

#### Implications of Public Cloud Resource Heterogeneity for Inference Serving

Sixth International Workshop on Serverless Computing (WoSC6) Dec 8, 2020



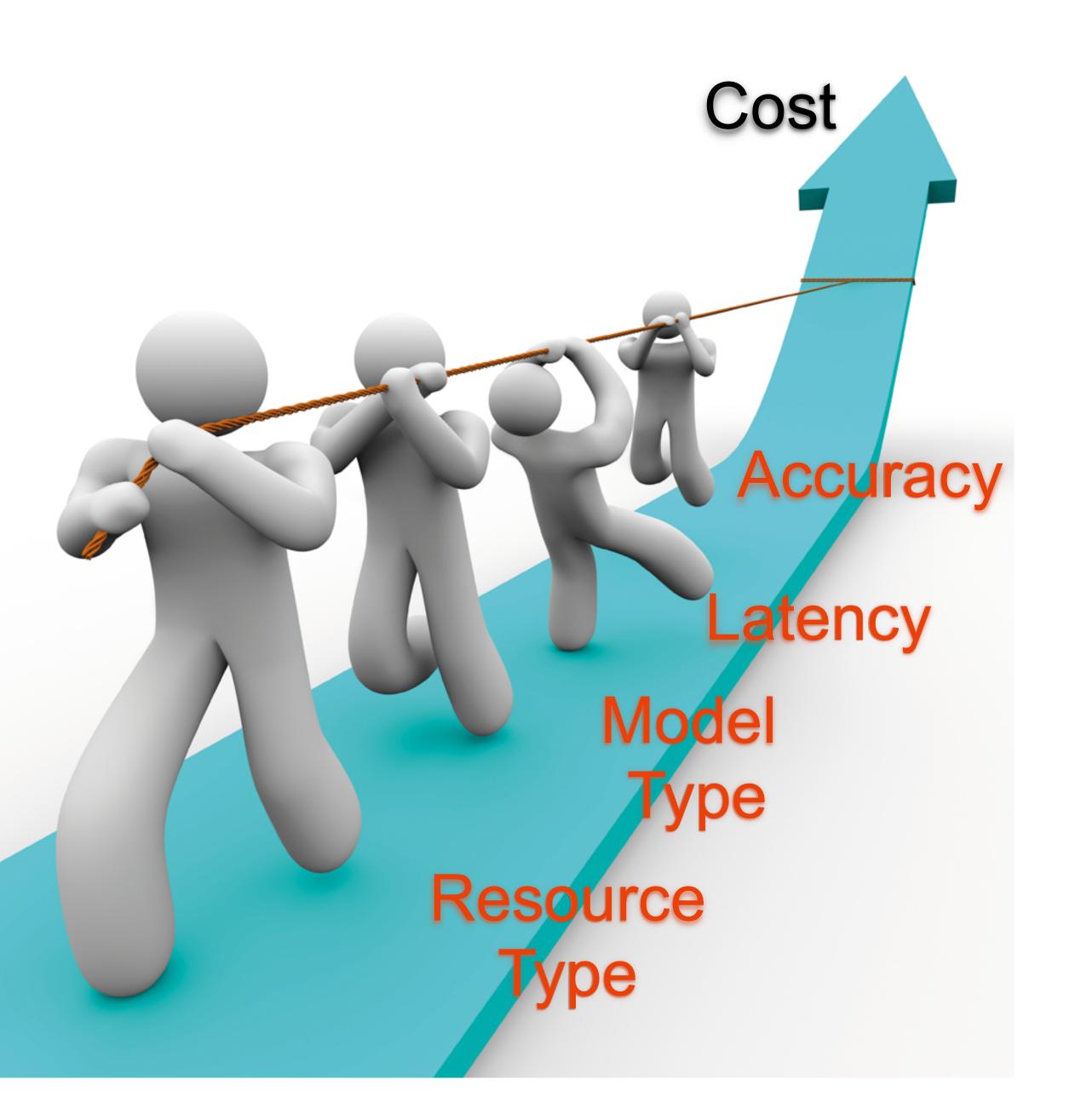
## **PennState**

Jashwant Raj Gunasekaran, Cyan Subhra Mishra, Prashanth Thinakaran, Mahmut Kandemir, Chita Das



## Model Serving in Public Cloud

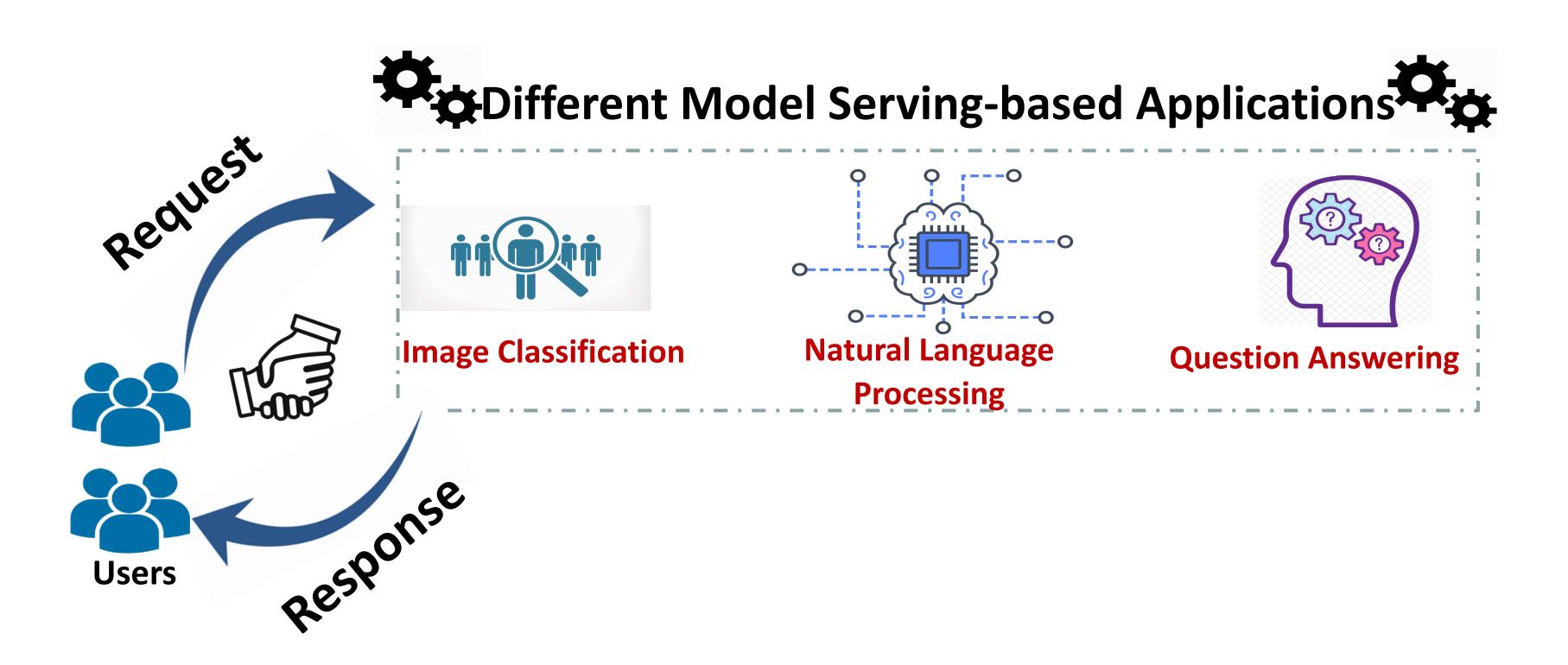






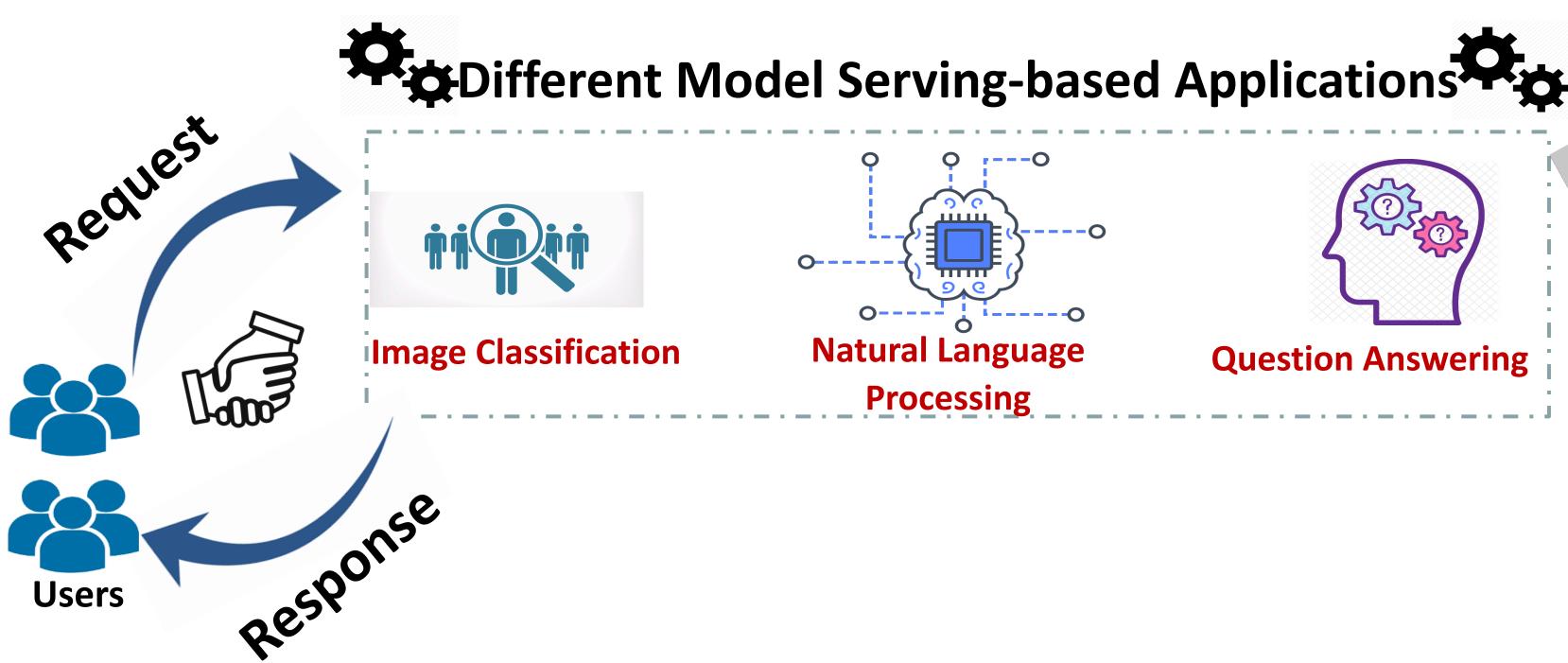






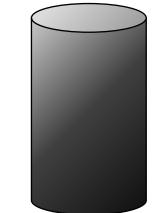






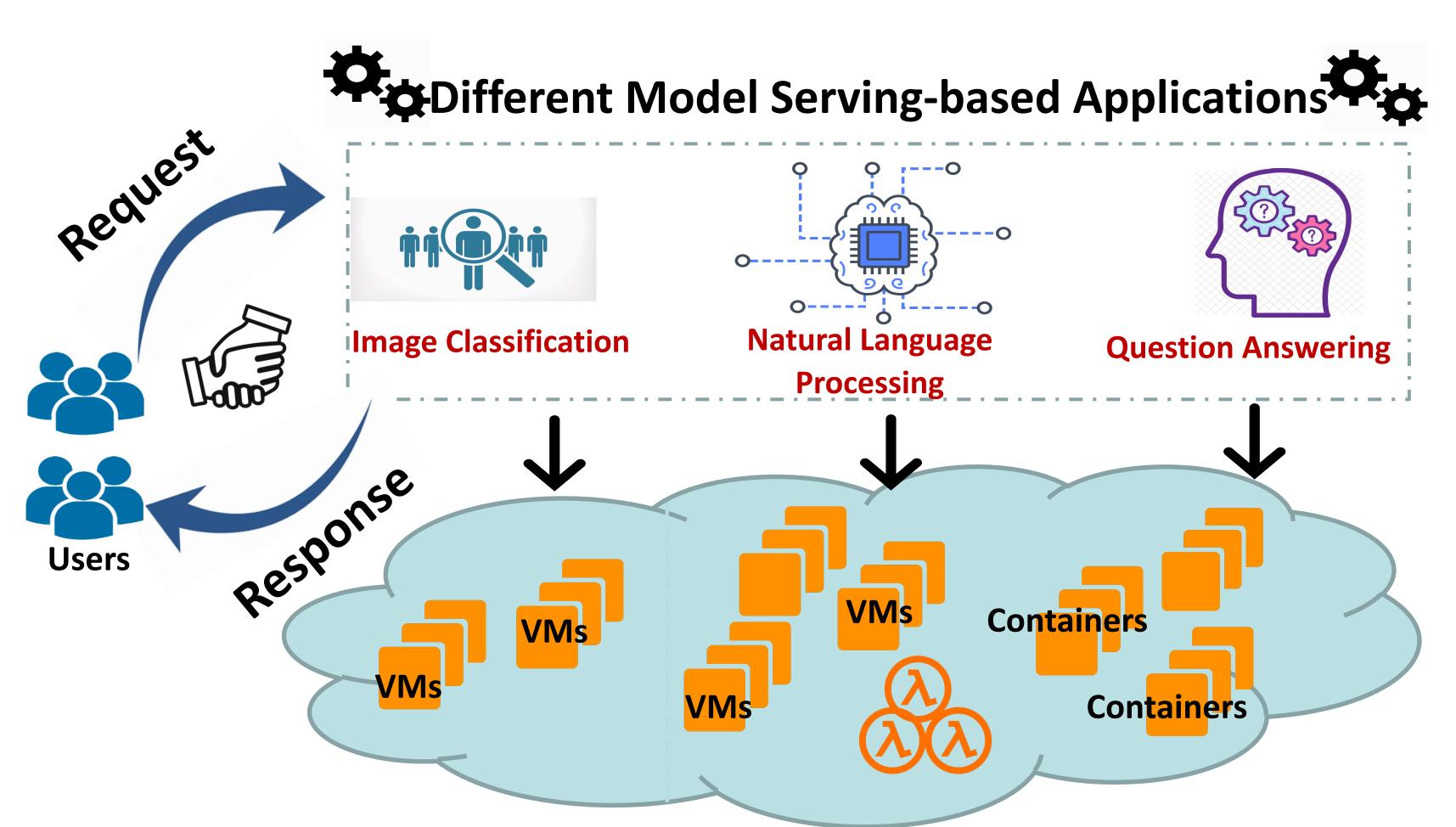


#### Training







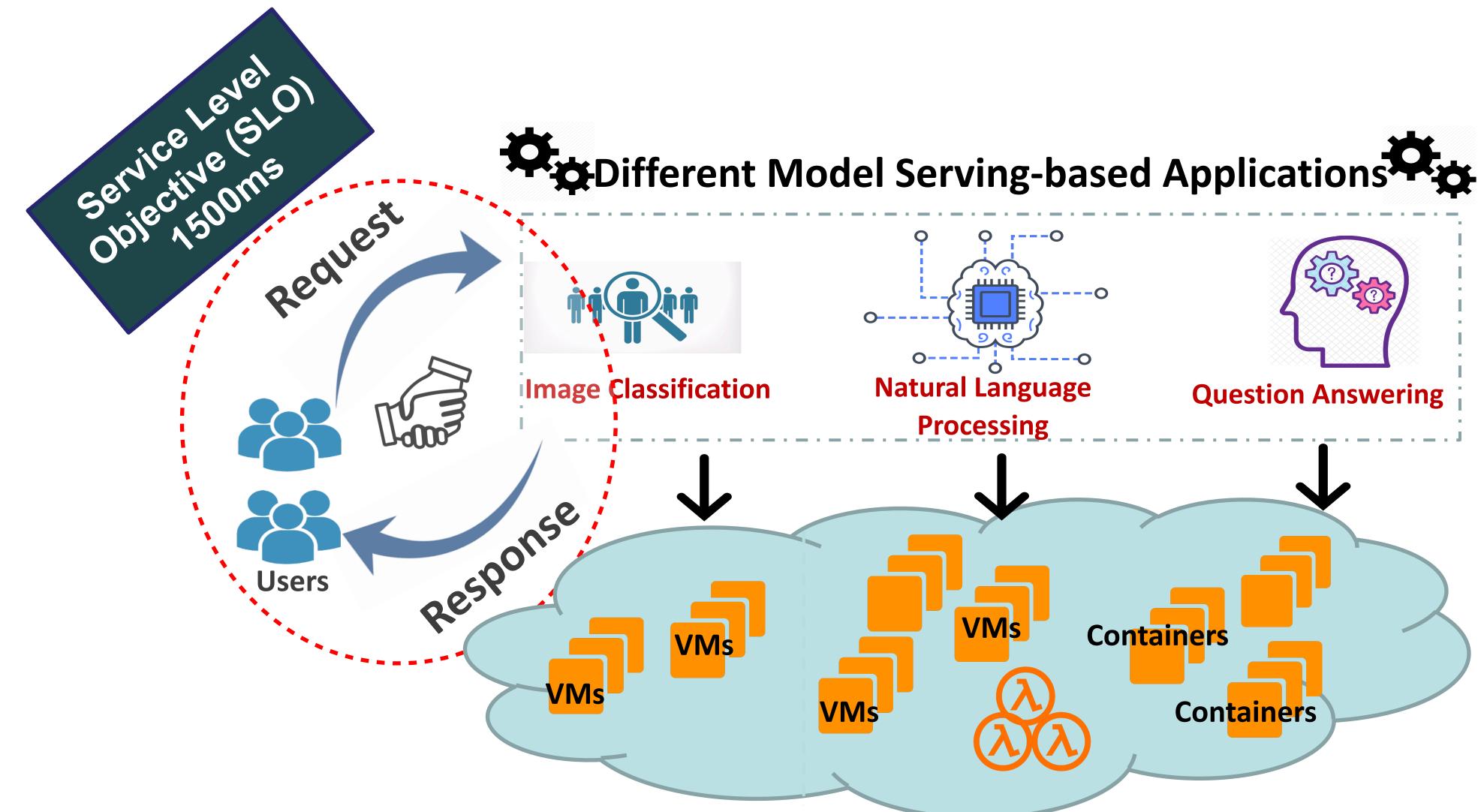






#### **Resources for Applications**





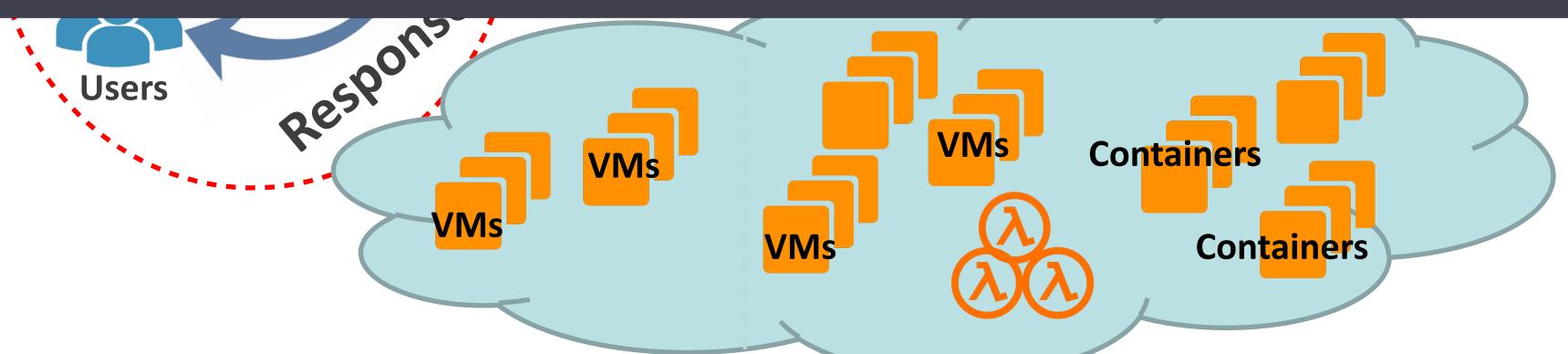


#### **Resources for Applications**



THEST

## How to optimize both model selection and resource selection?

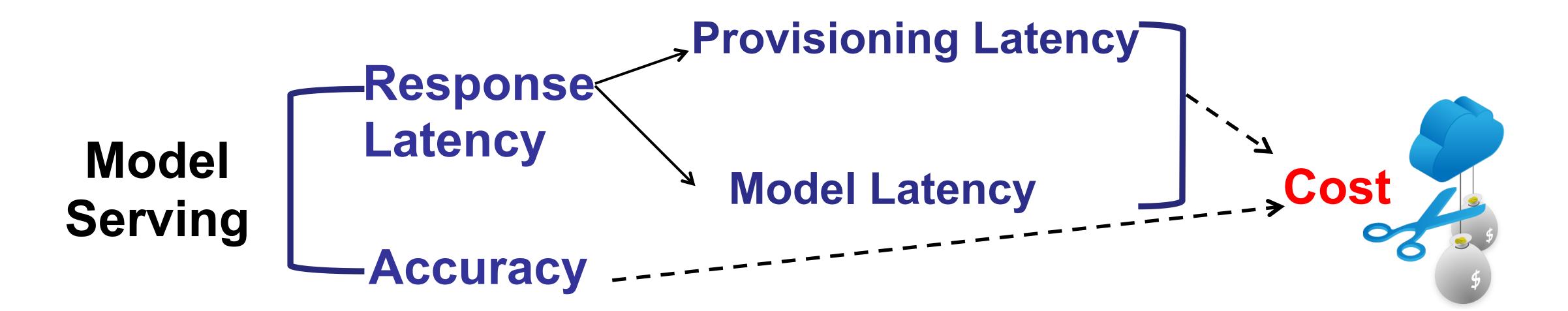






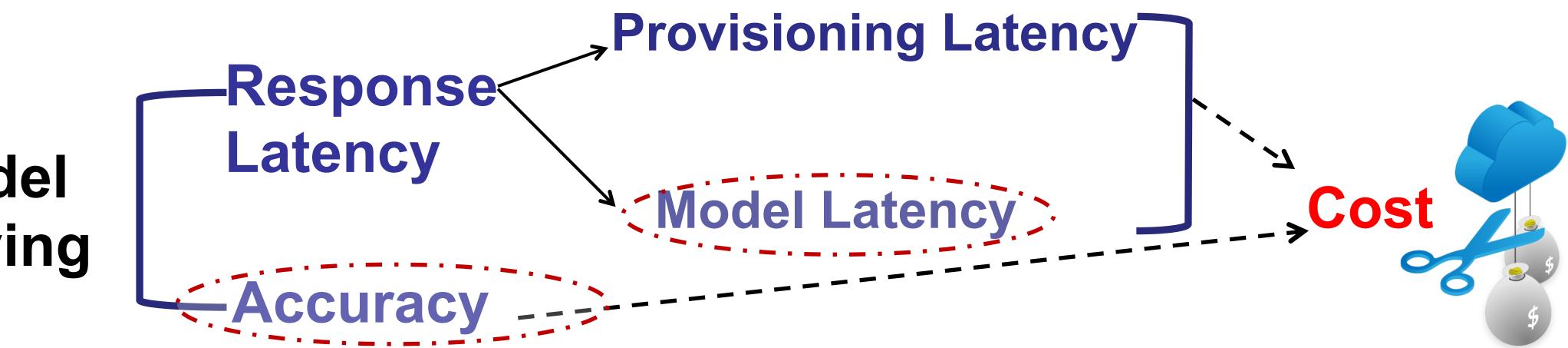
#### **Resources for Applications**







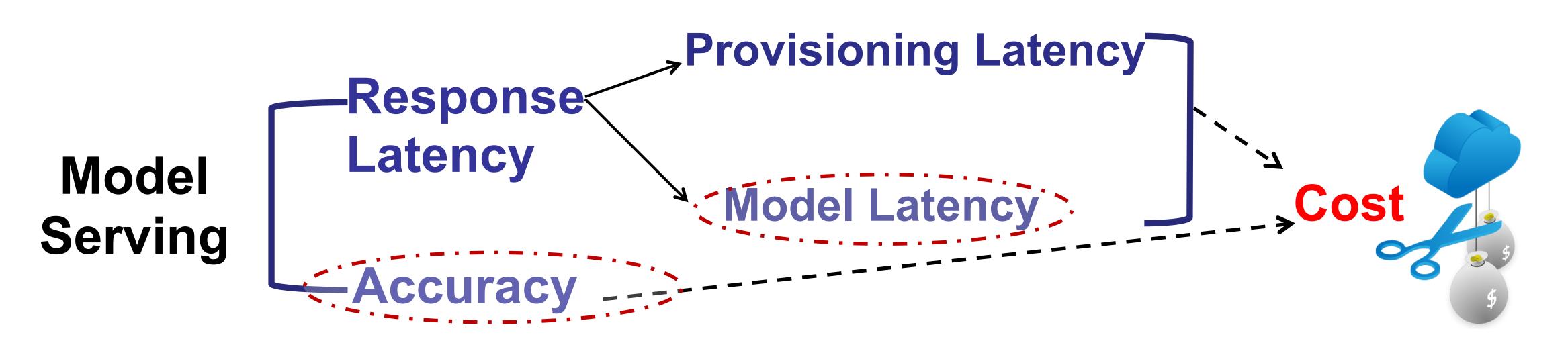




#### Model Serving







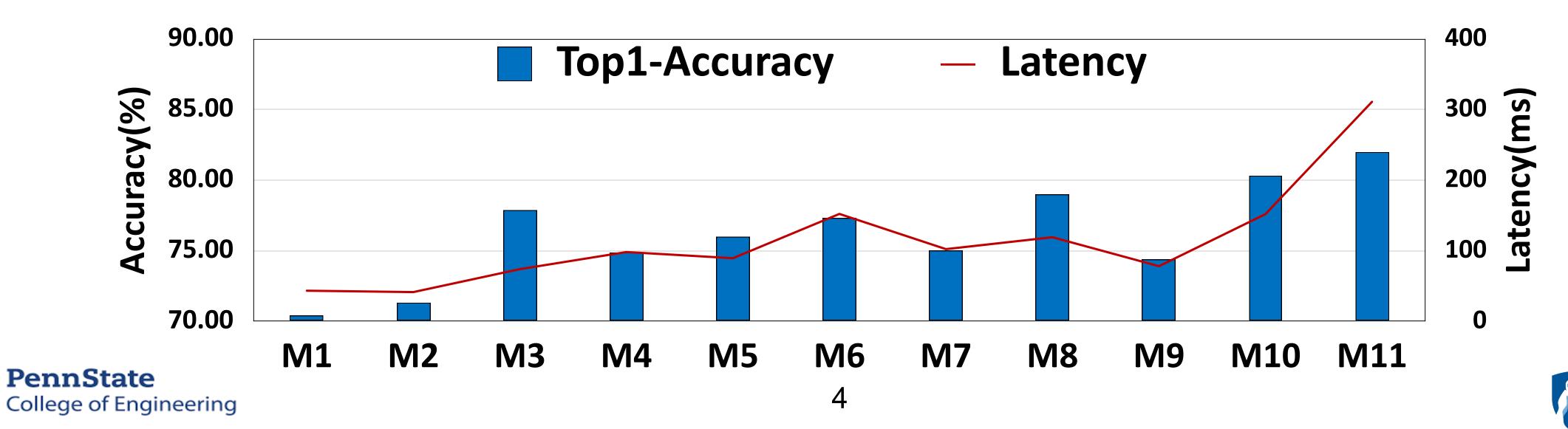
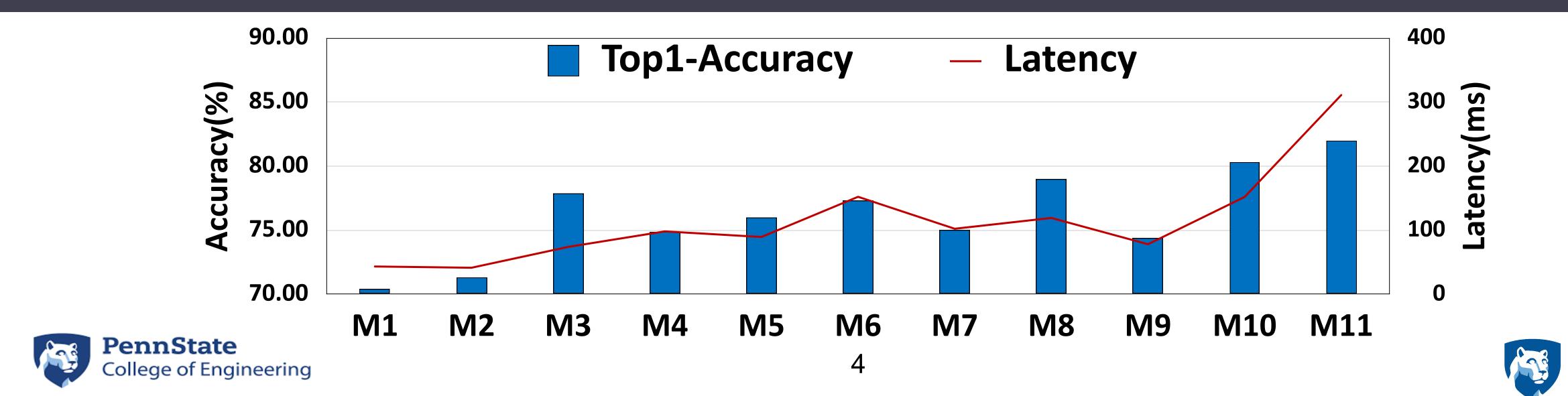


Image Classification using pretrained keras models



#### Response Latency Model Model Latency



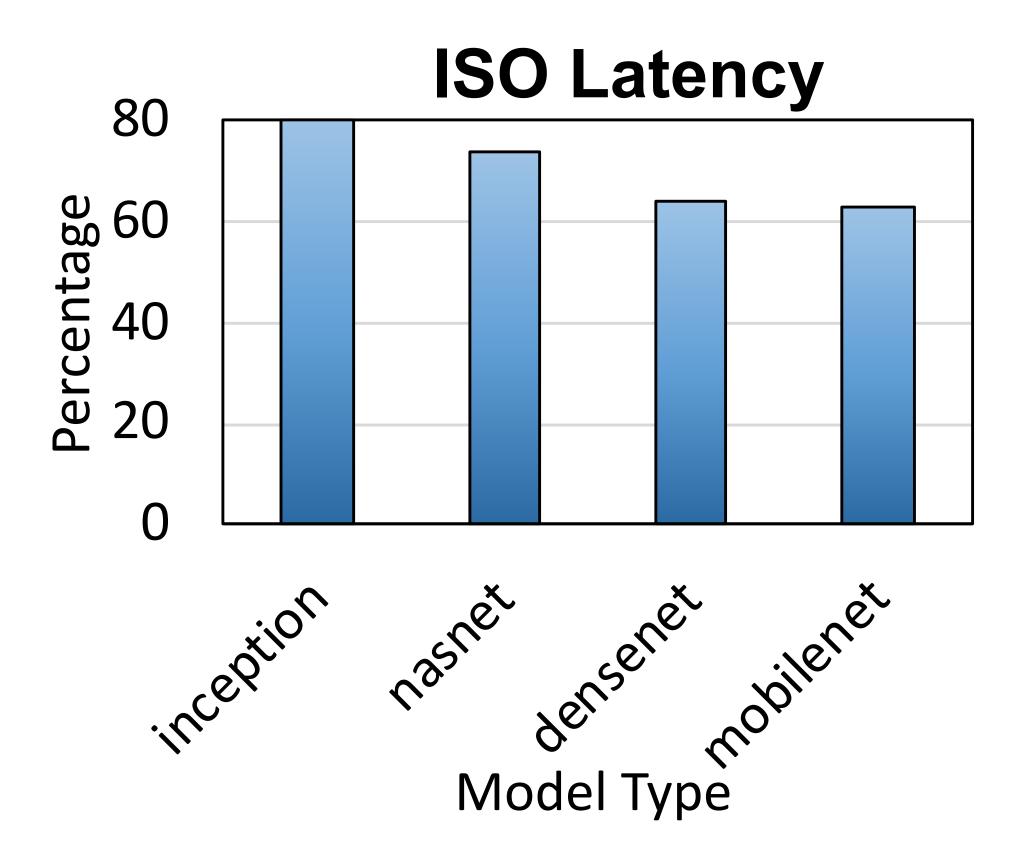
**Provisioning Latency** 

Model Serving Challenges?

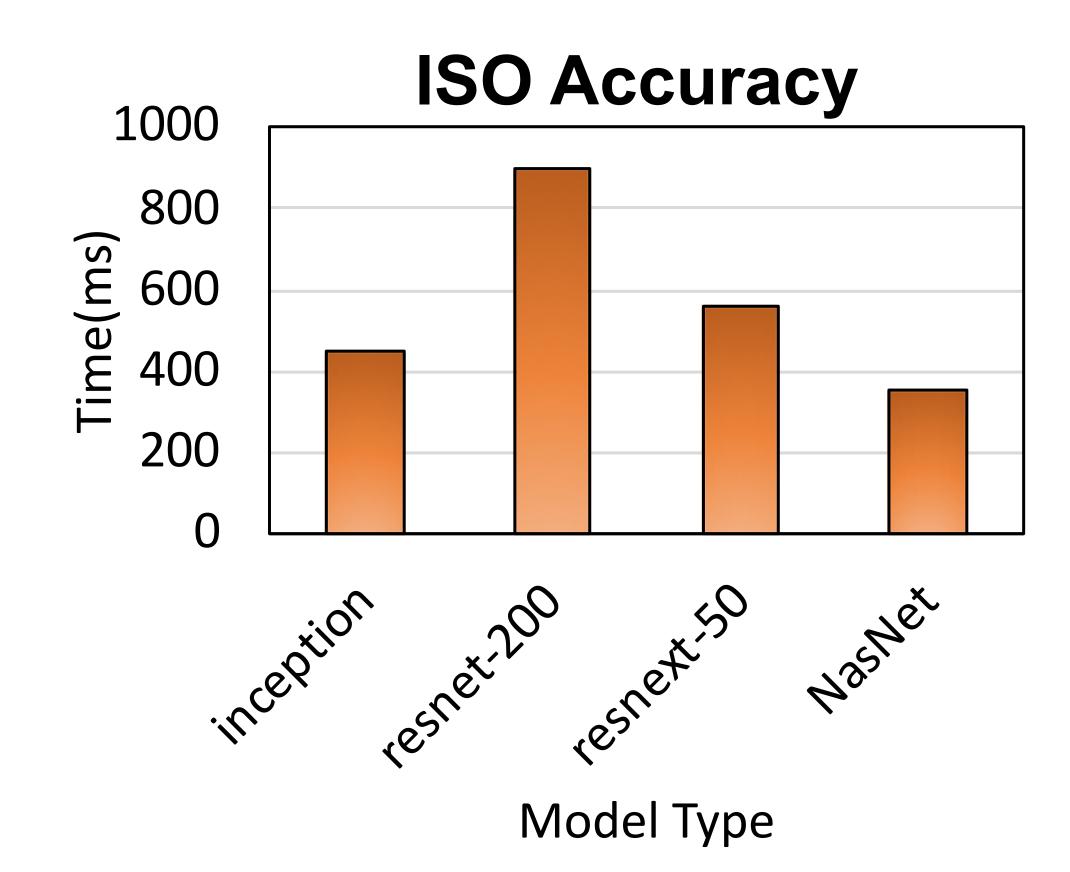


Cost

## Model Selection





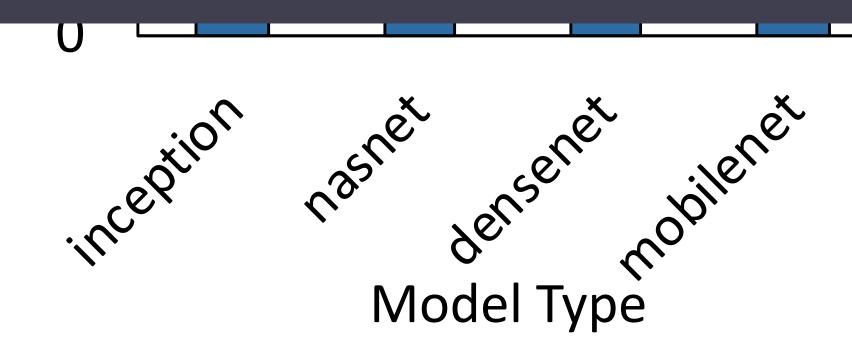




#### Model Selection

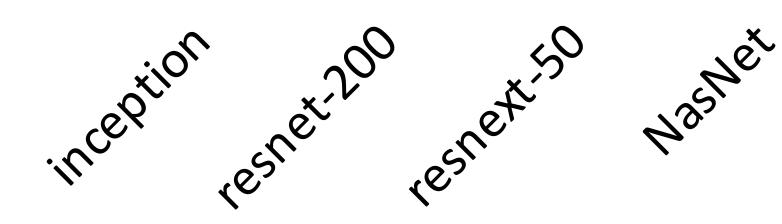
# 80 ISO Latency

#### What about resource selection?









Model Type



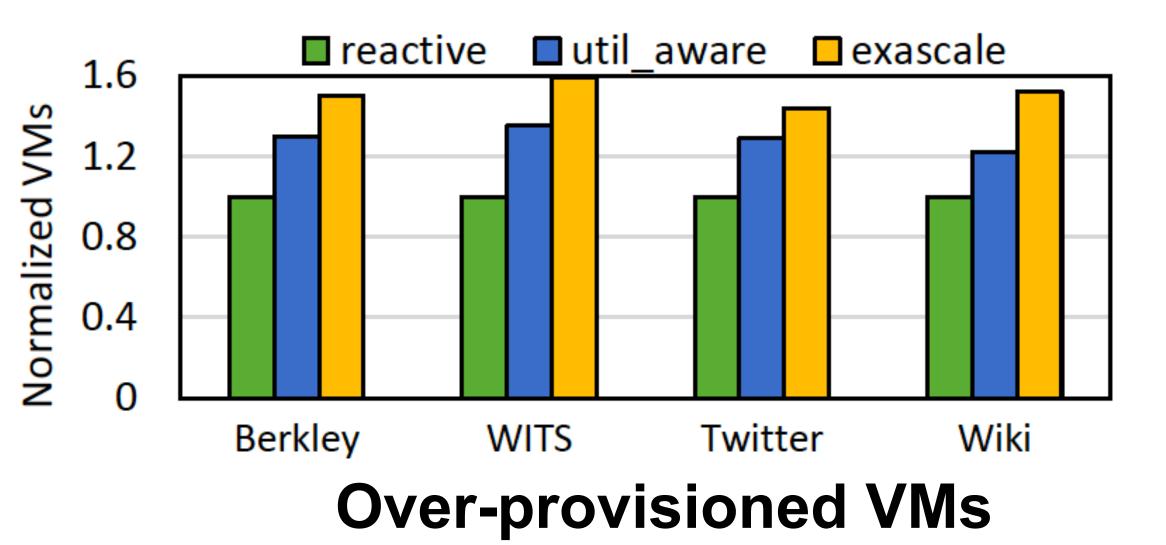






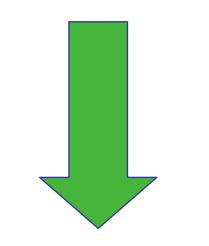






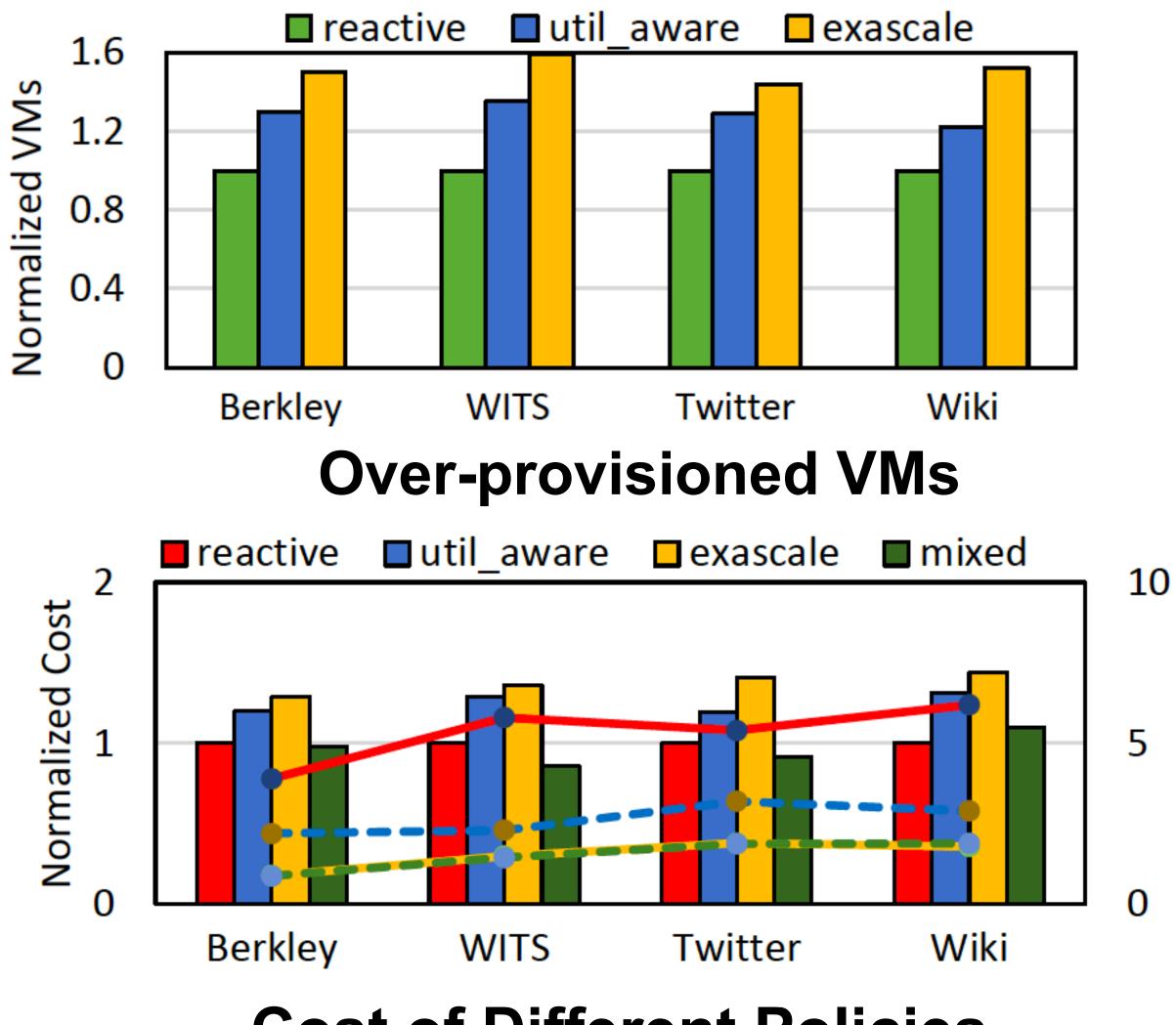


#### **SLO violations**





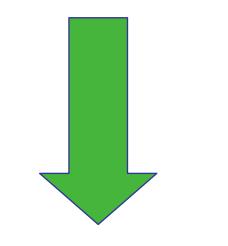




**Cost of Different Policies** 



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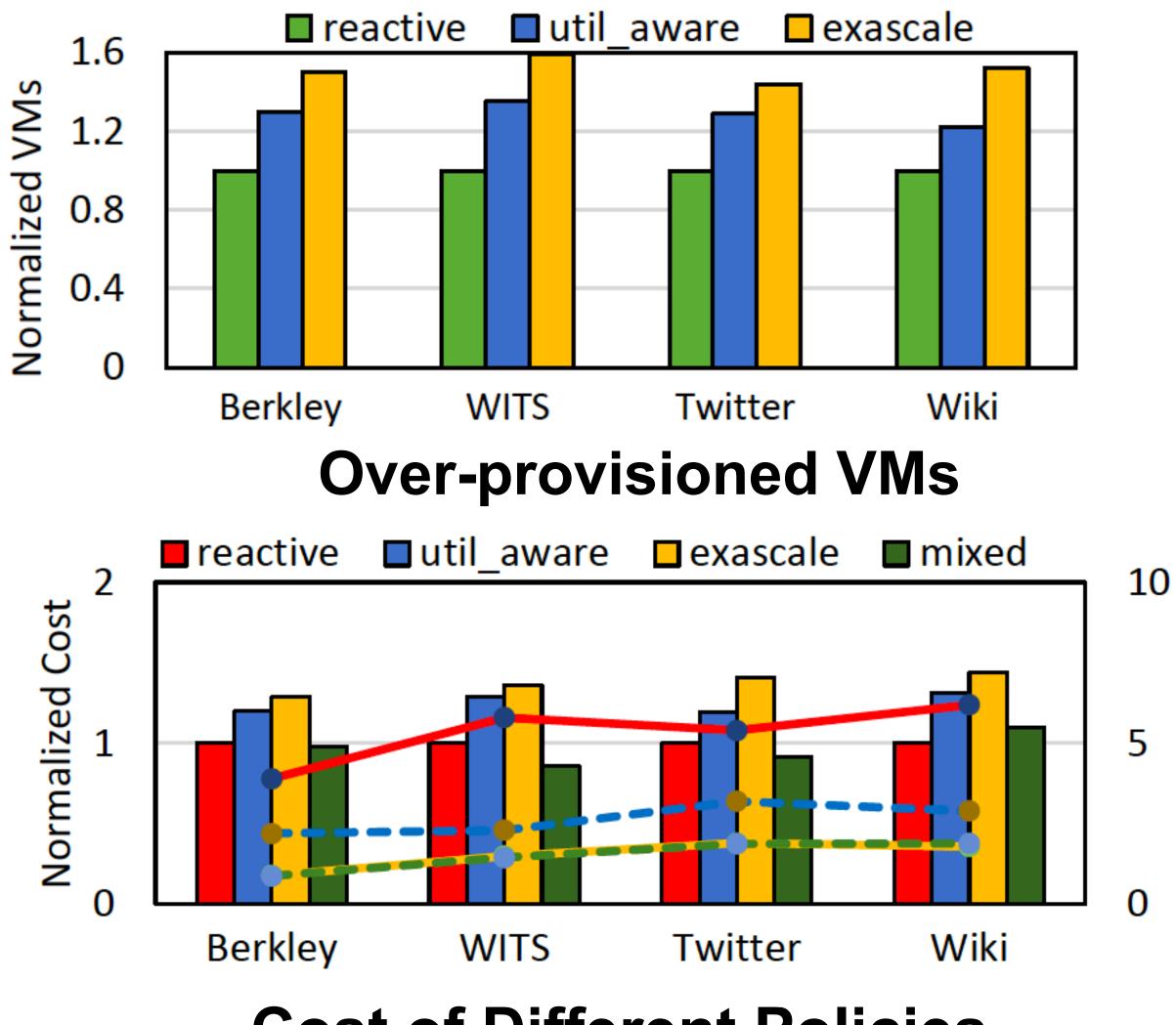








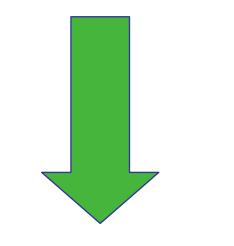




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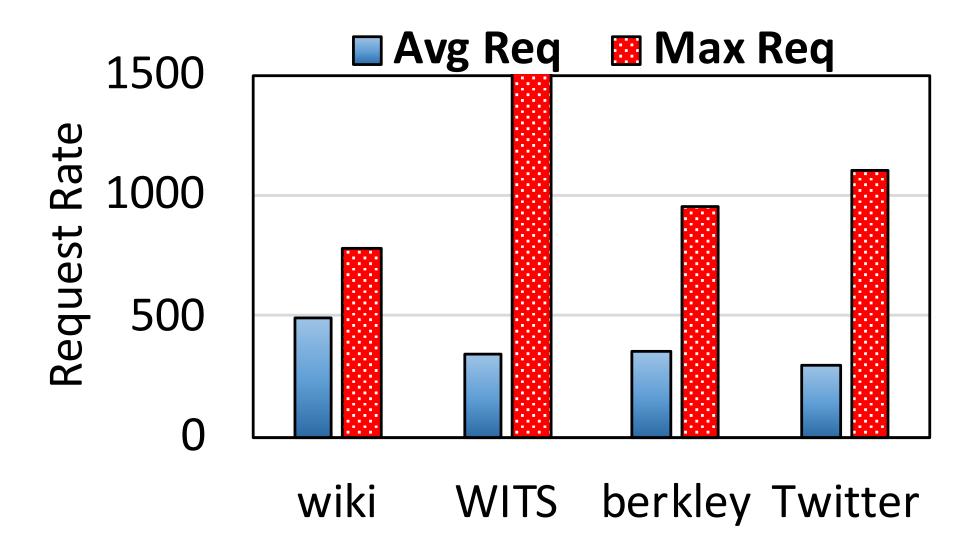








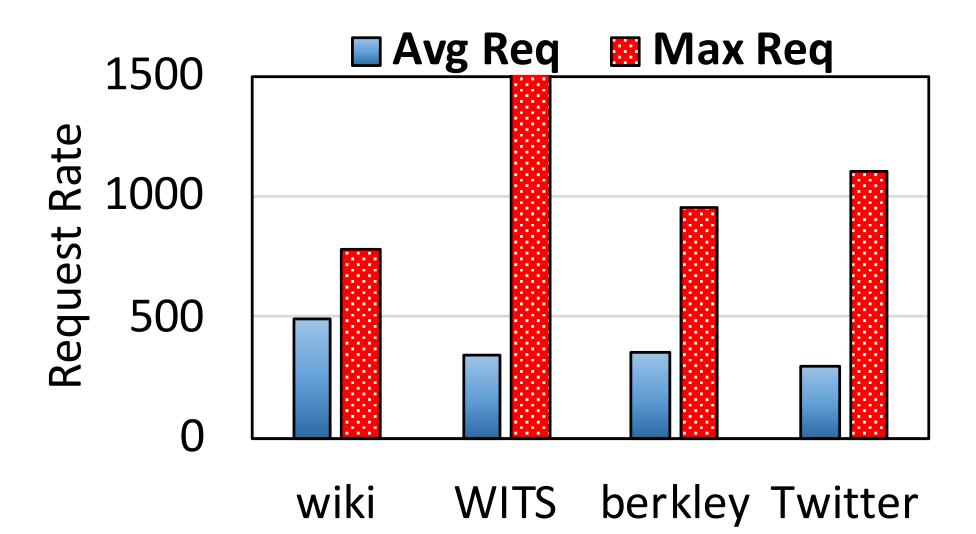
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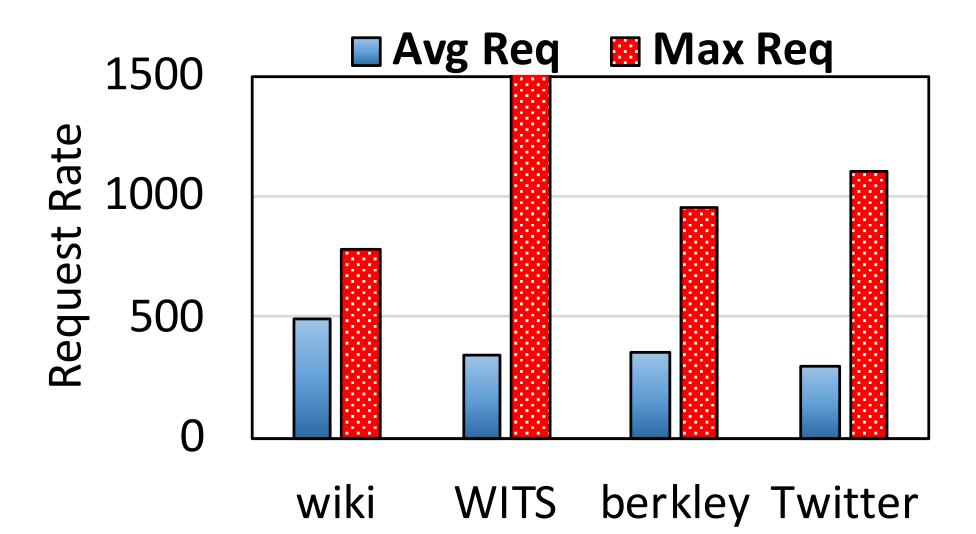








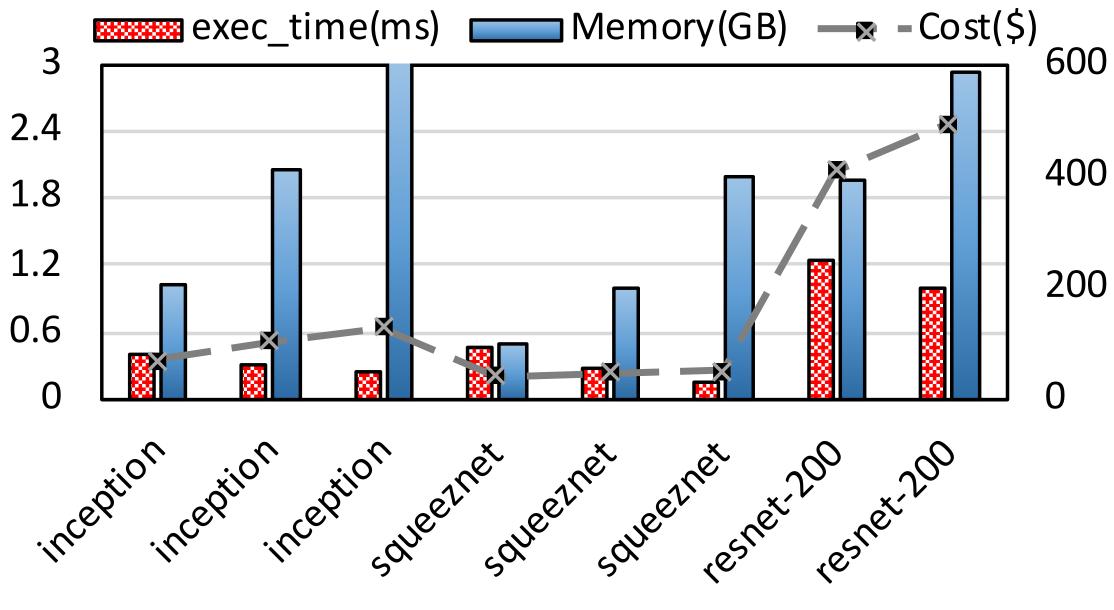
#### **Arrival rate variability**







#### **Serverless Function Configuration**

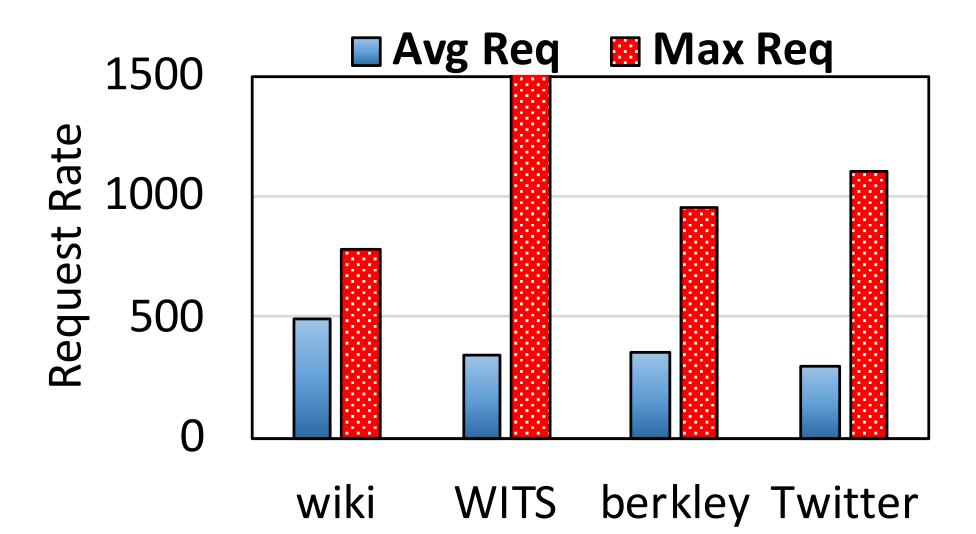


Model Type





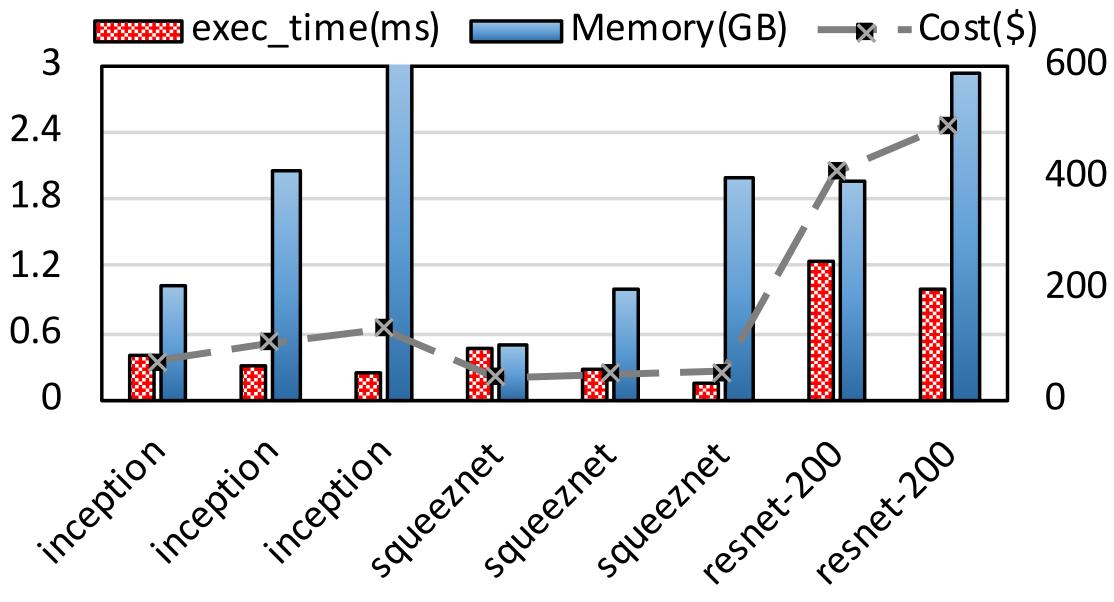
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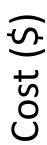


#### Model Type

#### Cost is 1.5x higher for 0.2x lower latency













 How to make the users oblivious of model selection from the extensive pool of models?





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- How to make the users oblivious of model selection from the extensive pool of models?
- How to right-size VMs and appropriately configure the serverless functions?
- •What is the right degree to combine serverless functions along with VMs for dynamic load?



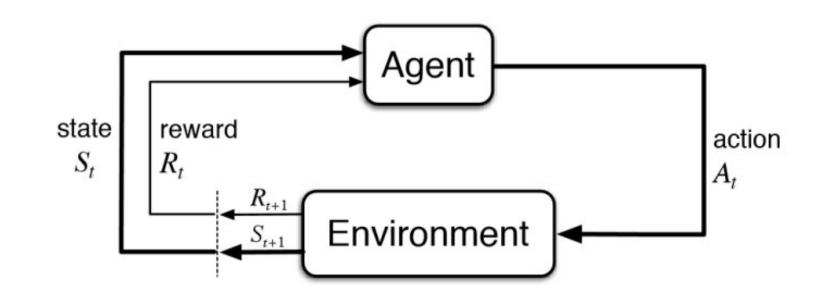






Feedback-driven learning based model selection.

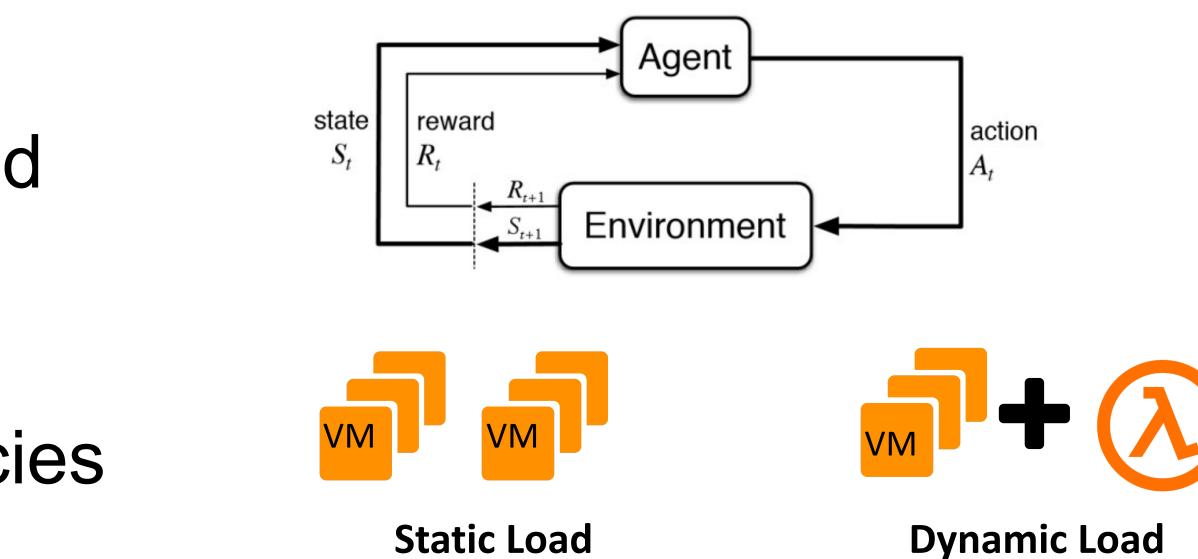






- Feedback-driven learning based model selection.
- Load-Based Procurement Policies



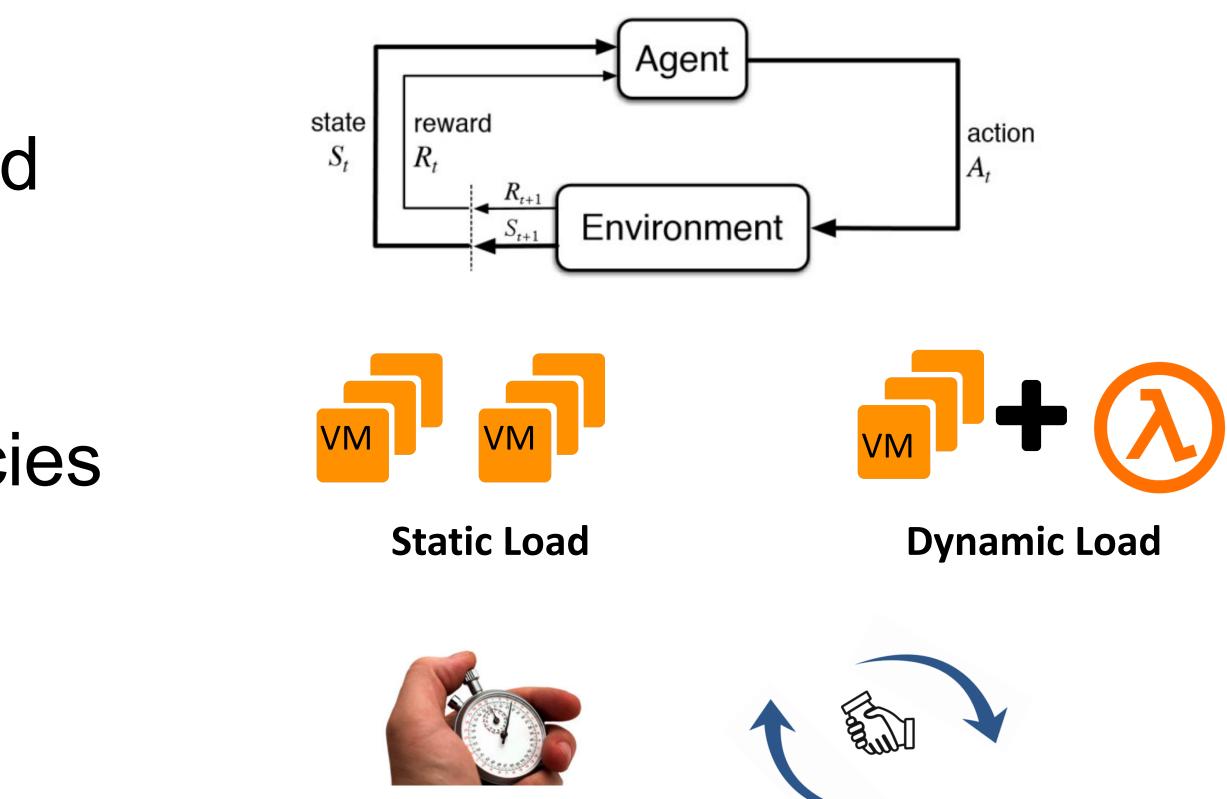






- Feedback-driven learning based model selection.
- Load-Based Procurement Policies
- Provisioning latency and SLO aware resource selection

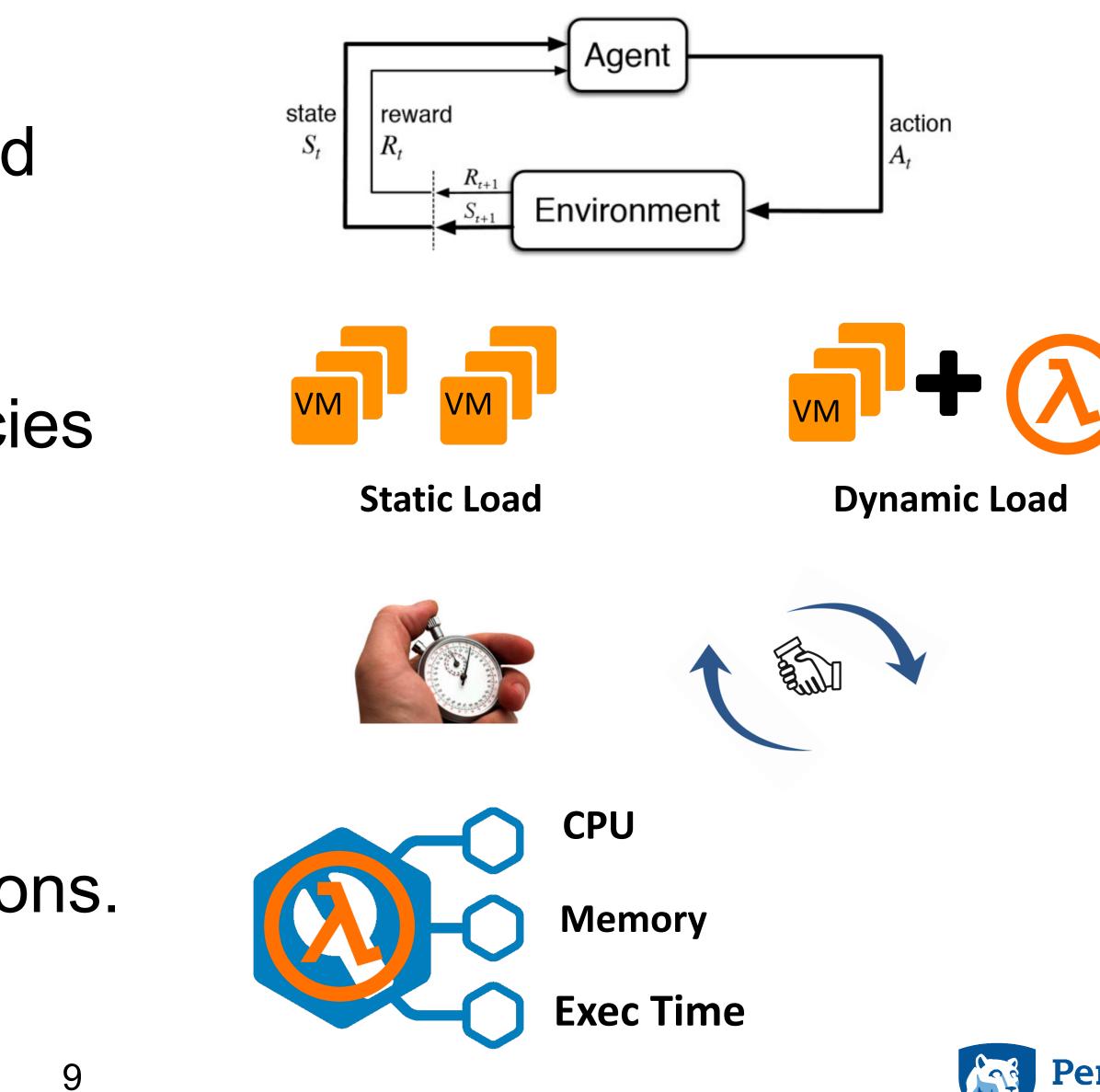






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- Dynamic serverless configurations.





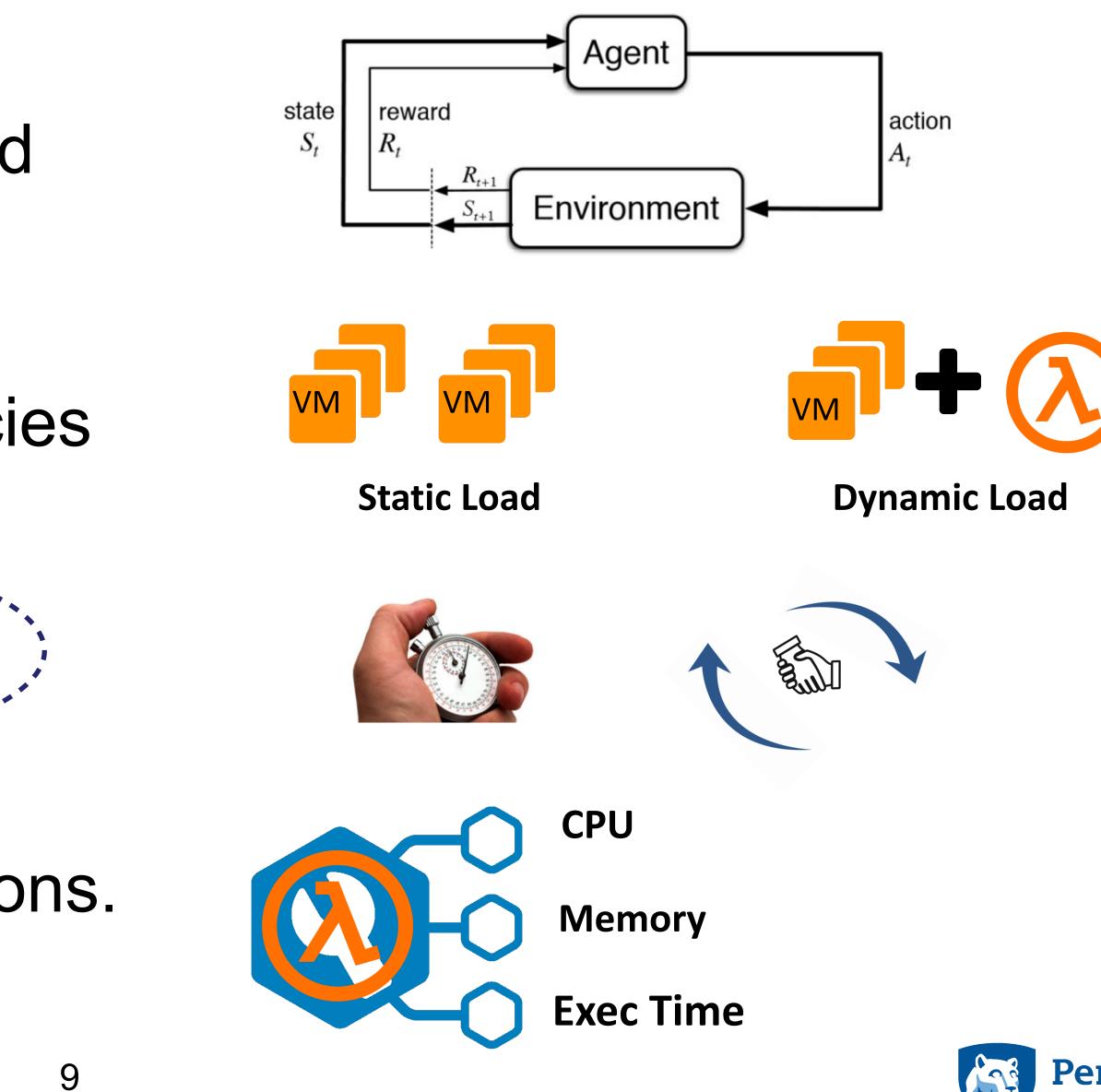


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Provisioning latency and SLO aware resource selection

Dynamic serverless configurations.







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- AWS resources.
- Pretrained ML models on imagenet dataset.







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Query Type	Memory Re- quired (GB)	Memory Al- located (GB)	Average Ex- ecution (ms)	Requests per vCPU for VMs
Caffenet	1.024	3.072	300	4
Googlenet	0.456	2.048	450	3
Squeezenet	0.154	2.048	130	6
Resnet-18	0.304	3.072	320	3
Resnet-200	1.024	3.072	956	1
Resnext-50	0.645	3.072	560	2



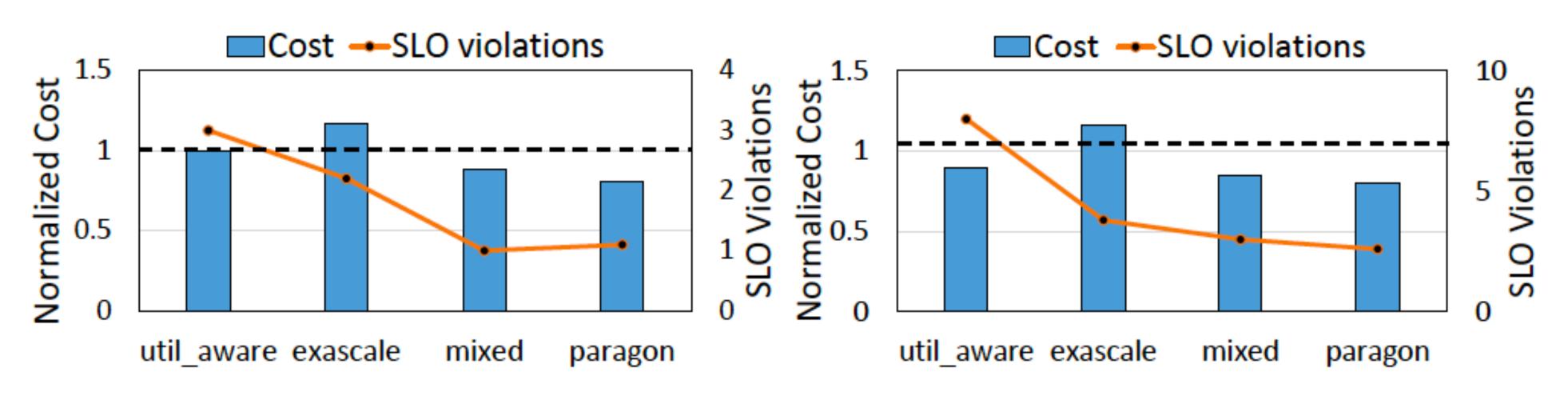
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## Initial Results



#### (a) Workload-1: Berkeley Trace. (b) Workload-1: WITS Trace.

# 60% less SLO Violations.10% reduction in deployment costs









