



Dynamically Adjusting Scale of a Kubernetes Cluster Under QoS Guarantee

Presenter: Li Lu

Qiang Wu, Jiadi Yu, Li Lu, Shiyou Qian, Guangtao Xue
Shanghai Jiao Tong University



- Huge electricity consumption
 - Data centers consume approximately **1.12%** of all electricity worldwide
 - A half of the operational expenses within a data center are consumed by the electricity cost
- Billing mechanism
 - Many cloud providers, such as Amazon, gradually support resource provisioning and billing in second manner
- Low cluster resource utilization
 - Cluster is generally designed to handle peak loads
 - During ordinary times, the load of a server is less than **50%** of peak and the CPU utilization of a server rarely goes beyond **40%**





Our work

- Target to widely-deployed web applications
- Find out a threshold of resource utilization
 - Guarantee QoS in a Kubernetes cluster
 - Determine the time when to scale up the cluster
- Design a system to scale up or down the cluster
 - Guarantee quality of service
 - Improve the cluster resource utilization

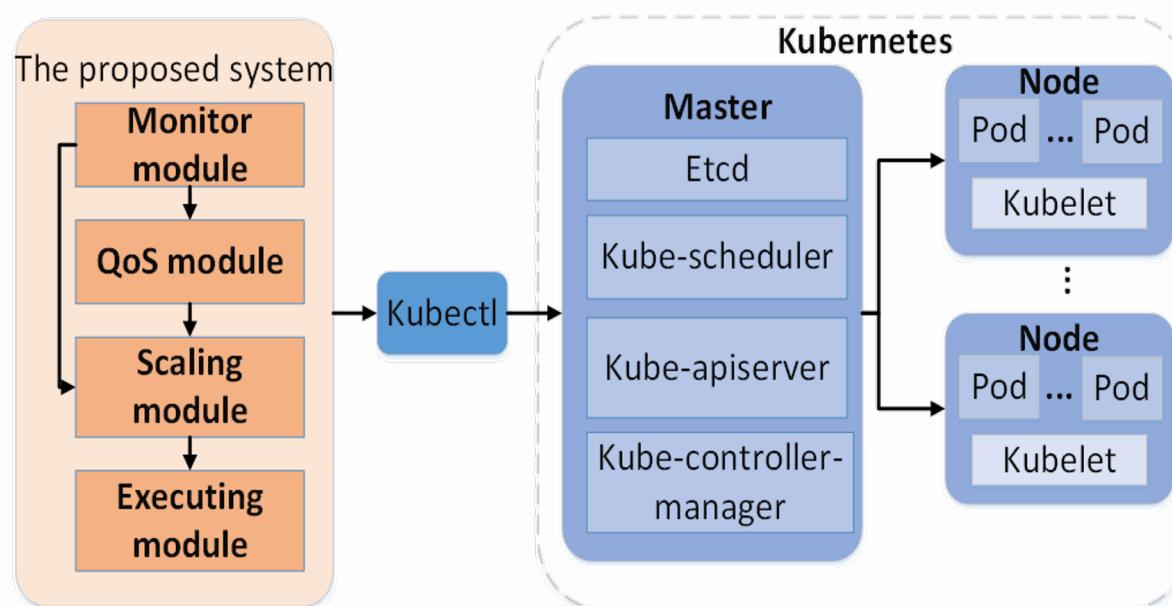


Outline

- **System design**
- **Evaluation**
- **Conclusion**

Overview

- Our system adopts a Monitor-Analyze-Plan-Execute (MAPE) model, include four modules:
 - Monitor module
 - QoS module
 - Scaling module
 - Executing module



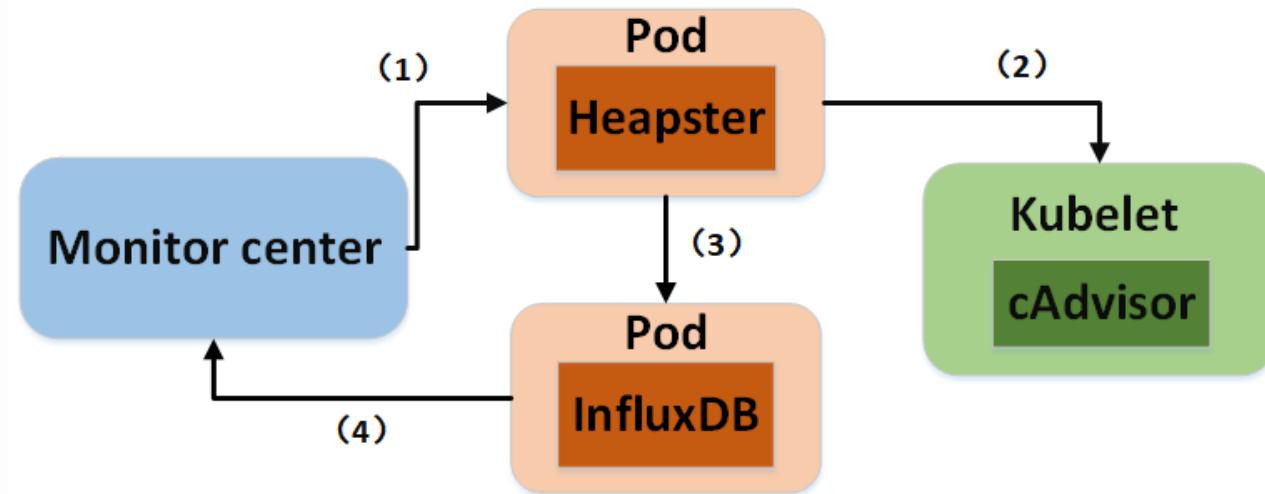
Monitor Module

➤ Goal:

Monitor module is used to monitor CPU utilization of a whole Kubernetes cluster.

➤ Workflow:

Monitor center --> Heapster --> cAdvisor --> Heapster --> InfluxDB --> monitor center





➤ Initialized Parts

- Run once to obtain the relationship between QoS and CPU utilization

➤ Goal:

- Obtain a proper threshold of CPU utilization
- Guarantee quality of service

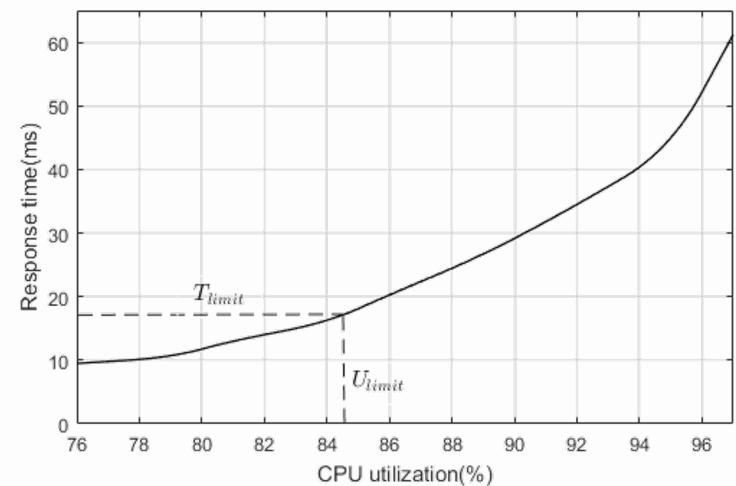
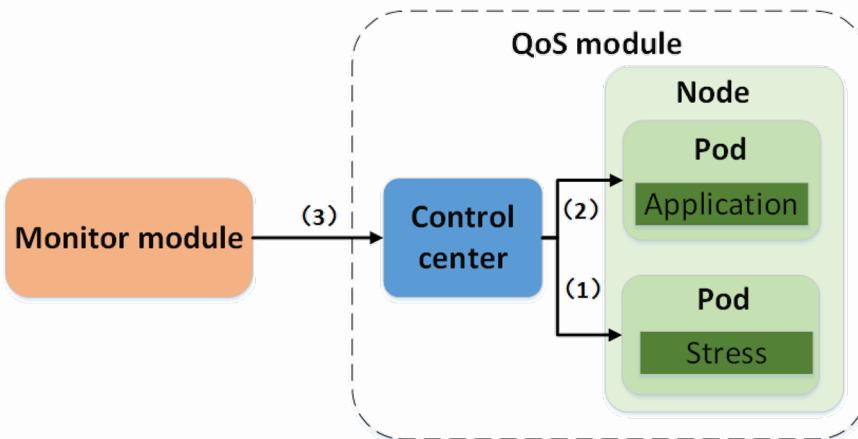
➤ Metrics of QoS: response time



➤ Workflow:

- Step1: The control center sends a HTTP request to the application, and then receives the response to calculate the response time.
- Step2: The control center gets CPU utilization from monitor module.
- Step3: The control center changes CPU utilization of the server, and then rerun step1;

➤ Thus, we get the relationship:





- The upper bound of response time:

$$T_{limit} = \alpha \times T_{normal}$$

T_{normal} is the response time whose relative CPU utilization is 40%

α is determined by users to meet their requirements

- Thus, we get the threshold of CPU utilization $U_{threshold}$:

$$U_{threshold} = \begin{cases} 90\% & U_{limit} \geq 90\% \\ U_{limit} & U_{limit} < 90\% \end{cases}$$

U_{limit} is the CPU utilization corresponding to T_{limit}



➤ Goal:

- Scale up or down according to the monitoring data from monitor module, while meet the QoS requirements by QoS module

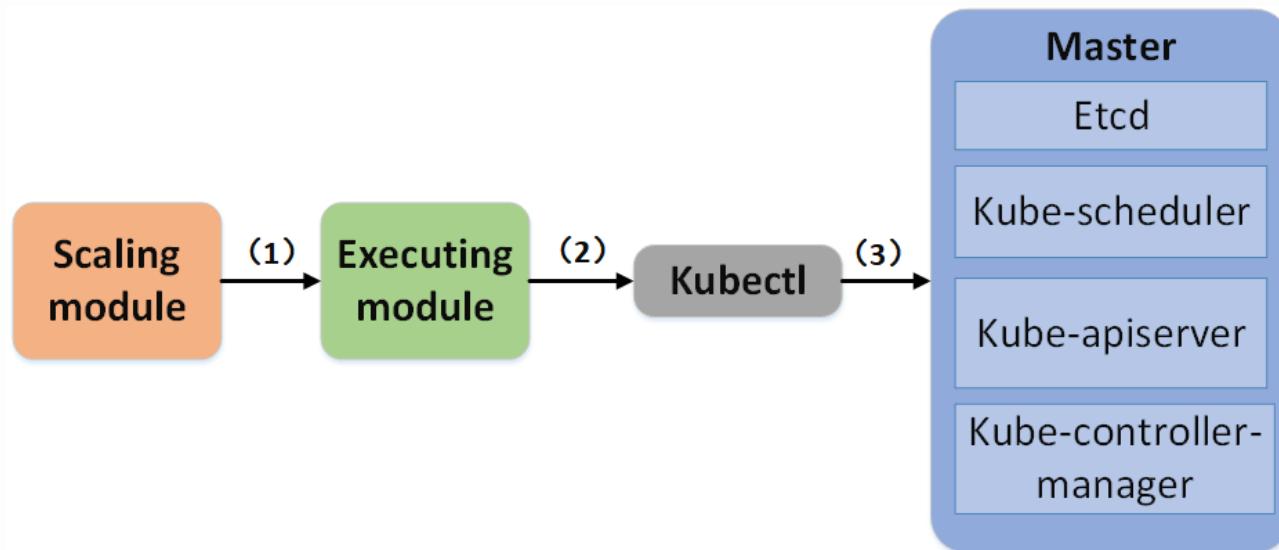
➤ Cluster Scaling Algorithm:

- If $U > U_{threshold}$,
 - $N_{add} = 2 * N_{add}$, if the cluster scaled up last time
 - $N_{add} = 1$, if the cluster don't scaled up last time
- If $U < 40\%$,
 - $N_{remove} = 2 * N_{remove}$, if the cluster scaled down last time
 - $N_{remove} = 1$, if the cluster don't scaled down last time

Executing Module

➤ Goal:

- Implement each operation for cluster scaling based on the output of the scaling module
 - Generate specific command for Kubectl to realize the scaling operation





Outline

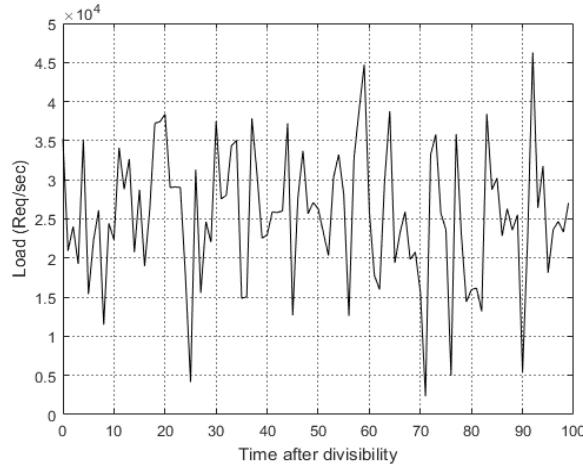
- **System design**
- **Evaluation**
- **Conclusion**

- **5 physical machines:**
4-cores Intel(R) Core(TM) i5-4460S 2.9 GHz CPU, 4 GB memory and 1 TB disk
- **CentOS Linux release 7.5**
- **Kubernetes v1.10 and Docker v18.06-ce**
Heapster v1.5.2 and InfluxDB v1.3.3
- **Testing application:**
Ticket Monster, deployed in Deployment manner with the HorizontalPodAutoscaler
- **Workloads:**
Apache JMETER , simulate the workload that users send HTTP requests to the Ticker Monster

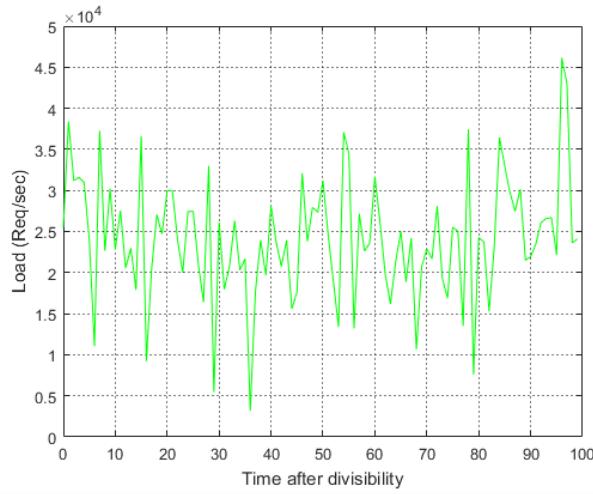


Workloads

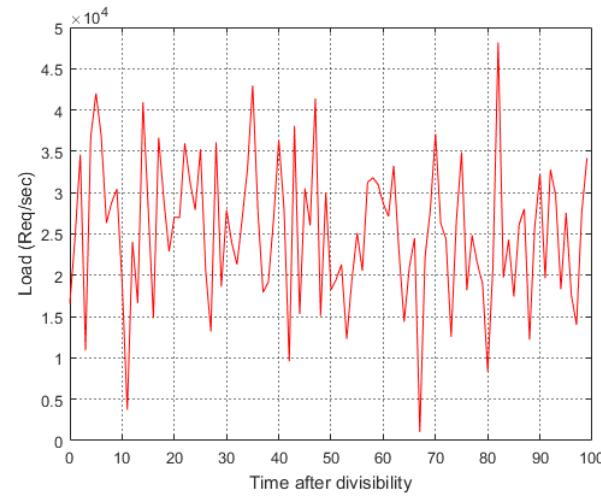
Workload Examples:



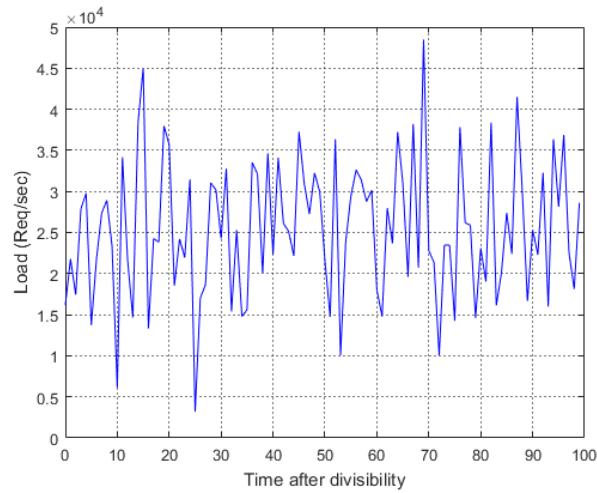
Load a



Load c



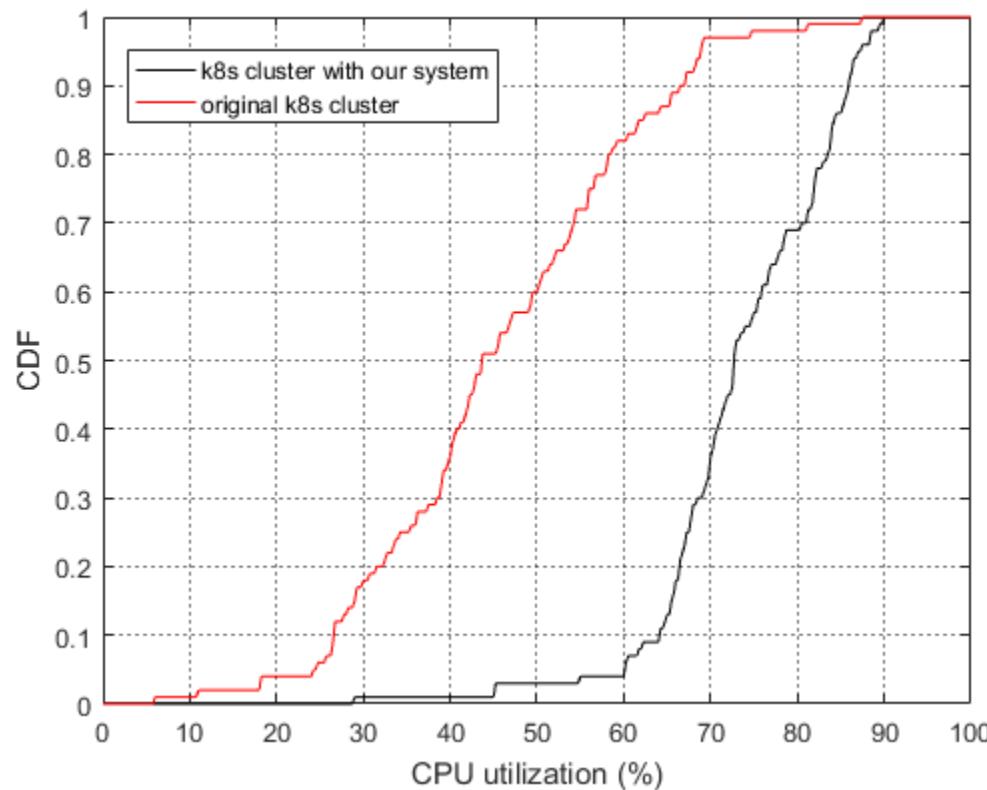
Load b



Load d

Improvement of CPU utilization

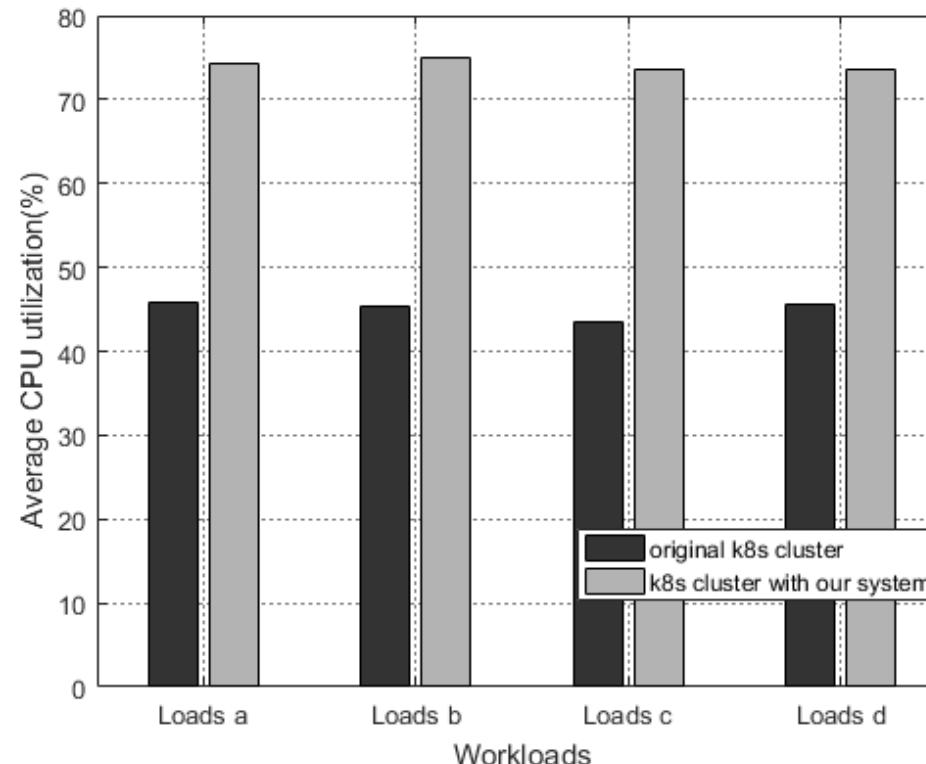
- CDF of CPU utilization of the original Kubernetes cluster and the Kubernetes cluster with our system:





Improvement of CPU utilization

- The average CPU utilization of the original Kubernetes cluster and Kubernetes cluster with our system under four different workloads:

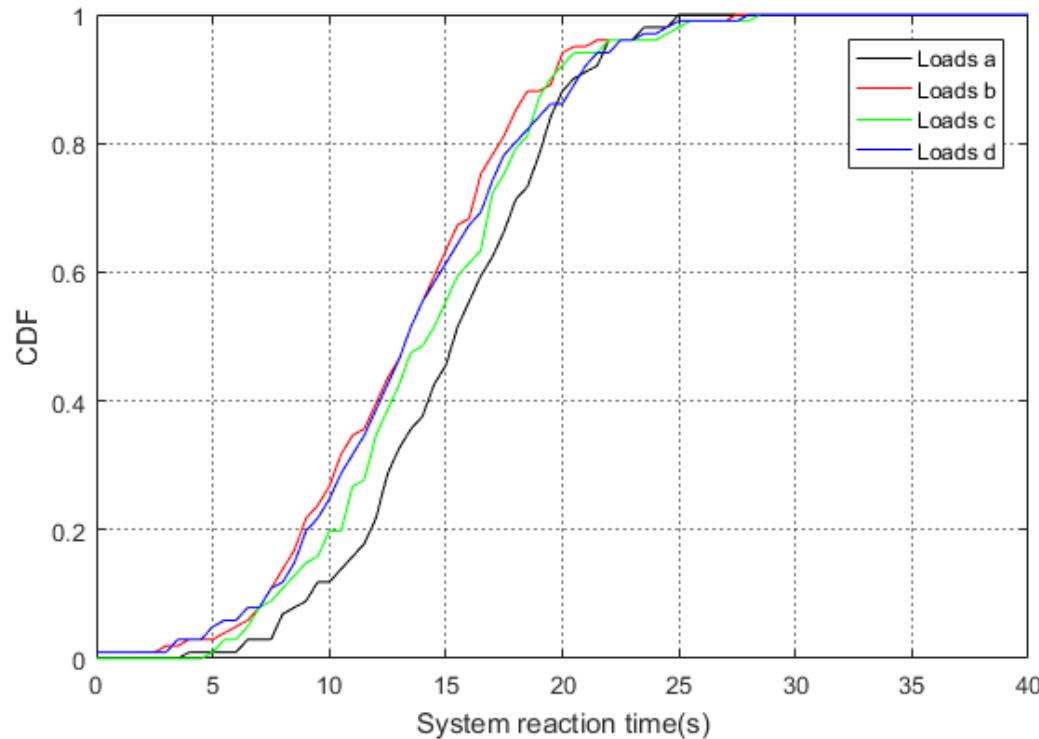


Improved by **28.99%**



Reaction time

- CDF of the system reaction time under four different workloads:

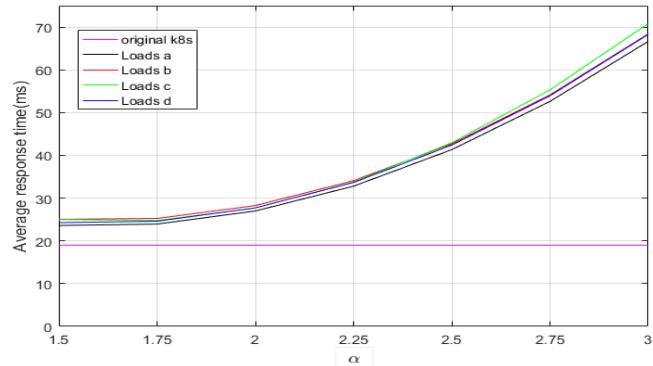


- The average reaction time of the system is about 15s.

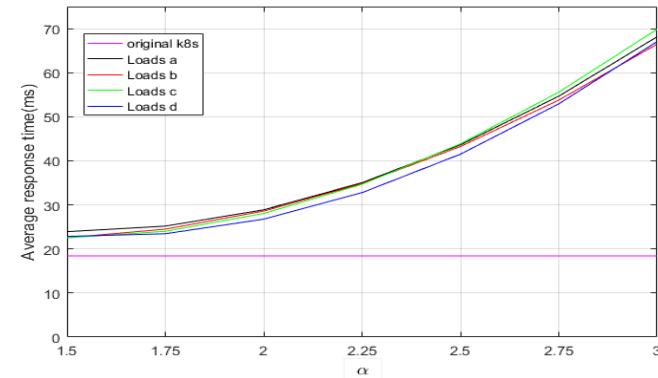


Parameter Selection

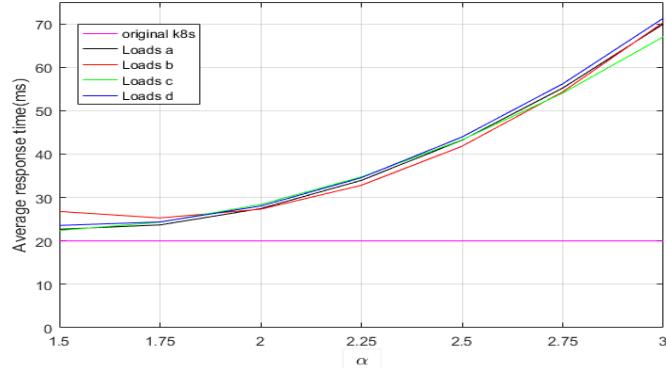
QoS coefficient α under different duration T_{dur}



$T_{dur}=10s$

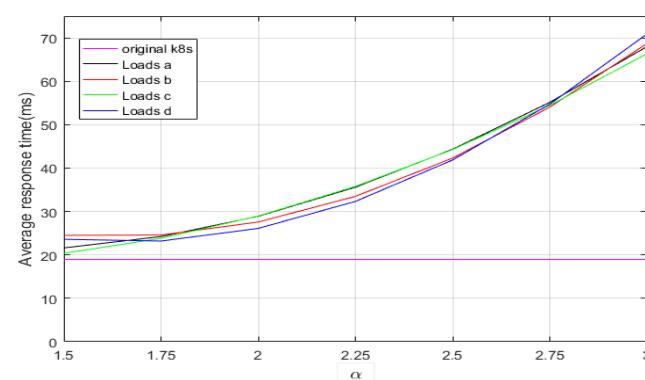


$T_{dur}=20s$



$T_{dur}=30s$

select $\alpha = 2$



$T_{dur}=40s$



Outline

- **System design**
- **Evaluation**
- **Conclusion**



Conclusion

- We propose a system, which dynamically adjusts scale of a Kubernetes cluster, to improve the resource utilization.
- The system can automatically derive a threshold of system resource utilization according to the specific application in a Kubernetes cluster, which promises QoS in a Kubernetes cluster.
- The experimental results show that CPU utilization of a Kubernetes cluster with our system is improved by 28.99% than that of a original Kubernetes cluster on average.



上海交通大学
SHANGHAI JIAO TONG UNIVERSITY



**Thank you !
Q & A**