

# Resource Central

Cortez et al., 2017

# Motivation

**Improve resource management in large cloud platforms.**

🤔 But what exactly does this mean?

- Balancing of disk IOPS loads
- Preventing physical resource exhaustion in oversubscribed servers
- Reducing cluster migration
- Prioritizing interactive workloads (over delay-insensitive ones)
- etc.

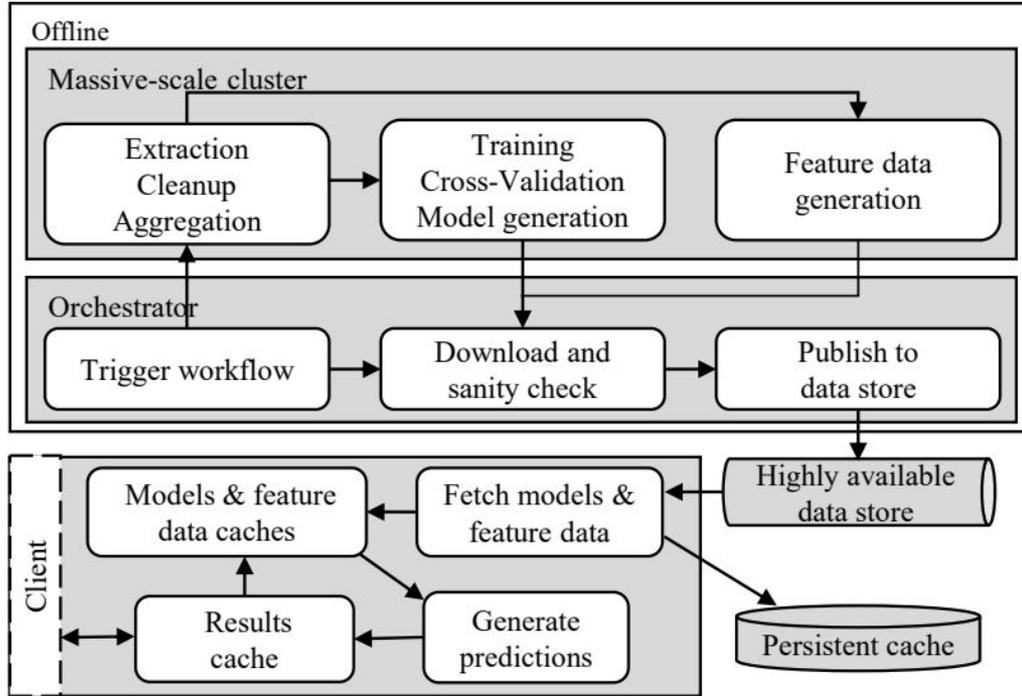
# Key Insights

1. If we **knew** certain characteristics of VMs, we could improve resource management in cloud platforms.
  - a. Average CPU utilization
  - b. P95 max CPU utilization
  - c. Deployment size in # of VMs
  - d. Deployment size in # of cores
  - e. Lifetime
  - f. Workload class
2. Empirical evidence shows that certain VM characteristics are fairly consistent over multiple lifetimes (so they can be **predicted**).
  - a. These characteristics are especially consistent on a per-subscription basis

 What other VM metrics might be useful to know?

 Are there any VM metrics that *can't* be predicted?

# Architecture



**Figure 9: RC architecture (pull version).**

*Discussion:* Consider the offline-online split pictured in this system diagram. What aspects of it do you agree or disagree with? Why?

*Discussion:* Would you add or remove any components? Why or why not?

# Modeling

| Metrics                   | Approach                            | #features | Model size | Feature data size |
|---------------------------|-------------------------------------|-----------|------------|-------------------|
| Avg CPU utilization       | Random Forest                       | 127       | 312 KB     | 376 MB            |
| P95 max CPU utilization   | Random Forest                       | 127       | 311 KB     | 376 MB            |
| Deployment size in #VMs   | Extreme Gradient Boosting Tree      | 24        | 305 KB     | 368 MB            |
| Deployment size in #cores | Extreme Gradient Boosting Tree      | 24        | 305 KB     | 368 MB            |
| Lifetime                  | Extreme Gradient Boosting Tree      | 127       | 329 KB     | 376 MB            |
| Workload class            | FFT, Extreme Gradient Boosting Tree | 34        | 152 KB     | 311 MB            |

**Table 1: Metrics, ML modeling approaches, model and full feature dataset sizes.**

*Discussion:* What are some strengths and weaknesses of their modeling approach? Aspects to consider include fidelity, performance, and ability to handle outliers.

*Discussion:* How exactly can the prediction results inform scheduling systems to improve resource utilization?

# Novelties & Strengths

- Performed a careful analysis of production VM workloads of a real cloud provider (i.e., Azure) and made this data public.
- Provided convincing evidence that certain VM characteristics can be predicted with high fidelity.
- Created a performant and robust prediction system (i.e., Resource Central).

🤔 What figure did you find most illuminating and why?

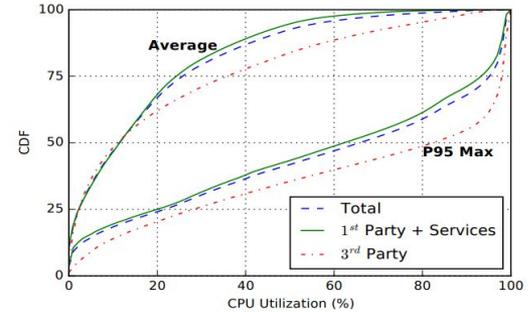


Figure 1: Average and P95 of max CPU utilizations.

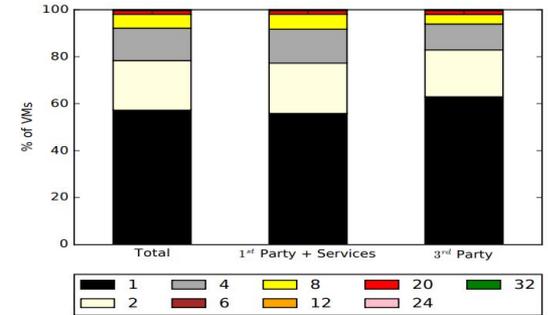


Figure 2: Number of virtual CPU cores per VM.

# Critique & Weaknesses

- IOPS analysis should differentiate between reads and writes, as the former is far cheaper than the latter.
- No graph or figure for the evaluation of how RC-informed scheduling (§6.2) can improve CPU utilization over baseline. Also, the cost of having >100% utilization is unclear, so it is difficult to evaluate the scheduling policies.
- Too much discussion of result caching, which is neither novel nor interesting.
- Interarrival times are remarked on and said to be easily modeled (§3.7), but are not used by Resource Central for predictions, even for VM Lifetime.

*Discussion:* What other deficiencies are present in the approach of the paper?  
What about in the way it was presented?

Thanks for your  
participation!