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UNIVERSITÀ DEGLI STUDI DI ROMA

Macroarea di Ingegneria

A QoS-Aware Broker for Multi-Provider Serverless Applications

Andrea Graziani
m. 0273395

Supervisor:	Valeria Cardellini
Tutor:	Gabriele Russo Russo

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Serverless Computing: Overview

My thesis is focused on "**serverless computing**".

It is a development paradigm according to which:

- The provider takes care of all aspect of server management.
- Cloud application are abstracted as a group of so-called **serverless functions**, which are computation units implementing a business functionality.

A serverless function is executed inside a containerized environment: the so-called **function instance**.

The FaaS platform **automatically** scales the number of function instances.

FaaS platforms impose a **limit** on the number of function instance runnable at the **same time** called **concurrency limit**.

A delay is observed when a new function instance is started by the provider: this event is called **cold start**.

Serverless Computing: Overview

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

- Generally, the **amount of memory allocated** to a serverless function.

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

Serverless Computing: Problems

- Lack of support for application whose functions are hosted on **multiple providers**.
- Lack of support for serverless function implementations abstraction, that is, for the so-called **concrete functions**.
- Fulfillment of **non-functional requirements** concerning the **quality of service** (QoS) levels that should be guaranteed for **multi-provider serverless applications**.
- Fulfillment of **functional requirements** concerning the **orchestration** of multi-provider applications.

Thesis Goals

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Goal # 1

To guarantee the **satisfaction of QoS levels** for multi-provider serverless applications.

It was necessary to develop:

1. An **analytical model** to evaluate aforementioned class of applications.
2. A **methodological way** to find the "best" configuration to satisfy QoS constraints.

Goal # 2

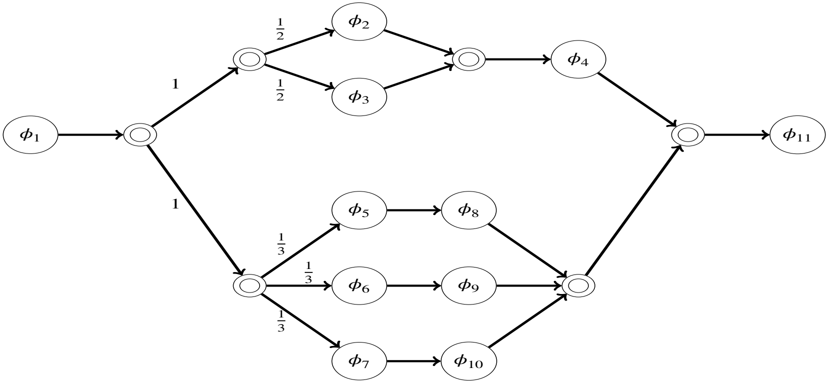
The **orchestration** for multi-provider serverless applications.

To achieve it, I had to build:

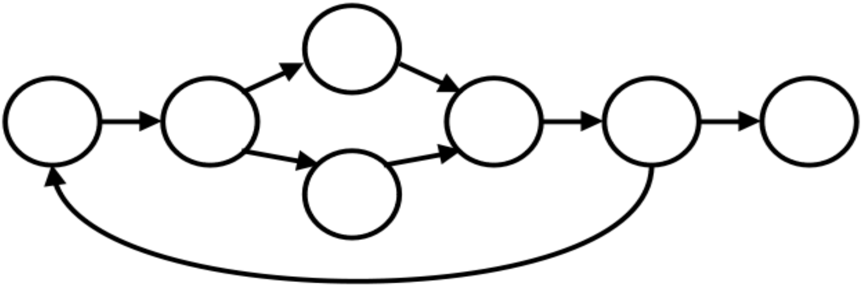
1. A **software framework**.
2. An extension to an already existent **representation scheme** to define a serverless application workflow.
   - Based on an existing language called **abstract function choreography language** (AFCL).

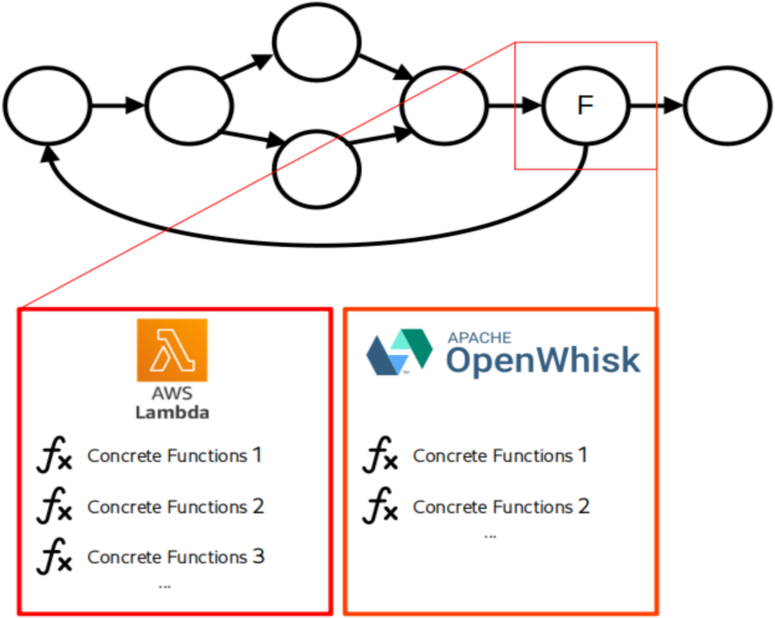
System Model

Serverless applications are abstracted to a **weakly connected weighted directed graph**.



Application Configuration

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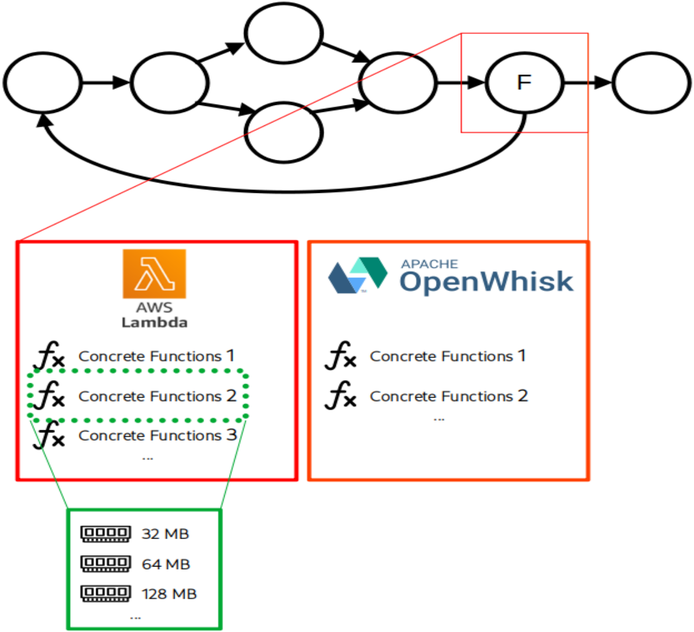
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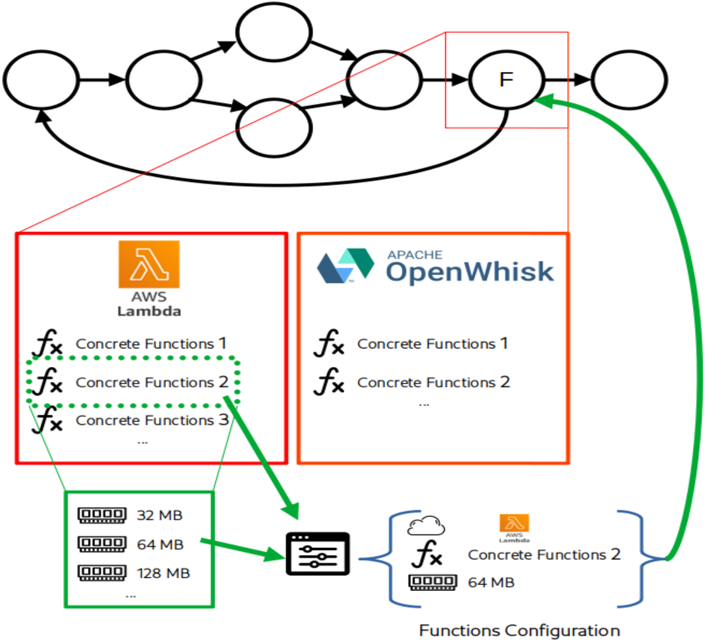
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Functions Configuration

System Model

- Estimations of **average response time** and **cost** under **any** possible configuration of all concrete functions is done using **exponential moving average** approach.
- An estimation of the **probability** according to which a request follows a cold start is required.
  - This is done using **Erlang-B** formula by modeling FaaS platform providers by **sets** of $M/G/K(t)/K(t)$ queueing systems.
    - $K(t)$: the number of function instances at time t.
- Performance estimations of the application depend on its workflow properties.

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** (MMKP).

$$\max \quad \sum_{i=1}^{|\mathcal{F}_{\mathcal{E}}(\mathcal{C})|} \sum_{j=1}^{|\mathbf{F}_{\phi_i} \times \mathbb{N}|} y_{\phi_{i_j}} \cdot p_{\phi_{i_j}}(t) \tag{5.10}$$

$$\text{subject to} \quad \sum_{i=1}^{|\mathcal{F}_{\mathcal{E}}(\mathcal{C})|} \sum_{j=1}^{|\mathbf{F}_{\phi_i} \times \mathbb{N}|} y_{\phi_{i_j}} \cdot c_{\phi_{i_j}}(t) \leq C \tag{5.11}$$

$$\sum_{\phi_i \in \mathcal{F}_{\mathcal{E}}(\mathcal{C}) \setminus \mathcal{F}_{\mathcal{E}}(\widetilde{\mathcal{P}})} \sum_{j=1}^{|\mathbf{F}_{\phi_i} \times \mathbb{N}|} y_{\phi_{i_j}} \cdot rt_{\phi_{i_j}}(t) + + \sum_{\phi_h \in \delta_{\mathcal{C}}} \sum_{j=1}^{|\mathbf{F}_{\phi_h} \times \mathbb{N}|} y_{\phi_{h_j}} \cdot rt_{\phi_{h_j}}(t) \leq RT \qquad \forall \delta_{\mathcal{C}} \in \Delta_{\mathcal{C}} \tag{5.12}$$

$$\sum_{i=1}^{|\mathcal{F}_{\mathcal{E}}(\mathcal{C})|} \sum_{j=1}^{|\mathbf{F}_{\phi_i} \times \mathbb{N}|} y_{\phi_{i_j}} \cdot a_{(\phi_{i_j}, \omega_P^{(l)})} \leq l - R(\mathbf{Q}_{\omega_P^{(l)}}, t) \qquad \forall \omega_P^{(l)} \in \widetilde{\mathbf{S}}_{\mathcal{C}} \tag{5.13}$$

$$\sum_{j=1}^{|\mathbf{F}_{\phi_i} \times \mathbb{N}|} y_{\phi_{i_j}} = 1 \qquad \forall i \in \mathbb{N} \cap [1, |\mathcal{F}_{\mathcal{E}}|] \tag{5.14}$$

$$y_{\phi_{i_j}} \in \{0, 1\} \qquad \begin{array}{r} \forall i \in \mathbb{N} \cap [1, |\mathcal{F}_{\mathcal{E}}|] \\ \forall j \in \mathbb{N} \cap [1, |\mathbf{F}_{\phi_i} \times \mathbb{N}|] \end{array}$$

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Cost Constraint

$$\sum_{\phi_i \in \mathcal{F}_{\mathcal{E}}(\mathcal{C}) \setminus \mathcal{F}_{\mathcal{E}}(\widetilde{\mathcal{P}})} \sum_{j=1}^{|\mathbf{F}_{\phi_i} \times \mathbb{N}|} y_{\phi_{i_j}} \cdot rt_{\phi_{i_j}}(t) +$$

$$+ \sum_{\phi_h \in \delta_{\mathcal{C}}} \sum_{j=1}^{|\mathbf{F}_{\phi_h} \times \mathbb{N}|} y_{\phi_{h_j}} \cdot rt_{\phi_{h_j}}(t) \leq RT \qquad \forall \delta_{\mathcal{C}} \in \Delta_{\mathcal{C}} \tag{5.12}$$

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Response Time Constraint

$$\boxed{\begin{aligned} &\sum_{\phi_i \in \mathcal{F}_{\mathcal{E}}(\mathcal{C}) \setminus \mathcal{F}_{\mathcal{E}}(\widetilde{\mathcal{P}})} \sum_{j=1}^{|\mathbf{F}_{\phi_i} \times \mathbb{N}|} y_{\phi_{i_j}} \cdot rt_{\phi_{i_j}}(t) + \\ &+ \sum_{\phi_h \in \delta_{\mathcal{C}}} \sum_{j=1}^{|\mathbf{F}_{\phi_h} \times \mathbb{N}|} y_{\phi_{h_j}} \cdot rt_{\phi_{h_j}}(t) \leq RT \qquad \forall \delta_{\mathcal{C}} \in \Delta_{\mathcal{C}} \end{aligned}} \tag{5.12}$$

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Capacity Constraint

$$\sum_{j=1}^{|\mathbf{F}_{\phi_i} \times \mathbb{N}|} y_{\phi_{i_j}} = 1 \qquad \forall i \in \mathbb{N} \cap [1, |\mathcal{F}_{\mathcal{E}}|] \tag{5.14}$$

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Heuristic Algorithm

I develop a custom heuristic algorithm based on **ant colony optimization** (ACO).

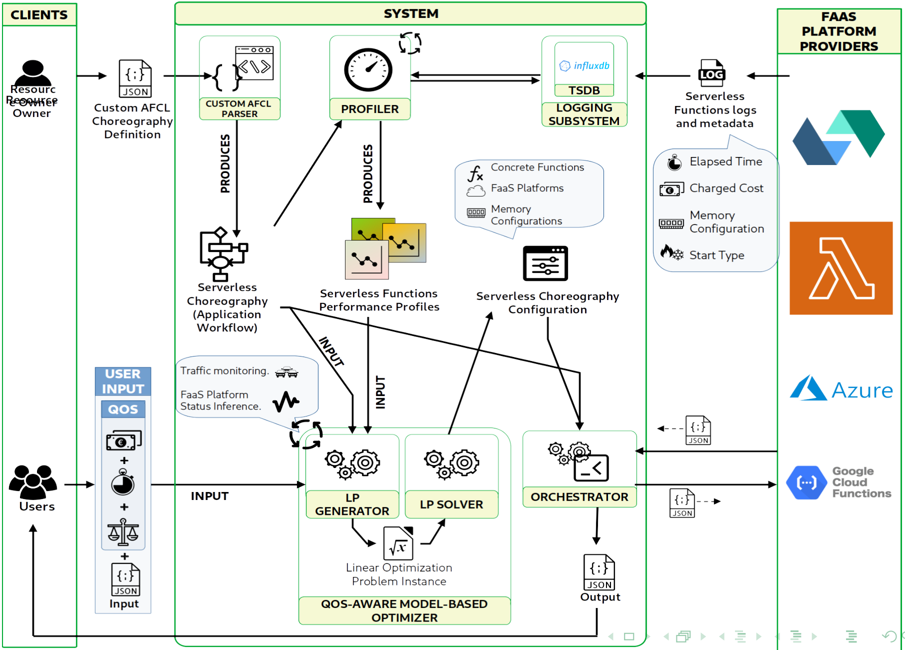
- It is based on a set of computational agents, called **artificial ants**, which **iteratively** construct a solution.
- At each iteration, each agent moves from a solution to another, applying a series of stochastic **local** decisions whose policy is based on following parameters:
  - **Attractiveness**.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

Pheromone trails are updated during **every iteration**.

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

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.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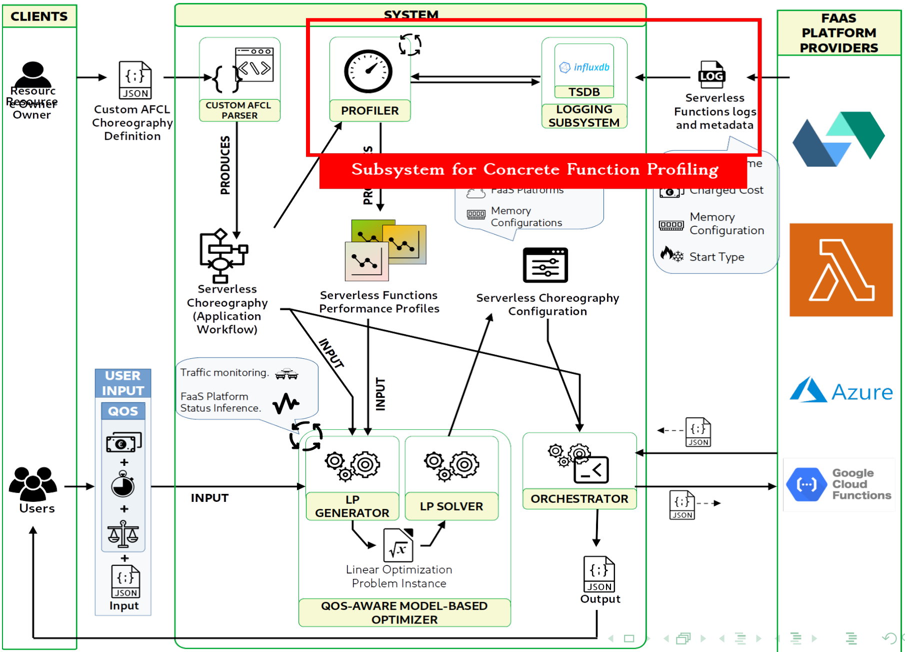
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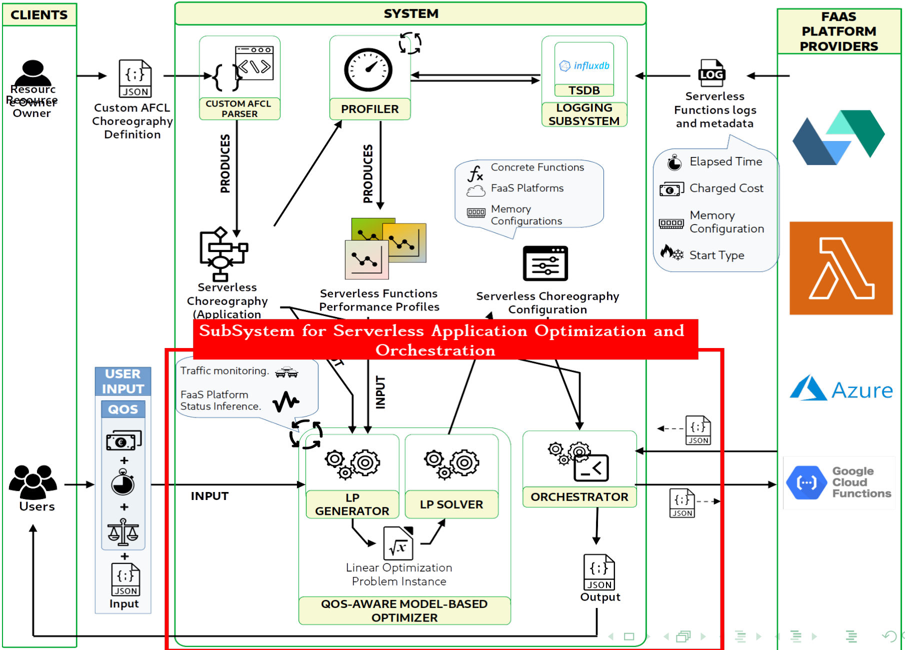
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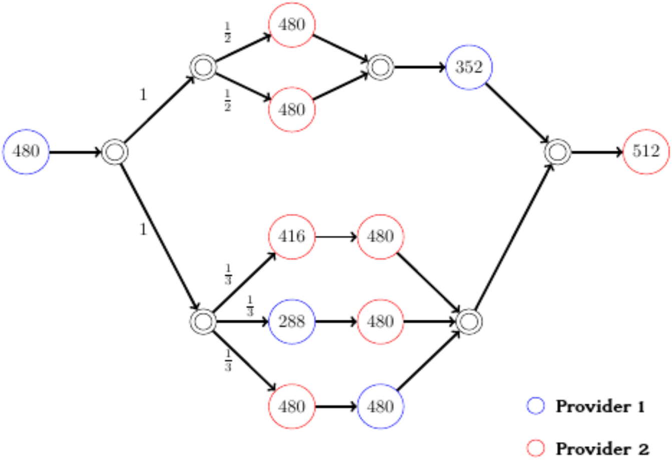
Validation Test

We validate our model through several experiments using an image-processing serverless application.

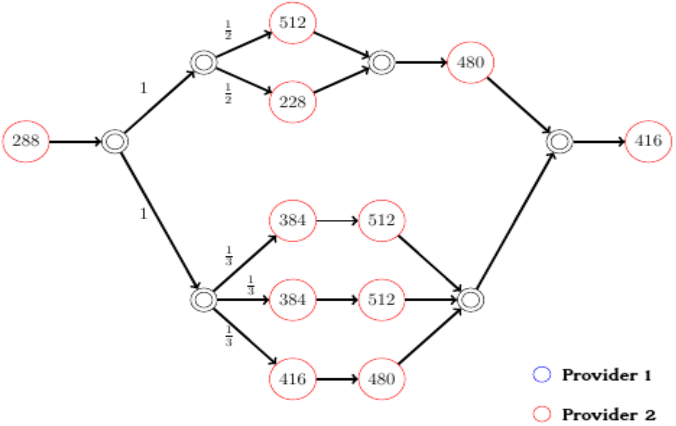
- Firstly, we check the respect of user specified QoS constraints in a static way thorough several **sequential invocations**.
- Then, we test the model in a dynamically way thorough **several concurrent and parallel invocations**.
  - That experiment aimed to **run out of capacity** on one FaaS provider and so force our prototype to schedule the concrete function on the other.

- Application configuration produced by our system **before** Provider 1 runs out of capacity...

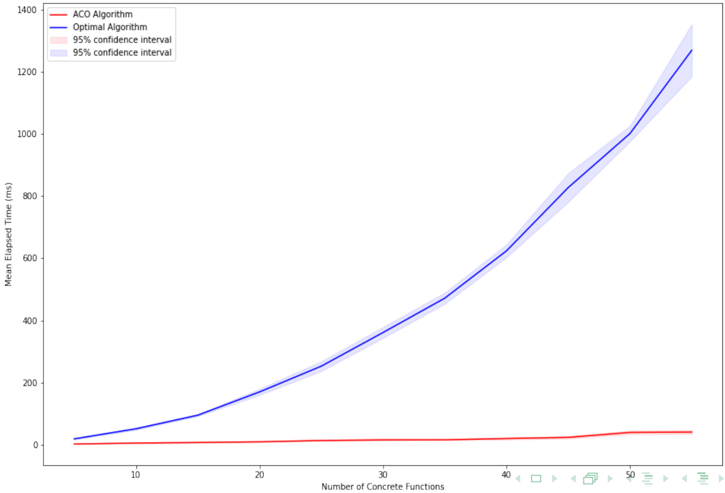


Validation Test

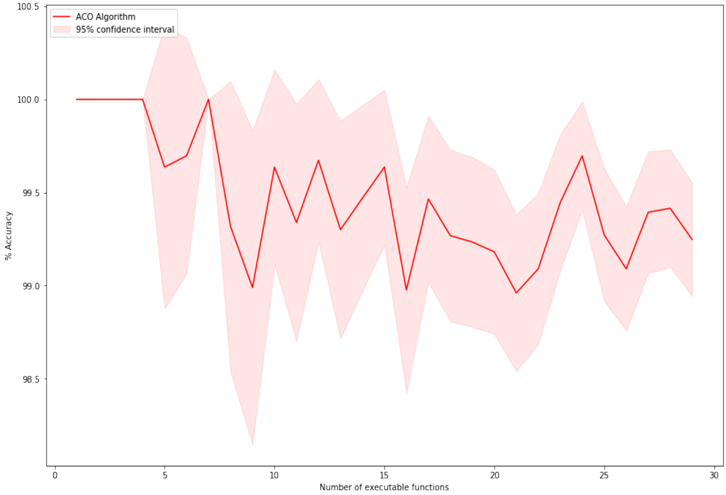
- …and **after**, when Provider 1 runs out of capacity.



- We compared **execution time** of our heuristic algorithm with that of optimal algorithm through several experiments.



- According to my results, **accuracy** is above **98%**.



Conclusions

1. I presented an analytical model to evaluate the performance of multi-provider serverless application.
2. I defined an optimization problem formulation to address the problem of finding a suitable configuration in order to meet user defined QoS constraints.
3. I developed a heuristic algorithm to rapidly solve it.
4. I validate proposed solution through test and experimental evaluations using my prototype.

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Thanks for your attention!

Questions?