

Bolt: I Know What You Did Last Summer... In the Cloud



Christina Delimitrou¹ and Christos Kozyrakis²

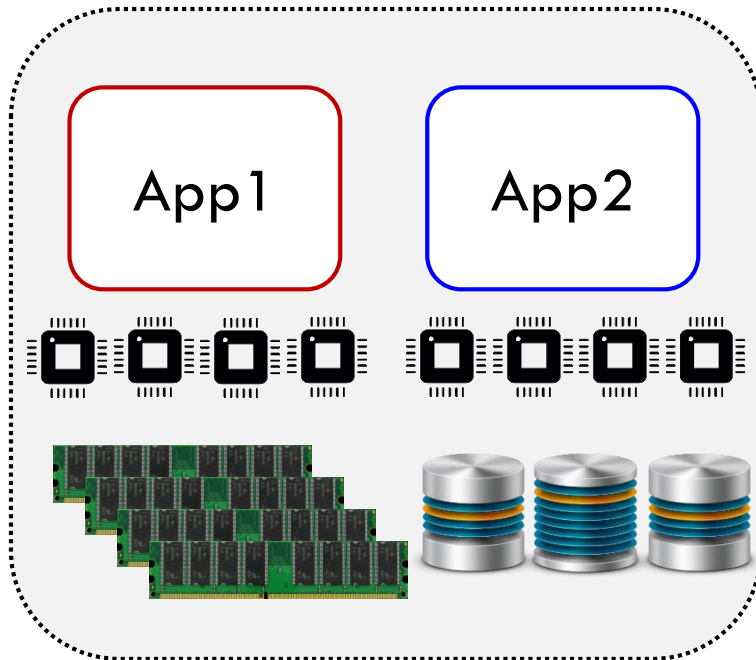
¹Cornell University, ²Stanford University

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Executive Summary

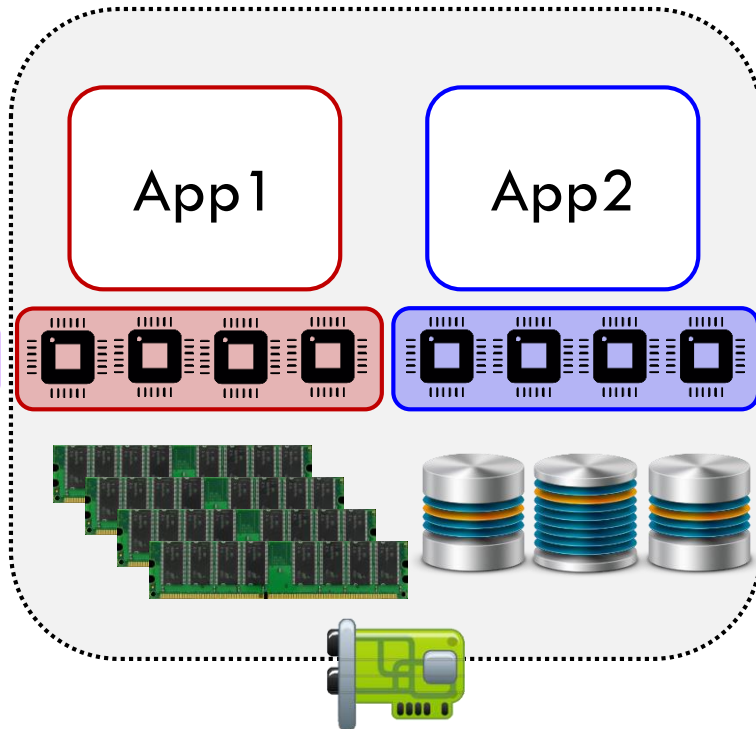
- Problem: **cloud resource sharing hides security vulnerabilities**
 - Interference from co-scheduled apps → **leaks app characteristics**
 - **Enables severe performance attacks**
- Bolt: adversarial runtime in public clouds
 - Transparent app detection (5-10sec)
 - Leverages **practical machine learning** techniques
 - DoS → **140x increase in latency**
 - **User study**: 88% correctly identified applications
 - Resource **partitioning** is **helpful** but **insufficient**

Motivation

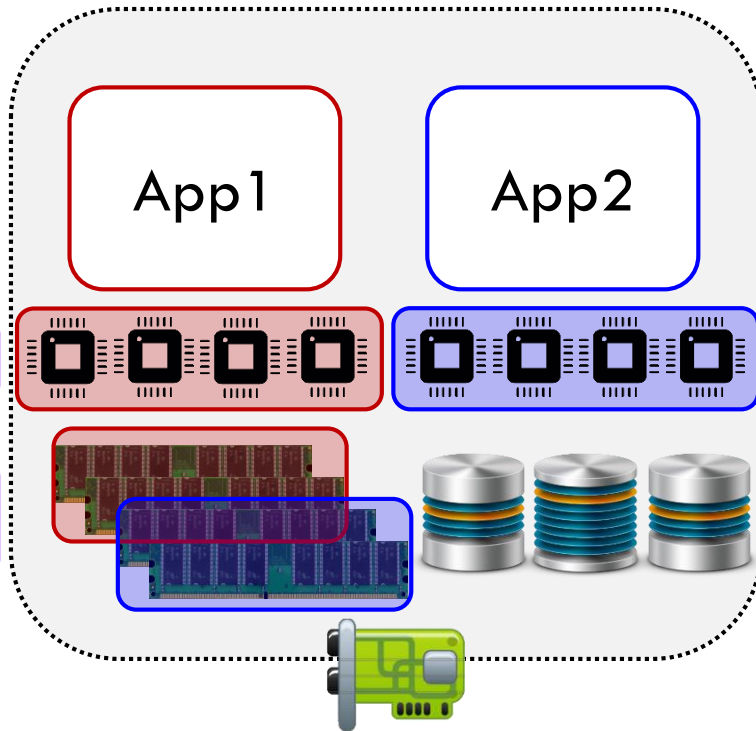


Google Cloud Platform

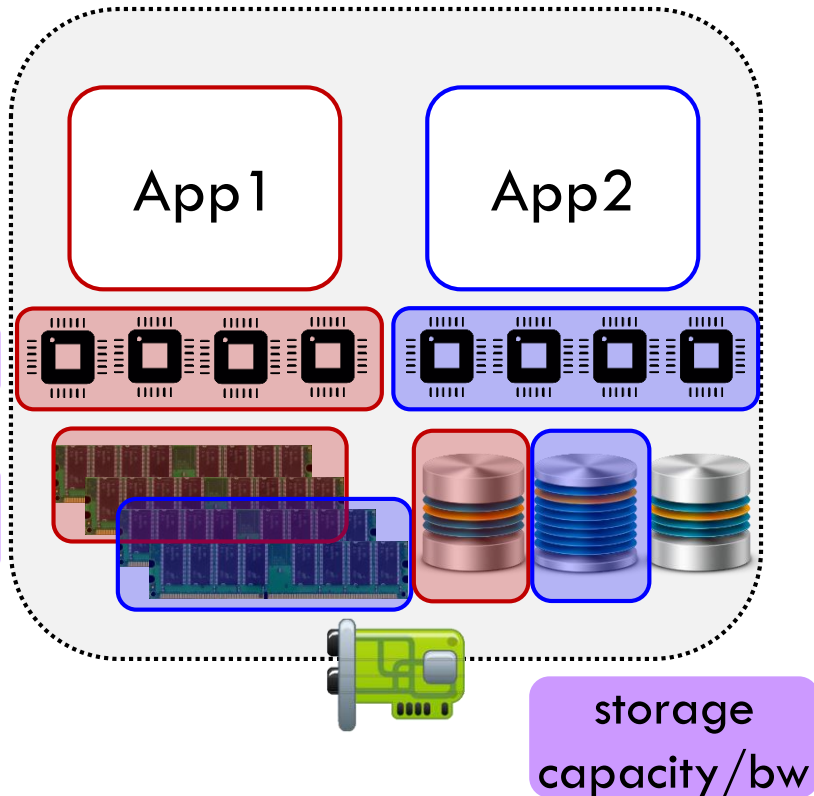
Motivation



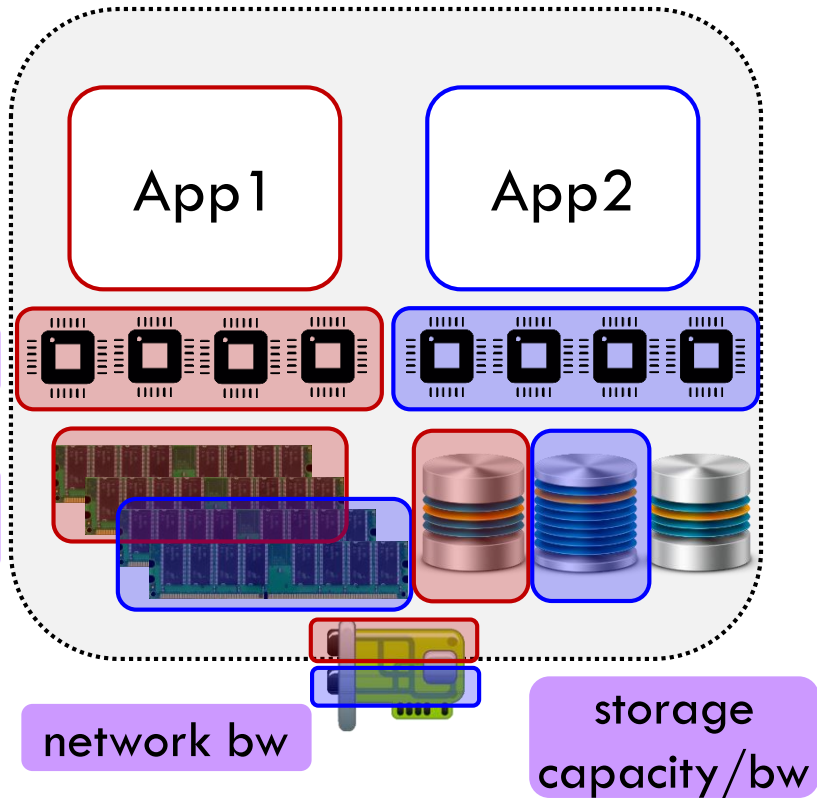
Motivation



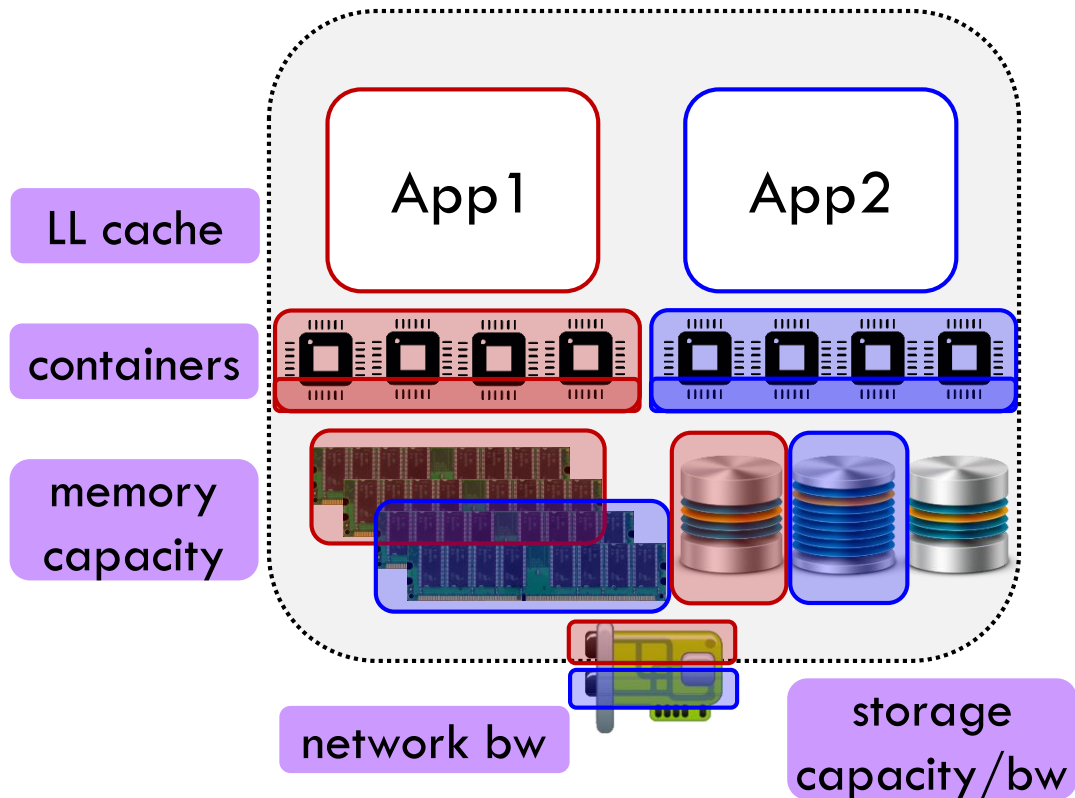
Motivation



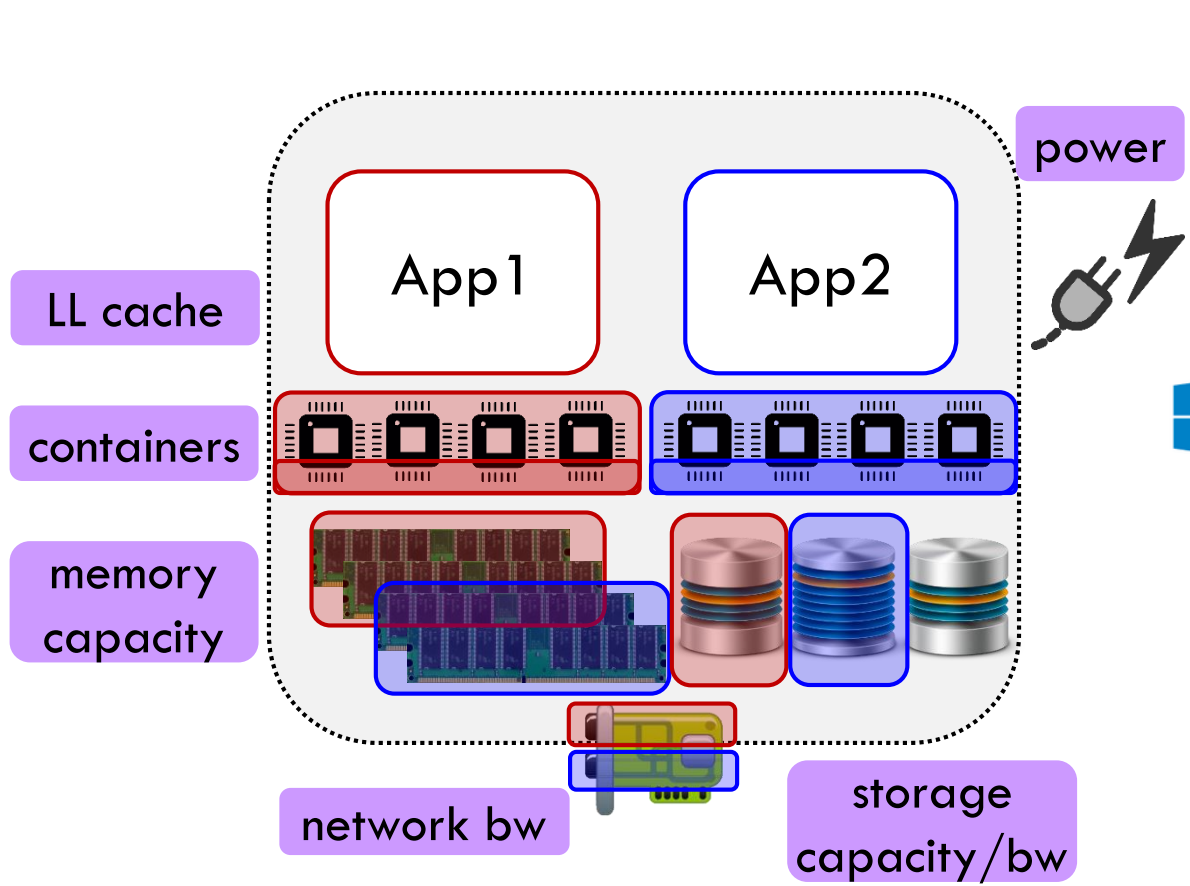
Motivation



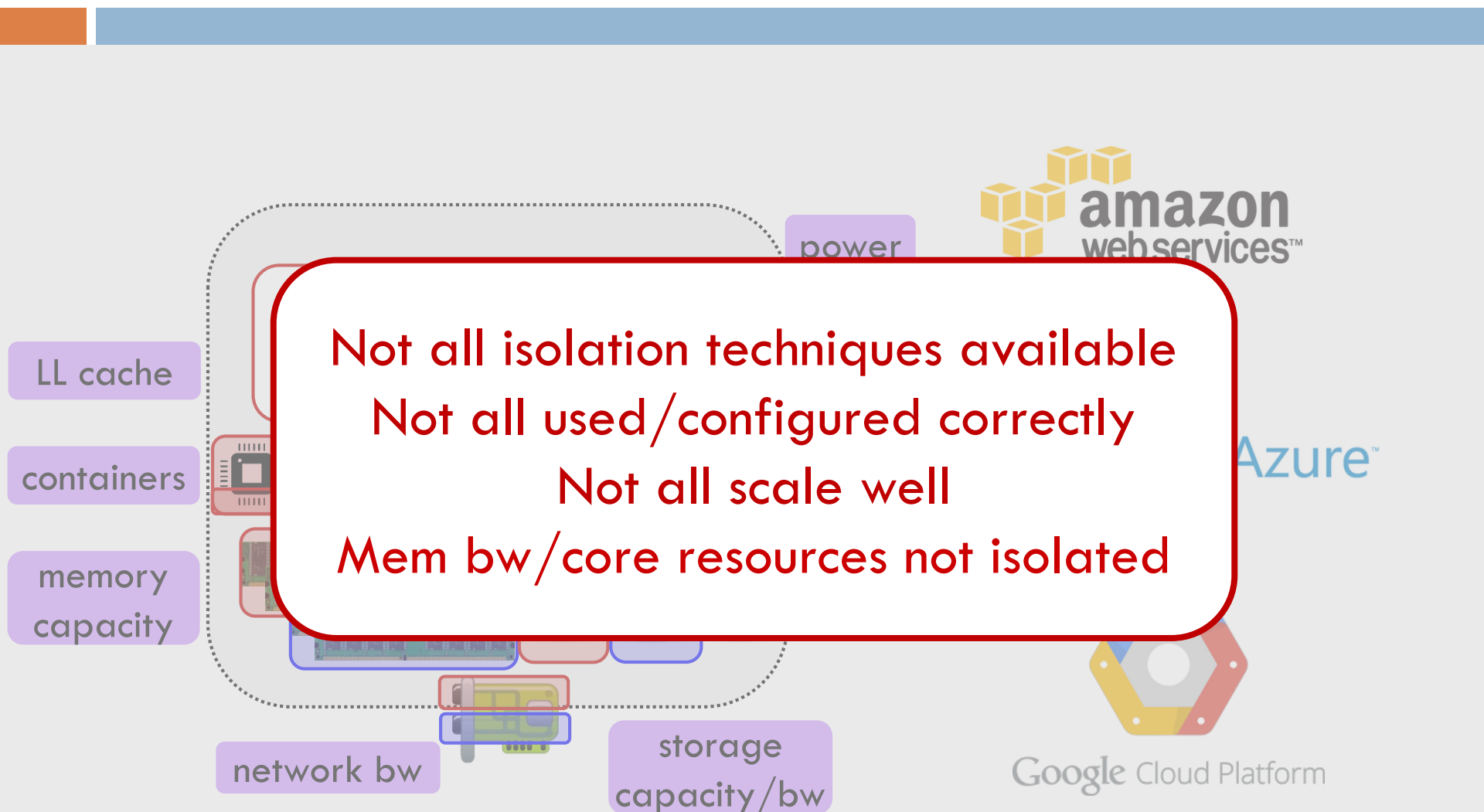
Motivation



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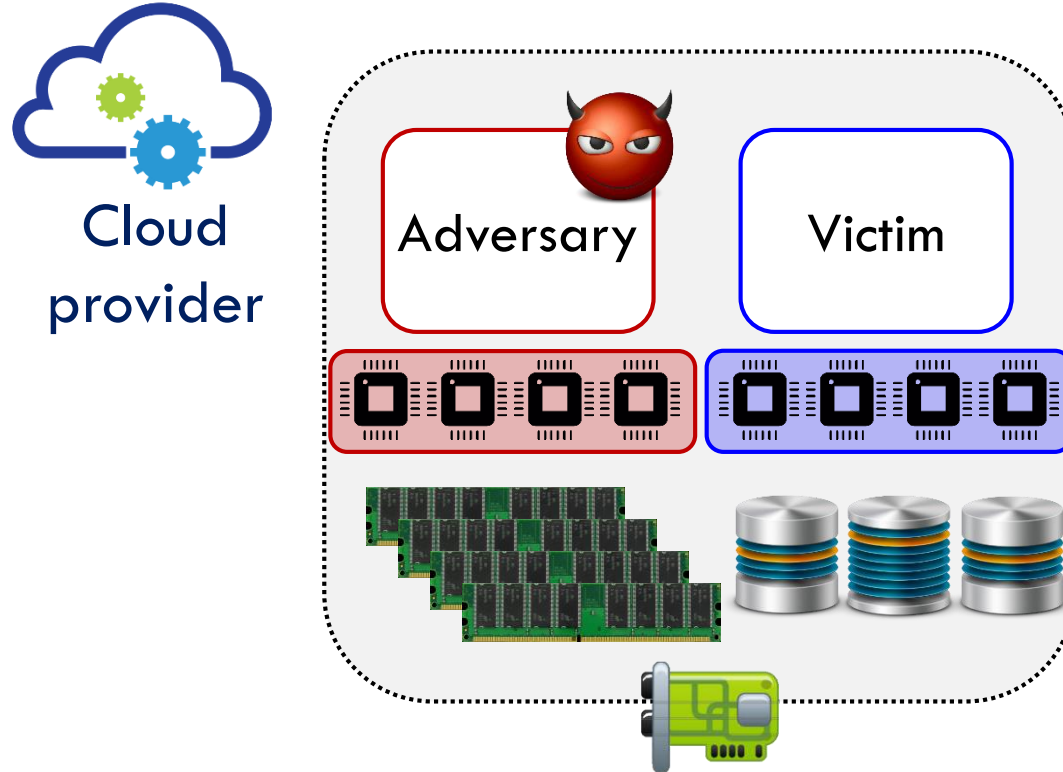
Motivation



Bolt

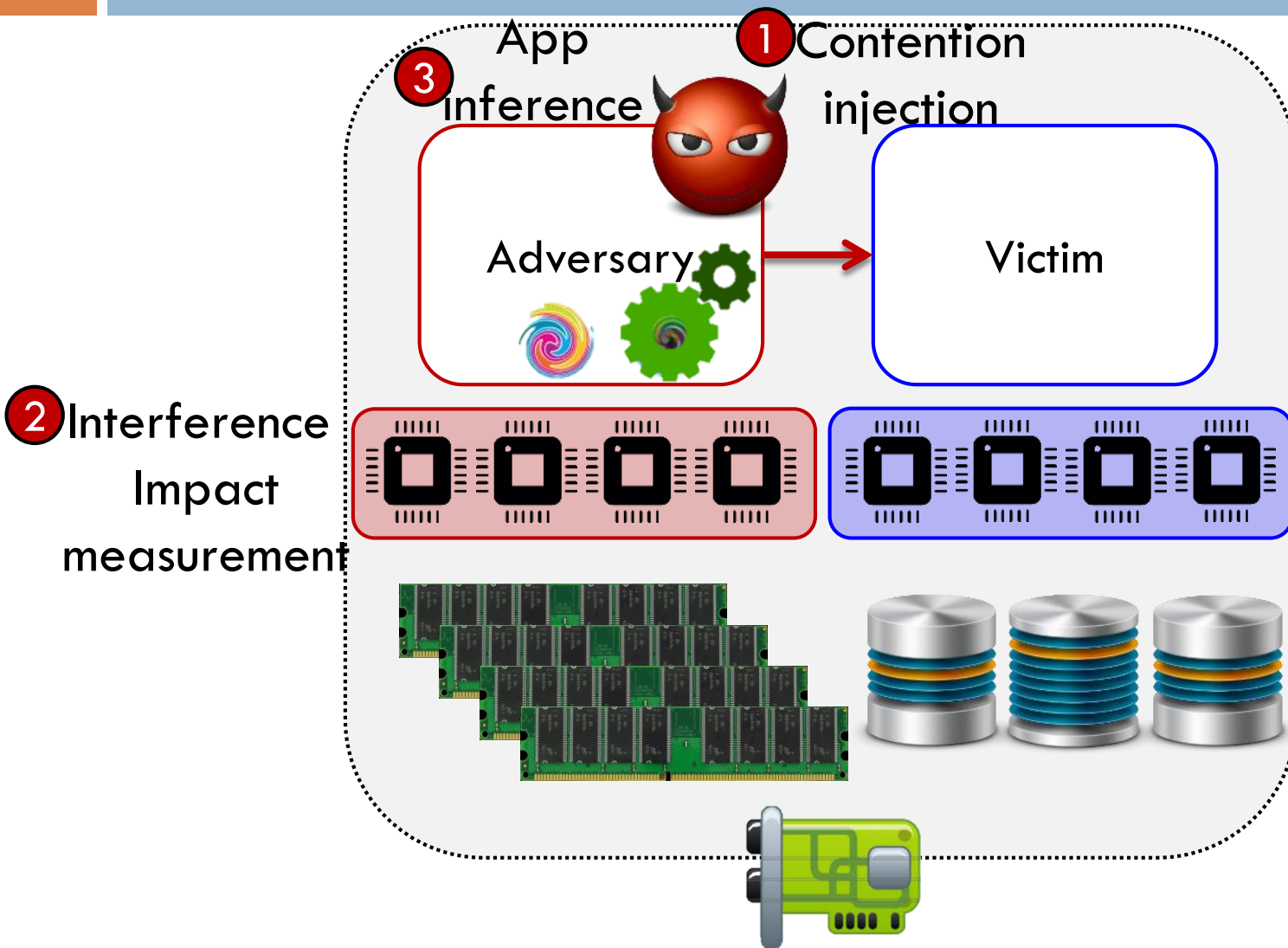
- **Key idea:** Leverage lack of isolation in public clouds to infer application characteristics
 - Programming framework, algorithm, load characteristics
- **Exploit:** enable practical, effective, and hard-to-detect performance attacks
 - DoS, RFA, VM pinpointing
 - Use app characteristics (sensitive resource) against it
 - Avoid CPU saturation → hard to detect

Threat Model

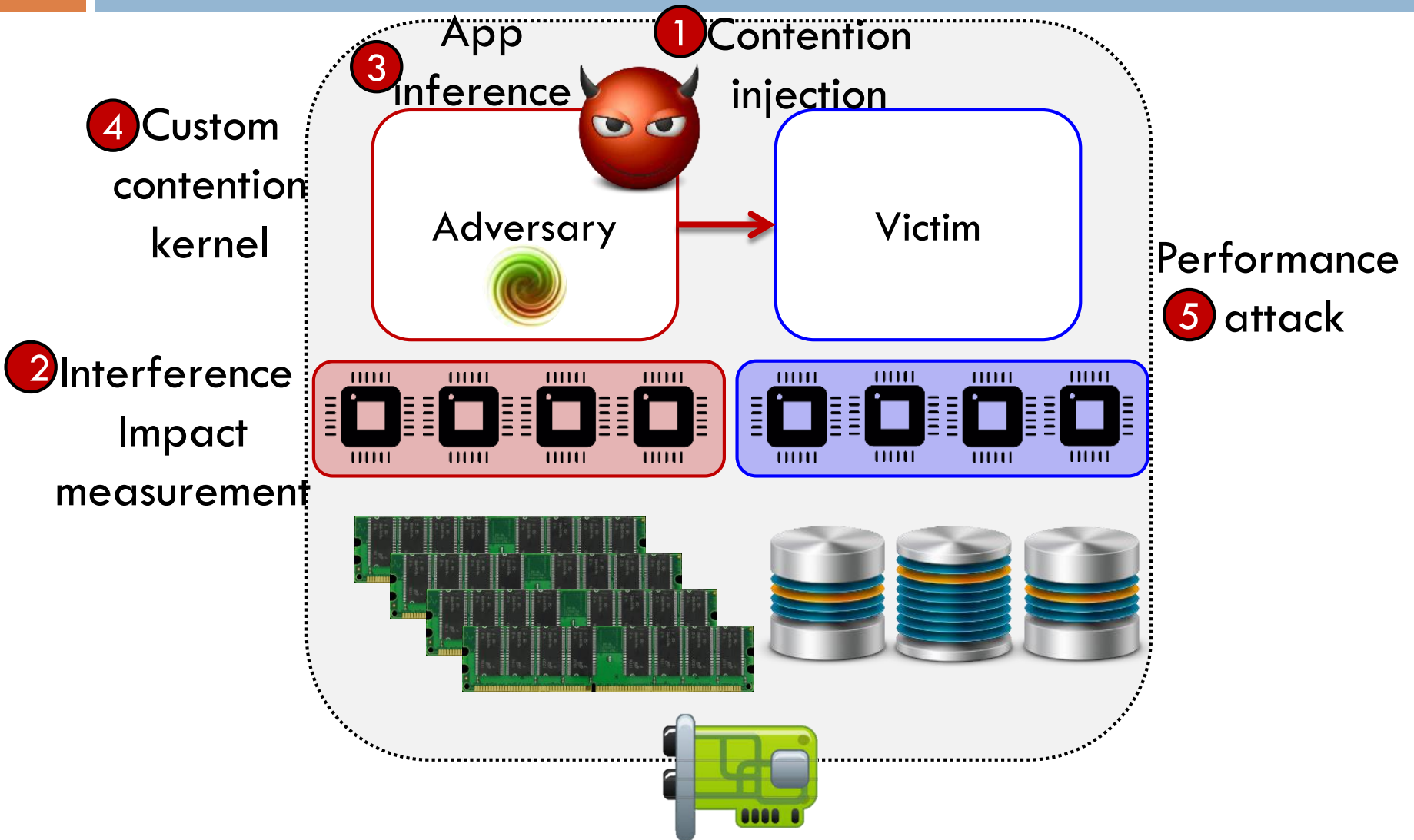


- ❑ Impartial, neutral cloud provider
- ❑ Active adversary but no control over VM placement

Bolt

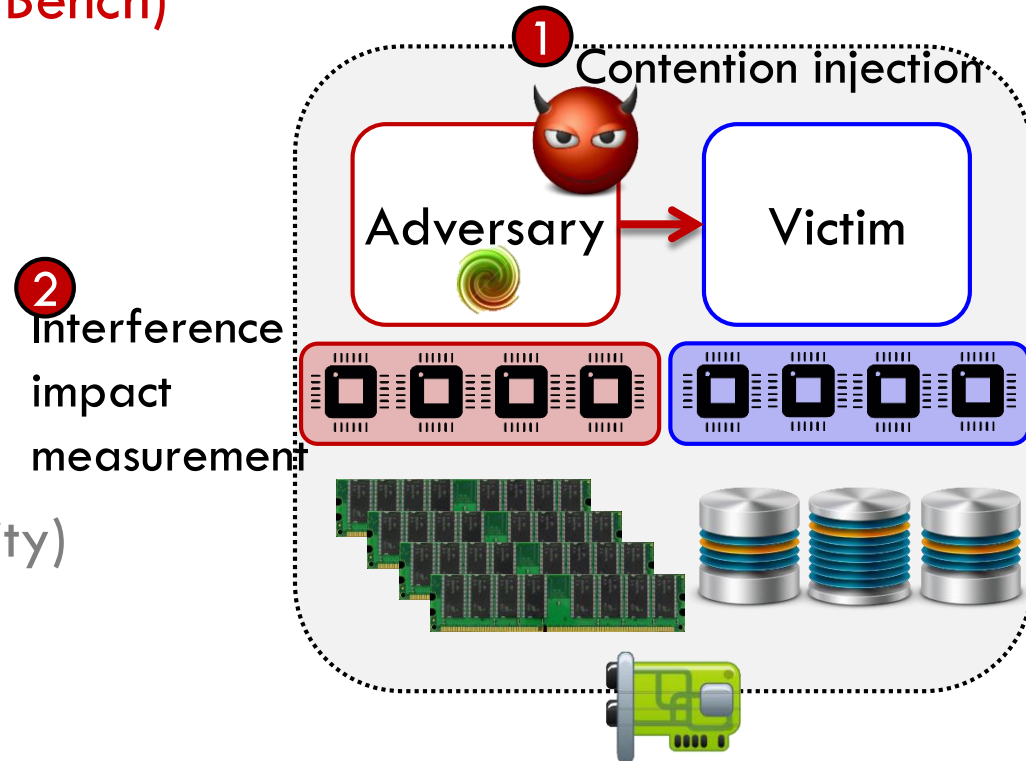


Bolt



1. Contention Measurement

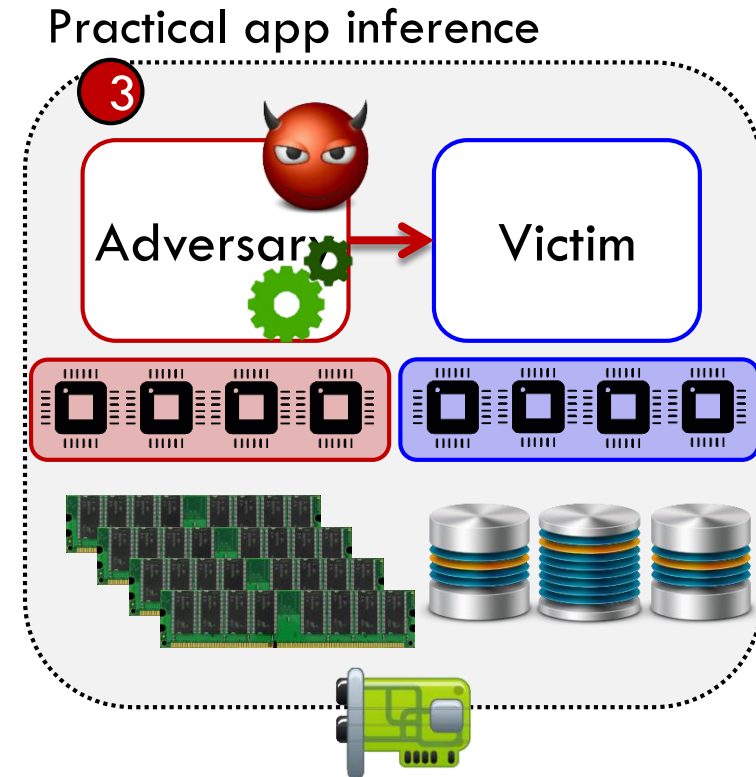
- Set of contentious kernels (iBench)
 - Compute
 - L1/L2/L3
 - Memory bw
 - Storage bw
 - Network bw
 - (Memory/Storage capacity)
- Sample 2-3 kernels, run in adversarial VM
- Measure impact on performance of kernels vs. isolation



2. Practical App Inference

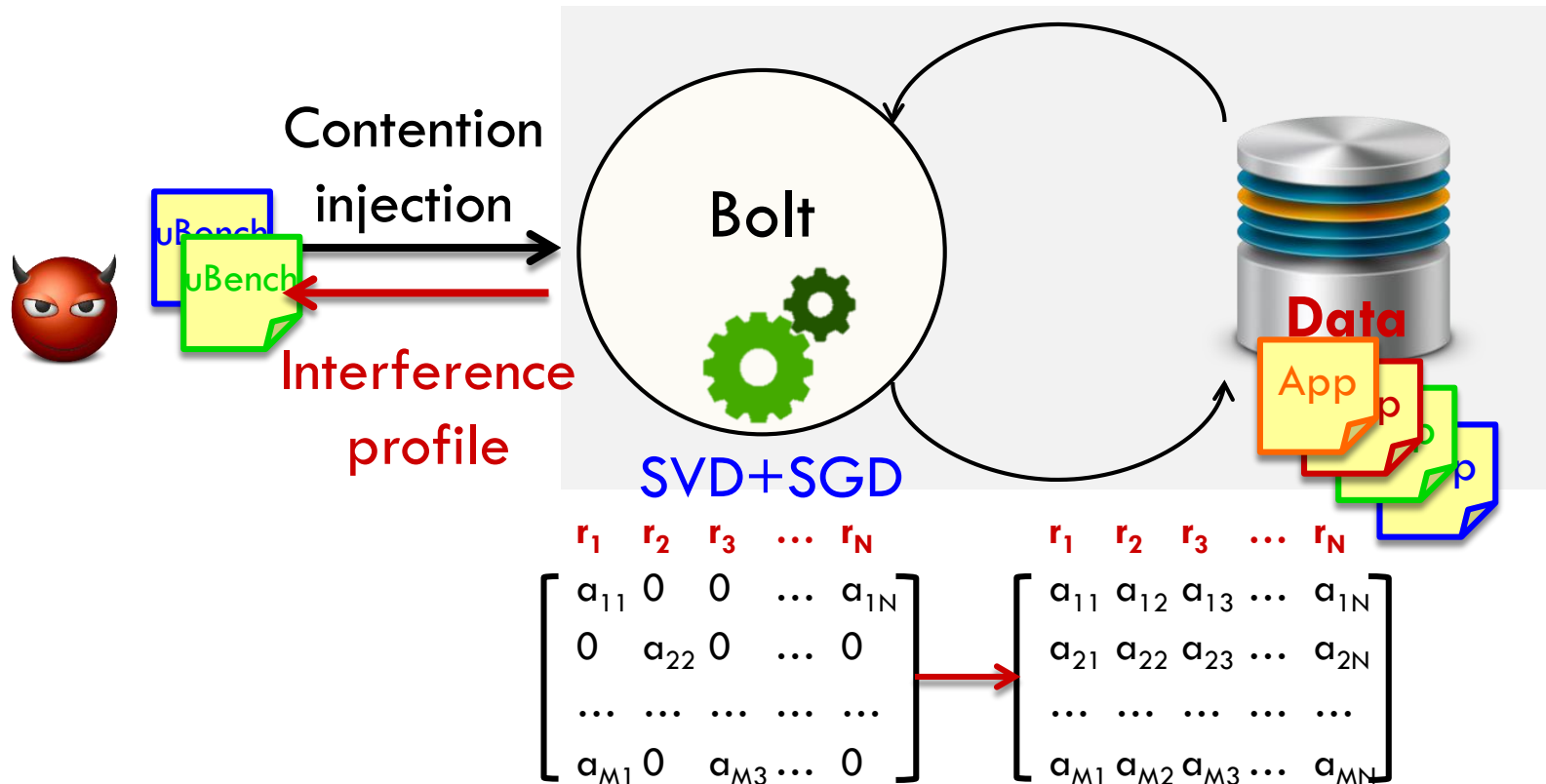
- Infer resource pressure in non-profiled resources
 - ▣ Sparse → dense information
 - ▣ SGD (Collaborative filtering)
- + □ Classify unknown victim based on previously-seen applications
 - ▣ Label & determine resource sensitivity
 - ▣ Content-based recommendation

Hybrid recommender



Big Data to the Rescue

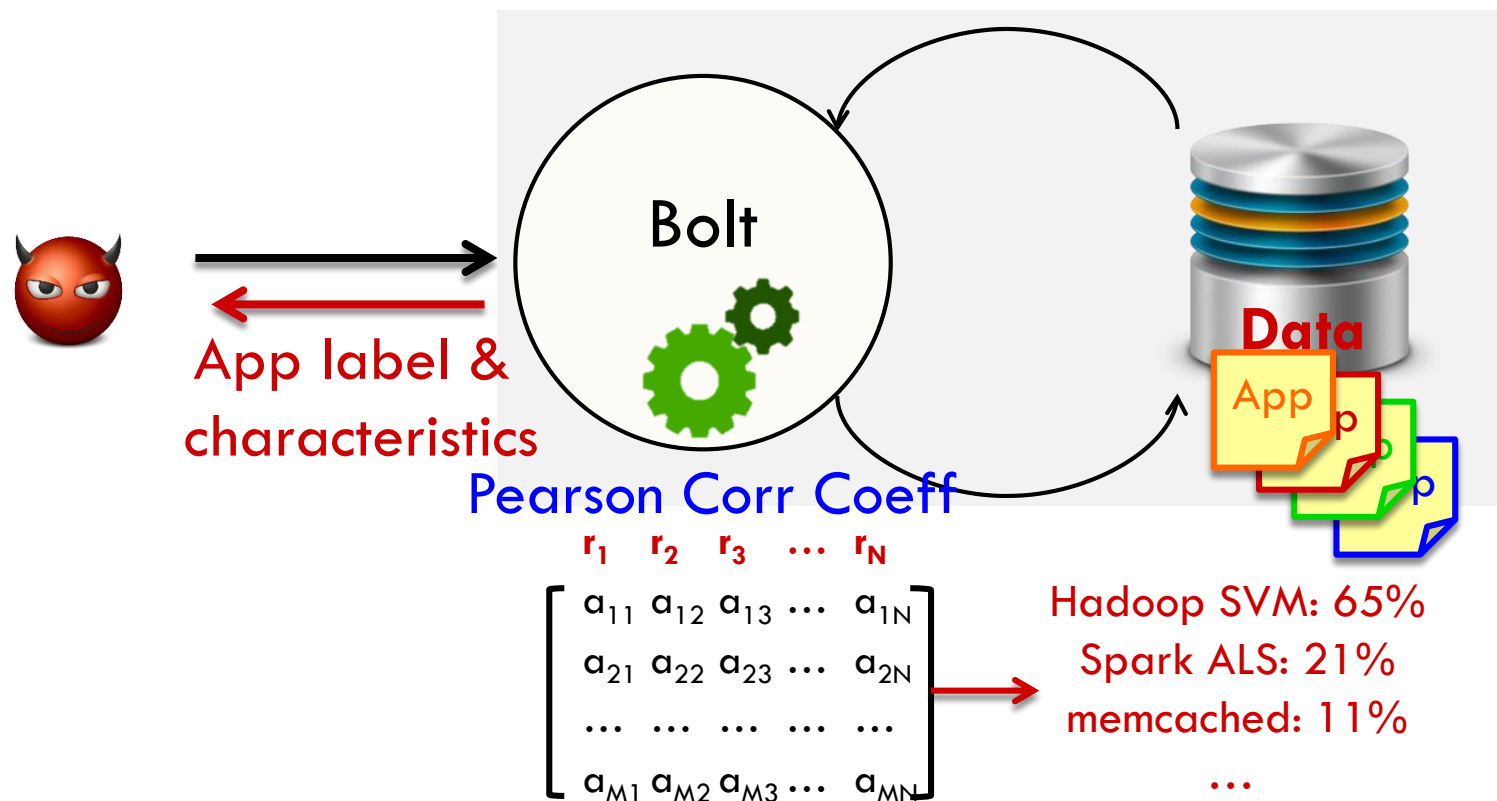
- Infer pressure in non-profiled resources
 - Reconstruct sparse information
 - Stochastic Gradient Descent (SGD), $O(\text{mpk})$



Big Data to the Rescue

2. Classify and label victims

- Weighted Pearson Correlation Coefficients
- Output:** distribution of similarity scores to app classes



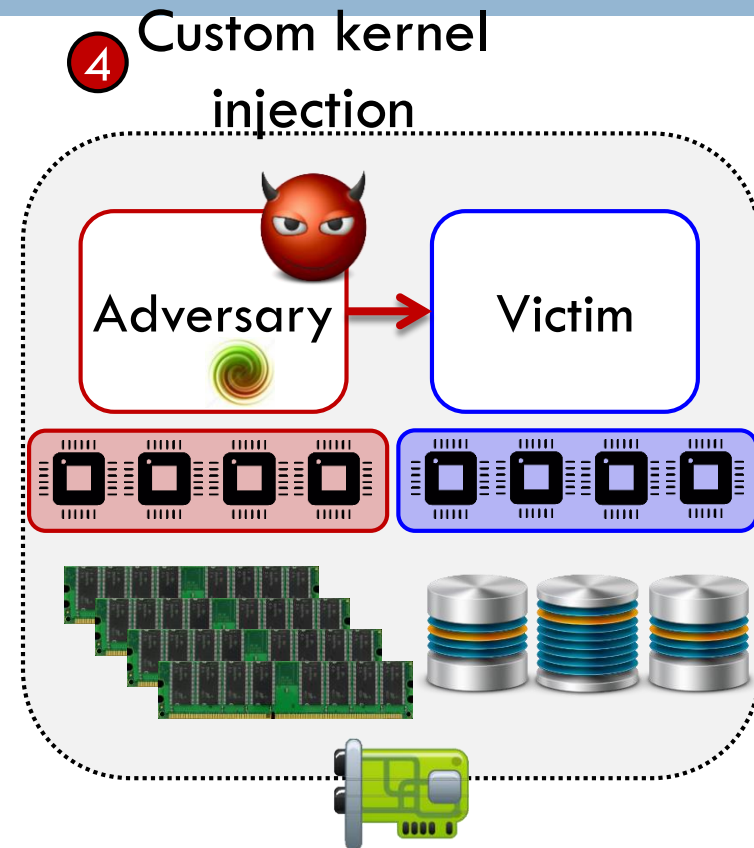
Inference Accuracy

- 40 machine cluster (420 cores)
- **Training apps**: 120 jobs (analytics, databases, webserver, in-memory caching, scientific, js) → **high coverage of resource space**
- **Testing apps**: 108 latency-critical webapps, analytics
- **No overlap in algorithms/datasets between training and testing sets**

| Application class | Detection accuracy (%) |
|---|------------------------|
| In-memory caching (memcached) | 80% |
| Persistent databases (Cassandra, MongoDB) | 89% |
| Hadoop jobs | 92% |
| Spark jobs | 86% |
| Webserver | 91% |
| Aggregate | 89% |

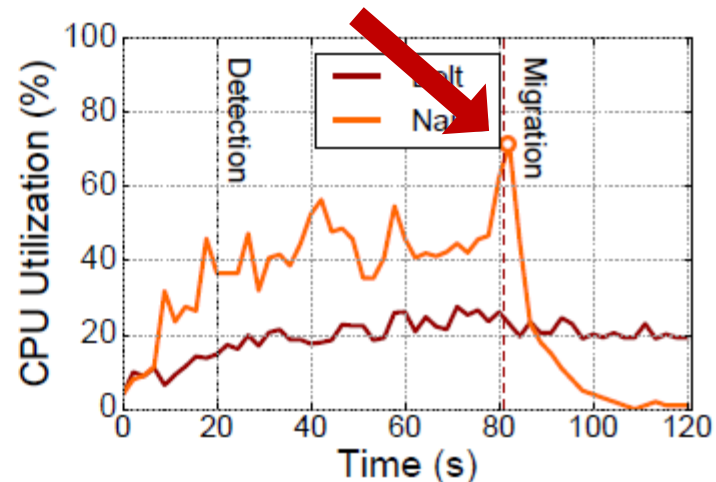
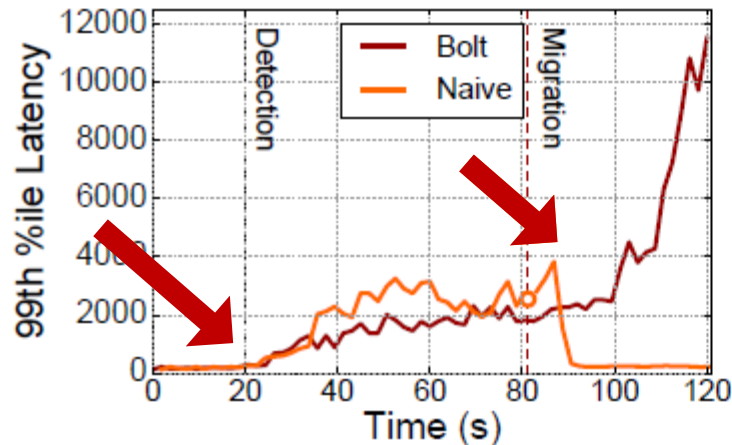
3. Practical Performance Attacks

1. Determine the resource bottleneck of the victim
2. Create custom contentious kernel that targets critical resource(s)
3. Inject kernel in Bolt
 - Several performance attacks (DoS, RFAs, VM pinpointing)
 - Target specific, critical resource → low CPU pressure



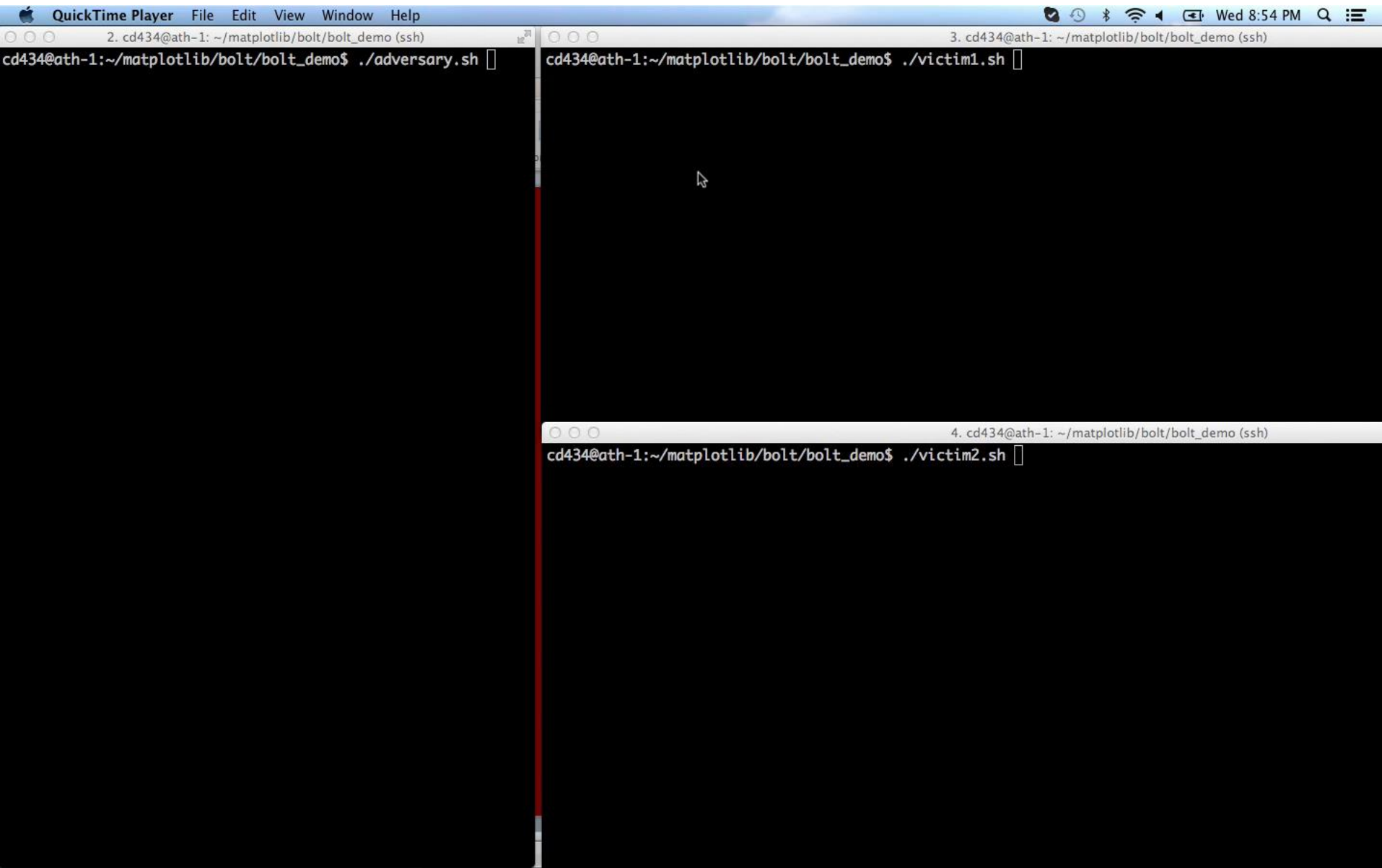
3. Practical DoS Attacks

- Launched against same 108 applications as before
- On average **2.2x** higher **execution time** and up to **9.8x**
- For interactive services, on average **42x** increase in **tail latency** and up to **140x**



- Bolt does not saturate CPU
- Naïve attacker gets migrated

Demo

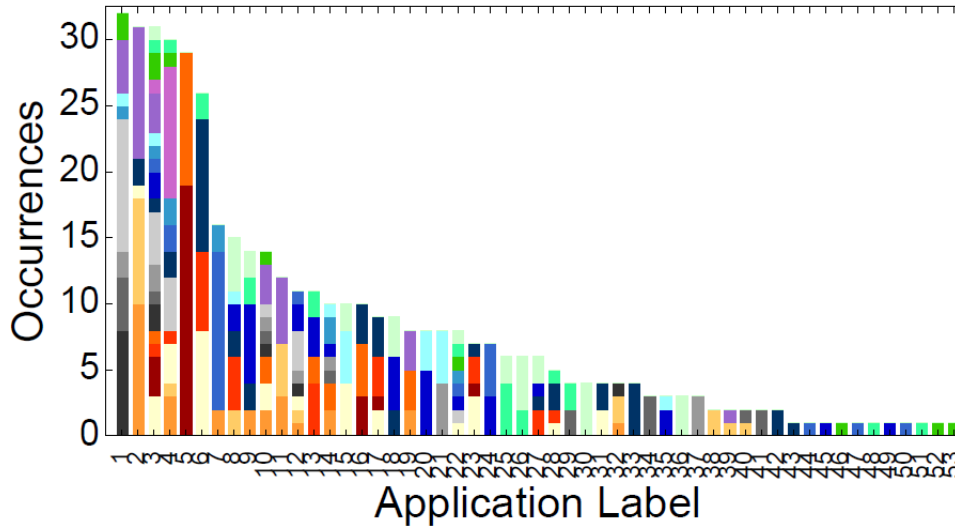


User Study

- 20 independent users from Stanford and Cornell
- Cluster
 - ▣ 200 EC2 servers, c3.8xlarge (32vCPUs, 60GB memory)
- Rules:
 - ▣ 4vCPUs per machine for Bolt
 - ▣ All users have equal priority
 - ▣ Users use thread pinning
 - ▣ Users can select specific instances
 - ▣ **Training set:** 120 apps incl. analytics, webapps, scientific, etc.

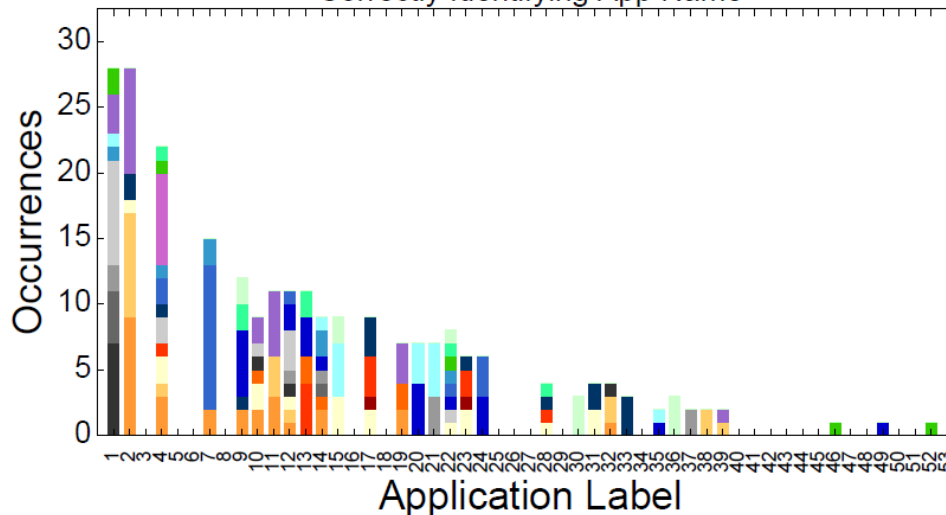
Accuracy of App Labeling

Ground Truth



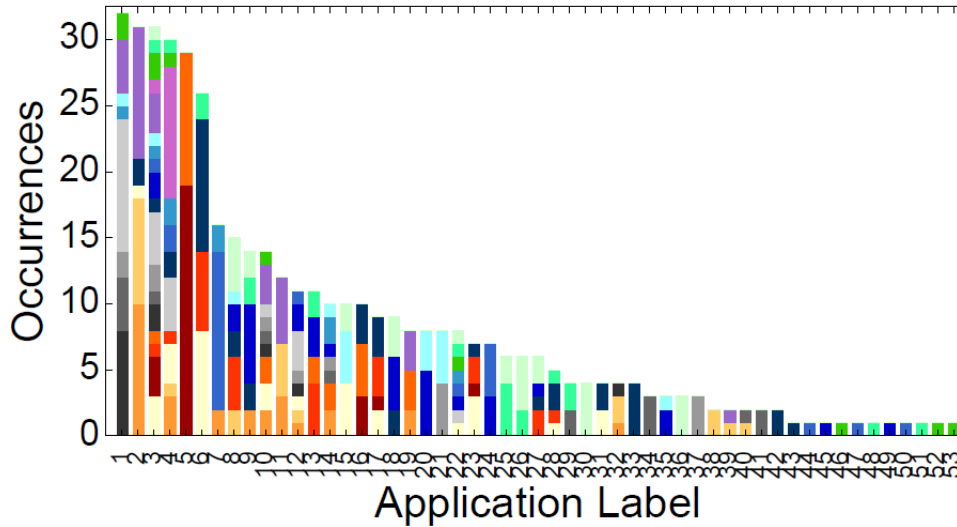
53 app classes
(analytics, webapps,
FS/OS, HLS/sim,
other...)

**Correct
app labels
63%**



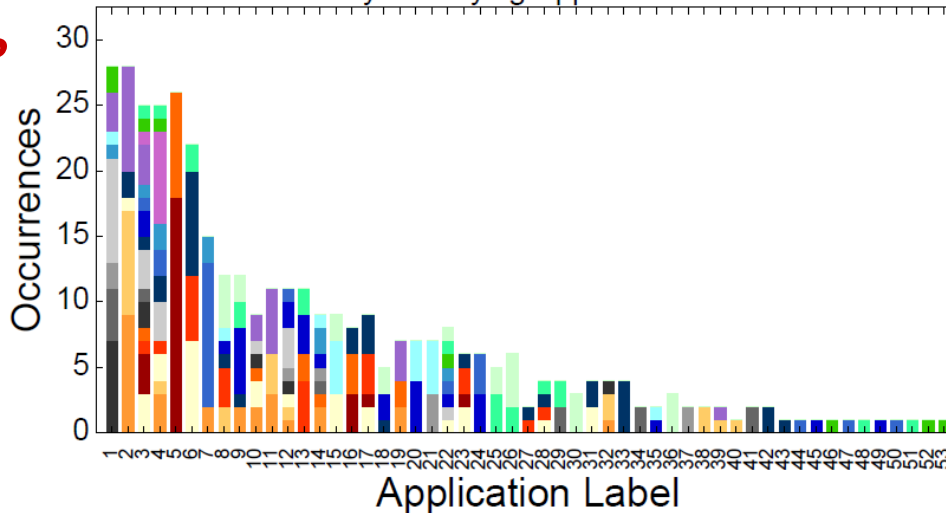
Accuracy of App Characterization

Ground Truth



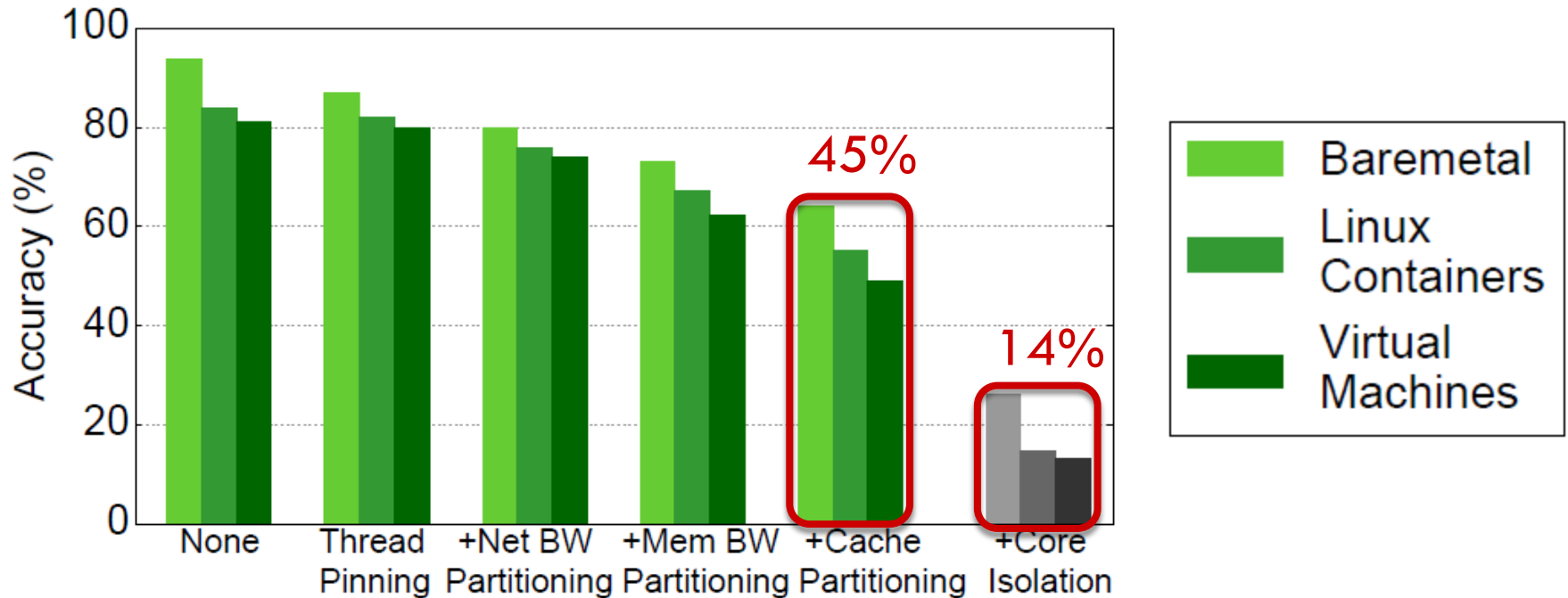
Correctly Identifying App Characteristics

**Correct
app characteristics
88%**



**Performance
attack results
in the paper**

The Value of Isolation



- Need more scalable, fine-grain, and complete isolation techniques

Conclusions

- **Bolt: highlight the security vulnerabilities from lack of isolation**
 - ▣ Fast detection using online data mining techniques
 - ▣ Practical, hard-to-detect performance attacks
 - ▣ Current isolation helpful but insufficient

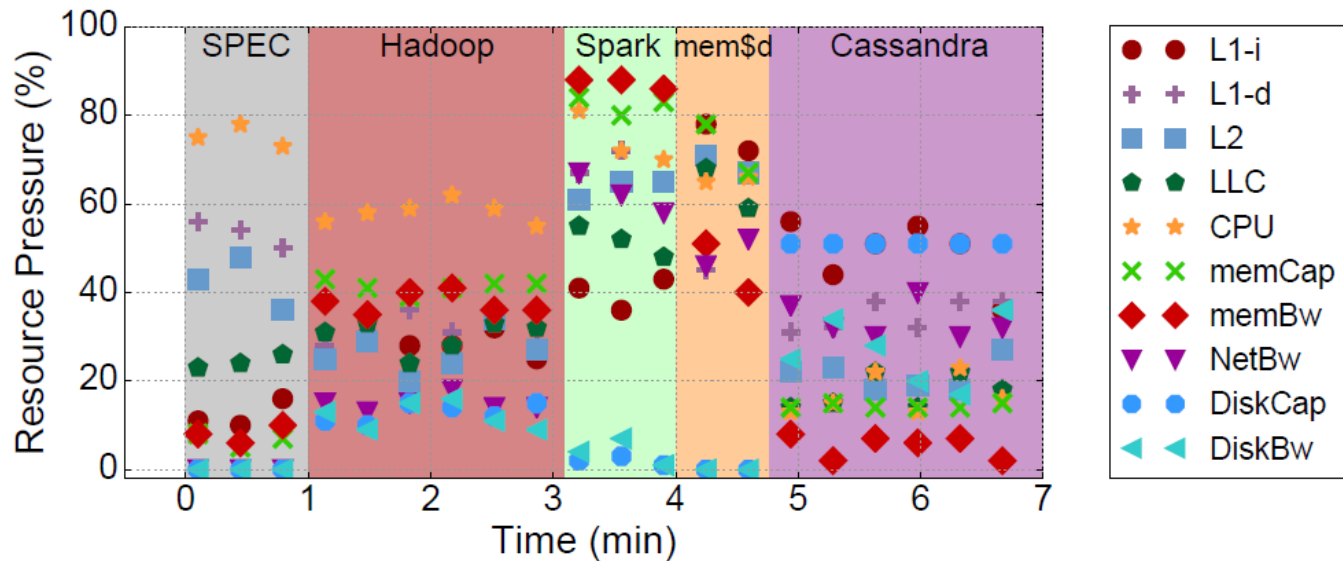
- **In the paper:**
 - ▣ Sensitivity to Bolt parameters
 - ▣ Sensitivity to applications and platform parameters
 - ▣ User study details
 - ▣ More performance attacks (resource freeing, VM pinpointing)

Questions?

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 - ▣ Fast detection using online data mining techniques
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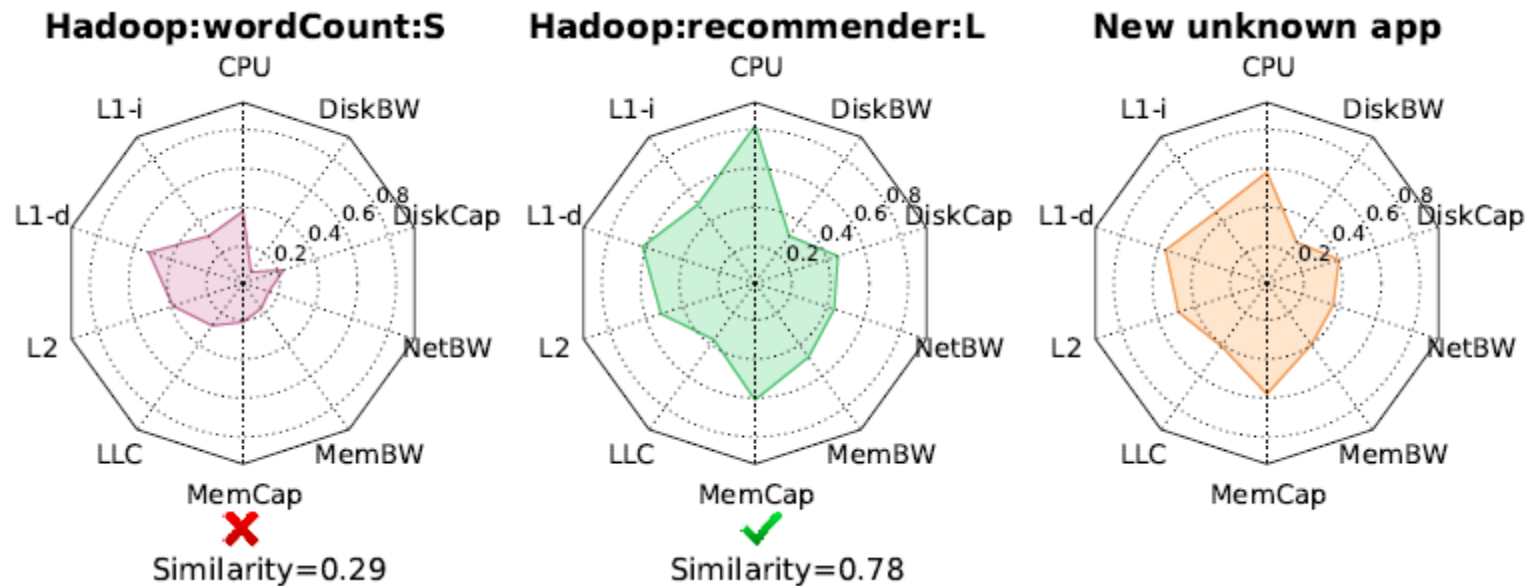
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Evolving Applications



- Cloud applications change behavior
- Users use the same cloud resources for several apps over time
- Bolt periodically wakes up, checks if app profile has changed; if so, reprofile & reclassify

Inference Within a Framework



- Within a framework, dataset and choice of algorithm affect resource requirements
- Bolt matches a new unknown application to apps in a framework by distinguishing their resource needs