



16th International Workshop on  
**Performance Modeling, Benchmarking and  
Simulation of High Performance Computer  
Systems**

# **Characterizing the Impact of GPU Power Management on an Exascale System**

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Phillipe Navaux

**Arthur Lorenzon**

Bruno Alvarez

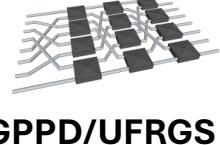
Jordà Polo

Antigoni Georgiadou

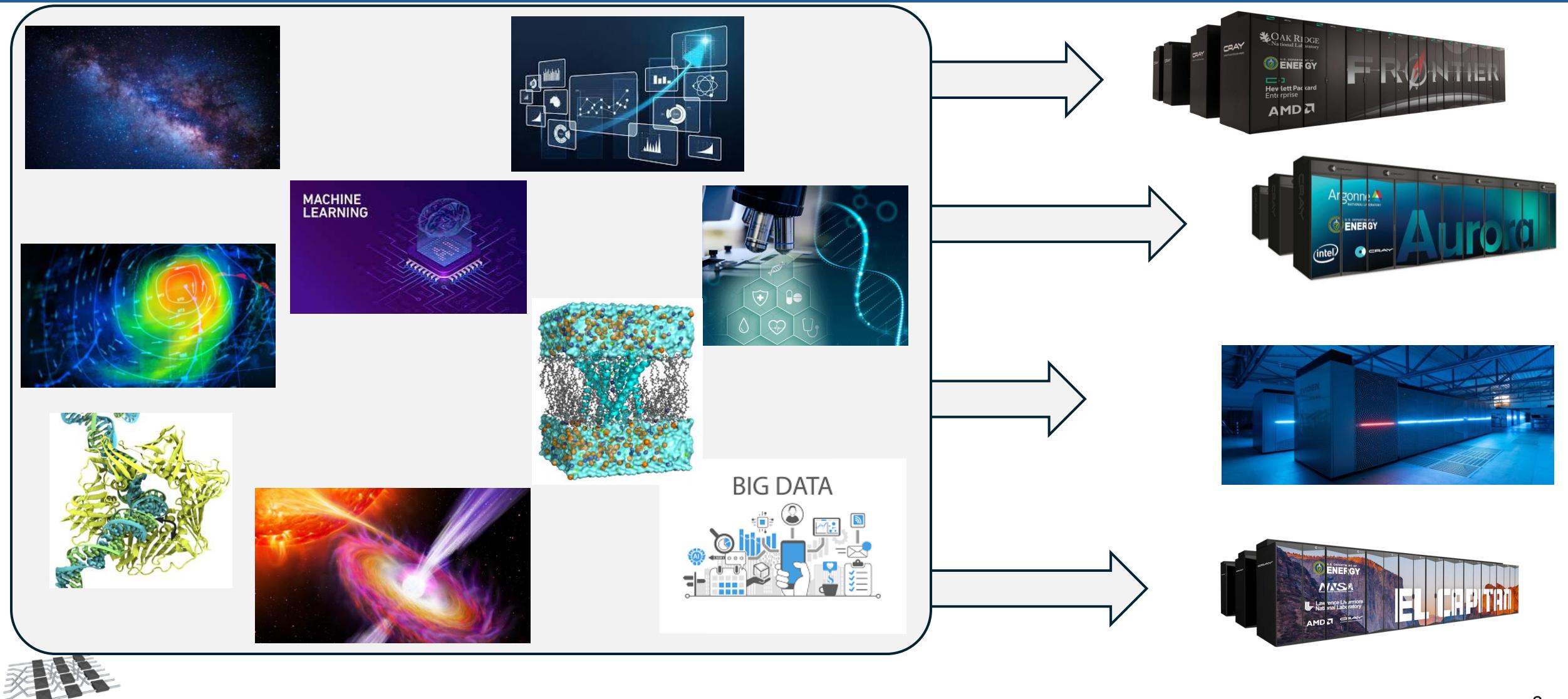
James B. White III

Woong Shin

Bronson Messer



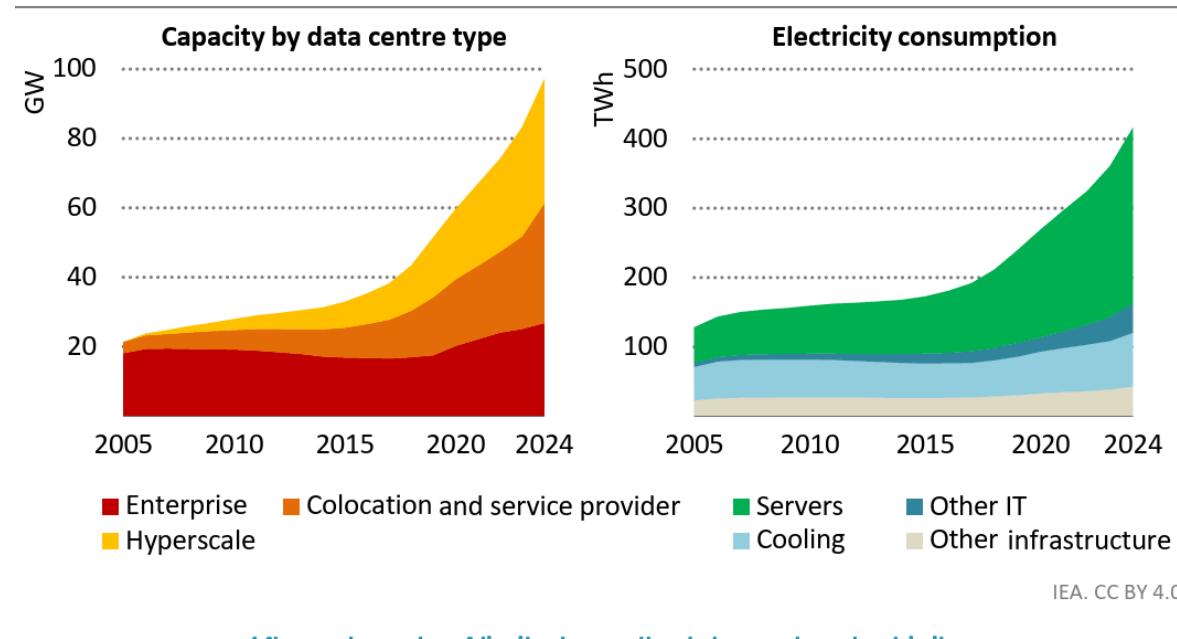
# Motivation



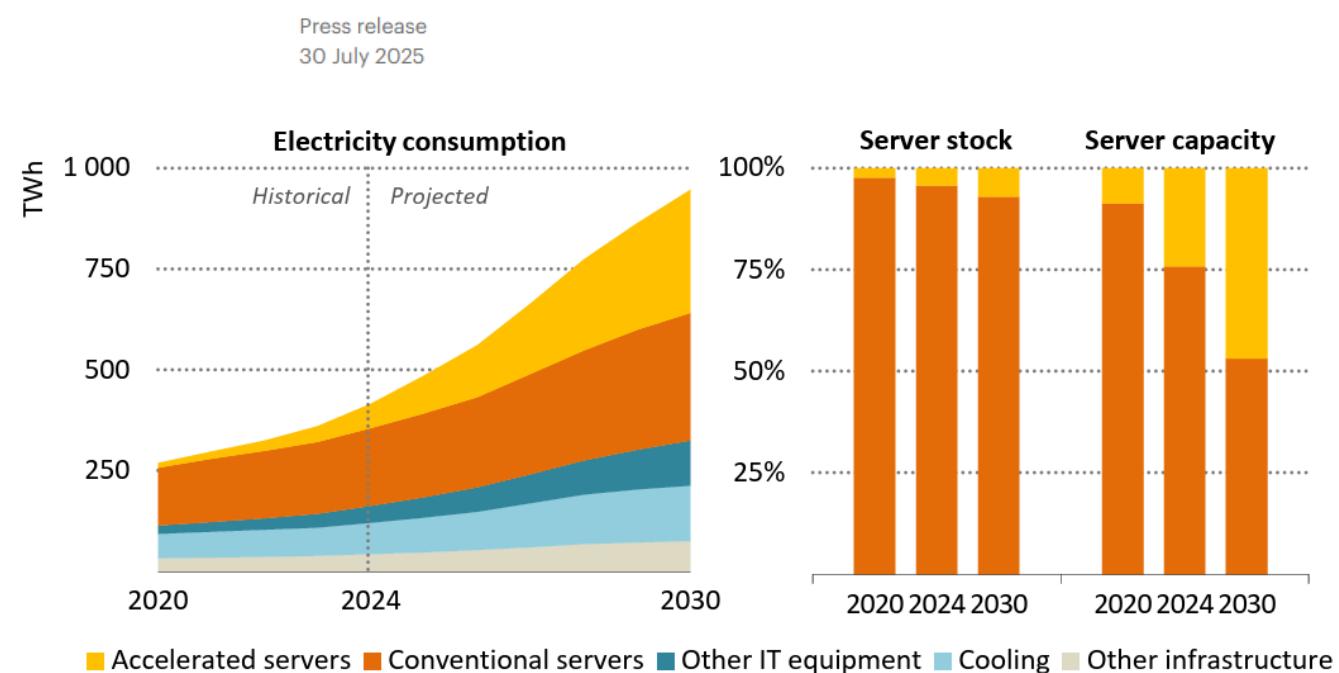
# Motivation

**Global electricity demand to keep growing robustly through 2026 despite economic headwinds**

**Figure 2.3 ▷ Total data centre electricity consumption by equipment type and data centre type, 2005-2024**



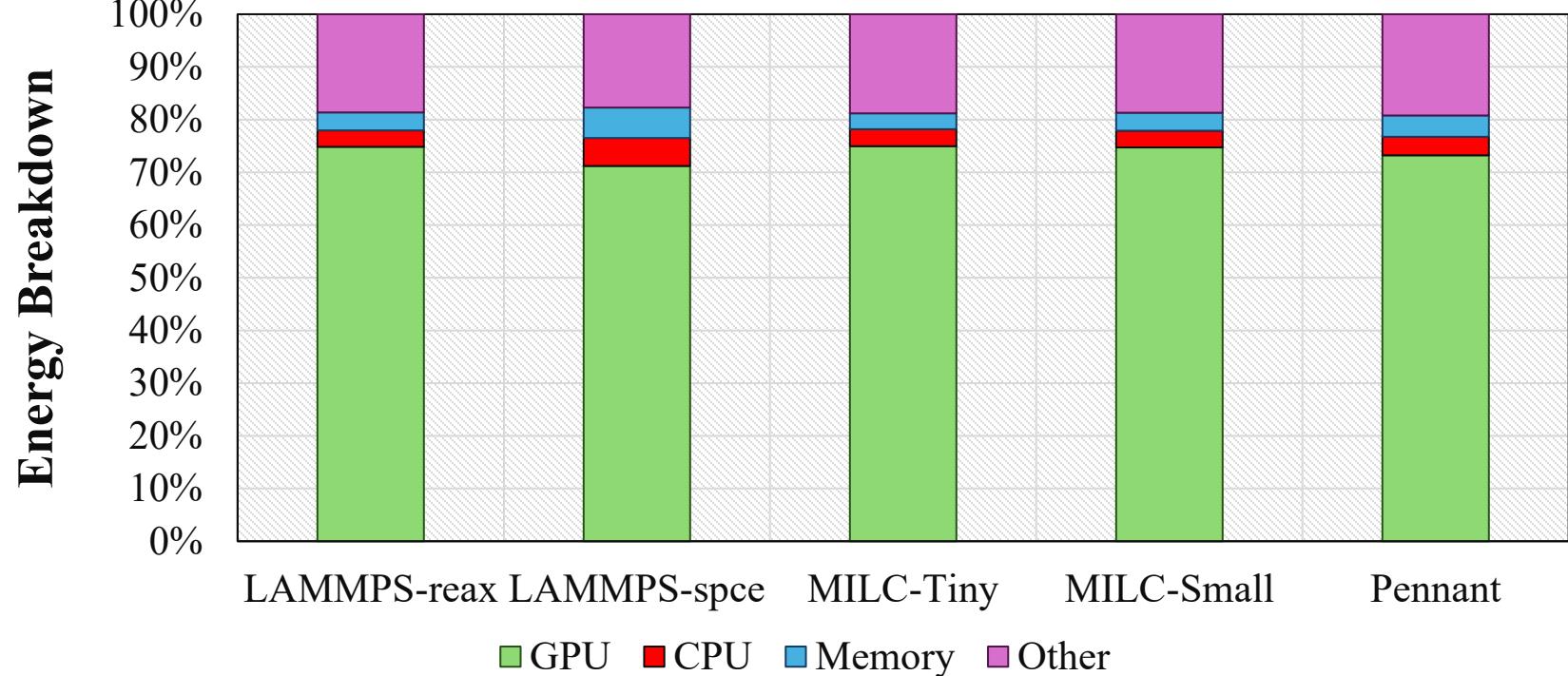
After a decade of limited growth, data centre electricity consumption began to accelerate again after 2015



Around 70% of the growth in electricity demand from servers between 2025 and 2030 comes from accelerated servers

# Motivation

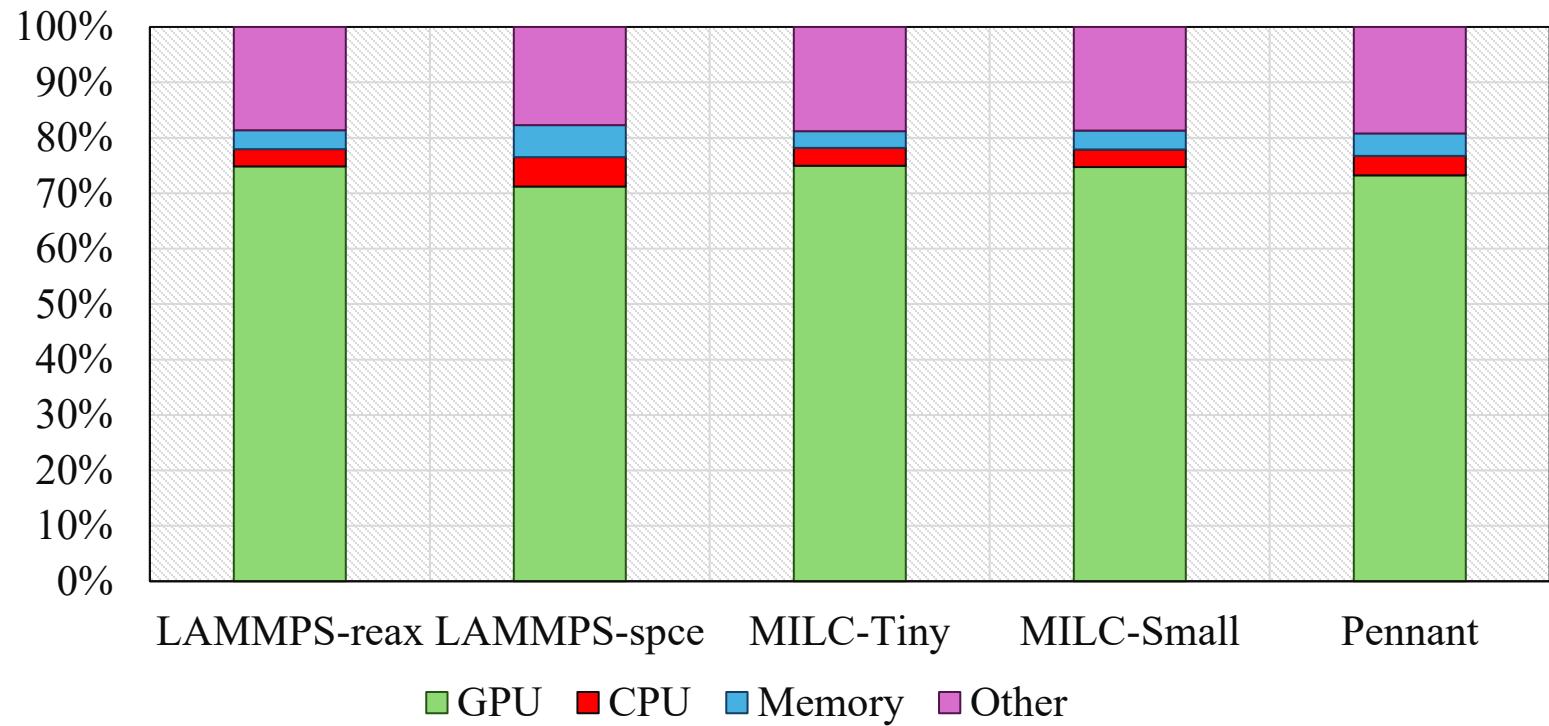
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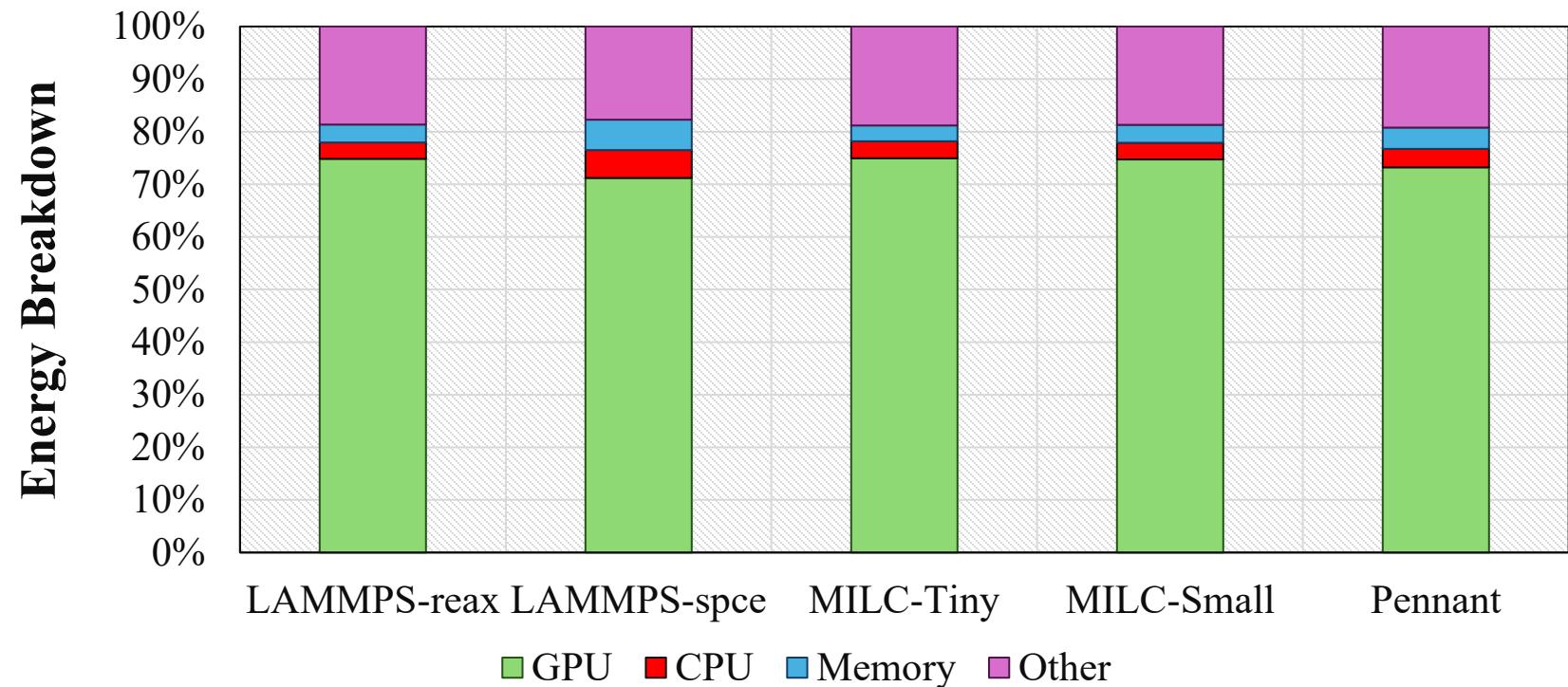
Energy Breakdown



Improving GPU energy efficiency  
is key to sustainable HPC

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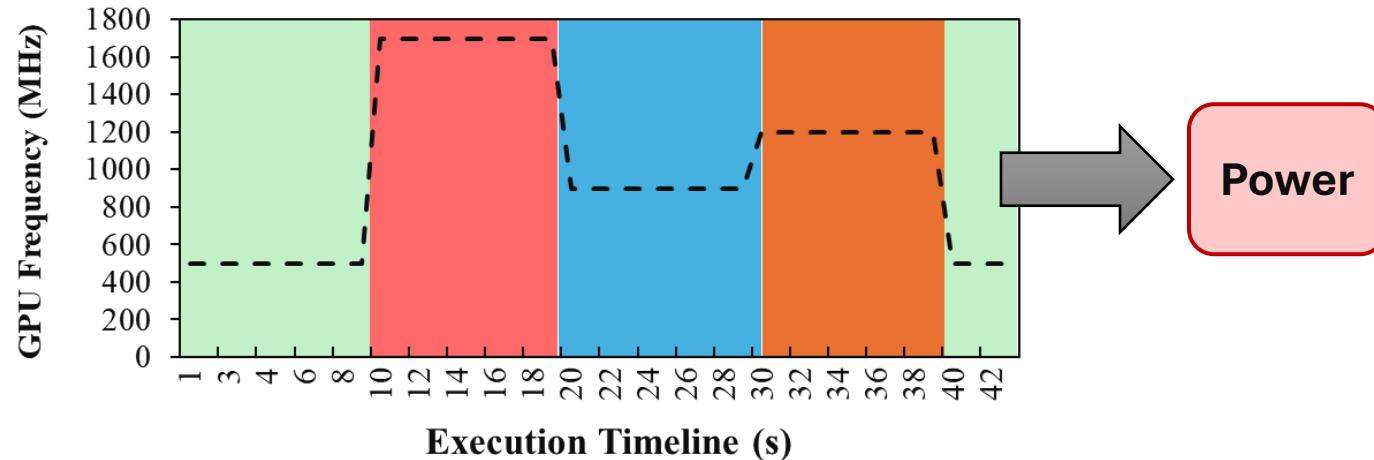


Improving GPU energy efficiency  
is key to sustainable HPC

Power management strategies  
can be employed to optimize  
energy efficiency

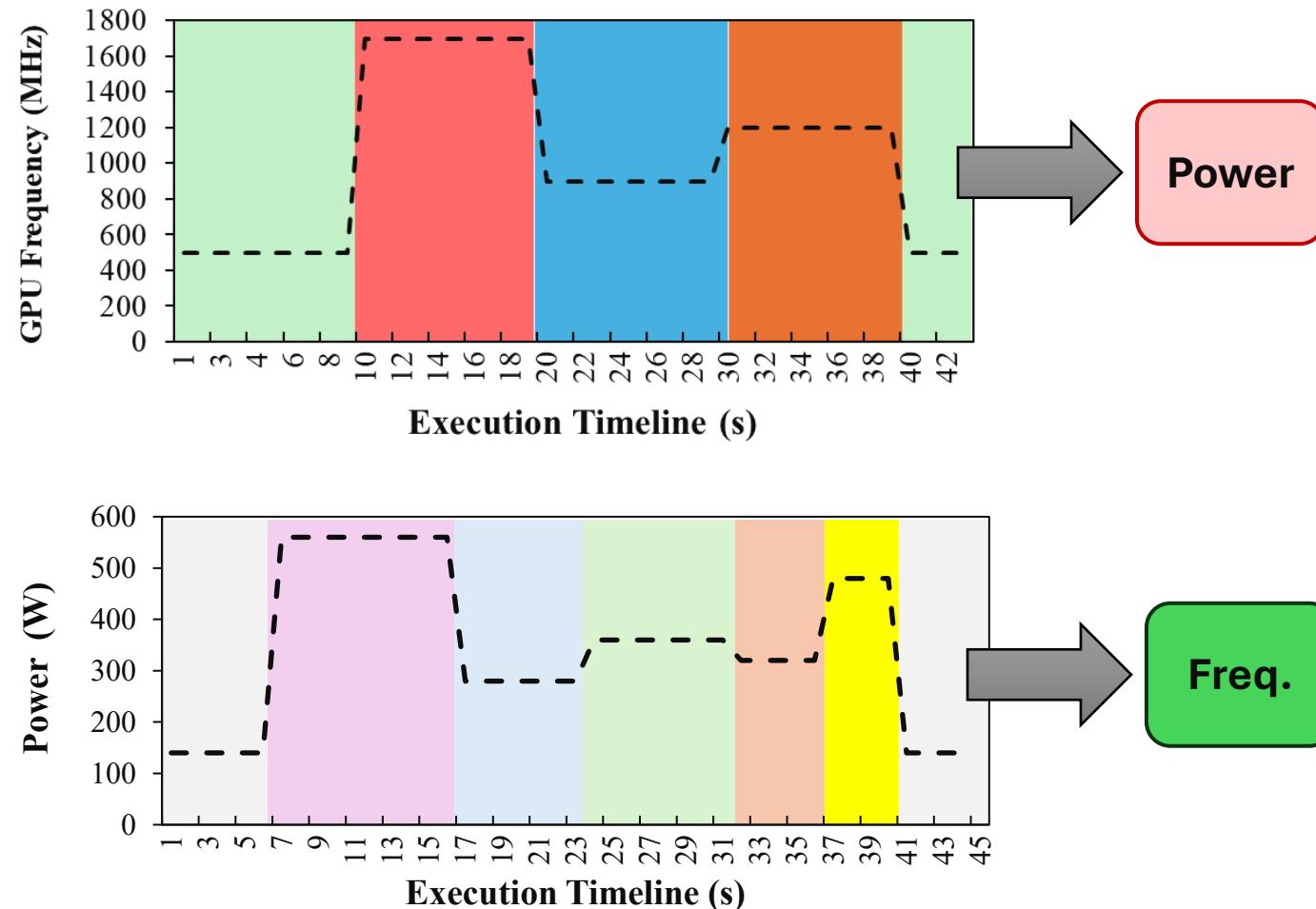
# Motivation: Power Management

- **Frequency Capping (DVFS):**
  - Limits GPU clocks to a fixed upper bound
  - Proactive/static control
  - Predictable performance, easy to reproduce



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- **Frequency Capping (DVFS):**
  - Limits GPU clocks to a fixed upper bound
  - Proactive/static control
  - Predictable performance, easy to reproduce
- **Power Capping:**
  - Sets a maximum GPU power budget
  - Reactive/adaptive control
  - Frequency adjusts with workload intensity
  - Less predictable

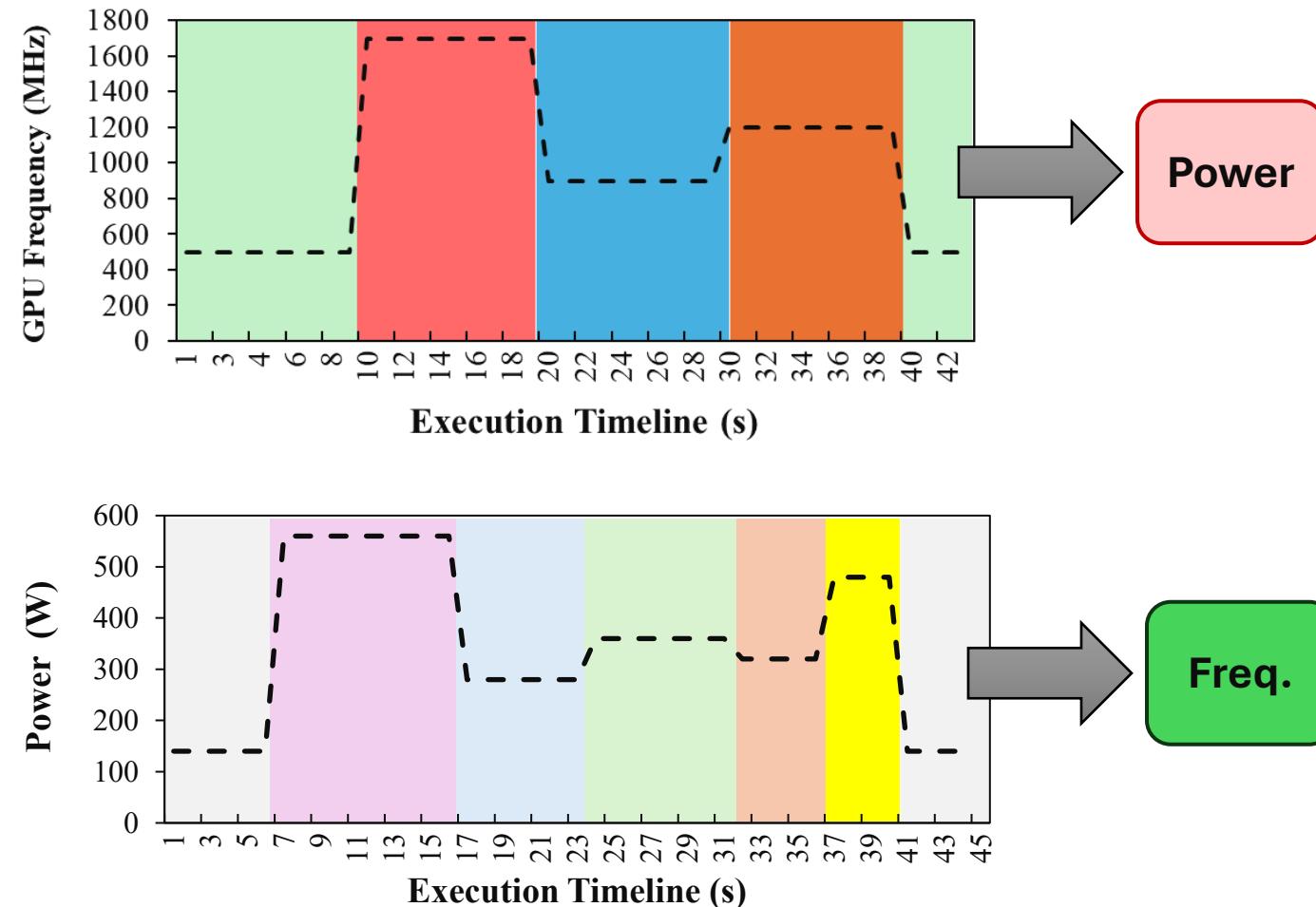


# Motivation: Power Management

- **Frequency Capping (DVFS):**

Which power management strategy delivers the best performance-energy efficiency for a given application?

- Frequency adjusts with workload intensity
- Less predictable



# Contributions

- Performance and energy benchmarking of power and frequency management strategies
- Characterize how these techniques affect:
  - Runtime
  - Energy-to-solution
  - Energy efficiency

# Agenda

- Methodology
- Evaluation
- Concluding Remarks

# Agenda

- Methodology

# Methodology: Benchmarks

- Seven GPU-accelerated applications:
  - **Cholla**: Memory-intensive astrophysics hydrodynamics code
  - **HACC**: Cosmology simulation sensitive to memory bandwidth and compute throughput
  - **Kripke**: Particle-transport proxy stressing memory access and spatial patterns
  - **LAMMPS**: Compute-intensive molecular dynamic code for materials simulation
  - **Pennant**: Unstructured-mesh hydrodynamics proxy sensitive to memory and compute balance
  - **PortUrb**: Urban flow simulation limited by memory bandwidth and large-array operations
  - **QuickSilver**: Monte Carlo transport proxy with irregular control flow and latency-bound access

# Methodology: Benchmarks

- Benchmarks exhibit different behaviors of FLOPs/byte, FLOPs/s, and L2 Cache HIT

	FLOPs/byte	FLOPs/s	L2 Cache Hit (%)
Cholla	0.62	6.58E+11	37.51
HACC	215.04	4.67E+12	84.02
Kripke	0.10	3.90E+10	62.68
LAMMPS	3.41	5.71E+12	51.78
Pennant	0.67	5.57E+11	45.11
PortUrb	11.45	1.08E+12	61.32
QuickSilver	1.83	1.61e+10	74.72

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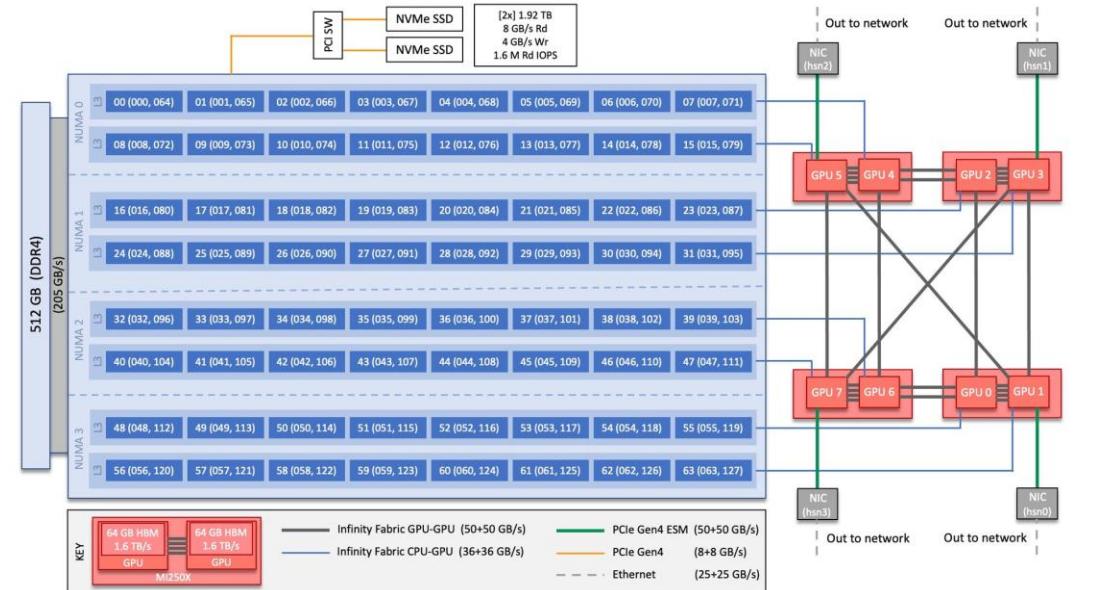
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# Methodology: Target Architecture

- 1 – 32 nodes from Frontier Supercomputer
  - 1x 64-core AMD Optimized 3rd Gen EPYC CPU
  - 4x AMD MI250X, each with 2 GCDs (total of 8 GCDs per node)
- GPU Operating Frequency:
  - 31 levels: 500 MHz, 540 MHz, ... 1700 MHz.
- GPU Power Capping:
  - 21 levels: 140W, 160W, ... 560W.
- Compilation process
  - AMD ROCm hipcc 6.2.4
  - Flags → `-O3` and `--ofload-arch=gfx90a`
  - Modules → `craype-accel-amd-gfx90a` and `rocm/6.2.4`



# Methodology: Evaluated Metrics

- **Performance (Perf)**
  - FOM (when available)
  - Total execution time
- **Performance per Watt (Perf/Watt)**
  - Ratio between performance and the average GPU power draw during execution
- **Getting power, energy, and other system-level metrics:**
  - *Omnistat*, an open-source, low-overhead monitoring tool
  - Configured using a runtime control file to enable the collection of core GPU metrics via the system-management interface
  - Aggregated metrics on a per-Slurm-job basis across all compute hosts.
  - Metrics collected every 0.1s
  - 10 executions, with std deviation < 2%

# Agenda

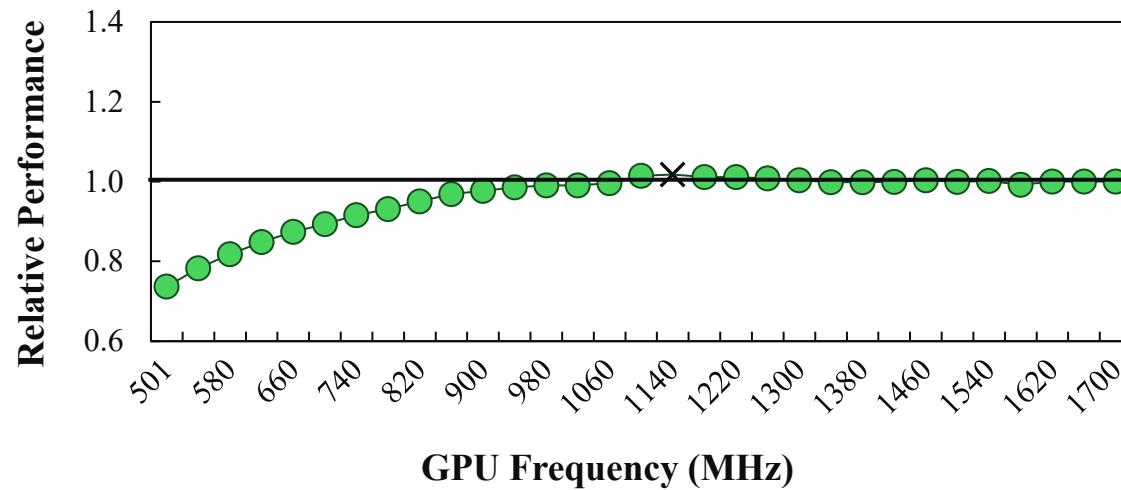
- Methodology
- Evaluation

# Evaluation

- What applications benefit more from Frequency Capping?

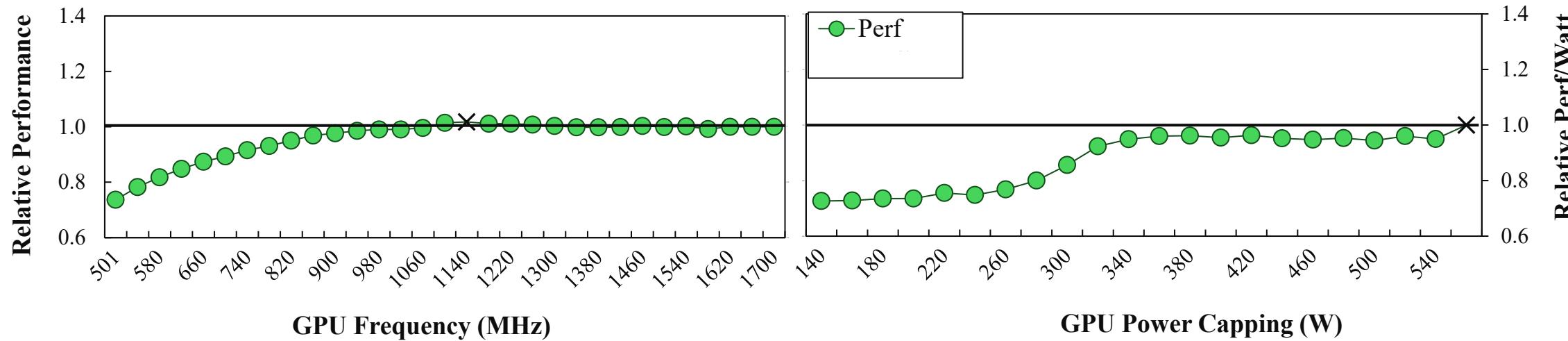
# Evaluation: Apps that benefit more from Freq. Capping

Cholla  
1 node



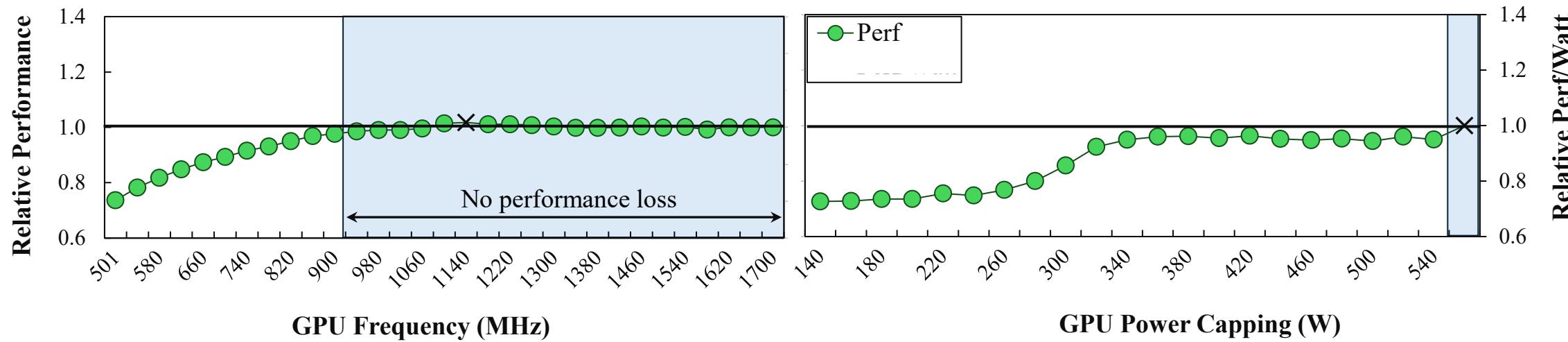
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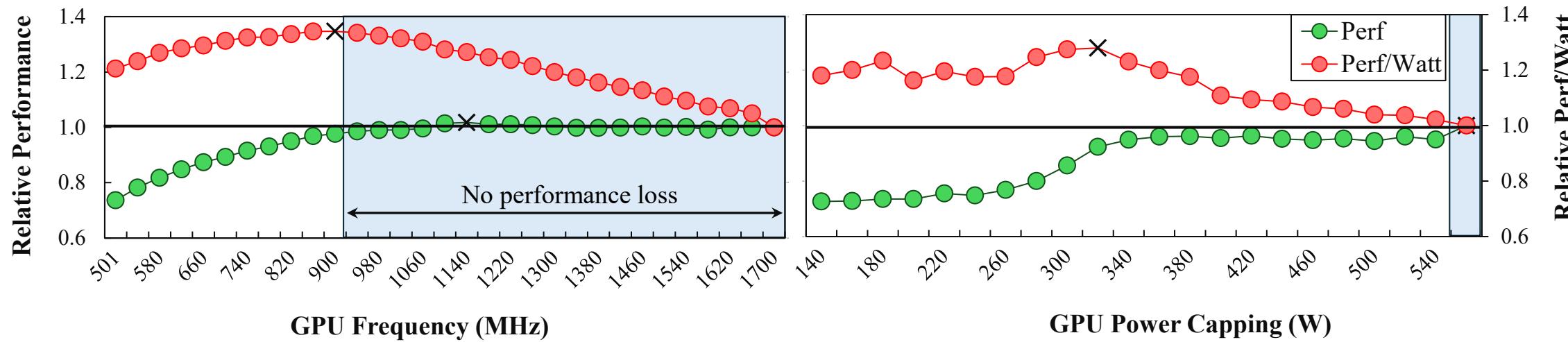
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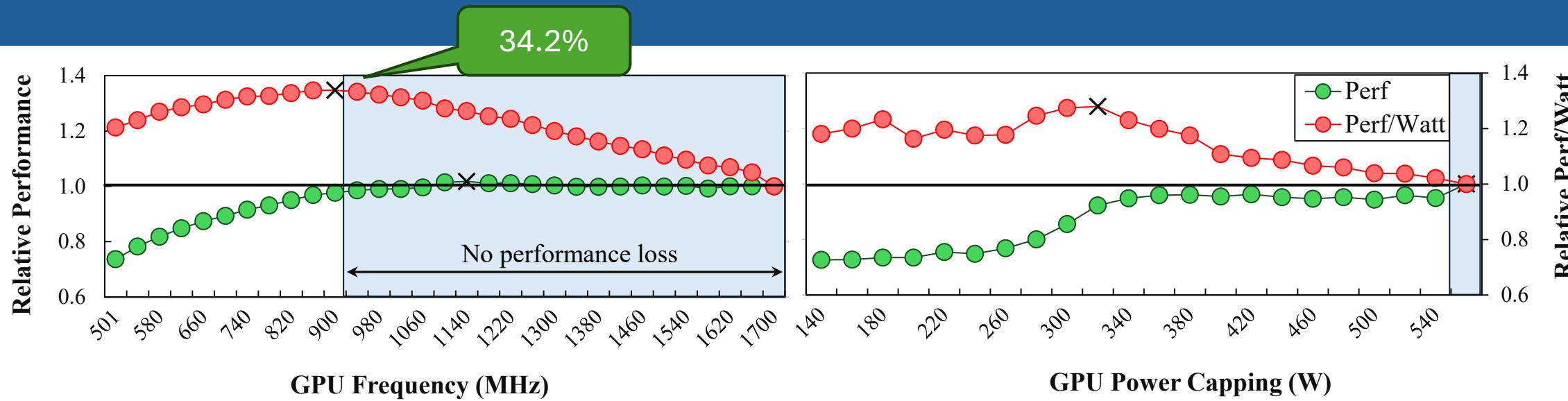
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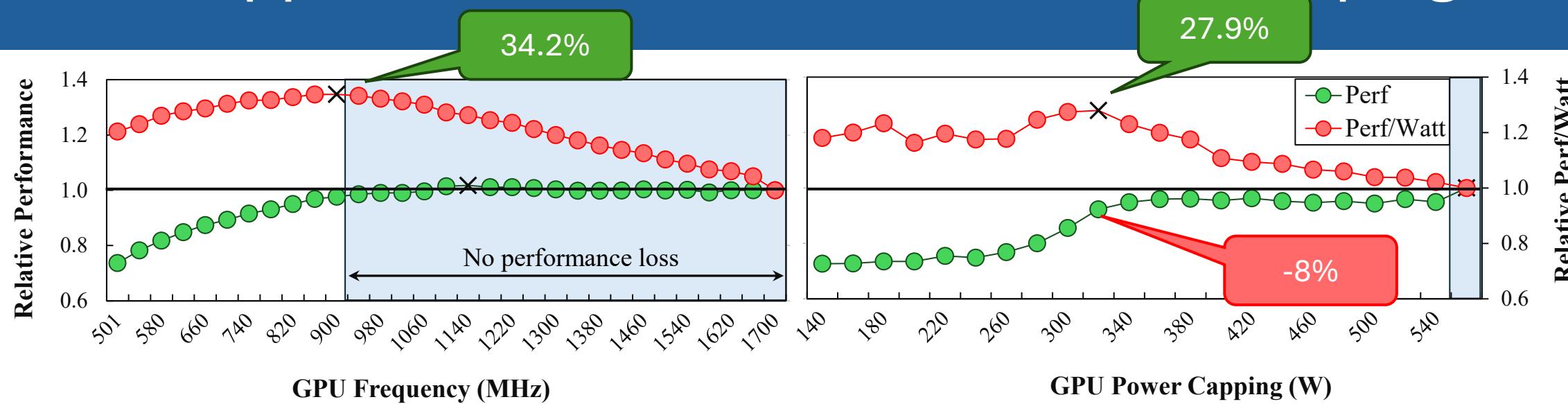
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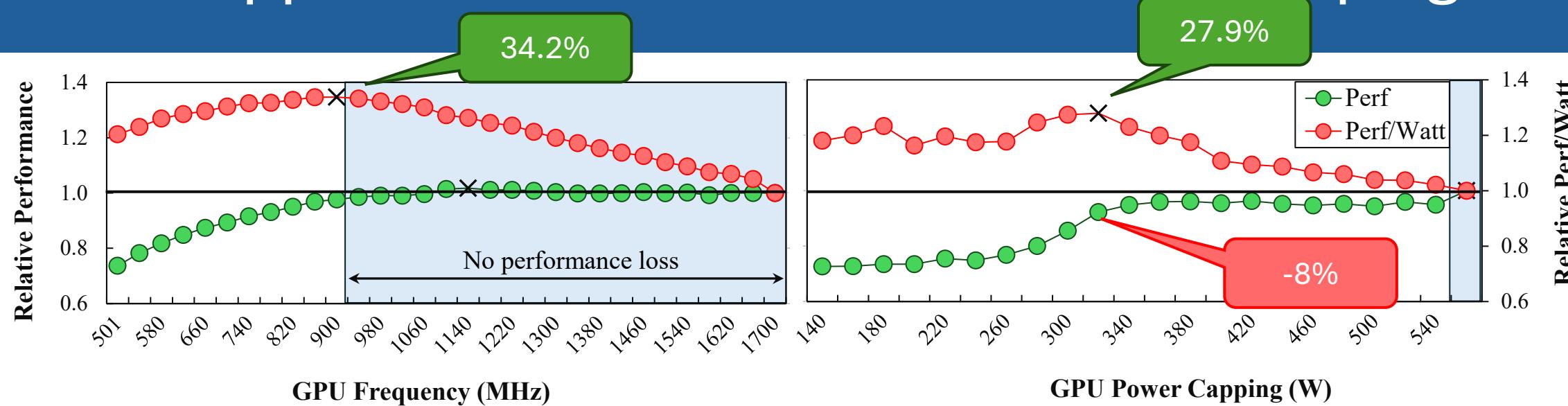
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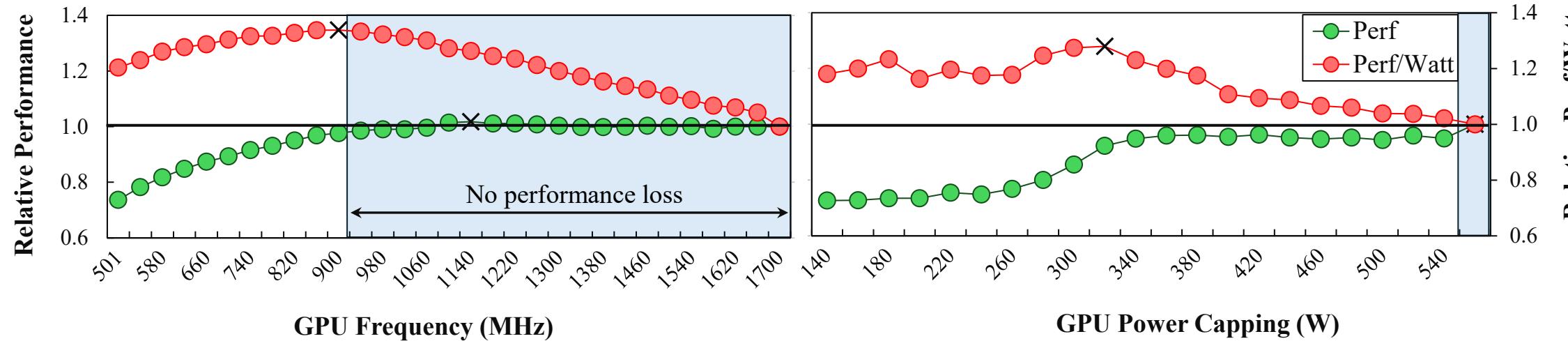
Cholla  
1 node



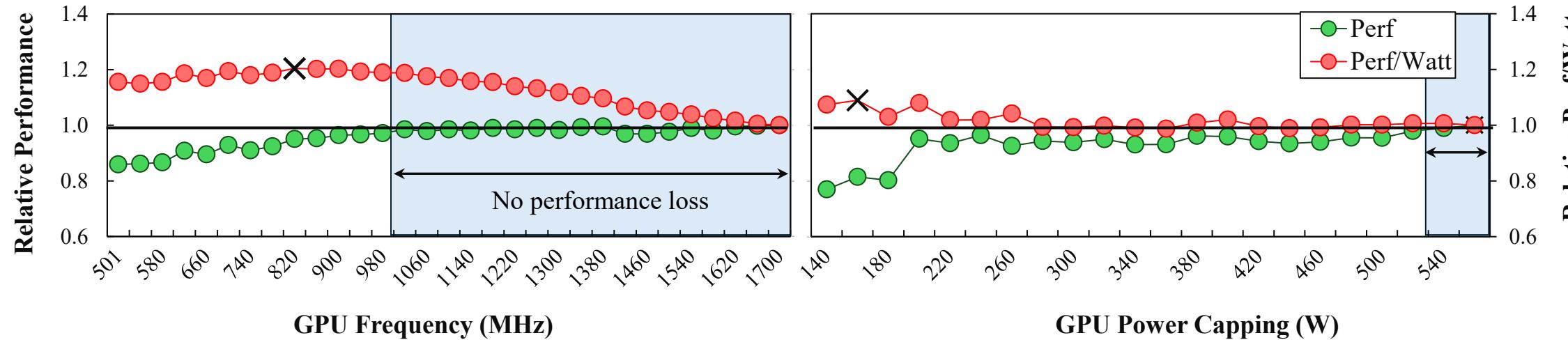
Energy efficiency increases when operating below peak frequency, as memory-controller and leakage power drop  
Frequency capping matches power draw to memory bandwidth limits, eliminating unnecessary dynamic power

# Evaluation: Apps that benefit more from Freq. Capping

Cholla  
1 node

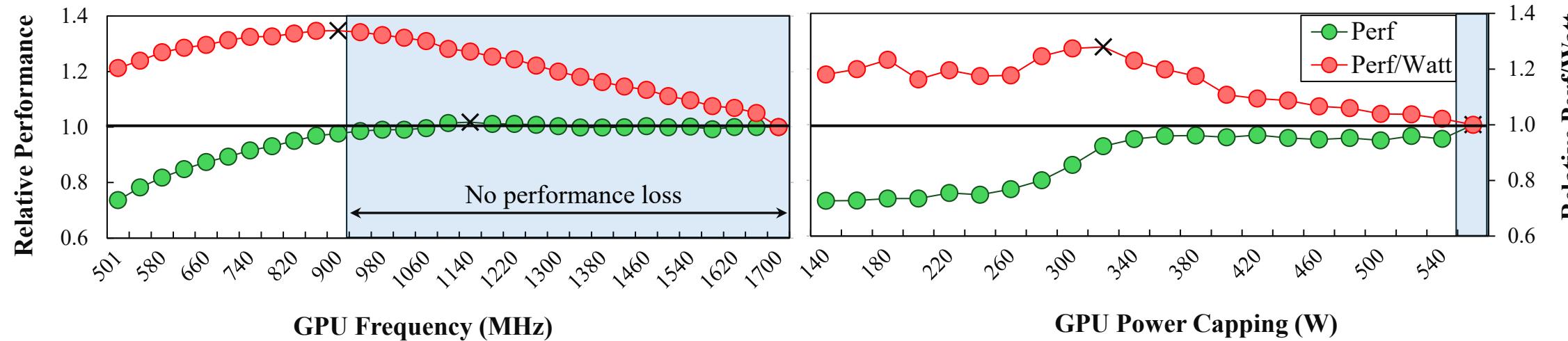


Cholla  
32 node

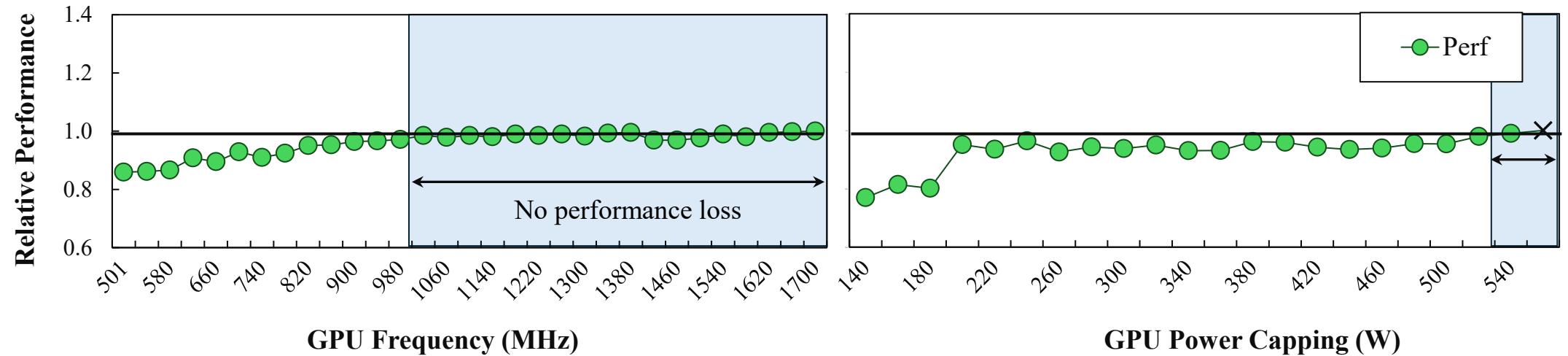


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Cholla  
1 node

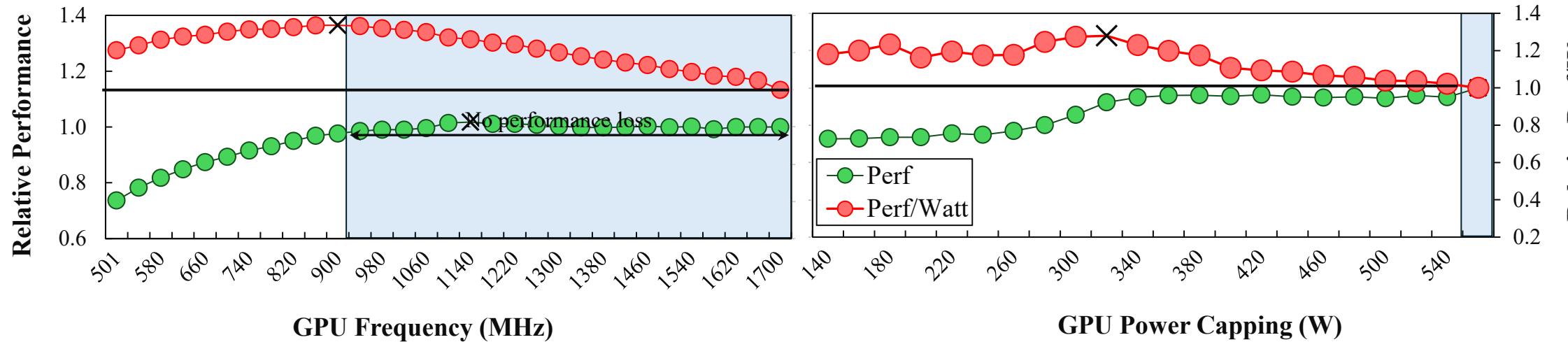


Cholla  
32 node

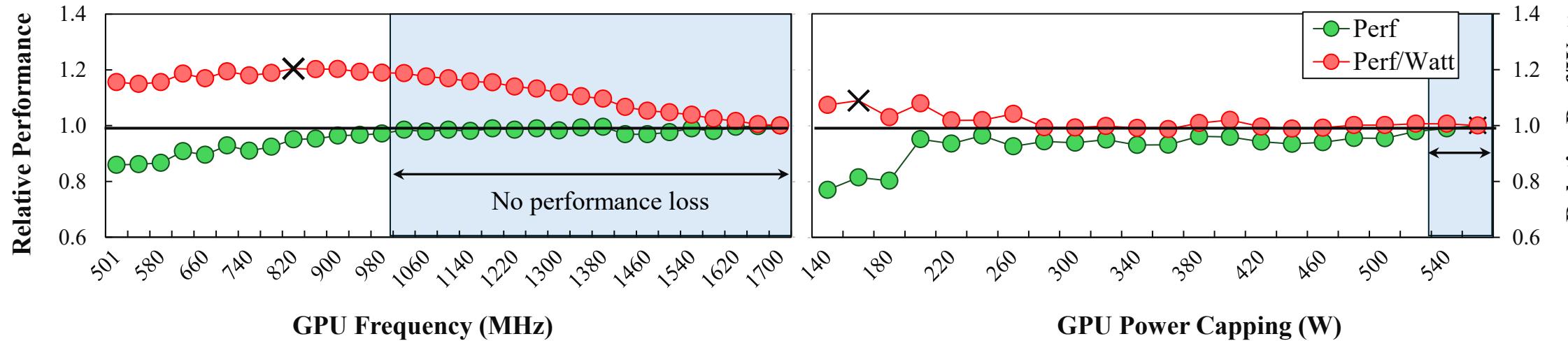


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1 node

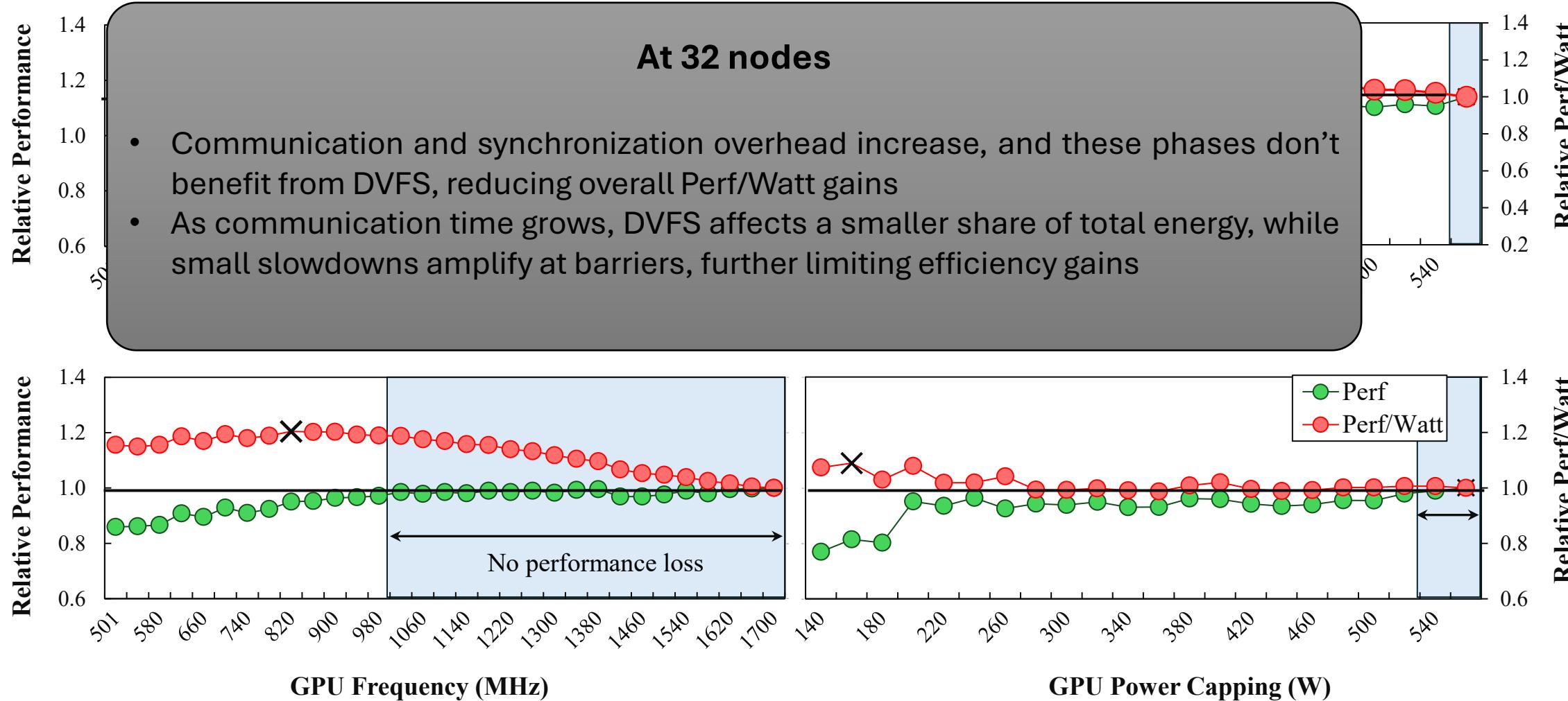


Cholla  
32 node



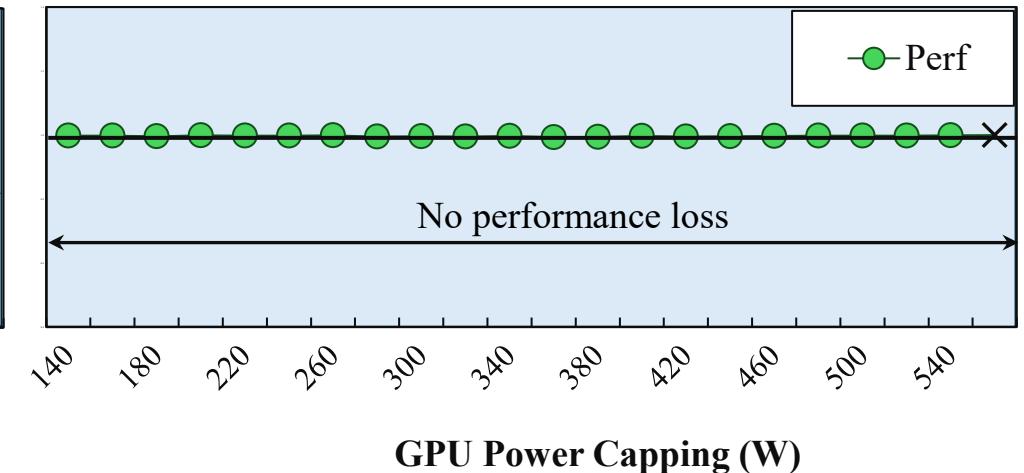
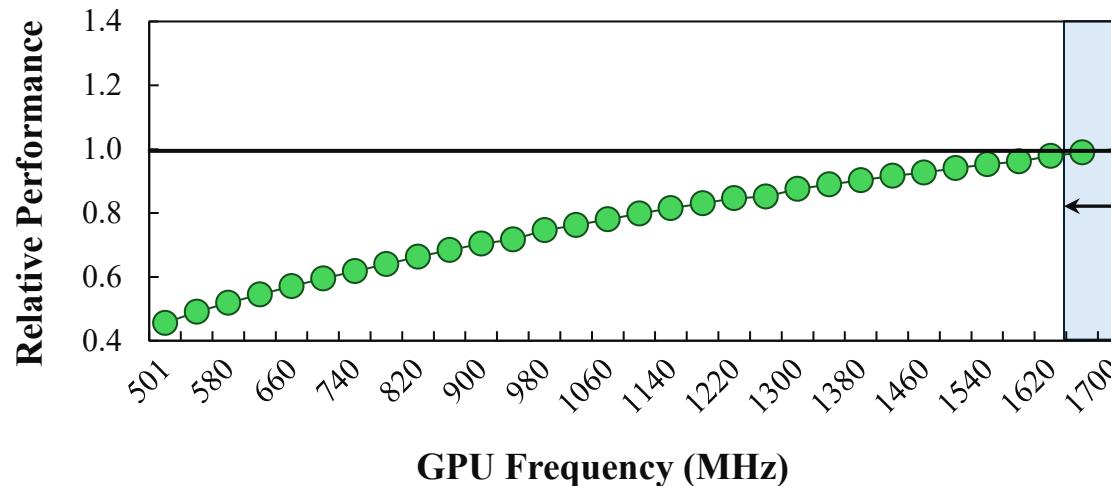
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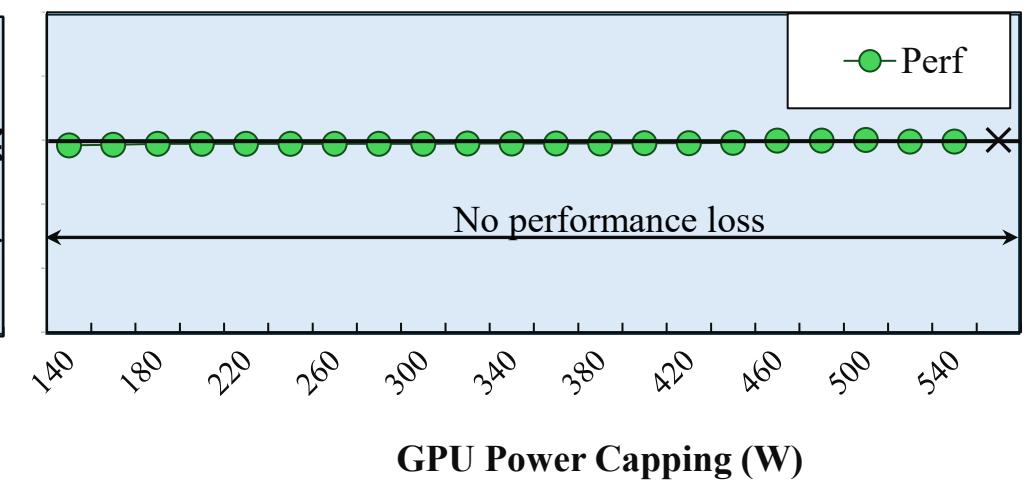
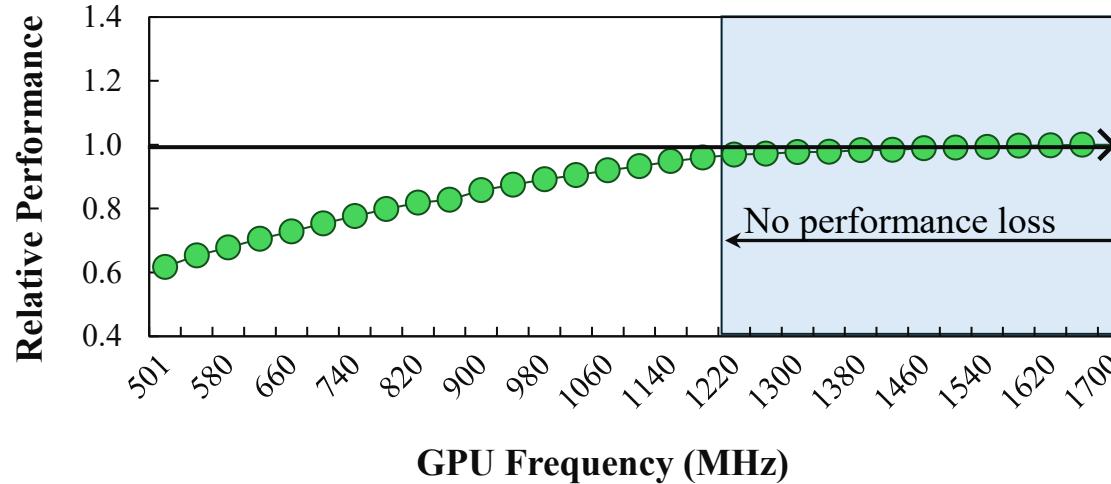


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LAMMPS  
1 node

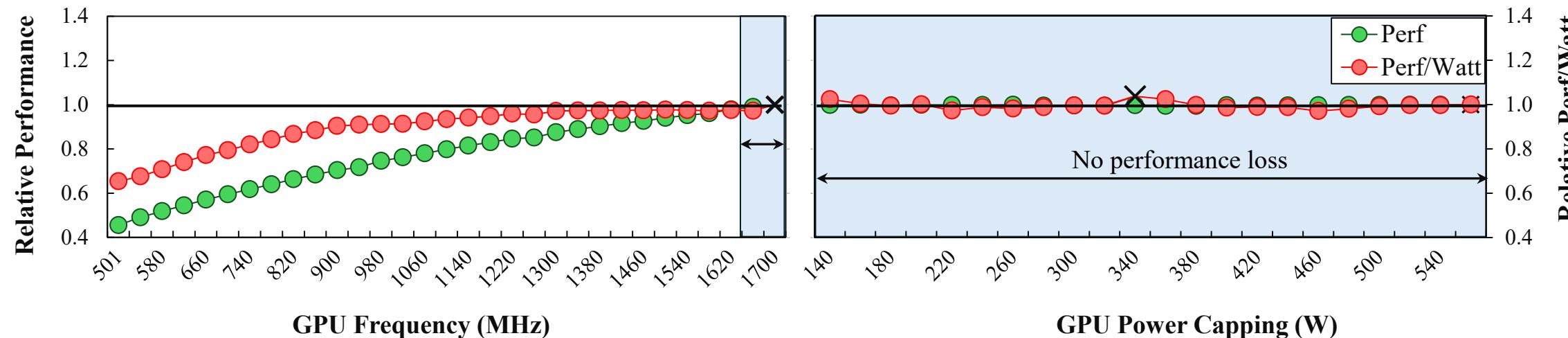


LAMMPS  
32 node

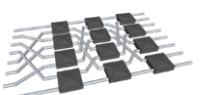
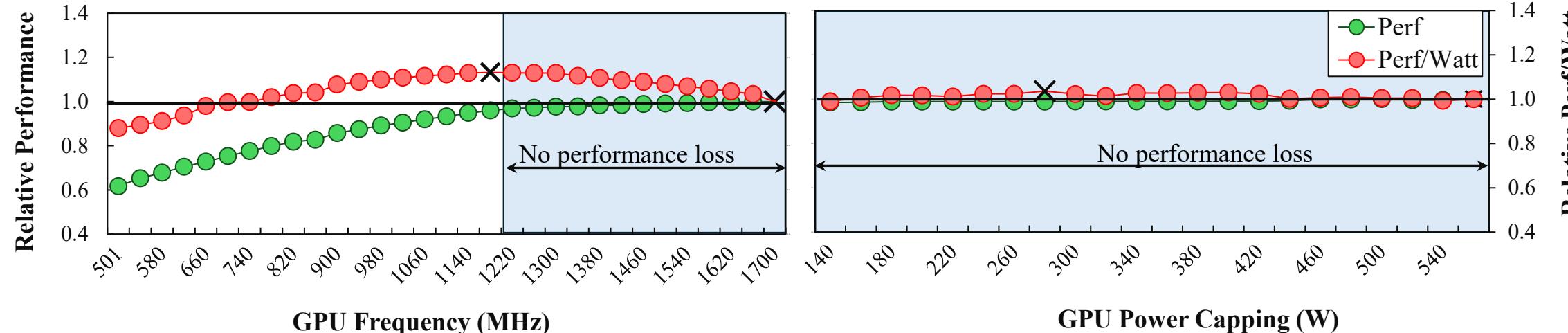


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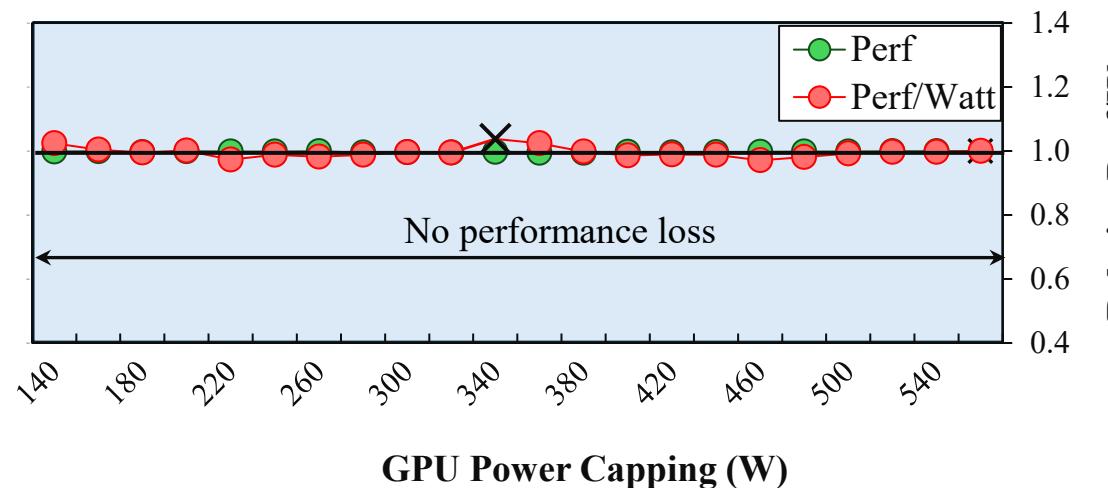
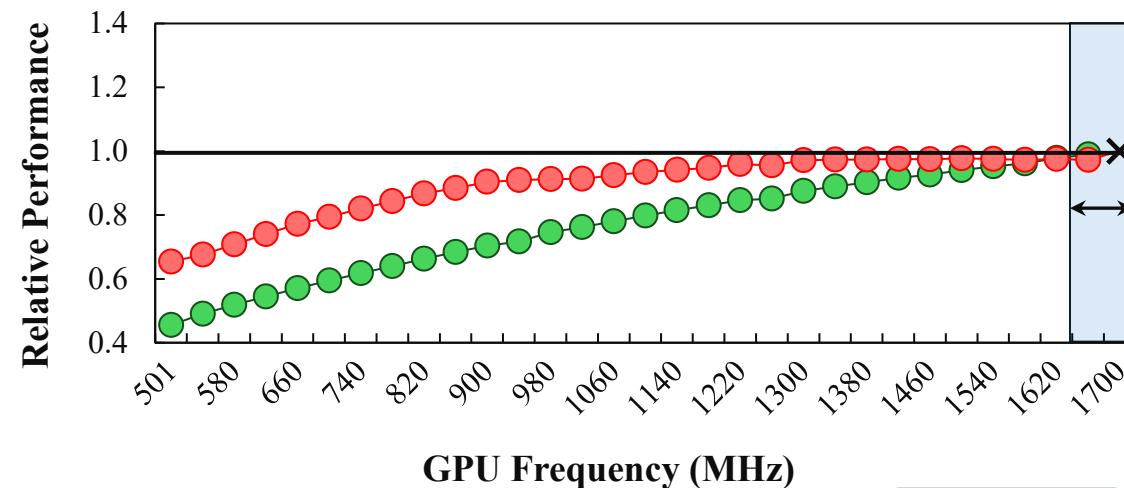


LAMMPS  
32 node

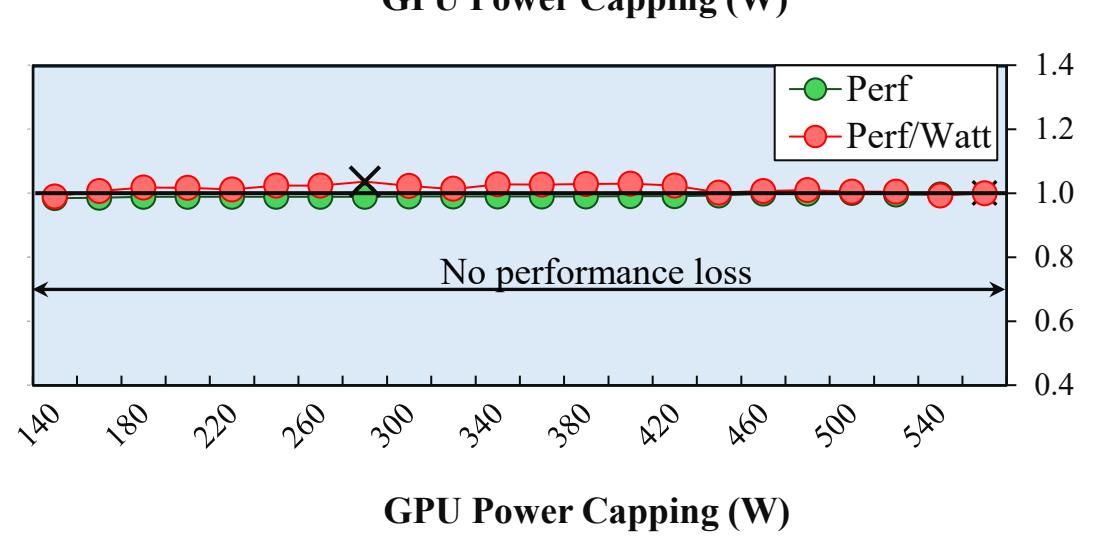
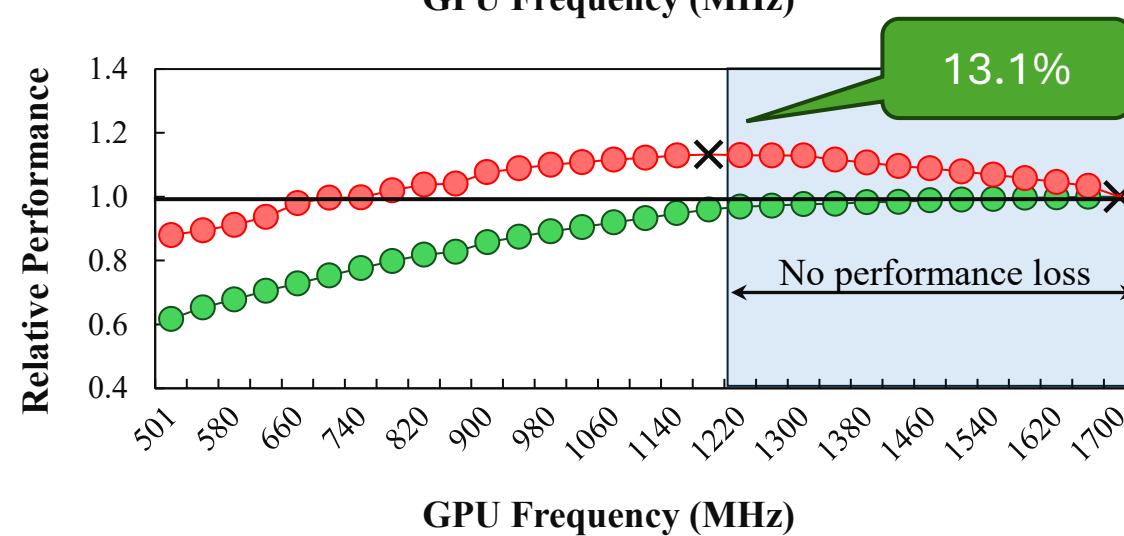


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LAMMPS  
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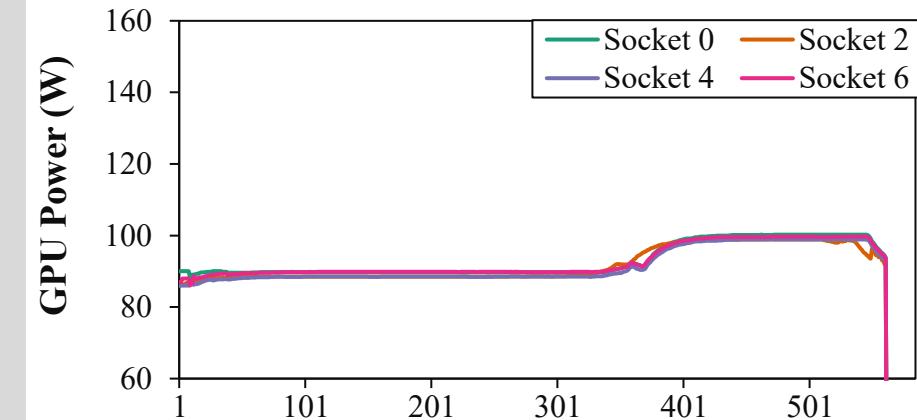
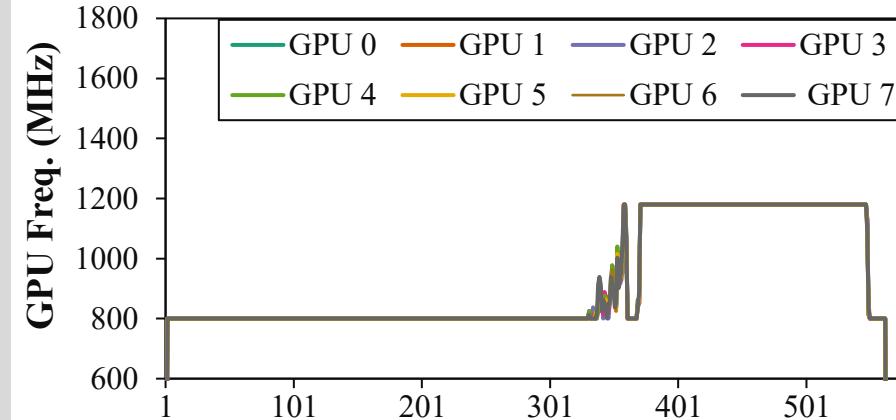


At scale, LAMMPS is bandwidth limited: fixed lower clocks cut power without hurting runtime, so DVFS beats power caps on Perf/Watt

# Evaluation: Apps that benefit more from Freq. Capping

LAMMPS  
32 node

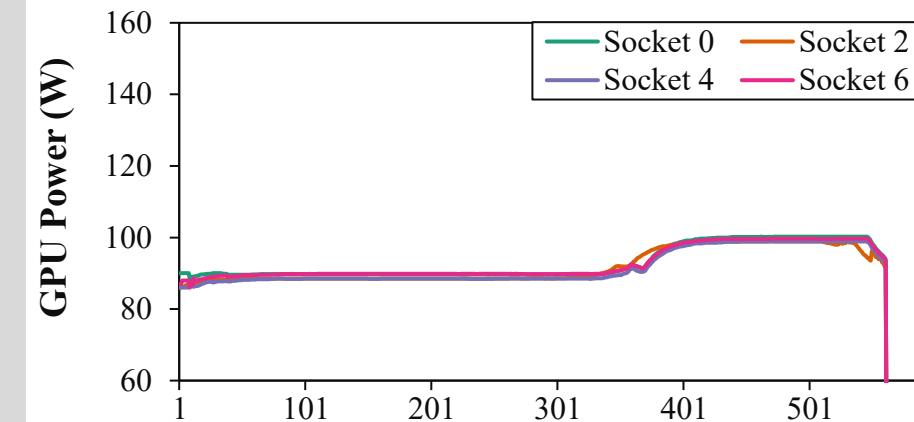
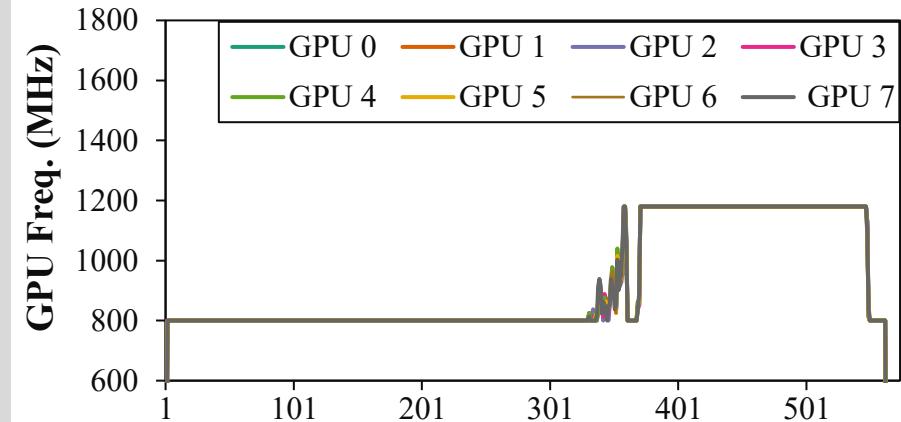
Freq. Capping  
at 1180MHz



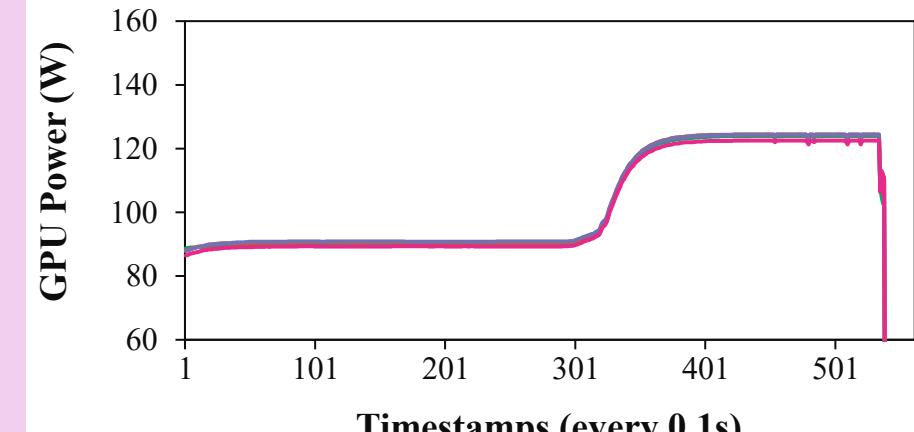
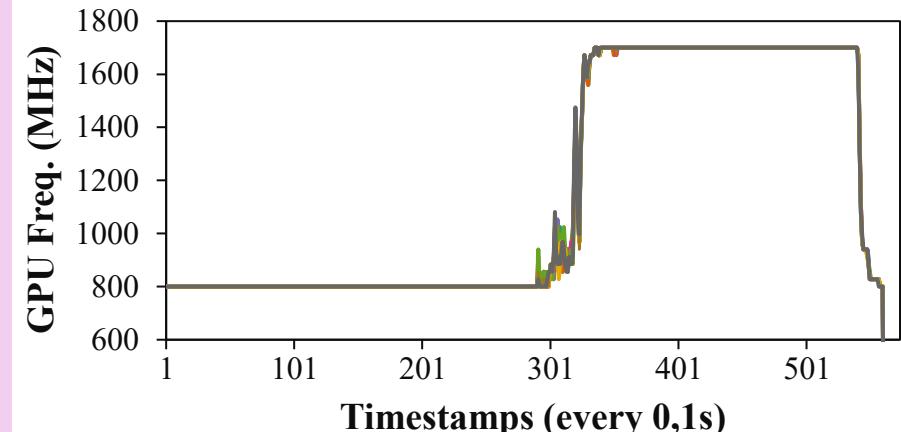
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LAMMPS  
32 node

Freq. Capping  
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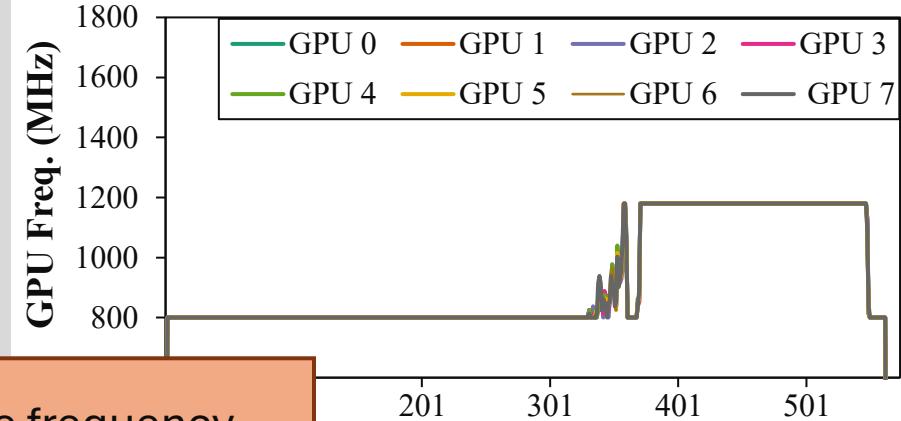
Power  
Capping at  
280 W



# Evaluation: Apps that benefit more from Freq. Capping

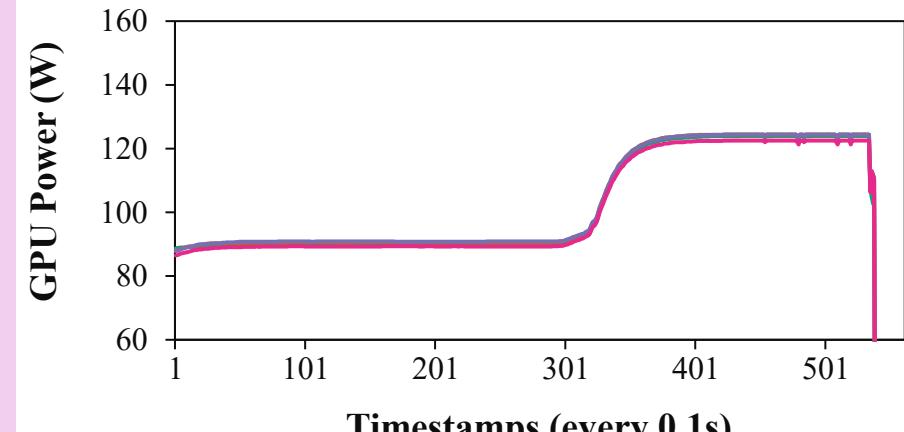
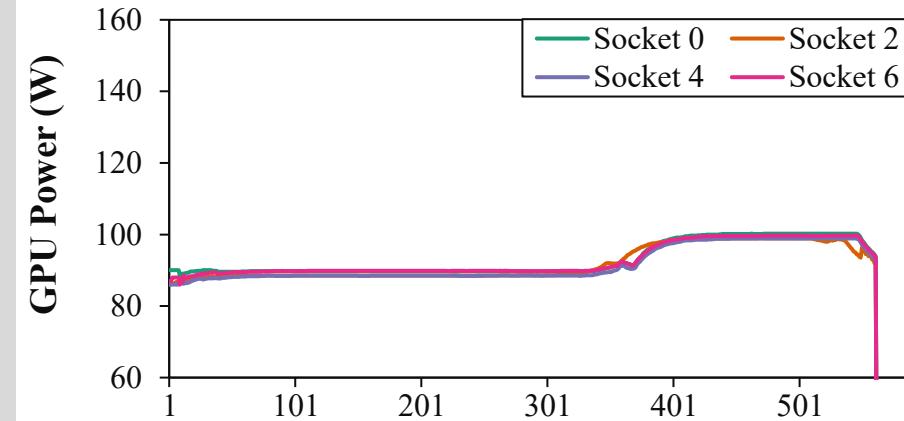
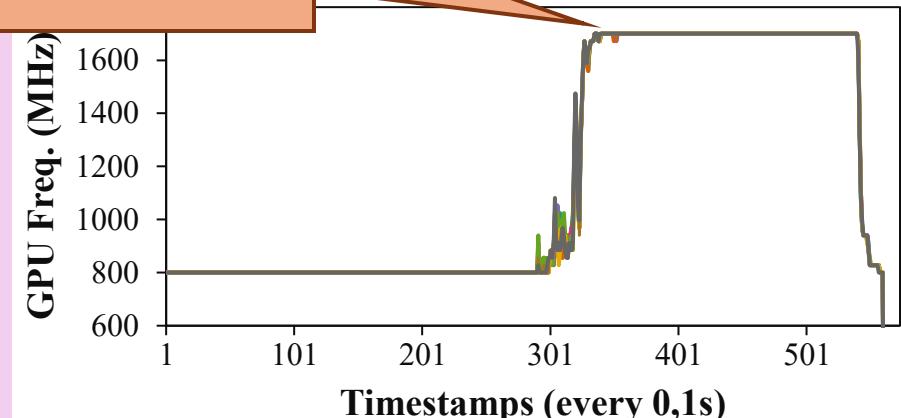
LAMMPS  
32 node

Freq. Capping  
at 1180MHz



Allowing GPU to increase frequency  
yielded no performance gain, only higher  
power and energy waste.

Power  
Capping at  
280 W

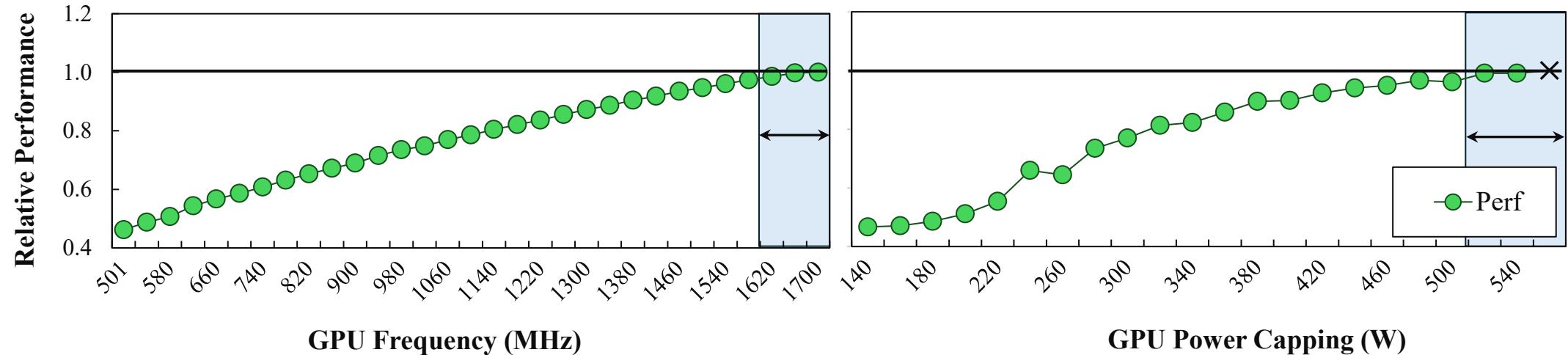


# Evaluation

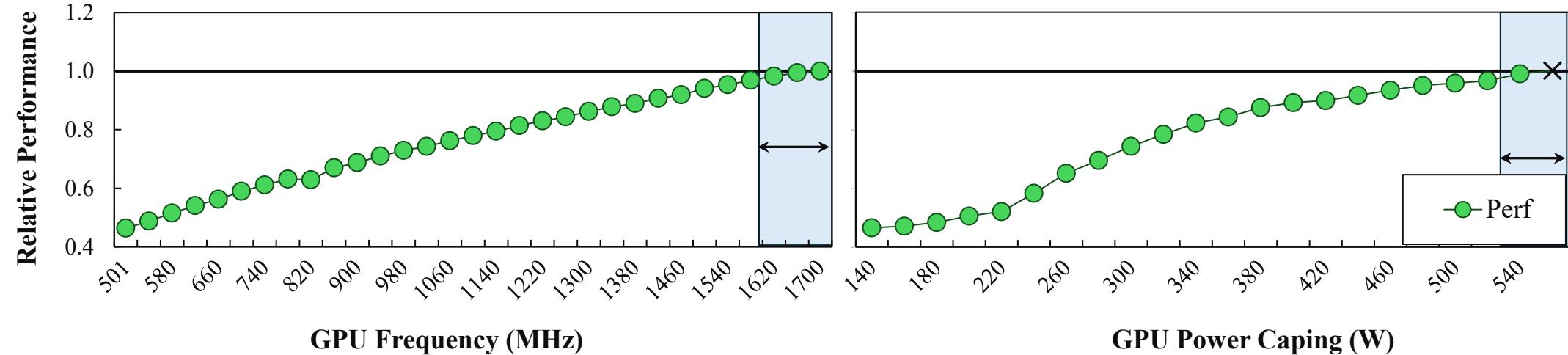
- What applications benefit more from Frequency Capping?
- What applications benefit more from Power Capping?

# Evaluation: Apps that benefit more from Power Capping

HACC  
1 node

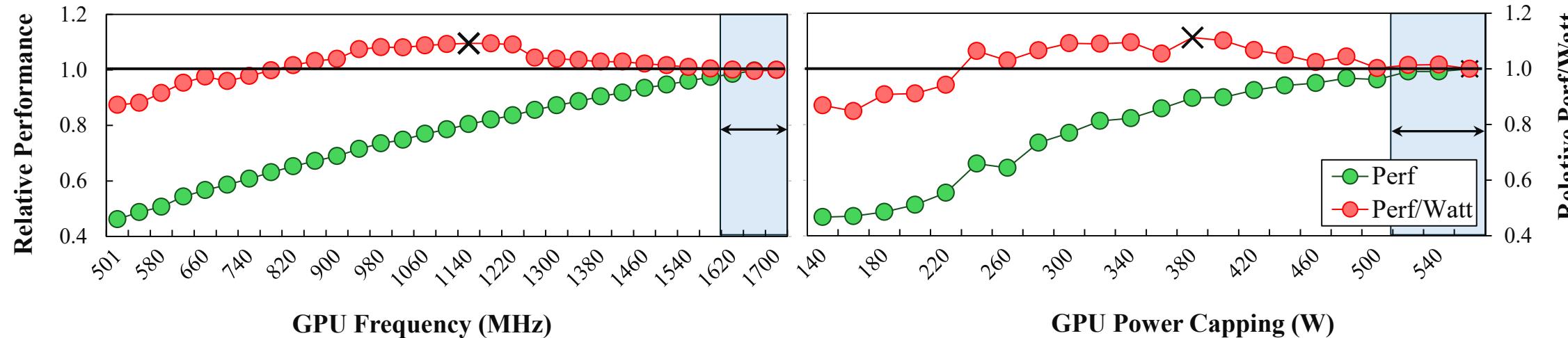


HACC  
32  
node

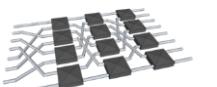
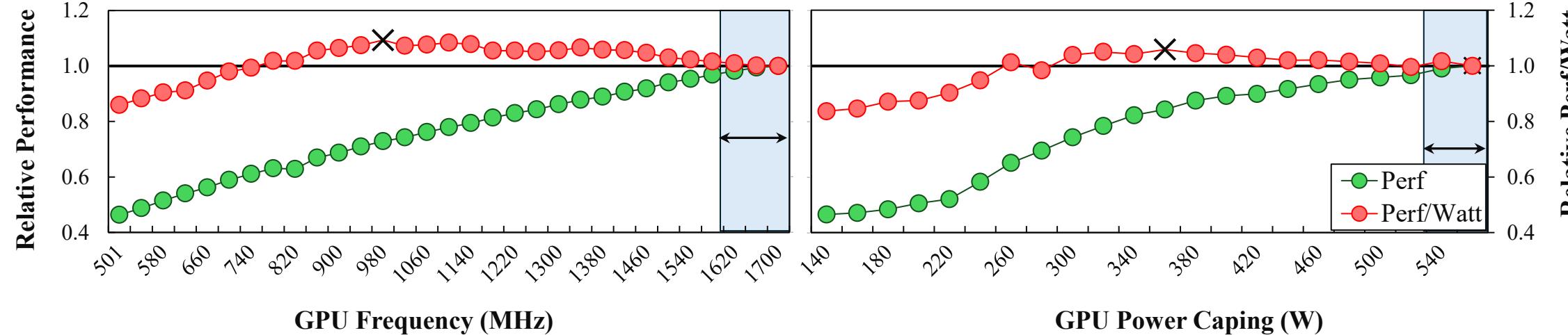


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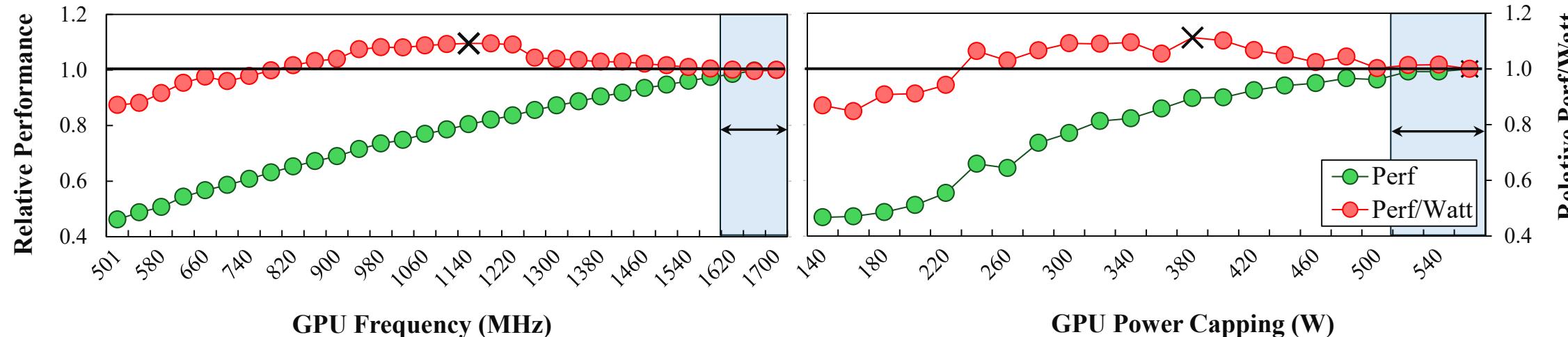


HACC  
32  
node

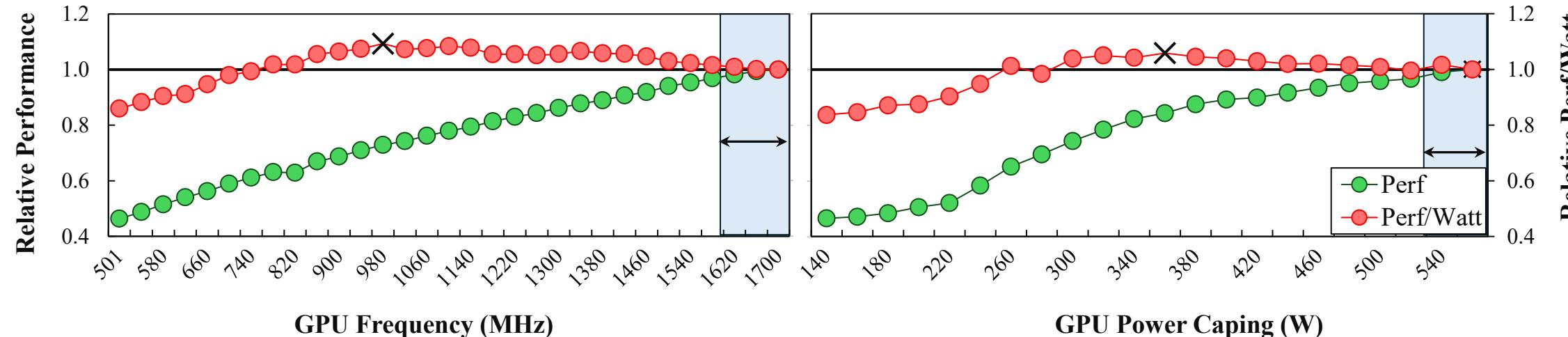


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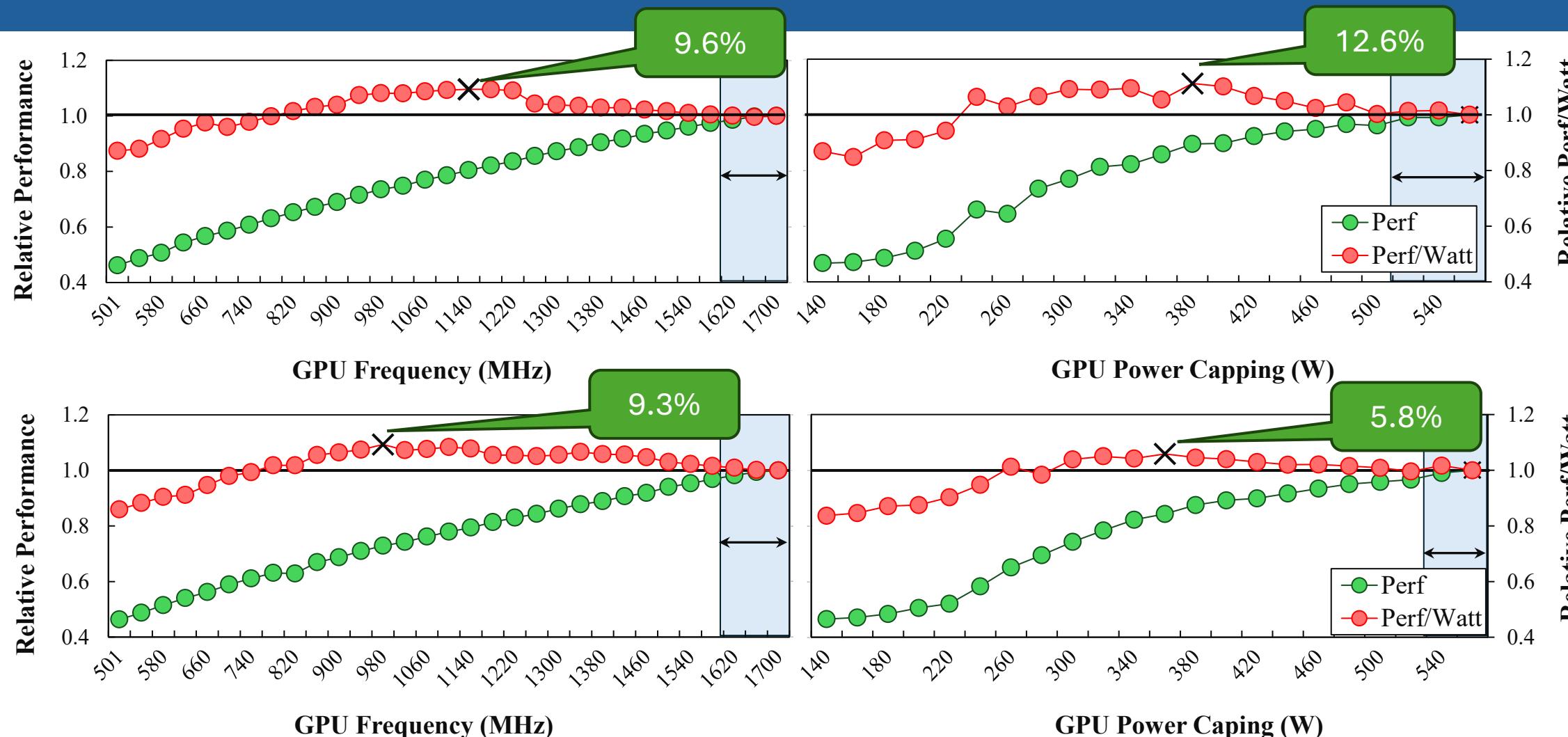


HACC  
32  
node



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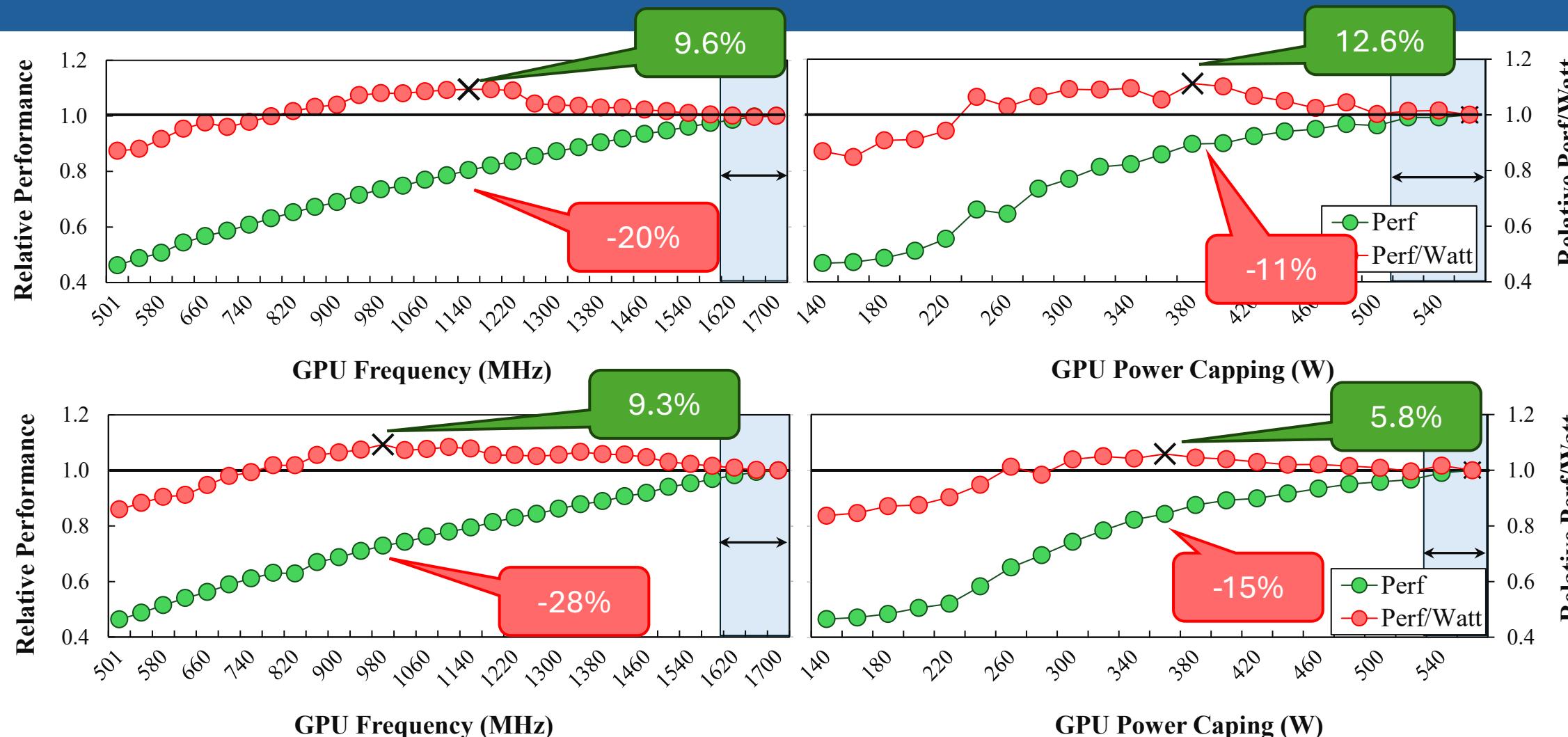
HACC  
1 node



HACC keeps GPU utilization stable at scale, so power and frequency capping drive the hardware to similar voltage-frequency states, resulting in equivalent performance-energy behavior

# Evaluation: Apps that benefit more from Power Capping

HACC  
1 node

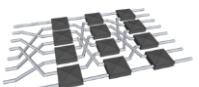
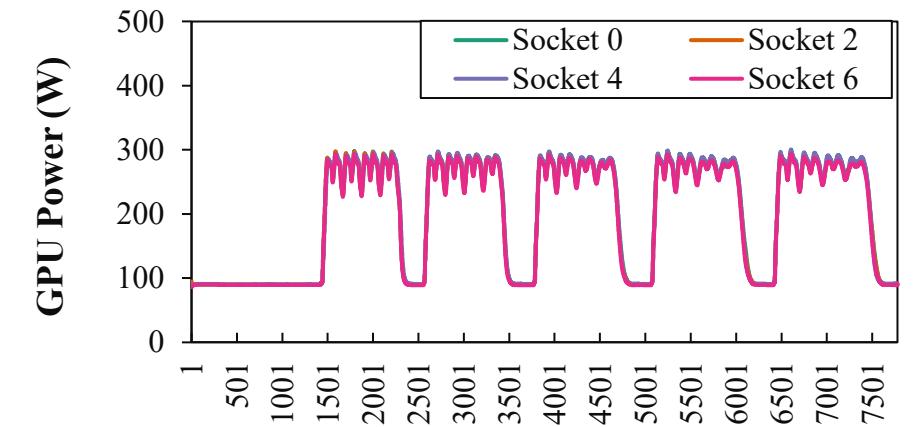
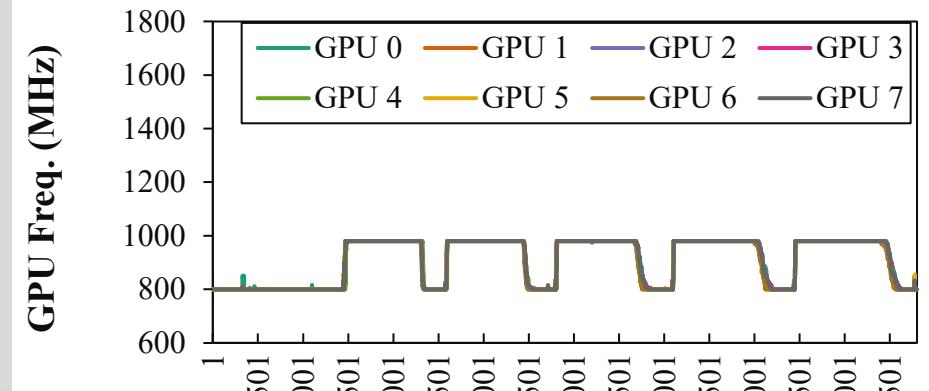


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# Evaluation: Apps that benefit more from Power Capping

HACC  
32  
node

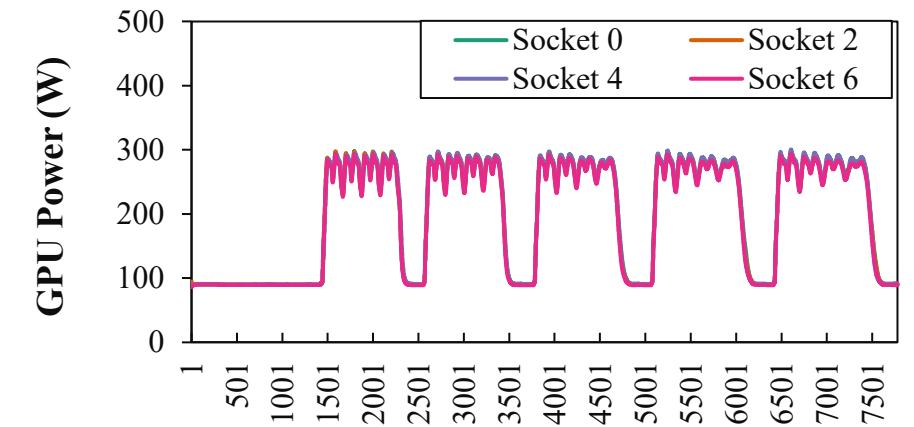
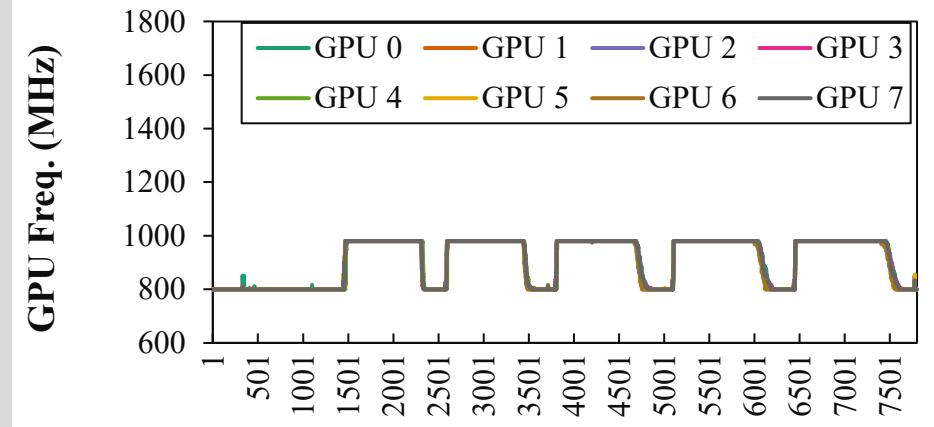
Freq. Capping  
at 980MHz



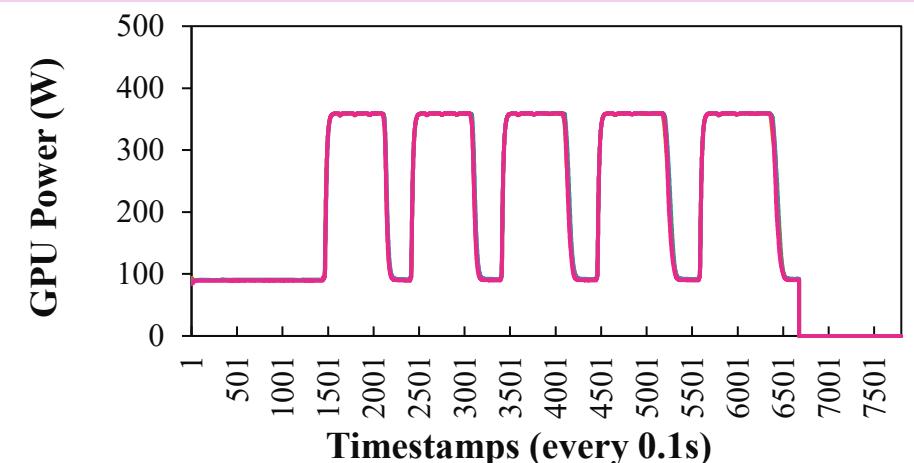
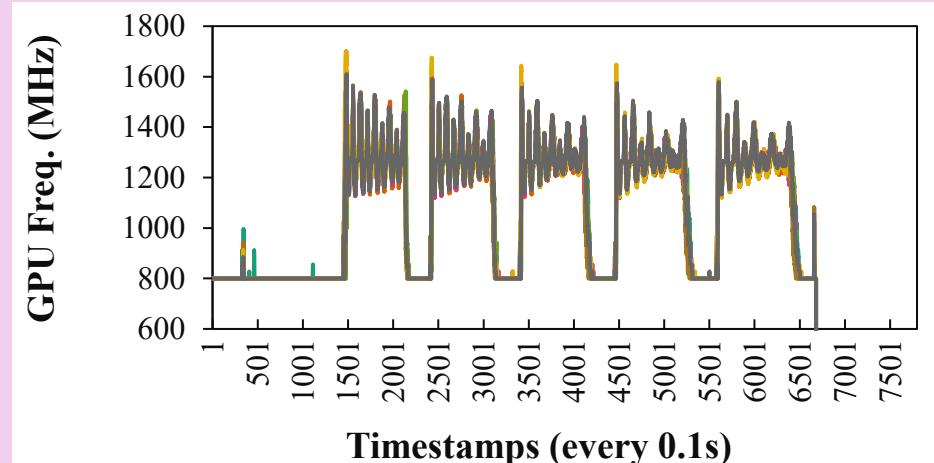
# Evaluation: Apps that benefit more from Power Capping

HACC  
32  
node

Freq. Capping  
at 980MHz



Power  
Capping at  
360 W



# Evaluation: Apps that benefit more from Power Capping

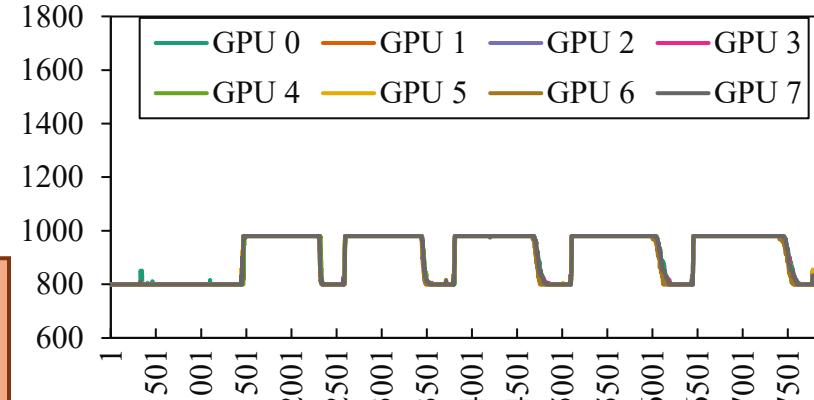
HACC  
32  
node

Freq. Capping  
at 980MHz

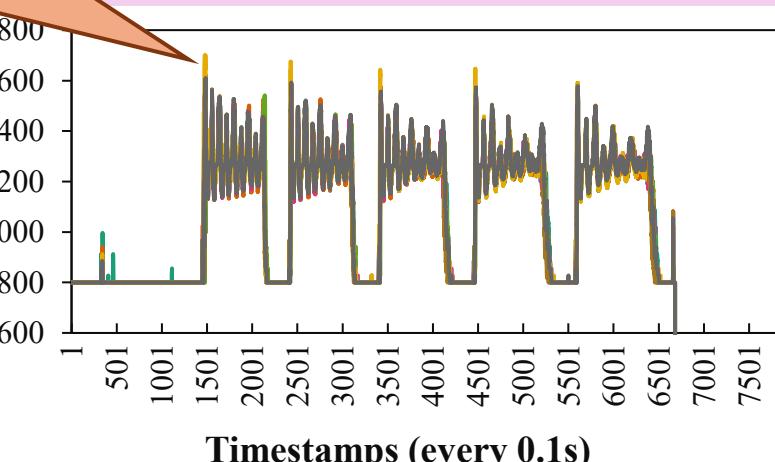
Allowing the GPU to boost frequency during burst kernels improved performance while reducing overall energy consumption

Power Capping at 360 W

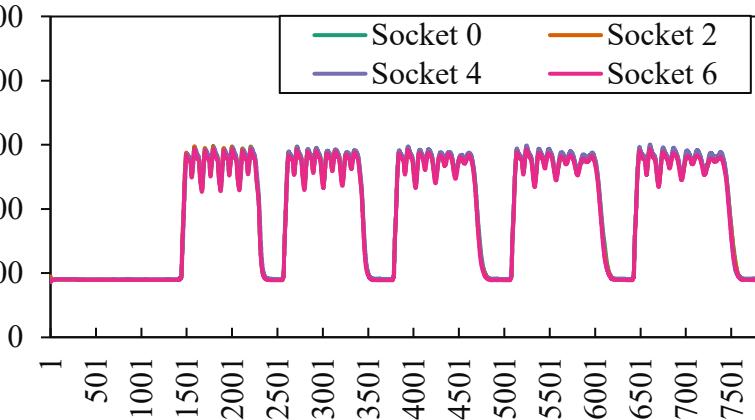
Freq. (MHz)



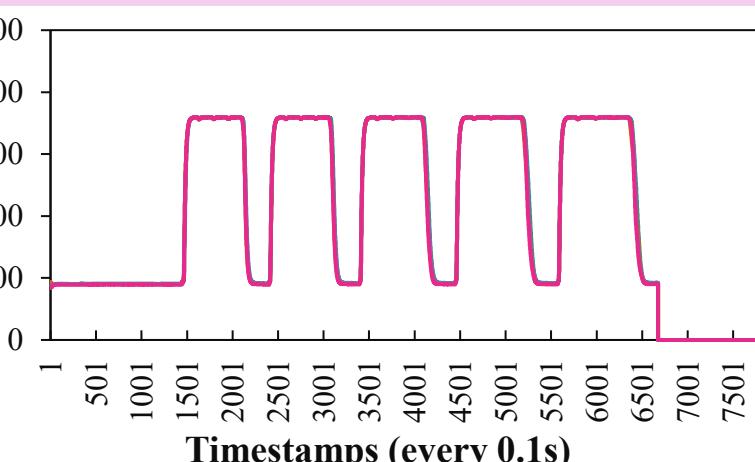
GPU Freq. (MHz)



GPU Power (W)



GPU Power (W)

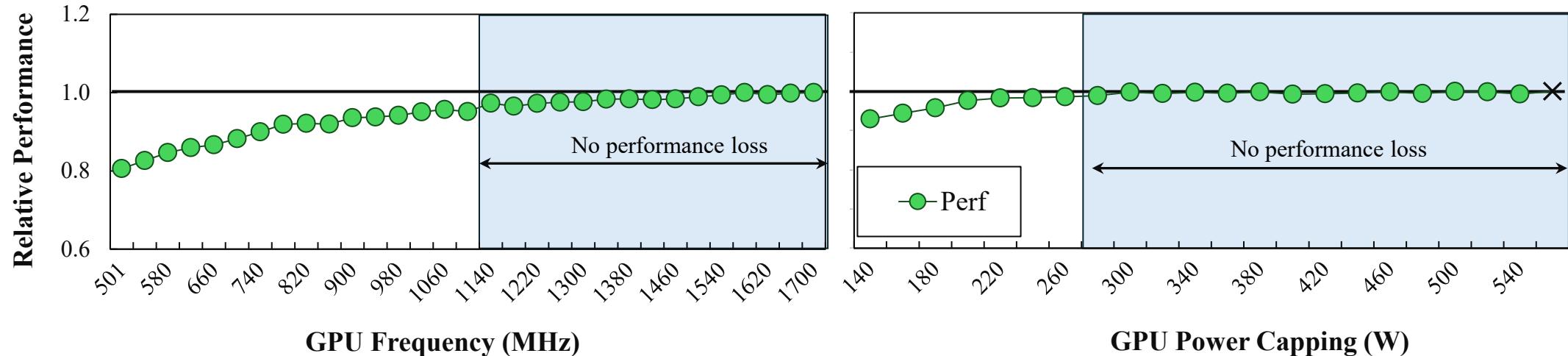


# Evaluation

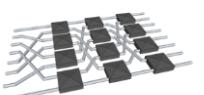
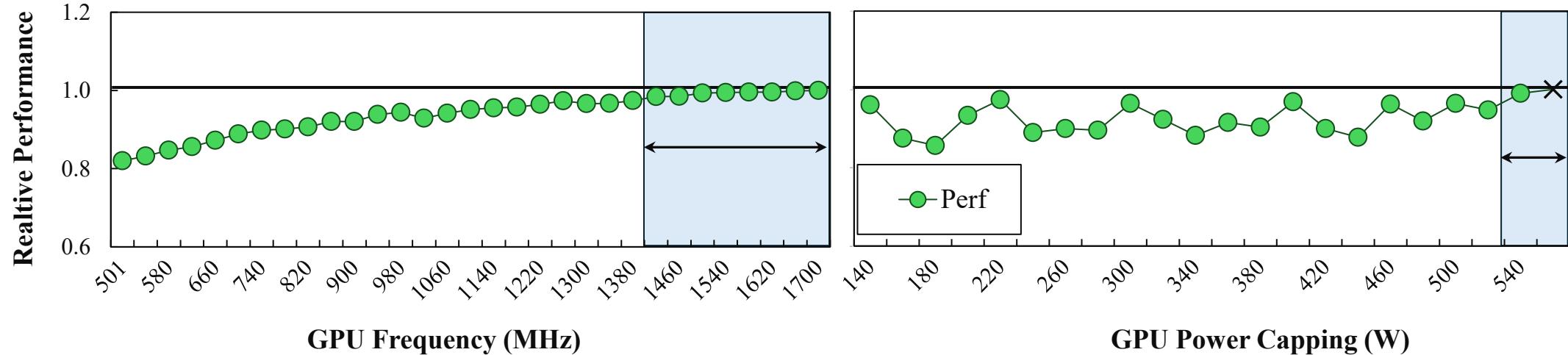
- What applications benefit more from Frequency Capping?
- What applications benefit more from Power Capping?
- What applications are impacted in similar ways?

# Evaluation: Apps with similar behavior

Kripke  
1 node

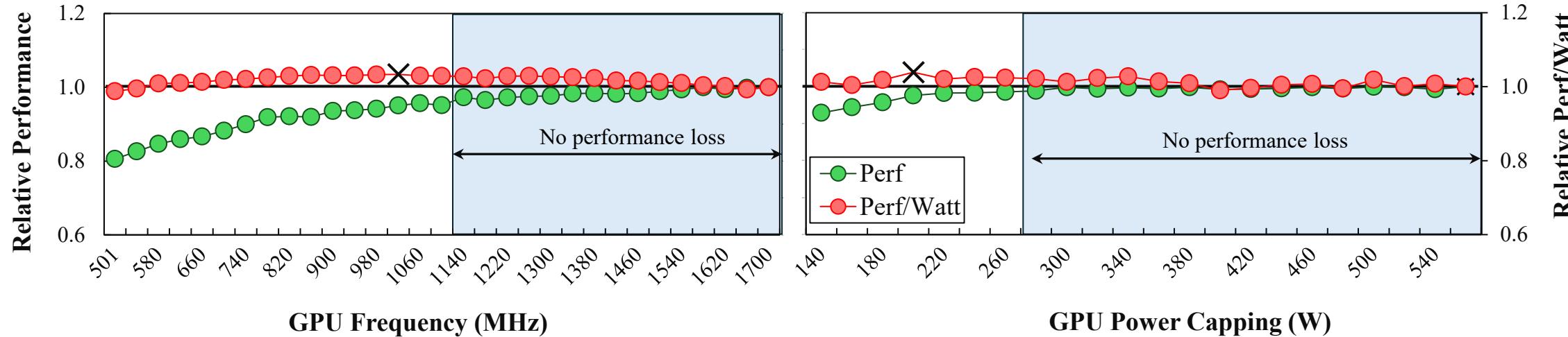


Kripke  
32  
node

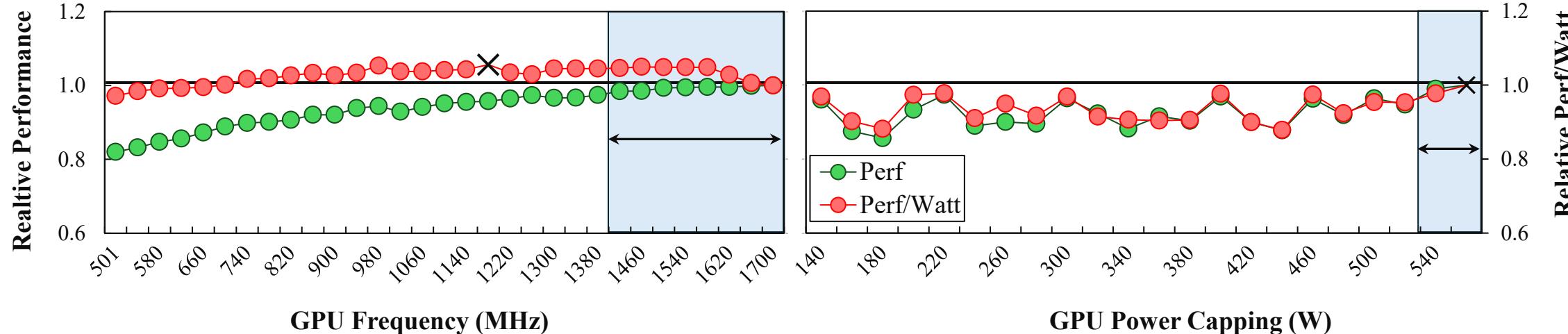


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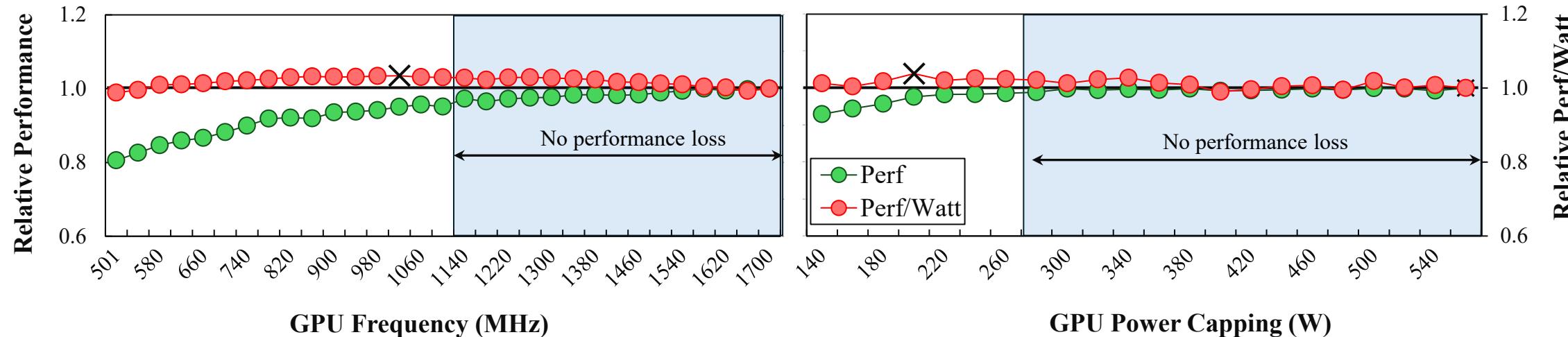


Kripke  
32  
node

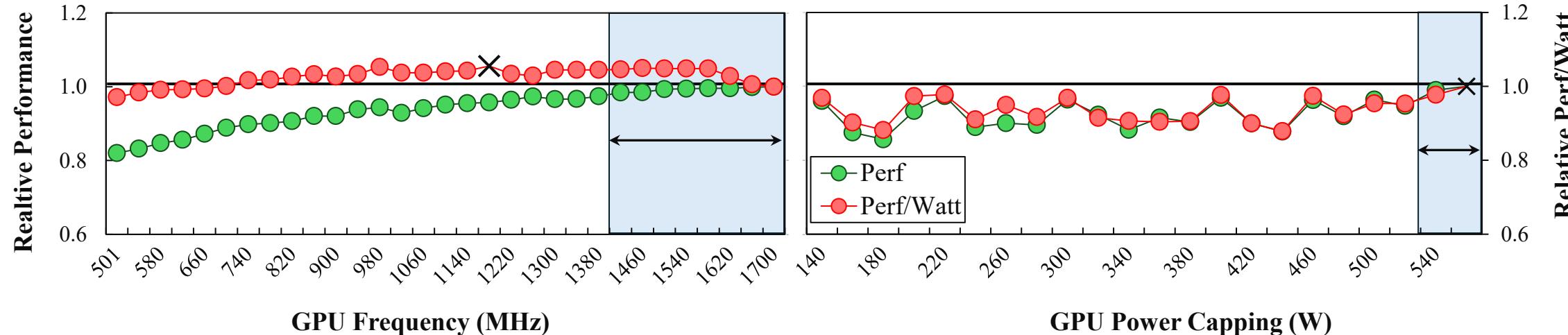


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Kripke  
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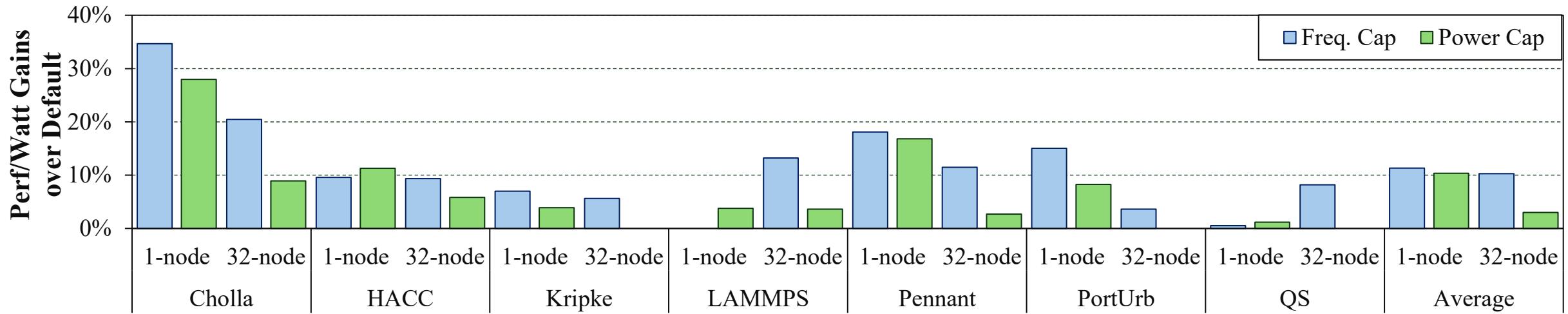


Low AI (0.1 FLOPs/byte) → performance is limited by memory access latency rather than compute throughput  
Lowering frequency or constraining power yields comparable effective frequencies, power draw, and Perf/Watt

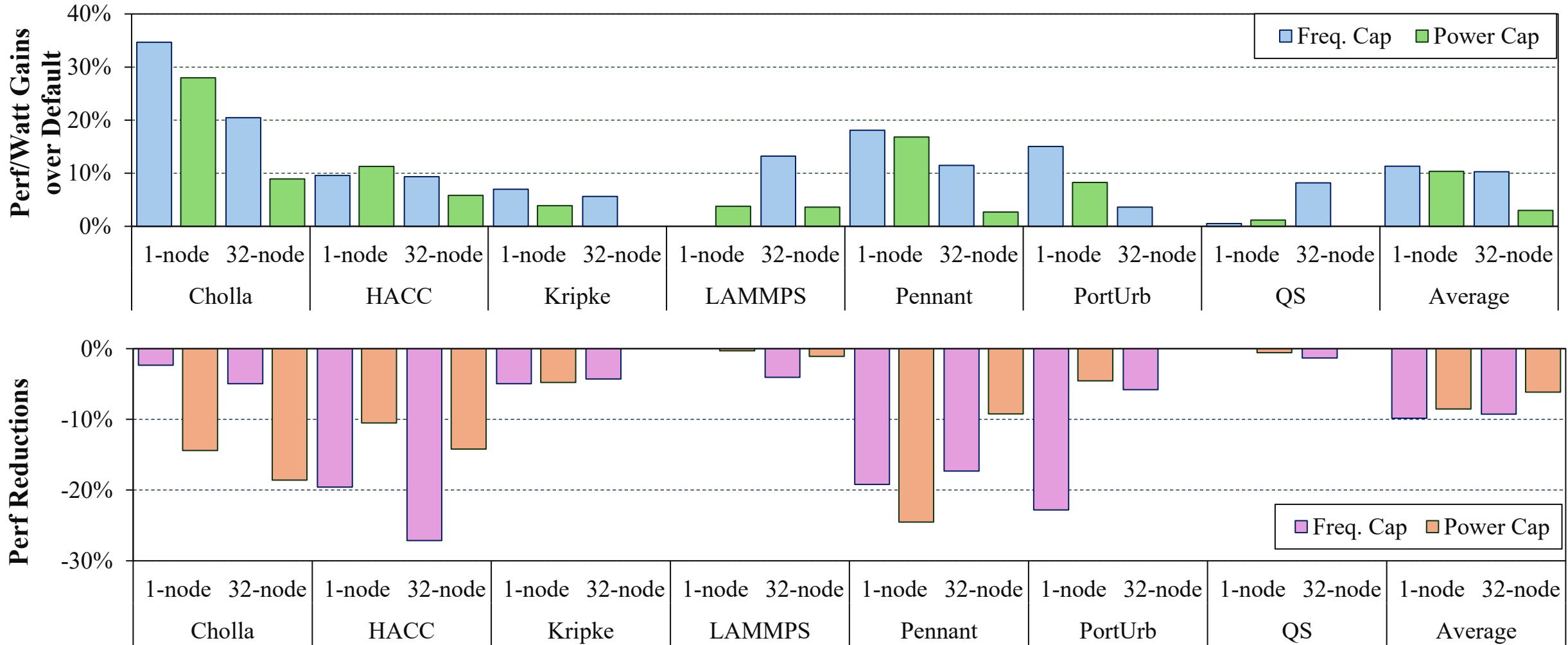
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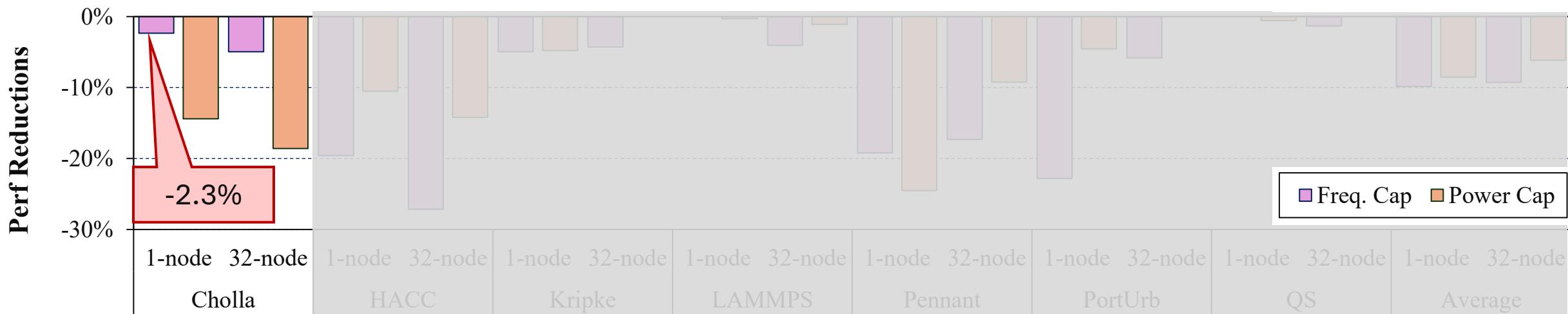
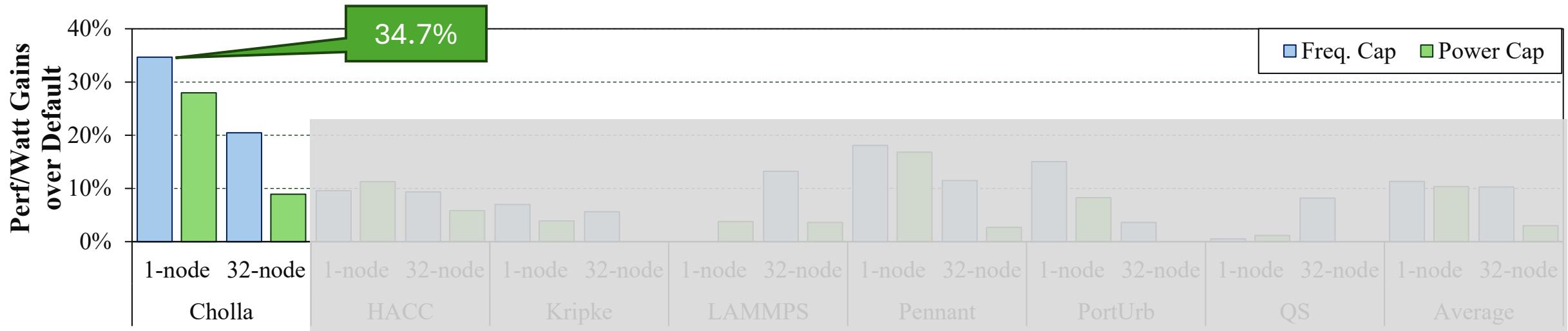
# Evaluation: What is the impact on the Performance?



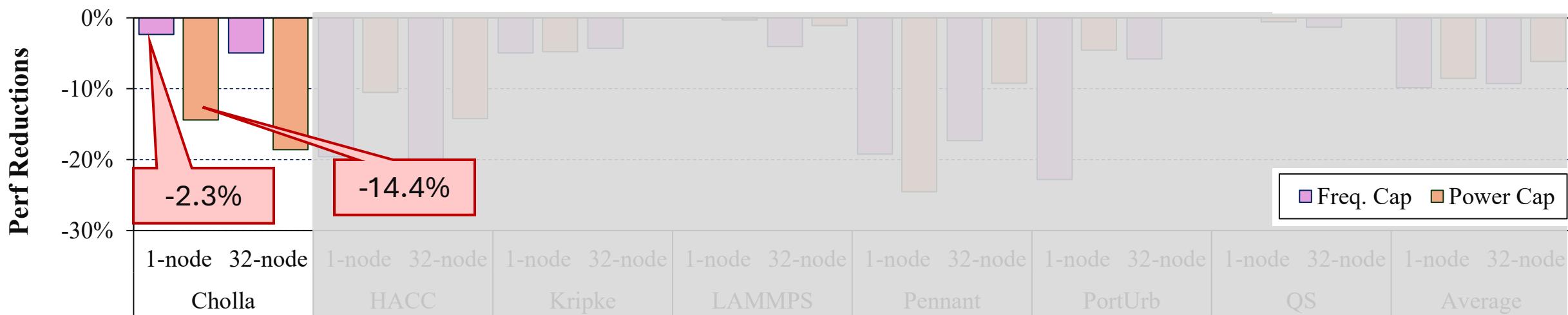
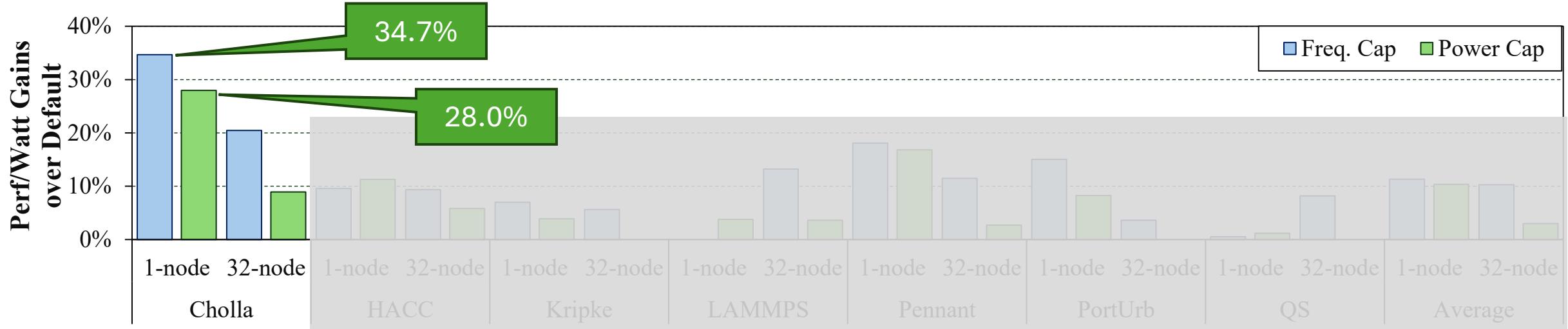
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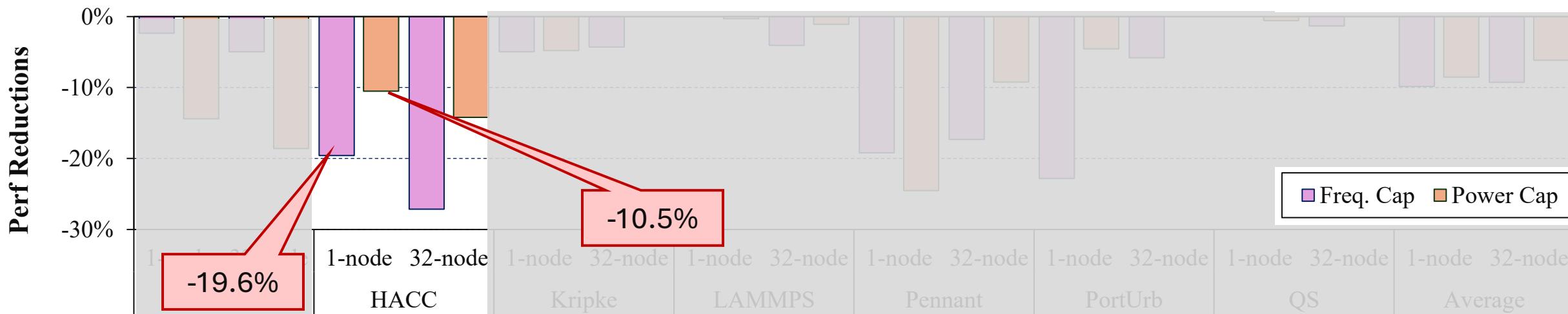
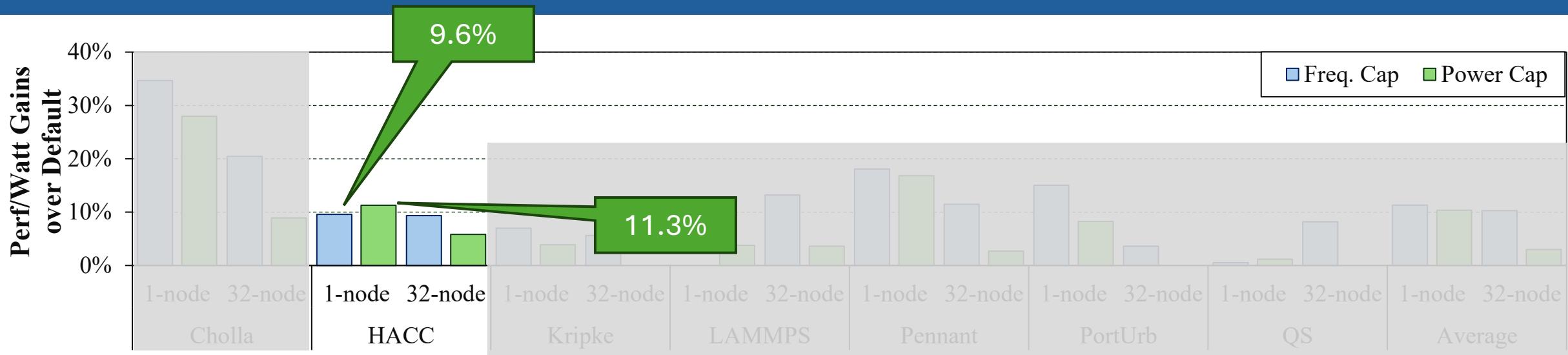
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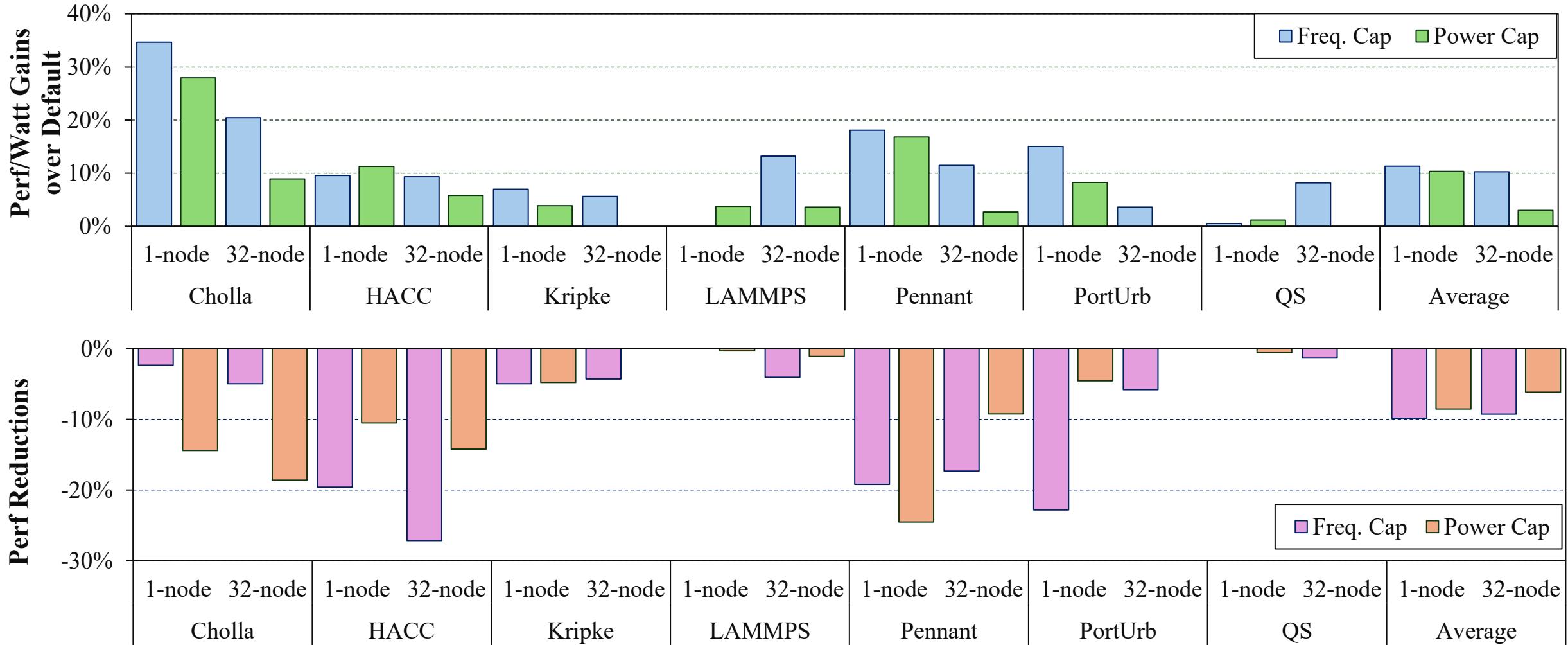
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- Discussion

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- Power management is a non-invasive, cost-effective path toward sustainable Exascale HPC.

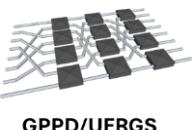
# (Ongoing) Future Works

- Understand the impact of hardware metrics on the power profile of each application
- Devise a model to automatically define the best power management technique
- Model a tool to change GPU frequency/power capping according to the active kernel
  - Implications on MPI communication, barriers, etc. etc..

# Acknowledgments

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# **Characterizing the Impact of GPU Power Management on an Exascale System**

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**Arthur Lorenzon**

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