

Scalability results of finetuning: Performance, Efficiency, Convergence and Generalization

Course 3 "Introduction to Large Language Models at Scale"

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Model training experiments

We conducted a series of **training experiments** to *investigate* the **impact** of hyperparameter choices and computational scaling on the **performance** of large language models (LLMs). Specifically, we trained the **meta-llama-3.2-1B** and **meta-llama-3.2-3B** models while varying key hyperparameters, including



- batch size = (4, 8, 16, 32)
- number of training epochs = (10, 30, 50)
- number of GPUs = (1, 2, 3, 4)
- learning rate = $5e-6$
- weight decay = 0.01
- warmup ratio = 0.1
- gradient accumulation = 4 steps
- maximum sequence length = 256 tokens

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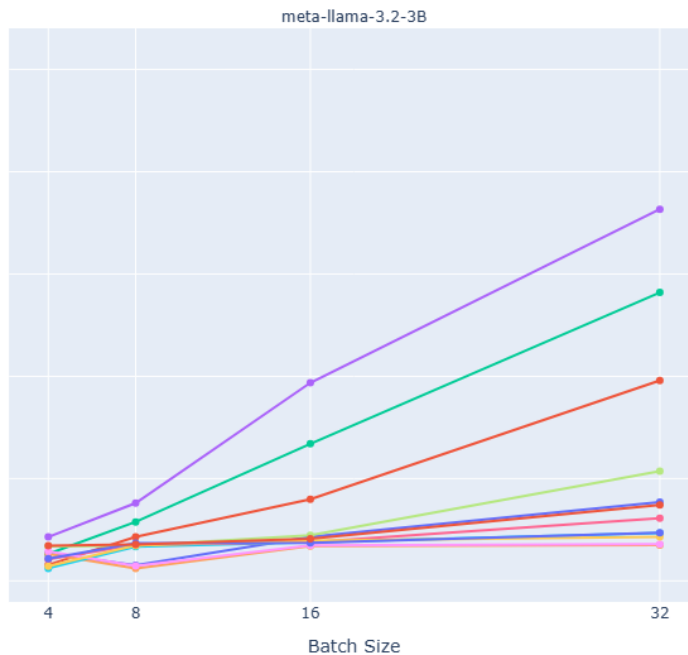
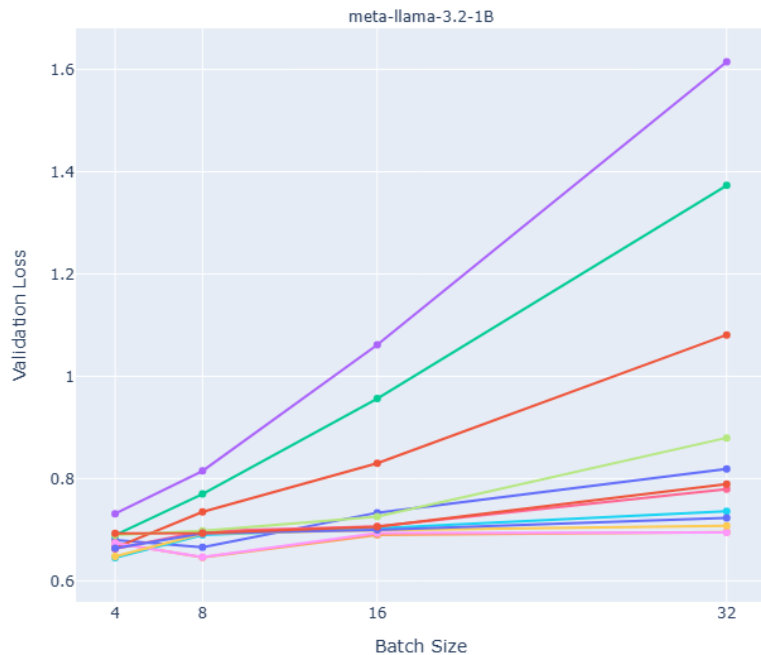
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Validation Loss vs Batch Size (GPUs - Epochs)

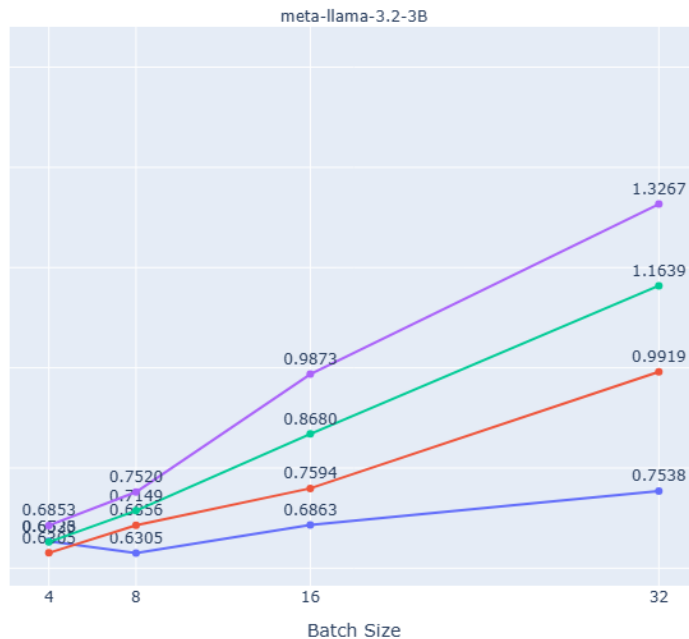
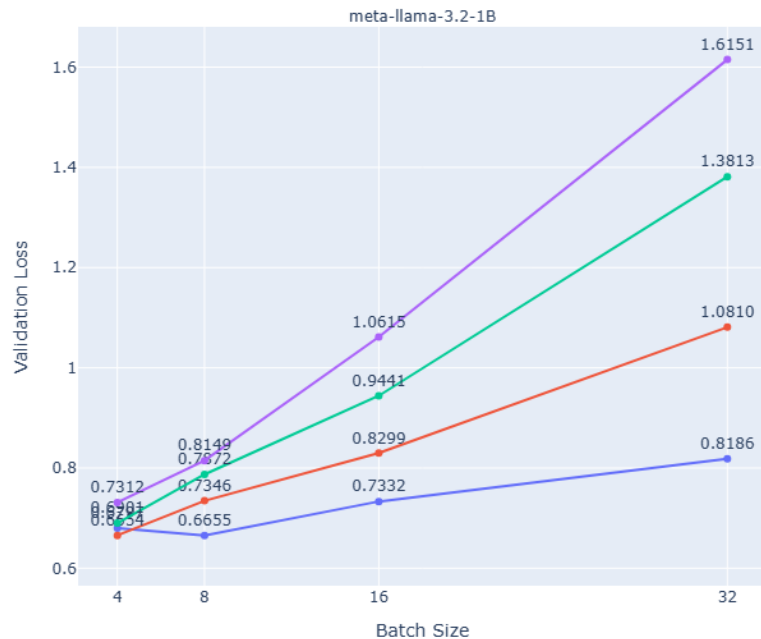


GPUs - Epochs

- GPUs: 1 | Epochs: 10
- GPUs: 2 | Epochs: 10
- GPUs: 3 | Epochs: 10
- GPUs: 4 | Epochs: 10
- GPUs: 1 | Epochs: 30
- GPUs: 2 | Epochs: 30
- GPUs: 3 | Epochs: 30
- GPUs: 4 | Epochs: 30
- GPUs: 1 | Epochs: 50
- GPUs: 2 | Epochs: 50
- GPUs: 3 | Epochs: 50
- GPUs: 4 | Epochs: 50

Performance 10 epochs

Validation Loss vs Batch Size (GPUs - Epochs)

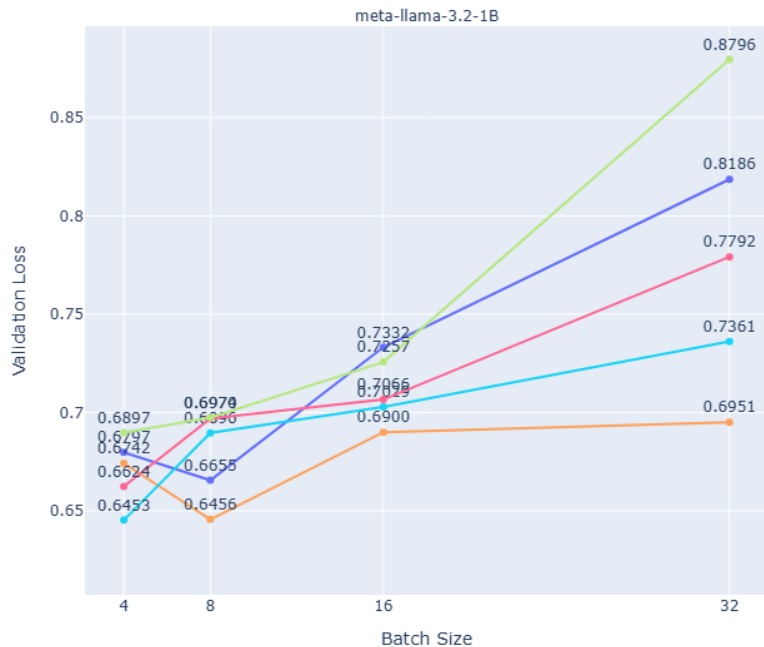


GPUs - Epochs

- GPUs: 1 | Epochs: 10
- GPUs: 2 | Epochs: 10
- GPUs: 3 | Epochs: 10
- GPUs: 4 | Epochs: 10
- GPUs: 1 | Epochs: 30
- GPUs: 2 | Epochs: 30
- GPUs: 3 | Epochs: 30
- GPUs: 4 | Epochs: 30
- GPUs: 1 | Epochs: 50
- GPUs: 2 | Epochs: 50
- GPUs: 3 | Epochs: 50
- GPUs: 4 | Epochs: 50

Performance 30 epochs

Validation Loss vs Batch Size (GPUs - Epochs)

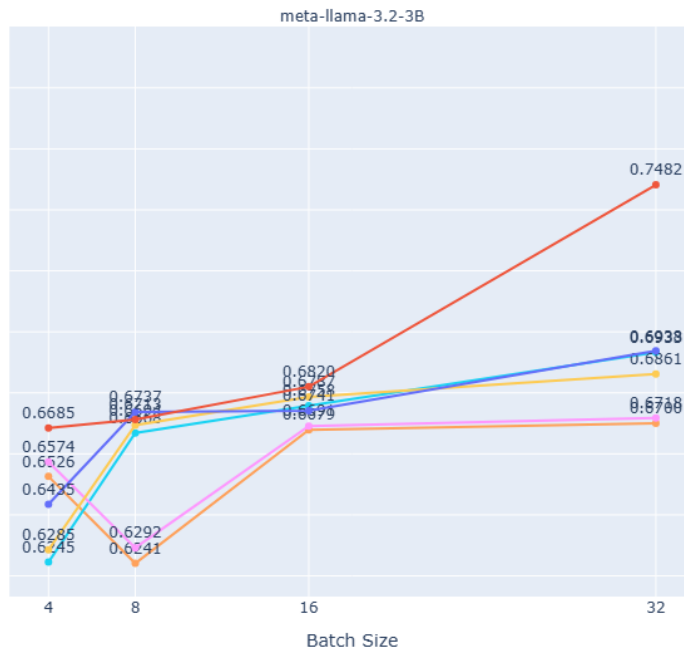
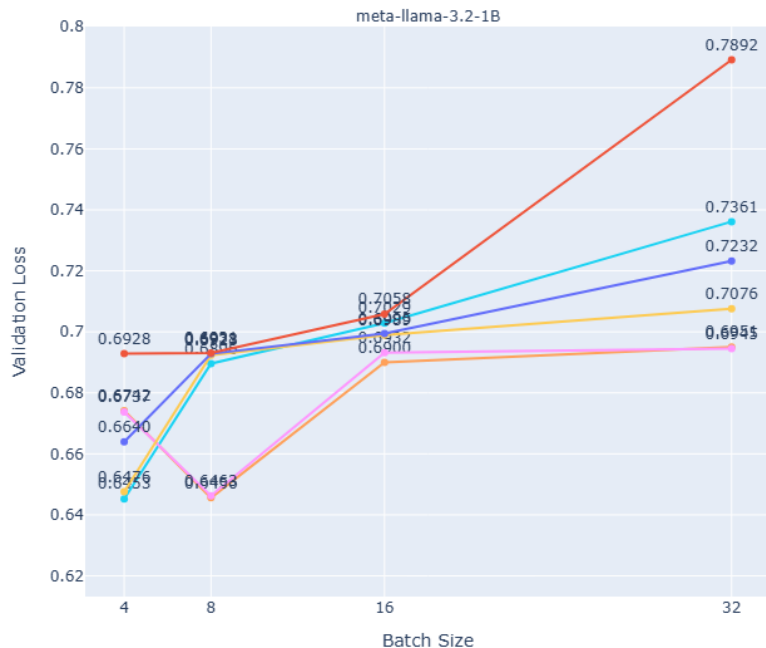


GPUs - Epochs

- GPUs: 1 | Epochs: 10
- GPUs: 2 | Epochs: 10
- GPUs: 3 | Epochs: 10
- GPUs: 4 | Epochs: 10
- GPUs: 1 | Epochs: 30
- GPUs: 2 | Epochs: 30
- GPUs: 3 | Epochs: 30
- GPUs: 4 | Epochs: 30
- GPUs: 1 | Epochs: 50
- GPUs: 2 | Epochs: 50
- GPUs: 3 | Epochs: 50
- GPUs: 4 | Epochs: 50

Performance 50 epochs

Validation Loss vs Batch Size (GPUs - Epochs)

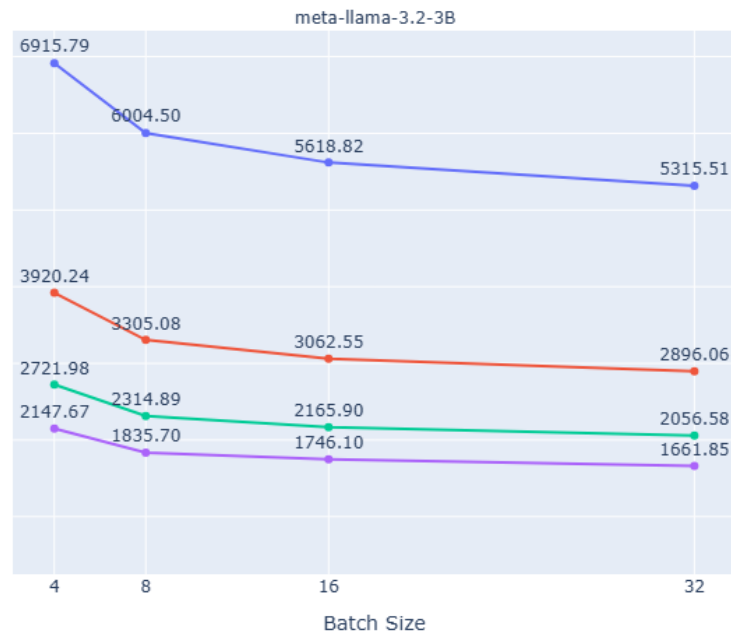
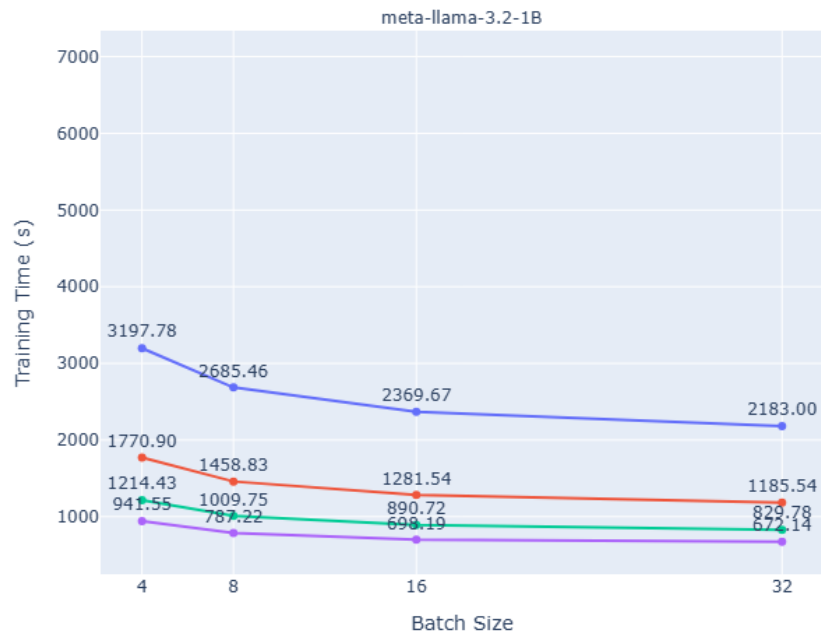


GPUs - Epochs

- GPUs: 1 | Epochs: 10
- GPUs: 2 | Epochs: 10
- GPUs: 3 | Epochs: 10
- GPUs: 4 | Epochs: 10
- GPUs: 1 | Epochs: 30
- GPUs: 2 | Epochs: 30
- GPUs: 3 | Epochs: 30
- GPUs: 4 | Epochs: 30
- GPUs: 1 | Epochs: 50
- GPUs: 2 | Epochs: 50
- GPUs: 3 | Epochs: 50
- GPUs: 4 | Epochs: 50

Performance

Training Time vs Batch Size per Model/GPUs - 30 Epochs



GPUs

GPUs: 1

GPUs: 2

GPUs: 3

GPUs: 4

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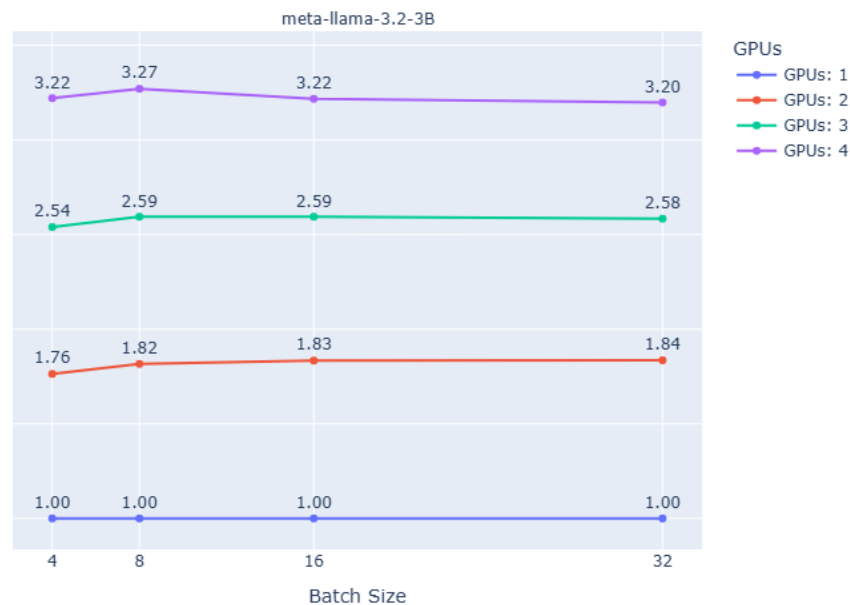
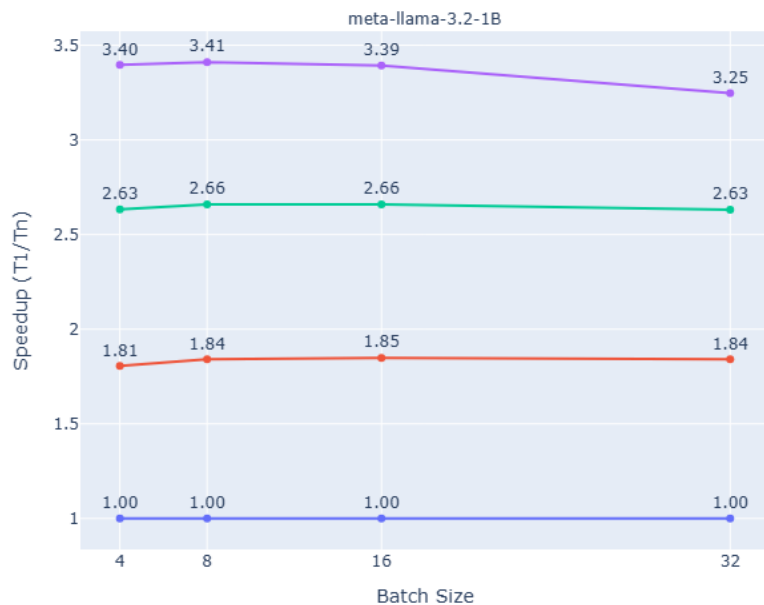
GPU Utilization

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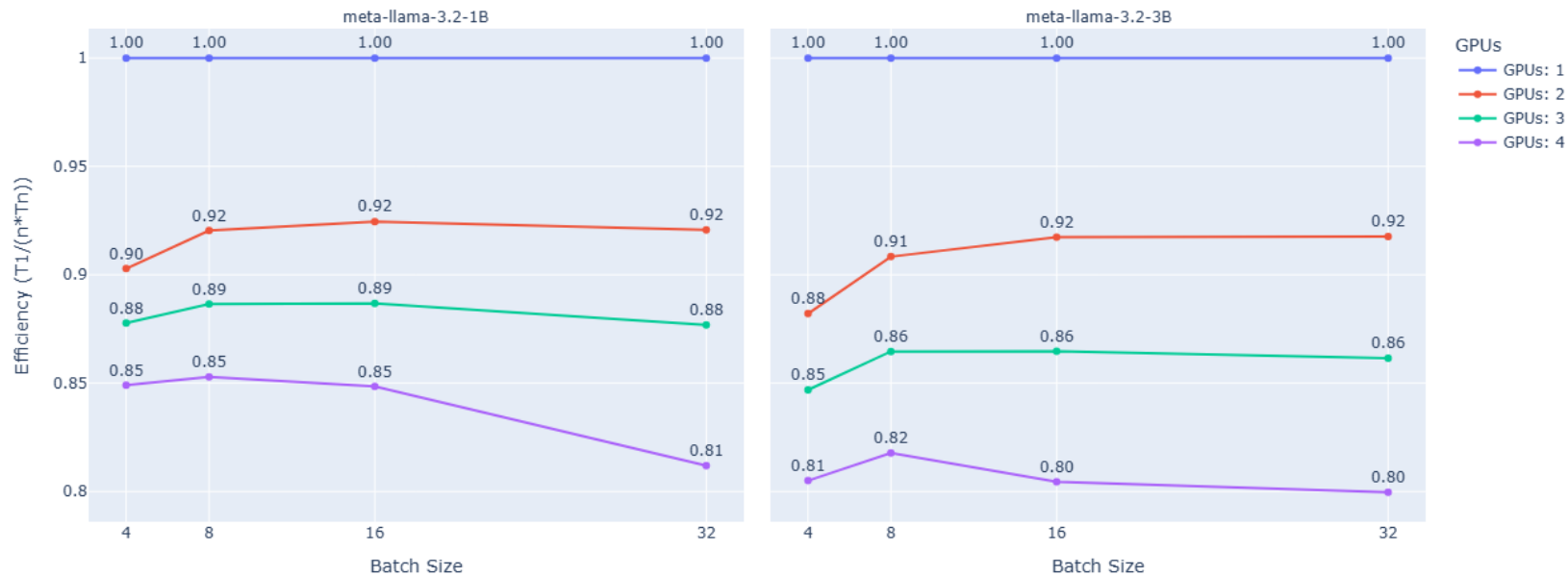
Efficiency

Speedup vs Batch Size per Model/GPUs - 30 Epochs



Efficiency

Efficiency vs Batch Size per Model/GPUs - 30 Epochs



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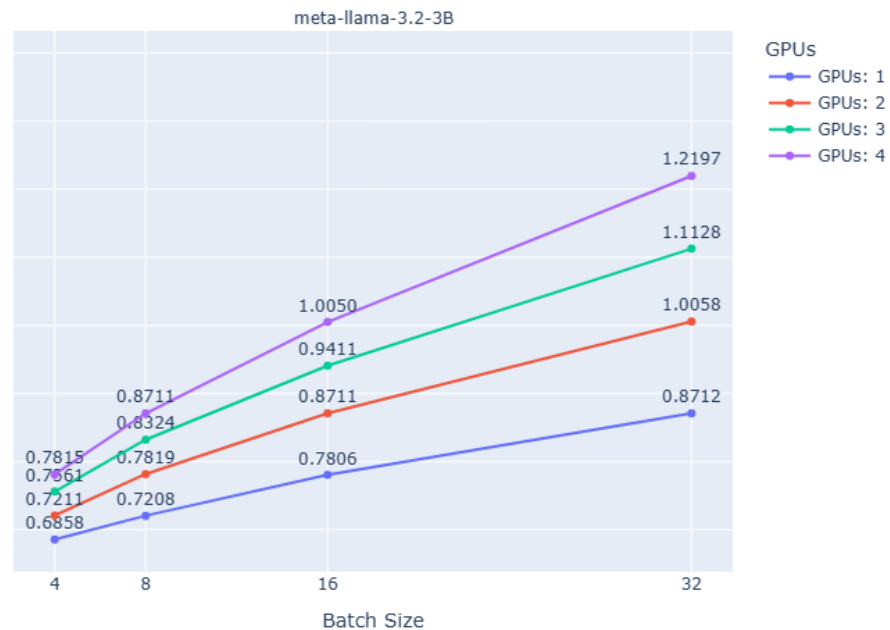
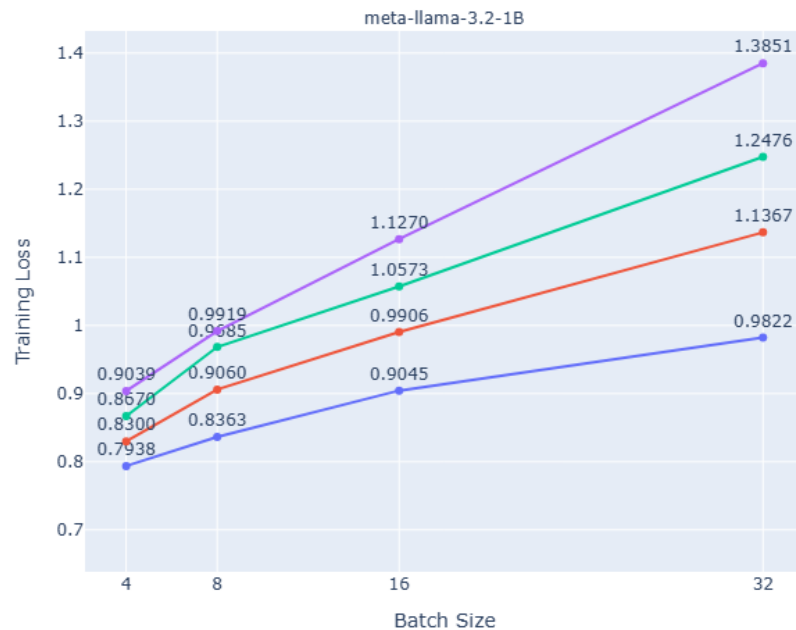
GPU Utilization

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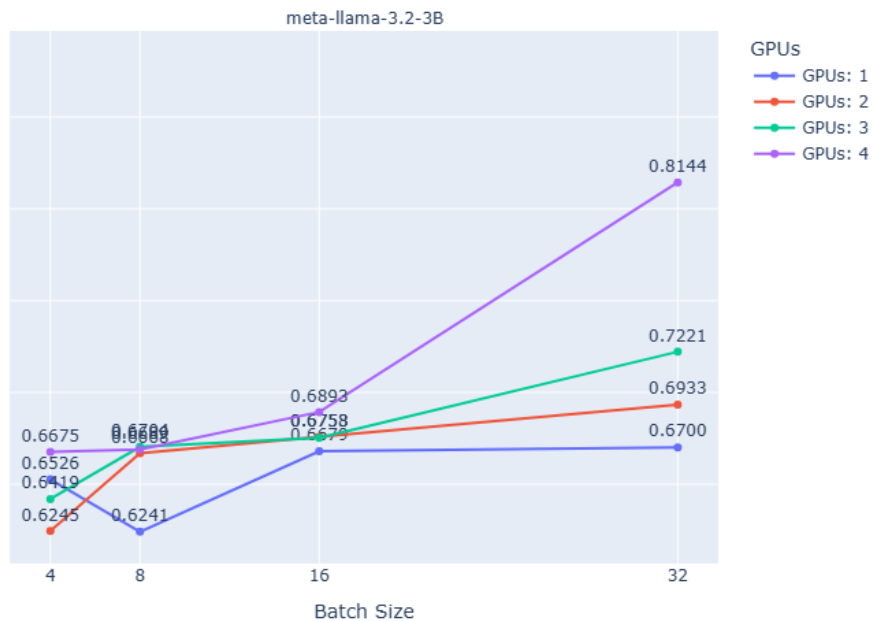
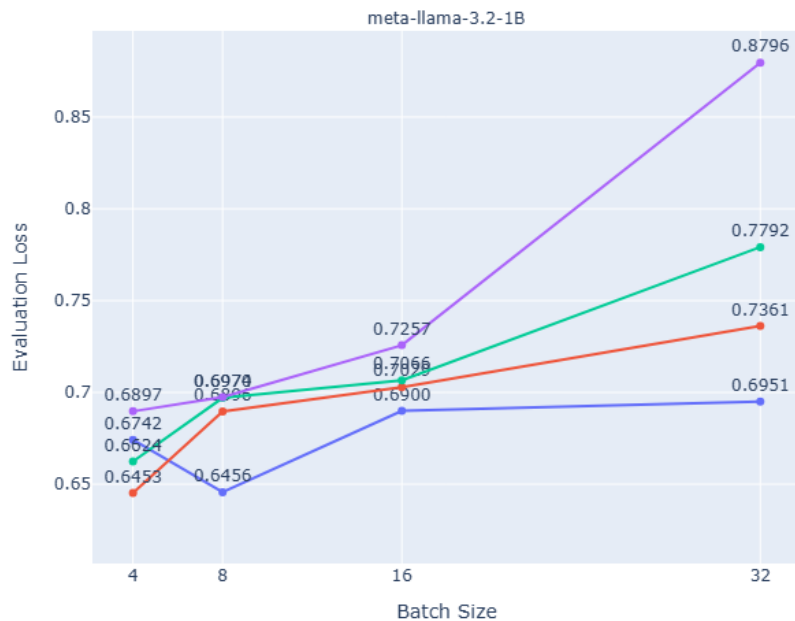
Convergence

Train Loss vs Batch Size per Model/GPUs - 30 Epochs



Convergence

Eval Loss vs Batch Size per Model/GPUs - 30 Epochs



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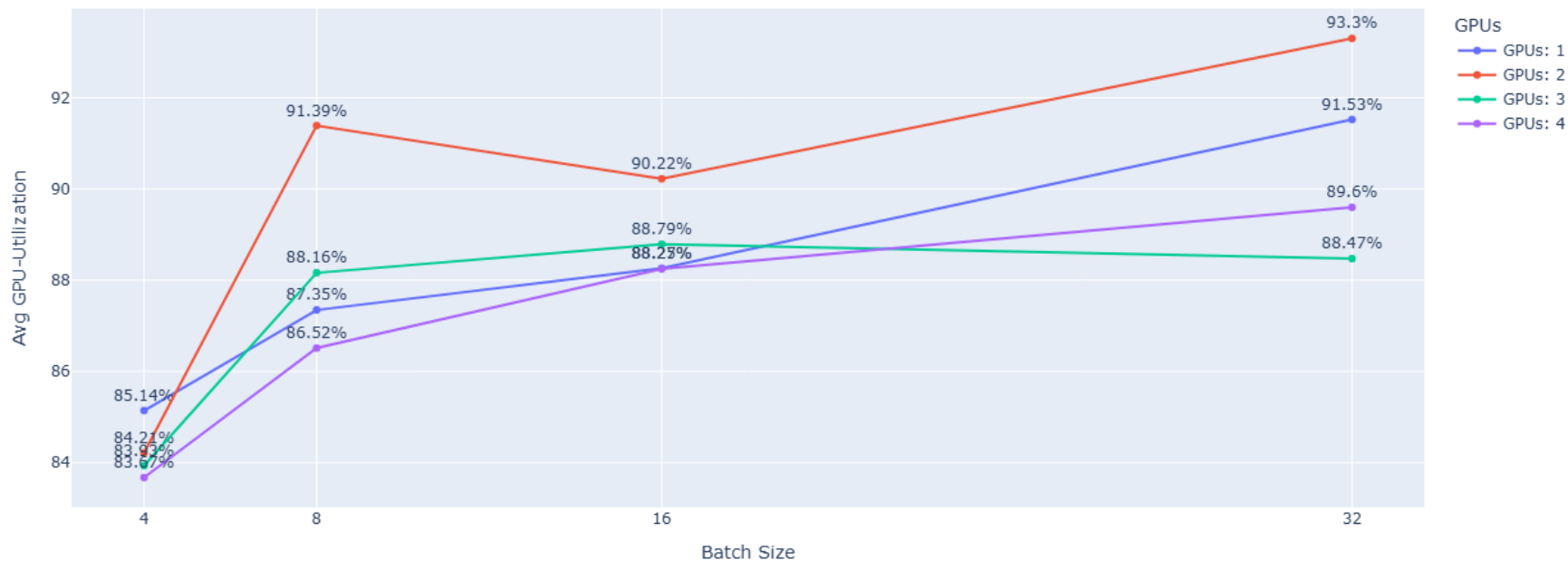
GPU Utilization

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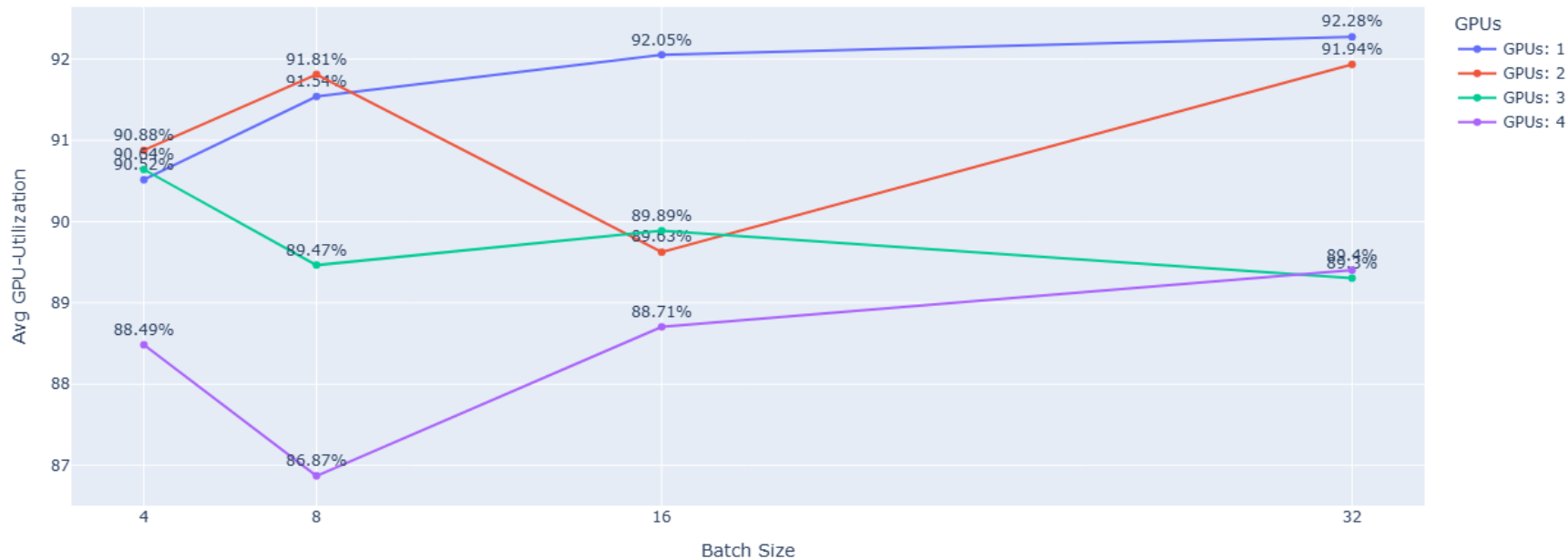
GPU Utilization – nvidia-smi 1B

Model: meta-llama-3.2-1B - Average GPU Utilization (%) vs Batch Size - 30 Epochs



GPU Utilization – nvidia-smi 3B

Model: meta-llama-3.2-3B - Average GPU Utilization (%) vs Batch Size - 30 Epochs



nvidia-smi

The amount of memory currently in use versus the total available memory (in MiB).

The current power draw versus the maximum power limit (in Watts).

NVIDIA-SMI 580.95.05				Driver Version: 580.95.05		CUDA Version: 13.0		
GPU ID	Name	Temp	Perf	Persistence-M Pwr:Usage/Cap	Bus-Id	Disp.A Memory-Usage	Volatile GPU-Util	Uncorr. ECC Compute M. MIG M.
0	NVIDIA A100-SXM4-80GB	48C	P0	76W / 500W	00000000:01:00.0	Off 79539MiB / 81920MiB	16%	0 Default Disabled
1	NVIDIA A100-SXM4-80GB	43C	P0	91W / 500W	00000000:41:00.0	Off 79539MiB / 81920MiB	100%	0 Default Disabled
2	NVIDIA A100-SXM4-80GB	50C	P0	99W / 500W	00000000:81:00.0	Off 79539MiB / 81920MiB	100%	0 Default Disabled
3	NVIDIA A100-SXM4-80GB	43C	P0	95W / 500W	00000000:C1:00.0	Off 79539MiB / 81920MiB	100%	0 Default Disabled

The percentage of the GPU's processing capacity being utilized for computation.

```
Tue Dec 16 00:56:04 2025
```

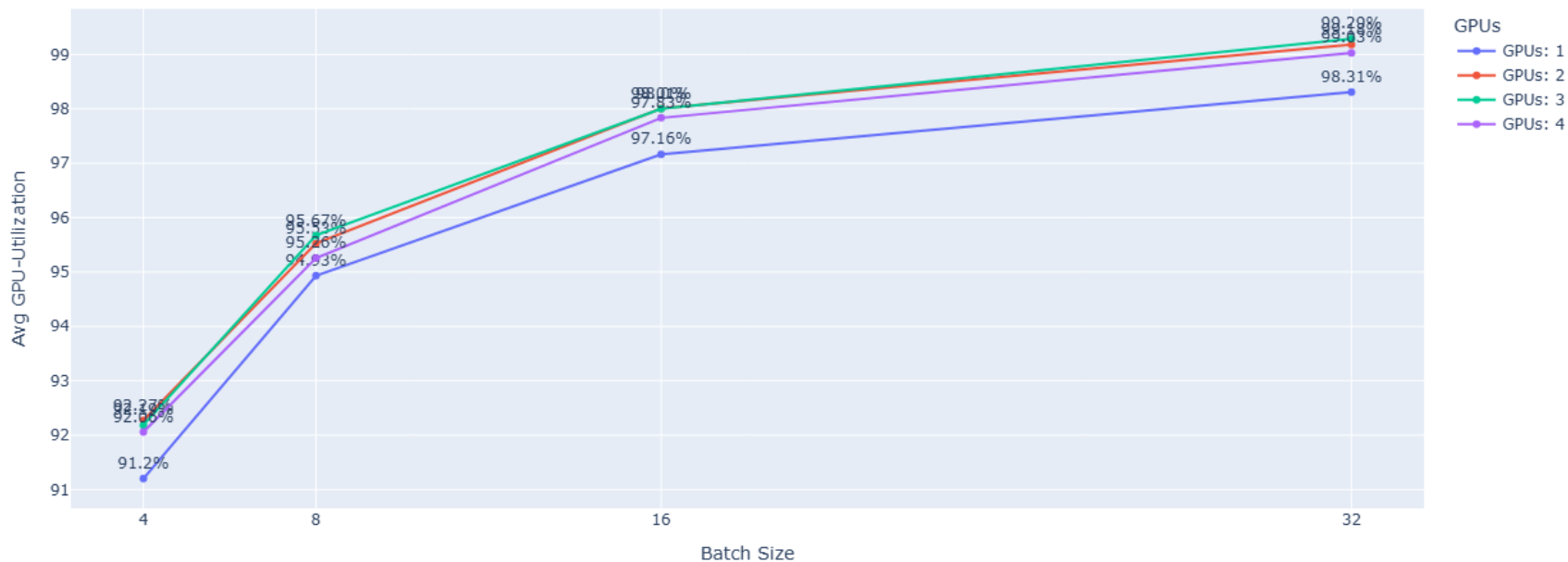
NVIDIA-SMI 580.95.05			Driver Version: 580.95.05			CUDA Version: 13.0		
GPU Fan	Name Temp	Perf Perf	Persistence-M Pwr:Usage/Cap	Bus-Id	Disp.A Memory-Usage	Volatile GPU-Util	Uncorr. Compute MIG M.	ECC M.
0 N/A	NVIDIA 33C	A100-SXM4-80GB P0	On 71W / 500W	00000000:01:00.0	Off 2179MiB / 81920MiB	3%	Default	0 Disabled
1 N/A	NVIDIA 31C	A100-SXM4-80GB P0	On 68W / 500W	00000000:41:00.0	Off 2179MiB / 81920MiB	3%	Default	0 Disabled
2 N/A	NVIDIA 34C	A100-SXM4-80GB P0	On 83W / 500W	00000000:81:00.0	Off 2179MiB / 81920MiB	8%	Default	0 Disabled
3 N/A	NVIDIA 31C	A100-SXM4-80GB P0	On 83W / 500W	00000000:C1:00.0	Off 2179MiB / 81920MiB	1%	Default	0 Disabled

Processes:							
GPU	GI	CI	PID	Type	Process name	GPU Memory	Usage
ID	ID	ID					
0	N/A	N/A	184275	C	.../pytorch/2.9.0/bin/python3.13	2168MiB	

NVIDIA-SMI 580.95.05				Driver Version: 580.95.05			CUDA Version: 13.0		
GPU Fan	Name Temp	Perf	Persistence-M Pwr:Usage/Cap	Bus-Id	Disp.A Memory-Usage	Volatile GPU-Util	Uncorr. Compute	ECC M.	
0 N/A	NVIDIA 48C	A100-SXM4-80GB P0	76W / 500W	00000000:01:00.0	Off 79539MiB / 81920MiB	16%	0	Default Disabled	
1 N/A	NVIDIA 43C	A100-SXM4-80GB P0	91W / 500W	00000000:41:00.0	Off 79539MiB / 81920MiB	100%	0	Default Disabled	
2 N/A	NVIDIA 50C	A100-SXM4-80GB P0	99W / 500W	00000000:81:00.0	Off 79539MiB / 81920MiB	100%	0	Default Disabled	
3 N/A	NVIDIA 43C	A100-SXM4-80GB P0	95W / 500W	00000000:C1:00.0	Off 79539MiB / 81920MiB	100%	0	Default Disabled	

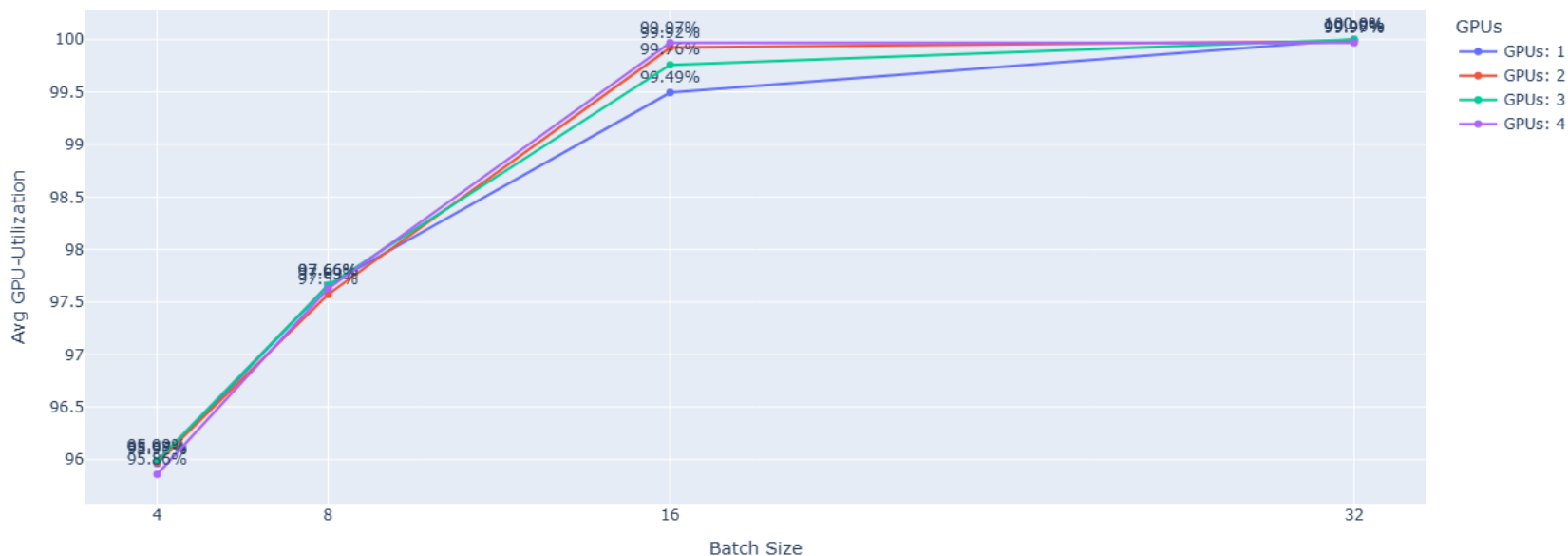
GPU Utilization – pynvml 1B

Model: meta-llama-3.2-1B - Average GPU Utilization (%) vs Batch Size Epochs per Model/GPUs - 30 Epochs



GPU Utilization – pynvml 3B

Model: meta-llama-3.2-3B - Average GPU Utilization (%) vs Batch Size Epochs per Model/GPUs - 30 Epochs



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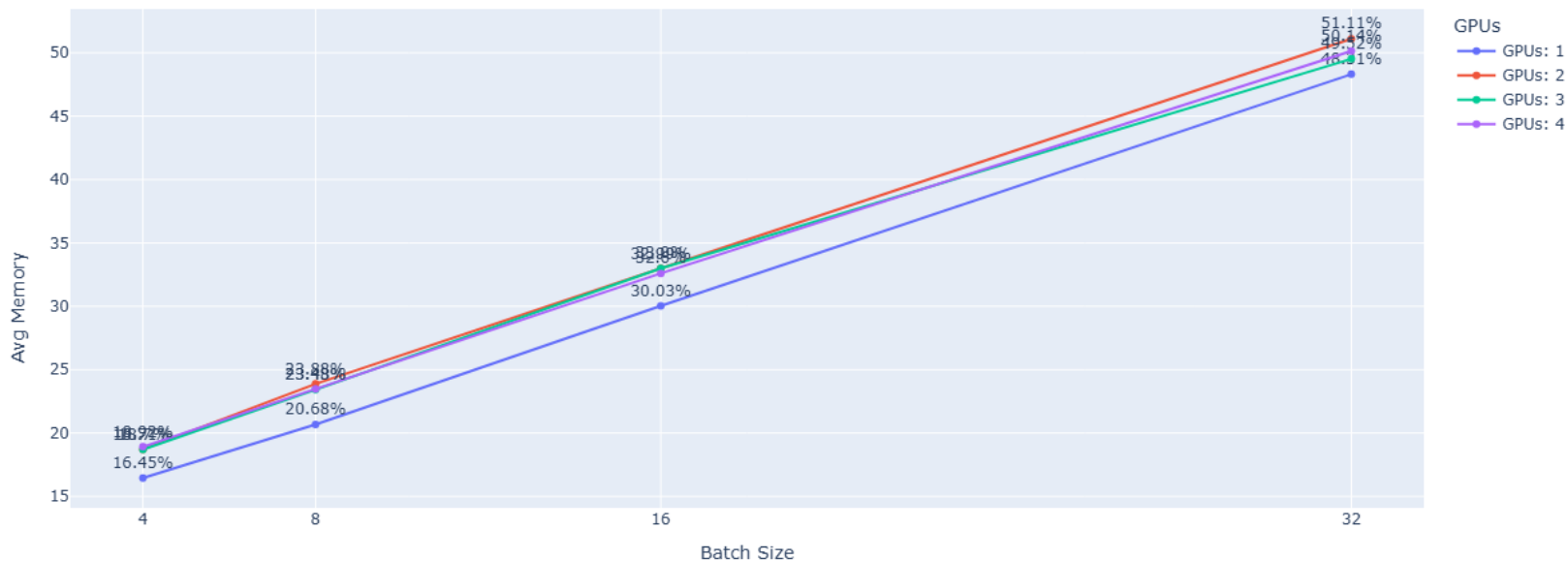
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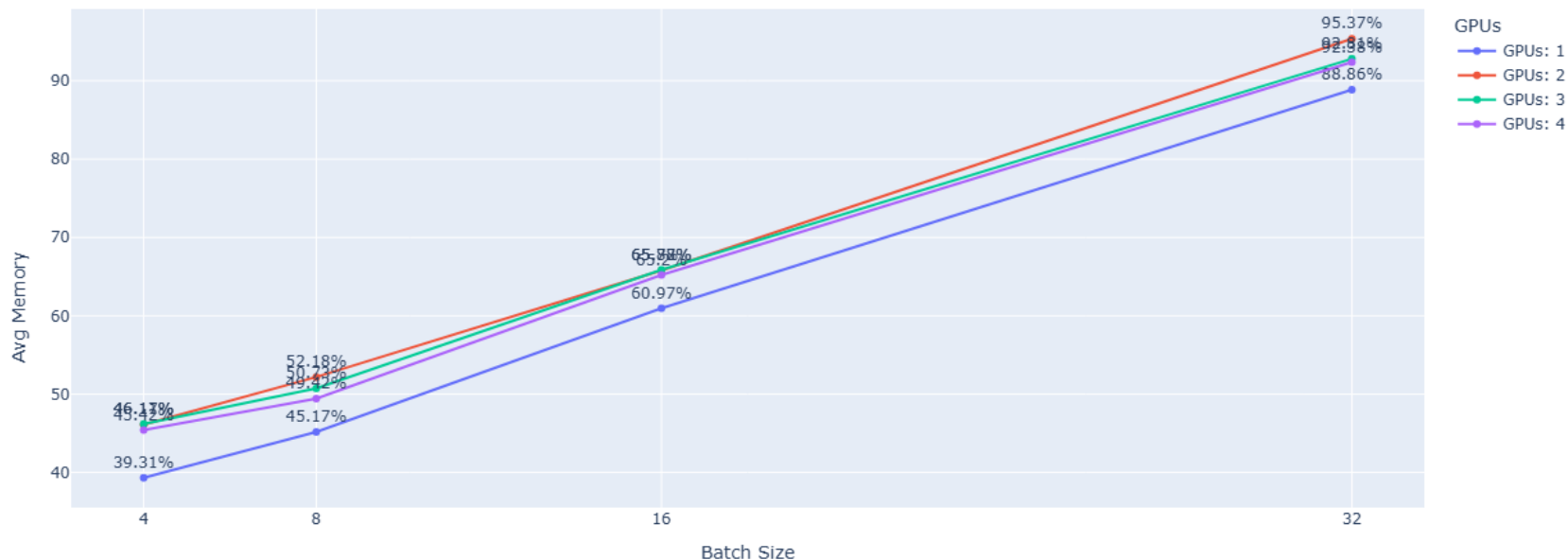
Memory Usage – nvidia-smi 1B

Model: meta-llama-3.2-1B - Average Memory Usage (%) vs Batch Size - 30 Epochs



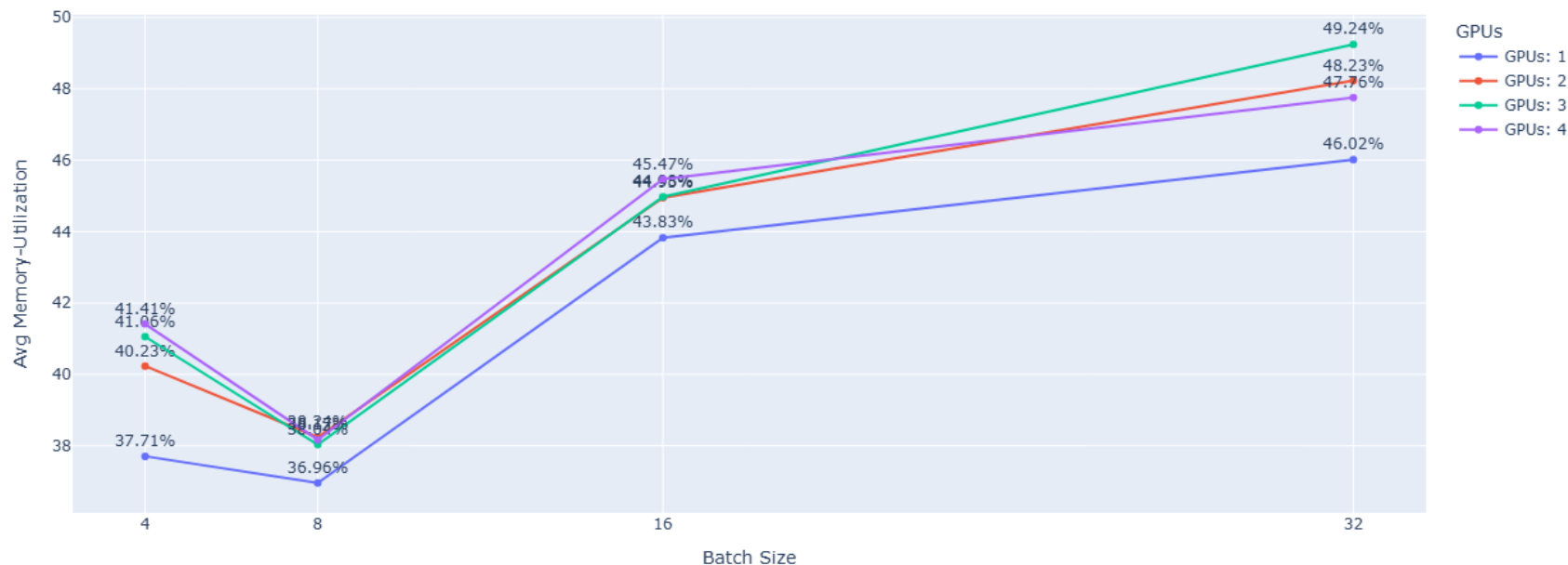
Memory Usage – nvidia-smi 3B

Model: meta-llama-3.2-3B - Average Memory Usage (%) vs Batch Size - 30 Epochs



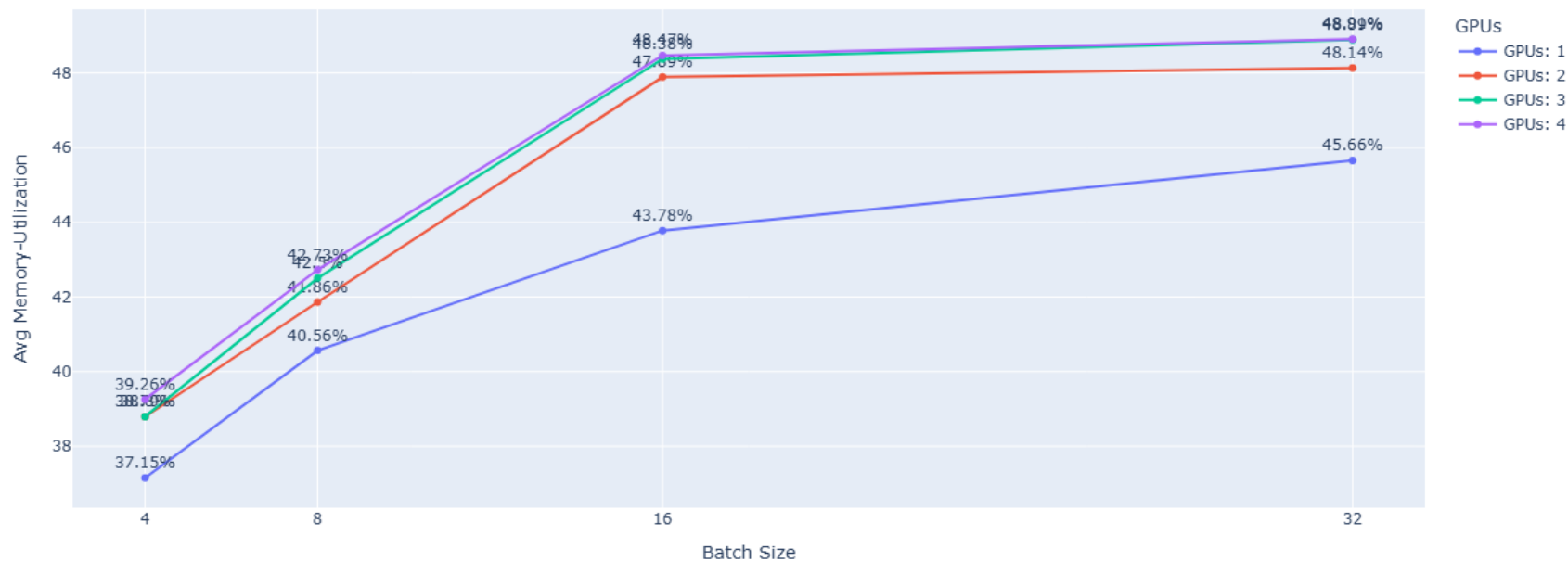
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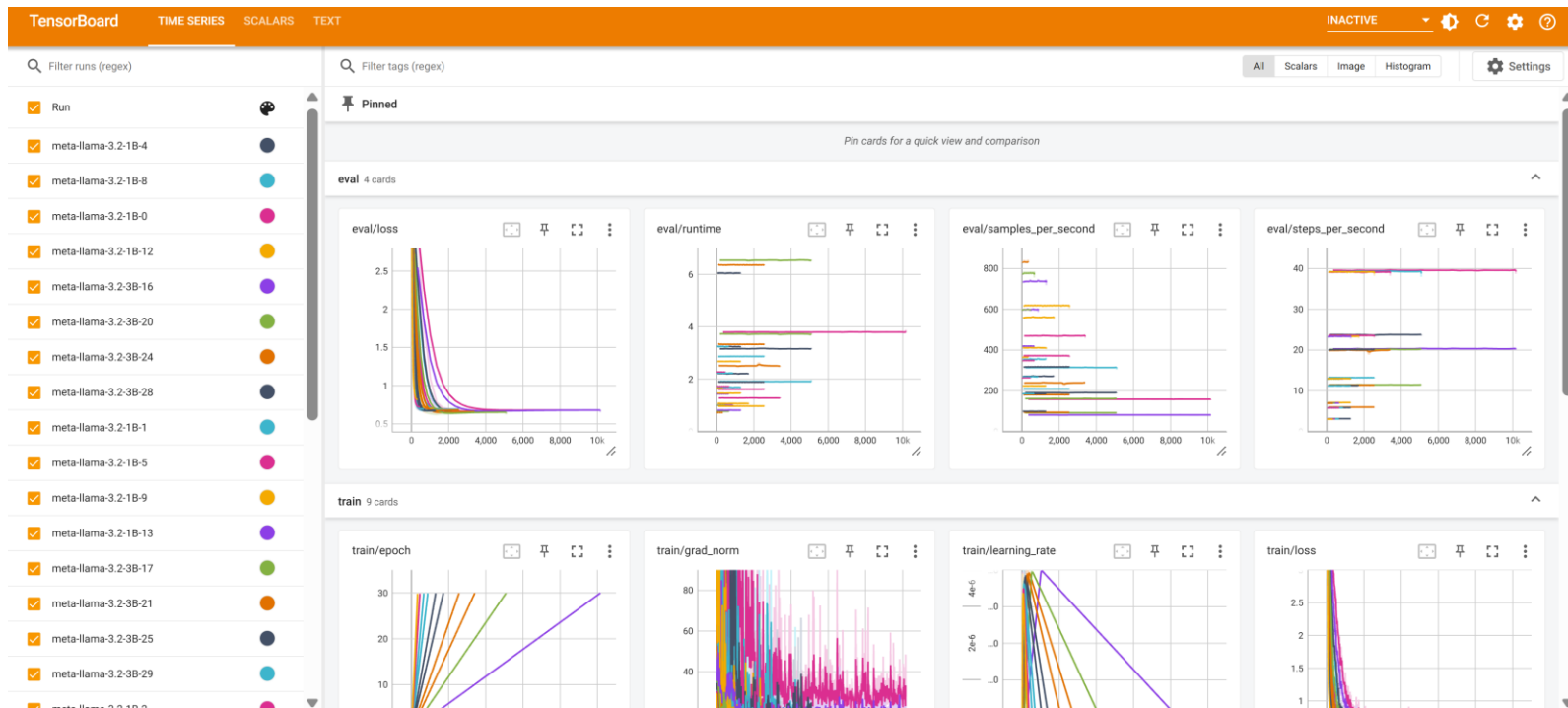
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Tensorboard - Parameters

`report_to="tensorboard"` By setting the value to "tensorboard", you instruct the training system to send all gathered metrics (like loss, accuracy, runtime, etc.) to TensorBoard. The system then generates the necessary **event files** that the TensorBoard application reads and visualizes.

`logging_strategy="steps"` With the value set to "steps", the system records the metrics **after a certain number of training steps** are completed. **Note:** This is usually accompanied by another parameter (e.g., `logging_steps=X`) `logging_steps=10`,

`logging_dir=tensorboard_log_dir` Specifies the local directory where the log files are stored.

`logging_first_step=True` Ensures that metrics are logged **immediately** after the very **first training step** is completed.

```
# 1. Get the unique run name from the output directory (e.g., 'meta-llama-3.2-1B-0')
# This name includes the model and the unique Task ID (which maps to GPU/Batch combo).
unique_run_name = os.path.basename(cfg.output_dir)

# 2. Define the final path for the logs
# Logs will be saved, e.g., to './tb_logs/meta-llama-3.2-1B-0'
tensorboard_log_dir = os.path.join("./tb_logs", unique_run_name)

# Training configuration
training_args = TrainingArguments(
    output_dir=cfg.output_dir,
    per_device_train_batch_size=cfg.batch_size,
    per_device_eval_batch_size=cfg.batch_size,
    num_train_epochs=cfg.num_epochs,
    learning_rate=cfg.learning_rate,
    weight_decay=cfg.weight_decay,
    gradient_accumulation_steps=cfg.gradient_accumulation_steps,
    warmup_ratio=cfg.warmup_ratio,
    fp16=False,
    bf16=True,
    report_to="tensorboard",
    logging_strategy="steps",
    logging_dir=tensorboard_log_dir,
    logging_steps=10,
    save_total_limit=2,
    logging_first_step=True,
    eval_strategy="epoch",
    save_strategy="epoch",
    load_best_model_at_end=True,
    metric_for_best_model="eval_loss",
    greater_is_better=False,
)
```

Thank you for your attention.

Questions?



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και Τεχνητής Νοημοσύνης