

Scalability and Performance Analysis of a Web Application

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Introduction & project's goal

Step 1

Create a web application that allows for searching a movie in the database and provides the information on the movies, the main actors and directors and the average rating, length etc. Design your application with the tools that you prefer.

Step 2

Create a query set.

Let us assume that the probability that a film is searched is proportional to the number of rating that it has received.

Create a list of 10,000 queries of movie titles (where entries can be duplicated) sampled according to this rules.

Step 3

Perform a load test to assess the scalability of your web app in closed-loop.

How many users can it handle?

Use the queries that you previously created as input of your load test.

Step 4

Design the system in such a way that its scalability increases.

To this aim, identify the bottleneck in your implementation (either experimentally or from its model).

Then, test your designed system using JMT to study the expected response time as function of the number of users. What is the optimal number of users?

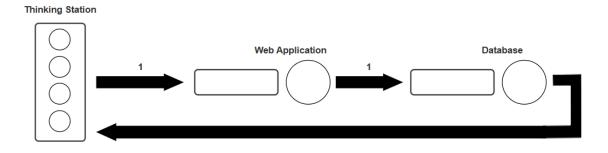
The goal:

We are interested in computing the speed of the system, the throughput in job/seconds as well as the response time, measured in seconds, which is crucial from the point of view of a user. Another important feature to take into account is the system's scalability focusing on its load factor when the number of jobs grows over time and consequently we need to identify the so-called bottleneck responsible to limit its performance.



Analytical model

In order to perform load testing we decided to structure the system under test in an **interactive single-class closed-system** depicted in the following way:



Being a closed system queueing network the web server is going to serve a fixed number of users placed in the **thinking station**, in which they send parallel and independent requests to the **web application station** that processes them.

The **database station** produces a response for each one and send back to each related user which will consult it for a certain amount of time, called thinking time, before doing another request.

We assume for the sake of simplicity that the web application and the database stations are single server with constant speed while users are statistically identical (single-class). In addition, each station is always visited by every user so the relative visit ratio for each one is equal to 1.

The throughput with a N amount of users X(N) of such a system is defined by the following formula:

$$X(N)=N/E(T[N])$$

where E(T[N]) is the system time



System architecture

Our web application is made up of a *javaScript* file, given that our goal was solely to test performance, the back-end logic was implemented within this file: the application was designed to send 10,000 queries to the database.

Node.js enabled the JavaScript file to interact with the database, which was set up locally using **PostgreSQL**.

In accordance with the project guidelines, the data was sourced from the IMDB website.



Load testing setup

In order to perform load testing in our application we relied on Apache **Jmeter** load tester software.

We created a test plan, which defines the overall settings for the test, then added to it two distinct thread groups:

- one to simulate server-to-database requests
- one to simulate user-to-server requests

Now, we'll take a closer look at both.

The former one simulates the interaction between the web application and the database and it contains two main components:

- Configuration element, that establishes a connection to the database
- **JDBC sampler**, in which we've defined a query the web application use to retrieve data from the database and show it in the web application.

Worth mention that we've decided to set a maximum number of connections (into the configuration element) of 50, this because if the default value (1) was kept, we've seen a error rate too high regarding the requests.

The latter simulates HTTP interactions between users and the server and it contains two main components:

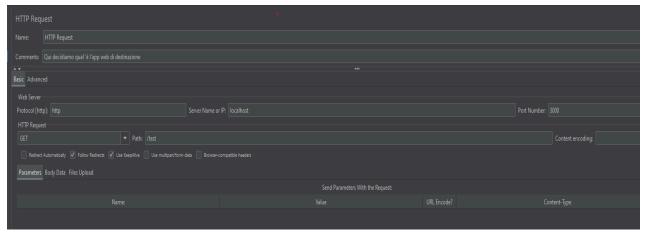
- HTTP sampler, which allows sending HTTP/HTTPS requests to a web server and also parse embedded resources (e.g., images) when instructed, simulating realistic HTML content retrieval.
- Flow Control Action sampler configured for Pause only to add a Constant Throughput Timer that is the key factor in evaluating system behavior under a controlled load. This timer regulates the number of requests sent per minute or second (it dynamically adjust pacing to achieve a defined throughput), here we've set a random delay with an average and deviation to achieve a more realistic user interactions

Right below is possible to see all the settings of the tests plan above described:





The thread group used for the server-to-database requests. Here we've only set the number of queries that a user sends: 10000, as indicated in the project specification.



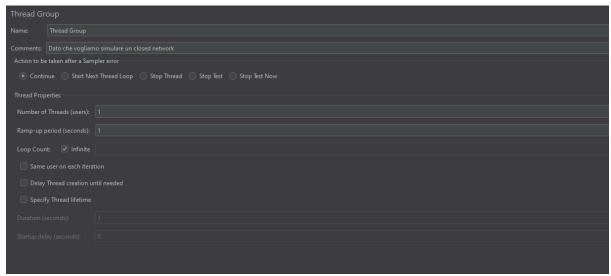
The JDBC configuration element, where it's possible to see the database's connection settings.

We've used a local database in the configuration, in fact the URL specified is a local one



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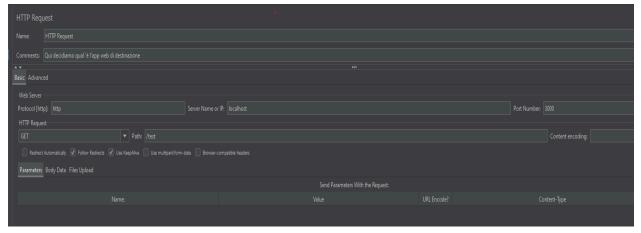
The JDBC sampler.



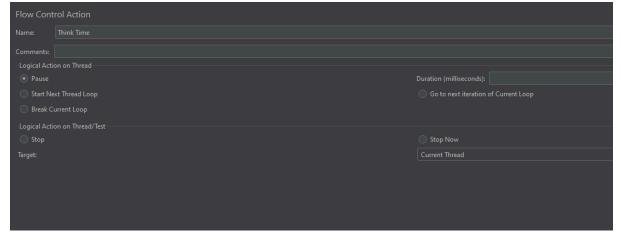
The thread group used for the user-to-server requests.

We've put the flag to "Loop count: Infinite" to simulate a real situation where the users are not limited in the number of requests they can make.





Here we see the HTTP Sampler, where we specified the protocol, the server name, the port (of the web application server), the type of request and the path to the resource that allows to the users to get the data.



Here we see Flow Control Action.



Uniform Random Timer					
Name:	Pause				
Comments:					
-Thread Dela	y Properties				
Random Delay Maximum (in milliseconds):		100			
Constant Delay Offset (in milliseconds):		1000			
Constant	relay Offset (III Hilliseconus).	1000			

Here we see the Constant Throughput Timer, where we've set the two delays.



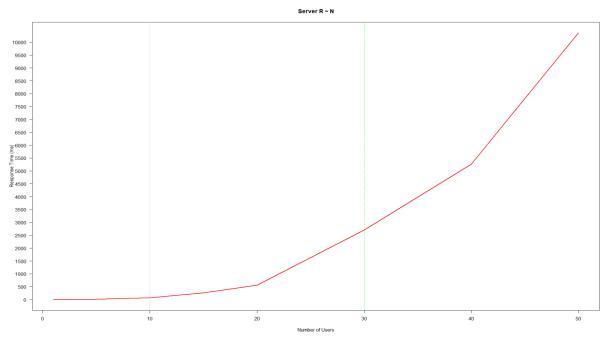
jMeter analysis

The results of **jMeter** obtained with the increase of the number of customer. To better readability we've used rStudio to make the plots. Analyzing the results we've found that

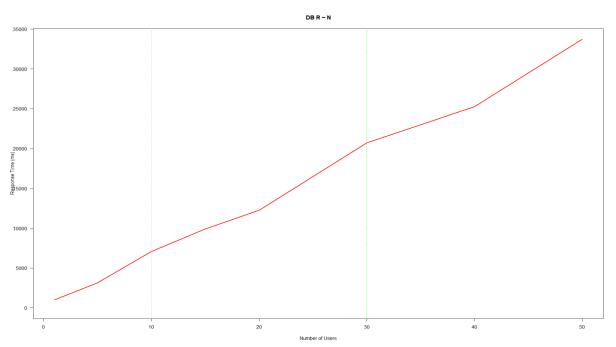
- the low load is with a number of customers below 10
- the moderate load is with a number of customers below 30
- over 50 customer we have heavy load

As indicated by the dashed lines in the graphs that follows:



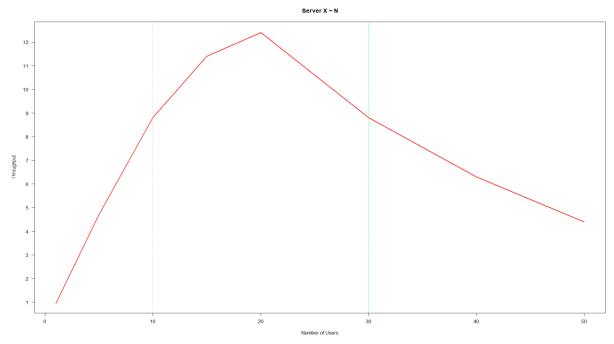


The response time of the server in the system not optimized.

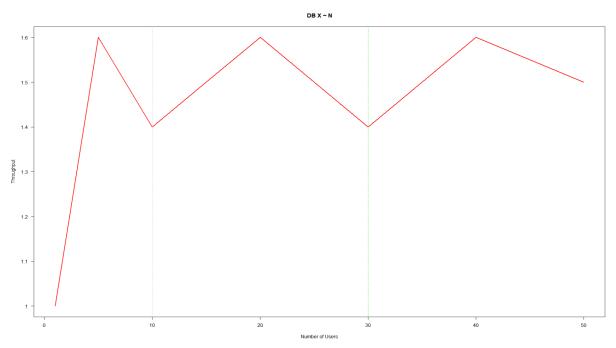


The response time of the database in the system not optimized.





The throughput of the server in the system not optimized.



The throughput of the database in the system not optimized.



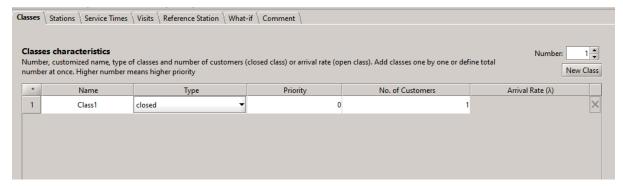
Model analysis by JMT

The settings in JMT touch various sections, let's dive into each of them:

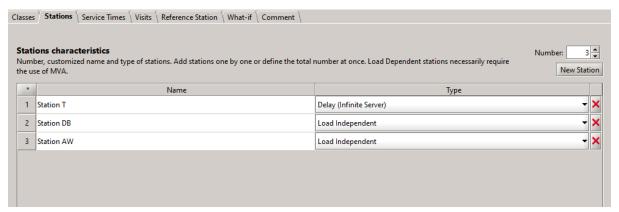
- Classes, here we've set the network type and the types/classes of the requests, respectively: closed, single-class
- Stations, here the number and type of stations are set, respectively: thinking station, web-application station, database station
- Service Times, here the results obtained with jMeter are used to set each station service time
- Visits, here the visit ratios are set as one visit per station each visit to the reference station
- Reference station, here which stations is used as reference station: the thinking station
- What-if, here we set the bound of the experiment and the parameter to increase, respectively: 1 to 50 customer, number of customer

Right below the settings we've used described above:





The classes.

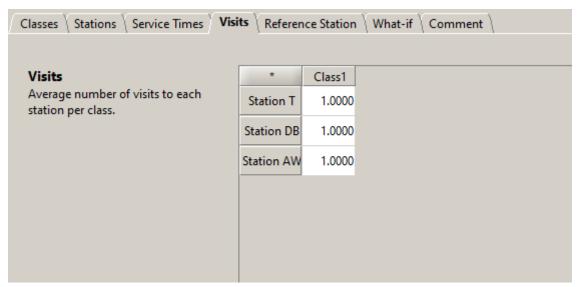


The stations.

Classes \ Stations \ Service Times \ Visits \ Reference Station \ What-if \ Comment \					
Service Times	*	Class1			
Input service times of each station for each class.	Station T	1.0500			
If the station is "Load Dependent" you can set the service times for each	Station DB	0.9830			
number of customers by double-click on "LD Settings" button.	Station AW	0.0090			
Press "Service Demands" button to enter service demands instead of service times and visits. MULTICLASS MODELS: when for a station the per-class service times are different, the results are correct ONLY IF its scheduling discipline is assumed Processor Sharing (PS) and not FCFS (See BCMP Theorem).					

The service times.

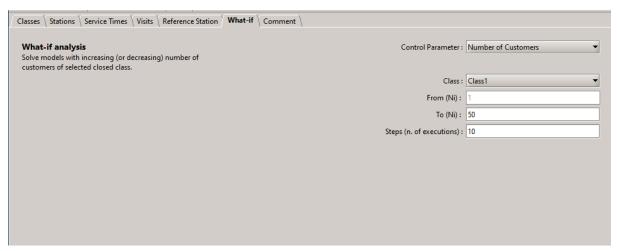




The visit ratio.



The reference station.





Bottleneck analysis, traffic equations, forced flow law, plots

In order to spot the bottleneck we determine the average service demand for each station that is the total amount of service required by a customer to each station i for each visit it does at the reference station, assuming single server systems and independent service time distribution from the number of visits performed by a job at queue:

$$E[D_i]=E[V_i]*1/\mu_i$$

and then the highest one determines the bottleneck station.

In our system we have that:

each station i is visited with visit-ratio: $E[V_i] = 1$, i = 1,2,3

service demand web application: $E[D_{app}] = E[V_{app}]^*1/\mu_{app} = 1^*0.0090 = 0.0090$ service demand database: $E[D_{DB}] = E[V_{DB}]^*1/\mu_{DB} = 1^*0.9830 = 0.9830$

Therefore the database station is the bottleneck!



Optimal number of users

After having identified the bottleneck, what happens to our system when the number of customers increases?

is going to be the first to have its utilization equal to 1, because the jobs tend to accumulate at that station and as a consequence limiting the system performances, namely the throughput.

In fact, the forced flow law states that the system throughput X_{sys} , which corresponds to the reference station one, is equal to $\mu_{bottleneck}/E[V_{bottleneck}]=1/0.9830=1.0173$ job/s

which imposes the following asymptotic limits to our interactive system:

Lower bound for the expected response time:

when a single job is in the system, we have that:

$$E[R] \ge MAX(E[D_{system}], N^*E[D_{bottleneck}] - E[Z])$$

where $E[D_{system}]=E[D_{app}]+E[D_{DB}]$

Upper bound for the throughput:

while for the throughput:

$$X \le MIN(N/(E[D_{system}]+E[Z]), 1/[D_{bottleneck}])$$

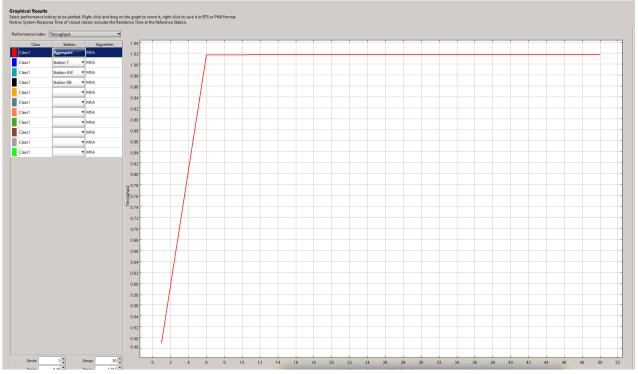
Using one of the two bounds we can compute the optimal number of users:

$$N_{\text{opt}} = (E[D_{\text{system}}] + E[Z])/E[D_{DB}] = ((0.009 + 0.9830) + (1.050))/(0.9830) = 2.0773 \sim 2 \text{ users}$$

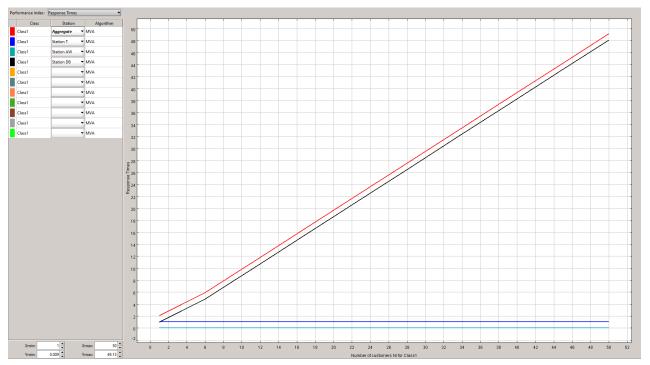
which is the best tradeoff between system throughput,response time and utilization. The reader might think that the optimal number of users is too low given the plots below. However, this number represents the point where the system starts to saturate not where it becomes unstable.

Between 2 and 6 users, the system remains stable but experiences increased latency.





Throughput of the system not optimized.



Response time of the system not optimized.

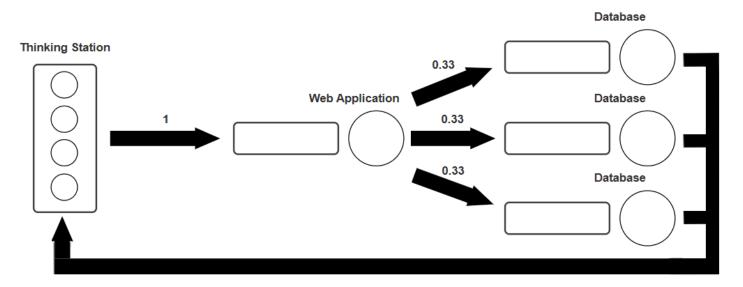


System optimization, analysis and results

Following the literature, we have attempted (and succeeded) in reducing the bottleneck effects by increasing the number of stations of the same type as the bottleneck station.

We kept all settings the same for the additional stations, except for the visit ratio, which was set to 0.33 for each database station.

This value is derived by assigning an equal probability for a request to be routed to one of the three available stations.



Let's compute again the bottleneck and the optimal number of users

the web application has relative visit-ratio equal to $E[V_{app}] = 1$ while,now,each database have $E[V_{DBi}] = \frac{1}{3}$ i=1,2,3

service demand web application: $E[D_{app}] = E[V_{app}]^*1/\mu_{app} = 1^*0.0090 = 0.0090$

service demand i-th database: $E[D_{DBi}] = E[V_{DBi}]^*1/\mu_{DBi} = 0.9830/3 = 0.3276$ i=1,2,3

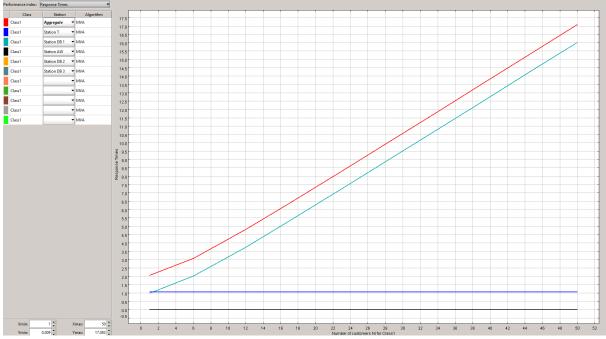
still the databases are the bottleneck but now the service demand is lower which speeds up our system so that:

$$X_{\text{system}} = \mu_{\text{bottleneck}} / \text{E[V}_{\text{bottleneck}}] = (1/0.9830)*3 = 1.0173*3 = 3.052 \text{ job/s}$$

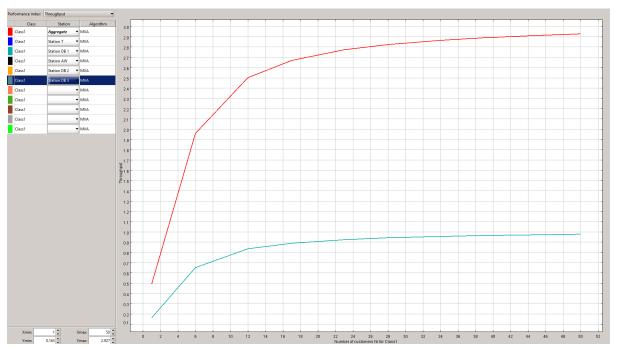
 N_{opt} = (E[D_{system}] + E[Z])/E[D_{DB}] = ((0.009 + 0.3276+0.3276+0.3276) + (1.050)) / (0.3276) = 6.2332 ~ 6 users

Right below the results in graphs:





Throughput system after optimization.



Throughput system after optimization.

As the reader can see, the throughput with three database stations increases by nearly a factor of three; similarly, the response time decreases by approximately the same factor.