



Accelerating Serverless Computing by Harvesting Idle Resources

Hanfei Yu¹, Hao Wang¹, Jian Li², Xu Yuan³, Seung-Jong Park¹

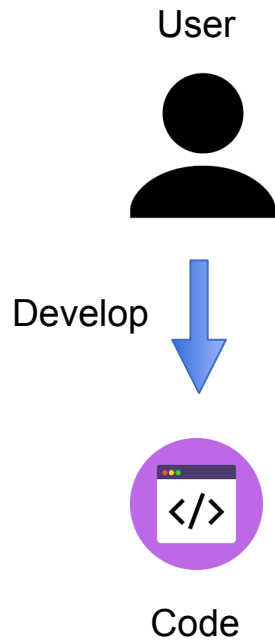
¹Louisiana State University, ²SUNY-Binghamton University, ³University of Louisiana at Lafayette



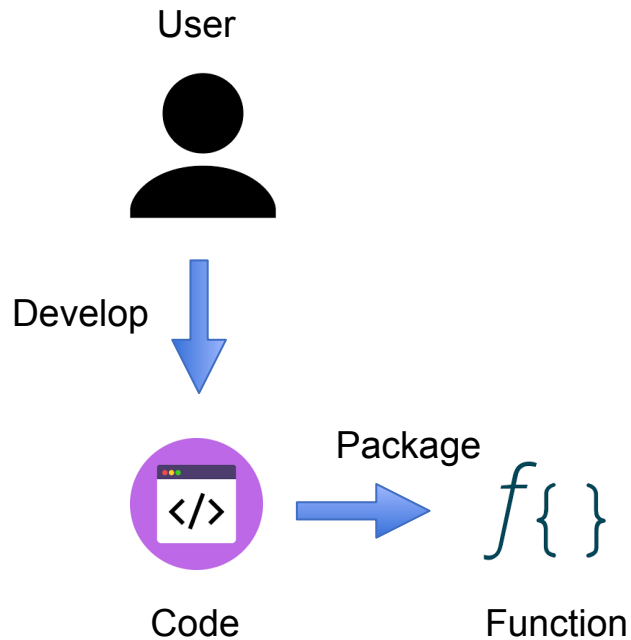
State University of New York



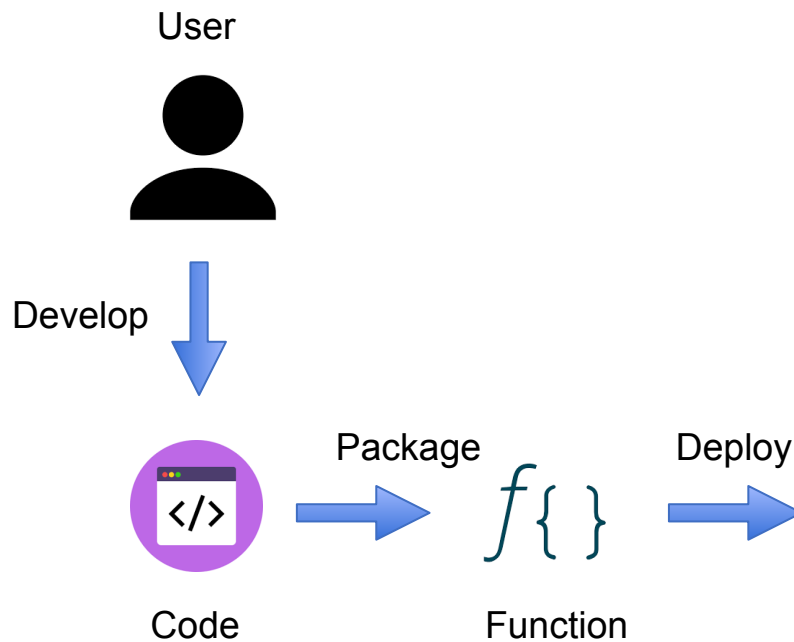
Serverless Computing



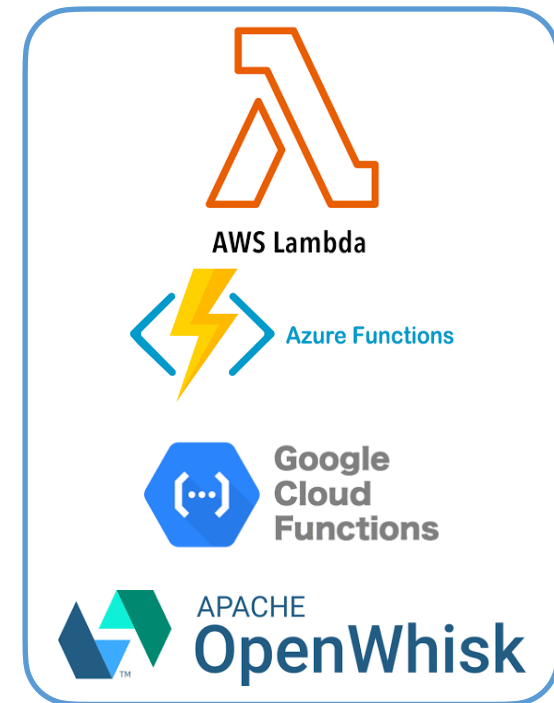
Serverless Computing



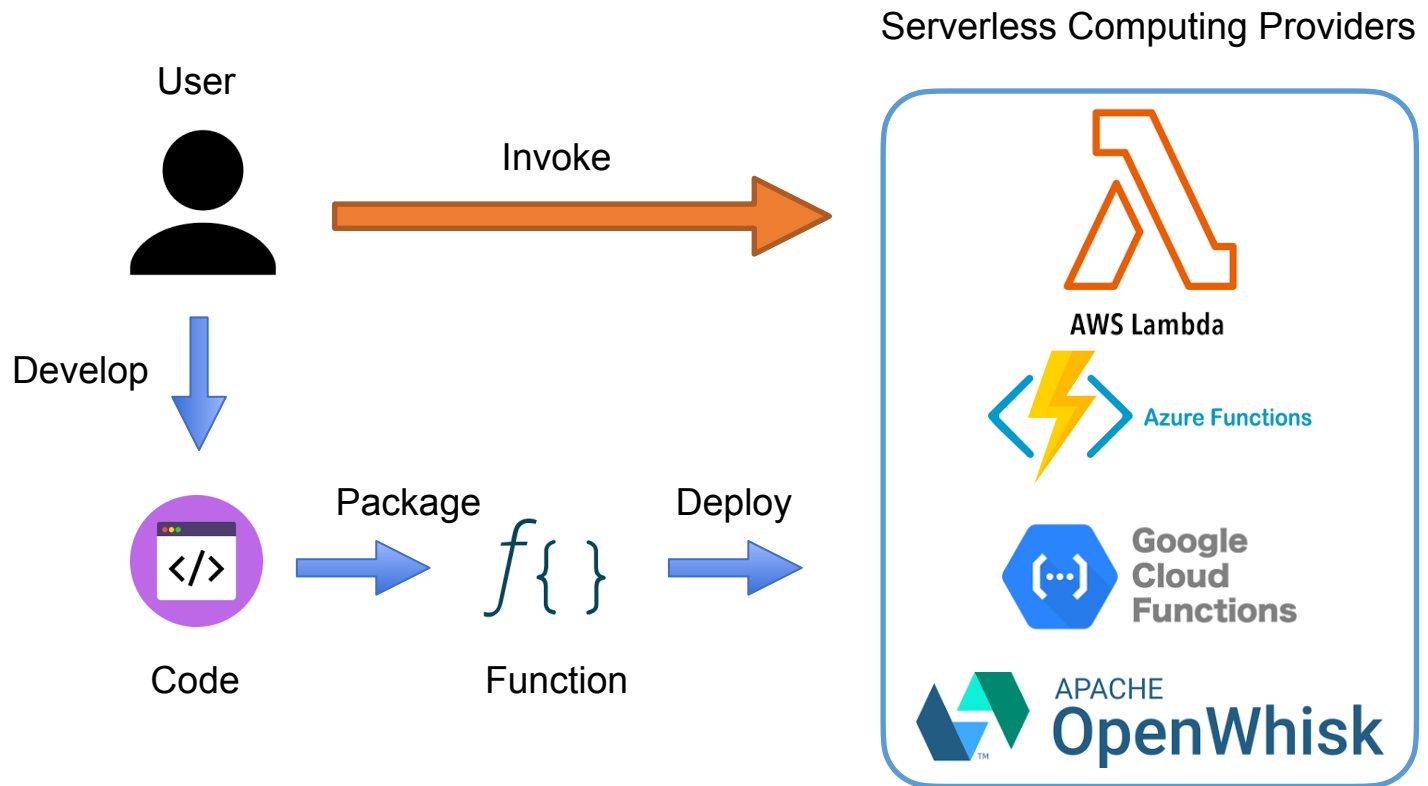
Serverless Computing



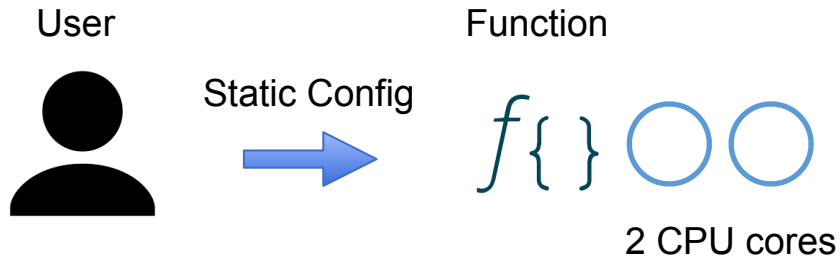
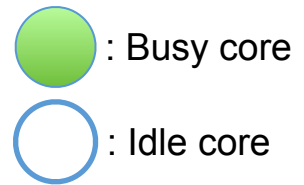
Serverless Computing Providers



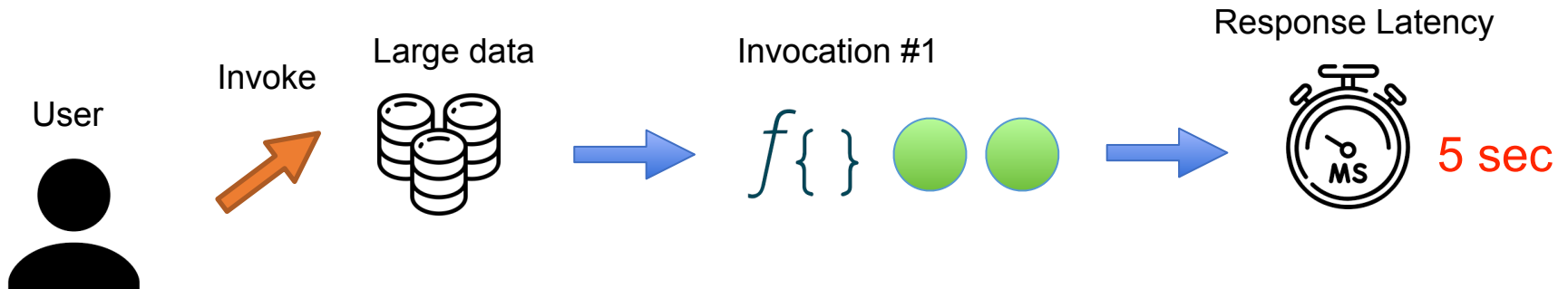
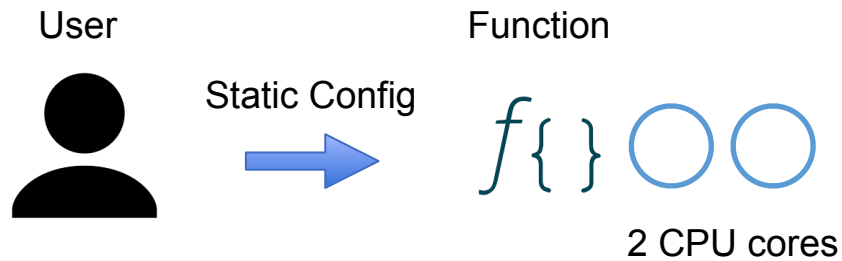
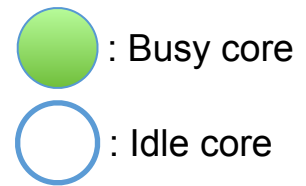
Serverless Computing



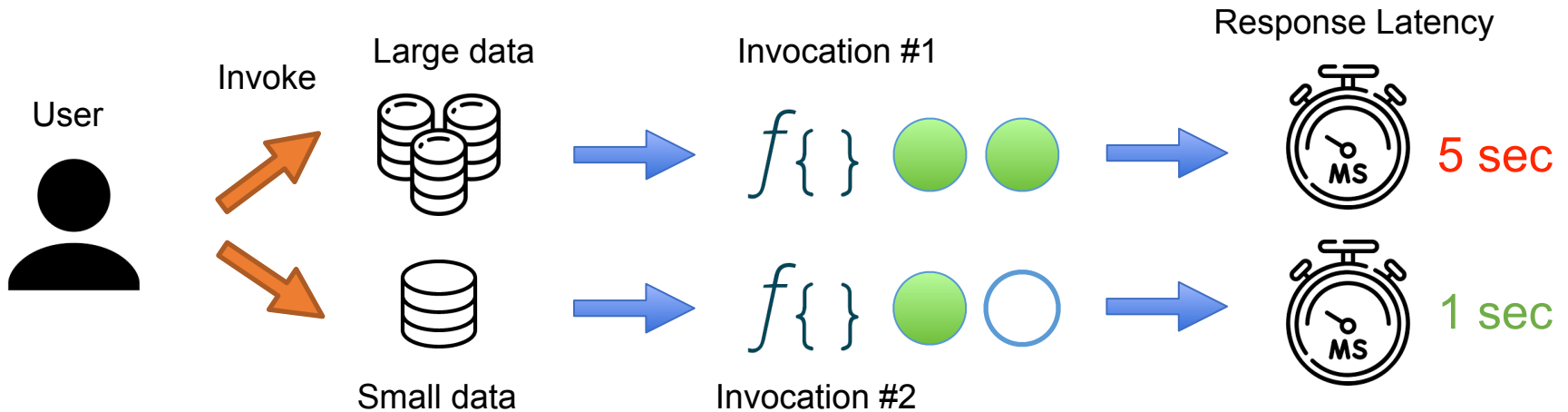
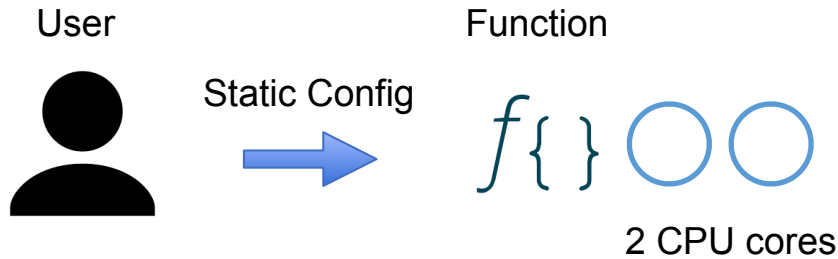
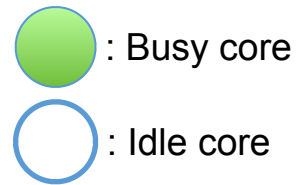
Function Static Configuration



Varying Input Data Size

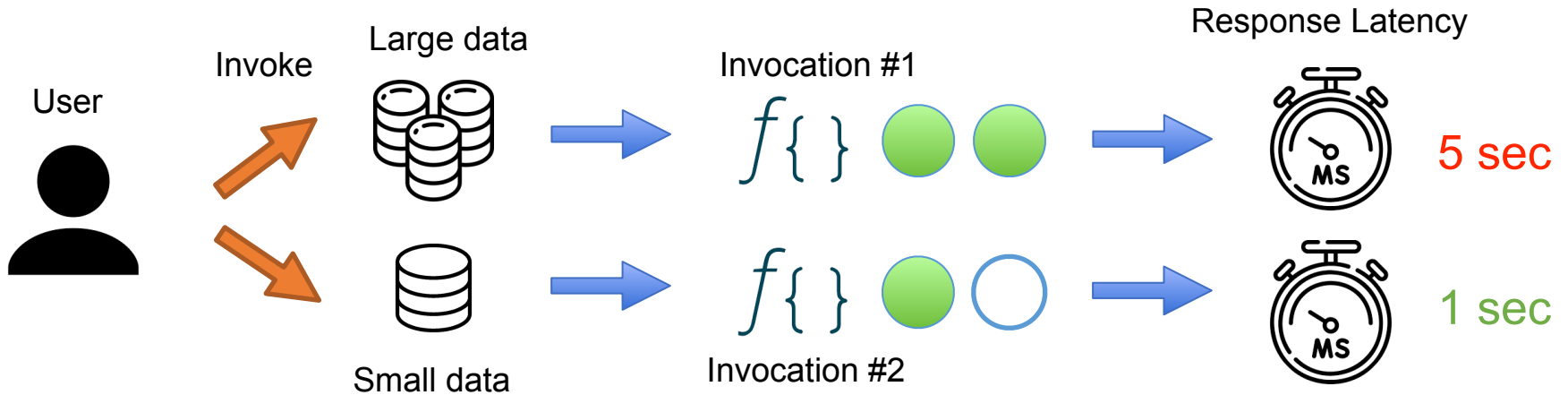


Varying Input Data Size



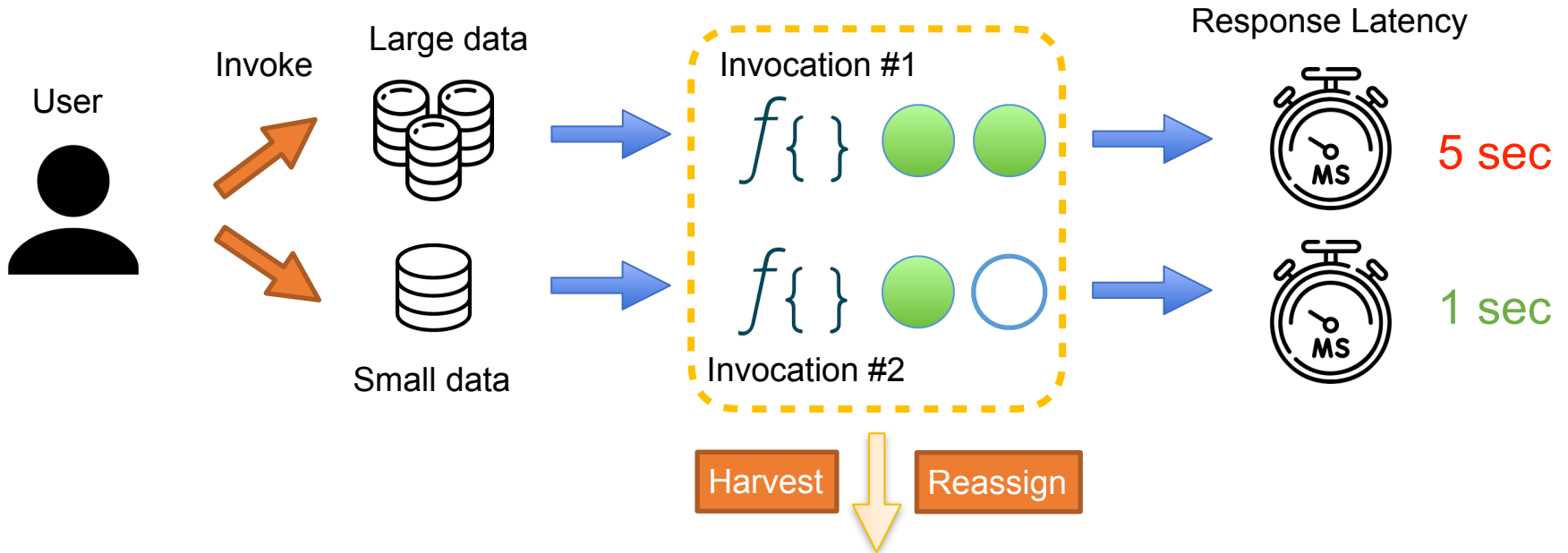
Harvesting & Acceleration

● : Busy core
○ : Idle core



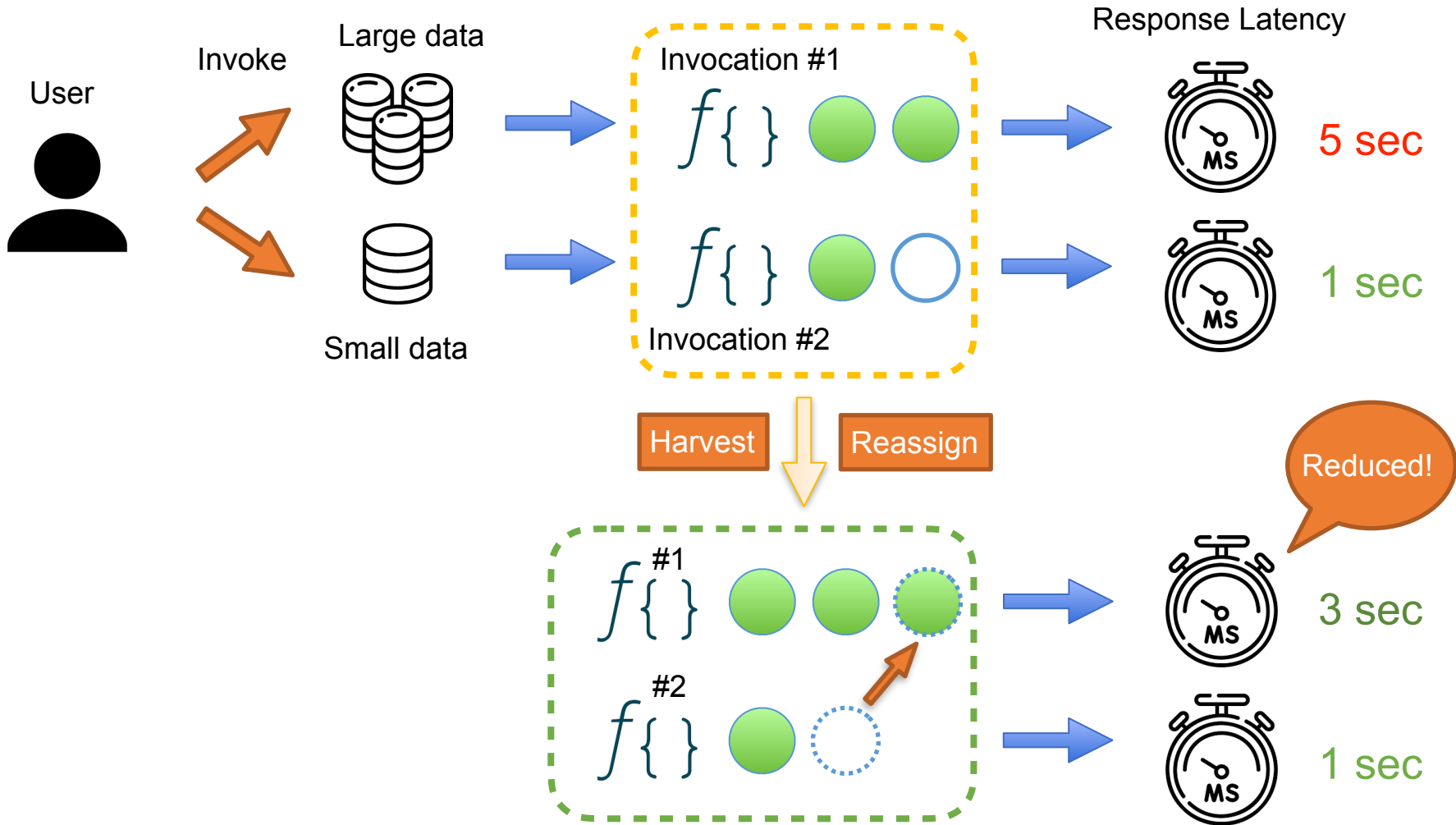
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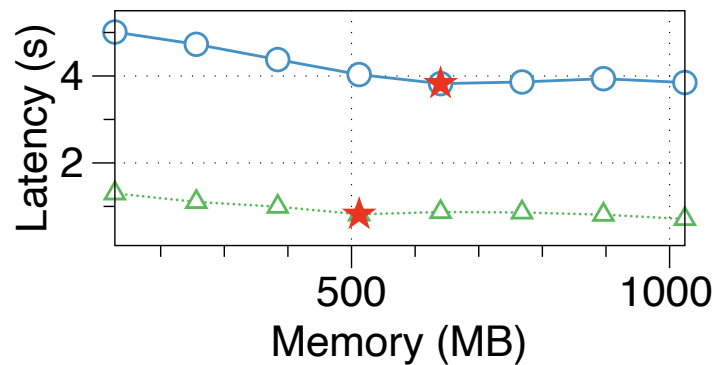
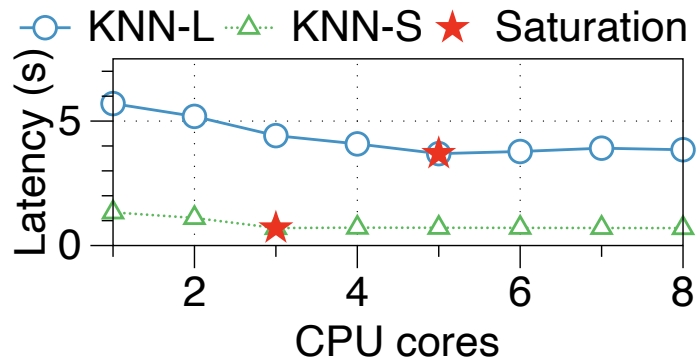
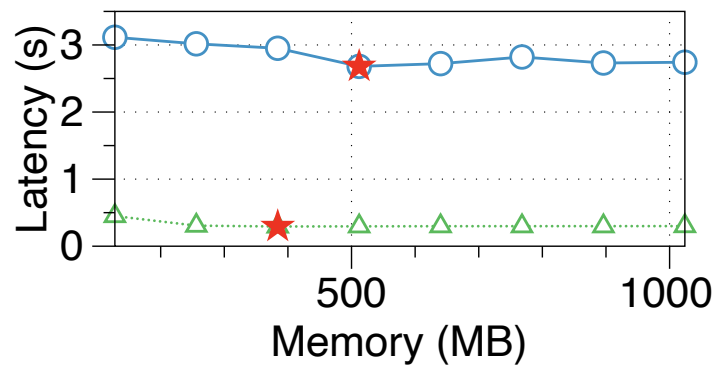
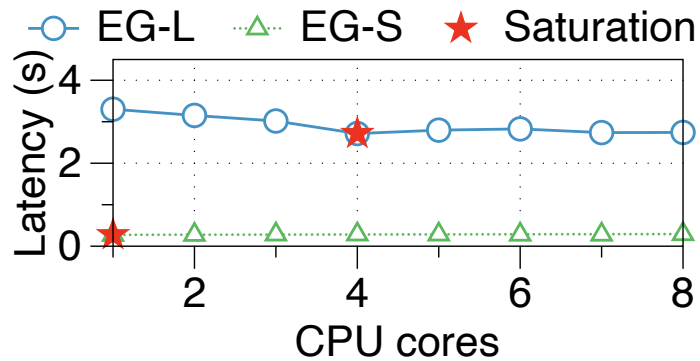


Harvesting & Acceleration

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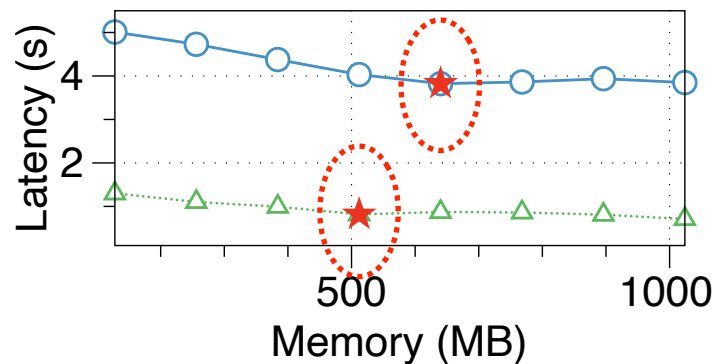
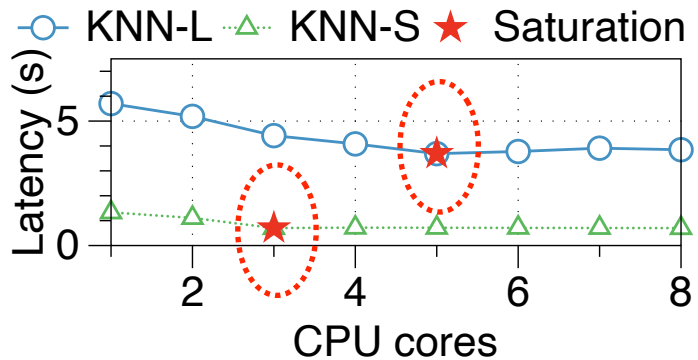
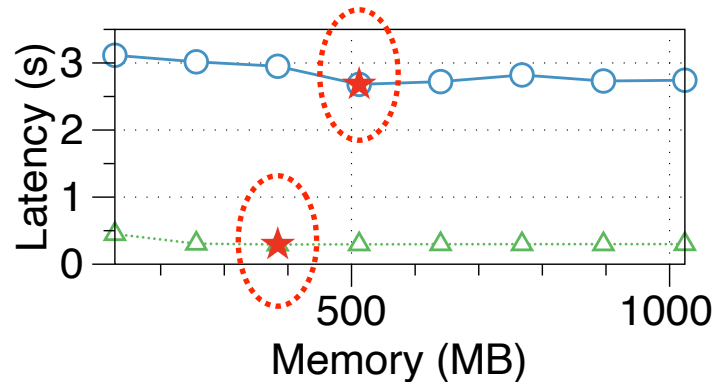
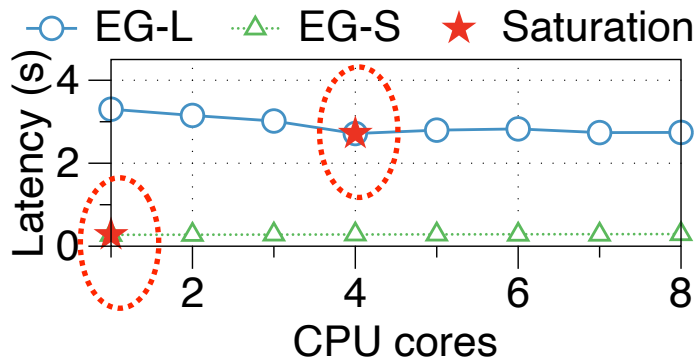
Realistic Applications



EG: email generation
KNN: K nearest neighbors

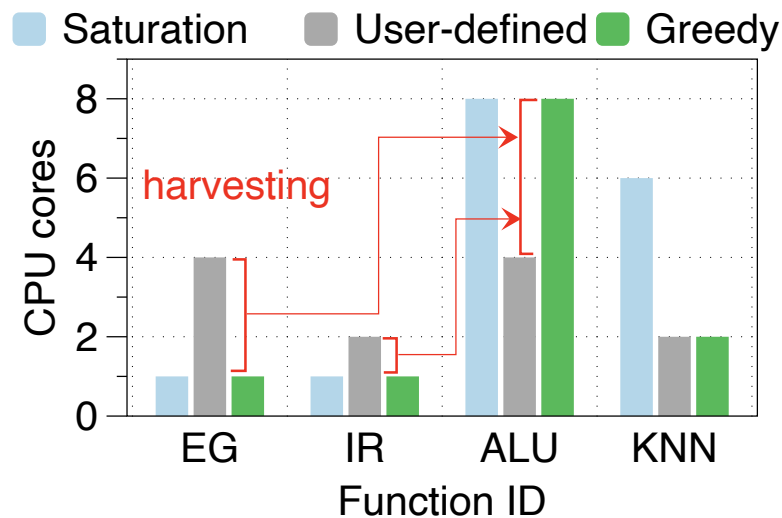
Realistic Applications

Performance stops growing when supplying more resources!

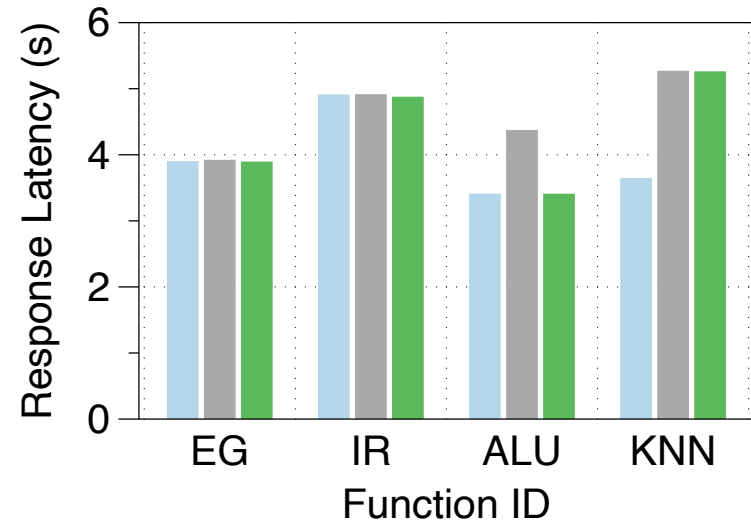


EG: email generation
KNN: K nearest neighbors

Realistic Harvesting & Acceleration



(a) CPU allocation

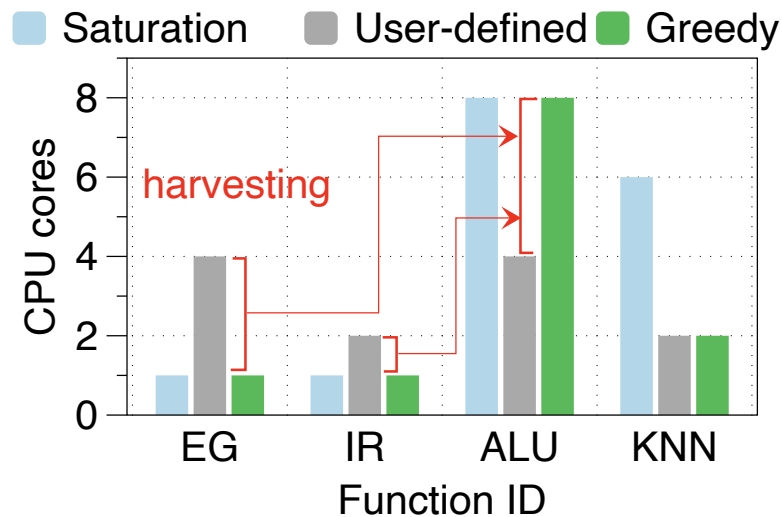


(b) Function response latency

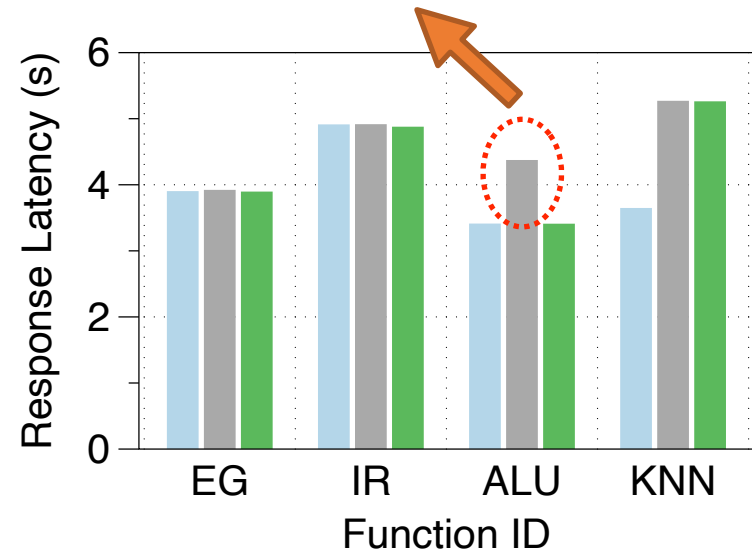
EG: email generation
IR: image recognition
ALU: arithmetic logic units
KNN: K nearest neighbors

Realistic Harvesting & Acceleration

Latency can be reduced with supplying harvested resources!



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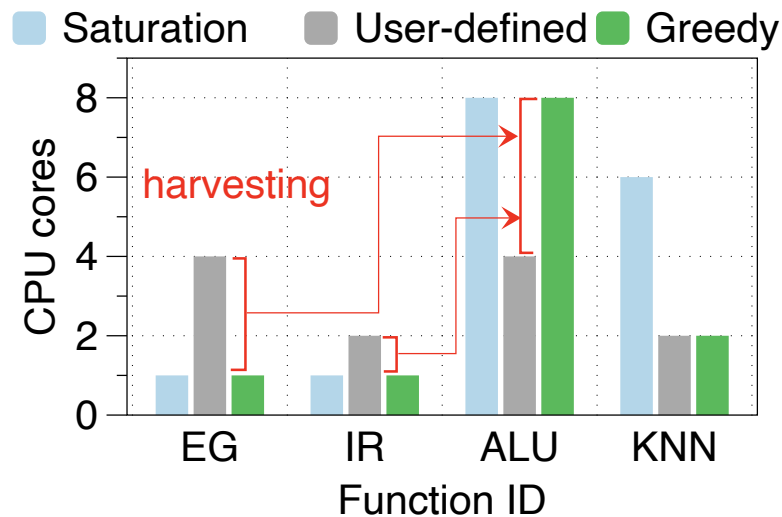


(b) Function response latency

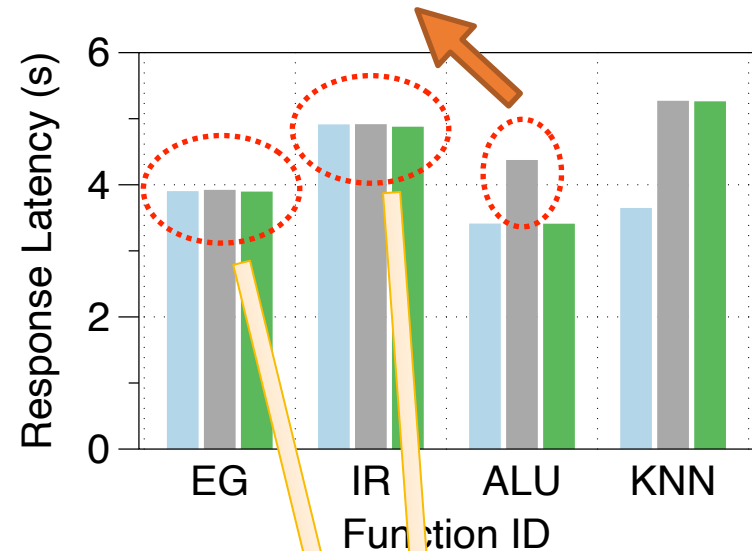
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Realistic Harvesting & Acceleration

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(a) CPU allocation



(b) Function response latency

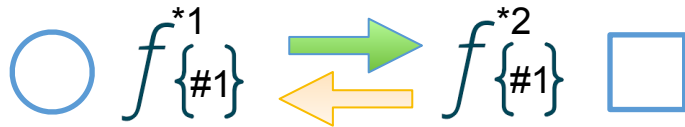
EG: email generation
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Careful harvesting does not degrade performance

General Rebalance Cases

- Idle core
- Idle memory
- # Function index
- * Invocation index

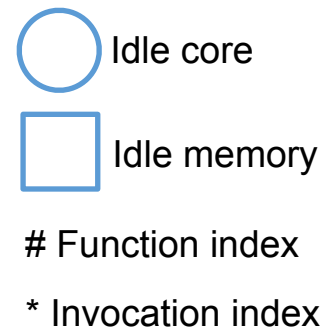
Case 1



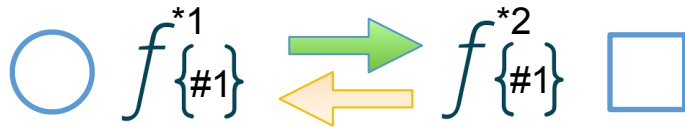
Within the same function
Both donator and receiver

Serverless Cluster

General Rebalance Cases

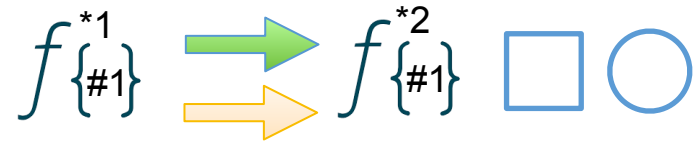


Case 1



Within the same function
Both donator and receiver

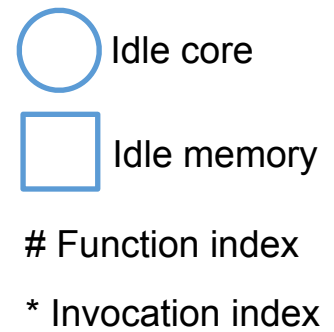
Case 2



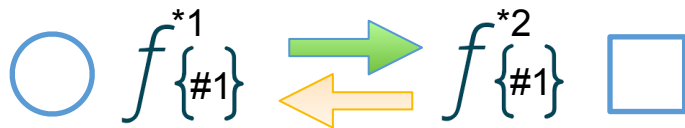
Within the same function
One donator and one receiver

Serverless Cluster

General Rebalance Cases

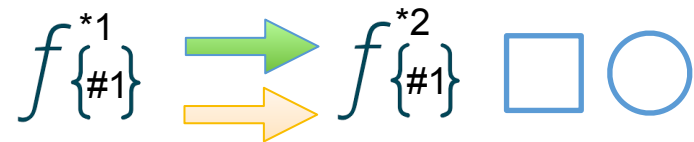


Case 1



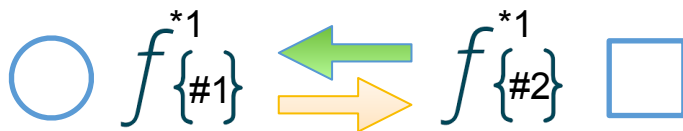
Within the same function
Both donator and receiver

Case 2



Within the same function
One donator and one receiver

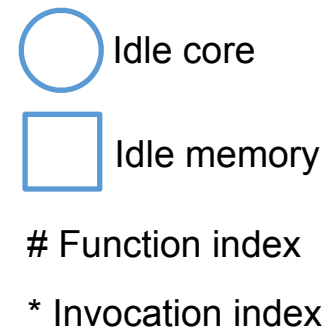
Case 3



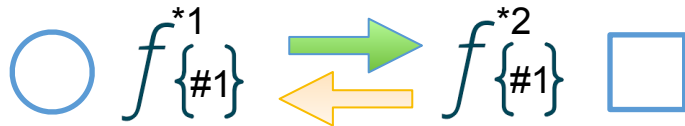
Between two functions
Both donator and receiver

Serverless Cluster

General Rebalance Cases

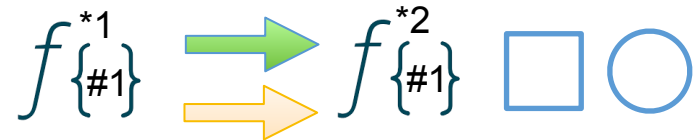


Case 1



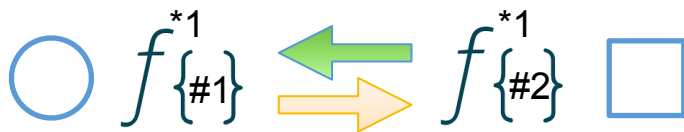
Within the same function
Both donator and receiver

Case 2



Within the same function
One donator and one receiver

Case 3



Between two functions
Both donator and receiver

Case 4



Between two functions
One donator and one receiver

Serverless Cluster

Dynamic Decisions

Perspective of a serverless platform:

- Varying **functions**

Dynamic Decisions

Perspective of a serverless platform:

- Varying **functions**
- Varying **invocations** per functions

Dynamic Decisions

Perspective of a serverless platform:

- Varying **functions**
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Dynamic Decisions

Perspective of a serverless platform:

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- Varying **invocations** per functions
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- Every invocation requires an **allocation decision**

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A series of sequential allocation decisions

Dynamic Decisions

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A series of sequential allocation decisions



Markov Decision Process (MDP)

Deep Reinforcement Learning

Perspective of a serverless platform:

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A series of sequential allocation decisions

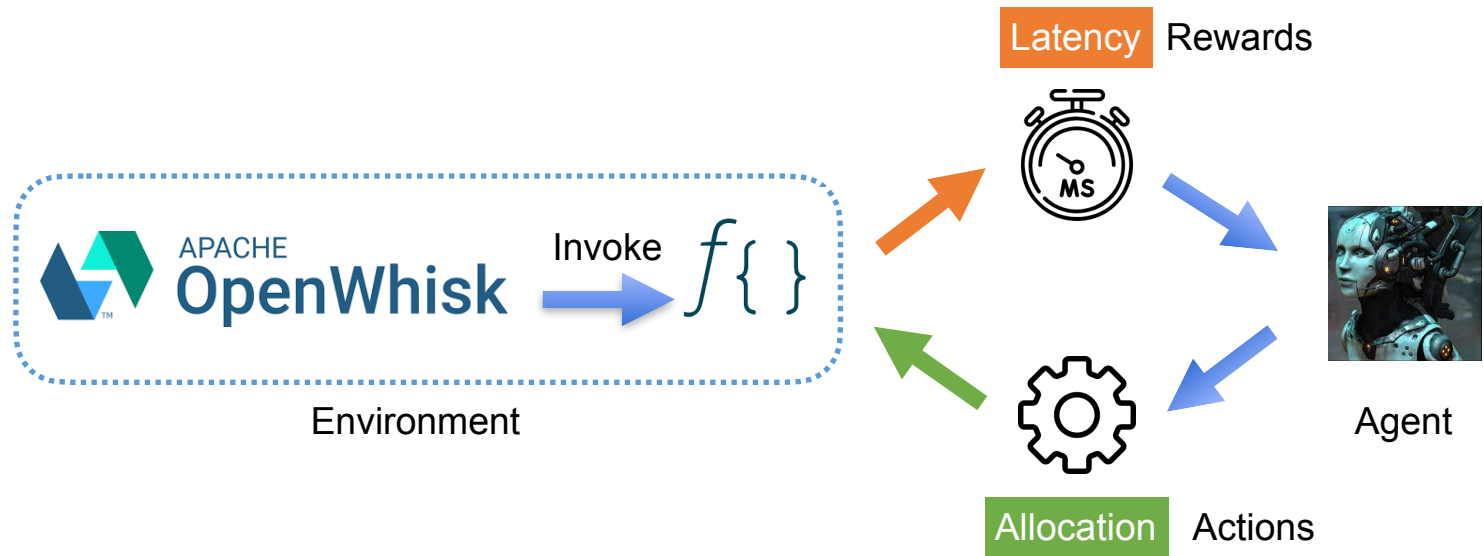


Markov Decision Process (MDP)



Deep Reinforcement Learning

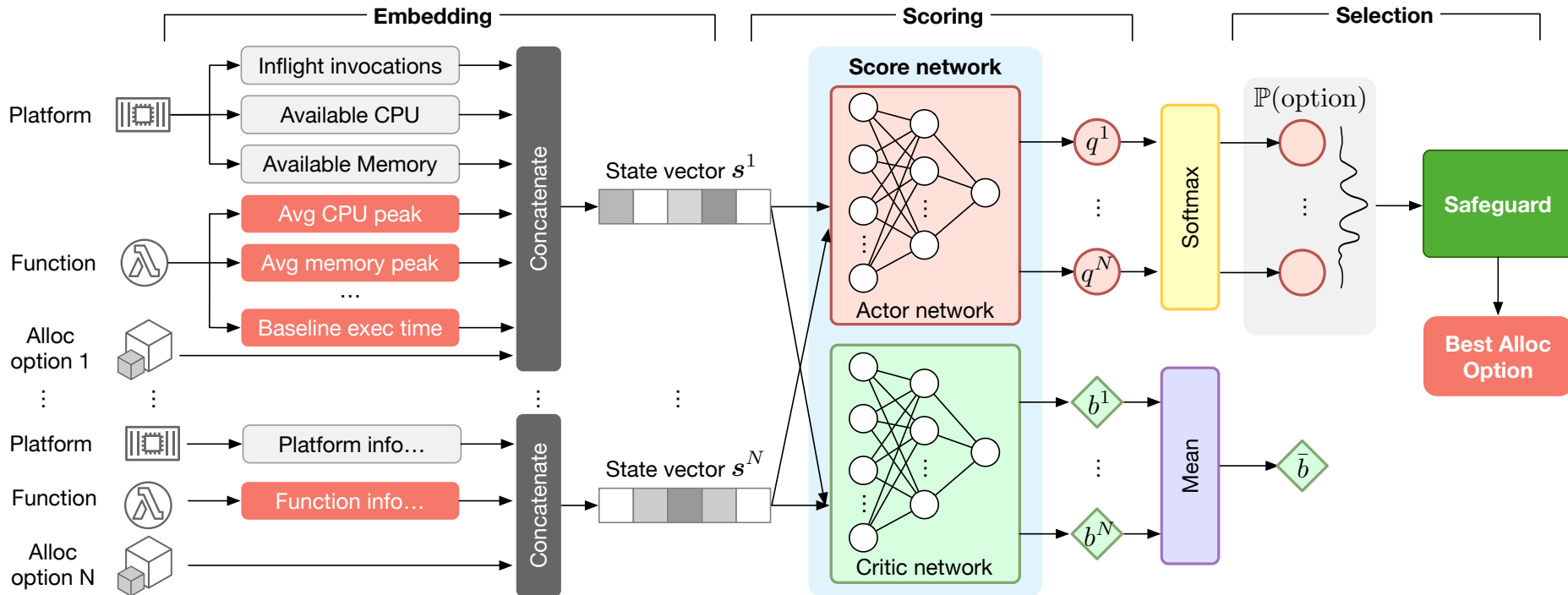
Deep Reinforcement Learning



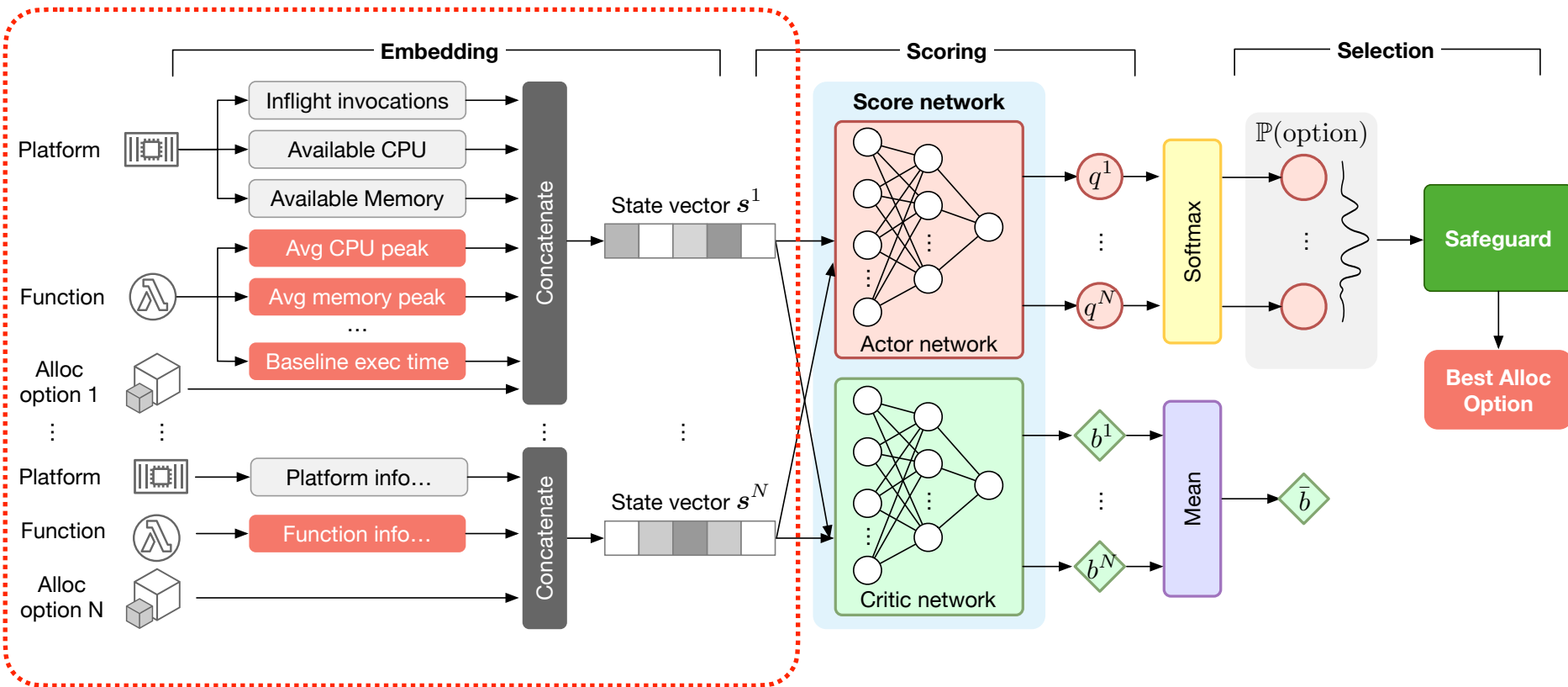
Freyr



Freyr Workflow

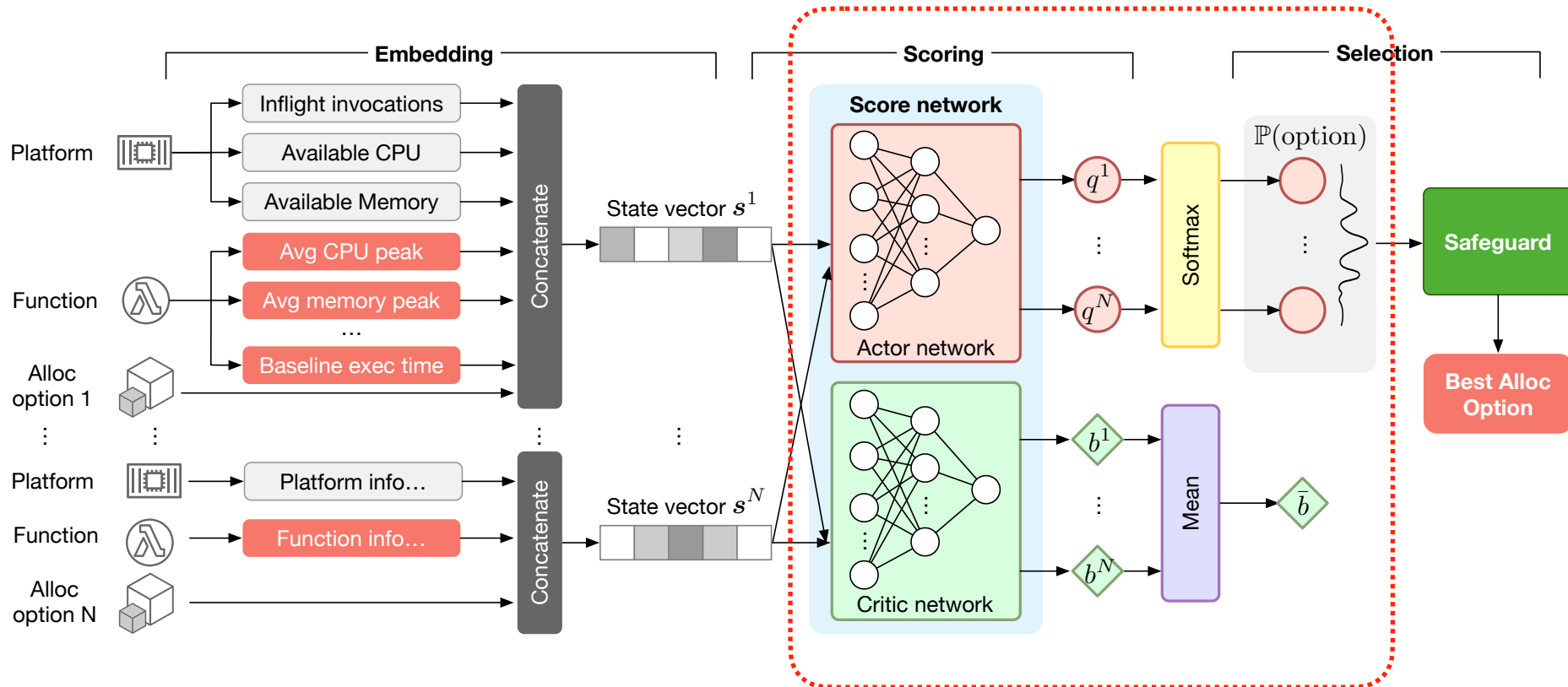


Freyr Workflow



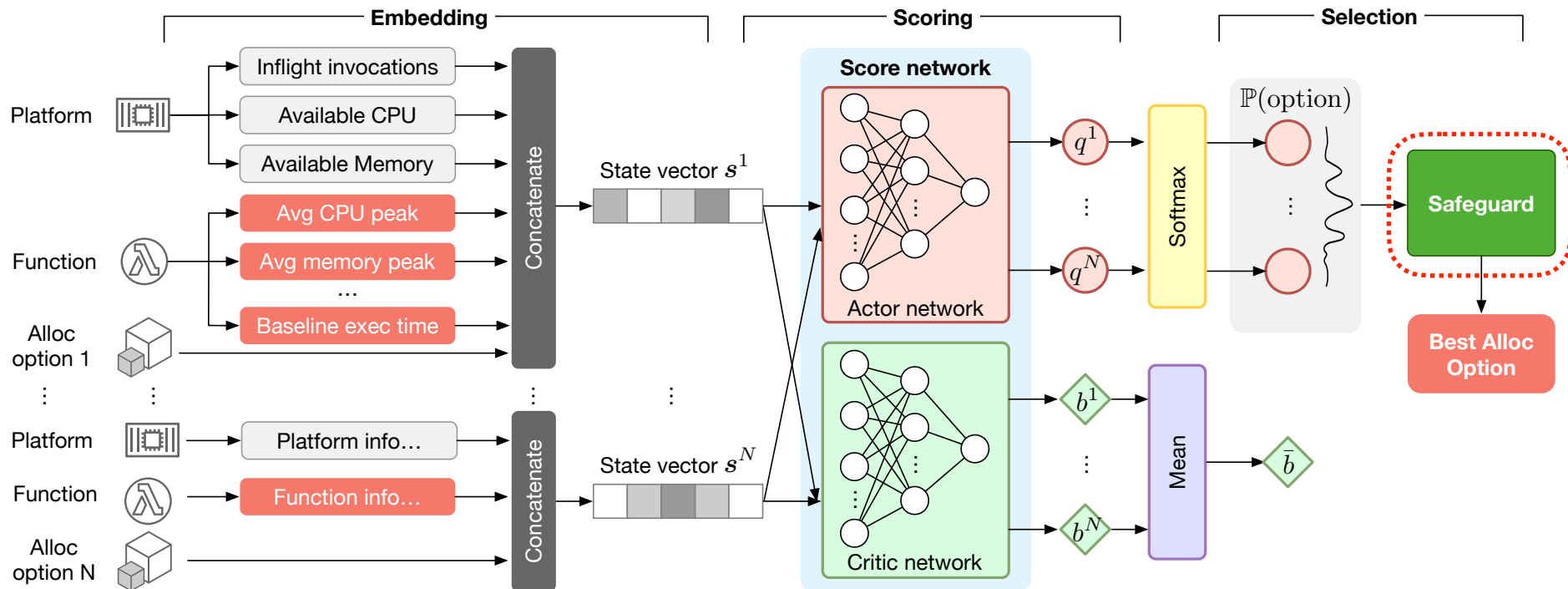
State information from the platform and the function

Freyr Workflow



Proximal Policy Optimization (PPO)

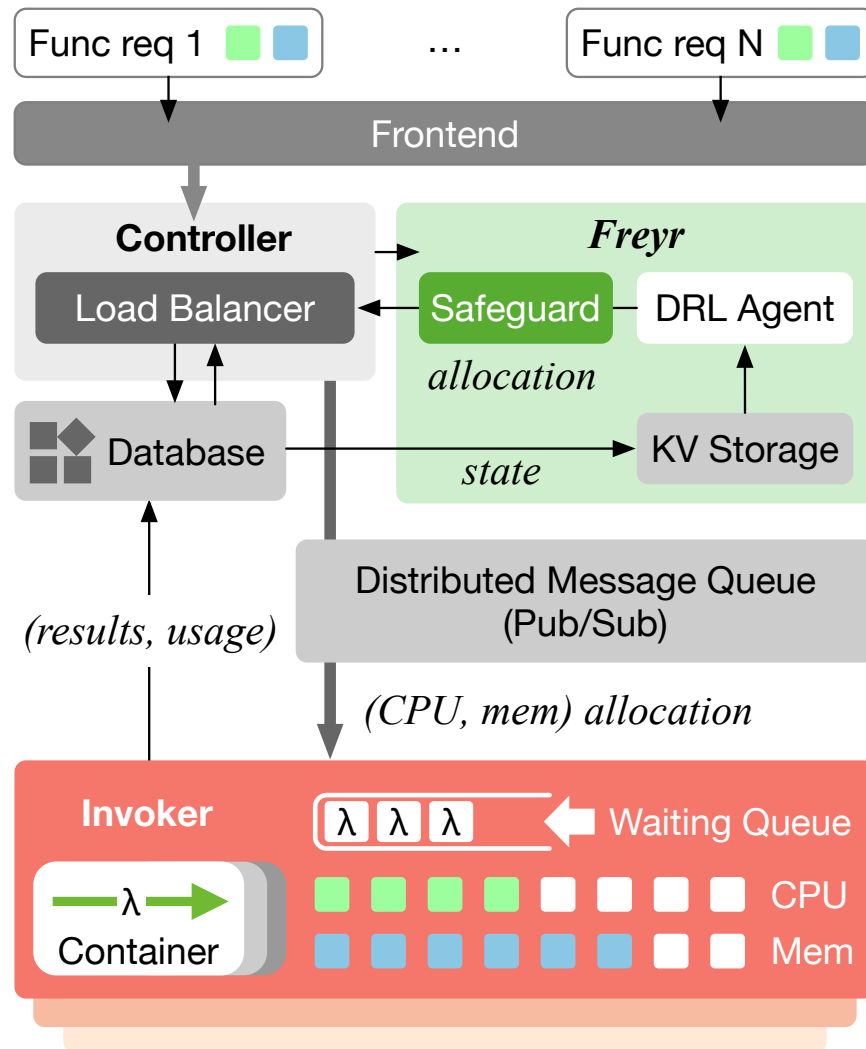
Freyr Workflow



Safeguard

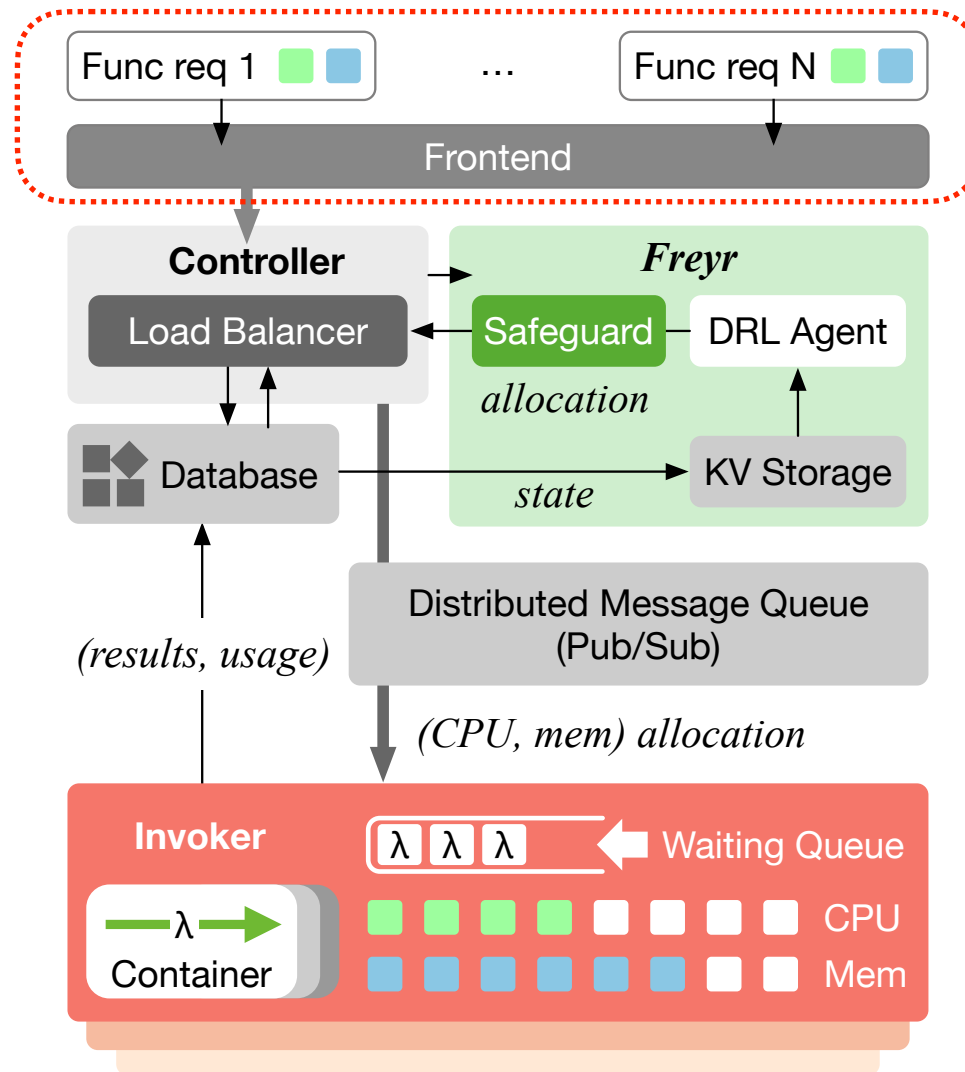
- Filter invalid allocation options
- Return resources when detecting a potential full usage

Freyr Architecture



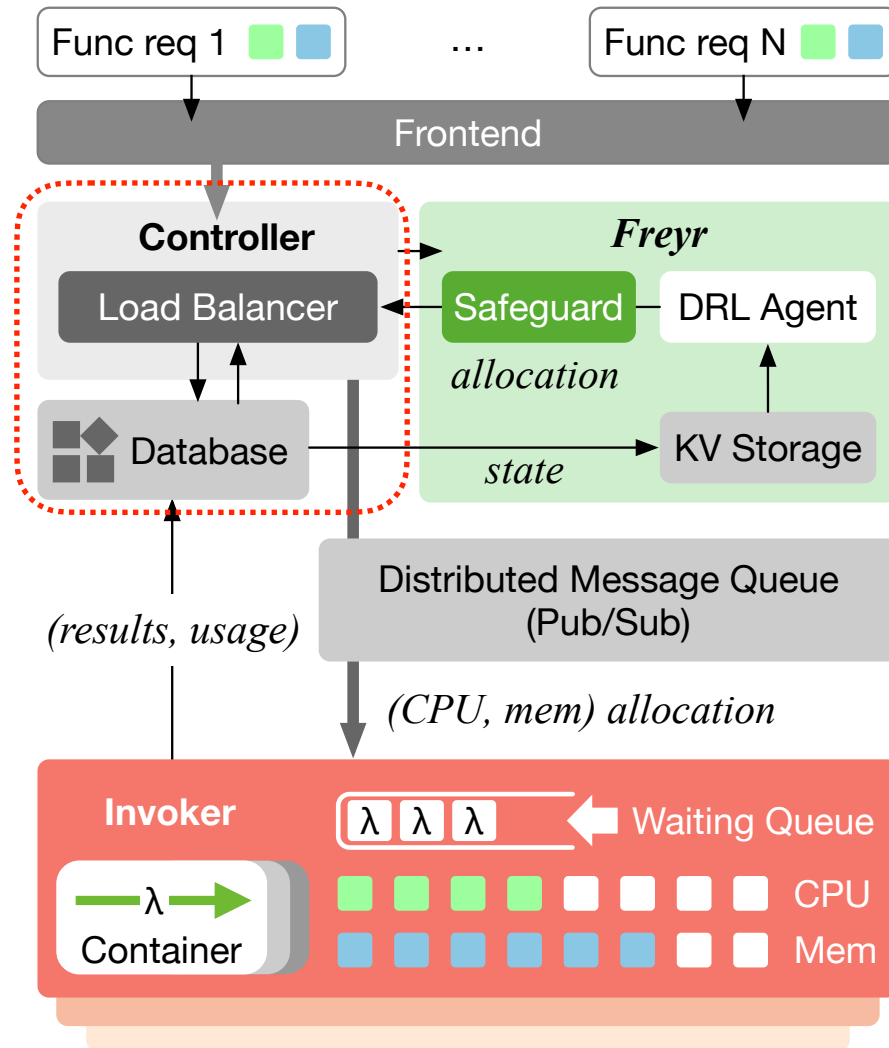
Freyr Architecture

Frontend receives function invocations from users

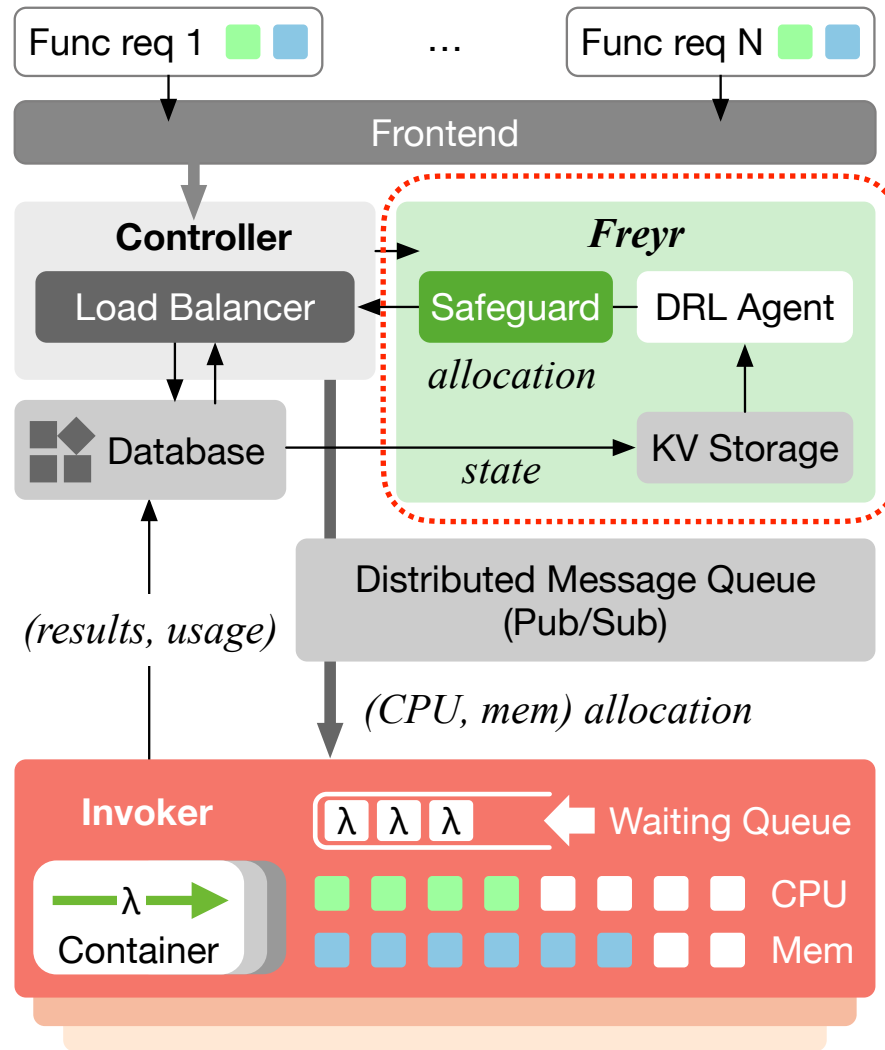


Freyr Architecture

Controller collects and sends states to the DRL agent

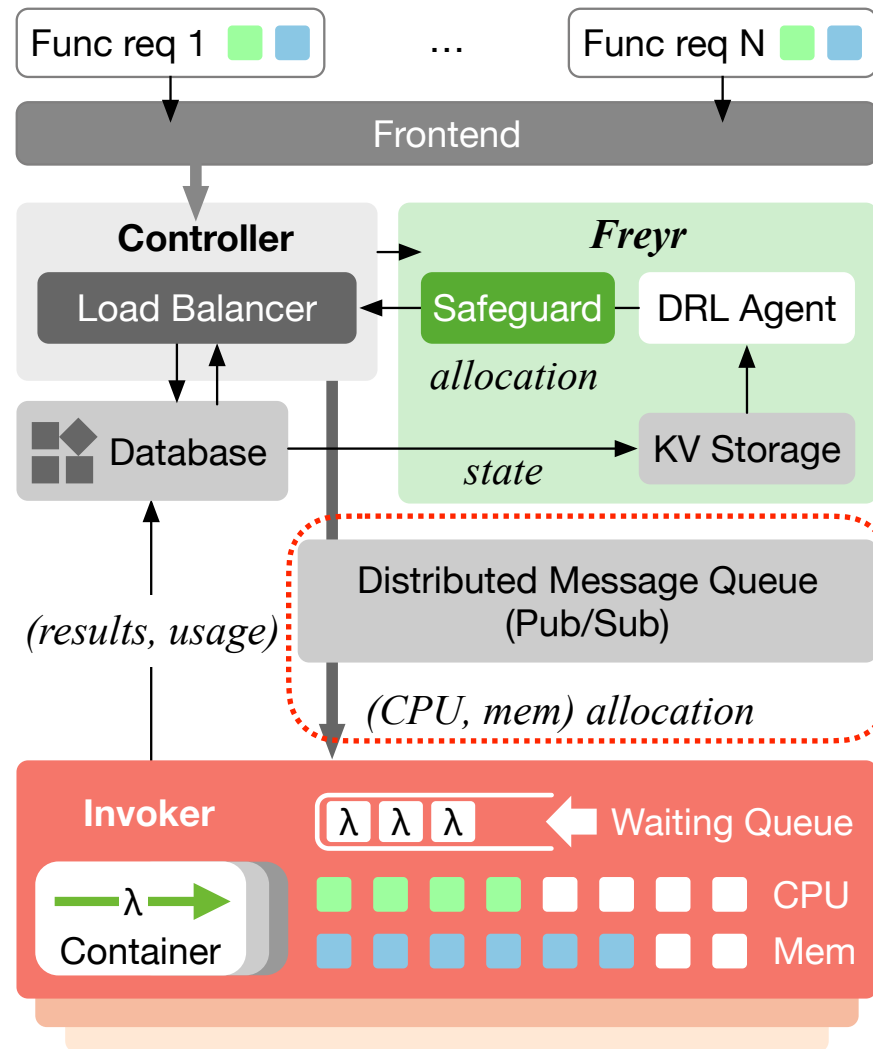


Freyr Architecture



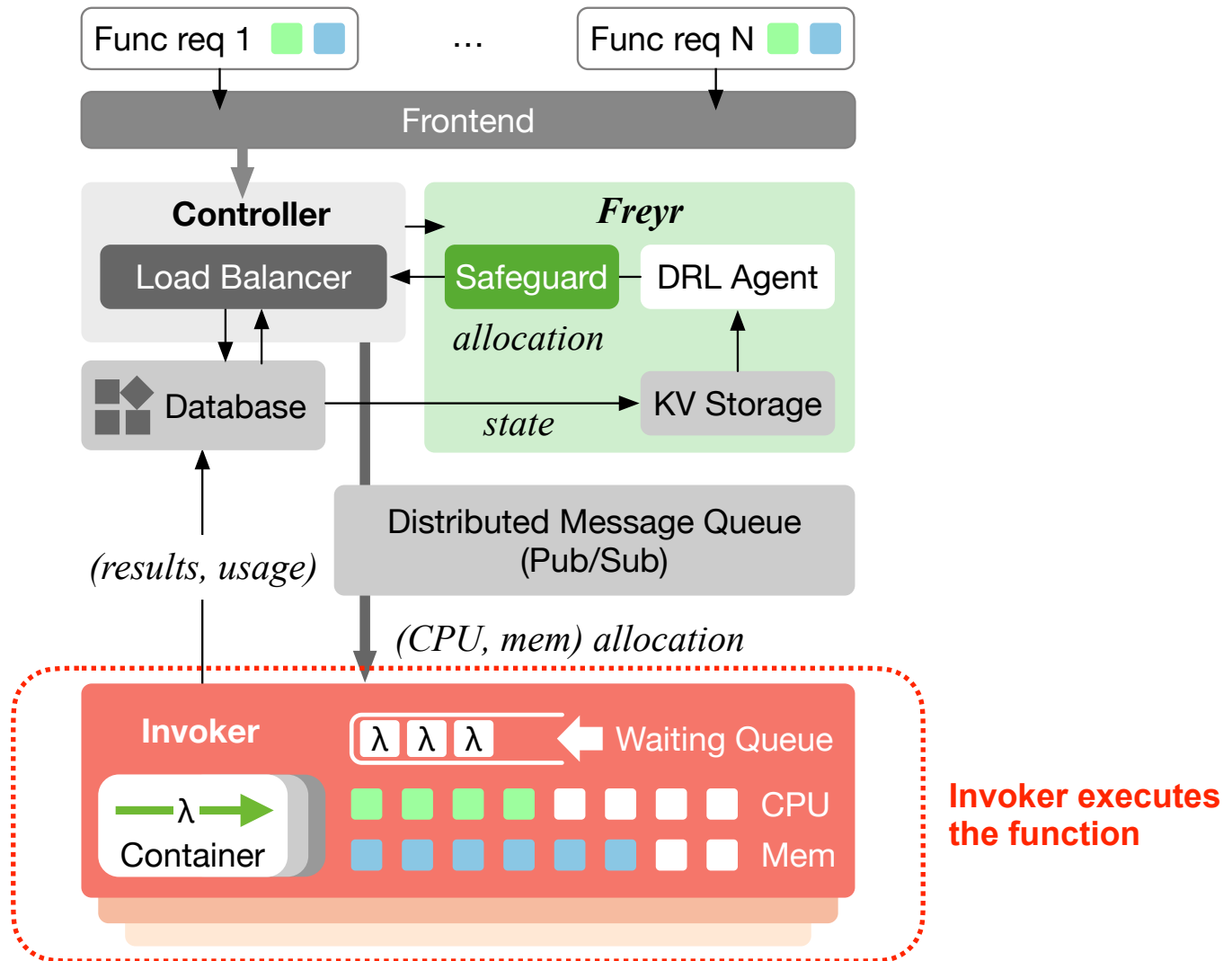
Agent predicts an allocation and sends it back to controller

Freyr Architecture

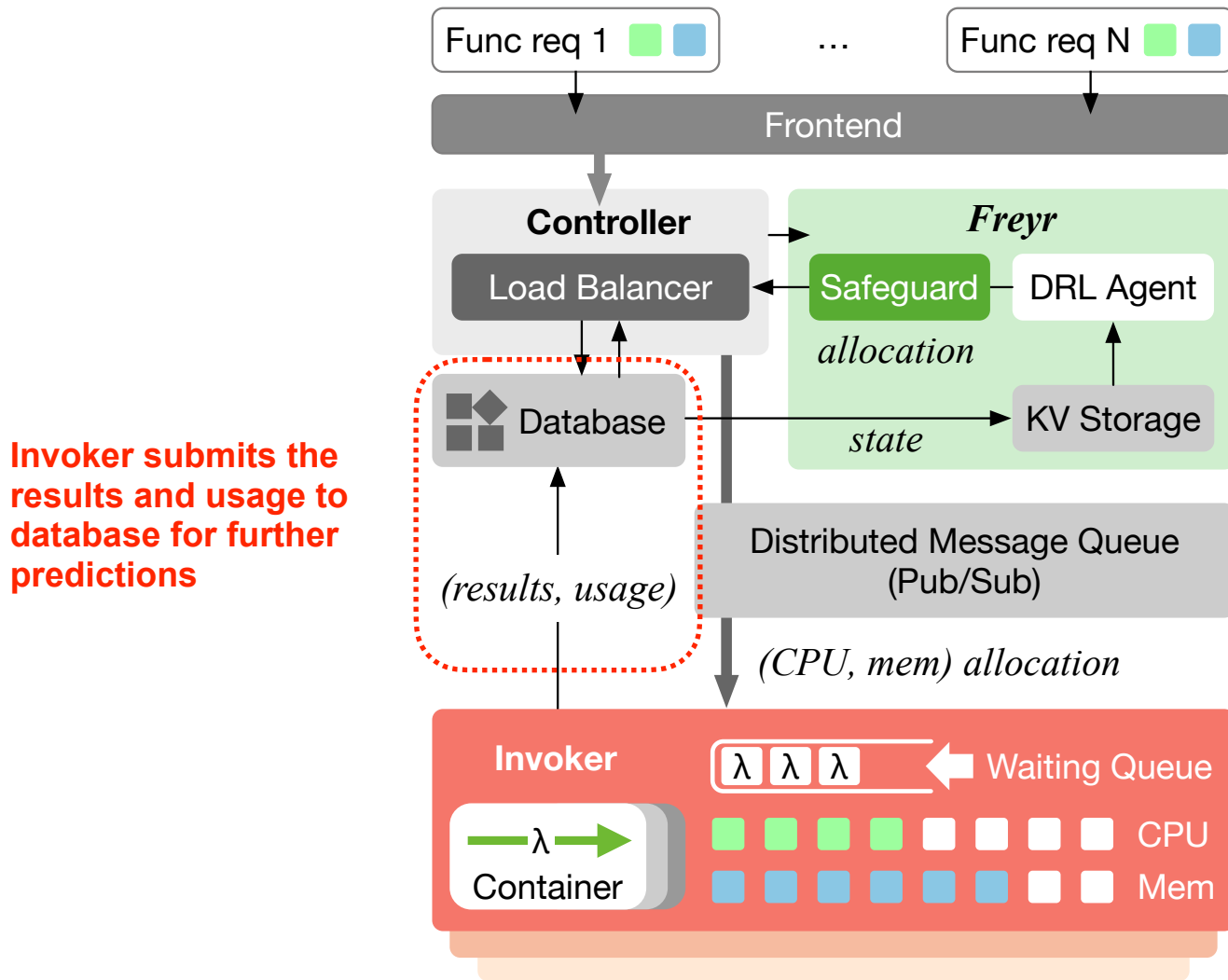


Controller then forwards the function invocation with its decision to an Invoker

Freyr Architecture



Freyr Architecture



Experiment

Setup

- 13 VMs, each with 8 CPUs and 32 GB memory
- One user client, one frontend, one controller
- 10 Worker nodes

Baselines

- Fixed RM
- Greedy RM
- ENSURE

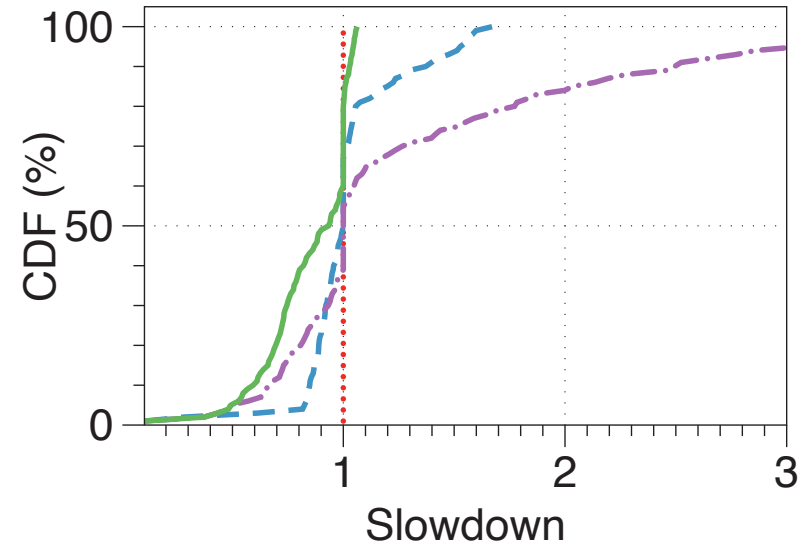
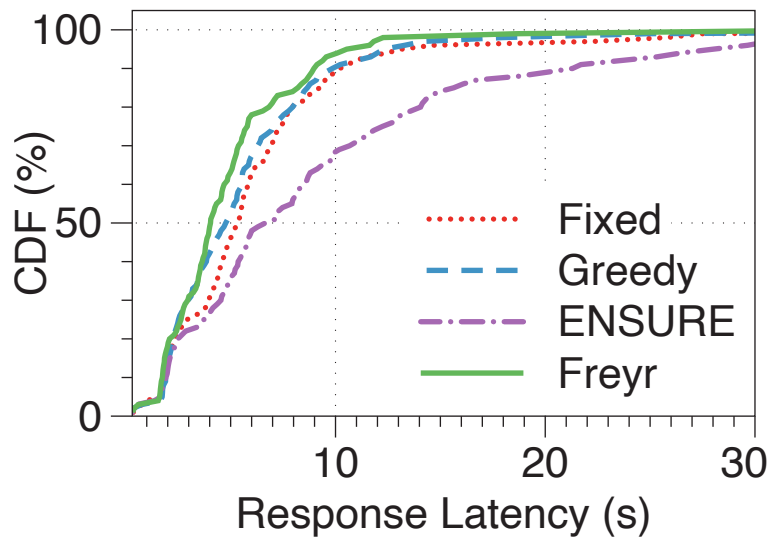
Fixed RM: default OpenWhisk as well as in existing serverless platforms

Greedy RM: heuristic

ENSURE: Suresh, Amoghavarsha, et al.

"Ensure: Efficient scheduling and autonomous resource management in serverless environments."
(ACSOS 2020)

Function Execution Speedup

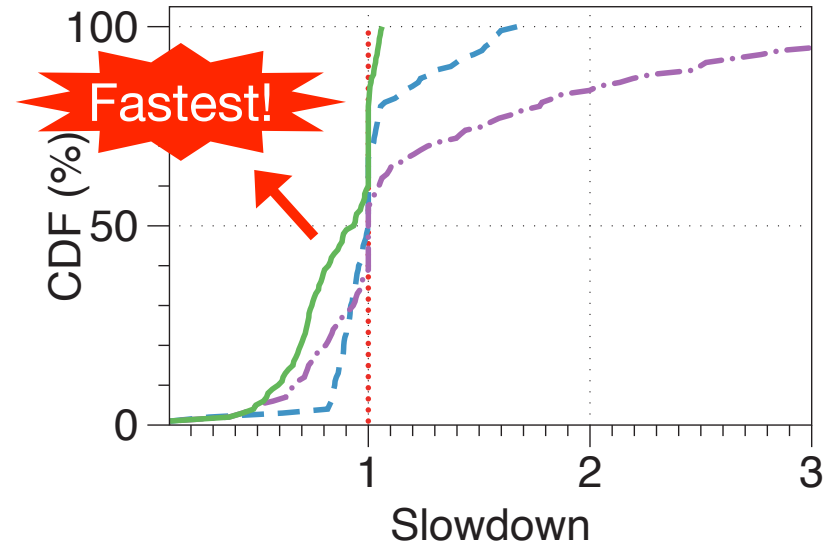
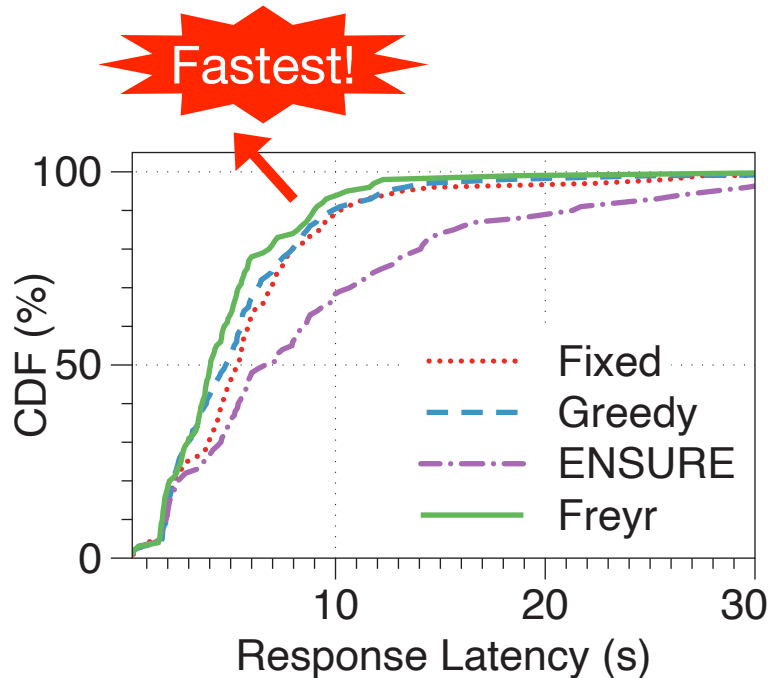


Response latency: function invocation end-to-end latency

Slowdown: relative performance compared to user-defined resources.

Larger than 1.0 means degradation, less than 1.0 means speedup

Function Execution Speedup

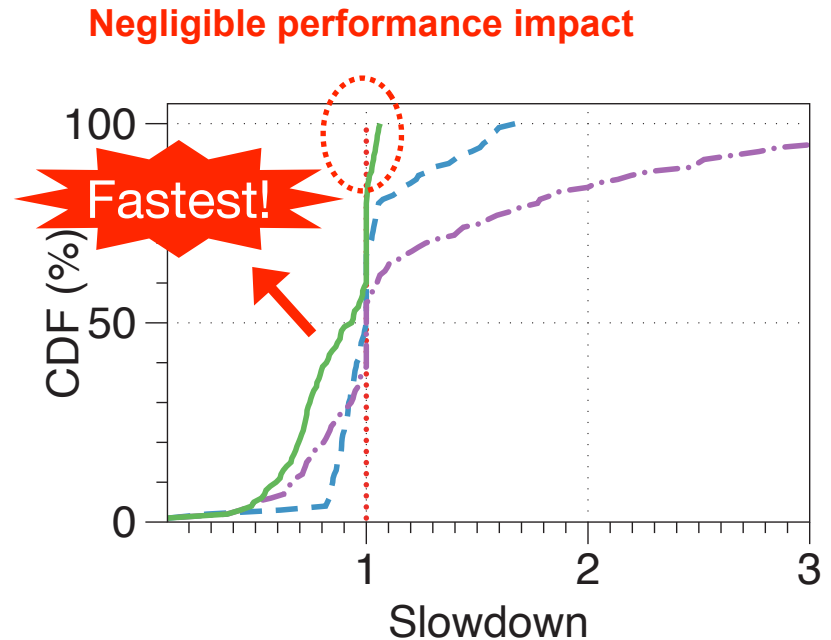
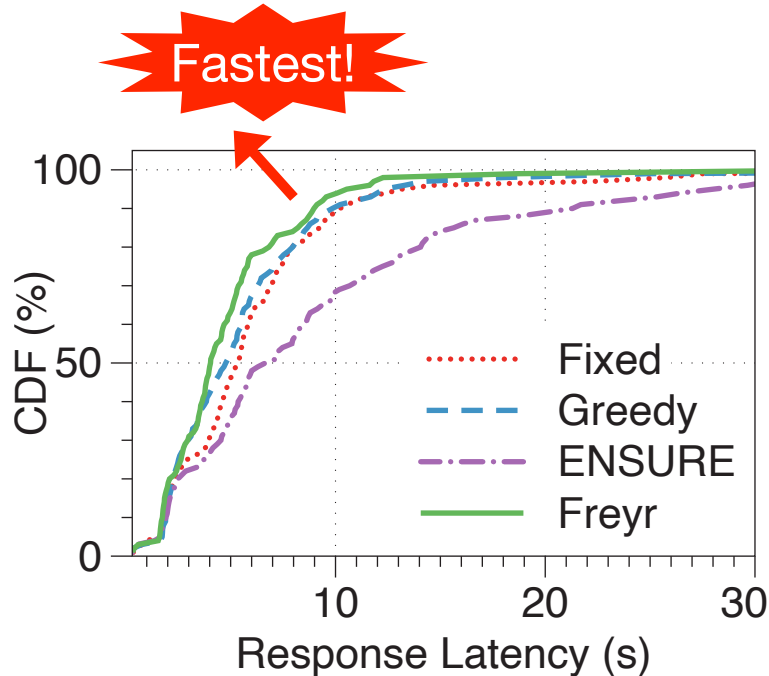


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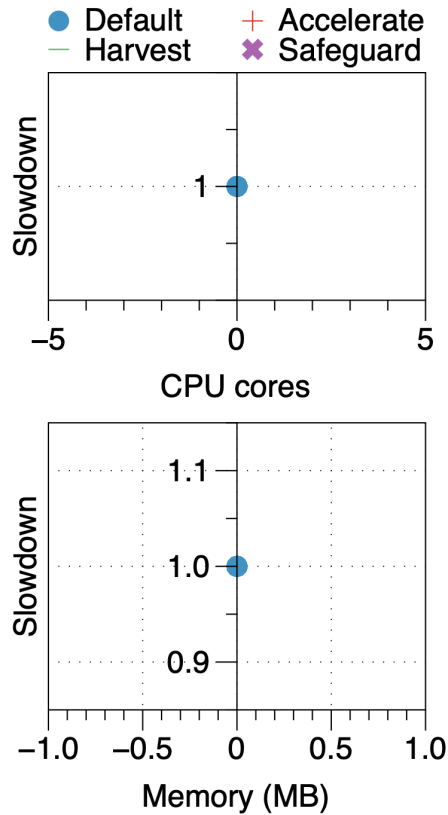
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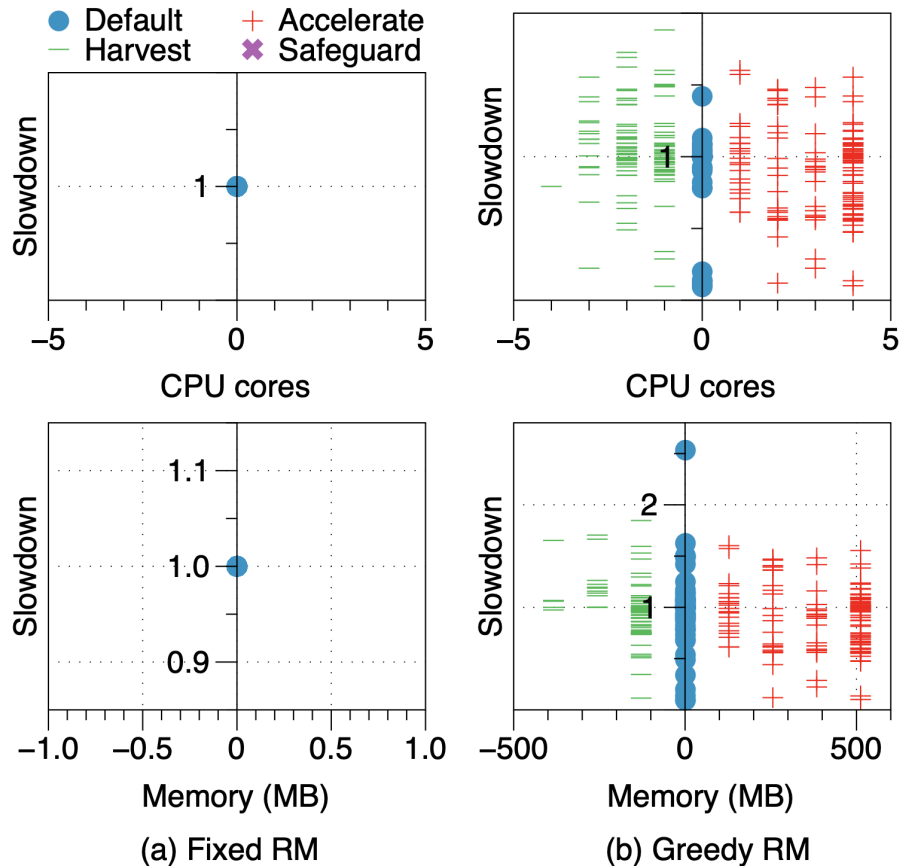
Resource Allocation



(a) Fixed RM

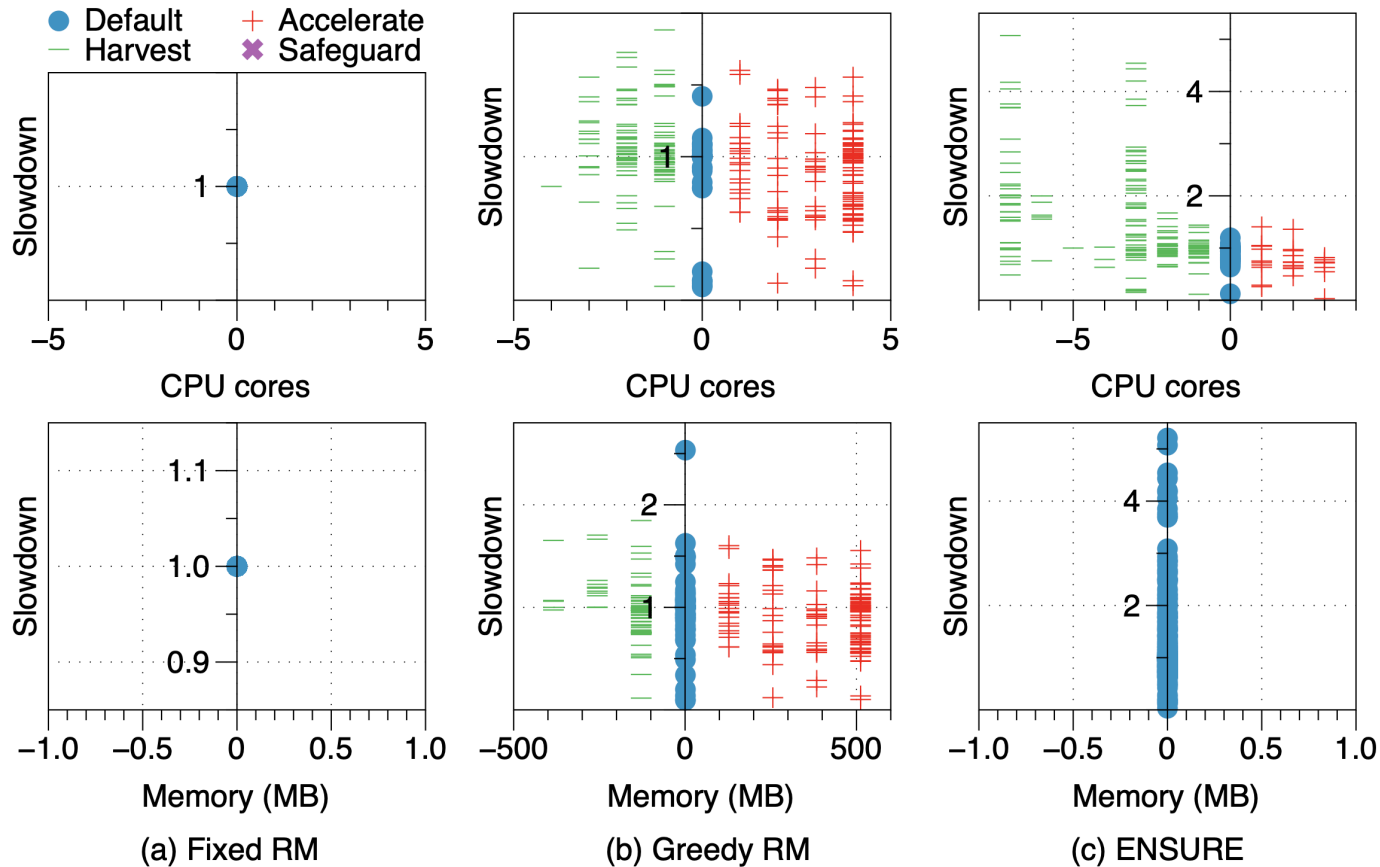
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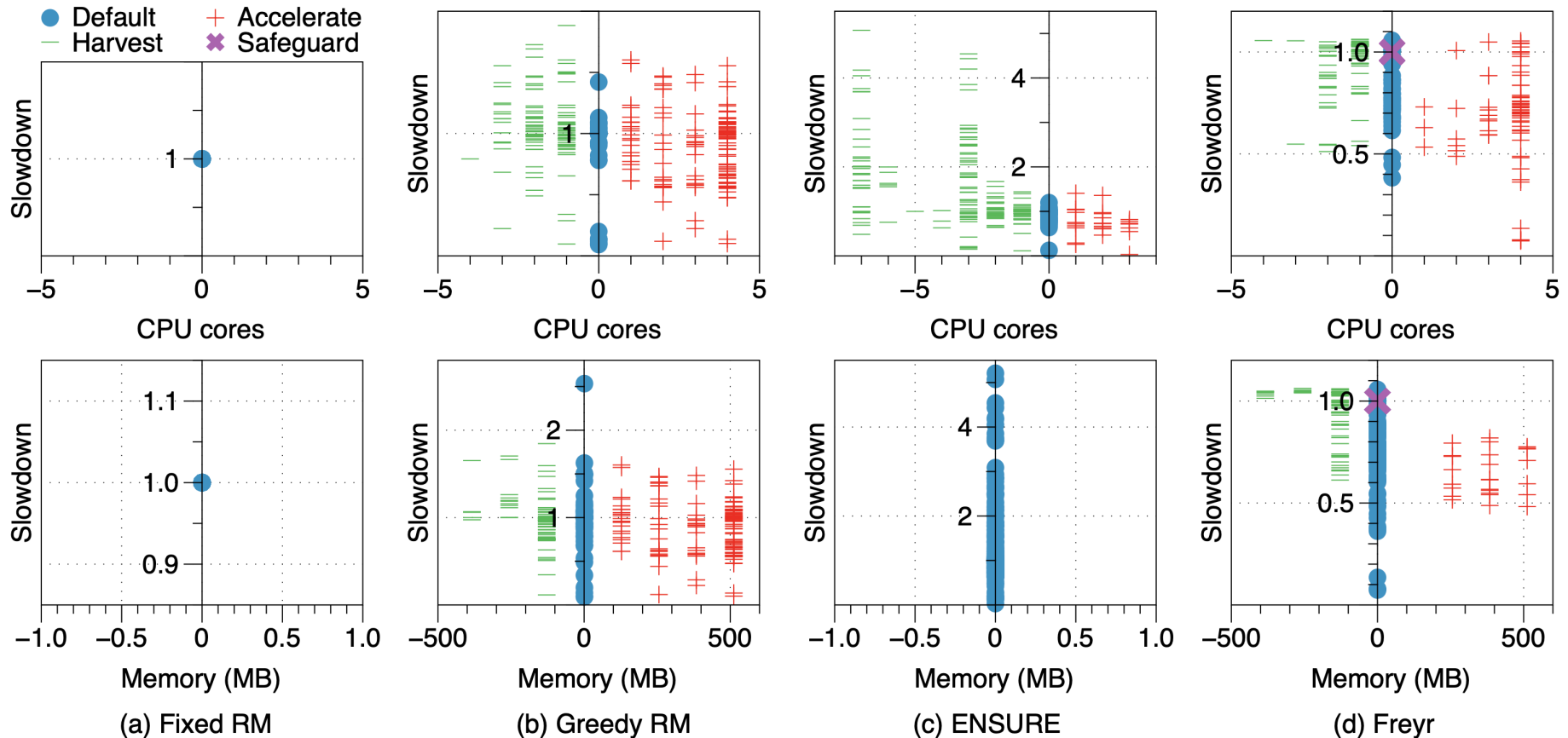
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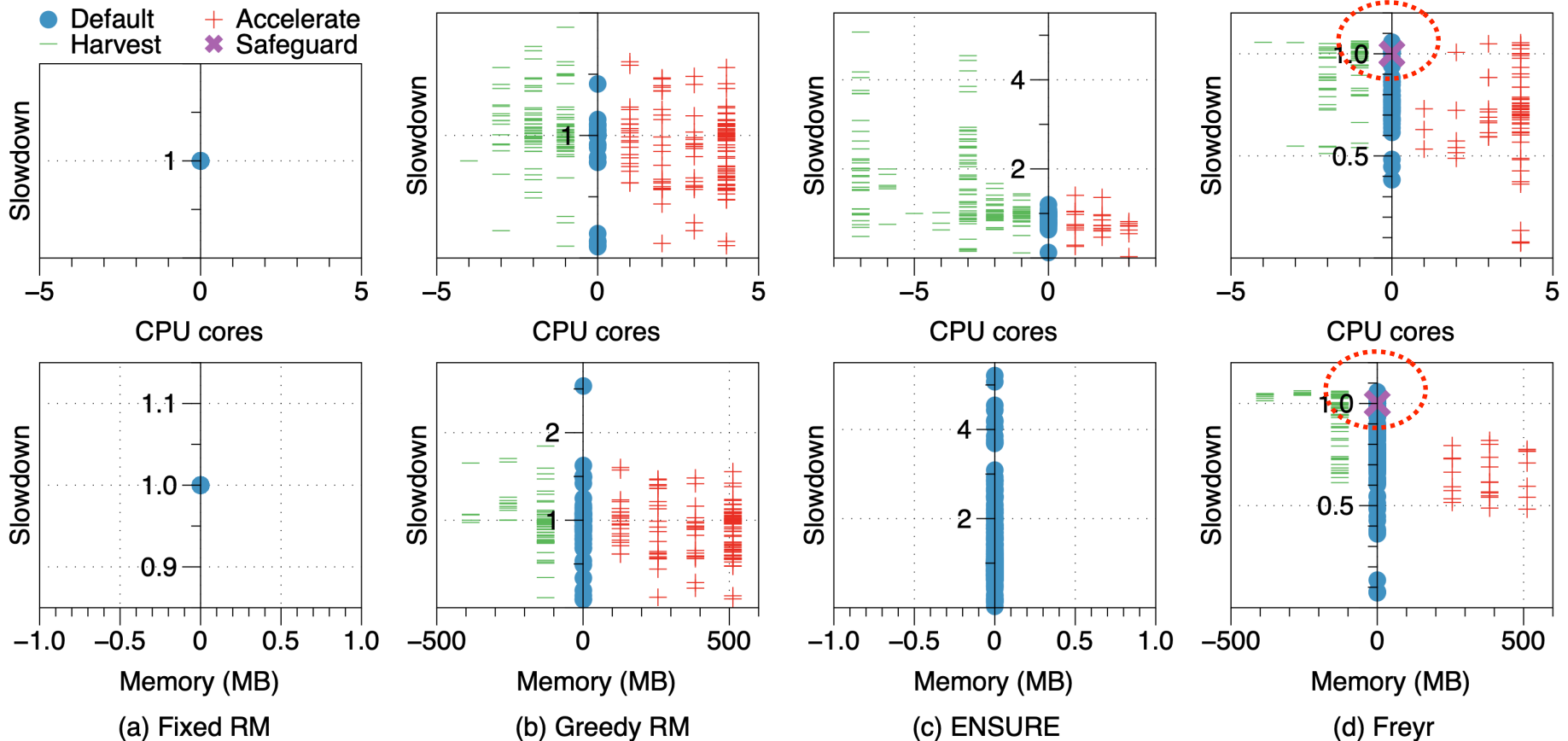
Resource Allocation



Slowdown: relative performance compared to user-defined resources. Larger than 1.0 means degradation, less than 1.0 means speedup.

Resource Allocation

Safeguard guarantees SLOs of harvested function invocations!



Slowdown: relative performance compared to user-defined resources. Larger than 1.0 means degradation, less than 1.0 means speedup.

Thank You

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IntelliSys Lab: <https://intellisys.haow.ca>



State University of New York

