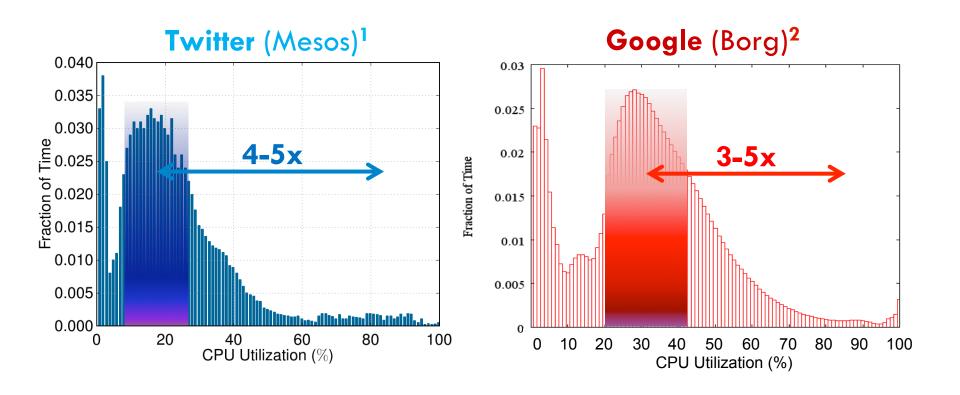
Leveraging Approximation to Improve Resource Efficiency in the Cloud

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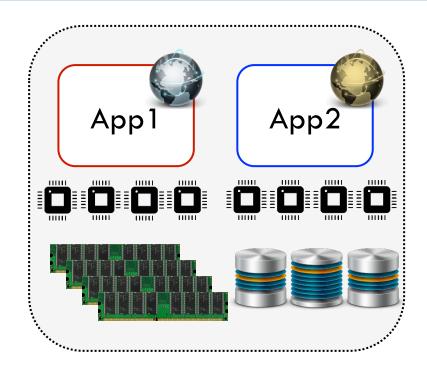
Datacenter Underutilization



¹ C. Delimitrou and C. Kozyrakis. Quasar: Resource-Efficient and QoS-Aware Cluster Management, ASPLOS 2014.

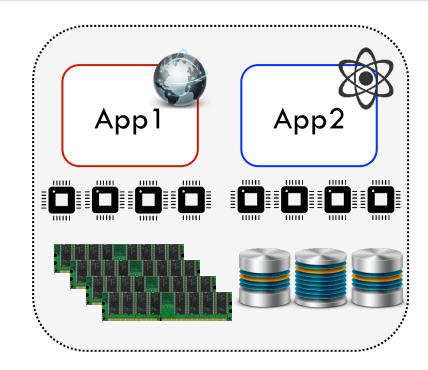
² L. A. Barroso, U. Holzle. The Datacenter as a Computer, 2013.

A Common Approach



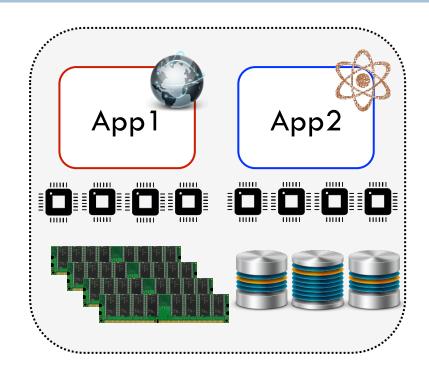
- Co-schedule multiple cloud services on same physical platform
- Often leads to resource interference, especially when sharing cores

A Common Cure



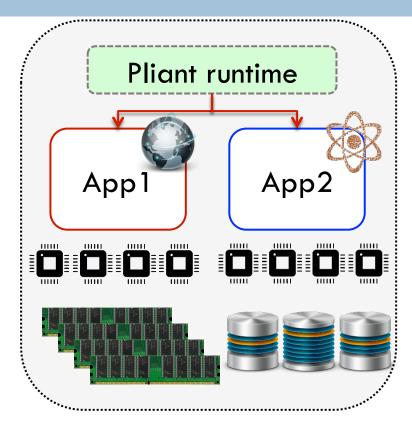
- Co-schedule one high priority and one/more best-effort apps
- Performance is non-critical for best effort jobs
- Disadvantage: assume best-effort apps are always low priority

Approximate Computing Apps to the Rescue



- Approximate computing apps can absorb a loss of resources as loss of output quality instead of a loss in performance
- Advantage: performance of all co-scheduled applications is highpriority

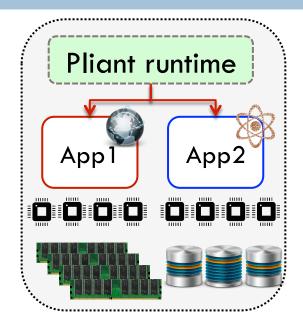
Pliant



- Enables latency-critical & approximate apps to share resources (including cores) without penalizing their performance
- Tunes degree and type of approximation based on measured interference

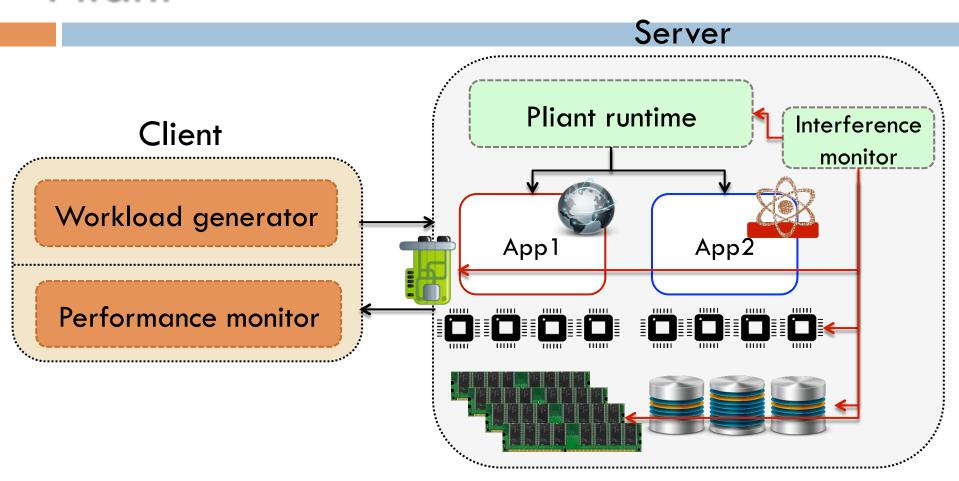
Challenges

- 1. Identify opportunities for approximation
 - ACCEPT (precision, loop perforation, sync elision), algorithmic exploration
- Lightweight profiling to determine when to employ approximation
 - End-to-end latency/throughput & perf counters



- 3. Determine what resource(s) to constrain?
 - Based on measured interference
- 4. Determine what type of approximation & to what extent?
 - Based on interference and performance impact

Pliant



DynamoRIO for switching between precise/approximate versions

- Initial implementation, overheads high but not prohibitive
- Looking into Petabricks and LLVM

Adaptive Approximation

Incremental approximation:

- Employ the minimum amount of approximation (quality loss) to restore the performance of the interactive service
- Several versions for each type of approximation, choose online

□ Interference-aware approximation:

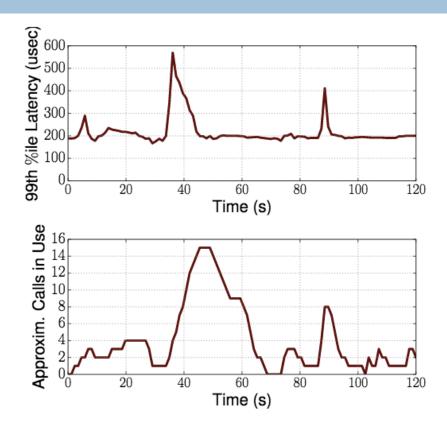
- Choose the type of interference that minimizes pressure in the bottlenecked resource
- Example:
 - High memory interference → prioritize algo tuning
 - High CPU interference → prioritize sync elision, loop perforation

Methodology

- Latency-critical interactive services: memcached & nginx
 - Open-loop workload generator & performance monitor
 - Facebook traffic pattern
- Approximate computing apps: PARSEC, SPLASH, Spark MLlib

System: 2 2-socket, 40-core servers, 128GB RAM each

Evaluation



- memcached sharing physical cores with PARSEC
- \Box Latency $\uparrow \rightarrow$ Degree of approximation \uparrow

Conclusions

- Approximate computing: opportunity to improve cloud efficiency without loss in performance
- Pliant: cloud runtime to co-schedule interactive services with approximate computing apps
 - Incremental and interference-aware approximation
 - Preserves QoS for interactive service with minimal loss in quality for approximate computing application
- Current work:
 - DynamoRIO → Petabricks/LLVM
 - Add cloud approximate computing application
 - Explore design space exploration & optimization problem
 - Leverage hardware isolation techniques

Questions?

- Approximate computing: opportunity to improve cloud efficiency without loss in performance
- Pliant: cloud runtime to co-schedule interactive services with approximate computing apps
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