



Characterizing and Orchestrating VM Reservation in Geo-distributed Clouds to Improve the Resource Efficiency

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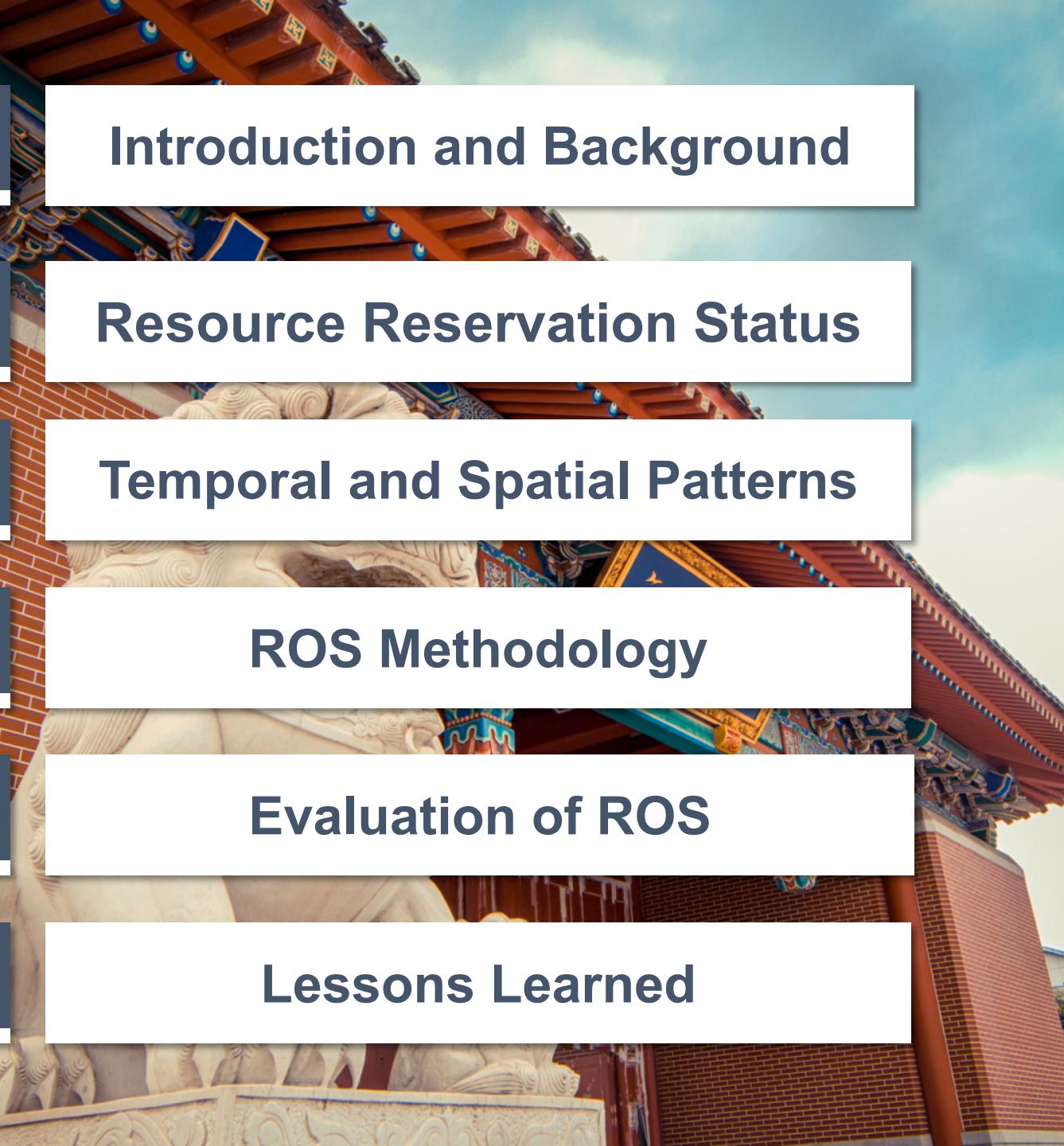
上海交通大学
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HUAWEI CLOUD



Content

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- The background of the slide shows a close-up view of a traditional Chinese building's facade, featuring intricate carvings on the eaves and vibrant red and gold colors.
- 1 Introduction and Background
 - 2 Resource Reservation Status
 - 3 Temporal and Spatial Patterns
 - 4 ROS Methodology
 - 5 Evaluation of ROS
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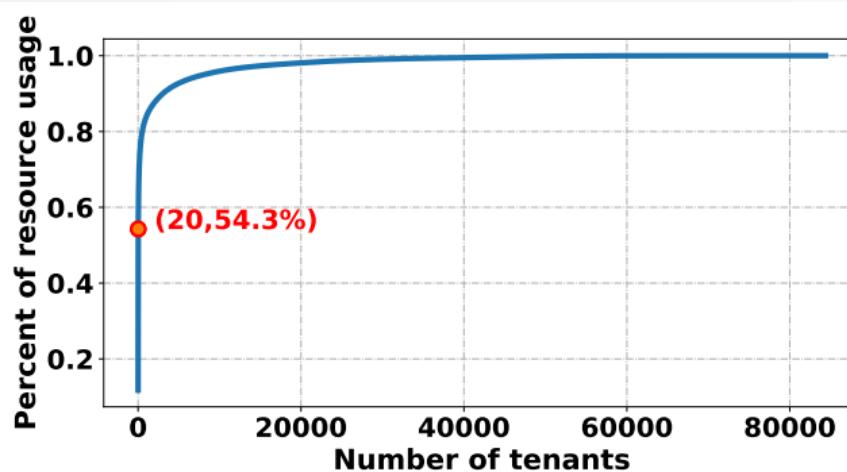
Introduction and Background



Large Tenants and Resource Reservation

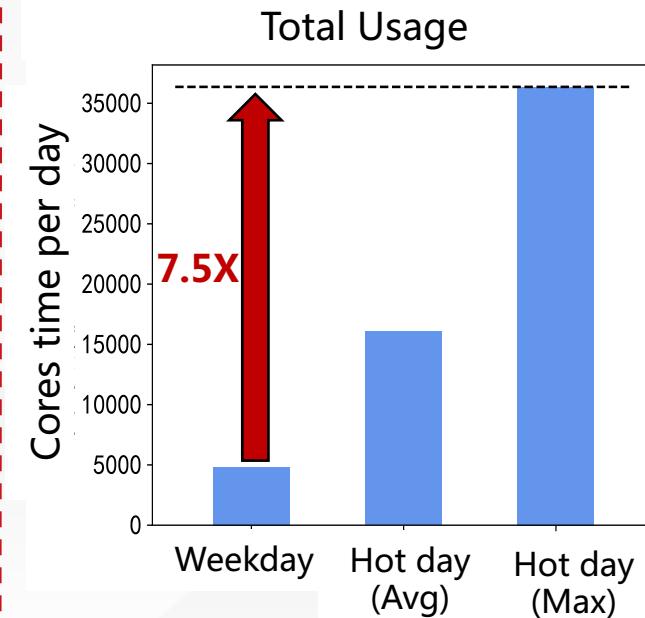


4 month data, about 85k tenants
top 20 dominates 54.5% of resource
Obvious **head effect**

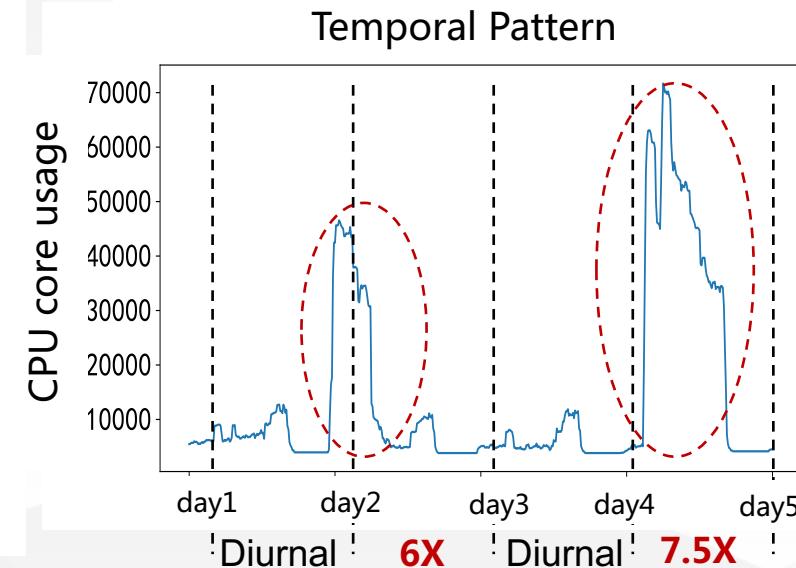


Aggregated resource usage

Bursty Resource usage can be 7.5X
Unplanned load burst
Service of **Reserved VMs** (AWS, Azure, Google, Huawei)



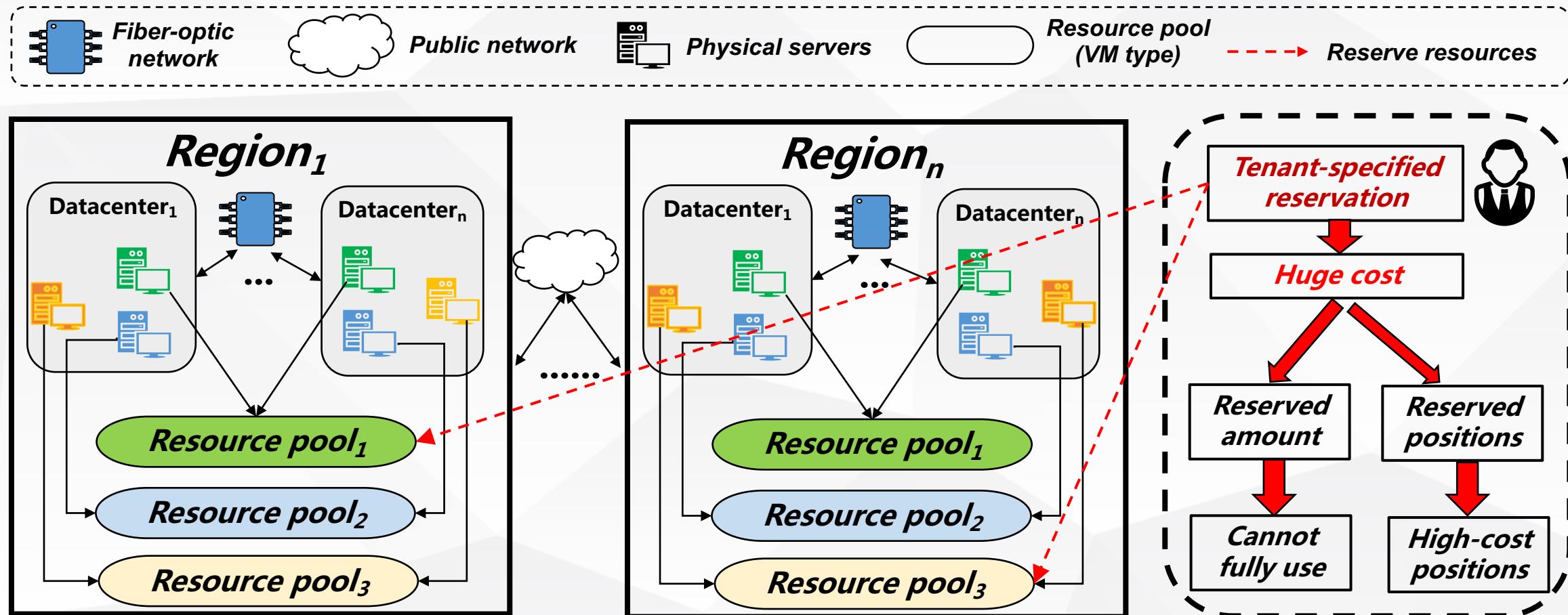
An example top tenant in Huawei Cloud



Geo-distributed Clouds and Reservation



Cloud Regions, Datacenter, Resource Pools, VM types, Tenant-specified reservation



□ Geo-distributed Trace: 2021, 4 month, top 20 tenants, 17 regions, sample every 10 mins





Resource Reservation Status

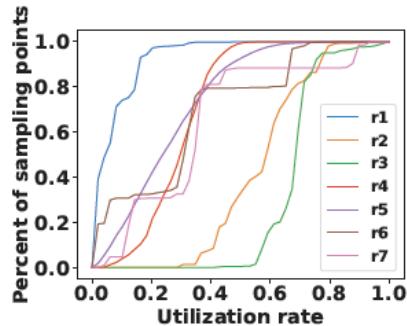


Utilization of Reserved Resources

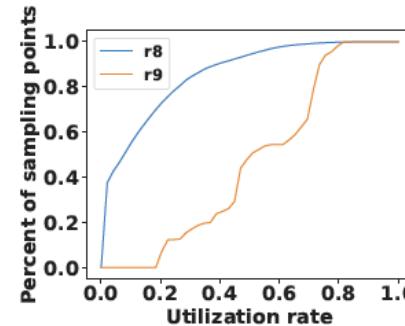


Tenant-specified resource reservation

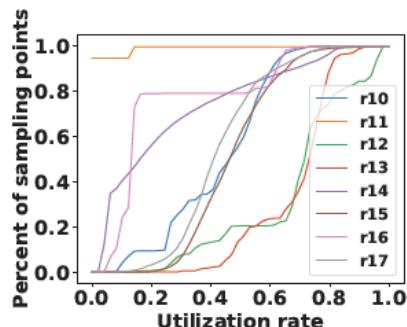
Utilization rate mainly distributed in 20%-50%, regions differ from each other



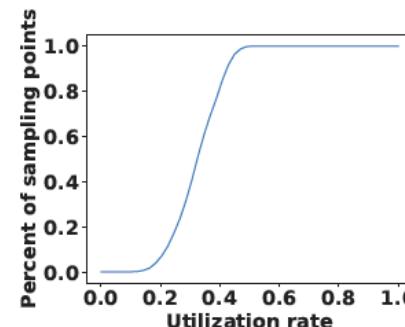
(a) Regions in *Region Set 1*.



(b) Regions in *Region Set 2*.



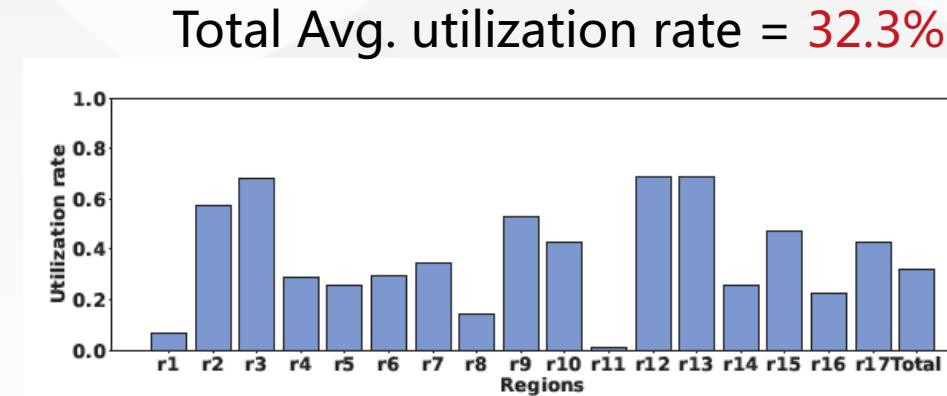
(c) Regions in *Region Set 3*.



(d) Overall resource utilization rate.

Distribution of resource reservation rate

“Tenant-specified reservation” causes huge resource waste



Total Avg. utilization rate = 32.3%

Average resource utilization rate

Note:

Region Set 1 : r1-r7

Region Set 2 : r8-r9

Region Set 3 : r10-r17



Resource Usage Distribution

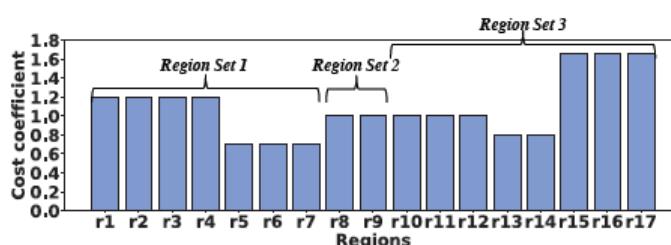


17 regions, 7 VM types, different **cost coefficients**

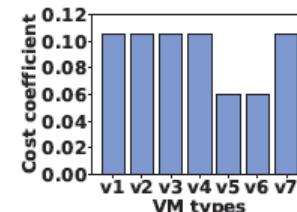
3 region set, regions in same *Region set* are **Geographically Close**, 7 VM types can **exchange**

Regions: Only **30.2%** and **14.0%** of resources distributed in low-cost.

VM types: Only **19.6%** of resource distributed in low-cost.

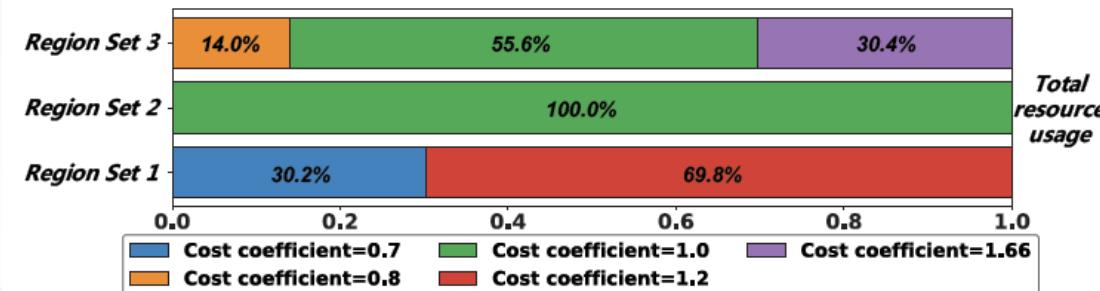


(a) Regions.

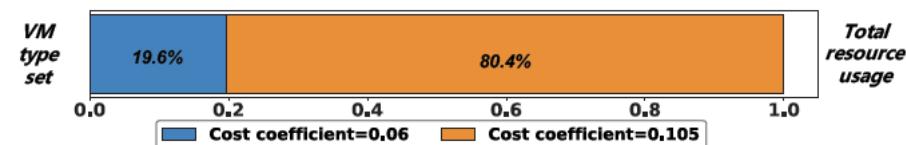


(b) VM types.

Cost coefficients of regions and VM types



Resource usage distribution of regions



Resource usage distribution of VM types

“Tenant-specified reservation” not tends to low-cost region/VM type



Main Takeaways



What can we learn from current reservation status?



- Low utilization rate, **on-demand reservation** can reduce waste;
- Not aware of region/VM type costs, potential to lead **to low cost**;
- Define tenant "**Acceptable Spatial Ranges**" to ensure SLO for tenants
- Acceptable Spatial Ranges:
 - reserved resource can **exchange** within some regions/VM types;
 - decide **positions** and **resource amount** in corresponding ranges.

“On-demand + low-cost reservation” needs to explore “temporal and spatial patterns” of large tenants



Temporal and Spatial Patterns



Temporal Usage Patterns



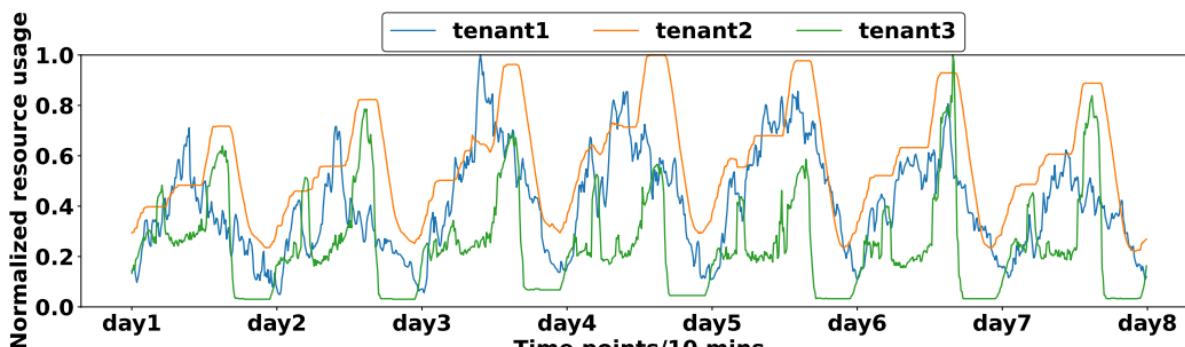
Diurnal usage pattern:

day and night mode in short and long term

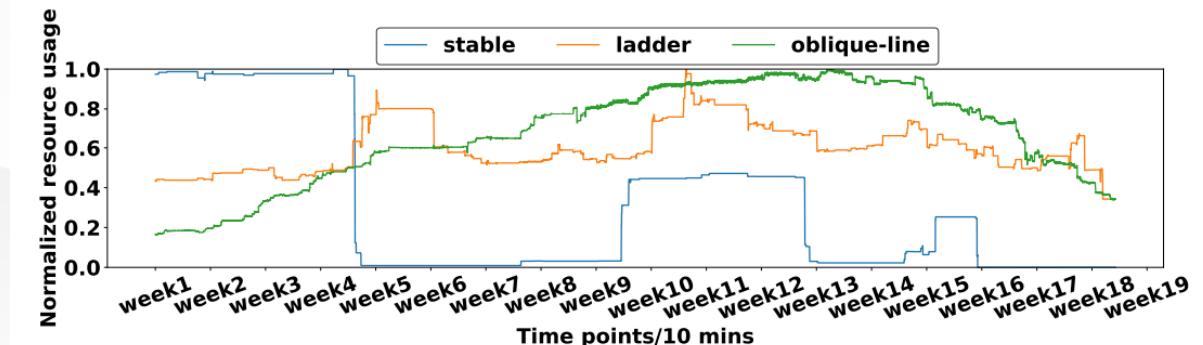
Time series prediction methods: LSTM/ARIMA

Persistent usage pattern:

stable usage in pattern in long term, (1) stable;(2)ladder;(3)oblique-line;
average values, linear regression analysis methods.



Diurnal pattern



Persistent pattern

Predictable patterns use corresponding prediction methods



Temporal Usage Patterns



Bursty usage pattern:

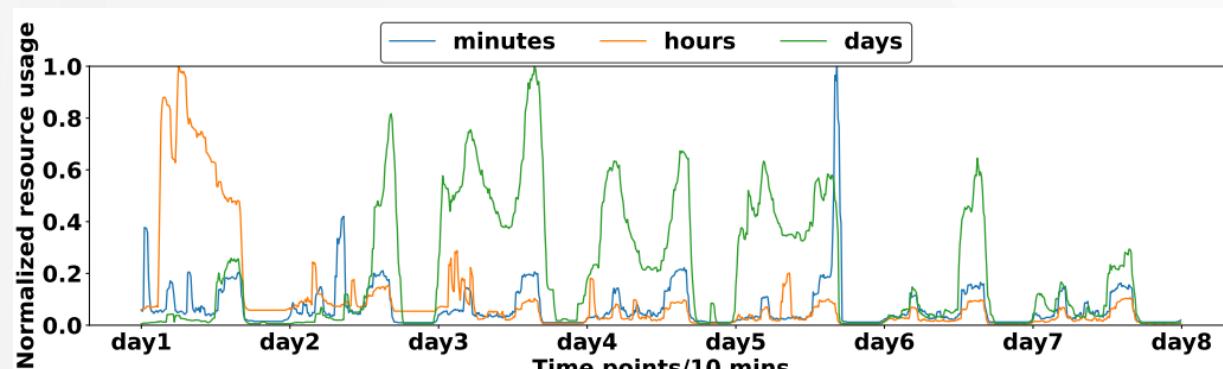
Diurnal pattern in normal time, existing **bursty** resource usage;

Bursty amount can be **3X-7X**; Bursty duration can be **minutes, hours, and days**;

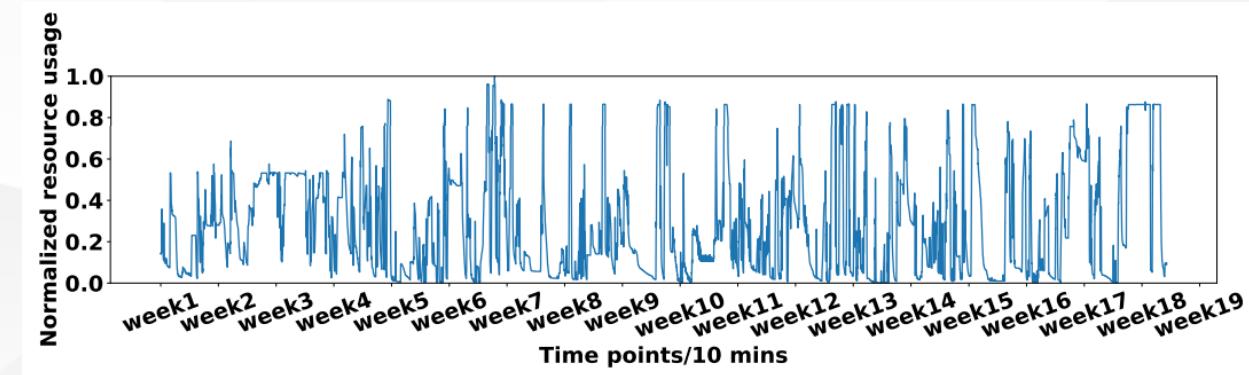
Online schedule and compensate for bursty time

Irregular usage pattern:

random in short and long term; Unpredictable, **online schedule and compensate**.



Bursty tenants



Irregular pattern

Unpredictable patterns rely on online schedule and compensate



Spatial Usage Patterns

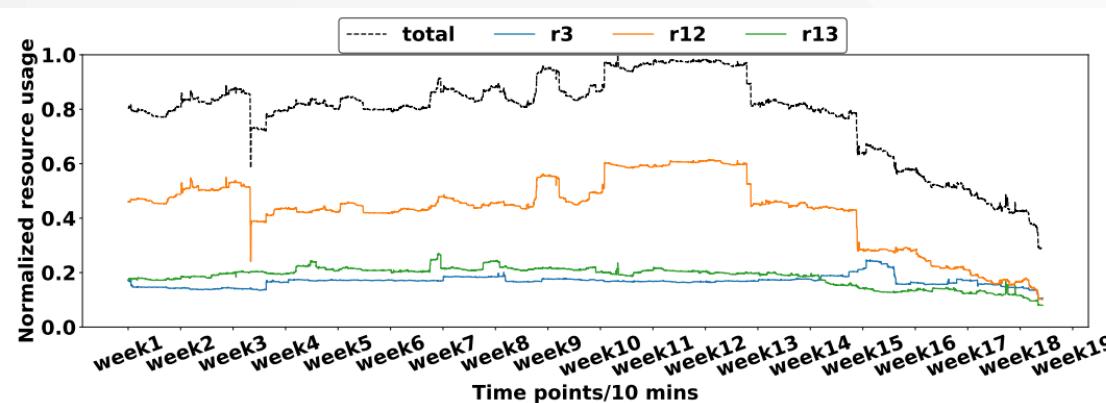


Using a single major region/VM type:

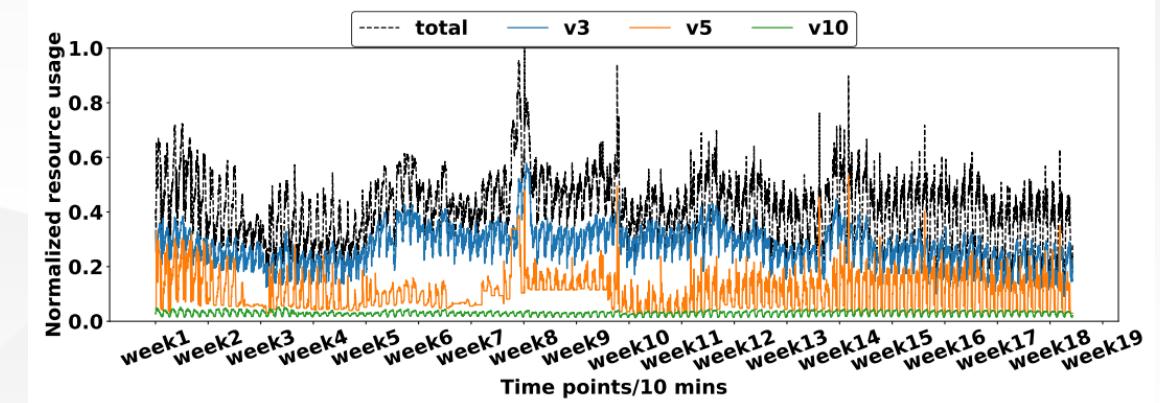
Focus on its **temporal patterns** for resource usage prediction.

Using multiple regions/VM types with stable usage division:

Stable division and **similar patterns** between regions/VM types;
“predicted temporal usage \times stable division” to predict usage on regions/VM types.



Multiple regions in stable mode



Multiple VM types in stable mode

Mainly consider temporal patterns for prediction and reservation

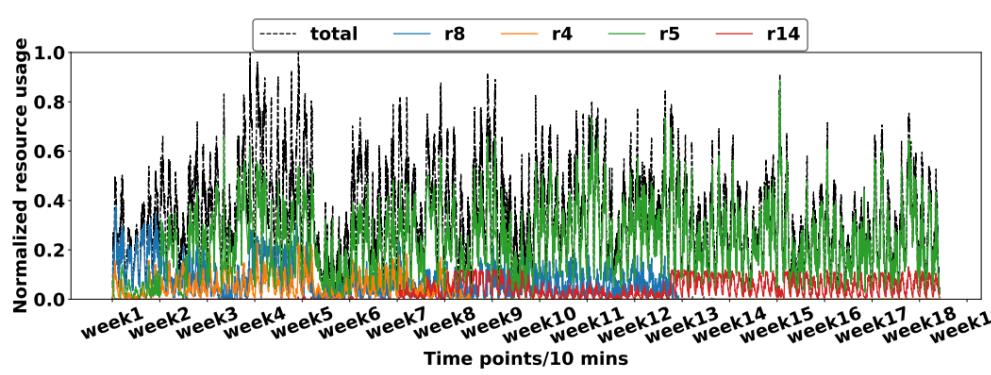
Spatial Usage Patterns



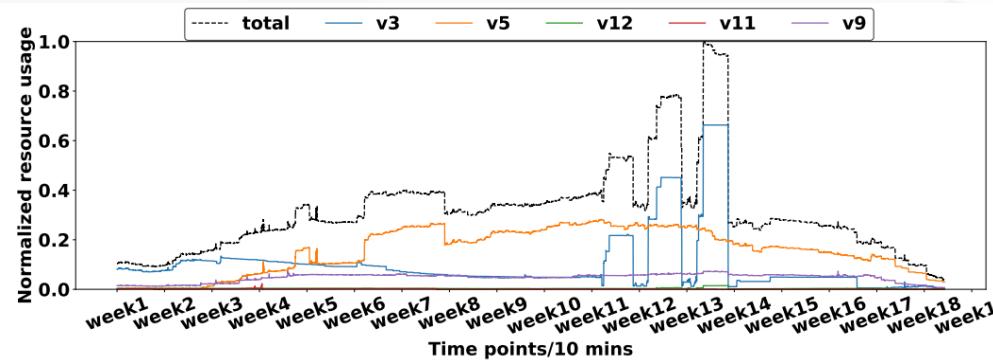
Using multiple regions/VM types with dynamic division:

Dynamic division and different patterns between regions/VM types;

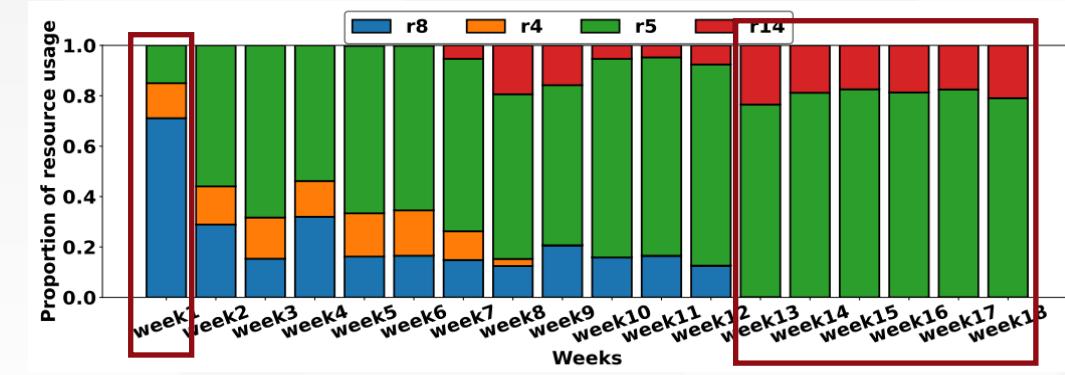
Combine tenants "Acceptable spatial ranges", different from temporal patterns.



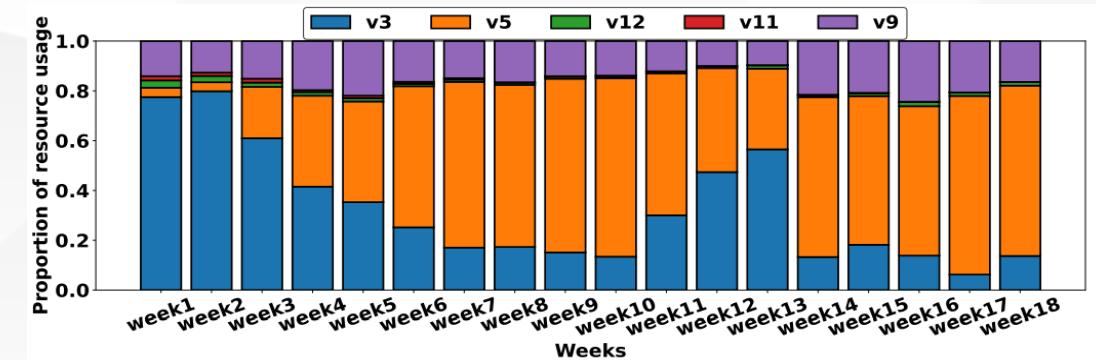
Multiple regions in dynamic mode



Multiple VM types in dynamic mode



Breakdown of dynamic regions



Breakdown of dynamic VM types

Prediction and reservation on tenants "Acceptable spatial ranges"

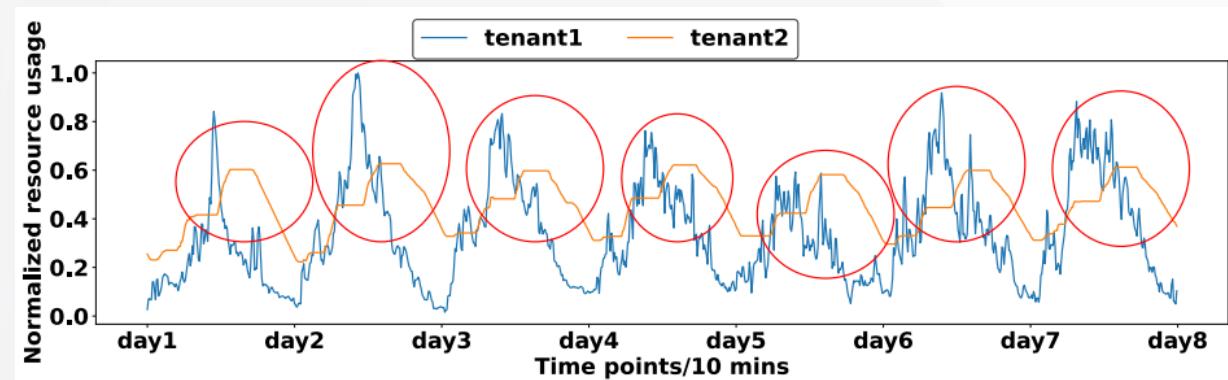
Potentials to Improve Resource Reservation Efficiency

Temporal potential:

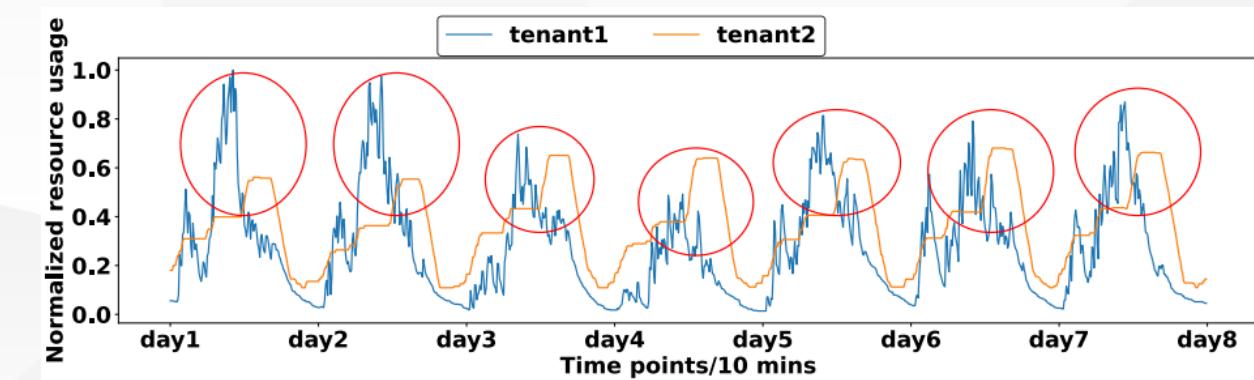
Diurnal tenants have **complementary** usage, temporal peak shaving reduces reservation;
The example can reduce **8.4%** of resource reservation.

Spatial potential:

Tenants on different spatial (region/VM type) have **complementary** usage;
Further reduce resource reservation, the example can reduce **13.2%**.



Temporal Potential of Peak Shaving
(tenant1 and tenant2 on region $r4$ VM type $v5$)



Spatial Potential of Peak Shaving
(tenant1 on $r14$ $v5$ & tenant2 on $r10$ $v1$)

Temporal + Spatial peak shaving can reduce resource reservation



Main Takeaways



What can we learn from temporal and spatial patterns?

- Predictable tenants needs **corresponding** prediction methods;
- Unpredictable tenants needs online schedule and **compensate**;
- Tenants have obvious **spatial patterns**:
 - (1) **different usage patterns** on regions/VM types;
 - (2) **dynamic usage division** on different regions/VM types.
- Predict/Reserve on tenants “Acceptable spatial ranges” .
- Temporal and Spatial potentials of **peak shaving**.



ROS: (1) Prediction; (2) Orchestration; (3) Schedule and Compensation





ROS Methodology

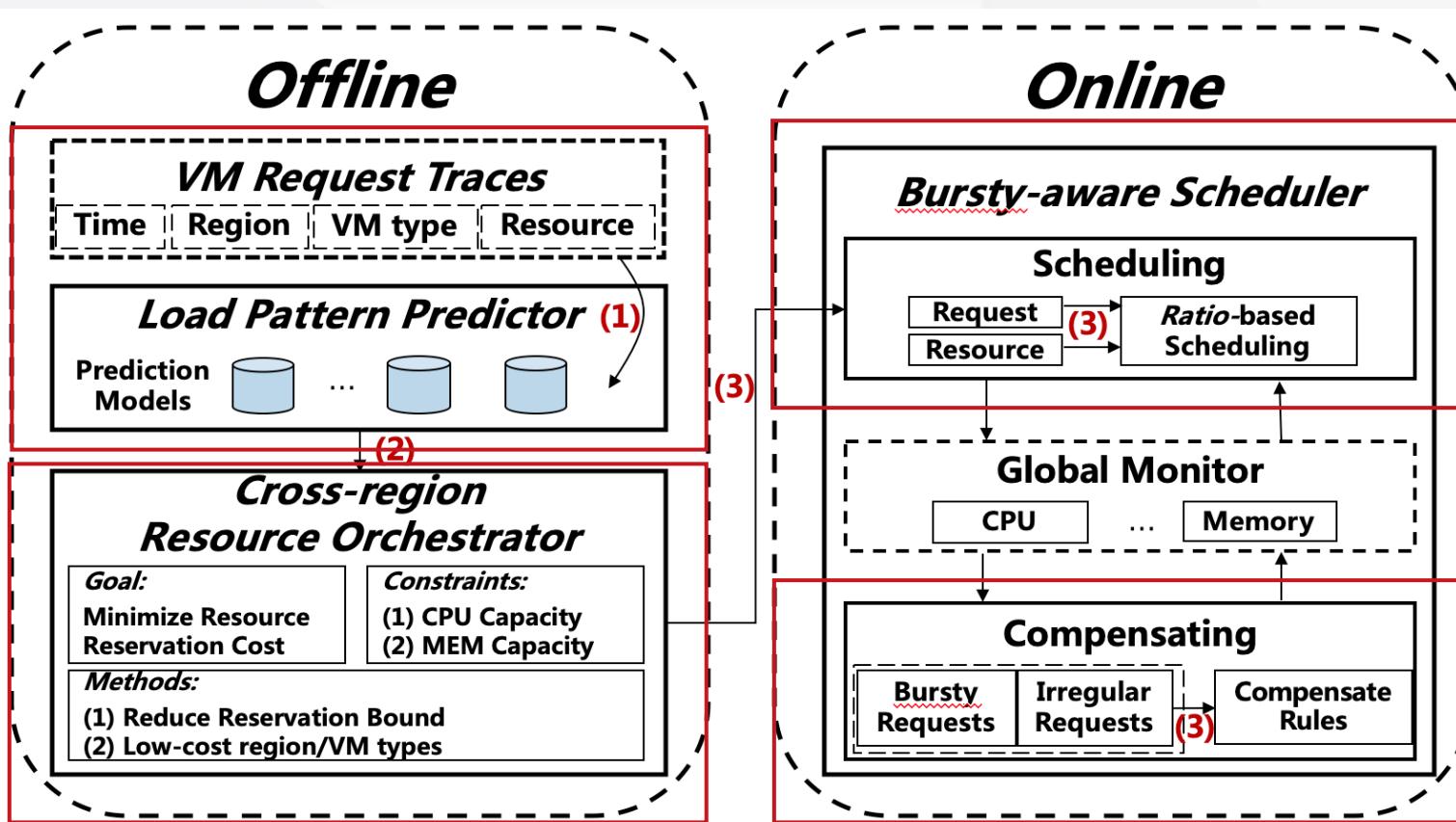


System Overview



ROS: Resource orchestration and VM scheduling policy

- **Load Pattern Predictor:** identify patterns of tenants, and predict resource usage
- **Cross-region Resource Orchestrator:** Orchestrate resource reservation to optimize the total cost
- **Bursty-aware Scheduler:** Scheduling VMs, and compensating for bursty with **cost-minimized rules**





Cross-region Resource Orchestration



Optimization model for orchestration

- Input: Predicted resource usage of tenants
- Output: orchestrate matrix *Ratio, reserved lines* of regions
- Peak shaving between tenants and cost coefficients of resource types

Note:

resource type: one VM type on one region

Cost coefficients Temporal and Spatial peak shaving

$$\text{Minimize: } C_{total} = \sum_{j=1}^n C_j \times \max\left(\sum_{i=1}^m (f_i(t) \times \text{ratio}_{ij})|_{t=0}^{24h}\right)$$

$$\text{s.t. } \begin{cases} \forall i, \quad \sum_{j=1}^n \text{ratio}_{ij} = 1 & i = 1 \cdots m \\ \forall j, \quad \max\left(\sum_{i=1}^m \text{CPU}_{ij}(t)|_{t=0}^{24h}\right) < \text{Cap}_{CPU}^j & j = 1 \cdots n \\ \forall j, \quad \max\left(\sum_{i=1}^m \text{MEM}_{ij}(t)|_{t=0}^{24h}\right) < \text{Cap}_{MEM}^j & j = 1 \cdots n \end{cases}$$

Goal: Optimize cost and reservation

- ① Satisfy SLO and resource demand
② Capacity of region/VM type





Evaluation of ROS





Evaluation Setup



Trace dataset:

- VM requests of large tenants during 2021.04-07

Spatial ranges:

17 regions divide into 3 region sets based on geographical positions

- *Region Set 1* : 7 regions
- *Region Set 2* : 2 regions
- *Region Set 3* : 8 regions

Typical VM types :

- v1, v2, v3, v4, v5, v6, v7
- Tenant workloads can be switched inside the 7 VM types

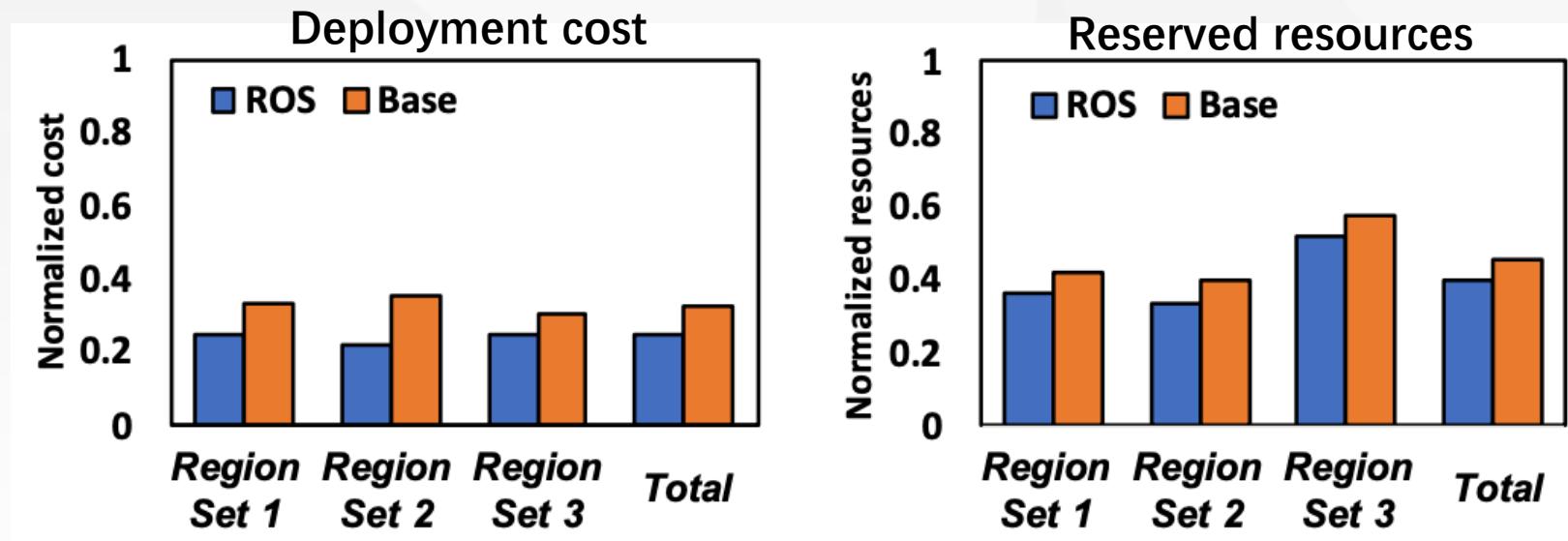
Baselines:

- Tenant-specified reservation
- Geo-distributed capacity planning (Narayanan et al.)



Reducing the Deployment Cost

- Compared with “**Tenant-specified**” strategy:
ROS can reduce **75.4% of deployment cost** and **60.1% of resource reservation**.
- Compared with “**Geo-distributed capacity planning**”:
ROS can reduce **24.7% of deployment cost** and **12.2% of resource reservation**.



Overall normalized cost and reserved resources

“Low-cost regions/VM types” and “Complementary patterns”



Lessons Learned





Lessons Learned



- Tenants do not understand their resource demand, prediction and low-cost region/VM type advices can reduce cost and improve utilization;
- It is better to consider the resource reservation from the cloud provider side, which can orchestrate multiple tenants together ;
- Tenants accept some low-cost regions, so cloud providers can build DCs on them, and improve utilization through adaptive orchestrator;
- Since tenants have tendency and tolerance to VM types, cloud providers can provide VM types accordingly to further improve resource efficiency





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The end

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