BeeHive: Sub-second elasticity for web services with Semi-FaaS execution

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Web Application and Dynamic Workload

Request bursts are long-term enemies for web applications



Slower response

Unavailability



- Dynamic workload demands <u>rapid</u> and <u>cost-efficient</u> burst handling
 - Reserving computation resource -> high cost
 - On-demand scaling -> slow response

https://www.theverge.com/2018/7/16/17577654/amazon-prime-day-website-down-deals-service-disruption https://www.nytimes.com/2015/12/01/technology/target-paypal-website-cyber-monday.html

Serverless Computing

- Serverless computing (e.g., Function-as-a-Service) is a new cloud-computing paradigm
 - Developers write fine-grand functions and submit them to FaaS platforms
 - FaaS platforms invoke functions on-demand and bill developers according to resource usage and execution time
 - Rapid auto-scaling, pay-as-you-go billing model, no management labor



Scaling with FaaS

- FaaS provides <u>rapid-scaling</u> and <u>cost-efficient</u> computing resources for web applications to handle request bursts
 - Provide more computation resources on demand rapidly(rapid-scaling)
 - Fine-grand configuration and billing to eliminate the cost (cost-efficient)

Scaling solution	Min. Running Time	Conf. & Bill granularity	Preparation Time
Reserve resource (AWS Reserved Burstable)	1 year	GB, Years	
On-demand virtual machine (AWS On-demand EC2)	1 min	GB, Seconds	~40s
On-demand container (AWS Fargate & ECS)	1 min	GB, Seconds	~40s
FaaS (AWS Lambda)	1 ms	MB, Milliseconds	<1s

Problem: How to run existing web applications with FaaS functions

Strawman 1: Direct Execution

- Directly run existing web applications in FaaS
- Stateful applications vs. stateless functions
 - FaaS platform manages functions under the stateless assumption
 - Web applications contain complex local state like user session
 - May cause unrecoverable state loss



FaaS Platform

Strawman 2: Application Refactor

- Refactor (part of) existing application to fit FaaS functions
- Manual rewriting
 - Most code (99.6% of the jar file) are framework (e.g., spring) code
 - Tightly coupled user code and framework code
 - <u>Too complex to manually refactor code</u>

Static analysis

- Java is a highly dynamic language, especially in the web application case
- Deep invocation depth (>20), complex polymorphism (31 implementations for 1 interface), dynamically generated classes (287 for one request)
- Hard to perform static analysis

Our Solution: Offloading-based Semi-FaaS

- Semi-FaaS: automatically extract time-consuming code snippets and offload their execution to FaaS at runtime
 - Partial: keep the state at the *server*, extract and offload part of the application to FaaS (direct execution)
 - Automatic: atomically slice and run logic with FaaS (manual rewrite)
 - Dynamic: Analyze at runtime based on dynamic profiling (static analysis)



*We implemented Semi-FaaS on Java Virtual Machine (JVM), but Semi-FaaS be extended to other language runtimes



Semi-FaaS Execution



Semi-FaaS Execution



Fallback-based Offloaded Execution



Problem: Fallbacks slow down offloaded execution

Frequent fallbacks hurt performance

Native invocation

Network access

Missing code or data

Handling Native Invoca

public interface Packageable {
 public void pack();
 public void unpack();

- Web applications rely on native invocation neaving
 - For reflection, access system resources, acceleration, etc.
 - E.g., a simple request can trigger 220k+ native invocation
 - Native invocations are not offloadable since they may rely on the hidden native state (e.g., JVM-internal state, OS-related state)
- Packageable: a new interface to pack hidden native state
 - Define how to pack/unpack hidden native state
 - Modify the JDK library to implement packageable interface for specific classes

Proxy-based Connection Management

- Network operation in web applications
 - Web application depends on the network to access external services (e.g., DB)
 - A simple request requires 80+ DB access
 - Network connections are hard to migrate due coupled with OS
- Proxy-based network connection migration



Shadow Execution

- Incomplete initial closure introduces frequent fallback
 - Dynamic nature of web applications makes it hard to traverse all runnable code
 - Closure completes itself with fallbacks during execution
- Shadow execution to hide overhead during warmup



The Beehive Runtime

- A modified JVM supporting semi-FaaS execution, with
 - Offload function selection based on runtime profiling data
 - Fallback detecting and handling
 - Memory consistent among endpoints following Java Memory Model
 - Memory management among endpoints
 - Optional fault tolerance mechani
- Enables unmodified application changing their underline JVI



Experimental Setup

Environment: AWS cloud

- DB: m4.10xlarge (40 vCPUs/2.40GHz, 160GB DRAM) EC2 instance
- Server: m4.xlarge (4 vCPUs/2.30GHz, 16GB DRAM) EC2 instance

Applications

- Thumbnail: micro-benchmark making thumbnail of images
- Pybbs: open-source forum application with 24692 classes
- Springblog: open-source blog application 18493 classes

(All mentioned data are average of all three applications by default)

Experimental Setup

- Scaling methods
 - Burstable: <u>Reserved resource</u> (reserved burstable EC2 instance)
 - **EC2**: <u>On-demand VM</u> (on-demand EC2 instance)
 - **Fargate**: <u>On-demand container</u> (AWS Fargate)
 - BeehiveO: Local FaaS platform (functions running on EC2 cluster managed by OpenWhisk)
 - BeehiveL: <u>Commercial FaaS platform</u> (functions running on AWS Lambda)



Centralized server acting as dispatcher and state manager becomes the bottleneck

Vanilla & Beehive-Single: No scaling
BeehiveO: Scale with Functions running on sufficient
m4.large instances managed by OpenWhisk
BeehiveL: Scale with Functions running on sufficient
AWS Lambda with 1GB memory

Beehive atomically scales to higher throughput (9.41x (O) & 9.11x (L)) Lower throughput (worst 7.14%) with the same resource due to management overhead

*Sringblog case for better presentation, refer to our full paper for performance result of all applications

Fast Scaling

Beehive handles burst faster with acceptable overhead
 Request burst happens



11.25x(O) & 6.43x(L) faster than scaling with EC2
6.32x(O) & 3.61x(L) faster than scaling with Fargate
Acceptable tail-latency slowdown (15.0%(O) & 31.0%(L) compared with EC2)

Reach stabilized latency in **668.56ms** on AWS Lambda with **function cache** (**two orders of magnitude better**)

On-demand Cost

Beehive enjoys on-demand billing provided by FaaS



When **burst infrequently**, Beehive **costs less** compared with **reserving resources** (**3.56x(L)** at a 10% burst rate)

Beehive **always costs more** compared with other **on-demand scaling methods** due to execution overhead, while reacting to burst faster

Conclusion

- FaaS is suitable for web applications to handle request burst
 - Challenging to leverage FaaS by direct execution or code refactoring
- Semi-FaaS: Fallback-based automatic computation offload at runtime with FaaS
 - Partial, automatic, dynamic way to leverage FaaS for computation offloading
 - Packageable, network proxy, shadow execution to eliminate performance overhead caused by fallbacks
- Beehive: Runtime supporting Semi-FaaS execution model
 - Automatically scale to higher throughput
 - Faster reaction to request burst compared to on-demand scaling
 - Lower cost compared to reserving idle resources in advance



Thanks!