**Engineering Large Software Systems Notes**

*Created: 2024-09-04*

*Updated: 2024-09-25*

**References**

* Engineering Large Software Systems course at the University of Toronto

**Prerequisites**

1. Design pattern theory (eg. factory, builder, observer, strategy, etc.)

* Observer: whenever action occurs, observers will be notified, aka. listeners.
* Factory: a class that can generate more classes.
* Strategy: similar to factory but dealing with functions.

2. Testing code

* Unit testing
* Integration testing

3. Code smells

* Anti patterns when writing code

4. Code design principles (eg. SOLID)

* S: single responsibility principle, every module should focus on one task.
* O: open-close principle, open for extension closed for modification.

5. Git usage

* Git merge vs rebase

## 1. Large Software Systems

***Large***: Numerous contributors, impacts many stakeholders.

***Software Engineering***: six step process of creating software through SDLC, e.g., agile, waterfall. 1. requirements engineering. 2. system design. 3. implementation. 4. testing. 5. deployment. 6. maintenance.

### 1.0. SDLC Models

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| **Agile** | **Waterfall** |
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| Iterative development process. | Defined requirements at the beginning. |
| Client can change requirements. | Scope does not change. |

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| **Iterative & Incremental Model** | **V Model** |

### Requirements Engineering

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| ***Requirements Engineering***: Discovering and documenting requirements necessary for project success through talking with client, interviews, surveys.  ***Known Requirements***: What users told us.  ***Overlooked Requirements***: What users didn’t tell us yet.  ***Emergent Requirements***: What will surface while building product.  ***Functional Requirements***: What a system should do; system features.  ***Non-Functional Requirements***: Requirements that enable a system to operate, maintain, and extend in quality.   * ***Operational NFRs***: performance of a system. * ***Structural NFRs***: code quality of the system. | Figure: Relationship between types of system requirements. |

### System Design

***System Design***: Defining software architecture of a system; solving structural and not code problems.

* Creating software architecture documents
* Architecture risks
* How to calculate cost for infrastructure
* Increase in **NFRs** aims to increase quality and performance of system at cost of capital and time.
* ***Vertical Scaling***: improving existing infrastructure by switching to better hardware.
* ***Horizontal Scaling***: adding/expanding more infrastructure e.g., more servers. Requires additional component to manage/integrate new additions e.g., load balancer server, Kubernetes. Creates network overhead.
  + Horizontal scaling costs more than vertical scaling because increasing spec is generally proportional to cost while adding a unit incurs extra cost with integration with the rest of the infrastructure.
  + Not practical to use only vertical scaling because there’s an upper limit to how good specs are.
  + Determine scaling method by first estimating load, expected requests per second (RPS).
* ***Scalability***: ability for application to respond well to heavy load.
* ***Elasticity***: ability for application to scale up with sudden demand.

**Deciding on Tradeoffs / Operational NFRs**

* ***Consistency***: every read returns most recent write or error.
* ***Availability***: every request to non-failing node must result in response even in server failure and update/maintenance (security updates, infrastructure version updates).
  + Metric: time-based availability. **Availability = uptime / (uptime + downtime)**.
    - Periodic request to **health check endpoint** to confirm server responsive to requests, API and DB connections/latency, cache warmed.
    - **Service-Level Agreement** (SLA), e.g., 99.9% availability allows only 43 min downtime/month. Reducing downtime involves automated recovery and alerting.
  + Metric: aggregated availability. **Availability = successful requests / total requests**.
* ***Partition tolerance***: system continues to operate even if some nodes fail. E.g., a system depending on a single database is not partition tolerant; solution: source-replica DB replication (lacks consistency in CAP assuming asynchronous writes to DB. ADIC principle. Synchronous writes loses availability but gains consistency.

A diagram of different types of relationships

Description automatically generated

Figure: ***CAP Theorem***: can only satisfy 2/3 of these characteristics.

* Low ***latency***: response time to request.
* High **performance**: improve performance by improving data retrieval (e.g., caching) and compute time.
  + ***Caching***: method to increase data retrieval performance. Retrieve from RAM instead of disk. Perform computation in advance instead of on request. E.g., memcached, redis.
  + ***Read-through caching***: data loaded into cache on demand.
  + ***Write-through caching***: write operation writes to DB and cache.

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| Cache | Pro | Con |
| Read-through | Fast for frequently accessed data and reduces load on DB. | Slow initial read if not cached.  Infrequent calls does not benefit from performance boost. |
| Write-through | Fast for accessing recently written data. | Slower write performance.  Writing too much. |
| Read-through and write-through | Fast access for most recent data. | Performance overhead. Complexity in implementation and maintenance. |

* + Increase in performance with caching leads to a:
    - decrease in code readability, freshness of data (data synchronization).
    - increase in maintenance cost (more code (bugs), infrastructure cost (reddis), memory usage.
* Size of input / output.
  + Decreasing size of output:
    - Reduces transport time.
    - Increases time to compress and decompress.
* Maintainability
* Cost

**Structural NFRs**

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| TEMP  Considerations to ensure the code quality of the system to enable it to be operated and maintained in quality.  Learning Outcomes:   * The tradeoffs between configuration and customziation * The significance of backward compatibility and its drawbacks * Deployment strategies and when to use them * The importance of tagging releases and semantic versioning |

* ***Upgradability***: ease at which a system can be upgraded.
  + Problems: keeping track of versions, dependents of upgraded code need to change, backward compatibility (supporting older versions in a version upgrade).
  + Obstacles: API specification changes, DB schema changes (data model refactoring), data migrations (add/delete rows/columns).
  + Semantic versioning: Major.minor.patch where major is backwards incompatible, minor and patch are backwards compatible, minor is substantial new functionality, patch is bug fix.
* Deployability: deployment strategies, given that multiple replicas exist and a load balancer server controls traffic to the connected servers.
  + Recreate: 1. Package new version 2. Shutdown old version (show maintenance page). While offline perform data migration, DB schema updates, etc. 3. Spin up new version.
  + Ramped: 1. Package new version 2. Shutdown one old replica. 3. Spin up one new replica. 4. Repeat.
  + Canary: manually perform step 3 in ramped. Then observe before rolling out more releases.
    - Todo: <https://sre.google/workbook/canarying-releases/>
  + Blue-green (red-black): 1. Package new version. 2. Spin up all new replicas. 3. Load balancer redirect traffic to new replicas. 4. Shut down all old replicas.

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|  | Pro | Con |
| Recreate | Simple, low cost  Don’t need to sync data since no data can be written when the service is unavailable.  Fast to fix urgent bug fixes, security vulnerability. | Website down time. |
| Ramped | Automated, risk controlled.  No downtime. | Slow to spin up shut down one by one  Possible race conditions if load balancer not perfect; requests sent to a service that’s shutting down  Need to sync data written to new and old replicas.  Will not be able to handle as many requests (solution: n+1 replicas). |
| Canary | Quick addressing of errors if they arise (find bugs quickly). | Even slower than ramped.  Need to sync data written to new and old replicas. |
| Blue-green | Efficient.  No downtime.  Don’t need to worry about different behaviour with different versions  Easy to roll back to old. version if there are bugs, advantage over ramped.  Fast to fix urgent bug fixes, security vulnerability. | Costly to duplicate.  Need to sync data written to new and old replicas. |

* Compatibility

TODO: Configurability and Customizability

### 1.3. Implementation

***Implementation***: Writing the code based on identified requirements and adherence to system design.

**Costs of Implementing Software**

1. Labor: developers, architects, management
2. Infrastructure: production and test environments
3. Maintenance: documentation, change management, tech debt

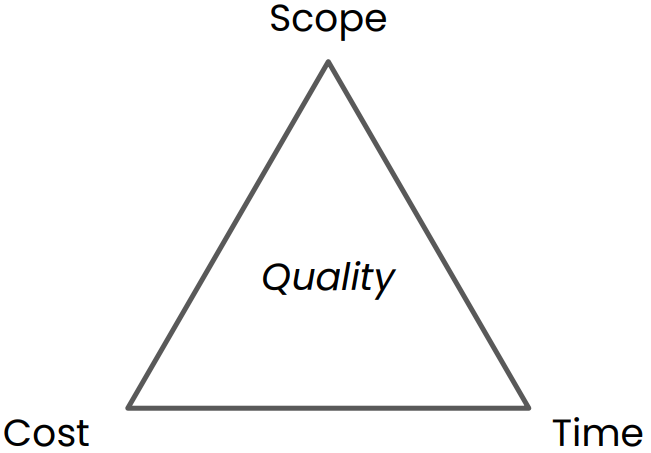


Figure: interdependence of scope, cost, and time on software quality.

**Evaluating the Success of a Project**

1. delivering on time
2. on scope
3. at cost.

See Appendix A – The CHAOS Report (1995)

### 1.4. Testing

### 1.5. Deployment

***Deployment***: Making software available to users.

### 1.6. Maintenance

***Maintenance***: Ensuring software continuously satisfy users.

### 1.7. Contributing to Open Source

***Open/Closed source software***: whether source code of a software program is public.

* Advantage of closed source: security by obscurity.
* Open source doesn’t mean free. Depends on licence.
* Free and open source software (FOSS) is both free and open source.
* Value of open source software: makes a developer’s life easier, e.g., git, TensorFlow, React.
* Why contribute? Get hired faster, access to industry talent, work on your craft.

**Software Licences**

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| <https://www.gnu.org/philosophy/categories.en.html> |  |

**Contribute to Which Project?**

* GitHub stars,
* “used by” count
* Low barrier to contribute
  + Large number of contributors
  + Large rate of contributions over time (commits over time)
  + Good documentation (e.g., readme)
  + Simple development setup (Time to hello world)
  + Streamlined (easy to understand) design
  + Healthy project

**How to Contribute?**

* Look for contributor documentation
* Fork repo. Why fork? So your changes can’t affect changes in parent repository. Then PR.

**Issue Hunting**

* Discover gaps in project by being a user
* GitHub issues: filter by tags such as “help wanted”, “good first issue”.
* Messaging forums: discord, reddit, GitHub discussions

## Appendix A – The CHAOS Report (1995)

<https://www.csus.edu/indiv/r/rengstorffj/obe152-spring02/articles/standishchaos.pdf>

* Investigation into the success of software development projects of large, medium, and small companies found that majority projects do not deliver on-time, on-budget, and on-scope.
* When a project fails, it’s important to investigate, study, report, and share the cause.
* Delivering software in smaller time intervals early and often increases success rate.
* A successful project has the following characteristics.

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|  | Criteria | Importance |
| 1 | User involvement | 19 |
| 2 | Executive management support | 16 |
| 3 | Clear statement of requirements | 15 |
| 4 | Proper planning | 11 |
| 5 | Realistic expectations | 10 |
| 6 | Smaller project milestones | 9 |
| 7 | Competent staff | 8 |
| 8 | Ownership | 6 |
| 9 | Clear vision & objectives | 3 |
| 10 | Hard-working, focused staff | 3 |

## Appendix B – Google Site Reliability Engineering Book

<https://sre.google/sre-book/table-of-contents/>

### B.3. Embracing Risk

<https://sre.google/sre-book/embracing-risk/>

[TODO]

### B.21. Handling Overload

<https://sre.google/sre-book/handling-overload/>

* The responsibilities of load balancer policies are to distribute work to prevent system overload.
* Overload is inevitable if usage grows. In handling these situations, “redirect when possible, serve degraded results when necessary, and handle resource errors transparently when all else fails.”
* The **queries per second** metric cannot accurately capture resource requirement when measuring capacity. Instead, measure capacity in terms of available resources, e.g., available CPUs, memory. Consequently, cost of a request can be measured in normalized CPU time.
* **Global overload** occurs when all global resources are full and when it does it’s important to deliver errors to only affected customers (the app making the request). This is done by allocating certain CPU consumption times to each app: gmail, calendar, android etc.
* Rejecting requests takes up resources and in certain cases more than performing the request. **Adaptive throttling** works on the client side to restrict the requests sent if within the last two minutes the number of requests exceeds multiples of accepted requests, .
* Requests to backend are associated with 4 levels of **criticality**: CRITICAL\_PLUS (serious user-visible impact), CRITICAL (user-visible impact), SHEDDABLE\_PLUS (partial unavailability expected), SHEDDABLE (frequent partial unavailability expected).
* In **task-level overload protection**, the executor load average is calculated by finding number of active (running/ready) threads in the process. Spikes in requests get smoothed out while sustained increase starts getting rejected. Can use memory pressure instead of CPU usage/CPUs reserved.

[CRITICAL TODO: **Handling Overload Errors**]

### B.22. Addressing Cascading Features

<https://sre.google/sre-book/addressing-cascading-failures/>

[CRITICAL TODO]