# Introduction to MESOS

By: Morteza Zakeri Cluster and Grid Computing Course Instructor: Dr. Mohsen Sharifi School of Computer Engineering Iran University of Science and Technology Winter 2017





### Outline

- Cluster Computing
  - Applications
  - Backgrounds and Terminology
  - Problems
- Sharing Frameworks
- Mesos
  - Goals to Design
  - Key Elements
  - Architecture

### Outline

- Resource Offer
- Implementation and API
- Evaluation (Benchmarks)
- Limitations
- Related Works
- Conclusion
- References

### **Cluster computing main applications**

- High Performance Computing (HPC)
  ➤Large internet services
  - E.g. social networks, online shopping, ...
  - Data-Intensive scientific applications
    - E.g. physics, astronomy, molecular sciences, ...



#### M4 globular cluster

### Terminology

- Cluster
  - A collection of distributed machines that collaborate to present a single integrated computing resource to the user.
- Node
  - Each computing element (single machine itself) within cluster.
- Framework
  - A software system that manages and executes one or more jobs on a cluster.

## Terminology

- Job
  - A unit of work run on the cluster nodes.
- Slot
  - Nodes are subdivided into "slots".
  - (In some framework such as Hadoop and Dryad)
- Task
  - Jobs are composed of short tasks that are matched to slots.
  - (In some framework such as Hadoop and Dryad)

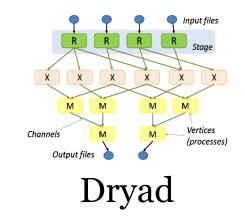
### Terminology

- *Elastic* framework
  - Can scale its resources up and down during execution tasks (dynamically).
  - Such as Hadoop and Dryad.
- *Rigid* framework
  - Can start running its jobs only after it has acquired a fixed quantity of resources.
  - Can not scale (up/down) dynamically.
  - Such as MPI.

### Background

- Diverse array of cluster computing frameworks:
  - Hadoop/ Map-Reduce (multi-terabyte data-sets)
  - Dryad (machine learning)
  - Pregel (graph computation)
  - MPI









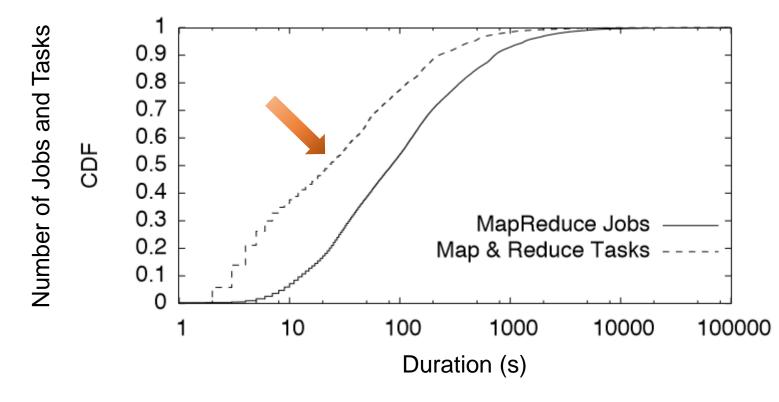
### Problem

- Rapid innovation in cluster computing frameworks.
  There is no single framework that optimal for all applications.
- Clusters typically run single framework at a time.

### Problem

- Hadoop data warehouse at Facebook:
  - Include 2000-node Hadoop cluster,
  - Uses a **fair scheduler** for Hadoop,
  - Takes advantage of the **fine-grained** nature of the workload,
  - Allocate resources at the level of tasks and to optimize data locality.
  - $\checkmark$
  - This cluster can only run Hadoop jobs!
- Can it run an MPI program efficiently?

### Problem



CDF of job and task durations in Facebook's Hadoop data warehouse

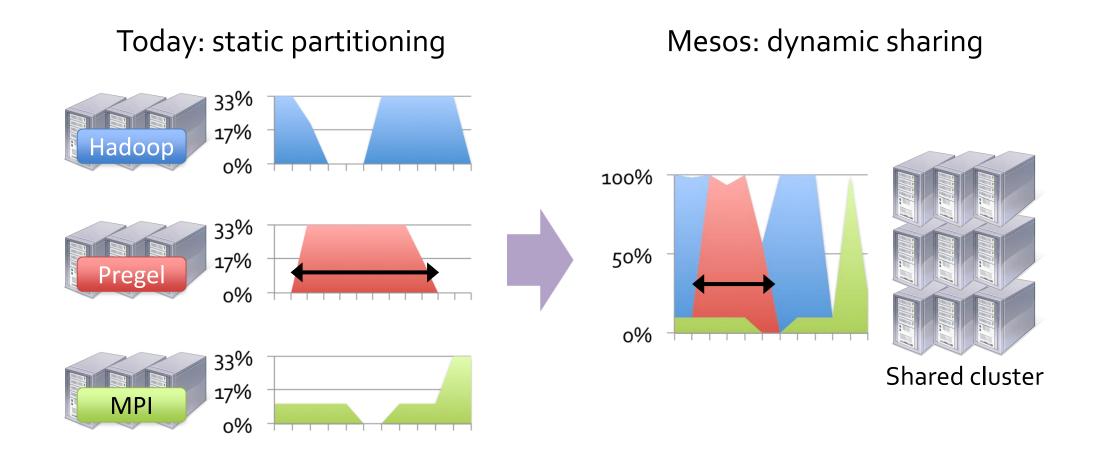
### Motivation to Sharing

- We want to run multiple frameworks in a single cluster.
- Sharing improves cluster utilization,
- Allows applications to share access to large datasets.
  - may be too costly to replicate across distinct clusters!

### Common solutions for sharing a cluster

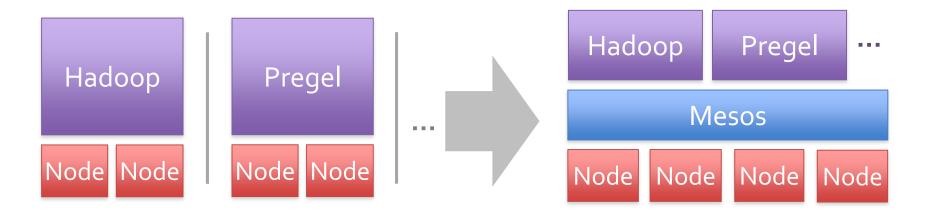
- **1. Statically partition** the cluster and run one framework per partition.
- 2. Allocate a set of VMs to each framework.
- Disadvantages:
  - No high utilization,
  - No efficient data sharing.

### **Better solution: Mesos!**



### What is Mesos?

- A platform for sharing clusters between multiple diverse cluster computing frameworks.
  - E.g. Hadoop + Pregel + MPI + ...



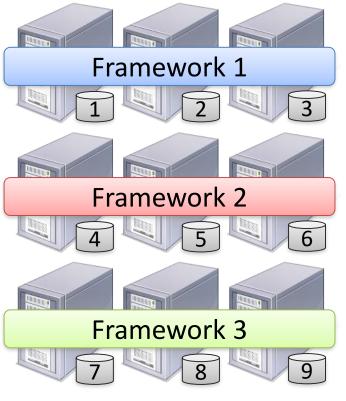
### Goals to Design Mesos

- High utilization
- Efficient data sharing
- Support diverse frameworks (current and future)
- Scalability to 10,000's of nodes
- **Reliability** in face of failures (robust to failures)

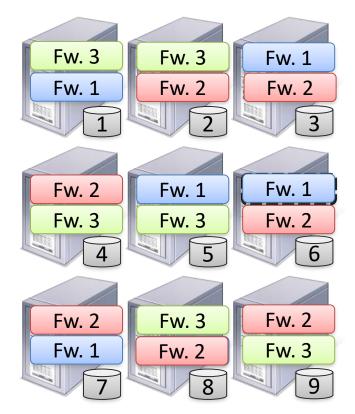
### **Design Philosophy**

- Small microkernel-like core
- Implements fine-grained sharing
  - Allocation at the level of tasks within a job
- Simple, scalable application-controlled **scheduling mechanism**.

### **Fine-Grained Sharing**



Storage System (e.g. HDFS)



Storage System (e.g. HDFS)

### Scheduling Mechanism: Challenges

- How to build a scalable and efficient system that supports a wide array of both current and future frameworks?
- 1. Each framework have different scheduling needs.
- 2. Scheduling system must scale to large number of cluster nodes.
  - Running hundreds of jobs with millions of tasks.
- 3. System must be fault-tolerant and highly available.

### **Scheduling Mechanism: Approaches**

### Centralized Scheduling

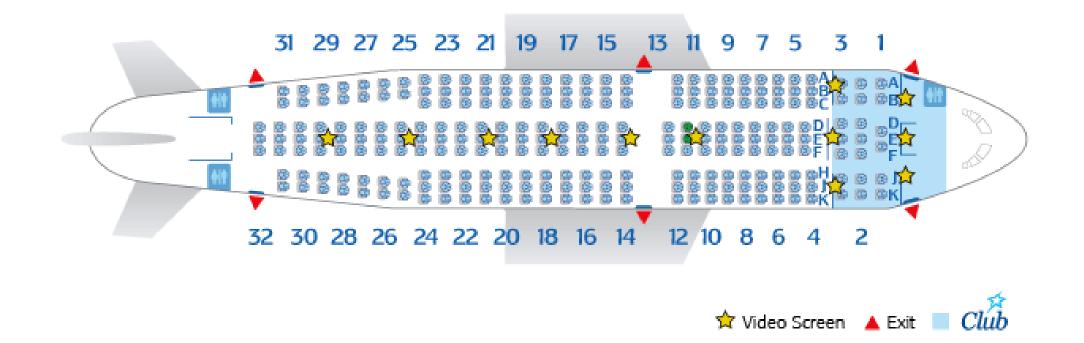
- Frameworks express needs in a specification language,
- Global scheduler matches them to resources.
- + Can make optimal decisions.
- – Complexity,
- Difficult to scale and to make robust,
- – Future frameworks may have unanticipated needs.

### **Scheduling Mechanism: Approaches**

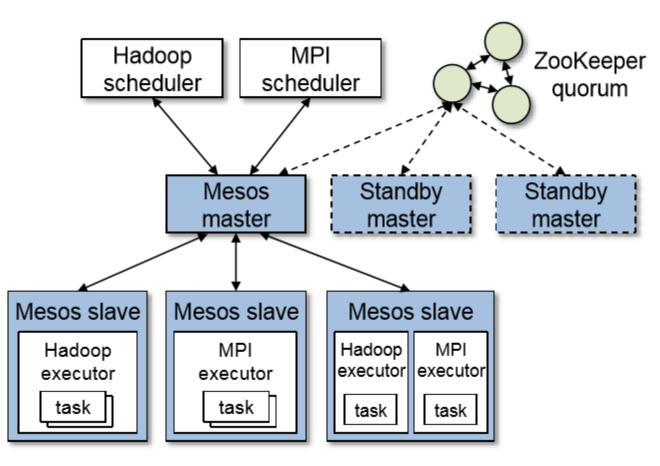
#### • Mesos Approach: Resource Offer

- Offer available resources to frameworks,
- Delegating control over scheduling to them.
- + Keeps Mesos simple, lets it support future frameworks,
- - Decentralized decisions might not be optimal!

### **Resource Offer**



### **Mesos Architecture**



Mesos architecture diagram, showing two running frameworks (Hadoop and MPI).

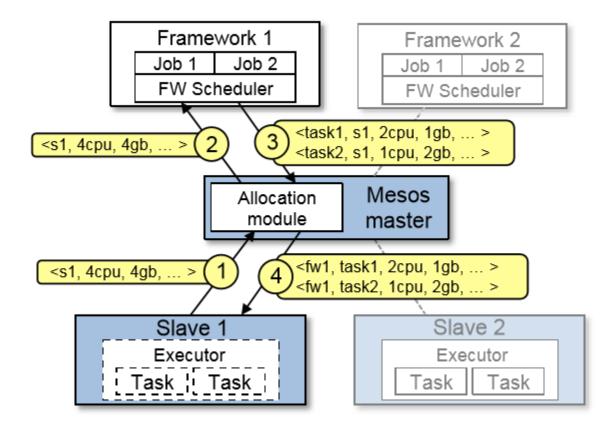
### Mesos Architecture: Resource Offer

- The **master** implements fine-grained sharing across frameworks using resource offers.
- Each resource offer is a list of free resources on multiple slaves.
- The master decides **how many** resources to offer to each framework according to an organizational policy.
- Mesos lets organizations define their own policies via a pluggable allocation module.

### Mesos Architecture: Resource Offer

- Each framework running on Mesos consists of two components:
  - A scheduler that registers with the **master** to be offered resources.
  - An executor process that is launched on slave nodes to run the framework's tasks.
- Frameworks' schedulers select which of the offered resources to use.

### Resource Offer: Example



### **Optimization: Filters**

• Let frameworks short-circuit rejection by providing a predicate on resources to be offered

- E.g. "nodes from list L" or "nodes with > 8 GB RAM"
- Could generalize to other hints as well

### Isolation

- Isolate resources using OS container technologies
  - Linux Containers (LXC)
  - Solaris Projects
- Limit the CPU, memory, network bandwidth, and (in new Linux kernels) I/O usage of a process tree.
- Not perfect, but much better than no isolation!

### Fault Tolerance

- Master's only state is the list of active slaves, active frameworks, and running tasks.
- Run multiple masters in a **hot-standby** configuration
  - Using **ZooKeeper**.



- To deal with scheduler failures, Mesos allows a framework to register multiple schedulers.
  - when one fails, another one is notified by the Mesos master to take over.

### Implementation

- About 10,000 lines of C++ code.
- Runs on Linux, Solaris and OS X.
- Frameworks ported: Apache Hadoop, MPICH2, Torque
- New specialized framework: Spark
  - for iterative jobs
  - up to **20**× faster than Hadoop
- & Open Source :)

### **Organizations using Mesos**

- **Twitter** uses Mesos on > 100 nodes to run ~12 production services (mostly stream processing).
- UCSF medical researchers are using Mesos to run Hadoop and eventually non-Hadoop apps.
- **Berkeley** machine learning researchers are running several algorithms at scale on Spark.
- Read more at:
  - <u>http://mesos.apache.org/documentation/latest/powered-by-mesos/</u>

### Mesos API

Scheduler Callbacks	Scheduler Actions
resourceOffer(offerId, offers) offerRescinded(offerId) statusUpdate(taskId, status) slaveLost(slaveId)	replyToOffer(offerId, tasks) setNeedsOffers(bool) setFilters(filters) getGuaranteedShare() killTask(taskId)
Executor Callbacks	Executor Actions
launchTask(taskDescriptor) killTask(taskId)	sendStatus(taskld, status)

### **Evaluation**

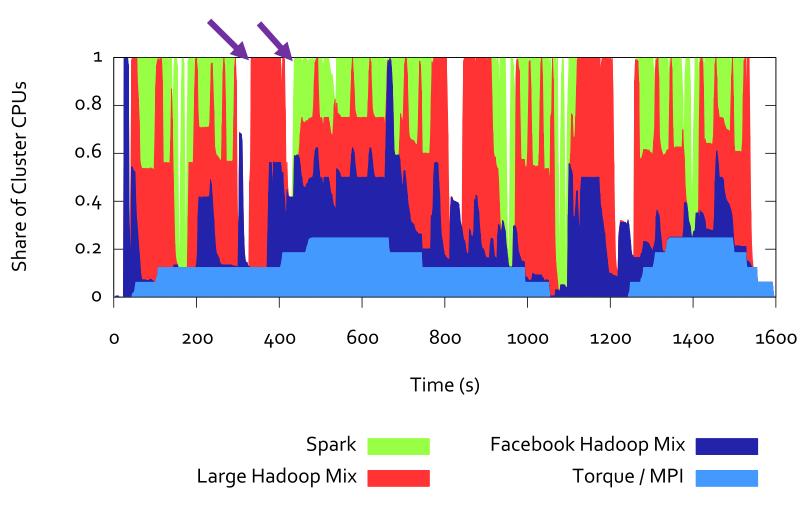
- Mesos is evaluated through a series of experiments on the Amazon Elastic Compute Cloud (EC2).
  - 96 Nodes
  - Nodes with 4 CPU cores and 15 GB of RAM.
- Four workloads in two scenarios:

96-node Mesos cluster using fair sharing

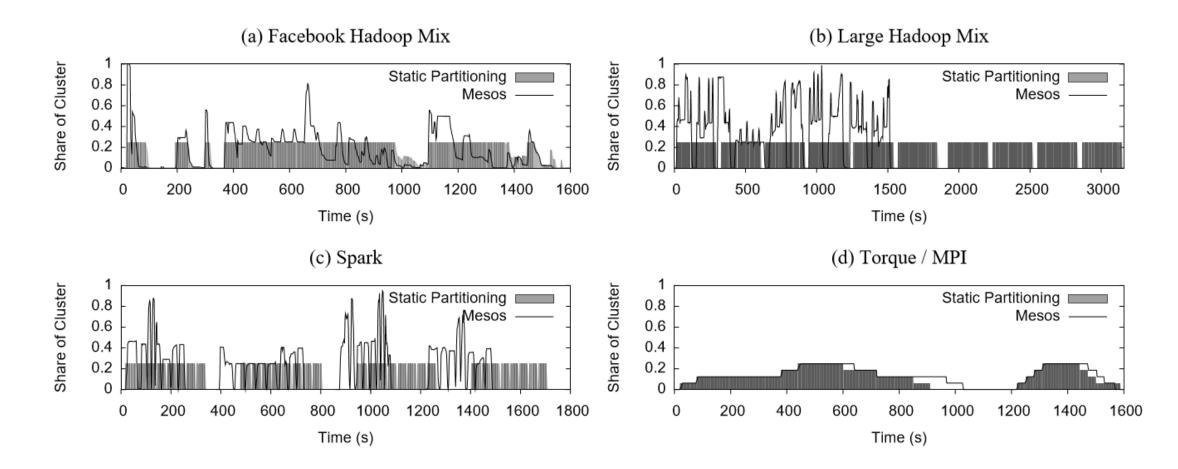
#### **V.S.**

> 4 static partitions of the cluster (24 nodes per partition)

### **Dynamic Resource Sharing**



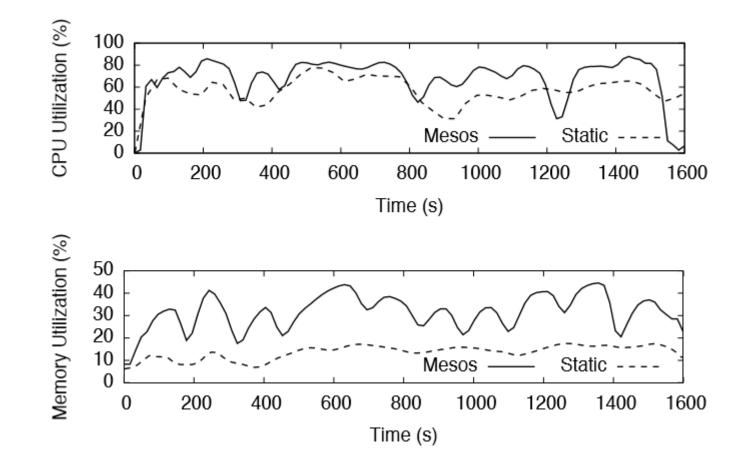
### Static vs Mesos: Resource Allocation



### Static vs Mesos: Speedup

Framework	Speedup on Mesos
Facebook Hadoop Mix	1.14 ×
Large Hadoop Mix	2.10 ×
Spark	1.26 ×
Torque / MPI	0.96× 🔶

#### Static vs Mesos: Utilization



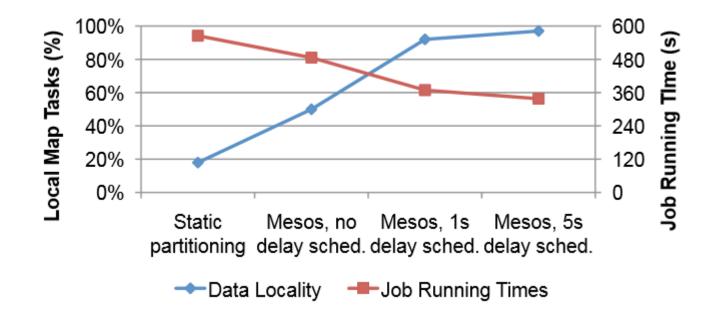
#### Mesos Overhead

- Ran two benchmarks using MPI and Hadoop on an EC2 cluster with **50 nodes**.
  - 2 CPU cores and 6.5 GB RAM
- The MPI job (LINPACK benchmark):
  - ~50.9s without Mesos  $\rightarrow$  ~51.8s with Mesos
- Hadoop job (WordCount job):
  - ~160s without Mesos  $\rightarrow$  ~166s with Mesos
- Result
  - In both cases, the overhead of using Mesos was less than 4%

## Data Locality with Resource Offers

- Ran 16 instances of Hadoop on a shared HDFS cluster
  - Using 93 EC2 nodes
  - 4 CPU cores and 15 GB RAM
- Used **delay scheduling** in Hadoop to get locality (wait a short time to acquire data-local nodes)
- Running the Hadoop instances on Mesos improves data locality.

#### Data Locality with Resource Offers



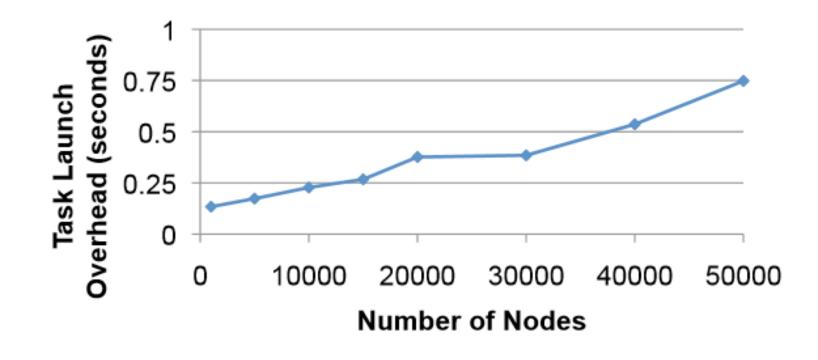


• Mesos only performs **inter-framework** scheduling (e.g. fair sharing), which is easier than **intra-framework** scheduling

#### • Result:

- Scaled to 50,000 emulated slaves
- 200 frameworks
- 100K tasks (30s len)

## **Scalability**



## **Results Analysis**

- Resource offer works well when:
  - Frameworks can scale up and down elastically
  - Task durations are homogeneous
  - Frameworks have many preferred nodes
- Otherwise Mesos may not be useful.
  - E.g. in the case of Torque / MPI
- Programming against Mesos is not easy :)

## **Related Work**

- HPC schedulers (e.g. Sun Grid Engine, Torque, ...)
  - Coarse-grained sharing for inelastic jobs (e.g. MPI jobs).
- Virtual machine clouds
  - Coarse-grained sharing similar to HPC schedulers.
- Condor
  - Centralized scheduler based on matchmaking

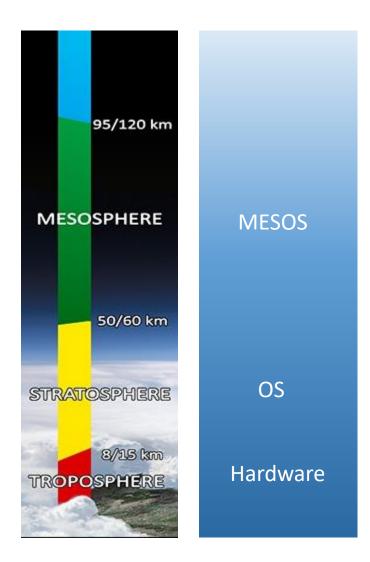
## Conclusion

- Mesos shares clusters efficiently among diverse frameworks.
- Two design elements:
  - Fine-grained sharing, at the level of tasks,
  - **Resource offers,** a scalable mechanism for application-controlled scheduling.
- Enables co-existence of current frameworks and development of new specialized ones.
- In use at Twitter, UC Berkeley, ...

## Conclusion

• What do you think about this?

 $\odot$ 



#### References

- [1] Hindman, B., Konwinski, A., Matei, Z., Ghodsi, A., D.Joseph, A., Katz, R., ... Stoica, I. (2011). Mesos: *A platform for fine-grained resource sharing in the data center*. *Proceedings of the ...*, 32. <u>https://doi.org/10.1109/TIM.2009.2038002</u>
- [2] *Apache Mesos*. (n.d.). Retrieved from <u>http://mesos.apache.org/</u>,2017
- [3] *Apache Hadoop*. (n.d.). Retrieved from <u>https://hadoop.apache.org/</u>, 2017
- [4] Zaharia, M., Borthakur, D., Sen Sarma, J., Elmeleegy, K., Shenker, S., & Stoica, I. (2010). *Delay scheduling*. Proceedings of the 5th European Conference on Computer Systems EuroSys '10, 265. <u>https://doi.org/10.1145/1755913.1755940</u>

# Thank you for your attention!

Any QUESTIONS? → m-zakeri@live.com