# Prediction-Based Power Oversubscription in Cloud Platforms

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  - Massive underutilization due to provisioning peak power for each server

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Power capping impact on workload performance (baseline: un-capped performance)

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Oversubscription is currently limited by performance impact of capping

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Solution: Criticality-aware per-VM power capping and oversubscription

- Provide power safety while protecting performance of critical VMs
- Strategy for criticality-aware oversubscription

Γ		
	servers	
	Chassis Manager	

ML System (Resource Central [SOSP'17])

Machine Learning (ML) and prediction serving system. Add algorithms and models to predict VM criticality and resource demand (e.g., p95 CPU)

servers	
Chassis Manager	

ML System (Resource Central [SOSP'17])

> VM Scheduler (Protean [OSDI'20])

VM placement with rules to tightly pack VMs on servers. Add rules for distributing power via criticality and utilizationaware VM placement.

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Chassis power draw > limit

• Start power capping on servers

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Per-VM power management using ML models, enhanced VM placement and per-VM power capping can increase oversubscription by 2x



Critical

Non-critical

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  - 99% precision and recall for user-facing workloads
- Static overrides
  - "Always-throttle" list of VMs (e.g., internal non-production)
  - "Do-not-throttle" list of VMs (e.g., all third-party, gaming)



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- Per-VM enables additional harvesting while protecting perf of critical VMs
  - Trade-off: Increased perf degradation for non-critical VMs
  - Relaxed perf requirement of workloads on non-critical VMs (e.g., internal non-production)

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  - **Constraints**: # capping events and perf (frequency) reduction for critical and non-critical VMs

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Approach	Harvested power (%)	Savings (\$10/W)
Traditional (no oversubscription)	0	0
State-of-the-art (w/ full-server capping)	6.2%	\$79.4M
Predictions for internal and non-premium external VMs	12.1%	\$154.9M

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Per-VM capping allow us to be selective and increase the amount of **oversubscription by 2x!** 

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  - 3. Server support for per-VM capping

#### Conclusions

- Limited power oversubscription on cloud platforms to restrict performance impact
- Solution: Prediction-based per-VM power capping
  - Algorithm and ML models for predicting performance criticality and VM utilization
  - Criticality- and utilization-aware VM placement
  - On-server criticality-aware power management system
  - Strategy for criticality-aware oversubscription
- Main result: Increase oversubscription by 2x while protecting critical workloads

# Thank you!