

Cloud Index Tracking: Enabling Predictable Costs in Cloud Spot Markets



Supreeth Shastri and **David Irwin**

University of Massachusetts Amherst

Spot Servers are gaining significance in the cloud

Servers that may **terminate anytime** after an **advance warning** period

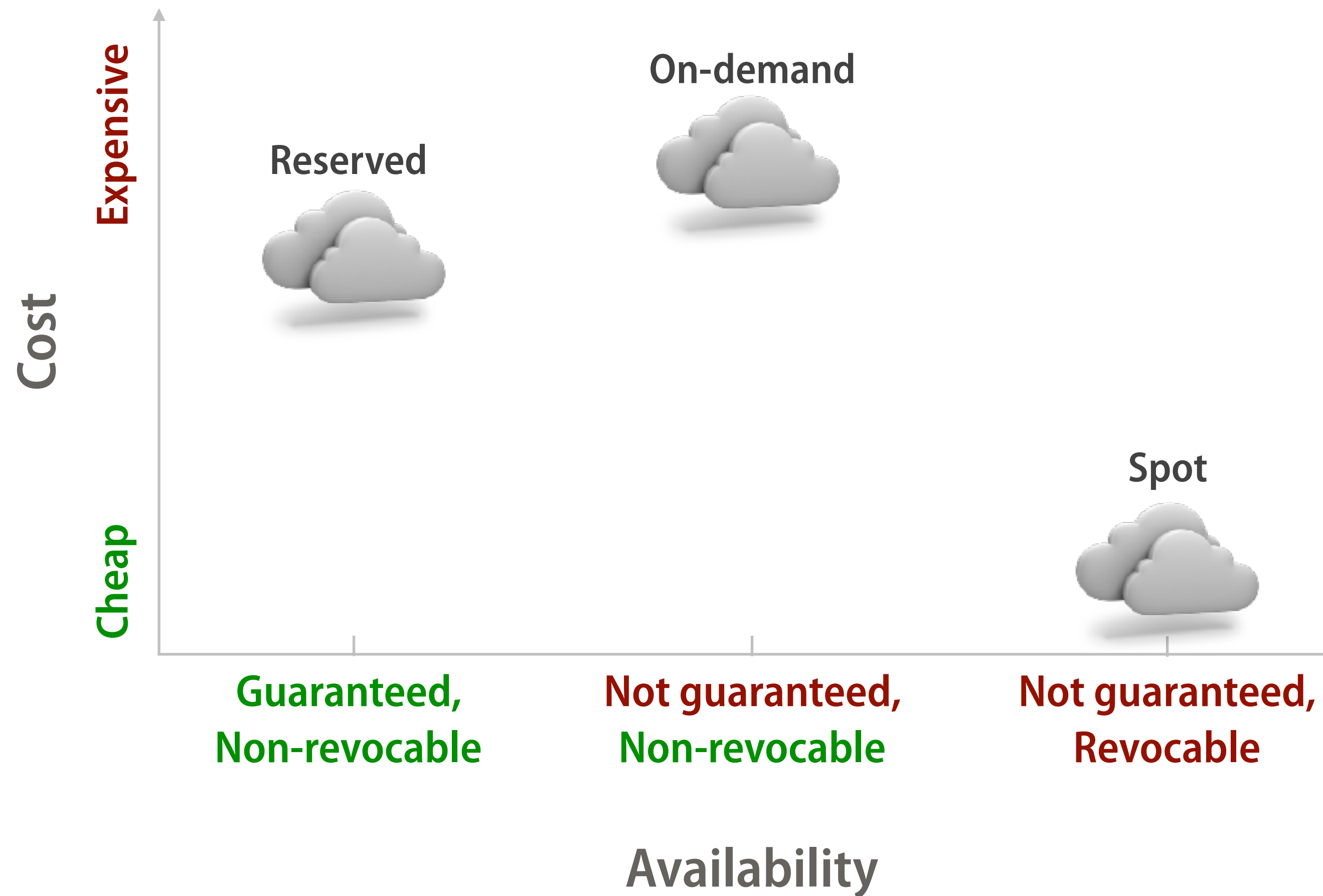
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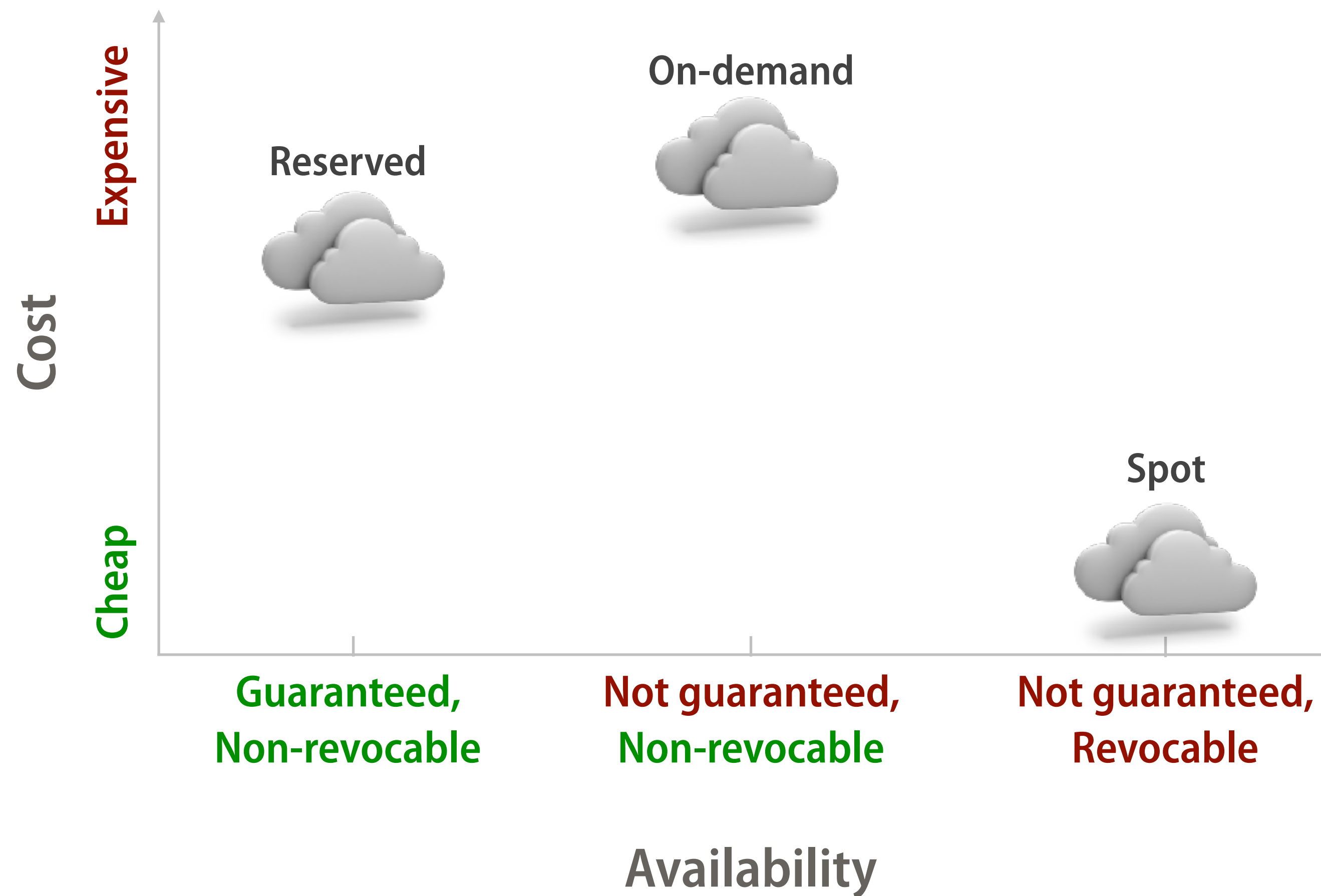
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Spot instances helped scale our clusters up by **4X** during the discovery of the **Higgs Boson**



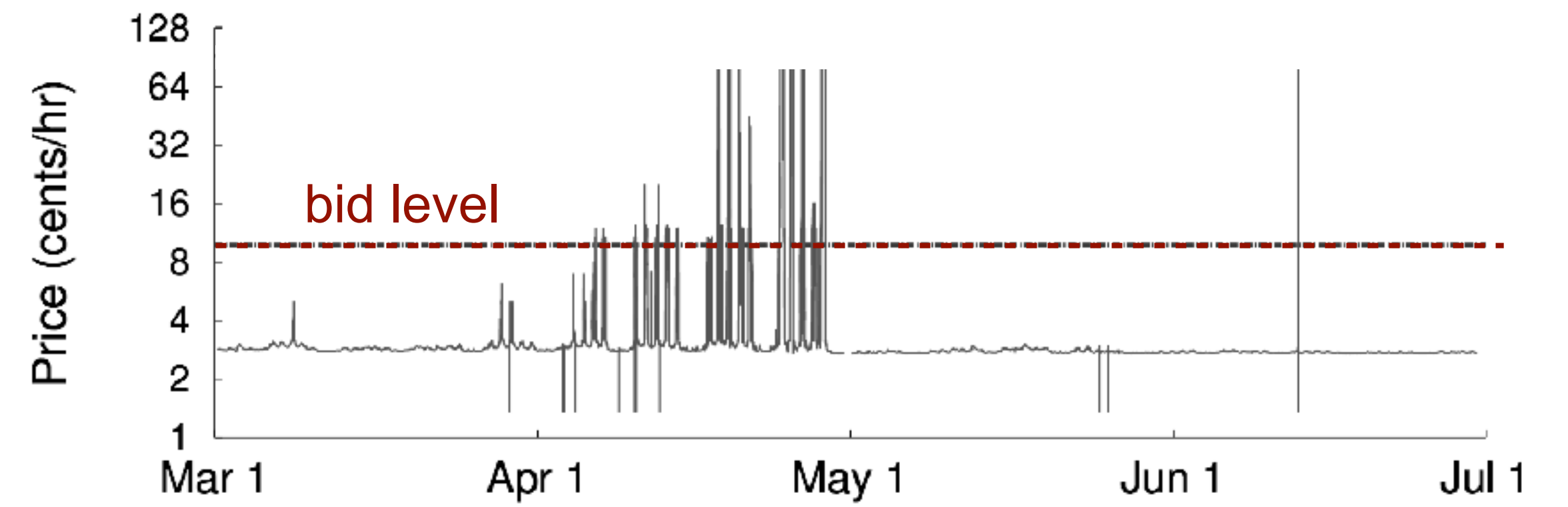
Researchers built the **largest** HPC cluster in the cloud with **1.1 million** vCPUs on EC2 spot



Spot server pricing

while low on average, it is characterized by

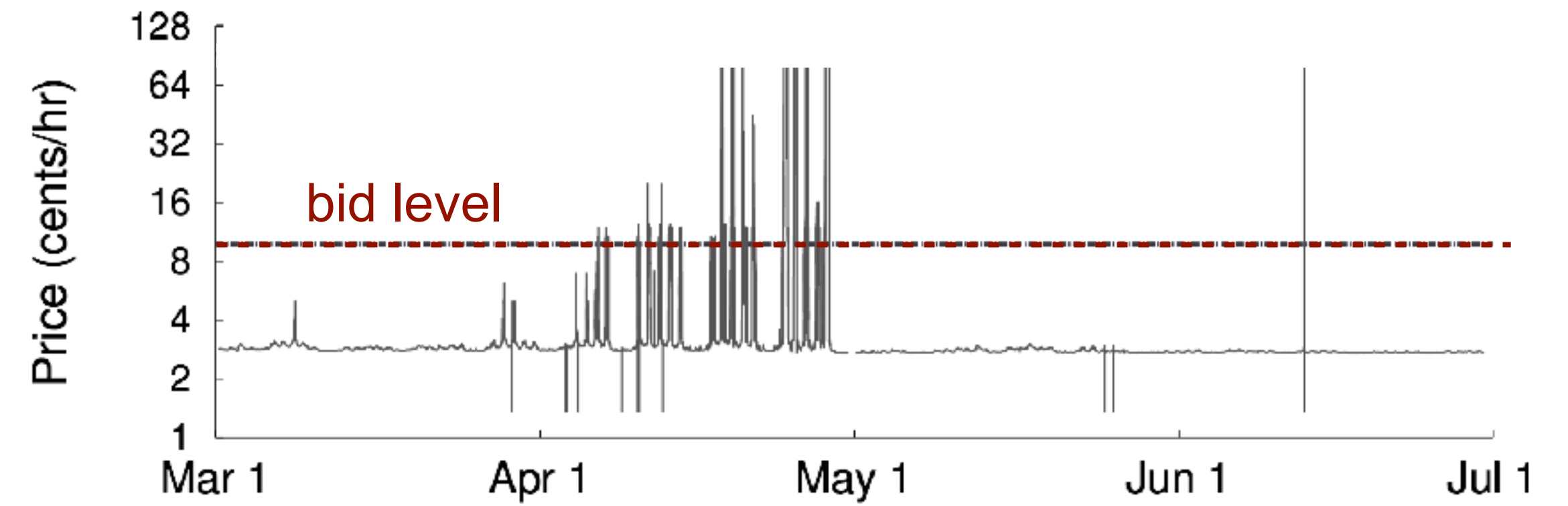
variability and *deliberate revocations*



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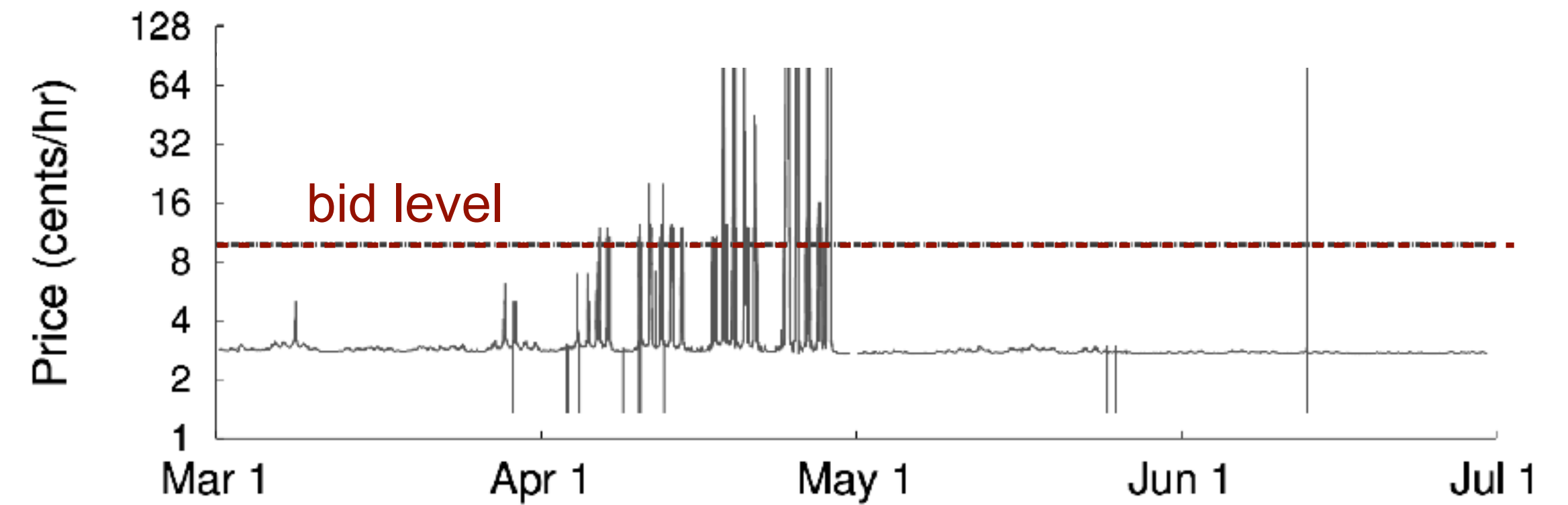


Predicting Spot Prices is an Active Area of Research

Ability to compare servers, plan IT budgets, and avoid disruptive revocations

Spot server pricing

while low on average, it is characterized by *variability* and *deliberate revocations*



Predicting Spot Prices is an Active Area of Research

Ability to compare servers, plan IT budgets, and avoid disruptive revocations

2015

Bid [**SIGCOMM**]

SpotOn [**SoCC**]

Cumulon [**VLDB**]

2016

No-bid [**HotCloud**]

Flint [**Eurosys**]

BOSS [**Infocom**]

2017

Prob-Guarantee [**SC**]

Proteus [**EuroSys**]

Exosphere [**SIGMETRICS**]

2018

LSTM [**HPDC**]

Tributary [**ATC**]

Predicting Spot Prices is Important

Prior work models individual spot server prices based on their **historical spot price data**

Difficult
Accurately Predicting Spot Prices is ~~Important~~

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Accurately Predicting Spot Prices is ~~Important~~ *Difficult*

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68

Hardware
config

•

2

OS
types

•

2-5

Zones
(datacenters)

•

14

Regions
(country, state)

•

2

Time
commitments

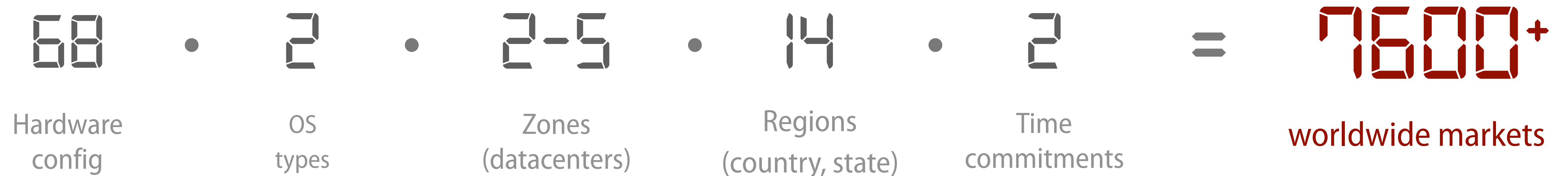
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$$\begin{array}{ccccccccc} 68 & \cdot & 2 & \cdot & 2-5 & \cdot & 14 & \cdot & 2 & = & 7600^+ \\ \text{Hardware} & & \text{OS} & & \text{Zones} & & \text{Regions} & & \text{Time} & & \text{worldwide markets} \\ \text{config} & & \text{types} & & \text{(datacenters)} & & \text{(country, state)} & & \text{commitments} & & \end{array}$$

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**One size fits all
model is unlikely**

**No visibility into
market internals**

**Limited correlation
with external variables**

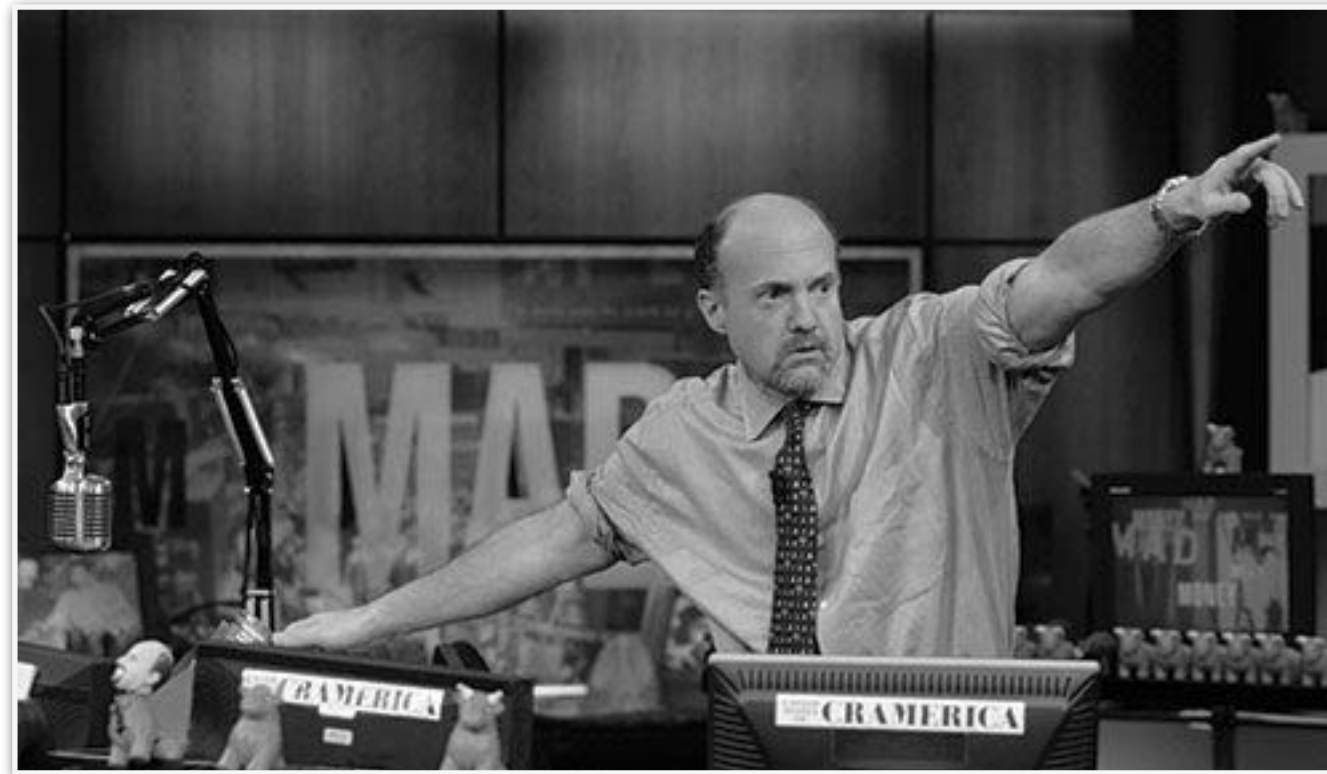




vs.

U.S. FUTURES		
DOW JONES INDUS FUT 3/17 (D./H7)		
19,904.00	▲	60.00 (0.30%)
S&P 500 3/17 (ES/H7)		
2,280.75	▲	6.25 (0.27%)
NASDAQ 100 3/17 (NQ/H7)		
5,116.50	▲	20.20 (0.40%)

Key Insight: A Market-based Index *for CLOUD*



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Rather than focusing exclusively on predicting **individual servers**, cloud users should make decisions based on **broader market indices**



Cloud Index

intuition for our hypothesis

index **construction** methodology

validation on Amazon EC2

Index-tracking

techniques for **predictability**

design of index-tracking by server hopping

performance **evaluation**

Underlying Characteristics of Large Cloud Platforms

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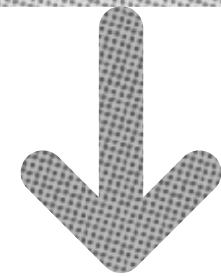
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Spot markets originating from the same physical machine family are not free from mutual interference

Underlying Characteristics of Large Cloud Platforms

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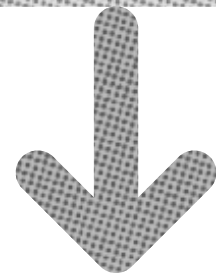


Not all spot markets could be **individually** modeled and predicted

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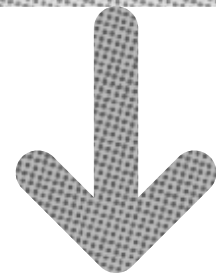
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Aggregate idle VM capacity in public cloud datacenters tends to be stable
[SoCC 2014, SOSP 2017]

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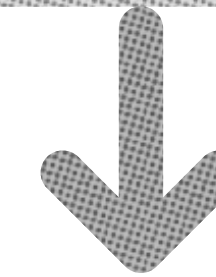
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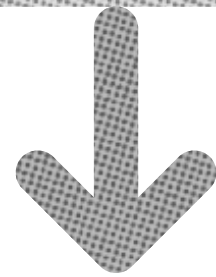


If idle capacity were priced like commodity, its clearing price will be **stable** and **predictable**

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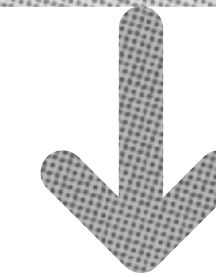
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We hypothesize that observing spot markets at **aggregate levels** (say, server family or datacenter levels) should lead to **stable prices**

Constructing a Market Index *for CLOUD*

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Characterizing an individual server **i**

Price = P_i , Memory = M_i GB

Compute = C_i ECUs

$$P_i^{\text{norm}} = \frac{P_i}{\sqrt{(C_i \cdot M_i)}}$$

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Characterizing a group of servers

Average of normalized prices

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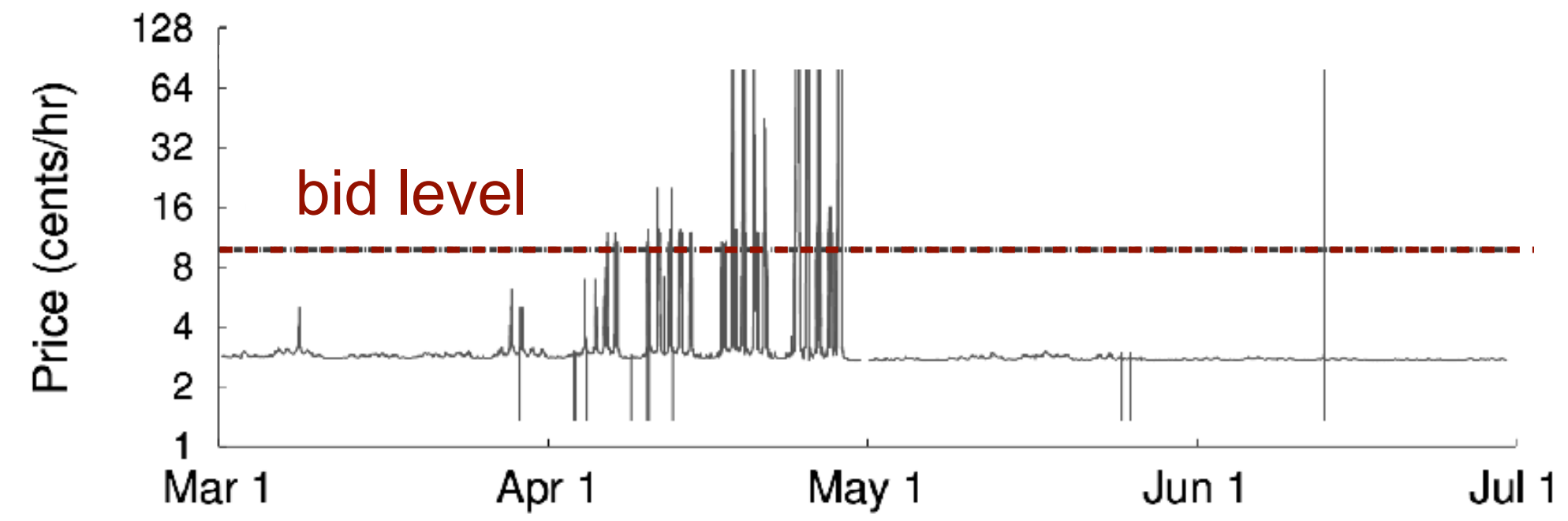
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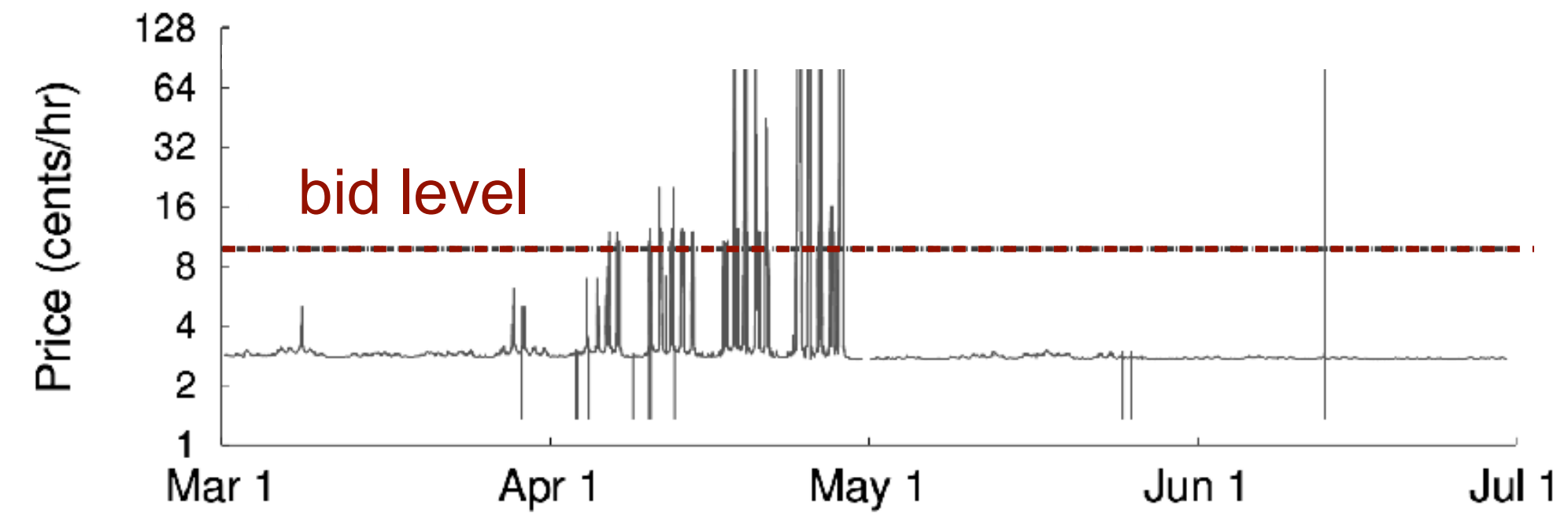
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Cloud index value represents the **average price per unit of compute time**
for the selected group of servers

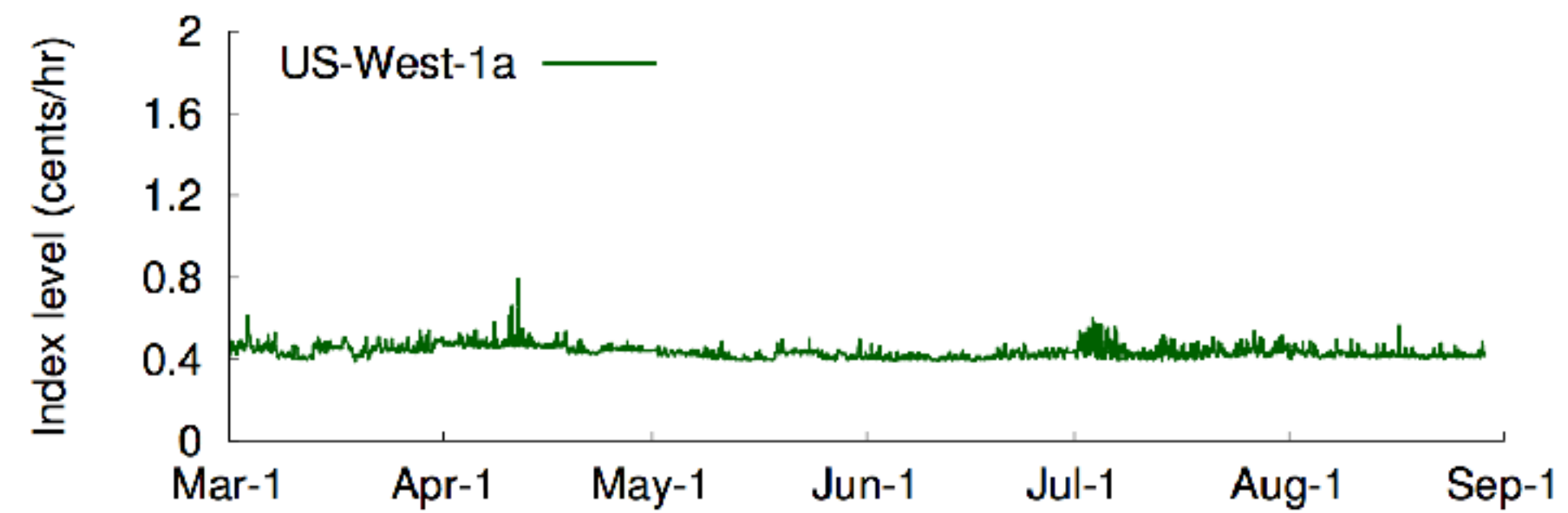
Individual Server Level



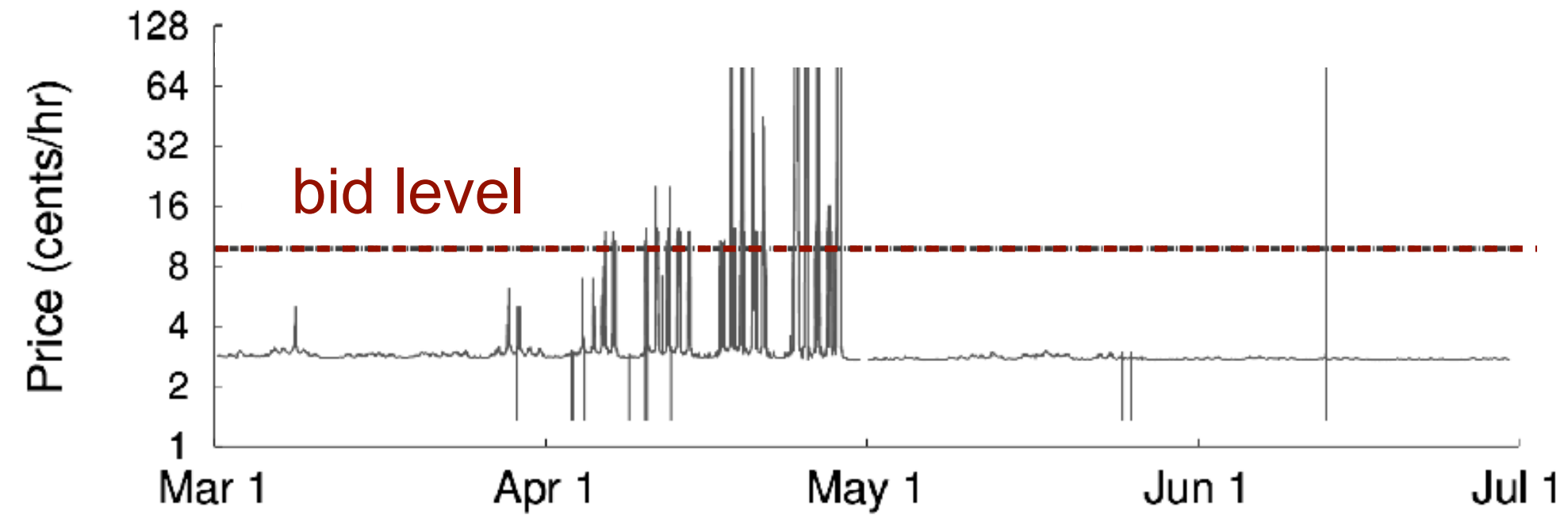
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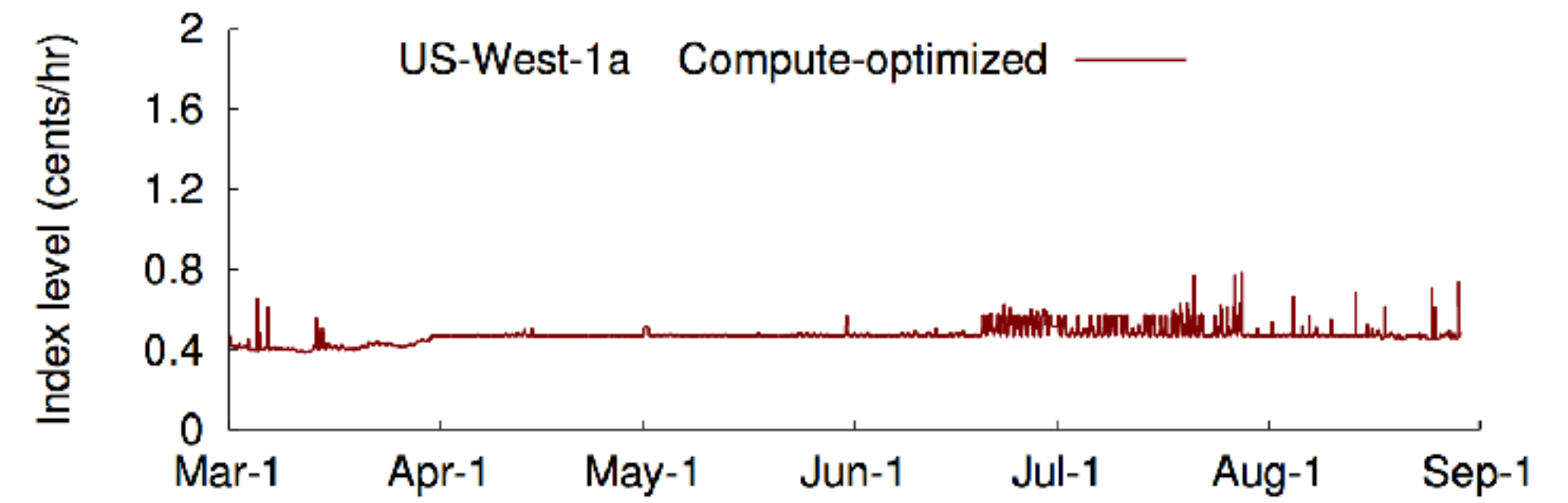
Datacenter Level (US-West-1a)



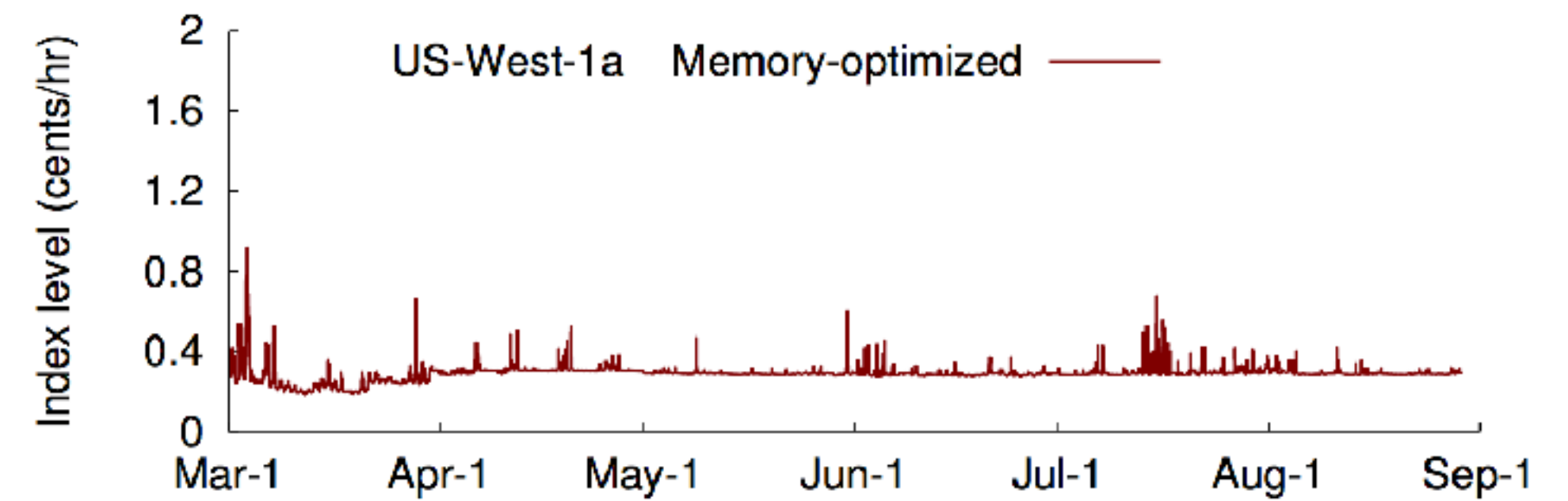
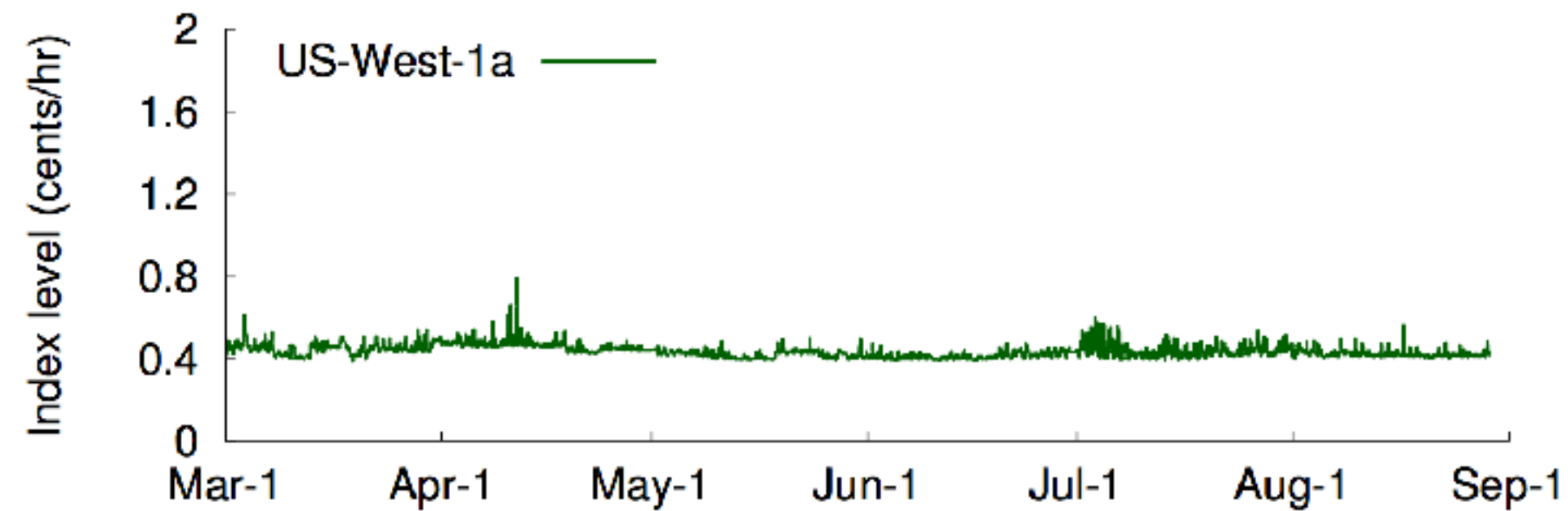
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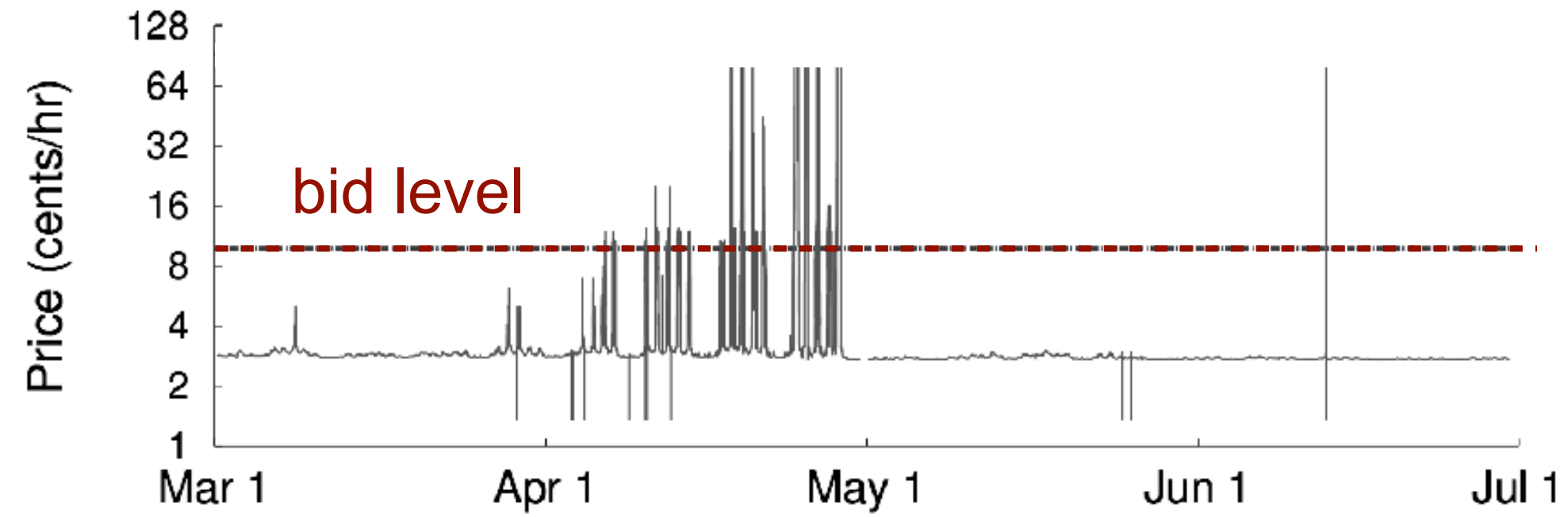
Server Family Level (US-West-1a)



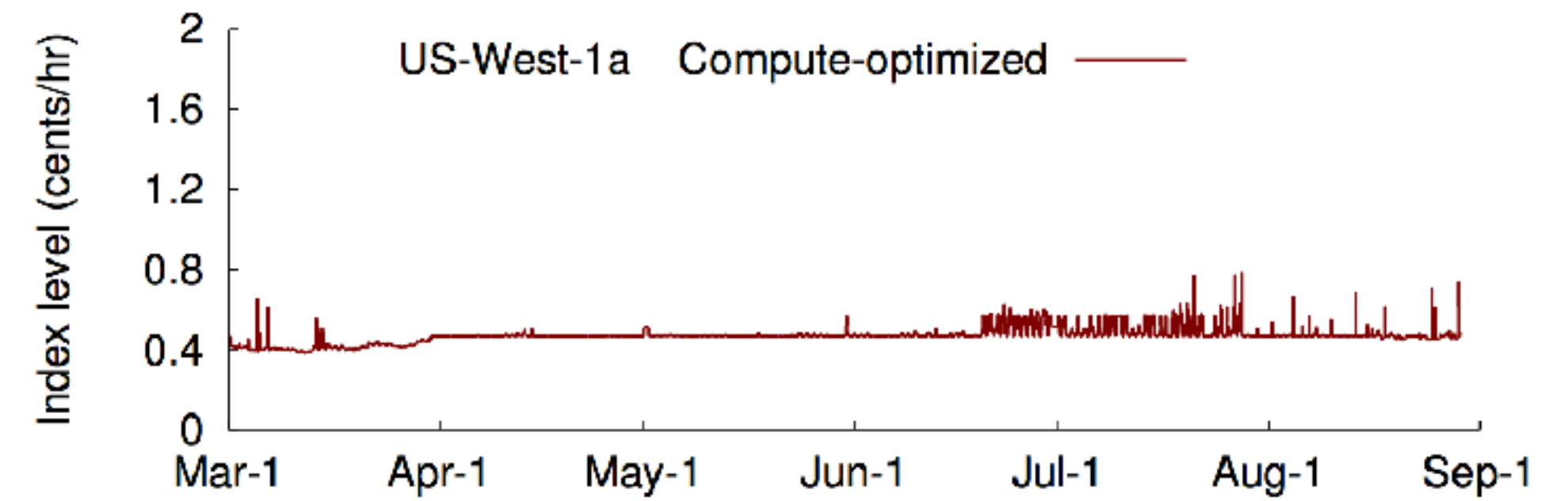
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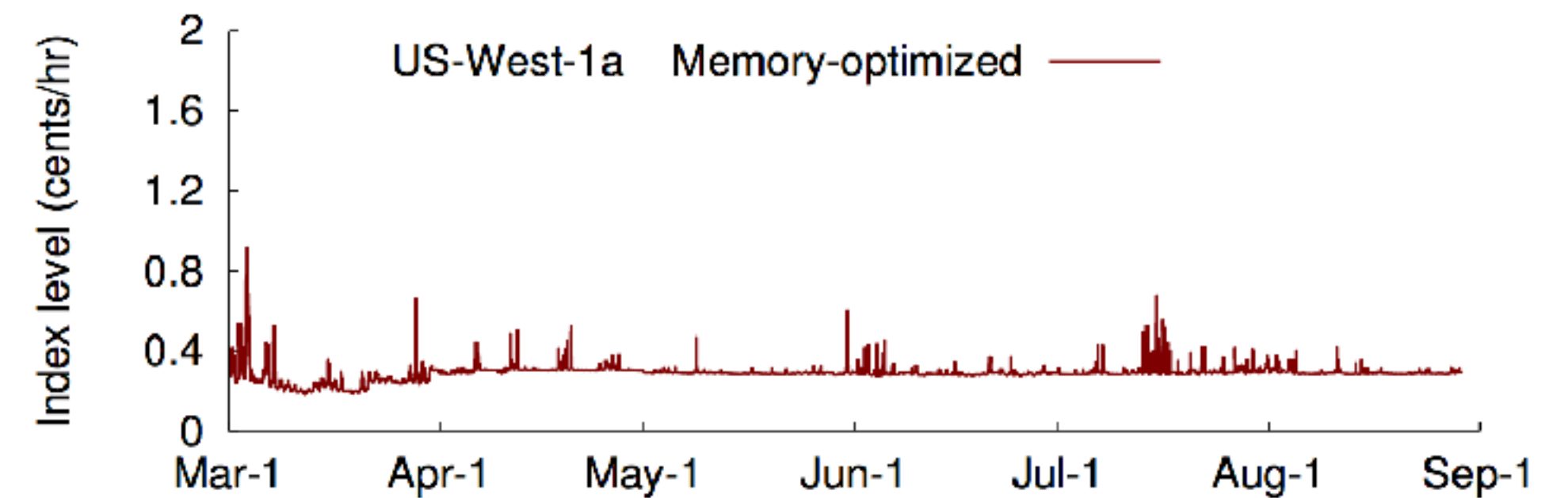
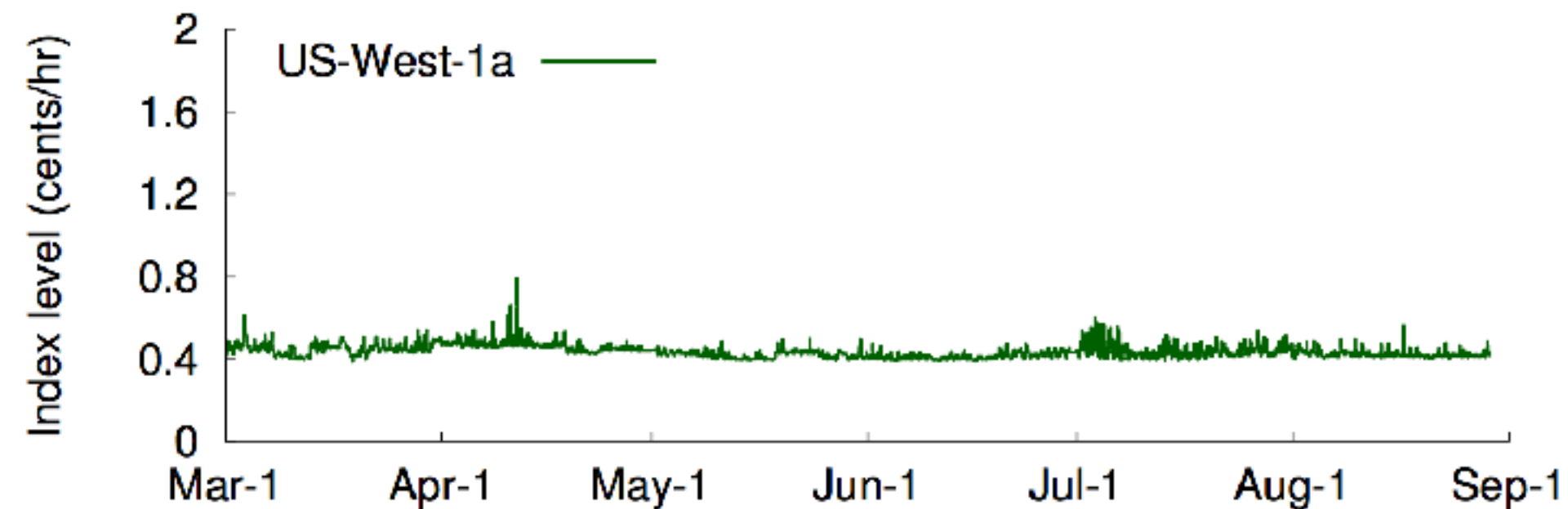
Individual Server Level



Server Family Level (US-West-1a)



Datacenter Level (US-West-1a)



Price prediction is more accurate and stable at **datacenter-** and **server family level** than **individual** level



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Index-tracking

techniques for **predictability**

design of index-tracking by server hopping

performance **evaluation**

Design elements

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Index-tracking in financial markets



Investments that match the returns of an index.

Construct a portfolio such that its constituent items are same as those present in the index.

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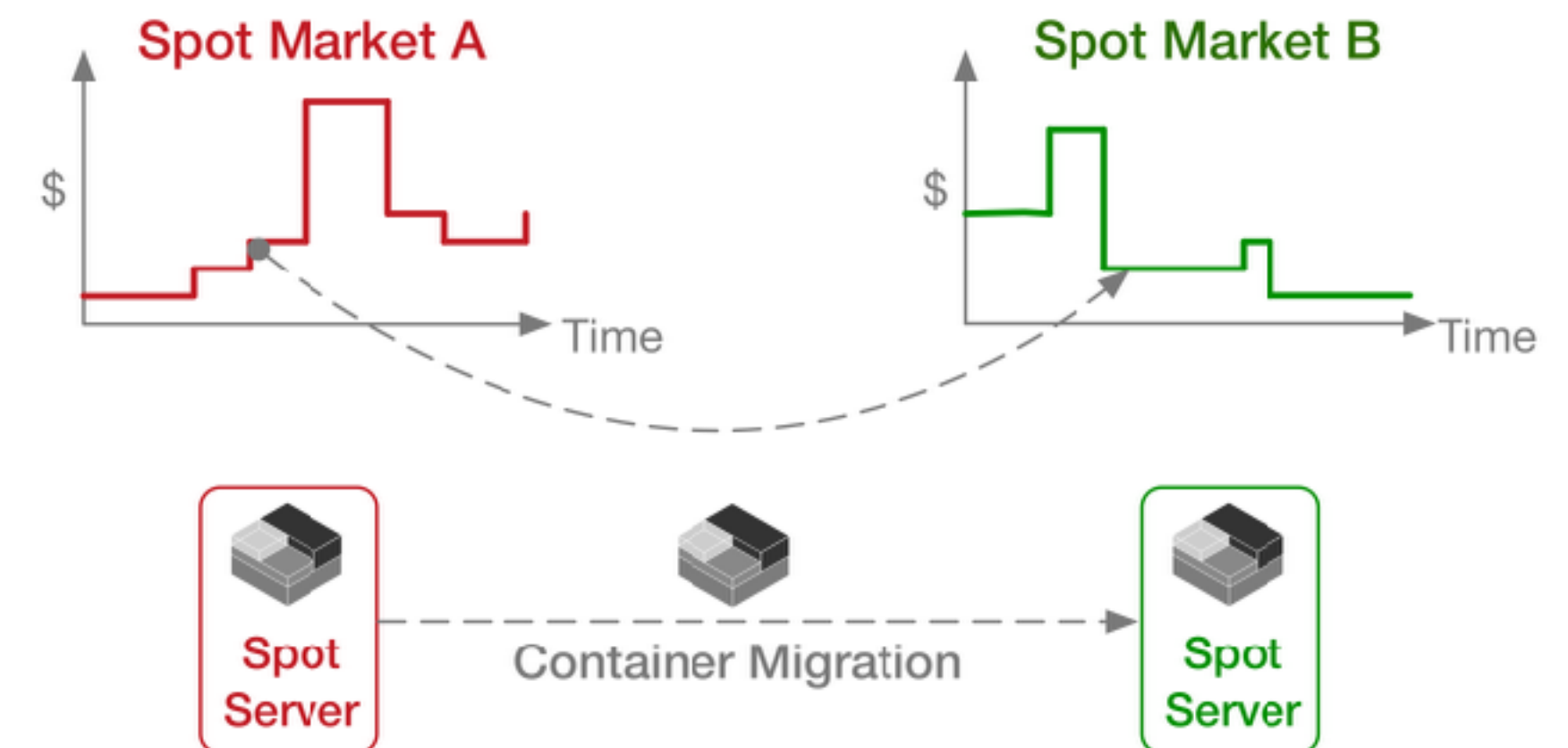
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Server hopping in cloud markets



A container that automatically hops spot VMs as market conditions change [SoCC 2017].

Increasing cost-efficiency, lowers revocations

Index Tracking by Server Hopping

Achieving index-level cost-efficiency despite market volatility

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Achieving index-level cost-efficiency despite market volatility

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Determine a broad set of **candidate markets**, and then compute its market index

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Host the application on a server that meets the **index-level cost-efficiency**

Index Tracking by Server Hopping

Achieving index-level cost-efficiency despite market volatility

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Determine a broad set of **candidate markets**, and then compute its market index

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Host the application on a server that meets the **index-level cost-efficiency**

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If market conditions violate the index invariant, then **transparently hop** to a better server

Index Tracking by Server Hopping

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Host the application on a server that meets the **index-level cost-efficiency**

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If market conditions violate the index invariant, then **transparently hop** to a better server

Server Choice

Select a server that shows best balance between **risk** (price volatility) vs. **reward** (cost-efficiency)

$$\text{Sharpe ratio} = \frac{(\mathbb{I} - \hat{P}_i)}{\text{std-dev}(\mathbb{I} - \hat{P}_i)}$$

\mathbb{I} = Index-level, and \hat{P}_i = Spot server's normalized efficiency

LXC based prototype for EC2 spot markets

<https://umass-sustainablecomputinglab.github.io/cloudIndex/>

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Evaluation

- ▶ Does index-tracking achieve **predictable expenses**?
- ▶ How does cost-availability of index-tracking **compare to others**?

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We compare **three systems** for running **two classes** of applications on EC2 spot markets

**Spot server with
static prediction
(SpotFleet)**

vs.

**Spot server with
cost-based hopping
(HotSpot)**

vs.

**Spot server with
index-tracking**

Long-running Single-node App

E.g., IoT sinks, crypto miners, p2p file trackers

Bulk-synchronous Parallel Jobs

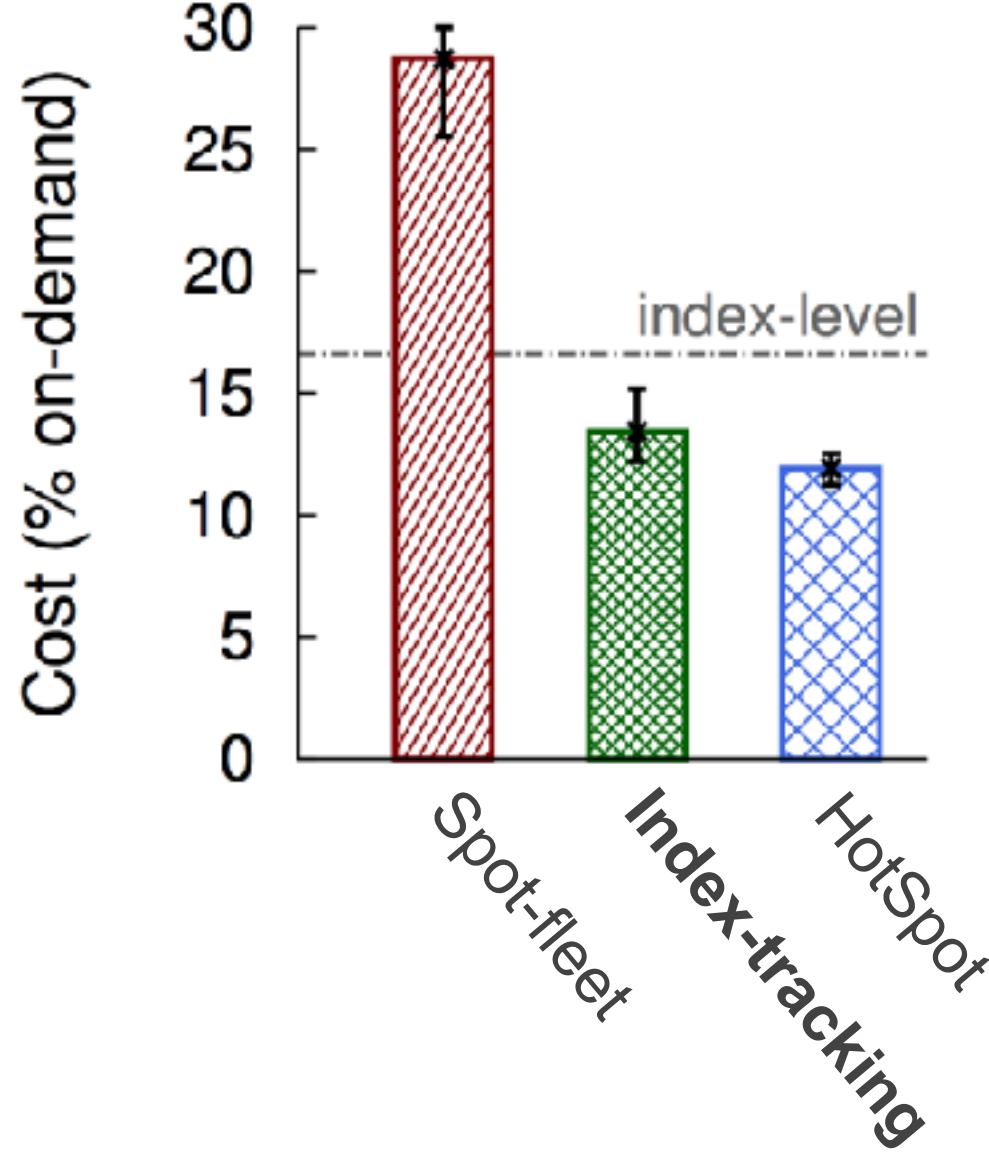
MapReduce type workload from Google traces

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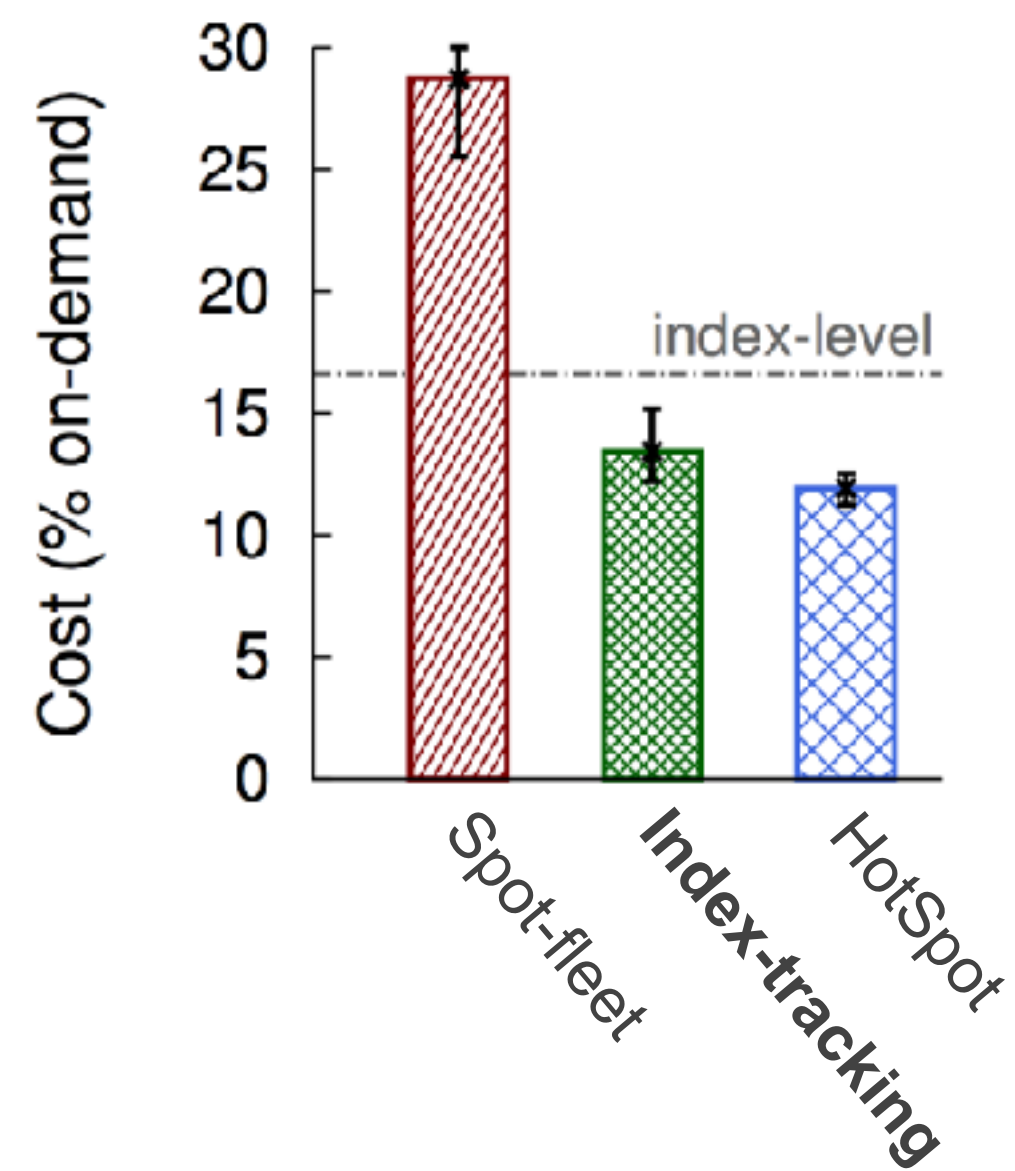
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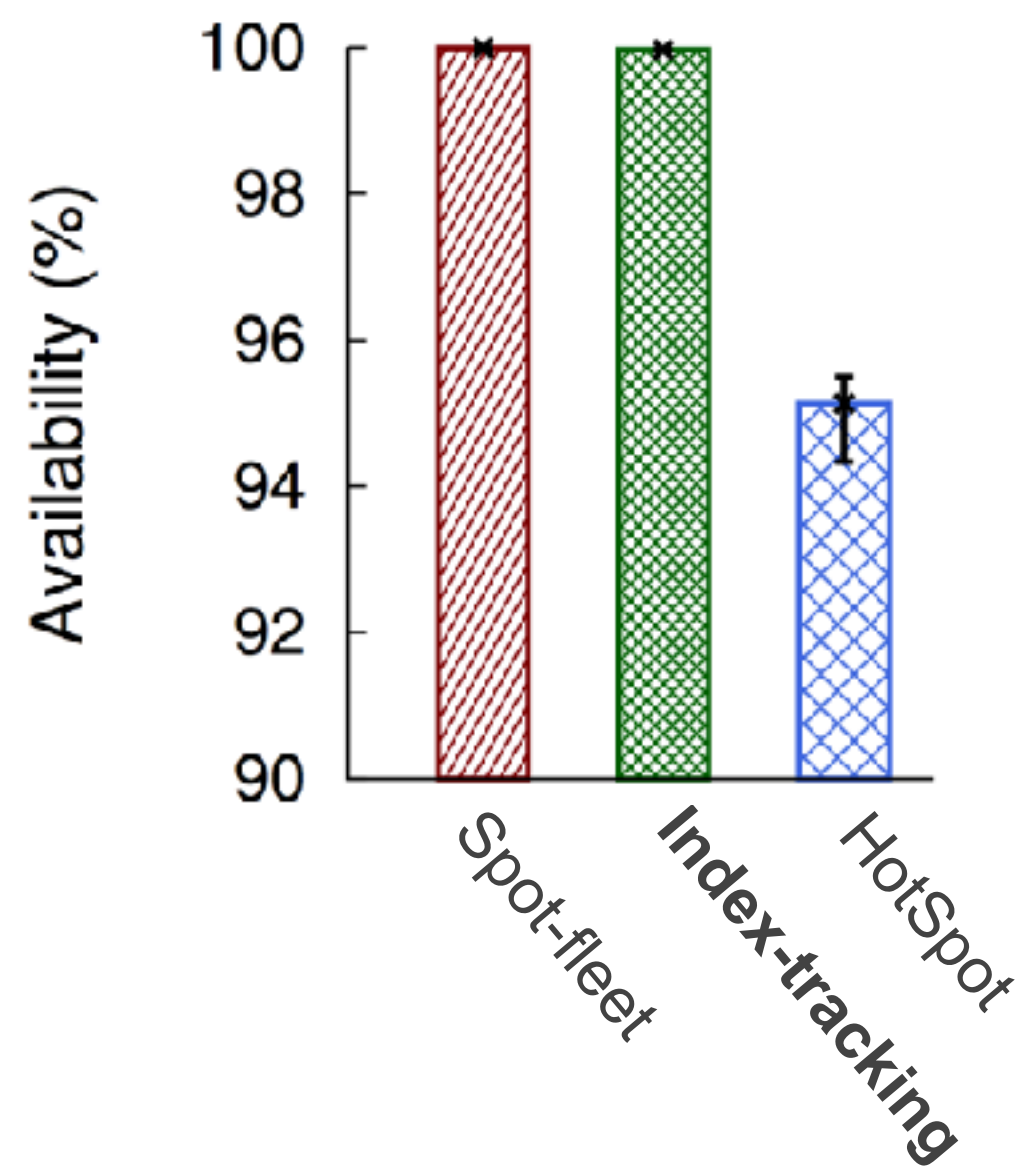
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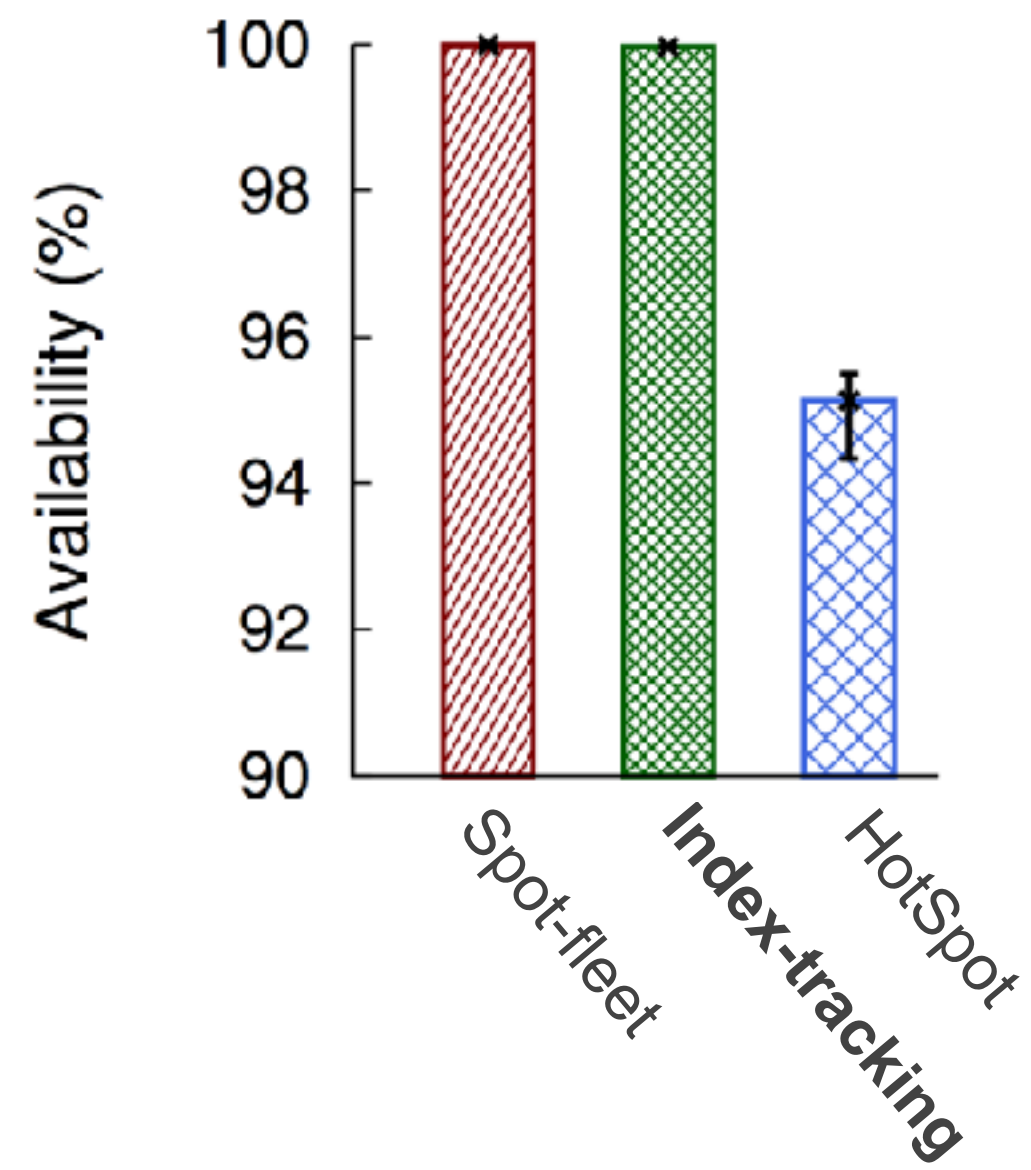
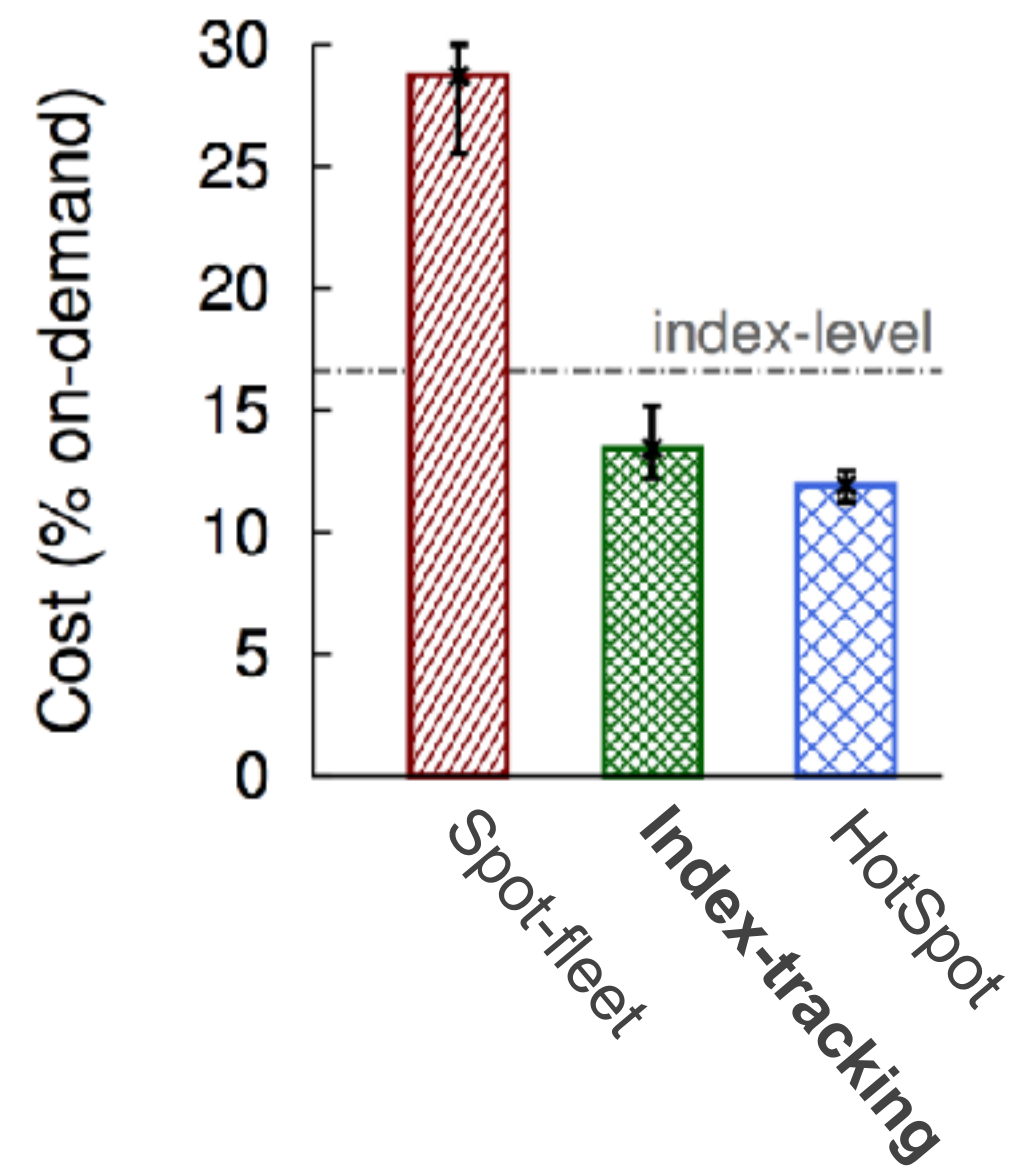
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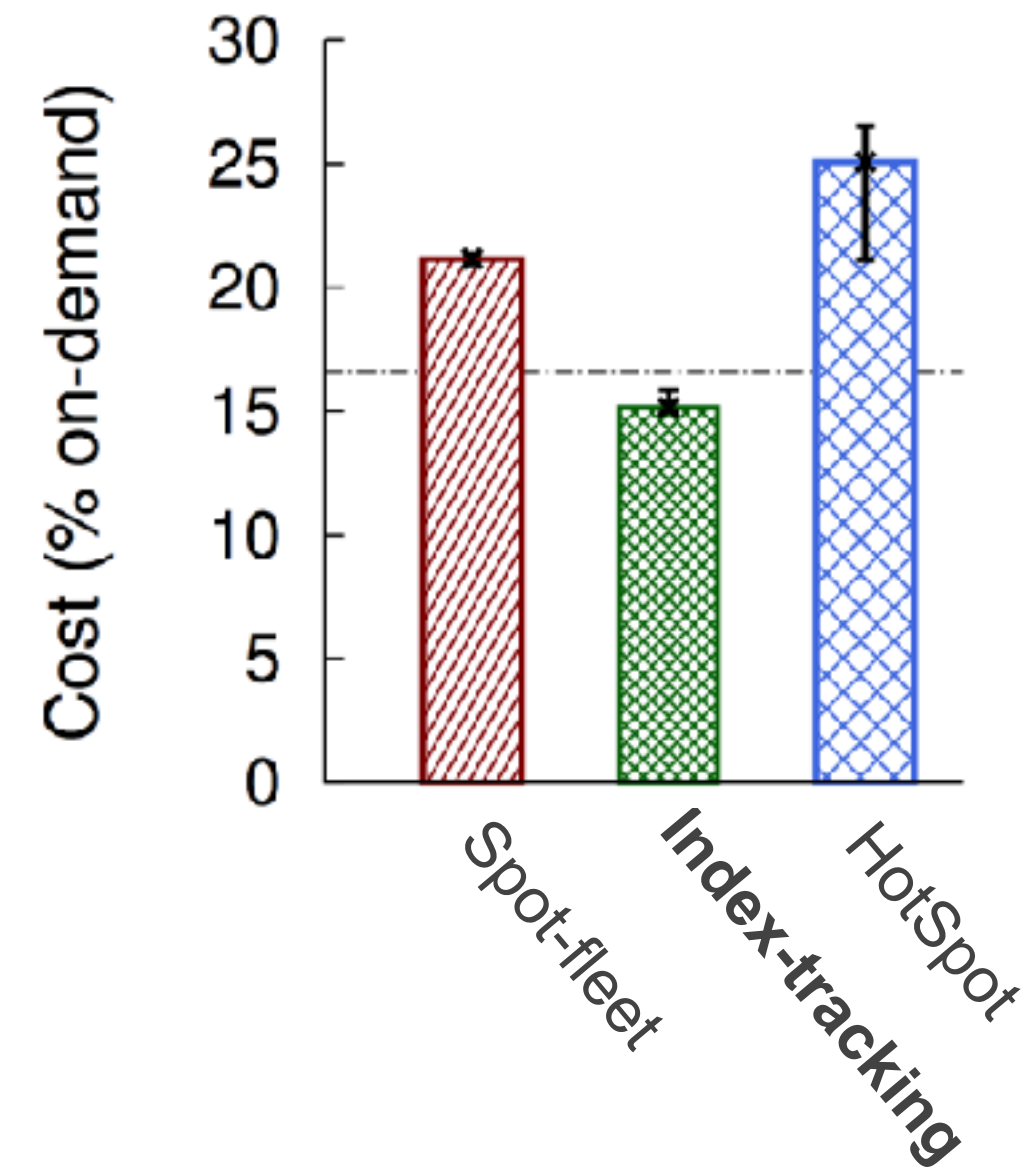
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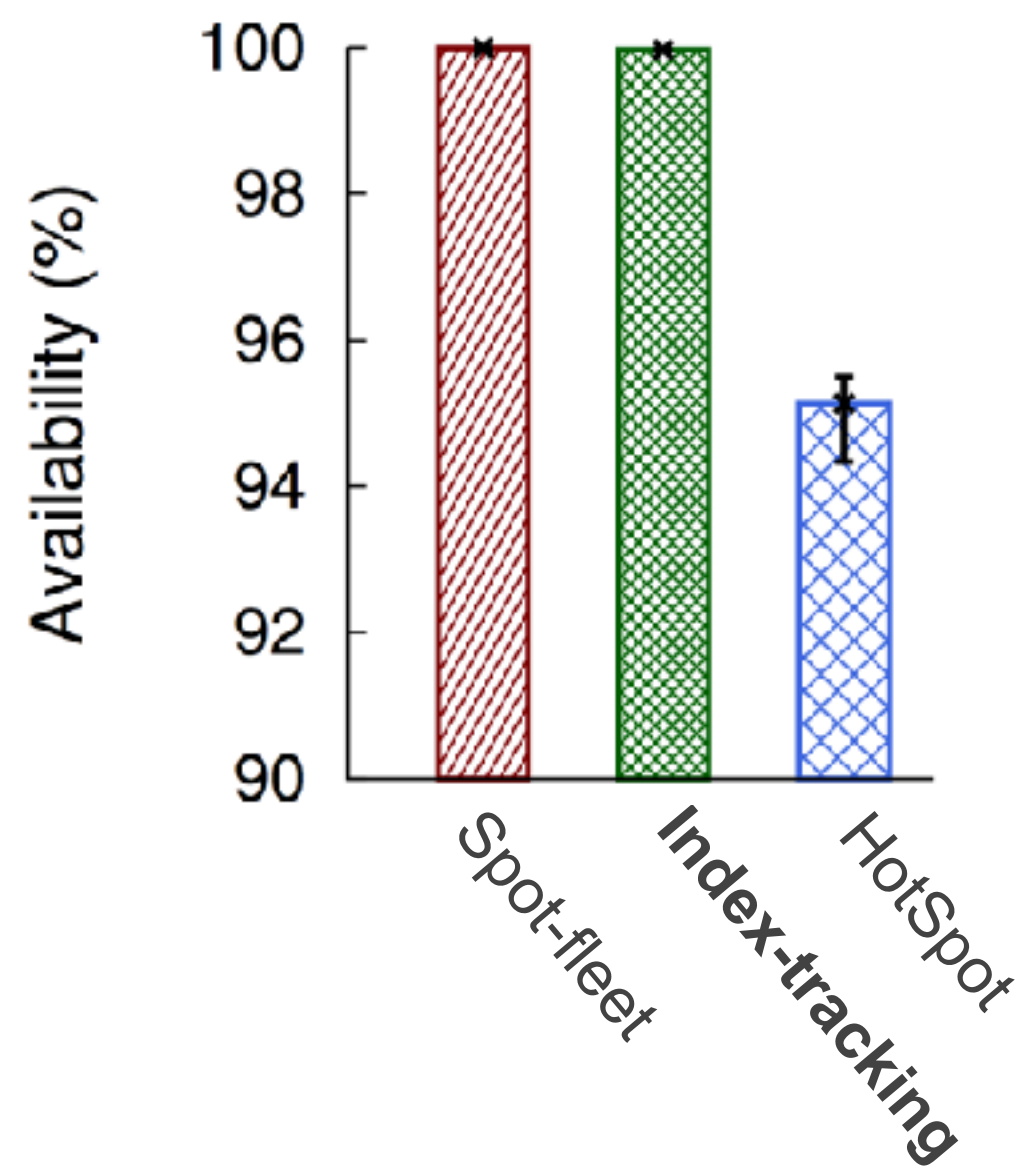
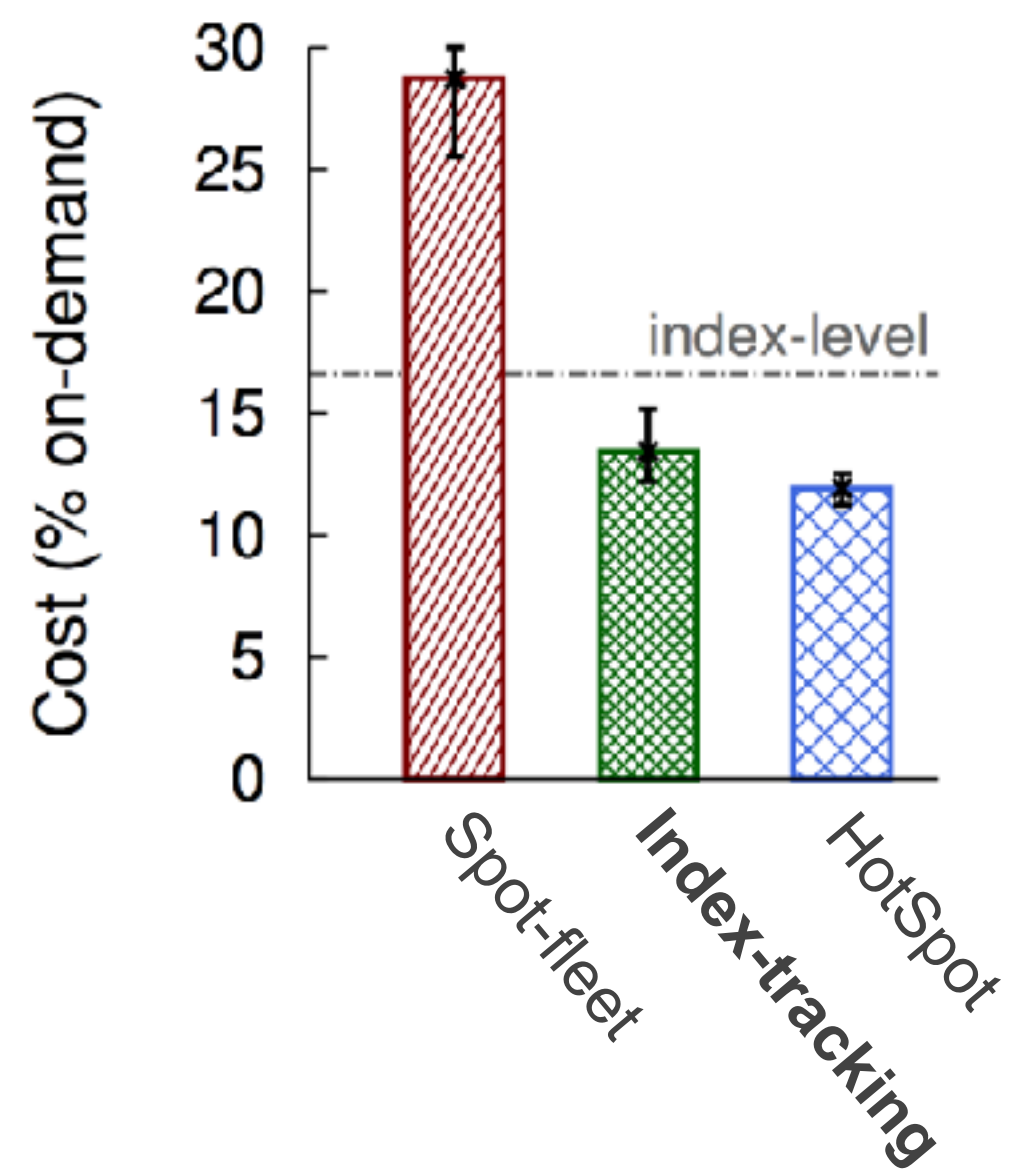
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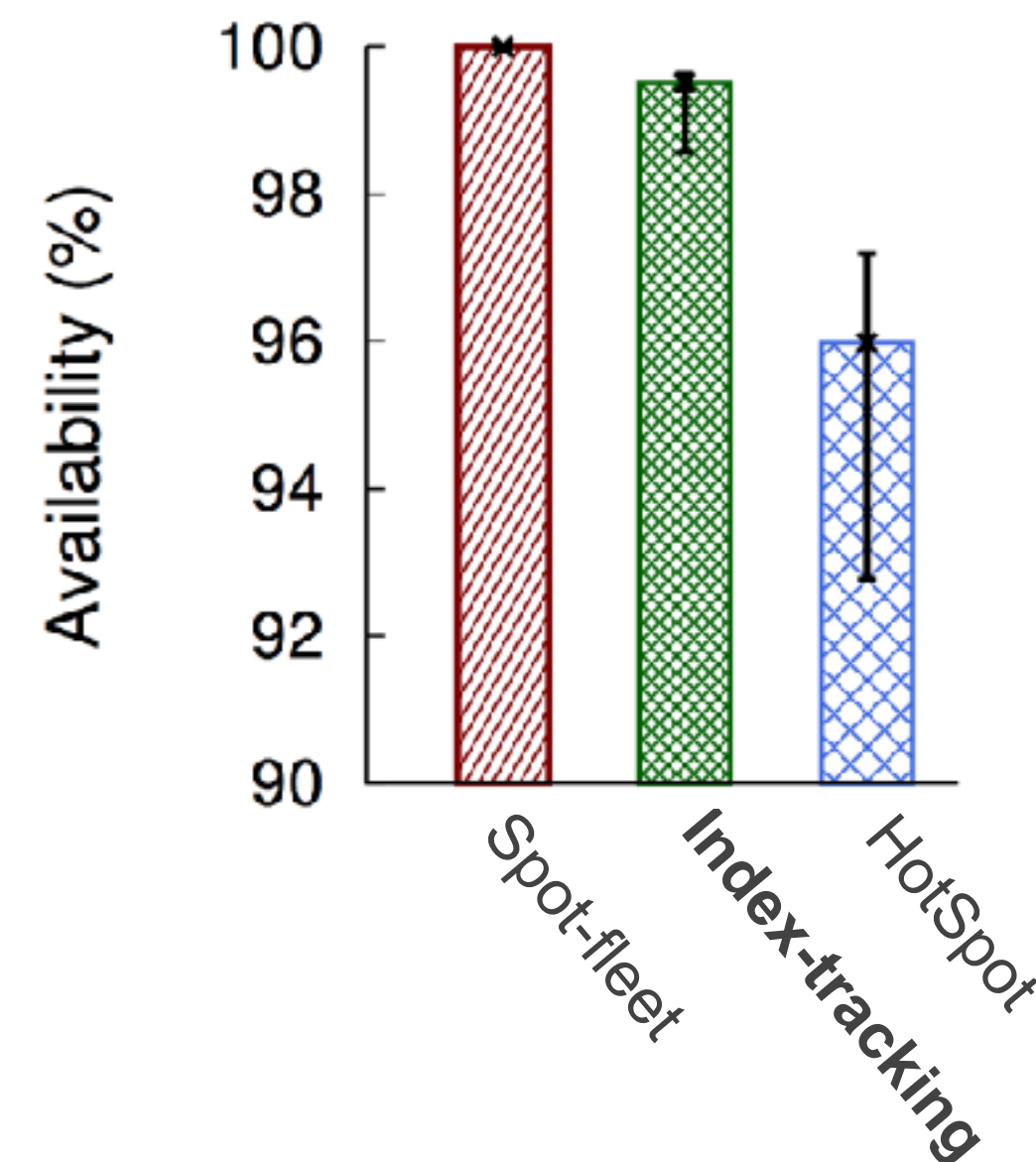
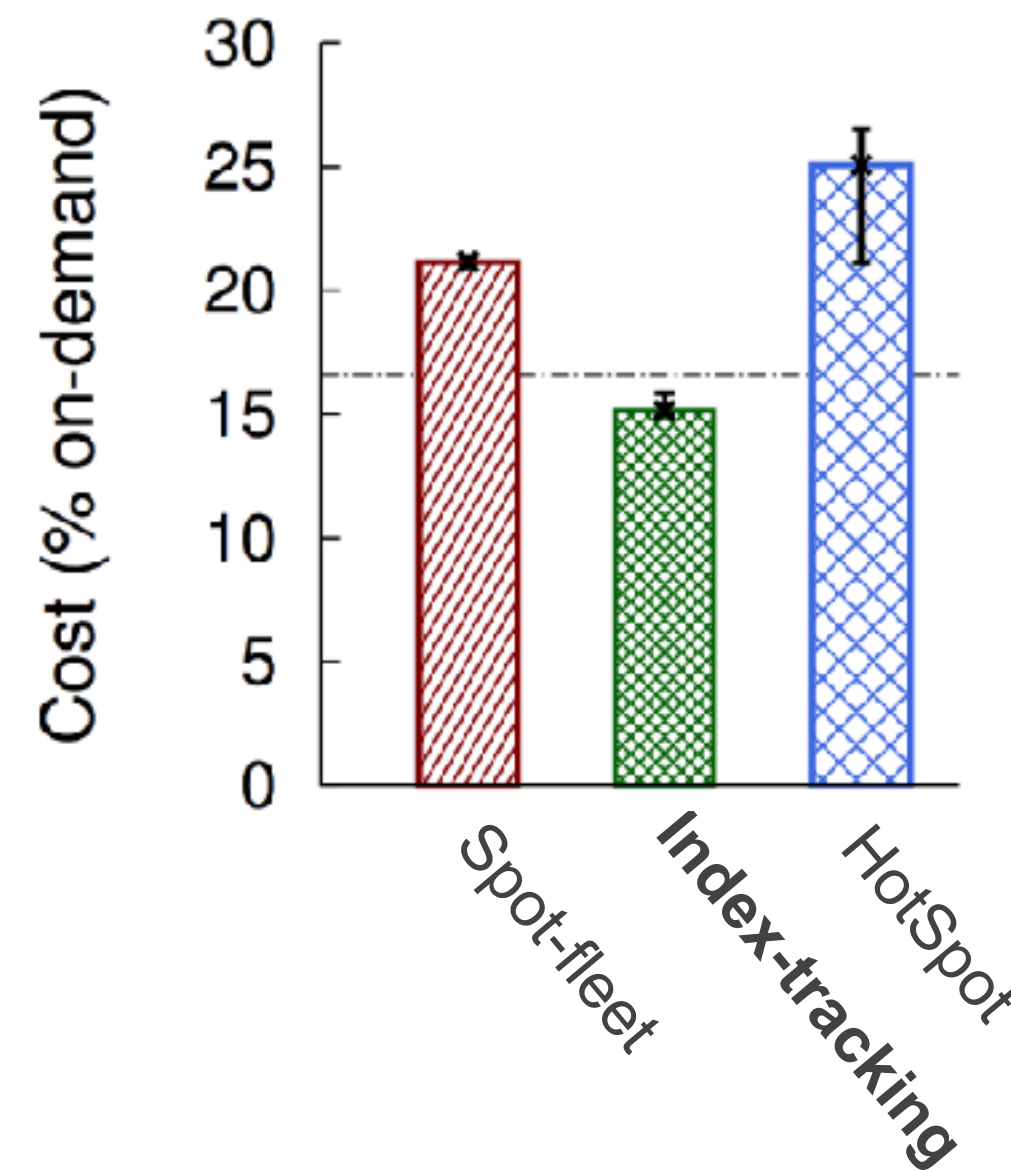
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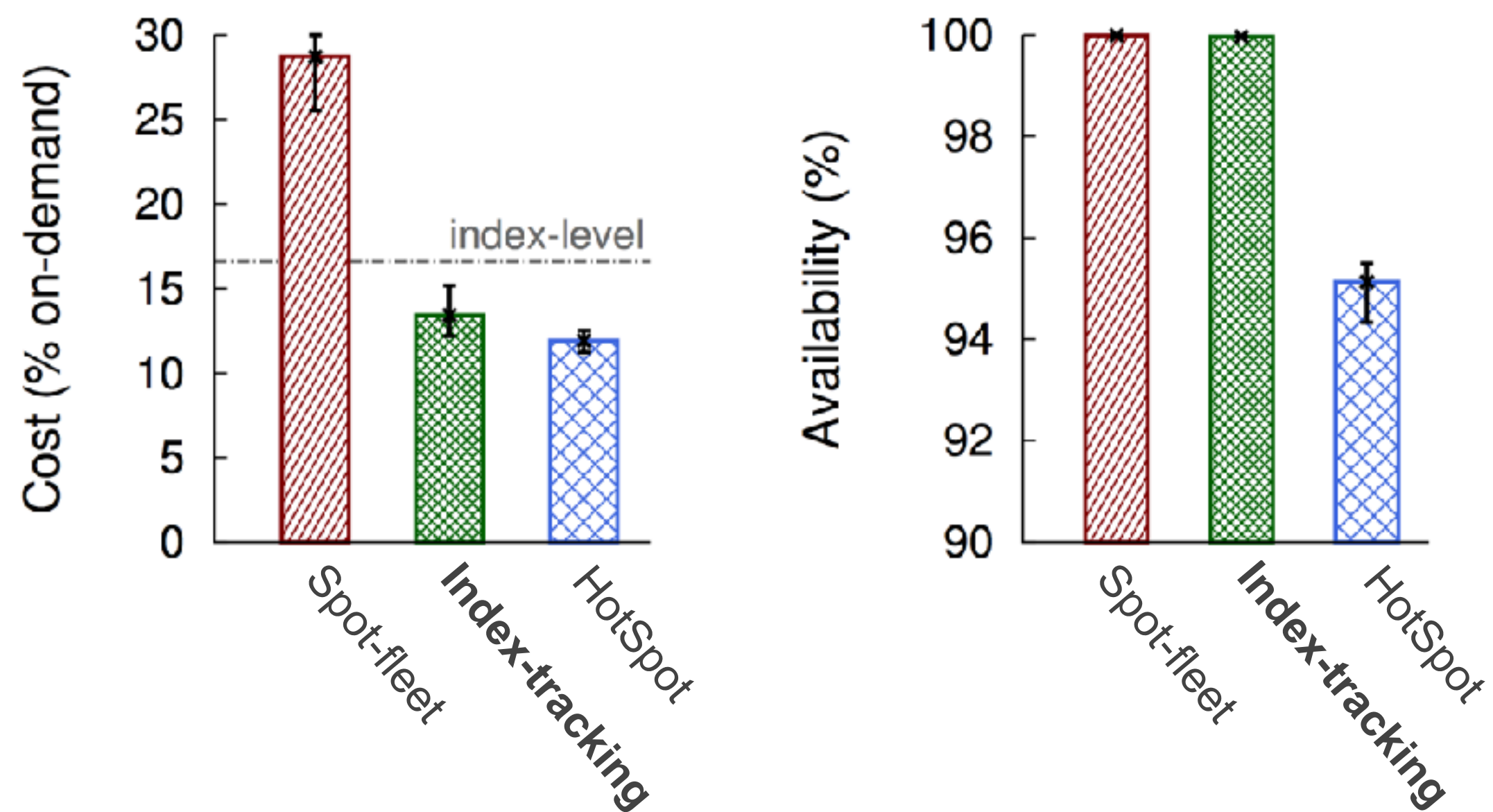
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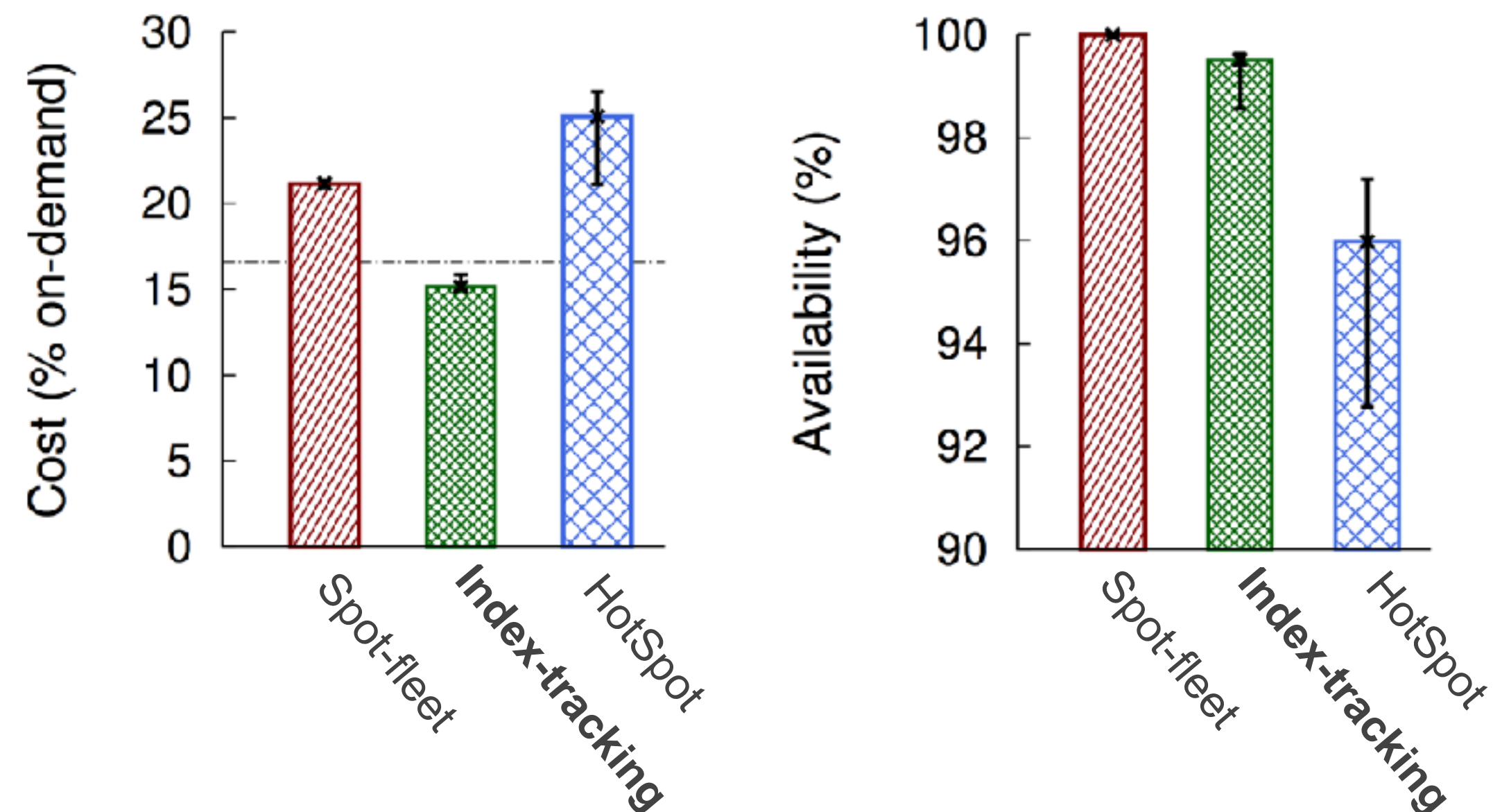
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Index-Tracking not only meets the **predicted cost-efficiency** but also achieves the **best cost-availability tradeoff** compared to other approaches.

Conclusion

Spot server markets enable inexpensive computing at scale but expose users to cost uncertainty

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Cost Uncertainty



Affects app performance
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Prior work focuses on
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Propose market-based
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Evaluations

Index-level
cost-efficiency

vs. other approaches

Achieves predictable costs
with higher availability
across applications