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Enterprise



# EVEREST: AN EFFECTIVE AND VERSATILE RUNTIME ENERGY SAVING TOOL

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# INTRODUCTION

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- Power and energy consumption continue to increase worldwide, especially with surge in AI
  - Data center energy consumption has grown 20-40% annually
  - Electricity consumption from data centres, artificial intelligence (AI) and the cryptocurrency sector could double by 2026\*
- Power/energy saving opportunities exist for CPUs and GPUs
  - Using Dynamic Voltage and Frequency Scaling (DVFS) at run-time
  - Requires workload characterization to quantify impact on performance
  - Huge opportunity in GPUs
  - Lack of tools to manage CPU/GPUs considering performance/power/energy tradeoffs

\*Executive summary – Electricity 2024 – Analysis – IEA: <https://www.iea.org/reports/electricity-2024/executive-summary>

# REQUIREMENTS

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- Optimize (reduce) power and/or energy consumption with minimal performance impact
  - Provide a method to allow the specification of a maximum allowed performance loss
- User and application agnostic
  - Users should not need to provide any information about their code
  - Applications should need to be changed
- Hybrid architecture support and vendor agnostic
  - Should work not only on CPUs but also on hybrid (CPU+GPU) architectures
  - Should work on devices from different silicon vendors (Intel, AMD, NVIDIA)
- Should not interfere with applications
  - Method should generate low overhead



# BACKGROUND (CPUS VS. GPUS)

| Feature                     | CPUs | GPUs |
|-----------------------------|------|------|
| Memory BW                   |      |      |
|                             |      |      |
|                             |      |      |
|                             |      |      |
| Frequency-<br>Power Profile |      |      |
|                             |      |      |
| Other<br>Opportunities      |      |      |
|                             |      |      |



# BACKGROUND (CPUS VS. GPUS)

| Feature                 | CPUs   | GPUs  |
|-------------------------|--|---|
| Memory BW               | Use DDR memory, does not feature high BW   | Use HBM now with massive BW                   |
|                         | Latest high-end CPUs provision ~5 GB/s/core  | Latest GPUs feature HBM3, >6x BW vs. CPUs     |
|                         | Applications often memory BW bound   | High memory BW reduces application bottleneck |
|                         | <b>Implication:</b> While a significant number of routines are memory bound on CPUs and can benefit from reduced clocks, GPUs need a different line of action. |   |
| Frequency-Power Profile |  |   |
|                         |  |   |
| Other Opportunities     |  |   |
|                         |  |   |



# BACKGROUND (CPUS VS. GPUS)

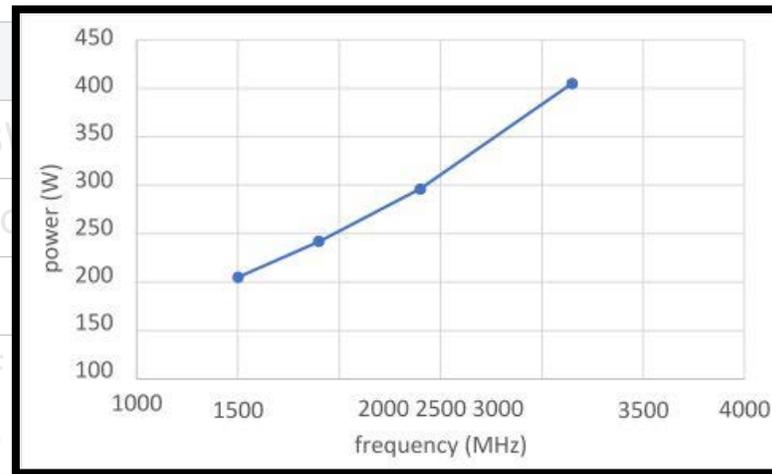
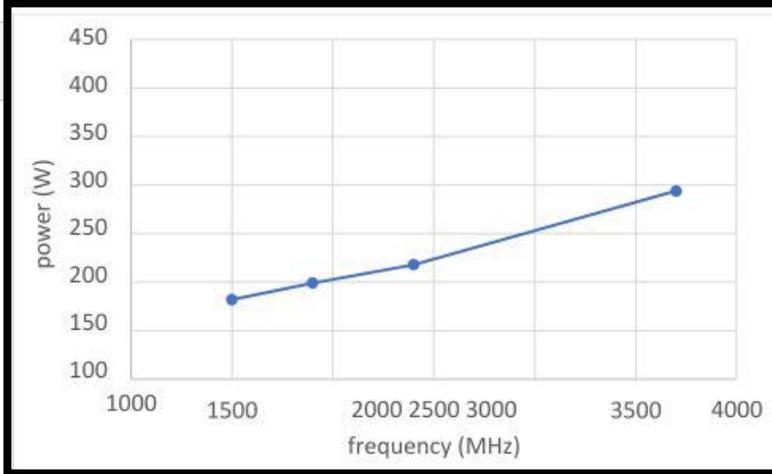
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| Frequency-Power Profile | Power increases linearly for memory bound apps, and super-linearly for compute bound apps with frequency  | GPUs designed for maximum throughput; power increases super-linearly at higher frequencies |
|                         | <b>Implication:</b> unlike CPUs, GPUs are highly energy-inefficient at the top-end of their frequency range – something that could be exploited for considerable energy savings |  |
| Other Opportunities     |   |  |
|                         |   |  |



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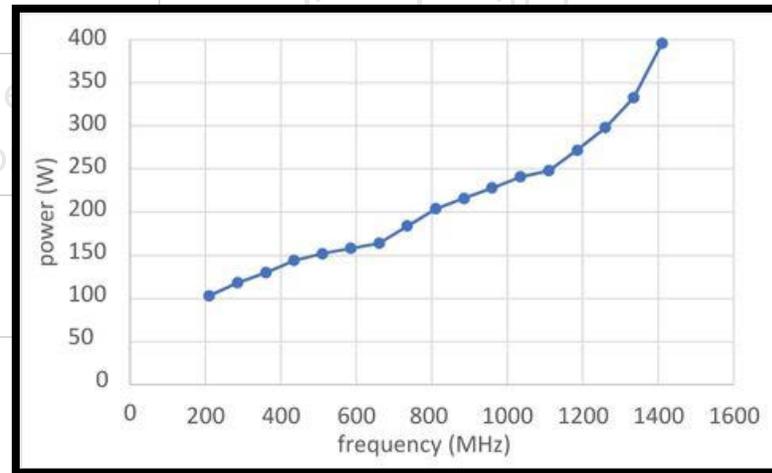
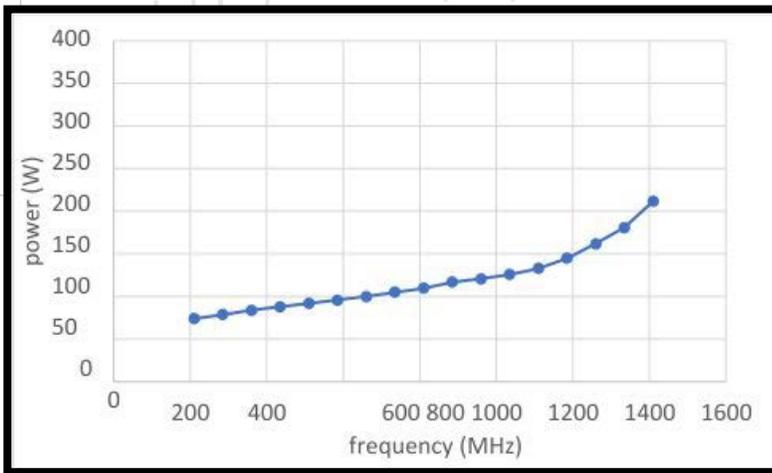
Feature

Memory BW



**CPU frequency-power profile: memory bound (lbm, left) vs. compute bound (imagick, right) application**

Other Opportunities



**GPU frequency-power profile: HPC (GROMACS, left) vs. ML (BERT, right) application**

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| Other Opportunities     | Compute bound applications rarely access data beyond the L2 cache   | GPUs might spend considerable time ‘waiting’ for work from CPUs                            |
|                         | <b>Implication:</b> Additional benefit possible from lowering uncore freq for compute bound phases on CPUs and lowering core freq in applications with low utilization on GPUs. |  |

**EVEREST**

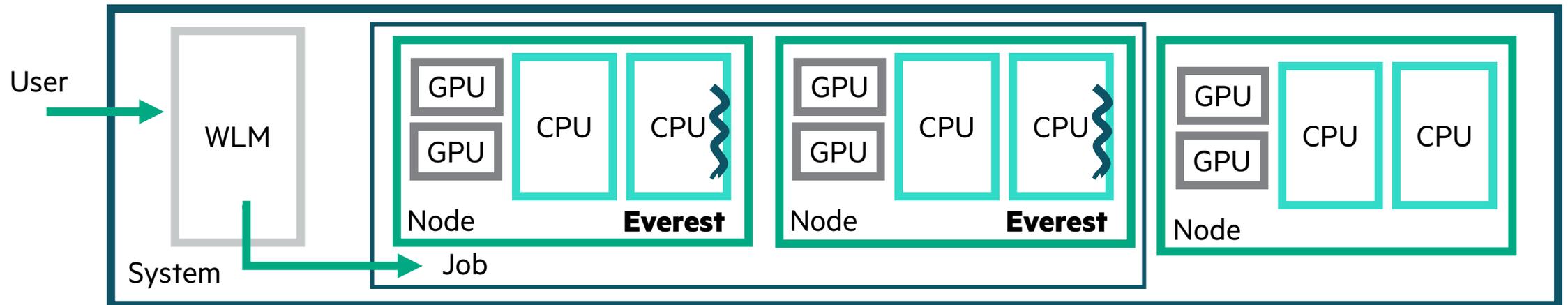
**A PROOF OF CONCEPT (POC) PROTOTYPE FOR DYNAMIC  
ENERGY OPTIMIZATION OF WORKLOADS**



# EVEREST, AN EFFECTIVE AND VERSATILE RUNTIME ENERGY SAVING TOOL

EVeREST dynamically characterizes workloads with a lightweight and portable algorithm and uses DVFS to achieve power/energy savings while meeting a specified performance guarantee.

- Relies on only 2 metrics that are standard across all architectures
  - CPUs: Instructions Per Second (IPS)
  - GPUs: GPU Utilization



# CPU APPROACH

- Goal: Predict application phase performance at different frequencies
- CPU Observations
  - When fully compute-bound, performance will vary proportionally with frequency
  - When fully memory-bound, performance does not change with frequency
  - Express this relationship as a formula:

$$\%CB = 100\% * \frac{\frac{IPS_{high}}{IPS_{low}} - 1}{\frac{Freq_{high}}{Freq_{low}} - 1}$$

$$\%MB = 100\% - \%CB$$

- Thus, when measuring IPS at a high frequency and at a low frequency, one can determine the compute- and memory-boundedness of an individual function (sensitivity analysis)



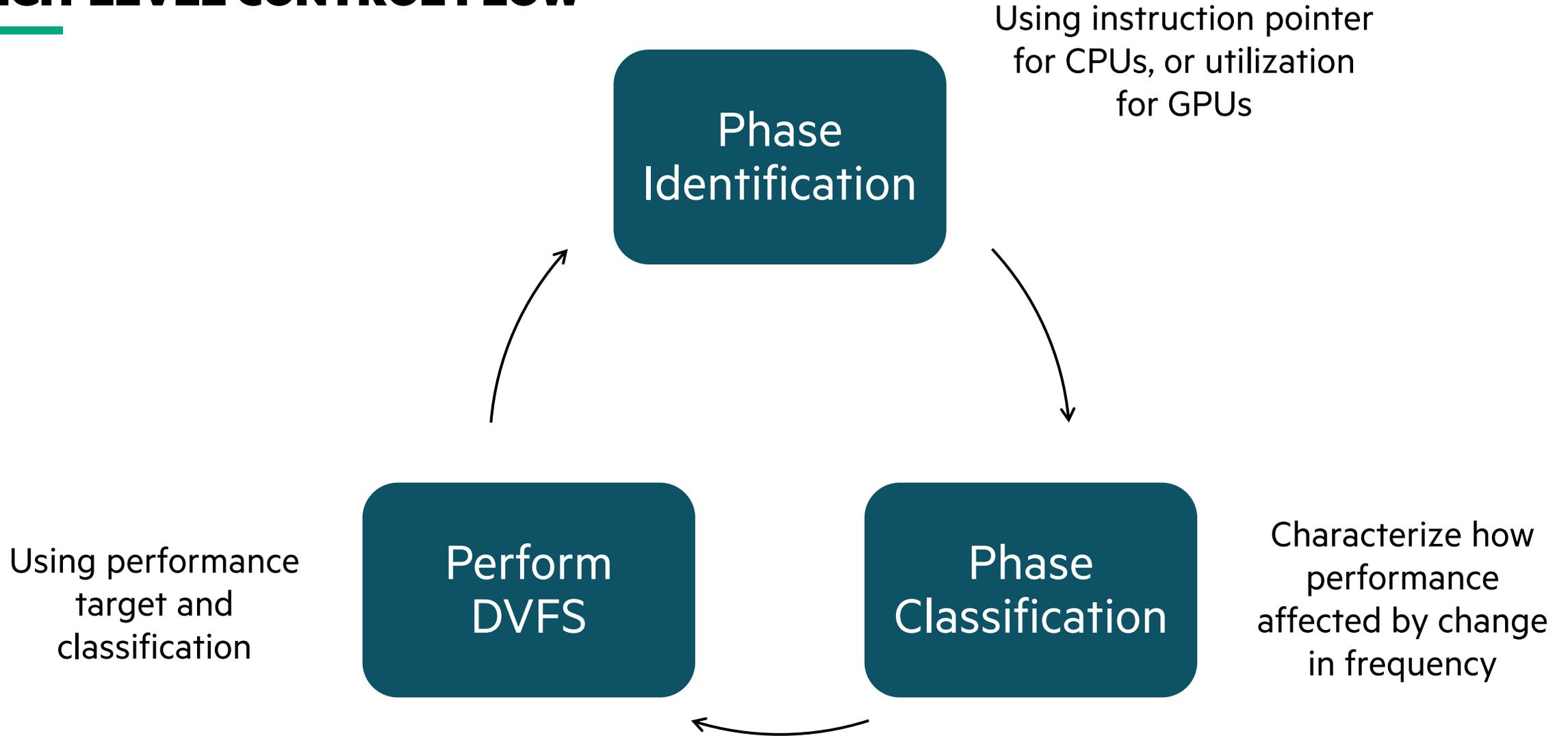
# GPU APPROACH

- Profiling on GPUs associated with significant overhead (1.5x to >3x)
  - Stack walking, kernel serialization
  - Need alternative to directly measure performance used for CPUs
- GPU Observations
  - GPU utilization is a metric directly available without profiling
  - In simple terms, Utilization can be expressed as: (kernel runtime **K**, application wall clock time **WCT**)

$$Utilization = \frac{K}{WCT}$$

- Many applications overlap GPU kernel execution with CPU code or memory transfers between device and host.
  - Application performance may become limited by either the CPU or the memory transfer time and not the GPU.
  - When clock reduces, K increases. If Utilization also increases proportionally to K, it implies WCT is independent of GPU clock.
- Thus, like CPUs, if we measure Utilization at a high frequency and at a low frequency, then we can predict application performance.

# HIGH-LEVEL CONTROL FLOW



# EVEREST RESULTS



# RESULTS

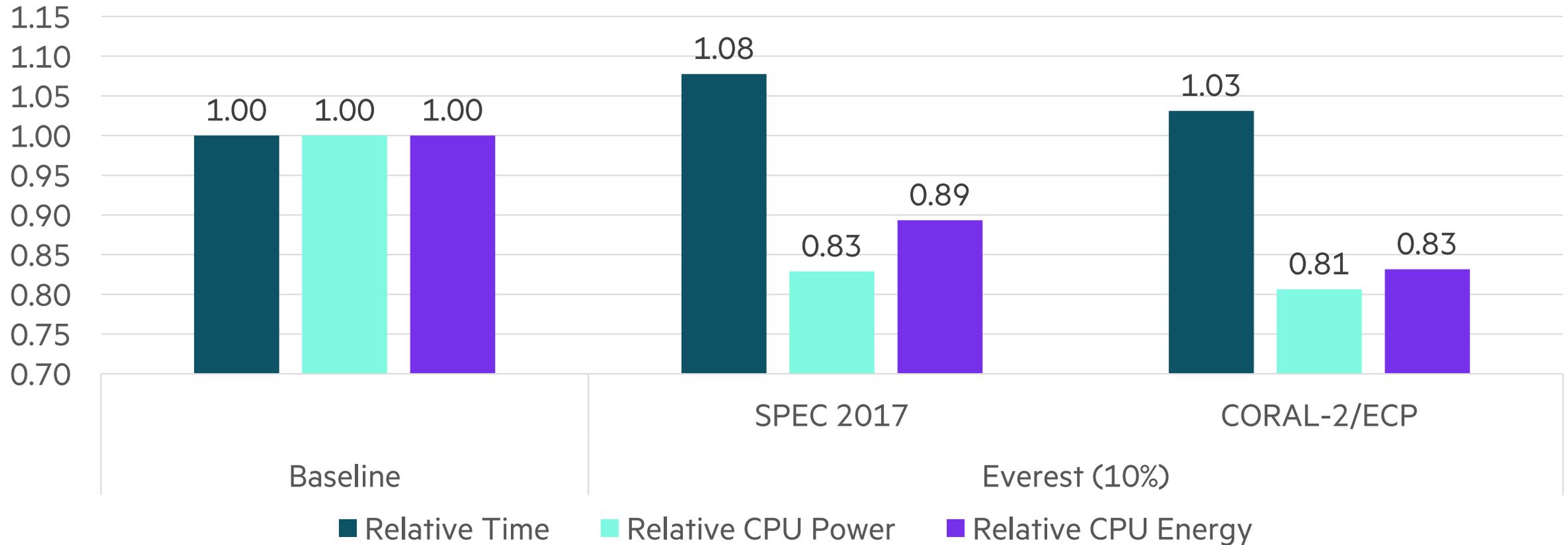
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- Evaluation on latest generation CPUs and GPUs – Intel Sapphire Rapids, AMD Genoa, NVIDIA A100
  - 27 CPU apps: 22 from SPEC 2017, 5 from CORAL-2/ECP
  - 6 (9) GPU apps: 3 (6) from HPC, 3 from AI/ML
- Evaluated at different levels of acceptable performance loss (5%, 10%, and 20%)
- Usage:
  - User submits job with additional parameter for acceptable performance loss
  - `srun ... --use-everest:pd ...`
    - Users can specify the maximal performance reduction they are willing to incur
    - Does not require modifying the application source
    - Does not depend on a specific compiler and MPI



# CPU RESULTS

## CPU Summary (Geomean)

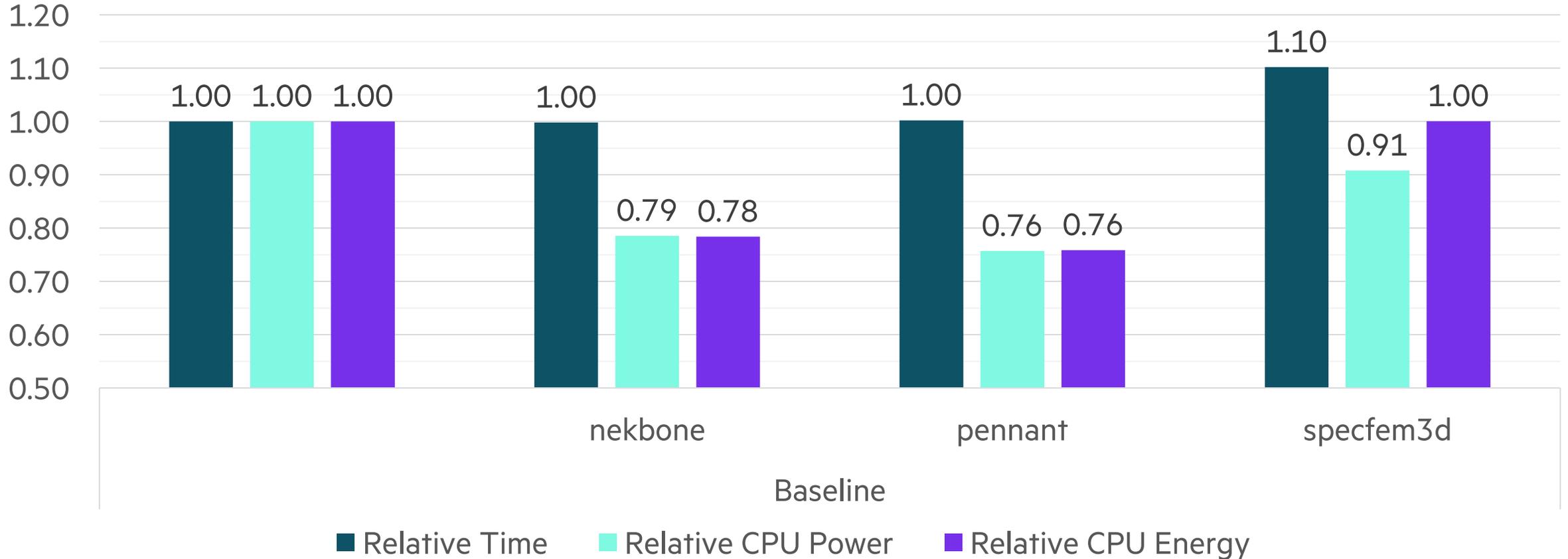


Memory-bound applications provide opportunities for 20-30% energy savings at minimal performance loss, while compute-bound applications can still achieve power savings proportional to the acceptable performance loss.



# CPU HIGHLIGHTS - MPI WORKLOADS

CORAL-2/ECP Applications

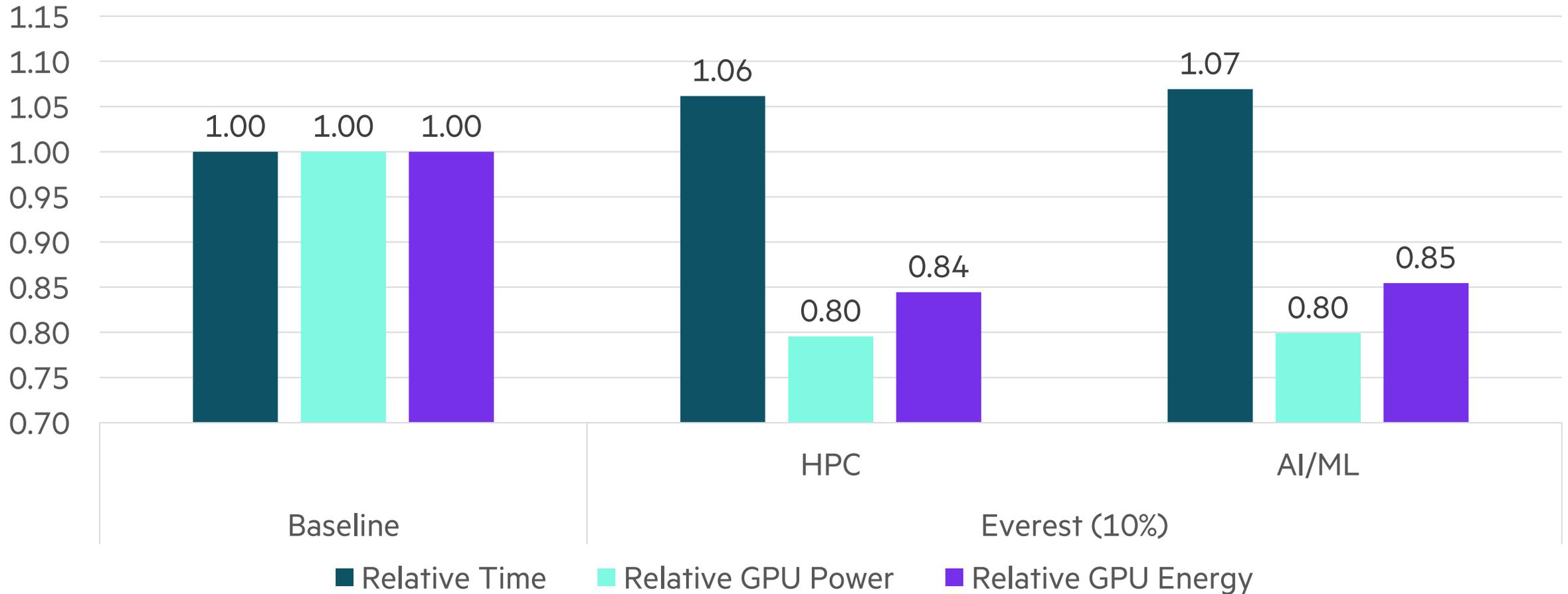


Everest can exploit any opportunities to save power and energy during intensive communication phases.



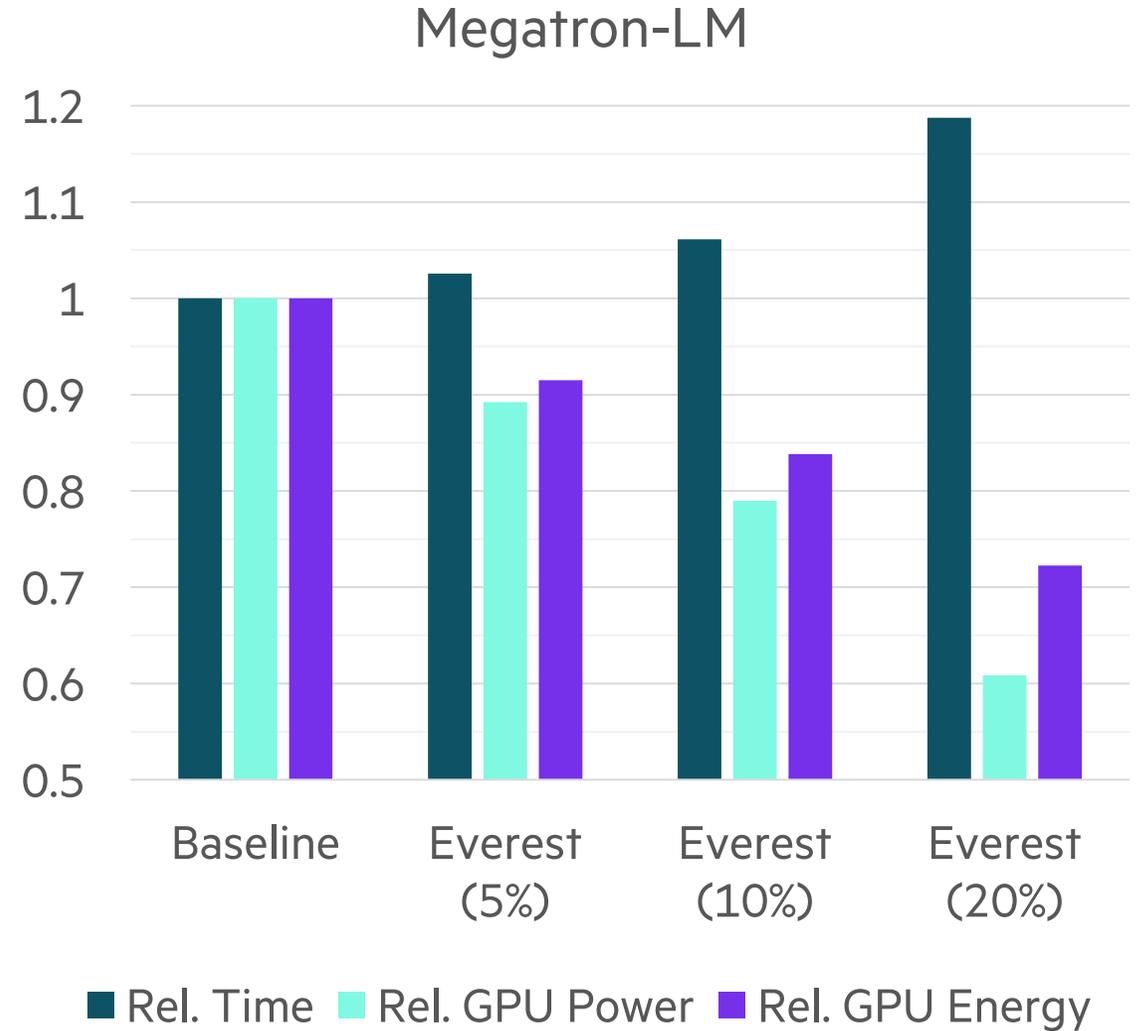
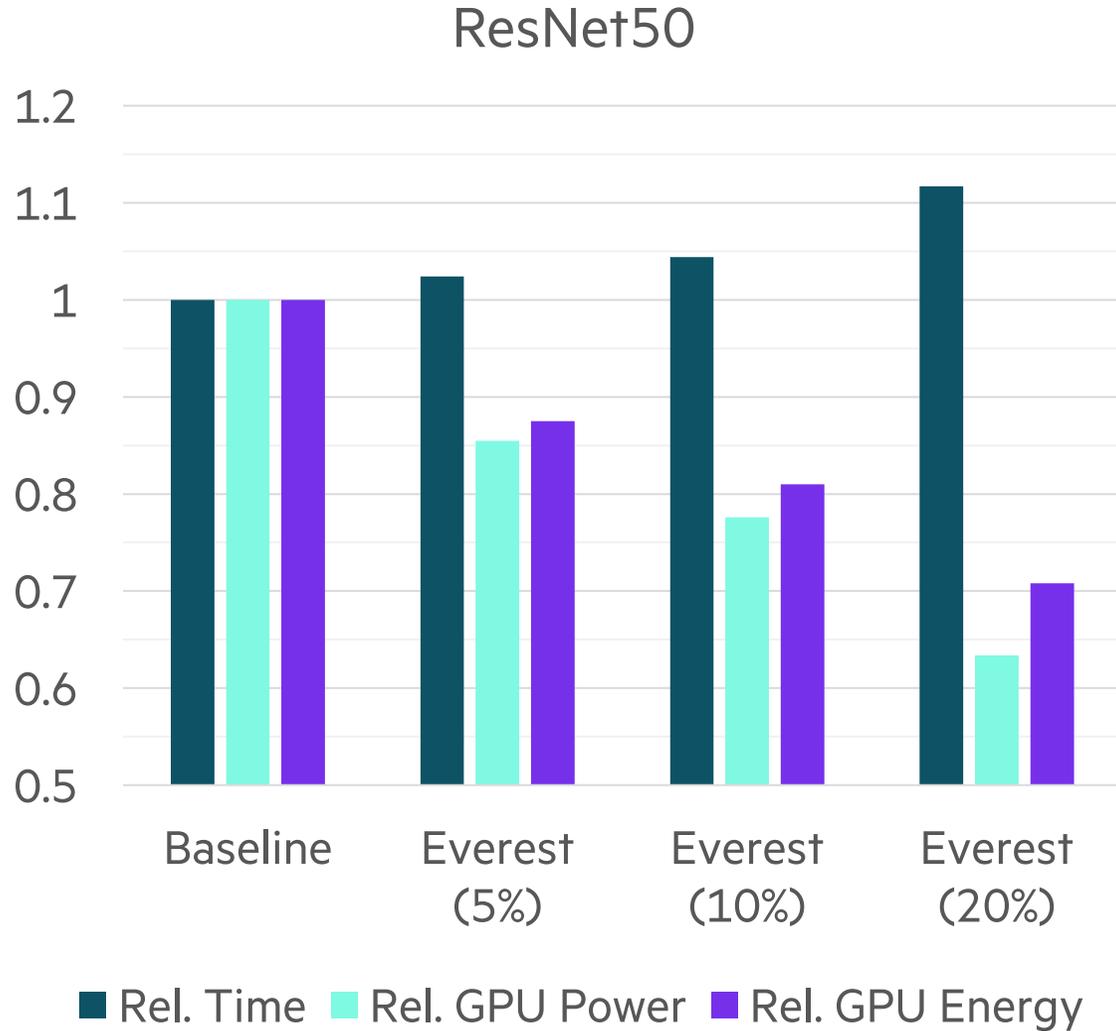
# GPU RESULTS

## GPU Summary (Geomean)



Everest can provide significant power and energy savings for GPUs.

# GPU HIGHLIGHTS - AI WORKLOADS



# CONCLUSION

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- Lightweight solution for dynamic optimization of application according to power/performance/energy tradeoffs
  - Huge opportunity with GPUs
- Compute vendor agnostic
  - Works for CPUs and GPUs of different vendors
- Portable
  - runtime-only, integration with user code not required
- Phase awareness
  - Can extract maximum power/energy savings without requiring user input
- Opportunity for collaboration and influencing product roadmap



**THANKS**

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