

A QoS-Aware Broker for Multi-Provider Serverless Applications

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Goals



My thesis is focused on "Serverless Computing".

Two main goals:

- The orchestration of a serverless workflow supporting:
 - Multiple FaaS Providers
 - Multiple Functions Implementations.
- The fulfillment of non-functional requirements concerning the quality of service (QoS) levels that should be guaranteed.



What's meant by serverless computing?

An applications service model where:

- Administration tasks (provisioning, monitoring, scaling, etc.) are directly managed by the provider.
- Small-granularity billing pricing model: pay-as-you-go.
- Cloud application are abstracted as a group of so-called "Serverless Functions".



What is a serverless function?

A computation unit implementing a business functionality.

- Stateless
- Event-Driven
- Short-Lived

Serverless Application

A software system made up of a serverless functions set.



A serverless function is executed inside a containerized environment: the so-called "Function Instance".

The FaaS platform automatically scales the number of function instances.

When a request comes in, only one of the following events can occur:

- Warm start.
- Cold start.



Concurrency Level

The limitation, imposed by FaaS platforms, on the **number** of function instance runnable at the **same time**.

Providers apply that restriction differently:

- Global (Per-Account) Concurrency Model.
 - AWS Lambda.
 - IBM Cloud Functions.
- Local (Per-Function) Concurrency Model.
 - Google Cloud Functions.



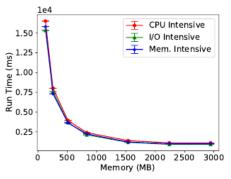


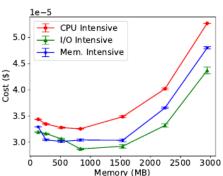
To invoke a serverless function, users have to specify a so-called serverless function configuration.

To invoke a **serverless application**, users have to specify a configuration for **all** its functions.



Configuration parameters **significantly** affect the **cost** and **response time** of serverless functions.





Serverless Computing: Problems



What current problems or limitations of the serverless computing paradigm I tried to solve?

- No support for application whose functions are hosted on multiple providers.
- No support for application whose functions have more than one implementation (Concrete Functions).
- Vendor lock-in issues.
- The fulfillment of quality of service (QoS) levels.

Serverless Computing: Problems



Why is so difficult to guarantee QoS levels?

- Lack of an analytical model to predict application performance.
- Lack of a methodological way to find a suitable configuration for a serverless application.

Related Works



Solutions already exist, but:

- They are unaware of the current status of FaaS platforms.
- No support for Multi-Provider, Multiple-Implementations serverless applications.
- Many do not support generic workload applications.

My Contributions



My contributions respect to the current state of art:

- An Analytical Model to evaluate applications performance.
 - Supports generic workflow (branch, loop, parallel)
 - Multi-Provider support.
 - Support for Multi-Implementation for functions.
 - FaaS-Status-Aware.
- A Methodological Way to find the "best" configuration to satisfy QoS constraints.
 - By solving an Optimization Problem (LP).

My Contributions



- A custom Heuristic Algorithm to resolve the aforementioned optimization problem.
 - Based on Ant Colony Optimization algorithm (ACO) family.
- A QoS-Aware Software Framework for the orchestration for multiprovider, multiple-implementations serverless applications.
- An extension to an already existent Representation Scheme to define a serverless application workflow.
 - Based on an existing language called Abstract Function Choreography Language (AFCL).

The Prototype



Main features of our software framework:

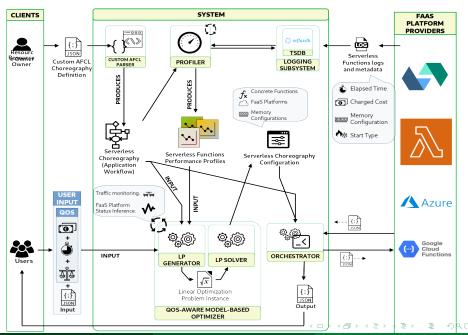
- Client-Server architecture.
- Cloud-native application.
- Includes a set of adapters to interact with following FaaS providers:
 - AWS Lambda.
 - Apache OpenWhisk.
- REST architectural style.

The Prototype



Main logical entities of our system:

- A Logging subsystem.
- An Orchestrator.
- A profiler.
- A QoS-aware model-based optimizer.
- A Custom AFCL Parser.



The Prototype



The use of an unique representation scheme for serverless workflows mitigates portability limitations and vendor lock-in issues.

Serverless application are managed using identical operations based on standard HTTP methods providing **Access Transparency**.



Serverless applications are abstracted to a Weakly Connected Weighted Directed Graph.

- Each vertex models an abstract serverless function, that is a computational unit required by business logic.
 - To each abstract function corresponds a set containing all concrete function implementing it.
- Each edge represents the calling relationship between two abstract functions.
 - Edge weights represent the so-called transition probability.



To evaluate the performance of a **Concrete Function**:

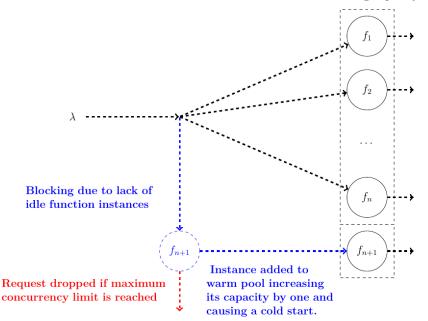
- Estimations about the Average Response Time (Cost) in case of cold (warm) start are required.
 - To compute aforementioned estimations, Exponential Moving Average based approach is adopted.
- An estimation of the probability according to which a request follows a cold start (Cold Start Probability) is required.



To compute Cold Start Probability:

- Any FaaS platform provider is modeled by set of $M/G/K(t)_{\mathbb{C}_{max}}/K(t)_{\mathbb{C}_{max}}$ queueing systems.
 - K(t): the number of function instances at time t.
 - C_{max} : the concurrency limit.
- Aforementioned set depends on the concurrency model adopted by providers.
- Erlang-B formula is used.

Warm Pool having capacity n





I formalized algorithms and equations to evaluate the performance of both a serverless application and of an abstract function.

Aforementioned algorithm depends on:

- The topological properties of the graph representing the application.
- The actual value of transition probabilities of all edges.

The Optimization Problem



- To achieve our goal consisting in finding the best configuration to guarantee QoS constraints, we have to solve an optimization problem.
 - It is based on Multi-Dimensional Multi-Choice Knapsack Problem Formulation.

The Optimization Problem



Formally, the configuration vector $\mathbf{x}_{\mathscr{C}}$ we want to find is such that:

$$\mathbf{x}_{\mathscr{C}} \stackrel{\text{def}}{=} \left\{ x_{\phi_{1}}, \dots, x_{\phi_{k}} \right\}$$

$$\in \left\{ \left\{ \bigcup_{j=1}^{|\mathbf{F}_{\phi_{1}}|} f_{\phi_{1_{j}}} \times \mathbf{M}_{f_{\phi_{1_{j}}}} \right\} \times \dots \times \left\{ \bigcup_{j=1}^{|\mathbf{F}_{\phi_{k}}|} f_{\phi_{k_{j}}} \times \mathbf{M}_{f_{\phi_{k_{j}}}} \right\} \right\}$$

$$= \left\{ \left\{ \bigcup_{j=1}^{k} f_{\phi_{i_{j}}} \times \mathbf{M}_{f_{\phi_{i_{j}}}} \right\}$$

$$\subseteq \left\{ \left\{ \mathbf{F}_{\phi_{i}} \times \mathbb{N} \right\} = \mathbf{X}_{\mathscr{C}} \right\}$$

$$(1)$$

The Optimization Problem



The number of all possible configuration grow very fast.

- An application with:
 - 6 functions.
 - 46 possible memory choice.
 - Only one implementation for each function.

lead to $9.47 \cdot 10^9$ different configurations.

- An application with:
 - 6 functions.
 - 46 possible memory choice.
 - Two implementation for each function.

lead to $6.06 \cdot 10^{11}$ different configurations.



$$max \sum_{i=1}^{|\mathcal{F}_{\mathcal{E}}(\mathcal{C})|} \sum_{j=1}^{|\mathbf{F}_{\phi_{i}} \times \mathbb{N}|} y_{\phi_{i_{j}}} \cdot p_{\phi_{i_{j}}}(t)$$

$$\text{subject to} \sum_{i=1}^{|\mathcal{F}_{\mathcal{E}}(\mathcal{C})|} \sum_{j=1}^{|\mathbf{F}_{\phi_{i}} \times \mathbb{N}|} y_{\phi_{i_{j}}} \cdot c_{\phi_{i_{j}}}(t) \leq C$$

$$\sum_{i=1}^{|\mathbf{F}_{\phi_{i}} \times \mathbb{N}|} \sum_{j=1}^{|\mathbf{F}_{\phi_{i}} \times \mathbb{N}|} y_{\phi_{i_{j}}} \cdot rt_{\phi_{i_{j}}}(t) +$$

$$+ \sum_{\phi_{i} \in \mathcal{F}_{\mathcal{E}}(\mathcal{C}) \setminus \mathcal{F}_{\mathcal{E}}(\tilde{\mathcal{P}})} \sum_{j=1}^{|\mathbf{F}_{\phi_{i}} \times \mathbb{N}|} y_{\phi_{i_{j}}} \cdot rt_{\phi_{i_{j}}}(t) \leq RT$$

$$\forall \delta_{\mathcal{C}} \in \Delta_{\mathcal{C}}$$

$$(5.12)$$

$$\sum_{i=1}^{|\mathcal{F}_{\mathcal{E}}(\mathcal{C})|} \sum_{j=1}^{|\mathbf{F}_{\phi_{i}} \times \mathbb{N}|} y_{\phi_{i_{j}}} \cdot a_{(\phi_{i_{j}}, \omega_{P}^{(l)})} \leq l - R(\mathbf{Q}_{\omega_{P}^{(l)}}, t)$$

$$\forall \omega_{P}^{(l)} \in \tilde{\mathbf{S}}_{\mathcal{C}}$$

$$\sum_{i=1}^{S_{\mathcal{C}}(0)} \sum_{j=1}^{S_{\mathcal{G}}(0)} y_{\phi_{i_j}} \cdot a_{(\phi_{i_j}, \omega_P^{(l)})} \le l - R(\mathbf{Q}_{\omega_P^{(l)}}, t)$$

$$\forall \omega_P^{(l)} \in \widetilde{\mathbf{S}}_{\mathcal{C}}$$

$$(5.13)$$

$$\sum_{j=1}^{|\mathbf{F}_{\phi_i} \times \mathbb{N}|} y_{\phi_{i_j}} = 1 \qquad \qquad \forall i \in \mathbb{N} \cap [1, |\mathcal{F}_{\mathcal{E}}|]$$

 $\forall i \in \mathbb{N} \cap [1, |\mathcal{F}_{\mathcal{E}}|]$ $\forall i \in \mathbb{N} \cap [1, |\mathbf{F}_{\phi_i} \times \mathbb{N}|]$

$$y_{\phi_{i_j}} \in \{0,1\}$$

(5.14)

The Heuristic Algorithm



I develop a custom heuristic algorithm based on Ant Colony
Optimization (ACO); it is called Pre-provisioned Colony Optimization
Algorithm with Lazy Pheromone Update.

- It is based on a set of computational agents, called artificial ants, which iteratively construct a solution.
- At each iteration, each agent moves from a solution to another, applying a series of stochastic local decisions whose policy is based on following parameters:
 - Attractiveness.
 - Pheromone Trail.



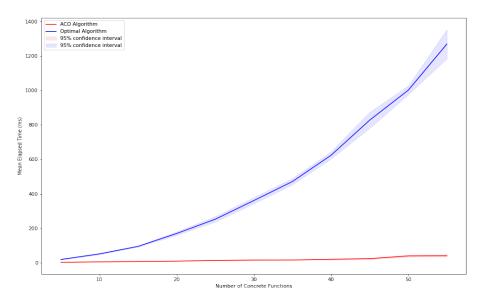
The Heuristic Algorithm

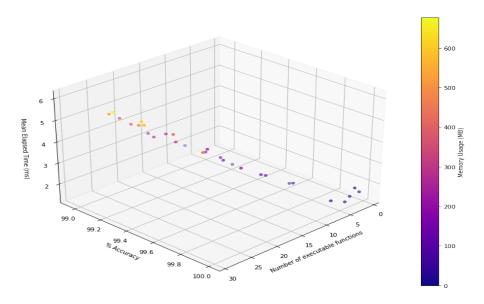


Pheromone trails are updated during every iteration.

Pheromone trails are used to decide which solutions should be preferred during **next iterations**.

- A Lazy Approach for pheromone trails update is used.
- A Pre-Provisioning Tactic is used to anticipates data needs assuring a lower latency.





Conclusions



- I presented an analytical model to compute the performance of a generic serverless workflows having multiple concrete functions hosted on multiple FaaS providers.
- I defined an optimization problems formulation to address the problem of finding a suitable configuration in order to meet user defined QoS constraints
- We developed a heuristic algorithm to rapidly solve it.
- We verified the validity of both the proposed algorithms and the analytical model through test and experimental evaluations developing a prototype supporting AWS and OpenWhisk.



Thanks for your attention! Questions?