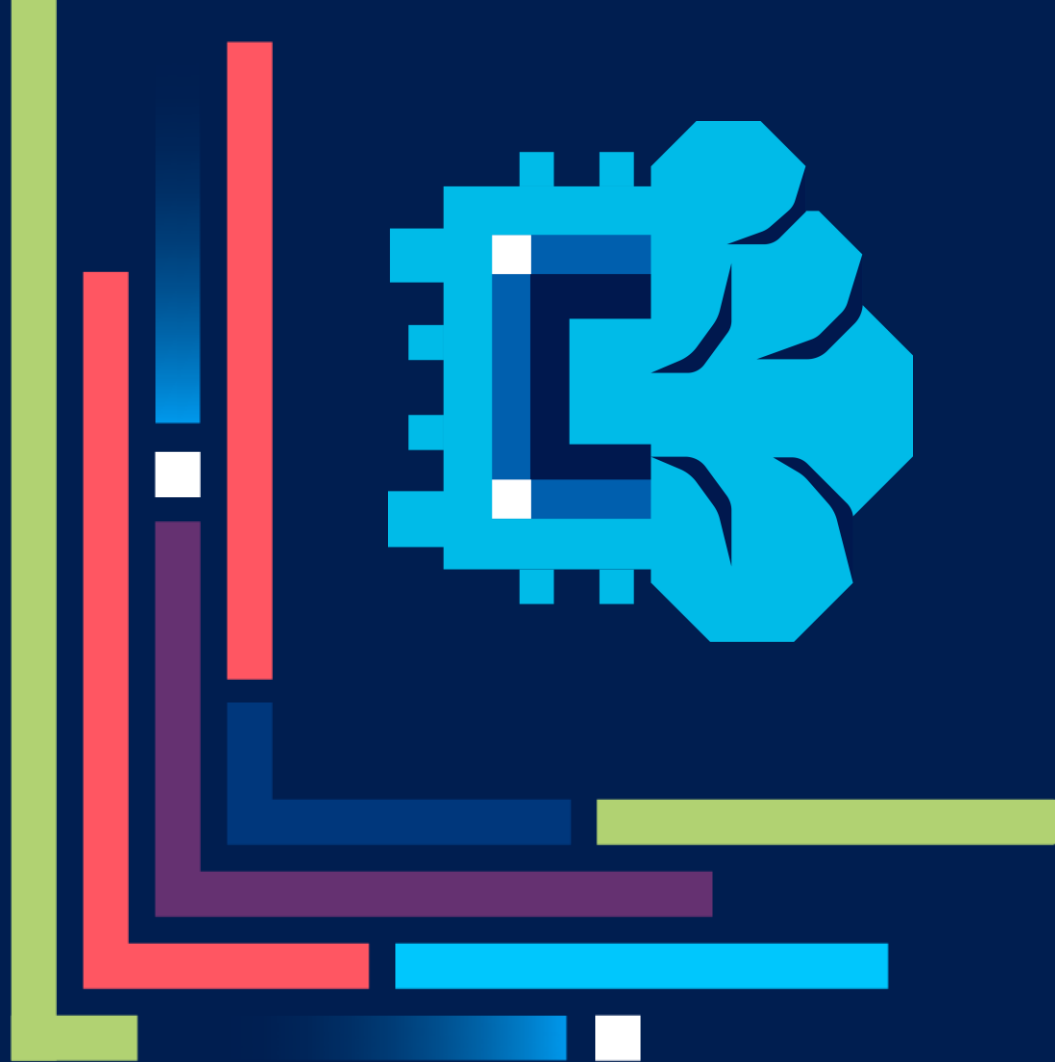
Bringing AI Everywhere

Accelerating GenAI & LLM with Intel Gaudi Solutions

Leo Zhao

AI Software Evangelist

March 27th, 2024





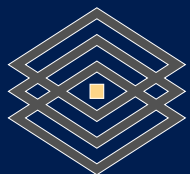
Brief of Intel® Gaudi® Solution



GenAI and LLMs require...



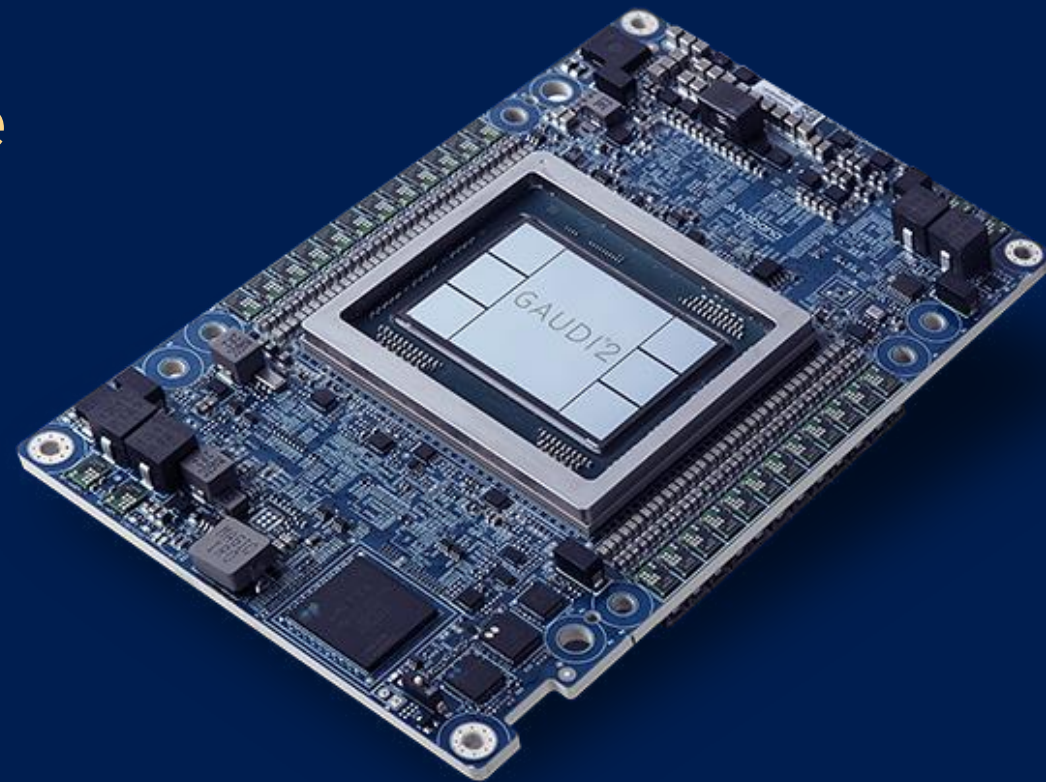
Deep learning performance



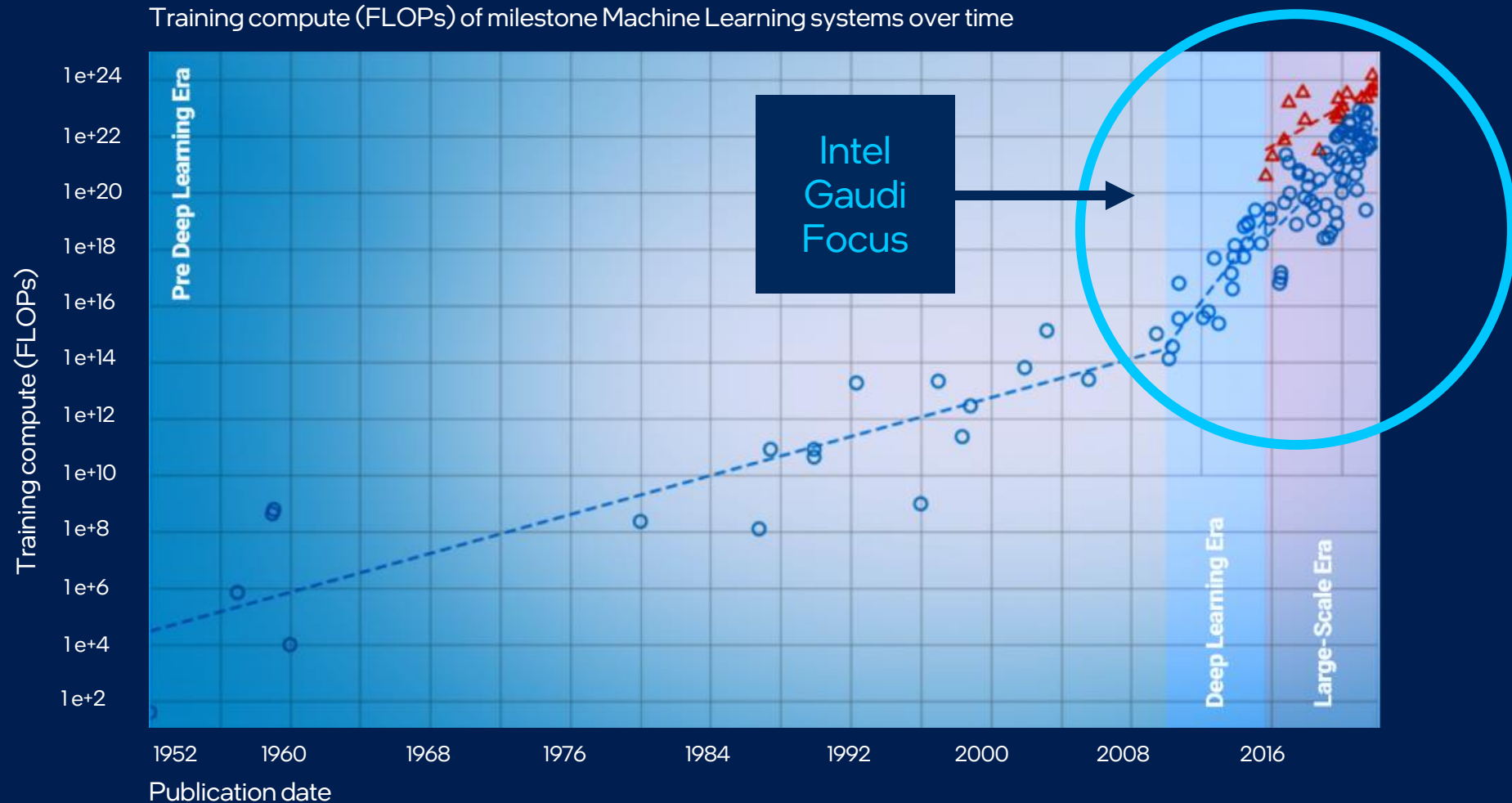
Flexible, efficient scalability



Developer ease of use



Intel® Gaudi® AI Accelerator: Built for the Deep Learning and Large-Scale Eras



Intel® AI Portfolio

Software



Deep Learning
Acceleration



Dedicated Deep Learning Training and Inference

General
Acceleration



Cloud Gaming, VDI, Media Analytics, Real-Time Dense Video

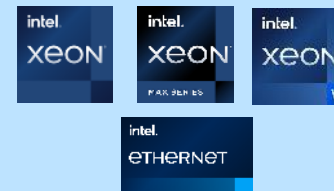


Parallel Compute, HPC, AI for HPC

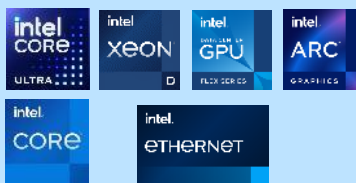
General
Purpose



Real-Time, Medium Throughput, Low Latency, and Sparse Inference



Medium- to Small-Scale Training and Fine Tuning



Edge and Network AI Inference



Client AI Usages

Intel® Gaudi® AI Accelerator Roadmap

Intel Gaudi
Accelerator



(16nm)

Available Now

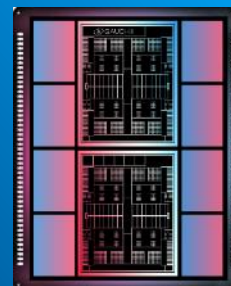
Intel Gaudi 2
Accelerator



(7nm)

Available Now

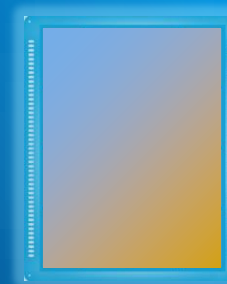
Intel Gaudi 3
Accelerator



(5nm)

2024

Intel next-gen
accelerator



2025

Intel® Gaudi® 2 Accelerator: Architected for GenAI and Large Language Models

Heterogenous AI-custom compute engine

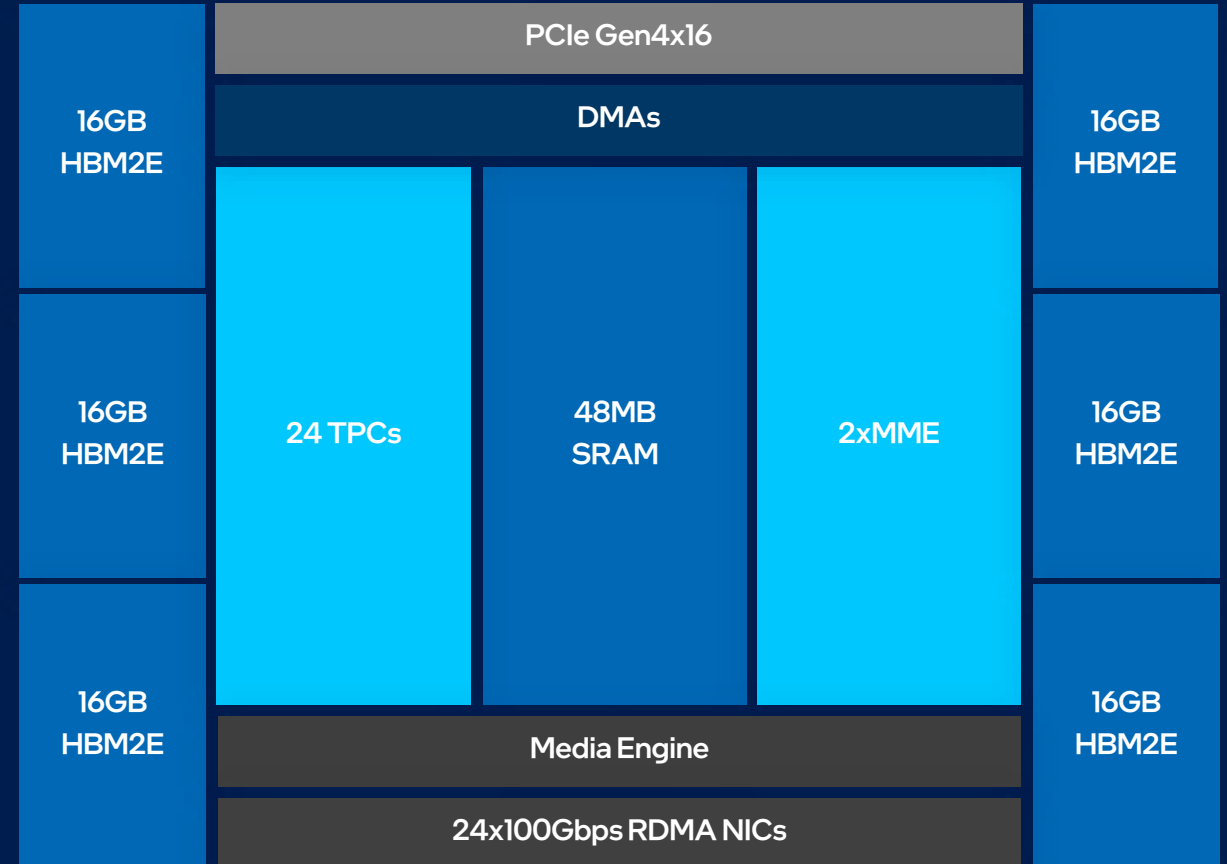
- 24 Tensor processor cores
- Dual matrix multiplication engines

Leading on-chip memory capacity

- 96 GB HBM2E
- 48 MB SRAM

Massive, flexible on-chip networking

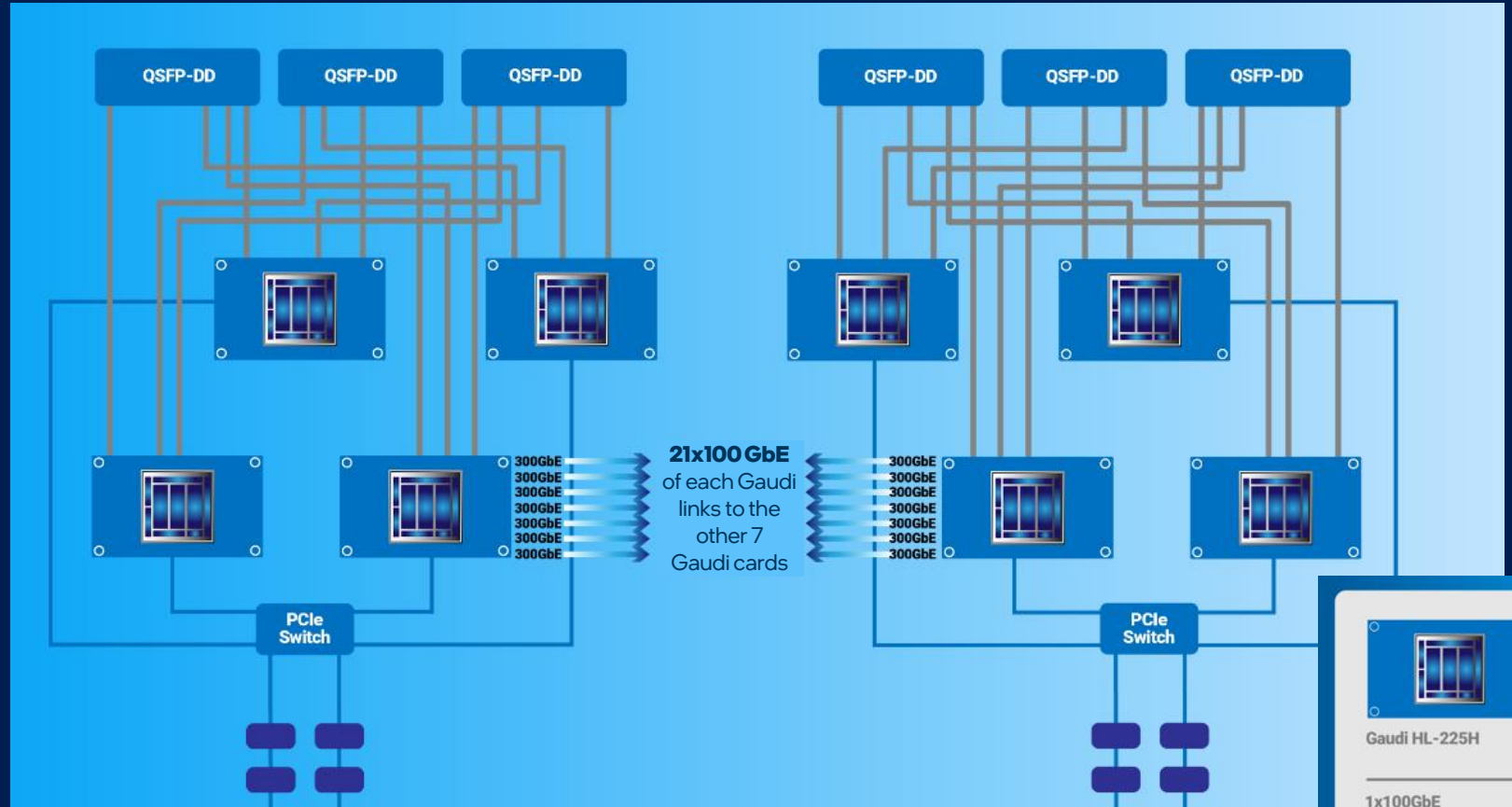
- 24x 100 GbE RoCE ports on every Intel Gaudi2 chip



Intel® Gaudi® 2 Server: Designed for Flexible, Efficient Scalability

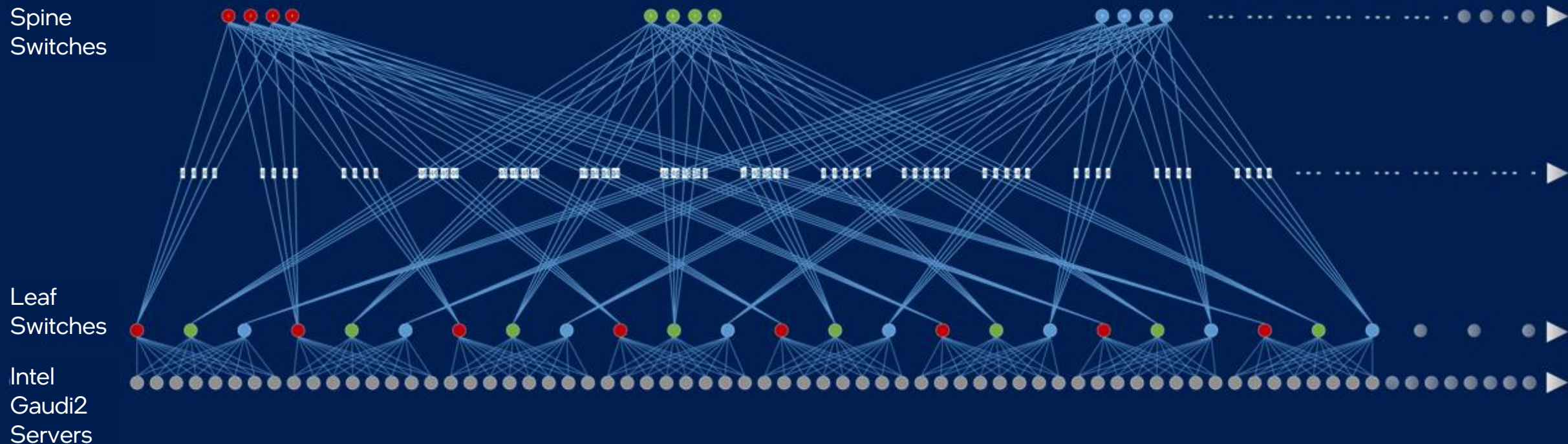
HLS-Gaudi 2 Reference Server featuring...

- 8 Intel Gaudi 2 mezzanine cards
- 24x 100 GbE ports per card
 - 21 for all-to-all connectivity to other 7 Intel Gaudi processors within the server
- Three to scale out
- Through six QSFP-DD ports
- Dual-socket Host CPU: Intel® Xeon® Scalable processor



Dual-socket Intel Xeon Scalable Processor Host CPU

Near-Infinite System Scale-Out Possibilities to Meet Demands for the Largest GenAI Models





Intel® Gaudi® Performance in MLPerf

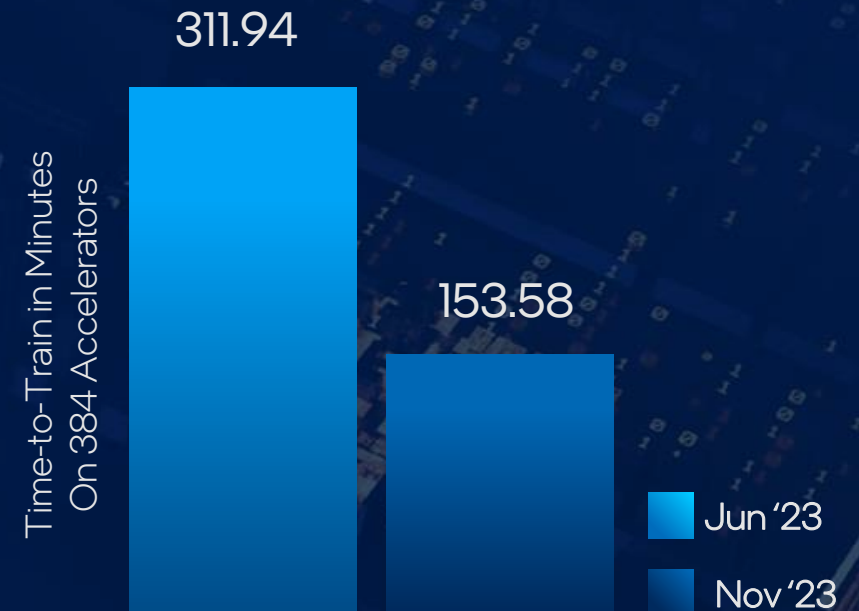
Intel® Gaudi® 2 Accelerator Performance Doubled with FP8

We projected for customers
+90% performance gain with FP8

Delivered
more than **2x** 

Committed to credible, reliable performance projections and delivering on them

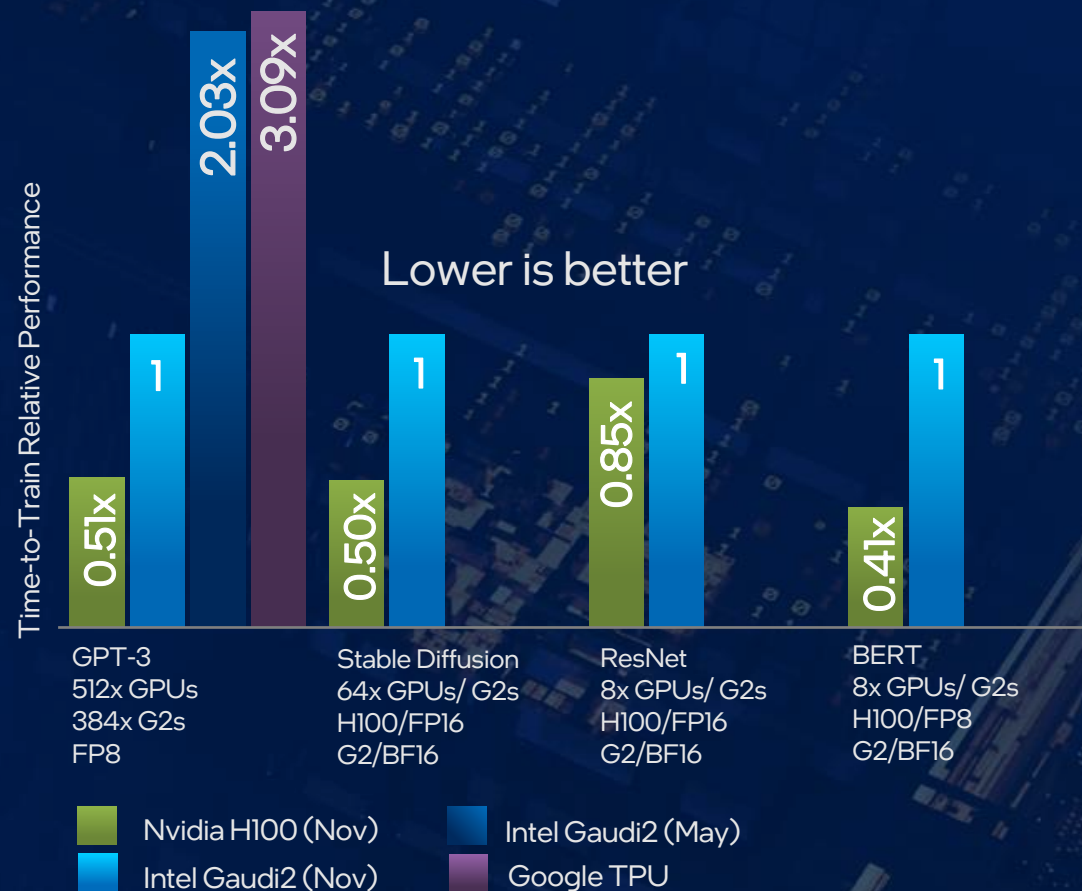
MLPerf Training 3.1 GPT-3 Benchmark



40-50% Better Price-Performance on GPT-3 vs. H100

Based on the most recent MLPerf Training benchmark and estimated range of server pricing, Intel Gaudi2 AI accelerator delivers between 40 and 50% better price performance than Nvidia H100.

- 1 of only 2 merchant silicon submissions for GPT-3
- H100/FP8 outperformed Gaudi2/BF16 on BERT
- Intel® Gaudi® 2 ResNet result near H100 submission

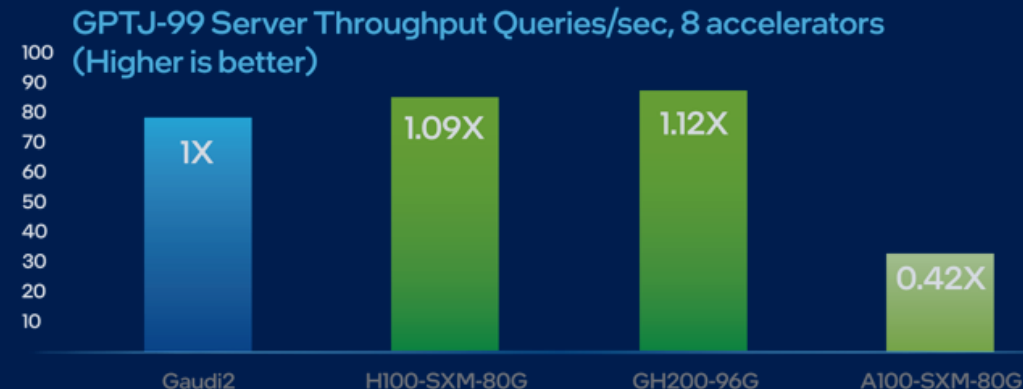


Near-Parity GPT-J Inference Performance vs. H100

Intel® Gaudi® 2 with FP8
achieved accuracy of

99.9%

- Intel Gaudi 2 throughput:
-9% (server) and
-28% (offline) vs H100
- Vs. A100: 2.4x (Server) and 2x (Offline)

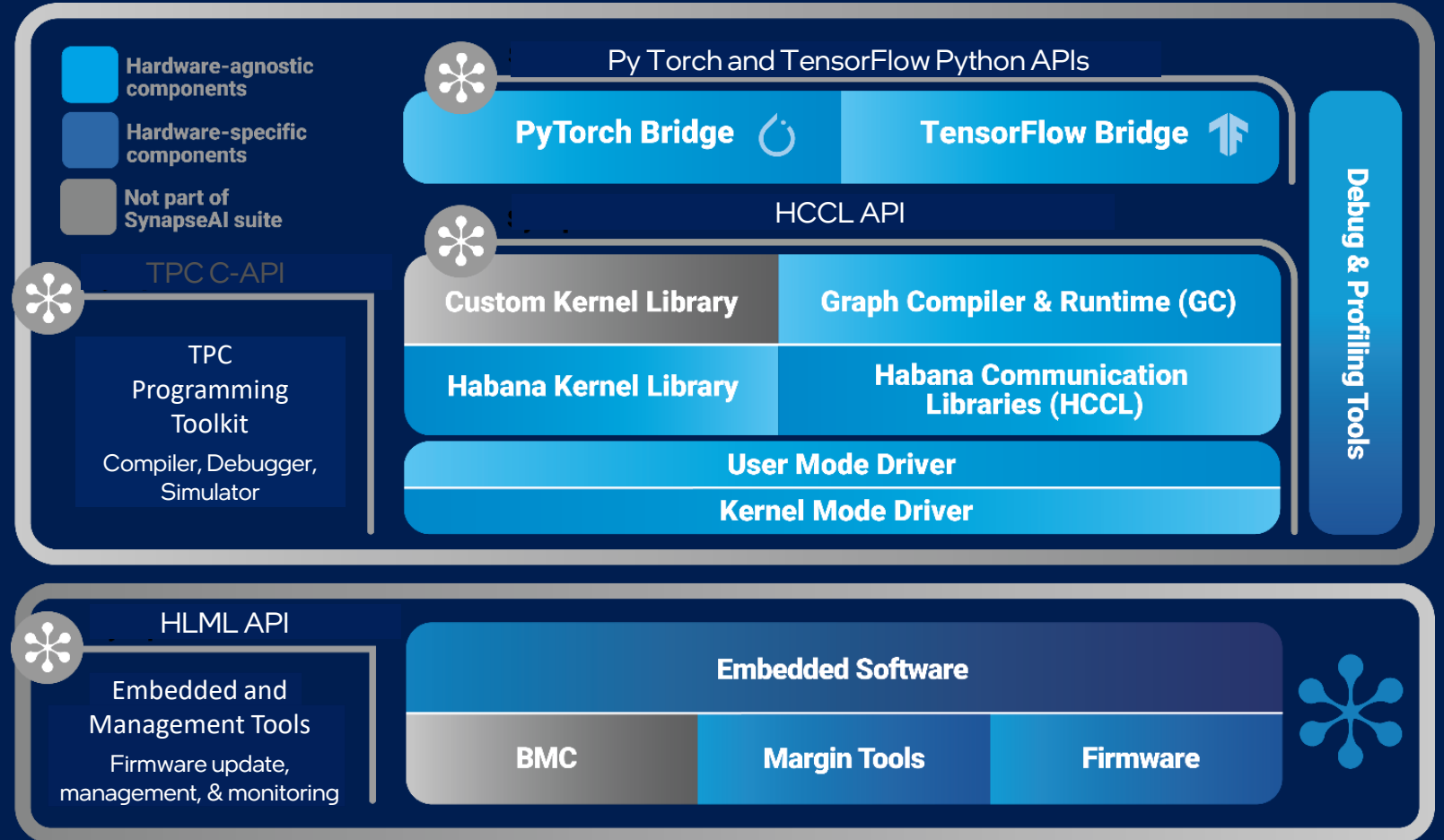


Intel® Gaudi® Software Ecosystem

Software Optimized for Intel® Gaudi® AI Accelerator Performance and Usability

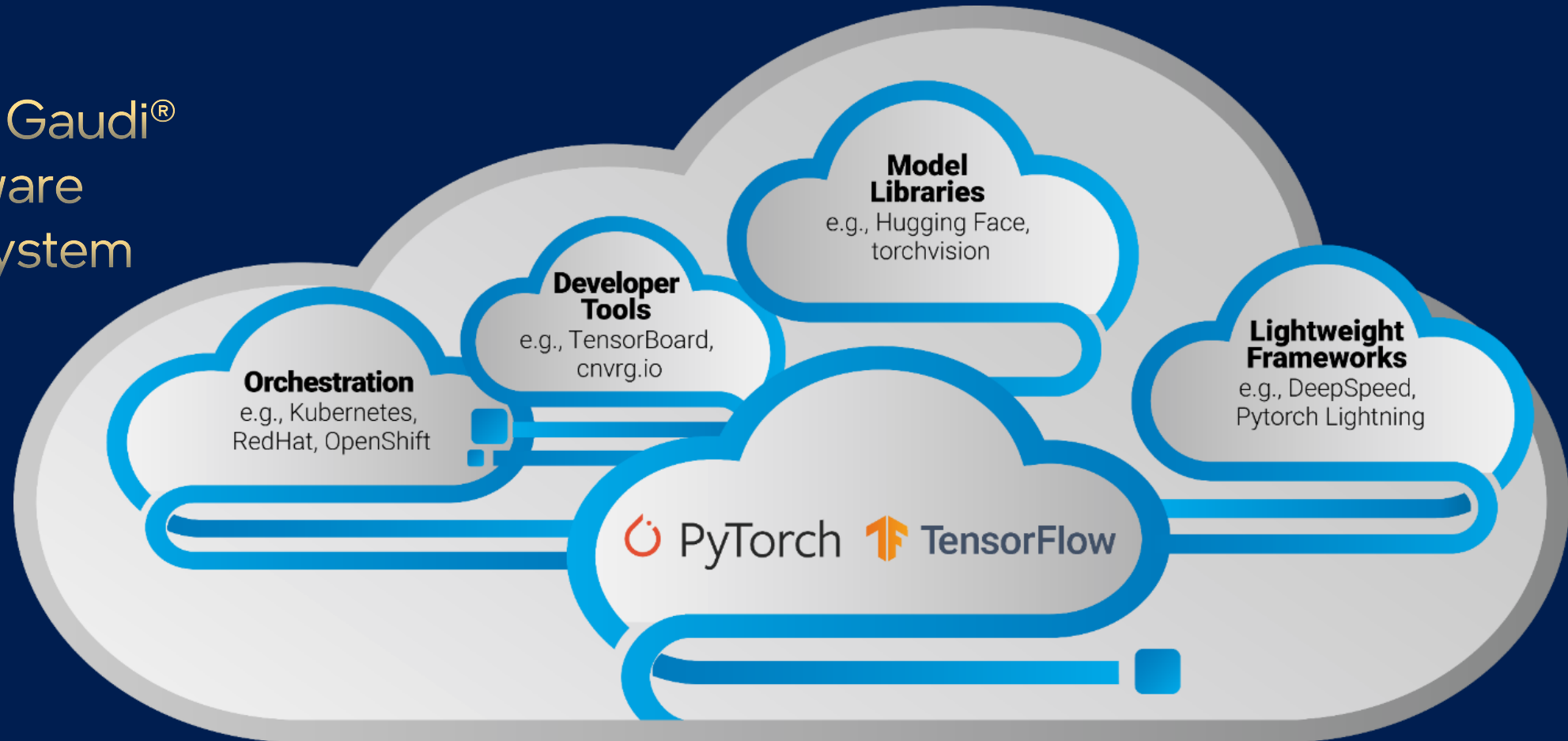


Will evolve to merge
with OneAPI,
delivering forward
compatibility for
Gaudi software to
Falcon Shores



Access to Thousands of AI Models to Ease Development Process

Intel® Gaudi®
Software
Ecosystem



Developer Resources

Intel® Gaudi® Developer Site

habana Developer Home Resources Documentation Catalog Forum Explore More

Home » Resources » Habana Model Performance Data

Habana Model Performance Data

See the latest performance data for Gaudi2 training, Gaudi2 inference, Gaudi training and Gaudi inference. For information on models and containers currently integrated with Habana's Synapse AI software suite visit the Habana catalog.

TRAINING INFERENCE

Gaudi2 MLPerf™ 3.0 Training Performance

These performance numbers have been generated with the latest version of SynapseAI and are improvements over the officially submitted numbers on MLCommons website.

Framework Version	Model	# HPU	Precision	Time To Train
PyTorch 2.0.1	MLPerf 3.0 - GPT3	256	bf16	442.5 min
PyTorch 2.0.1	MLPerf 3.0 - BERT	64	bf16	2.2 min
PyTorch 2.0.1	MLPerf 3.0 - BERT	8	bf16	13.3 min
PyTorch 2.0.1	MLPerf 3.0 - ResNet	8	bf16	16.4 min
PyTorch 2.0.1	MLPerf 3.0 - 3D U-Net	8	bf16	21.3 min
TensorFlow 2.12.1	MLPerf 3.0 - ResNet	8	bf16	15.9 min
TensorFlow 2.12.1	MLPerf 3.0 - BERT	8	bf16	14.5 min

Gaudi2 Reference Models Training Performance

Show 25 entries Search:

Framework Version	Model	# HPU	Precision	Throughput	Accuracy	Time To Train
Select Framework	Filter Model					
DeepSpeed 0.9.4	Megatron-DeepSpeed BLOOM 13B	64	bf16	64.37 sent/sec		
DeepSpeed 0.9.4	Megatron-DeepSpeed LLaMA 13B	64	bf16	55.12 sent/sec		
Lightning 2.0.4	Stable Diffusion	64	bf16	6820.62 img/sec		
Lightning 2.0.4	Stable Diffusion	8	bf16	1202.97 img/sec		
Lightning 2.0.4	Stable Diffusion	1	bf16	151.88 img/sec		

Intel Gaudi GitHub

Model-References Public

README.md

Please visit this page for performance information.

This repository is a collection of models that have been ported to run on Habana Gaudi AI accelerator. They are intended as examples, and will be reasonably optimized for performance while still being easy to read.

Computer Vision

Models	Framework	Validated on Gaudi	Validated on Gaudi2
ResNet50, ResNeXt101	PyTorch	Training	Training, Inference
ResNet50 for PyTorch Lightning	PyTorch Lightning	Training	Training
ResNet152	PyTorch	Training	-
MobileNetV2	PyTorch	Training	-
UNet 2D, UNet3D	PyTorch Lightning	Training, Inference	Training, Inference
SSD	PyTorch	Training	Training
GoogLeNet	PyTorch	Training	-
Vision Transformer	PyTorch	Training	-
DINO	PyTorch	Training	-
YOLOX	PyTorch	Training	-

Audio

Models	Framework	Validated on Gaudi	Validated on Gaudi2
Wav2Vec2ForCTC	PyTorch	Inference	Inference
Hubert	PyTorch	-	Training

Generative Models

Models	Framework	Validated on Gaudi	Validated on Gaudi2
V-Diffusion	PyTorch	Inference	-
Stable Diffusion	PyTorch Lightning	Training, Inference	Training, Inference
Stable Diffusion FineTuning	PyTorch	Training	Training
Stable Diffusion v1.5	PyTorch	Inference	Inference
Stable Diffusion v2.1	PyTorch	Inference	Inference

Intel Gaudi Optimum Library on Hugging Face Hub

Validated Models

The following model architectures, tasks and device distributions have been validated for Optimum Habana:

In the tables below, ✓ means single-card, multi-card and DeepSpeed have all been validated.

Optimum Habana

Optimum Habana is the interface between the Transformers and Diffusers libraries and Habana's Gaudi processor (HPU). It provides a set of tools enabling easy model loading, training and inference on single- and multi-HPU settings for different downstream tasks. The list of officially validated models and tasks is available here. Users can try other models and tasks with only few changes.

What is a Habana Processing Unit (HPU)?

HPUs offer fast model training and inference as well as a great price-performance ratio. Check out this blog post about BERT pre-training and this article benchmarking Habana Gaudi2 versus Nvidia A100 GPUs for concrete examples. If you are not familiar with HPUs and would like to know more about them, we recommend you take a look at our conceptual guide.

Install

To install the latest stable release of this package:

```
pip install --upgrade-strategy eager optimum[habana]
```

	LoRA	LoRA	
StableLM	✗	Single card	text generation
Falcon	✗	Single card	text generation
CodeGen	✗	Single card	text generation
MPT	✗	Single card	text generation
TS	✓	✓	summarization translation question answering

Train on Gaudi2

Native PyTorch

based model migration

Tools:

- GPU migration tools
- HPU graph
- FP8

References:

- Habana Model References

Megatron & DeepSpeed

based model migration

Tools:

- 3D Parallelism
- DP Zero
- LLM pretrain
- FP8

References:

- Habana Megatron-DeepSpeed
- Habana DeepSpeed

Huggingface

based model migration

Tools:

- HF API compatibility
- One step train
- DP Zero
- FP8

References:

- Optimum Habana

Inference on Gaudi2

Model / Framework based	Tools	References
Native PyTorch	GPU migration tools DeepSpeed Tensor parallelism FP8 quantization tool HPU graph	Habana Model References
Huggingface	HF API compatibility One step inference DeepSpeed Tensor parallelism FP8 quantization tool HPU graph	Optimum Habana
vLLM / TGI	To be released soon	Habana vLLM / TGI-gaudi
Ray.io	To be released soon	

Easily Get Started with PyTorch Models

```
import torch
import torch.nn as nn
import torch.optim as optim
import torch.nn.functional as F
import torchvision
import torchvision.transforms as transforms
import os

# Import Habana Torch Library
import habana_frameworks.torch.core as htcore

# neural network model
class SimpleModel(nn.Module):
    ...

# training loop
def train(net, criterion, optimizer, trainloader, device):
    ...
    loss.backward()

# API call to trigger execution
htcore.mark_step()

optimizer.step()

# API call to trigger execution
htcore.mark_step()

def main():
    ...

# Target the Gaudi HPU device
device = torch.device("hpu")
```

*Minimal
code to
start
using Gaudi*

Migrating Python APIs with GPU dependencies

```
import torch
import torch.nn as nn
import torch.optim as optim
import torch.nn.functional as F
import torchvision
import torchvision.transforms as transforms
import os

# Import GPU Migration Package:
import habana_frameworks.torch.gpu_migration

# Import Habana Torch Library
import habana_frameworks.torch.core as htcore

# neural network model
class SimpleModel(nn.Module):
    ...

# training loop
def train(net, criterion, optimizer, trainloader, device):
    ...
    loss.backward()
    # API call to trigger execution
    htcore.mark_step()
    optimizer.step()
    # API call to trigger execution
    htcore.mark_step()

def main():
    ...
    # Target the Gaudi HPU device
    device = torch.device("hpu")
```

Simplifies replacing Python API calls that have dependencies on GPU libraries with HPU-specific API calls

Specific API calls from following Python libraries are mapped to equivalents in SynapseAI:

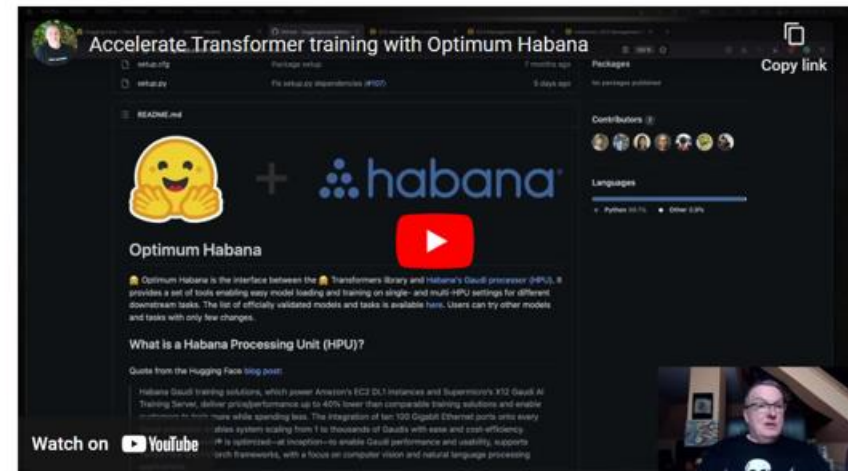
- torch.cuda
- torch APIs with GPU related parameters. For example, torch.randn(device="cuda").
- pytorch_lightning
- apex
- pynvml

Getting Started With Hugging Face on Gaudi

```
from optimum.habana import GaudiConfig, GaudiTrainer, GaudiTrainingArguments
from transformers import BertTokenizer, BertModel
...
tokenizer = BertTokenizer.from_pretrained("bert-base-uncased")
model = BertModel.from_pretrained("bert-base-uncased")
gaudi_config = GaudiConfig.from_pretrained("Habana/bert-base-uncased")
args = GaudiTrainingArguments(
    output_dir="/tmp/output_dir",
    use_habana=True,
    use_lazy_mode=True,
)
trainer = GaudiTrainer(
    model=model,
    gaudi_config=gaudi_config,
    args=args,
    tokenizer=tokenizer,
)
trainer.train()
```

Supported Model Architectures include:

- NLP: BERT, ALBERT, DistilBERT, RoBERTa, T5, GPT2, BLOOM, LLaMA, MPT, Mixtral,...
- Vision: ViT, SWIN, Stable-diffusion-diffusers
- Audio: Wav2vec2

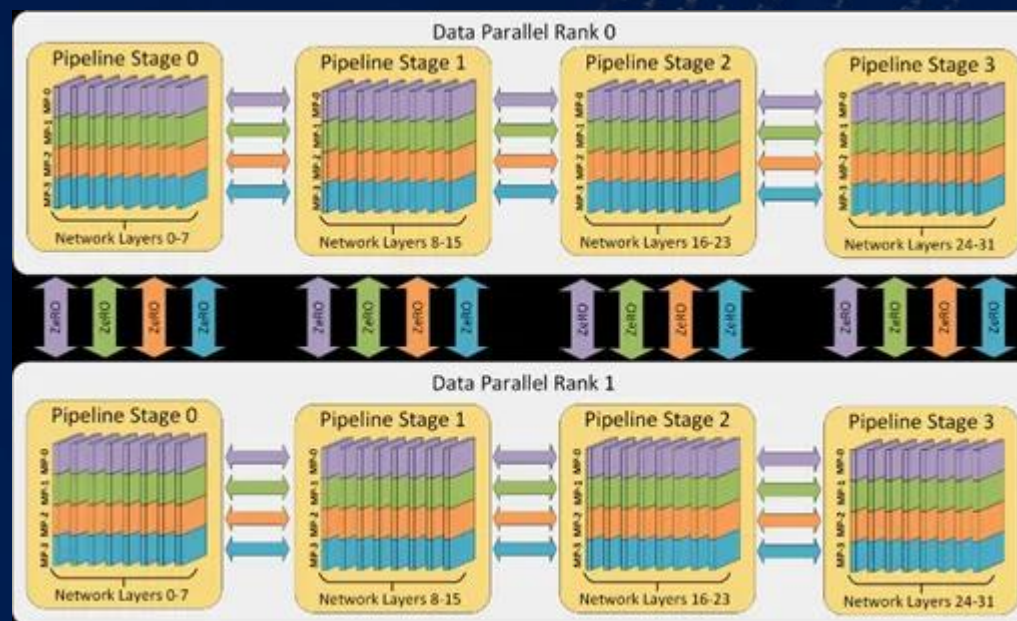


Optimum-Habana documentation: https://huggingface.co/docs/optimum/habana_index

Train models with 4D parallelism on Gaudi2

By using Megatron-DeepSpeed, train LLaMA on large scale Gaudi2 systems.

- Data Parallelism
- Tensor Parallelism
- Pipeline Parallelism
- Sequence Parallelism



Example: `HL_HOSTSFILE=scripts/hostsfile HL_SEQ_PARALLEL=1 HL_MICRO_BATCH=1 HL_NUM_NODES=32
HL_PP=8 HL_TP=8 HL_DP=4 scripts/run_llamav2.sh`

FP8 Training with Intel Gaudi Transformer Engine

1. Import TE and use TE modules in your model, e.g. a `te.Linear`

```
import torch
import habana_frameworks.torch.hpex.experimental.transformer_engine as te

# Set dimensions.
in_features = 768
out_features = 3072
hidden_size = 2048

# Initialize model and inputs.
model = te.Linear(in_features, out_features, bias=True)
inp = torch.randn(hidden_size, in_features, device="hpu")
```

2. Wrap the forward pass of the training with `fp8_autocast`

```
from habana_frameworks.torch.hpex.experimental.transformer_engine import recipe

# Create an FP8 recipe. Note: All input args are optional.
fp8_recipe = recipe.DelayedScaling(margin=0, interval=1)

# Enable autocasting for the forward pass
with te.fp8_autocast(enabled=True, fp8_recipe=fp8_recipe):
    out = model(inp)

loss = out.sum()
loss.backward()
```

https://docs.habana.ai/en/latest/PyTorch/PyTorch_FP8_Training/index.html



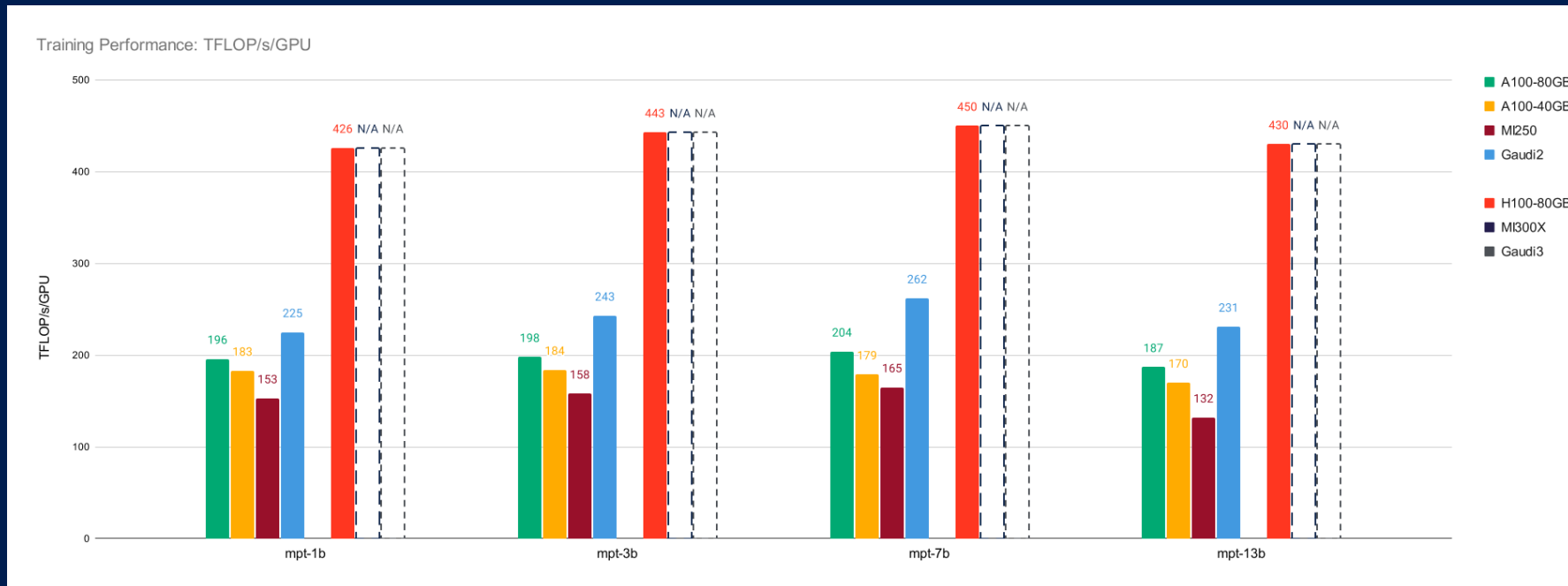
Customer examples in GenAI & LLM

Intel Gaudi 2 Training New MPT Model from MosaicML: Outperforms H100 for price-performance

3rd party evaluation by Databricks' MosaicML:

<https://www.databricks.com/blog/llm-training-and-inference-intel-gaudi2-ai-accelerators>

- Delivers 260 TFLOPs-per-Gaudi2: ~55% training performance relative to H100
 - With its cost advantage over H100, Gaudi 2 outperforms H100 for price-performance training MPT

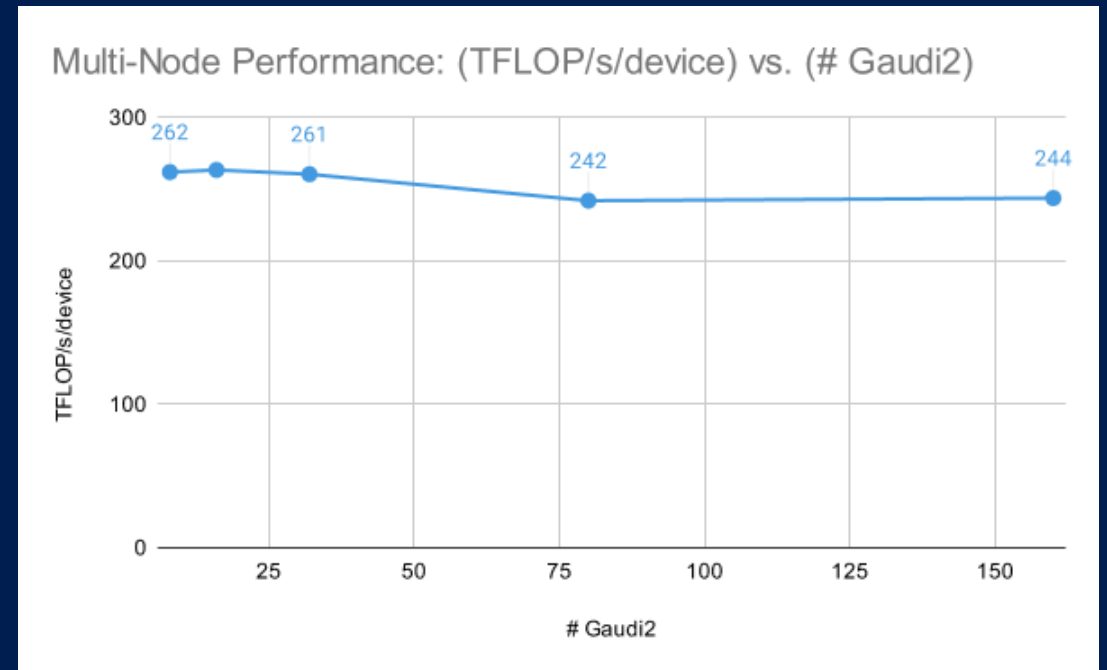
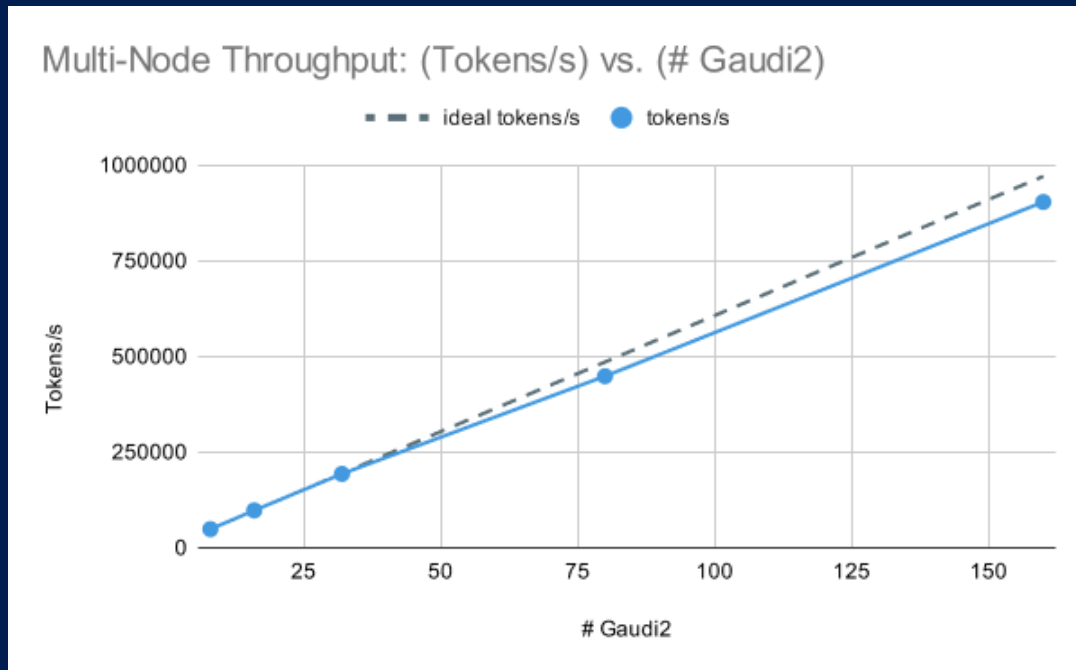


“The MI300x and Gaudi 3 are not yet profiled but are expected to be competitive with H100.”

MosaicML Evaluation: Gaudi2 near-linear scaling training MPT

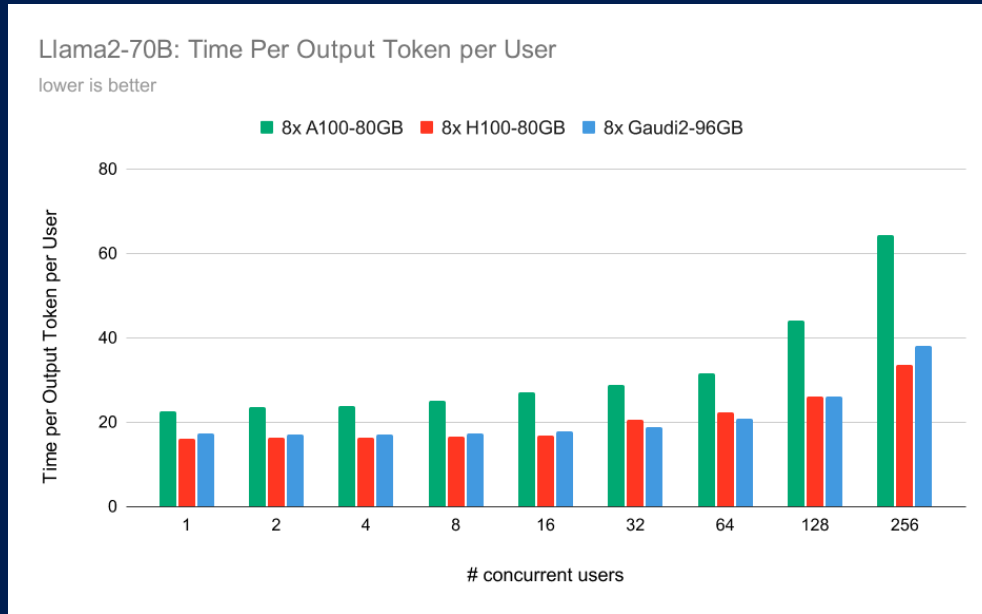
“MPT-7B multi-node training performance: As we scale from 8xGaudi2 to 160xGaudi2, we see a near-linear increase in throughput and nearly constant TFLOP/s/device. Note that the global train batch size is held constant at 1920 samples, so these plots demonstrate strong scaling.”

<https://www.databricks.com/blog/llm-training-and-inference-intel-gaudi2-ai-accelerators>

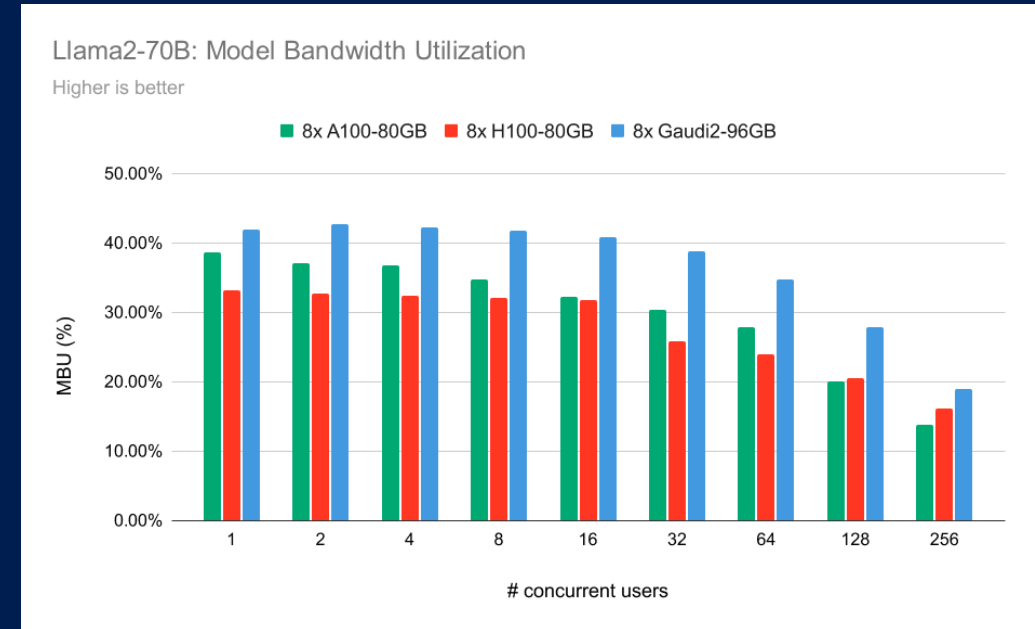


MosaicML Evaluation: Gaudi2 Inference of LLama2-70B

<https://www.databricks.com/blog/llm-training-and-inference-intel-gaudi2-ai-accelerators>



“The TPOT for Gaudi2 is better than H100 under most of the user loads.”



For decoding latency, the most expensive phase of LLM inference:

With 2450 GB/s of HBM2 memory bandwidth, Gaudi 2 matches H100 decoding latency with 3350 GB/s of HBM3 bandwidth

Stability.ai Cluster Built on Intel® Gaudi® 2 AI Accelerator

“Today, we’re excited to announce that we just secured a design win that’s a pretty big deal. A large AI supercomputer will be built entirely on Intel’s Xeon processors and 4,000 Intel Gaudi2 AI Hardware accelerators. Stability AI is now our anchor customer. Given the magnitude of the Stability AI build out, it will be a top 15 AI supercomputer in the world.”

– Pat Gelsinger

intel®
GAUDI

Stability.ai: Benchmarking Compute Solutions

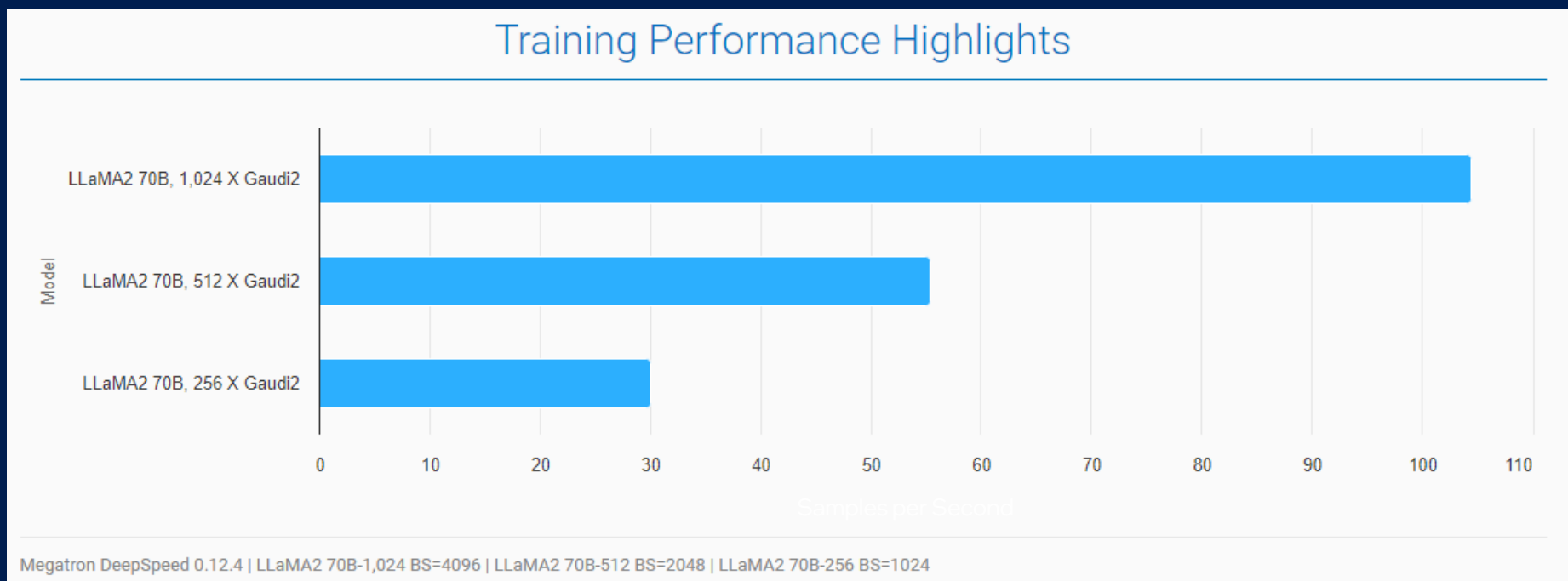
- Training throughput of 2B MMDDiT on Stable Diffusion 3

Device	Attention	# Nodes	# Accelerators (total)	Batch Size per Accelerator	Total Batch Size	Images / sec (100-MA)
Gaudi2	FusedSDPA	2	16	32	512	1,254
Gaudi2	FusedSDPA	2	16	16	256	927
H100-80GB	xFormers	2	16	16	256	595
A100-80GB	xFormers	2	16	16	256	381

Device	# Nodes	# Accelerators (total)	Batch Size per Accelerator	Total Batch Size	Images / sec	Images / sec / device
Gaudi2	32	256	16	4096	12,654	49.4
A100-80GB	32	256	16	4096	3,992	15.6

LLaMA2-70B Training on Gaudi2

- Train LLaMA2-70B on 256x, 512x, 1024x Gaudi2



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Thank You!

