

SeBS: a Serverless Benchmark Suite for Function-as-a-Service Computing

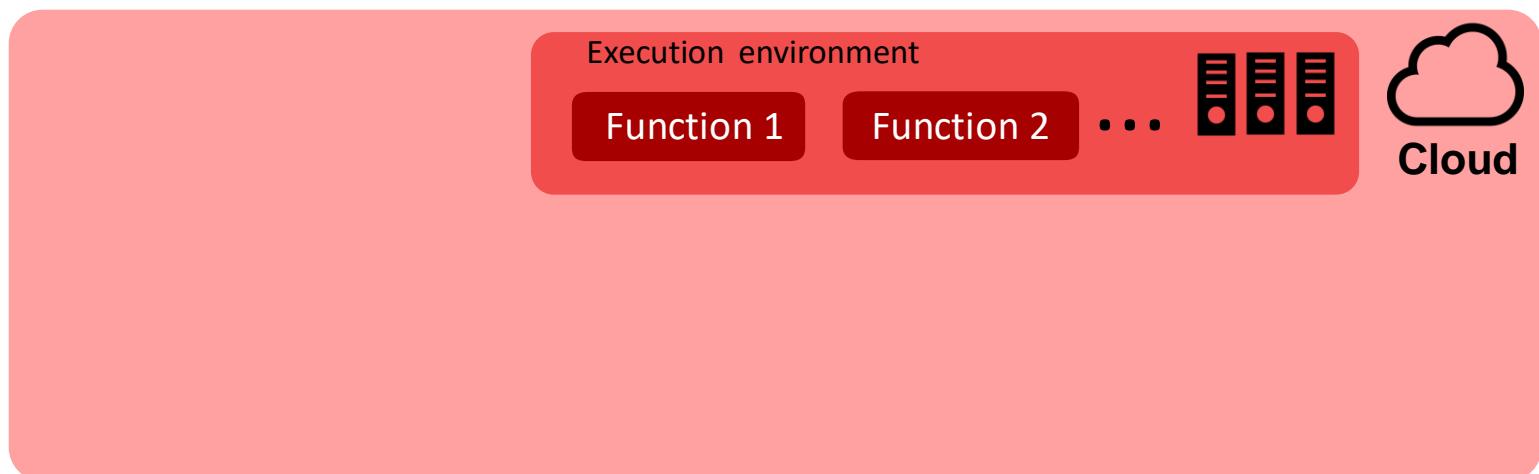
Marcin Copik, Grzegorz Kwasniewski, Maciej Besta, Michał Podstawski, Torsten Hoefer



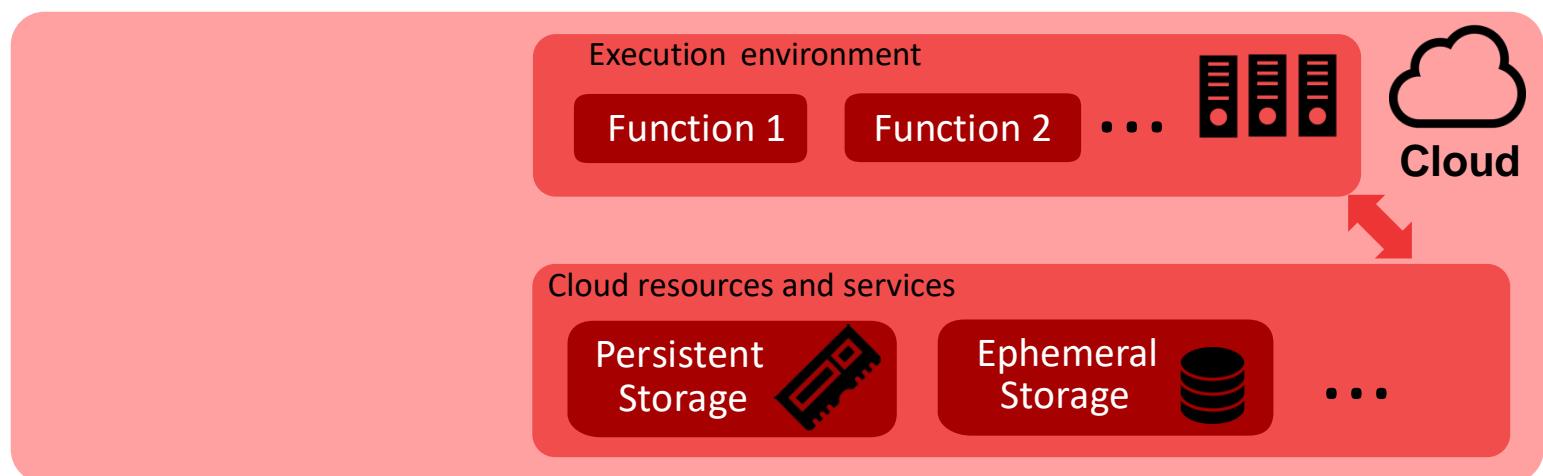
What is serverless?



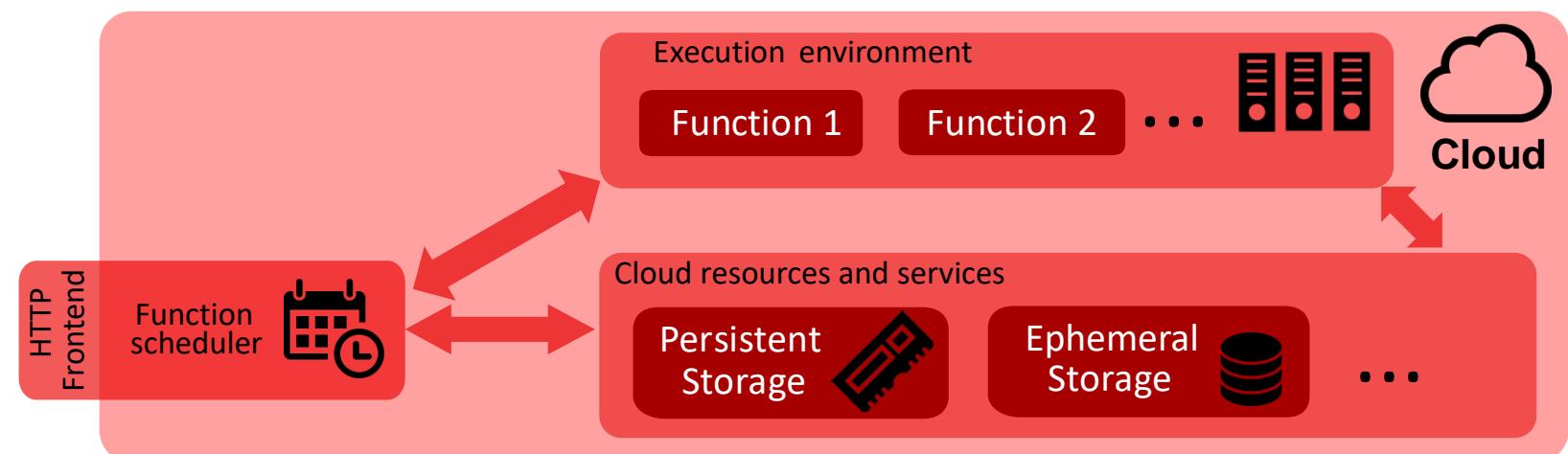
What is serverless?



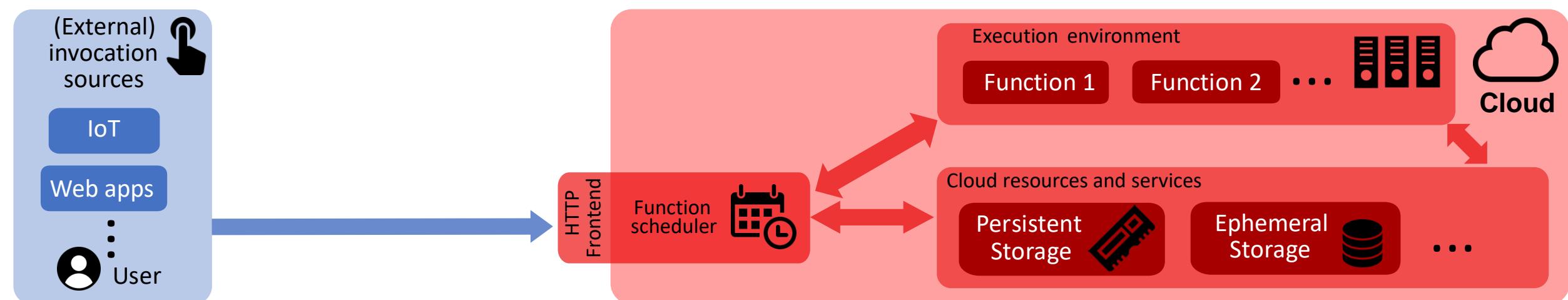
What is serverless?



What is serverless?



What is serverless?



Serverless – why & why not?



Serverless – why & why not?



User

- 👍 Pay-as-you-go billing
- 👍 Massive parallelism
- 👍 Simplified deployment
- 👍 Architecture agnostic



Cloud Provider

Serverless – why & why not?



User

- 👍 Pay-as-you-go billing
- 👍 Massive parallelism
- 👍 Simplified deployment
- 👍 Architecture agnostic



Cloud Provider

- 👍 Higher machine utilization
- 👍 Fine-grained scheduling

Serverless – why & why not?



User

- 👍 Pay-as-you-go billing
- 👍 Massive parallelism
- 👍 Simplified deployment
- 👍 Architecture agnostic

- 👎 High computing cost
- 👎 Variable performance
- 👎 Vendor lock-in
- 👎 Black-box platform



- 👍 Higher machine utilization
- 👍 Fine-grained scheduling

Serverless – why & why not?



User

- 👍 Pay-as-you-go billing
- 👍 Massive parallelism
- 👍 Simplified deployment
- 👍 Architecture agnostic

- 👎 High computing cost
- 👎 Variable performance
- 👎 Vendor lock-in
- 👎 Black-box platform



Cloud Provider

- 👍 Higher machine utilization
- 👍 Fine-grained scheduling

- 👎 Handling heterogeneity
- 👎 Micro-architecture effects

Commercial serverless systems

Policy

AWS Lambda

Azure Functions

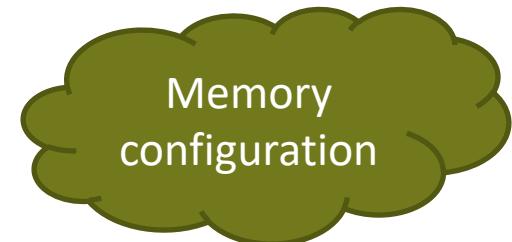
Google Cloud Functions

Commercial serverless systems



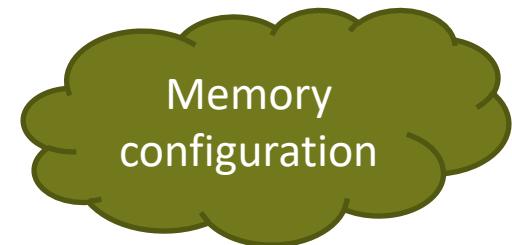
Policy	AWS Lambda	Azure Functions	Google Cloud Functions
Languages (native)	Python, Node.js, C#, Java, C++, etc.	Python, TypeScript, C#, Java, etc.	Node.js, Python, Java, Go.

Commercial serverless systems

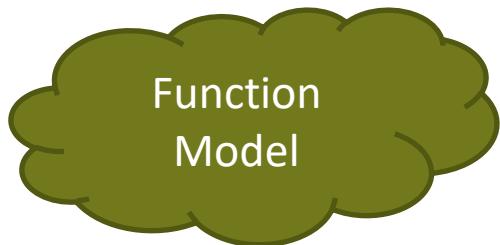


Policy	AWS Lambda	Azure Functions	Google Cloud Functions
Languages (native)	Python, Node.js, C#, Java, C++, etc.	Python, TypeScript, C#, Java, etc.	Node.js, Python, Java, Go.
Memory Allocation	Static, 128 – 3008 MB.	Dynamic, up to 1536 MB.	Static, 128, 256, 512, 1024, 2048 MB.

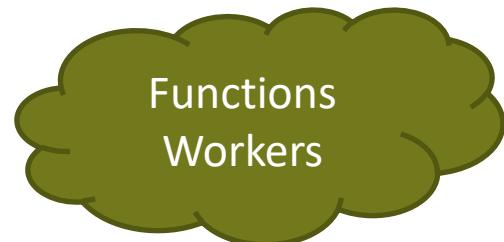
Commercial serverless systems



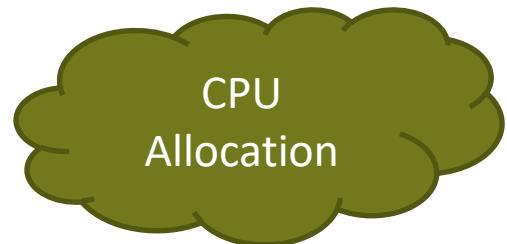
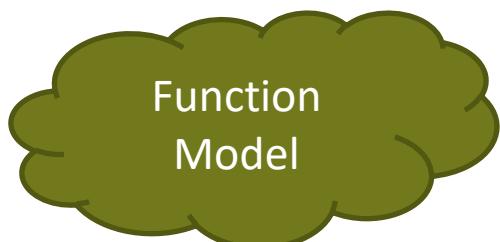
Policy	AWS Lambda	Azure Functions	Google Cloud Functions
Languages (native)	Python, Node.js, C#, Java, C++, etc.	Python, TypeScript, C#, Java, etc.	Node.js, Python, Java, Go.
Memory Allocation	Static, 128 – 3008 MB.	Dynamic, up to 1536 MB.	Static, 128, 256, 512, 1024, 2048 MB.
Concurrency Limit	1000 Functions	200 Function Apps.	100 Functions.



Commercial serverless systems



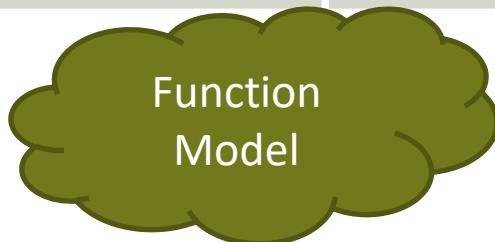
Policy	AWS Lambda	Azure Functions	Google Cloud Functions
Languages (native)	Python, Node.js, C#, Java, C++, etc.	Python, TypeScript, C#, Java, etc.	Node.js, Python, Java, Go.
Memory Allocation	Static, 128 – 3008 MB.	Dynamic, up to 1536 MB.	Static, 128, 256, 512, 1024, 2048 MB.
Concurrency Limit	1000 Functions	200 Function Apps.	100 Functions.
CPU Allocation	Proportional to memory, 1 vCPU on 1792 MB.	Unknown.	Proportional to memory, 2.4 GHz CPU at 2048 MB.



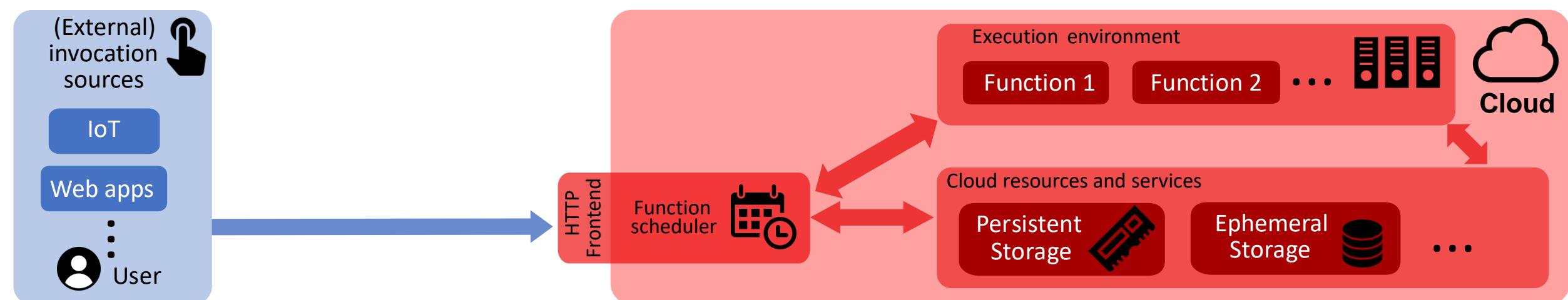
Commercial serverless systems



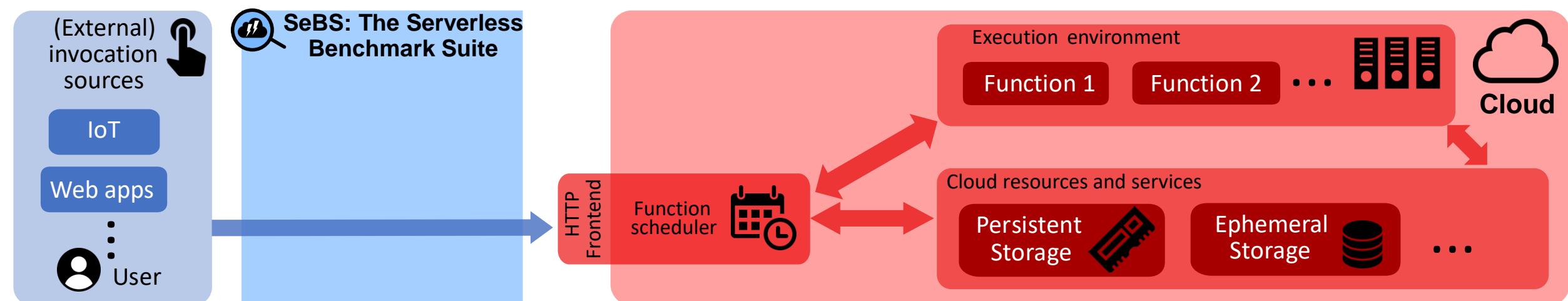
Policy	AWS Lambda	Azure Functions	Google Cloud Functions
Languages (native)	Python, Node.js, C#, Java, C++, etc.	Python, TypeScript, C#, Java, etc.	Node.js, Python, Java, Go.
Memory Allocation	Static, 128 – 3008 MB.	Dynamic, up to 1536 MB.	Static, 128, 256, 512, 1024, 2048 MB.
Concurrency Limit	1000 Functions	200 Function Apps.	100 Functions.
CPU Allocation	Proportional to memory, 1 vCPU on 1792 MB.	Unknown.	Proportional to memory, 2.4 GHz CPU at 2048 MB.
Billing	Duration and declared memory.	Average memory use, duration.	Duration, declared CPU and memory.
Deployment	zip package up to 250 MB.	zip package, Docker image.	zip package, up to 100 MB.
Time Limit	15 minutes	10 min / 60 min / unlimited.	9 minutes.



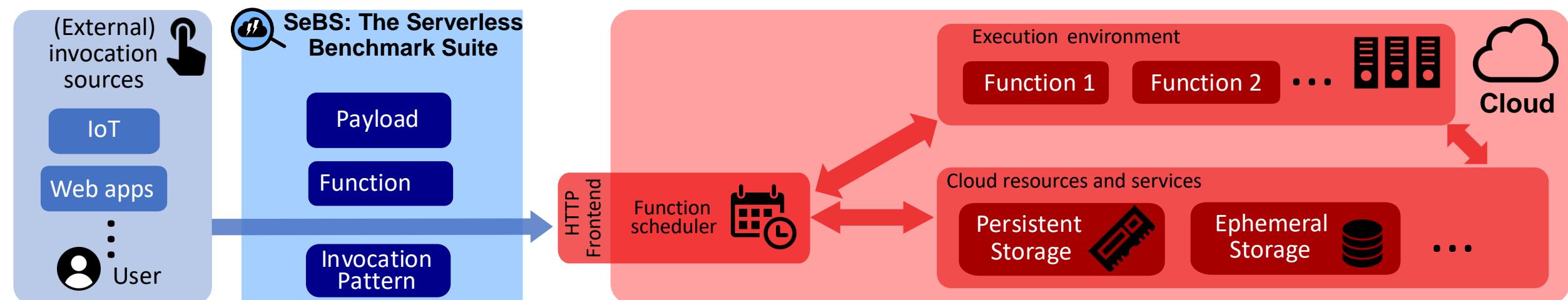
What is serverless?



What is serverless?



What is serverless?



Benchmark Applications

Type	Name	Language
------	------	----------

Benchmark Applications

Type	Name	Language
Webapps	uploader	Python, Node.js

Benchmark Applications

Type	Name	Language
Webapps	uploader	Python, Node.js
Multimedia	thumbnailer	Python, Node.js, C++

Benchmark Applications

Type	Name	Language
Webapps	uploader	Python, Node.js
Multimedia	thumbnailer	Python, Node.js, C++
Utilities	compression	Python

Benchmark Applications

Type	Name	Language
Webapps	uploader	Python, Node.js
Multimedia	thumbnailer	Python, Node.js, C++
Utilities	compression	Python
Inference	image-recognition	Python, C++

Benchmark Applications

Type	Name	Language
Webapps	uploader	Python, Node.js
Multimedia	thumbnailer	Python, Node.js, C++
Utilities	compression	Python
Inference	image-recognition	Python, C++
Scientific	graph-bfs	Python

Results and Insights

Results and Insights

Results and Insights

	Results, methods, and insights
	
	
.	

Results and Insights

Results, methods, and insights	
	
\$	
	

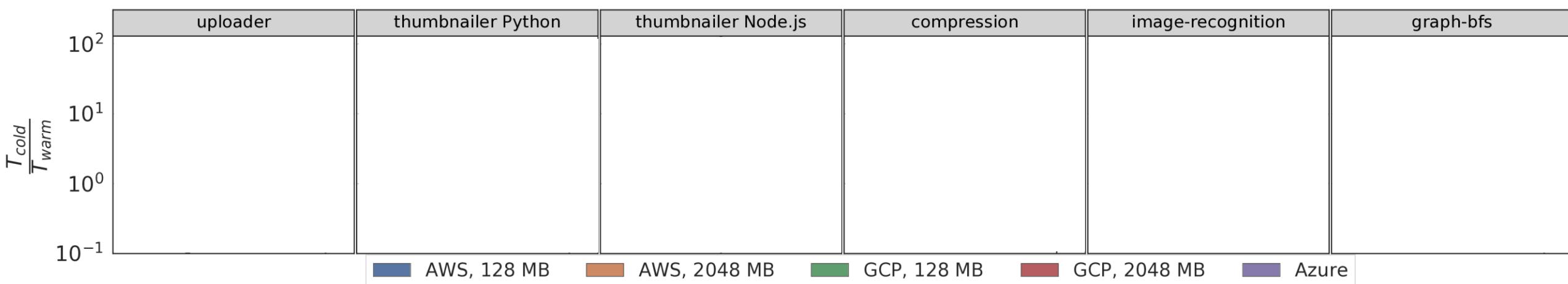
Results and Insights

Results, methods, and insights	
	
\$	
	
	

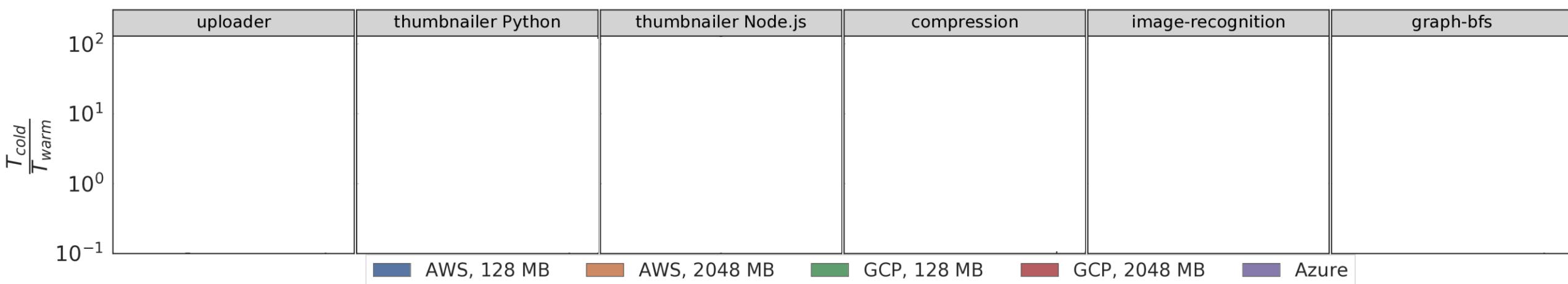
Results and Insights

	Results, methods, and insights
	<p>High-memory allocations increase cold startup overheads on GCP.</p> <p>GCP functions experience reliability and availability issues.</p>
	
	
	

Performance Analysis: Cold Startups



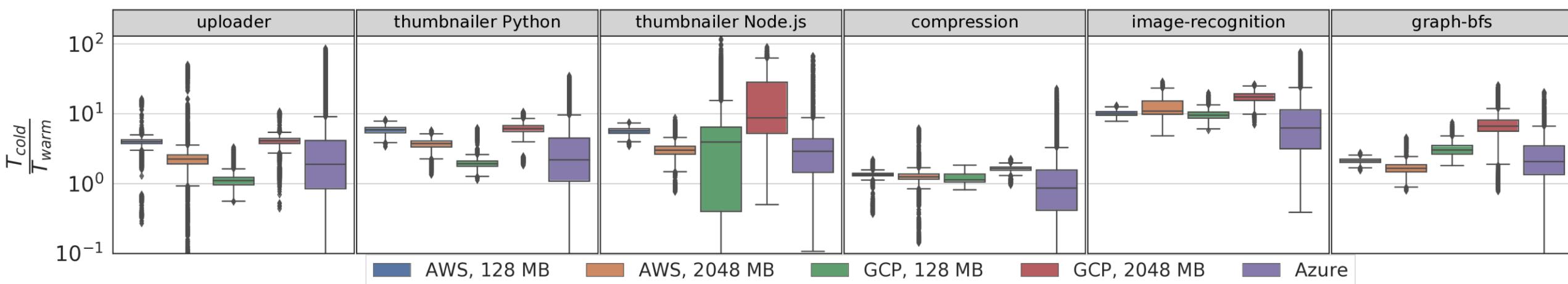
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

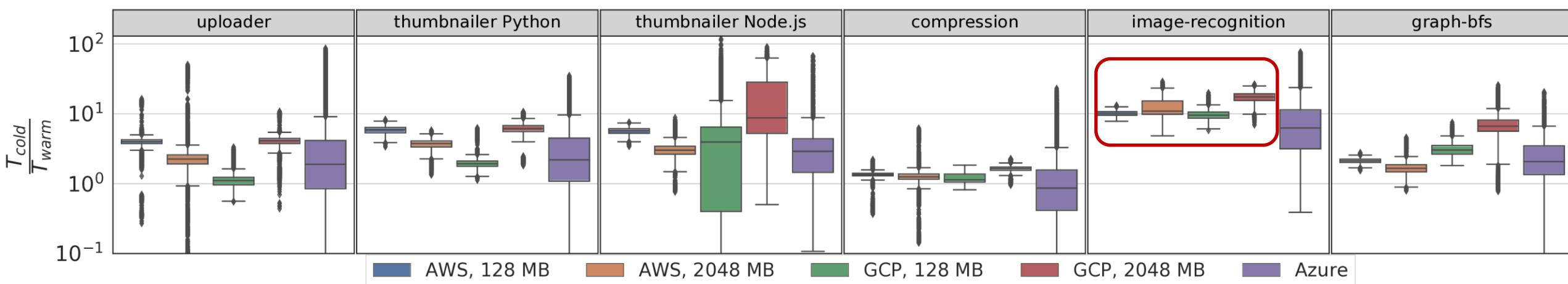
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

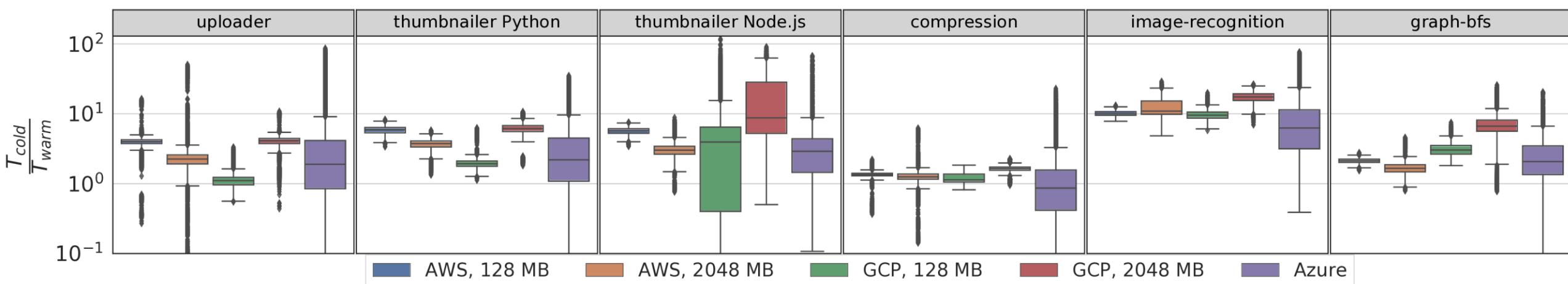
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

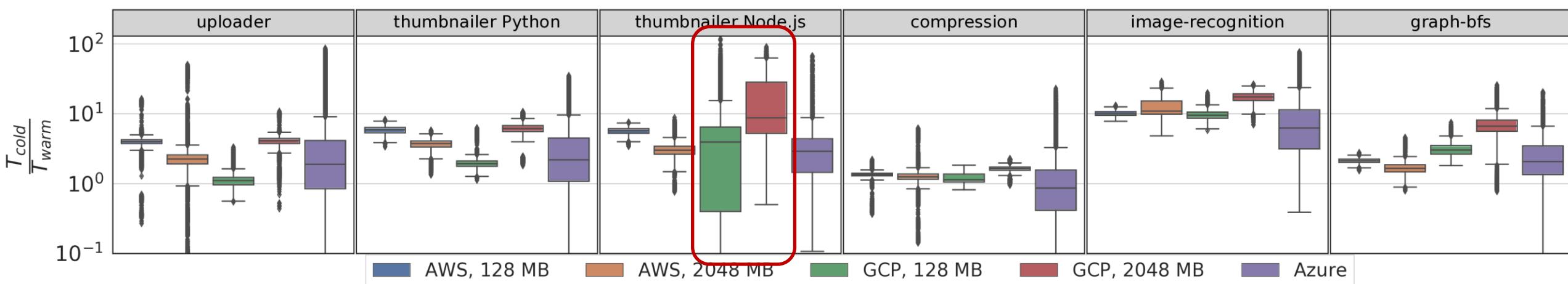
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

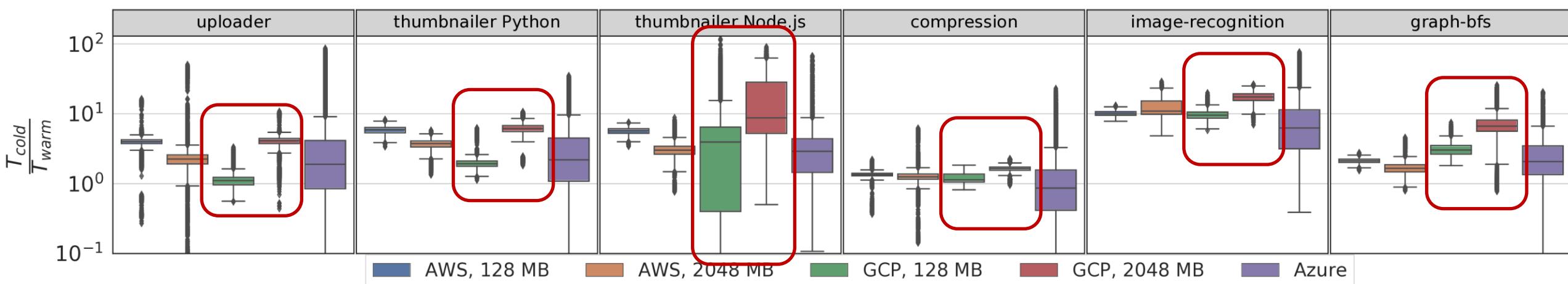
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

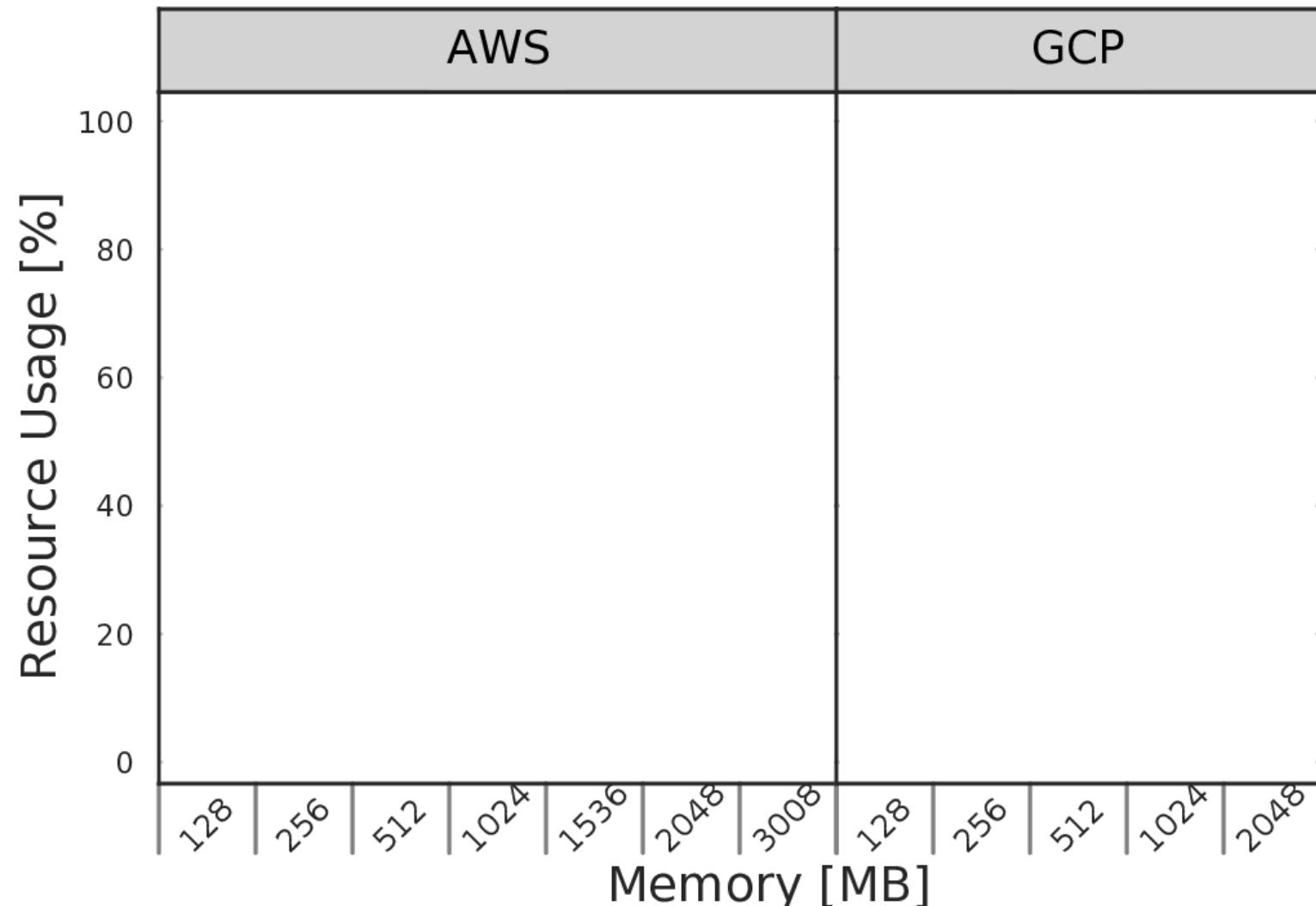
Results and Insights

	Results, methods, and insights
	<p>High-memory allocations increase cold startup overheads on GCP.</p> <p>GCP functions experience reliability and availability issues.</p>
	
	
	

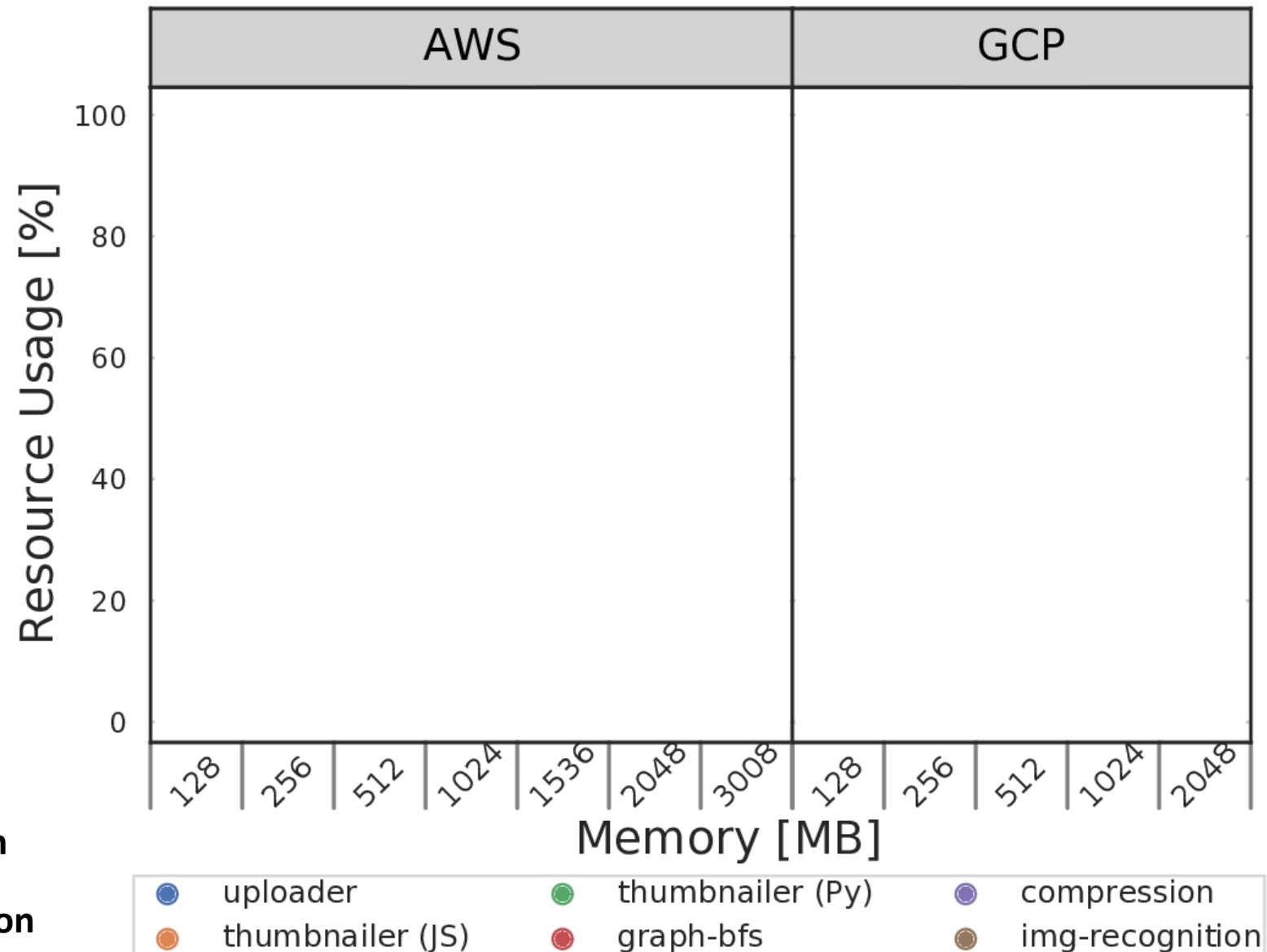
Results and Insights

Results, methods, and insights	
	High-memory allocations increase cold startup overheads on GCP. GCP functions experience reliability and availability issues.
	Resource underutilization due to high granularity of pricing models.
	
	

Cost Analysis: Resource Usage

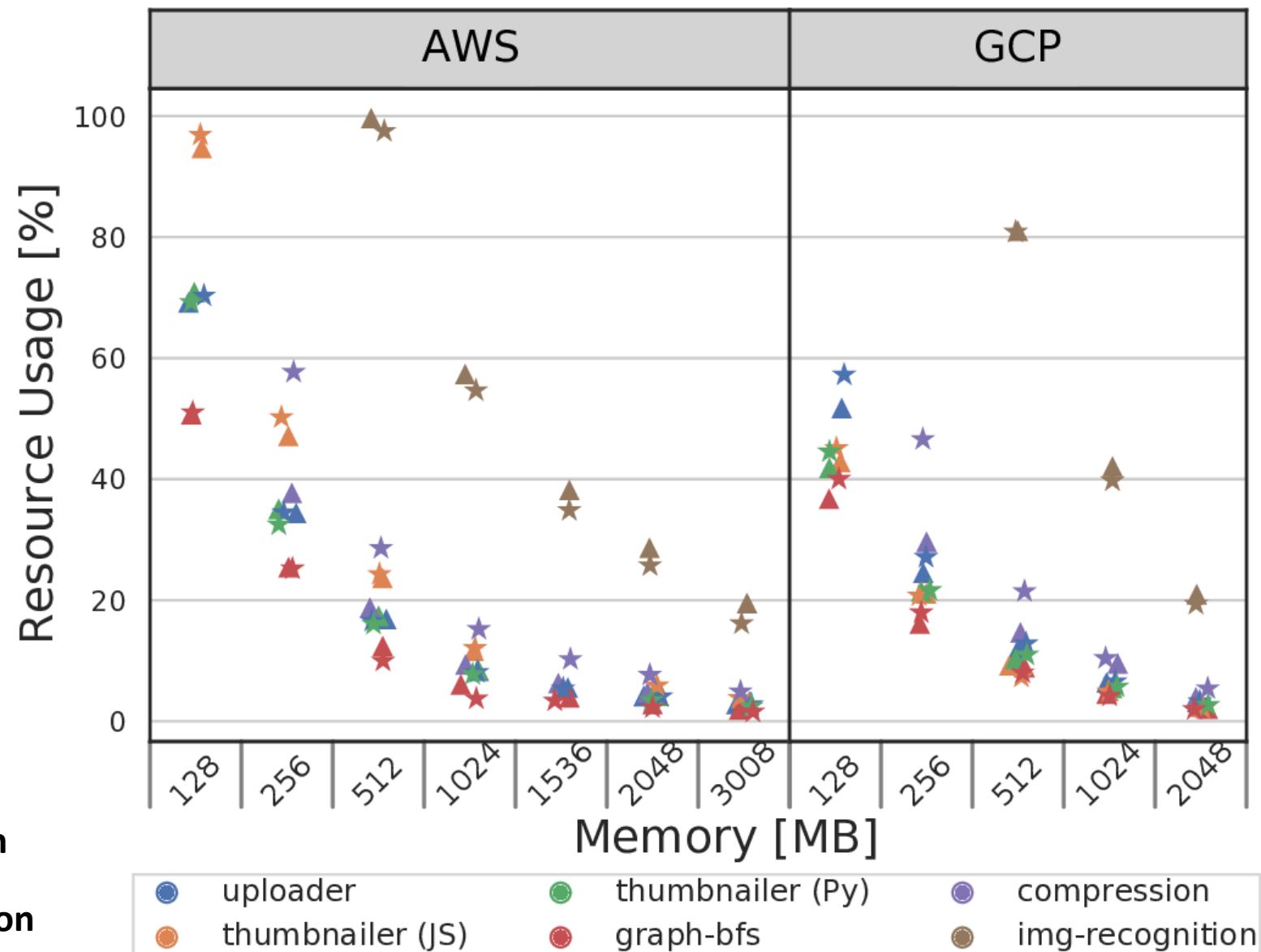


Cost Analysis: Resource Usage



▲ Cold Execution
★ Warm Execution

Cost Analysis: Resource Usage



▲ Cold Execution
★ Warm Execution

Results and Insights

Results, methods, and insights	
	High-memory allocations increase cold startup overheads on GCP. GCP functions experience reliability and availability issues.
	Resource underutilization due to high granularity of pricing models.
	
	

Results and Insights

Results, methods, and insights	
	<p>High-memory allocations increase cold startup overheads on GCP.</p> <p>GCP functions experience reliability and availability issues.</p>
	<p>Resource underutilization due to high granularity of pricing models.</p>
	
	<p>AWS Lambda container eviction is agnostic to function properties.</p> <p>Analytical models of AWS Lambda container eviction policy.</p>

FaaS Analysis: Eviction Modeling

FaaS Analysis: Eviction Modeling

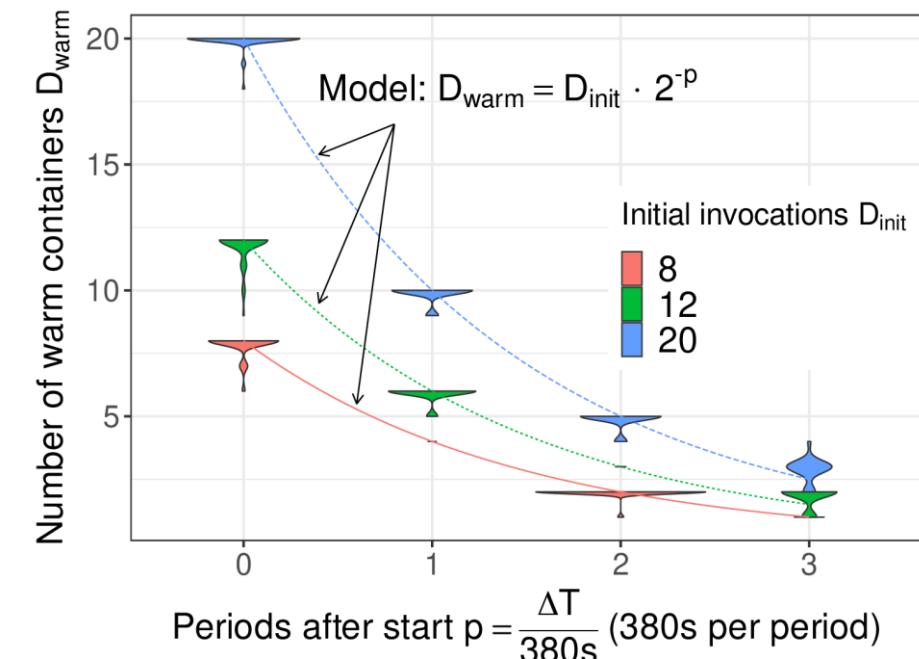
Configuration:

- **Time between invocations:** 1 – 1600s.
- **# of function instances:** 1 -20
- **Memory:** 128 – 1536 MB
- **Package size:** 8 kB, 250 MB
- **Duration:** 1 – 10s
- **Language:** Python, Node.js

FaaS Analysis: Eviction Modeling

Configuration:

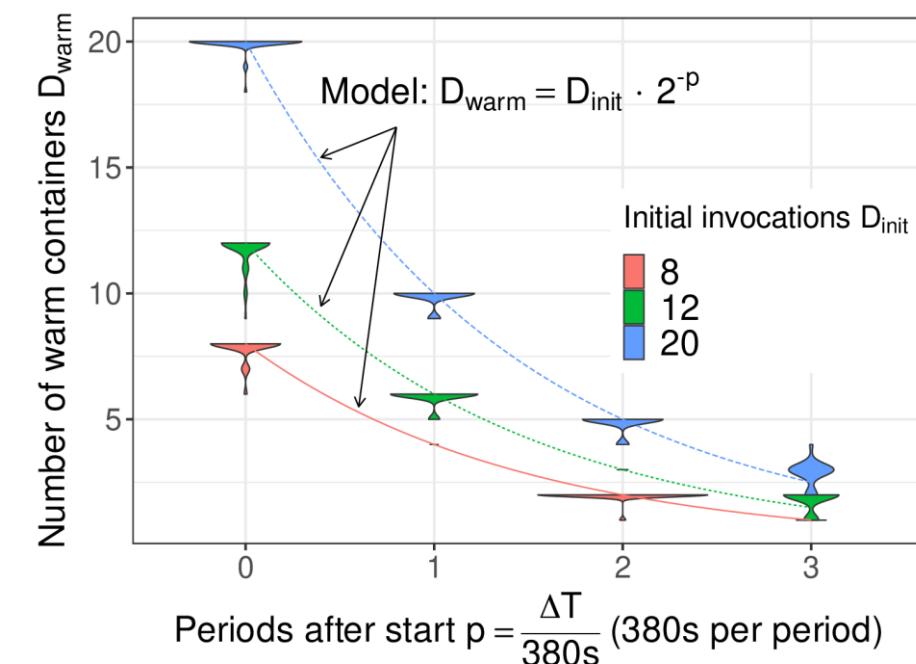
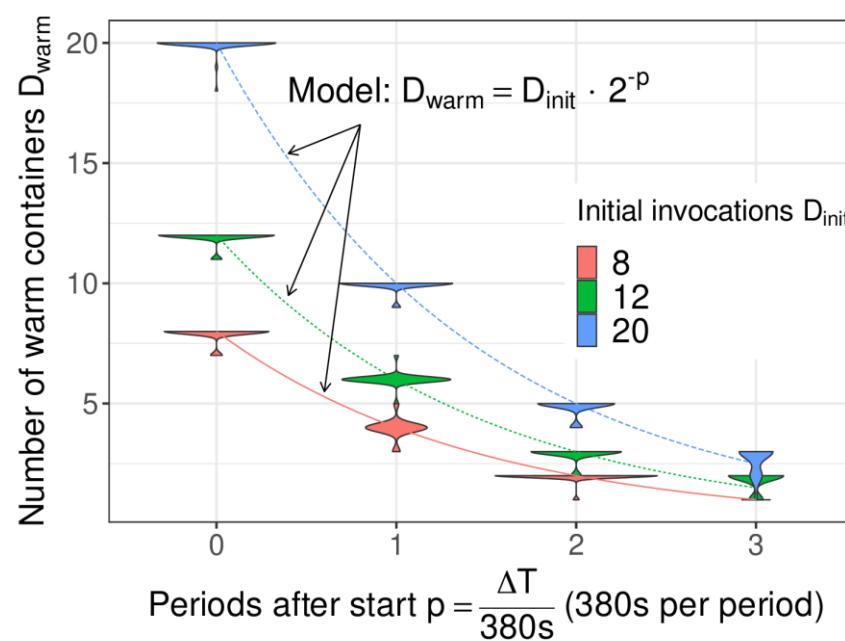
- Time between invocations: 1 – 1600s.
- # of function instances: 1 -20
- Memory: 128 – 1536 MB
- Package size: 8 kB, 250 MB
- Duration: 1 – 10s
- Language: Python, Node.js



FaaS Analysis: Eviction Modeling

Configuration:

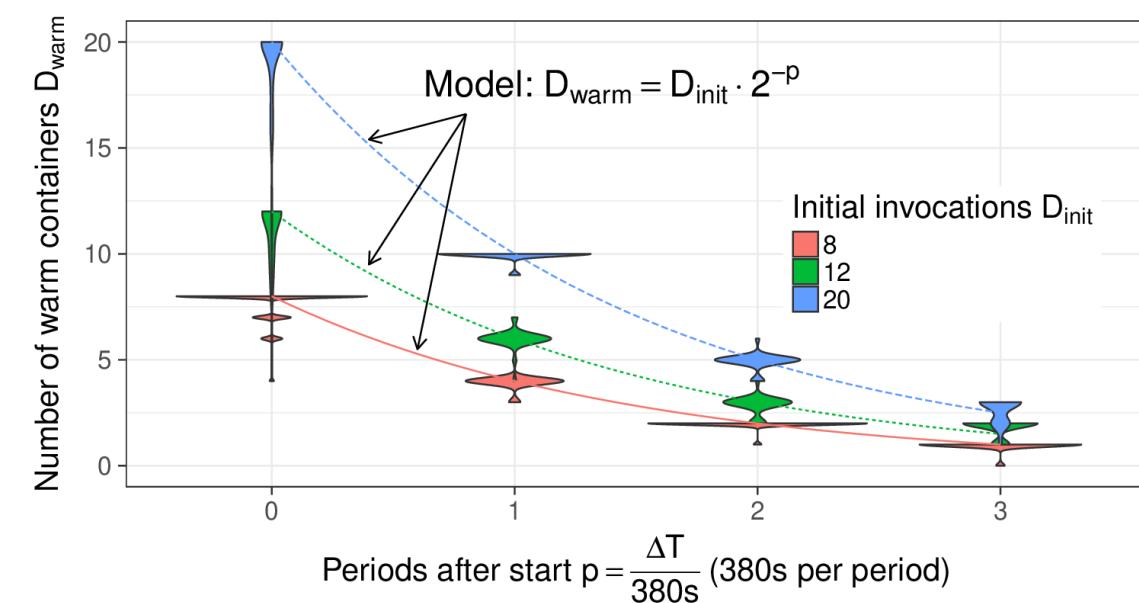
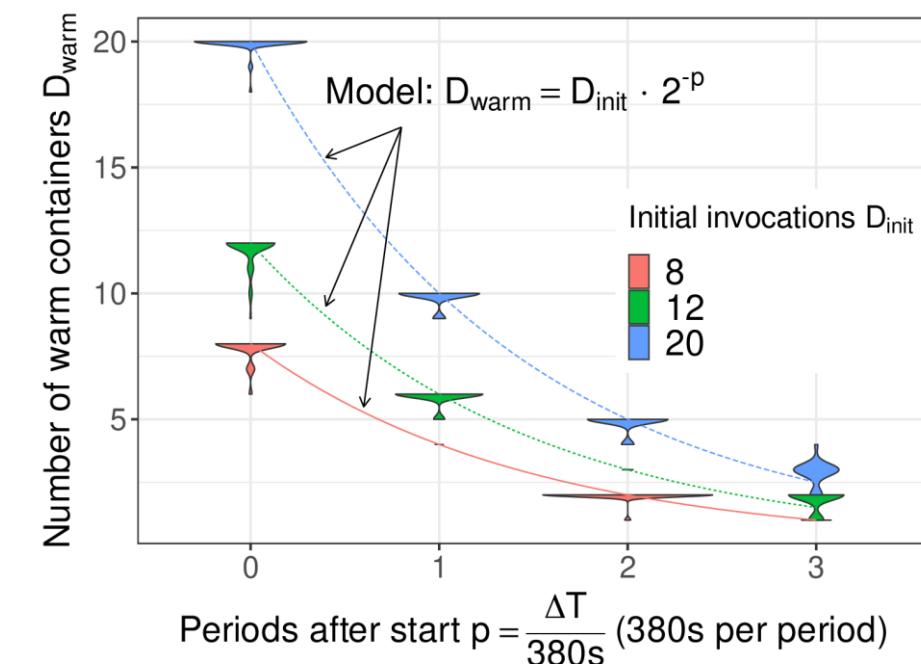
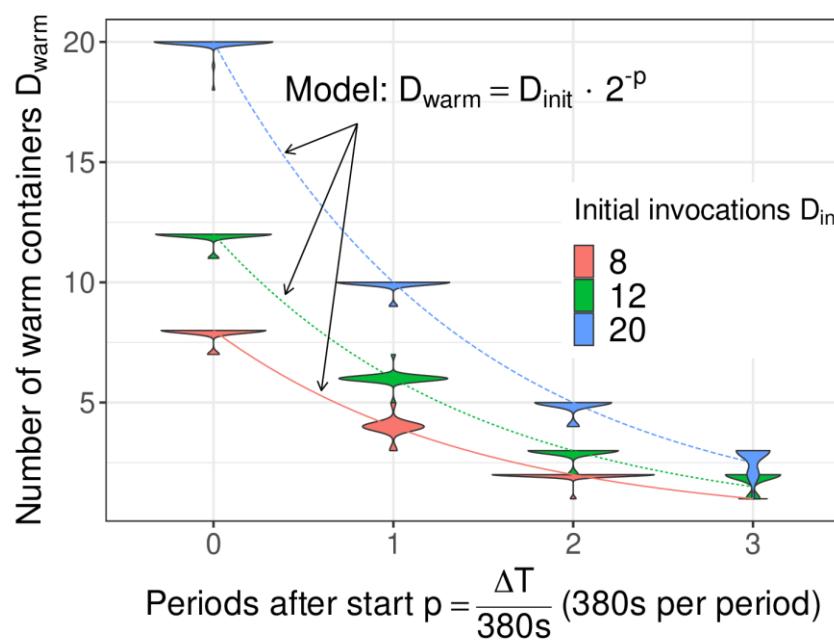
- Time between invocations: 1 – 1600s.
- # of function instances: 1 -20
- Memory: 128 – 1536 MB
- Package size: 8 kB, 250 MB
- Duration: 1 – 10s
- Language: Python, Node.js



FaaS Analysis: Eviction Modeling

Configuration:

- Time between invocations: 1 – 1600s.
- # of function instances: 1 -20
- Memory: 128 – 1536 MB
- Package size: 8 kB, 250 MB
- Duration: 1 – 10s
- Language: Python, Node.js



Results and Insights

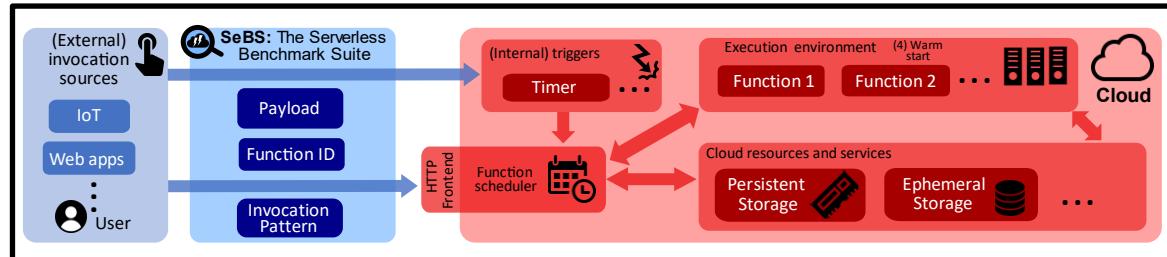
Results, methods, and insights	
	<p>High-memory allocations increase cold startup overheads on GCP.</p> <p>GCP functions experience reliability and availability issues.</p>
	<p>Resource underutilization due to high granularity of pricing models.</p>
	
	<p>AWS Lambda container eviction is agnostic to function properties.</p> <p>Analytical models of AWS Lambda container eviction policy.</p>

Results and Insights

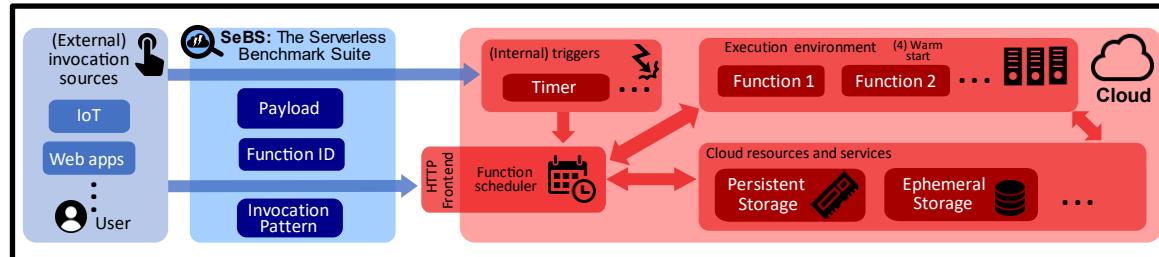
Results, methods, and insights	
	High-memory allocations increase cold startup overheads on GCP.
	GCP functions experience reliability and availability issues.
	I/O-bound functions experience very high latency variations.
	AWS Lambda achieves the best performance on all workloads.
	Irregular performance of concurrent Azure Function executions.
	AWS Lambda performance is not competitive against VMs assuming comparable resources.
	Resource underutilization due to high granularity of pricing models.
	Break-even analysis for IaaS and FaaS deployment.
	High costs of Azure Functions due to unconfigurable deployment.
	The function output size can be a dominating factor in pricing.
	Accurate methodology for estimation of invocation latency.
	Warm latencies are consistent and depend linearly on payload size.
	Highly variable and unpredictable cold latencies on Azure and GCP.
	AWS Lambda container eviction is agnostic to function properties.
	Analytical models of AWS Lambda container eviction policy.

Summary

Summary

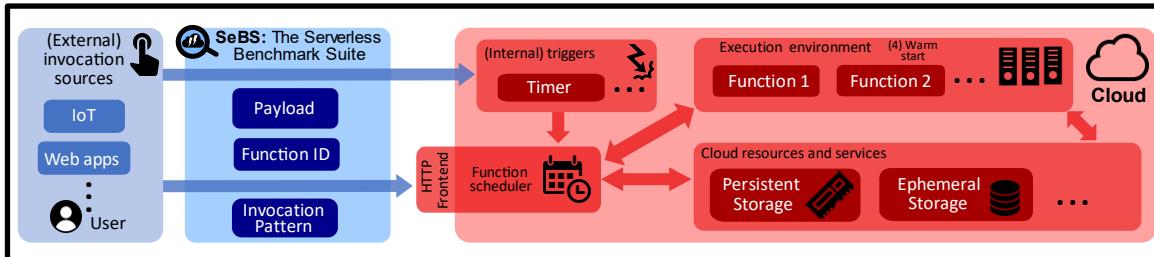


Summary



Type	Name	Language
Webapps	uploader	Python, Node.js
Multimedia	thumbnailer	Python, Node.js, C++
Utilities	compression	Python
Inference	image-recognition	Python, C++
Scientific	graph-bfs	Python

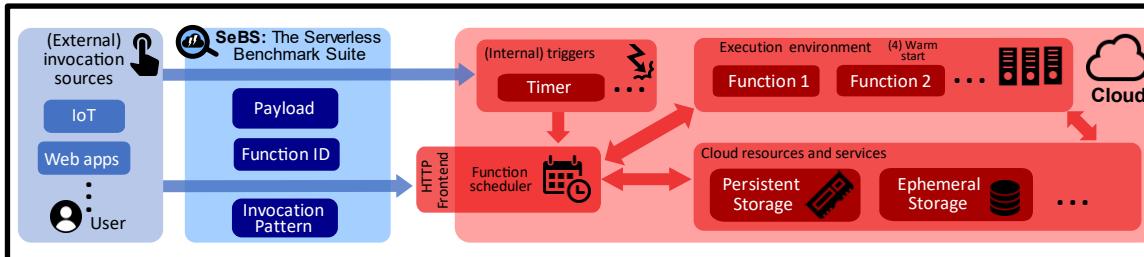
Summary



Type	Name	Language
Webapps	uploader	Python, Node.js
Multimedia	thumbnailer	Python, Node.js, C++
Utilities	compression	Python
Inference	image-recognition	Python, C++
Scientific	graph-bfs	Python

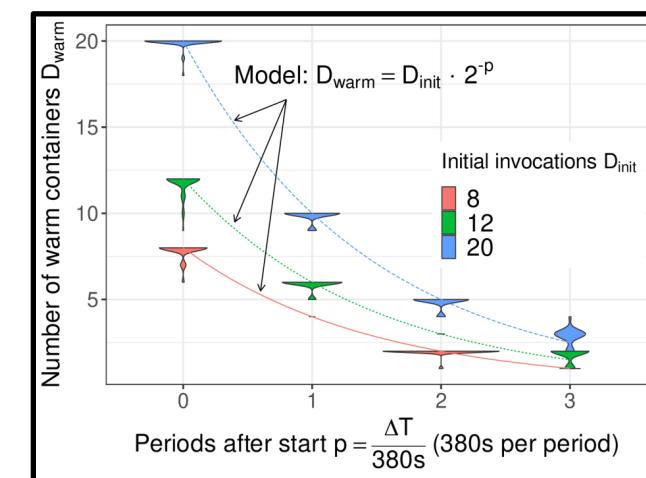
Results, methods, and insights	
	High-memory allocations increase cold startup overheads on GCP. GCP functions experience reliability and availability issues. I/O-bound functions experience very high latency variations. AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions. AWS Lambda performance is not competitive against VMs assuming comparable resources.
	Break-even analysis for IaaS and FaaS deployment. Resource underutilization due to high granularity of pricing models. High costs of Azure Functions due to unconfigurable deployment. The function output size can be a dominating factor in pricing.
	Accurate methodology for estimation of invocation latency. Warm latencies are consistent and depend linearly on payload size. Highly variable and unpredictable cold latencies on Azure and GCP.
	AWS Lambda container eviction is agnostic to function properties. Analytical models of AWS Lambda container eviction policy.

Summary

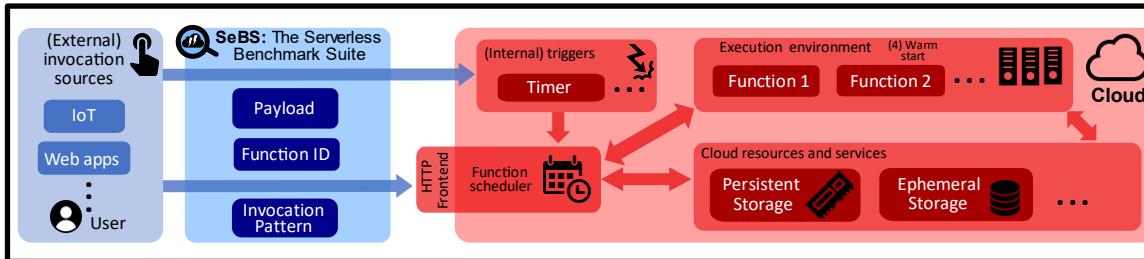


Type	Name	Language
Webapps	uploader	Python, Node.js
Multimedia	thumbnailer	Python, Node.js, C++
Utilities	compression	Python
Inference	image-recognition	Python, C++
Scientific	graph-bfs	Python

Results, methods, and insights	
	High-memory allocations increase cold startup overheads on GCP. GCP functions experience reliability and availability issues. I/O-bound functions experience very high latency variations. AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions. AWS Lambda performance is not competitive against VMs assuming comparable resources.
	Break-even analysis for IaaS and FaaS deployment. Resource underutilization due to high granularity of pricing models. High costs of Azure Functions due to unconfigurable deployment. The function output size can be a dominating factor in pricing.
	Accurate methodology for estimation of invocation latency. Warm latencies are consistent and depend linearly on payload size. Highly variable and unpredictable cold latencies on Azure and GCP.
	AWS Lambda container eviction is agnostic to function properties. Analytical models of AWS Lambda container eviction policy.

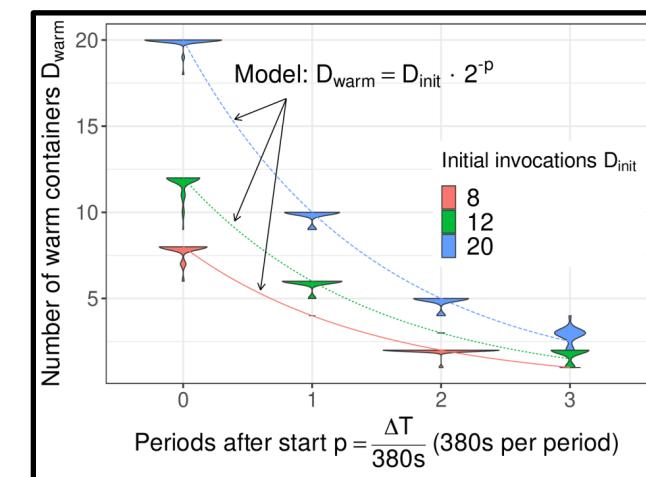


Summary

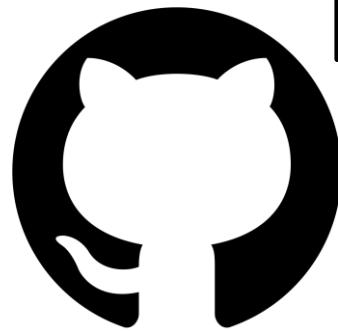


Type	Name	Language
Webapps	uploader	Python, Node.js
Multimedia	thumbnailer	Python, Node.js, C++
Utilities	compression	Python
Inference	image-recognition	Python, C++
Scientific	graph-bfs	Python

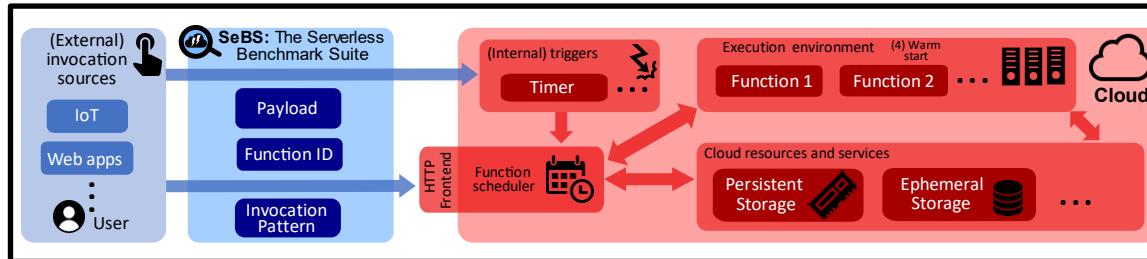
Results, methods, and insights	
	High-memory allocations increase cold startup overheads on GCP. GCP functions experience reliability and availability issues. I/O-bound functions experience very high latency variations. AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions.
	Break-even analysis for IaaS and FaaS deployment. Resource underutilization due to high granularity of pricing models. High costs of Azure Functions due to unconfigurable deployment. The function output size can be a dominating factor in pricing.
	Accurate methodology for estimation of invocation latency. Warm latencies are consistent and depend linearly on payload size. Highly variable and unpredictable cold latencies on Azure and GCP.
	AWS Lambda container eviction is agnostic to function properties. Analytical models of AWS Lambda container eviction policy.



[spcl/serverless-benchmarks](https://github.com/spcl/serverless-benchmarks)

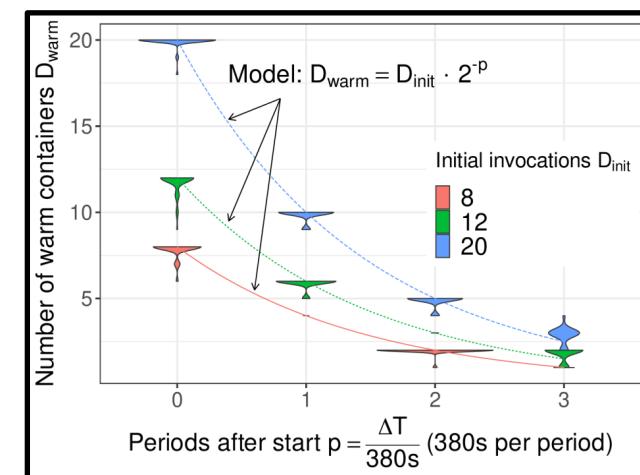


Summary



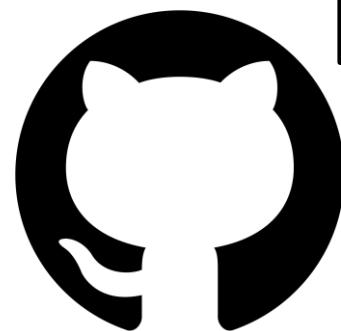
Type	Name	Language
Webapps	uploader	Python, Node.js
Multimedia	thumbnailer	Python, Node.js, C++
Utilities	compression	Python
Inference	image-recognition	Python, C++
Scientific	graph-bfs	Python

Results, methods, and insights	
	High-memory allocations increase cold startup overheads on GCP. GCP functions experience reliability and availability issues. I/O-bound functions experience very high latency variations. AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions.
	Break-even analysis for IaaS and FaaS deployment. Resource underutilization due to high granularity of pricing models. High costs of Azure Functions due to unconfigurable deployment. The function output size can be a dominating factor in pricing.
	Accurate methodology for estimation of invocation latency. Warm latencies are consistent and depend linearly on payload size. Highly variable and unpredictable cold latencies on Azure and GCP.
	AWS Lambda container eviction is agnostic to function properties. Analytical models of AWS Lambda container eviction policy.



[spcl/serverless-benchmarks](https://github.com/spcl/serverless-benchmarks)

OpenWhisk
 C++ Functions
 Serverless Workflows



Q&A

Future Work

Questions

Q&A

Future Work

Questions



spcl/serverless-benchmarks

Q&A

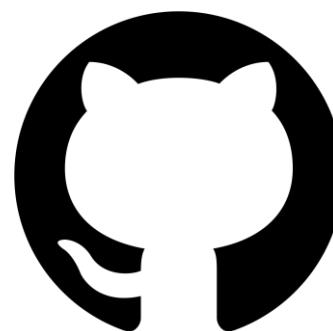
Future Work

✓ OpenWhisk

⚙️ C++ Functions

Serverless Workflows

Questions



spcl/serverless-benchmarks

Q&A

Future Work

✓ OpenWhisk

⚙️ C++ Functions

Serverless Workflows

Questions

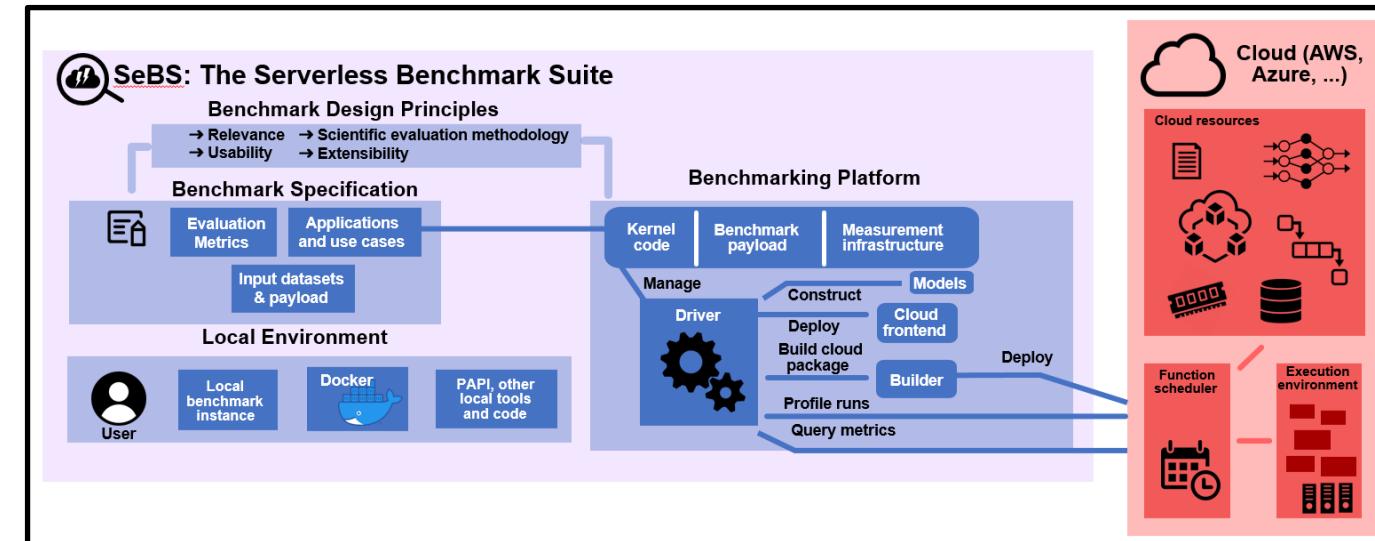
- What are the requirements for a good benchmark suite?
- How can we measure function invocation latency accurately?
- How much performance do we lose when switching from IaaS to FaaS?



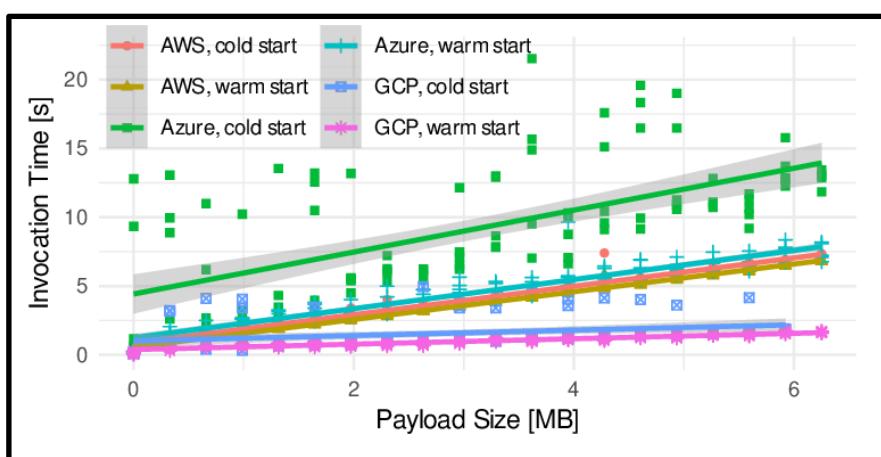
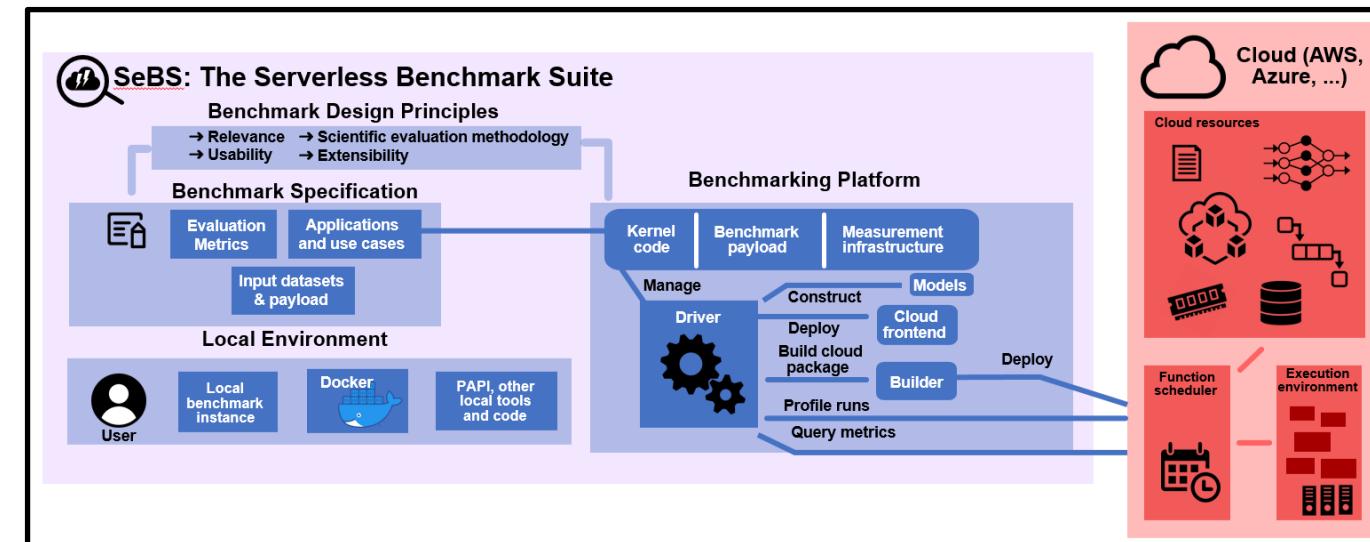
spcl/serverless-benchmarks

SeBS in details...

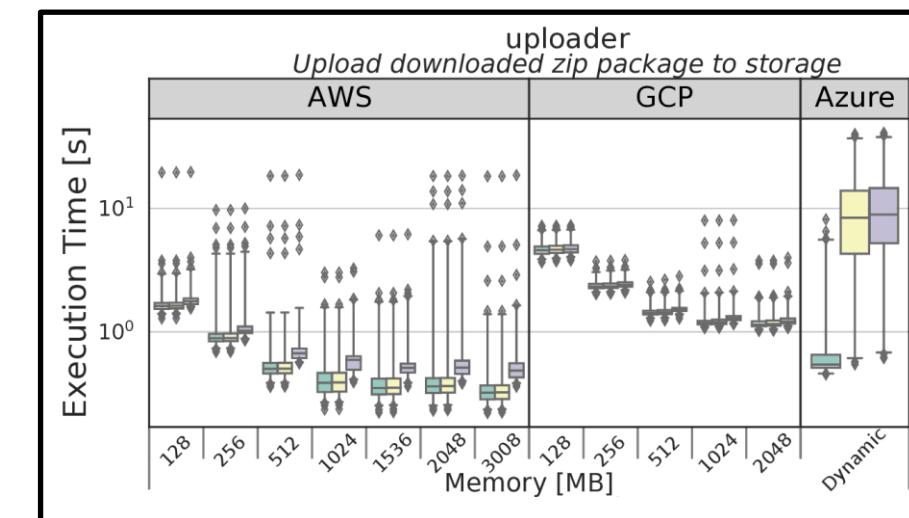
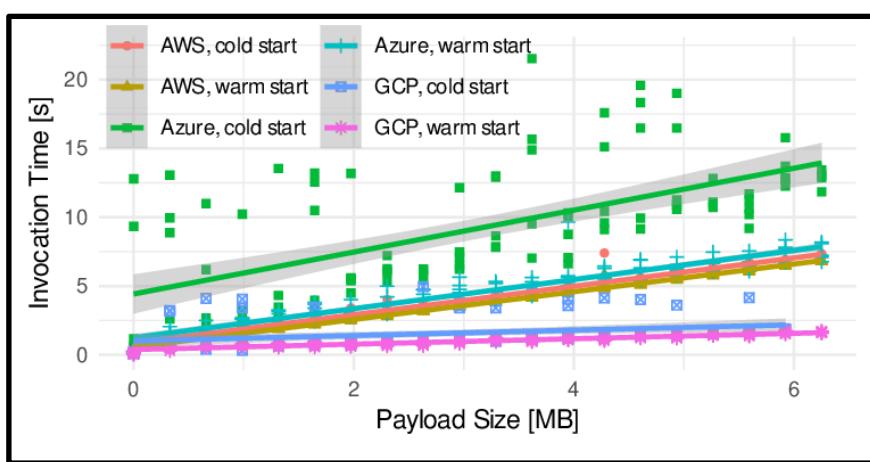
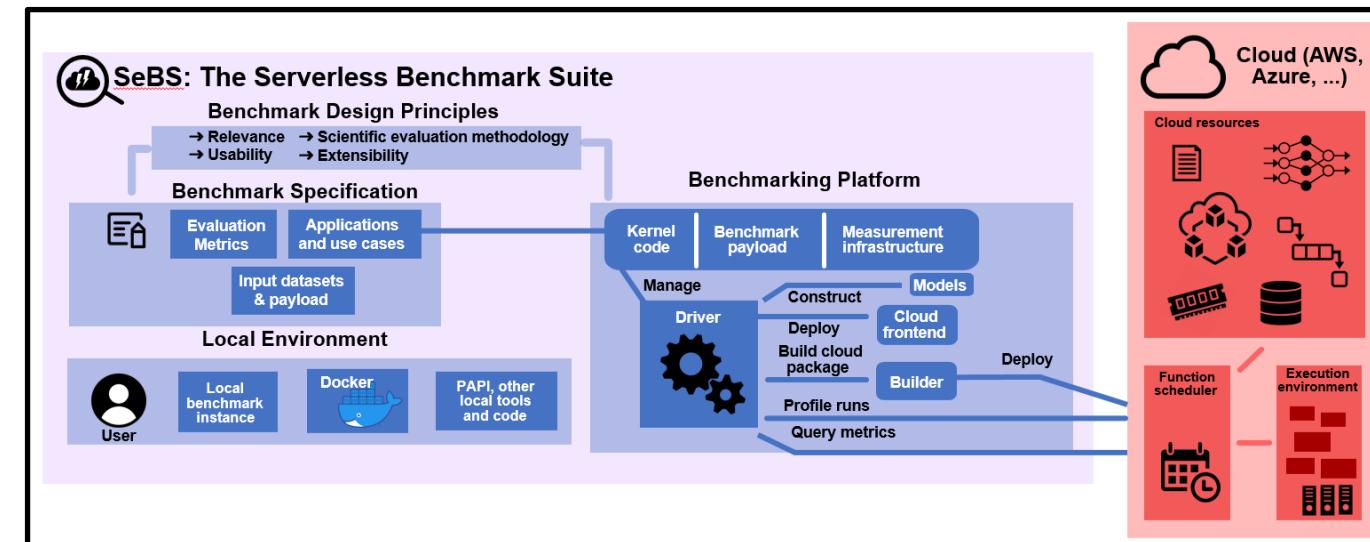
SeBS in details...



SeBS in details...



SeBS in details...



How to build a benchmark?

How to build a benchmark?

Benchmarking in the Cloud: What It Should, Can, and Cannot Be

Enno Folkerts¹, Alexander Alexandrov², Kai Sachs¹,
Alexandru Iosup³, Volker Markl², and Cafer Tosun¹

¹ SAP AG, 69190 Walldorf, Germany
firstname.lastname@sap.com

² TU Berlin, Germany
firstname.lastname@tu-berlin.de

³ Delft University of Technology, The Netherlands
A.Iosup@tudelft.nl

Abstract. With the increasing adoption of Cloud Computing, we observe an increasing need for Cloud Benchmarks, in order to assess the performance of Cloud infrastructures and software stacks, to assist with provisioning decisions for Cloud users, and to compare Cloud offerings. We understand our paper as one of the first systematic approaches to the topic of Cloud Benchmarks. Our driving principle is that Cloud Benchmarks must consider end-to-end performance and pricing, taking into account that services are delivered over the Internet. This requirement yields new challenges for benchmarking and requires us to revisit existing benchmarking practices in order to adopt them to the Cloud.

How to build a benchmark?

Benchmarking in the Cloud: What It Should, Can, and Cannot Be

Enno Folkerts¹, Alexander Alexandrov², Kai Sachs¹,
Alexandru Iosup³, Volker Markl², and Cafer Tosun¹

¹ SAP AG, 69190 Walldorf, Germany
`firstname.lastname@sap.com`

² TU Berlin, Germany

`firstname.lastname@tu-berlin.de`

³ Delft University of Technology, The Netherlands
`A.Iosup@tudelft.nl`

Abstract. With the increasing adoption of Cloud Computing, we observe an increasing need for Cloud Benchmarks, in order to assess the performance of Cloud infrastructures and software stacks, to assist with provisioning decisions for Cloud users, and to compare Cloud offerings. We understand our paper as one of the first systematic approaches to the topic of Cloud Benchmarks. Our driving principle is that Cloud Benchmarks must consider end-to-end performance and pricing, taking into account that services are delivered over the Internet. This requirement yields new challenges for benchmarking and requires us to revisit existing benchmarking practices in order to adopt them to the Cloud.

How is the Weather tomorrow? Towards a Benchmark for the Cloud

Carsten Binnig Donald Kossmann Tim Kraska Simon Loesing

Systems Group, Department of Computer Science, ETH Zurich
`{firstname.lastname}@inf.ethz.ch`

How to build a benchmark?

Benchmarking in the Cloud: What It Should, Can, and Cannot Be

Enno Folkerts¹, Alexander Alexandrov², Kai Sachs¹,
Alexandru Iosup³, Volker Markl², and Cafer Tosun¹

¹ SAP AG, 69190 Walldorf, Germany
firstname.lastname@sap.com

² TU Berlin, Germany

firstname.lastname@tu-berlin.de

³ Delft University of Technology, The Netherlands
A.Iosup@tudelft.nl

Abstract. With the increasing adoption of Cloud Computing, we observe an increasing need for Cloud Benchmarks, in order to assess the performance of Cloud infrastructures and software stacks, to assist with provisioning decisions for Cloud users, and to compare. We understand our paper as one of the first systematic topic of Cloud Benchmarks. Our driving principle is that benchmarks must consider end-to-end performance and take account that services are delivered over the Internet. This yields new challenges for benchmarking and requires us to adapt our benchmarking practices in order to adopt them to the

How is the Weather tomorrow? Towards a Benchmark for the Cloud

Carsten Binnig Donald Kossmann Tim Kraska Simon Loesing
Systems Group, Department of Computer Science, ETH Zurich
{firstname.lastname}@inf.ethz.ch

Scientific Benchmarking of Parallel Computing Systems

Twelve ways to tell the masses when reporting performance results

Torsten Hoefler
Dept. of Computer Science
ETH Zurich
Zurich, Switzerland
htor@inf.ethz.ch

Roberto Belli
Dept. of Computer Science
ETH Zurich
Zurich, Switzerland
bellir@inf.ethz.ch

How to build a benchmark?

Benchmarking in the Cloud:
What It Should, Can, and Cannot Be

Enno Folkert
Alexandru Io-

¹ SA
f

firs
³ Delft Uni

Jóakim v. Kistowski
University of Würzburg
joakim.kistowski@uni-wuerzburg.de

Abstract. With the serve an increasing n Hewlett-Packard Company
performance of Cloud klaus.lange@hp.com
provisioning decisions for Cloud users, and to compa
We understand our paper as one of the first systematic
topic of Cloud Benchmarks. Our driving principle is t
marks must consider end-to-end performance and pr
account that services are delivered over the Internet.
yields new challenges for benchmarking and requires us
benchmarking practices in order to adopt them to the

How to Build a Benchmark

Jeremy A. Arnold
IBM Corporation
arnoldje@us.ibm.com

John L. Henning
Oracle
john.henning@oracle.com

Karl Huppler
karl.huppler@gmail.com

Paul Cao
Hewlett-Packard Company
paul.cao@hp.com

¹ Tim Kraska Simon Loesing
Computer Science, ETH Zurich
e}@inf.ethz.ch

Scientific Benchmarking of Parallel Computing Systems

Twelve ways to tell the masses when reporting performance results

Torsten Hoefler
Dept. of Computer Science
ETH Zurich
Zurich, Switzerland
htor@inf.ethz.ch

Roberto Belli
Dept. of Computer Science
ETH Zurich
Zurich, Switzerland
bellir@inf.ethz.ch

Benchmarking Goals

Benchmarking Goals

Principles

- ✓ Usability
- ✓ Portability
- ✓ Extensibility
- ✓ Scientific

Benchmarking Goals

Principles

- ✓ Usability
- ✓ Portability
- ✓ Extensibility
- ✓ Scientific

Applications

- ✓ Realistic workloads
- ✓ Single implementation
- ✓ Varying computational characteristics

Benchmarking Goals

Principles

- ✓ Usability
- ✓ Portability
- ✓ Extensibility
- ✓ Scientific

Applications

- ✓ Realistic workloads
- ✓ Single implementation
- ✓ Varying computational characteristics

Metrics

- ✓ Cloud time
- ✓ User time
- ✓ Resource utilization
- ✓ I/O

Benchmarking Goals

Principles

- ✓ Usability
- ✓ Portability
- ✓ Extensibility
- ✓ Scientific

Applications

- ✓ Realistic workloads
- ✓ Single implementation
- ✓ Varying computational characteristics

Metrics

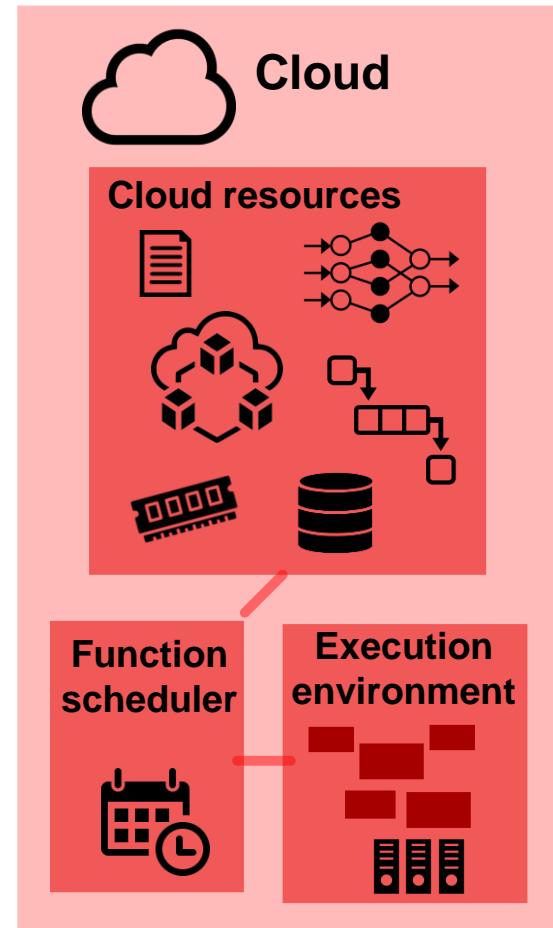
- ✓ Cloud time
- ✓ User time
- ✓ Resource utilization
- ✓ I/O

Experiments

- ✓ Performance and cost
- ✓ FaaS vs IaaS
- ✓ Invocation overhead
- ✓ Container eviction

How to build a benchmark?

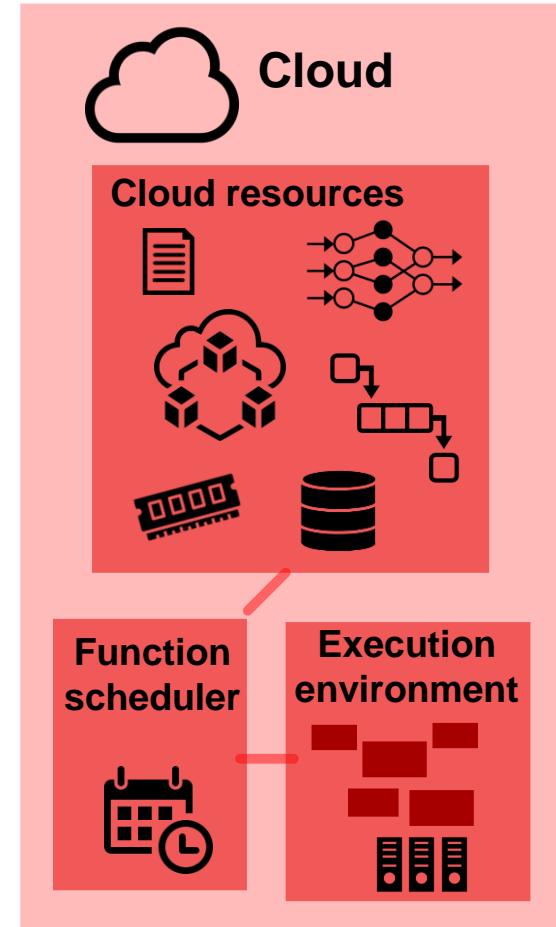
How to build a benchmark?



How to build a benchmark?



SeBS: The Serverless Benchmark Suite



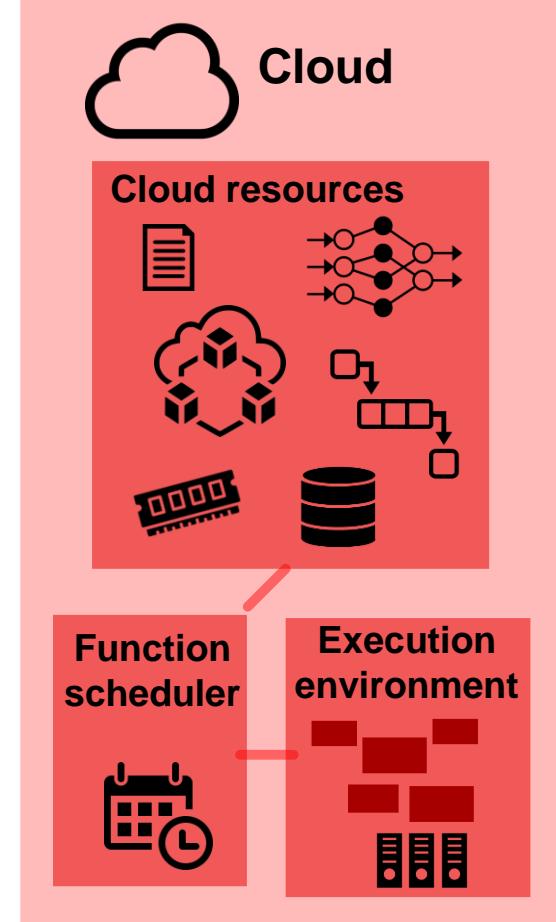
How to build a benchmark?



SeBS: The Serverless Benchmark Suite

Benchmark Design Principles

- Relevance → Scientific evaluation methodology
- Usability → Extensibility



How to build a benchmark?



SeBS: The Serverless Benchmark Suite

Benchmark Design Principles

- Relevance → Scientific evaluation methodology
- Usability → Extensibility

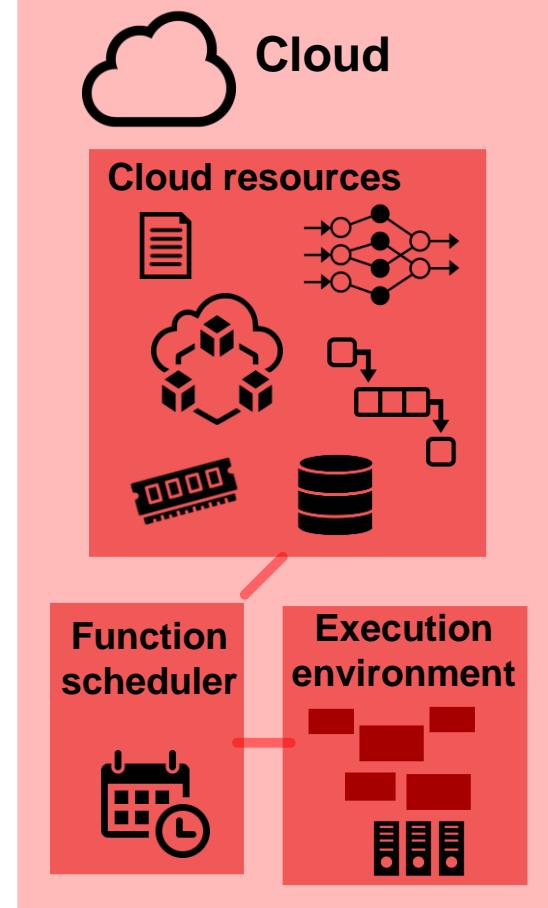
Benchmark Specification



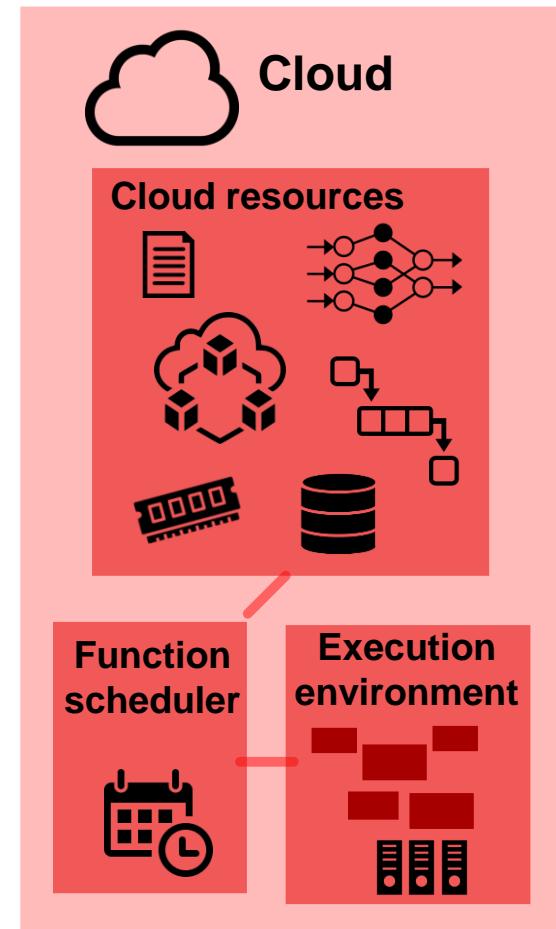
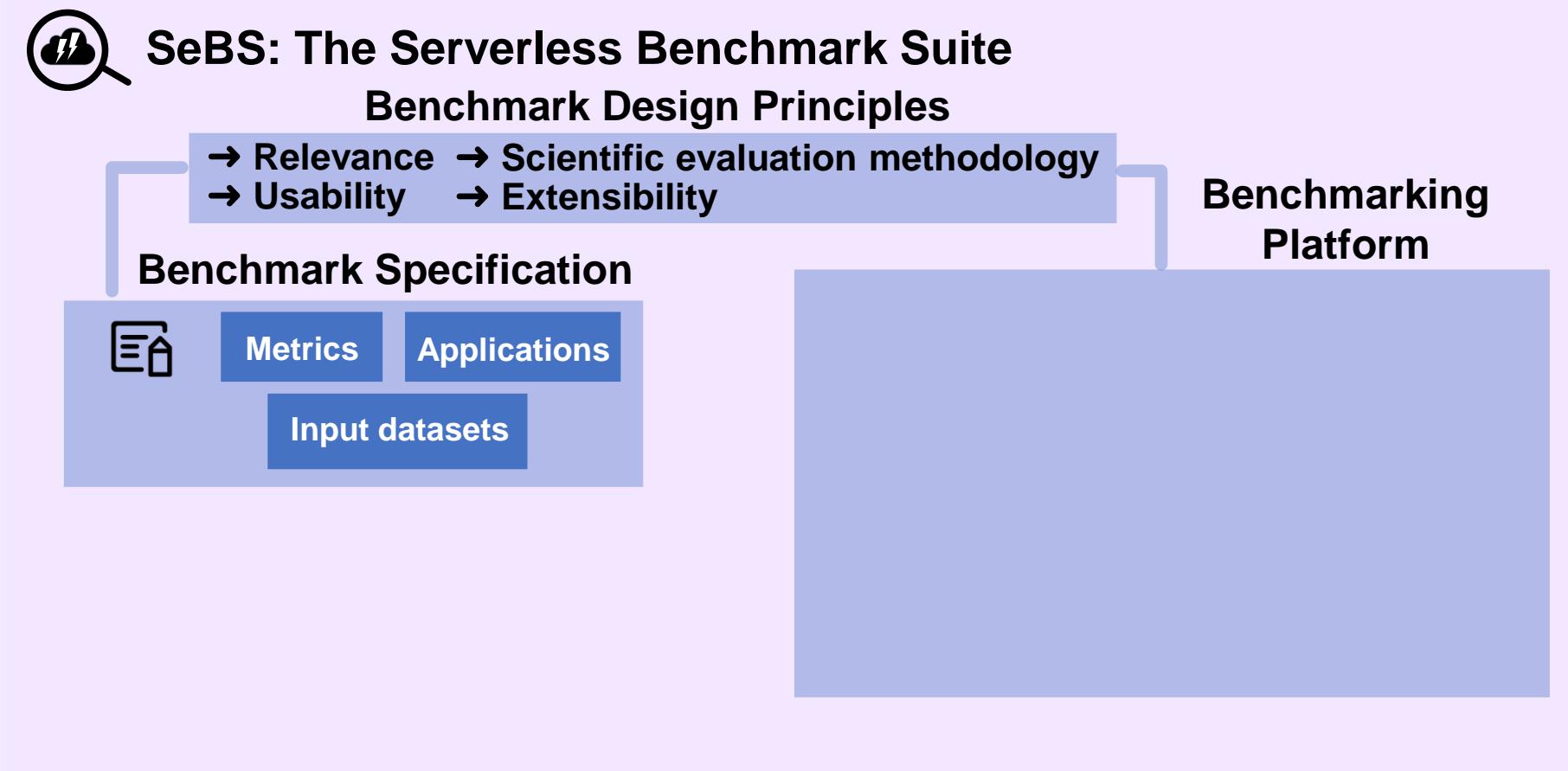
Metrics

Applications

Input datasets



How to build a benchmark?



How to build a benchmark?



SeBS: The Serverless Benchmark Suite

Benchmark Design Principles

- Relevance → Scientific evaluation methodology
- Usability → Extensibility

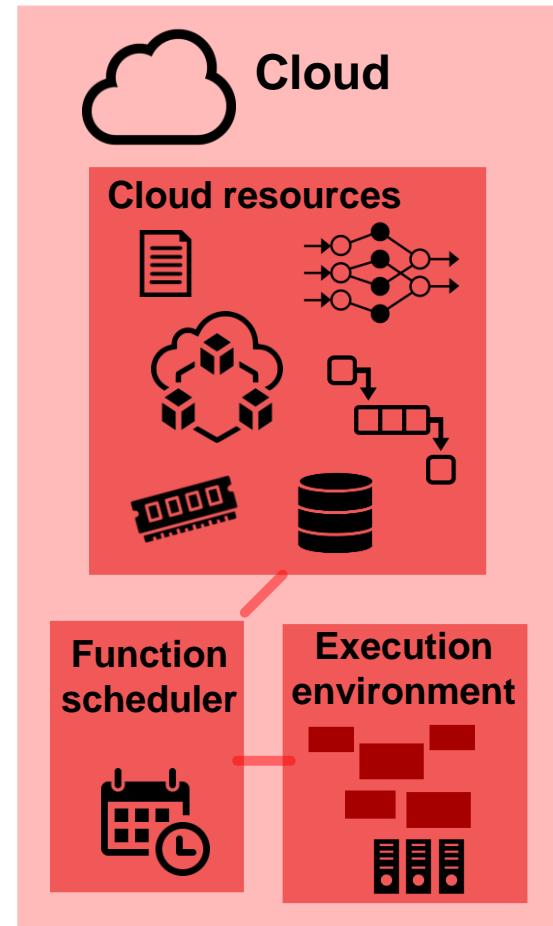
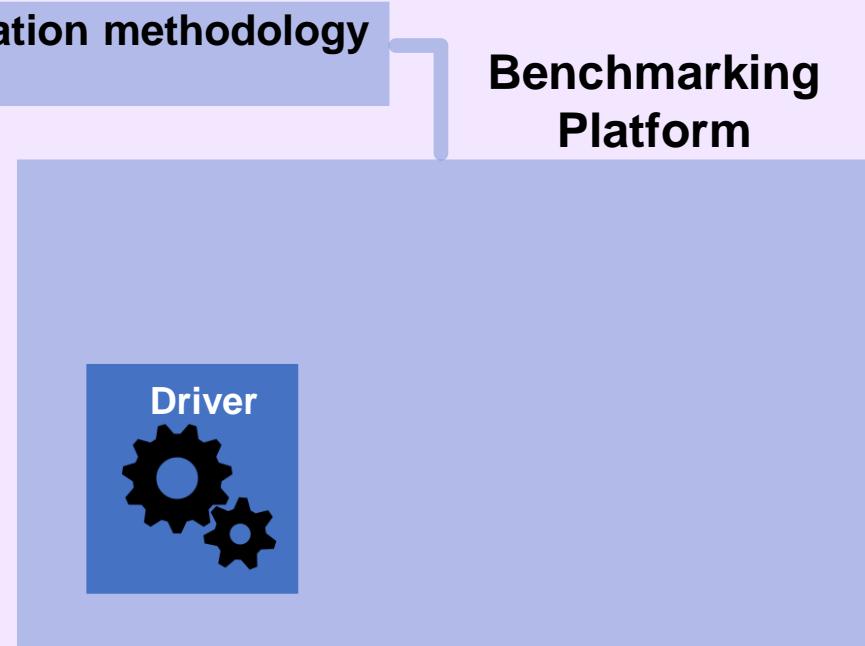
Benchmark Specification



Metrics

Applications

Input datasets



How to build a benchmark?



SeBS: The Serverless Benchmark Suite

Benchmark Design Principles

- Relevance → Scientific evaluation methodology
- Usability → Extensibility

Benchmark Specification



Metrics

Applications

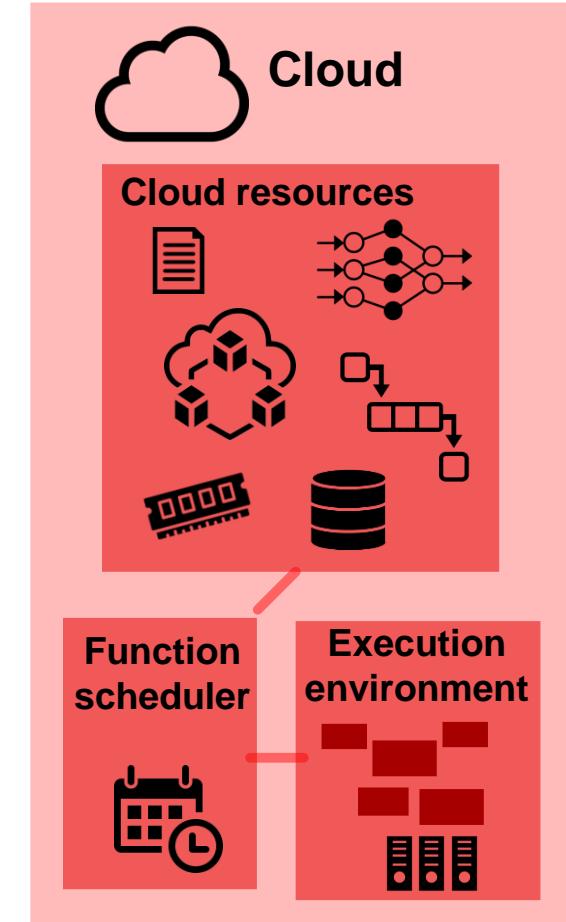
Input datasets

Benchmarking Platform

Kernel code | Benchmark payload | Measurement infrastructure

Manage

Driver



How to build a benchmark?



SeBS: The Serverless Benchmark Suite

Benchmark Design Principles

- Relevance → Scientific evaluation methodology
- Usability → Extensibility

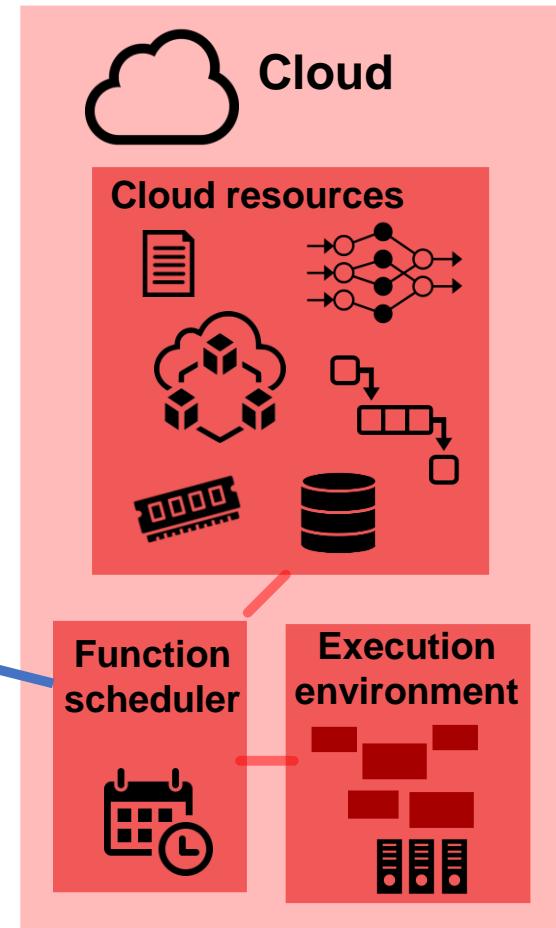
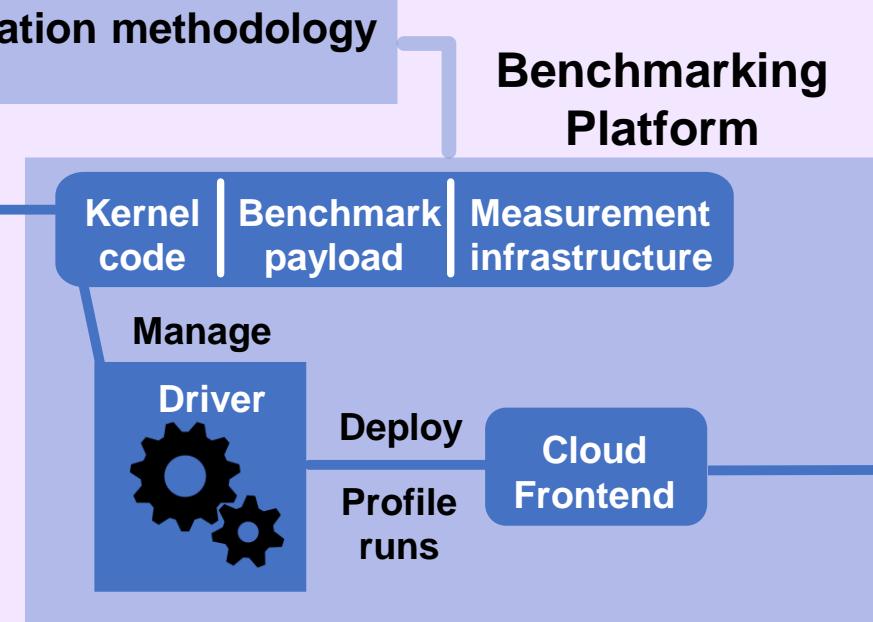
Benchmark Specification



Metrics

Applications

Input datasets



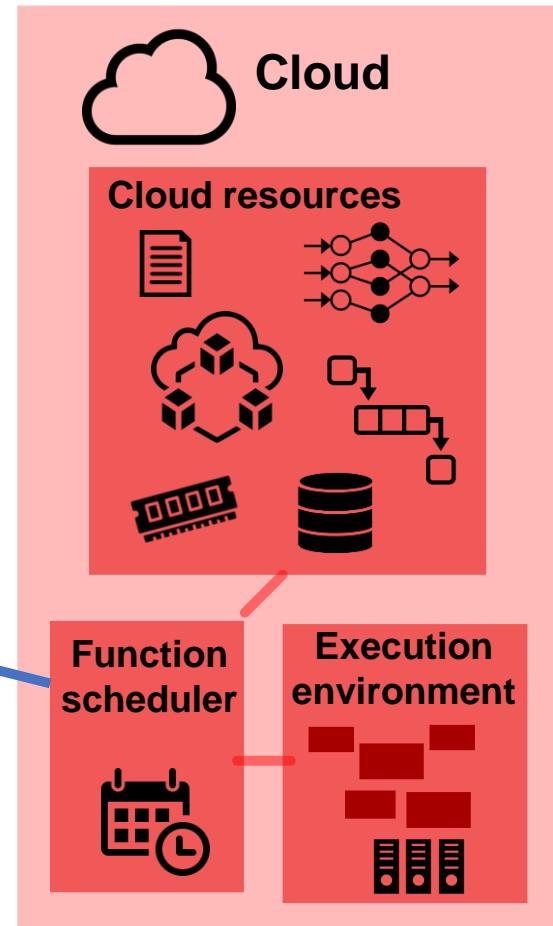
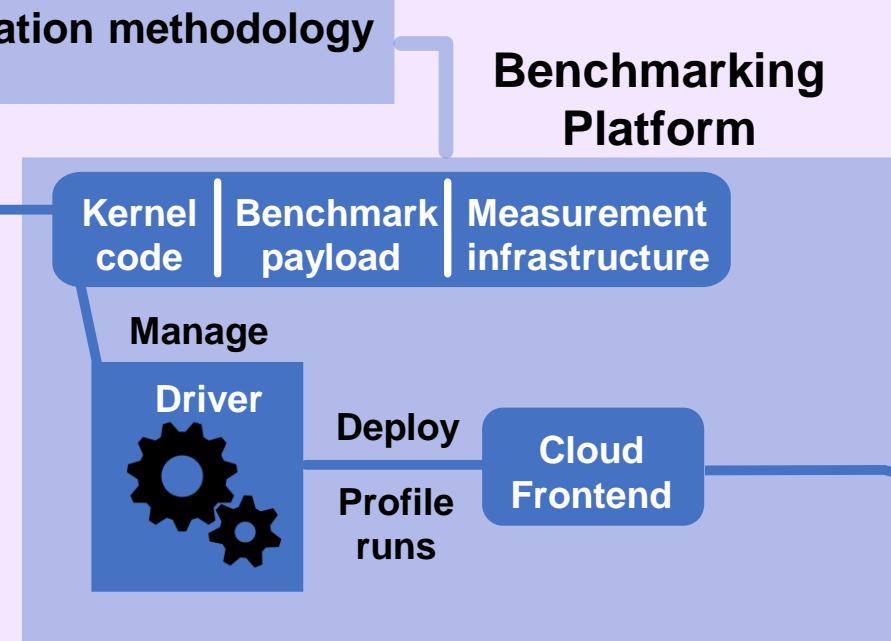
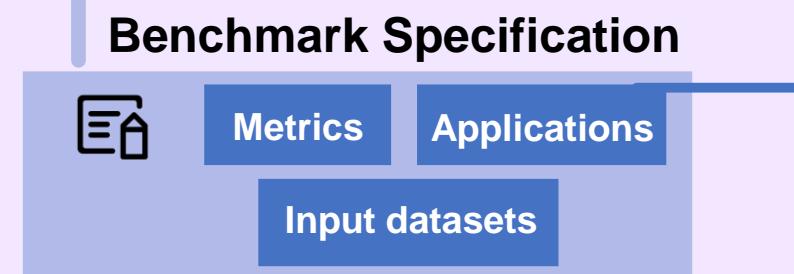
How to build a benchmark?



SeBS: The Serverless Benchmark Suite

Benchmark Design Principles

- Relevance → Scientific evaluation methodology
- Usability → Extensibility



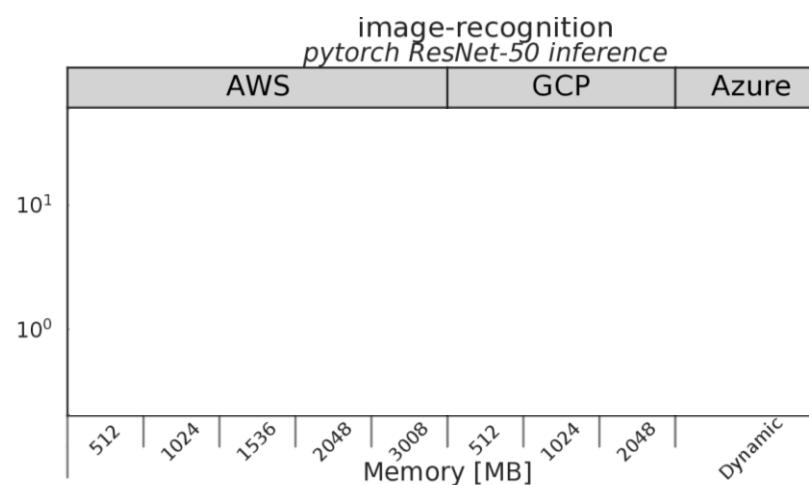
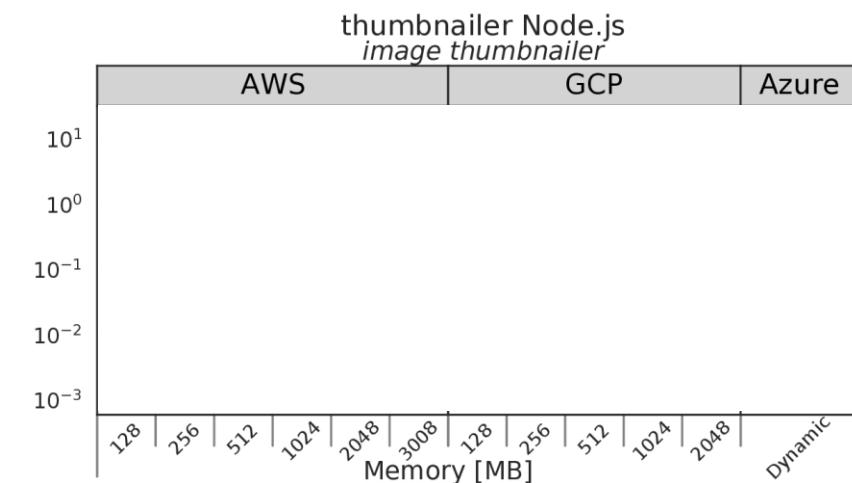
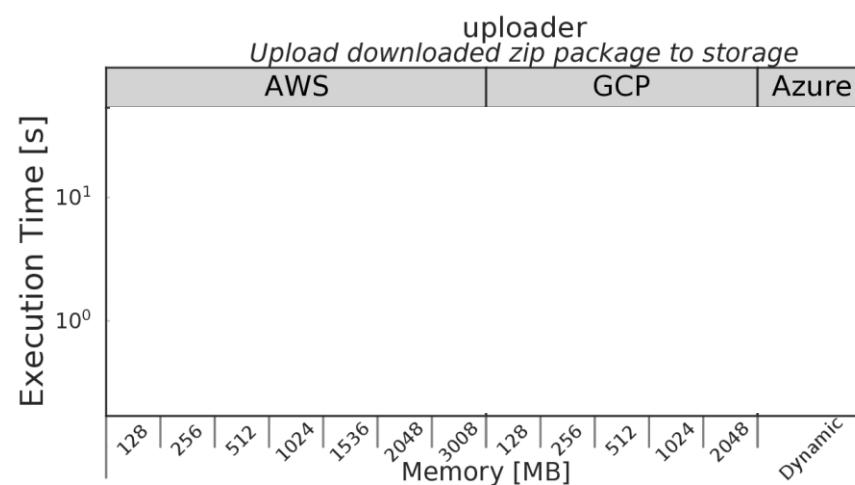
Results and Insights

Results, methods, and insights	
	
\$	
	
	

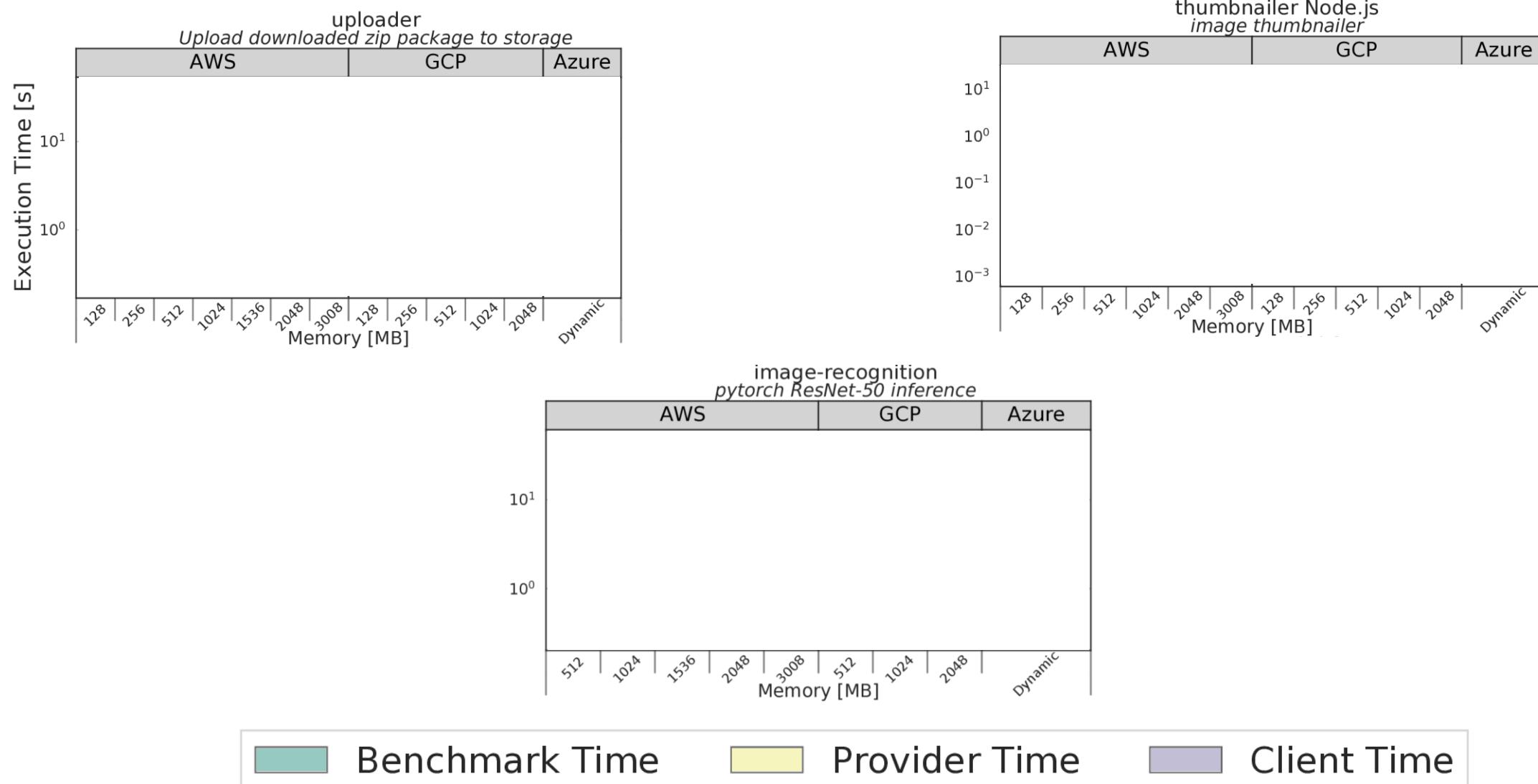
Results and Insights

	Results, methods, and insights
	AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions. I/O-bound functions experience very high latency variations.
	
	
	

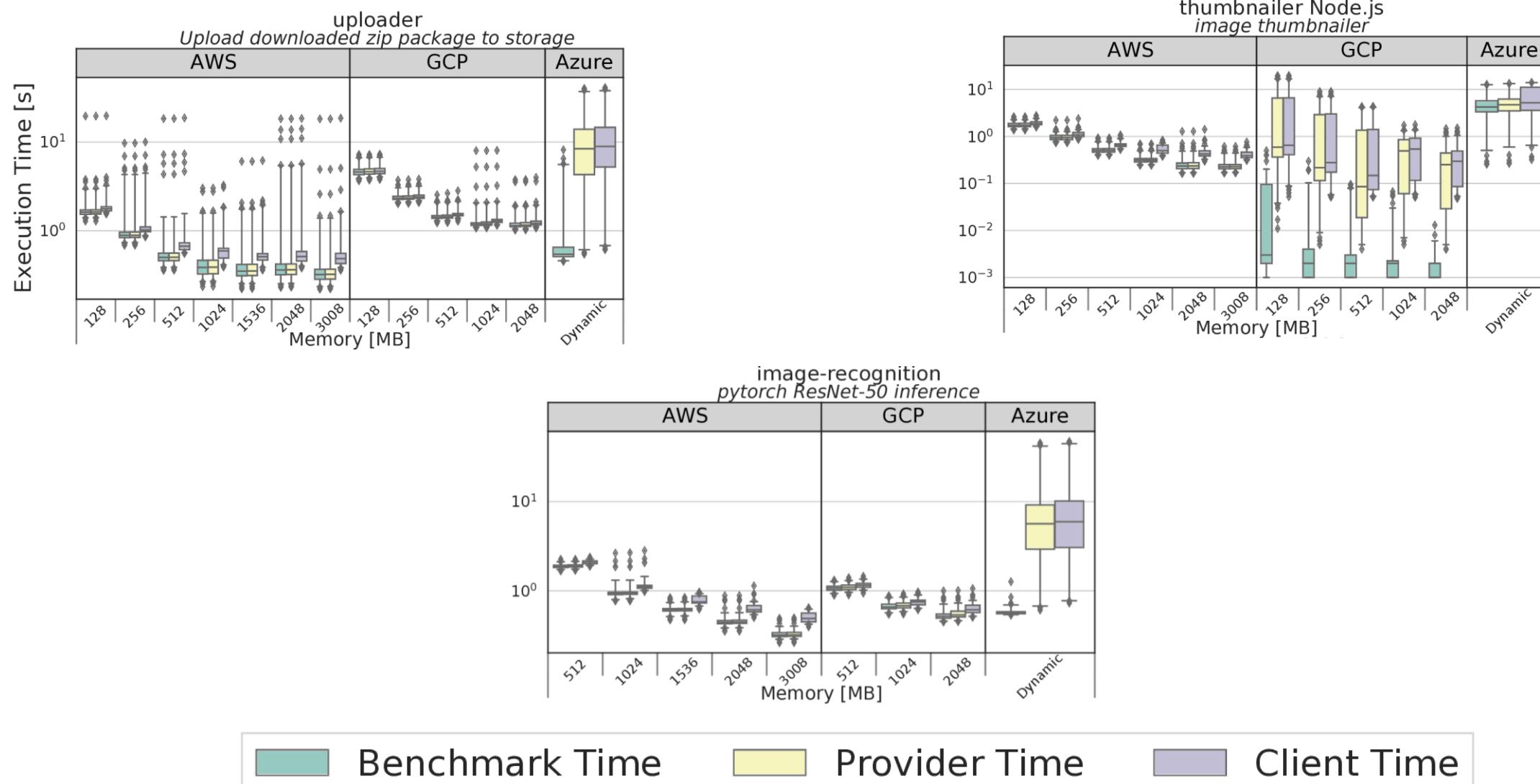
Performance Analysis: Warm Startups



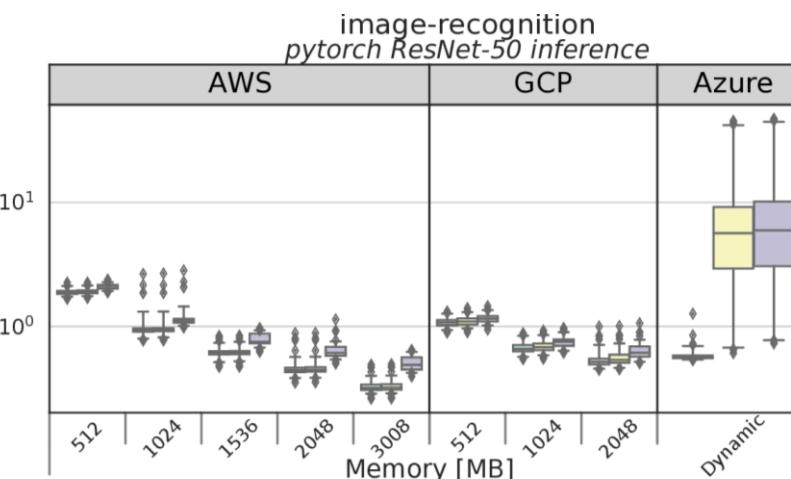
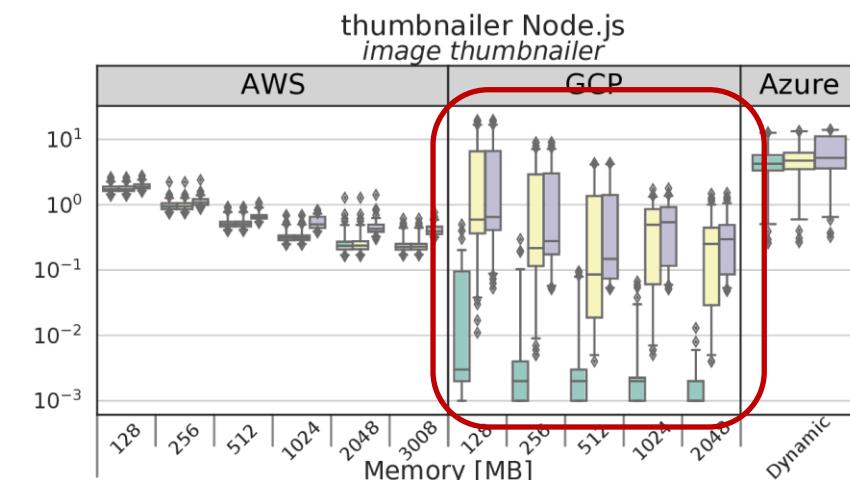
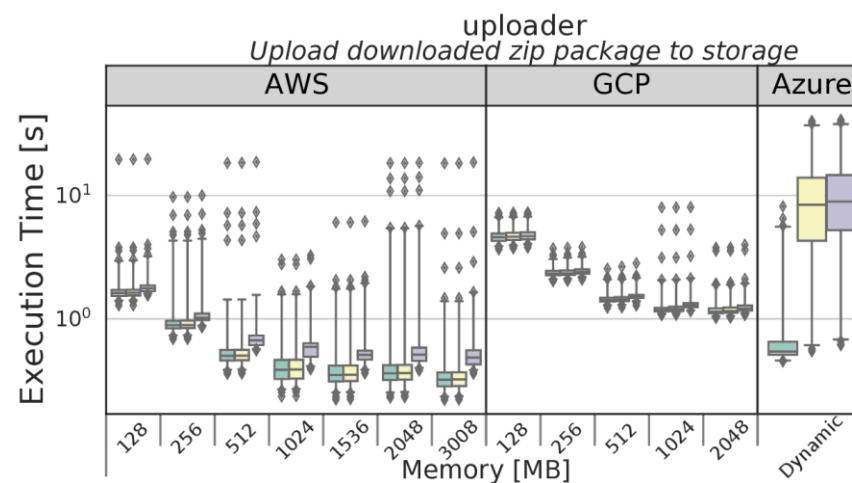
Performance Analysis: Warm Startups



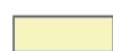
Performance Analysis: Warm Startups



Performance Analysis: Warm Startups



Benchmark Time

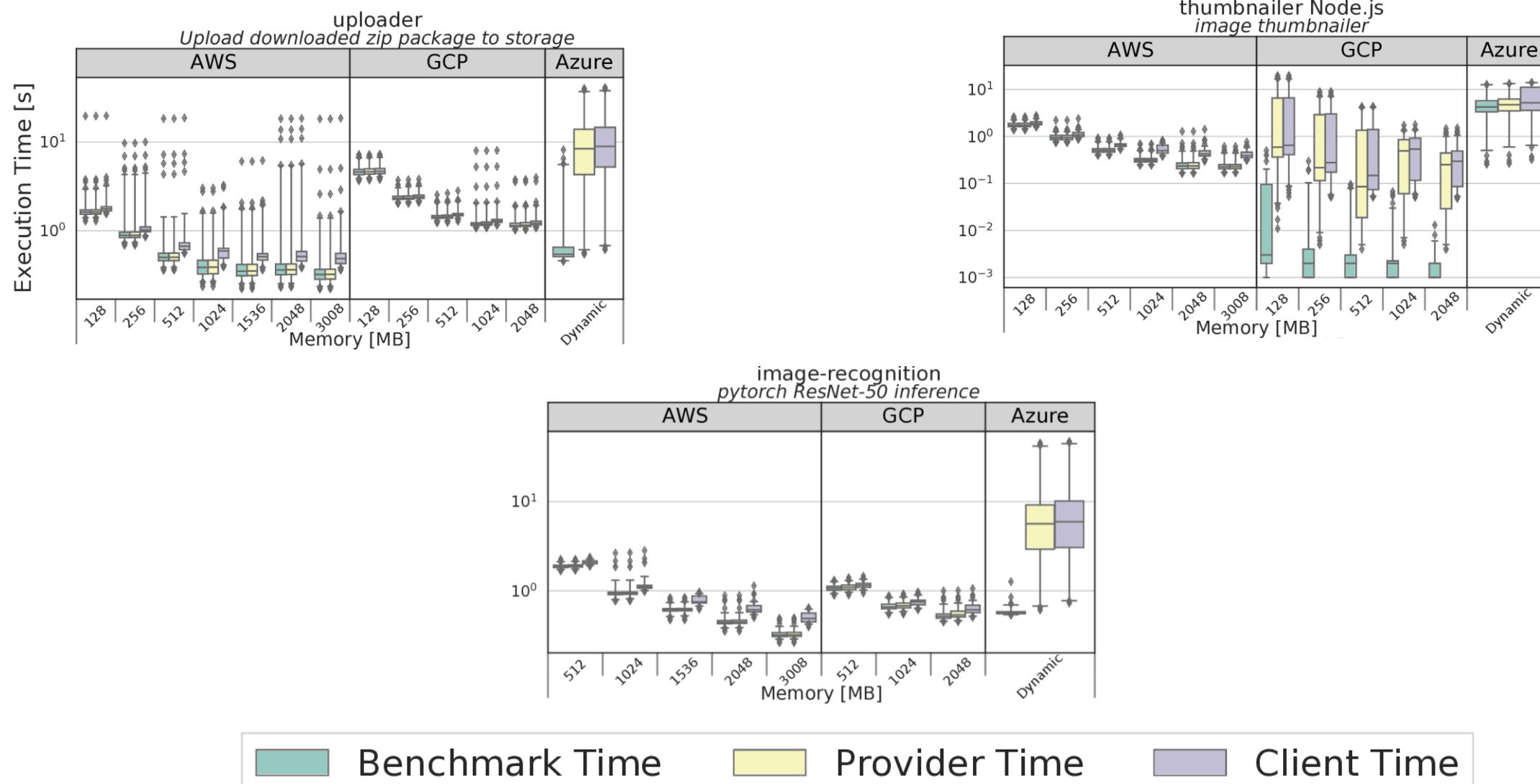


Provider Time

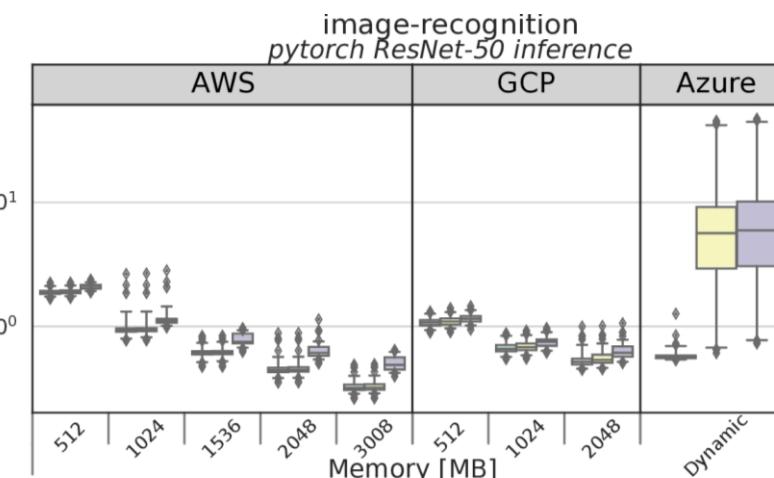
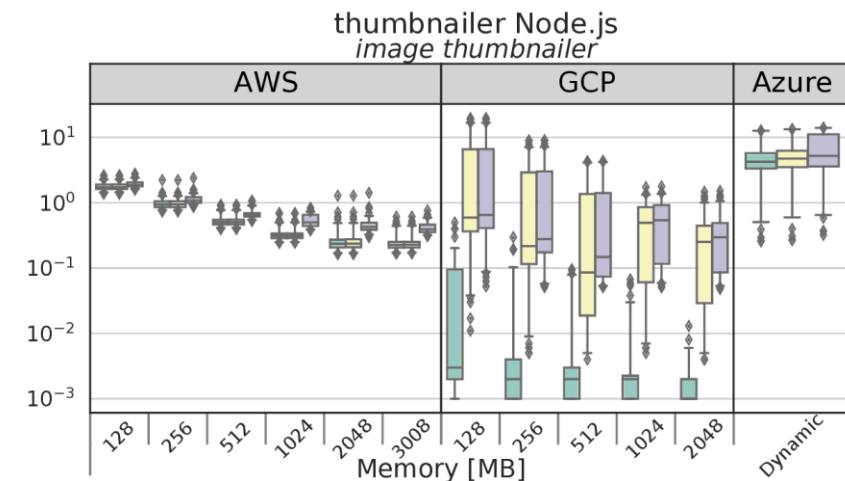
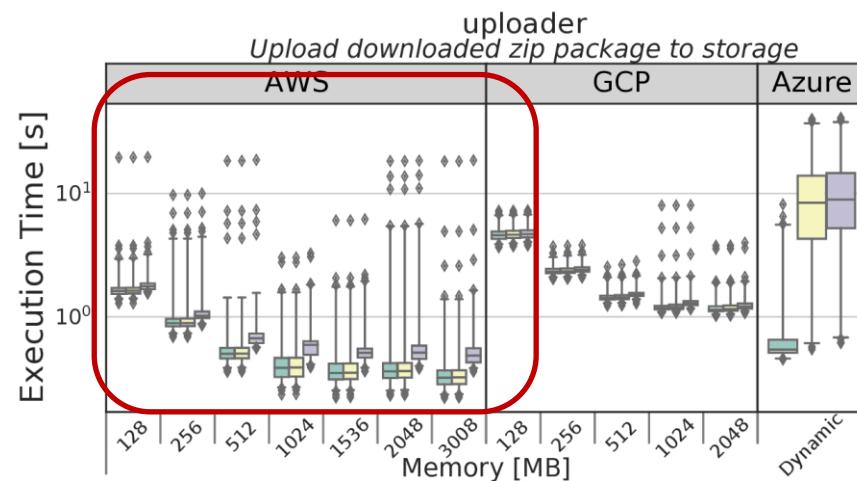


Client Time

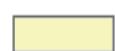
Performance Analysis: Warm Startups



Performance Analysis: Warm Startups



Benchmark Time

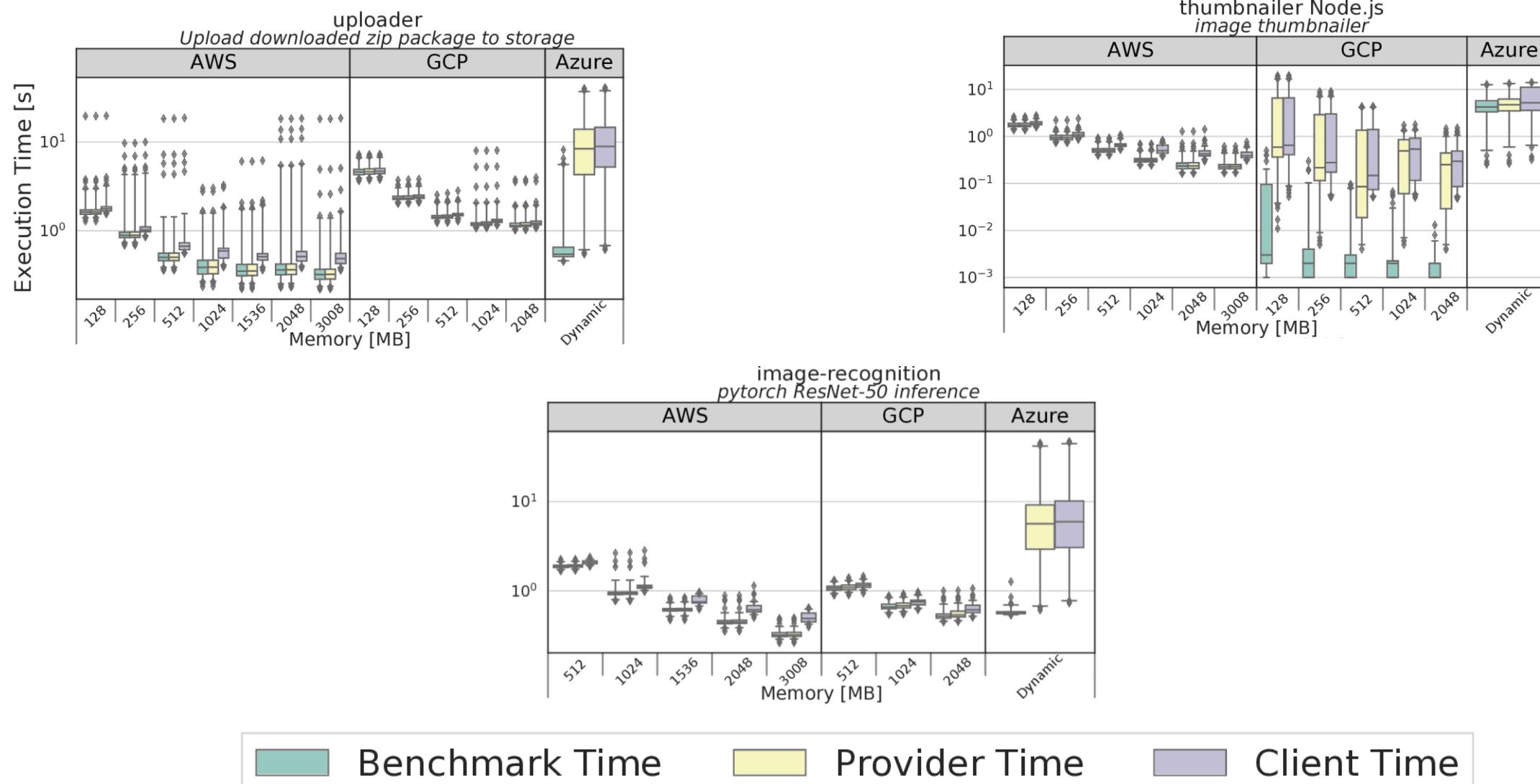


Provider Time

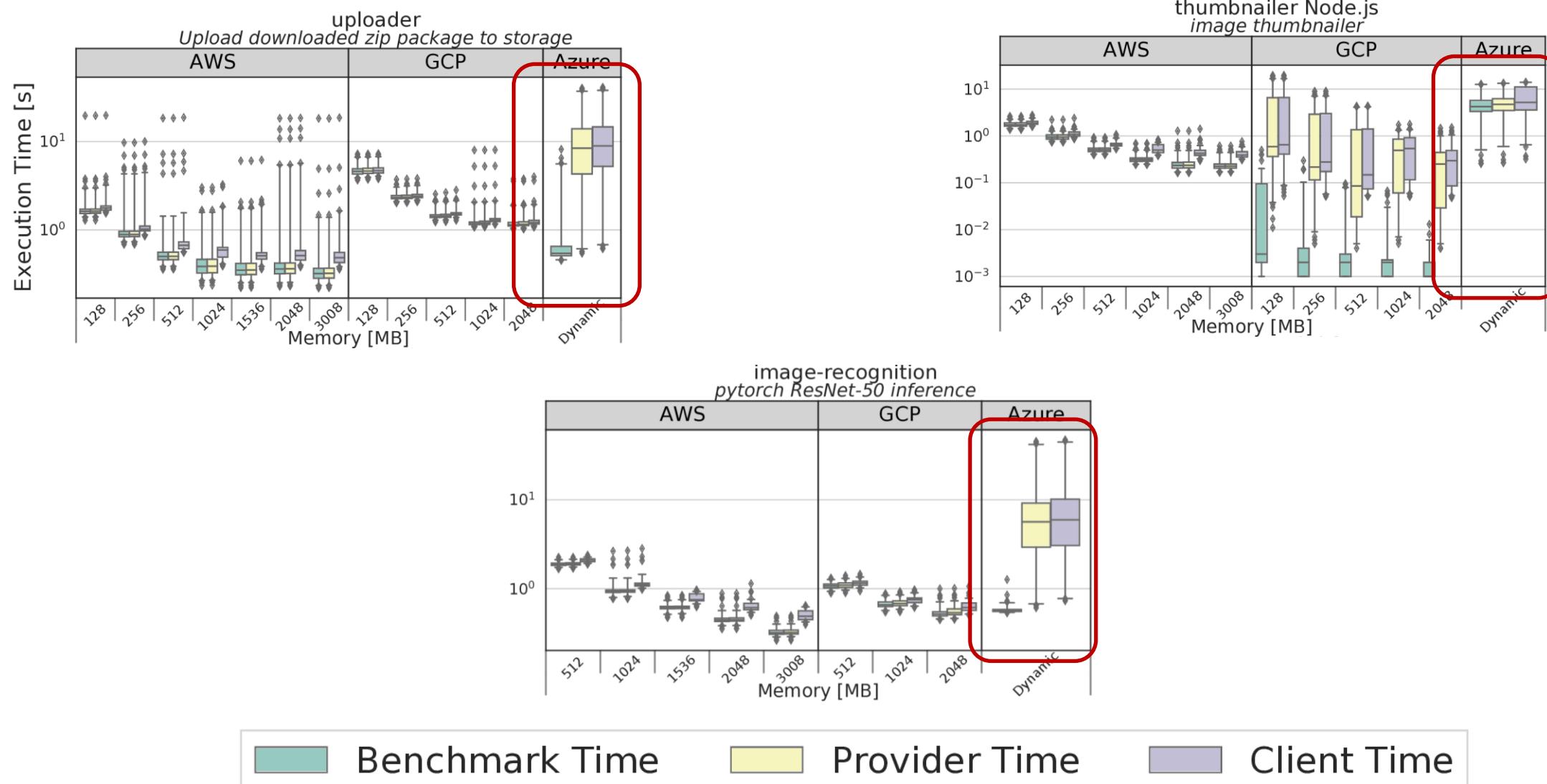


Client Time

Performance Analysis: Warm Startups



Performance Analysis: Warm Startups



Results and Insights

	Results, methods, and insights
	AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions. I/O-bound functions experience very high latency variations.
	
	
	

Results and Insights

Results, methods, and insights	
	AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions. I/O-bound functions experience very high latency variations.
	AWS Lambda performance is not competitive against VMs assuming comparable resources.
	
	

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS [s]						
FaaS [s]						
Overhead						
Memory [MB]						

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Results and Insights

Results, methods, and insights	
	AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions. I/O-bound functions experience very high latency variations.
	AWS Lambda performance is not competitive against VMs assuming comparable resources.
	
	

Results and Insights

Results, methods, and insights	
	AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions. I/O-bound functions experience very high latency variations.
	AWS Lambda performance is not competitive against VMs assuming comparable resources.
	Break-even analysis for IaaS and FaaS deployment.
	
	

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth- First Search
IaaS							
FaaS							

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth- First Search
IaaS							
FaaS							

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.0116/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Cloud Storage [req/h]						
FaaS							

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.0116/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Cloud Storage [req/h]	11371	27503	18819	1284	15312	117153
FaaS							

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.0116/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Cloud Storage [req/h]	11371	27503	18819	1284	15312	117153
FaaS	Eco 1M [\$]						
	Eco Break-Even						
	Perf 1M [\$]						
	Perf Break-Even						

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.0116/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Cloud Storage [req/h]	11371	27503	18819	1284	15312	117153
FaaS	Eco 1M [\$]	3.54	2.29	3.75	32.1	15.8	2.08
	Eco Break-Even	3275	5062	3093	362	733	5568
	Perf 1M [\$]	6.67	3.34	10	50	19.58	2.5
	Perf Break-Even	1740	3480	1160	232	592	4640

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.0116/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Cloud Storage [req/h]	11371	27503	18819	1284	15312	117153
FaaS	Eco 1M [\$]	3.54	2.29	3.75	32.1	15.8	2.08
	Eco Break-Even	3275	5062	3093	362	733	5568
	Perf 1M [\$]	6.67	3.34	10	50	19.58	2.5
	Perf Break-Even	1740	3480	1160	232	592	4640

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.0116/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Results and Insights

Results, methods, and insights	
	AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions. I/O-bound functions experience very high latency variations.
	AWS Lambda performance is not competitive against VMs assuming comparable resources.
	Break-even analysis for IaaS and FaaS deployment.
	

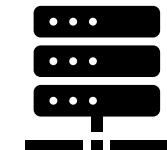
Results and Insights

Results, methods, and insights	
	AWS Lambda achieves the best performance on all workloads. Irregular performance of concurrent Azure Function executions. I/O-bound functions experience very high latency variations.
	AWS Lambda performance is not competitive against VMs assuming comparable resources.
	
	Break-even analysis for IaaS and FaaS deployment.
	Accurate methodology for estimation of invocation latency. Warm latencies are consistent and depend linearly on payload size. Highly variable and unpredictable cold latencies on Azure and GCP.
	

FaaS Analysis: Invocation Overhead

 User

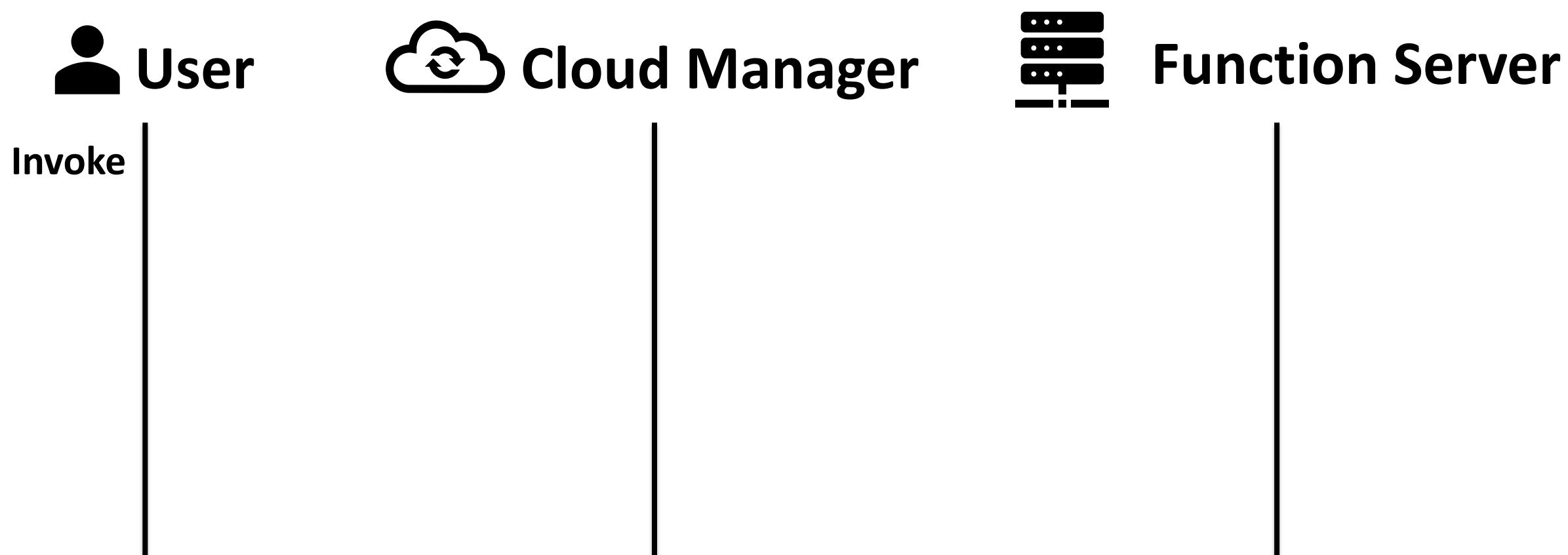
 Cloud Manager



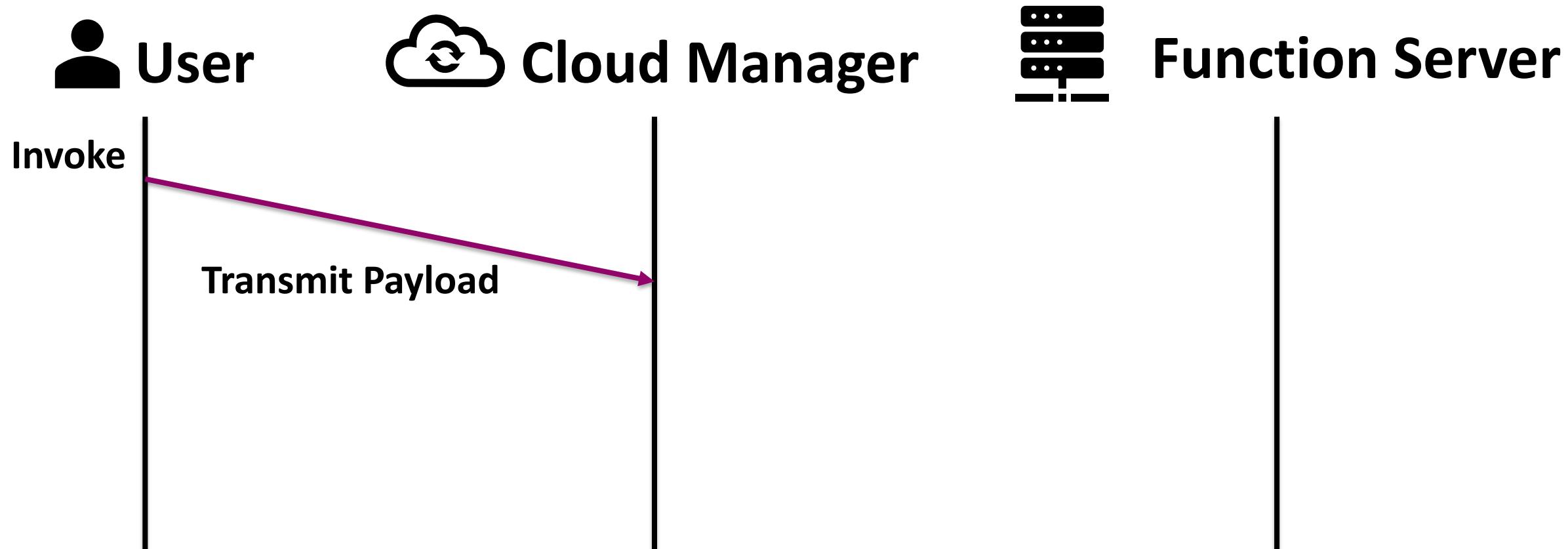
Function Server



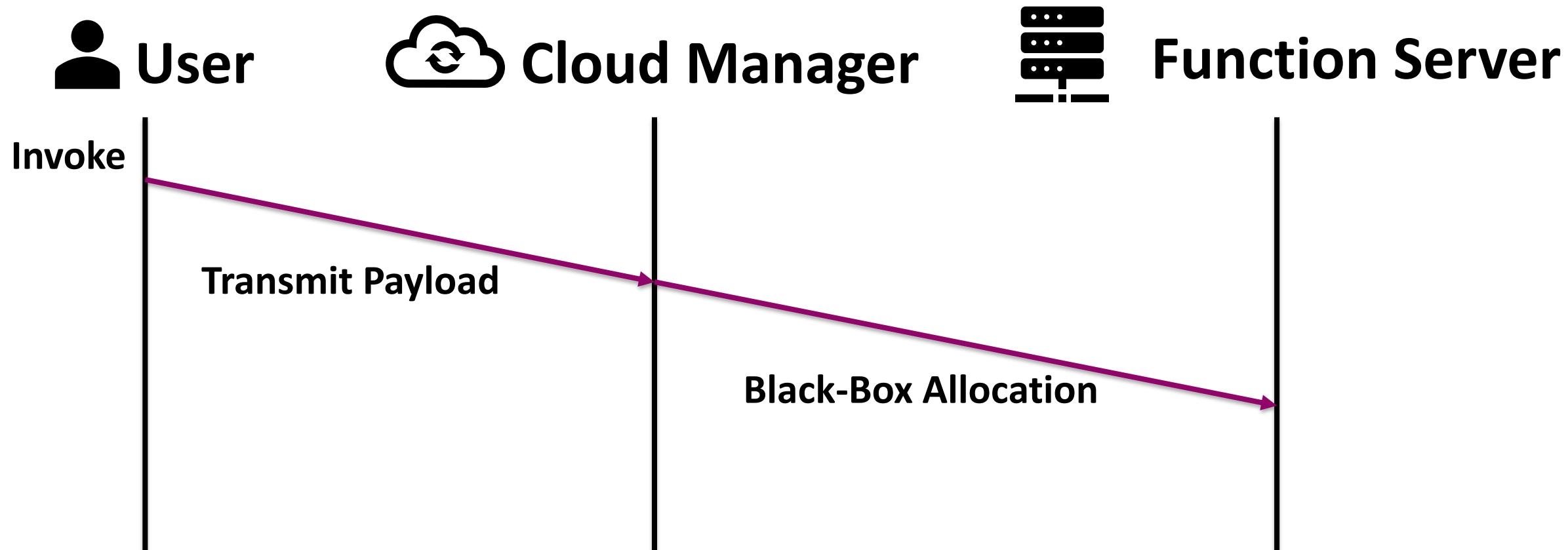
FaaS Analysis: Invocation Overhead



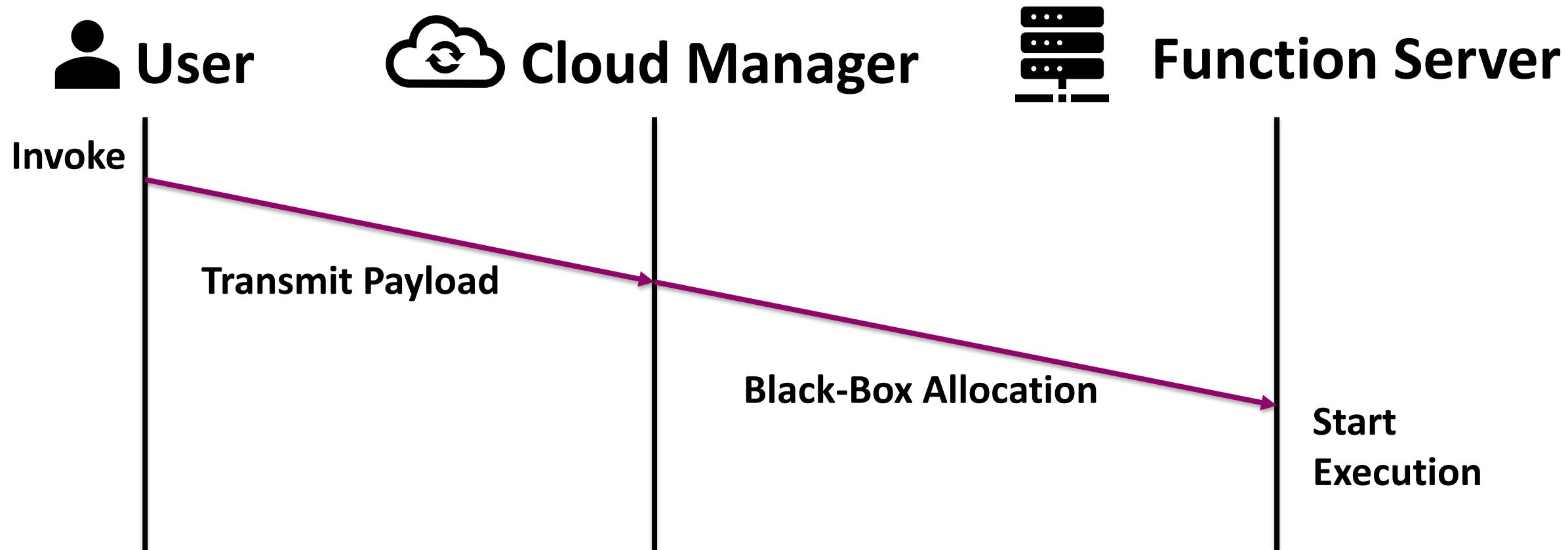
FaaS Analysis: Invocation Overhead



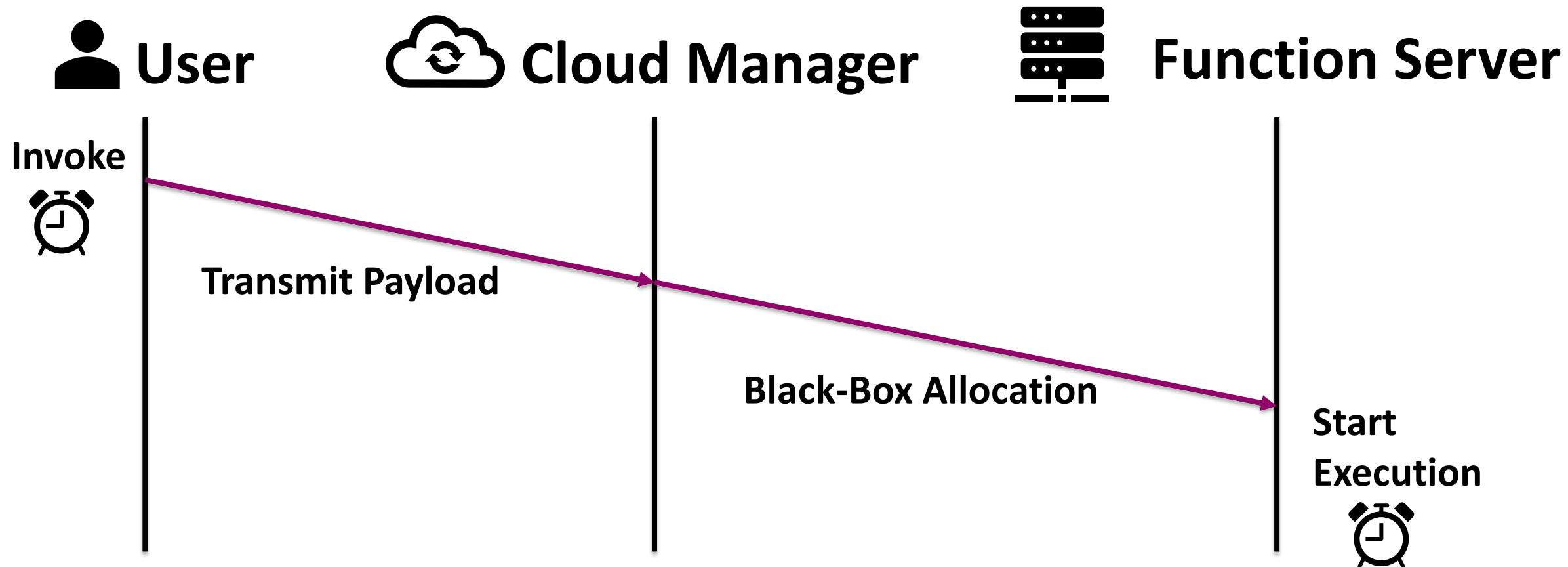
FaaS Analysis: Invocation Overhead



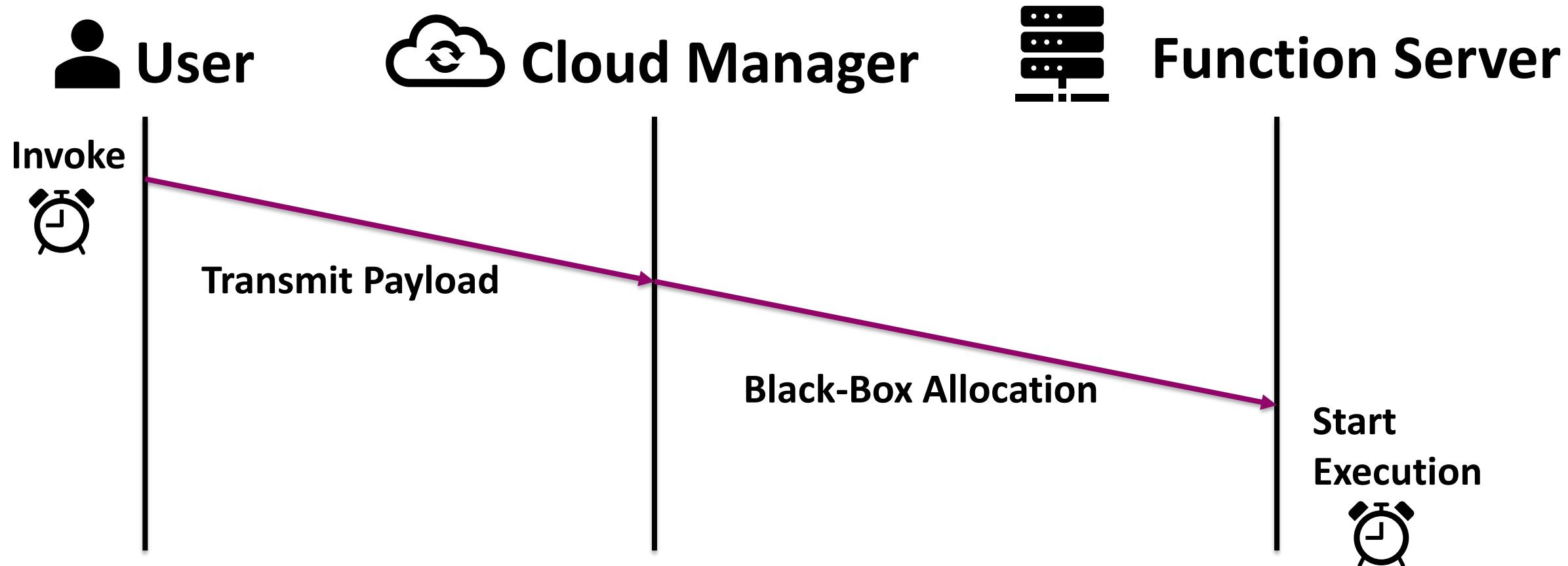
FaaS Analysis: Invocation Overhead



FaaS Analysis: Invocation Overhead



FaaS Analysis: Invocation Overhead



Solution: apply clock-drift estimation protocols!

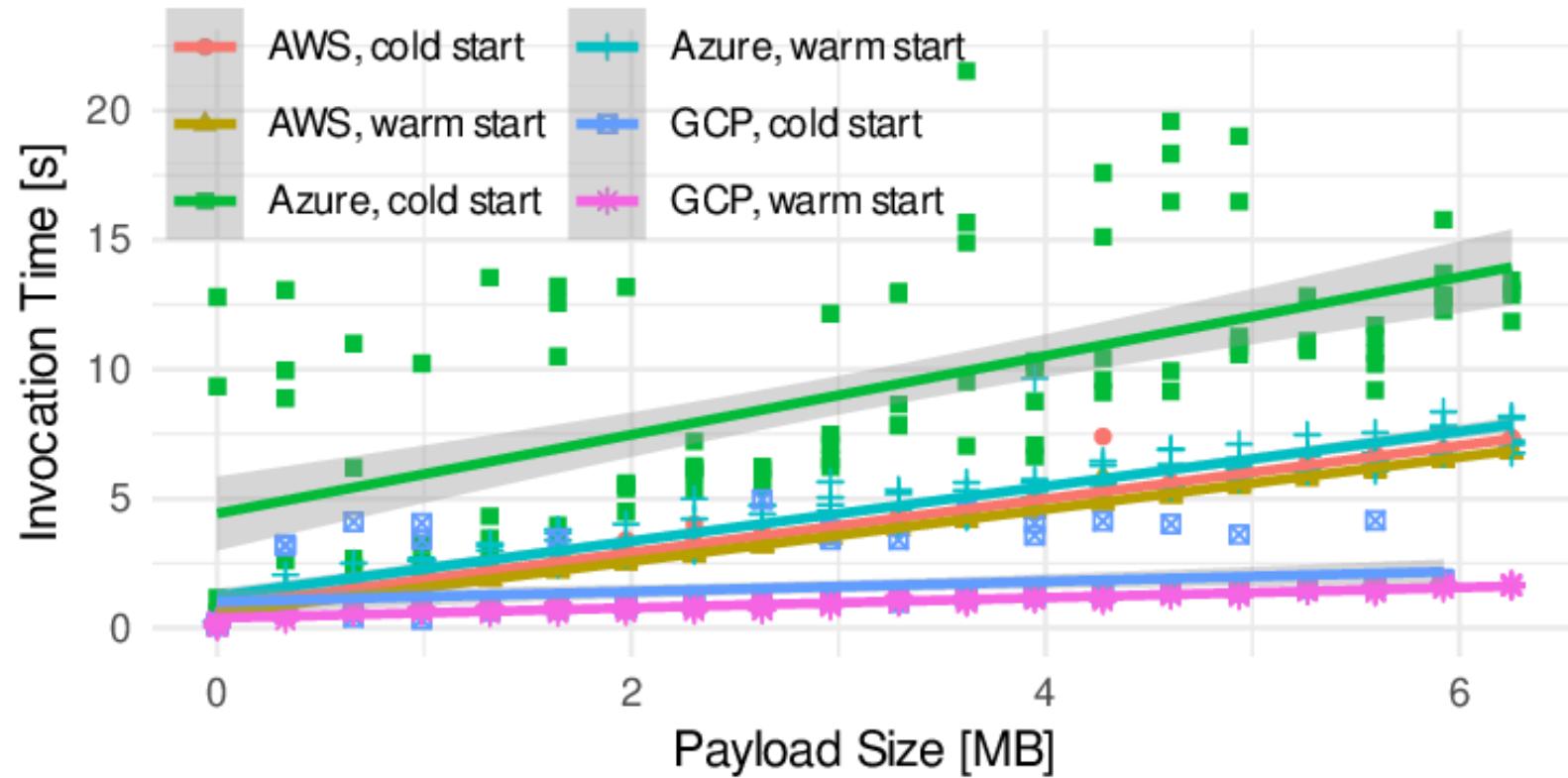
FaaS Analysis: Invocation Overhead

FaaS Analysis: Invocation Overhead

Configuration:

- Compare timestamps on client and function side.
- Clock drift estimation protocol.
- Payload: 1 kB – 5.9 MB

FaaS Analysis: Invocation Overhead



Configuration:

- Compare timestamps on client and function side.
- Clock drift estimation protocol.
- Payload: 1 kB – 5.9 MB

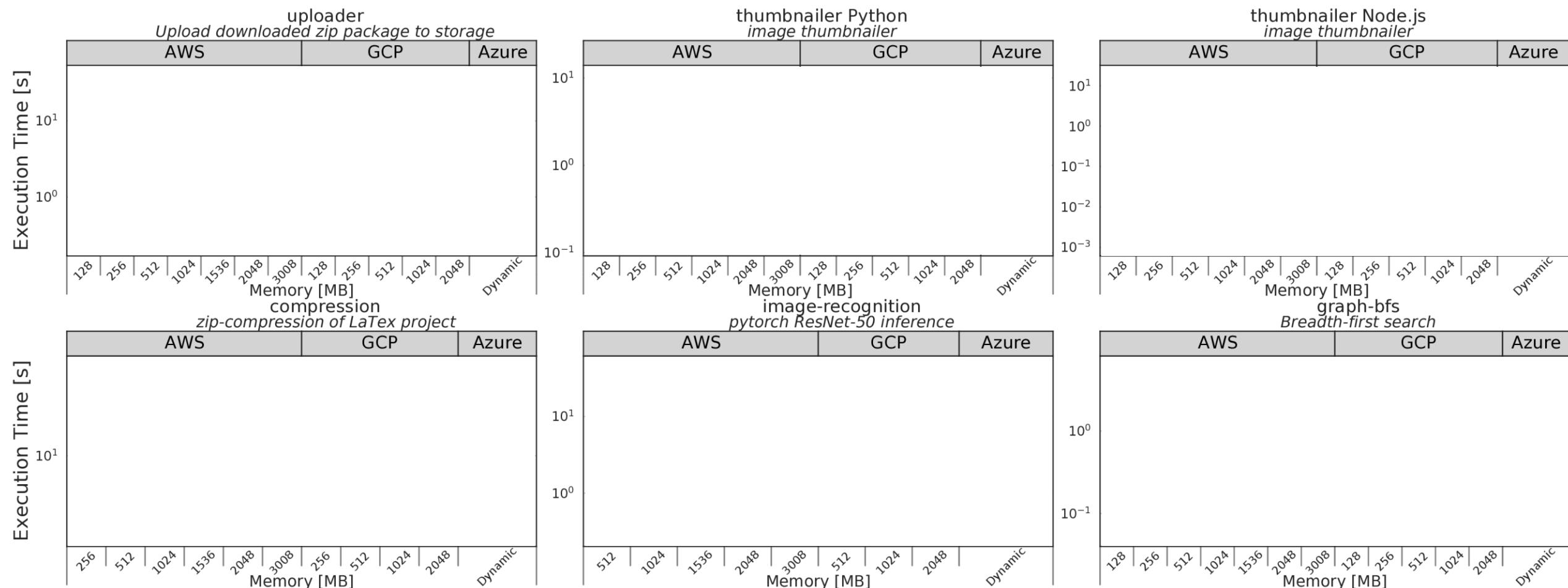
Summary

Summary

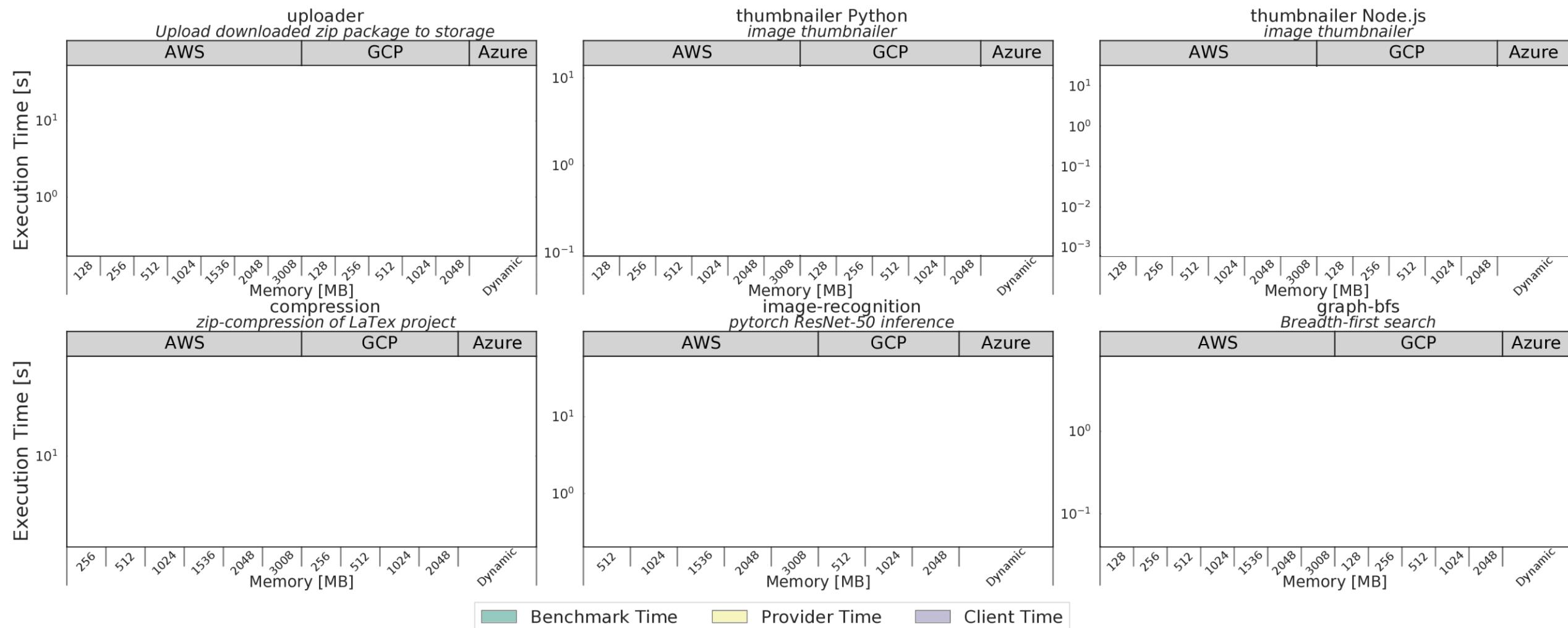


spcl/serverless-benchmarks

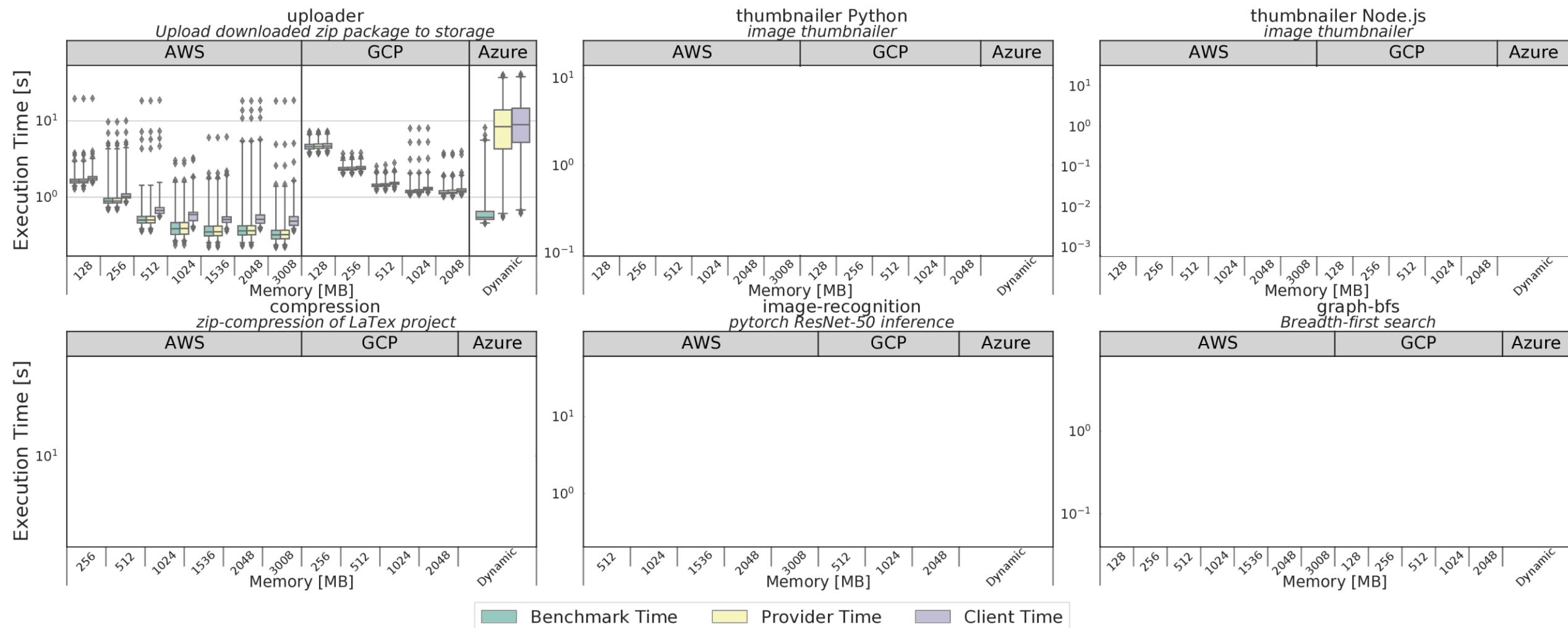
FaaS Analysis



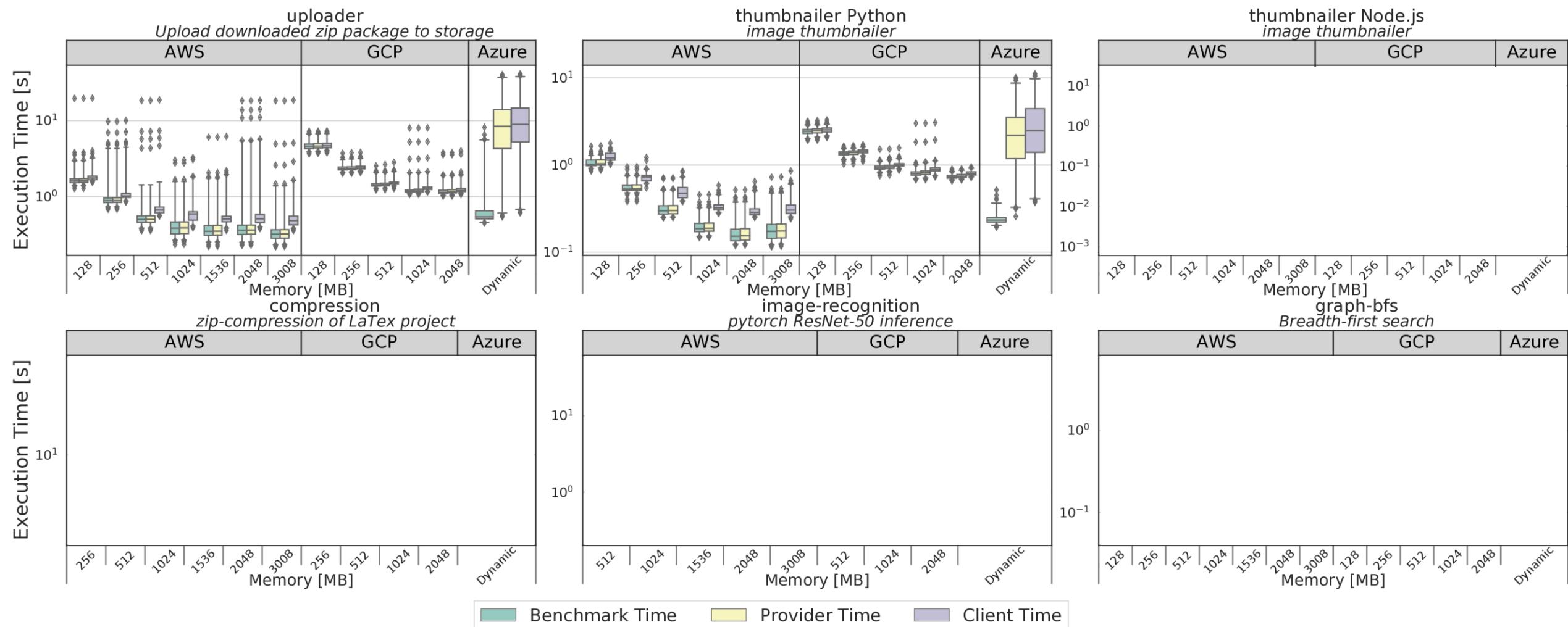
FaaS Analysis



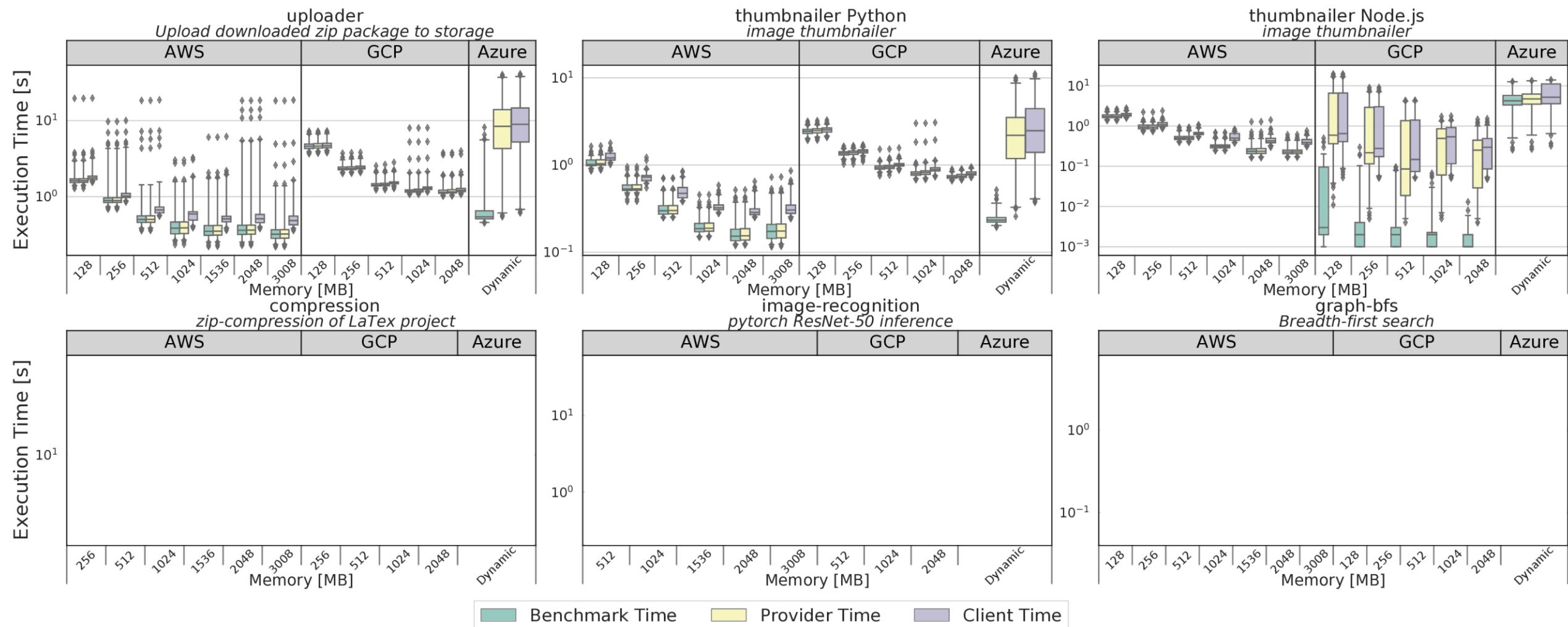
FaaS Analysis



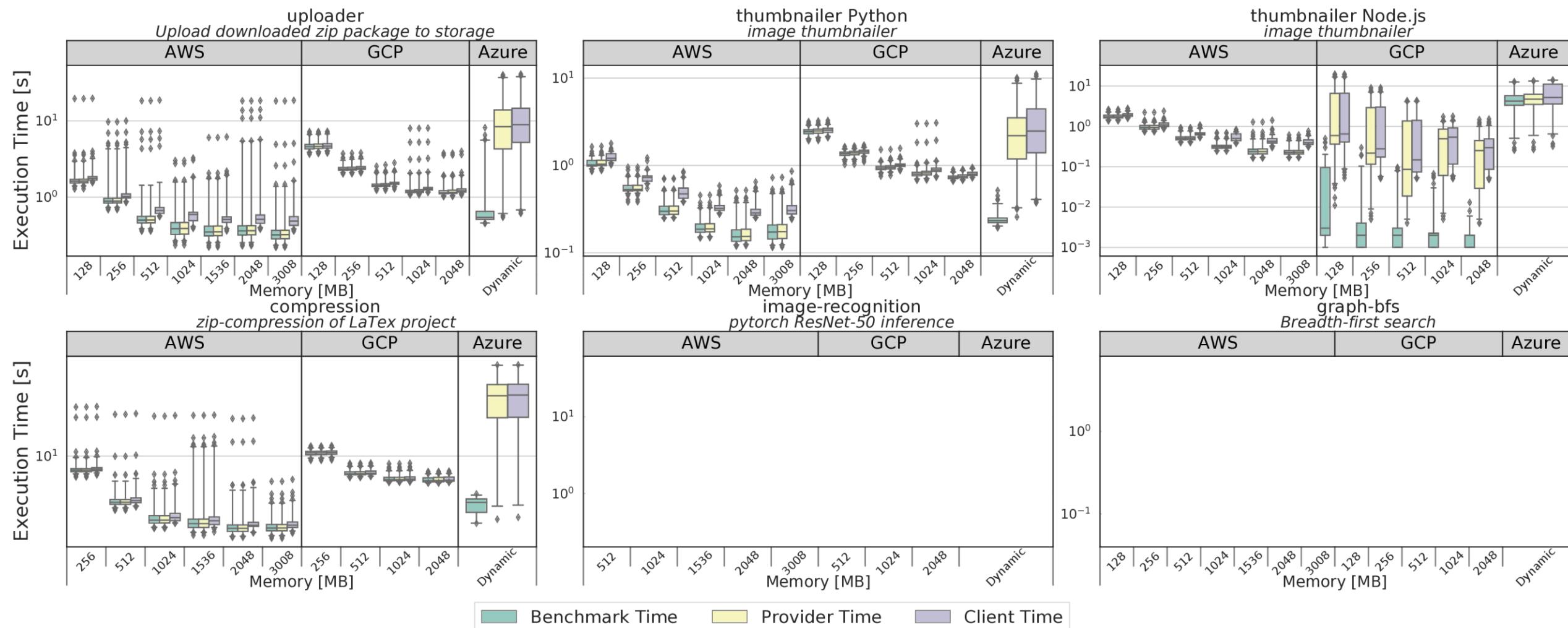
FaaS Analysis



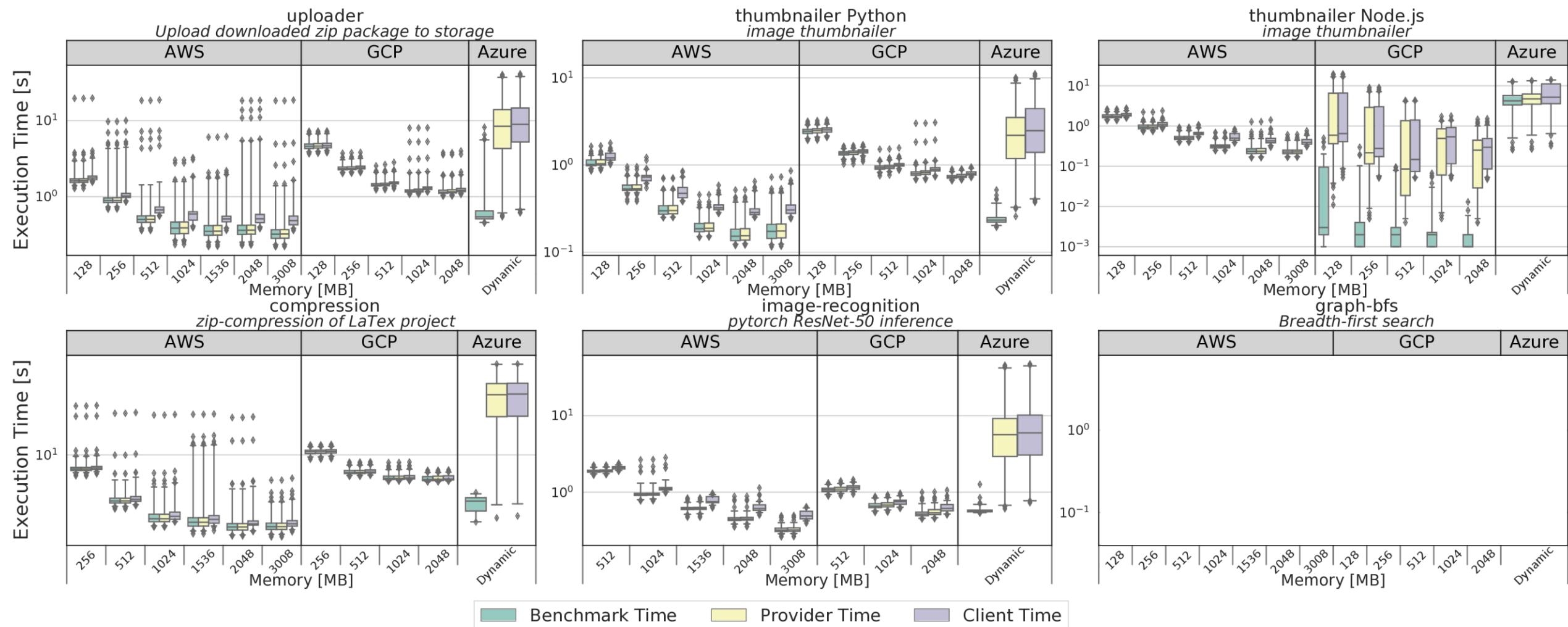
FaaS Analysis



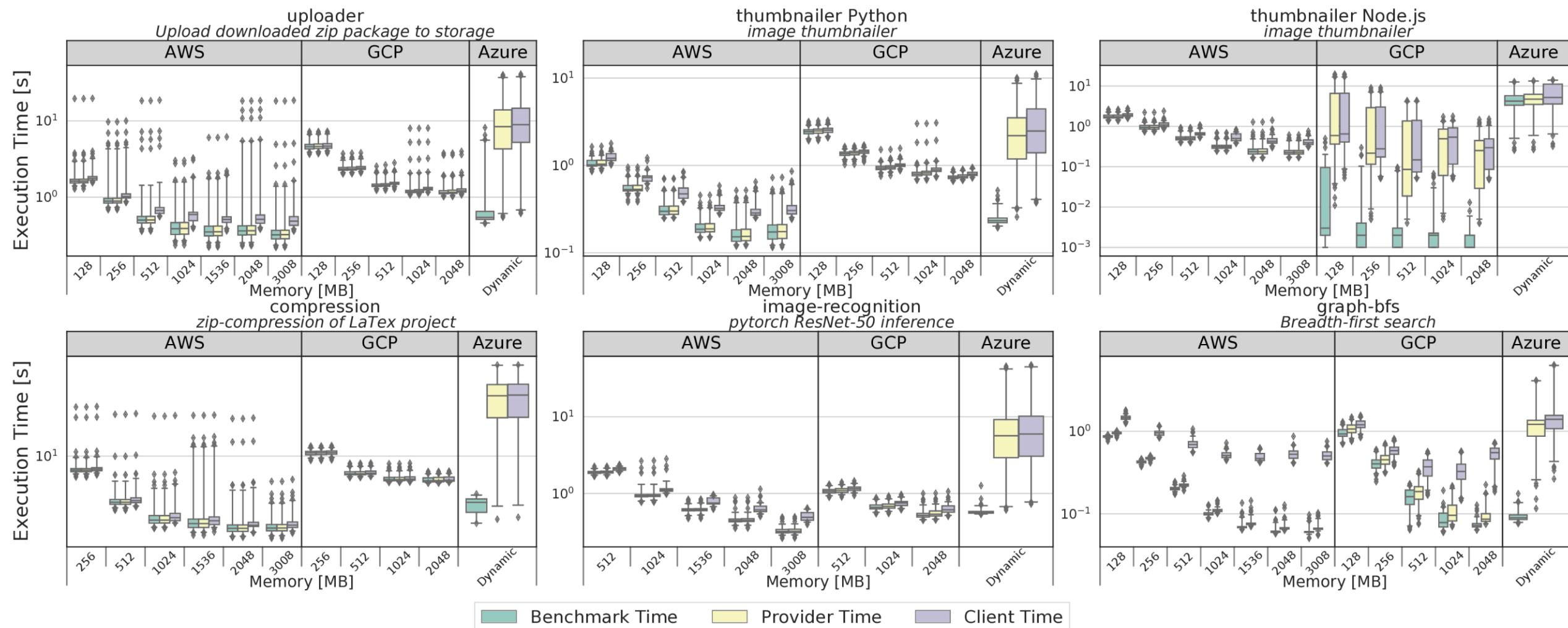
FaaS Analysis



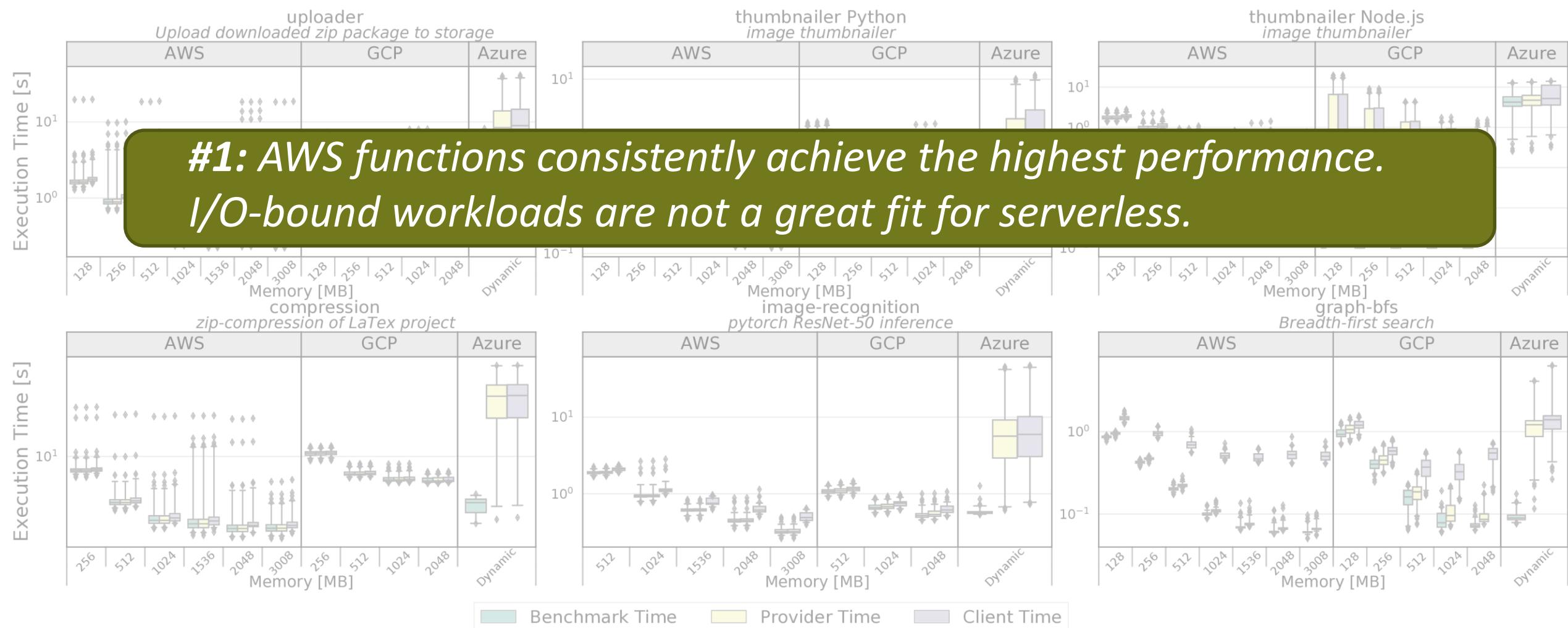
FaaS Analysis



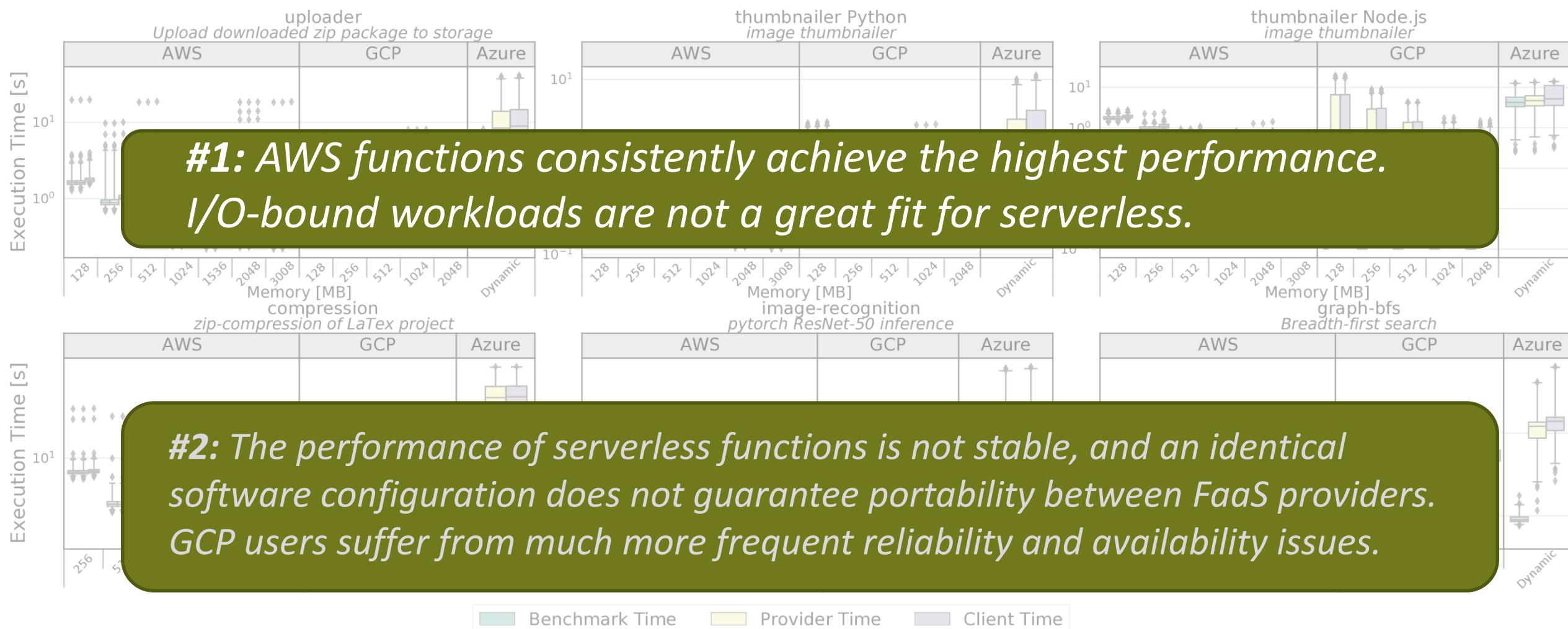
FaaS Analysis



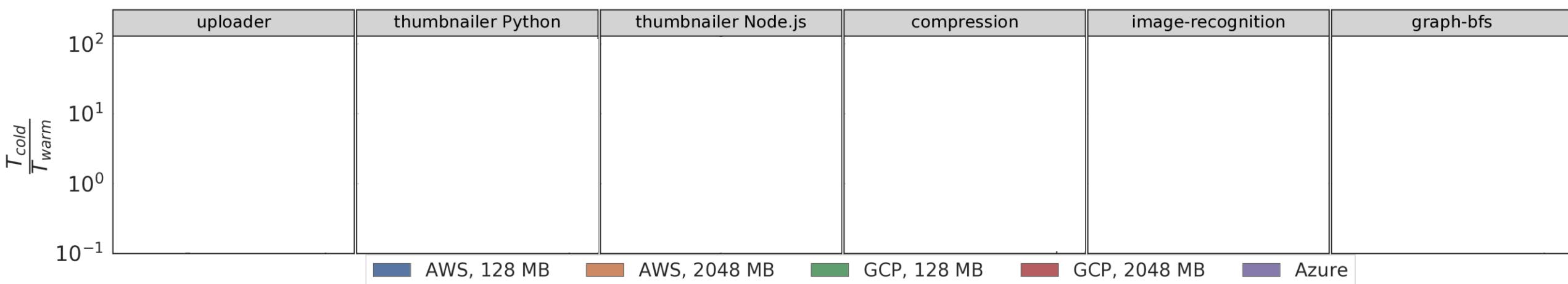
FaaS Analysis



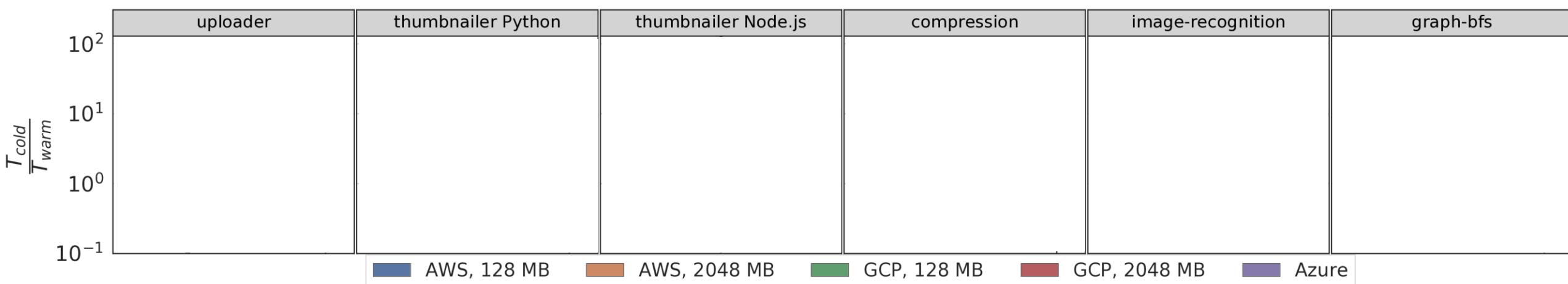
FaaS Analysis



Performance Analysis: Cold Startups



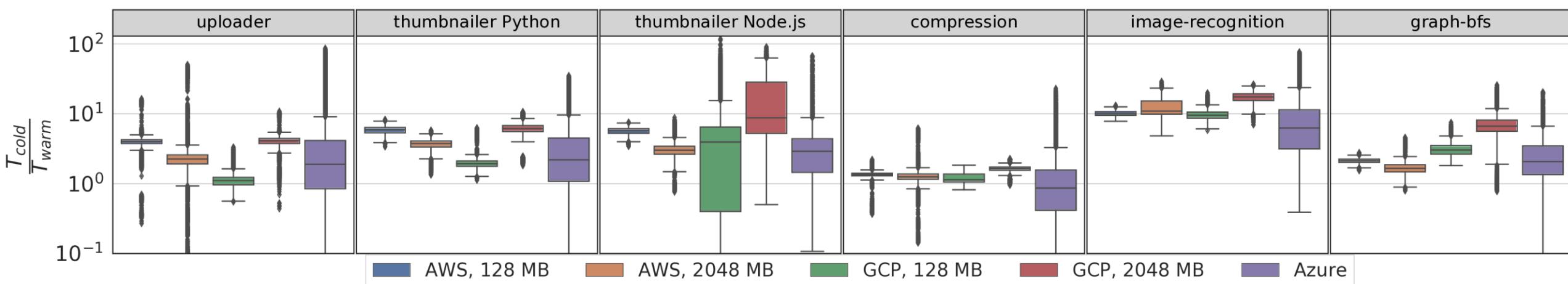
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

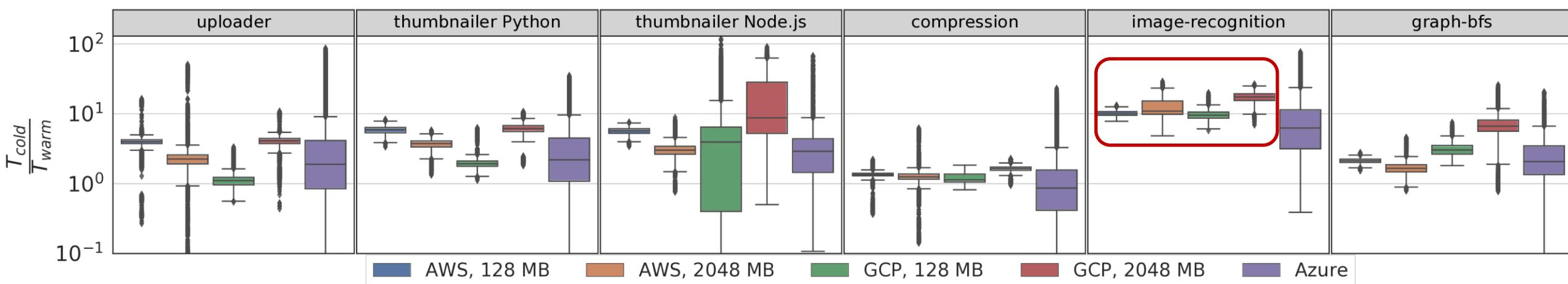
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

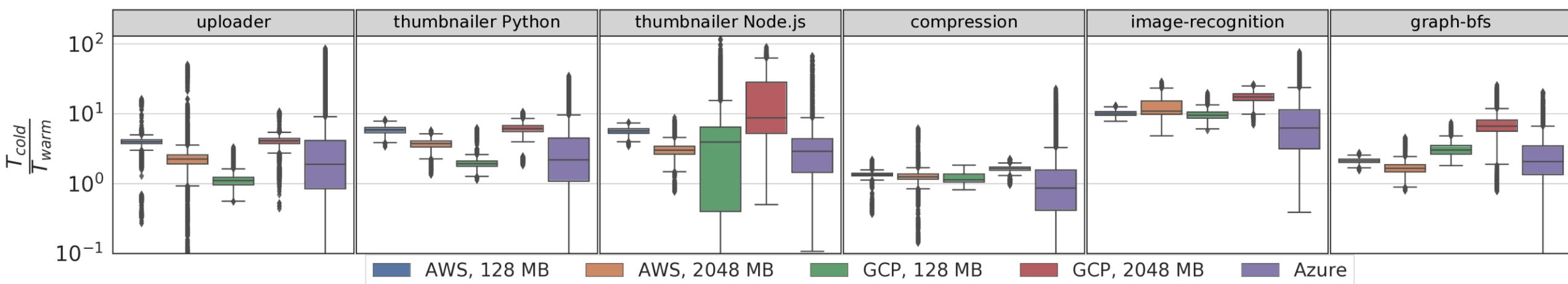
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

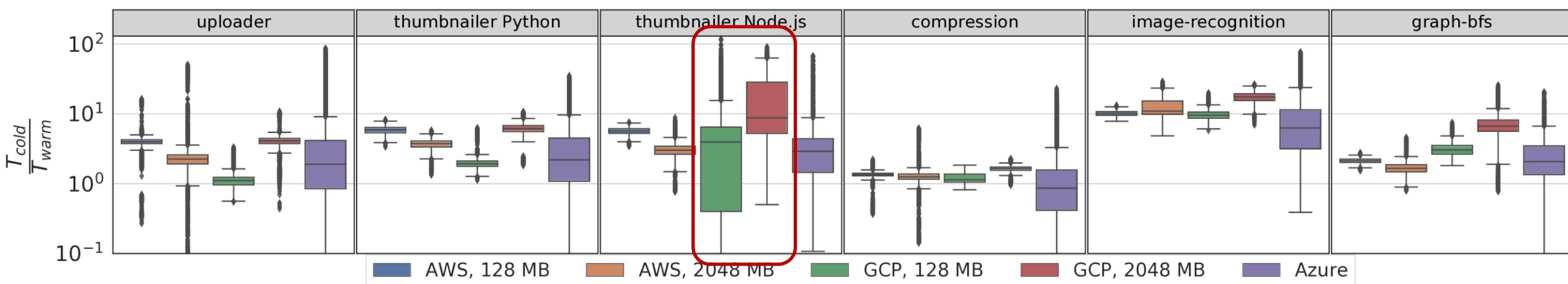
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

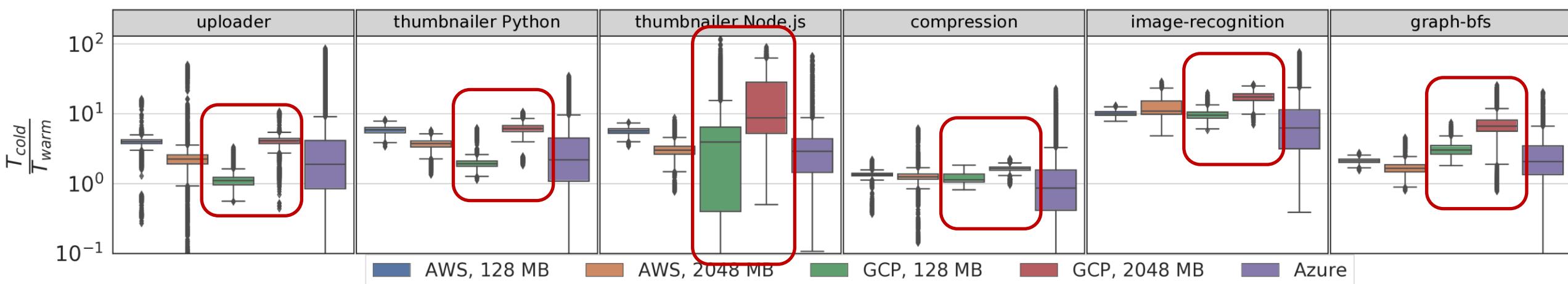
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

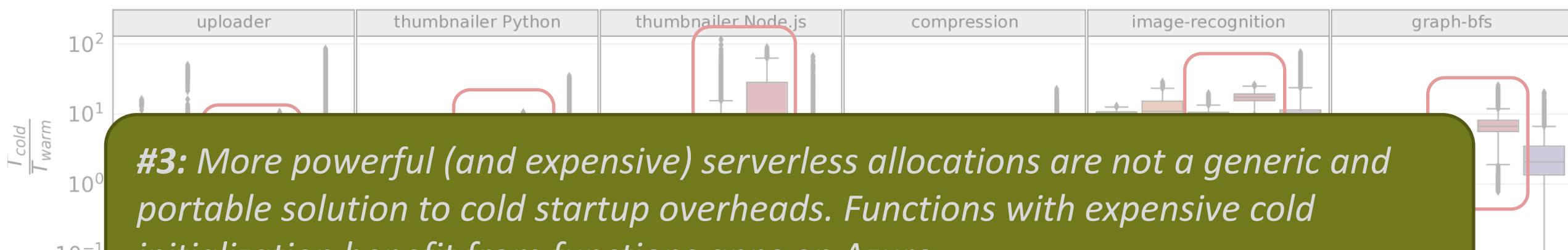
Performance Analysis: Cold Startups



Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

Performance Analysis: Cold Startups



#3: More powerful (and expensive) serverless allocations are not a generic and portable solution to cold startup overheads. Functions with expensive cold initialization benefit from functions apps on Azure.

Configuration:

- **Measurements:** 200 warm and 200 cold executions.
- **Estimation:** all N^2 combinations of N warm and N cold executions.
- **Azure Functions:** mixed cold and warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS, Local Storage [s]						
IaaS, S3 [s]						
FaaS [s]						
Overhead, Local Storage						
Overhead, S3						
Memory [MB]						

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS, Local Storage [s]						
IaaS, S3 [s]						
FaaS [s]						
Overhead, Local Storage						
Overhead, S3						
Memory [MB]						

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS, Local Storage [s]	0.216	0.045	0.166	0.808	0.203	0.03
IaaS, S3 [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead, Local Storage	1.79x	4.14x	1.43x	3.65x	1.58x	2.49x
Overhead, S3	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS, Local Storage [s]	0.216	0.045	0.166	0.808	0.203	0.03
IaaS, S3 [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead, Local Storage	1.79x	4.14x	1.43x	3.65x	1.58x	2.49x
Overhead, S3	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS, Local Storage [s]	0.216	0.045	0.166	0.808	0.203	0.03
IaaS, S3 [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead, Local Storage	1.79x	4.14x	1.43x	3.65x	1.58x	2.49x
Overhead, S3	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS, Local Storage [s]	0.216	0.045	0.166	0.808	0.203	0.03
IaaS, S3 [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead, Local Storage	1.79x	4.14x	1.43x	3.65x	1.58x	2.49x
Overhead, S3	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS, Local Storage [s]	0.216	0.045	0.166	0.808	0.203	0.03
IaaS, S3 [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead, Local Storage	1.79x	4.14x	1.43x	3.65x	1.58x	2.49x
Overhead, S3	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS, Local Storage [s]	0.216	0.045	0.166	0.808	0.203	0.03
IaaS, S3 [s]	0.316	0.13	0.191	2.803	0.235	0.03
FaaS [s]	0.389	0.188	0.253	2.949	0.321	0.075
Overhead, Local Storage	1.79x	4.14x	1.43x	3.65x	1.58x	2.49x
Overhead, S3	1.23x	1.43x	1.24x	1.05x	1.37x	2.4x
Memory [MB]	1024	1024	2048	1024	3008	1536

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Performance Analysis: IaaS vs FaaS

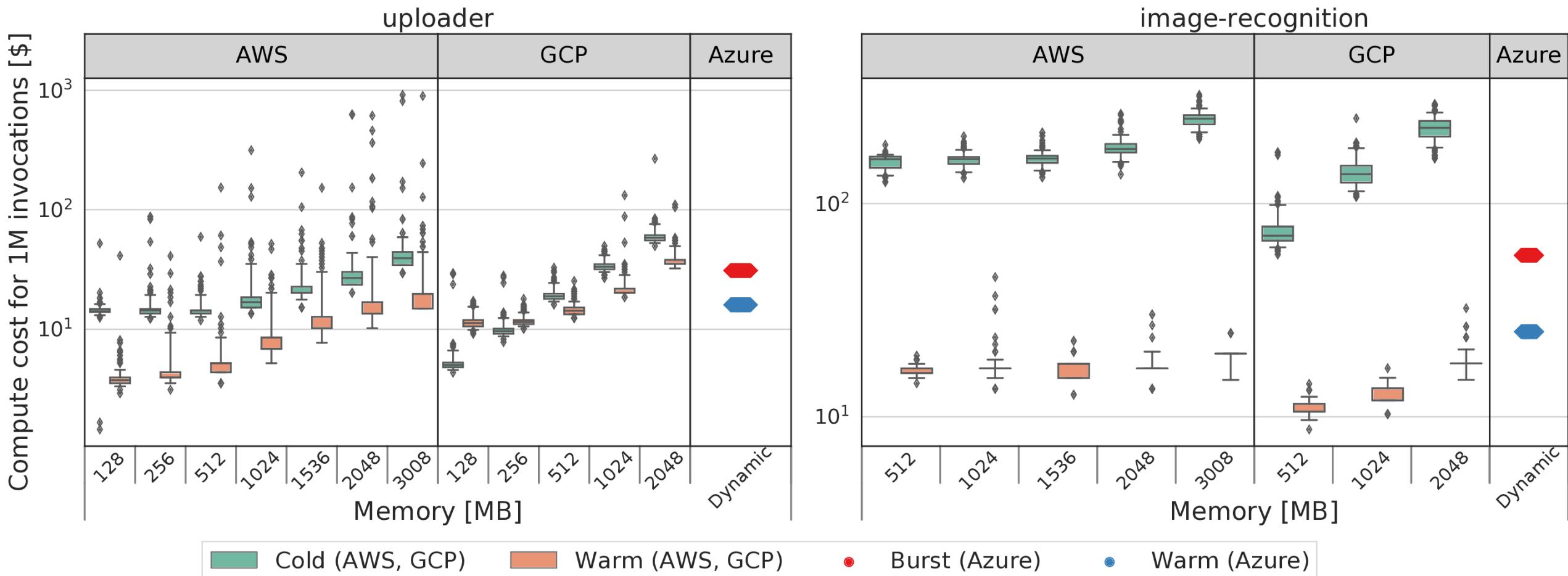
	Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS, Local Storage [s]	0.216	0.045	0.166	0.808	0.203	0.03
IaaS, S3 [s]	0.216	0.13	0.191	2.803	0.235	0.03
FaaS, Local Storage [s]	0.216	0.13	0.191	2.803	0.235	0.03
Overhead, Local Storage [%]	0	0	0	0	0	0
Overhead, S3 [%]	0	0	0	0	0	0
Memory [MB]	1024	1024	2048	1024	5008	1550

#4: Performance overheads of FaaS executions are not uniformly distributed across application classes. The transition from a VM-based deployment to serverless architecture is currently accompanied by significant performance losses.

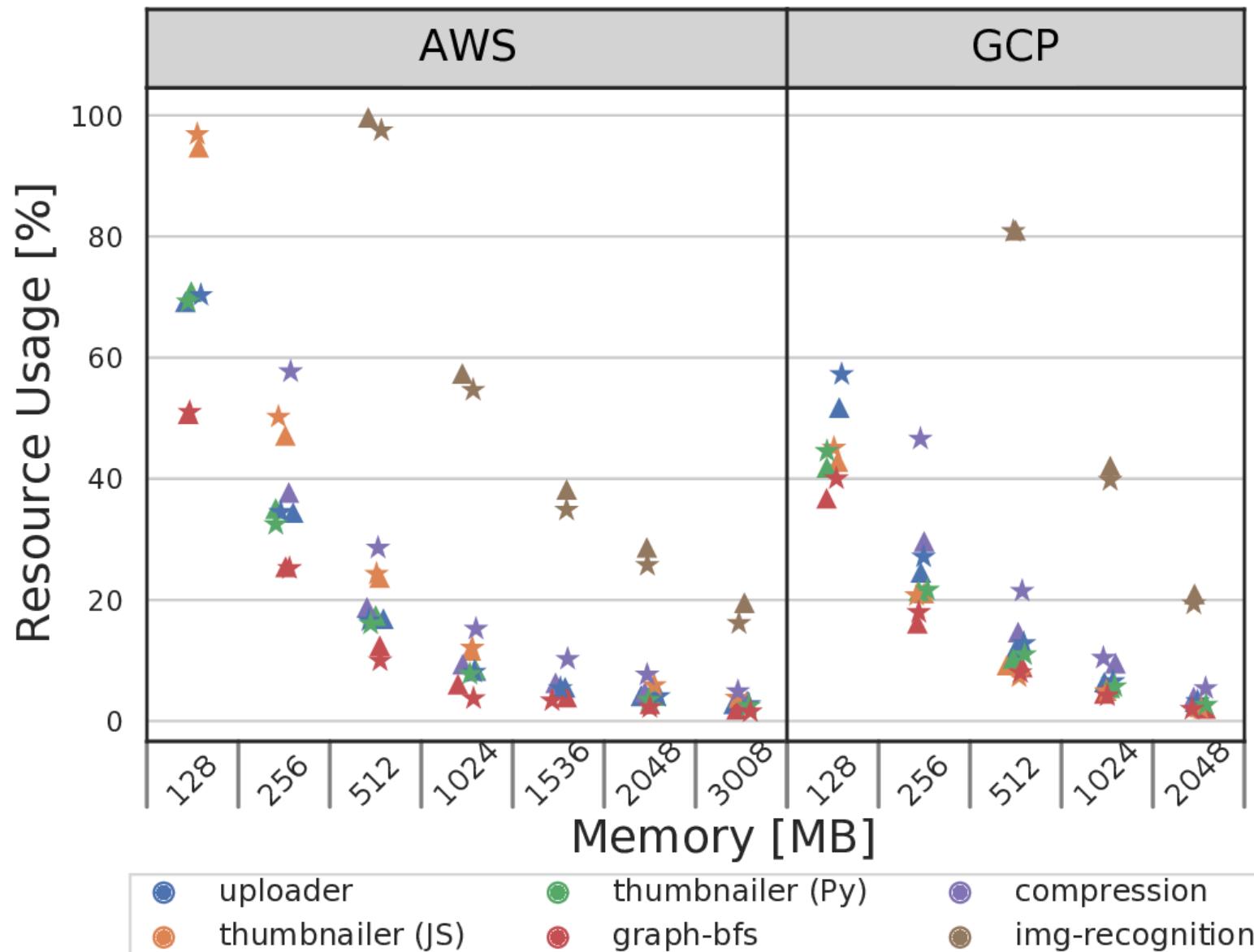
Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.
- **Measurements:** 200 warm executions.

Cost Analysis: Eviction Modeling



Cost Analysis: Eviction Modeling



Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth- First Search
IaaS							
FaaS							

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth- First Search
IaaS							
FaaS							

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.011 6/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Local Storage [req/h]						
FaaS	Cloud Storage [req/h]						
	Eco 1M [\$]						
	Eco Break-Even						
	Perf 1M [\$]						
	Perf Break-Even						

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.011 6/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Local Storage [req/h]	16627	79282	21697	4452	17658	119272
	Cloud Storage [req/h]	11371	27503	18819	1284	15312	117153
FaaS	Eco 1M [\$]	3.54	2.29	3.75	32.1	15.8	2.08
	Eco Break-Even	3275	5062	3093	362	733	117153
	Perf 1M [\$]	6.67	3.34	10	50	19.58	2.5
	Perf Break-Even	1740	3480	1160	232	592	4640

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.011 6/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Local Storage [req/h]	16627	79282	21697	4452	17658	119272
FaaS	Cloud Storage [req/h]	11371	27503	18819	1284	15312	117153
Eco 1M [\$]	3.54	2.29	3.75	32.1	15.8	2.08	
Eco Break-Even	3275	5062	3093	362	733	117153	
Perf 1M [\$]	6.67	3.34	10	50	19.58	2.5	
Perf Break-Even	1740	3480	1160	232	592	4640	

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.011 6/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

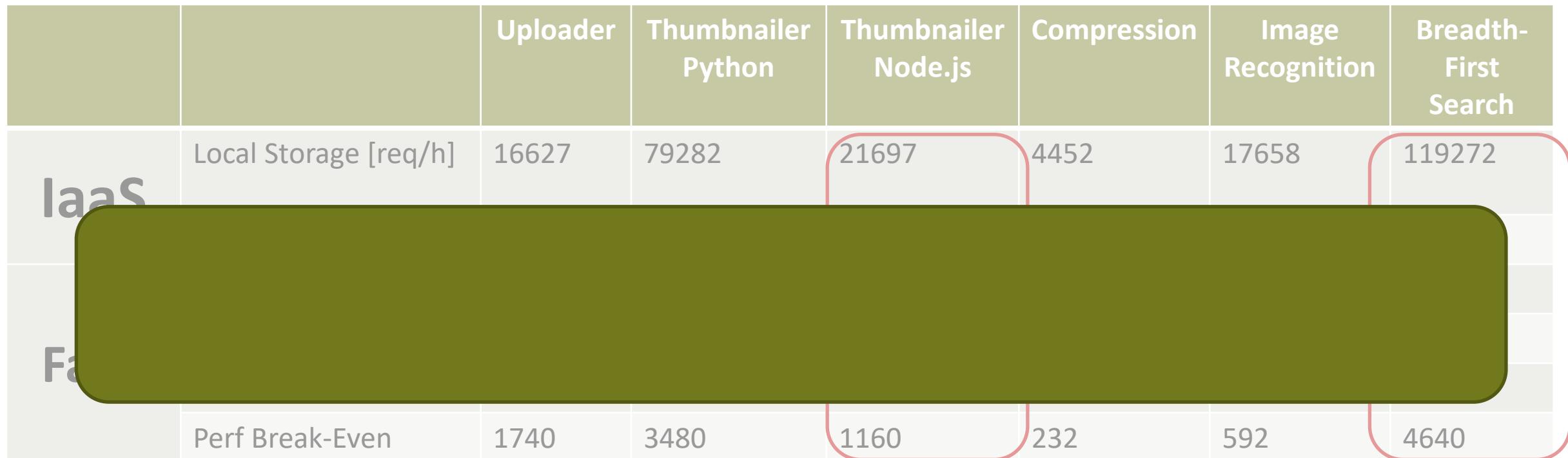
Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Local Storage [req/h]	16627	79282	21697	4452	17658	119272
	Cloud Storage [req/h]	11371	27503	18819	1284	15312	117153
FaaS	Eco 1M [\$]	3.54	2.29	3.75	32.1	15.8	2.08
	Eco Break-Even	3275	5062	3093	362	733	117153
	Perf 1M [\$]	6.67	3.34	10	50	19.58	2.5
	Perf Break-Even	1740	3480	1160	232	592	4640

Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.011 6/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS



Configuration:

- **IaaS:** AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.011 6/h.
- **FaaS:** AWS Lambda.
- **Local storage:** Minio as Docker container.

Cost Analysis: FaaS vs IaaS

		Uploader	Thumbnailer Python	Thumbnailer Node.js	Compression	Image Recognition	Breadth-First Search
IaaS	Local Storage [req/h]	16627	79282	21697	4452	17658	119272
FaaS	Perf Break-Even	1740	3480	1160	232	592	4640

#7: The IaaS solution delivers better performance at a lower price, but only in systems where high utilization is achieved.

Configuration:

- IaaS: AWS t2.micro instance with 1 vCPU and 1 GB RAM, 100% utilization, \$0.011 6/h.
- FaaS: AWS Lambda.
- Local storage: Minio as Docker container.