## Heterogeneous Resource Scheduling Using Apache Mesos for Cloud Native Frameworks

Sharma Podila

Senior Software Engineer
Netflix

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### Agenda

- Context, motivation
- Fenzo scheduler library
- Usage at Netflix
- Future direction

#### Why use Apache Mesos in a cloud?

Resource granularity

Application start latency

#### A tale of two frameworks

Reactive stream processing

Container deployment and management

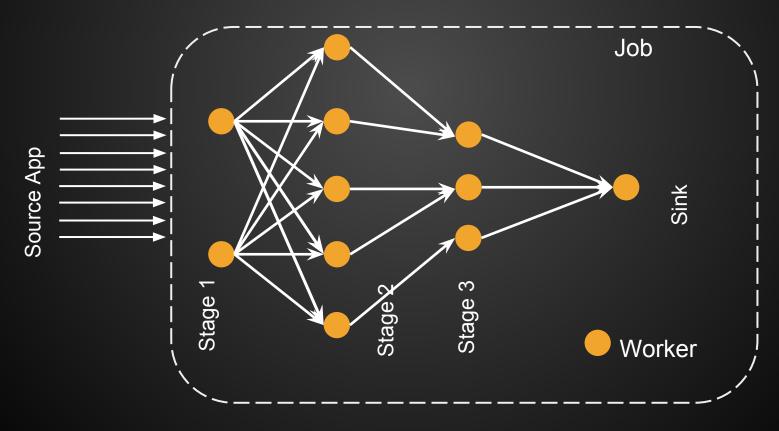
#### Reactive stream processing, Mantis

- Cloud native
- Lightweight, dynamic jobs
  - Stateful, multi-stage
  - Real time, anomaly detection, etc.
- Task placement constraints
  - Cloud constructs
  - Resource utilization
- Service and batch style



### Mantis job topology

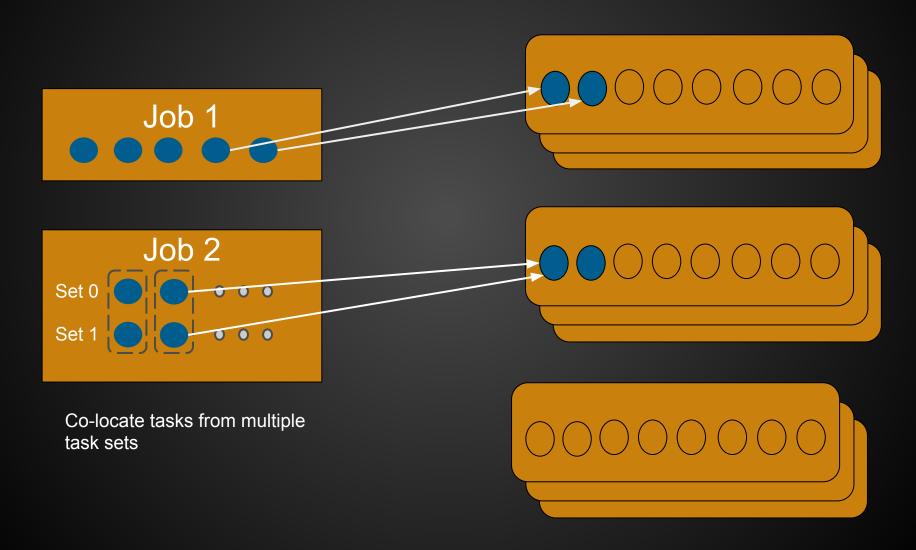
A job is set of 1 or more stages
A stage is a set of 1 or more workers
A worker is a Mesos task



#### Container management, Titan

- Cloud native
- Service and batch workloads
- Jobs with multiple sets of container tasks
- Container placement constraints
  - Cloud constructs
  - Resource affinity
  - Task locality

## Container scheduling model



## Why develop a new framework?

## Easy to write a new framework?

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What about scale?

Performance?

Fault tolerance?

Availability?

#### Easy to write a new framework?

What about scale?

Performance?

Fault tolerance?

Availability?

And scheduling is a hard problem to solve

# Long term justification is needed to create a new Mesos framework

#### Our motivations for new framework

 Cloud native (autoscaling)

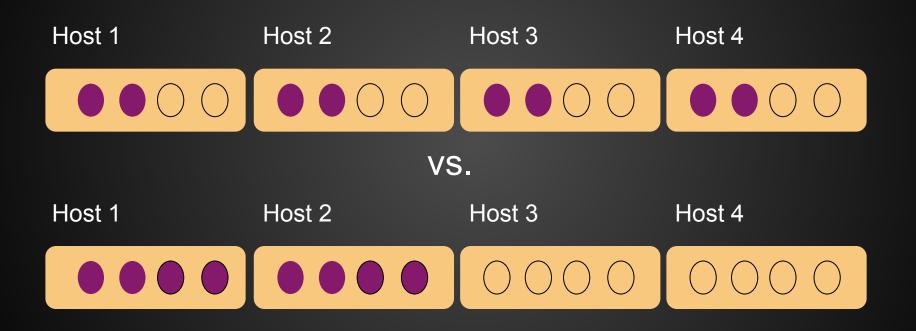
#### Our motivations for new framework

 Cloud native (autoscaling)

 Customizable task placement optimizations (Mix of service, batch, and stream topologies)

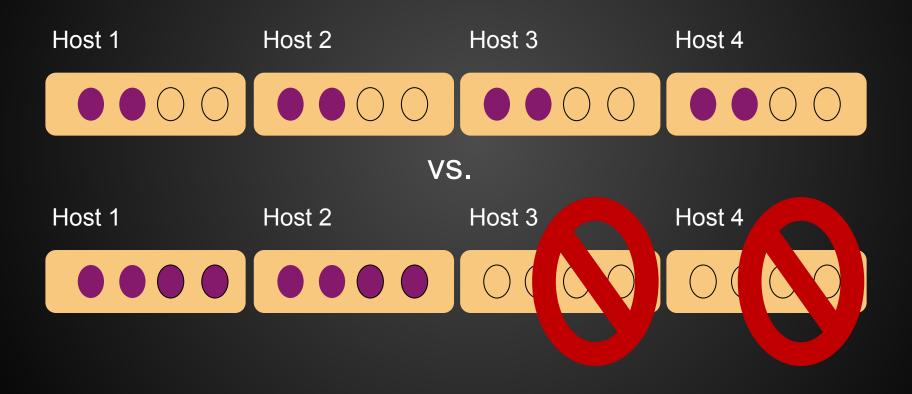
#### Cluster autoscaling challenge

For long running stateful services



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For long running stateful services



API for users to interact

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Be connected to Mesos via the driver

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Compute resource assignments for tasks

API for users to interact

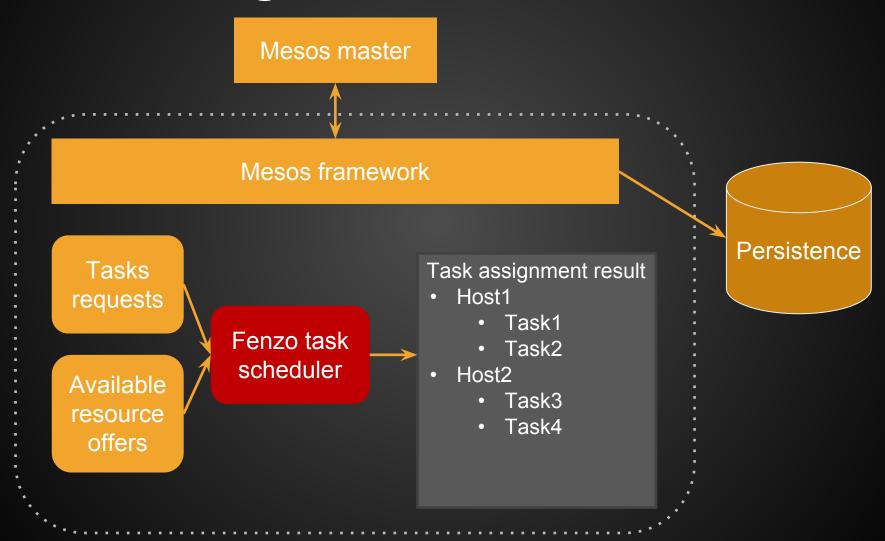
Be connected to Mesos via the driver

Compute resource assignments for tasks

#### Fenzo

A common scheduling library for Mesos frameworks

## Fenzo usage in frameworks



#### Fenzo scheduling library

Heterogeneous resources

Heterogeneous task requests

Autoscaling of cluster

Visibility of scheduler actions

Plugins for Constraints, Fitness

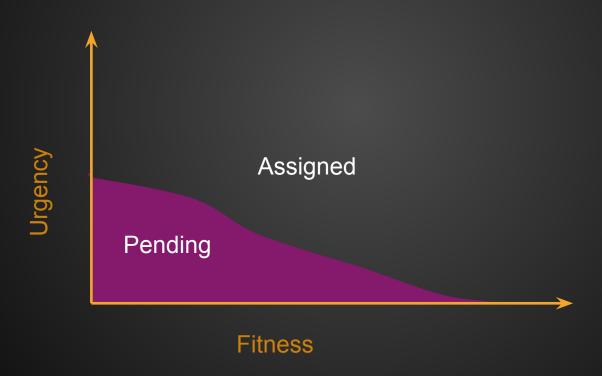
High speed

# Announcing availability of Fenzo in Netflix OSS suite

### Fenzo details

## Scheduling problem

N tasks to assign from M possible slaves



### Scheduling optimizations



<sup>&</sup>lt;sup>1</sup> Assuming tasks are not reassigned

#### Scheduling strategy

```
For each task

On each host

Validate <u>hard constraints</u>

Eval <u>fitness</u> and <u>soft constraints</u>

Until fitness good enough, and

A minimum #hosts evaluated
```

#### Task constraints

Soft Hard

#### Task constraints

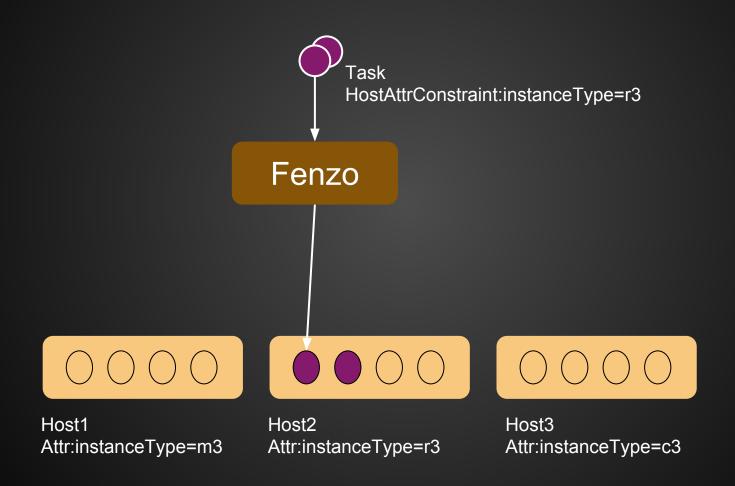
Soft

Hard

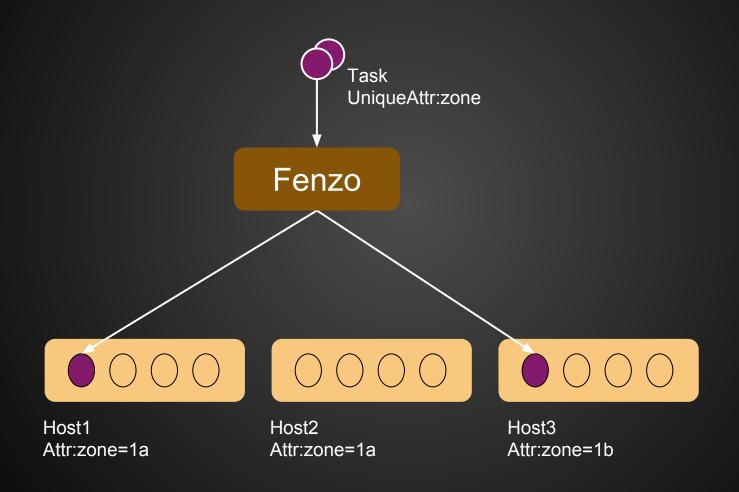
Extensible

#### **Built-in Constraints**

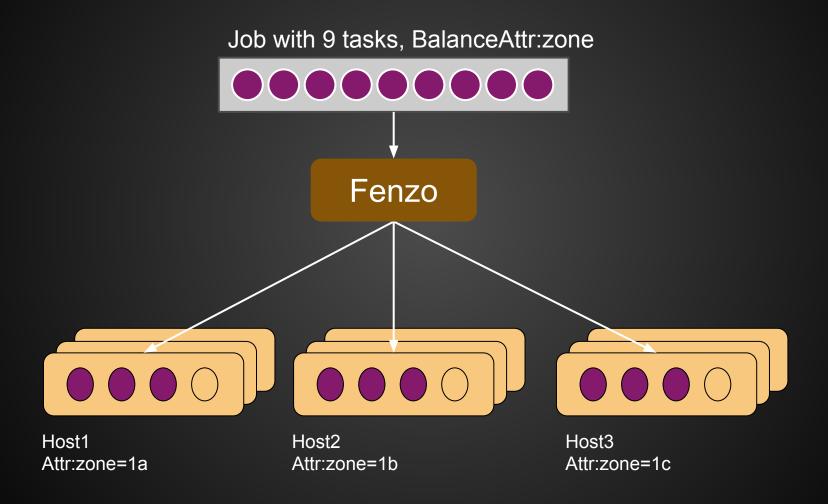
#### Host attribute value constraint



#### Unique host attribute constraint



#### Balance host attribute constraint



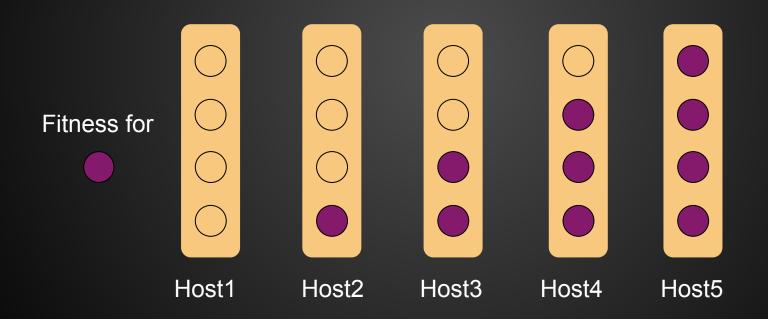
#### Fitness evaluation

Degree of fitness

Composable

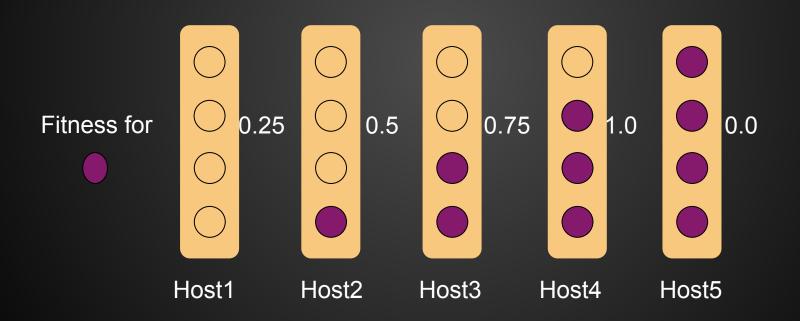
#### Bin packing fitness calculator

fitness = usedCPUs / totalCPUs



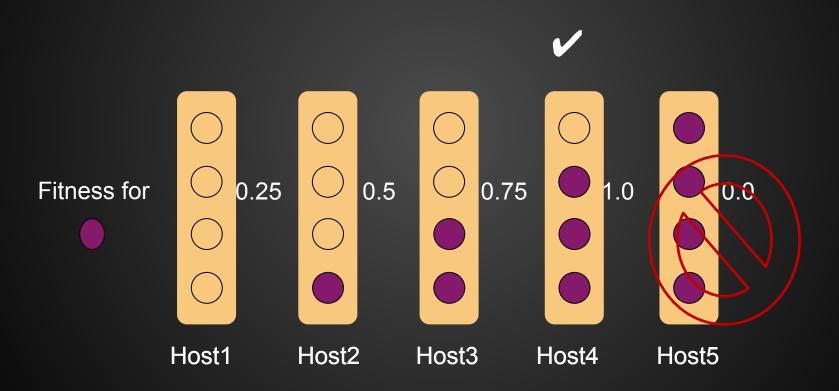
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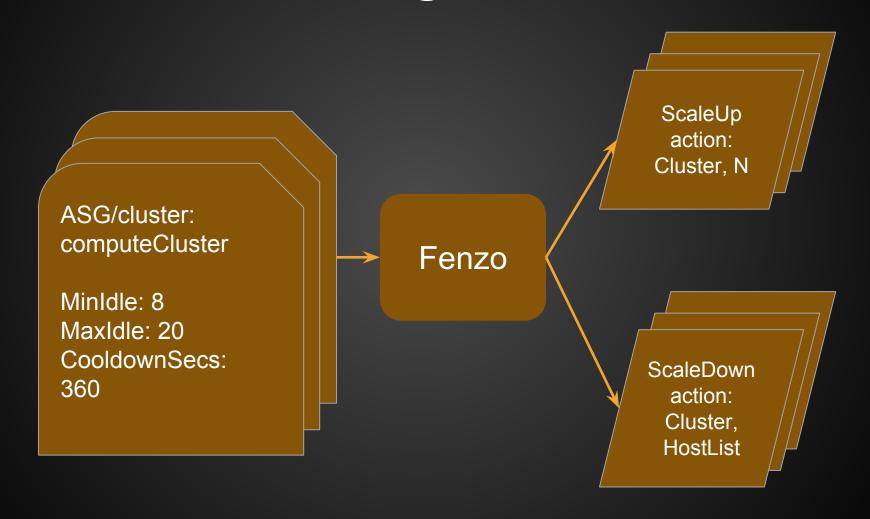
fitness = usedCPUs / totalCPUs



#### Composable fitness calculators

```
Fitness
= ( BinPackFitness * BinPackWeight +
RuntimePackFitness * RuntimeWeight
) / 2.0
```

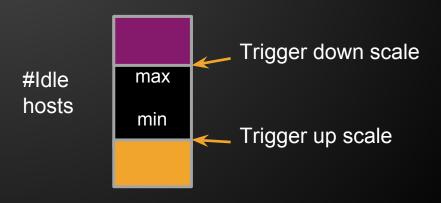
#### Cluster autoscaling in Fenzo



#### Rules based cluster autoscaling

- Set up rules per host attribute value
  - E.g., one autoscale rule per ASG/cluster, one cluster for networkintensive jobs, another for CPU/memory-intensive jobs
- Sample:

Cluster Name	Min Idle Count	Max Idle Count	Cooldown Secs
NetworkClstr	5	15	360
ComputeClstr	10	20	300

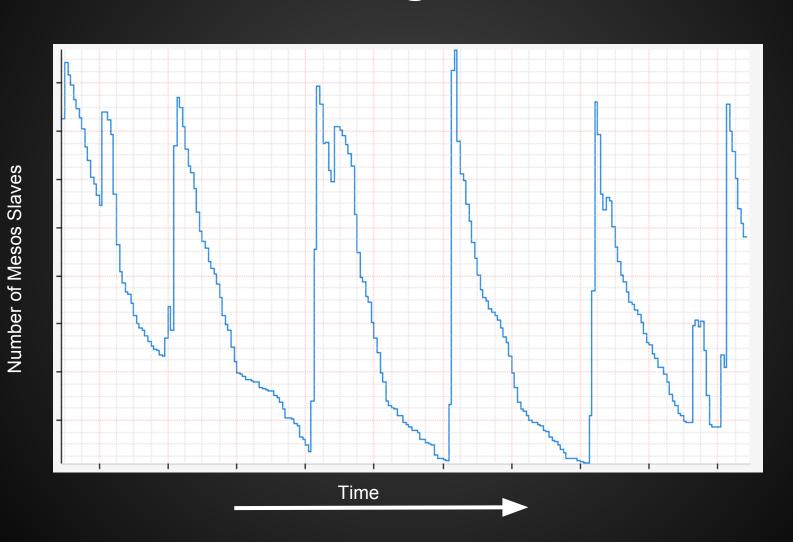


#### Shortfall analysis based autoscaling

- Rule-based scale up has a cool down period
  - What if there's a surge of incoming requests?
- Pending requests trigger shortfall analysis
  - Scale up happens regardless of cool down period
  - Remembers which tasks have already been covered

### Usage at Netflix

#### Cluster autoscaling



#### Scheduler run time (milliseconds)

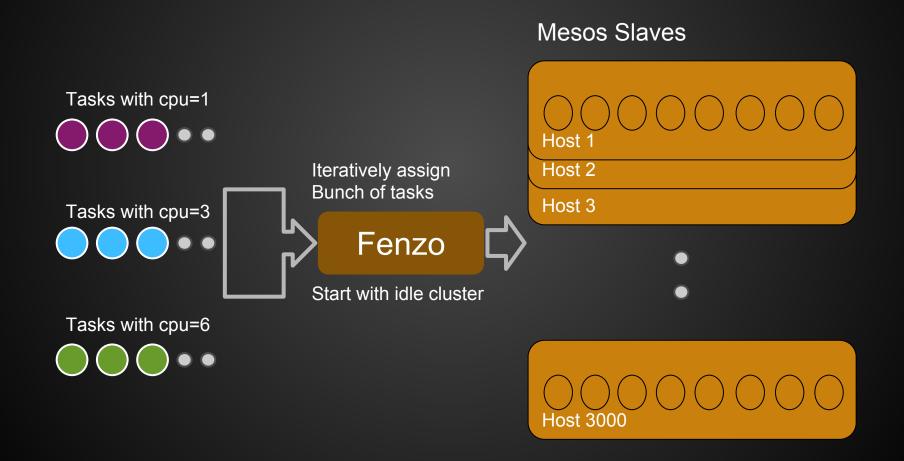
Scheduler run time in milliseconds (over a week)					
Average	2 mS				
Maximum	38 mS				

Note: times can vary depending on # of tasks, # and types of constraints, and # of hosts

#### **Experimenting with Fenzo**

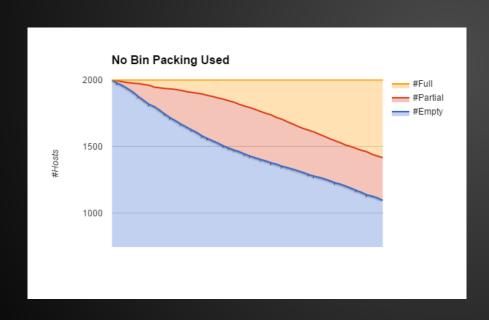
Note: Experiments can be run without requiring a physical cluster

#### A bin packing experiment



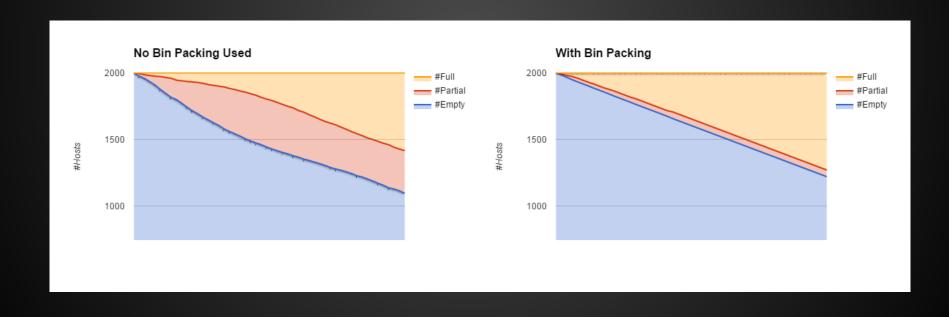
#### Bin packing sample results

Bin pack tasks using Fenzo's built-in CPU bin packer



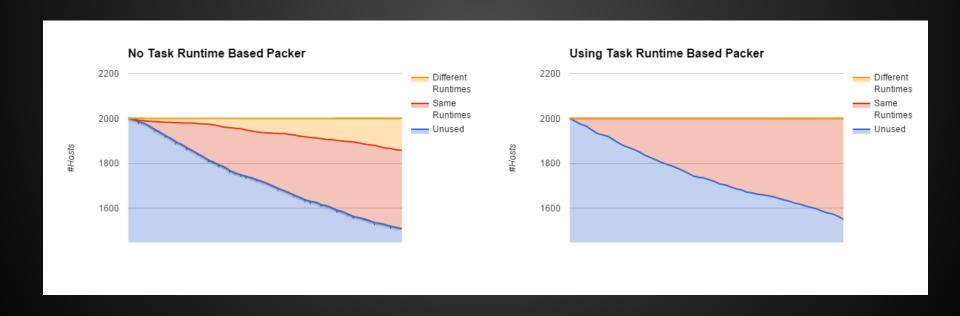
#### Bin packing sample results

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#### Task runtime bin packing sample

Bin pack tasks based on custom fitness calculator to pack short vs. long run time jobs separately



#### Scheduler speed experiment

Hosts: 8-CPUs each

Task mix: 20% running 1-CPU jobs, 40% running 4-CPU, and 40% running 6-CPU jobs

**Goal**: starting from an empty cluster, assign tasks to fill all hosts

**Scheduling strategy**: CPU bin packing

# of hosts	# of tasks to assign each time	Avg time	Avg time per task		Max time	Total time
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# of hosts	# of tasks to assign each time	Avg time	Avg time per task	Min time	Max time	Total time
1,000	1	3 mS	3 mS	1 mS	188 mS	9 s
1,000	200	40 mS	0.2 mS	17 mS	100 mS	0.5 s

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1,000	200	40 mS	0.2 mS	17 mS	100 mS	0.5 s
10,000	1	29 mS	29 mS	10 mS	240 mS	870 s
10,000	200	132 mS	0.66 mS	22 mS	434 mS	19 s

#### **Accessing Fenzo**

Code at

https://github.com/Netflix/Fenzo

Wiki at

https://github.com/Netflix/Fenzo/wiki

#### **Future directions**

- Task management SLAs
- Support for newer Mesos features
- Collaboration

#### To summarize...

## Fenzo: scheduling library for frameworks

Heterogeneous resources

Heterogeneous task requests

Autoscaling of cluster

Visibility of scheduler actions

Plugins for Constraints, Fitness

High speed

# Fenzo is now available in Netflix OSS suite at

https://github.com/Netflix/Fenzo

#### Questions?

### Heterogeneous Resource Scheduling Using Apache Mesos for Cloud Native Frameworks

Sharma Podila

spodila @ netflix . com

