

Final Report on: Pretix Redundant Operation

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Introduction

This report aims to describe the work developed for the final phase of the practical assignment of the discipline of Computational Infrastructures Management (*I*) from the Msc. degree in Informatics Engineering of the University of Aveiro at the Department of Electronics, Telecommunications and Informatics. It is assumed that the reader has knowledge about the previous report, and it is here included: the characterization of the service level agreements for each service; the description of the load balancing mechanisms among software components; the redundancy strategy; the listing of metrics and computational resources monitored, along with the respective defined alarms; and other relevant aspects such as horizontal scalability and component fault tolerance.

The service provided is Pretix, an online shop, box office and ticket outlet already successfully used by other service providers for conferences, festivals, exhibitions, workshops and more. The previously delivered work focused on the product presentation, distributed installation, and resource and performance analysis. All code developed is publicly accessible in our GitHub repository:

<https://github.com/FilipePires98/GIC>.

1 System Architecture

1.1 Infrastructure

To provide the Pretix Ticketing Software to the public, it was required to set up an infrastructure containing their Django-based web application, an instance of a Web Server Gateway Interface (WSGI) hosting the application and of a reverse-proxy for the web application deployment in production mode, as well as a database management system (DBMS) (where PostgreSQL was used) for handling disk storage and a caching server (where Redis was chosen) for in-memory storage and asynchronous queuing of tasks.

Our strategy is thoroughly described in the previous report, which already considered minimum redundancy to ensure availability. It is achieved through the usage of Docker to isolate each component for greater control and easier configuration, and Docker Swarm for the orchestration of the entire infrastructure. Nevertheless, as it was a still primitive solution, naturally several upgrades were applied, mostly internally to each container. The latest architecture is visually presented in Figure 1. This solution considers quality attributes whose assurance is described in the sections below.

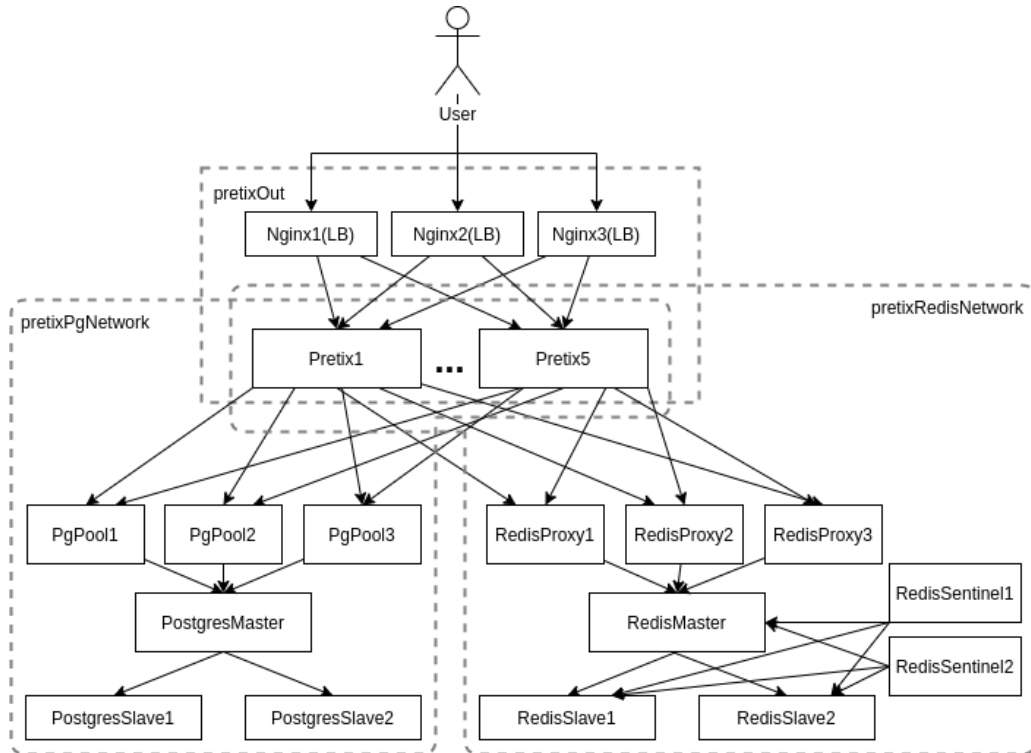


Figure 1: Infrastructure architecture diagram deployed in Docker Swarm.

1.2 Load Balancing

The knowledge gathered so far regarding the operation of Pretix affirms that user activity peaks are to be expected when in production. This has many implications and requires the implementation of mechanisms that maximize the infrastructure's efficiency.

With this in mind, we have come up with mechanisms for load balancing based on proxy servers. These servers, when strategically placed, act as intermediaries for requests seeking resources from the system itself. This not only reduces the complexity of requests made to our services, but also provides additional benefits with regards to security and load balancing, as they allow a controlled and intelligent distribution of requests among components.

An NGinX proxy was installed between the end users and the web application. With the help of Docker Swarm, replicas of both the proxy and the web application were deployed. This is transparent for the user and it works as if he/she is sending requests to one single access point.

The storage clusters also resort to proxies. For the PostgreSQL server cluster, we adopted Pgpool-II. As our database is replicated in master/slave mode, Pgpool is able to take advantage of the replication feature in order to reduce the load on each PostgreSQL server by evenly distributing queries among available servers and consequently improving the overall throughput. Pgpool provides other useful features such as connection pooling, where it maintains established connections to the servers and reuses them whenever a new connection with the same properties comes in, and automated fail over, mentioned in greater detail in section 2.3.

For the Redis server cluster, the adopted solution was the well-known HAProxy. The instances installed spread requests evenly across the Redis cluster. Here, a master/slave mode was also used, meaning that the proxies interact with master instances that then delegate work among slaves.

Load tests executed during development proved that such additions had a significant impact on the performance of the system. The usage of networks, seen in Figure 1, where the proxies mentioned above are inserted help orchestrating and logically grouping components.

1.3 Redundancy

In order for us to provide Pretix, it is only required to maintain one instance of each component running. However, this is highly inadvisable as it is susceptible to failure from any point, i.e. if any of the system components fails, the entire system goes down. This is referred to as a redundancy measure of N .

Ensuring actual redundancy of a system requires at least a measure of $N+1$, where each component can be replaced by another if required. In most cases this is achievable through the replication feature of Docker Swarm. Pretix's stack has K replicas of NGinX, of the Pretix web

application, of Pgpool proxy and of HAProxy, where K can be defined on deployment with a value greater than 2.

Redundancy was also the reason why a master/slave mode was adopted on both clusters. By instantiating 1 master and K slaves, it is possible to fulfill the requirements implicit to N+1 since if any slave fails the others can replace it and if the master fails a slave is elected the new master. Fortunately, for the PostgreSQL cluster this reelection process is done automatically, but for the Redis cluster the use of sentinels is required: sentinels observe the behavior of masters and slaves, detect failures and reassign roles when needed; here we insist on providing N+1 redundancy by having multiple instances of sentinels as well.

1.4 Scalability

Any project intended to be successful must consider the scalability of the product in the future. Ours is no exception since, if the ticketing platform becomes popular, the infrastructure must be prepared to respond to greater loads of income traffic.

There were 2 means to ensure scalability: vertical or horizontal. As we had access to limited computational resources and adding more physical RAM or CPUs was not an option, the only alternative left was to support horizontal scaling.

The idea here is to make possible the deployment of new replicas of any component and integrate them to the existing infrastructure without significant negative impact on insertion and in a completely transparent way to the users. This is exactly what we achieved. Additionally, each component identifies itself to the monitoring server on initialization so that the data related to what is being managed is constantly up-to-date.

Scaling a portion of the stack can be done both manually or in an automated way. By default we support manual scaling, but we found it was logical to implement automated mechanisms to scale up or down according to the online users and the resources used. This was achieved through the monitoring of meaningful metrics collected from each container instance, as we present in section 2.3.

2 Infrastructure Management

2.1 Service Level Agreement

As service providers, we are supposed to fulfill a commitment with our clients, whether it is formally defined or not. Obviously it is a good practice to elaborate an agreement where particular aspects of the service are agreed between the service provider and the service user - this is called a Service Level Agreement (SLA). We describe ours in this section.

Considering the scope of our project, we've defined the following commitment key points:

- Pretix is to be provided through web browsers, by accessing a particular address, and it will only be accessible to those within the university's virtual private network.
- Its usage is fully dependent on the usability of the Pretix service itself and no complexity should be added by the infrastructure we provide with regards to this.
- We are responsible for providing free access to the ticketing platform through our servers and keeping data integrity even when the service is unavailable
- The installations hosting our deployment belong to a third party (the university) and thus we are not accountable for possible availability issues related to networking and machine up time. Nevertheless, if this is not considered, we ensure an availability of 100% under normal circumstances and of 70% for high activity peaks ¹, with a maximum response time of 50 seconds for ticket purchasing requests.
- No support is guaranteed for issues about features related to the internal implementation of the ticketing software.

It is worth mentioning that our analysis to the (free version of the) Pretix product showed that the software had strong limitations on maximum load capacity and response time that, according to the authors, could not be solved through horizontal scaling.

Focusing on the infrastructure itself and the computational resources, we have defined a list of Service Level Objects (SLO) - the metrics to be observed - representative of the stack's state, and a list of Service Level Indications (SLI) - the thresholds or functions applied to the metric values - that will help us react to situations where the SLA compliance is threatened. These SLOs revolve around: memory capacity, operation latency, data transfer throughput, availability percentage, recovery time and reliability.

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2.2 Monitoring

In a production environment it's mandatory to establish a monitoring system to facilitate the inference of the service stack and all its components status. It further helps in identifying prob-

¹Over 50 requests per minute.

lems and enables the manager to define alerts he/she considers to be critical and notify him/her if the monitored metric overcomes a specific threshold, possibly violating the established SLA. This kind of system also allow the definition of reactions for specific SLA violation.

To implement a monitor system on our service we were guided to create a independent virtual machine to host all the monitoring systems, decision we quickly understood to be optimal. This way the monitor infrastructure will not create any load on the service facility, leaving all the resources to the service fulfillment. It also allows the possibility of deploying the system services without the monitoring platform enables, a scenario not advisable but possible.

We decided to tackle the monitoring aspect in two distinct fronts: the log collection and the metric collection. Both were handled in parallel and will be further detailed in the following subsections.

2.2.1 Log Collection

Log collection, as the name implies, consists in gathering in a centralized fashion all the logs generated by all services composing our full system. There are a lot of options to achieve this, some that we considered and investigated. To analyze the logs themselves in a user-friendly and feature rich platform we choose the Elasticsearch stack(ELK), comprised of Elasticsearch itself, Logstash and Kibana. The purpose of each of those components are:

- Elasticsearch - text based database, optimized to search and filter through textual data,, such as logs.
- Logstash - textual data(logs) collector/receiver; stack entryptpoint.
- Kibana - GUI data analyzer, enabling the creation of dashboards, alerts, ...

One of the approaches we tested the centralize all logs but ended up not being used in the final product was the rsyslog approach. This technology allows the definition of a server and multiple clients, each one assigned to each service, that gathers the system logs of the machine and sends them to the predefined rsyslog server. Although enabling a centralized registry of all logs of the system's services, some integration problems were encountered but after some analysis we found this option not to be optimal, since the rsyslog when receiving exterior logs will persist them in the system logs file of the hosting machine. Since this entity would be hosted by the monitoring virtual machine, every log received would be replicated since the rsyslog persists them and the Elasticsearch would also persist them due to it's indexing process.

The approach we settle on ended up being much cleaner and simpler in our opinion. After some research we found that one of the logging drivers provided by the docker platform was syslog, that enables the redirecting of all logs to a syslog server specified in the driver options.

This was exactly aligned with our scenario, since all our deployment was made in docker's swarm mode and we wanted to centralize all services logs. We also found that Logstash allows the creation of a server that can receive any type of message(since filters can be applied to the incoming messages), enabling the direct connection between the docker syslog driver and the ELK stack.

After the logs arrive at the Logstash, the internal ELK flow was configured so that Elasticsearch indexes the information and Kibana is able to access it. The system is then capable of collecting all logs and present them in the Kibana interface, enabling a centralizes analysis and further alert definition.

Further processing over collected logs can be made if the information is context specific and allows these type of processing, such as counters or other metrics. In our case we concluded that this was not necessary since we were able to obtain metrics from all the services necessary, information we processed with the technologies described in the following subsection.

2.2.2 Metric Collection

Metric collection is very similar to log collection in the sense that the objective is to centralize all generated information from all the system's composing services, but this time the data collected are metrics, most of the time numeric data like counters, gauges and indicators about the service process itself or even context specific ones.

Obviously, there are many approaches and technologies that allow metric collection. After some investigation we found some that are commonly used for this purpose, such as Nagios, Zabbix, Munin, TICK stack, Prometheus and more. We opted to implement the Prometheus one, since our main service(Pretix) already exported metrics in the Prometheus format. For the remaining services we needed to install parallel metric exporters compatible with Prometheus. Since our infrastructure was deployed using docker in swarm mode, the addition of exporters was made by changing the Dockerfiles and respective entrypoints for each service. The mapping between service and installed exporter was:

- Nginx -
- Pg Master/Slave -
- HA Proxy -
- Redis Master/Slave/Sentinel -

In addition to all those exporters, we also installed another one in the monitoring virtual machine itself responsible for monitoring the hardware itself, like the CPU, memory and disk

usage. For that task we used the SNMP protocol, which is a internet standard protocol for collecting and organizing information about managed devices on IP networks and for modifying that information to change device behavior. The protocol support was installed a priori on every compute node, being only necessary request it's metrics when necessary. Since the deployment is made using the Docker's Swarm mode, it was impossible for us to know where our services would be deployed in the Swarm computing network. With that in mind we had two options: either we statically indicated every computing node to be monitored, or we dynamically discovered the ips of the machine hosting our services. We concluded that the first option was not feasible since it would cause unnecessary stress in machines and would not scale well, since if there were 1000 computational nodes we would need to indicate the 1000 ips. We opted to dynamically discover the machine ips, which implied that we needed to create a custom ip collector.

This custom component had an independent service and was created using Python's Flask library, which enabled us to quickly develop the necessary endpoints and respective logic. At initialization, each service sends a HTTP Post to our custom service with its name, replication number and id. Our service will then register or update it's ip and modify the respective Prometheus target files for each type of exporter.

Prometheus is the engine behind the monitoring and alerting, but it's as necessary a powerful, usable and efficient user interface. Grafana is commonly integrated with Prometheus, being the GUI indicated by the engine itself. That was the reason we deployed Grafana, enabling us to explore the collected metrics, construct meaningful dashboards and configuring critical alerts.

2.3 Alerts and Automation

Collecting metrics and logs means nothing if there's no objective. For that reason we defined the SLA described in section 2.1, enabling us to know when the infrastructure and system are performing as expected or not. When the SLA is violated the system should notify the infrastructure manager so that a action may be performed. In more advanced scenarios, automatic reactions may be specified so that the system can adjust to certain problems, creating some desirable flexibility.

For our context and service, we defined both alerts and automatic responses. Alerts were configured in every metric considered critical and that it's behavior may translate in the status of the service, as explained in section 2.1. All the generated alarms and notifications were made by a Microsoft Teams channel. All those alarms were also sent to a custom service responsible for receiving, processing and if necessary reactive to the incoming alarms. This service was the same as the one responsible for collecting the computing machines for efficiency reasons. When receiving an alarm, the component will process it and if it's one that can be corrected

in an automatic fashion, a reaction will be executed, in our case, using the Docker's Python SDK to automatically scale up or down the service that triggered the alarm. The alarms and respective reaction are:

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3 Additional Remarks

3.1 Pretix Quickstart

During development, a need for a faster basic setup of Pretix emerged for debugging purposes. As many of our tests assumed the existence of organizers and events already inserted on the databases, each time these had to be cleared the objects had to be manually created before proceeding with some implementation process.

In order to respond to this need, we developed a Python script capable of completely dispose of the manual labour. This script, which we called `quickstart.py` and placed under a dedicated folder, uses Selenium (16) to create an organization inside Pretix, generate a secret API key for sending REST requests (storing it in a text file), create a default event and prepare it for load testing with Locust. It also contains configurable variables to personalize its output.

Such quickstart tool was not only found useful for the obvious advantage in development speed it brought to us, but was also considered a good help tool for any person that wishes to quickly experiment our infrastructure without having to know much about Pretix itself.

3.2 Documentation

A great portion of the code we deal with isn't of our authorship, so the documentation it contains is the documentation we get. Nevertheless, the code we developed has specific purposes and, although usually very much self explanatory due to its nature, should be easy to interpret by any person. Our goal is to ensure that what we deliver can be easily reused or even continued by other developers.

With this in mind, we made significant efforts on ensuring that all code developed by us followed a common structure with the same coding style. Also, we made sure all source files have a description of their purpose and relevant information on the first lines, and contain comments on key points of the code explaining snippets considered of greater complexity/importance.

3.3 Assignment Contributions

Regarding the work distribution among developers, a close-contact strategy was defined where each worked on a cluster component or piece of software according to a predefined plan. The cluster strategy and respective details were decided in conjunction, as well as the key objectives and tasks to be achieved before the final delivery deadline.

Pretix' features were explored by both developers in order to better comprehend the web platform and the REST API. Then, João focused on the deployment of our solution on the department's infrastructure while Filipe focused on creating and executing the load tests. The

analysis of the results were carefully conducted in conjunction as well, in order to maintain a common knowledge for better workflows. It is needless to say that bug and error solving was made along the development phase by both developers any time it was required. Once performance benchmarking and bottleneck identification were completed, this report and the code documentation became our primary concern, with both contributing equally.

Conclusions

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