# **Owl**: Performance-Aware Scheduling for Resource-Efficient Function-as-a-Service Cloud

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# FaaS Gaining Popularity

A new report from Datadog has found that serverless computing could be entering the **mainstream** with **over half** of all organizations using serverless ...

—— TechCrunch¹, Jun. 2022

"

AWS Lambda ..., and more than **a million** customers are using it today, according to AWS.

—— Protocol<sup>2</sup>, Aug. 2022

"

<sup>1.</sup> Datadog finds serverless computing is going mainstream, https://tcrn.ch/3D5GhHB

<sup>2.</sup> Amazon's Werner Vogels: Enterprises are more daring than you might think, https://bit.ly/3F8Xtij

# FaaS Gaining Popularity

A new report from Datadog has found that serverless computing could be entering the **mainstream** with **over half** of all organizations using

Problem: How to serve functions efficiently?

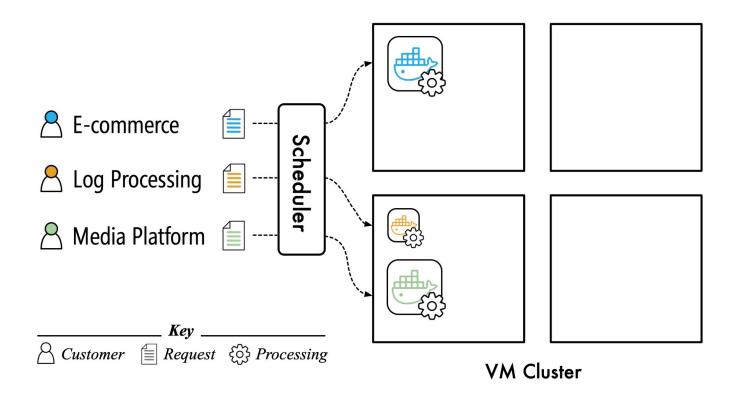


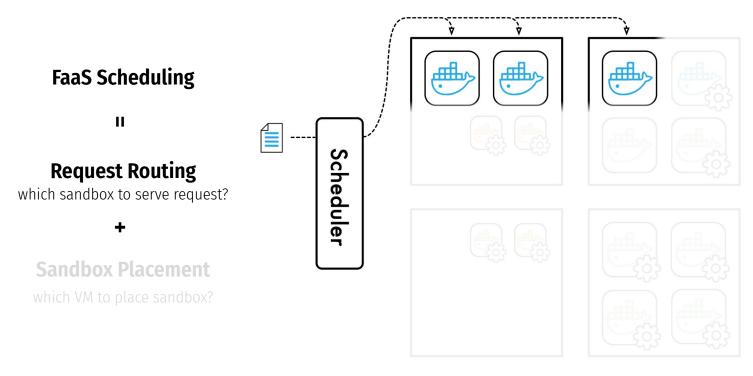
—— Protocol<sup>2</sup>, Aug. 202

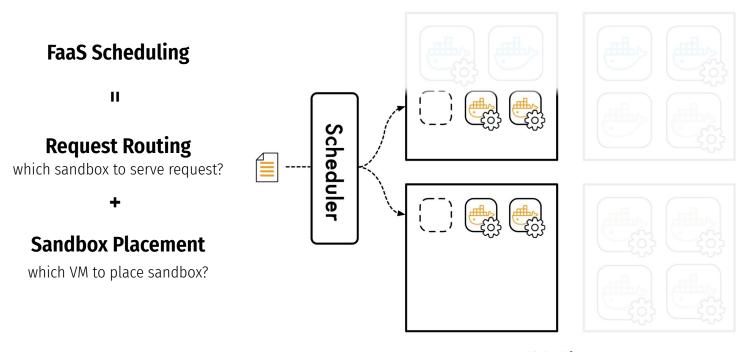
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### **FaaS Scheduling**

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### **Request Routing**

which sandbox to serve request?

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### **Sandbox Placement**

which VM to place sandbox?

#### **Model\***

- 1. Each **VM** has a memory capacity.
- 2. Each **sandbox** has a memory size.

#### Goal

Pack sandboxes onto VMs.

<sup>\*</sup> similar as bin-packing

### Status Quo

### **FaaS Scheduling**

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### **Request Routing**

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### Setting\*

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### Goal

Pack sandboxes onto VMs.

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### **State of the Practice**

П

### **Most-Recently Used**

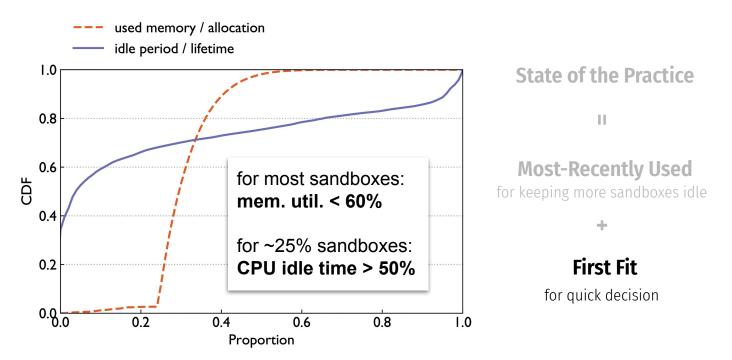
for keeping more sandboxes idle

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### **First Fit**

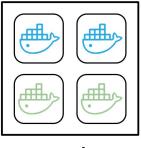
for quick decision

### Resource Inefficiency



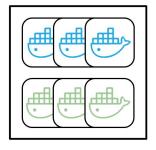
<sup>\*</sup> data collected from an one-day production trace

### **Naive Overcommitment**



**—**Sandbox Overcommitment→

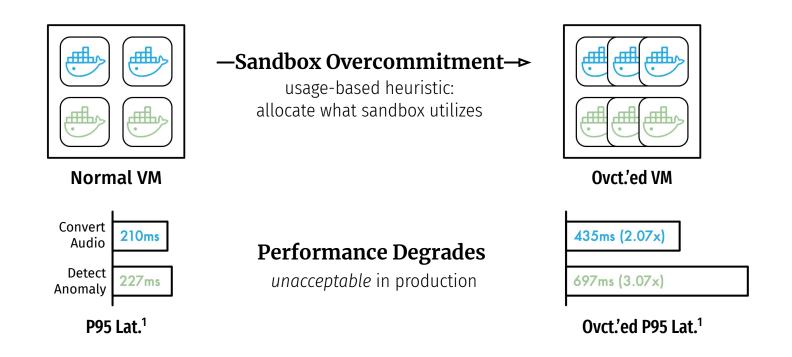
usage-based heuristic: allocate what sandbox utilizes



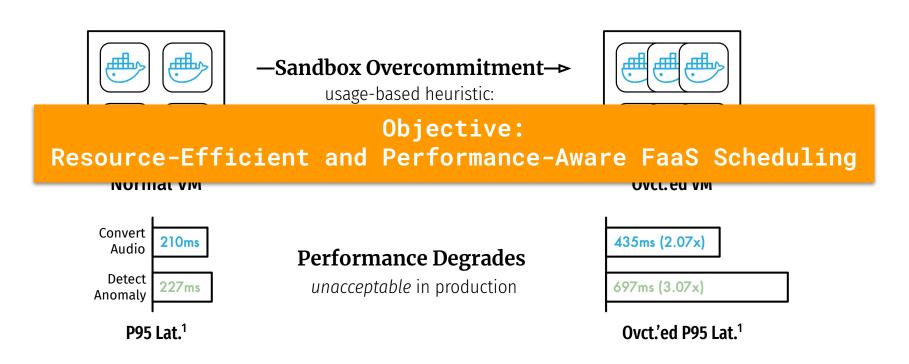
Normal VM

Ovct.'ed VM

### Naive Overcommitment Falls Short



### Naive Overcommitment Falls Short

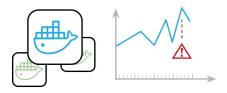


### Outline

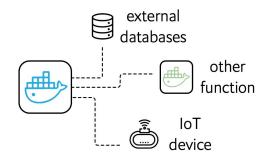
- 1. Background and Motivation
- 2. Collocation Profiling
- 3. Profile-Guided Overcommitment
- 4. Performance-Monitored Overcommitment
- 5. Owl and its Evaluation

# Restrictions for Profiling

No latency degradation.



No **offline** invocation (b/c of side-effect).



# Restrictions for Profiling

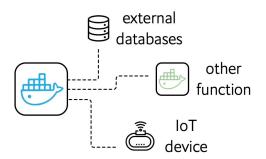
No latency **degradation**.



Ideal profiling:

in-production performance-protected

No **offline** invocation (b/c of side-effect).



# New Technique: Collocation Profiling

**Key Question** 

How many sandboxes can a VM host?

in-production performance-protected

### Collocation Profiling

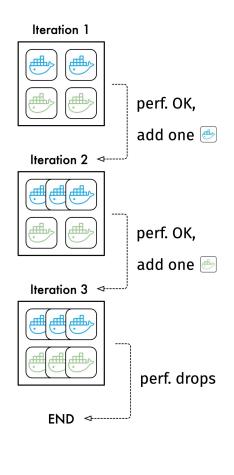
#### **Key Question**

How many sandboxes can a VM host?

#### **Procedure**

- 1. **Saturate** sandboxes with requests.
- 2. Iteratively add more sandboxes ...
- 3. ... until perf. starts dropping.





# Collocation Profiling

#### **Key Question**

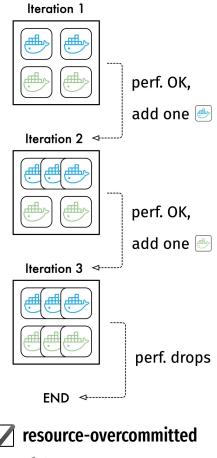
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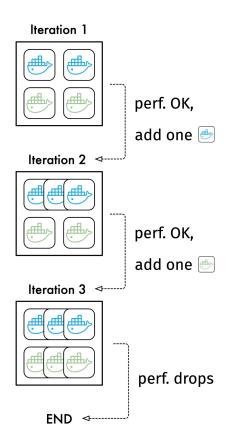




# Collocation Profiling

Collocation <sup>2</sup>	Util. <sup>1</sup>
<b>ĕ</b> ×3 <b>ĕ</b> ×3	1.48
<b>⊕</b> ×3 <b>⊕</b> ×5	1.42
<b>⊕</b> ×3 <b>⊕</b> ×3	1.26
<b>€</b> ×5	1.22
<b>*</b> 5	1.24
<b>ĕ</b> ×7	1.20

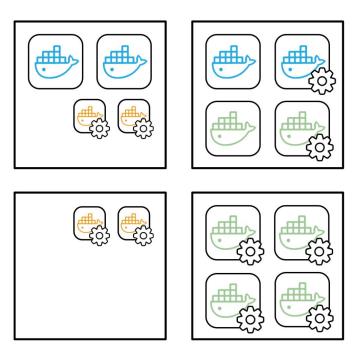
- 1. Util. = Allocated Memory / Total VM Memory (> 1 means overcommitment)
- 2. Limited to two functions b/c of complexity.



Collocation	Util.	
<b>⊛</b> ×3 <b>⊛</b> ×3	1.48	
<b>⊕</b> ×3 <b>⊕</b> ×5	1.42	
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<b>€</b> ×7	1.20	

### **Key Question**

How to place sandboxes using profiles?



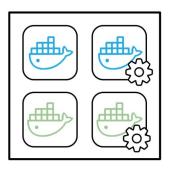
**VM Cluster** 

Collocation	Util.
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### **Greedy Algorithm**

Collocating sandboxes with highest util.







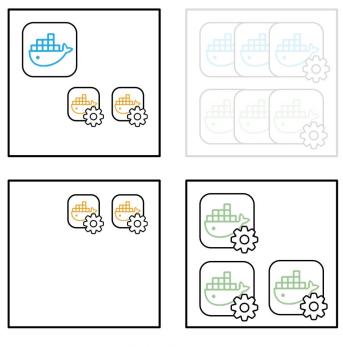


**VM Cluster** 

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### **Greedy Algorithm**

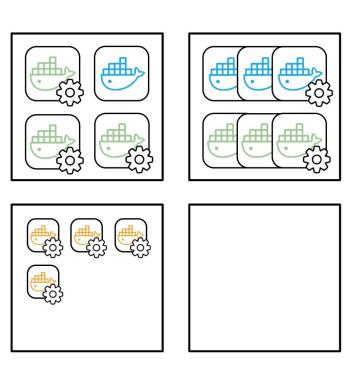
Collocating sandboxes with highest util.



**VM Cluster** 

#### **Offline**

Collocating sandboxes with highest util.



**VM Cluster** 

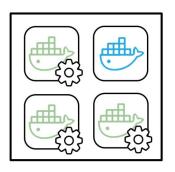
#### **Offline**

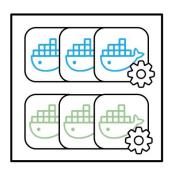
Collocating sandboxes with highest util.

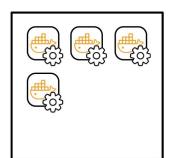
#### **Online**

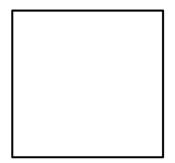
- 1. Periodically update placement.
- 2. Only include VMs with **sandbox change**.

[more details and optimizations in paper]





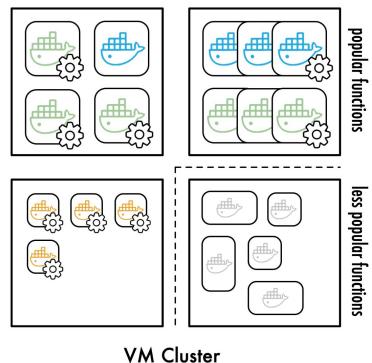




#### **Problem**

profiling requires continuous requests

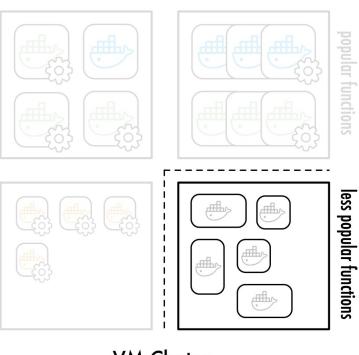
⇒ only applies to popular functions



### Performance-Monitored Overcommitment

#### **Solution**

- 1. Usage-based overcommitment.
- 2. Keep **monitoring** performance.
- 3. **Remedy** degradation (e.g., sandbox migration).

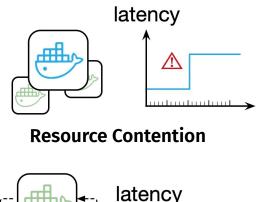


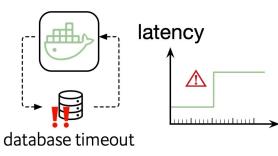
**VM Cluster** 

### Problem: External Degradation

#### **Solution**

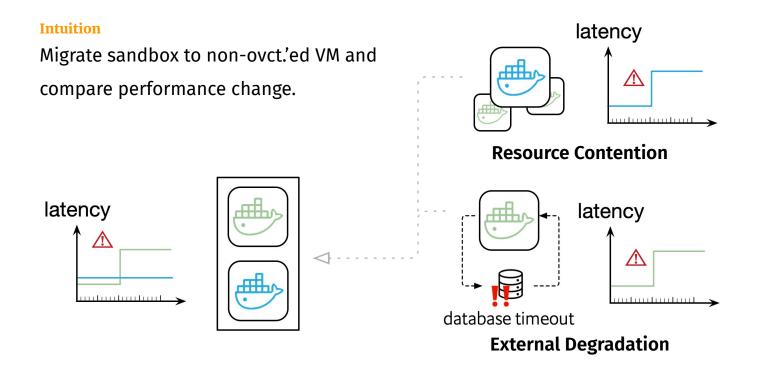
- 1. Usage-based overcommitment.
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**External Degradation** 

# New Technique: Comparative Validation



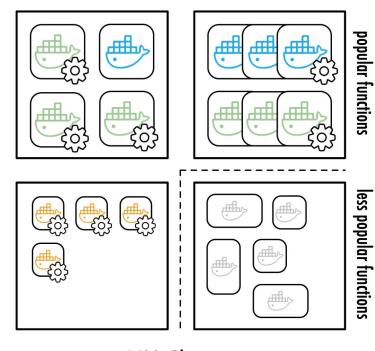
# Putting it all Together

#### **Popular Functions**

- 1. Profile Collocations
- 2. Collocate Sandboxes
- 3. Consolidate Idle Ones

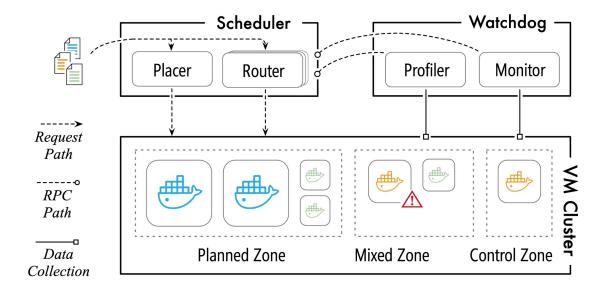
#### **Less Popular Functions**

- 1. Monitor Performance
- 2. Remedy Degradation
- 3. Validate Cause

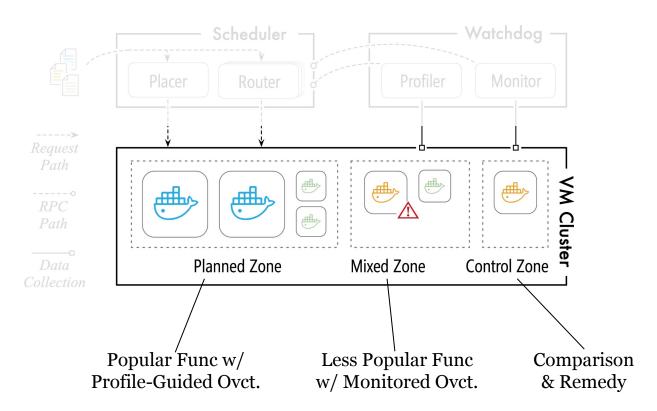


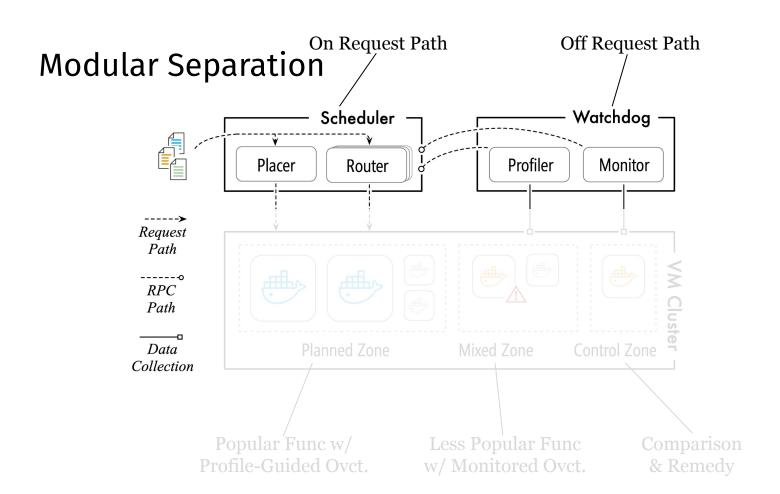
**VM Cluster** 

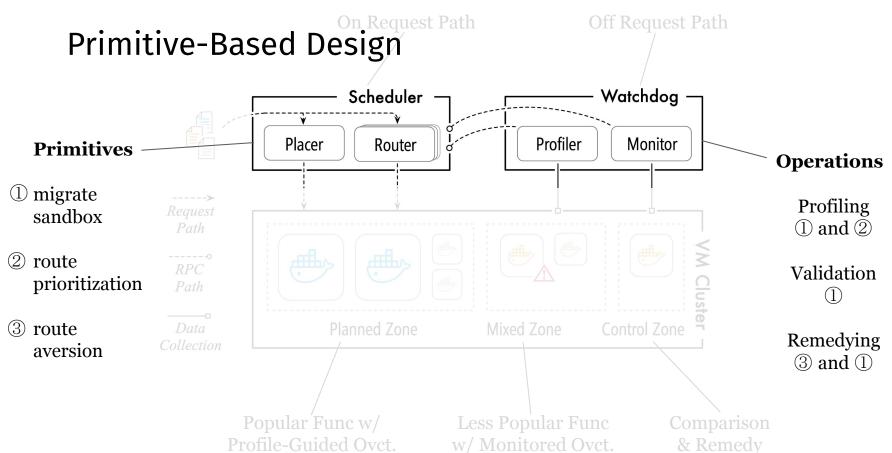
# **Implementation**



### **Cluster Zoning**







# Evaluation Highlight

### **Setting**

- three 20-minute production workloads
- · four variant schedulers with different techniques enabled

More results in the paper.

- (1) scheduling latency
- (2) microbenchmarking
- (3) large-scale validation

Results	VM Cost	Function Performance
① Baseline	-	not degraded
② Naive Overcommitment	decrease by 44.26% from ①	degraded
③ /w Profile Guidance	increase by 9.24% from ②	not degraded
④ /w Consolidation	decrease by 8.53% from ③	not degraded
Overall	decrease by 43.8% from ①	not degraded

# Open-Source Benchmark

Abbreviation	Function	Memory Size	Actual Usage	Lanaguage	Dependencies
QV	Query Vacancy	256 MiB	~70MiB	JavaScript	Key-Value Store
RS	Reserve Spot	256 MiB	~70MiB	JavaScript	Key-Value Store, Message Queue
AL	Anonymize Log	1024 MiB	~20MiB	Rust	Message Queue
FL	Filter Log	1024 MiB	~20MiB	Rust	Message Queue
DO	Detect Object	3072 MiB	~1700MiB	Python	Model Serving Framework
CI	Classify Image	2560 MiB	~500MiB	Python	Model Serving Framework
GMM	Get Media Meta	128 MiB	~20MiB	Python	Object Store
CA	Convert Audio	256 MiB	~100MiB	Python	Object Store
ID	Ingest Data	768 MiB	~10MiB	C++	SQL Database
DA	<b>Detect Anomaly</b>	768 MiB	~10MiB	C++	SQL Database

https://github.com/All-less/faas-scheduling-benchmark

### Conclusion

Performance-Aware and Resource-Efficient Scheduling in Public FaaS

- 1. Profile-Guided Overcommitment
- 2. Performance-Monitored Overcommitment