Phantora: Maximizing Code Reuse in Simulation-based Machine Learning System Performance Estimation

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Meta

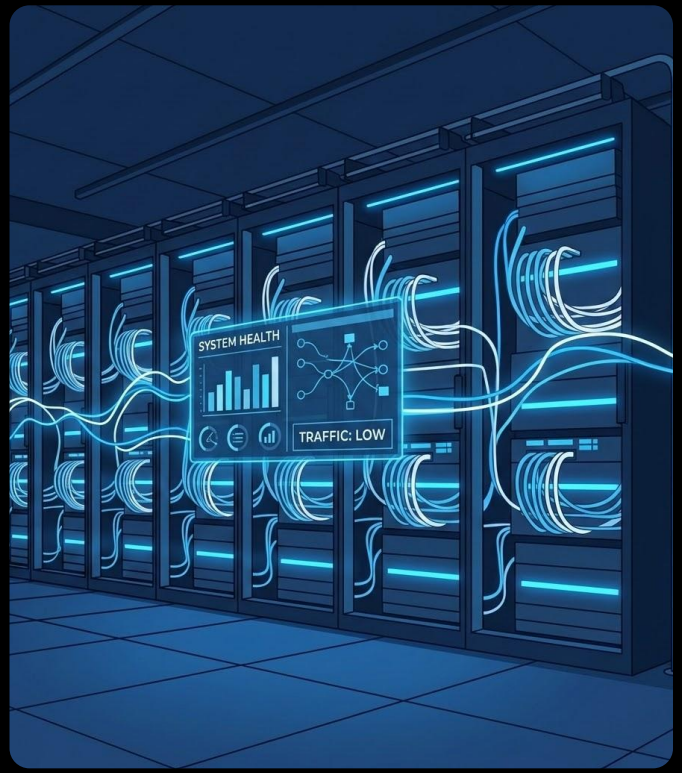
Performance Prediction is Critical

Economic Constraint

Real-world tuning costs enormous resources

Resource Planning

Provisioning decisions before hardware availability



Performance Modelling is Difficult

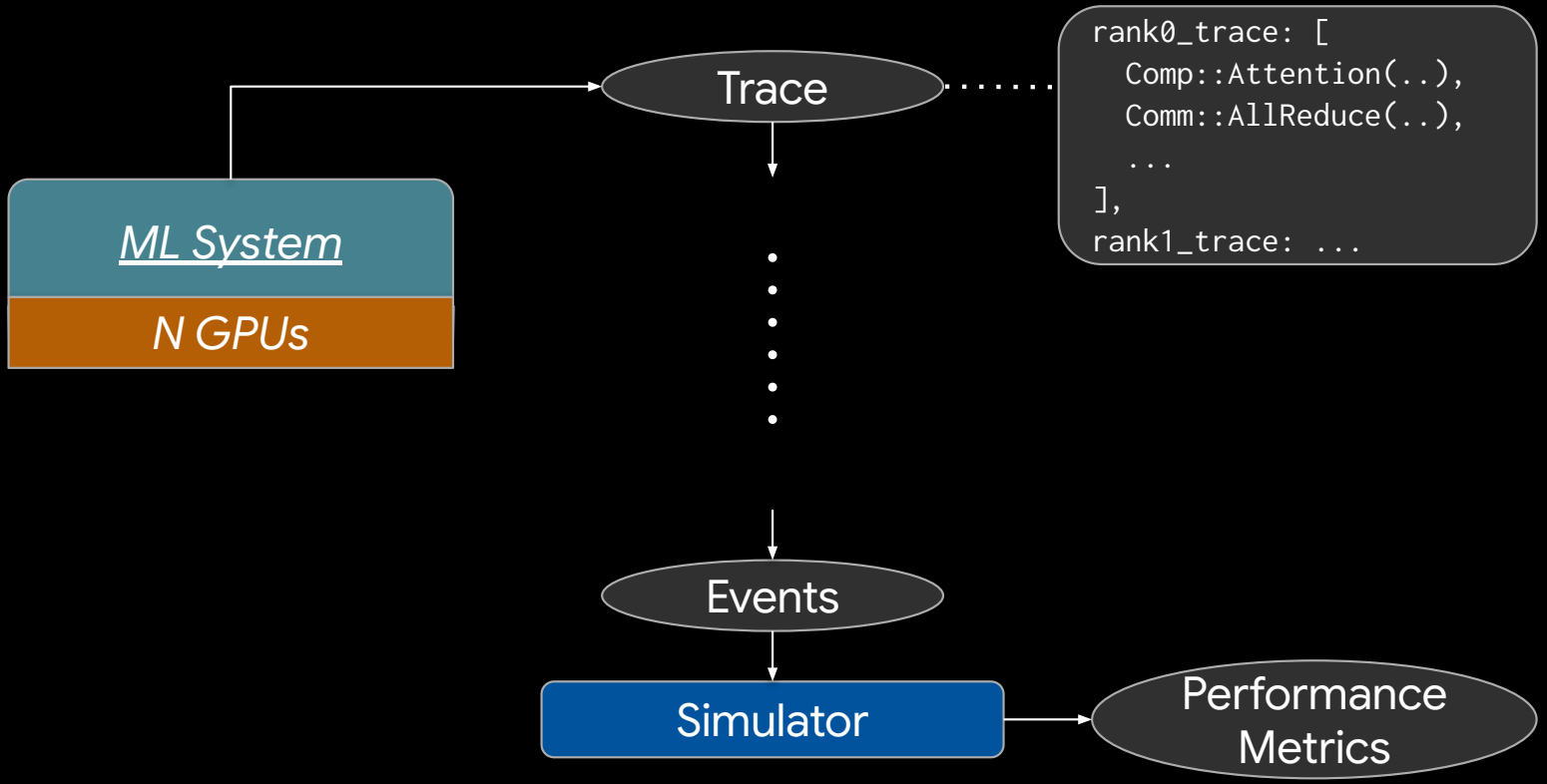
Complex configuration space

- Batch size
- Parallelization

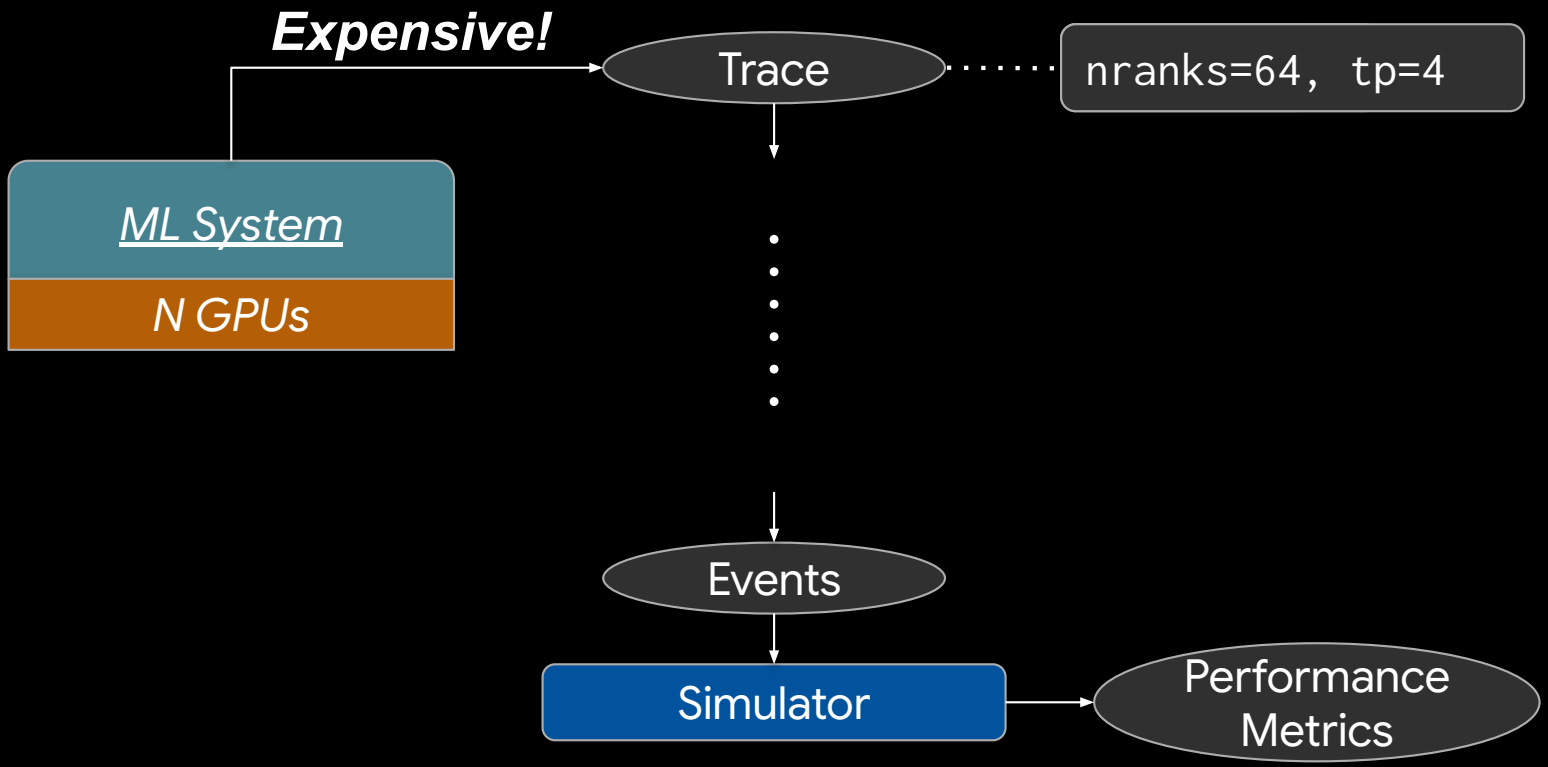
Framework-specific features

e.g., Selective activation recomputation

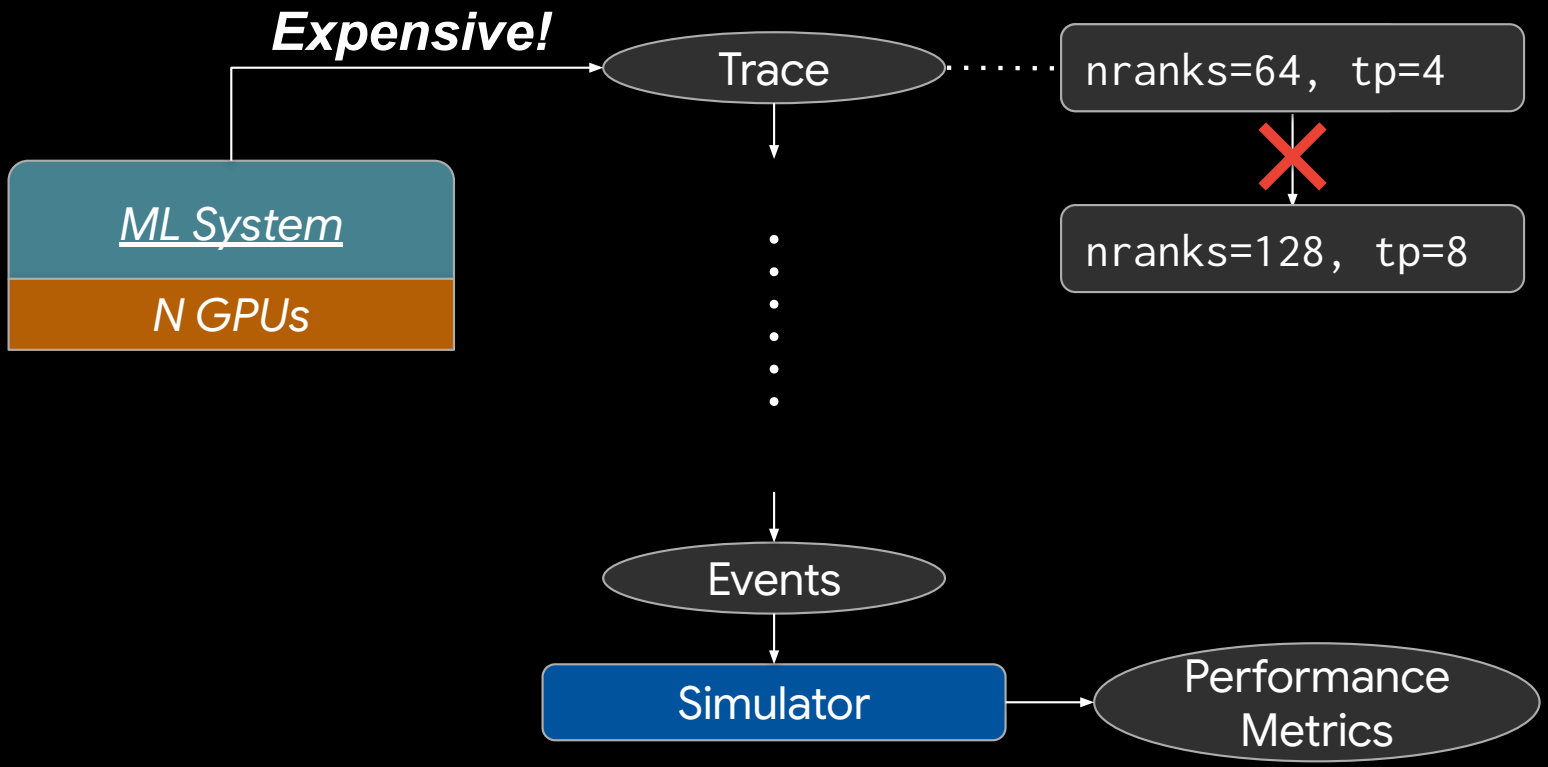
Trace-based Simulation



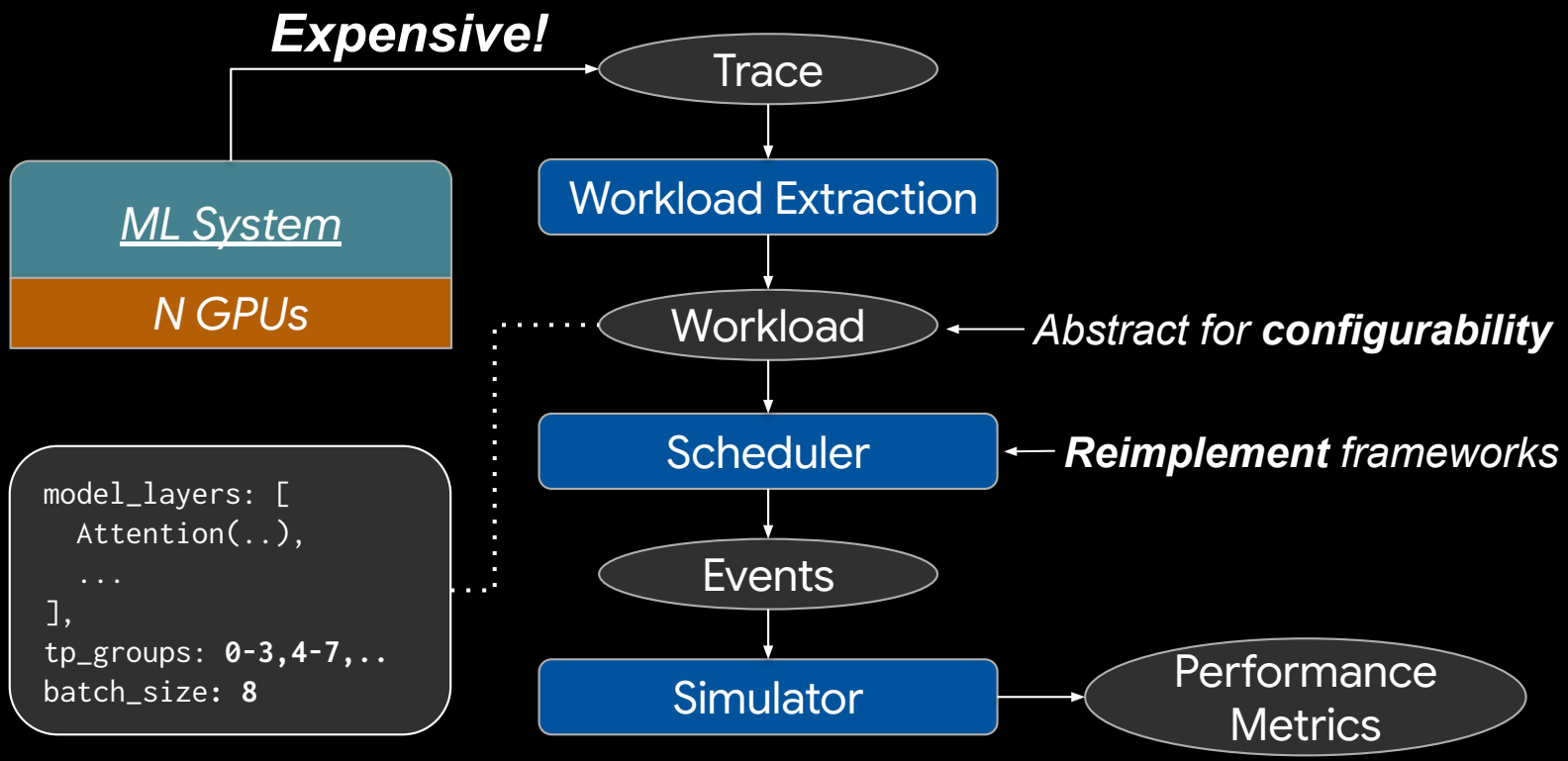
Trace-based Simulation



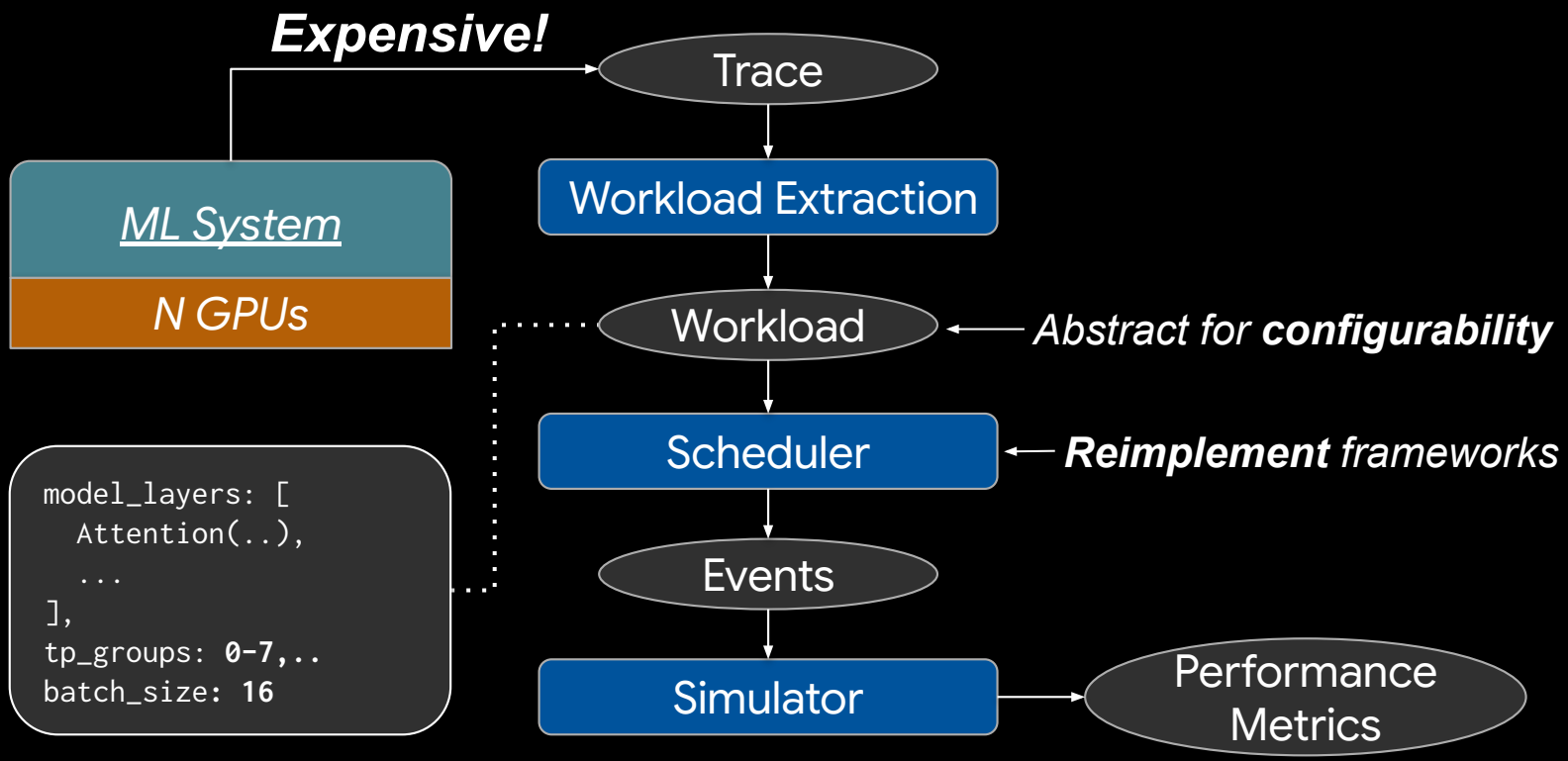
Trace-based Simulation



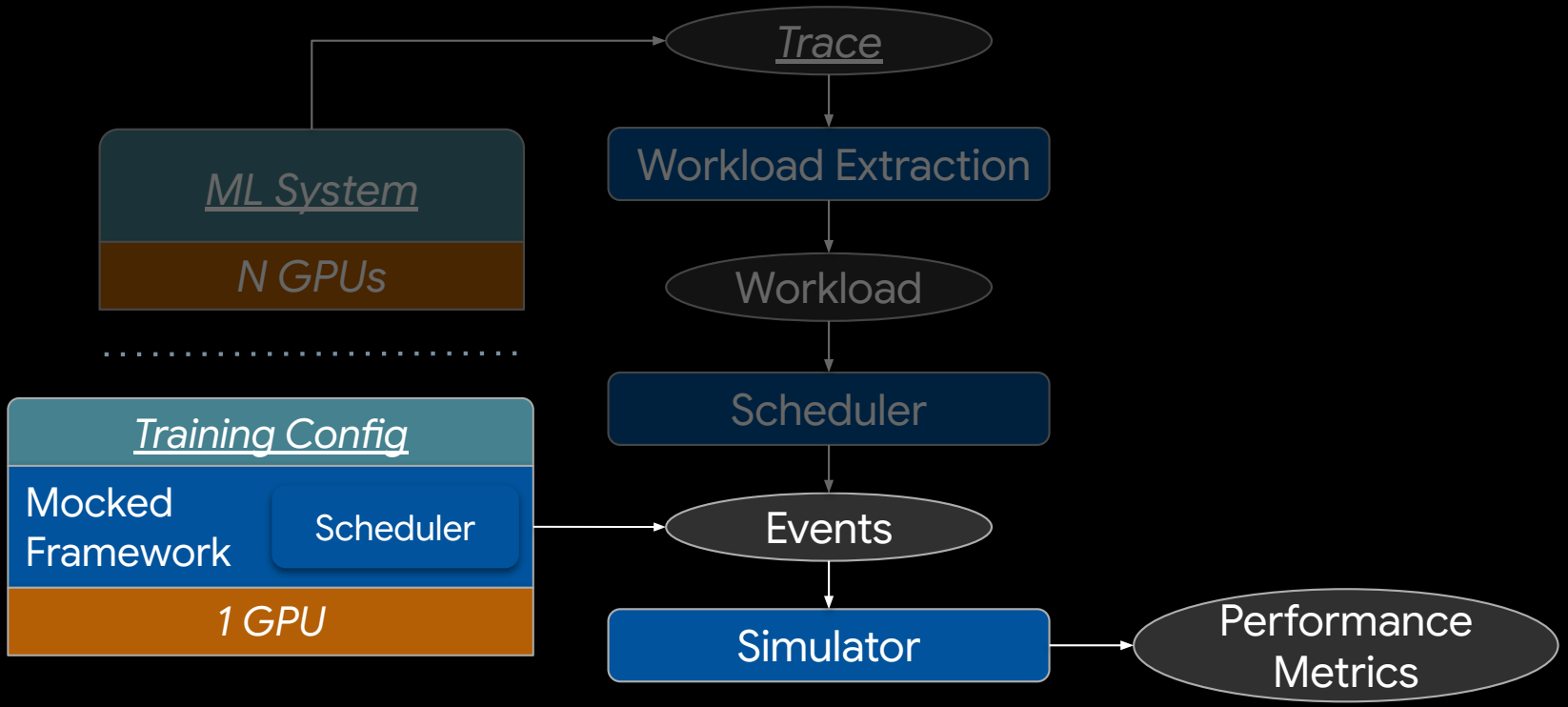
Reimplementation in Trace-based Simulation



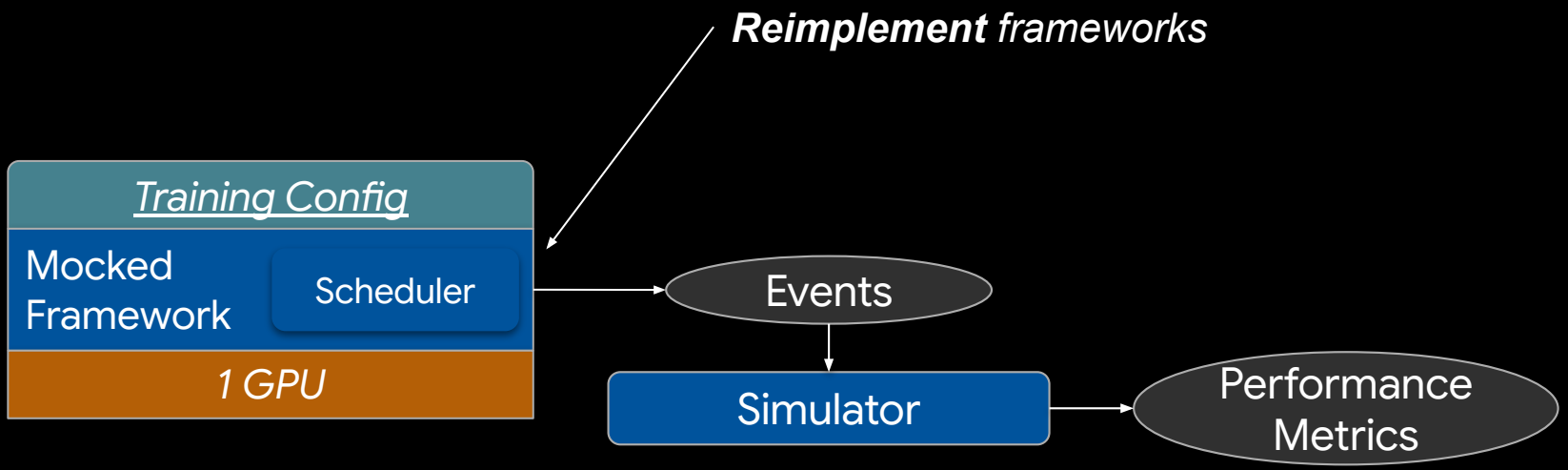
Reimplementation in Trace-based Simulation



Mocked Frameworks (SimAI, NSDI '25)



Reimplementation in Mocked Frameworks

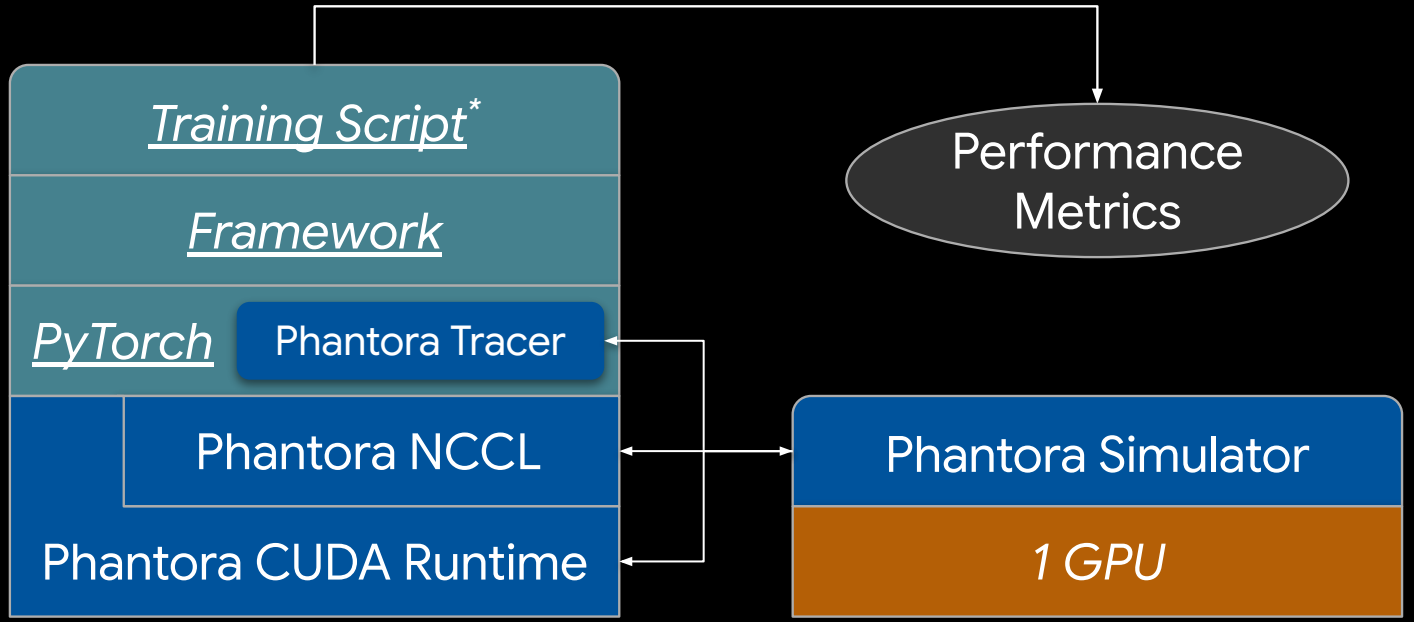


Can we directly ***reuse*** the framework's code for simulation?

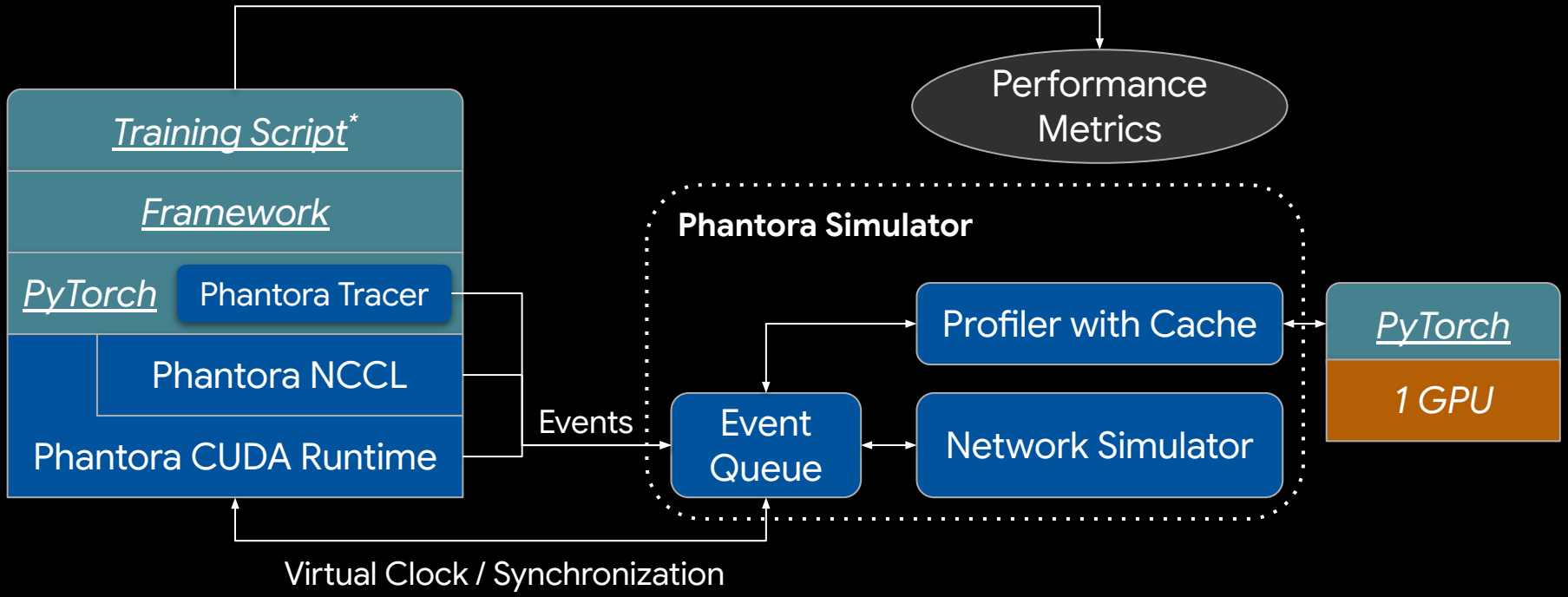
Can we directly ***reuse*** the framework's code for simulation?

- Less maintenance effort
- Capture specific framework features

Phantora: Hybrid Simulation



Phantora: Hybrid Simulation



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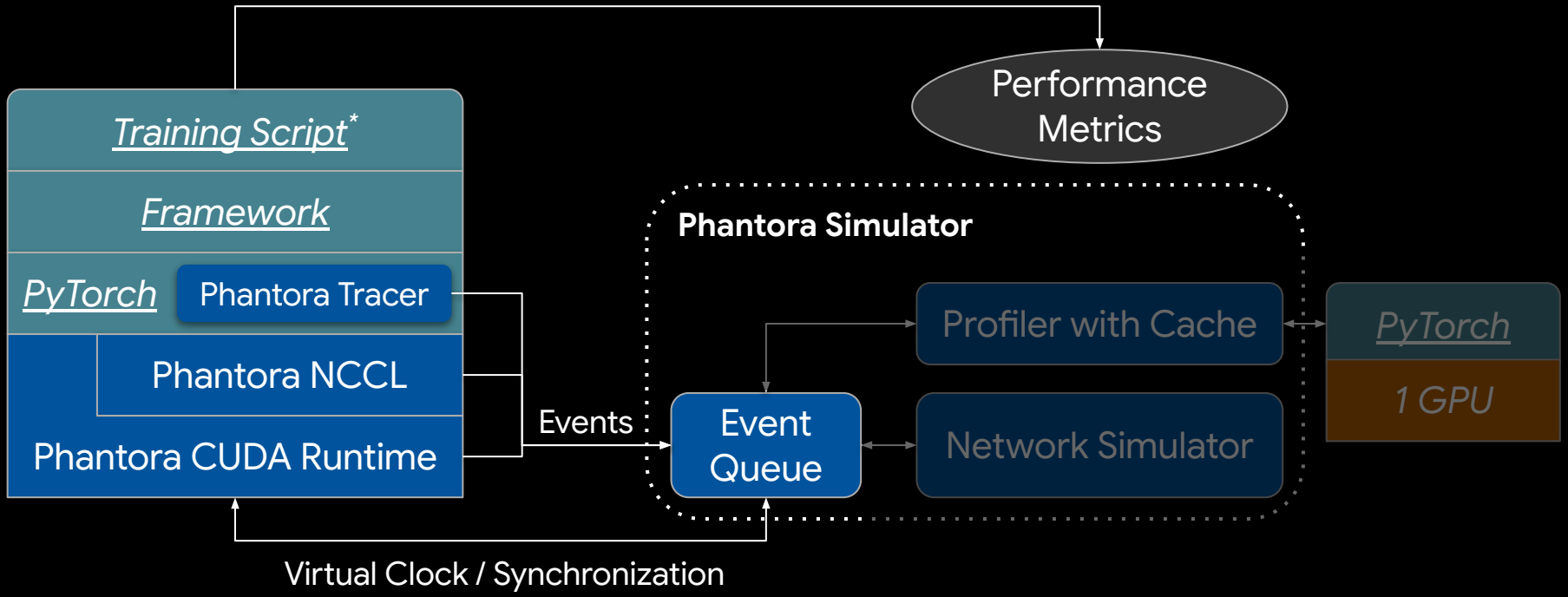
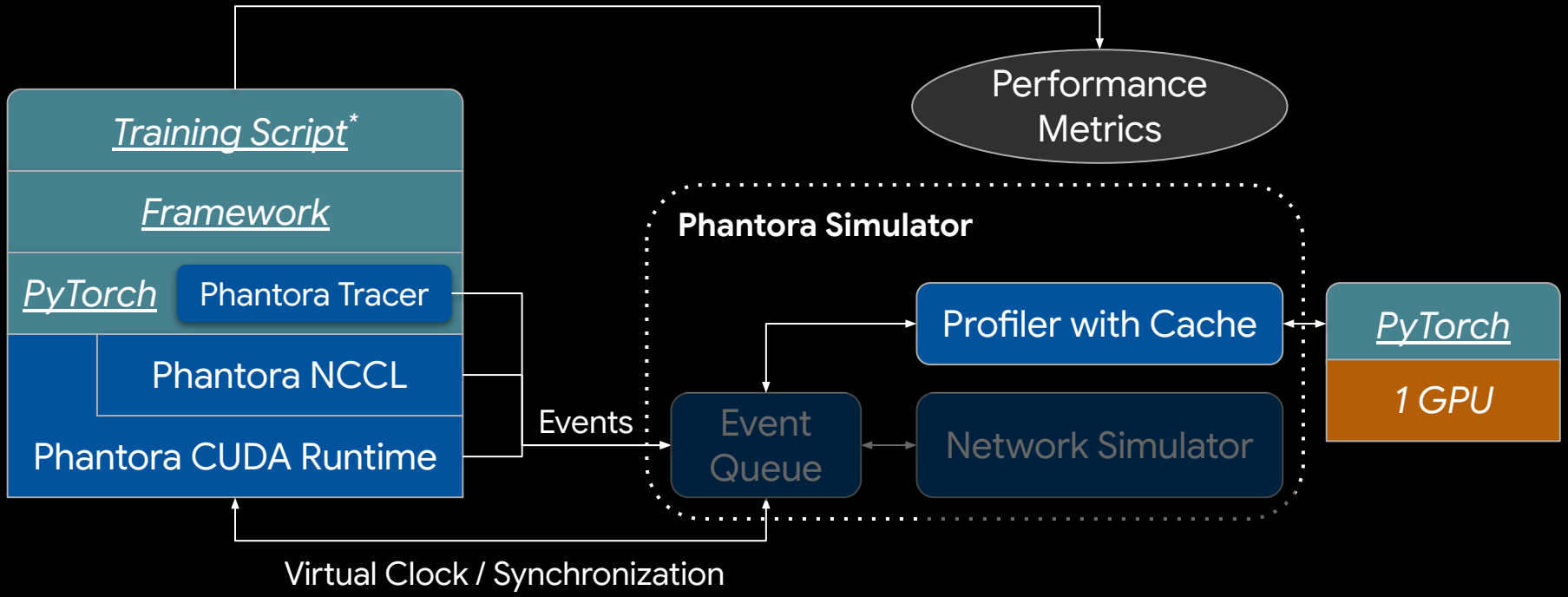
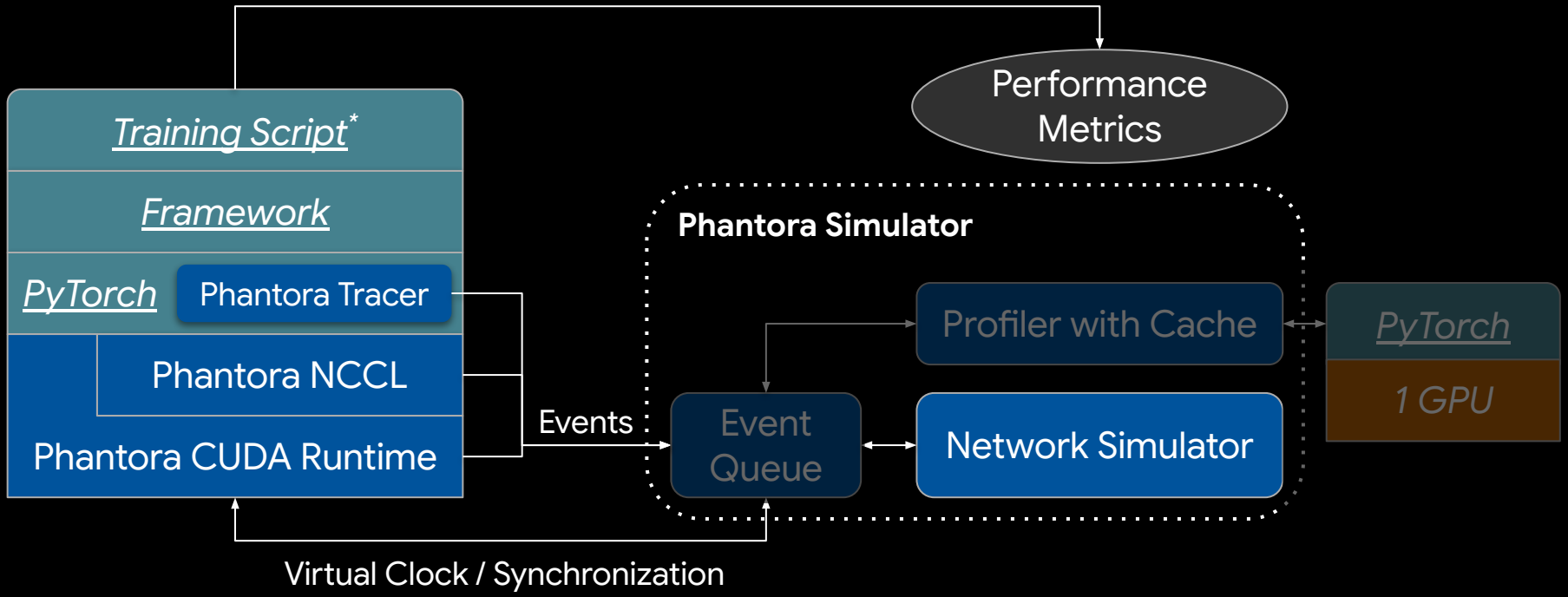


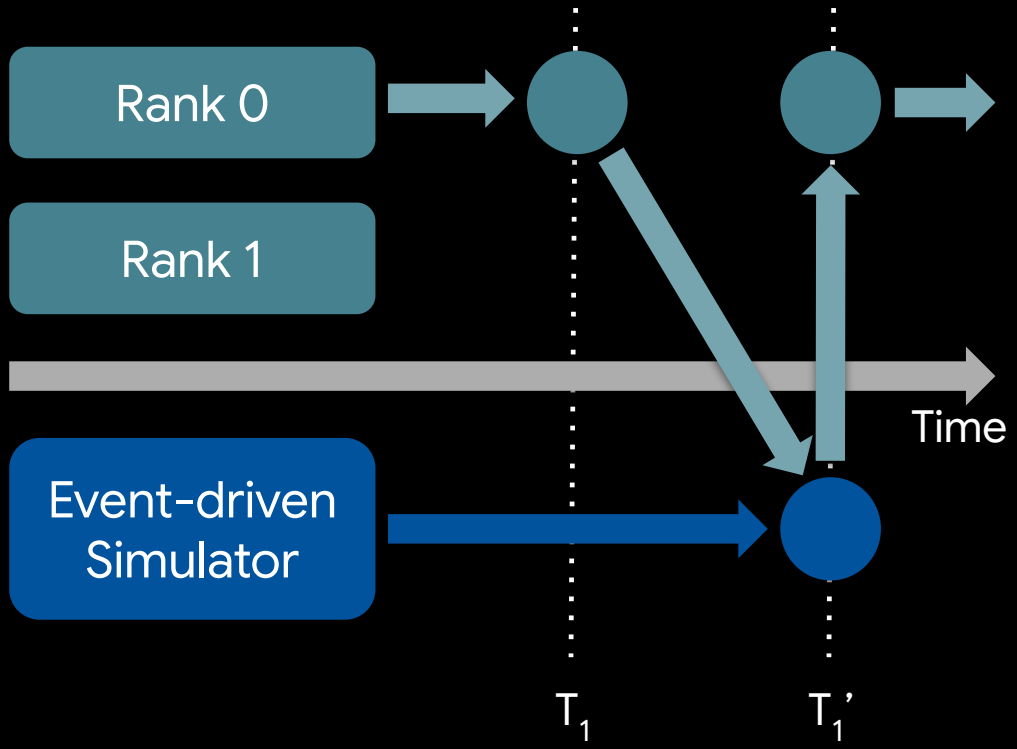
Phantora: Hybrid Simulation



Phantora: Hybrid Simulation



Problem: Past Events



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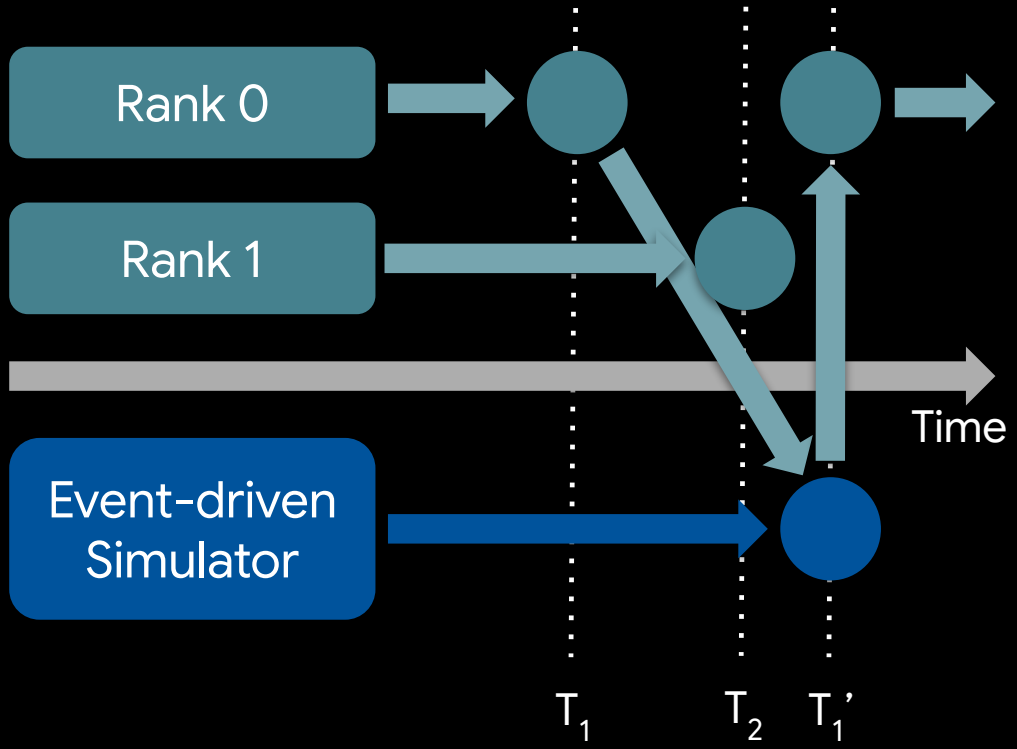

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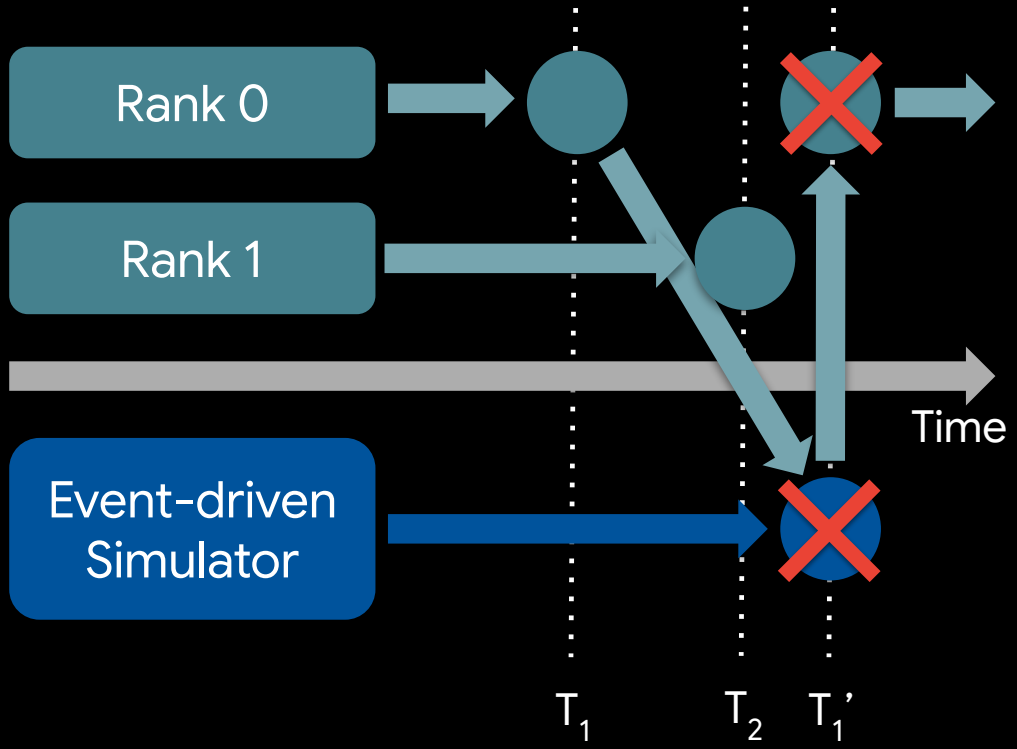
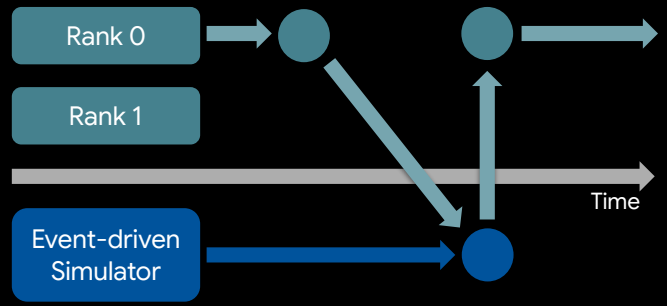
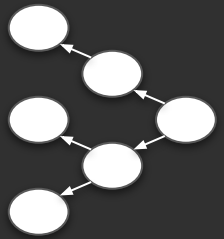


Solution: Rollback



Event Queue with Dependency Graph

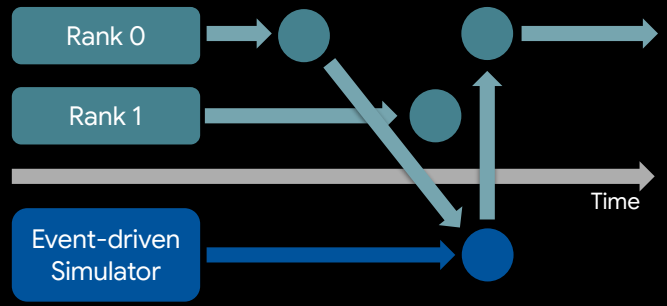


Network Simulator

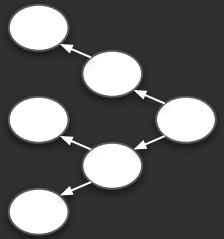
Throughput History (T_1, T_1')

Network Event Queue

Solution: Rollback



Event Queue with Dependency Graph



(1) Network event at T_2 between T_1 and T_1'

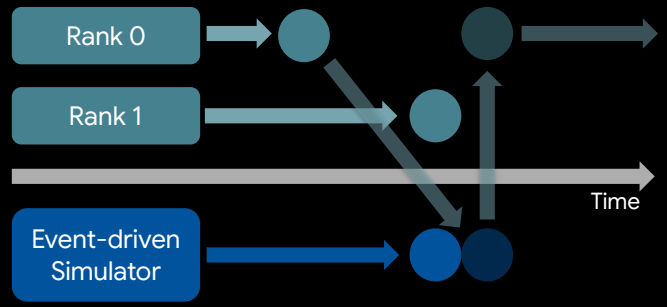
T_2

Network Simulator

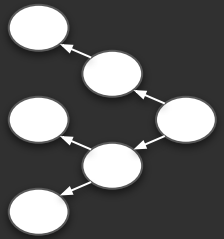
Throughput History (T_1, T_1')

Network Event Queue

Solution: Rollback



Event Queue with Dependency Graph



(1) Network event at T_2 between T_1 and T_1'

T_2

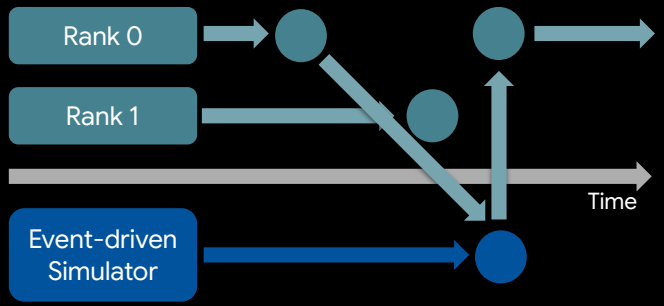
Network Simulator

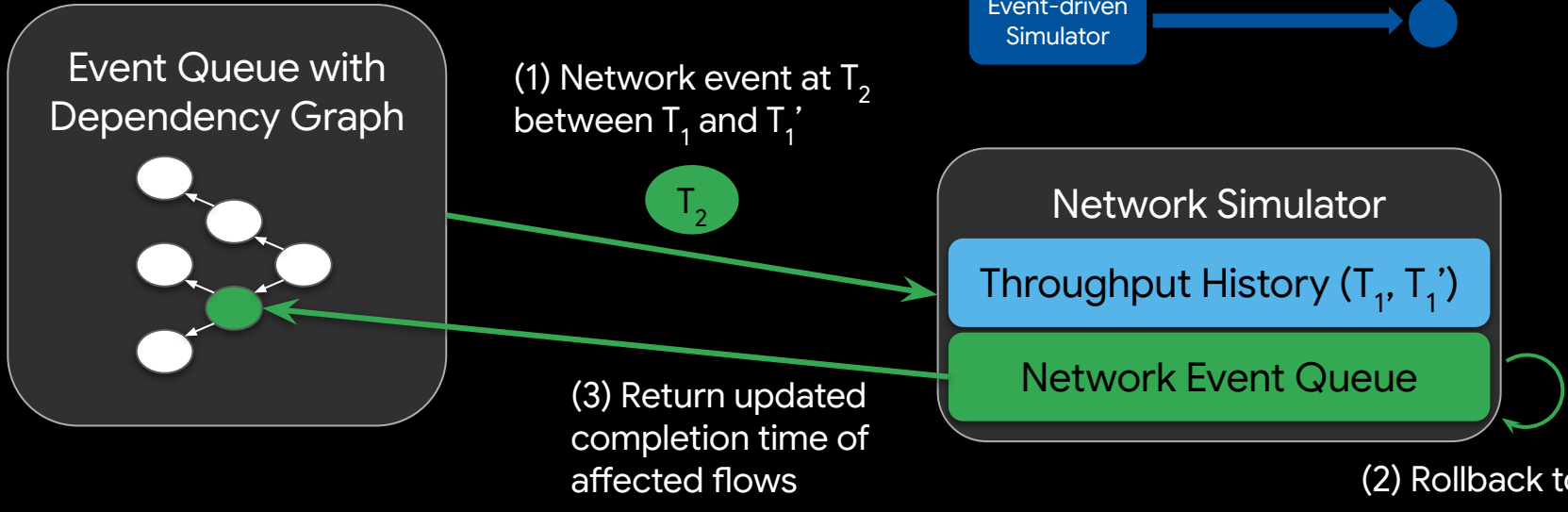
Throughput History (T_1, T_1')

Network Event Queue

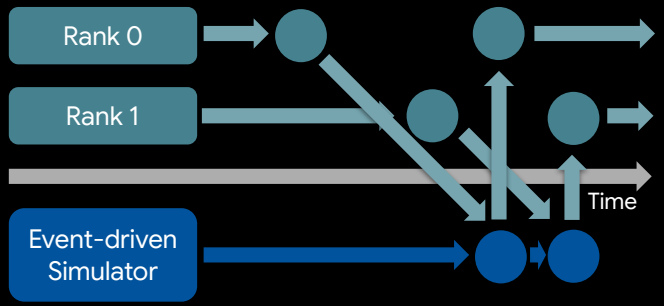
(2) Rollback to T_2

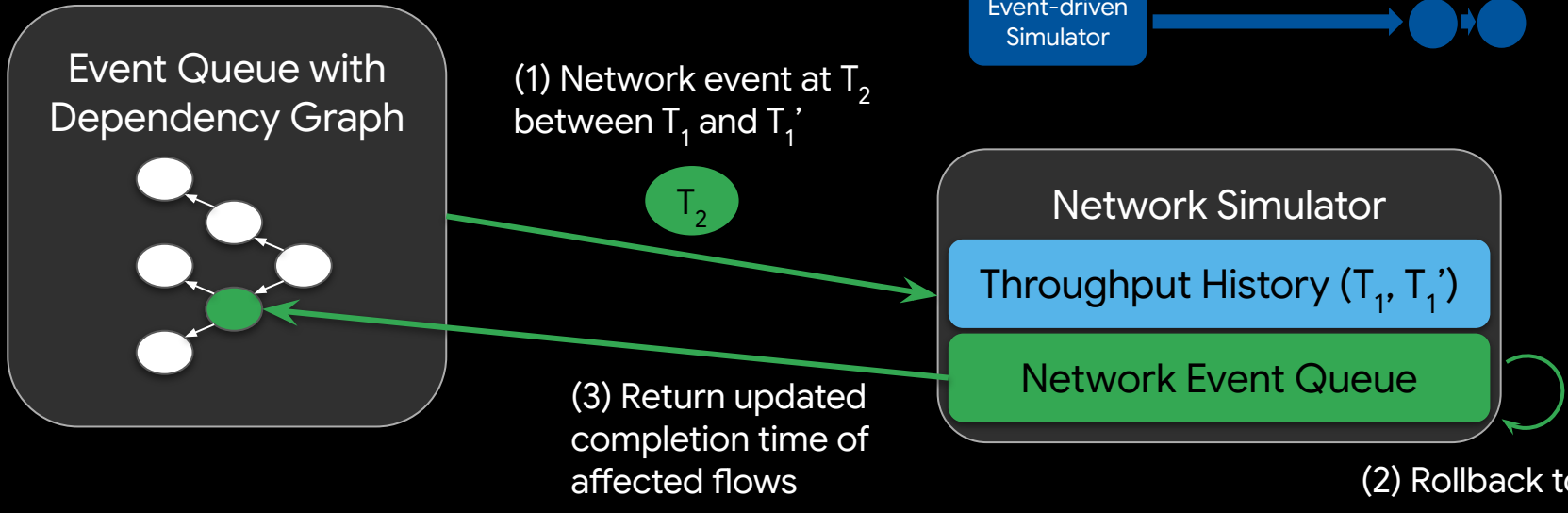
Solution: Rollback



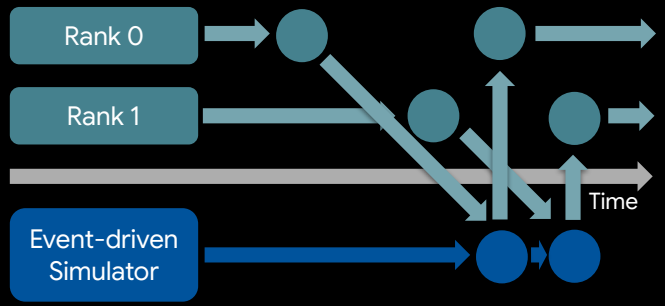


Solution: Rollback





Solution: Rollback



Event Queue with Dependency Graph

(1) Network event at T_2 between T_1 and T_1'

T_2

Network Simulator

Throughput History (T_1, T_1')

Network Event Queue

(2) Rollback to T_2

(3) Return updated completion time of affected flows

(4) Update previous events' start & completion time

Emulation of NCCL & CUDA Runtime behaviors

GPU Memory Tracking

Track cudaMalloc & cudaFree, returning errors when exceeding configured capacity.

CUDA Streams & Events

Precise management of dependencies and synchronizations, including overlaps.

NCCL Communications

Handling complex synchronizations required for NCCL setup and collective operations.

Small User Effort: Unmodified frameworks

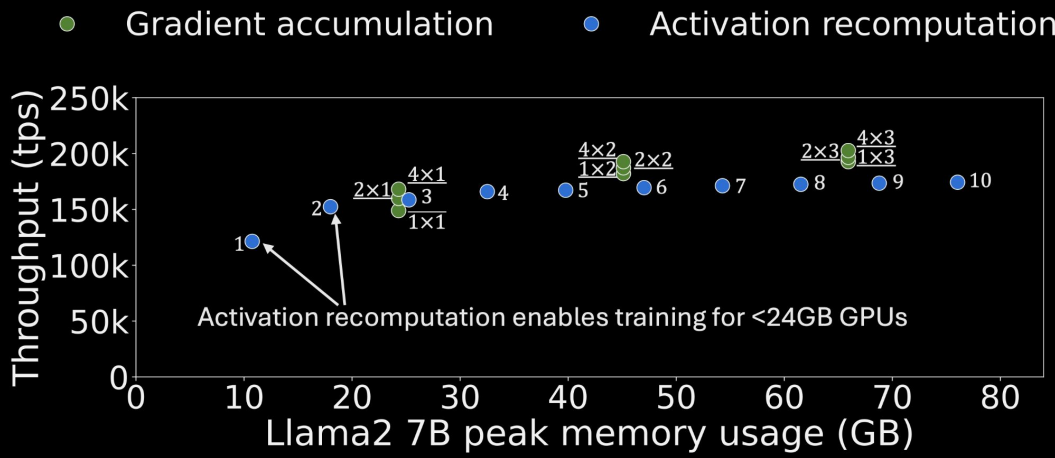
```
## install-requirements.sh
pip install megatron-core deepspeed torchtitan

## train.py
+from phantora_utils import (
+  enable_function_tracer,
+  disable_function_tracer,
+)

......

if __name__ == “__main__”:
+   enable_function_tracer()
    train()
+   disable_function_tracer()
```

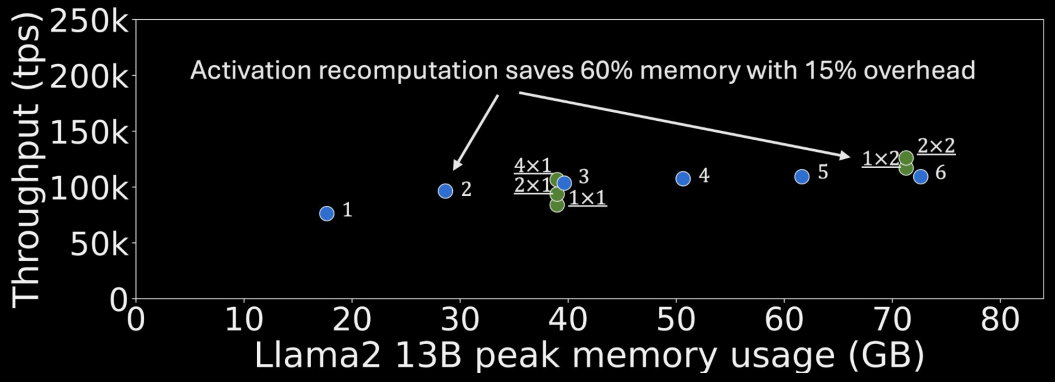
Small User Effort: Logging as-is

```
logger.info(
    f"step: {train_state.step:2}  "
    f"loss: {global_avg_loss:7.4f}  "
    f"memory: {gpu_mem_stats.max_reserved_gib:5.2f}GiB"
    f"({gpu_mem_stats.max_reserved_pct:.2f}%)  "
    f"wps: {round(wps):,}  "
    f"mfu: {mfu:.2f}%{color.reset}"
)
```

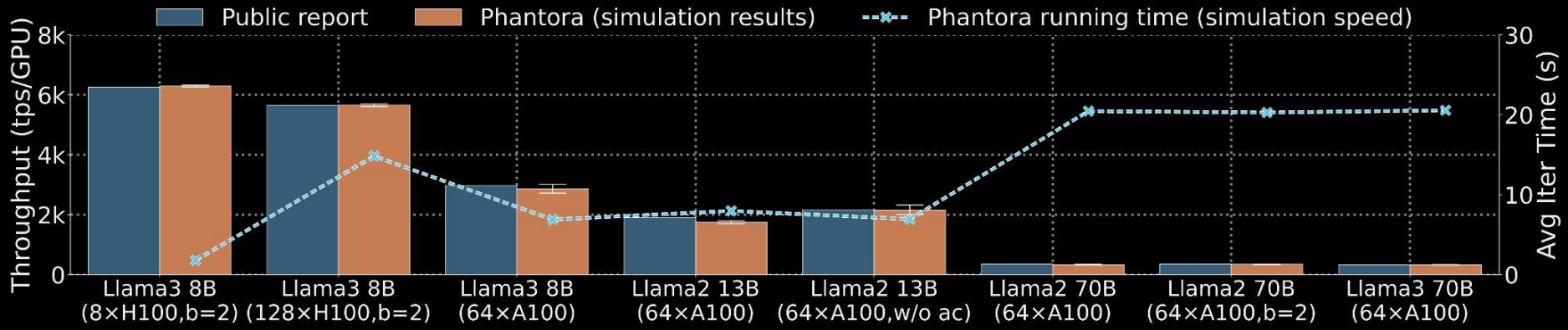
```
[rank0]:2024-09-19 14:46:00,990 - root - INFO - step:  5  loss:  0.0000
memory: 41.56GiB(51.95%)  wps: 2,876  mfu: 53.38%
[rank0]:2024-09-19 14:46:37,206 - root - INFO - step: 10  loss:  0.0000
memory: 41.56GiB(51.95%)  wps: 2,887  mfu: 53.60%
[rank0]:2024-09-19 14:47:13,114 - root - INFO - step: 15  loss:  0.0000
memory: 41.56GiB(51.95%)  wps: 2,902  mfu: 53.87%
[rank0]:2024-09-19 14:47:47,622 - root - INFO - step: 20  loss:  0.0000
memory: 41.56GiB(51.95%)  wps: 2,904  mfu: 53.90%
[rank0]:2024-09-19 14:48:22,964 - root - INFO - step: 25  loss:  0.0000
memory: 41.56GiB(51.95%)  wps: 2,905  mfu: 53.92%
[rank0]:2024-09-19 14:48:58,821 - root - INFO - step: 30  loss:  0.0000
memory: 41.56GiB(51.95%)  wps: 2,904  mfu: 53.90%
[rank0]:2024-09-19 14:49:34,490 - root - INFO - step: 35  loss:  0.0000
memory: 41.56GiB(51.95%)  wps: 2,886  mfu: 53.57%
[rank0]:2024-09-19 14:50:30,248 - root - INFO - step: 40  loss:  0.0000
memory: 41.56GiB(51.95%)  wps: 2,894  mfu: 53.72%
```

Case Study: Tuning Megatron config under limited VRAM



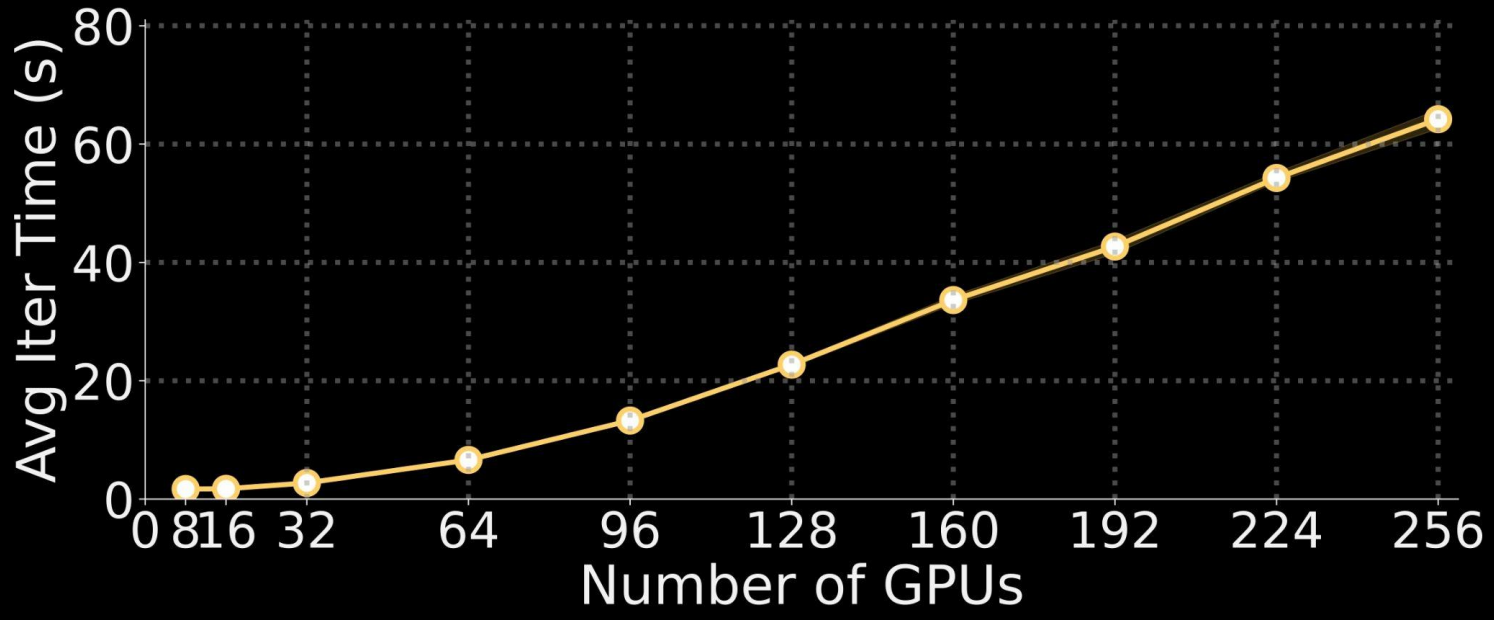


Evaluation: Accuracy and Speed



Phantora achieves 2.9% average error and 8.5% maximum error simulating TorchTitan's benchmark with 1×H200 or 1×A100

Evaluation: Scalability



Speed of Phantora (Megatron, Llama2 7B, simulated with 1×H200)

Discussions

- Tuning existing systems and exploring new system ideas (parallelizations, fault tolerance mechanism, hardware co-design, ...)
- Further improve network simulator: Other methods should be able to apply as Phantora still uses standard discrete event simulation.
- Value-dependent performance is challenging to incorporate, especially RL training performance, which heavily depends on response lengths.

Conclusion

- Current ML system performance simulation methods requires framework reimplementation. Direct reuse of framework code is achievable via hybrid simulation.

- Phantora provides high-fidelity estimation with small user effort and almost complete support of framework features.

- Phantora is publicly available at https://github.com/QDelta/Phantora.