

Toward Resource-Efficient Cloud Systems: Avoiding Over-Provisioning in Demand-Prediction Based Resource Provisioning

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Outline

- 1 Introduction
- 2 System Design
- 3 Experiment
- 4 Conclusion

1 Introduction

2 System Design

3 Experiment

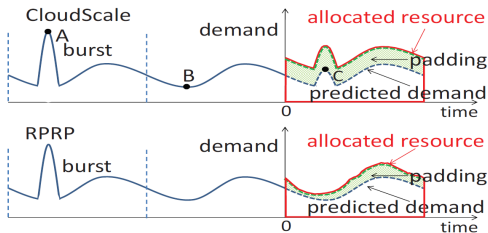
4 Conclusion

Introduction

- In cloud systems, cloud providers abstract resources in physical machines into virtual machines and sell them to the tenants.
- To ensure resource provisioning for guaranteeing SLOs¹, clouds can use *demand-prediction based resource provisioning schemes*.
- Achieving the tradeoff between the penalties associated with *SLO violations* and *high resource utilization* requires an accurate demand prediction methodology.

¹SLO: Service Level Objectives

Previous Work - CloudScale



- CloudScale predicts the demand at a time period based on a historical record.
- Padding: using the high-frequency spectrum or the average of the latest prediction error.
- Online Adaptive: to handle underestimation, raising the resource allocation by $\alpha > 1$ until an error is corrected.

RPRP¹

- RPRP excludes bursts in demand prediction and specifically handles bursts to avoid resource over-provisioning.
- Algorithm
 - *burst-exclusive prediction algorithm*
 - *load-dependent padding algorithm*
 - *responsive padding algorithm*
- Algorithm 1 and 2 aim to exclude bursts, and algorithm 3 aims to handle bursts.

¹RPRP: Resource-efficient Predictive Resource Provisioning system

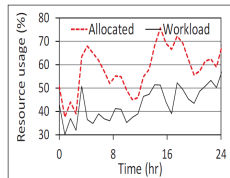
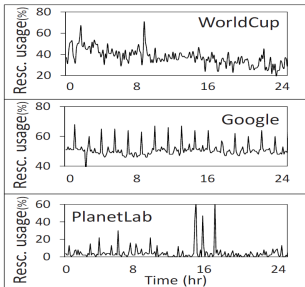
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Objective

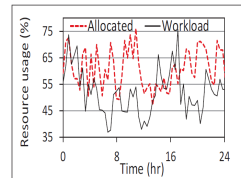
- Denote a VM's records:
 - *workload demand*: $D = \{d_{t_1}, \dots, d_{t_i}, \dots, d_{t_N}\}$
 - *allocated resource*: $A = \{a_{t_1}, \dots, a_{t_i}, \dots, a_{t_N}\}$
 - *utilized resource*: $U = \{u_{t_1}, \dots, u_{t_i}, \dots, u_{t_N}\}$
 - *resource capacity*: C
- And from the historical records, we have:
 - *predict demand*: $P = \{p_{t_{N+1}}, p_{t_{N+2}}, \dots, p_{t_{N+T}}\}$
 - *allocated resource*: $A = \{a_{t_{N+1}}, a_{t_{N+2}}, \dots, a_{t_{N+T}}\}$
- Goal: determine allocated resource A such that
 - $d_{t_i} \leq a_{t_i} \leq C$
 - and meanwhile to minimize $a_{t_i} - d_{t_i}, \forall t_i > t_N$

Algo.1: Burst-exclusive Prediction

- Trace analysis and CloudScale prediction + padding.



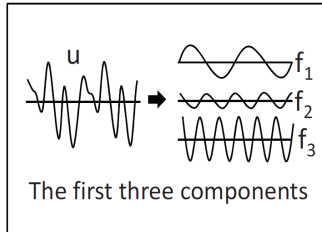
(a) Low burst density.



(b) High burst density.

Algo.1: Burst-exclusive Prediction

- RPRP relies on FFT to exclude the burst.
- FFT is applicable for predicting workload demand in repeated periodic patterns P based on the historical utilization series U .



Algo.2: Load-dependent Padding

Algo.3: Responsive Padding

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Experiment

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Conclusion

Future Work