

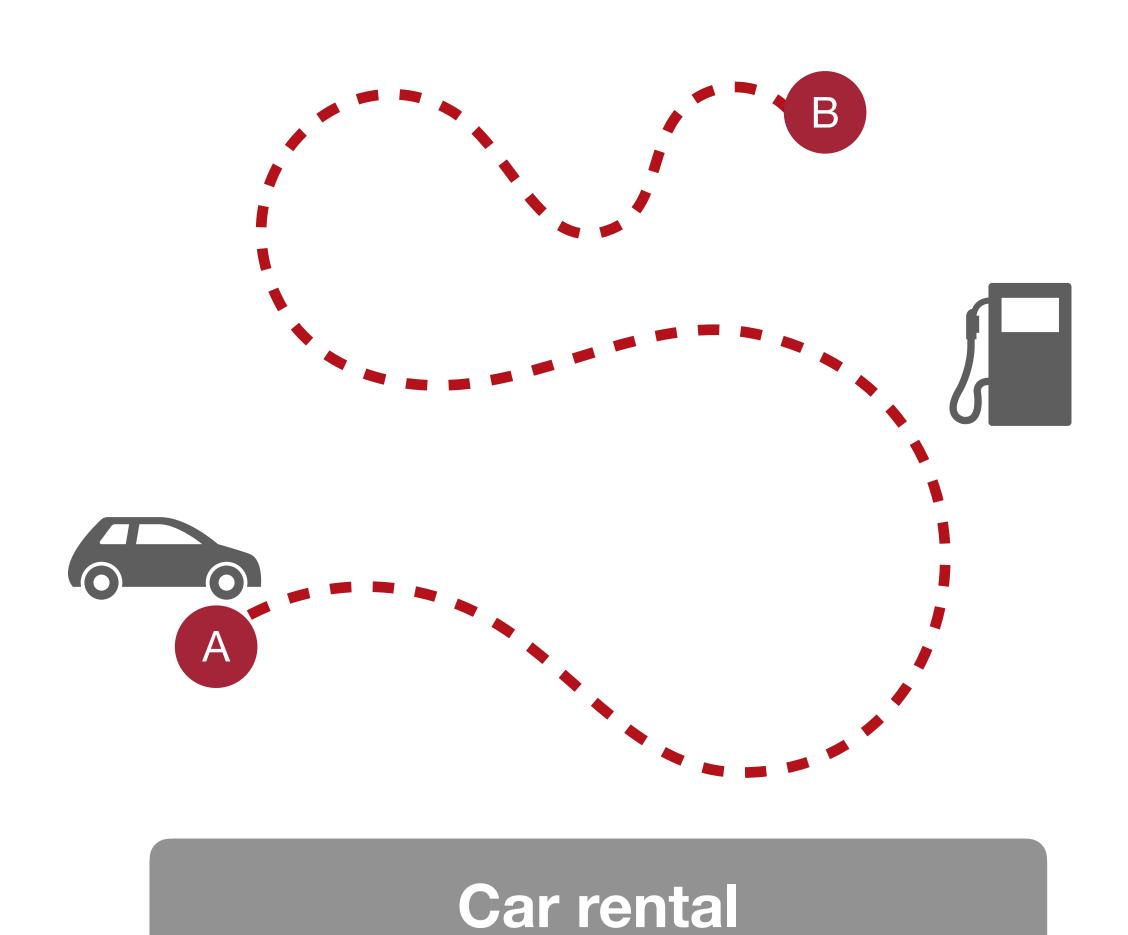
Pre-Warming Is Not Enough: Accelerating Serverless Inference With Opportunistic Pre-Loading

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Shanghai Jiao Tong University¹, Stevens Institute of Technology², Tianjin University³

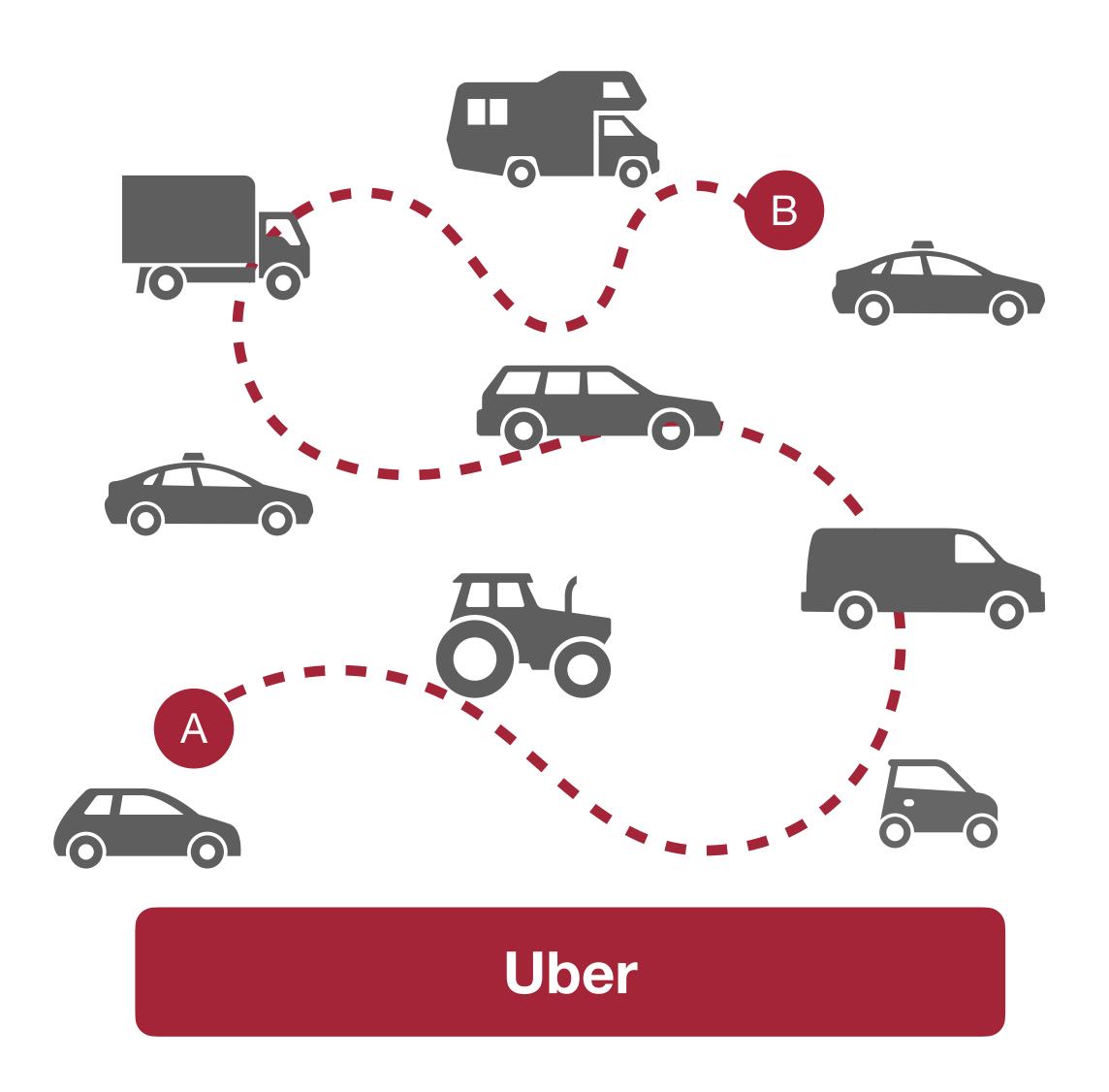


Cloud

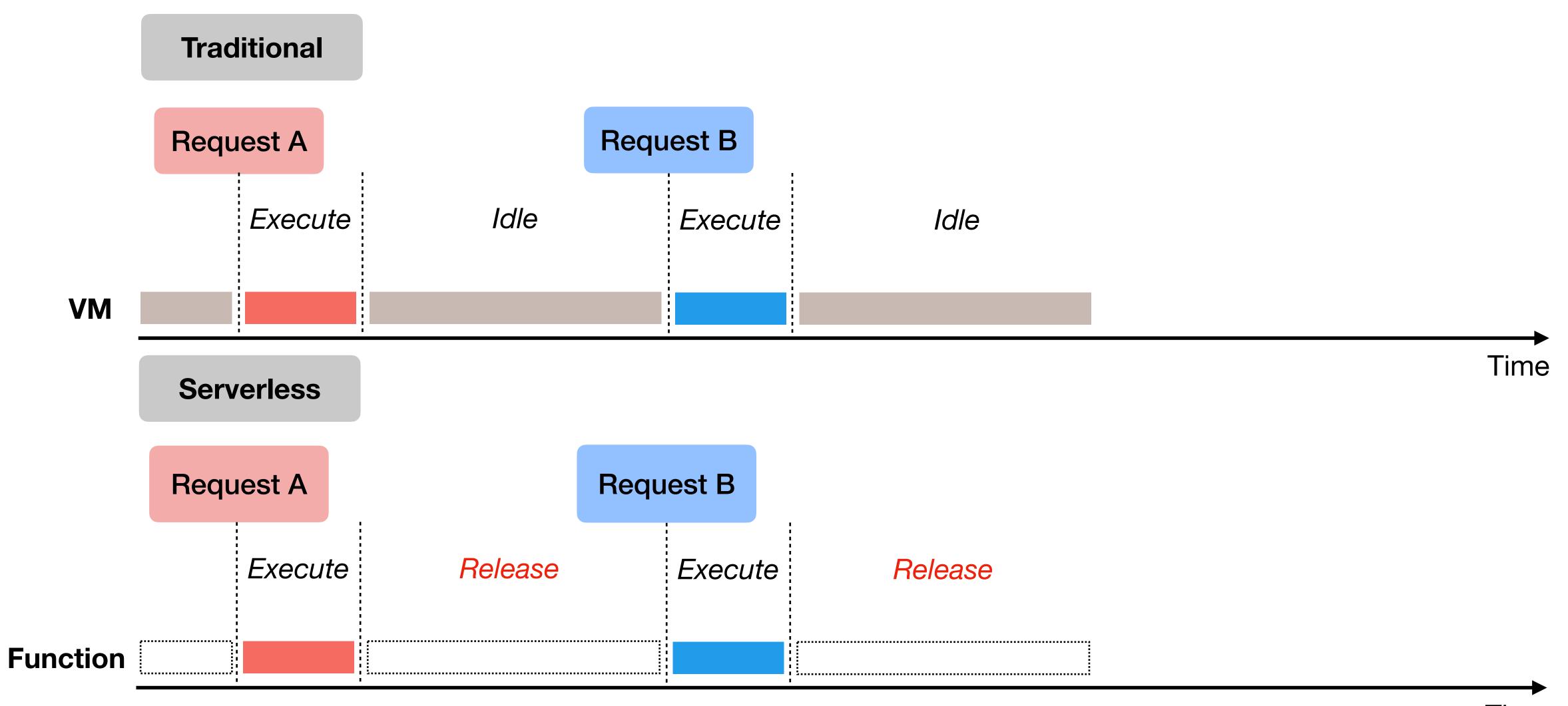


From *A Serverless Vision for Cloud Computing* by Prof. Ana Klimovic

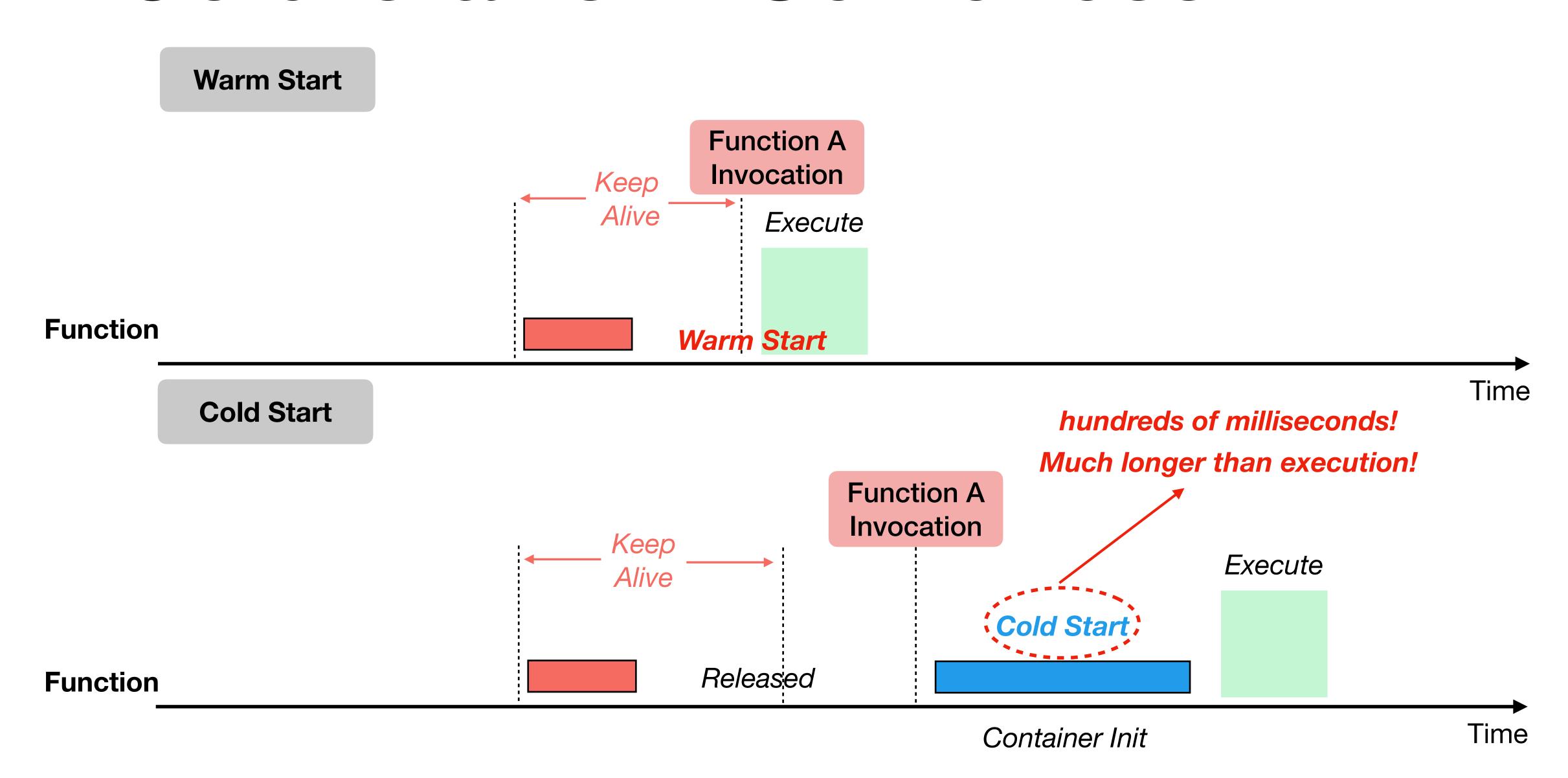
Serverless



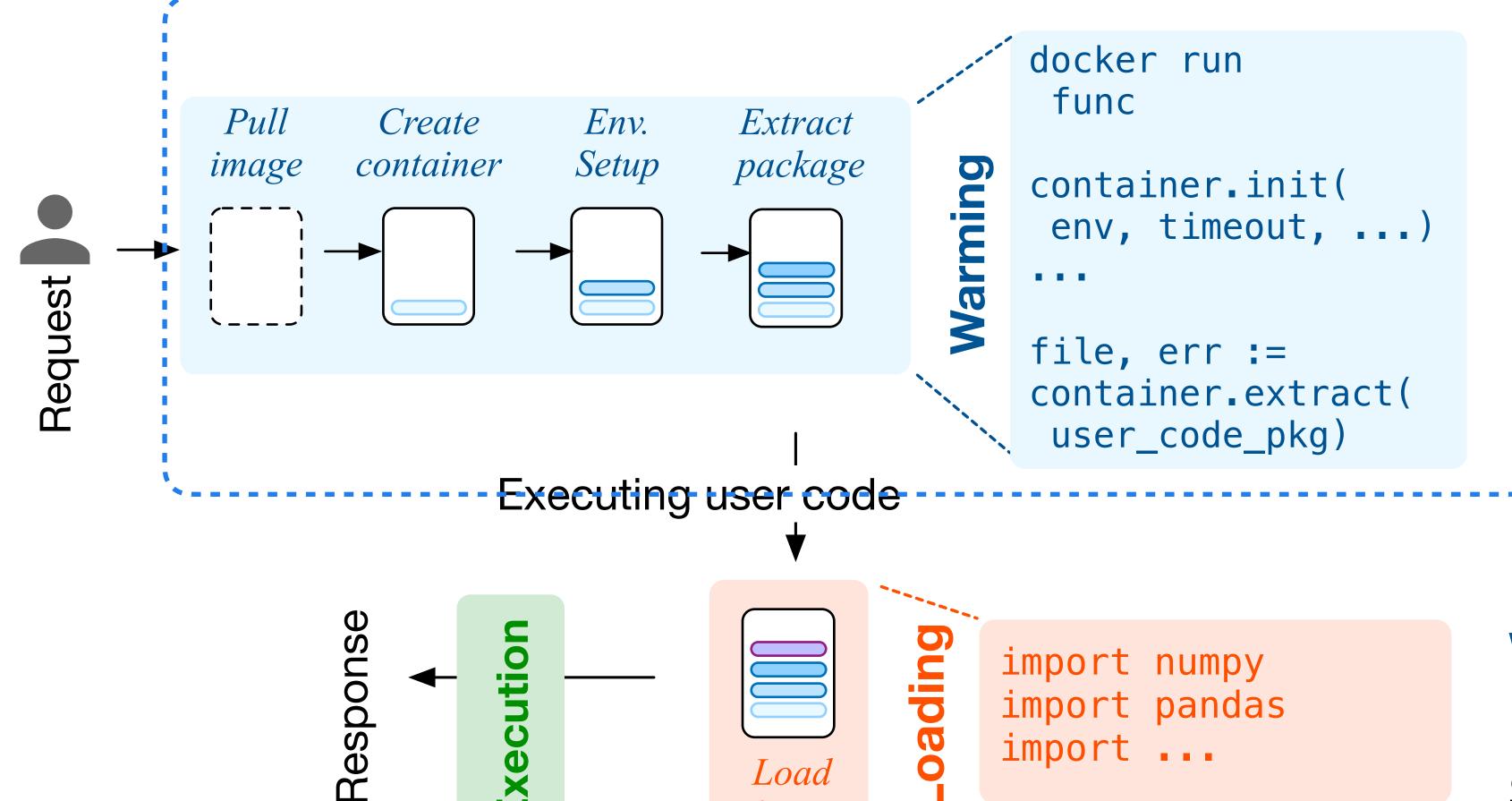
Traditional Cloud vs. Serverless



Cold-starts in Serverless



Serverless Cold Start



Libraries

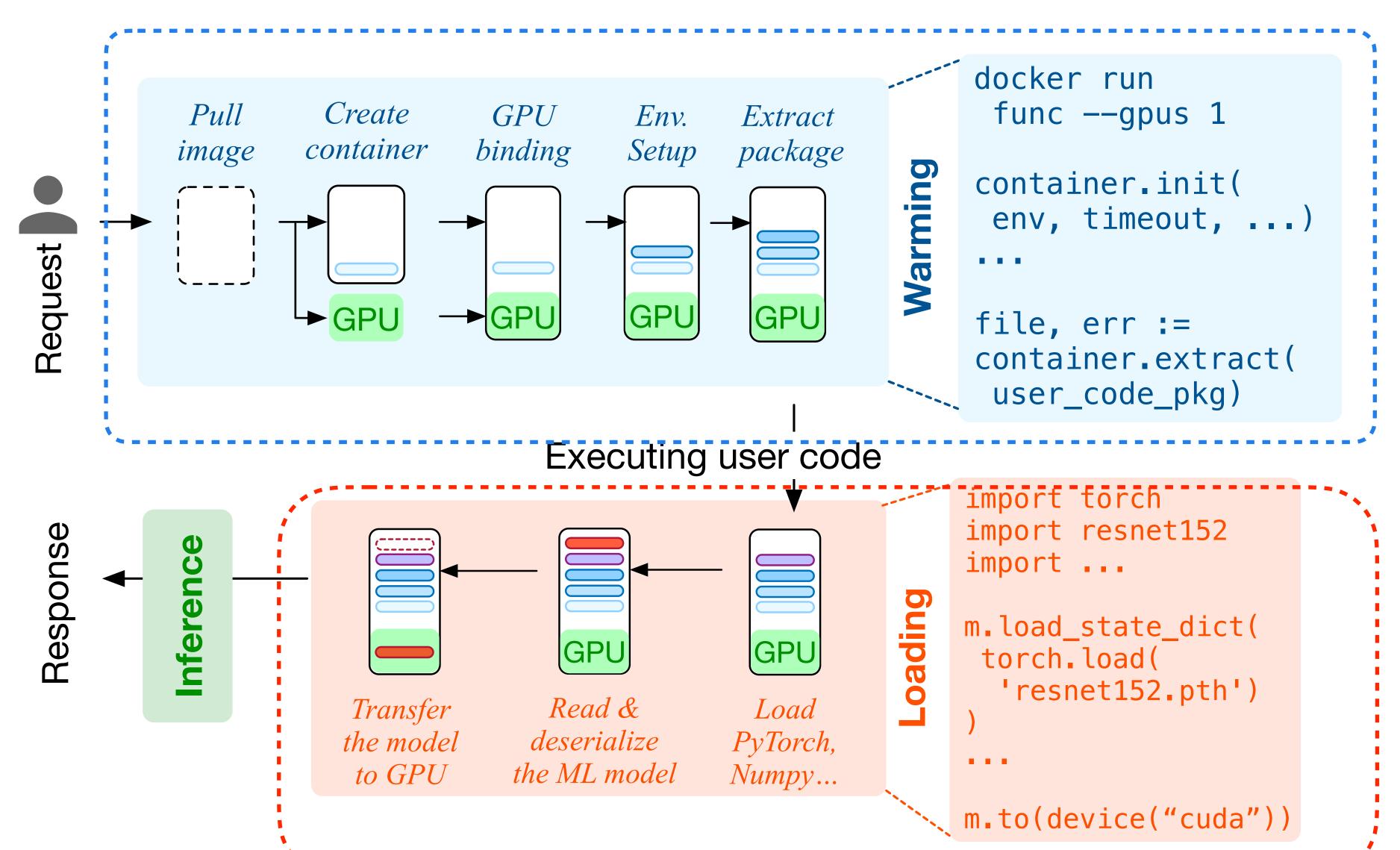
Pre-Warming:

Warm containers before request arrives

Warming >> Loading [

Solution: Pre-Warming

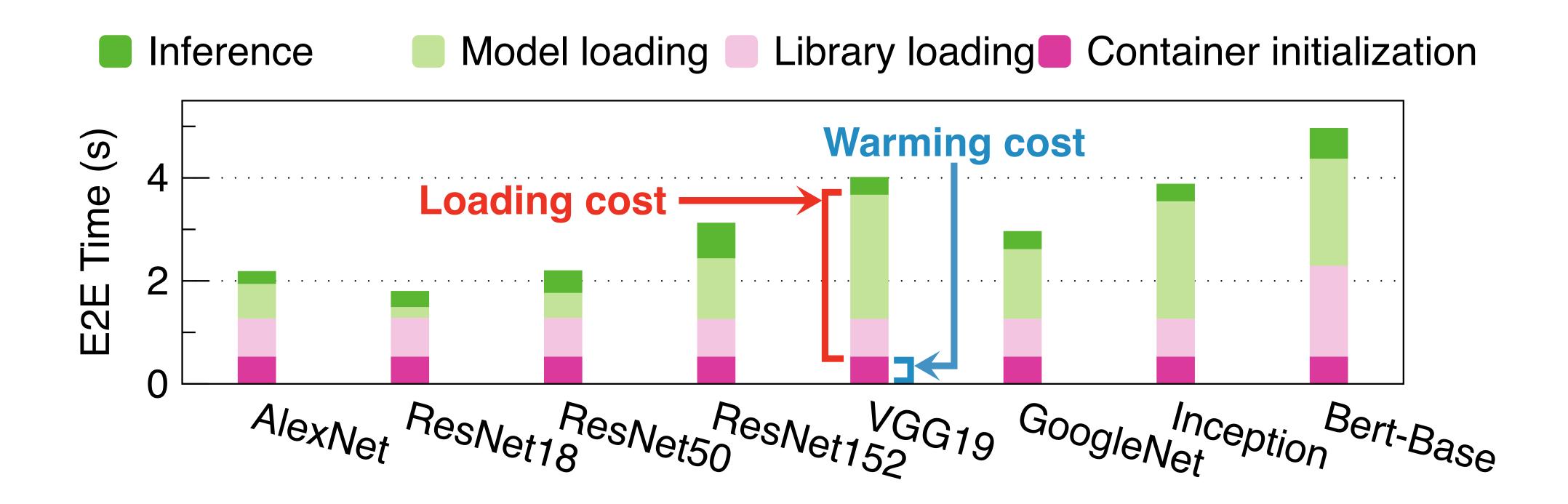
ML Inference's Cold Start



Pre-Warming's Scope:
Complete warming
before request arrives

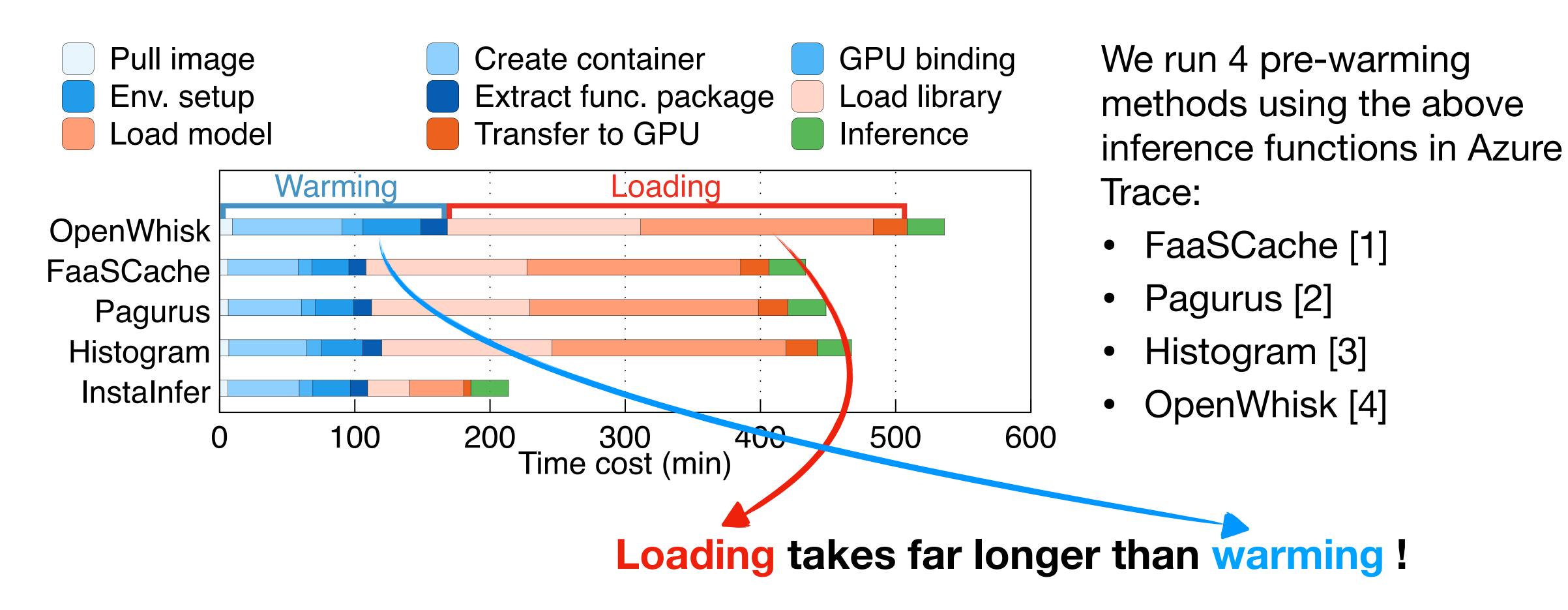
Loading is out of prewarming's scope.

Existing Works:



In a cold start, loading takes far longer than warming!

Limitations of Existing Works

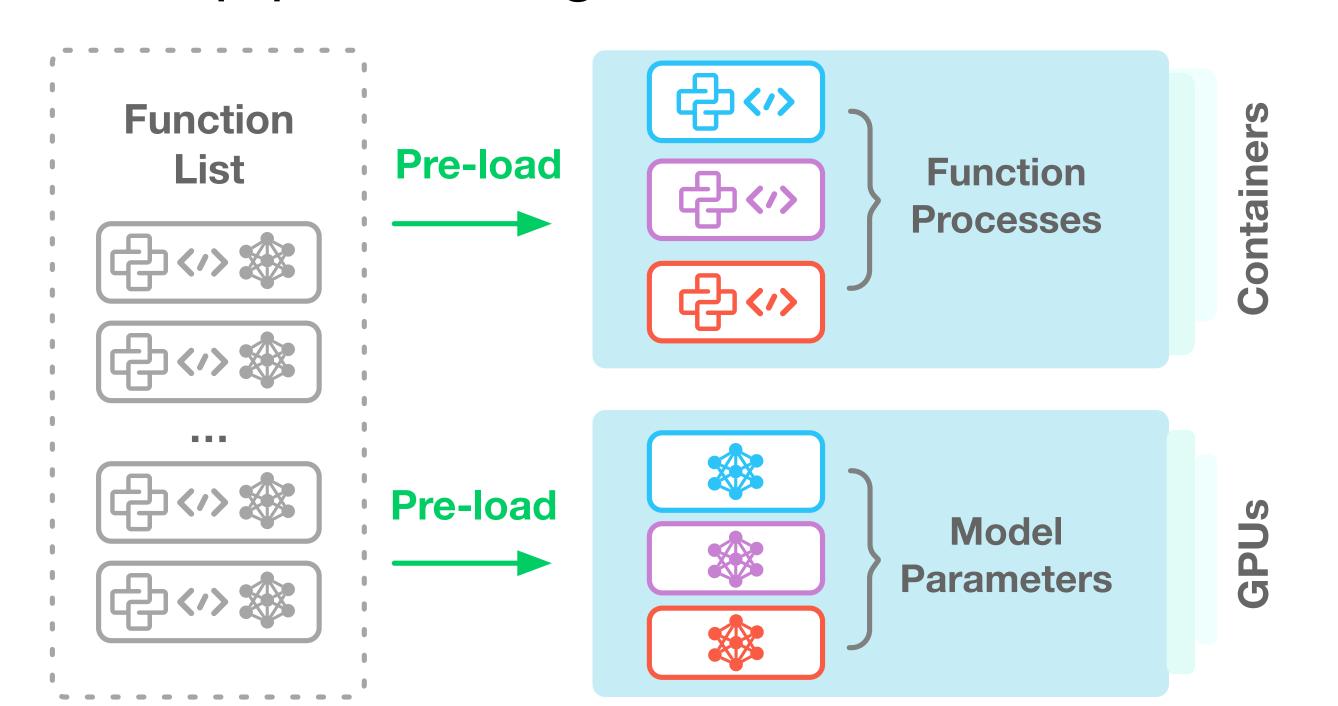


- [1] FaaSCache: Fuerst, Alexander, et al. "SeBS: A Serverless Benchmark Suite for Function-as-a-Service Computing." ASPLOS'21
- [2] Pagurus: Li, Zijun, et al. "Help Rather Than Recycle: Alleviating Cold Startup in Serverless Computing..." ATC'22
- [3] Histogram: Shahrad, Mohammad, et al. "Serverless in the Wild: Characterizing and Optimizing the Serverless..." ATC'20
- [4] OpenWhisk: Apache OpenWhisk. https://openwhisk.apache.org.

Our Contribution:

Opprtunistic Pre-loading:

- Pre-load inference functions in idle containers and transfer the model to GPU instances.
- Keep pre-loading until all idle instances are full.





Design Goals

Minimizing Loading Latency



Pre-load multiple functions within a container

Zero Wastage



Only utilize existing container/GPU

Compatible with pre-warming

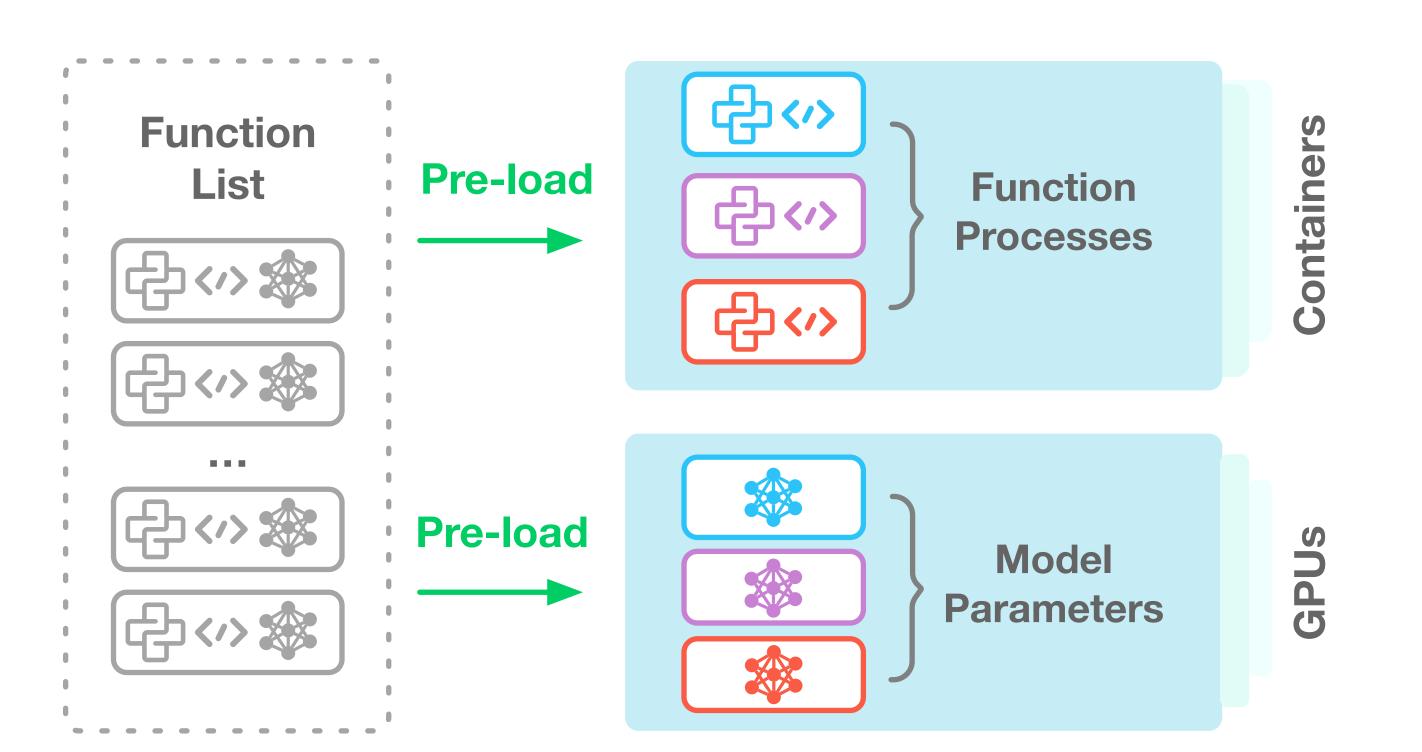


Never affect container creation & removal

Opportunistic Pre-loading

So many inference functions, only limited containers.

- Instance level: Pre-load multipe functions within a container/GPU until full.
- Platform level: Pre-loading until all idle instances are full.



Opportunistic:

- Only use idle instances.
- Never create new one.

Zero Wastage

 Never affect the lifetime of prewarmed instances.

Compatible with Pre-warming

Challenges

When to pre-load / off-load a function?

Based on its request's arrival rate

Estimated using the prediction model of pre-warming methods.

How to minimize loading latency within the limited instances?

Priority-based Bin-Packing

$$E(f) = P_{arrive} \times T_{loading}$$

Challenges

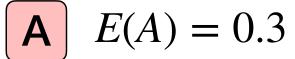
When to pre-load / off-load a function?

Based on its request's arrival probability

Estimated using the prediction model of pre-warming methods.

How to maximize acceleration within the limited instances? Priority-based Bin-Packing

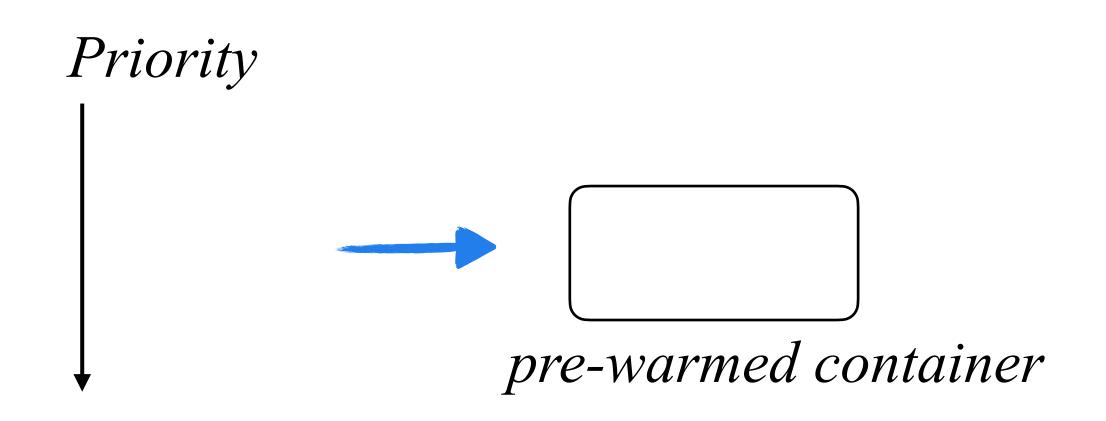
$$E(f) = P_{arrive} \times T_{loading}$$



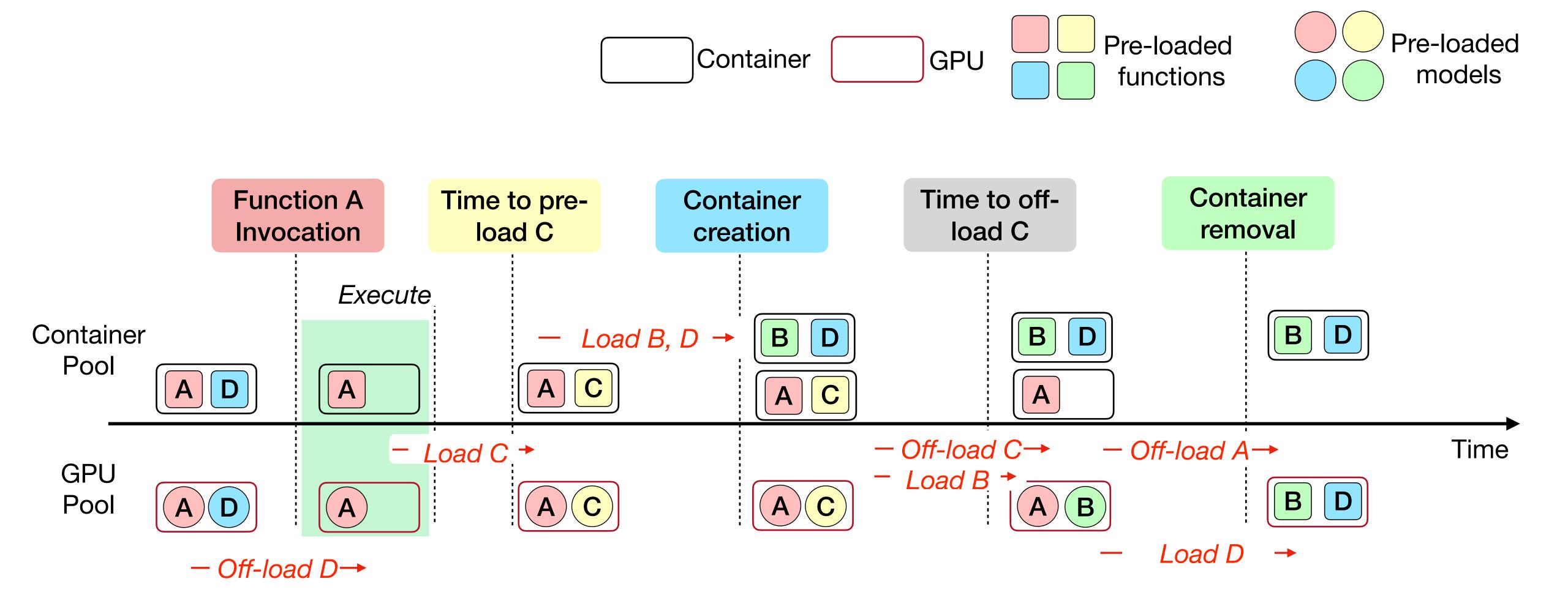
$$E(B) = 1$$

$$E(C) = 0.8$$

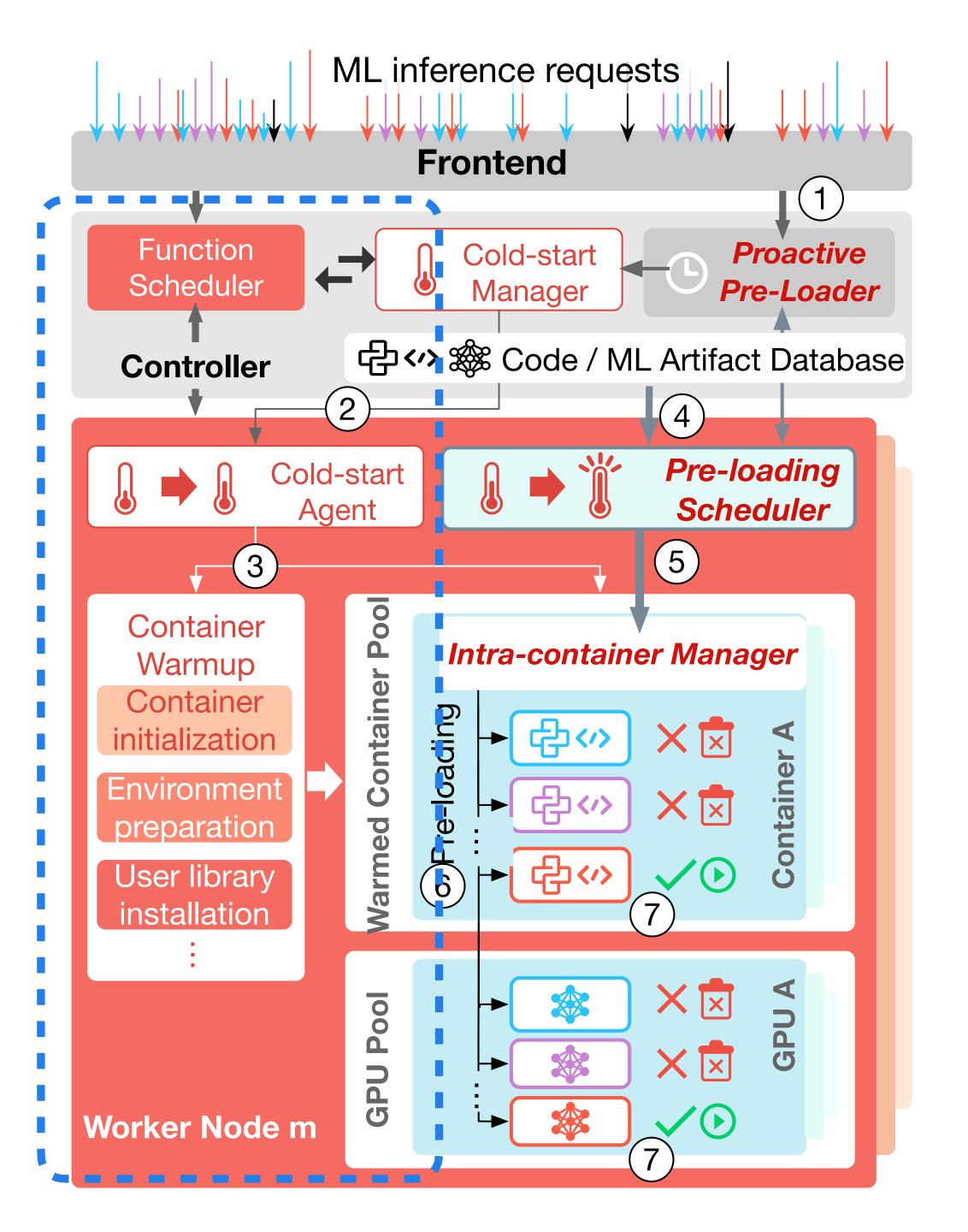
D
$$E(D) = 1.5$$



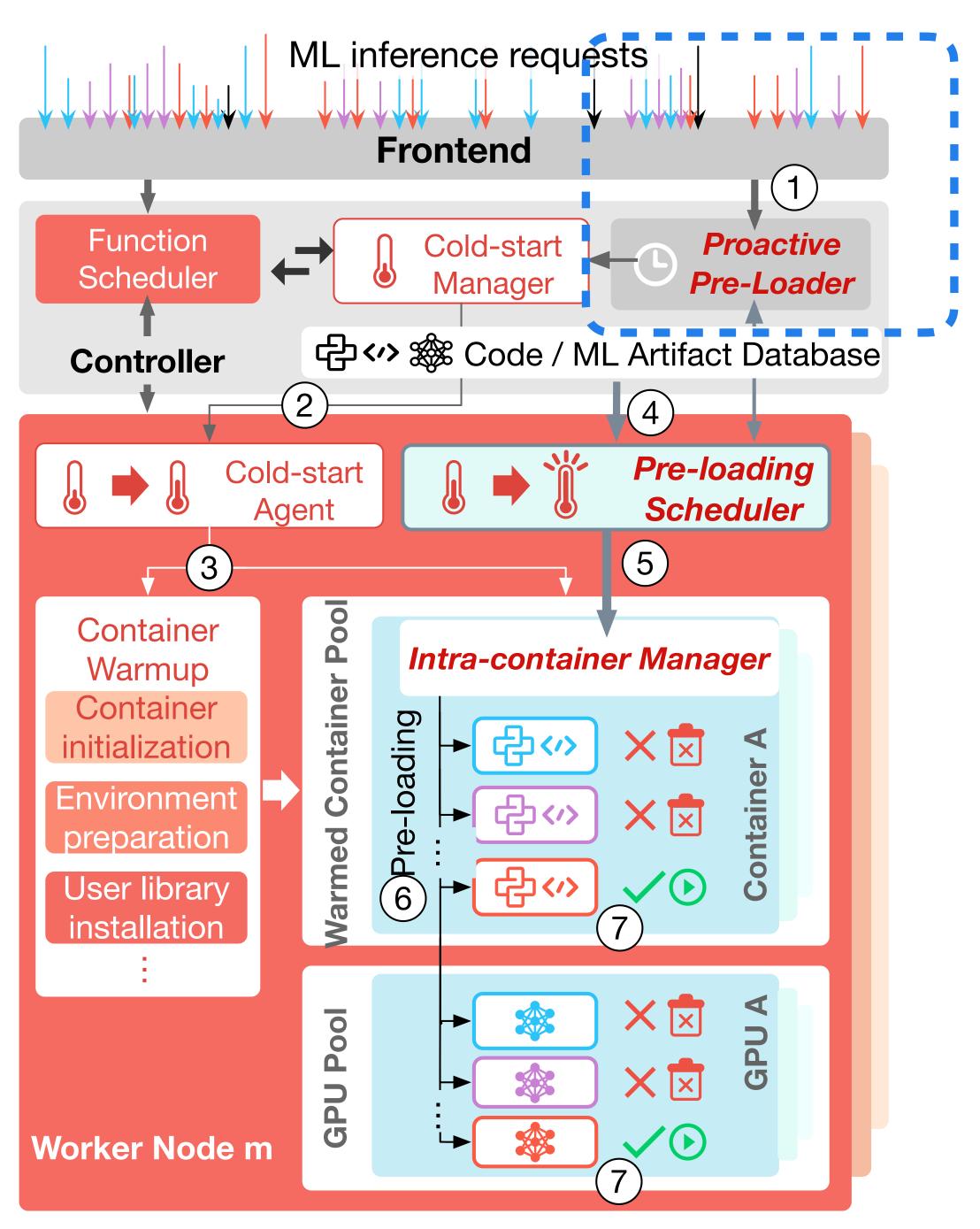
Instalnfer Workflow



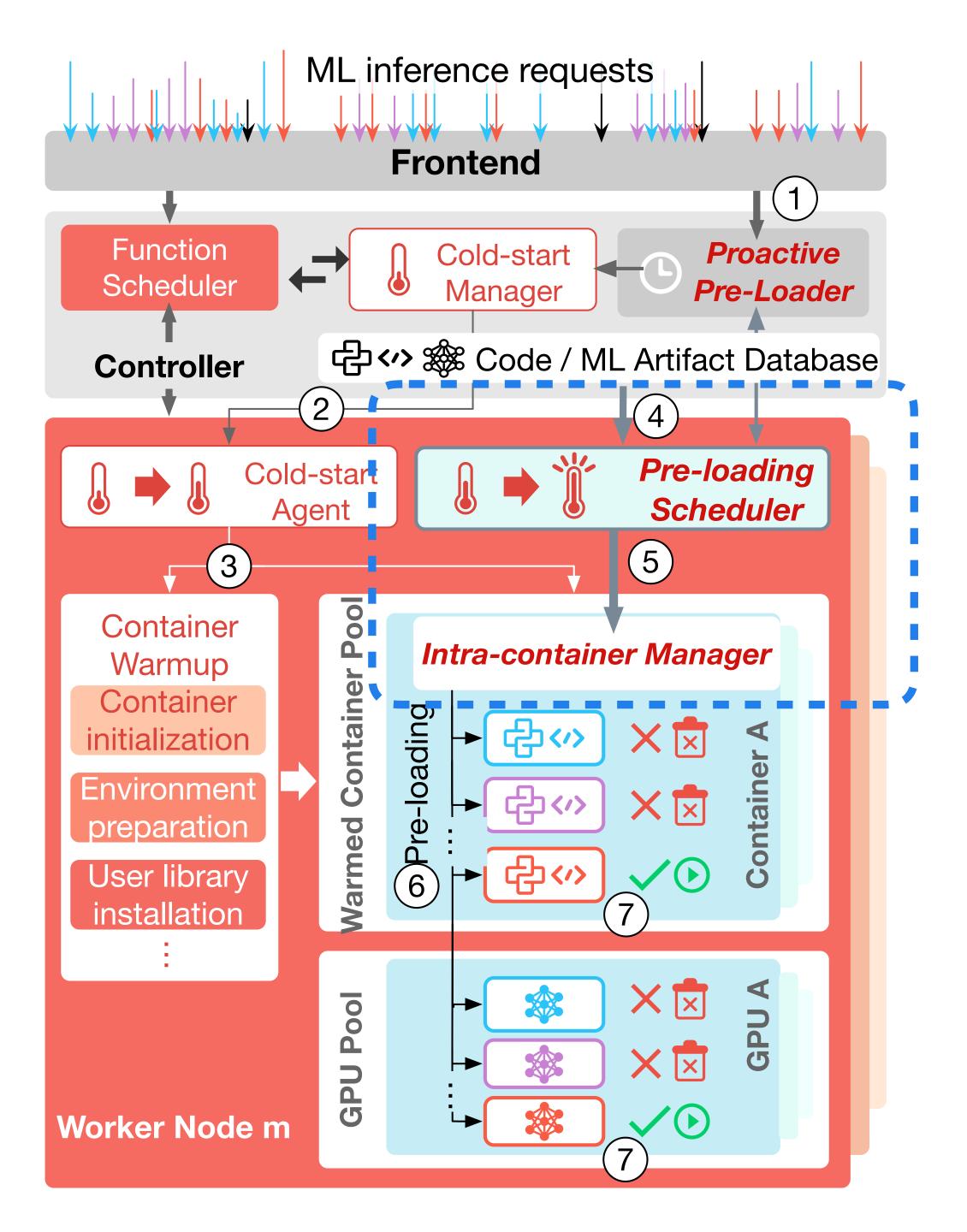
The pre-warming mechanism creates idle containers.



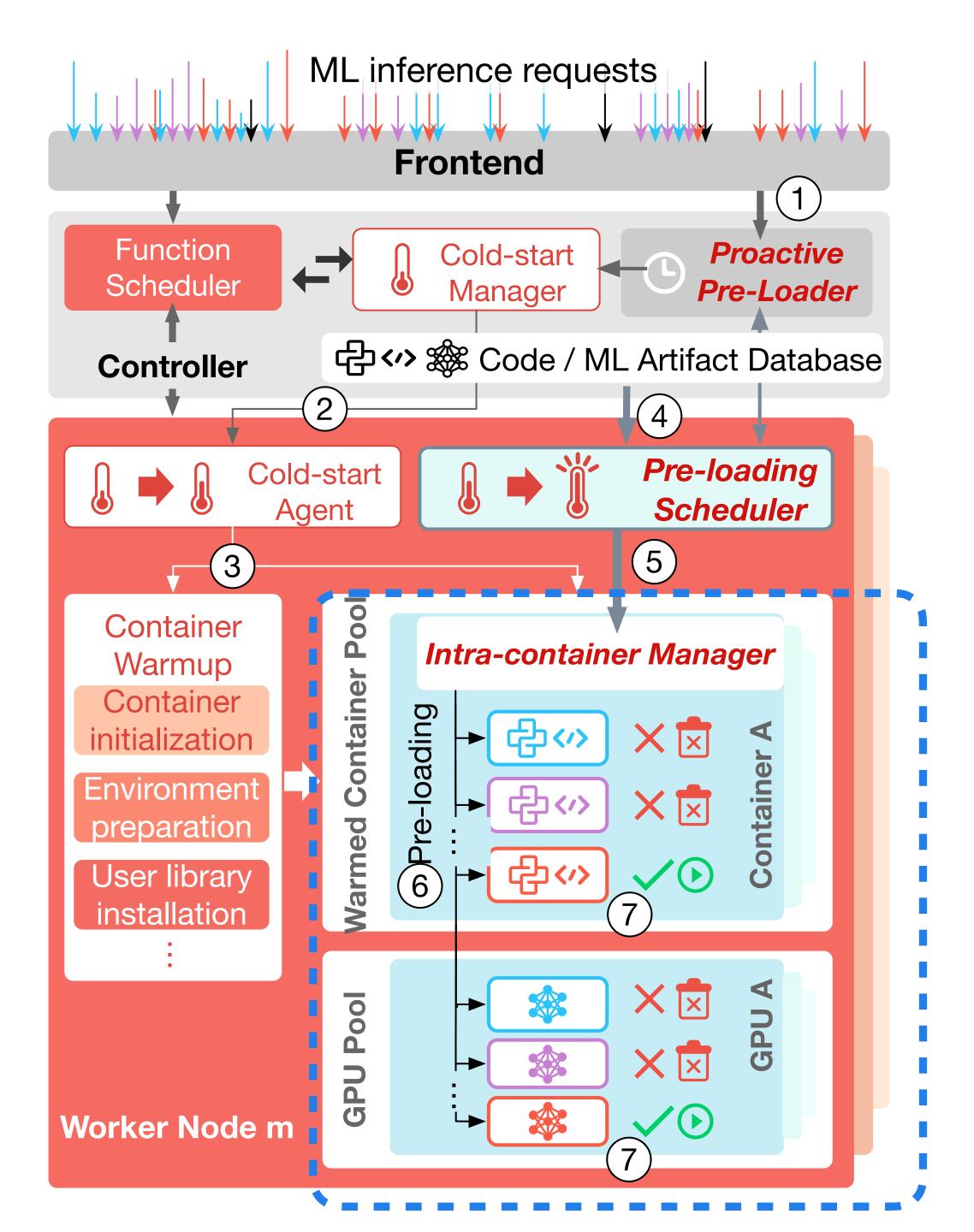
Records past invocations to decide when to load each function.



Operates bin-packing, pre-loading functions into instances.

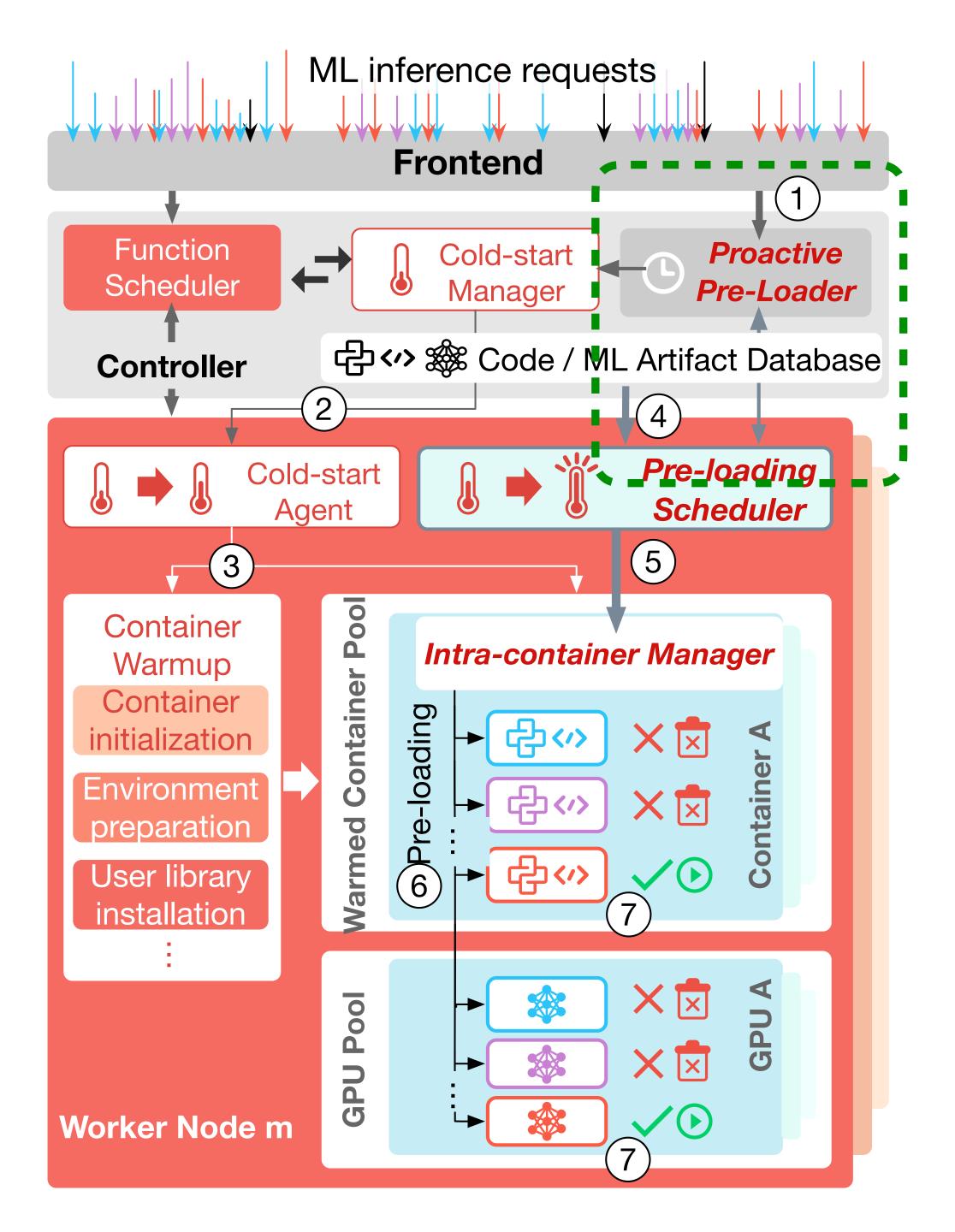


Operate loading and off-loading; Guarantee security



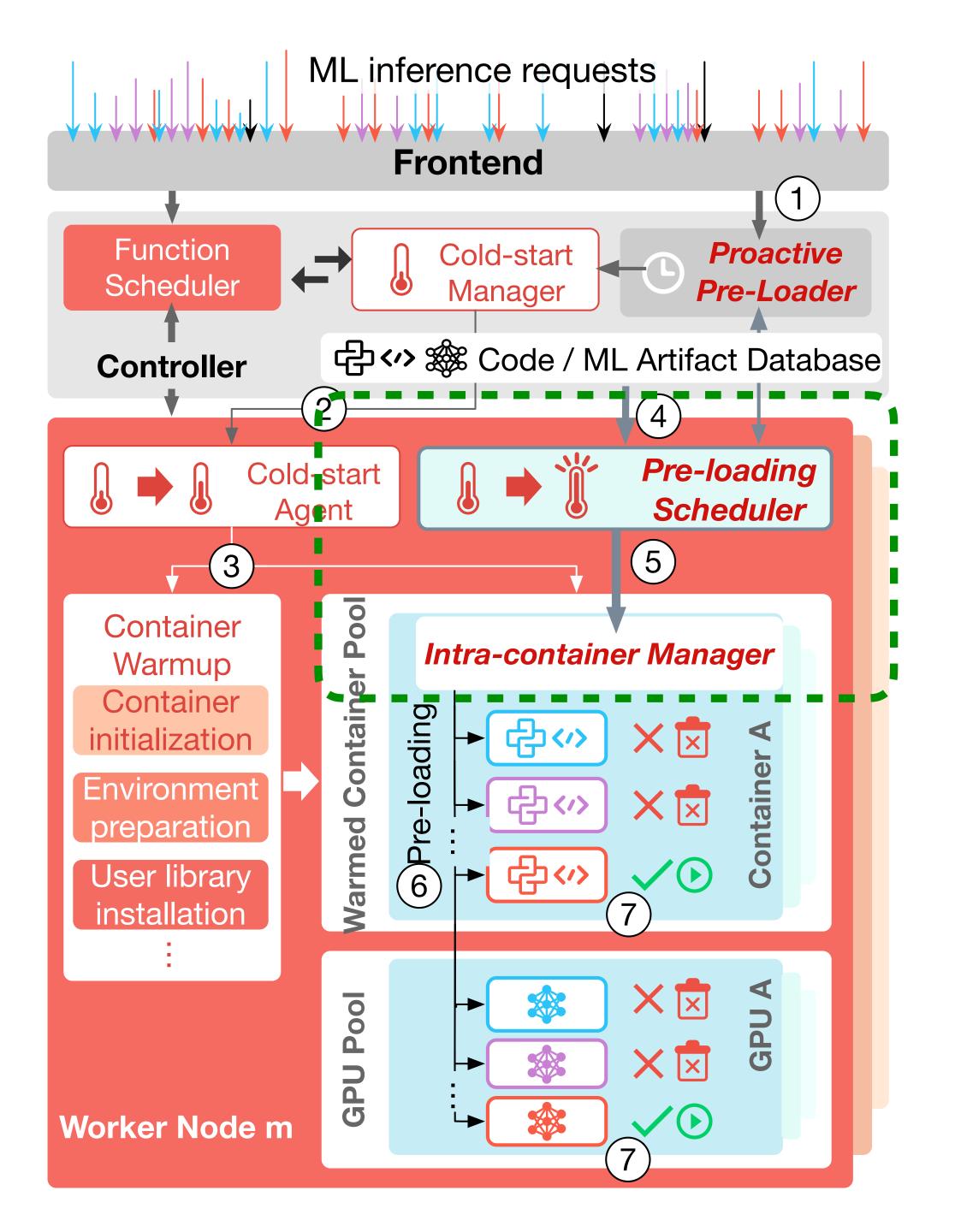
Invocation Reaction Workflow

Choose a worker node with an instance that loads the function.



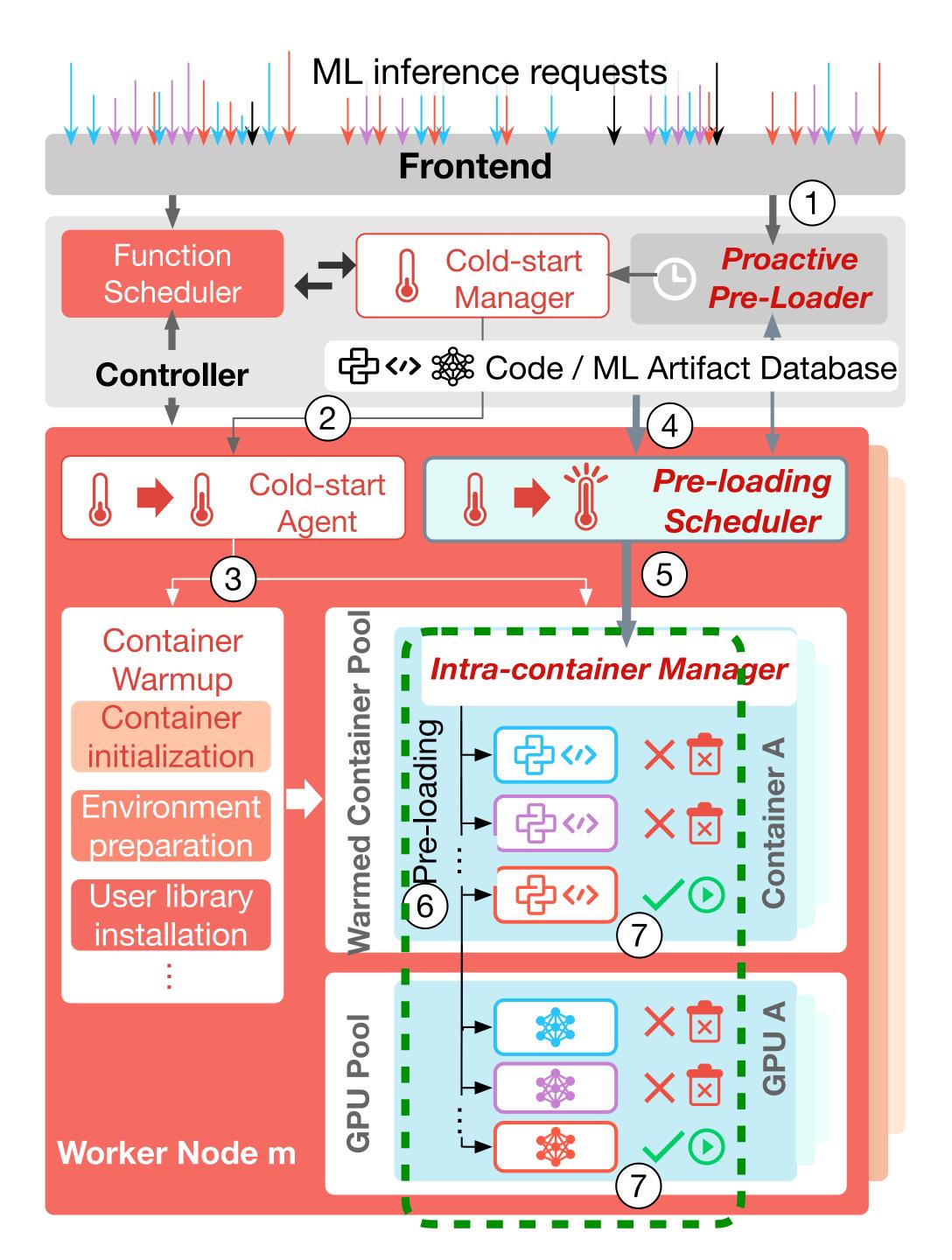
Invocation Reaction Workflow

Send the invocation to the container that loaded the function



Invocation Reaction Workflow

The container terminates all other functions and starts inference.



Implementation

Instalnfer is prototyped on top of Docker and Apache OpenWhisk

Proactive Pre-Loader

OpenWhisk's Load Balancer

Pre-Loading
Scheduler

OpenWhisk's Container Pool

Intra-Continaer
Manager

Docker Container's Proxy

GPU Support

Nvidia MPS &

Nvidia Container

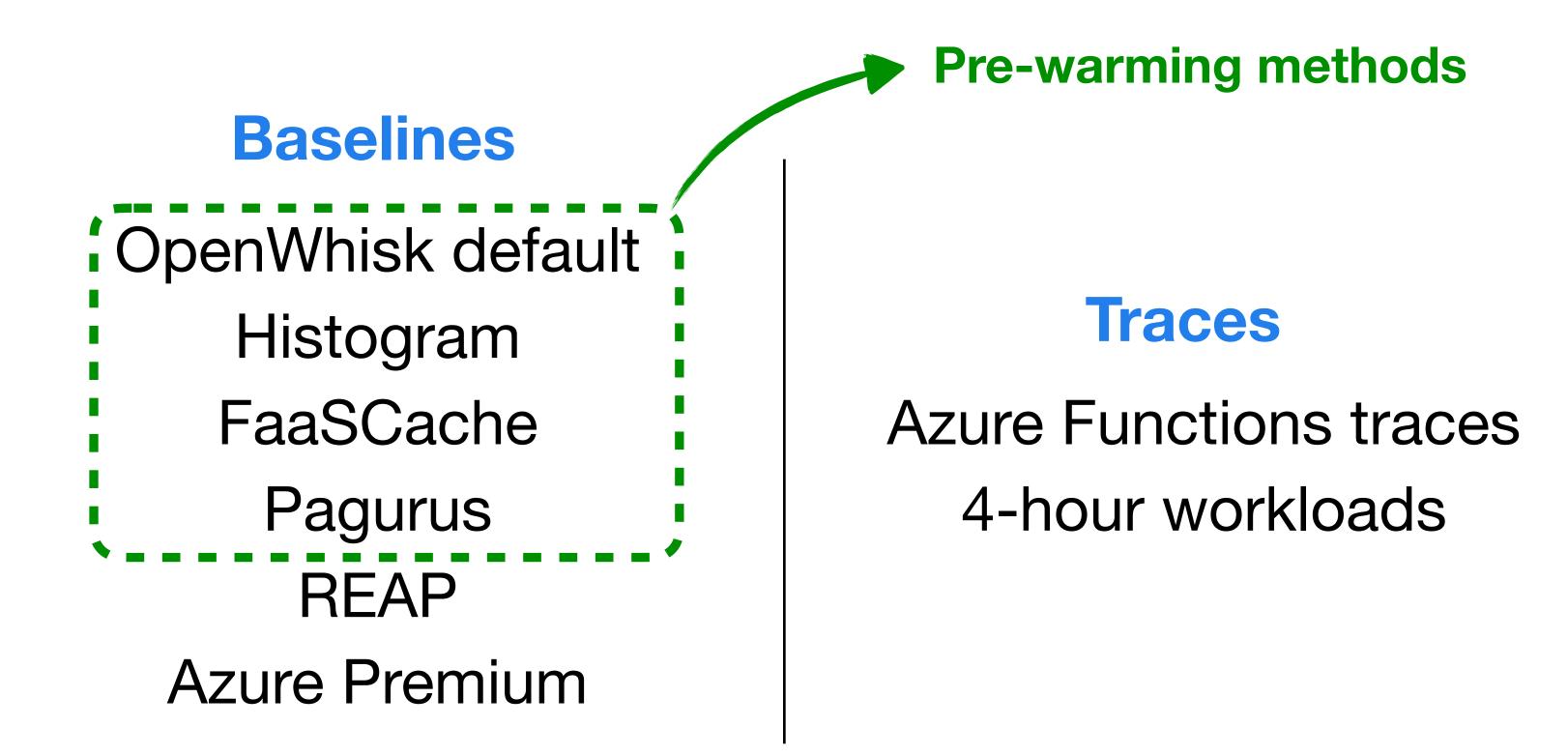
Testbed

5 nodes
160 AMD EPYC CPU cores
640 GB Memory
4 Nvidia A10 GPUs

Metrics

Function response latency
Memory cost

Evaluation



Histogram: Shahrad, Mohammad, et al. "Serverless in the Wild: Characterizing and Optimizing the Serverless..." ATC'20
FaaSCache: Fuerst, Alexander, et al. "SeBS: A Serverless Benchmark Suite for Function-as-a-Service Computing." ASPLOS'21
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Azure Premium: Microsoft. Azure Functions warmup trigger. 2023

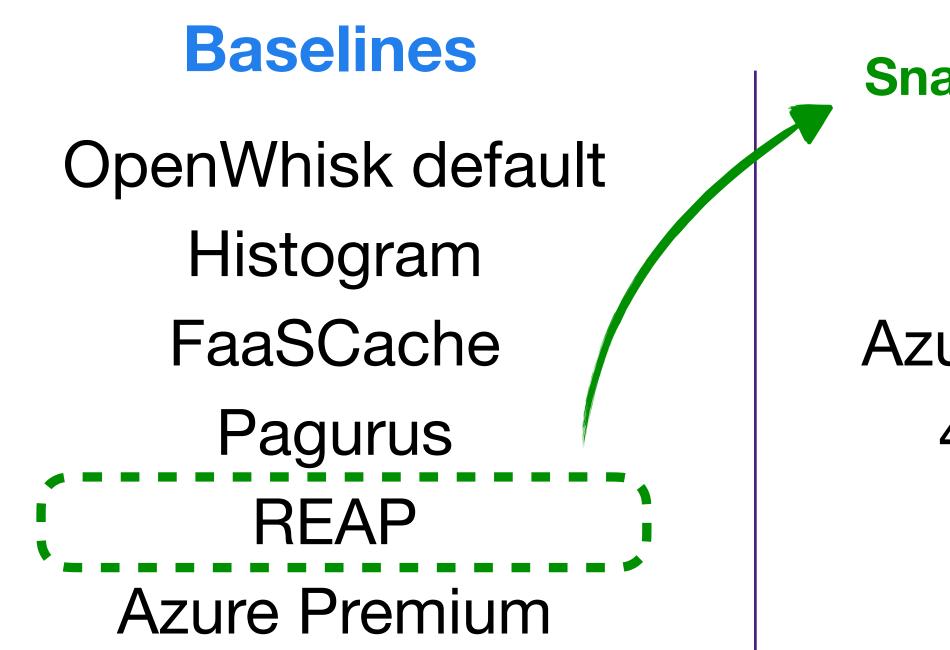
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Snapshot method

Traces

Azure Functions traces
4-hour workloads

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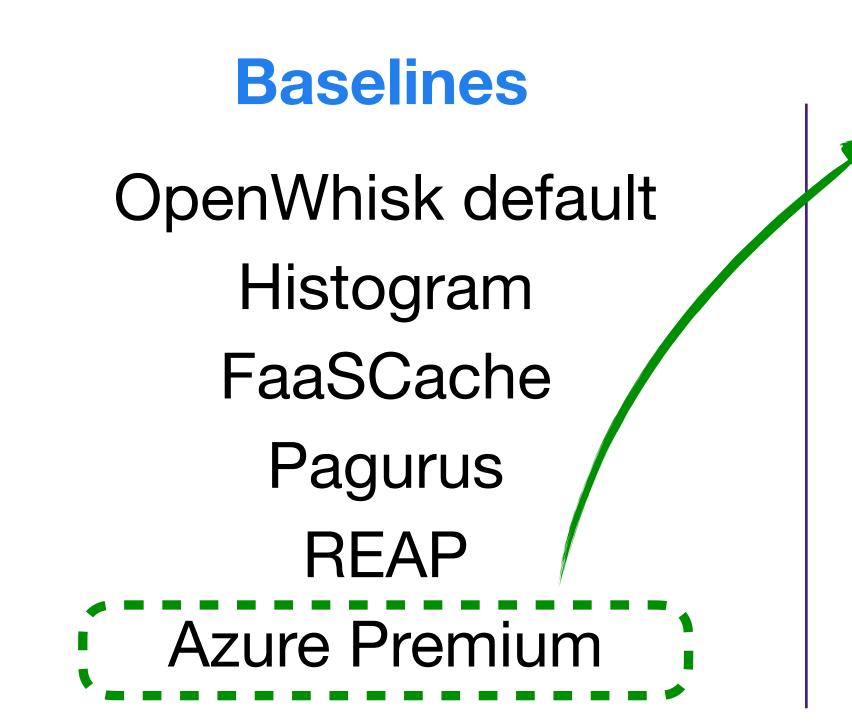
Testbed

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160 AMD EPYC CPU cores
640 GB Memory
20 A10 GPUs

Metrics

Function response latency
Memory cost

Evaluation



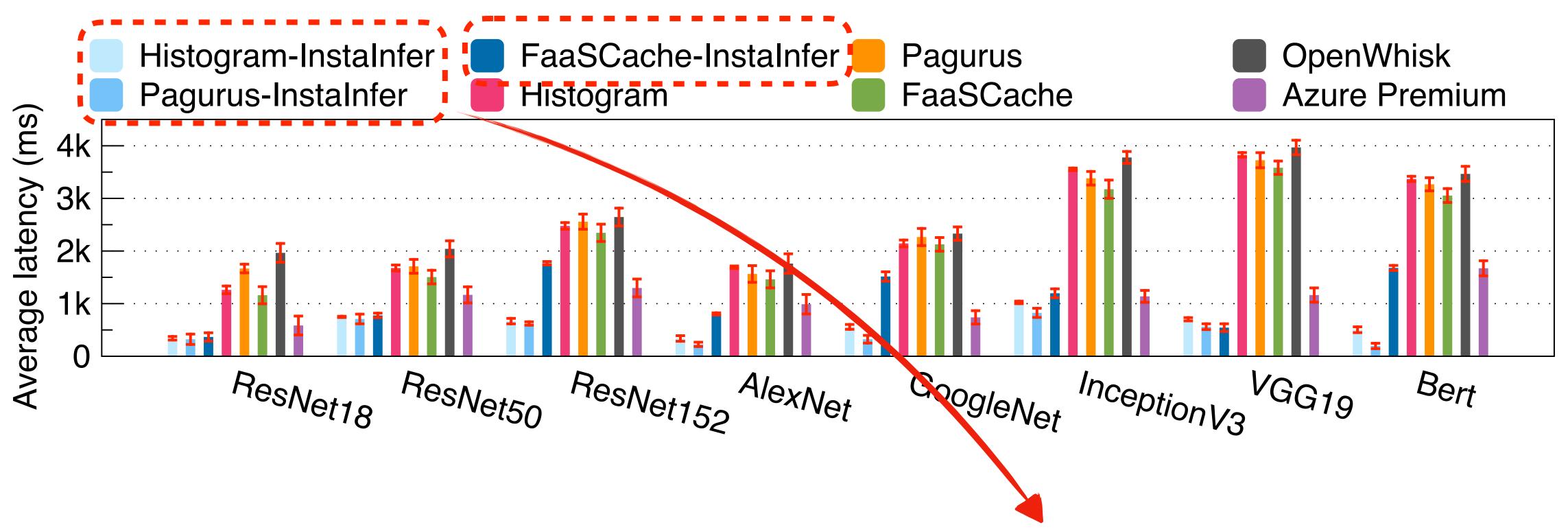
Pre-loading method

Traces

Azure Functions traces
4-hour workloads

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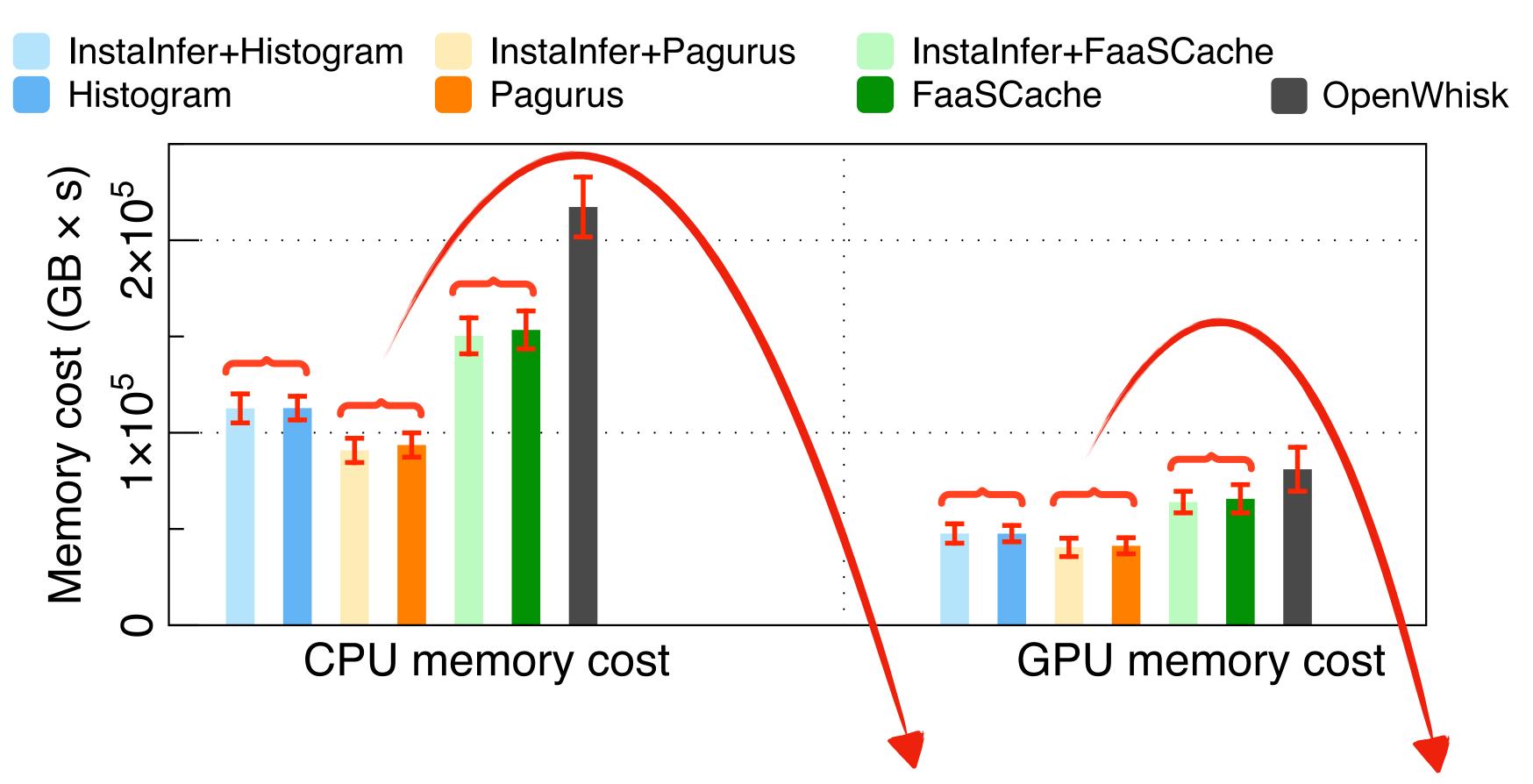
End-to-end Latency



Instalnfer is compatible with each pre-warming baseline

Instalnfer achieves better invocation latency than other baselines

Memory Cost



Instalnfer does not introduce additional memory cost and GPU cost after combining with pre-warming baselines

Opprtunistic pre-loading

Instalnfer

Minimize loading latency

93%

Function loading latency reduction

No additional cost

<1%

Memory waste

Transparent to providers

4

Compatible pre-warming methods



Instalnfer Code Repo:

https://github.com/IntelliSys-Lab/InstaInfer



Jianxun Li < lijx@sjtu.edu.cn>









Q & A

Security & Privacy

- **1. User layer:** Only functions belonging to the same user can be preloaded on the same machine.
- 2. Process layer: Deletes all other functions's data when a function is invoked
- 3. OS layer: Each pre-loaded functions' code, librairies, and data are in an unique non-root Linux user.

Proviacy: We treat each function's package is a blackbox. (Developers are asked to modify two lines of code)

Modification Guide

Developers are asked to modify two lines of code:

Original:

```
model.load_state_dict(torch.load(model_path))
# inference...
```

Instalnfer:

```
model.load_state_dict(InstaInfer_load_model(model_path))
sys.stdin.readline() # wait for request
# inference...
```