

Geographically-aware scaling for real-time persistent websocket applications.

Master's Project in Software Engineering



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Abstract

This section summarises the content of the thesis for potential readers who do not have time to read it whole, or for those undecided whether to read it at all. Sum up the following aspects:

- relevance and motivation for the research
- research question(s) and a brief description of the research method
- results, contributions and conclusions

Kent Beck [[JBB⁺93](#)] proposes to have four sentences in a good abstract:

1. The first states the problem.
2. The second states why the problem is a problem.
3. The third is the startling sentence.
4. The fourth states the implication of the startling sentence.

Chapter 1

Introduction

"Does this solution scale?"

— *Every business person ever*

Everyone these days, probably more than ever, wants to build solutions that scale. Together with the globalization of our lives and businesses grows the need for global, distributed, scalable software systems.

To address this needs, more and more providers offer Cloud Computing solutions [2.1.5](#), giving rise to the new architecture model called *Infrastructure as a Service (IaaS)*. This makes building scalable software systems easier, since the system engineers do not have to design and maintain their proprietary server infrastructure anymore. This also allows to build systems that are more efficient in terms of resource usage - the existing *IaaS* solutions offer scaling up and down capabilities in response to changes in demand for system's services.

As more and more systems are deployed to the cloud, a need for measuring their properties becomes more and more ubiquitous.

In order to reason about scalability-related properties of the system, one needs to define this notion considered by many as vague. For reasoning about cloud scalability, I find the following definition most relevant:

[WG06] Scalability is an ability of a system to handle increased workload by repeatedly applying a cost-effective strategy for extending a systems capacity.

Furthermore, to have a scientific discussion about scalability, quantifiable models and metrics are necessary. Selection of an appropriate scalability measurement model was a big part of my initial study.

1.1 Initial Study

Initial study pointed me to the relevant scalability vectors discussed by researchers and engineers regarding cloud computing. Each of them has a dedicated section in the Background Chapter [2.1](#):

- General Cloud Scalability ([2.1.5](#))
- Data Layer Scalability ([2.1.6](#))
- WebSocket Scalability ([2.1.7](#))

The study yielded multiple scalability measurement models. I analyzed the work of Jogalekar and Woodside [JW00], Srinivas and Janakiram [SJ05], Cooper et al. [CSE⁺10] and multiple others.

Some of the proposed models are not applicable since they were designed for different types of scalability - not for a cloud-based distributed system, but for instance for multiprocessor or multithreading scalabilities.

The models I selected to use for measurements in this research were proposed by Pattabhiraman et al. [GPBT11]. They were designed specifically for the task I am performing - measuring properties of a distributed system deployed in a cloud environment. Their models are also customizable - one can plug in metrics one deems relevant, which allowed us to select metrics that are representative of the business goal of the System Under Test 4.

Selection and application of literature-based models is described in full in Scalability Measurements chapter 7.

1.2 Problem Statement

The problem that I study in this research concerns application stack decomposition of the System Under Test 4. It can be generalized to scalability of systems with similar characteristics:

- utilizing real-time, uncacheable data,
- utilizing user-generated data,
- broadcast to user base distributed globally,
- broadcast using the WebSocket protocol 2.3,
- deployed in the cloud,
- dynamically scalable (up and down).

This regards the size and geographical placement of WebSocket server instances, database and cache server instances (with data distribution), routing between them and internal layer communication.

I propose and compare three architectures 5.6 to find which satisfies scalability criteria best.

More specifically, the host company is in need of scaling their system. The initial research showed little indication of which architecture decomposition will provide lowest client-experienced latencies, best allocated resources utilization and lowest operating cost. On top of that, they do not possess any tools or frameworks that would allow them to measure their system internally and externally. Neither are they in possession of a framework to generate the load test on which the scalability of the system could be measured.

1.2.1 Research Questions

1. What's the architecture decomposition stack that:
 - (a) guarantees data delivery to geographically distributed users with minimal, consistent and manageable latency?
 - (b) scales up and down most effectively in response to demand changes?¹
 - (c) guarantees lowest cost and best degree of utilization of the employed resources within a data center?
2. Does architecture with geographical awareness of its users' distribution provide better Quality of Service than the baseline architecture?

The answer to these questions is sought among the 3 tested architecture, described in detail later 5.6. Answers to these questions are evaluated using the models and a set of metrics described in detail in a dedicated chapter 7, section 7.4.

¹This proved to be untestable as the research progressed, because of the nature of DNS protocol, described in detail 7.4.1.

1.2.2 Solution Outline

In order to answer the Research Questions 1.2.1, a solution consisting of the following steps was developed:

1. **Preparing the load test suite.** It is used to simulate geographical distribution of users and load they generate as well as to collect latency-related metrics of WebSocket handshakes and message delivery. Described in details in a Load Testing Framework section 5.5.2. [TODO] diagram of the distributed load test suite
2. **Preparing the metric collection suite.** It is used to collect data from the application servers. This consists of collecting Amazon's CloudWatch² metrics and our custom metrics (e.g. process memory usage). Metric collection process is described in details in a dedicated section 7.2. [TODO] diagram of the metric collection suite (custom metrics + cloudwatch automatic +)
3. **Feeding the collected metric into scalability meters.**
4. **Automatic scalability management.** This is performed using Amazon's CloudWatch Alarms³ triggering Auto Scaling Groups⁴ capacity changes. In order to include our custom metrics in orchestrating the scalability process, a module based on Amazon's Python library (Boto⁵) was built.

The solution is used to generate realistic user load and benchmark the performance of the system in 3 different versions of varying geographic distributions 5.6.

1.2.3 Research Method

Software Engineering is a relatively difficult field to investigate. Most of the problems are of design (rather than pure scientific) nature. West Churchman coined a specific term for these kind of problems [Chu67]. He calls them **wicked**, since they are resistant to resolutions.

Easterbrook et al. [ESSD08] claim it is often difficult to identify the true underlying nature of the research problem and thus the best method to research it. In their work, they name and compare five most common classes of research methods to select from:

- controlled experiments,
- case studies,
- survey research,
- ethnographies,
- action research.

They help to select the method by first establishing the type of research question being asked:

- existence question - *Does X exist?*,
- description and classification question - *What is X like?*,
- descriptive-comparative question - *How does X differ from Y?*.

The questions of this research are of the last type. The authors [ESSD08] suggest pinpointing up-front what will be accepted as a valid answer to the research question. I answer this question in a separate section 7.7.

²<http://aws.amazon.com/cloudwatch/>

³<http://docs.aws.amazon.com/AmazonCloudWatch/latest/DeveloperGuide/AlarmThatSendsEmail.html>

⁴<http://aws.amazon.com/autoscaling/>

⁵<http://aws.amazon.com/sdk-for-python/>

The detailed description of each of the methods helped me to settle for the **controlled experiment** as a research method for this research. It is well-suited for testing a hypothesis where manipulating independent variables has an effect of dependent ones, which is exactly the case of my research. The manipulated variable is architecture decomposition, and the measured ones are determined by scalability measurement models described in Scalability Measurements 7 chapter.

1.2.4 Research Difficulty

The experiments is performed in a shared cloud environment, which is inherently unpredictable and changing. Non-deterministic network conditions play a significant role here. Aside from that, one has to acknowledge the fact that control over hardware in a cloud environment is by definition given up by the researcher.

Designing and executing a test, yielding statistically relevant results, where all the necessary variables are controlled (within reasonable boundaries) is a huge challenge. Bornholt [Bor14] mentions that even the linking order in the process of generating binaries can have an effect of the performance of the benchmarked system!

The focus of the project is on scaling a websocket application. This is a relatively new technology, thus finding proper scientific coverage is not trivial.

Some of the necessary development and data collection tools are relatively low-level and require thorough understanding of UNIX operating systems, on which the servers are deployed. The same is applicable to tuning the operating system for optimal TCP-related performance 6.1.2.

Simulating the geo-location of clients with reasonable accuracy also proves problematic.

Overall, its a complex project requiring knowledge of multiple aspects of both hardware (TCP, Web-Socket protocol, networks, operating system), software (architecture, cloud computing, data replication/sharding).

1.2.5 Hypothesis

The correct geographical decomposition of the stack of The System Under Test 4 can lead to improved performance in comparison with the Baseline Architecture 5.6.1.

The geographically distributed architectures 5.6.2 and 5.6.3 [TODO insert diagrams] should provide lower user-experienced request handling and broadcast latencies than the baseline architecture 5.6.1. Some of the latencies that are unavoidable due to geographical global distances, in case of these architectures, can be pushed down to lower levels of the systems. This means that the data retrieval / synchronization can happen asynchronously, in the background, rather than synchronously on every user's request.

On the other hand, the distributed versions of the architecture add complexity in multiple system layers. Implications of these will probably increase the cost of system utilization (due to increased internal data transfer).

1.3 Contributions

The research will result in the following contributions:

1. Architecture decomposition analysis for system's meeting the criteria described in the Problem Statement section 1.2.
2. Case Study on Amazon Web Services⁶ and usability of their cloud stack for applications using WebSocket protocol.
3. Open sourced pieces of code for:
 - metrics collection (both from the load test suite and the servers),
 - auto-scaling execution,

⁶<http://aws.amazon.com/>

- constructing scalability models and populating them with collected metrics,
- remote execution of load tests on a fleet of client-simulating servers.

These are described in details in Experiment Deliverables section 5.8.

1.4 Related Work

There is two key papers this research is based on. First of them is *Multi-Cloud Provisioning and Load Distribution for Three-Tier Applications* by Grozev and Buyya [GB14]. The second one is *SaaS performance and scalability evaluation in clouds* by Pattabhiraman et al [GPBT11].

Multi-Cloud Provisioning and Load Distribution for Three-Tier Applications

In this work, the authors tackle the problem of deploying applications across multiple clouds while satisfying system's nonfunctional requirements (in their understanding, application deployed in Amazon's *us-east-1* region and *eu-west-1* region would be deployed to multiple clouds). They propose algorithms for resource providing and load distribution. They focus on resource utilization and cost and legal regulations regarding serving users and servers geographical location and include them in their algorithms.

In order to satisfy the requirements, the authors had to develop their own routing stack. Both Amazon's services involved in the routing process – Route 53⁷ and Elastic Load Balancing⁸ – do not consider application's regulatory requirements when selecting a data center site. Moreover, they do not consider the cost and degree of utilization of the employed resources within a data center.

This is how they perform cloud selection to serve a user:

1. As a first step, the user authenticates with one of their entry point servers. At this point, the entry point has the users identity and geographical location (extracted from the IP address).
2. As a second step, the entry point broadcasts the users identifier to the admission controllers of all data centers. They call this step *matchmaking broadcast*.
3. The admission controllers respond to the entry point whether the users data are present and if they are allowed (in terms of legislation and regulations) to serve the user. In the response, they also include information about costs within the data center.
4. Based on the admission controllers responses, the entry point selects the data center to serve the user and redirects him or her to the load balancer deployed within it. The entry point filters all clouds that have the users data and are eligible to serve him or her. If there is more than one such cloud, the entry point selects the most suitable with respect to network latency and pricing. If no cloud meets the data location and legislative requirements, the user is denied service.
5. After a data center is selected, the user is served by the Application Server and Database servers within the chosen cloud.

Below I list most relevant takeaways from their work.

Sticky Load Balancing - it is a technique of load balancing that routes same users to the same backend server instances. They utilize it in their work - I cannot in this research. Amazon's stack does not offer sticky load balancing for protocols other than HTTP.

Cloud Server Instance Termination - it is not beneficial to terminate a running VM ahead of its next billing time. It is better to keep it running until its billing time in order to reuse it if resources are needed again.

⁷<http://aws.amazon.com/route53/>

⁸<http://aws.amazon.com/elasticloadbalancing/>

Cost Minimization - this is not an automated feature cloud providers offer. Authors advice on how to include it in a cloud controlling stack. It is an important business goal in my research.

Latency Minimization and Approximation - authors, just as myself, treat latency as one of the key objectives in the process of optimizing their cloud stack. They approximate latencies and users geographic location using tools PingER⁹ and GeoLite¹⁰. PingER is used to approximate latencies between user and a cloud and GeoLite to map users IP addresses to geographical coordinates. In my research, I do not need to use tools like that since Amazon's stack offers this in sufficient capacity.

Why is this work relevant? The authors are attempting to solve the same problem - how to distribute and route users between multiple clouds given arbitrary conditions. They work with a similar technology stack as I do.

SaaS performance and scalability evaluation in clouds

Pattabhiraman et al realized that cloud computing and its measurement provides a new set of challenges when it comes to measuring performance, as opposed to measuring performance of traditional software systems. They list key points for measuring cloud applications - validating and ensuring the elasticity of scalability and evaluating utility service billings and pricing models. Both are important in my case. They raise an interesting question regarding the latter:

How to use a transparent approach to monitor and evaluate the correctness of the utility bill based on a posted price model during system performance evaluation and scalability measurement?

The authors divide the performance indicators into three groups and propose formal quantifiable meters for each of them:

1. computing resource indicators (CPU, disk, memory, networks, etc.),
 - Computing Resource Allocation Meter (CRAM),
 - Computing Resource Utilization Meter (CRUM),
2. workload indicators (connected users, throughput and latency, etc.),
 - System Load Meter (SLM),
3. performance indicators (processing speed, system reliability, availability)
 - System Performance Meter (SPM).

They also propose model which allow to combine the above: System Capacity Meter (SCM), System Effective Capacity Meter (SEC), Effective System Scalability (ESS), and Effective Scalable Range (ESR). Values that are fed into these meters are customizable. The meters are described in detail in the Scalability Measurements chapter 7.

Authors perform a case study with Amazon's AWS stack. The case study consists of performed tests and measurements taken on scaling up and scaling down their application. It provides some warnings for me to consider:

- Cloud limitations need to be taken into account. For instance, memory usage data is not collectible from their server instances using their services.
- One needs to be aware of hidden costs that cloud providers excel at introducing.
- One needs pay attention to inconsistencies in performance and scalability data, especially data collected using third-party tools and.

More related work is described in the following chapter - Background 2.

⁹<http://www-iepm.slac.stanford.edu/pinger/>

¹⁰<http://dev.maxmind.com/geoip/legacy/geolite/>

1.5 Document Outline

In this section I outline the structure of the thesis - all chapters with short summaries.

1. **Background** - contains a short summary of the literature study. Describes researched work in the relevant fields [2](#).
2. **Benchmarking** - contains a description and guidelines regarding the benchmarking process, which constitutes a big part of this research. Lists typical benchmarking mistakes and what I do to avoid them [3](#).
3. **The System Under Test** - contains a description and guidelines regarding the benchmarking process, which constitutes a big part of this research. Lists typical benchmarking mistakes and what I do to avoid them [4](#).
4. **Experiment Outline and Design** - contains a description of the conducted experiment, its goals and deliverables [5](#).
5. **Experiment Implementation** - contains a full description of the compared architectures and cloud stack on which the research is performed, including the technological details [6](#).
6. **Scalability Measurements** - contains a description of literature-based scalability metrics and models. Lists the models and metrics selected for measurements in this research, as well as all metrics collectible from the AWS cloud environment [7](#).
7. **Experiment results** - contains the data collected in the experiments [8](#).
8. **Evaluation** - contains evaluation of the experiment and analysis of the results [9](#).
9. **Conclusions** - contains a list of conclusions derived from the analysis of the experiment results [10](#).
10. **Further Work** - contains a description of relevant, testabl variables I did not have enough time to cover in this research. [11](#)

Chapter 2

Background

2.1 Scalability

Scalability seems to be a notion that everyone intuitively grasps, but has difficulties when it comes to clear explanations. In my literature study I came across a few definitions, which might help with that, of which three can be found below:

[WS04] Scalability is a measure of an application systems ability to, without modification, cost-effectively provide increased throughput, reduced response time and/or support more users when hardware resources are added.

[WG06] Scalability is an ability of a system to handle increased workload (without adding resources).

In light of this definition we would talk about a scalability failure if one of the following occurred:

- Address space was exceeded,
- Memory was overloaded,
- Available network bandwidth was exceeded,
- etc.

[WG06] Scalability is an ability of a system to handle increased workload by repeatedly applying a cost-effective strategy for extending a systems capacity.

According to this definition, one could determine system scalability failure if a given resource got overloaded or exhausted an adding capacity to this resource would not result in a proportional ability to handle additional demand, e.g.:

- an additional processor will not contribute to meeting the higher demand if handing of that processor entails an overhead).
- a newly added server instance might not contribute to handling a higher user demand if slows down the routing process.

2.1.1 Scalability Trade-offs

Scalability is generally desired in the software systems, yet, as all architectural software decisions come at a cost [Hei10]; one must be aware of the trade-offs usually associated with it. Weinstock and Goodenough [WG06] and Tsai et al. [THBG12] point these out:

- performance and scalability (non-scalable system will often demonstrate degrading performance with increasing demand, but scalable systems require performance sacrifice on lower usage levels),

- cost and scalability (designing a system to be scalable up-front entails additional costs),
- operability and scalability (it is difficult for humans to operate large systems),
- usability and scalability (it may be possible to increase servable demand with limiting system's service scope - e.g. removing personalization and displaying generic, cacheable data),
- data consistency and scalability (higher scalability can be achieved if system allows for data inconsistencies).
- levels of scalability mechanisms and automated migration (automated migration has to be aware of every level of scalability mechanism; the more of them the more complex the migration),
- workload support and database access (different workloads need support for different database access mechanisms, such as row-oriented storages, column-oriented storages or hybrid approaches).

2.1.2 Need for Scalability

As the globalization and internetization progresses, more and more systems are expected to be capable of serving millions of globally distributed users. A single server instance often cannot live up to that task and thus application needs to be divided and distributed in multiple smaller chunks. This division happens on different application layers, and different parts of the system have to communicate and synchronize with each other.

In case of many software systems (the system under test 4 included), user traffic, and with it the need for system services and resources, varies significantly. A need to be able to scale up and down dynamically in response to traffic arises from this.

Huge parts of the internet are shifting towards real-time. This trend is giving rise to new technologies for exchanging messages between clients and servers in the client-server architecture. Traditionally, client would send a request to a server and receive a response. This is hugely inefficient when there is a need for continuous bidirectional exchange of messages. As an improvement, new mechanisms for server-client communication have been introduced recently 2.3.2 to enhance the process and reduce overhead. One of them - websockets - are a huge next step on this path, but introduce new challenges. One of them is scalability of applications which make use of this technology.

2.1.3 Existing Approaches

Scalability has been studied in parallel computing, distributed computing, databases, and service-oriented architectures (SOA) [THBG12]. Scalable design principles for application servers include: divide-and-conquer, asynchrony, encapsulation, concurrency, parsimony, system partitioning, service-based layered architecture, pooling and multiplexing, queuing and background processes, data synchronization, distributed session tracking, and intelligent load distribution.

There exist multiple scalability vectors for web applications. Different decompositions of application stack can be applied to achieve the goal of scalability. A few traditional approaches of tackling an issue like that exist already [Lei08]:

Scaling up. Increasing volume of system allocated resources (a stronger machine, or with a better configuration, including more computing resource, more memory, higher disk bandwidth and larger disk space [THBG12]). As internet is unreliable and so called "middle-mile bottlenecks" exist, a web application end-user latency and throughput experience is not fully deterministic. Traffic levels fluctuate tremendously, so the need to provision for peak traffic levels means that expensive infrastructure will sit underutilized most of the time. It is easy to implement, but costly, even extremely when you start pushing at current hardware limits.[Qve10] Because the resources of a single machine cannot be increased infinitely, and increase of the cost is not proportional to the increase of the resources, scale-up is not feasible in an Internet scale [THBG12].

Scaling out. Scaling out horizontally - increasing the number of units of resources comprising the system. Cost of hardware can be reduced dramatically this way. In a web application, one can

deploy multiple instances of servers. Different types of balancing can be applied to distribute traffic among them: application layer balancing, business load balancing, and anticipating load. The overhead of parsing requests in the application layer is high thus limiting scalability compared to load balancing in the transport layer. Client state needs to be stored in a layer shared between the web servers.[\[Qve10\]](#) Scale-out is usually adopted in a cloud [\[THBG12\]](#).

Content Delivery Networks. These only handle static assets. Communication in our system [4](#) happens mainly over websockets, which are not supported by CDNs. CDNs can be divided into Big Data Center CDNs and Highly Distributed CDNs, which put application data within end-user ISPs.[\[Lei08\]](#)

Peer to peer networks. An architecture different from client - server, where users communicate with each other directly. It handles adding and removing nodes to and from the network dynamically very well.

2.1.4 Scalability Factors

There exist multiple scalability factors one has to be aware of when designing and implementing scalability of a software system[\[THBG12\]](#).

Levels of Scalability Mechanisms

Most applications deployed in the cloud adopt a tiered application - typically the storage, application and presentation tiers. Cross-tier scaling is not very common because of the high cost of duplicating the whole application stack. Hybrid approach is more popular - the whole stack is duplicated but single tier scaling within the stack is leveraged.

Scalability mechanisms for each tier might be different, since each of them has its own intrinsic properties, objectives and constraints.

Automated Migration

For many systems, automatic migration of data is critical for scalability. This migration can be performed online (the application continues operating throughout the whole process) or offline (the application is taken down for maintenance).

Workload Support

Different workloads require different scalability mechanisms. Online Analytical Processing (OLAP) applications have to be prepared to scale for high volume of read operations, whereas Online Transaction Processing (OLTP) applications have to be prepared to scale for a lot of write operations. Mixed workloads (as is the case for The System Under Test [4](#)) require an architecture ensuring no biases towards either type of operations.

Recovery and Fault-tolerance

A scalable system needs to detect failures of nodes and scale down to exclude the failed node from operation. Similarly, when the failed node comes back to the system, it should scale up including the recovered node. In a fully scalable system, this process should be autonomous.

Software Architecture

In a scalable system, architecture should be divided into decoupled modules which can be scaled independently. Authors [\[THBG12\]](#) use Service-Oriented Architecture as a model example of an architecture that many scalable systems adopt.

Database Access

Accessing data is often time consuming. This is why data layer is likely to be a bottleneck of a SaaS application. Indirect access (accessing the data layer through an exposed API) allows data tier and other tiers to scale independently using their own preferred mechanisms.

2.1.5 Cloud Scalability

Cloud computing has become virtually ubiquitous. All biggest internet services are either deployed to a cloud, or run their proprietary cloud systems. A lot of companies running proprietary clouds also make them available for hosting to the external clients. Netflix¹ and Spotify² (both deployed to Amazon's AWS) are examples of the first approach, with Microsoft(Azure)³, Google(Google Cloud Platform)⁴ and Amazon(Amazon Web Services)⁵ being the example of the second. These services are typically billed on a resource utility basis.

General cloud capabilities and the cloud stack I am working with for the scope of this project is described in a dedicated chapter 6.4.

One of the biggest advantages of deploying one's architecture to the cloud that aforementioned companies offer is the ability to dynamically scale up and in response to application traffic changes.

As Grozev and Buyya [GB14] put it, to fully facilitate cloud capabilities, software engineers need to design for the cloud, not only to deploy to it.

2.1.6 Data Layer Scalability

Data layer scalability is an important part of system scalability. In a distributed system, multiple system agents share the data. There exist multiple strategies for operating on a shared, distributed data layer.

Data layer is often a performance bottleneck because of requirements for transactional access and atomicity - it is hard to scale out when system uses a relational data store [GB14].

CAP Theorem put forward by Eric Brewer [GL02] states that it is impossible for a distributed system to provide all of the following guarantees:

- consistency,
- availability,
- partition tolerance.

Therefore, a trade-off needs to be made, depending on stakeholder priorities, which of the three to give up.

ACID stands for Atomicity, Consistency, Isolation, Durability and is a set of properties guaranteeing reliable processing of database transactions. It has been a guideline in designing multiple database systems. The term was originally coined by Haerder and Reuter in 1983 [HR83].

BASE stands for Basically Available, Soft State, Eventually Consistent [LD12]. Its a complement of ACID. Author claims we lack precise metrics to measure BASE aspects and thats why every system implements eventual consistency differently.

Most importantly for this research, the author [LD12] explains *MySQL Cluster* - it performs much like an ACID database but with the performance benefits of a cluster. Aside from that, *MySQL Replication* can be put to use. It can be configured at multiple topologies, not only the basic master-slave; consistency differs per configuration.

¹<https://aws.amazon.com/solutions/case-studies/netflix/>

²<https://aws.amazon.com/solutions/case-studies/spotify/>

³<https://azure.microsoft.com/>

⁴<https://cloud.google.com/>

⁵<https://aws.amazon.com/>

Many solutions and strategies for dealing with database scalability have emerged, including *NoSQL* and *NewSQL* databases, data replication and sharding [Amz02].

Cooper et al touch on that subject in their work [CSE+10]. They claim that in scaling out the database layer one should aim for elasticity (dynamically adding capacity to a running system) and high availability. These are hard to achieve using traditional database systems. They show that new protocols are being developed to address that issue, such as that: *two-phase commit protocol* (provides atomicity for distributed transactions) and *paxos*.

The authors [CSE+10] also give an overview of different database systems, including *PNUTS*, *BigTable*, *HBase*, *Cassandra*, *Sharded MySQL*, *Azure*, *CouchDB*, and *SimpleDB*.

In their work [CSE+10], we can find enumeration of classic data-related scalability trade-offs:

- read performance and write performance,
- latency and durability,
- synchronous and asynchronous replication,
- data partitioning (column and row-based storage).

The technology stack I have been working with in the scope of this project consists of *MySQL* as persistent storage and *Redis* as a key-value cache. It is described in details in a dedicated chapter 4.4.

MySQL Scalability

MySQL White paper [Ora15] gives us a good overview of how scalability works in case of MySQL Cluster. Authors suggest starting the design of the scaling process by identifying which characteristic the application possesses: lots of write operations, real-time user experience, 24x7 user experience or agility and ease-of-use. The System Under Test 4 falls into the first and second categories.

They claim to support (among others) auto-sharding for write-scalability, active / active geographic replication and online scaling and schema upgrades. Geographic replication offers distribution of clusters across remote data centers, which they claim helps reduce latency (which is extremely important case of The System Under Test 4).

Authors show how MySQL Cluster is optimized for real-timeness:

- data structures are optimized for in-memory access,
- persistence of updates runs in the background,
- all indexed columns are stored in memory.

According to the white paper [Ora15], MySQL Cluster is very well-suited for on-line, on-demand scaling.

MySQL can also be used to scale in unconventional ways. Ruffin et al, in Social-Data Storage-Systems [RBR11], mention that Twitter uses MySQL as a key value store.

They show how MySQL can be scaled horizontally by both sharding and data replication. They also indicate that the more structured the RDBMS data, the harder it is to scale horizontally. MySQL is optimized for writes, since only one record in one table is touched, whereas reads can prove expensive if they contain joins, especially spreading across multiple cluster nodes. Facebook and Twitter solved it by putting a cache on top of MySQL [RBR11].

Redis Scalability

Single Redis installation is said to be of limited scalability, because of the fact that for good performance the whole data set should fit into memory [RBR11]. Redis offers cluster⁶ and replication⁷ solutions.

In a Redis Cluster, data is automatically sharded (not replicated) among multiple nodes. One has control over sharding; it is possible to ensure that certain data points end up on the same (or on

⁶<http://redis.io/topics/cluster-spec>

⁷<http://redis.io/topics/replication>

a given) node. Redis Cluster does not guarantee strong consistency - under certain conditions the Cluster can lose writes that were acknowledged to the client. Support for synchronous writes exists, through the command *WAIT*, which highly (but not entirely) decreases the likelihood of lost writes. Nevertheless, using it is discouraged unless absolutely necessary.

Redis can also be scaled for reads using a simple replication in a single master - multiple slaves topology.

2.1.7 Websocket Scalability

An introduction into push-base communication over the internet can be found in Agarwals work - Toward a Push-Scalable Global Internet [Aga11]. The key message in the article is that push message delivery on the World Wide Web is not scalable for servers, intermediate network elements, and battery-operated mobile device clients. And yet, most of modern day websites have highly dynamic content updated even up to multiple times a minute (very much so for The System Under Test 4).

Most of internet communications happens over HTTP Running over TCP. Real-time message delivery requires an always-on connection from the server to the client. HTTP proxies have limited memory and TCP ports, are shared among multiple users. Servers need to be provisioned in order to maintain active TCP connections from large populations of user clients. These all provide challenges that need to be dealt with in scalable applications [Aga11].

A number of TCP ports available (and file descriptors available to a server process) on a server instance is an inherent scalability limitation of an application using the WebSocket protocol. Hardware limitations are different in an HTTP-based communication.

On a more positive note, Cassetti and Luz [CL12] claim that overhead introduced by the websocket protocol and websocket API is rather small as compared to other communication methods. Furthermore, one data intensive applications can achieve superior bandwidth and performance when using websockets.

My own research has proven preparing websocket-based communication scalability to be difficult and tedious. Most of the cloud scalability stack that is available and that I have been working with was designed and built to support http-based – and not websocket-based – communication. On top of that, one needs to tweak the host operating system kernel in multiple ways to increase the number of concurrent websocket connections the system can maintain. Details of the tuning process are described later 6.1.2.

The Websocket Protocol is described in details in a dedicated section 2.3.

2.1.8 Measuring Scalability

Measuring scalability proves a challenge, since there is no single physical quantity or unit which the community has accepted as a scalability measure.

Measuring and researching in a cloud environment is difficult, because, by definition, the researcher has no control over the hardware that his system is running on. Clouds are also inherently non-deterministic. Nevertheless, as Sobel et al. describe in their work [SSS⁺08], even in cloud computing environments, where researchers have little control over network topology or hardware platform, understanding the performance bottlenecks and scalability limitations imposed by the offered infrastructure is valuable.

As this is a crucial topic to my research, this topic is explored in detail in a dedicated chapter 7.

2.2 Geographical Distribution

With the rise of global internet services, single applications have to server users who are distributed around the whole globe. With the distribution of the users, comes distribution of the data. Multiple solutions exist that, through data distribution, aim at improving the *Quality of Service*.

The topic is nicely introduced by Tom Leighton in his article *Improving performance of the internet* [Lei08]. He provides a few arguments why it is important to keep data as close to end users as possible. Apart from obvious latency benefits, by doing this, one reduces the chances of suffering from a big *middle mile* provider outage. He introduces a scale to reason about internet locality.

Scale Name	Range	Latency Range	Typical Packet Loss
local	less than 100 miles	1ms	0.6%
regional	500 - 1000 miles	15ms	0.7%
cross-continent	3000 miles	50ms	1.0%
multi-continent	6000 miles	100ms	1.4%

Table 2.1: Internet Locality

As the table 2.1 shows, the longer data must travel through the middle mile, the more it is subject to congestion, packet loss, and poor performance [Lei08]. This is why companies try to locate servers close to end users (on a scale that cannot be reproduced in this research - requires finer control over infrastructure). The lowest granular scale I am able to work with within this research is *regional*. Leighton writes that big websites have at least two geographically dispersed mirror locations to improve performance, reliability and scalability.

In his work, he includes a set of guidelines to consider when performing geographical scalability:

- reduce transport layer overhead (I am achieving this with usage of the websocket protocol),
- prefetch embedded content,
- assemble pages at the edge (even on an end client machine),
- offload computations to the edge,
- ensure significant redundancy in all systems to facilitate failover,
- use software logic to provide message reliability.

Ed Howorka in his interesting paper *Colocation beats the speed of light* [How15] about trading system geographical distributions focuses on the best placement of servers for traders trading on multiple exchanges. His paper demonstrates that traders gain nothing by positioning their computer at the midpoint between two financial exchanges. He claims that every algorithm on a central machine talking to surrounding servers (users in case of the System Under Test 4) can be replaced by colocated servers (located in geographical proximity to users). Furthermore, he implies and sets out to prove that server colocation provides a better solution for high-speed applications (as opposed to using a big, centralized server located in the middle).

2.3 WebSocket Protocol

WebSockets is an independent, TCP-based communication protocol. The protocol was standardized by the Internet Engineering Task Force (IETF) in 2011⁸ and has been gaining popularity ever since. All major web browsers and mobile operating system support it network⁹.

2.3.1 Technical details

Every WebSocket communication has to start with an opening handshake. WebSocket is a protocol related to HTTP, in a sense that its handshake is interpreted by HTTP servers as an *Upgrade request*. Websocket also operates on the same ports as HTTP (80 and 443), so as not to get the communication blocked by all the internet's intermediaries configured e.g. to allow exclusively HTTP traffic.

A sample handshake included in the client's request is depicted below. It is a simple GET request. A set of random bytes - **Sec-WebSocket-Key** - needs to be included.

The server takes the **Sec-WebSocket-Key**, appends a globally unique identifier (GUID) string, computes a SHA1 hash from it and Base64-encodes it. The computed value is sent in the response header confirming the upgrade as **Sec-WebSocket-Accept**:

⁸<https://tools.ietf.org/html/rfc6455>

⁹<http://caniuse.com/#feat=websockets>

```

1 GET HTTP/1.1
2 Upgrade: websocket
3 Connection: Upgrade
4 Host: echo.websocket.org
5 Origin: http://www.websocket.org
6 Sec-WebSocket-Key: i9ri'Af0gSsKwUlmLjIkGA==
7 Sec-WebSocket-Version: 13
8 Sec-WebSocket-Protocol: chat

```

Listing 2.1: WebSocket Upgrade Client Request

```

1 HTTP/1.1 101 Web Socket Protocol Handshake
2 Upgrade: WebSocket
3 Connection: Upgrade
4 Sec-WebSocket-Accept: Qz9Mp4/YtIjPccdpbvFEm17G8bs=
5 Sec-WebSocket-Protocol: chat
6 Access-Control-Allow-Origin: http://www.websocket.org

```

Listing 2.2: WebSocket Upgrade Server Response

That leads to opening a websocket (*ws://*) communication channel between the client and the server. Secure websocket connections (*wss://*) are also possible.

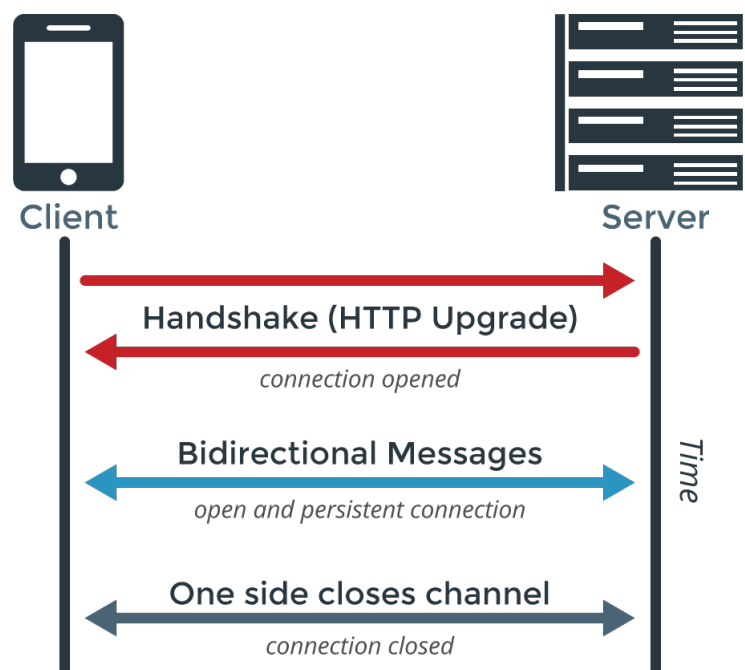


Figure 2.1: WebSocket Protocol

Typically mentioned benefits of using WebSocket protocol include reduced overhead in comparison with HTTP communication. After the initial handshake, only application-specific data travels between client server, as opposed to HTTP where every request/response contains headers, cookies, content type, content length, user-agent, server id, date, last-modified etc.

WebSocket protocol has an another advantage over HTTP on a TCP protocol level (on which both are based) - TCP connection only needs to be opened once per WebSocket communication, whereas for HTTP it is opened for every request (which introduces overhead).

What is more, HTTP servers are often configured to persist in a log form the start and completion of every HTTP request. This entails additional CPU cycles and costly I/O operations. Unless an

application is configured to log its proprietary data, this additional logging cost is much smaller in case of WebSockets (since there is only one initial request).

Configuring a server for high-scalability WebSocket communication is no easy task, as I have learned the hard way during this research. The whole process is described in details in a dedicated section [6.1.2](#).

2.3.2 Usage and Origins

Multiplayer online games and real-time applications with a lot of user generated content are types of application that benefit most from popularization of the WebSocket protocol.

One of the biggest advantages of websockets is that the server can easily send unsolicited messages to the client, rather than simply respond to client's requests (as is the case with HTTP protocol). Before websockets, other solutions were being developed to enable bi-directional client-server communication, yet none of them has gained such popularity and is considered as elegant.

Before websockets, the applications mentioned above had to resort to other, less efficient ways to facilitate bi-directional client-server communication.

Agarwal presents us with an overview of protocols previously used for the same kind of communication [[Aga11](#)]:

AJAX - asynchronous JavaScript and XML, a request/response model; when originally introduced, it was a huge improvement, since the page didn't have to be refreshed anymore to get new data. Still used widely, but for a slightly different purpose than the WebSocket Protocol.

Long Polling - consists of the client sending a single request and the server waits until it can provide the response (or times out), which does not create much traffic but uses a lot of server resources. Connections are artificially kept open and clients need to reconnect periodically.

Short Polling - also called AJAX-based timer, consists of the client repeatedly (timer is used to regulate that) sending the same request to the server (polling), which creates a lot of useless traffic.

webRTC - Web Real-Time Communication, enables audio, video and text communication between users using web browsers ^{[10](#)}.

Server-Sent Events - a stream of events generated by the server, to which a client subscribes. Communication is impossible in the other direction^{[11](#)}.

¹⁰<http://www.webrtc.org/>

¹¹https://developer.mozilla.org/en-US/docs/Server-sent_events/Using_server-sent_events

Chapter 3

Benchmarking

"If you're not keeping score, you're just practicing."

— Vince Lombardi, American Football Player

As this project concerns benchmarking different architecture decompositions of The System Under Test 4, it is important to define and describe the benchmarking process.

Benchmarking - the term is most commonly defined as an execution of a set of operations against a given object / program, in order to determine its relative performance. Usually, it consists of a set of standardized tests. The same term is often used to describe benchmarking programs themselves. It is crucial for researchers, tool developers and users [BLW15].

As James Bornholt describes in his article [Bor14], computer science research is among the most non-deterministic. A huge risk of omitted-variable bias exists. This is due to the fact that computer systems have a tremendous number of dynamic parts (both in hardware and software). It is virtually impossible for researchers and other experts to have complete knowledge of every bit of the system they are working with. Controlling all the involved variables is insurmountable.

Furthermore, many of the system parts factors are non-deterministic; e.g. networks and multi-threading. Because of that, researchers need to be very careful when attempting to draw statistically meaningful conclusions from their research.

Benchmarking a distributed system is even more complex [Bor14]. If it is hosted in a cloud environment, the control over the hardware is in the hands of the cloud provider, not the researcher. This is the situation I am in performing this research.

Bornholt [Bor14] lists some of the many environmental factors that can have influence on the performance of a computer system. The most striking example is the *linking order*, which can significantly bias the results of a benchmark. As UNIX systems are highly customizable, it is important to keep the values of environmental variables as consistent as possible across benchmarking sessions.

Author also states a need for the scientific community to unify and standardize the tools for benchmarking systems. A lot of researchers rebuild their infrastructures and build their scripts anew on every project. This is where I hope to contribute to the scientific and open source communities with releasing all the possible parts of the tooling build for the purpose of this project (described in details later 5.8).

For a benchmark to be dependable, results need to be reproducible [Bor14]. That means the same results can be obtained rerunning the benchmark later on a machine with the same hardware and the same software versions. Reproducibility of experimental results requires measurements to be reliable. A measurement can be called reliable, if the method guarantees accuracy (small systematic and random measurement error, i.e., no bias or volatile effects, resp.) and sufficient precision.

Beyer et al [BLW15] list well-established benchmarks for specific tasks:

SPECi - The Standard Performance Evaluation Corporation¹. They provide benchmarks for CPUs, Graphic Cards, Java Client/Servers, Mail Servers and others.

¹<https://www.spec.org/>

TPC - Transaction Processing Performance Council². They provide benchmarks for Transaction Processing (OLTP), Decision Support, Virtualization and Big Data.

nlrpBENCH - Natural Language Requirements Processing³. They provide benchmarks for Requirements Engineering.

In case of the system being benchmarked in this research [4](#), it is hard to reuse any of these since the performance depends on a custom set of operations spreading across multiple system layers.

Authors of "Benchmarking cloud serving systems with YCSB" [[CSE+10](#)] point out that it is important to have random distributions for benchmarking loads - *uniform*, *zipfian*, *latest*, *multinomial*, etc.

From my own experience, I have drawn that against a complex system like our utilizing a wide stack of technologies, benchmarking tool should do a few "dry runs" in order to give caches a chance to fill up, internal tables to get populated or cloud components to be warmed up (e.g. Elastic Load Balancers in case of our stack [6.4](#)).

One also has to be aware of what the potential congestions are. For instance, without properly tuning the operating system's TCP stack, one has to wait a few minutes between load testing runs to let all the TCP sockets return to a usable state (that is exit the TIME_WAIT state, unless the operating system is tuned to reuse such sockets). Details of the process of tuning are described in detail later [6.1.2](#). In the process of testing, I have also realized our API Server takes a substantial amount of time to handle disconnections by huge number of clients (test suite disconnects all of them at the same time). That means there needs to be a non-zero pause between load test suite reruns.

3.1 Benchmarking Crimes

Gernot Heiser prepared a list [[Hei10](#)] of *benchmarking crimes* which should be avoided when attempting to benchmark a system.

Selective benchmarking

1. Not covering the full evaluation space.
2. Not evaluating potential performance degradation.
 - (a) **Progressive Criterion** - performance actually does improve significantly in area of interest.
 - (b) **Conservative Criterion** - performance does not significantly degrade elsewhere.

For a benchmark to be reliable, both criteria need to be demonstrated. I trust the models that have been selected to reason about system properties [7.4](#) are living up to that task.

As an economist Milton Friedman used to put it, *there's no such thing as free lunch*⁴.

In this context, we can interpret it in the following way. Techniques improving performance in some capacity usually entail an additional cost (extra bookkeeping, caching). As Heiser writes: "This is really at the heart of systems: it's all about picking the right trade-offs" [[Hei10](#)]. In our case the potential improved performance comes at a cost of increased data transfers between data centers (and therefore costs) and simply increased complexity of certain system layers.

3. Subsetting a benchmark without strong justification. A warning sign that one might sin in this way:
 - "we picked a representative subset", "typical results are shown" - reads as "we cherry picked the data to fit our expected results".

²<http://www.tpc.org/>

³<http://nlrp.ipd.kit.edu/>

⁴<http://www.amazon.com/Theres-Such-Thing-Free-Lunch/dp/087548297X>

If benchmarking is performed on a subset of a system, a strong justification needs to be as to why a given subset of functionalities / components was selected. I am benchmarking a real-life system use case.

4. Selective data set hiding deficiencies. Luckily, more and more statistical tools are being developed to detect that ⁵.

Using the same dataset for calibration and validation

The way to avoid this is to calibrate the system first (using calibration workload). Then, one should use evaluation workload to show how accurate the model is. Both workloads need to be disjoint.

Providing o indication of significance of data

An example of this fallacy would be providing raw averages, without any indication of variance. At least standard deviations must be quoted. If in doubt, Heiser recommends using student's t-test to check significance [Hei10].

Benchmarking of simplified simulated system

I am performing the tests on the full system deployed in the exact same environment as the production system.

Using inappropriate and misleading benchmarks

The tests are being performed on different versions of the proprietary system. The business goal of the host organization of the research is to find the best architecture - the benchmarking process is supposed to help with this. There is no incentive to be dishonest.

Unfair benchmarking of competitors

The aforementioned applies.

Providing relative numbers only

Heiser suggests this reads as "I am covering that the results are really bad or irrelevant" [Hei10]. In the case of this research, some results need to stay relative since this is what the selected scalability measurement meters 7.4 require. Absolutes might not make sense in this context.

Providing no proper baseline

Often the state-of-the-art solution can be used as a baseline. I believe the solution we picked as a baseline (the simplest architecture decomposition) is a good one. The research leading up to the test did not yield any data to prove that statement wrong.

Using arithmetic mean for averaging across benchmark scores

Wallace and Fleming [FW86] point out that this is an incorrect way to approaching averaging. The proper way to do this (i.e. arrive at a single figure of merit) is to use the *geometric mean of the normalised scores*.

⁵http://dataskeptic.com/epnotes/ep55_detecting-cheating-in-chess.php

Chapter 4

The System Under Test

A system under test can be best described as a *real-time, dynamic, parimutuel prediction market*. It is a platform with a publicly exposed RESTful and WebSocket APIs ¹, through which system clients can connect their websites and mobile applications to access functionality. The scope of this project concerns the WebSocket API.

The core functionality of the platform is to allow users to make predictions on (near) future events. Users can also ask their own questions regarding these events and send them out globally or to their friends. The events can literally be anything - users' imagination is the limit. A simple use case of the system would be a live football match or an e-game match broadcast on a streaming service like *twitch.tv*². Questions can also be automatically generated based on an incoming stream of events regarding the e-game/sports match. The stream can consist of an "admin" inputting the events manually through the system's admin interface, or, if a game/match has an automatic stream of data available, it can also be connected to the platform.

What is important for this research, is that questions (and predictions users make on them) are organized as parimutuel pools. That means that every prediction in the pool affects the future distribution of winnings. Pennock describes it in details in his work on dynamic parimutuel markets [Pen04].

In order to keep the users informed on what the current distribution is (and thus what are the multipliers for each of the available options), the updated distribution needs to be broadcast as quickly as possible to all interested users (and reach them around the same time). In a sense, every user is therefore both a consumer and a producer of data. In case when users are globally distributed this provides a set of data-distribution and latency related challenges on which I focus in this research.

4.1 Sample Use Case

As mentioned above, one of the simplest use cases of the platform is a live football match. A client, who has connected his application to the platform can now offer users real time questions and placing predictions on them.

A case illustrating the need for the research well would be *UEFA Champions League*³ final, for instance between *Real Madrid* and *FC Barcelona*. These events generate viewership in the range of hundreds of millions ⁴.

Let's assume we have 10 million users connected to the platform throughout the game. Most of them come from Spain and the rest of Europe (65%). Since the time-zone suits football-loving South Americans well, a lot of users are connected from there too (20%). The platform also hosts some North American fans (5%), with the rest comprised of hard-core Asian (9%) and Australian (1%) fans.

¹<http://instamrkt.github.io/slate/>

²<http://www.twitch.tv/>

³<http://www.uefa.com/uefachampionsleague/>

⁴<http://www.uefa.com/uefachampionsleague/news/newsid=2111684.html>

The company providing a real-time data feed for the game have their servers located in a UK-based data center.

This raises a few interesting questions that are explored within this research:

1. Where should WebSocket server instances, which distribute messages, be spun up for minimal user-platform latencies and optimal general system performance?⁵
2. How big should the instances be?
3. What should be the geographic distribution of the instances? How should users be routed to them?
4. What should be the data-layer architecture? Should the data be centralized or distributed? Distributing using replication, sharding, or some other clustering technique?

4.2 Users of the System

The users have a possibility to connect to the platform from any mobile and desktop device that supports WebSocket communication protocol. That means they can use the platform during live games in stadiums (more and more of which provide good network connectivity), watching sports on TV or e-gaming streams from their couch or in bars and other gatherings with their friends.

That means the platform has to be capable of working efficiently not only on cable and strong wi-fi networks, but also on 4g and 3g networks to facilitate mobile usage. As the experts at AT&T inform⁶, latency is more of an issue on wireless networks than on wired ones. This is because wireless network is more constrained in size and scope than a wired one. The wireless connections there need to be managed more carefully. Most usage is expected to come from mobile networks (which often drop), so reconnecting them quick to the right instance is of huge importance.

Due to the nature of the platform, the demand is extremely lumpy and comes in a few-hour-long spikes for the big events.

4.3 Basic Architecture

The most basic architecture, in its non-distributed form, of the main WebSocket API system components is depicted on the figure 4.1.

The system consists of the following units:

Event Server - receives and processes the external game/match related stream of events. Notifies the WebSocket API server so that it can broadcast the updates to the users.

Resolver - upon an arrival of a new event, collects all the outstanding pools related to it, resolves accordingly, calculates winnings, updates accounts. Notifies the WebSocket API server so that it can broadcast the updates to the users.

WebSocket API Server - handles all end-user communication. Receives requests, sends out response and broadcasts game, prediction pool and all other updates. User's subscribe to game-specific channels in order to receive updates.

4.4 Technology Stack

The platform is written mostly in Python 2.7.6. Data storages used by the platform consist of Redis 2.8.10 as a caching solution and Mysql 2.8.6 as persistent storage. A sample client application is written in Ampersand.js⁷. Cloud stack used to which the application is deployed is fully described later 6.4.

⁵Optimal as defined in the Scalability Measurements chapter 7.

⁶<http://developer.att.com/application-resource-optimizer/docs/best-practices/multiple-simultaneous-tcp-connections>

⁷<http://ampersandjs.com/>



Figure 4.1: Basic System Architecture

4.5 Web Server

We use [nginx](http://nginx.org/)⁸ as a web server solution, which handles encrypted communication and forwards the traffic to the backend servers. If there are multiple backend servers deployed on a single machine, nginx handles the load balancing between them (using a simple round-robin routing policy).

4.6 Message Handling and Message Publishing

The API server serves two main purposes: handling and responding to user requests, and broadcasting messages with state updates that these requests (and the system internally) generates.

These two functionalities is what my load testing focuses on. Request handling simply happens on a WebSocket protocol level, whereas publishing is done using Redis' pub-sub functionality. Users subscribe to different games - the pub-sub channels. Every channel has a listening thread, that picks up on the message and distributes it to clients connected over WebSockets, as shown in the code snippet [INSERT CODE SNIPPETS ON PUBLISHING (HOW SELECTION IS DONE BY EACH API SERVER)].

4.7 Unique Aspects of The System

There are multiple unique aspects of the system that make this research interesting:

- Critical data is communicated using the WebSocket protocol.
- Users receiving data shared locally should receive it at the same time as data shared globally (with as small a latency as possible). This creates a need for globally consistent and manageable latency between end user and the system.
- Connected users (who serve both as consumers and producers of data) and sources of data are geographically changing. Users geographical center of mass is changing for each peak of demand.

⁸<http://nginx.org/>

- Demand is extremely lumpy. This creates a need for being able to quickly the system up and down.
- New data is generated constantly by the users so caching the content for geographical distribution is difficult.

Chapter 5

Experiment Outline and Design

5.1 Goal of the experiment

The goal of the project is to design a scalability framework for a real-time persistent websocket distributed application (the system). The core researched topic is whether a systems awareness of clients geographical distribution can improve the system performance according to selected metrics, in comparison with traditional approach [5.6.1](#).

The goal of the project is to see if the proposed architecture decompositions [5.6.2](#), [5.6.3](#) can perform better (quantitatively, according to the selected meters) than the baseline architecture in serving geographically dispersed clients. Approaches used to scale simple http applications cannot always be translated to websocket applications since the communication protocol differs. Websockets put a different kind of strain on the server machines since these need to keep the connection opened on a port for a prolonged period of time rather than simply open, server and close (as is the case with http). Along the way, an answer needs to be found what level of decoupling provides best performance on each layer of a stateless persistent system. One of the properties of a system of that kind is that key value stores come under heavy load since this is where the state resides. A good solution for distributing (sharding / replicating) these also needs to be found. Same goes for persistent storage.

5.2 Problem To Be Solved

The host company is facing the problem of scaling up their system. It has been designed in a way that should support scalability - all processes are decoupled and communicate over TCP sockets, using the ZMQ library [TODO reference zmq here]. Most internal messaging happens using pub-sub architecture. Thanks to that, new nodes can be added to the system in order to increase its capacity.

Even though the system has been designed for scalability, it has not been battle-tested. The scalability process is performed in this research; multiple system limitations and have been identified [6.1.1](#) and solutions proposed and implemented.

Our (the host company) preliminary research has not yielded definitive answers as to which level of geographic architecture decomposition will result in best system performance. The key metrics influencing these performance are user-experienced WebSocket handshaking, message handling and broadcasting latencies, utilization of the allocated cloud resources and utilization cost.

Furthermore, the organization does not possess any tools to measure application performance. This research provides the organization with a set of tools to measure any component of the system (in terms of resource usage and processing time), store the metrics and visualize them in real-time in a comprehensible way.

On top that, my research delivers the design and implementation of the load testing suite, including cloud client operating system images, orchestration scripts for simulating the load testing from the clients and collecting the metrics onto a master machine.

[TODO image - sample graphite dashboard]

5.3 Question to be answered

1. What's the architecture decomposition stack that:
 - (a) guarantees data delivery to geographically distributed users with minimal, consistent and manageable latency?
 - (b) guarantees lowest cost and best degree of utilization of the employed resources within a data center?
2. Does architecture with geographical awareness of its users' distribution provide better Quality of Service than the baseline architecture?

The answer to these questions is sought among the 3 tested architecture, described in detail below 5.6. In short - I propose 3 architectures of increasing level of geographical distribution (= decreasing level of centralization) and then use the load test and measurements suite to assess which performs best.

Answers to these questions are evaluated using the models and a set of metrics described in detail in a dedicated chapter 7, section 7.4.

5.4 Research Method

The research method I selected is *Controlled experiment* 1.2.3. The *controlled* part is somewhat tricky in the cloud environment. The intrinsic property of cloud computing is that the researcher has to give up control over hardware to the cloud provider, thus fully controlling it cannot be in scope of a project such as mine. Nevertheless, I did my best to keep constant all variables I have influence over and measure and control all the rest.

Variables constant across benchmark runs against different architectures include:

- Number of connecting users ([TODO] exact number),
- Pace at which users connect ([TODO] exact number),
- Number of requests that need to be handled (with some random, controlled variance),
- Number of updates that need to be broadcast (with some random, controlled variance),
- The length of the benchmark execution,
- Selected instance types (with controlled variance, different decompositions require different components)

Variables over which I have no full control:

- allocated network bandwidth for cloud instances,
- spikes in resource allocation on shared cloud instances (e.g. Allocated CPU cycles) (this happens mostly for micro instaces ¹, I use bigger instances),
- general global network conditions (benchmarks are run at the same time of a day, at times [to my best knowledge] with no major global events which could disrupt network conditions).

5.5 Experiment Outline

In this section I list all the steps necessary for the experiment to be prepared and executed.

5.5.1 Pre-experiment preparations

This consists of two main parts: preparing the load test suite and the deployed system.

¹http://docs.aws.amazon.com/AWSEC2/latest/UserGuide/concepts_micro_instances.html

Load Test Suite

The following steps are necessary to deploy the load test suite:

1. copy the load test client instance image to all AWS regions,
2. for each region, start the desired number of instances using the image.

The System

This varies slightly depending on the deployed architecture; below I list the general outline:

1. prepare the clean server (and database if necessary) instance image and distribute it across relevant regions (clean means no unnecessary data stored and no redundant services running),
2. start the instances,
3. clean redis and mysql and populate with data necessary for the benchmark run,

Once all the client and server instances are brought up and reachable (verifiable through AWS API), the load test can begin.

5.5.2 Load Testing

Simulated clients come from instances located in all available Auto-Scaling AWS regions ², except for eu-central-1 (Frankfurt) 6.1.7.

The load testing process is enumerated below:

1. The single master node orchestrating the suite collects IP addresses of the client instances,
2. it connects to all the client instances (using a library utilizing ssh) and issues a *start test* command,
3. it sleeps for a specified amount of time,
4. it connects to clients again and kills the client processes,
5. it sleeps for a small amount of time to let client processes close all connections, process collected metrics and save them to a temporary file,
6. it connects to clients again and copies the file to master node where metrics are aggregated.

5.5.3 Users Distribution

Since this is going to differ for a type of game served on the platform, I decided to use the most basic distribution - the same number of users connecting from each of the 8 available regions: North Virginia, Oregon, North California, Ireland, Singapore, Sydney, Tokyo and Sao Paulo.

5.5.4 Used Benchmarking Scenario

The benchmarking scenario is the same for each tested architecture. As mentioned before, clients connect from 8 globally-distributed regions, 15 instances in each region, each with the same number of clients. Each instance connects one client a second, so in total $8 * 15 = 120$ are connected every second until the full load is reached. [TODO] specify the full load here.

Once all the users have connected, the benchmark runs for exactly 90 minutes and then all the clients disconnect.

Upon connecting, each users lists available games (which there is one during the benchmark run), *thinks* between 0 and 15 seconds, joins a game (to subscribe to all game updates) and lists available prediction pools for the game (which there is also only one for the benchmark run). Once this is

²<http://docs.aws.amazon.com/general/latest/gr/rande.html>

completed, each users places a prediction every (0,30) seconds (once the prediction confirmation is received, the counter is reset). This continues until the benchmark run is finished.

The server responds to each user request and sends out an unsolicited pool distribution update every 3 seconds.

```
1 var _ = require('underscore');
2
3 // max delays in milliseconds
4 var MAX_GAME_SELECTION_DELAY = 15000;
5 var MAX_PREDICTION_PLACING_DELAY = 30000;
6
7 // called upon reception of the list of active games by the client
8 var handleActiveGames = function(msg) {
9   // ... handle reception of list of active games
10
11   // simulate user thinking and selection of a game
12   setTimeout(function() {
13     game = selectGame();
14     listAvailablePoolsForGame(game.game_id);
15     subscribeToGame(game.game_id);
16   }, _.random(1000, MAX_GAME_SELECTION_DELAY));
17 };
18
19 // called upon reception of the list of active pools for game by the client
20 // and upon confirmation of prediction placing
21 var handlePoolsList = function(msg) {
22   // ... handle reception of list of active pools
23
24   // simulate user thinking and placing prediction
25   setTimeout(function() {
26     pool = selectPool();
27     placePredictionInPool(pool);
28   }, _.random(1000, MAX_PREDICTION_PLACING_DELAY));
29 };
```

Listing 5.3: Code simulating users joining games and placing predictions.

Why is this benchmark representative?

It's the expected load during a football (soccer) game. The load is *expected*, since the system has not been run at this scale yet - that's partly why the host company is in need of a project like this one.

User's load is representative of that scenario - users connect over a few minutes before the game starts, then stay connected over 90 minute game load and then disconnect once the game is finished.

5.6 Compared Architectures

This section contains simple descriptions and diagram overviews of the compared architectures. Implementation details can be found in the following chapter [6.3](#).

5.6.1 Baseline Architecture

The baseline architecture consists of the main three components (the websocket server, redis and mysql) on one massive EC2 instance located in the eu-west-1 AWS datacenter (Ireland). Clients from all regions connect directly to the websocket server on this EC2 instance.

[TODO add diagram]

The technical specification of the instances, number of processes and other technical details can be found in the next chapter [6.3.1](#).

5.6.2 Proposed Architecture I

The first proposed architecture improvement consists of the moving the websocket servers closer to users. This means each of the 8 regions has its own websocket server instance - 8 times smaller than one massive instance in case of the baseline architecture.

Database instances are still not distributed and are both located in Ireland. Clients are routed through Amazon's route53 based on their latency or geo-location.

[TODO add diagram]

The technical specification of the instances, number of processes and other technical details can be found in the next chapter [6.3.2](#).

5.6.3 Proposed Architecture II

The second proposed architecture adds the distribution of the data layer. Read replicas are placed, together with WebSocket servers, in each of the regions. Database (both mysql and redis) writes still are routed to the master node located Ireland.

[TODO add diagram]

The technical specification of the instances, number of processes and other technical details can be found in the next chapter [6.3.3](#).

5.7 Results Evaluation

Results are evaluated using literature-based scalability meters [[GPBT11](#)]. I selected (and slightly modified - added cost factor) the ones [7.4](#) which best facilitate reasoning about factors most important to the host organization - client latencies, cloud resources utilization and operating cost. The tested architectures are compared using these quantifiable meters.

The selected meters allow to compare deployed architectures in general scalability terms, but also provide insight into specific aspects of the system, so that analysis can be broken down and architectures can be compared in terms of e.g. latency or cost exclusively.

[TODO list these meters here]

5.8 Experiment deliverables

The project delivers the following:

1. open sourced code for:
 - using the scalability meters,
 - managing the AWS-based region-distributed server suite (built on top of boto ³),
 - remote execution of WebSocket load testing and latency-related metric collection,
2. a framework for comparison multiple aspects of different variants of The System Under Test [4](#).
3. a thorough analysis of the deployed architectures,
4. improved system scalability (multiple limitations and suboptimal design decisions have been identified in the process [6.1.1](#)).

³<https://boto.readthedocs.org/en/latest/>

Chapter 6

Experiment Implementation

"At scale, everything breaks."

— Urs Holzle, Google Fellow

6.1 Baseline Architecture Preparations

In order to obtain meaningful benchmark results, I had to perform multiple optimizations of our application code. This part of the project proved extremely timely (a lot of debugging time was not budgeted for when preparing the project timeline).

I also managed to identify main resource bottlenecks. Both in cases of the client simulating process and the server, it is the CPU.

Since Node.js is single-threaded, I prepared scripts that allow to run multiple processes in parallel to utilize all cores available to the client [6.2.1](#).

As far as the WebSocket server is concerned, I have identified 2 core thread types (among many more that the server consists of) - one for request handling, another one for pushing redis-published messages through the user's WebSocket. Since TCP is a 1-1 communication protocol, a UDP-like broadcast to all WebSockets is not possible. For every channel that users subscribe to (games are channels on the domain level), the system needs a thread that listens to a redis channel and iterates over a list of WebSockets of users subscribed to that channel. These threads, and the message handling threads are by far the two most resource (CPU-bound) intensive.

This allowed us to calculate the optimal number of WebSocket server processes that should be running on each machine - the number of machine CPU cores divided by 2. Since the WebSocket servers are fully stateless, any number can be run in parallel.

[TODO - sync vs async db?]

6.1.1 Discovered System Limitations

Listing Pools

The First discovered limitation was that the response with list of prediction pools for game included *every* prediction made in this pool. Predictions were included so that the clients can look through them and see what predictions other users (their friends in particular) have made.

Each prediction JSON object, as depicted in snippet [6.4](#), is around 135 bytes. With a workload generated by the benchmark the number of predictions quickly goes into the thousands and tens of thousands, which already means a single message size measured in *Megabytes*. This is obviously not feasible so we decided not to include all the predictions and prepare a separate request for predictions of your friends.

```

1 {
2     "prediction_id": "c34bcc05-58ea-4294-9310-9f8cbd121e29",
3     "placed_at": 1421937458,
4     "amount": 1.0,
5     "user_id": 1
6 }

```

Listing 6.4: Single prediction JSON object.

Message Publishing

The message publishing included in the benchmarking scenario happens when users place predictions. Each prediction influences the distribution of the pool and this information is broadcast to all users subscribed to a game to which the pool belongs.

I have discovered two huge system limitations when it comes to message publishing.

In the beginning, the distribution update was broadcast after *every prediction*. For a load of 24000 users, predicting (simplifying) on average every 15 seconds, 1600 updates *per second*. This is not sustainable - neither the publishing threads that had to push the message out through a list of sockets, nor the client processes could keep up with such messaging pace (or a pace nowhere near that, since publishing to a few thousand users takes hundreds of milliseconds). More importantly, humans are only capable of registering and reacting to changes as often as once every few hundred milliseconds (tested empirically ¹). As the platform is supposed to serve as a second-screen fun, the users should not be disturbed even at that pace. Thus, we decided to batch the prediction updates and send them once every 3 seconds (per pool). This improved the performance tremendously.

At first, updating the last broadcast timestamp in redis (*line 6* from the listing 6.5) was located below line 14. When running multiple WebSocket server instances, this caused a race condition (retrieving data concerning hundreds of predictions from redis and formatting them takes up to tens of milliseconds), so up to 3 broadcast messages would be sent within a few milliseconds. This problem got solved by moving it up the if clause. The check is still not atomic, but get and set are now separated only with one integer comparison (within the if clause, *line 5*, listing 6.5) so the race condition is not expected to happen at any reasonable scale. Nevertheless, this will be made atomic eventually.

```

1 # WebSocket Handler class
2 # place prediction request handling
3
4 last_broadcast = db.get_pool_last_distribution_broadcast_timestamp(pool_id) or 0
5 if ((int(placed_at) - int(last_broadcast)) > im_configs.BATCH_FREQUENCY):
6     db.set_pool_last_distribution_broadcast_timestamp(pool_id, epoch_millis())
7
8     new_predictions = db.get_predictions_to_broadcast_for_pool(pool_id)
9     distribution = self.prepare_distributions_structure(pool_id)
10
11     broadcast_message = im_messages.PoolDistributionUpdatedPublish(
12         pool_id, new_predictions, distribution
13     )
14     self.do_publish(broadcast_message, game_id)
15 else:
16     # persist for later broadcast
17     db.add_prediction_ids_to_broadcast_for_pool(pool_id, prediction_id)

```

Listing 6.5: Message batching.

The second big issue was that because publishing of distribution updates was batched, every broadcast message included details of every new prediction (few hundred) made within the time window between batched broadcasts. This message would be broadcast to **every user every 3 seconds**. This was generating a tremendous, unnecessary data load. Since the purpose of this was the same as

¹<http://www.humanbenchmark.com/tests/reactiontime>

in including new predictions when listing pools, the same solution was applied. Users request their friend data separately now. Identifying this was a complete game changer - it reduced the application data load 300-fold and allowed to increase number of sustainable users in the system by 2 orders of magnitude.

6.1.2 Kernel tuning

Operating systems are not tuned by default to sustain a huge number of concurrent users, nor are they tuned to support quick connects / disconnects and TCP sockets reusal. Because of that, I needed to dig into unix kernel and tcp stack handling and prepare the operating system setup on which I based the images used in the project.

Listing 6.6 shows all the *sysctl* changes that were introduced to */etc/sysctl.conf*. Sysctl ² is used to modify kernel paramaters at runtime. The changes are described in details below.

```
1 # WebSocket scalable config
2 fs.file-max = 1655360
3
4 net.core.netdev_max_backlog=65536
5 net.ipv4.tcp_max_syn_backlog=65536
6 net.core.somaxconn=16384
7
8 net.ipv4.tcp_slow_start_after_idle=0
9
10 net.ipv4.tcp_mem=383865 511820 2303190
11 net.ipv4.tcp_rmem=8192 873700 8388608
12 net.ipv4.tcp_wmem=4096 655360 8388608
13
14 net.ipv4.tcp_fin_timeout=5
15
16 net.ipv4.tcp_max_orphans=262144
17
18 net.ipv4.tcp_synack_retries=2
19 net.ipv4.tcp_syn_retries=2
20
21 net.ipv4.tcp_tw_reuse=1
22
23 net.ipv4.ip_local_port_range=1025 65535
24
25 vm.overcommit_memory=1
```

Listing 6.6: Sysctl.conf updates.

fs.file-max Maximum number of open file descriptors that can be opened by a single process. With the default value, both nginx and the WebSocket server is incapable of serving more than 1024 users.

net.core.netdev_max_backlog Maximum number of packets that can be queued on interface input. If kernel is receiving packets faster than can be processed, this queue increases (high value is of this parameter is not necessary so as not to drop messages).

net.ipv4.tcp_max_syn_backlog Maximum number of remembered connection requests, which are still did not receive an acknowledgment from connecting client.

net.core.somaxconn Max length of the listen queue backlog (processes like nginx need this value to be high for high performance). Withuot setting this value, TCP listen queue overflows happen. *"possible SYN flooding on port"* in */var/log/messages* is an indication that this value is too low.

net.ipv4.tcp_slow_start_after_idle This value should be set to 0 to ensure connections don't go back to default TCP window size after being idle for too long.

²<http://linux.die.net/man/8/sysctl>

net.ipv4.tcp_mem *min pressure max*, Total memory pages allocated for TCP sockets. If this memory reaches pressure value, kernel starts putting pressure on TCP memory. More than *max* will not be allocated. Messages "*maximum number of memory pages allocated to the TCP was reached*" indicates these values are too low.

net.ipv4.tcp_rmem *min default max*, Size of receive buffer for each TCP connection. Kernel auto tunes these values between the min-max range.

net.ipv4.tcp_wmem *min default max*, Size of send buffer for each TCP connection. Kernel auto tunes these values between the min-max range.

net.ipv4.tcp_fin_timeout Time a connection should spend in FIN_WAIT_2 state. By default it is set to 60s, lower values tell kernel to free memory faster and move sockets to TIME_WAIT, which allows for their faster reusal.

net.ipv4.tcp_max_orphans Maximum number of sockets which have been closed and no longer have a file handle attached to them. Needs to be high for high performing servers.

net.ipv4.tcp_synack_retries How many times to retry creating the TCP connection. Lower values allow to fail faster.

net.ipv4.tcp_syn_retries as above.

net.ipv4.tcp_tw_reuse If set to 1, kernel allows to reuse TCP sockets in TIME_WAIT state. Similar to `tcp_tw_reuse`, changing which is highly discouraged since enabling it introduces problems with clients behind NATs and stateful firewalls (the kernel can simply close their sockets).

net.ipv4.ip_local_port_range Port range available to processes - the broader more ports available. Without adding virtual IPs, the absolute maximum of TCP ports the server can have open is 65535³. This is a TCP-protocol related limitation, since the port number is represented with an unsigned 16-bit integer ($2^{16} = 65535$). To go above that limit, multiple virtual IP addresses needed to be created on a network interface.

vm.overcommit_memory regulates the heuristics of memory allocation during process forks. Setting to 1 is necessary for high-performing redis.

Linux TCP stack

In short, the TCP stack needs to be tuned to sustain multiple number of concurrent open TCP connections and to be able to reuse closed sockets quickly. This is achieved using the `sysctl` flags described above.

Maintaining the TCP stack in itself is not very expensive in terms of resources. Maintaining idle connections is almost memory-free. Nevertheless, it is possible to leak memory / file descriptors from an application and this, at scale, can become expensive. Descriptor usage can be monitored at unix systems using `lsof` command.

³The number of client connections that a server can support has nothing to do with ports, since a server is (typically) only listening for WS/WSS connections on one single port. But, if the machine also has an HAProxy or a Load Balancer that forward connections (which is my case - nginx), this limit holds.

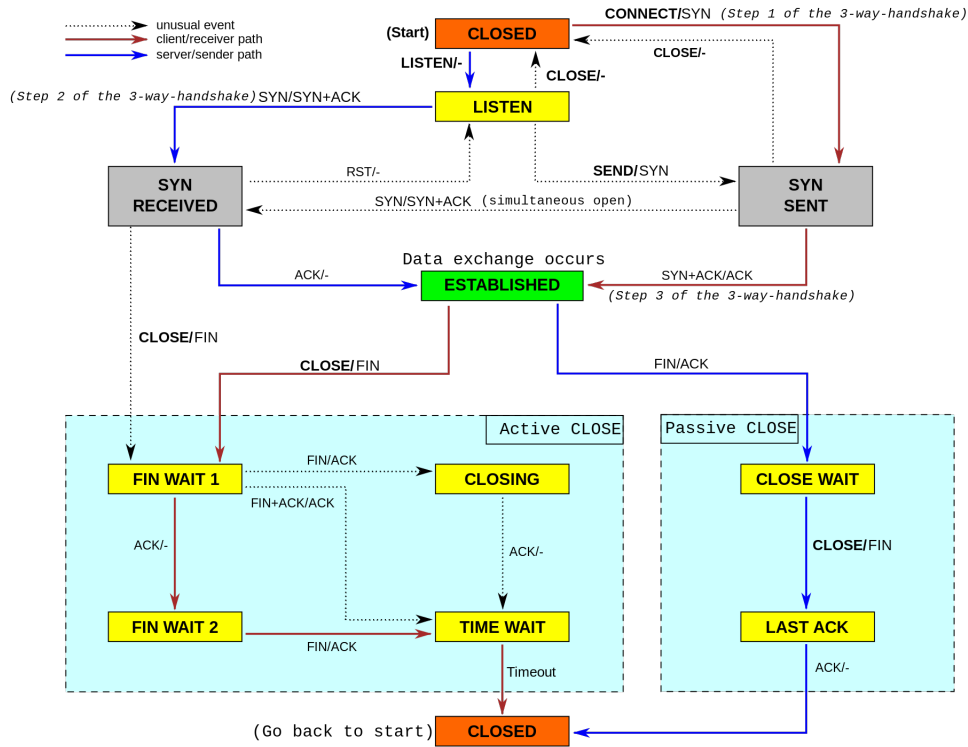


Figure 6.1: Possible states of a TCP socket.

So as to be able to re-run the benchmark quickly (this was very important during the preparation process), sockets in `TIME_WAIT` needed to be made reusable using the `net.ipv4.tcp_tw_reuse` flag. Sockets end up in this state after active close (son on the initiating side - more oftenthe client in case of The System Under Test 4), as depicted on figure 6.3.

Even though this explanation is relatively brief, researching and applying this information took weeks.

Real-Time Insights

In order to have real-time insight into the system and the state of the host Operating System, I prepared a suite to measure and visualize the relevant metrics.

The setup I used is the following:

Graphite⁴ a scalable, real time graphing tool. I used it to display both our system internal metrics and operating system parameters.

Statsd⁵ a tool aggregating metrics and preparing appropriate output formats for tools like graphite. I used it to display both our system internal metrics and operating system parameters.

Custom Scripts to collect operating system data. This is slightly different in case of the deployed cloud architecture and is described below 6.1.6.

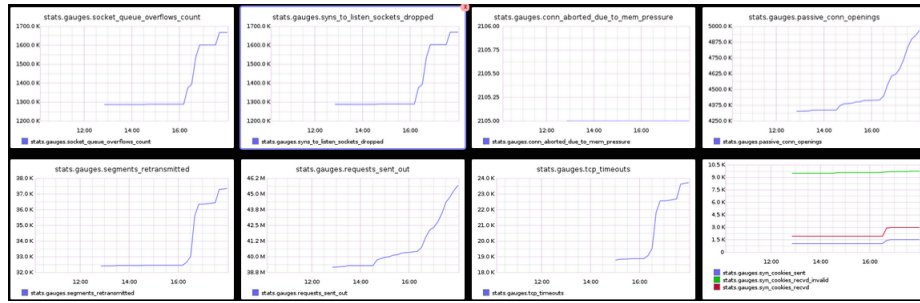


Figure 6.2: Monitored TCP-related values.

Measured operating system parameters were selected so that they indicate bottlenecks and problematic, TCP-related areas ⁶⁷. The commands used to obtain them are the following:

ss -s TCP socket summary, collectibles: counts of established, closed, orphaned, time_wait and syn_received sockets,

netstat -s network summary, collectibles: counts of SYN cookies sent and received (also invalid), connections aborted due to memory pressure, SYNs to listen sockets dropped, socket queue overflows, failed connection attempts, passive connection openings, segments retransmitted, requests sent out and TCP timeouts.

The monitoring helped multiple times in identifying the bottlenecks. Increases in socket queue overflow counts and SYNs to listen sockets dropped (overflows and drops are rarely a good thing) helped us identify why the sockets were failing to open and which TCP parameters we needed to tune.

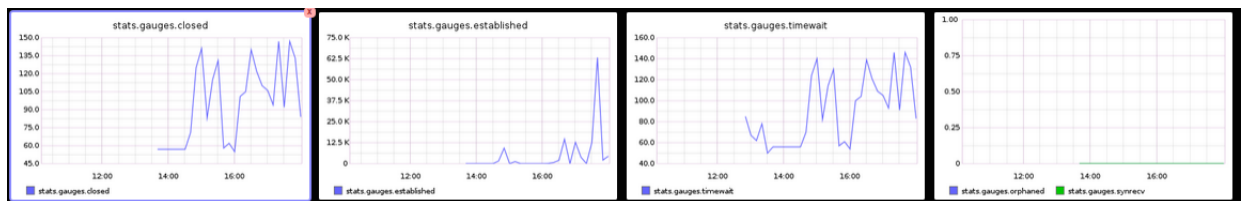


Figure 6.3: Monitored TCP sockets.

A very similar setup is also deployed in the cloud - a dedicated instance is running, to which all server instances send metrics in real-time.

6.1.3 Nginx Optimizations

On each WebSocket server instance, we are running an nginx process to end the encrypted communication and load balance traffic before all server processes running on a given machine.

The following optimizations were made to improve nginx performance. The fragment of the configuration can be found in listing 6.7.

worker_processes auto sets the number of nginx processes to the number of available CPU cores automatically,

worker_connections sets the number of maximum connections opened by a worker process (cannot be higher than the number of allowed file descriptors per process),

tcp_nodelay setting to off disables Nagles algorithm and thus allows sockets not to buffer data but send it right away (and effectively reduce latencies up to 0.2 seconds⁸),

⁶⁷<http://www.slideshare.net/chartbeat/tuning-tcp-and-nginx-on-ec2>

⁷<http://engineering.chartbeat.com/2014/01/02/part-1-lessons-learned-tuning-tcp-and-nginx-in-ec2/>

⁸https://t37.net/nginx-optimization-understanding-sendfile-tcp_nodelay-and-tcp_nopush.html

tcp_nopush setting to off deactivates the TCP_CORK option in the Linux TCP stack and allows to send smaller data chunks,

keepalive_timeout defines how long to keep an idle keep-alive connection open (in seconds),

proxy_read_timeout defines how long to work for a backend to server the connection / request (high values necessary for potentially highly congested servers),

backlog sets the max length of the queue of pending connections,

deferred instructs the kernel not to wait for the final ACK packet and not to initiate the connection until the first packet of real data has arrived,

worker_rlimit_nofile sets the limit of file descriptors that can be opened by nginx process (2 are necessary for each WebSocket connection - incoming and forwarded).

round-robin routing had to be added to the WebSocket server upstream so that incoming connections are divided between all server processes.

```
1 worker_processes auto;
2 worker_connections 165535;
3 tcp_nopush off;
4 tcp_nodelay off;
5 keepalive_timeout 1000;
6 proxy_read_timeout 1000;
7 worker_rlimit_nofile 30000;
8
9 upstream api_server_dev {
10     server 127.0.0.1:9998;
11     server 127.0.0.1:9997;
12     server 127.0.0.1:9996;
13     server 127.0.0.1:9995;
14     server 127.0.0.1:9994;
15     server 127.0.0.1:9993;
16     server 127.0.0.1:9992;
17     [...]
18 }
19
20 [...]
21 # inside server section
22 listen 443 backlog=16384 deferred;
23 [...]
```

Listing 6.7: Nginx configuration modifications.

6.1.4 Mysql Optimizations

Most of MySQL optimizations are still ahead, since I haven't really stretched it to its limits yet. We have encountered and solved a massive bottleneck at one point. MySQL would periodically, regularly stall for a very long time.

I managed to pinpoint the issue down to the frequency with which the log was flushed to disk. With the default value of the *innodb_flush_log_at_trx_commit* parameter (1), this happens on every transaction commit. Since changing the parameter value to 2, the flushing only happens once per second and the performance has vastly improved.

The second parameter I modified is *max_connections* so that all the server processes with their multiple threads could connect without problems against (we hit the default limit rather early).

There is definitely potential for further improvements in this scope.


```

1 max_connections = 500
2 innodb_flush_log_at_trx_commit = 2

```

Listing 6.8: Mysql configuration modifications.

6.1.5 Redis Optimizations

[1772] 24 Jul 10:09:09.009 # Can't save in background: fork: Cannot allocate memory [1772] 24 Jul 10:09:15.020 * 1 changes in 900 seconds. Saving...

ResponseError: MISCONF Redis is configured to save RDB snapshots, but is currently not able to persist on disk. Commands that may modify the data set are disabled. Please check Redis logs for details about the error.

Explain overcommitting here shortly (redis.io/topics/faq)

sysctl: vm.overcommit_memory = 1

LAB NOTES FROM 24.07

6.1.6 Cloud baseline preparations

make sure to describe client simulation in details and issues that you battled with there - like cpu maxing, having to divide into multiple processes etc.

client machines reaching different resource limits at different loads, because of the increased publishing?

started with t1.micro - had to move to m1.small, because we were reaching memory limits on these that run longer than 3 minutes, kernel was silently killing the process and metrics would fail to get reported from the clients

Jul 24 11:36:28 ip-172-31-24-5 kernel: [39921901.888692] Out of memory: Kill process 2992 (node) score 673 or sacrifice child"

super hard to estimate the necessary client resource, which we had to over-provision anyway - clients reaching hardware limits affects reported measurements and renders the test run useless.

run previously on smaller scale (to save costs)

Adjust the ports opened for cloud instances - redis publishing (communication going the other way round)

setting up custom cloudwatch metric collection - 7.3. what we thought would work didn't.

running as cronjob had to dig into perl script which cached the instance id - so all server instances were reporting metrics for the first instance the scripts were deployed on. (verified from the EC2 instance using the command `wget http://169.254.169.254/latest/meta-data/instance-id`), had to update our node scripts for collecting data pushed through and received on the sockets - the node library we use [REF] had a different internal representation so had to dig in to report data properly.

Image Preparations

Both servers and clients are spun up using AMIs (describe what they are [amazon source]).

Both needed to be meticulously prepared to ensure smooth deployment of the test suite and tested architectures.

The client has been prepared with the right environment so that the test suite master can ssh into it (using node sequest) and run the test command [put it here].

The server has all the necessary services that need to be automatically started up: - /etc/init.d starts redis (if applicable) - /etc/init.d starts mysql (if applicable) - /etc/init.d starts supervisord - supervisord conf has the processes that need to be started (with an autostart option) [PASTE sample api server config here] - api server configurations has the correct host names for data stores - redis and mysql

based on the same optimized image all unnecessary services closed: clients all (mysqld, redis, supervisord, nginx) switched off; redis all but redis distributed servers no redis / mysql

Distributing Resources Across Cloud Regions

Aws stack is divided into Regions (e.g. Ireland, N. Virginia, Sao Paulo) and these into availability zones. All the service stack you set up for a region. No easy way to copy between regions - businesses built around that.

[TODO]:

copy code snippets for deploying client groups and orchestrating them and servers

6.1.7 Excluding Frankfurt Region

Had to be excluded, because the AMIs we use are not compatible with the region's infrastructure. We are using ParaVirtual images, whereas Frankfurt supports HVM [describe with sources].

Observations

splitting up redis and mysql onto separate instances causes CPUs to max out 100% (handling the networks stack?) and data traffic to spike (obvsly)

backup by table from lab notes 21.07

6.2 Load Testing Framework

all the analyzed tools, list also in lab notes from may 11 [FOR EACH WHY WASN'T SATISFACTORY] jmeter, thor (low extensibility which we needed), autobahn (max available connections but no further control), gatling-websocket, <http://www.opensourcetesting.org/performance.php> tsung - looked promising, highly scalable erlang, but hard to understand what was going on, community was not active enough, wsbench

6.2.1 Load Test Suite Preparations

Split up into processes utilizing all the cores, necessary to go with different users INSERT NODE PROCESS HERE

6.2.2 Making Load Test Process Bulletproof

I needed that. If the process crashes no results are reported and we are in the dark.

Clients would randomly not report results, dug in and discovered 3 things: - writing on a closing socket would cause the process to crash - not a very nice way (no time though) so i am doing this, cannot use try catches in async node env: [TODO] make it into a snippet: `process.on('uncaughtException', function(err) console.log('Exception:', err));` - i was using sequest which would timeout when orchestrating 120 client machines

6.3 Compared Architectures Implementation Details

selection - similar computing allocated resources [lab notes table]

instance selection limited by our image type, not all supported (our image paravirtual, some require newer hardware virtual machine HVM) [<https://www.opswat.com/blog/aws-2015-why-you-need-switch-pv-hvm>]

[TODO] describe selected instance types

we picked CPU intensive cause by far that's the first hardware limit we hit

6.3.1 Baseline Architecture Implementation Details

ONE MASSIVE INSTANCE WITH WEBSOCKET SERVER + REDIS + MYSQL NO NETWORK OVERHEAD 12/13 SERVER PROCESSES

c3.8xlarge instance, 32 CORES, 60GB, 140 GB SSD Storage with 4000 IOPS*

6.3.2 Proposed Architecture I Implementation Details

Region-distributed API servers 1 API-server instance per one of 8 regions (m3.xlarge, 4cores, 15gb RAM, 2x40SSD) 2 processes per instance

1 big redis instance (m2.4xlarge, 8 cores, 86.4 GB mem, 2 x 840 GB storage) 1 big mysql RDS instance (r3.4xlarge, 16 cores, 122 GB mem), 100 GB disk 1000 IOPS**

6.3.3 Proposed Architecture II Implementation Details

Region-distributed API servers + redis cluster + MySQL Read Replicas

1 API-server instance per one of 8 regions (m3.2xlarge, 4cores, 30gb RAM, 2x40SSD) 2 processes per instance

Redis cluster - each node sitting on the same node as API server.

Mysql Read Replica nodes in each region - db.r3.large (2 cores, 15 GB) Writes still go to the master in main region.

*each server instance has a 140gb instance with 4000 provisioned iops; that's a minimum size since amazon requires 30:1 iops to size ratio. based an image on that so had to put it for every instance in every configuration to be able to reuse image. not that important, since we don't use much disk space ** ratios different for rds

6.4 Cloud architecture setup

selected aws

multiple availability zones very important in a shared cloud environment - often because another customer is getting hit by a DDoS.[GB14]

In Amazons environment, Grozev and Buyya suggest using Elasticache service [GB14]. We cannot use this, also described in 6.4.

[TODO] insert diagram Route53 -> Elb -> Nginx -> Api Server instance

6.4.1 Route53

dns mapped to instance ips / load balancer dns names lbr, geo, weighed rr their internal heuristics take network conditions from past weeks [VERIFY], rather amibiguous For each end users location, Route 53 will return the most specific Geo DNS record that includes that location. In other words, for a given end users location, Route 53 will first return a state record; if no state record is found, Route 53 will return a country record; if no country record is found, Route 53 will return a continent record; and finally, if no continent record is found, Route 53 will return the global record.

health checks - can be used custom-defined, but for us we can use load balancing health checks (will only route if healthy incstances behind it) [http://docs.aws.amazon.com/Route53/latest/DeveloperGuide/dns-failover-complex-configs.html]

http://docs.aws.amazon.com/Route53/latest/DeveloperGuide/Tutorials.html

6.4.2 Autoscaling

instances behind each groups. automatically scalable, connected to external redis and mysql instances.

only within one region, multiple availability zones http://docs.aws.amazon.com/AutoScaling/latest/DeveloperGuide/G http://docs.aws.amazon.com/AutoScaling/latest/DeveloperGuide/how-as-works.html#arch-AutoScalingMultiAZ

Auto Scaling attempts to distribute instances evenly between the Availability Zones that are enabled for your Auto Scaling group. Auto Scaling does this by attempting to launch new instances in the Availability Zone with the fewest instances. alarm - object that watches over a single metric (e.g. avg cpu use of ec2 instances in auto scaling group over specified time period) (OK, ALARM, INSSUFICIENT_DATA), can trigger scaling up / down policy on an autoscaling group policy variants (change ExactCapacity, ChangeInCapacity, PercentChangeInCapacity) recommendation one policy for scaling

out, another policy for scaling in <http://docs.aws.amazon.com/AutoScaling/latest/DeveloperGuide/as-scale-based-on-demand.html> <http://docs.aws.amazon.com/AmazonCloudWatch/latest/DeveloperGuide/AlarmThatSendsEmail.html> http://docs.aws.amazon.com/AmazonCloudWatch/latest/DeveloperGuide/US_AlarmAtThresholdEC2.html

Alarms

avg / min / max / sum / sample count (for whole group) of cpu / disk io / network io is \geq / $<$ / $>$ / $=$ than x for at least X consecutive periods of 5 min / 15 min / 1 hour / 6 hours

Scaling Policies

execute policy when alarm add / remove / set to X instances / and then wait Y (cooldown period)

Launch configurations

6.4.3 Load balancing

elb uses dns name to server a pool of lb instances in the backend (you might exceed one instance connection limit), that's why they are not provided a static ip routing: request count-based for http(s), for others (tcp which is a workaround for websockets) they use tcp [<http://docs.aws.amazon.com/ElasticLoadBalancing/latest/DeveloperGuide/how-elb-works.html#request-routing>] The client uses DNS round robin to determine which IP address to use to send the request to the load balancer. No control over routing without load balancing groups. For example, if you have ten instances in Availability Zone us-west-2a and two instances in us-west-2b, the traffic is equally distributed between the two Availability Zones. As a result, the two instances in us-west-2b serve the same amount of traffic as the ten instances in us-west-2a. Instead, you should distribute your instances so that you have six instances in each Availability Zone. To distribute traffic evenly across all back-end instances, regardless of the Availability Zone, enable cross-zone load balancing on your load balancer. However, we still recommend that you maintain approximately equivalent numbers of instances in each Availability Zone for higher fault tolerance. [<http://docs.aws.amazon.com/ElasticLoadBalancing/latest/DeveloperGuide/how-elb-works.html#request-routing>] only 64k connections possible per ELB instance TODO: VERIFY THAT stickiness only works for HTTP/HTTPS protocols Once you have a testing tool in place, you will need to define the growth in the load. We recommend that you increase the load at a rate of no more than 50 percent every five minutes. Both step patterns and linear patterns for load generation should work well with Elastic Load Balancing. If you are going to use a random load generator, then it is important that you set the ceiling for the spikes so that they do not go above the load that Elastic Load Balancing will handle until it scales (see Pre-Warming the ELB). [<https://aws.amazon.com/articles/1636185810492479#pre-warming>]

I won't use this - dns prevented us from using dynamic scale up/down 7.4.1. On top of that, one cannot simply run a benchmark against ELBs - they need to be prewarmed (by default very small ELB node instances are assigned so one needs to gradually increase the load). On top of that, if they don't receive traffic for 12 hours they scale down completely and the whole process needs to be repeated (source: our conversation with their premium support, which btw costs 250\$/hour). So I decided to skip it. Which is possible, since without dynamic scaling up and down, one can simply plug in ec2 instance ips into route53 [that's what I ended up doing].

6.4.4 Data Layer

Technologies we used described in 4.4. We used Mysql and redis as a caching solution.

Amazon Relational Database Service

Does not support mysql cluster as of July 2015. Support coming but not released yet. Offers mysql read replicas. With a single master. That obviously scales only reads You can create a MySQL Read Replica in a different region than the source DB instance to improve your disaster recovery capabilities, scale read operations into a region closer to end users, or make it easier to migrate from a data center in one region to a data center in another region.

Elasticache

doesn't support crossregion (you cannot connect to an elasticache cluster from outside of the region)

The above explain why, for a scaled out solution we had to roll with our custom setup. We were suggested using technologies like DynamoDB and Kinesis, but out of scope.

6.5 System Scaling Delay

Delays: up to 2 minutes to get metrics, up to ????? minutes for instance to be accessible to clients + ???? for DNS changes to propagate (TTL set to 60 seconds but TODO VERIFYF) Unfortunately there are a large number of (misbehaving) DNS servers out there that dont properly obey TTLs on records and will still serve up stale records for an indefinite amount of time. [<http://engineering.chartbeat.com/2014/01/02/part-1-lessons-learned-tuning-tcp-and-nginx-in-ec2/>]

DNS PROTOCOL INHERENT LIMITATIONS PREVENTED US FROM TESTING HOW FAST AND DOWN ARCHITECTURE SCALES

Chapter 7

Scalability Measurements

In this chapter I describe most relevant models and metrics from scientific literature and the subsets I decided to use in this research [7.4](#).

7.1 Scalability Meters

Models I find most useful were proposed by Pattabiraman et al [[GPBT11](#)]. As mentioned in Related Work section [1.4](#), they divide the performance indicators into three groups and propose formal quantifiable meters for each of them:

1. computing resource indicators (CPU, disk, memory, networks, etc.),
 - Computing Resource Allocation Meter (CRAM),
 - Computing Resource Utilization Meter (CRUM),
2. workload indicators (connected users, throughput and latency, etc.),
 - System Load Meter (SLM),
3. performance indicators (processing speed, system reliability, availability)
 - System Performance Meter (SPM).

They also propose model which allow to combine the above: System Capacity Meter (SCM), System Effective Capacity Meter (SEC), Effective System Scalability (ESS), and Effective Scalable Range (ESR).

7.1.1 Scalability Meters

The geometric interpretation of each of these meters is a polygon. The quantifiable value of the meter is the area of the polygon. Sample meter is plotted on the figure [7.2](#).

The value of each metric plugged into the meter is represented by the radial distance between the center of the polygon and the vertex. Since all the metrics that serve as inputs to the meter have equal weights, the central angle (as shown of figure [7.1](#)) between two consecutive radial distances is always going to be equal. For n metrics in a meter, each central angle will have a value of $2\Pi/n$.

The area of a polygon can be computed as a sum of areas of the triangles that constitute it. The sides of each of these triangles consist of the side of the polygon and 2 radii.

Area of a triangle can be computed using the formula:

$$A_t = 0.5 * a * b * \sin(\alpha_{ab})$$

where a and b are to sides of the triangle and α_{ab} is the angle between them.

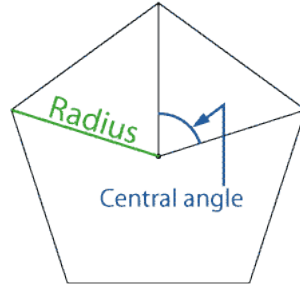


Figure 7.1: Polygon's radius and central angle

In case of these meters, α_{ab} is always equal to $2\Pi/n$. The area of a triangle limited by two radii with values of two metrics x_i and x_{i+1} would look like this:

$$A_t = 0.5 * x_i * x_{i+1} * \sin(2\Pi/n)$$

To compute the area of the polygon one needs to sum all the areas of the triangles formed by radii, and thus the formula is the following:

$$A_p = 0.5 * \sin(2\Pi/n) * \sum_{i=1}^n x_i * x_{i+1}$$

where x_i is a value of an input metric and n is the count of input metrics.

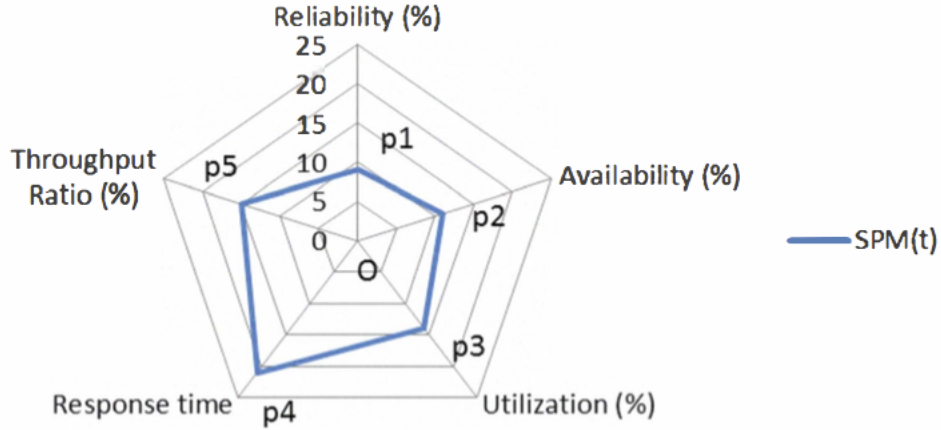


Figure 7.2: Sample Meter Geometric Interpretation (SPM)

Computing Resource Allocation Meter (CRAM)

$$CRAM(S, t) = 0.5 * \sin(2\Pi/n) * \sum_{i=1}^n a_i * a_{i+1}$$

S - represents the measured system,
 t - represents the time over which total resource allocation for the system is evaluated,
 a_i - represents the values of different types of *allocated* computing resources to S by a cloud (network traffic boundary, cache allocated, memory allocated, cpu allocated, etc.),
 n - represents the count of different types of allocated computing resources to S by a cloud.

Computing Resource Utilization Meter (CRUM)

$$CRUM(S, t) = 0.5 * \sin(2\Pi/n) * \sum_{i=1}^n u_i * u_{i+1}$$

S - represents the measured system,
 t - represents the time over which total resource allocation for the system is evaluated,
 u_i - represents the values of different types of computing resources *utilized* by S in a cloud (network traffic boundary, cache allocated, memory allocated, cpu allocated, etc.),
 n - represents the count of different types of computing resources utilized by S in a cloud.

System Performance Meter (SPM)

$$SPM(S, t) = 0.5 * \sin(2\Pi/n) * \sum_{i=1}^n p_i * p_{i+1}$$

S - represents the measured system,
 t - represents the time over which system performance is evaluated,
 p_i - represents the values of different system performance metrics of the S in a cloud (throughput ratio, reliability, availability, utilization, response time, etc.),
 n - represents the count of different performance metrics measured in S .

System Load Meter (SLM)

$$SCM(S, t) = 0.5 * \sin(2\Pi/3) * [CTL(t) * UAL(t) + UAL(t) * SDL(t) + CTL(t) * CDL(t)]$$

S - represents the measured system,
 t - represents the time over which load for the system is evaluated,
 $CTL(t)$ - represents system communication traffic load,
 $UAL(t)$ - represents system user access load,
 $SDL(t)$ - represents system data load.

System Capacity Meter (SCM)

$$SCM(S, t) = 0.5 * \sin(2\Pi/3) * [SL(t) * SP(t) + SL(t) * AR(t) + SP(t) * AR(t)]$$

S - represents the measured system,
 t - represents the time over which system capacity is evaluated,
 $SL(t)$ - represents system load based on SLM,
 $SP(t)$ - represents system performance based on SPM,
 $AR(t)$ - represents system resource allocation based on CRAM.

System Effective Capacity Meter (SEC)

$$SEC(S, t) = 0.5 * \sin(2\Pi/3) * [SL(t) * SP(t) + SL(t) * RU(t) + SP(t) * RU(t)]$$

S - represents the measured system,
 t - represents the time over which load for the system is evaluated,
 $SL(t)$ - represents system load based on SLM,
 $SP(t)$ - represents system performance based on SPM,
 $RU(t)$ - represents system resource utilization based on CRUM.

Since all the meters above use the values sampled over a period of time as input, for each of the mentioned meters one can talk about *min* and *max* values. This is useful for the last two meters.

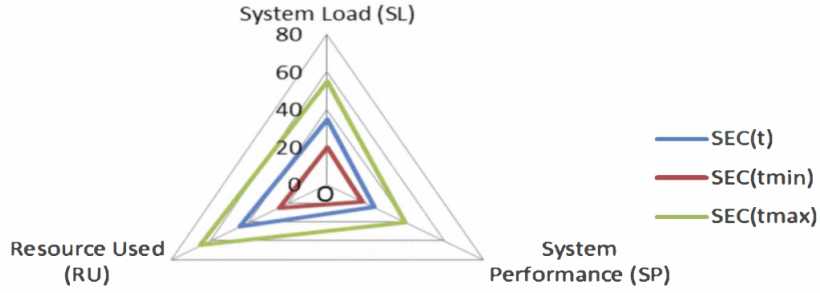


Figure 7.3: Sample System Effective Capacity Meter (SEC)

Effective Scalable Range (ESR)

$$ESS = [SL(t)/SL_{min}]/[SEC(t)/SEC_{min}]$$

t - represents the time over which the system is evaluated,
 $SL(t)$ - represents system load based on SLM,
 $SEC(t)$ - represents system effective capacity based on SEC.

Effective System Scalability (ESS)

$$ESR = (SEC_{max} - SEC_{min})$$

SEC - represents *max* and *min* system effective capacity based on SEC.

Using the last two models (ESR and ESS), one can compare scalability of two systems/architectures in absolute quantifiable terms.

The meters I decided to use in this research are described below [7.4](#).

7.1.2 Other Models

There is much more work related to the measuring scalability of distributed systems. Srinivas and Janakiram in their work [\[SJ05\]](#) mention similar metric evaluating scalability as a product of throughput and response time (or any value function) divided by the cost factor.

They propose a model considering scalability as a function of synchronization, consistency, availability, workload and faultload. It aims on identifying bottlenecks and hence improving the scalability. The authors also emphasize the fact of interconnectedness of synchronization, consistency and availability:

$$scalability = f(availability, synchronization, consistency, workload, faultload)$$

Jogalekar and Woodside [JW00] propose a strategy-based scalability metric based on productivity and cost effectiveness (a function of system's throughput and its Quality of Service). It separates evaluation of throughput or quantity of work from Quality of Service (which, according to the authors, can be any suitable expression). Productivity can be computed as follows:

$$F(k) = \lambda(k) * f(k) / C(k)$$

$F(k)$ - represents the productivity of system at scale k ,

$\lambda(k)$ - represents throughput in responses/s of the system at scale k ,

$f(k)$ - represents the average value of a responses of the system at scale k , calculated from its QoS,

$C(k)$ - represents the running cost per second of the system at scale k .

Based on that, one can compute scalability metric relating systems at two different scale factors as the ratio of their productivity figures:

$$v(k_1, k_2) = (F(k_1) / F(k_2))$$

7.2 AWS Available metrics

Since the System Under Test 4 is deployed to Amazon's Cloud 6.4, the metrics of the system need to be collected as a combination of metrics offered by Amazon's measurement service (CloudWatch¹ and our custom collection process (described here 7.3).

Metrics from CloudWatch are collectible by default every 5 minutes. They offer a paid *detailed mode*, with which one can collect metrics every minute. From what I have experienced, even in the detailed mode, the requested metrics show up in the CloudWatch API results with a delay as high as 2 minutes.

The CloudWatch API has its limitations - for instance, the allocation and utilization of memory for EC2 instances and processes is not available. This is something we are collecting in our custom metric collection process 7.3.

As Pattabhiraman et al. [GPBT11] warn, there exist additional costs which are not apparent from the start. If one exceeds 10 metrics, 10 alarms, 1,000,000 million requests, 1000 SNS email notifications, 5GB of incoming data logs or 5GB of data archiving per month then additional costs accrue. This is included in the cost estimates.

CloudWatch Alarms² can be configured based on these metrics. They can be applied to trigger AutoScaling actions that scale the deployed system in or out automatically.

A very important limitation for the case of this research is that data cannot be aggregated automatically across AWS regions³.

For each metric collected for a given time window (1 or 5 minutes), CloudWatch offers the following statistics: min, max, average, sum and sample count (count (number) of data points used for the statistical calculation).

¹<https://aws.amazon.com/cloudwatch/>

²<http://docs.aws.amazon.com/AWSEC2/latest/UserGuide/using-cloudwatch.html>

³http://docs.aws.amazon.com/AmazonCloudWatch/latest/DeveloperGuide/cloudwatch_concepts.html

Metrics can be aggregated over different domains - instances, AutoScaling groups, Elastic Load Balancers (all instances behind it) or availability zones. Metrics for groups of instances differ from metrics for single instances.

7.2.1 Instance Specific

The following metrics can be collected on an EC2 instance level:

- CPU utilization (% and absolute),
- Network in/out (in bytes),
- Disk read/write (number of operations and bytes),
- Custom user-defined metrics,
- Instance status checks data.

7.2.2 Cloud (group) Specific

Collected from Elastic Load Balancer for all instances connected.

7.2.3 Load Test Metric Collection

<http://stackoverflow.com/questions/3871886/ssl-and-load-balancing>

elb uses dns name to server a pool of lb instances in the backend (you might exceed one instance connection limit) in order for the full range of ELB machine IP addresses to be utilized, make sure each [client] refreshes their DNS resolution results every few minutes. One test client equals one public IP address. ELB machines seem to route all traffic from a single IP address to the same back-end instance, so if you run more than one test client process behind a single public IP address, ELB regards these as a single client. Use 12 test clients for every availability zone you have enabled on the ELB. These test clients do not need to be in different availability zones they do not even need to be in EC2 (although it is quite attractive to use EC2 instances for test clients). If you have configured your ELB to balance among two availability zones then you should use 24 test clients. Each test client should gradually ramp up its load over the course of a few hours. Each client can begin at (for example) one connection per second, and increase its rate of connections-per-second every X minutes until the load reaches your desired target after a few hours. - WE CANNOT DO THAT

[TODO] describe the process here - latency, handshaking, all aggregation, results snippet explain the difference between how broadcast latency is aggregated vs request handling

Tools described here [6.2](#)

7.3 Custom Metric Collection

locally: pidstat latencies by our custom client suite for network we looked at ntop, nethogs, nload, vnstat, wireshark but amazon provides this data.

`ss -s netstat -s`

plugged into statsd and graphite for real time monitoring

THE ABOVE DOESN'T WORK ON AMAZON LINUX.

perl scripts!!!! <http://docs.aws.amazon.com/AWSEC2/latest/UserGuide/mon-scripts.html> end up in cloudwatch

7.4 Selected Meters and Metrics

Computing Resource Utilization Meter Server CPU utilization, compiled over all server instances + RDS DB where applicable (Cloudwatch) Memory utilization, compiled over all server instances (Cloudwatch Custom) Disk utilization, compiled over all server instances (Cloudwatch)

System Performance Meter Handshaking latencies (Our suite) Request Handling latencies (Our suite) Broadcast latencies (Our suite) Dropped Connections (Our suite) Total Network In (Our suite / cloudwatch) Total Network Out (Our suite / cloudwatch) - include internal, billable transfer somehow

System Load Meter Concurrent users (Our suite) Data sent + data received (Our suite) Handled requests count (Our suite) Broadcast messages count (Our suite) Redis Hits (info — keyspace.hits) Mysql queries select * from information_schema.global_status s1 where variable_name = 'QUERIES';

System Effective Capacity Meter $SLM + SPM + CRUM + 1 / (\text{Computed Cost of running the architecture per unit of time})$

That's an update as to the proposed model.

We need a valuable that's inversely proportional to the cost value (e.g. how long you can run on x \$)

Effective Scalable Range and Effective System Scalability $SEC_{max} + SEC_{min} + SLM_{max} + SLM_{min}$

The benchmark is run against each of the deployed architecture setups and the resulting metrics are used to populated the described meters.

7.4.1 Excluded Metrics

I was forced to exclude availability - not enough resources (money and time) to run the tests long enough to run into any availability issues, since Amazon SLAs offer 99.9999% orders of availability for selected technologies.

Also forced to include Dynamic ScaleUp and Down. The way load test suite is setup is taht it hits a single Route53 url, which then distributes traffic between regions according to the specified policy. Dynamically adding / removing resources causes changes to that policy - that means global DNS records need to be updated. One can set TTL (time to live) which DNS records should respect and refresh their caches every TTL seconds. They don't and it's a known issue - I ran into this, had to wait even up to 30 minutes for 60s changes to propagate properly.

7.5 Controlling The Dynamic Cloud Environment

tl;dr - impossible.

But you can pick instances that are more stable and have better guarantees (rather than some weird spikes and collecting credits if you run under limit - that's what happens on t1.micro isntances).

We run repetetive tests to verify there is no huge influential differences in allocated resources.

One can also request dedicated rather than shared resources - outside of the scope of this project though, but offers finer control.

7.6 Time synchronization across the system

syncing clocks on machines using /etc/ntpd (polled from amazon servers)

7.7 Expected Answers

I am going to use the *Effective System Scalability* and *Effective Scalable Range* meters to perform a quantifiable comparison of the deployed architectures 5.6.

I expect to see differences in various metrics. Using all the other selected meters, I will be able to reason about singled out factors of scalability (such as cost or message handling latency).

Chapter 8

Experiment results

8.1 Baseline Architecture

8.2 Proposed Architecture I

latencies are blocking for the server in case of architecture II and III, whereas in I they happen outside of server handling and are parallelized. Db calls are blocking, so if the cross-region latency is a 100ms it all happens on server-db BLOCKING and paralyzes the communication; that's why all metrics blow up

[TODO diagram]

[TODO prove with time loops simple db.add_prediction()]

8.3 Proposed Architecture II

8.4 Latencies by region

They differ much - provide a comparison by region.

Chapter 9

Evaluation

9.1 Statistical Significance

<http://technology.stitchfix.com/blog/2015/05/26/significant-sample/> http://insightdatascience.com/blog/ab_testing-s
http://www.qubit.com/sites/default/files/pdf/mostwinningabtestresultsareillusory_0.pdf <http://www.evanmiller.org/h>
[not-to-run-an-ab-test.html](http://www.evanmiller.org/not-to-run-an-ab-test.html)

9.2 Threats to validity

ec2 whacky, control over hardware released to amazon (can be dedicated machines) shared environment, someone else might get ddosed or sth - you have no control over that network is inherently non-deterministic, load tests needed to actually test availability

9.3 Host Company Benefits

- measurement suite for further improvements - real time telemetry introduced into the system - system performance objectively improved along the process, bottlenecks identified - cloud management suite

Chapter 10

Conclusions

QUOTE: 9 women won't make a baby in a month
summary

10.1 Takeaways

endless process of digging deeper, deeper, deeper and deeper - can be never ending. That's why it's important to set up goals up front.

the journey continues (25 =, 24000 =, 10.000.000), nice logarithmic scale

pure hardware is rarely the limit (that is, your app is a limitation most of the time - it's hard to utilize resources 100% effectively)

Chapter 11

Further Work

varying user distributions per region dynamodb improvement to the models so that weighs can be added to the scalability meteres according to preferences. “

Chapter 12

Everything below that needs to be removed (except for bib), Front Matter

The first thing is to connect the class by saying:

```
\documentclass{uvamscse}
```

12.1 Title

Specify the title of the thesis with `\title` and `\subtitle` commands:

```
\title{MetaThesis}
\subtitle{A Thesis Template Leading by Example}
```

Any thesis can survive without a `\subtitle`, but the `\title` is mandatory.

12.2 Author

Introduce yourself with `\author` and `\authemail`:

```
\author{Vadim Zaytsev}
\authemail{vadim@grammarware.net}
```

Again, `\authemail` is not mandatory. If you need anything fancier, just put it inside `\author`.

```
\author{Vadim Zaytsev\footnote{Yes, that one.}}
```

The footnote would be printed on the bottom of the title page, and will be referred to by a symbol, not by a number as any footnotes within the main document body.

12.3 Date

By default, the date inserted in your PDF is the day of the build, e.g., “March 25, 2014”. If you want it to be formatted differently or be more vague or outright fake, use `\date`:

```
\date{Spring 2014}
```

The argument is just a string, the format is unrestricted:

```
\date{Tomorrow. Honestly.}
```



Figure 12.1: A hypothetical thesis title page without a cover picture (on the left), with an overly large one (in the centre) and with a tiny pic (on the right).

12.4 Host

If your hosting organisation is not the UvA, specify it with `\host`. The logo on the bottom of the title page will still be the UvA one, because this is the organisation guaranteeing your degree.

```
\host{Grammarware, Inc., \url{http://grammarware.github.io}}
```

NB: footnotes will not work, unless you know how to `\protect` them.

12.5 Cover picture

If the first page of your thesis looks too blunt, add a picture to it:

```
\coverpic{figures/terminal.png}
```

You can even specify the picture's width as an optional argument:

```
\coverpic[100pt]{figures/terminal.png}
```

How these three options look, you can see from [Figure 12.1](#).

12.6 Abstract

A thesis is fine without an abstract, if you do not feel like writing it and your supervisor does not feel like enforcing it. If you do want an abstract, make it with the `\abstract` command:

```
\abstract{This is not a thesis.}
```

The abstract is just like any other section of your thesis, so you can use any \LaTeX tricks there. If you think that the name “abstract” is too abstract for your abstract, you can still use `\abstract` without being too abstract:

```
\abstract[Confession]{I am a cenosillicaphobiac.}
```

Kent Beck [[JBB⁺93](#)] proposes to have four sentences in a good abstract:

1. The first states the problem.
2. The second states why the problem is a problem.
3. The third is the startling sentence.
4. The fourth states the implication of the startling sentence.

In practice, each of these “sentences” can be longer than an actual sentence, but it is in general a good rule of thumb to condense the summary of your thesis into these four tiny messages. Do not write too much, make it tweetable.

Chapter 13

Core Chapters

The structure of your thesis is up to you and your supervisor. Whatever you do, do not consider the guidelines below as dogmas.

13.1 Classic structure

Problem statement and motivation. You describe in detail what problem the research is addressing, and what is the motivation to address this problem. There is a concise and objective statement of the research questions, hypotheses and goals. It is made clear why these questions and goals are important and relevant to the world outside the university (assuming it exists). You can already split the main research question into subquestions in this chapter. This section also describes an analysis of the problem: where does it occur and how, how often, and what are the consequences? An important part is also to scope the research: what aspects are included and what aspects are deliberately left out, and why?

Research method. Here you describe the methods used to answer the research questions. A good structure of this section often follows the subquestions by providing a method for each. The research method needs a thorough motivation grounded in theory in order to be acceptable. As a part of the method, you can introduce a number of hypotheses — these will be tested by the research, using the methods described here. An important part of this section is validation. How will you evaluate and validate the outcomes of the research?

Background and context. This chapter contains all the information needed to put the thesis into context. It is common to use a revised version of your literature survey for this purpose. It is important to explicitly refer from your text to sources you have used, they will be listed in your bibliography. For example, you can write “A small number of programming languages account for most language use [MR13]”, where the following entry would be included in your bibliography:

[MR13] Leo A. Meyerovich and Ariel S. Rabkin. Empirical Analysis of Programming Language Adoption. In *Proceedings of the 2013 ACM SIGPLAN International Conference on Object Oriented Programming Systems Languages and Applications*, OOPSLA, pages 1–18. ACM, 2013. doi:10.1145/2509136.2509515.

Have a look at § 14 to learn more about citation.

Research. This chapter reports on the execution of the research method as described in an earlier chapter. If the research has been divided into phases, they are introduced, reported on and concluded individually. If needed, this chapter could be split up to balance out the sizes of all chapters.

Results. This chapter presents and clarifies the results obtained during the research. The focus should be on the factual results, not the interpretation or discussion. Tables and graphics should be used to increase the clarity of the results where applicable.

Analysis and conclusions. This chapter contains the analysis and interpretation of the results. The research questions are answered as best as possible with the results that were obtained. The analysis also discussed parts of the questions that were left unanswered.

An important topic is the validity of the results. What methods of validation were used? Could the results be generalised to other cases? What threats to validity can be identified? There is room here to discuss the results of related scientific literature here as well. How do the results obtained here relate to other work, and what consequences are there? Did your approach work better or worse? Did you learn anything new compared to the already existing body of knowledge? Finally, what could you say in hindsight on the research approach by followed? What could have done better? What lessons have been learned? What could other researchers use from your experience? A separate section should be devoted to “future work”, i.e., possible extension points of your work that you have identified. Even other researchers should be able to use those as a starting point.

13.2 Reporting on replications

Here are the guidelines to report on replicated studies [Car10]:

Information about the original study

Research question(s) that were the basis for the design

Participants, their number and any other relevant characteristics

Design as a graphical or textual description of the experimental design

Artefacts, the description of them and/or links to the artefacts used

Context variables as any important details that affected the design of the study or interpretation of the results

Summary of the results in a brief overview of the major findings

Information about the replication

Motivation for conducting the replication as a description of why the replication was conducted: to validate the results, to broaden the results by changing the participant pool or the artifacts.

Level of interaction with original experimenters. The level of interaction between the original experimenters and the replicators should be reported. This interaction could range from none (i.e. simply read the paper) to them being the same people. There is quite a lot of discussion of the level of interaction allowed for the replication to be “successful”, but this level should be reported even without addressing the controversy.

Changes to the original experiment. Any changes made to the design, participants, artifacts, procedures, data collected and/or analysis techniques should be discussed along with the motivation for the change.

Comparison of results to original

Consistent results, when replication results supported results from the original study, and

Differences in results, when results from the replication did not coincide with the results from the original study. Authors should also discuss how changes made to the experimental design (see above) may have caused these differences.

Drawing conclusions across studies

NB: this section contains portions of text repeated directly from Carver [Car10] and only slightly massaged. Do not do this for your thesis, write your own thoughts down.

13.3 L^AT_EX details

13.3.1 Environments

A L^AT_EX environment is something with opening and closing tags, which look like `\begin{name}` and `\end{name}`. Some useful environments to know:

<code>itemize</code>	bullet lists
<code>enumerate</code>	numbered lists
<code>description</code>	definition lists
<code>center</code>	centered line elements
<code>flushright</code>	right aligned lines
<code>flushleft</code>	left aligned lines
<code>tabular</code>	table
<code>longtable</code>	multi-page table (needs the <code>longtable</code> package)
<code>sideways</code>	rotates some text
<code>quote</code>	block quote
<code>verbatim</code>	unformatted text
<code>minipage</code>	compound box with elements inside
<code>boxedminipage</code>	compound box with elements inside and a border around it
<code>table</code>	floating table (needs to have <code>tabular</code> nested inside)
<code>figure</code>	floating figure
<code>sourcecode</code>	floating listing
<code>equation</code>	mathematical equation
<code>lstlisting</code>	pretty-printed syntax highlighted listing
<code>multline</code>	mathematical equation spanning over multiple lines
<code>eqnarray</code>	system of mathematical equations
<code>gather</code>	bundled mathematical equations
<code>align</code>	bundled and aligned mathematical equations
<code>array</code>	matrix
<code>CD</code>	commutative diagrams

13.4 Listings

```
1 define(Ps1,G1,G2)
2   ←
3   usedNs(G1,Uses),
4   ps2n(Ps1,N),
5   require(
6     member(N,Uses),
7     'Nonterminal ~q must not be fresh.',
8     [N]),
9   new(Ps1,N,G1,G2),
10  !.
```

Listing 13.9: Code in Prolog

```

module Syntax

imports Numbers
imports basic/Whitespace

exports
sorts
    Program Function Expr Ops Name Newline

context-free syntax
    Function+          → Program
    Name Name+ " =" Expr Newline+ → Function
    Expr Ops Expr      → Expr    {left,prefer,cons(binary)}
    Name Expr+         → Expr    {avoid,cons(apply)}
    "if" Expr "then" Expr "else" Expr → Expr {cons(ifThenElse)}
    "(" Expr ")"       → Expr    {bracket}
    Name               → Expr    {cons(argument)}
    Int                → Expr    {cons(literal)}
    "_"               → Ops     {cons(minus)}
    "+"               → Ops     {cons(plus)}
    "=="              → Ops     {cons(equal)}

```

Listing 13.10: Code in SDF

```

1 import types.*;
2 import org.antlr.runtime.*;
3
4 public class TestEvaluator
5     public static void main(String[] args) throws Exception {
6
7         // Parse file to program
8         ANTLRFileStream input = new ANTLRFileStream(args[0]);
9         FLLexer lexer = new FLLexer(input);
10        CommonTokenStream tokens = new CommonTokenStream(lexer);
11        FLParser parser = new FLParser(tokens);
12        Program program = parser.program();
13
14        // Parse sample expression
15        input = new ANTLRFileStream(args[1]);
16        lexer = new FLLexer(input);
17        tokens = new CommonTokenStream(lexer);
18        parser = new FLParser(tokens);
19        Expr expr = parser.expr();
20
21        // Evaluate program
22        Evaluator eval = new Evaluator(program);
23        int expected = Integer.parseInt(args[2]);

```

Listing 13.11: Code in Java

```

1  #!/usr/local/bin/python
2  # wiki: BGF
3  import os
4  import sys
5  import slpsns
6  import elementtree.ElementTree as ET
7
8  # root::nonterminal* production*
9  class Grammar:
10     def __init__(self):
11         self.roots = []
12         self.prods = []
13     def parse(self, fname):
14         self.roots = []
15         self.prods = []
16         self.xml = ET.parse(fname)
17         for e in self.xml.findall('root'):
18             self.roots.append(e.text)
19         for e in self.xml.findall(slpsns.bgf_('production')):
20             prod = Production()
21             prod.parse(e)
22             self.prods.append(prod)

```

Listing 13.12: Code in Python

Chapter 14

Literature

BIBTeX is a JSON-like format for bibliographic entries. Encode each source once as a BIBTeX entry, give it a name and refer to it from any place in your thesis. The bibliography at the end of the thesis will be compiled automatically from those entries that are referenced at least once, it will also be automatically sorted and fancyfied (URLs, DOIs, etc).

DOI is a digital object identifier, it is uniquely and immutably assigned to any paper published in a well-established journal or conference proceedings and can be used to refer to it. When used in a browser, it resolves to a publisher's website where paper can be obtained. Including DOIs in citations is considered good practice and lets the readers of your thesis get to the text of the paper in one click. Books do not have DOIs, only ISBNs; some workshop proceedings and most unofficial publications do not have DOIs. If you want to get a DOI assigned to your work such as a piece of code, upload it to [FigShare](#).

Keys in key-value pairs within each BIBTeX entry are never quoted, values usually are, but can also be included within curly brackets or left as is, which works fine for numbers (e.g., years). If you want to preserve the value from any adjustments (e.g., no recapitalisation in titles), use curly braces *and* quotes. Separate authors and editors by “and”, which will automatically be mapped to commas or left as “and”s as necessary.

14.1 Books

[GJ08] is just as good as the Dragon Book, but newer and has an awesome extended bibliography available for free.

```
@book{GruneJacobs,
  author   = "D. Grune and C. J. H. Jacobs",
  title    = "{Parsing Techniques: A Practical Guide}",
  series   = "Monographs in Computer Science",
  edition  = 2,
  publisher = "Springer",
  url      = "http://www.cs.vu.nl/~dick/PT2Ed.html",
  year     = 2008,
}
```

14.2 Journal papers

Not all TOSEM papers are hard to read [KLV05].

```
@article{GrammarwareAgenda,
  author    = "Paul Klint and Ralf L{\a}mmel and Chris Verhoef",
  title     = "{Toward an Engineering Discipline for Grammarware}",
  journal   = "ACM Transactions on Software Engineering Methodology (TOSEM)",
  volume    = 14,
  number    = 3,
  year      = 2005,
  pages     = "331--380",
}
```

14.3 Conference papers

There is no limit to how many grammars can be used in one paper, but the current record stands at 569 [Zay13].

```
@inproceedings{Micropatterns2013,
  author = "Vadim Zaytsev",
  title = "{Micropatterns in Grammars}",
  booktitle = "{Proceedings of the Sixth International Conference on Software Language Engineering (SLE 2013)}",
  year = 2013,
  editor = "Martin Erwig and Richard F. Paige and Eric Van Wyk",
  volume = "8225",
  series = "LNCS",
  pages = "117--136",
  address = "Switzerland",
  month = oct,
  publisher = "Springer International Publishing",
  doi = "10.1007/978-3-319-02654-1_7",
}
```

14.4 Theses

The seventh PhD student of Paul Klint was Jan Rekers [Rek92].

```
@phdthesis{Rekers92,
  author = "J. Rekers",
  title = "{Parser Generation for Interactive Environments}",
  school = "University of Amsterdam",
  year = 1992,
  url = "http://homepages.cwi.nl/~paulk/dissertations/Rekers.pdf",
}
```

There is also `mastersthesis` type with exactly the same structure for referring to Master's theses.

14.5 Technical reports

The original seminal work introducing two-level grammars was never published in any book or conference, but there is a technical report explaining it [vW65]. SMC, or *Stichting Mathematisch Centrum*, was the old name of CWI fifty years ago.

```
@techreport{Wijngaarden65,
  author = "Adriaan van Wijngaarden",
  title = "{Orthogonal Design and Description of a Formal Language}",
  month = oct,
  year = 1965,
  institution = "SMC",
  type = "{MR 76}",
  url = "http://www.fh-jena.de/~kleine/history/languages/VanWijngaarden-MR76.pdf",
}
```

14.6 Wikipedia

You do not refer to Wikipedia from academic writing, it works the other way around.

14.7 Anything else

You can refer to pretty much anything (websites, blog posts, software) through `misc` type of entry [\[Par08\]](#):

```
@misc{ANTLR,  
  author      = "Terence Parr",  
  title       = "{ANTLR}---ANother Tool for Language Recognition",  
  howpublished = "Software",  
  url         = "http://antlr.org",  
  year        = "2008"  
}
```

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