

# A Comprehensive Framework For DDoS Resiliency in the Cloud

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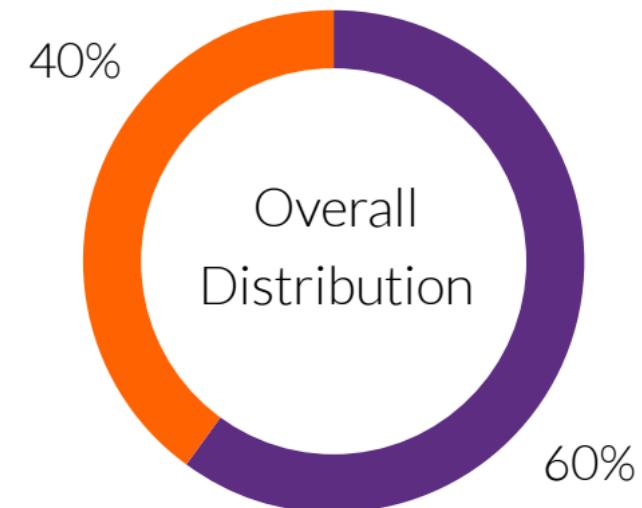
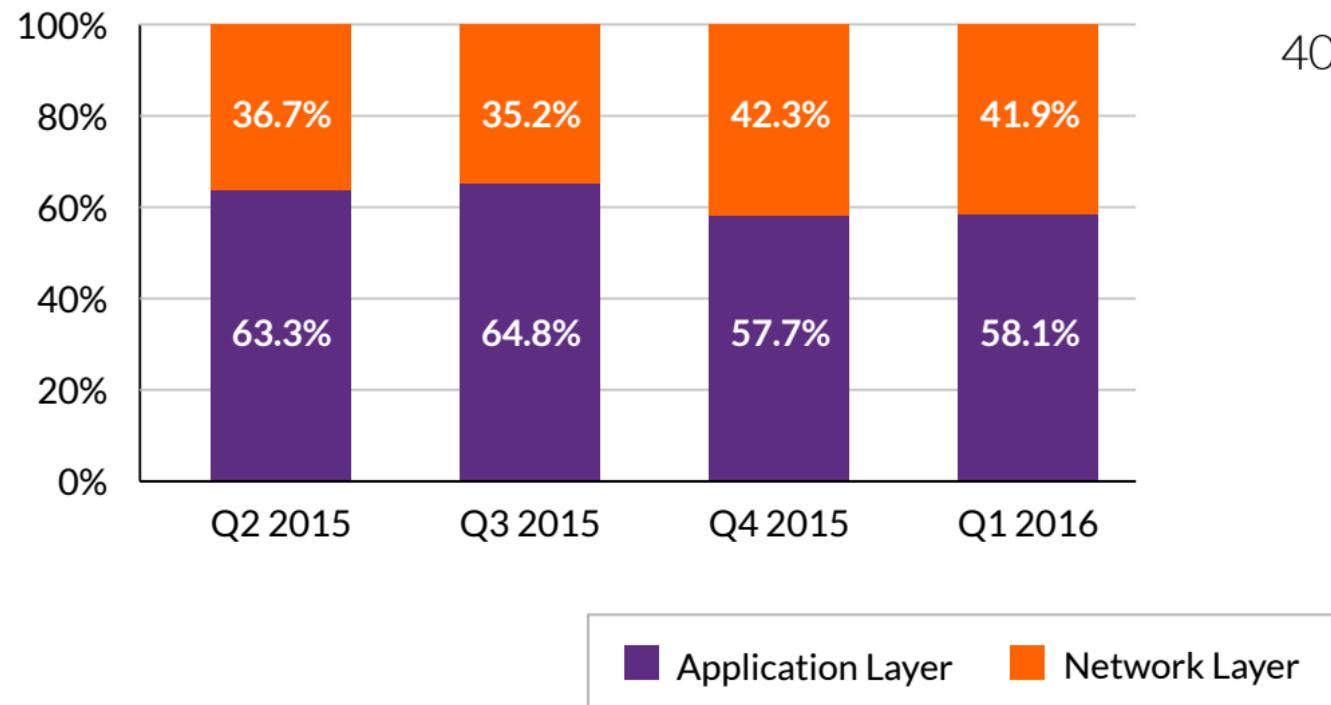
# Introduction

- DDoS attacks are becoming very effective
- Easy to launch attacks
  - DDoS as a Service
  - \$38 for one hour/month attack subscription service
- “*Subleasing*” botnets containing millions of hosts
  - User machines, mobile devices, IoT devices, etc.
- Largest recorded DDoS assault at 650 Gbps



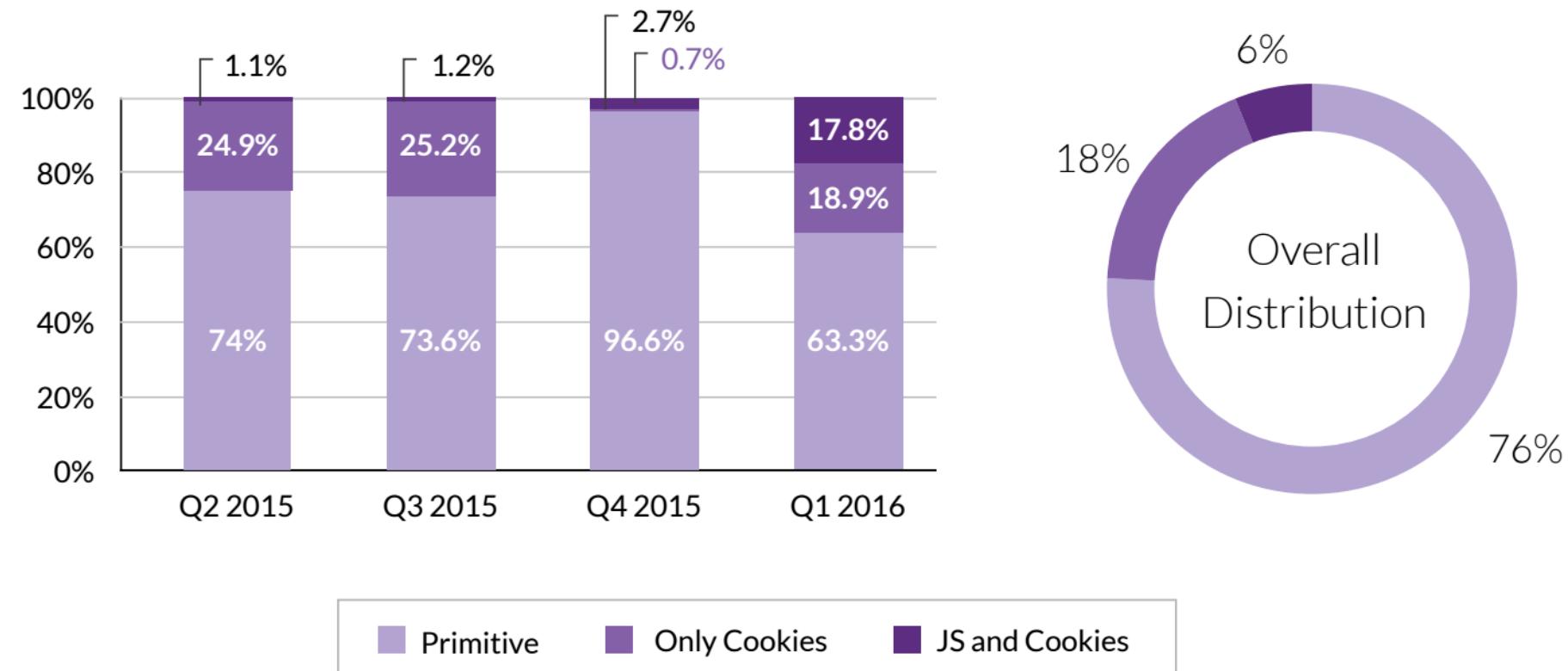
# Motivation

- DDoS is far from being solved



# Attack Capabilities

- Majority of botnets are still primitive





# The Cloud

- Cloud adoption rates are increasing
  - Cloud providers making it easier
- Provides flexibility and elasticity
  - Efficient use of ever-increasing capacity
- Lucrative targets for DDoS attacks
- Does profit drive security?
  - Telemetry infrastructure unexploited



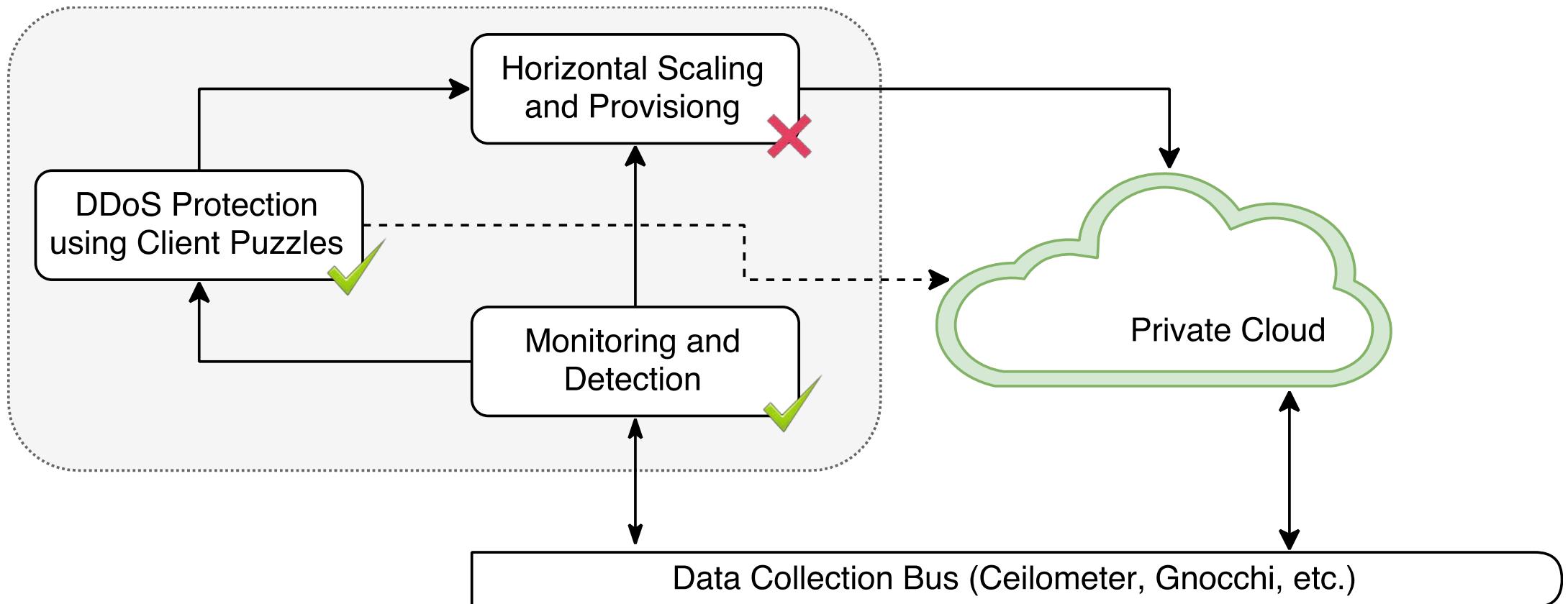
# Traditional Approaches

- Often very intrusive
  - Capabilities require changes to core routers
- Require cooperation between ISPs
  - Unlikely to happen without enforcement
- Require expensive classification of hosts
  - IP traceback is expensive and easily fooled
- Very few make use of the cloud
  - The ones that are often proprietary

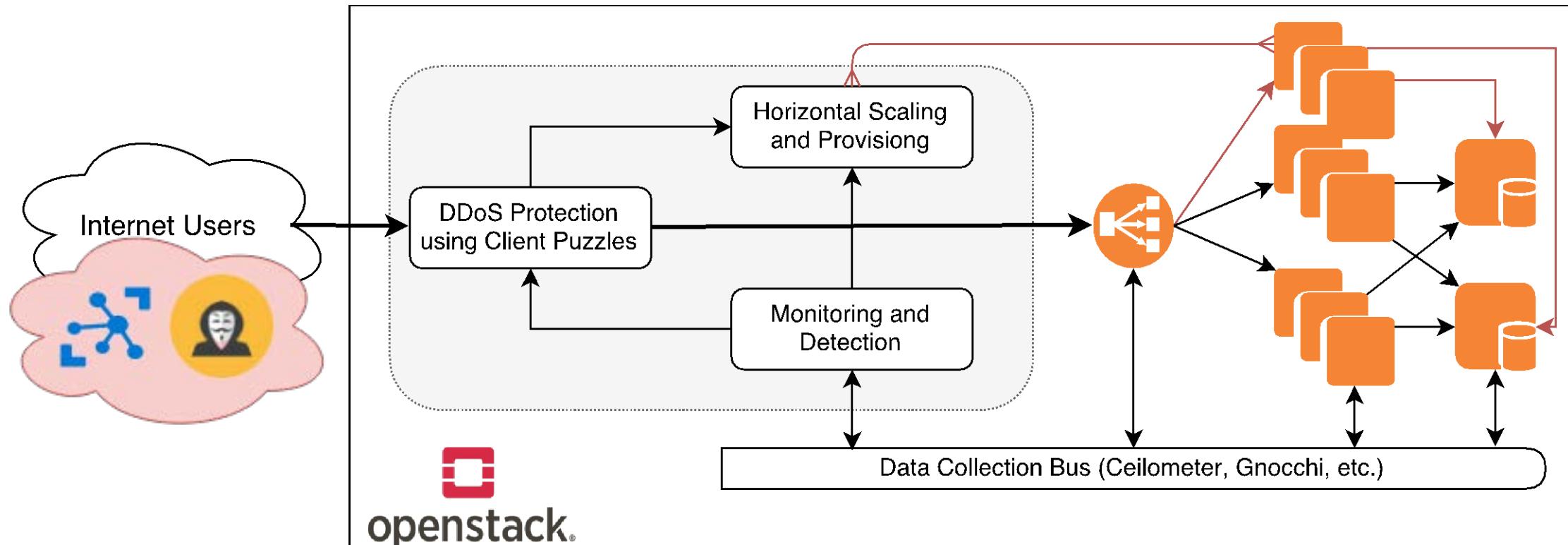
# Requirements

- We need to detect DDoS attacks
  - Distinguish between high load and DDoS
- We need to effectively respond to attacks
  - Trigger protection mechanism
  - Maintain operation
- We need to scale up horizontally if possible
  - High availability, Content Delivery Networks (CDN)

# Approach



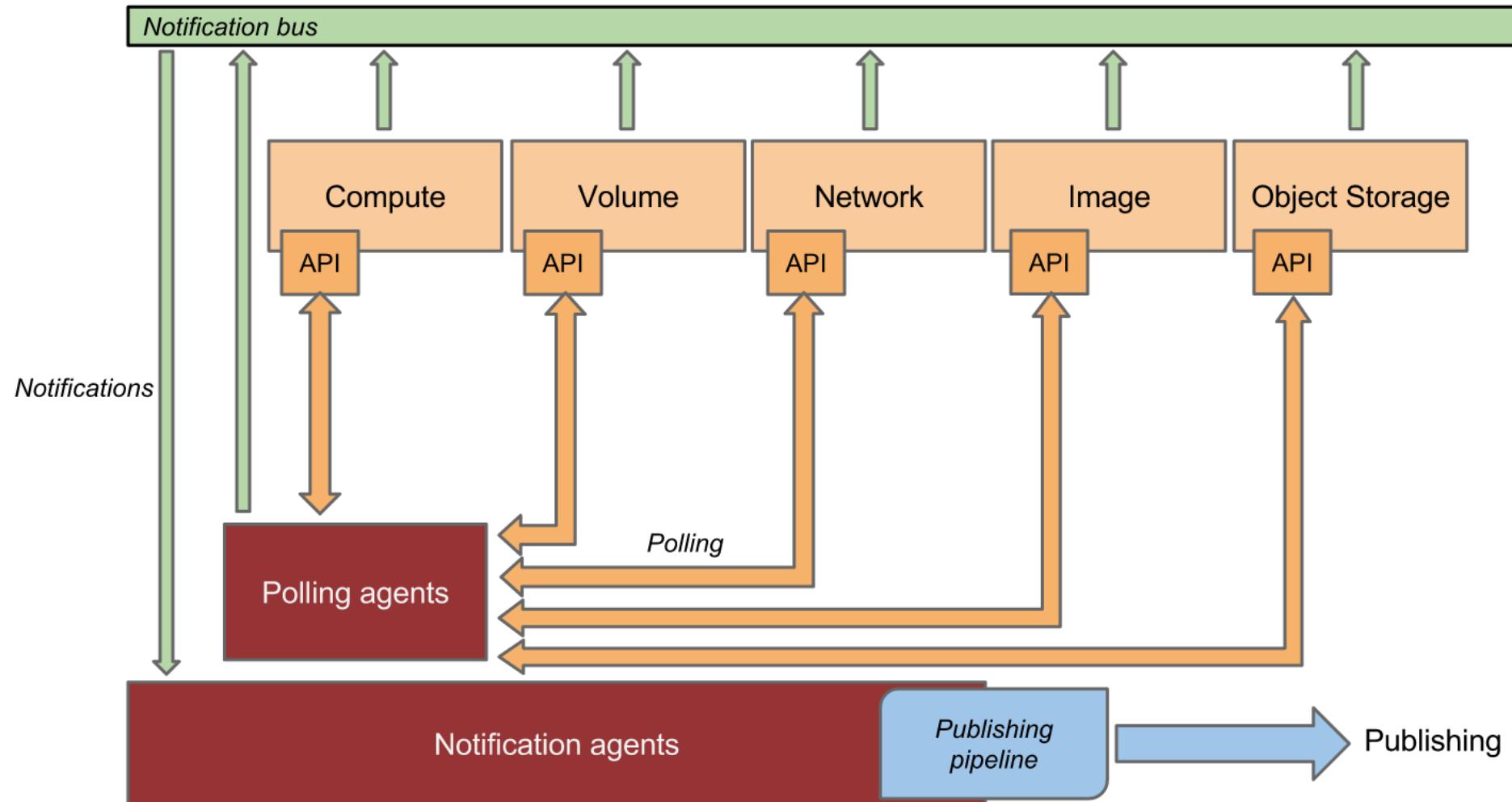
# System Model



# Data Collection

- Openstack, EC2, Azure provide strong telemetry infrastructure
  - Make use of it for security monitoring
- Openstack ceilometer
  - Complemented with Gnocchi for current and future versions
  - Time series database as a service

# Ceilometer



# Change Point Detection

- Statistical detection of abrupt changes in normal behavior
- Traditionally, observe
  - $\Delta_t = \{SYNReq_t - SYNRep_t\}, t = 0, 1, 2, \dots$
  - Focuses only on network information
  - Myopic to performance of server instances

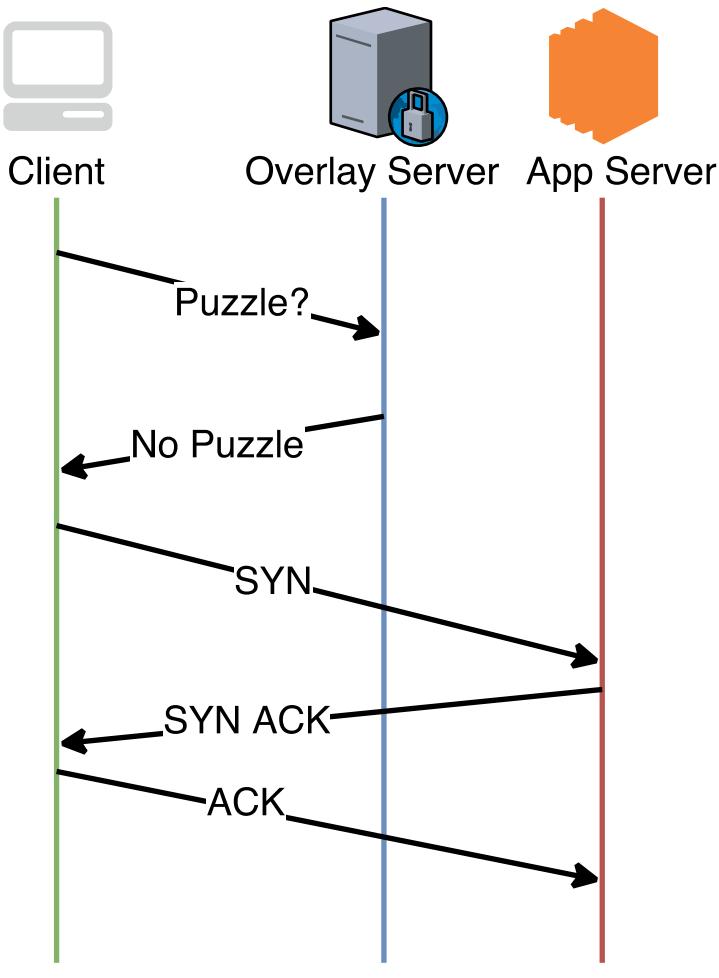
# In Our Approach

- Ceilometer data collection
  - Disk usage, CPU utilization, Memory utilization, Network utilization
  - [Apache logs]
- Define new sample vector
  - $\overrightarrow{\Delta_t} = \{disk_t, cpu_t, mem_t, req_t, rep_t\}^T$
  - Provides richer definition of normal behavior
  - Covers larger space of attacks

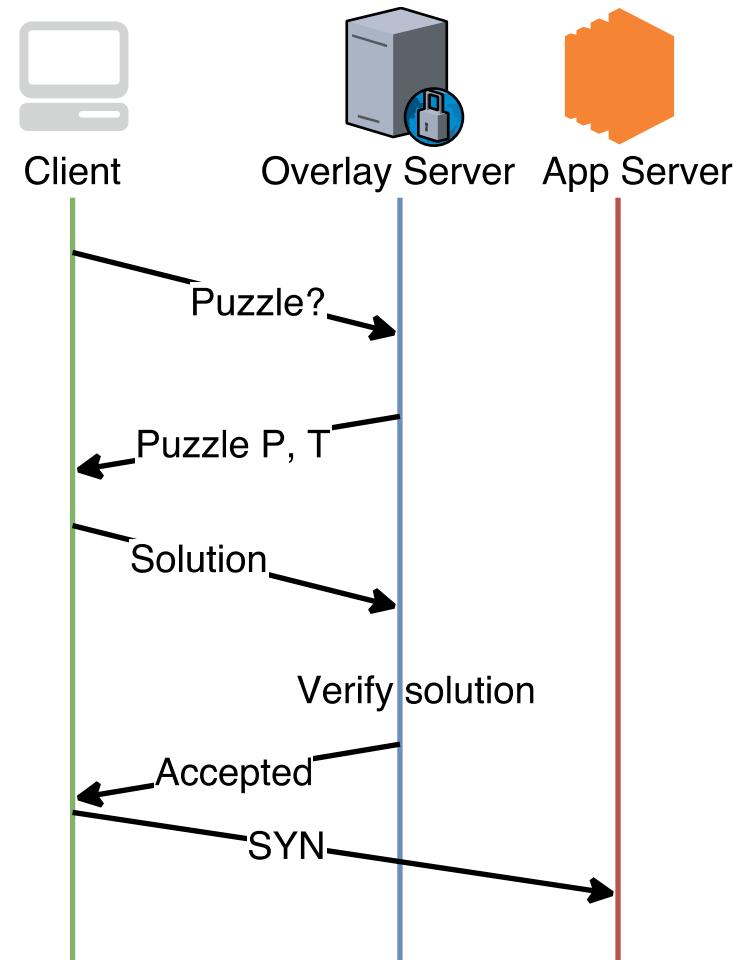
# Client Puzzles

- Efficient stateless proof of work
  - Receive service only after appropriate “*payment*”
- No puzzle solution required under regular load
- Non intrusive, no infrastructure change required
- Puzzle mechanism initiated upon attack detection
  - Use historical data to select puzzle complexity
  - Mechanism/Incentive design problem

# Puzzle Protocol



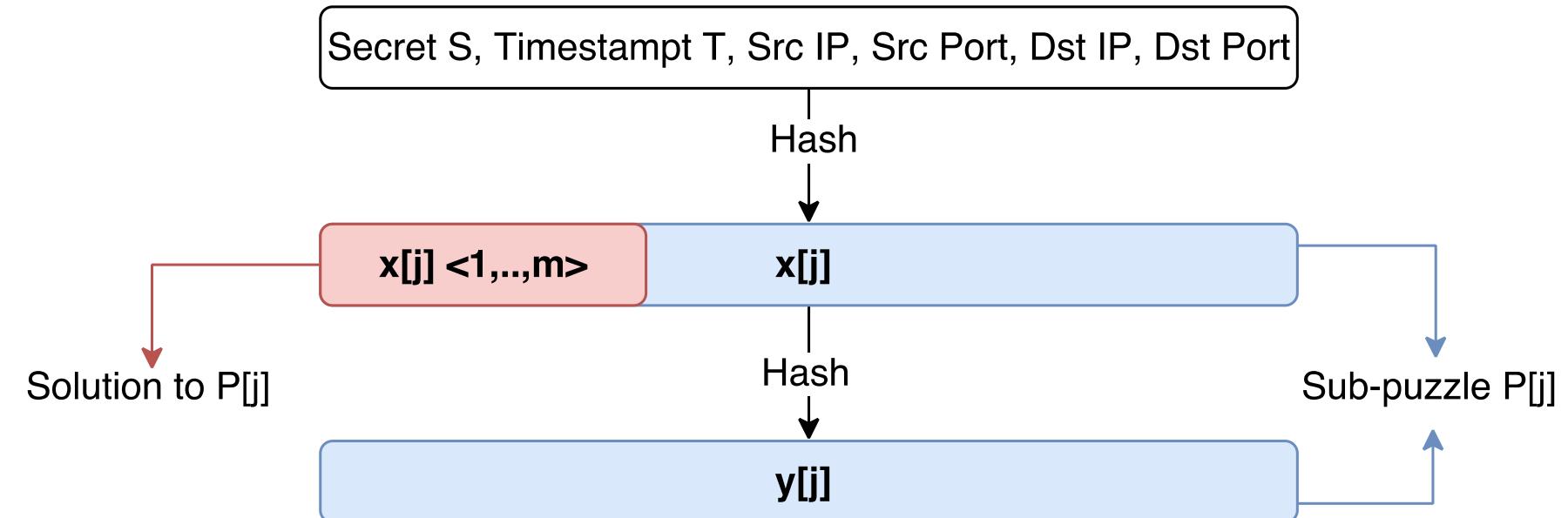
Under Regular Load



Under Attack

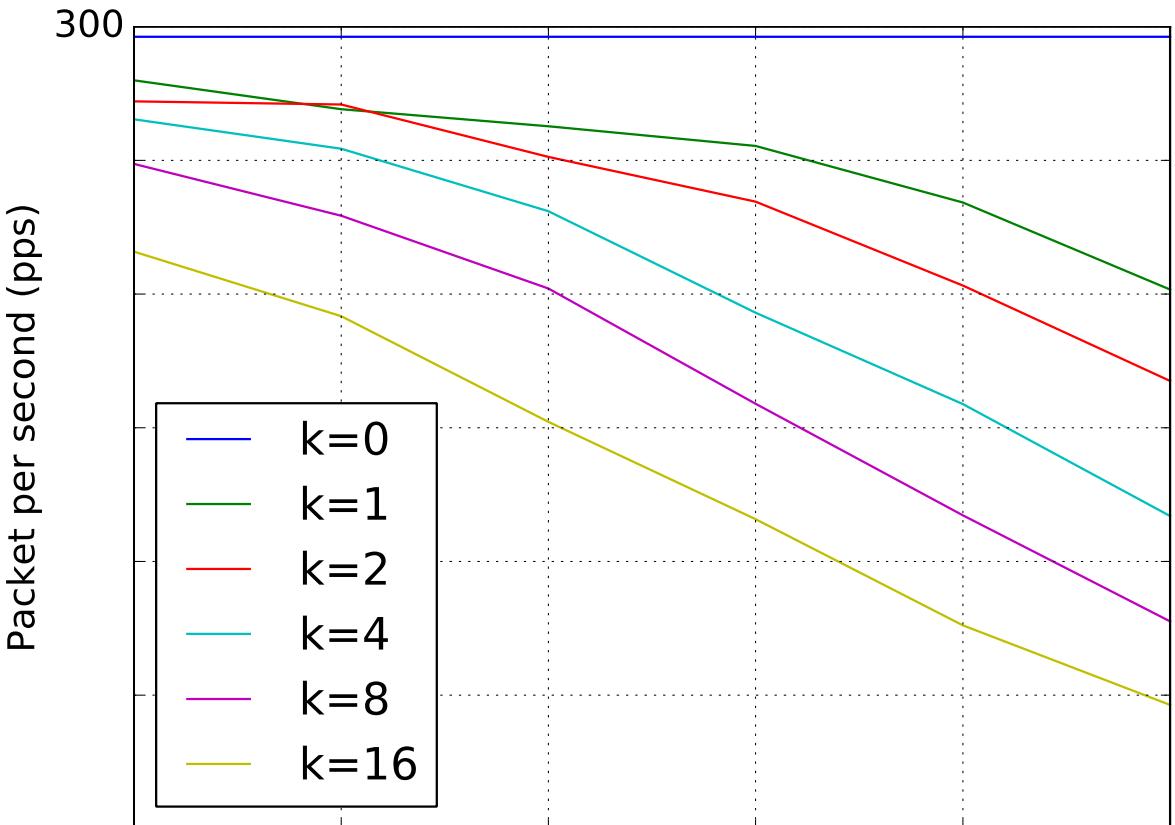
# Puzzle Construction

- Each puzzle is composed of  $k$  sub-puzzles
  - Each sub-puzzle of length  $m$  bits
- Solve by brute force



# Effectiveness

- Stateless mechanism to filter clients
- Serious rate limiting on the client side
- More complexity comes at lower cost for server



# Limitations

- Can be DDoS'ed
  - More efficient ways to generate and check puzzles
  - White/Black listing clients based on history
  - Still better than SYN flood attacks
- How to pick  $k$  and  $m$ 
  - Currently fixed for all users
  - Increased for everyone when attack intensifies
  - Mechanism/Incentive design problem

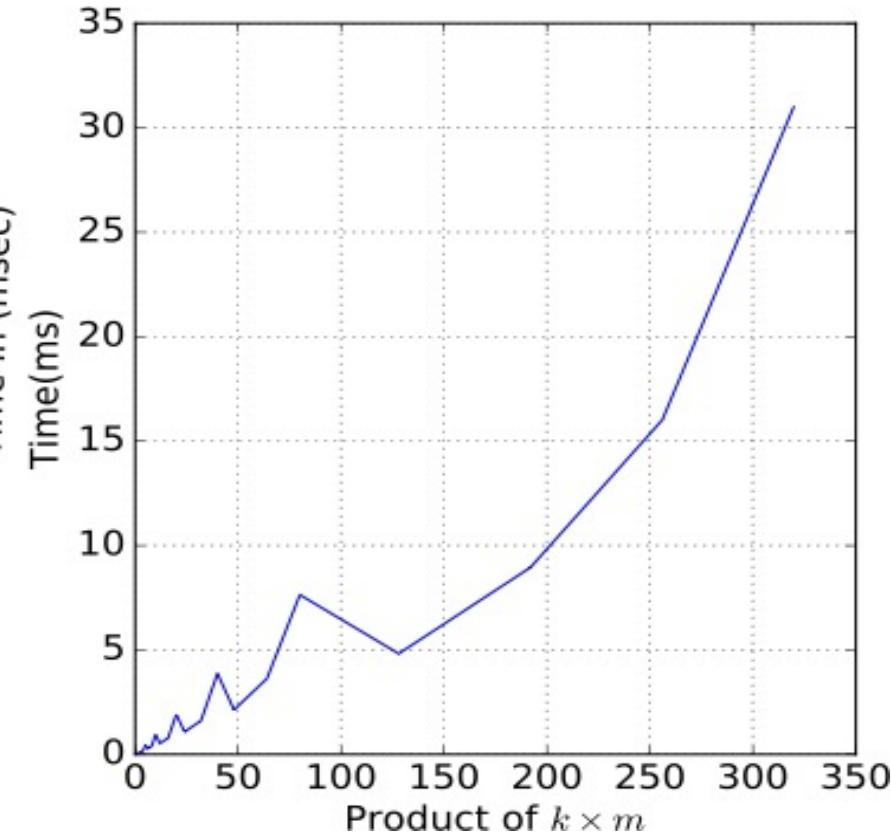
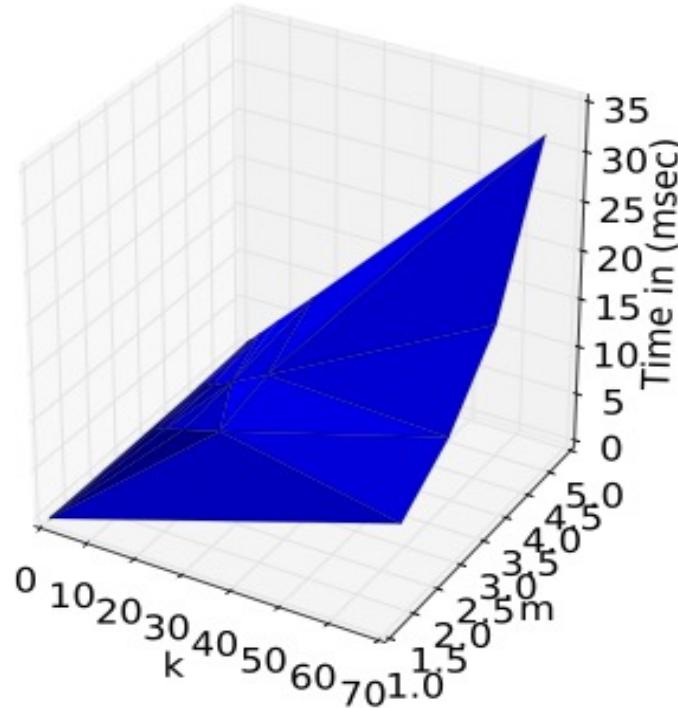
# Choosing $k$ and $m$

- Provide service for those who actually want it
- Motivation from network pricing
- $x_i$ : Request rate for client  $i$
- Define utility for each client

$$U_i(x_i, x_{-i}) = \log(1 + x_i) - \boxed{\text{PuzzleCost}_i} - \boxed{\text{ServiceDelay}}$$

# Time to Solve a Puzzle

- Expected number of hashes to solve a puzzle is  $k \times 2^{\{m-1\}}$





# Service Delay

- Model application service as an M/M/c queue
  - c is the number of server replicas
- Rate of arrivals  $\lambda = \frac{\sum_i x_i}{N}$ , N being the number of flows
- Service rate  $\mu$  is estimate from *ab* and other stress testing tools
- Analytical solution for expected wait time  $W$ 
  - Function of  $\rho = \frac{\lambda}{c\mu}$

# Service Time Estimate

```
This is ApacheBench, Version 2.3 <$Revision: 1706008 $>
Copyright 1996 Adam Twiss, Zeus Technology Ltd, http://www.zeustech.net/
Licensed to The Apache Software Foundation, http://www.apache.org/

Benchmarking 34.210.20.7 (be patient)

Server Software:      nginx/1.12.0
Server Hostname:     34.210.20.7
Server Port:          80

Document Path:        /
Document Length:     52757 bytes

Concurrency Level:    100
Time taken for tests: 44.630 seconds
Complete requests:   500
Failed requests:     0
Total transferred:   26532500 bytes
HTML transferred:    26378500 bytes
Requests per second: 11.20 [#/sec] (mean)
Time per request:    8926.041 [ms] (mean)
Time per request:    89.260 [ms] (mean, across all concurrent requests)
Transfer rate:        580.56 [Kbytes/sec] received
```



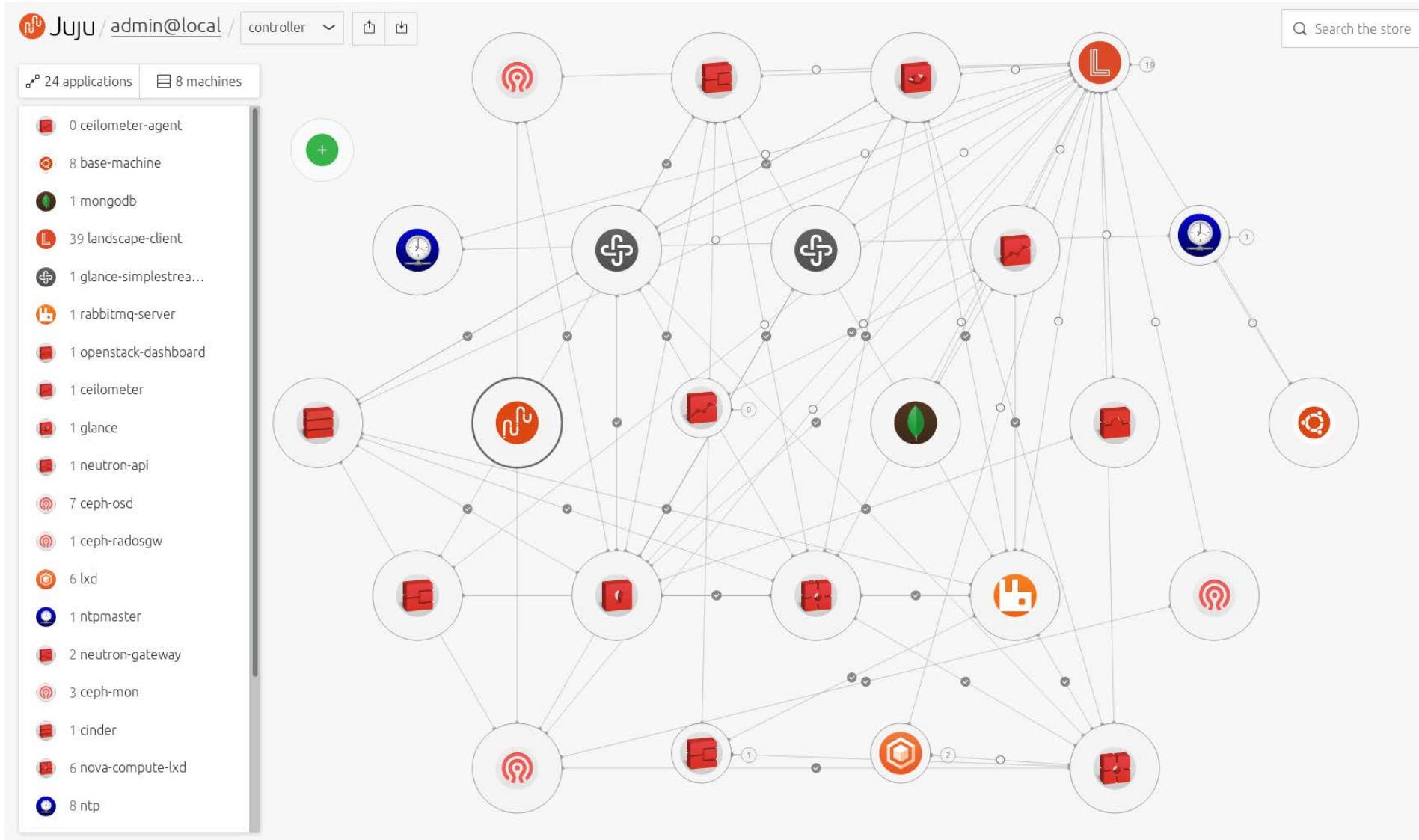
# Stackelberg Game (Mechanism Design)

- $U_i(x_i, x_{-i}) = \log(1 + x_i) - k_i 2^{\{m_i-1\}} x_i - W$
- $W$  is a function of  $x_i$  and  $x_{-i}$
- Solve for equilibrium rates  $x_i^*$ 
  - $U_i(x_i^*, x_{-i}^*) \geq U_i(x_i, x_{-i}^*) \quad \forall x_i, \forall i$
- The cloud's design problem is to find  $k_i^*$  and  $m_i^*$ 
  - $\underset{p_i, k_i \in \mathbb{N}}{\operatorname{argmax}} \sum_i \log(1 + x_i^*) - \sum_i (m_i \times k_i) x_i^*$

# Horizontal Scaling Ideas

- Adding more replicas naively
- Give networking and compute budgets
  - Solve optimization problem
- Think in terms of the queueing model
  - Scale up to keep  $\rho < 1$

# OpenStack Deployment



# Infrastructure and Applications

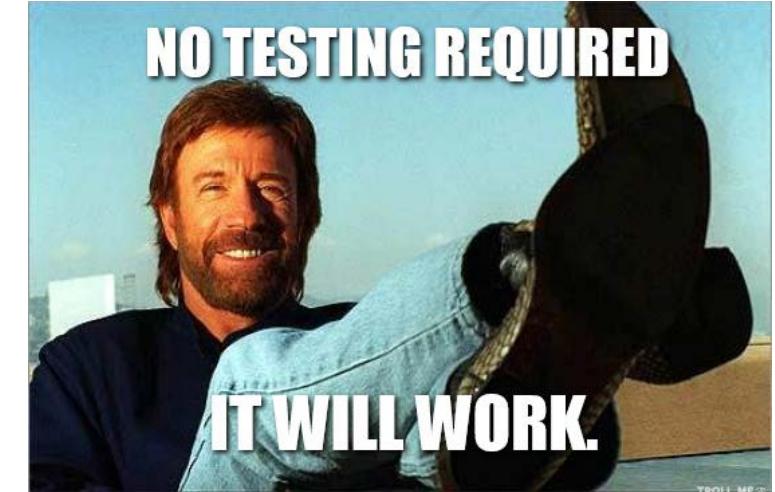
- Deployment composed of 4 servers and 6 commodity machines
- Running Wordpress server replicas with MySQL backend
- NginX load balancer

# Steps for Evaluation

- Evaluate change point detection mechanism
- Simulations to evaluate puzzle difficulty selection
- Comprehensive evaluation
  - Simulate normal and heavy load using stress testing tools
  - Simulate attack using botnet simulator

# Major Challenge

- We have tried multiple deployment tools
  - Each claim to be the “Chuck Norris” of deployment
- Lack of documentation for troubleshooting deployment errors
- Biggest success with Ubuntu Autopilot so far



# Conclusion

- Comprehensive design for resilient applications in a private cloud
- Uses telemetry infrastructure for monitoring
- Uses client puzzles for DDoS protection under attack
- Provides horizontal scaling to guarantee performance