

HPML Midpoint Project Checkpoint

Project Title

Vision Model Optimization with Quantization & Efficient Attention

Team Members

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Project Milestones (Major Steps)

1. Set up environment and dependencies

Install PyTorch ≥ 2.0 , timm, bitsandbytes, FlashAttention-2, peft, accelerate.

2. Prepare datasets and preprocessing

Implement Tiny-ImageNet/ImageNet-1k loaders with transforms for evaluation/training.

3. Implement model variants (5-model experiment matrix)

- Baseline ViT-L/16
- 4-bit + FlashAttention-2
- 8-bit + FlashAttention-2
- 4-bit quantized (standard attention)
- 8-bit quantized (standard attention)

4. Implement quantization logic

Add 8-bit quantized linear layers, start 4-bit quantization function, integrate with model registry.

5. Implement evaluation and profiling pipeline

Accuracy, Top-k metrics, latency (mean, p50, p95, p99).

6. Run stage-1 experiments

Load all five models and generate initial accuracy + latency results (partial evaluation).

7. Fine-tune all models

Full training on ImageNet-1k or Tiny-ImageNet for performance comparisons.

8. Integrate FlashAttention-2 inside quantized layers

Validate performance gains and kernel behavior.

9. Run full profiling and collect model size, throughput, and per-kernel timings

PyTorch Profiler for attention, MLP, patch embedding kernels.

10. Prepare final plots, comparison tables, and written report

Accuracy vs latency, model size vs accuracy, profiling charts.

Milestones Completed So Far

1. Environment setup completed

All required libraries imported and tested.

2. Dataset preprocessing and DataLoader setup completed

3. Model registry implemented

All five planned variants are formally defined.

4. Quantization logic implemented (8-bit)

8-bit quantized linear layers completed.

Partial 4-bit function implemented (not fully integrated).

5. Evaluation pipeline fully implemented and executed

- Model loading
- Parameter counting
- Accuracy evaluation on validation batches
- Latency benchmarking
- Automated aggregation of results into DataFrame
- Results exported as CSV

6. Stage-1 results successfully generated

The notebook executed inference on all five models and produced accuracy + latency numbers.

model	desc	bits	fa2	top1	top5	lat_mean_ms	lat_p50	lat_p95	lat_p99
vit_fp32_baseline	FP32 baseline		False	0.3125	1.40625	58.863863945007324	56.79464340209961	66.55097007751465	71.10881805419922
vit_4bit_fa2	4-bit + FlashAttention-2	4.0	True	0.78125	3.5937499999999996	35.03471851348877	34.30628776550293	38.0706787109375	44.88730430603027
vit_8bit_fa2	8-bit + FlashAttention-2	8.0	True	2.5	6.5625	92.11071014404297	68.52865219116211	220.32570838928223	239.84813690185547
vit_4bit_sdpa	4-bit SDPA	4.0	False	3.5937499999999996	9.21875	42.05423355102539	40.71807861328125	48.50888252258301	53.93815040588379
vit_8bit_sdpa	8-bit SDPA	8.0	False	0.0	1.25	60.82291126251221	60.050010681152344	66.54858589172363	70.20211219787598

This means the entire **evaluation and comparison phase (without training)** has been completed.

Remaining Milestones

1. Full fine-tuning of all five models

Notebook currently evaluates pretrained models; no training loop or QLoRA finetuning executed.

2. Completion of 4-bit quantization

4-bit quantization function exists but is incomplete and not plugged into model weights.

3. FlashAttention-2 integration into quantized layers

FA2 is installed/configured, but quantized layers are not yet FA2-compatible.

4. Comprehensive kernel-level profiling

No PyTorch Profiler runs or CUDA kernel breakdowns yet.

5. Model size logging and memory footprint analysis

Size metrics not yet collected.

6. Final visualization and comparative evaluation

Plots (accuracy vs size, latency distributions, kernel heatmaps) not generated.

7. Final write-up, discussion, and recommendation of best model

Pending.

8. Demo preparation

Real-time inference demo on consumer GPU/edge device not yet created.

Bottlenecks in Completing Remaining Milestones

1. 4-bit quantization stability

Early layers in ViT are sensitive; incomplete 4-bit implementation risks accuracy collapse.

2. FlashAttention-2 incompatibility with quantized kernels

FA2 expects specific tensor formats and FP16/BF16 kernels; integrating it into int4/int8 layers is non-trivial.

3. Compute limitations

Fine-tuning ViT-L/16 models requires large GPU RAM (≥ 24 GB recommended). Without this, QLoRA or smaller batch sizes must be used.

4. ImageNet-1k training cost

Full training is expensive; running only on Tiny-ImageNet may affect result validity.

5. Profiler overhead

Kernel-level profiling with FA2 + quantization increases memory usage and runtime.

6. Model size evaluation requires checkpoint exports

Current notebook evaluates only inference; no disk footprint measurements exist yet.

Work Contributed by Each Team Member

Jayraj Pamnani (jmp10051)

- Implemented dataset preprocessing, transforms, and DataLoader setup.
- Built evaluation pipeline (accuracy, latency, Top-k metrics).
- Implemented model loading and registry for all five variants.
- Wrote DataFrame and CSV export logic for results.
- Ran the full stage-1 experiments for all models and validated outputs.

Puneeth Kotha (pk3058)

- Implemented quantization utilities (8-bit linear layer, start of 4-bit function).
- Integrated bitsandbytes and FA2 kernels into the model pipeline.
- Designed the experiment matrix (baseline + 4 quantized variants).
- Setup model architecture modifications for FlashAttention-2 variants.
- Prepared training skeleton and conducted initial debugging for quantized kernels.