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„Service meshes as a tool for resilient microservice
deployment on example of Istio“

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Contents

1	Introduction	3
2	Core concepts	5
2.1	Containers	5
2.2	Container Orchestration	6
2.3	Service mesh	7
2.4	Istio	8
2.5	Resiliency	10
3	Project overview and setup	11
3.1	Microservices	11
3.2	Docker & Kubernetes	12
3.3	Setting up Kubernetes & Istio	12
3.4	Deploying microservices	14
3.5	Deploying custom ingress gateway	14
4	Case study	16
4.1	Load balancing	16
4.1.1	Round Robin	17
4.1.2	Random	18
4.2	Rate limiting	18
4.2.1	Global	19
4.2.2	Local	20
4.3	Canary Deployment	22
4.4	Circuit Breaking	24
5	Summary	27
6	Conclusion	28
7	Future work	29

Abstract

The purpose of this paper is to illustrate the way service mesh and its features can be applied and used to achieve resiliency in microservice-based systems on the example of the open-source project Istio. We will learn when service meshes turned relevant and why it happened, then conduct a brief review of the history of microservice architecture up to the modern days. Next, the problems that service meshes can solve with a focus on resiliency will be analyzed, and we will see how these problems were solved before the appearance of Istio.

Then, on the example of a simple distributed microservice app, such features as load-balancing, rate-limiting, circuit-breaking will be configured using Istio. We will discuss the role of these features in solving modern-day resiliency issues. Finally, we will dive into deeper detail and see how different policies work, learn how to configure global and local rate-limit, set up Istio to run canary deployments and circuit-breakers, and observe how the system behaves under these configs. This will be achieved by using specialized monitoring tools and logs.

In the end, a short progress report will be conducted. There, we will review the job done and shortly discuss the future of service mesh technology.

1 Introduction

Distributed microservice applications have become part of software development for quite a long time. With the lapse of time, it slowly replaced the 3-tier architecture for large-scale systems like social networks, banking, music streaming, etc. And it makes sense. Mentioned systems tend to scale not only in user-base, but also in the amount and quality of data these users consume. Have Facebook or Spotify stood monolithic, it would be a monstrous abomination of software and would require a way larger number of people to maintain it. Numerous Facebook or Spotify services have been developed by third parties and are used via APIs that interact and share data with core data centers. Nonetheless, the monolithic approach is not completely gone. Some applications are still monolithic (e.g. some isolated corporate backends).

However, in general, the monolithic approach became outdated with the development of cloud technologies. [1]

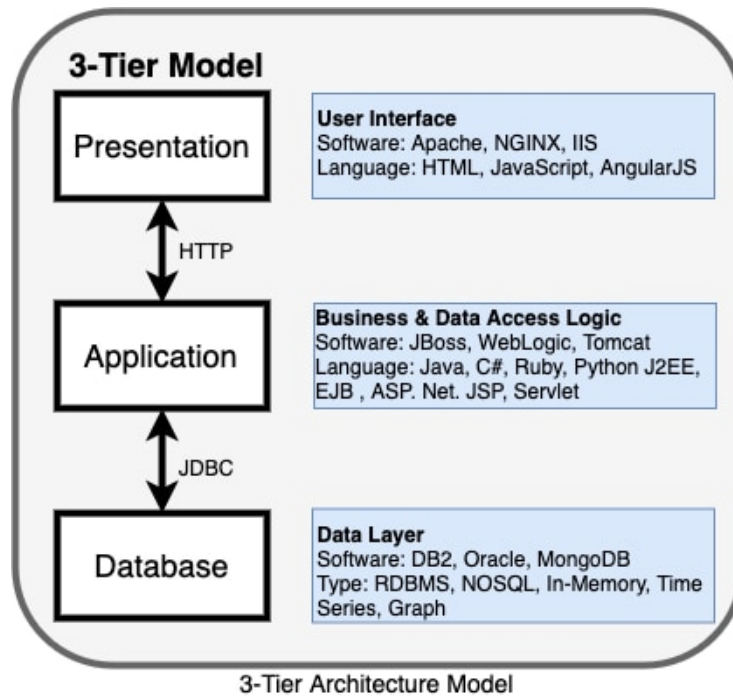


Figure 1: 3-tier monolithic app structure [1]

As seen in the figure above, the 3-tier model heavily relies on its tiers. It can only scale vertically until a certain point. When scaling reaches its critical mass, the system quickly becomes overwhelmed and hence, harder to maintain. Each tier has to be considered.

On the other hand, a microservice-based system relies less on vertical scaling. Its main strength is the ability to scale up horizontally and the fact that services are decoupled from each other and can communicate via APIs, event streams, or brokers in the cloud. This independence makes it possible to develop each service on the tech stack that is best suitable for the business logic that the service is responsible for, vastly increasing freedom of technology choice, and, as a consequence, making debugging and deployment way easier and faster. [23].

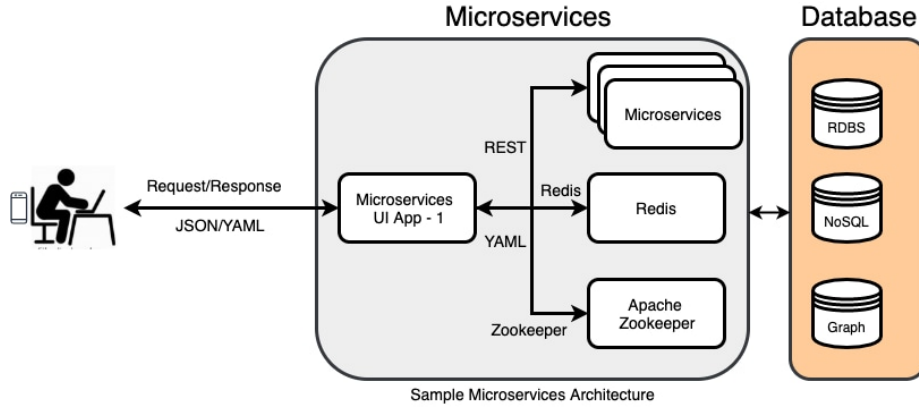


Figure 2: Microservice example [1]

One of the most significant sacrifices is performance though. Each tier of the monolithic app is located on the same piece of hardware, which makes data exchange extremely fast. In the case of microservices though, communication over API or brokers means an extra level of complexity for data exchange. Additionally, it becomes more and more challenging to manage a microservice system as it grows. This growing complexity requires a new management approach. This approach is known as DevOps. According to Amazon: "*DevOps is the combination of cultural philosophies, practices, and tools that increases an organization's ability to deliver applications and services at high velocity: evolving and improving products at a faster pace than organizations using traditional software development and infrastructure management processes.* [5]".

However, the scope of our interest lies not in the DevOps itself, but in the tools that DevOps teams use to make this philosophy work. Namely, containerization and container orchestration. And to be even more specific Docker and Kubernetes. These two technologies are the solid foundation of this paper and in order to understand the core idea of the paper's topic we first need to briefly discuss them in the section below..

2 Core concepts

2.1 Containers

Let's use another direct quote, this time from Citrix: "*Containerization is a form of virtualization where applications run in isolated user spaces, called containers, while using the same shared operating system (OS). A container is essentially a fully packaged and portable computing environment.* [3]"

The backbone of container is a virtual machine (VM). In other words we run a software that starts the VM with the app of our need anywhere we need, without caring about version discrepancy. It really doesn't matter. Containerized app can be run literally anywhere on a cloud or another computer.

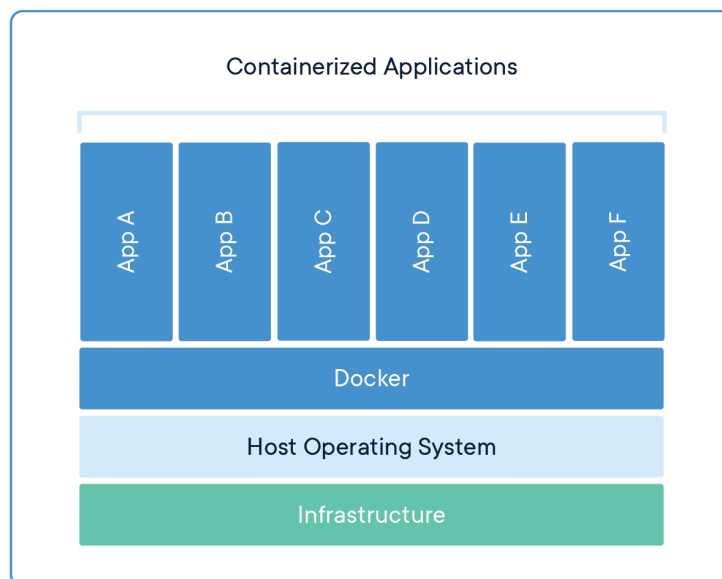


Figure 3: Structure of docker based containerized apps [6]

These containers are lightweight and have everything that our app needs to run: dependencies, runtime libraries, programming language versions, etc. This is great for abstracting our app for further easy deployment in the cloud environment of our choice. This makes containers best friends of microservices [10]. Most of the modern large scale applications use exactly this approach. For the below case study industry standard docker will be used.

However, as time showed, docker alone was not enough. Although it was a huge breakthrough back in time, containers very quickly reached the limit where managing them became a tedious task. Another concept was required to efficiently handle containers in the cloud environment.

2.2 Container Orchestration

One of the biggest cons of cloud-based systems is their complexity. The bigger the system grows, the harder it is to observe its individual elements and manage their lifecycles. The following aspects should be considered: resource allocation, container availability, container health, load balancing, traffic routing, and many other things. [18]. Doing it manually can be very demanding and time-consuming. Hence, the concept of container orchestration was introduced.

The main purpose of container orchestration is the automation of the above-mentioned tasks and providing tools for logging and debugging. To achieve this, special automation tools are required. One of such tools is an open-source container orchestration tool Kubernetes [13]. It is a framework that gives another higher level of abstraction and control over the system. It allows to dynamically scale apps by adding, removing containers upon the need to load balance them, and health checks them to timely prevent faults.

Kubernetes cluster is where things happen. It is a set of nodes where our containers are running. The backbone of the entire system. From the clusters control pane we have access and can manage system state, schedule container deployments across groups of machines, manage traffic, etc. [2]

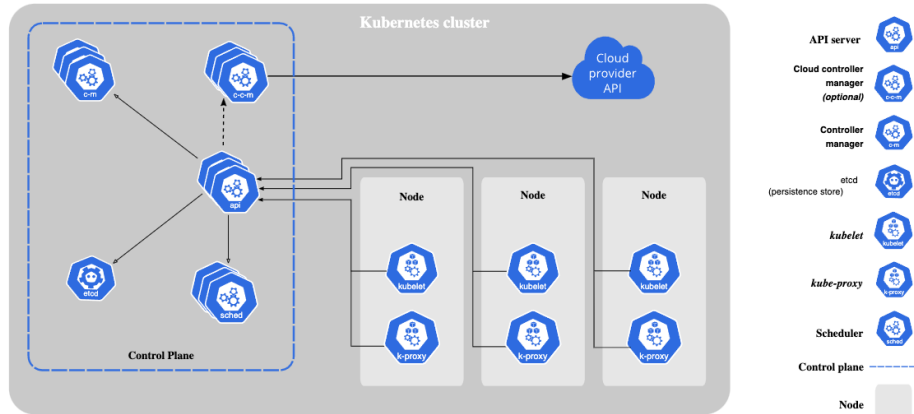


Figure 4: Kubernetes components [15]

However, things can get big even with orchestration. Kubernetes provides us with many tools for deployment automation. However, traffic management and security are still necessary to have some means for observing individual services, their versions, or even further fine-tune traffic distribution and security. These features are usually implemented on an individual service basis [17]. But what if our system has a decent amount of microservices that use different technologies? We will be having a hard time adapting each microservice to observe and control the mentioned data. Not only it is necessary to adapt each MS individually, some way to collect and process this data from each of them is also required. Lots of room for errors. But luckily, there is a solution to this situation.

2.3 Service mesh

What is a service mesh? To answer this question, let's first try to assume a typical situation, that leads to the migration of the system to a service mesh.

Imagine some distributed microservice app (for example social network) that is orchestrated by Kubernetes. With a lapse of time, its user base will grow, and it'll need to start being scaled. At this stage, just Kubernetes should be fine to solve scaling issues. But now assume that new features are constantly being added, and they require new microservices and even more scaling as the user base grows even bigger. It is still possible to manage the situation by just using Kubernetes features. However, at some point, the app becomes extremely popular, and we found ourselves in a situation where hundreds of thousands of replicas of different microservices that are scattered across numerous clusters that interact and share data must be observed, load-balanced, further, recovered from failures, etc. A small misconfiguration can lead to many wasted hours of debugging and even worse, could lead to services' downtime that can last longer than acceptable, which, considering the size huge of our application, can cost us some millions if not tenths of millions of outgoings. It could still be acceptable if our profit margin is high, but let's just make a rough simple, and approximate calculation.

Let's say that the app brings us 100 million per year and the maintenance budget is 10 million (10%). For now, pin this information in our heads and return to scaling our app, namely to the point where we mentioned wasted hours for debugging and downtimes. As the system gets more complex, the more(complex) Kubernetes configs will be required to handle possible failures and faults. Also, gathering metrics for each microservice will become a mess. It will be required to modify each microservice separately with additional code to retrieve the required data. And with 100% probability, these microservices are not written on just one programming language, since each MS is designed for its specific purpose on a tech stack that is suitable for it the most. Already feels daunting, isn't it? And now imagine debugging all of this clutter. It's possible with just Kubernetes, but hard. And complexity will grow exponentially as the app scales. Now, with all, we have from above, let's unpin the information about maintenance cost. All that was described above cost us 10 million per year. Can the costs be reduced? Yes. How? Service mesh. So let's answer the above question now.

As already mentioned, our system is a collection of distributed microservice apps, where each MS is responsible for some business rule that is relevant for the entire system. A service mesh is an additional infrastructure layer that can be added to our services. This layer allows us to transparently add capabilities like observability, traffic management, and security, without adding them to our code. The term "service mesh" describes both the type of software that is used to implement this pattern and the security or network domain that is created when that software is used. [11]. Think of it in that way - if a person is a microservice (since we all have some purpose, interact and share data), then our gadgets are service meshes in some way:

1. **Observability:** track our vital signs (e.g. fitness trackers & smart watches)
2. **Traffic management and load balancing:** based on our cars' location traffic lights can be adjusted accordingly to reduce car load on the streets.
3. **Security:** smoke detectors in our houses can call fire brigades in case of fire.

Same with the our microservices and service meshes. Each MS will get a "sidecar" or a proxy that transparently adds above mentioned capabilities. This gives us level of abstraction that is required to reduce the complexity of maintenance routines and drastically save us extra expenses. Now, just few config files are required to affect more than one service with additional capabilities without modifying each of them separately. Huge time and money saver. Since we have basic understanding of what a service mesh is let's see what's let's have a more closer look at Istio.

2.4 Istio

Istio is an open-source service mesh project based on EnvoyProxy. It creates an additional layer with proxies (sidecars) that are attached to any desired service and allows it to utilize the features mentioned in the section above. To understand how istio works, let us first briefly see what role envoy plays in it.

According to Envoy's documentation: "*Envoy is a high-performance C++ distributed proxy designed for single services and applications, as well as a communication bus and "universal data plane" designed for large microservice "service mesh" architectures. [7]*". So envoy gives us functionality to attach individual proxies to our microservices and provides means to organise these proxies in a mesh. In other words, envoy is the skeleton of Istio.

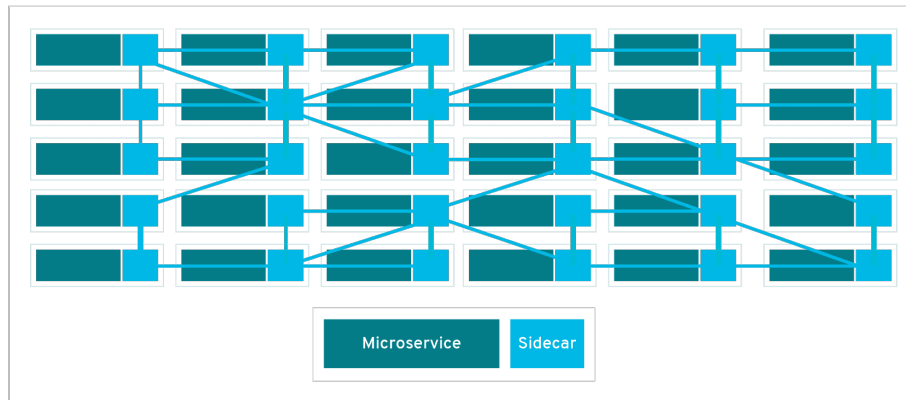


Figure 5: This is how proxies could look like when deployed [16]

Istio has two main components [12]:

1. **Data plane:** is responsible for communication between services. This is basically our mesh deployed via envoy proxies. It makes our network understand what traffic is flowing through it.
2. **Control plane:** is responsible for reading configs and "explaining" them to our network.

This is how a microservice looks before using istio

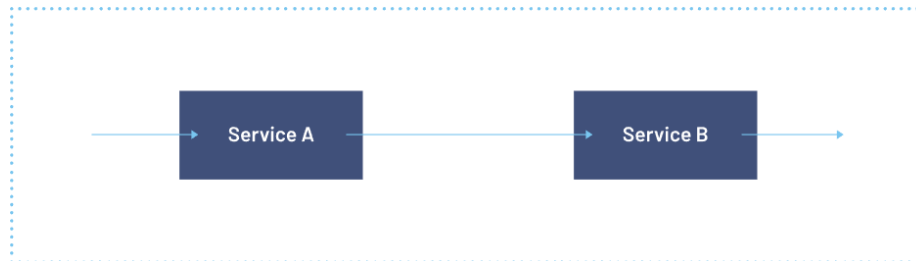


Figure 6: before istio [11]

and after:

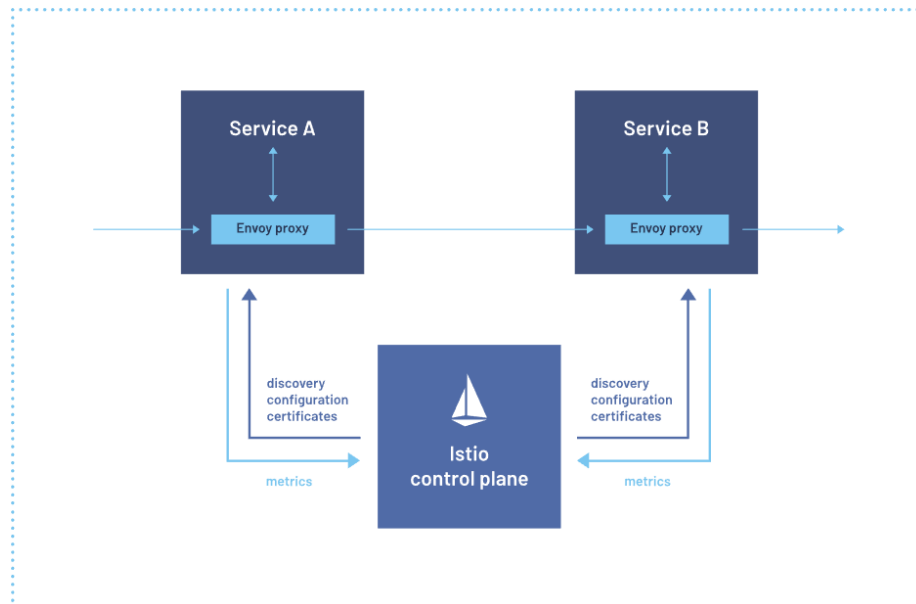


Figure 7: after istio [11]

Now that we have an understanding of what service meshes are, let's see how resiliency by utilizing Istios' features can be achieved.

2.5 Resiliency

Every system is prone to errors and faults. There is no guarantee that our app will be 100% up. As was mentioned in previous sections, the microservice approach means that there are more things to pay attention to and these things have to work in unison with one another. The more areas that need to be controlled, the more places appear where something can go wrong. And there is no way to consider absolutely everything. With so many dependencies and intersections, it is only a matter of time until something will go wrong. But it doesn't mean that nothing can be done about it. Yes, it's not possible to think through every single problem, but we can utilise mechanisms that let us observe services and see when and where something went out of hand to fix it timely. The goal is to maintain system stability at the highest possible level, as every second of downtime costs money. As was already noted, every single failure can't be prevented. But it can be assured that after every failure, the system is up and running as fast as possible. This is called resiliency. To call the system resilient, strong failure detecting and failure recovery mechanisms are required. Istio provides some strategies that can help to maintain the necessary amount of resiliency [20]:

1. Load balancing
2. Rate limiting
3. Circuit breaking
4. Timeouts and retries

In the scope of this case study, the focus will be on the first three and on reviewing resilient version deployment via a canary approach.

Monitoring tools are a great addition to the traffic observability of our microservices. By utilising them, we can visualise traffic flow and see where problems occur. Grafana, Kiali, Prometheus, Prometheus and Jaeger are great tools that allow us to observe data.

1. **Prometheus** is monitoring system and time series database for gathering metrics
2. **Grafana** is graphical representation of Prometheus data. It has customizable dashboards.
3. **Kiali** gives us the view of the data flow over our entire mesh.
4. **Jaeger** for distributed tracing

Without further ado let us go through our project setup and studying Istio features.

3 Project overview and setup

In the sections below, we will go step by step over our test project, its setup, and how to utilize Istio configs to simulate real environment situations on a smaller scale. This simplicity will allow us to get a general idea of how Istio features work, what they do, and how they affect the entire system. Each case will be analyzed and compared to real situations in big environments. Each config will be discussed, and it will be reviewed how they influence the system with the help of the above-mentioned monitoring tools. In the end, we will sum everything up and discuss how these Istio techniques can be used in conjunction with other techniques that were not covered in detail. So let's start.

3.1 Microservices

Our case study will be based on a small internet of things (IoT) app based on processing weather data of 2 types: air temperature and weather conditions (sunny, cloudy, etc.). The app consists of 3 main microservices that are responsible for the main data flow. There are 2 more microservices (frontend and CLI), but for the scope of our research, the main three will do. Main microservices are:

1. **Weather simulation (WS) service** is responsible for collecting (read simulating) weather data. Has two modes: air temperature and weather condition.
2. **Message Queue (MQ)** serves as an intermediary between IoT and other services. Other services can subscribe to it to fetch messages relevant to it as soon as these messages arrive in the MQ.
3. **Machine Learning (ML)** MS is subscribed for weather messages in MQ. It processes weather data and returns weather predictions. Can also detect weather anomalies.
4. **GUI app** (irrelevant) is subscribed to all types of messages and serves as a weather monitoring interface for weather station staff.
5. **CLI app** (irrelevant) accepts commands for MQ debugging.

All microservices are Spring Boot Java applications. The program is just a simulation of real-world weather sensors, which means that all data is randomly generated and used only as a foundation for our Istio playground. Machine learning predictions are also based on simple average calculation and random percentages for anomalies. The message queue is a simplified broker service that serves as binding between ML and WS.

3.2 Docker & Kubernetes

Each MS is compiled into a JAR file and wrapped in a separate docker container. For Docker environment simple *openjdk:11* will be used. Docker files will look as simple as:

```
FROM openjdk:11

WORKDIR /usr/app/iot

COPY ./ ./
CMD ["java", "-jar", "simulation.jar"]
```

Also, each MS is deployed to my personal Docker hub for a simpler Kubernetes configuration. For the local Kubernetes setup, Minikube is used. It has everything that is required for deploying our app and testing Istio. I am using *hyperkit* driver since I had problems setting up Istio with macOS native *darwin64* driver. According to my research, this issue is known and is currently being fixed.

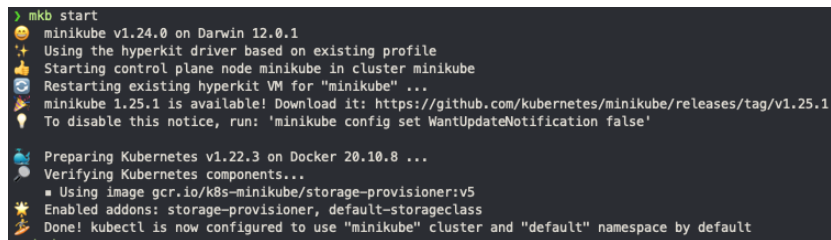
3.3 Setting up Kubernetes & Istio

The project can run on any system (Linux, MacOS, Windows) with recommended minimum of 8 GB RAM. Software requirements:

- Docker (Version for local my setup v20.10.12)
- Minikube (Version for my local setup v1.24.0)
- Postman (Version for my local setup v9.13.2) or curl

After above requirements are met we can proceed with the initial setup:

1. Open the terminal and run `$ minikube start` command. This will start Minikube with default driver for your system and create docker image where our Kubernetes will reside. Successful start should show something like this.



```
> mkb start
minikube v1.24.0 on Darwin 12.0.1
Using the hyperkit driver based on existing profile
Starting control plane node minikube in cluster minikube
Restarting existing hyperkit VM for "minikube" ...
minikube 1.25.1 is available! Download it: https://github.com/kubernetes/minikube/releases/tag/v1.25.1
To disable this notice, run: 'minikube config set WantUpdateNotification false'

Preparing Kubernetes v1.22.3 on Docker 20.10.8 ...
Verifying Kubernetes components...
  ■ Using image gcr.io/k8s-minikube/storage-provisioner:v5
  ■ Enabled addons: storage-provisioner, default-storageclass
Done! kubectll is now configured to use "minikube" cluster and "default" namespace by default
```

Figure 8: Successful *minikube start*. (*mkb* is a custom alias for minikube)

To make sure that Kubernetes is up and running execute `$ kubectl version` command. You should see something like that:

```
Client Version: version.Info{Major:"1", Minor:"22",
GitVersion:"v1.22.4", GitCommit:"b695d79d4f967c403a96986
f1750a35eb75e75f1", GitTreeState:"clean", BuildDate:
"2021-11-17T15:41:42Z", GoVersion:"go1.16.10", Compiler:
"gc", Platform:"darwin/amd64"}
Server Version: ...
```

2. Now let's install Istio using projects' getting started guide.¹ This will guide us through downloading and configuring Istio demo setup on the system. For now, the guide needs to be followed until part 2 of *Install Istio* section. In this part namespace label is being added, to instruct Istio to automatically inject Envoy sidecar proxies when we later deploy our application by executing command:

```
$ kubectl label namespace default istio-injection=enabled
```

Without it, Envoy sidecar proxies for each service must be added individually.

3. Lastly, we need to install monitoring tools. To do so, navigate to istio folder and run: `$ kubectl apply -f samples/addons`. This will deploy Kiali, Prometheus, Grafana and some other helpful tools to our cluster.

Now let us make sure that everything was installed correctly by running `$ kubectl get all -A`. In deployment section under `istio-system` we should see our monitoring tools listed.

NAMESPACE	NAME	READY ..
...		
istio-system	deployment.apps/grafana	1/1
istio-system	deployment.apps/istio-egressgateway	1/1
istio-system	deployment.apps/istiod	1/1
istio-system	deployment.apps/jaeger	1/1
istio-system	deployment.apps/kiali	1/1
istio-system	deployment.apps/prometheus	1/1
...		

¹<https://istio.io/latest/docs/setup/getting-started/>

3.4 Deploying microservices

At this point, we have our microservices ready and containerized, and our Kubernetes cluster up and running. It is time to deploy our weather simulation IOT app. To do so, we need to apply our Kubernetes config files. Config files are similar for all three microservices and contain settings for Service and Deployment, where service serves as a network gate for our pods. Deployment config will fetch a .jar file from the Docker hub, which will start the Spring boot app within our cluster. By default, we will have only one replica of each pod. We will increase the number of replicas further in the load balancing section. To deploy microservices, perform the following steps:

1. Clone the project from repository ²
2. Navigate to `../configs/IOT/` in terminal
3. Execute `$ kubectl apply -f .` (including dot)

This will deploy all 3 microservices to our cluster. Deployment may take some time. Execute `$ kubectl get pods` to make sure that everything was deployed correctly. The output should look like that:

NAME	READY	STATUS	...
machine-learning-v1-5db8b8b6b-lwpcj	2/2	Running	
machine-learning-v2-6468565977-5lx69	2/2	Running	...
message-queue-5d848945c8-wz2vd	2/2	Running	
simulation-7ff94c55cb-2vmxq	2/2	Running	...

As can be seen on the screenshot, machine-learning has two instances, v1 and v2. This is done on purpose to demonstrate canary deployment in one of the future sections. Also, note 2/2 status in the READY column by every pod. This indicates that our pod has something attached to them, namely our Envoy proxy sidecar. If we deployed these configs without installing Istio the status would be 1/1. This means that we did everything correctly. Alongside service logs, we now have the opportunity to view sidecar logs by just adding `istio-proxy` to our standard log command, but more on this in further sections.

3.5 Deploying custom ingress gateway

By default, Kubernetes Ingress is responsible for incoming traffic. Istio, however, adds up to this concept and provides extensive customization and flexibility to further fine-tune how we want our network traffic to behave. This concept is called Istio Gateway [9]. By default, Istio deploys `istio-ingressgateway` which is the global entrance to our mesh of sidecars. We can configure mesh-wide rules and policies for incoming traffic. We can restrict URLs, limit incoming connections, etc. However, we can also deploy our custom gateways for specific services

²<https://github.com/watty888/bachelor-project>

or groups of services. For our case, we will define such a gateway to allow traffic only to the specific URL of a specific service. To do so, we will add a virtual service that specifies the URL and path to the service in our cluster. This gateway will also help us to analyze Isitos' rate-limiting capabilities. To apply gateway and virtual service, navigate to `../configs` directory and execute `$ kubectl apply -f gateway.yml`.

Now, to make an API call to the simulation service, we don't need to access its port anymore. Instead, we need to do it through the ingress gateway port and then specify the URL. To access the ingress port, execute:

```
$ (kubectl -n istio-system get service istio-ingressgateway
-o\newline jsonpath='{.spec.ports[?(@.name=="http2")].nodePort}')
```

Output of this command is the required port. Now let's make first API call. **Host** is cluster ip. To obtain it run `$ minikube ip`. **Port** is the ingress port obtained earlier above. **Endpoint** is

`api/v1/performSimulation?duration=60000&type=weather&threads=1`.

Time to make a POST call in Postman. To test if everything works, change the duration to a few milliseconds to ensure that the request goes through (200 OK) and the endpoint is reachable. Then change it back to 1 minute and run the call again. While the request is running, let's start the Kiali dashboard:

1. In terminal navigate to istio folder
2. Execute `$ istioctl dashboard kiali` and navigate to Graph

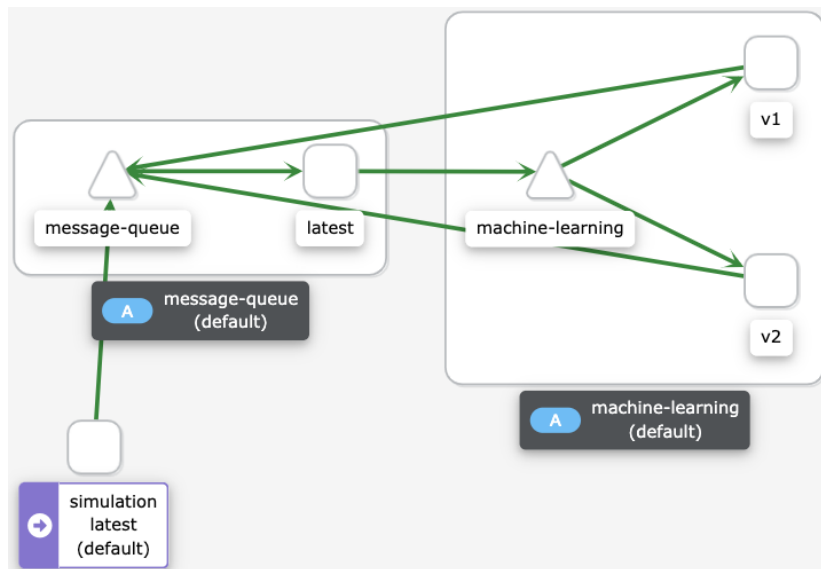


Figure 9: If you see this, then everything is configured correctly

4 Case study

Now that the system is configured, we can proceed with pure Istio stuff. In this section, we will apply different Istio configurations and see how they affect the system. Many of the features that are going to be tested, as was mentioned in the introduction section of the paper, are also achievable with pure Kubernetes. However, as it will soon become clear, service meshes didn't appear out of anywhere. It happened purely because of the limitation that Kubernetes has when it comes to monitoring microservices, gathering metrics, and fine-tuning some security aspects.

We will configure and study following features one by one:

- Load balancing
- Rate limiting
- Circuit breaking
- Canary deployment

Without further ado, let us proceed.

4.1 Load balancing

Load balancing is one of the key issues of microservice-based architectures. The need for it appears as the system scales. We require it to properly distribute traffic between our nodes in Kubernetes. But what if we need to manage traffic between microservices.

Istio utilizes Envoy's load balancing mechanism based on Destination Rules. Destination rules work in conjunction with virtual services. Let's cite the official Istio docu on these: " *You can think of virtual services as of how you route your traffic to a given destination, and then you use destination rules to configure what happens to traffic for that destination.*" [21]. That is exactly what we require now, except that virtual service is not required for this demo. Default ingress VS will be used automatically.

Let's see what we will be testing in the first place. The POST request to the simulation service will spam our message queue with messages with corresponding labels. This will trigger the machine-learning service to collect these messages. We are going to load balance this message collection. Namely, we will scale our machine learning service so that more than one services subscribe to these messages and see how the traffic will be organized among replicas.

Note that there are two versions of Machine Learning service v1 and v2. Let's scale v1 to 3 replicas and forget about v2 for now:

```
$ kubectl scale deployments/machine-learning-v1 --replicas=3
```

We should now see 5 pods of v1 if we run `kubectl get pods`.

NAME	READY	STATUS	...
machine-learning-v1-5db8b8b6b-f6q7f	2/2	Running	
machine-learning-v1-5db8b8b6b-k75m7	2/2	Running	
machine-learning-v1-5db8b8b6b-qrmzh	2/2	Running	...
machine-learning-v2-6468565977-ppx54	2/2	Running	
message-queue-5d848945c8-sq2jf	2/2	Running	
simulation-7ff94c55cb-ptdf2	2/2	Running	...

We are going to apply the destination rule to v1, which defines simple load balancing policies. By default, there are three policies, but custom ones can also be set up for more specific cases. For now, we only need to concentrate on the standard ones, which are:

- ROUND_ROBIN
- RANDOM

To test the policies, let us first open 3 parallel terminal tabs and run tracing logs for each of the v1 replicas sidecar proxies:

```
$ kubectl logs --follow machine-learning-v1-5db8b8b6b-<pod-id>
istio-proxy
```

Thereafter, let's load Machine Learning service by starting simulation for one minute in postman:

```
POST http://<minikube-ip>:<ingress-port>/api/v1/performSimulation?duration=60000&type=weather&threads=1
```

4.1.1 Round Robin

Round-robin is the default load balancing policy in Istio. This means that we don't need to configure anything for it. It is already there.

Now let's observe. We will notice that data started to flow as soon as we switch to our terminal tabs after sending a POST request. What also is important, is that each tab receives connections in consecutive order. This means that our round-robin is working.

```

% k (kubectl)
[2022-02-17T20:16:14.841Z] "POST /api/v1/events HTTP/1.1" 200 - via_upstream - "-" 117 0 11 10 "-" "Java/11.0.13" "c5dbd4d6-3047-995c-a8cb-3372fee4cd95" "message-queue:3001" ""
228.10:3001 172.17.0.14:44476 - default
[]

% k (kubectl)
[2022-02-17T20:16:14.625Z] "POST /api/v1/events HTTP/1.1" 200 - via_upstream - "-" 118 0 102 102 "-" "Java/11.0.13" "90029553-0d64-935b-b0e7-dffcccefb89" "message-queue:3001" ""
1.228.10:3001 172.17.0.13:36734 - default
[]

% - (-zsh)
228.10:3001 172.17.0.15:45698 - default
[2022-02-17T19:55:07.547Z] "POST /api/v1/events HTTP/1.1" 200 - via_upstream - "-" 118 0 13 12 "-" "Java/11.0.13" "0bd3cc33-b043-92d6-b6ef-d48bf1dfadfd" "message-queue:3001" ""
% - (-zsh)

```

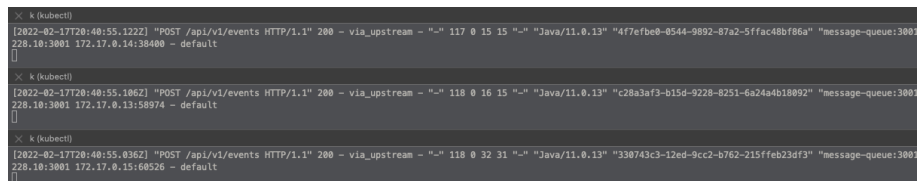
Figure 10: Round Robin request order: 2→3→1→2→3→1→2→3

4.1.2 Random

The random load balancing policy is pretty self-explanatory. Replicas will receive weather data for processing in random order. It's often used to evenly distribute requests when the load is high. We first need to apply the RANDOM policy to v1. To do so, run from the root folder:

```
$ kubectl apply -f configs/load-balancing/random.yaml
```

Now we can run the simulation again and observe logs. We will note now that there is no order anymore in how requests come.



```
k (kubectl)
[2022-02-17T20:48:55.122Z] "POST /api/v1/events HTTP/1.1" 200 - via_upstream - "-" 117 0 15 15 "-" "Java/11.0.13" "4f7efbe8-8544-9892-87a2-5ffac48bf86a" "message-queue:3801"
228.18:3801 172.17.0.14:38400 - default
k (kubectl)
[2022-02-17T20:48:55.186Z] "POST /api/v1/events HTTP/1.1" 200 - via_upstream - "-" 118 0 16 15 "-" "Java/11.0.13" "c28a3af3-b15d-9228-8251-6a24a4b18892" "message-queue:3801"
228.18:3801 172.17.0.13:58974 - default
k (kubectl)
[2022-02-17T20:48:55.836Z] "POST /api/v1/events HTTP/1.1" 200 - via_upstream - "-" 118 0 32 31 "-" "Java/11.0.13" "338743c3-12ed-9cc2-b762-215fdeb23df3" "message-queue:3801"
228.18:3801 172.17.0.15:68526 - default
```

Figure 11: Random request order: 2→1→1→2→3→1→2→2

We can now clean up our configs and proceed to the next part

```
$ kubectl delete -f config/load-balancing
```

4.2 Rate limiting

Rate limiting is a great security technique that can prevent different attack types like brute force, DoS/DDoS, web scrapping, etc. It does so by limiting the number of requests the service can accept. For example, in Instagram and Twitter any third-party application that integrates them can only refresh to look for new tweets or messages a certain amount of times per hour [22]. This makes sense, as otherwise, their servers would simply fall apart from the number of requests per second.

Kubernetes already supports rate limit on the node basis, but with Istio we can fine-tune it on the service level. In Istio rate limit is controlled by our gateway and proxies. There are two types of rate limiting in Istio - **global** and **local**. The global rate limit is applied to the entire mesh and restricts connection from the main ingress gateway. Local, on the other hand, is being applied on the local proxy level of a specific service. Hence, we can, for example, limit the number of requests our machine learning can handle and save it from overload.

Each of the strategies will be applied to the system to observe its behaviour.

4.2.1 Global

As mentioned above, global rate limiting controls the requests that come from outside our mesh. Let's set up the situation where our system only accepts a certain number of requests per minute to prevent it from hacker attacks. The configuration will proceed in four steps [19]:

1. Applying config map. This is the main config that has instructions to limit the number of requests to machine learning-service (1 per minute)
2. Deploying global rate limit service which implements Envoy's rate limit service protocol based redis but adapted to our needs³.
3. Applying an EnvoyFilter to the ingressgateway to enable global rate limiting using Envoy's global rate limit filter.
4. Applying another EnvoyFilter to the ingress-gateway that defines the route configuration on which to rate limit.

Let's apply all 4 configs by executing:

```
$ kubectl apply -f configs/ratelimit/global .
```

Now, run the simulation several times for few milliseconds and start observing:

```
$ POST http://<minikube-ip>:<ingress-port>/api/v1/performSimulation?
duration=10&type=weather&threads=1
```

Lastly let's observe istio-ingressgateway log and see what happens:

```
$ kubectl logs pod/istio-ingressgateway-78f69bd5db-dq5vt -n
istio-system
----

[2022-02-21T00:47:08.438Z] "POST /api/v1/performSimulation?
duration=1&type=weather&threads=1 HTTP/1.1" 200 - via_upstream -
-" 0 2 1387 1331 "172.17.0.1" "PostmanRuntime/7.29.0"
"140fd23f-0516-95e8-ac90-79fd440fc20a" "192.168.64.9:30046"
"172.17.0.8:3001" outbound|3001||simulation.default.svc.cluster.
local 172.17.0.4:53802 172.17.0.4:8080 172.17.0.1:34480 - -
[2022-02-21T00:47:13.654Z] "POST /api/v1/performSimulation?
duration=1&type=weather&threads=1 HTTP/1.1" 429 - request_rate_limited -
-" 0 0 4 - "172.17.0.1" "PostmanRuntime/7.29.0" "c204d442-db4d-
9b37-97b1-baf9f5e89438" "192.168.64.9:30046" "-" outbound|3001|
|simulation.default.svc.cluster.local - 172.17.0.4:8080 172.17.0.1:34480 - -
[2022-02-21T00:47:14.576Z] "POST /api/v1/performSimulation?
duration=1&type=weather&threads=1 HTTP/1.1" 429 - request_rate_limited
```

³<https://github.com/istio/istio/blob/release-1.13/samples/ratelimit/rate-limit-service.yaml>

```
- "-" 0 0 3 - "172.17.0.1" "PostmanRuntime/7.29.0" "821d1e99-3f27-
9b13-bce6-a61020d0a18b" "192.168.64.9:30046" "-" outbound|3001|
|simulation.default.svc.cluster.local - 172.17.0.4:8080 172.17.0.1:34480 - -
[2022-02-21T00:47:15.273Z] "POST /api/v1/performSimulation?
duration=1&type=weather&threads=1 HTTP/1.1" 429 - request_rate_limited
- "-" 0 0 2 - "172.17.0.1" "PostmanRuntime/7.29.0" "b9da7bd7-69bd-
9e5f-97bc-083b50ea47d8" "192.168.64.9:30046" "-" outbound|3001|
|simulation.default.svc.cluster.local - 172.17.0.4:8080 172.17.0.1:34480 - -
```

As can be seen, only the first request was a 200 Ok. All the others were blocked because of our 1 request per second policy.

4.2.2 Local

To configure the local rate limit on proxy an Envoy filter configuration is required. This filter enables rate limit for the Machine learning service, which is done by patching the `envoy.filters.http.local_ratelimit` envoy filter into the HTTP connection filter chain. The local rate limit filter's token bucket is configured to allow 10 requests/min. Remove previous filters and apply new ones:

```
$ kubectl apply -f configs/ratelimit/local .
```

This time we'll need Grafana to better observe internal traffic.

```
$ istioctl dashboard grafana
```

Start the simulation like before and check Machine Learning service log:

```
$ kubectl logs service/machine-learning -n default istio-proxy
----
[2022-02-21T01:01:43.323Z] "POST /api/v1/events HTTP/1.1" 429 -
local_rate_limited - "-" 0 18 0 - "-" "Java/11.0.13"
"f0c4a877-5245-98e2-ad94-a6a343872ddb" "machine-learning:3001" "-"
inbound|3001|| - 172.17.0.6:3001 172.17.0.7:51612 outbound_
3001_..machine-learning.default.svc.cluster.local -
[2022-02-21T01:01:43.309Z] "POST /api/v1/events HTTP/1.1" 200 -
via_upstream - "-" 118 0 19 19 "-" "Java/11.0.13"
"0fb88228-fa6e-9df6-a93a-1df2fb110cac" "message-queue:3001"
"172.17.0.7:3001" outbound|3001||message-queue.default.svc.cluster.local
172.17.0.6:42744 10.109.126.133:3001 172.17.0.6:59572 - default
[2022-02-21T01:01:43.307Z] "POST /api/v1/events HTTP/1.1" 200 -
via_upstream - "-" 117 0 24 24 "-" "Java/11.0.13"
"d3363e76-c5c5-9487-809e-590c89afeb76" "message-queue:3001"
"172.17.0.7:3001" outbound|3001||message-queue.default.svc.cluster.local
172.17.0.6:42302 10.109.126.133:3001 172.17.0.6:59570 - default
[2022-02-21T01:01:43.357Z] "POST /api/v1/events HTTP/1.1" 200 -
```

```

via_upstream - "-" 116 0 17 16 "-" "Java/11.0.13"
"2d35ed7a-3828-98d5-bfa9-d70b53d94e14" "message-queue:3001"
"172.17.0.7:3001" outbound|3001||message-queue.default.svc.cluster.local
172.17.0.6:42302 10.109.126.133:3001 172.17.0.6:59570 - default
[2022-02-21T01:01:45.581Z] "POST /api/v1/events HTTP/1.1" 429 -
local_rate_limited - "-" 0 18 0 - "-" "Java/11.0.13"
"221af186-b92c-9f32-a6e2-75a89a35e3f2" "machine-learning:3001"
 "-" inbound|3001|| - 172.17.0.6:3001 172.17.0.7:51652 outbound_
.3001_..machine-learning.default.svc.cluster.local -

```

It is clear from the log, that the local rate limit is working. We get our 429s and 200s, and `local_rate_limited` message. Now check Grafana.

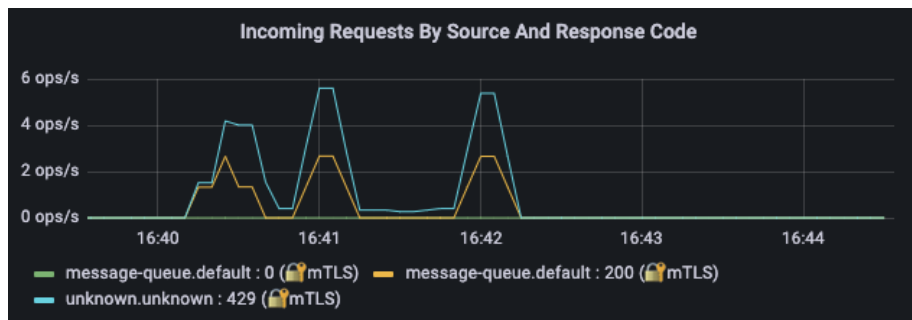


Figure 12: Incoming request into Machine Learning service

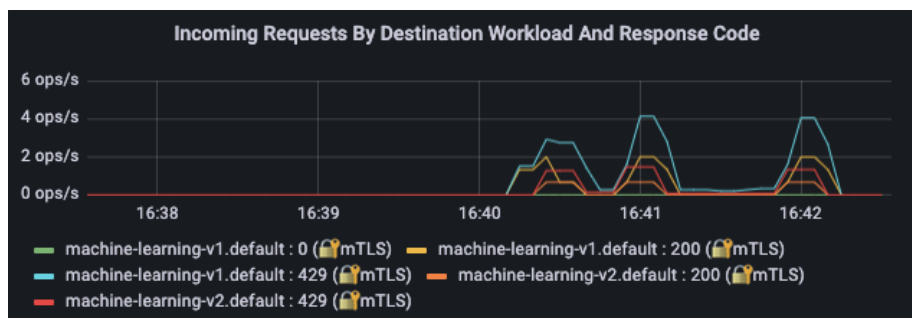


Figure 13: Requests distributed between v1 and v2

In figure 12 requests with statuses 200 and 429 can be seen. These were sent to the machine-learning service. Figure 13 shows the distribution of these requests among pods of the ML service.

4.3 Canary Deployment

In Kubernetes and Istio canary deployment is one of the ways to introduce new features gradually. Assume that a new feature has been added to our Machine Learning microservice. However, we are not yet confident enough that it will work as expected. It was decided to release version two of the service, but not at once. We limit the amount of traffic that version 2 accepts. Doing so will, foremost, allow us to always have stable version 1 for rollback if something goes wrong. As time goes by and version 2 proves to be stable, we can increase the amount of traffic towards it and slowly discontinue version 1.

This can be achieved in Kubernetes on a global, as well as on microservice level. But for microservices, it quickly becomes a tedious task as soon as the system grows big enough, since replicas should be taken into consideration. That's where Istio comes to the rescue. We can easily control traffic flow between versions of the service by utilising our sidecar proxies.

In the beginning, 2 versions of the machine-learning service were deployed. Assume that v1 is current version and v2 is the one that is to be rolled out. First, delete previous configs.

```
$ kubectl delete -f configs/ratelimit/local .
```

Now set up a simple canary deployment using Istio features, namely Virtual Service and Destination Rule. We first need to specify our host service, which is `machine-learning`. Thereafter, we specify two destinations. One for each version. By editing the `weight` property, the load for each version can be decided. For testing, set 90 for v1 and 10 for v2, respectively. The Virtual Service should look as simple as:

```
apiVersion: networking.istio.io/v1alpha3
kind: VirtualService
metadata:
  name: machine-learning
spec:
  hosts:
    - machine-learning
  http:
    - route:
        - destination:
            host: machine-learning
            subset: v1
            weight: 90
        - destination:
            host: machine-learning
            subset: v2
            weight: 10
```

Here we defined the host services and specified which versions should be affected and with which load.

Next, define the Destination Rule:

```
apiVersion: networking.istio.io/v1alpha3
kind: DestinationRule
metadata:
  name: machine-learning
spec:
  host: machine-learning
  subsets:
    - name: v1
      labels:
        version: v1
    - name: v2
      labels:
        version: v2
```

For convenience, both configs should be added into one file and applied:

```
$ kubectl apply -f configs/canary.yaml
```

We can now start our simulation for at least a few minutes and see what happens. Run Simulation in Postman run and open the Kiali dashboard and go to Graph:

```
POST http://<minikube-ip>:<ingress-port>/v1/performSimulation?
duration=100000&type=weather&threads=1
```

```
$ istioctl dashboard kiali
```

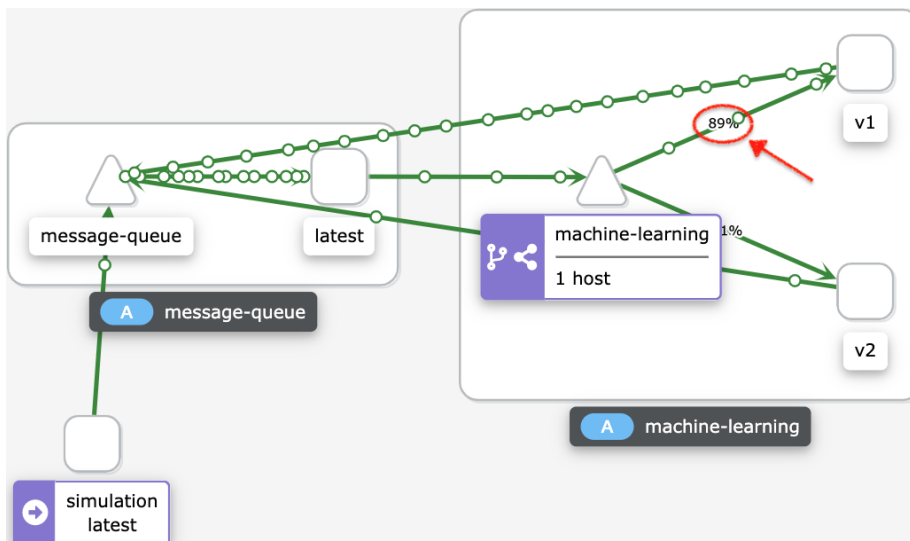


Figure 14: Making sure that our canary settings are working

In the right part, it can be seen that traffic distribution between versions of Machine Learning service is 89 and 11 percent. Due to small fluctuations, these values may vary slightly. According to observations, the amplitude is around 5-7%. In Grafana, traffic distribution will look like that:

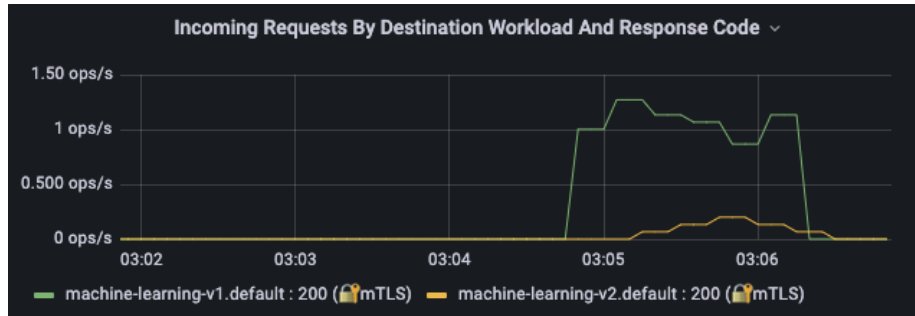


Figure 15: Traffic distribution between versions

Now clean up configs to prepare for the last concept review.

```
$ kubectl delete -f configs/canary.yaml
```

4.4 Circuit Breaking

Circuit Breaking is a popular pattern used in microservice-based systems for resiliency. Microservices interact with each other and share data, and some of them depend on the others in different ways and grades. It often happens that some services can get overwhelmed by requests from one service so that other services can never reach them. Which wastes resources and can lead to failures or even cascade failures. The job of circuit breaking is to detect such potential failures.

In Istio, setting up circuit breaking is as simple as the writing of one config. But to test the config, it will be necessary to access services inside the service mesh. To do so, we will utilise a simple load-testing client Fortio [8], which is already included in the Istio folder.

```
$ kubectl apply -f samples/httpbin/sample-client/
fortio-deploy.yaml
```

To make sure it worked execute `$ kubectl get pods`. Fortio should be on the list and have status Running.

NAME	READY	STATUS	...
fortio-deploy-687945c6dc-twtvx	2/2	Running	...
...

In this example, the message-queue service will be overloaded to test circuit breaking. First, prepare the config:

```

apiVersion: networking.istio.io/v1beta1
kind: DestinationRule
metadata:
  name: mq-circuit-breaker
spec:
  host: message-queue
  trafficPolicy:
    connectionPool:
      http:
        http1MaxPendingRequests: 10
        maxRequestsPerConnection: 3
      tcp:
        maxConnections: 1

```

In this config, the number of maximum pending requests is set to 10, which means that every further pending request will be blocked. Also, each connection may handle up to 3 requests. Max connection is the number of consecutive connections to the service. Every excess connection will be considered pending. Now apply the config and start testing.

```
$ apply -f configs/circuit-breaker.yaml
```

Next login to fortio pod so that requests could be sent through it. We can immediately test if it works:

```

$ export FORTIO_POD=$(kubectl get pods -l app=fortio -o
'jsonpath={.items[0].metadata.name}')

$ kubectl exec "$FORTIO_POD" -c fortio -- /usr/bin/fortio
curl -quiet http://message-queue:3001/api/v1/observations

```

This will produce the following output:

```

HTTP/1.1 200 OK
content-length: 0
date: Sun, 20 Feb 2022 18:10:18 GMT
x-envoy-upstream-service-time: 5
server: envoy

```

which means that fortio is configured correctly. Now run the test case. The service will be called with 20 concurrent connections (-c 20) and send 20 requests (-n 20):

```

23:07:47 I logger.go:127> Log level is now 3 Warning (was 2 Info)
Fortio 1.17.1 running at 0 queries per second, 2->2 procs,
for 20 calls: http://message-queue:3001/api/v1/observations
Starting at max qps with 20 thread(s) [gomax 2] for exactly
20 calls (1 per thread + 0)
23:07:47 W http_client.go:806> [16] Non ok http code 503 (HTTP/1.1 503)

```

```

23:07:47 W http_client.go:806> [15] Non ok http code 503 (HTTP/1.1 503)
23:07:48 W http_client.go:806> [7] Non ok http code 503 (HTTP/1.1 503)
23:07:48 W http_client.go:806> [8] Non ok http code 503 (HTTP/1.1 503)
23:07:48 W http_client.go:806> [9] Non ok http code 503 (HTTP/1.1 503)
23:07:48 W http_client.go:806> [6] Non ok http code 503 (HTTP/1.1 503)
23:07:48 W http_client.go:806> [11] Non ok http code 503 (HTTP/1.1 503)
23:07:48 W http_client.go:806> [3] Non ok http code 503 (HTTP/1.1 503)
Ended after 60.266846ms : 20 calls. qps=331.86
Aggregated Function Time : count 20 avg 0.026513552 +/- 0.01763
min 0.004410098 max 0.059791383 sum 0.530271044
# range, mid point, percentile, count
>= 0.0044101 <= 0.005 , 0.00470505 , 10.00, 2
> 0.007 <= 0.008 , 0.0075 , 15.00, 1
> 0.009 <= 0.01 , 0.0095 , 20.00, 1
> 0.011 <= 0.012 , 0.0115 , 25.00, 1
> 0.012 <= 0.014 , 0.013 , 35.00, 2
> 0.014 <= 0.016 , 0.015 , 45.00, 2
> 0.02 <= 0.025 , 0.0225 , 55.00, 2
> 0.025 <= 0.03 , 0.0275 , 60.00, 1
> 0.03 <= 0.035 , 0.0325 , 70.00, 2
> 0.035 <= 0.04 , 0.0375 , 75.00, 1
> 0.04 <= 0.045 , 0.0425 , 80.00, 1
> 0.045 <= 0.05 , 0.0475 , 85.00, 1
> 0.05 <= 0.0597914 , 0.0548957 , 100.00, 3
# target 50% 0.0225
# target 75% 0.04
# target 90% 0.0532638
# target 99% 0.0591386
# target 99.9% 0.0597261
Sockets used: 20 (for perfect keepalive, would be 20)
Jitter: false
Code 200 : 12 (60.0 %)
Code 503 : 8 (40.0 %)
Response Header Sizes : count 20 avg 75 +/- 61.24 min 0 max 125 sum 1500
Response Body/Total Sizes : count 20 avg 171.4 +/- 56.83 min 125 max 241 sum 3428
All done 20 calls (plus 0 warmup) 26.514 ms avg, 331.9 qps

```

From this log, it can be observed that out of 20 requests, only 12 were successful and 8 were blocked by our circuit breaker. Which means our configuration works. One can also specifically check the number of requests that were flagged for circuit breaking by querying the proxy:

```

$ kubectl exec "$FORTIO_POD" -c istio-proxy -- pilot-agent request GET
stats | grep message-queue | grep pending
----
cluster.outbound|3001||message-queue.default.svc.cluster.local
.circuit_breakers.default.remaining_pending: 10

```

```

cluster.outbound|3001||message-queue.default.svc.cluster.local
.circuit_breakers.default.rq_pending_open: 0
cluster.outbound|3001||message-queue.default.svc.cluster.local
.circuit_breakers.high.rq_pending_open: 0
cluster.outbound|3001||message-queue.default.svc.cluster.local
.upstream_rq_pending_active: 0
cluster.outbound|3001||message-queue.default.svc.cluster.local
.upstream_rq_pending_failure_eject: 0
cluster.outbound|3001||message-queue.default.svc.cluster.local
.upstream_rq_pending_overflow: 1021
cluster.outbound|3001||message-queue.default.svc.cluster.local
.upstream_rq_pending_total: 3371

```

As can be seen from the log 10 requests were flagged.
Let's now clean configs and proceed to summary:

```

$ kubectl delete -f circuit-breaker.yaml
$ kubectl delete -f gateway.yaml
$ kubectl delete -f IOT/

```

5 Summary

As can be seen from the above tests, Istio is a pretty comfortable tool to use and also a very powerful one. Although our example was based on a small microservice system, similar configurations could easily be applied to a much bigger scale and with even deeper fine-tuning. As was already mentioned, almost all the problems that were covered can be solved either on Kubernetes, or microservices level, but will require way more man-hours and hence more costs. The above examples illustrate, that with simple configs one can configure complex behaviors without almost any system overloads, as well as maintain a good level of resiliency.

Let us sum up what was learned through this study:

- How cloud-based applications are being built in modern days
- What are service meshes and how do they help software engineers to save time and money
- How to use Istios' load balancing feature to manage traffic between our replicas in a more convenient way
- How to apply two rate limit strategies: global for the entire mesh and local for the exact service
- How the canary deployment can be realized using Istios' virtual services and destination rules
- How circuit breaking can prevent services from cascade failures and increase resiliency

6 Conclusion

Each of the described Istio features, especially when combined appropriately, can give developers an exceptional toolset for solving modern-day Microservice and Cloud Computing problems. And we didn't even cover other features and concepts. What has been covered is just a slice of traffic management features and a bit of security. Istio has other powerful concepts like ⁴

- **Security** - Authorization and Authentication
- **Observability** - Telemetry and Monitoring features
- **Extensibility** - Web Assembly plugin system

Istio is a very young technology and evolves constantly. Each major update brings new features that further extend developers' arsenal. It is the logical consequence of software engineering's evolution. The systems get larger and more complex, hence new techniques that handle these complexities come to life. Such changes require automation, scalability, and continuous delivery. These are the most valuable aspects that have to be considered while evolving. [14]

In the beginning, microservices were a great alternative to hard-to-scale monolith architecture. However, with the lapse of time, even they did not escape the latter's fate. And here's a good example: Such a complex network itself



Figure 16: Microservice Death Star by Amazon and Netflix [4]

requires some sort of management, and we are not speaking about Kubernetes here. A higher level of control is required. A good analogy would be the complexity of programming language levels (from low to high). Each level is the logical result, the evolution, the higher level of abstraction above its predecessor.

⁴<https://istio.io/latest/docs/concepts/>

Each has a higher level of control and less low level. Like C has more control over memory than Java, Kubernetes has more control over microservices than Istio. But, in conjunction with one another, they make it possible to build more complex applications.

7 Future work

If I did further research on the topic, I would love to try service meshes on a real-life customer deployed system, whether it is a new setup or some maintenance project. It would be nice and useful to see how these features function on a way larger scale, to collect real-life metrics, see security features in action, etc. As was mentioned in conclusion, it is indeed intriguing where time will bring us. Little hardware development stagnation is over and technologies are rapidly improving again. It is only a matter of time until we will come to the point where the next complexity level jump occurs. And I wonder where these changes will bring us. Who knows, maybe with the lapse of time even service meshes will become so complex that the new solution will have to appear to manage it. It would be fascinating to theorise on this matter.

References

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