

Diagnosing the Interference on CPU-GPU Synchronization Caused by CPU Sharing in Multi-Tenant GPU Clouds

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Building Multi-Tenant GPU Clouds is Still a Challenge

Powered by virtualization, resources, like GPUs and CPUs, can be shared among instances!

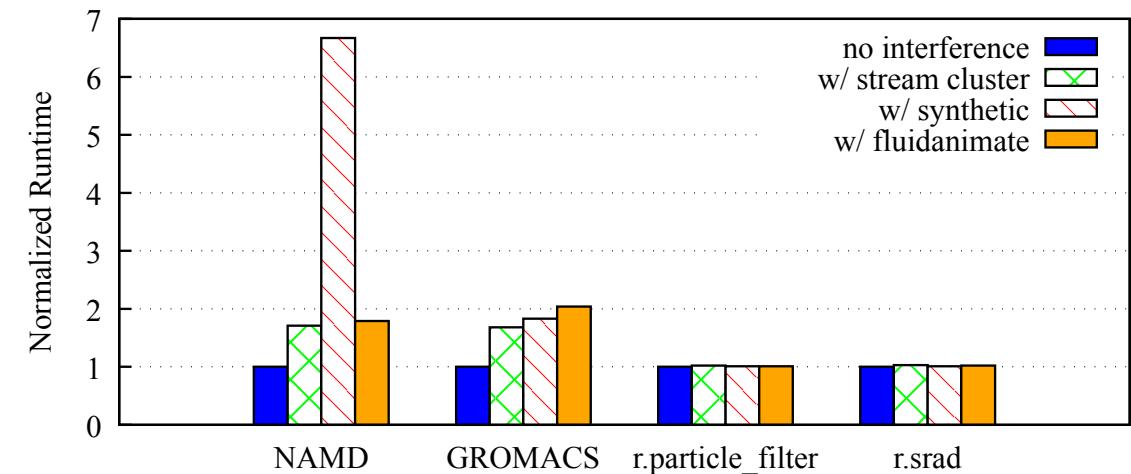
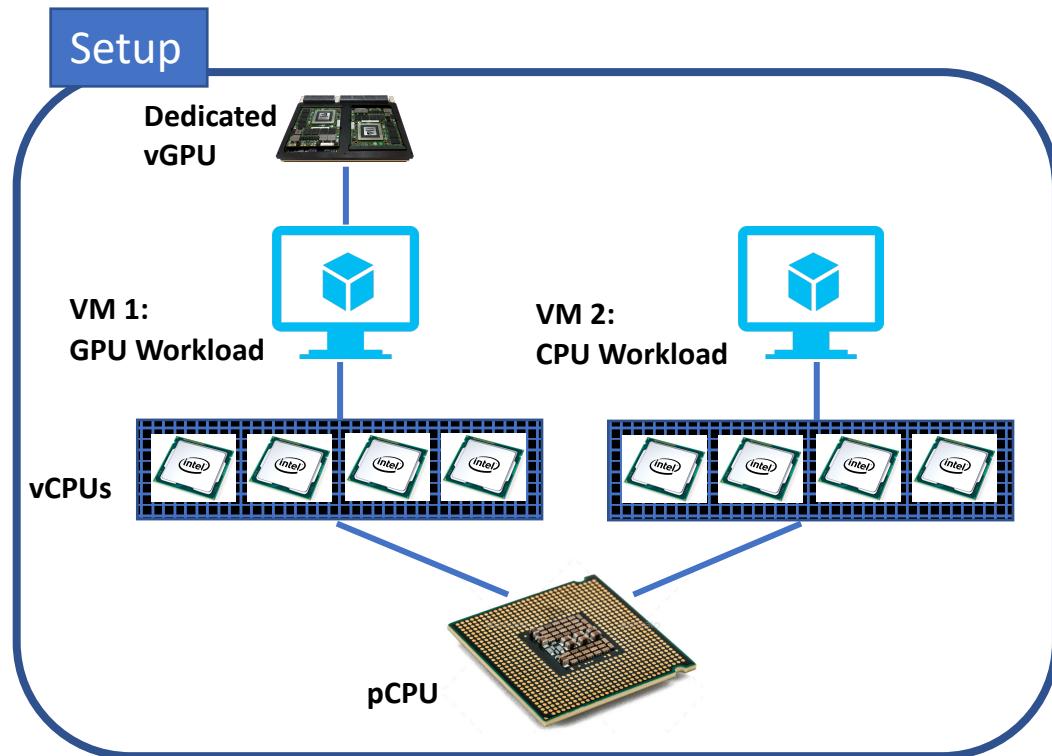
Workloads suffer from **poor performance isolation** and **low utilization** when instances **share GPUs**.

Zhang et al. [TPDS '13], Qi et al. [TACO '14], Xue et al. [USENIX ATC '16], Zhang et al. [TPDS '18], Lu et al. [TPDS '19]

How sharing CPUs among GPU instances could affect the workload performance?

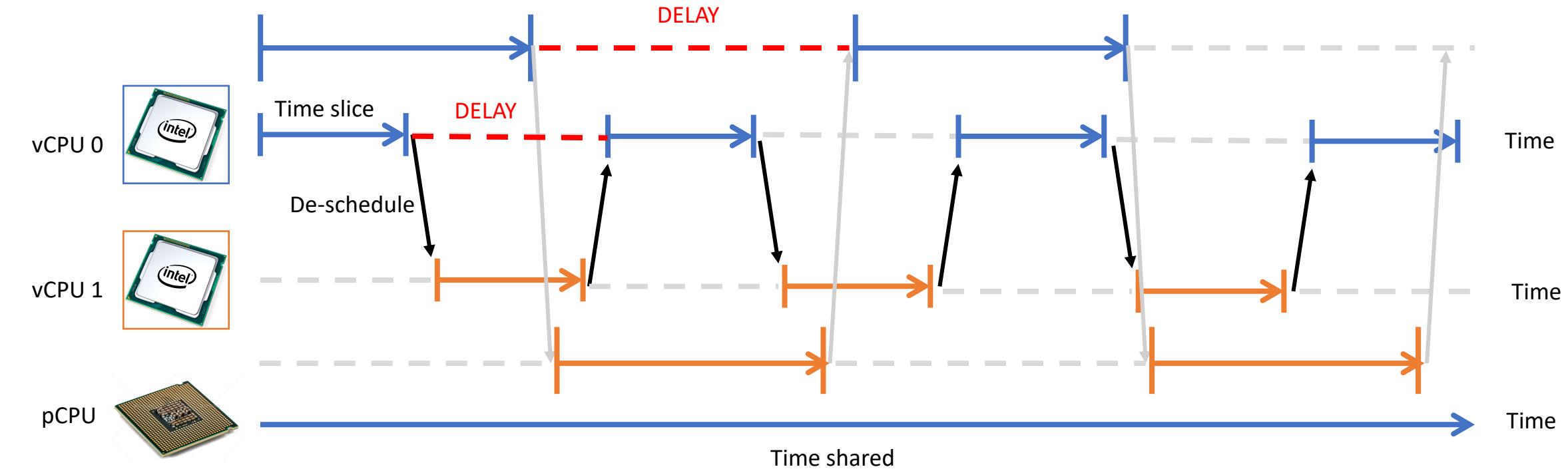
Examining Interference Caused by CPU Sharing

Workloads may suffer from **poor** and **unpredictable performance!**



Result of complex interplay between vCPU sharing and the characteristics of the GPU workload.

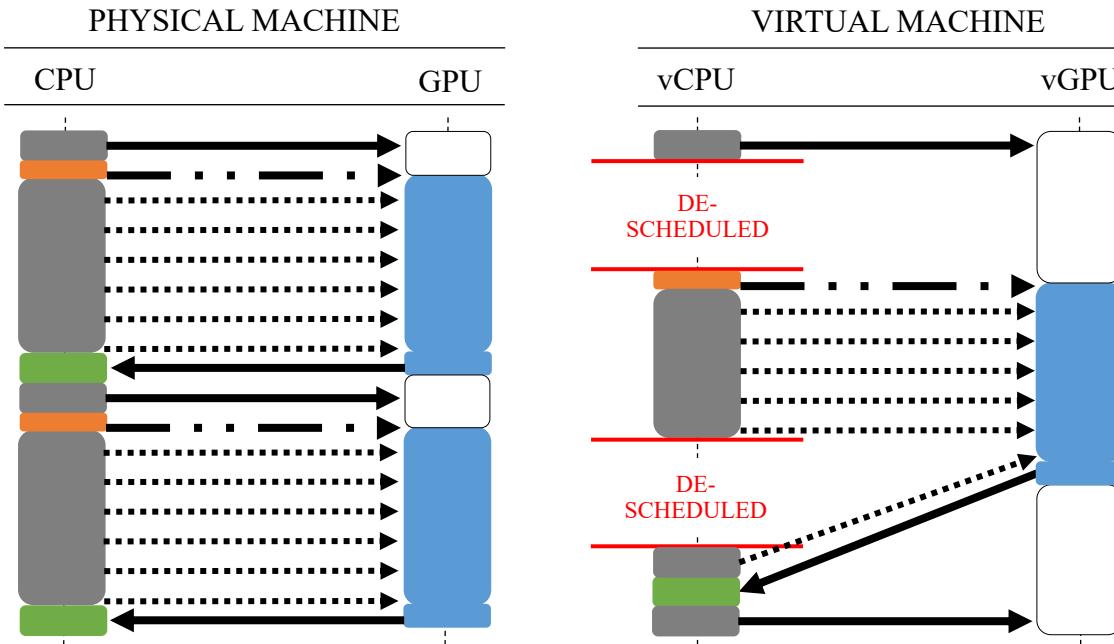
vCPU Discontinuity Caused by vCPU Sharing



How vCPU discontinuity interacts with GPU workloads remains understudied!

vCPU Discontinuity leads to Performance Degradation and vCPU Underutilization

Polling CPU-GPU synchronization



cudaEventQuery

CPU EXEC.

INV KERNEL

DATA SYNC

GPU EXEC.

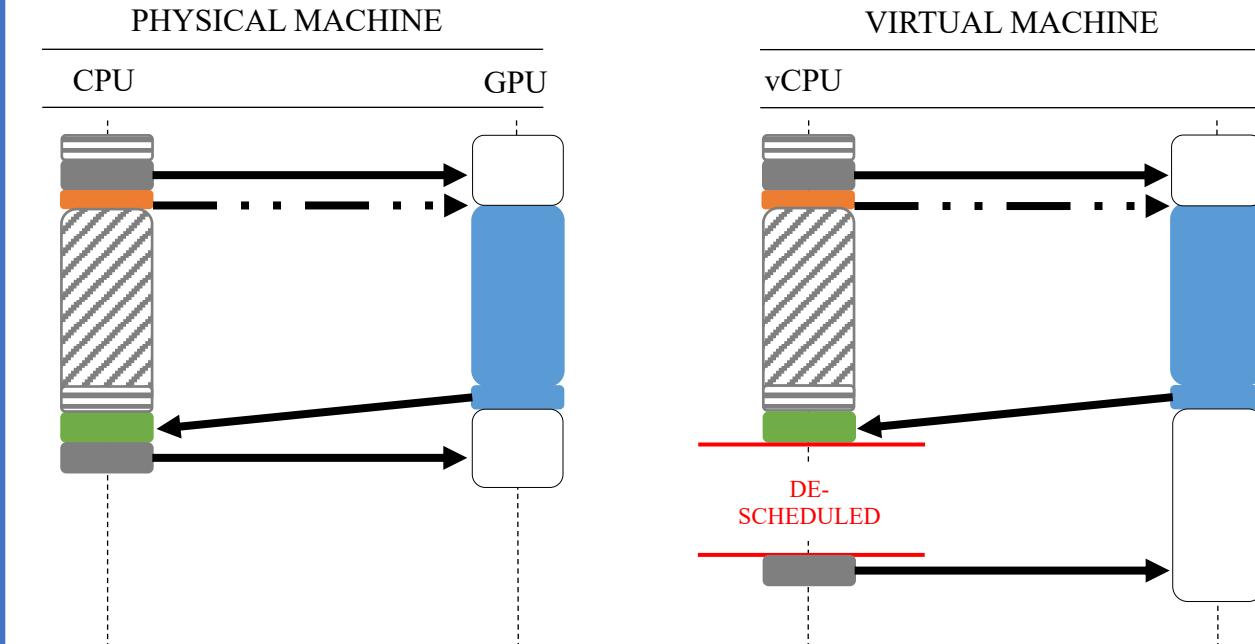
IDLE

BLOCK

WAKE LAT.

Copy data →
Poll

Blocking CPU-GPU synchronization



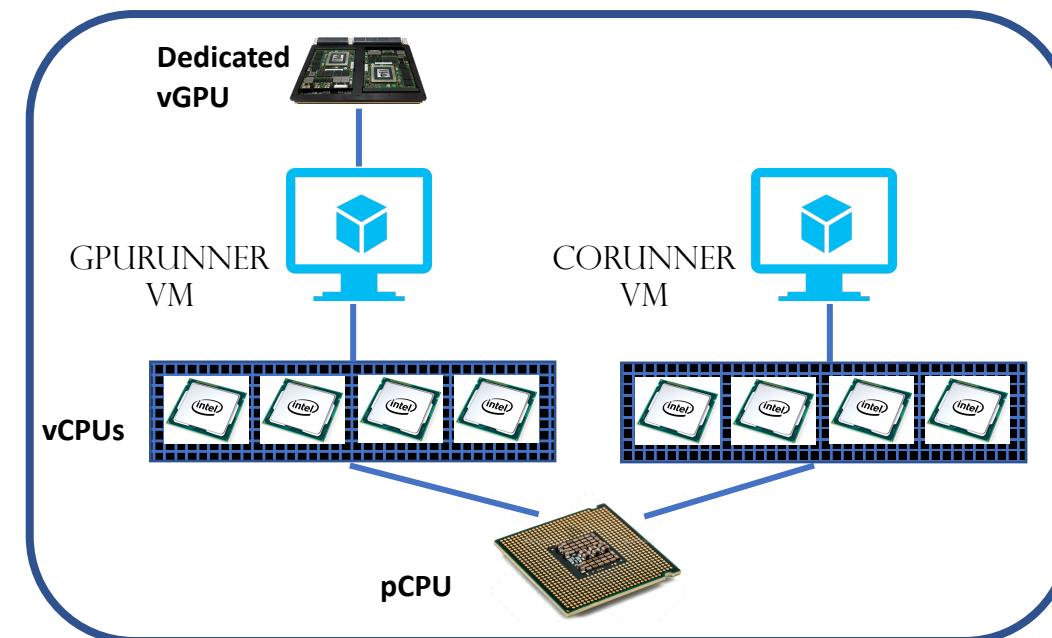
cudaEventSynchronize

Synchronization-intensive workloads are most vulnerable in the GPU cloud!

Measuring Performance Interference

Host system: HPE ProLiant DL385Gen10 server with 256GB memory using Ubuntu Server 20.04.2 (kernel 5.8.0), 4 Intel Xeon Gold 6138 20-core processors, and 2 NVIDIA Tesla P100 GPUs

Virtualized system: VMM is KVM, Ubuntu Server 20.04.2 (kernel 5.8.0), 1 GPU attached using PCI Passthrough with driver CUDA 10.1



Measuring Performance Interference

continued...

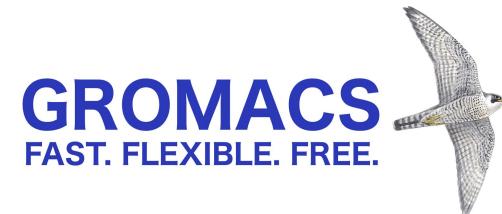
CPU-intensive Benchmarks:

- Synthetic program [while(1) loop]
- matmul



[stream_cluster, dedup, x264, fluidanimate]

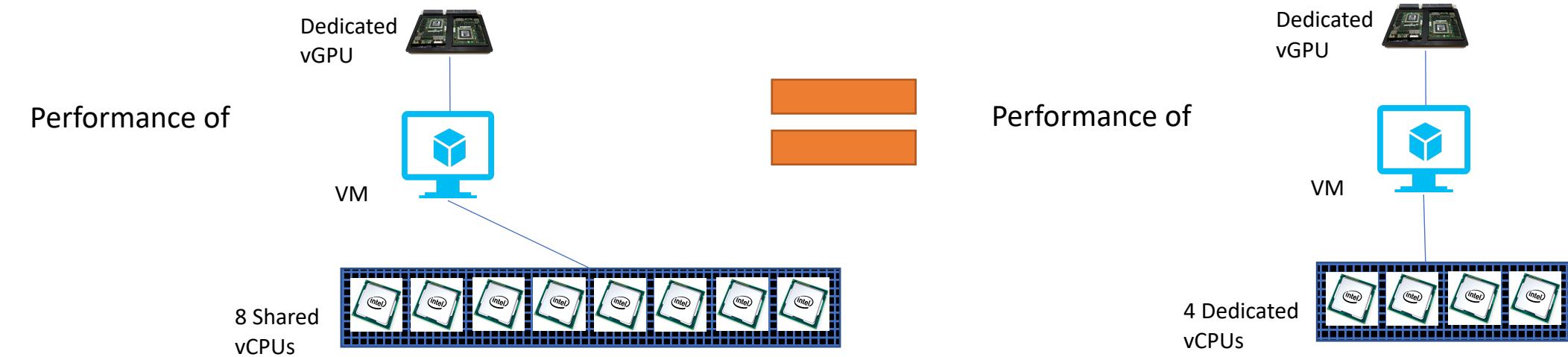
GPU-intensive Benchmarks:



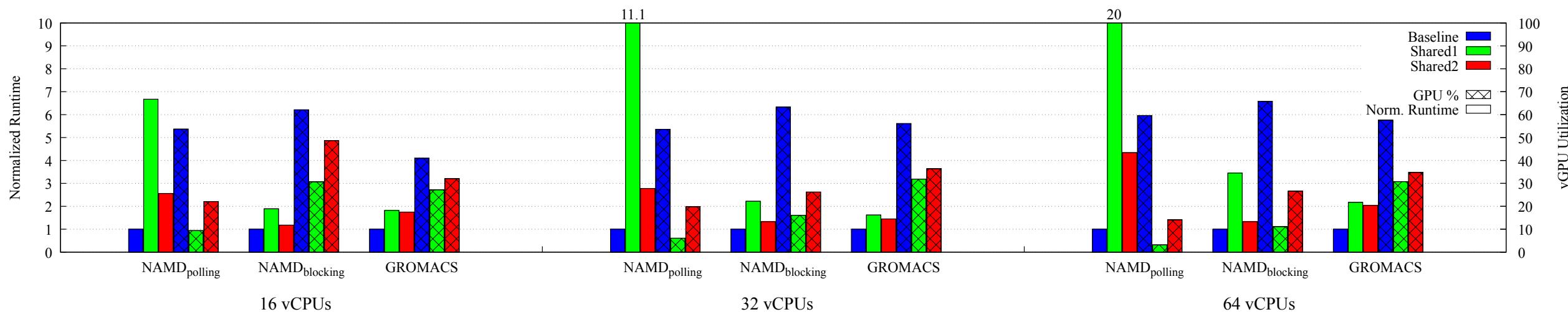
Measuring Performance Interference continued...

Experiment	corunner	GPUranner	Perf. Iso.	#vCPUs
Shared1	✓	✓	✗	16, 32, 64
Shared2	✓	✓	✓	16, 32, 64
Baseline	✗	✓	N.A	8, 16, 32

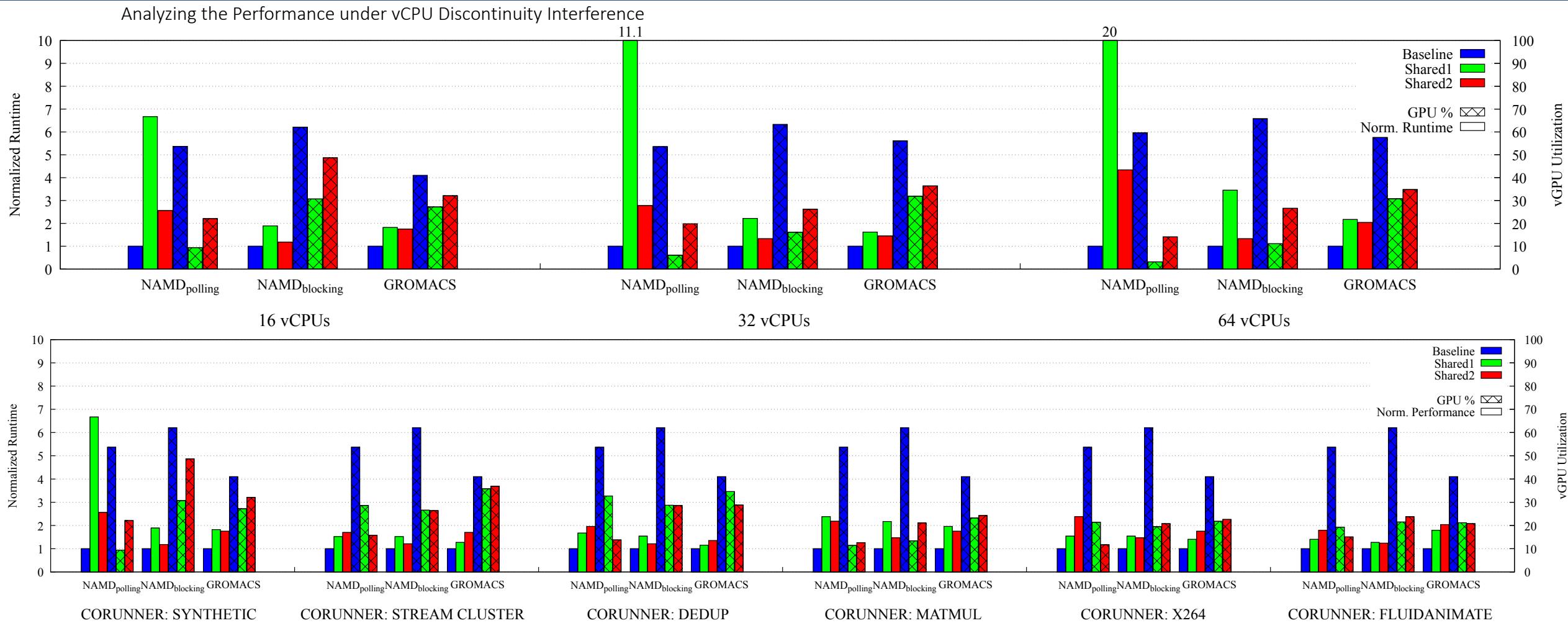
Assuming there is no degradation issue from sharing vCPUs:



Analyzing the Performance under vCPU Discontinuity Interference



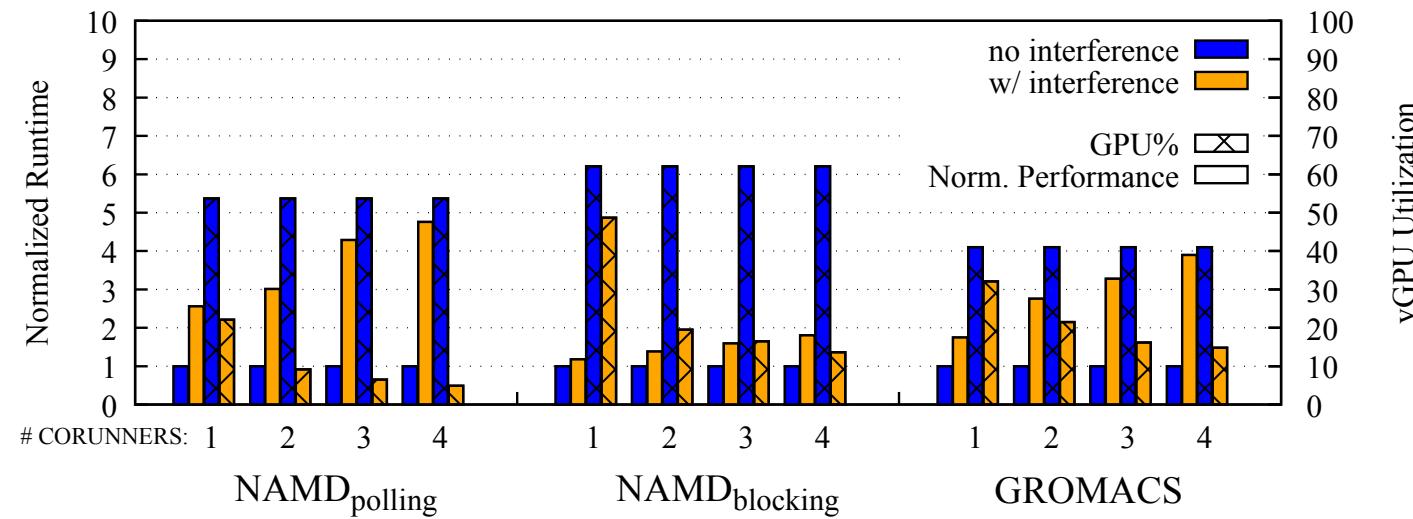
1. Degradation in Runtime Performance
2. vGPU Under-utilization
3. Increased Degradation Under Varied #vCPUs
4. Unpredictability in Performance
5. Increased Degradation Under Varied #VMs



1. Degradation in Runtime Performance
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Analyzing the Performance under vCPU Discontinuity Interference



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Diagnose: Case Study of **NAMD** Scalable Molecular Dynamics

Fine-grained profiling done using the performance analysis tool:

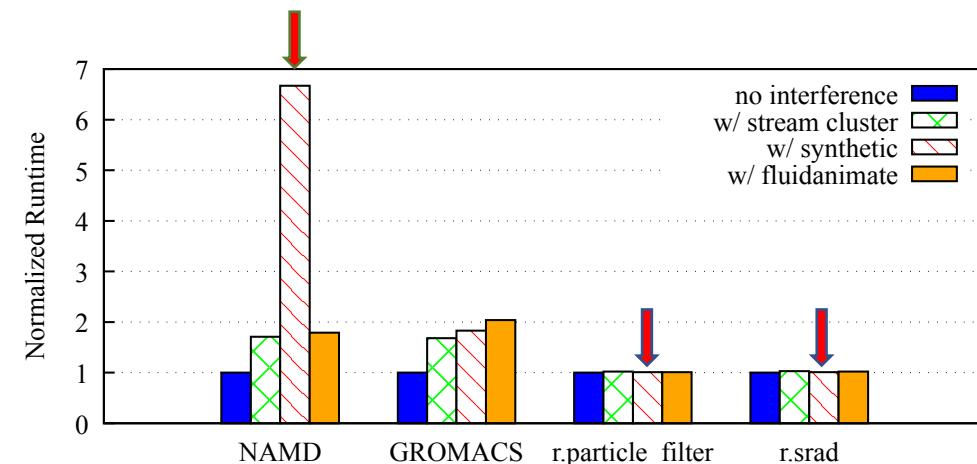


We found that the major sources of latencies are:

1. During kernel invocation
2. Receiving completed kernels

Workloads with more frequent number of kernels are more vulnerable to performance degradation!

r.particle_filter & r.srad	1 kernel / sec
NAMD	370 kernels / sec
synchronization-intensive workload	



Latency in Kernel Invocation

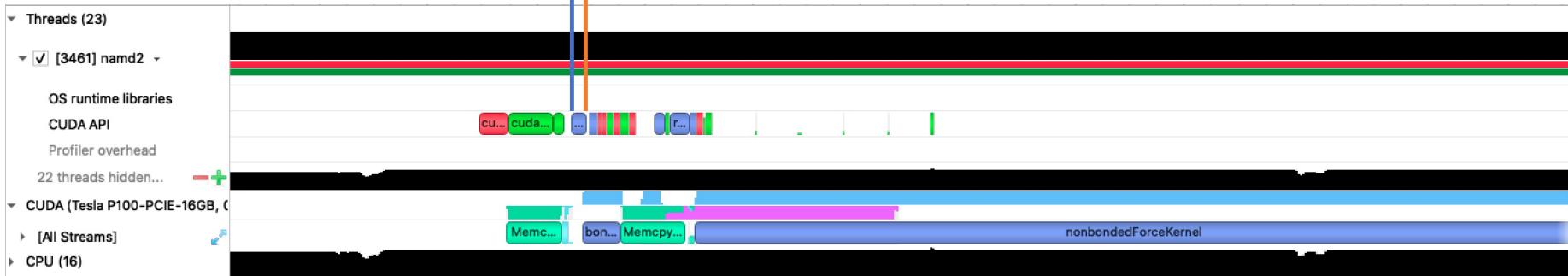
Polling CPU-GPU synchronization

vCPU DE-SCHEDULED

Latency $\approx 29.79\text{ms}$



Affected by vCPU Discontinuity:
95 kernels / sec



Normal:
171 kernels / sec

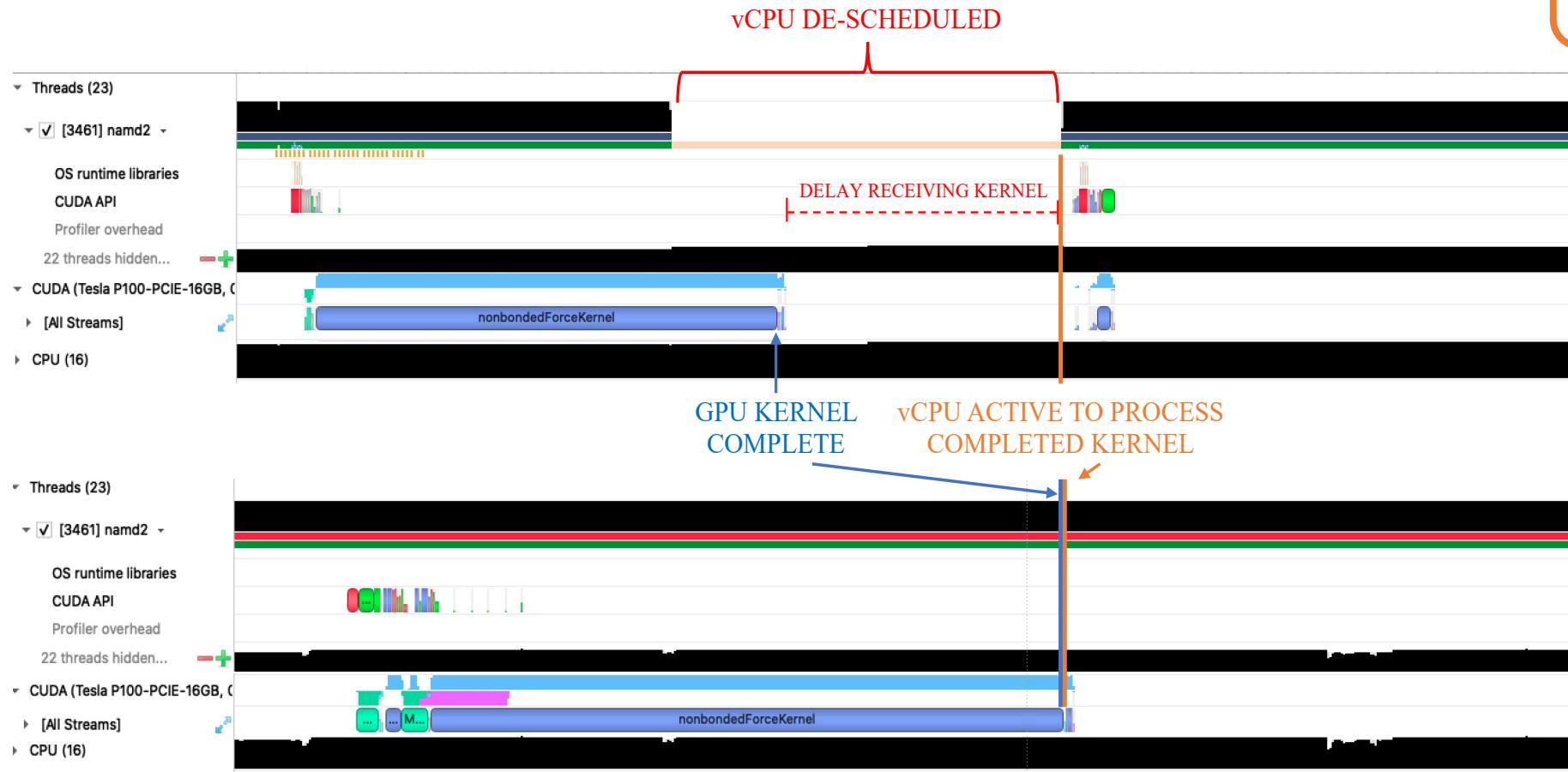
Blocking CPU-GPU synchronization

Negligible Latency.

Latency in Receiving Completed Kernels

Polling CPU-GPU synchronization

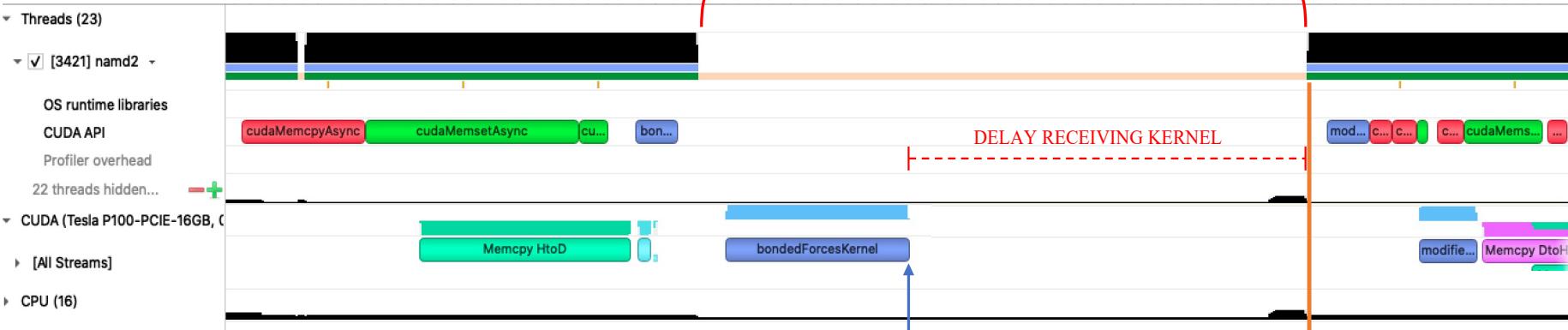
Latency $\approx 28.65\text{ms}$



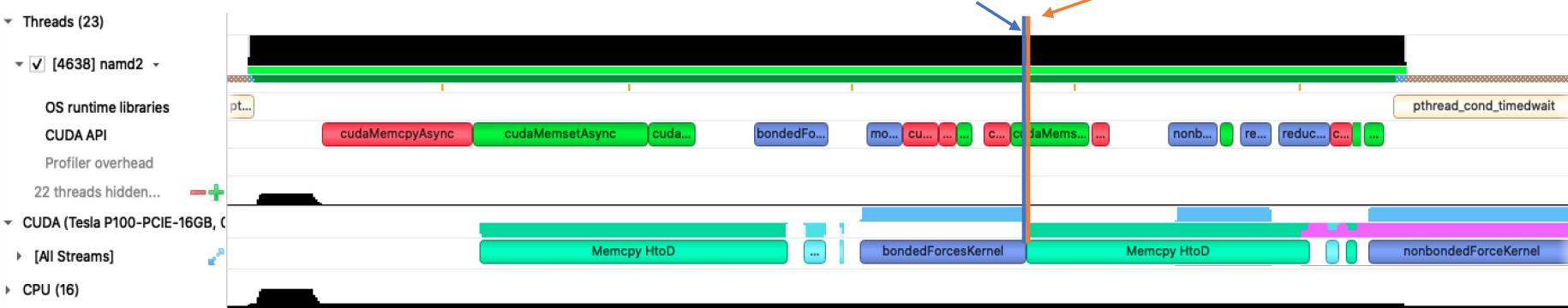
Latency in Receiving Completed Kernels continued...

Blocking CPU-GPU synchronization

Latency $\approx 13.58\text{ms}$



Affected by vCPU Discontinuity:
191 kernels / sec



Normal:
248 kernels / sec

Optimizing Performance in GPU Clouds

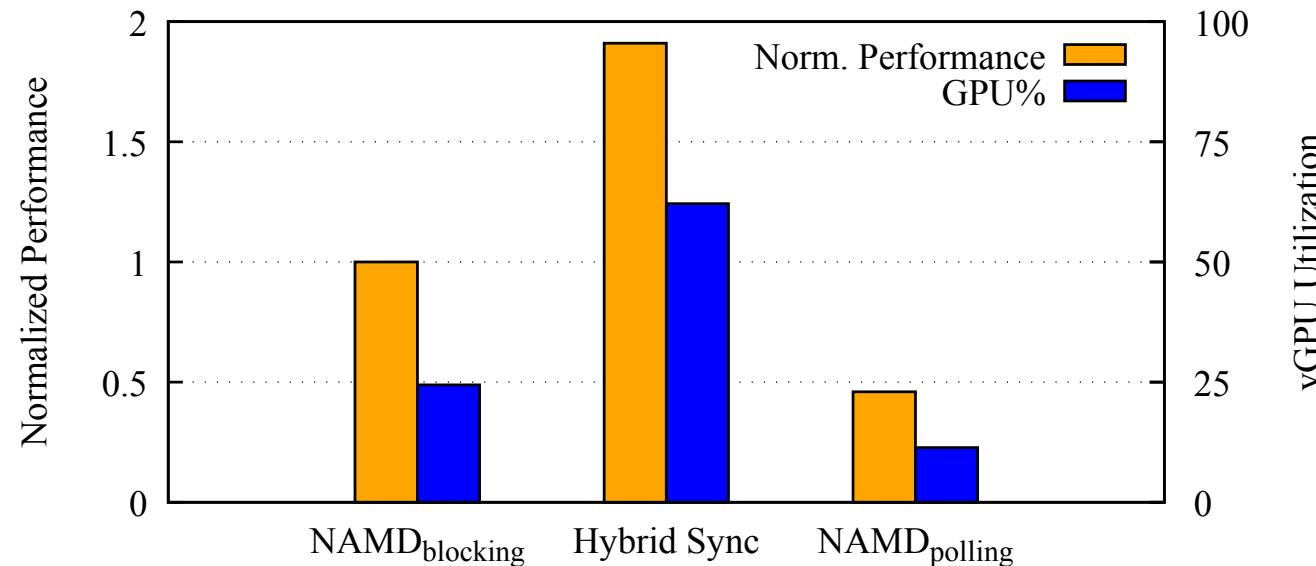
Continuous device communication substantially wastes vCPU time slice

Kernel Execution time is consistent and predictable ($\pm 2\%-5\%$ for NAMD)

*We propose a novel
polling-then-blocking CPU-GPU synchronization primitive*

This hybrid primitive reduces vCPU time slice waste and reduces kernel latency. It is portable and scalable to other applications!

Polling-then-Blocking Synchronization Results



	CPU-GPU Synchronization Techniques		
	Polling	Blocking	Hybrid
Invoking Kernel	High Latency (29.79ms)	Negligible Latency	Low Latency (5.42ms)
Receiving Kernel	High Latency (28.65ms)	High Latency (13.58ms)	Low Latency (1.49ms)

1 Execution Cycle \approx 4.5ms

Conclusion

We diagnosed the effect of vCPU discontinuity on GPU workloads

Performance interference caused by vCPU discontinuity:

- Inefficient CPU-GPU synchronization
- Increases vulnerability of vCPUs to be descheduled
- Unpredictable GPU workload performance
- ↑ # shared vCPUs, ↑ # VMs ⇒ ↑ GPU workload degradation

We proposed a *polling-then-blocking hybrid synchronization technique*

The diagnose should provide a guideline for users and cloud providers using multi-tenant GPU cloud instances sharing CPU resources!

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Thank you for your attention!
Any questions?

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