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Introduction to Data Engineering

Week 3



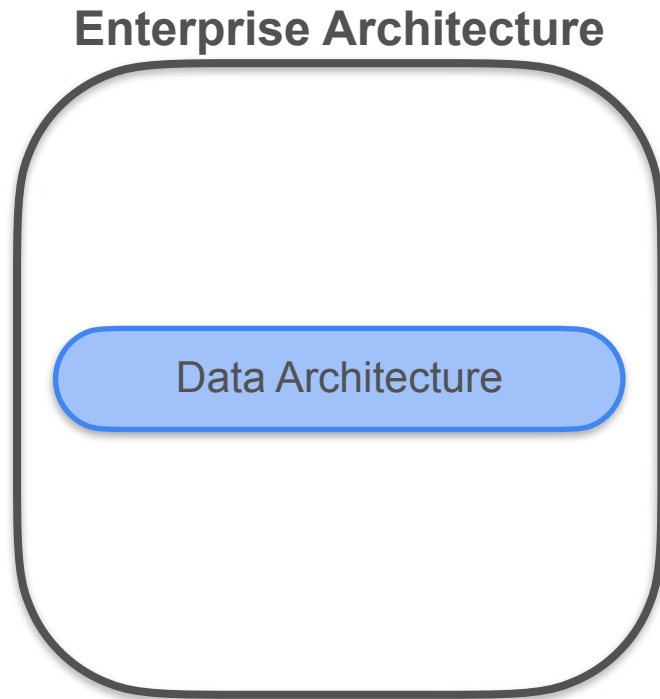
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Data Architecture

Week 3 Overview

Week 3 Overview

- Enterprise & data architecture

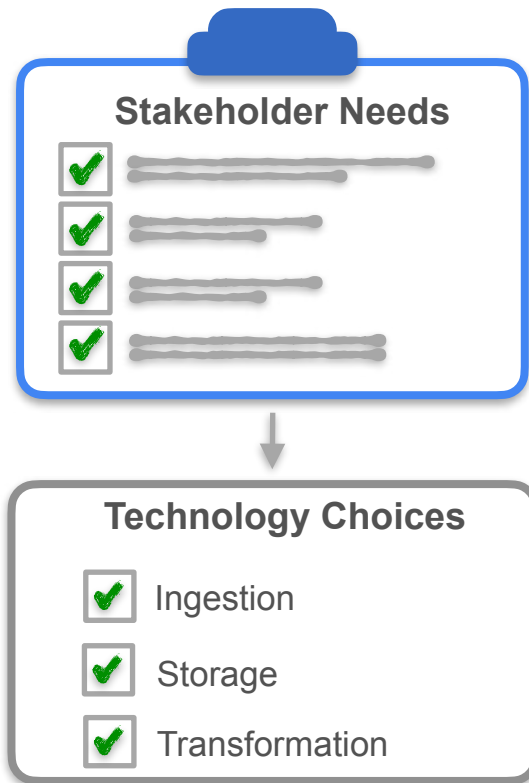


Week 3 Overview

- Enterprise & data architecture
- Specific architecture examples

Week 3 Overview

- Enterprise & data architecture
- Specific architecture examples
- Choosing technologies



Week 3 Overview

- Enterprise & data architecture
- Specific architecture examples
- Choosing technologies
- Guiding architectural principles



Think like an architect!

Week 3 Overview

- Enterprise & data architecture
- Specific architecture examples
- Choosing technologies
- Guiding architectural principles
- Trade-off evaluation



Cost



Performance



Scalability



Security

Week 3 Overview

- Enterprise & data architecture
- Specific architecture examples
- Choosing technologies
- Guiding architectural principles
- Trade-off evaluation
- AWS Well-Architected Framework





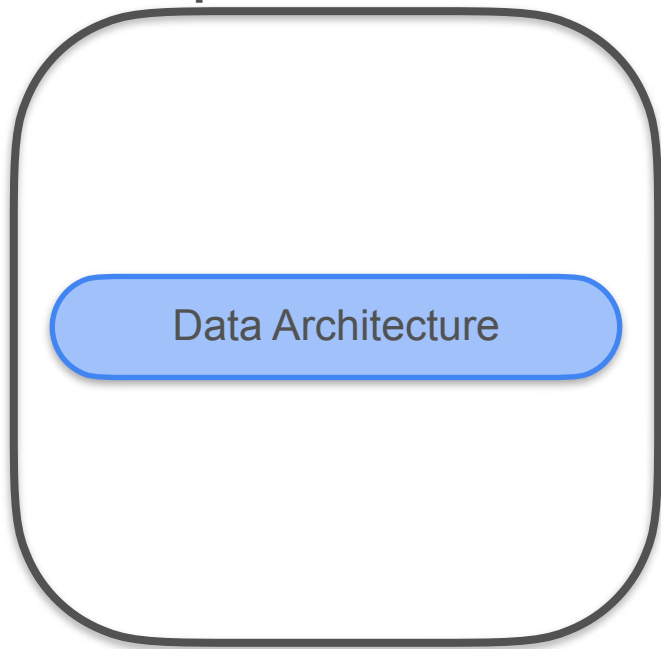
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Data Architecture

What is Data Architecture?

Enterprise Architecture

Enterprise Architecture



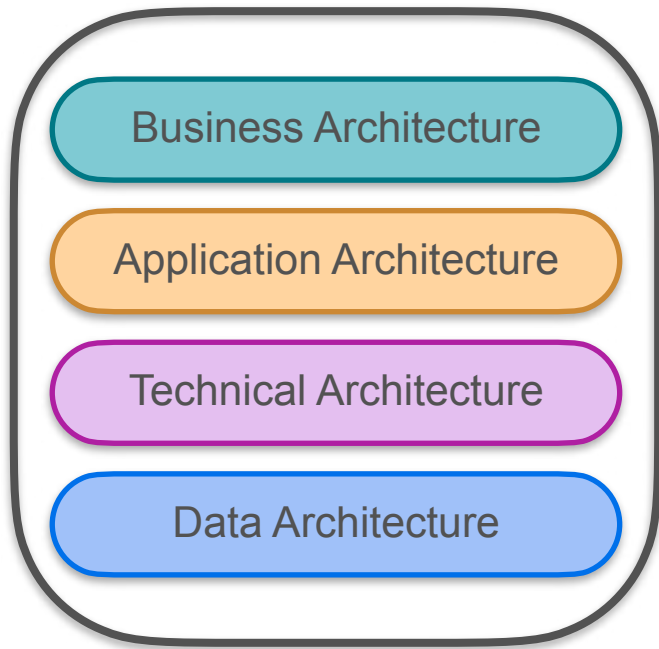
Data Architecture

“the design of systems **to support change in an enterprise**, achieved by flexible and reversible decisions reached through a careful evaluation of trade-offs”

“the design of systems **to support the evolving data needs of an enterprise**, achieved by flexible and reversible decisions reached through a careful evaluation of trade-offs.”

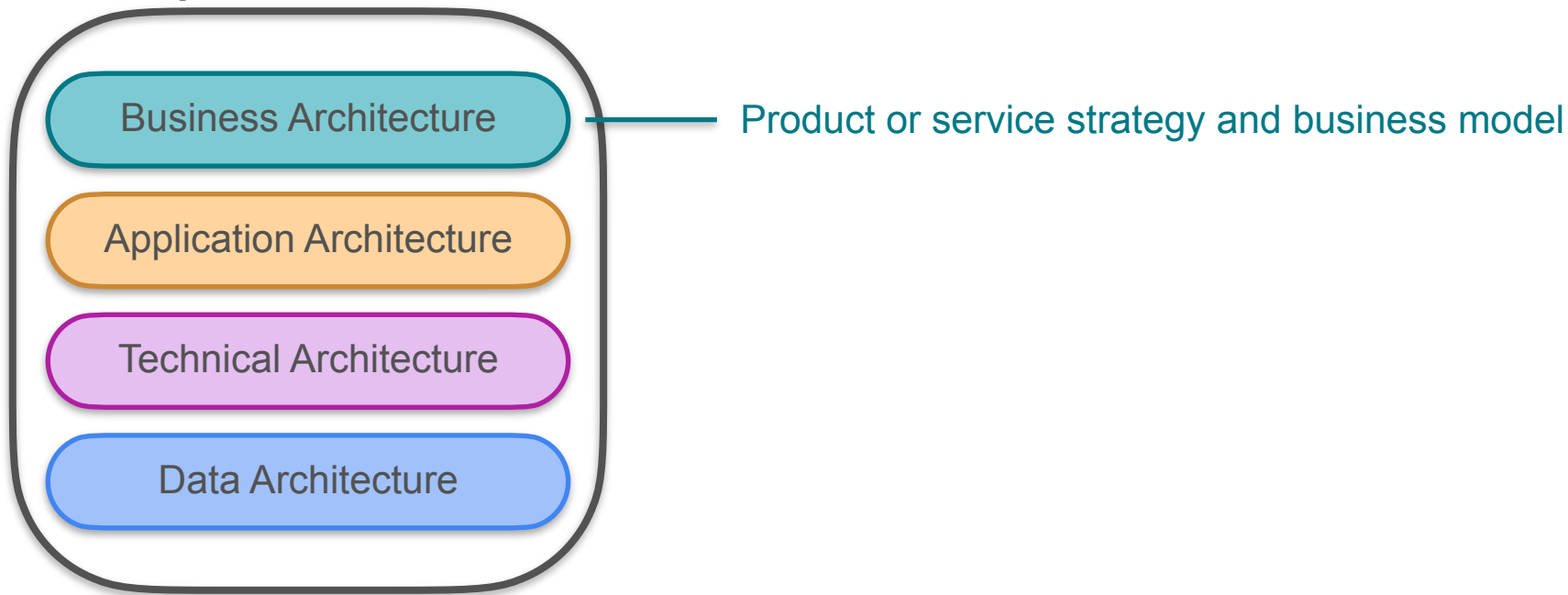
Enterprise Architecture

Enterprise Architecture



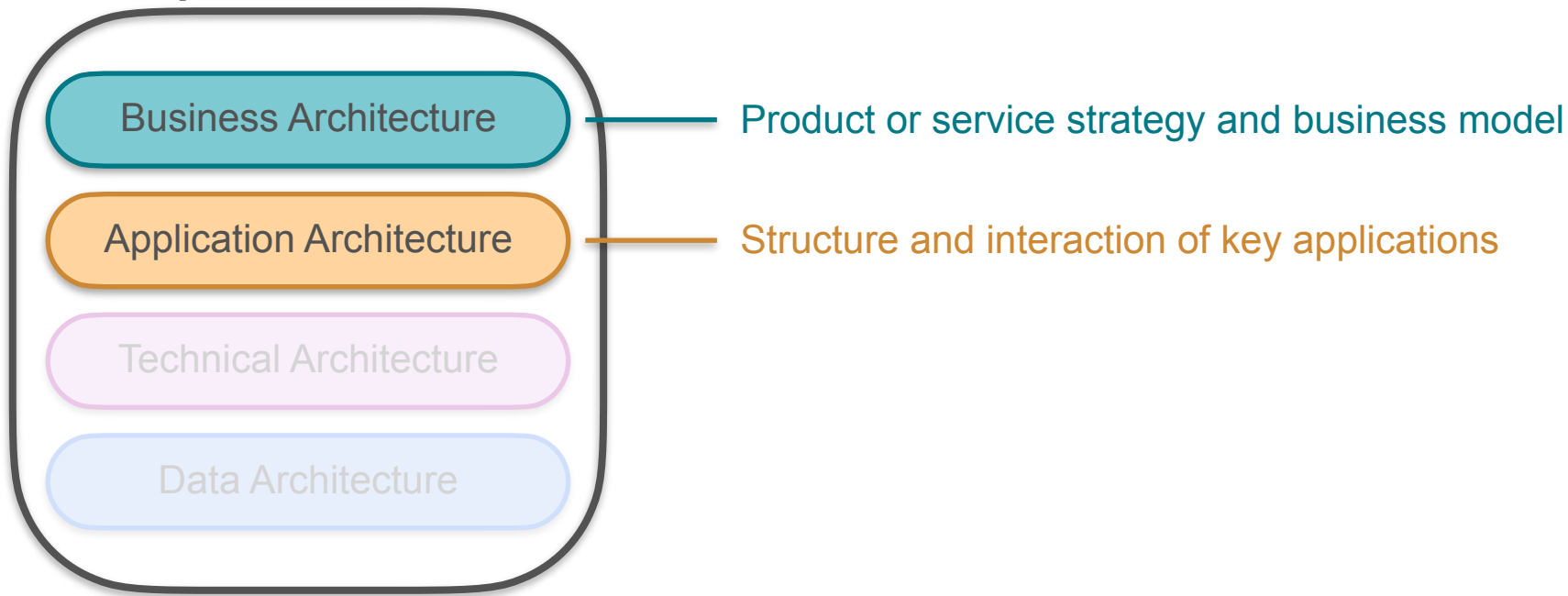
Enterprise Architecture

Enterprise Architecture



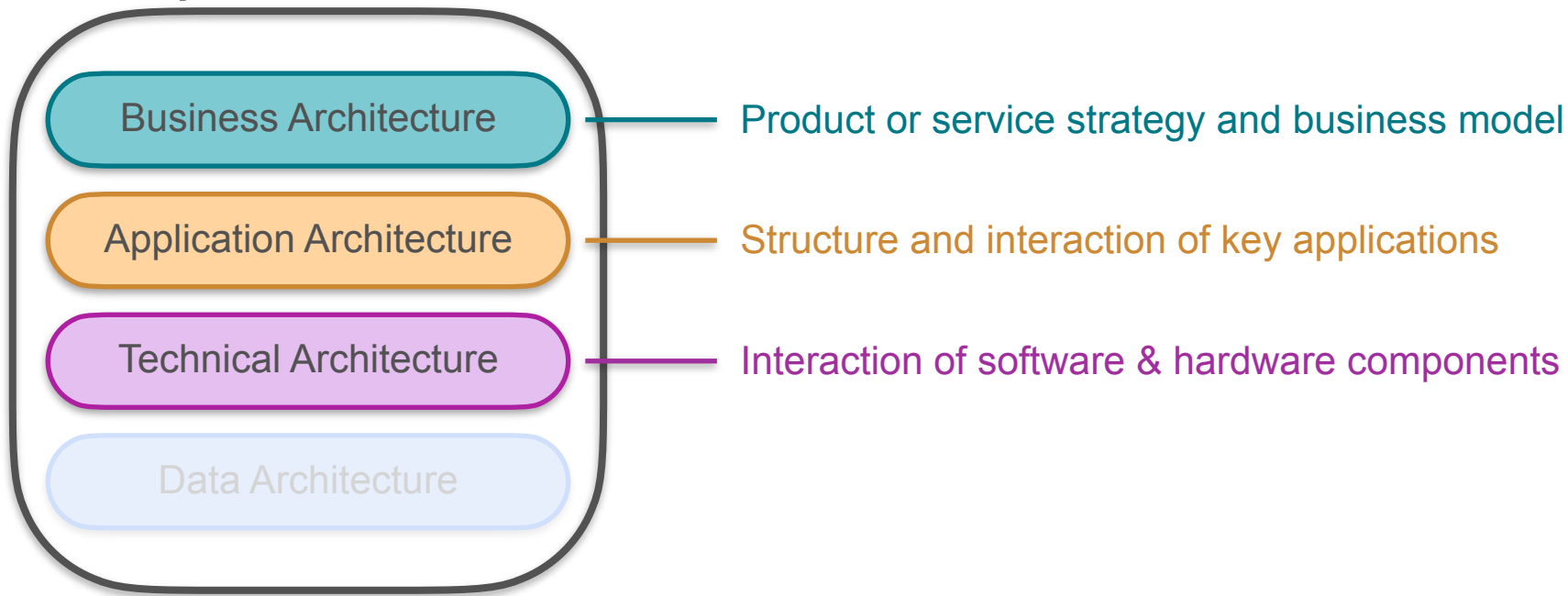
Enterprise Architecture

Enterprise Architecture



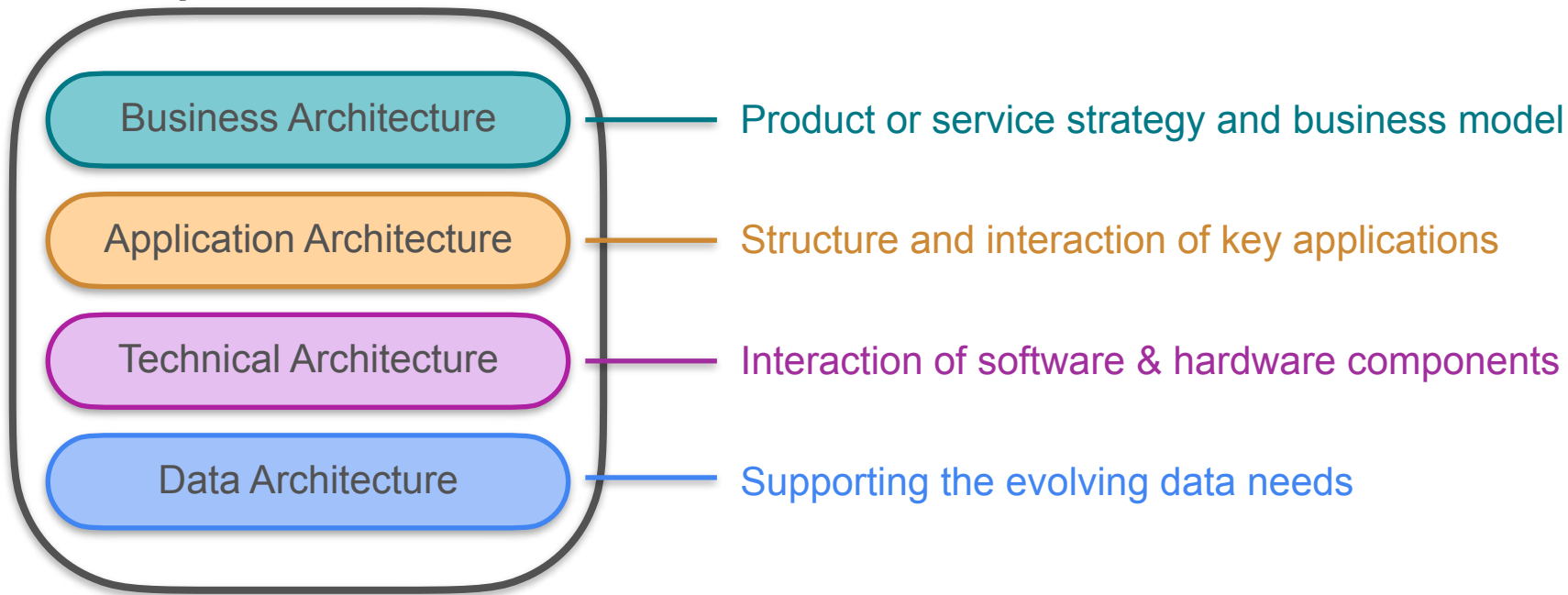
Enterprise Architecture

Enterprise Architecture



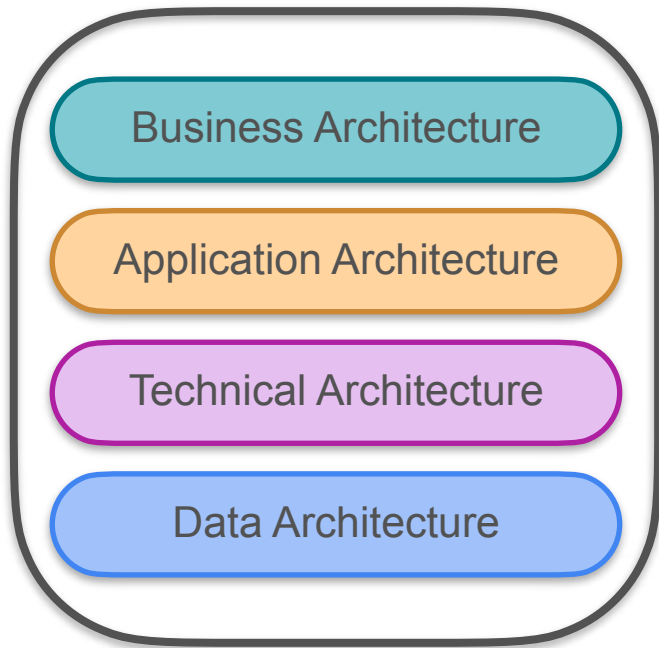
Enterprise Architecture

Enterprise Architecture

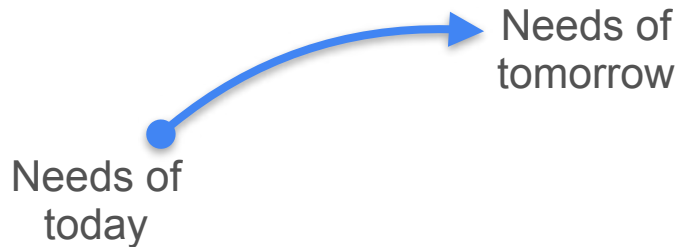


Enterprise Architecture

Enterprise Architecture



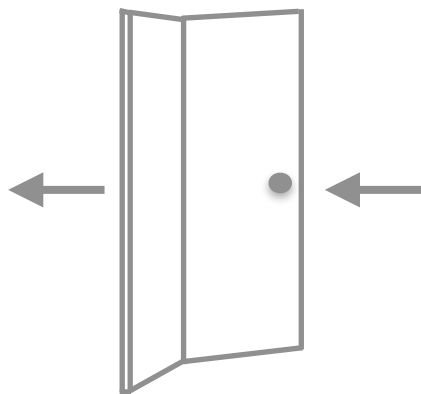
Change management



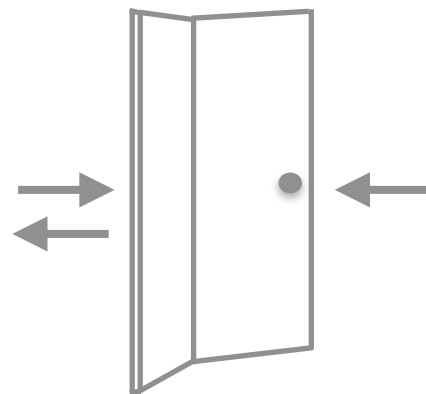
**Adapt to organizational
changes**

One-Way and Two-Way Door Decisions

Organization Decisions



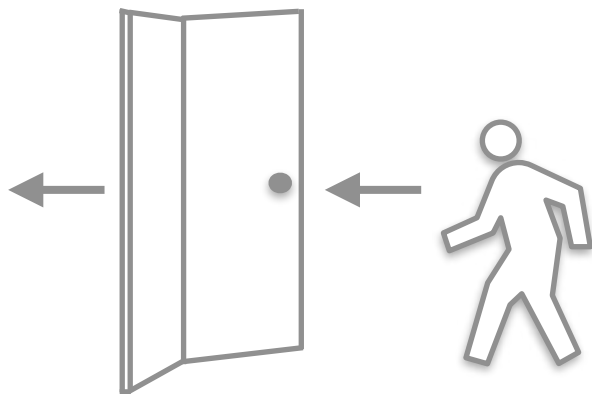
One-way



Two-way

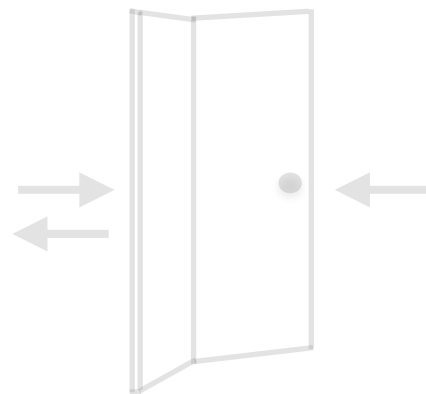
One-Way and Two-Way Door Decisions

Organization Decisions



One-way

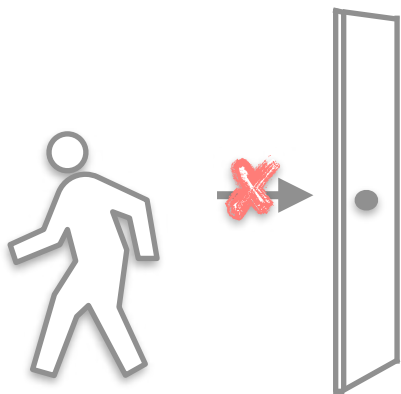
A decision that is almost impossible to reverse



Two-way

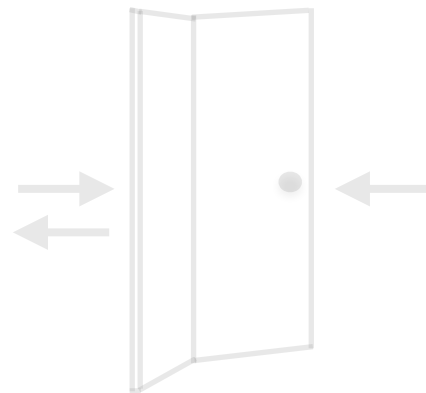
One-Way and Two-Way Door Decisions

Organization Decisions



One-way

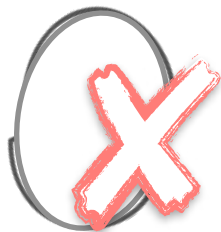
A decision that is almost impossible to reverse



Two-way

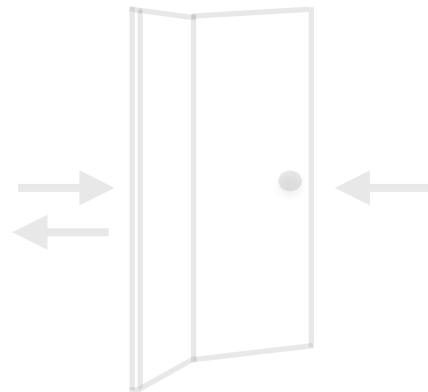
One-Way and Two-Way Door Decisions

Organization Decisions



One-way

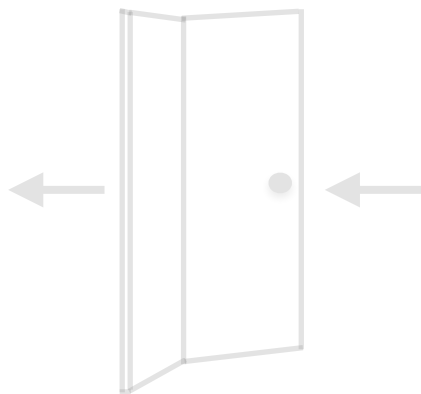
A decision that is almost impossible to reverse



Two-way

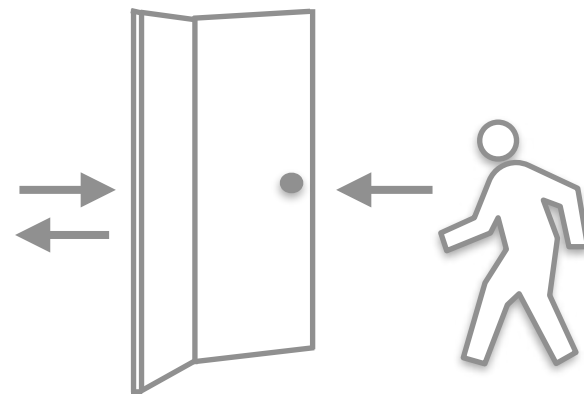
One-Way and Two-Way Door Decisions

Organization Decisions



One-way

A decision that is almost impossible to reverse

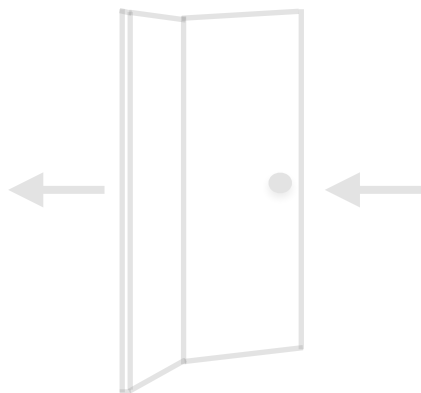


Two-way

An easily reversible decision

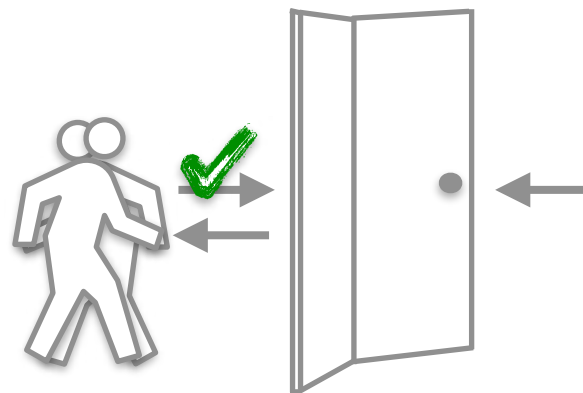
One-Way and Two-Way Door Decisions

Organization Decisions



One-way

A decision that is almost impossible to reverse



Two-way

An easily reversible decision

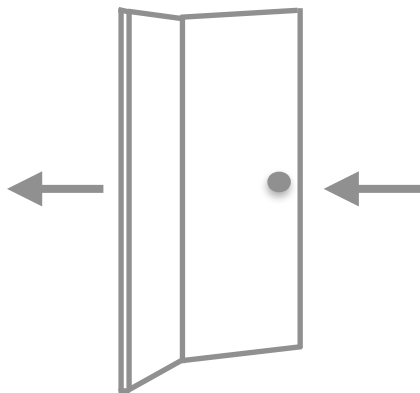
Two-Way Door Decision

S3 Object Storage Classes

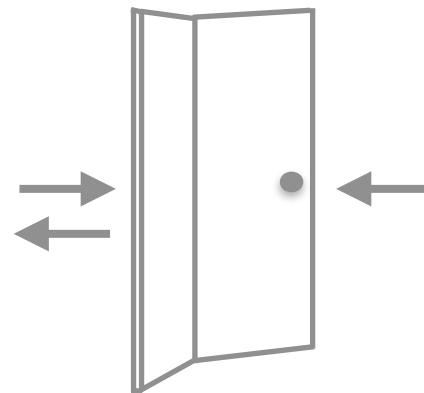


Reversible Decisions

Organization Decisions



One-way

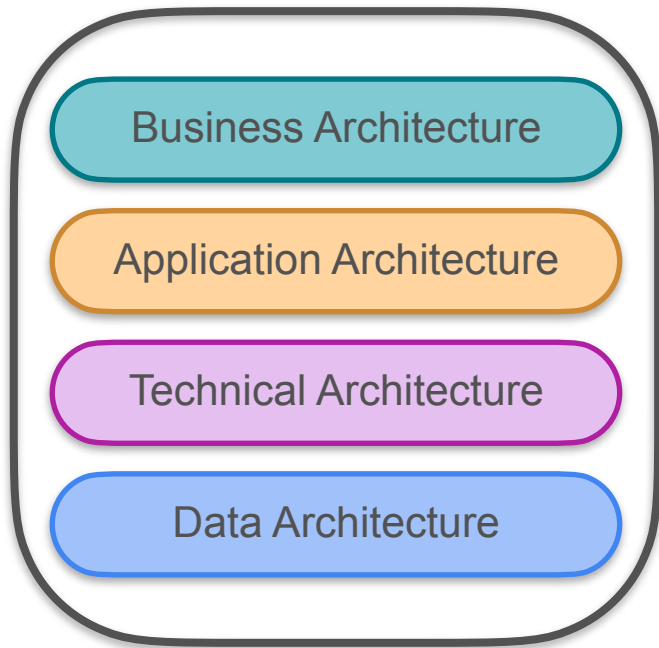


Two-way

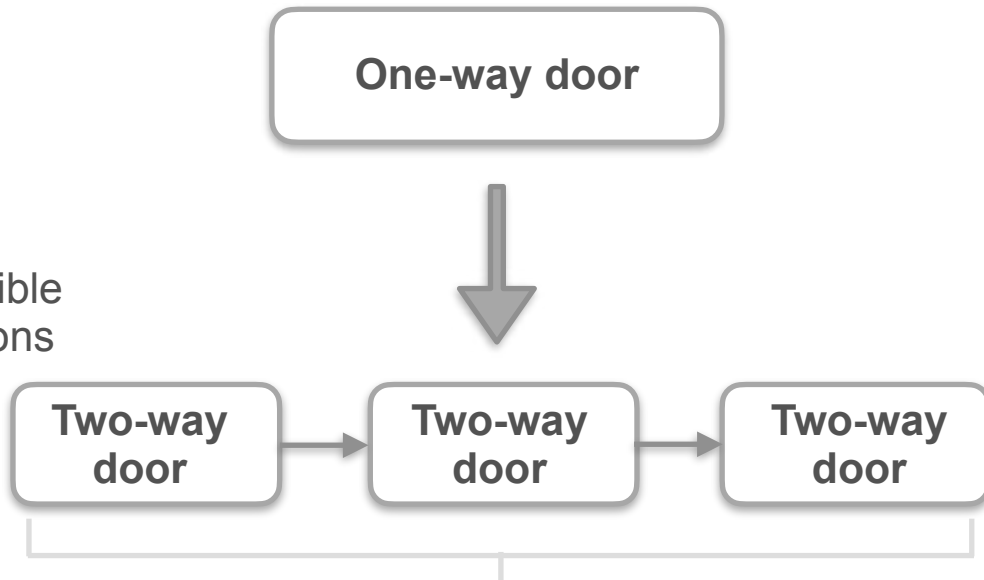
Reversible
Decisions

Reversible Decisions

Enterprise Architecture



Reversible
Decisions





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Data Architecture

Conway's Law

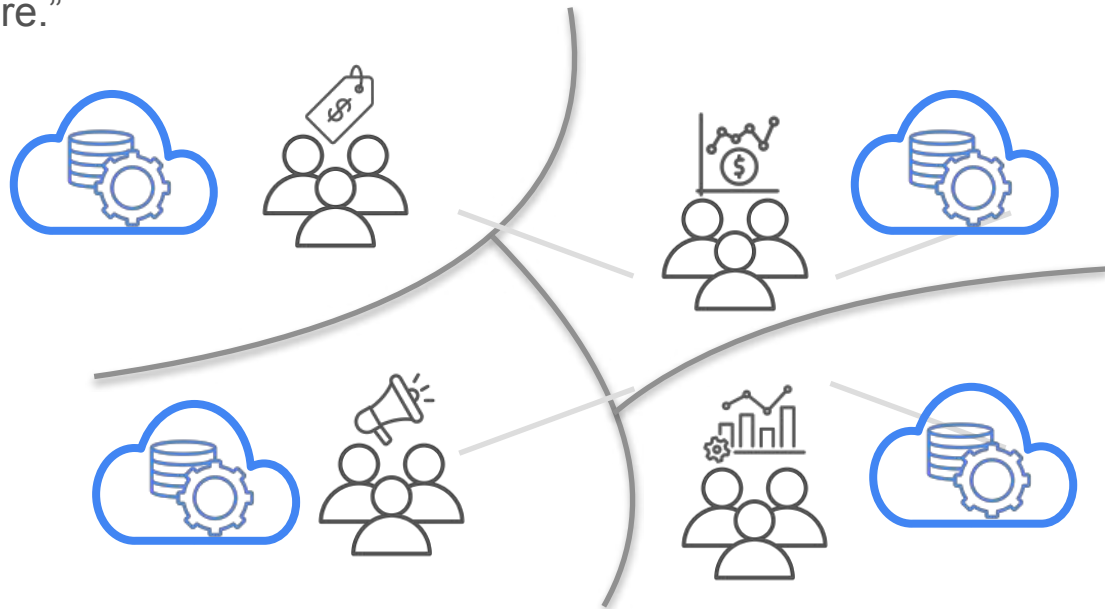
Conway's Law

“Any organization that designs a system will produce a design whose structure is a copy of the organization's communication structure.”

- Melvin Conway

Conway's Law

“Any organization that designs a system will produce a design whose structure is a copy of the organization's communication structure.”





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Data Architecture

Principles of Good Data Architecture

Principles of Good Data Architecture

1. Choose common components wisely
2. Plan for failure!
3. Architect for scalability
4. Architecture is leadership
5. Always be architecting
6. Build loosely coupled systems
7. Make reversible decisions
8. Prioritize security
9. Embrace FinOps

Principles of Good Data Architecture

- 1. Choose common components wisely
- 4. Architecture is leadership

How data architecture impacts other teams and individuals

- 5. Always be architecting
- 6. Build loosely coupled systems
- 7. Make reversible decisions

Data architecture is an ongoing process

- 2. Plan for failure!
- 3. Architect for scalability
- 8. Prioritize security
- 9. Embrace FinOps

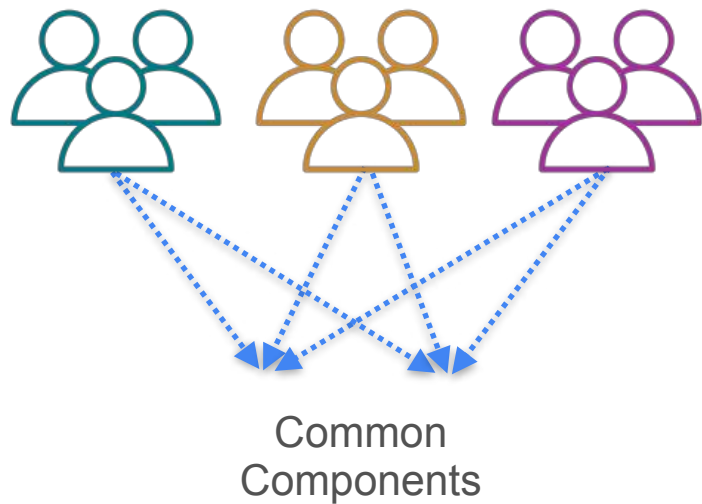
Unspoken but understood priorities

Principles of Good Data Architecture

1. Choose common components wisely
4. Architecture is leadership

**How data architecture impacts
other teams and individuals**

Common Components



Examples



Object Storage



Version-Control
Systems

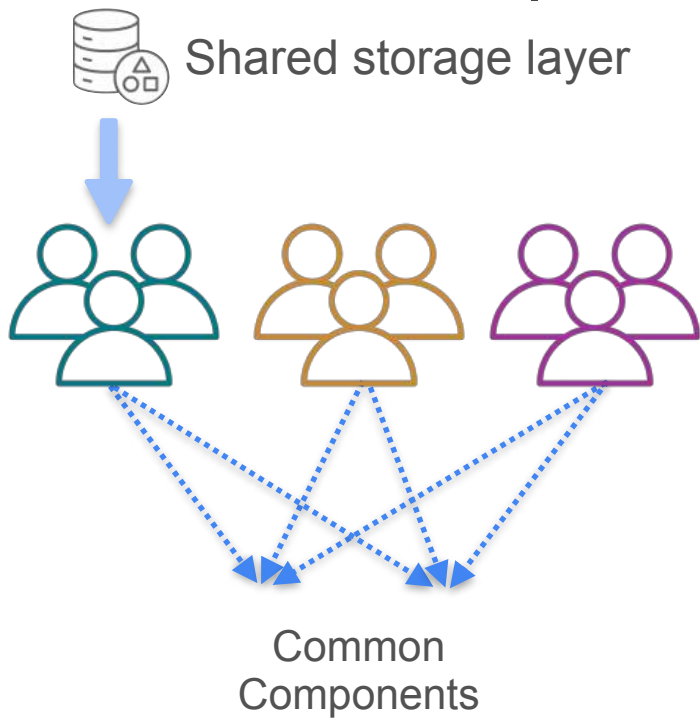


Observability &
Monitoring Systems



Processing Engines

Common Components



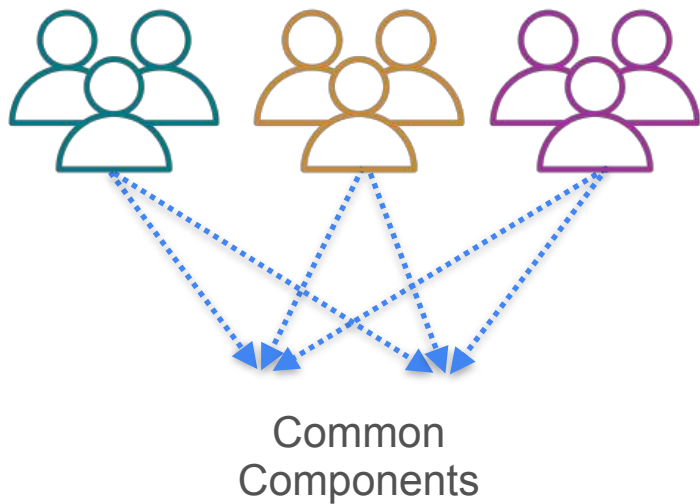
Common components:

- Facilitate team collaboration
- Break down silos

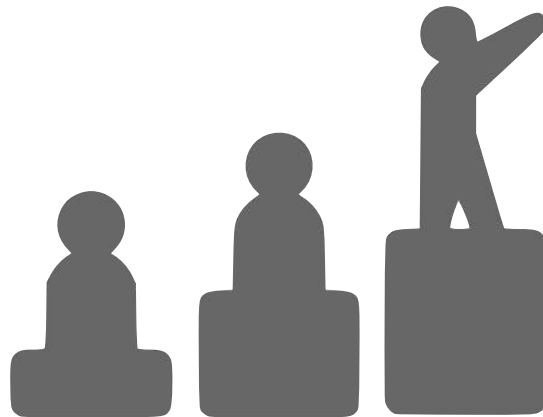
“Wise” Choice:

- Identify tools that benefit all teams
- Avoid a one-size-fits-all approach

Architecture Leadership



Architecture is leadership



Seek mentorship from data architects



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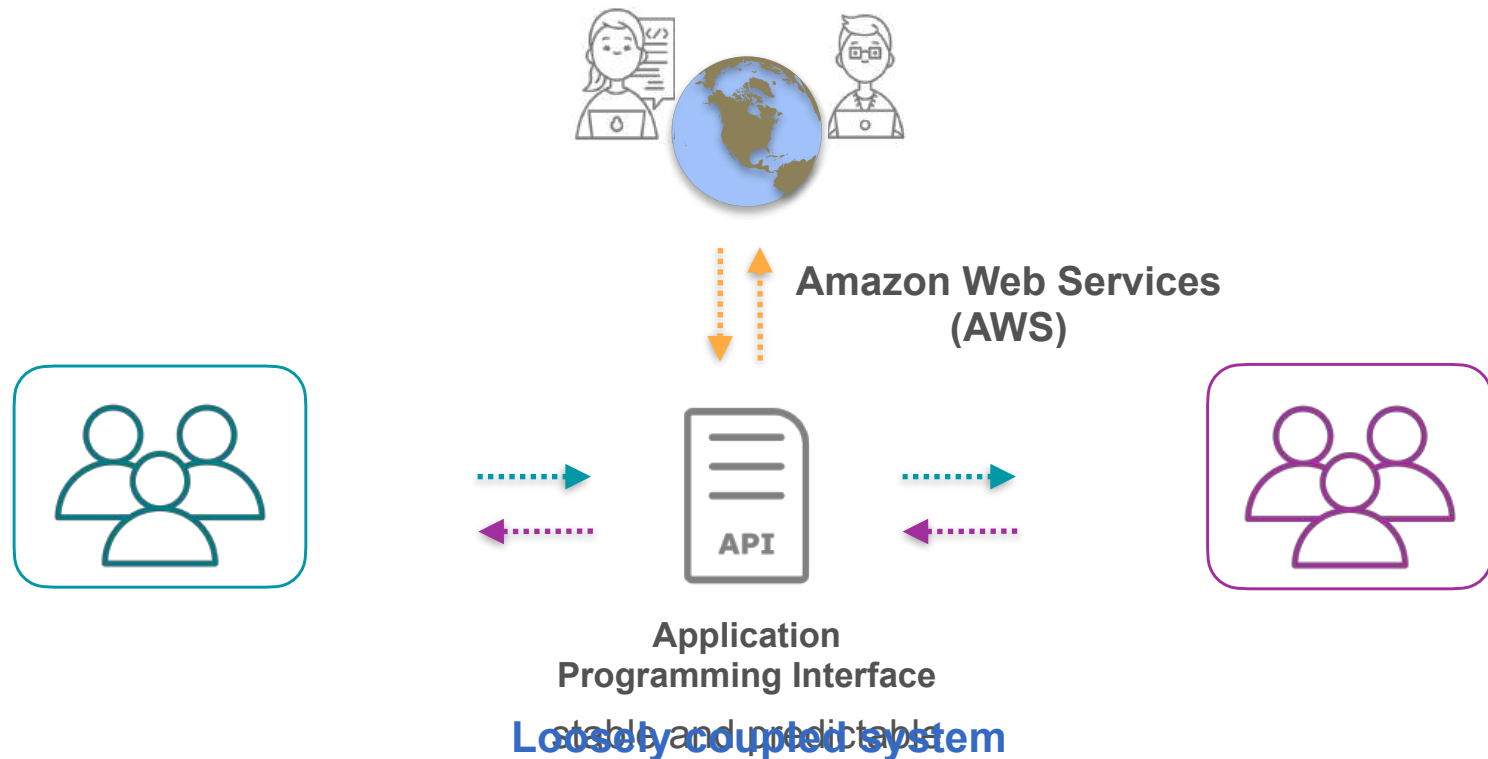
Data Architecture

Always Architecting

Application Programming Interface



Application Programming Interface



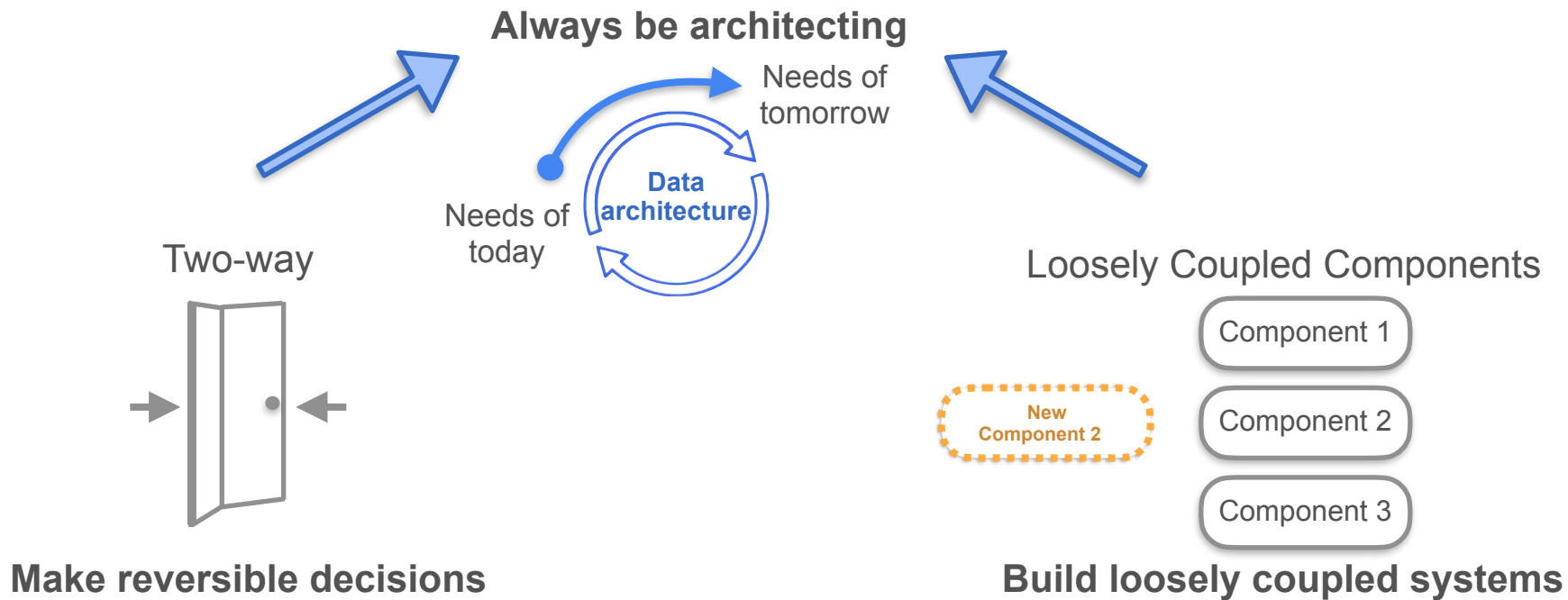
Always Architecting

Make reversible decisions

Build loosely coupled systems

Always be architecting

Always Architecting





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Data Architecture

When Your Systems Fail

When Your Systems Fail

Plan for failure

Architect for scalability

Prioritize security

Embrace FinOps

Plan for Failure

Practical and Quantitative Approach

System metrics:

Availability

Reliability

Durability

Plan for Failure

Availability

The percentage of time an IT service or a component is expected to be in an operable state.



Amazon S3

Examples of S3 Classes	S3 One Zone-IA	S3 Standard
Availability	99.5%	99.99%
Annual Downtime in hours	44-hour downtime	1-hour downtime

Plan for Failure

Reliability

The probability of a particular service or component performing its intended function during a particular time interval



Plan for Failure

Durability

The ability of a storage system to withstand data loss due to hardware failure, software errors, or natural disasters.



Amazon S3

Durability

99.999999999%

(11 nines)

Plan for Failure

Recovery Time Objective RTO

The maximum acceptable time for a service or system outage

For example: Consider the Impact to customers

Recovery Point Objective RPO

A definition of the acceptable state after recovery

For example: Consider the maximum acceptable data loss

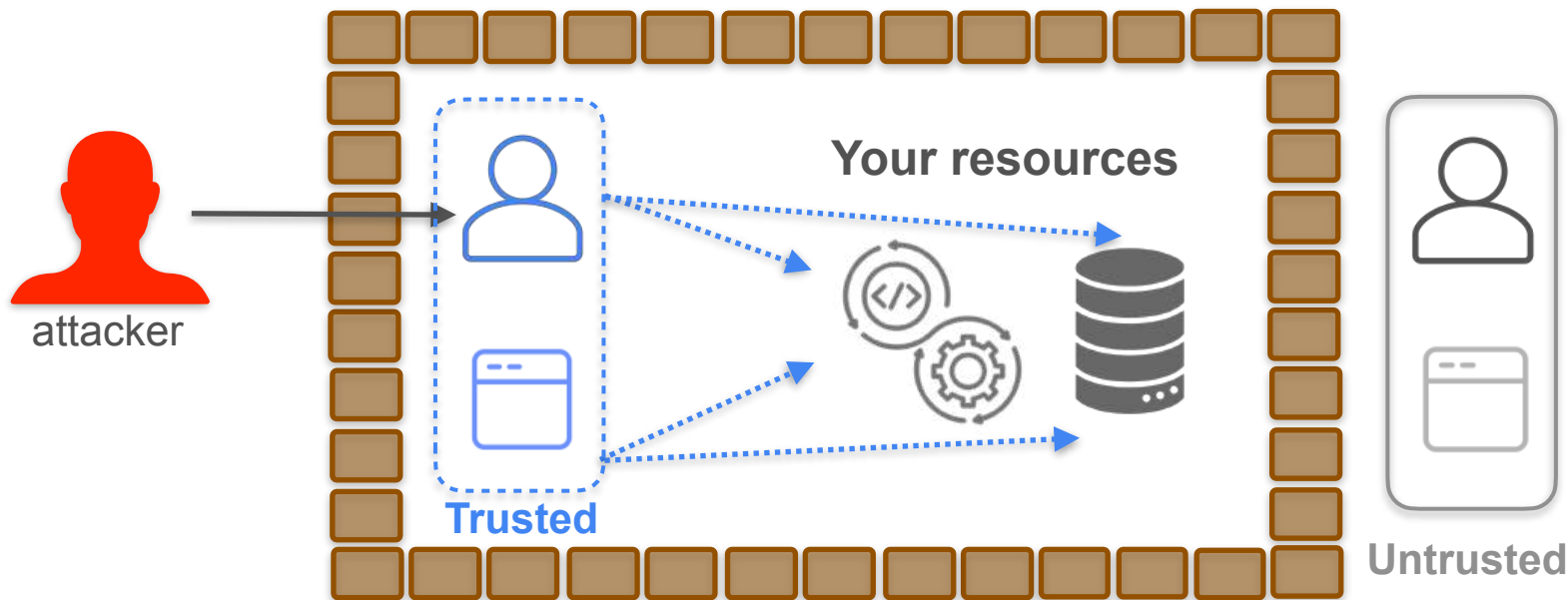
Prioritize Security

- Culture of security
- Principle of least privilege
- Zero-trust security



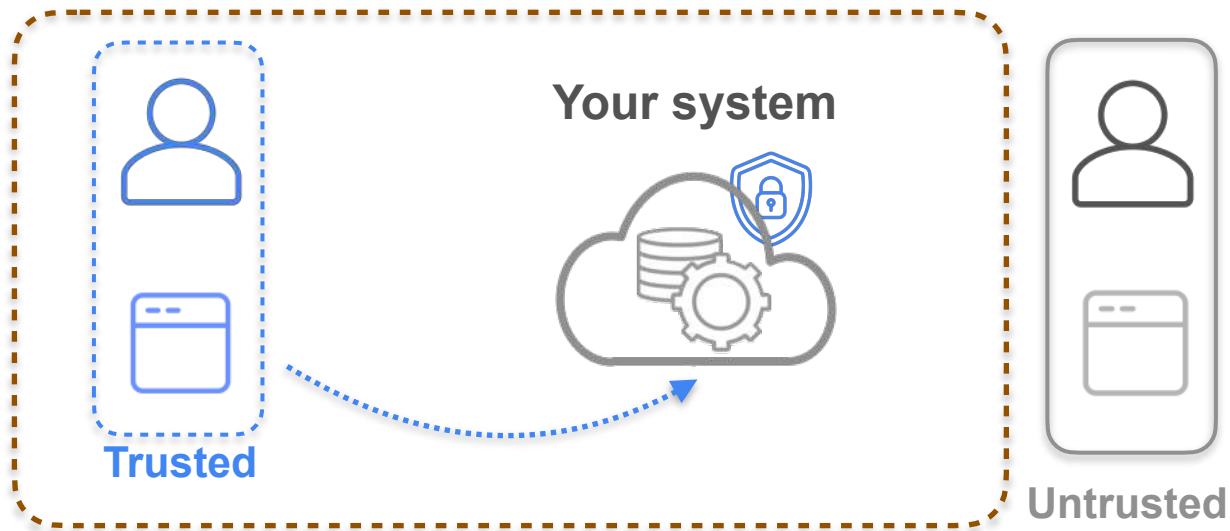
Prioritize Security

Hardened-Perimeter Approach



Prioritize Security

Zero-Trust Security



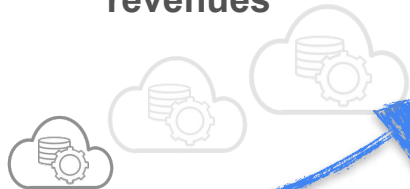
Architecting for Scalability & Embrace FinOps

Large Unforeseen Costs



OR

Missed Opportunities for revenues



Demand

Plan for Failure!



**Embrace
FinOps**

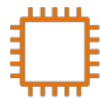
**Architect for
scalability**

Embrace FinOps

How to optimize a daily job in terms of cost and performance?



EC2 instance



EC2 instance



Spot Instance



Spot Instance



Spot Instance



Spot Instance



EC2 instance

Embrace FinOps

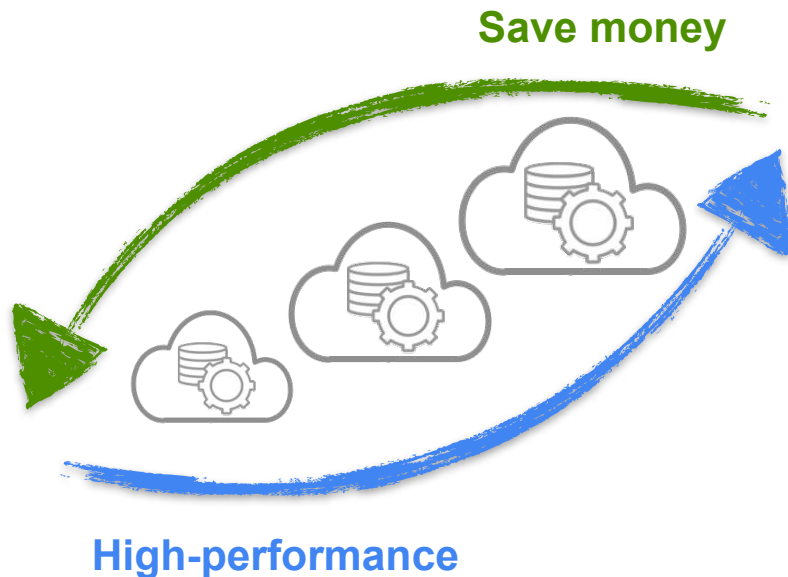
Pay-as-you-go models:

- Cost-per-query model
- Cost-per-processing-capacity model



Manage budgets,
priorities and efficiency

Readily Scalable:



When Your Systems Fail

Plan for failure

Architect for scalability

Prioritize security

Embrace FinOps



**Serve the needs of
your organization**



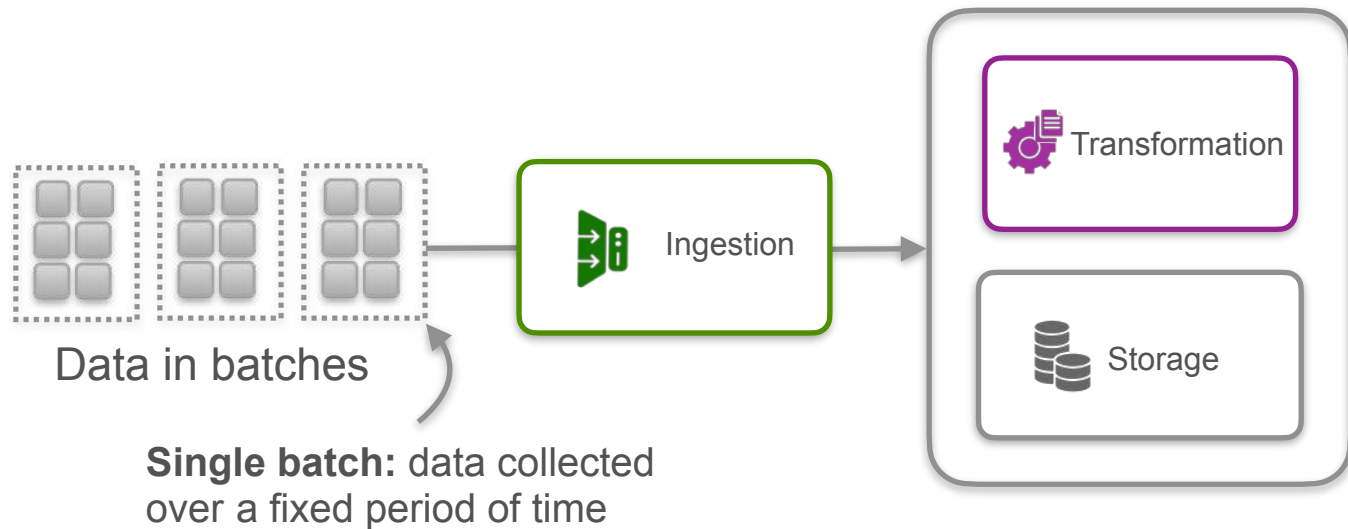
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Data Architecture

Batch Architectures

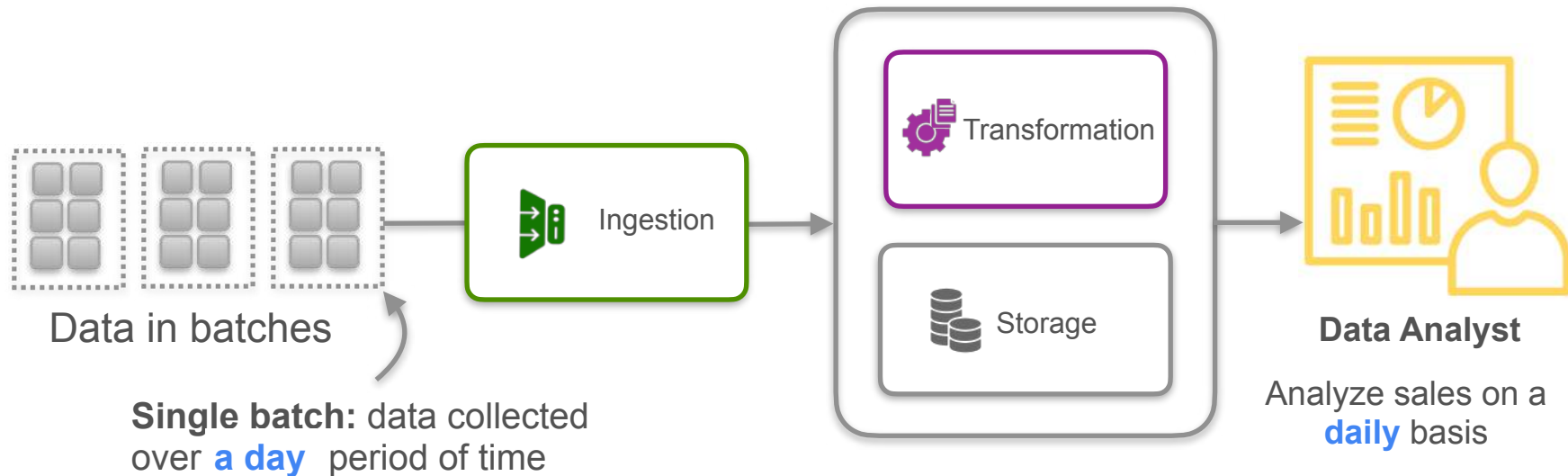
Batch Data Architecture

Real-time analysis is not critical



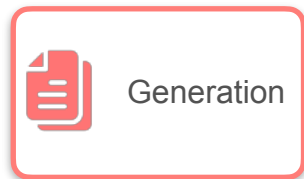
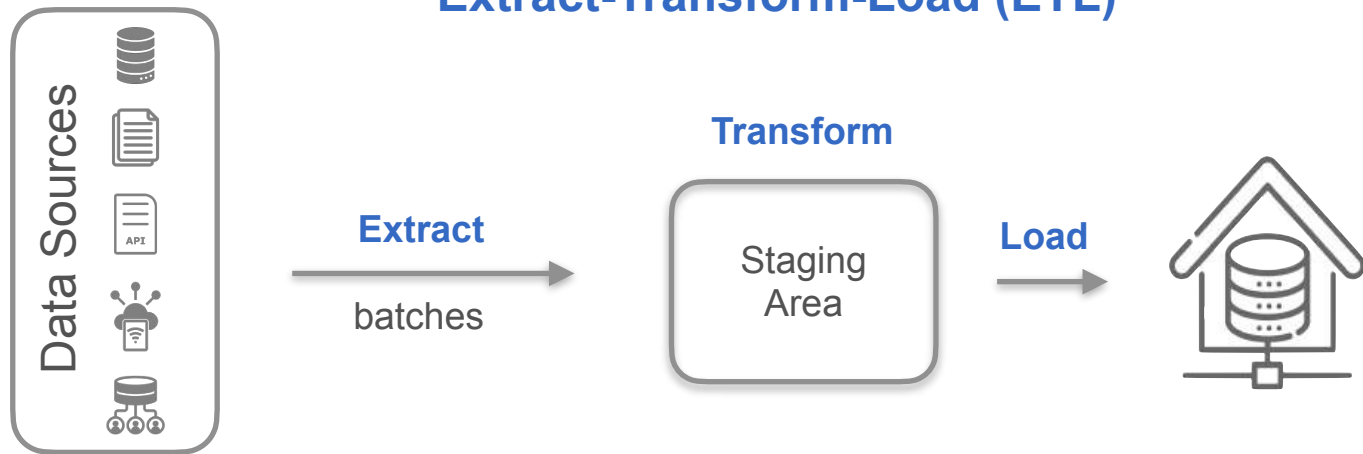
Batch Data Architecture

Real-time analysis is not critical



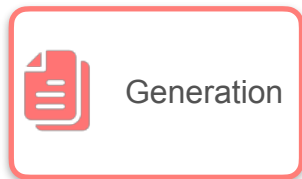
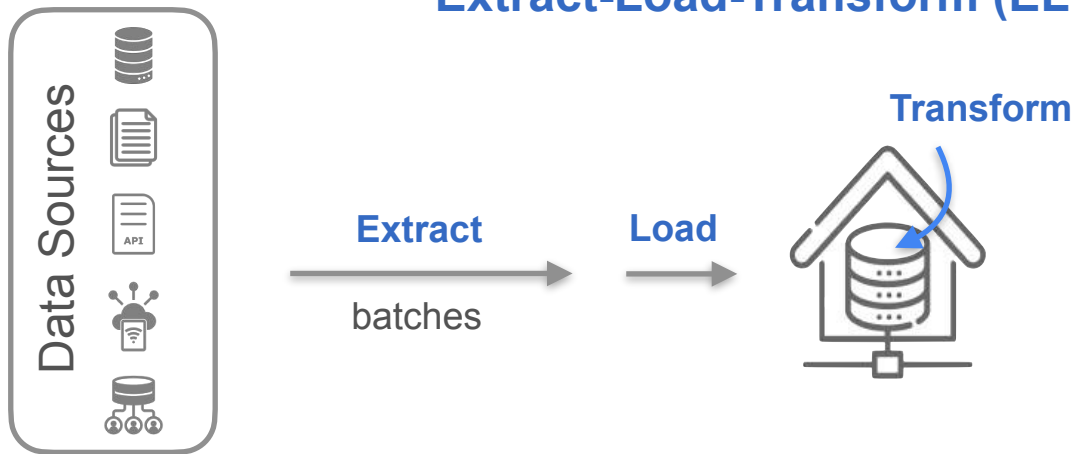
Example of Batch Architectures

Extract-Transform-Load (ETL)

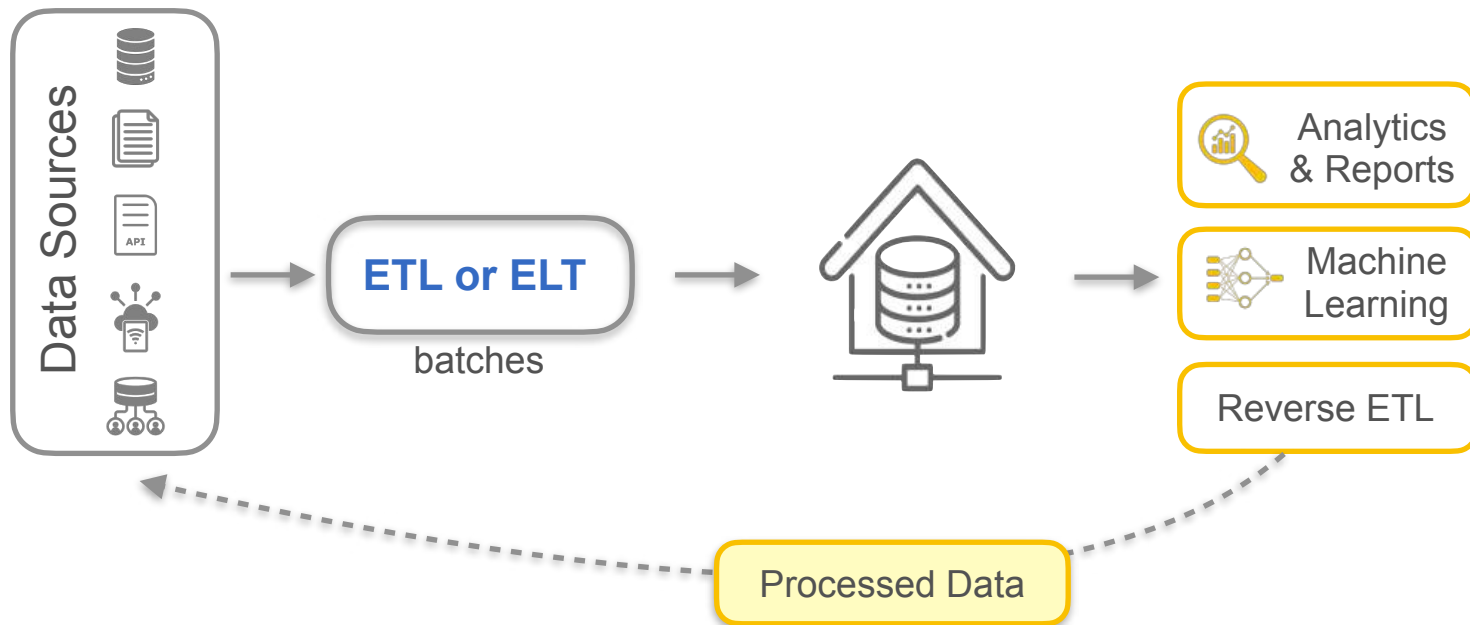


Example of Batch Architectures

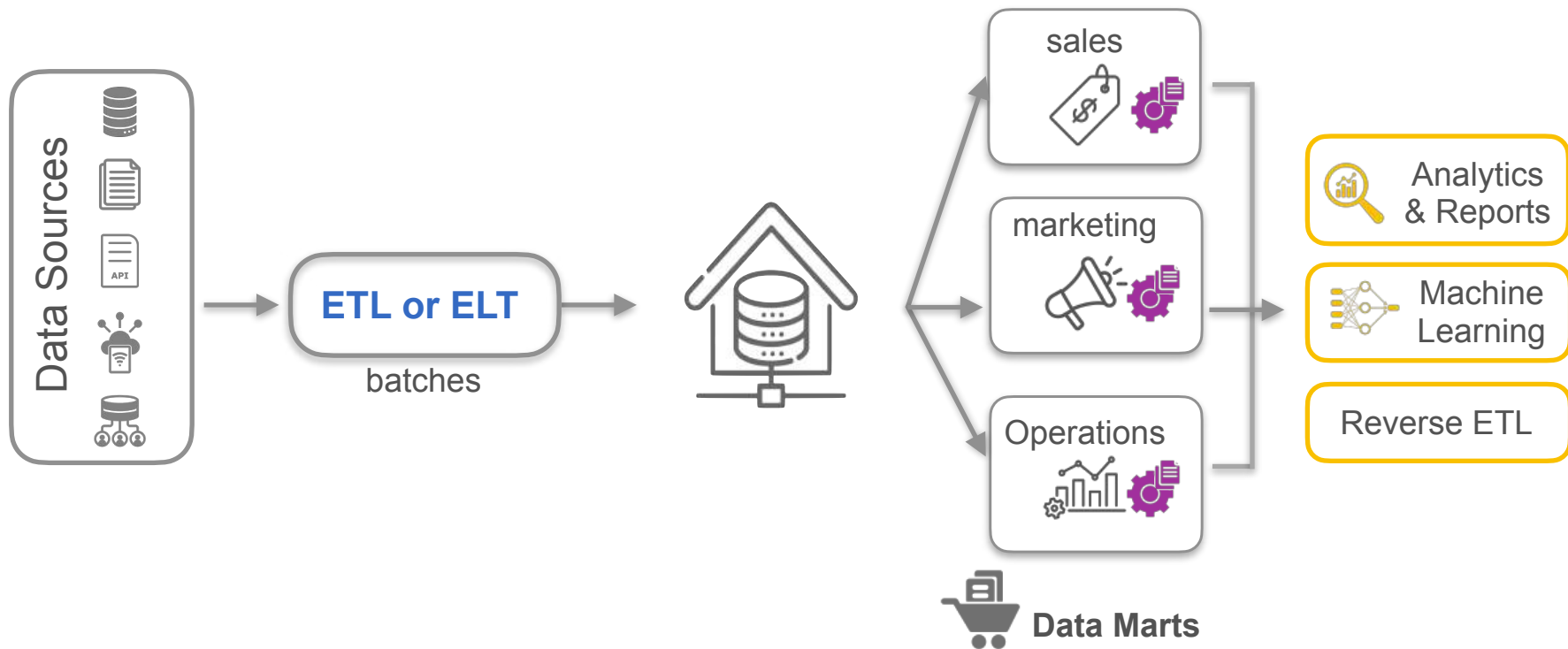
Extract-Load-Transform (ELT)



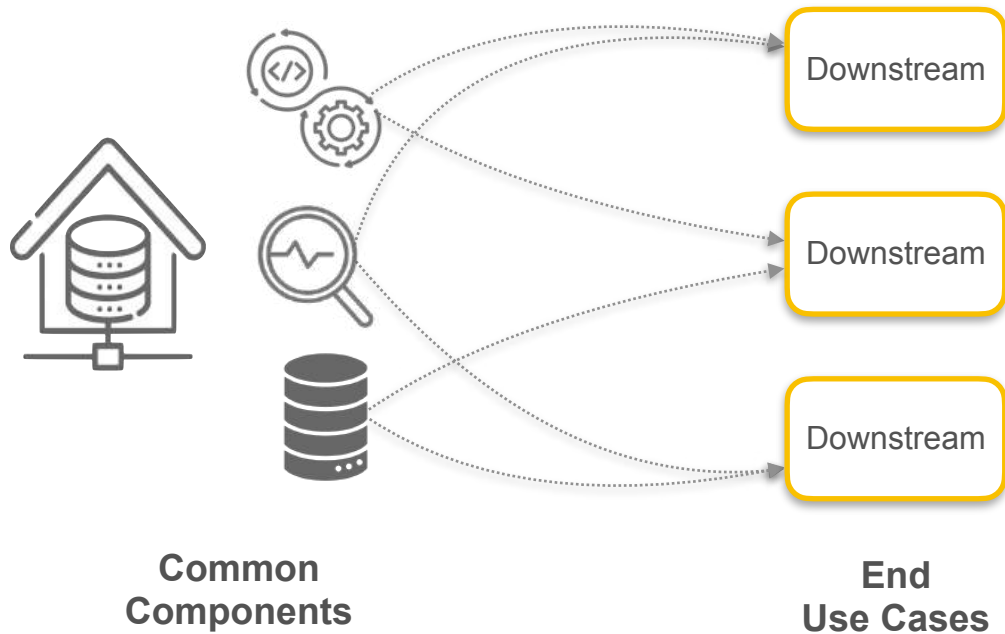
Example of Batch Architectures



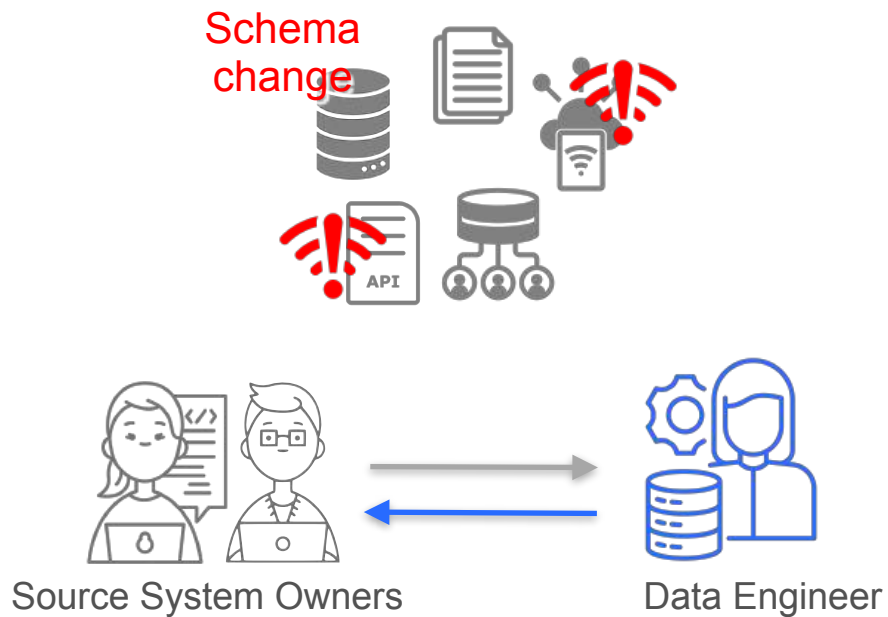
Example of Batch Architectures



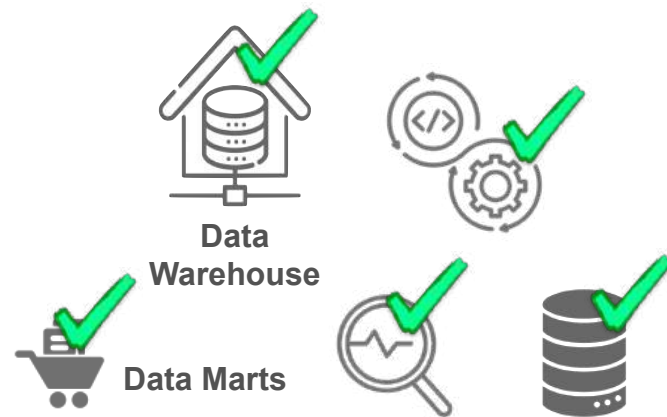
Choose Common Components



Planning for Failure

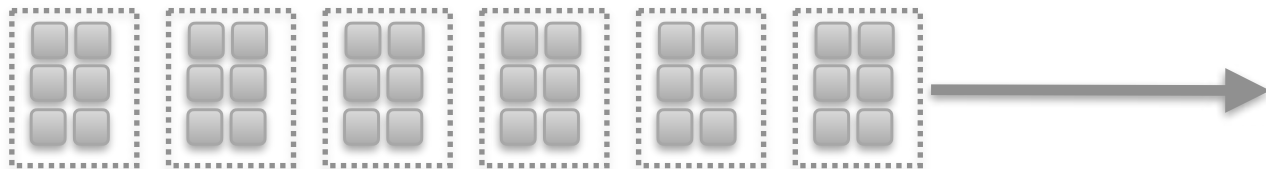


Availability and reliability specs



Make Reversible Decisions

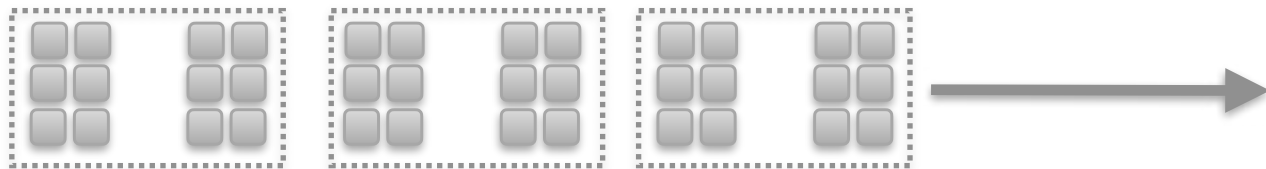
Build flexibility into your system



Ingest 1 day's
worth of data

Make Reversible Decisions

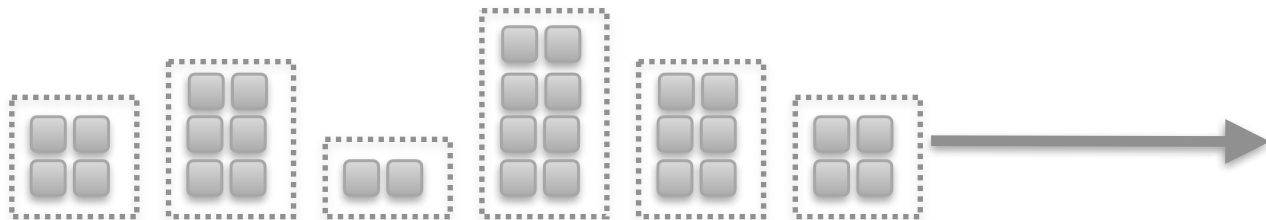
Build flexibility into your system



Ingest **2** day's
worth of data

Make Reversible Decisions

Build flexibility into your system



Ingest different
amounts of data

Embrace FinOps

Cost-Benefit Analysis

	Component 1	Component 2
Performance	High	Low
Cost	High	Low



Value for the
business



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Data Architecture

Streaming Architectures

Data - Stream of Events

**At its source,
data is a continuous
stream of events**



Event
Producer

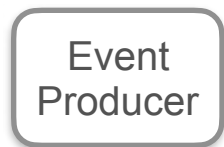
Data - Stream of Events

Batch Data Pipeline

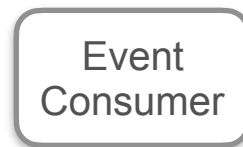


Data - Stream of Events

Streaming Data Pipeline

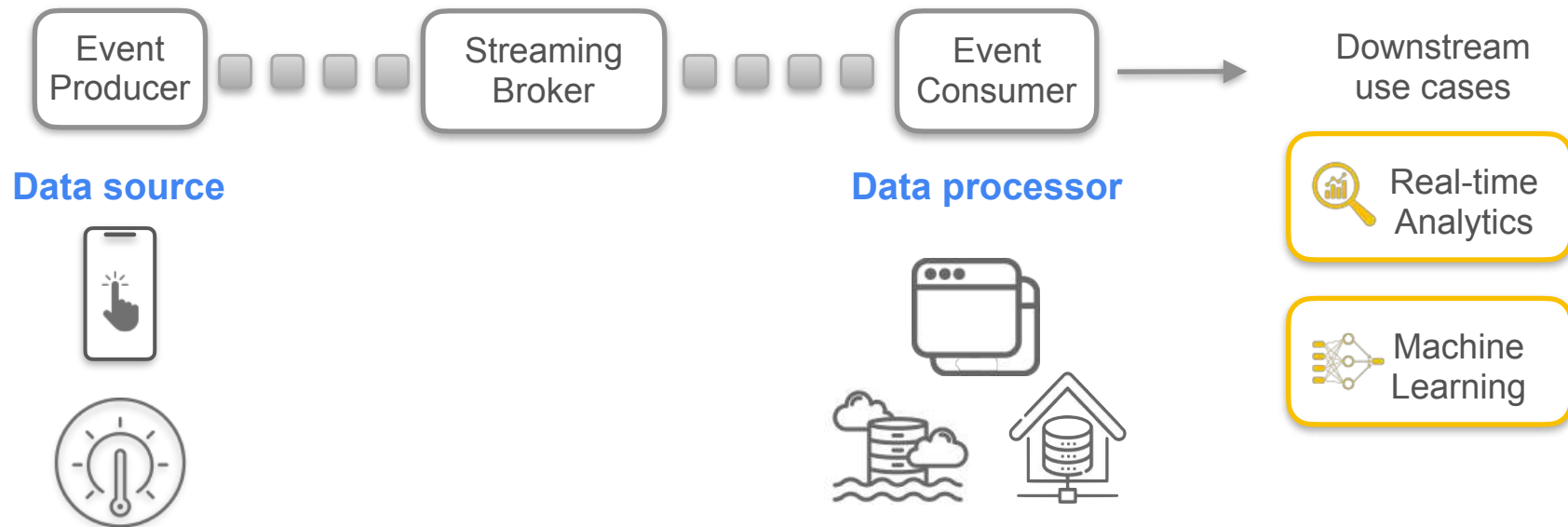


Continuous, near real-time
ingestion



Data available
shortly after it is
produced (<1s)

Streaming Systems



Streaming Frameworks



Event Streaming Platform

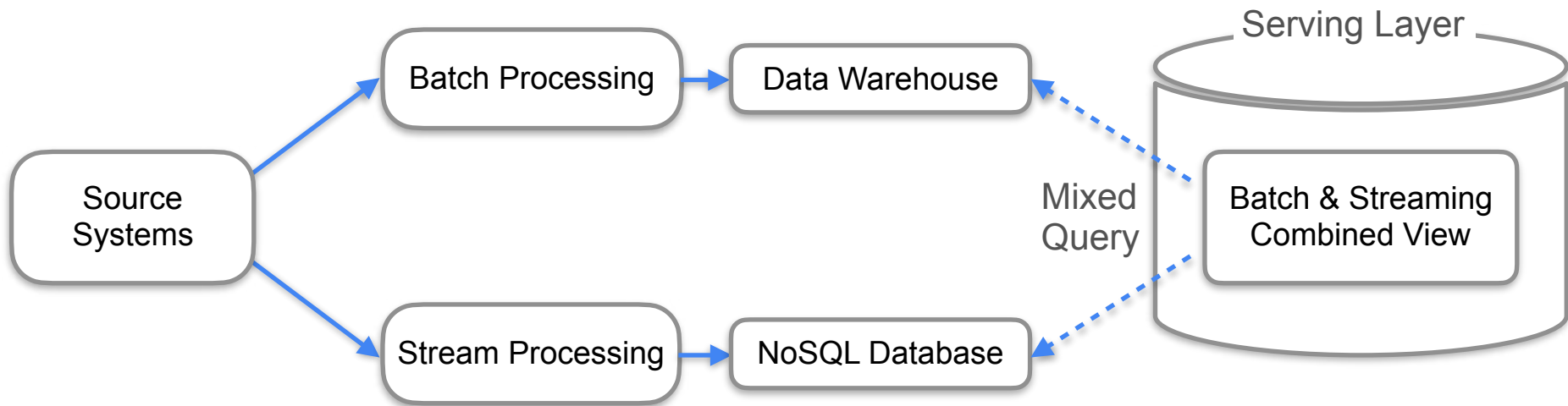


Streaming & Real-time Analytics



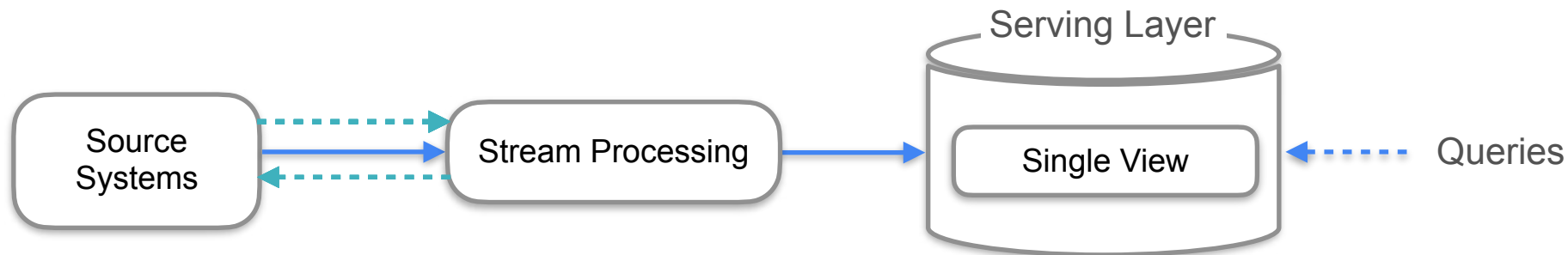
Lambda Architecture

Challenges: managing parallel systems with different code bases



Kappa Architecture

A true event-based architecture



- Retain some historical data
- Replay large chunks of retained data

Unified Batch & Streaming

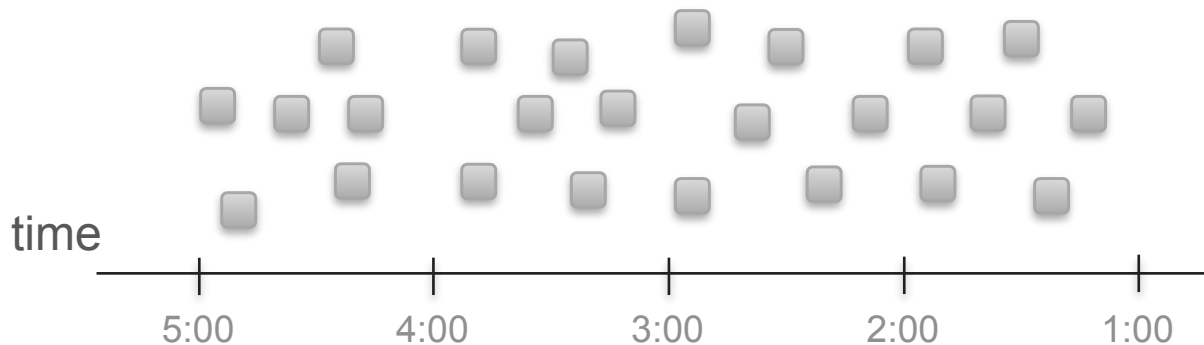
Unifying multiple code paths



Google Dataflow



Data viewed as events



Unified Batch & Streaming

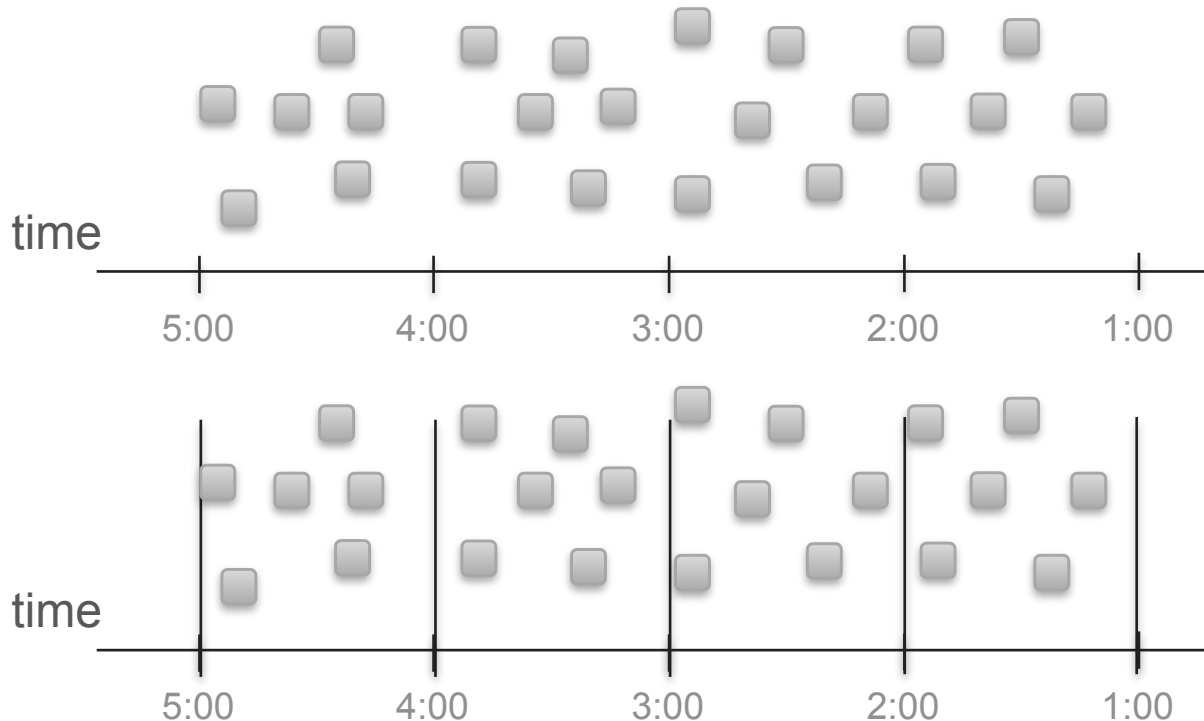
Unbounded data:
real-time event streams



**Same code to process
both types of data**



Bounded data:
data in batches



Unified Batch & Streaming



Flink



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Data Architecture

Architecting for Compliance

Architecting for Compliance

General Data Protection Regulation (GDPR)

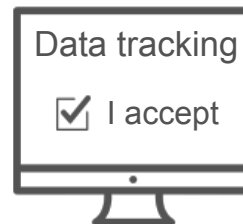
Enacted in European Union in 2018



Personal Data

- Personal Identifiable Information (PII)
- Other information collectively used to identify an individual

Right to have your data deleted



ID

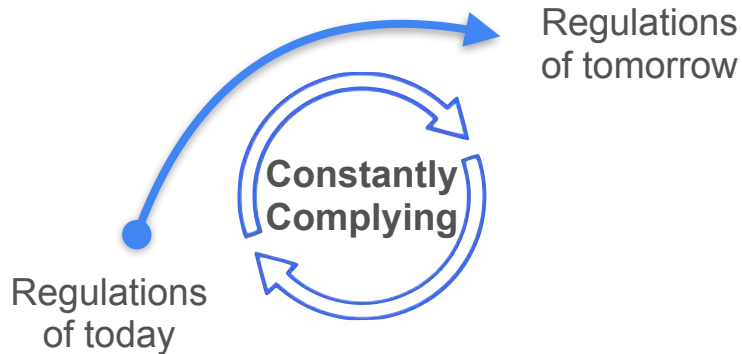
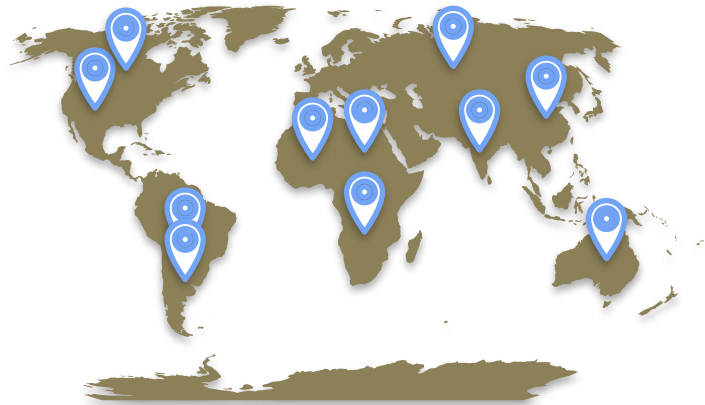
Last

First

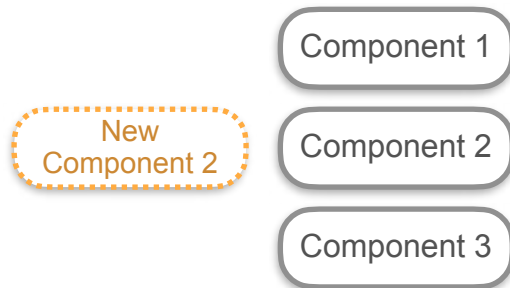
Card

Architecting for Compliance

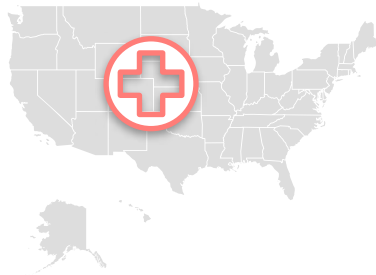
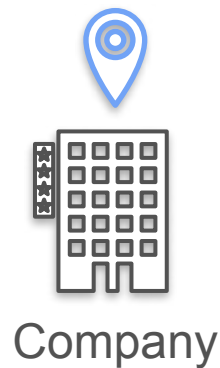
GDPR and similar regulations have been enacted globally



Loosely Coupled Components



Architecting for Compliance



Industry

**Health Insurance Portability
and Accountability Act
(HIPAA)**



Financial Data

Sarbanes Oxley Act in the U.S



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Choosing the Right Technologies

Choosing tools and technologies

Choosing Tools and Technologies

RDBMS



ORACLE

SAP ASE



NoSQL



Google
Cloud Bigtable



Azure
CosmosDB



MarkLogic



Aerospike



ArangoDB

Monitoring



amazon
Cloudwatch



New Relic



Datadog



APPDYNAMICS
part of Cisco



rubrik



solarwinds



WAVEFRONT
by vmware



cribl



Moogsoft



scalyr



unravel

pagerduty



SCALYR



ScienceLogic



MAGNITUDE



Grafana Labs

OpsRamp

Storage



amazon
S3



Google
Cloud Storage



PURE STORAGE



ALLUXIO



nimble storage



Qumulo



CLUMIO



money.com.io

Transformations



dbt



talend



alteryx



Fivetran



Stitch
A Talend Company



MATILLION



pentaho
A Hitachi Group Company



TRIFACTA



amazon
Glue



amazon
Athena



Azure Data
Factory



Paxata
a DataRobot company



StreamSets



UNIFI



tamr



Airbyte



Meltano



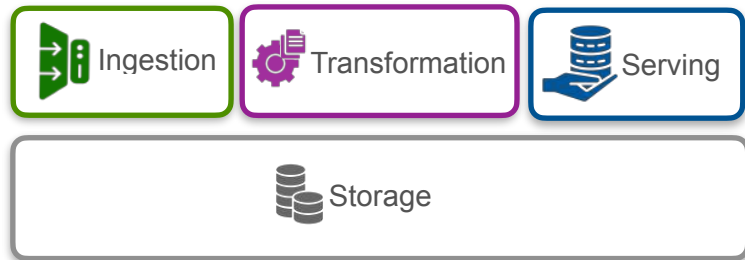
Narrator



Rivery

Choosing Tools and Technologies

Options of software solutions



**Open source
software**

**Managed open source
software**

**Proprietary
software**

Choosing Tools and Technologies

Keep in mind the end goal!

Deliver high-quality data products



What Why
When



Choosing Tools and Technologies - Considerations

Location



On-Premises



Cloud



Hybrid

Other Considerations



Cost
Optimization



Team's size &
capabilities



Build

vs



Buy



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Choosing the Right Technologies

Location

Location - On-Premises



On-Premises

Company **owns and maintains** the hardware and software for their data stack.

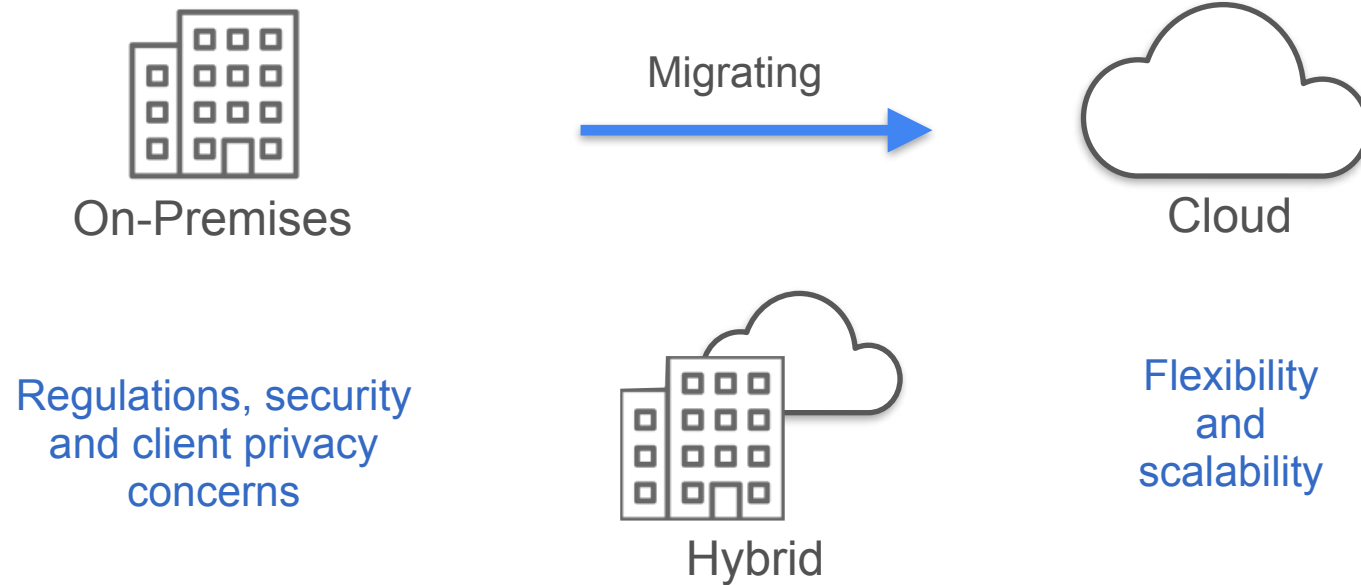
- Provisioning
- Maintaining
- Updating
- Scaling

Location - Cloud

- **Cloud provider** is responsible for **building and maintaining** the hardware in data centers
- You **rent** the compute and storage resources
- You don't need to maintain or provision any hardware



Location



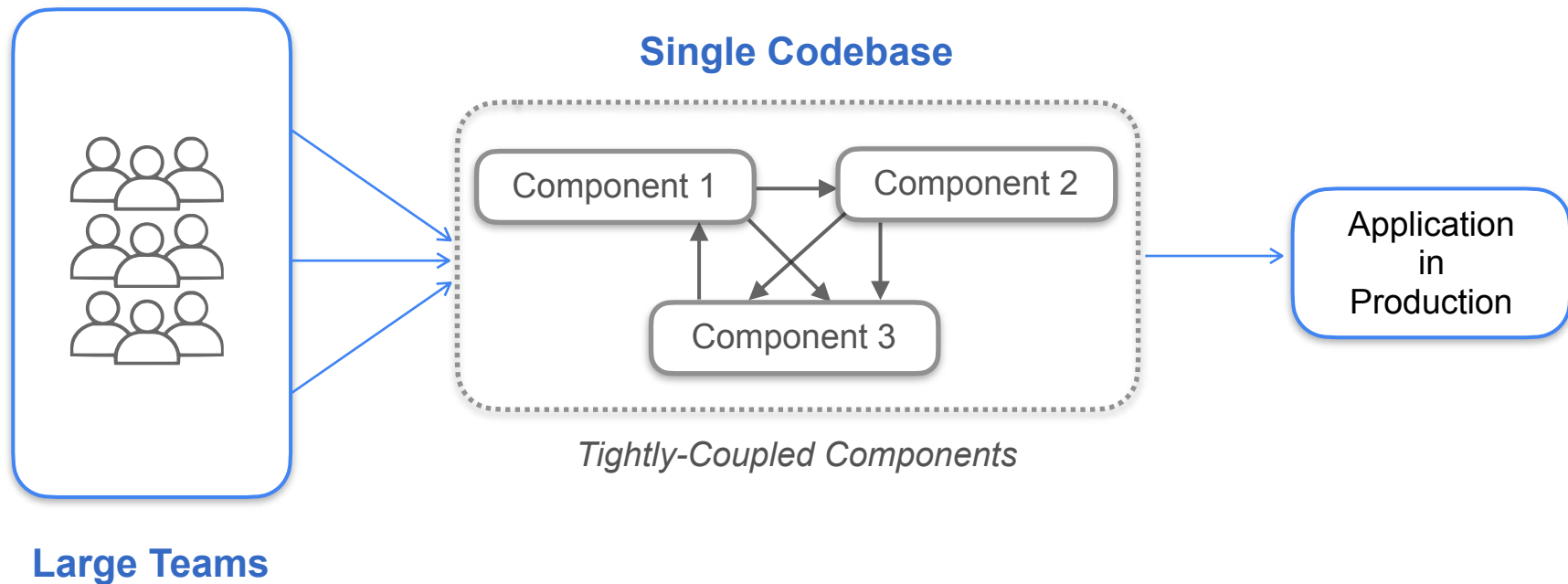


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Choosing the Right Technologies

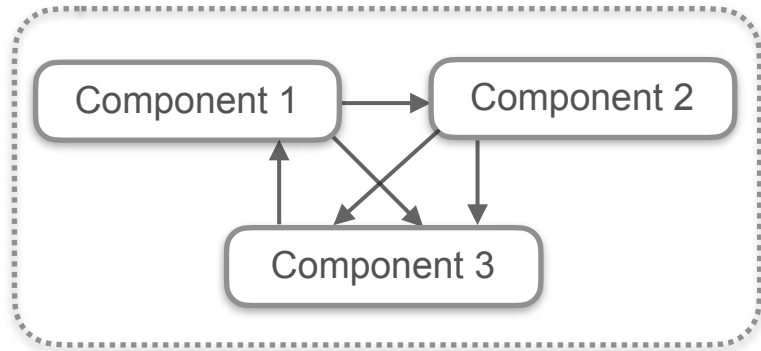
**Monolith versus
modular systems**

Monolithic Systems



Monolithic Systems

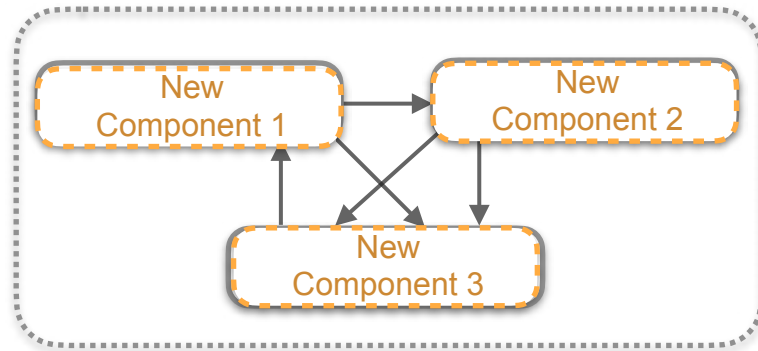
Single Codebase



Tightly-Coupled Components



Hard to maintain

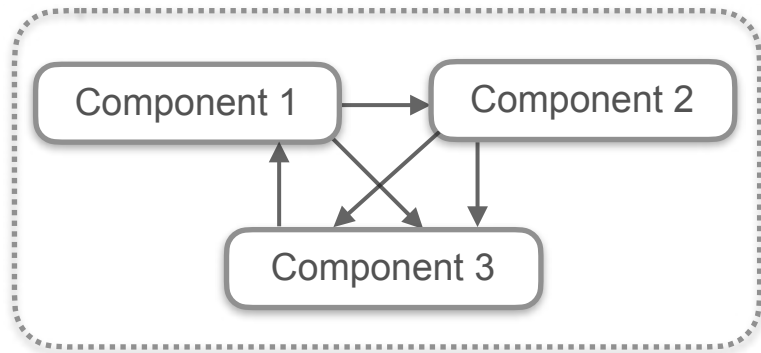


Entire application has to be re-written

- Easy to reason about and to understand
- Deal with one technology

Monolithic VS Modular Systems

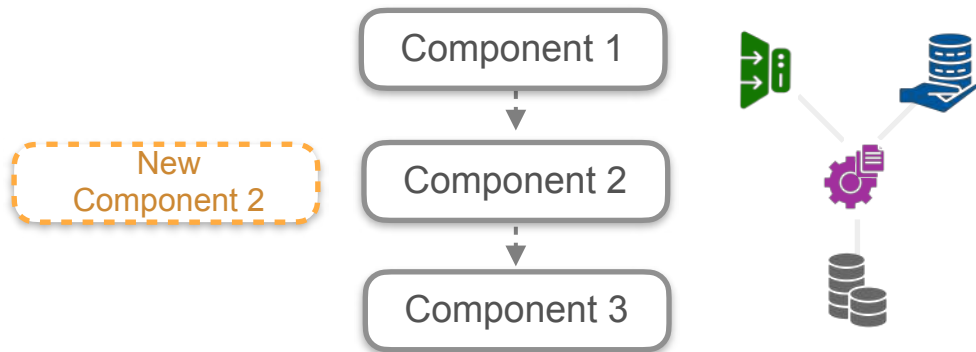
Monolithic



Tightly-Coupled Components

- Easy to reason about and to understand
- Deal with one technology

Modular



Loosely-Coupled Components

- Interoperability
- Flexible & reversible decisions
- Continuous improvement



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Choosing the Right Technologies

Cost Optimization and Business Value

Cost Optimization

**Total Cost of Ownership
(TCO)**

**Total Opportunity Cost of
Ownership (TOCO)**

FinOps

Cost Optimization

Total Cost of Ownership (TCO)

The total estimated cost of a solution, project or initiative over its entire lifecycle.



Hardware & Software



Maintenance



Training

Direct Costs

Easy to identify costs, directly attributed to the development of a data product.

- Salaries
- Cloud bills
- Software subscriptions

Indirect Costs (Overhead)

Expenses that are not directly attributed to the development of a data product.

- Network downtime
- IT support
- Loss of productivity

Cost Optimization

Total Cost of Ownership (TCO)

The total estimated cost of a solution, project or initiative over its entire lifecycle.



Hardware & Software

Capital Expenses (CapEx)

The payment made to purchase long-term fixed assets



Cost Optimization

Total Cost of Ownership (TCO)

The total estimated cost of a solution, project or initiative over its entire lifecycle.



Hardware & Software

Capital Expenses (CapEx)

The payment made to purchase long-term fixed assets



Operational Expenses (OpEx)

Expense associated with running the day-to-day operations.



Cost Optimization

Total Cost of Ownership (TCO)

The total estimated cost of a solution, project or initiative over its entire lifecycle.

On-premises

Capital Expenses (CapEx)

The payment made to purchase long-term fixed assets



Cloud

Operational Expenses (OpEx)

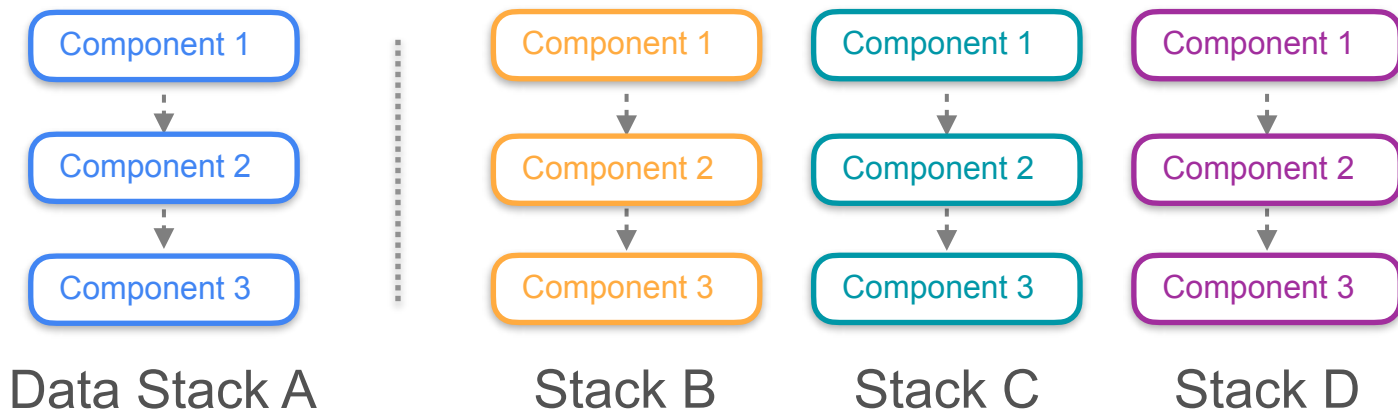
Expense associated with running the day-to-day operations.



Cost Optimization

Total Opportunity Cost of Ownership (TOCO)

The cost of lost opportunities that you incur in choosing a particular tool or technology.

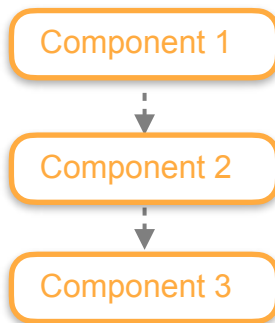
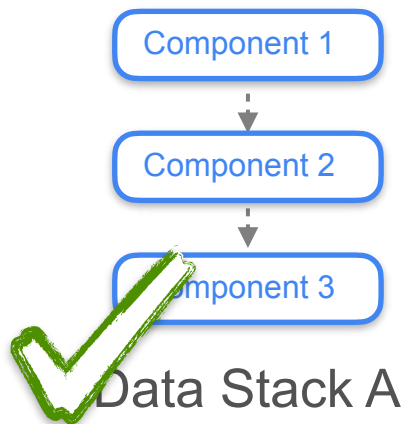


Cost Optimization

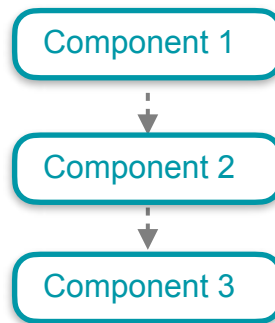
Total Opportunity Cost of Ownership (TOCO)

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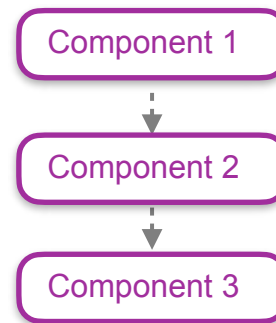
TOCO = 0



Stack B



Stack C



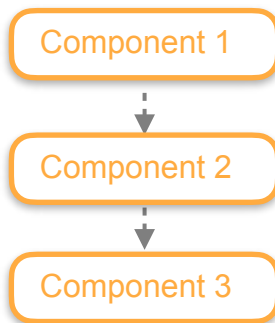
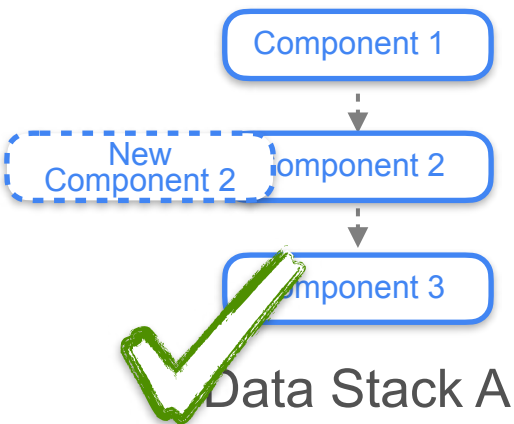
Stack D

Cost Optimization

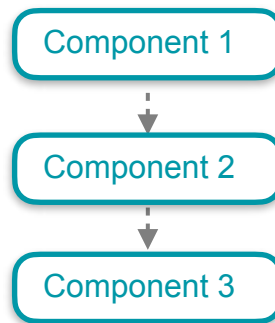
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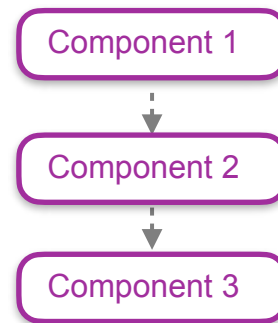
TOCO \neq 0



Stack B



Stack C



Stack D

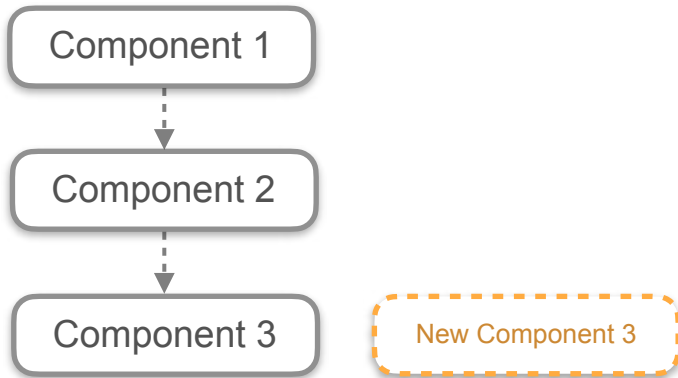
Cost Optimization

Total Opportunity Cost of Ownership (TOCO)

The cost of lost opportunities that you incur in choosing a particular tool or technology.

Minimize TOCO

Build flexible systems



Loosely-Coupled Components

Recognize components that are likely to change

- Immutable technologies
Object storage, Networking, SQL
- Transitory technologies
Stream processing, Orchestration, AI

Cost Optimization

FinOps



Minimize TCO and TOCO



Maximize revenue
generation opportunities



Cloud

OpEx-first

- Flexible, pay-as-you-go technologies
- Modular options



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Choosing the Right Technologies

Build vs Buy

Build Your Own Solution



Customize your own solution



Benefits:

- Get exactly the solution you need
- Avoid licensing fees
- Avoid being at the mercy of a vendor

Use Existing Solution



- Open-Source (*community*)
- Commercial Open-Source (*vendor*)
- Proprietary Non-Open-Source

Considerations



Your team

- Does your team have the bandwidth and capabilities to implement an open source solution?
- Are you a small team? Could using a managed or proprietary service free up your time?



Cost

- How much are licensing fees associated with managed or proprietary services?
- What is the total cost to build and maintain a system?



Business Value

- Do you get some advantage by building your own system compared to a managed service?
- Are you avoiding undifferentiated heavy lifting?



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Choosing the Right Technologies

**Server, Container, and
Serverless Compute Options**

Server

You set up and manage the server

- Update the OS
- Install / update packages
- Patch software
- Networking, scaling, and security

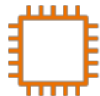
Virtual Server

Application

Dependencies

OS

Example:



EC2
instance

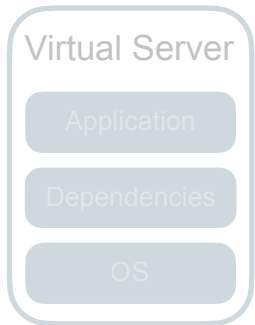
Container

Serverless

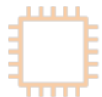
Server

You set up and manage the server

- Update the OS
- Install / update packages
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Example:

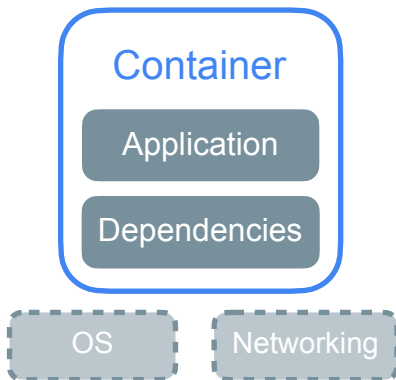


EC2
instance

Container

Modular unit that packages code and dependencies to run on a server

- Lightweight & portable
- You set up the application code and dependencies

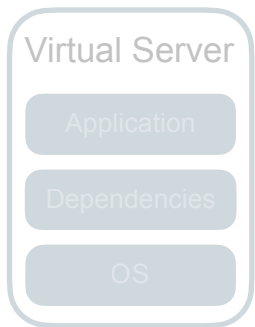


Serverless

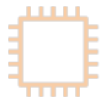
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You set up and manage the server

- Update the OS
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Example:

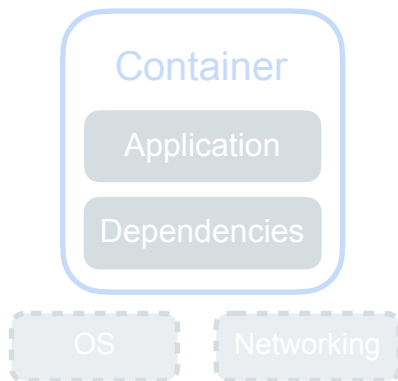


EC2
instance

Container

Modular unit that packages code and dependencies to run on a server

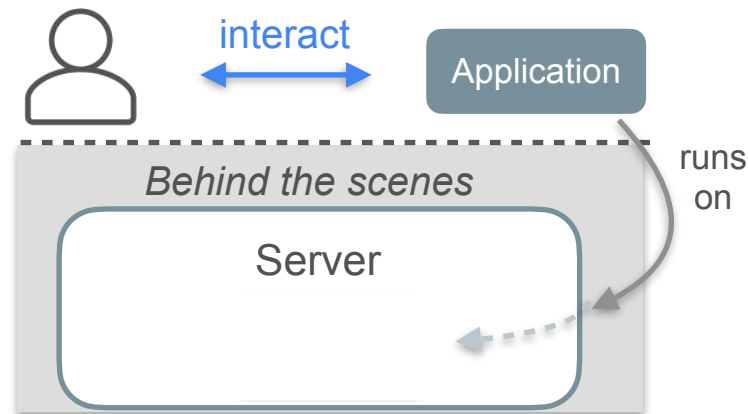
- Lightweight & portable
- You set up the application code and dependencies



Serverless

You don't need to set up or maintain the server

- Automatic scaling
- Availability & fault-tolerance
- Pay-as-you-go



Amazon Serverless Services

**Serverless
services you've
worked with**



Amazon Athena



AWS Glue

**Service that
popularized
serverless services**



AWS Lambda

Run code in response
to an event

**Advantages of
serverless services**

- Execute small chunks of code on as-needed basis
- Run services on as-needed basis
- Pay a little bit each time you run code or use a service

When To Use Serverless Services?

Cloud cost



Example:



AWS
Lambda

Expensive in a
high event rate
environment

Model & Monitor



- Event rates
- Event duration
- Cost per event

Limitations



- Limits on execution frequency, concurrency, and duration
- Use container if your application can't function within limits

Best for:

Simple & discrete tasks

Not good when:

Many parts require a lot of compute or memory power

Container orchestration



kubernetes

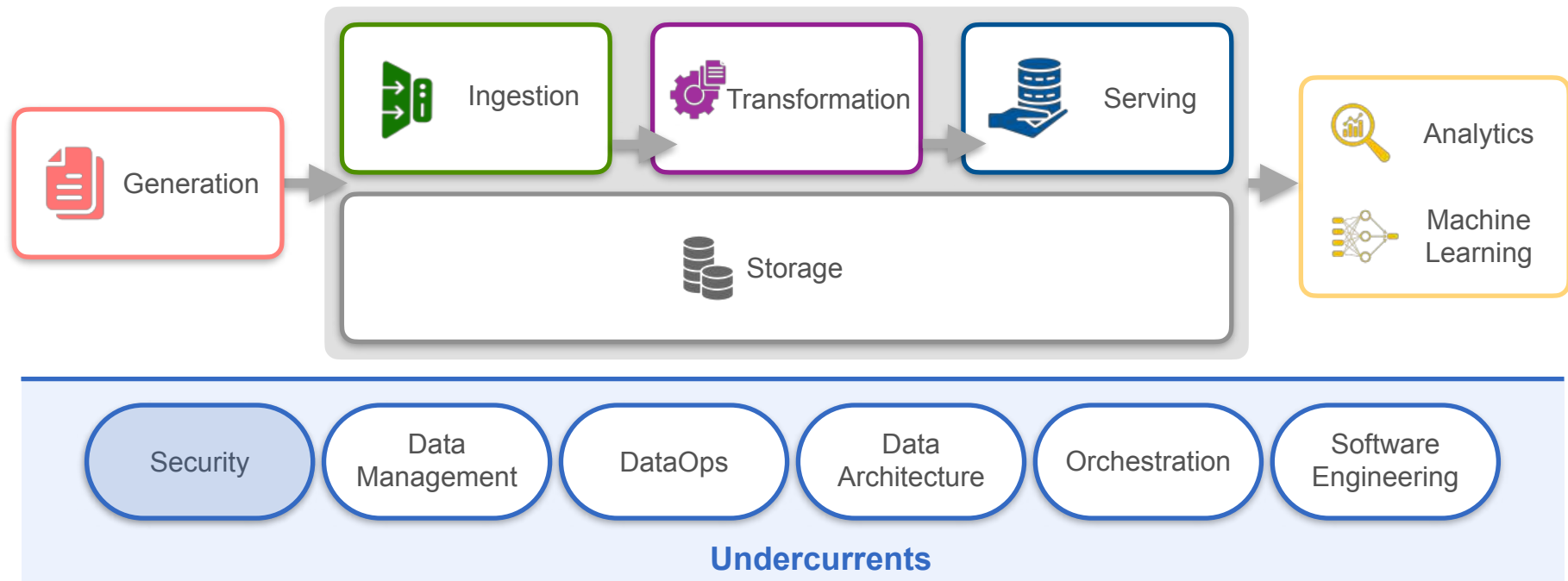


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Choosing the Right Technologies

How the Undercurrents Impact Your Decisions

The Undercurrents



Security

What are the security features of the tool?



Use tools from reputable sources.



Know where your tools come from!

Data Management

How are data governance practices implemented by the tool provider?



Data Breach



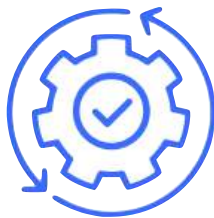
Compliance



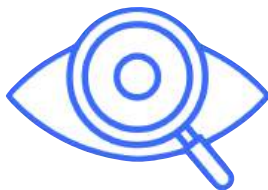
Data Quality

DataOps

What features does the tool offer in terms of automation and monitoring?



Automation



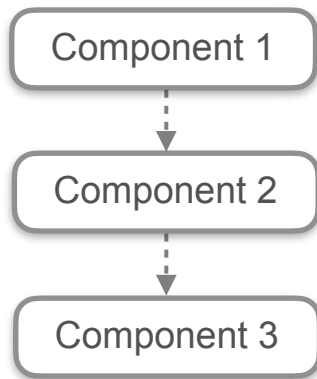
Monitoring



Service Level
Agreement (SLA)

Data Architecture

Does the tool provide modularity and interoperability?



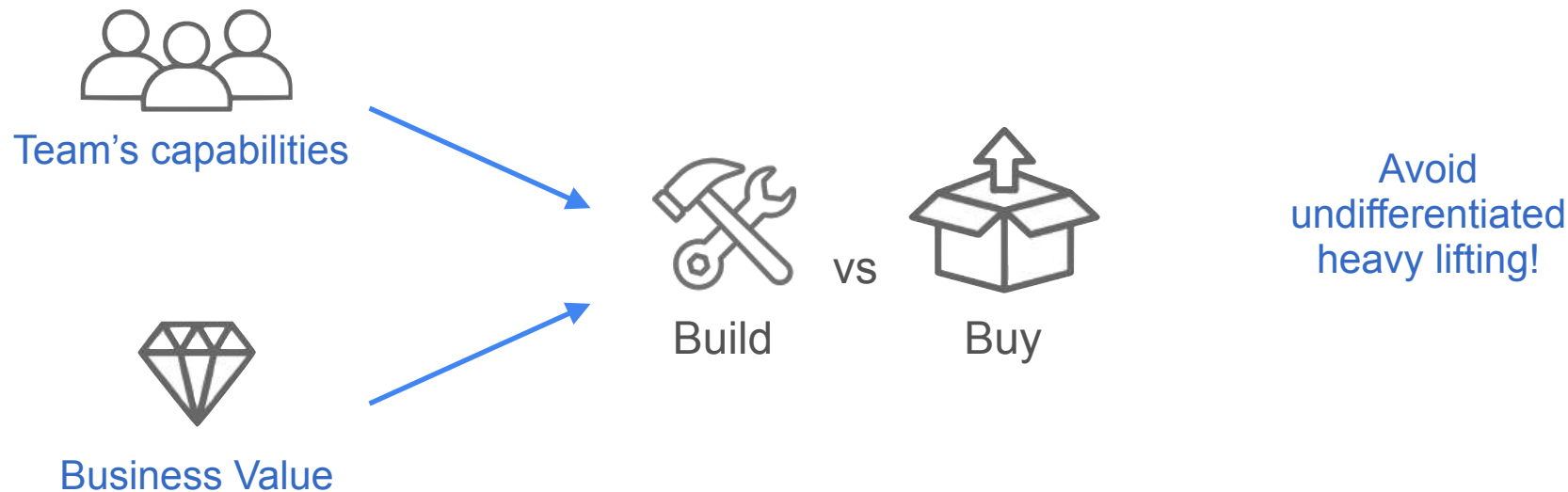
Loosely-Coupled Components

Orchestration



Software Engineering

How much do you want to do?





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Investigating your architecture (on AWS)

Intro to the AWS Well-Architected Framework

AWS Well-Architected Framework

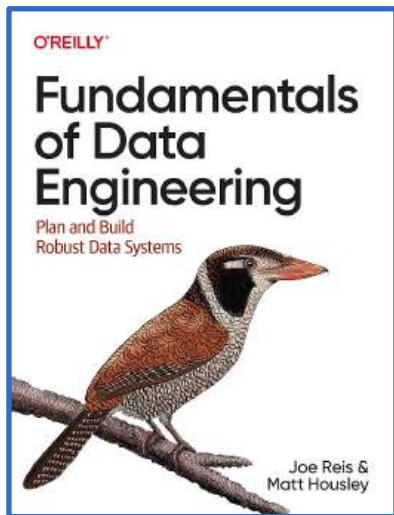


AWS Well-Architected

Set of principles and best practices that help you build:

scalable & robust architectures on AWS.

AWS Well-Architected Framework



Principles of good data architecture

Inspired by



AWS Well-Architected



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Investigating your architecture (on AWS)

The AWS Well-Architected Framework

AWS Well-Architected Framework



AWS Well-Architected

Operational Excellence

Performance Efficiency

Security

Cost Optimization

Reliability

Sustainability



AWS Well-Architected

Operational Excellence

- How you can develop and run your workloads on AWS more effectively
- Monitor your systems to gain insight into your operations
- Continuously improve your processes and procedures to deliver business value

Security

- How to take advantage of cloud technologies to protect your data, systems, and assets

Reliability

- Everything from designing for reliability to planning for failure and adapting to change

Performance Efficiency

- Taking a data-driven approach to building high-performance architecture
- Evaluating the ability of computing resources to efficiently meet system requirements
- How you can maintain that efficiency as demand changes and technologies evolve

Cost Optimization

- Building systems to deliver maximum business value at the lowest possible price point
- Use AWS Cost Explorer and Cost Optimization Hub to make comparisons and get recommendations about how to optimize cost for your systems

Sustainability

- Consider the environmental impact of the workloads you're running on the cloud
- Reducing energy consumption and increasing efficiency across all components of your system



AWS Well-Architected

Operational Excellence

Security

Reliability

Performance Efficiency

Cost Optimization

Sustainability

- Set of principles and questions
- Helps you design and operate reliable, secure, efficient, cost-effective, and sustainable systems in the cloud
- Helps you think through the pros and cons of different architecture choices

Data Analytics Lens

AWS Well-Architected Framework



Introduction

Workload context checklist

Design principles by pillar

▼ Pillars of the Well-Architected Framework

▶ Operational excellence

▶ Security

▶ Reliability

▶ Performance efficiency

▶ Cost optimization

▶ Sustainability

▶ Scenarios

Conclusion

Contributors

Document revisions

Notices

AWS Glossary

Design principles arranged by pillar

[PDF](#) | [RSS](#)

These are the design principles outlined in this paper organized by pillar of the AWS Well-Architected Framework.

Operational excellence

- 1 – Monitor the health of the analytics application workload
- 2 – Modernize deployment of the analytics jobs and applications

Security

- 3 – Designing data platforms for governance and compliance
- 4 – Implement data access control
- 5 – Control the access to workload infrastructure

Reliability

- 6 – Design resilience for analytics workload



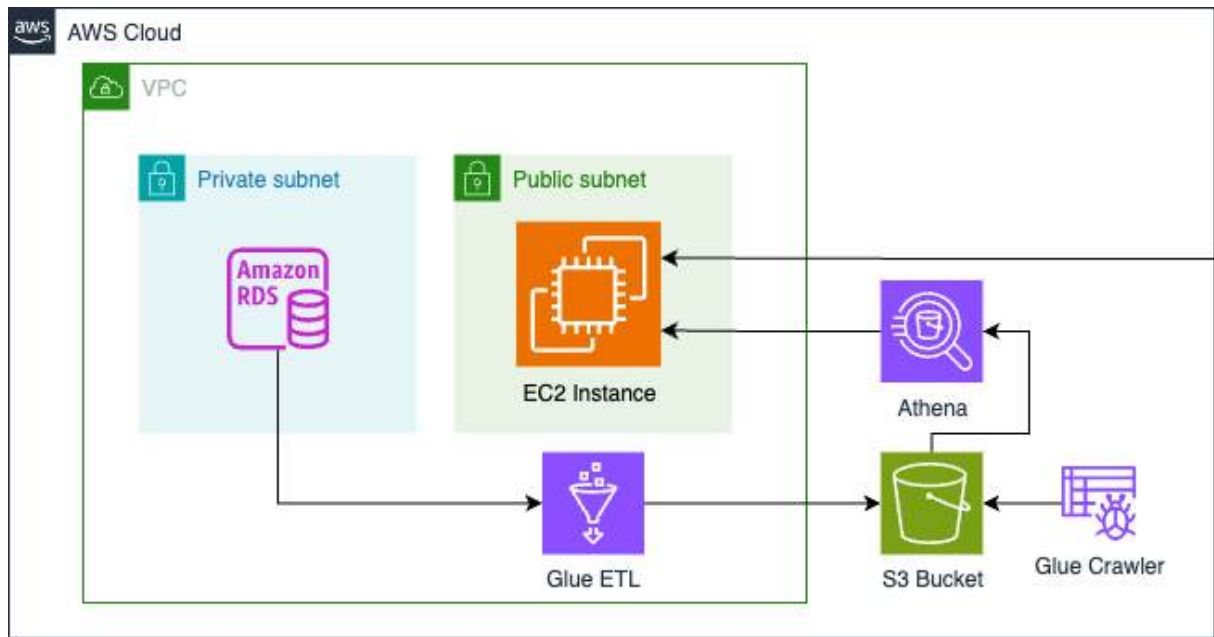


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Lab Walkthrough

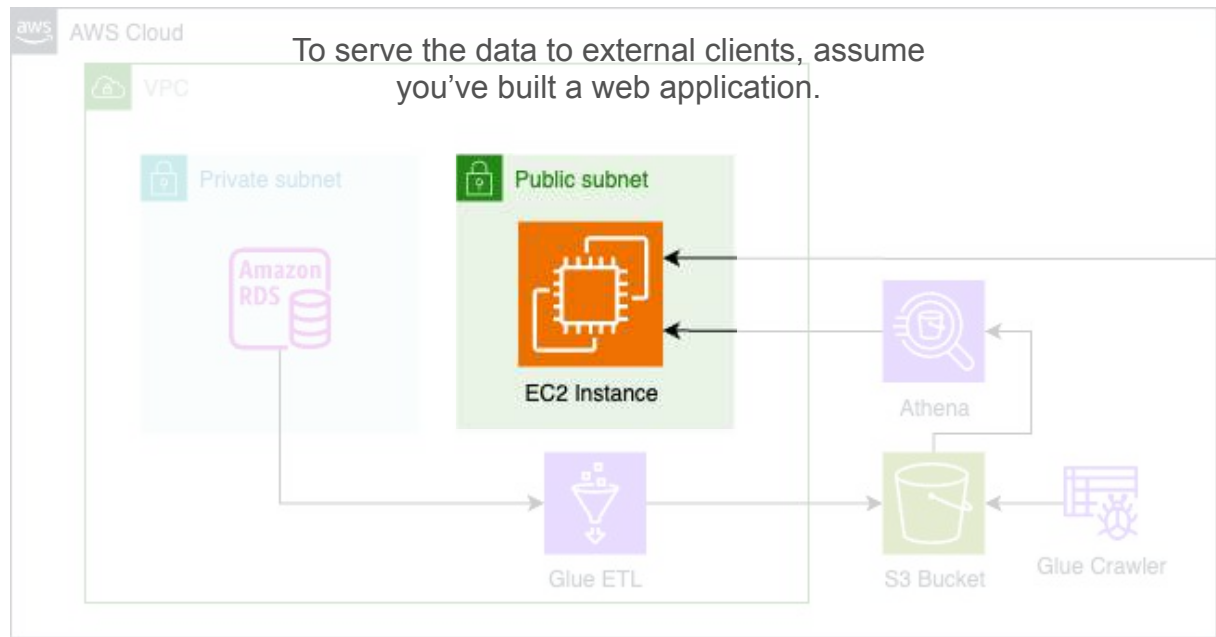
Introduction to the Lab

Week 3 Lab



- Serve embedded dashboards to clients
- Share analytics data through third-party data sharing platforms

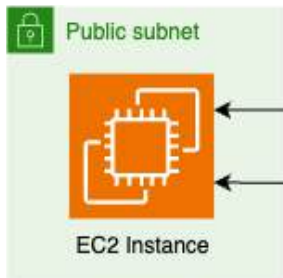
Week 3 Lab



- Serve embedded dashboards to clients
- Share analytics data through third-party data sharing platforms

Week 3 Lab

To serve the data to external clients, assume you've built a web application.



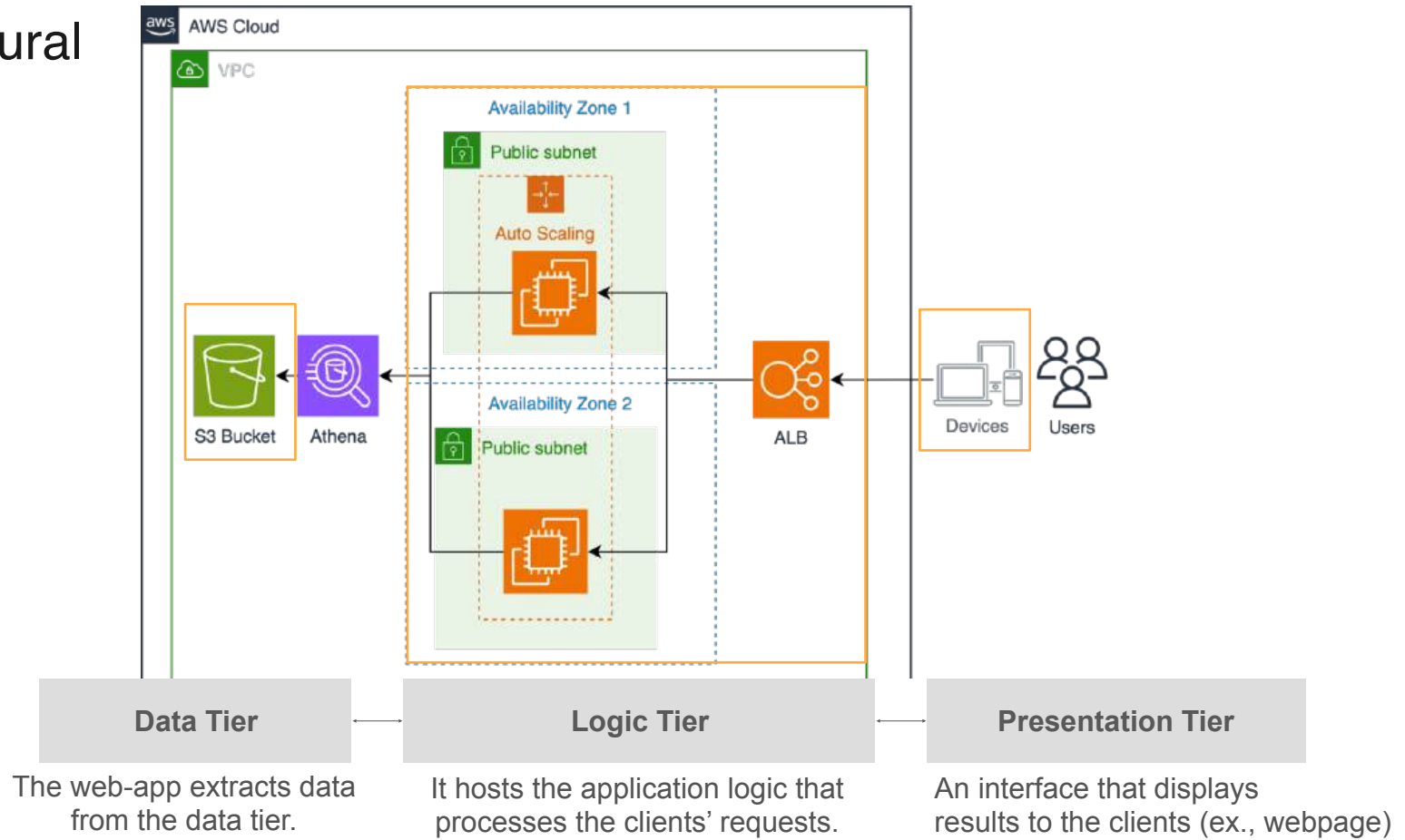
In this lab, ensure that the web app:

- Capable of scaling to meet the clients' needs
- Uses computing resources efficiently
- Designed in a secure and reliable way

Principles of good data architecture.

AWS well-architected framework.

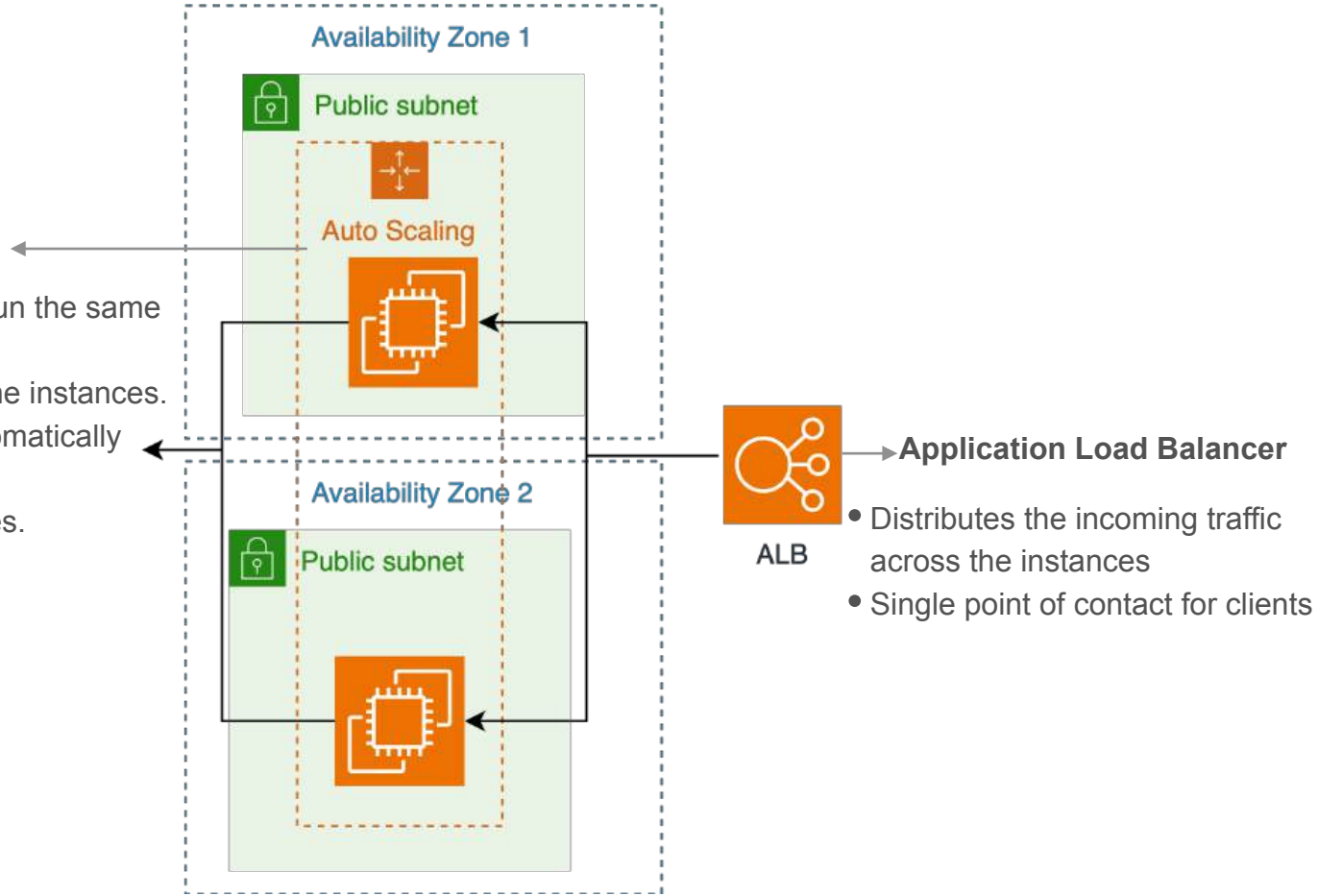
Architectural Diagram



Logic Tier

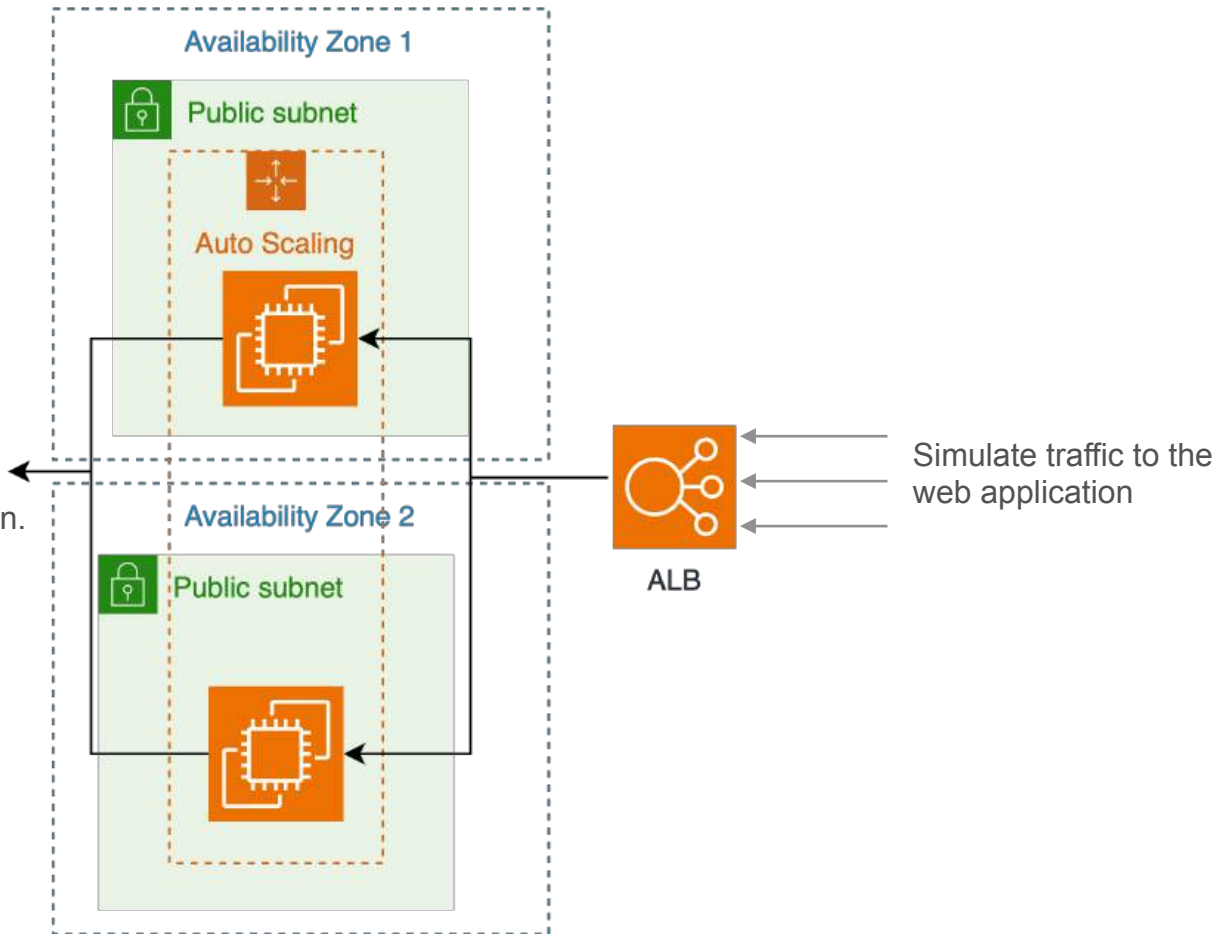
EC2 Auto Scaling Group

- Collection of EC2 instances that run the same logic.
- Requests are distributed across the instances.
- The number of instances can automatically increase or decrease.
- Configured to start with 2 instances.



Week 3 Lab

- Assess the scalability of the app.
- Monitor computing resources and network activity.
- Configure EC2 instances to enable performance efficiency and cost optimization.
- Adjust the security options of the load balancer.





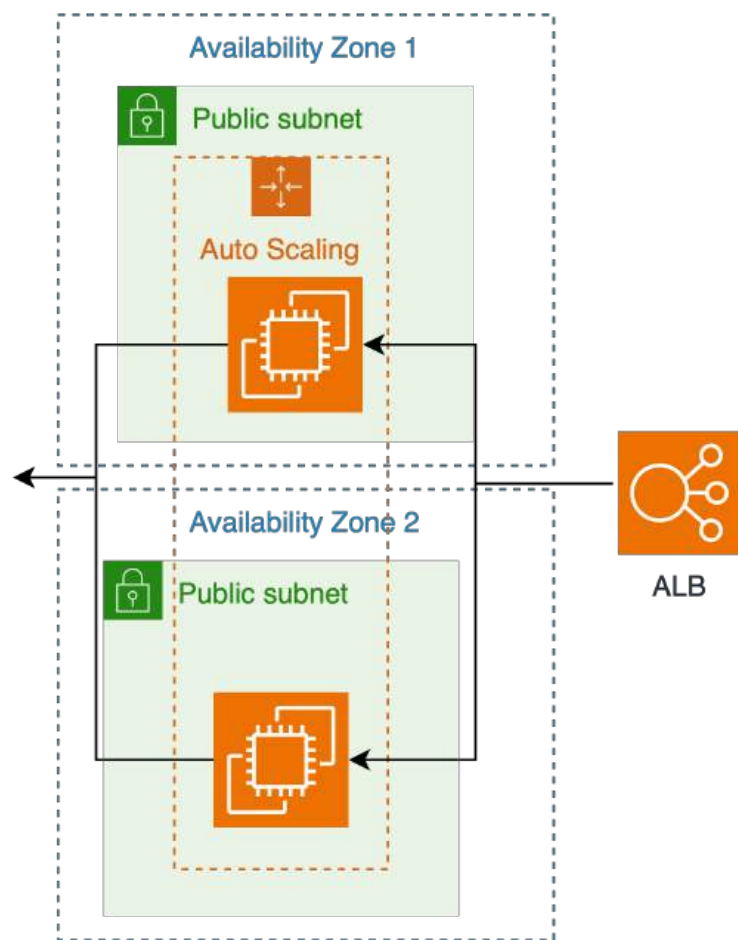
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Lab Walkthrough

Monitoring the Web App

Monitoring the Web App

3. Getting the address of the web application
4. Monitoring CPU usage and networking activity

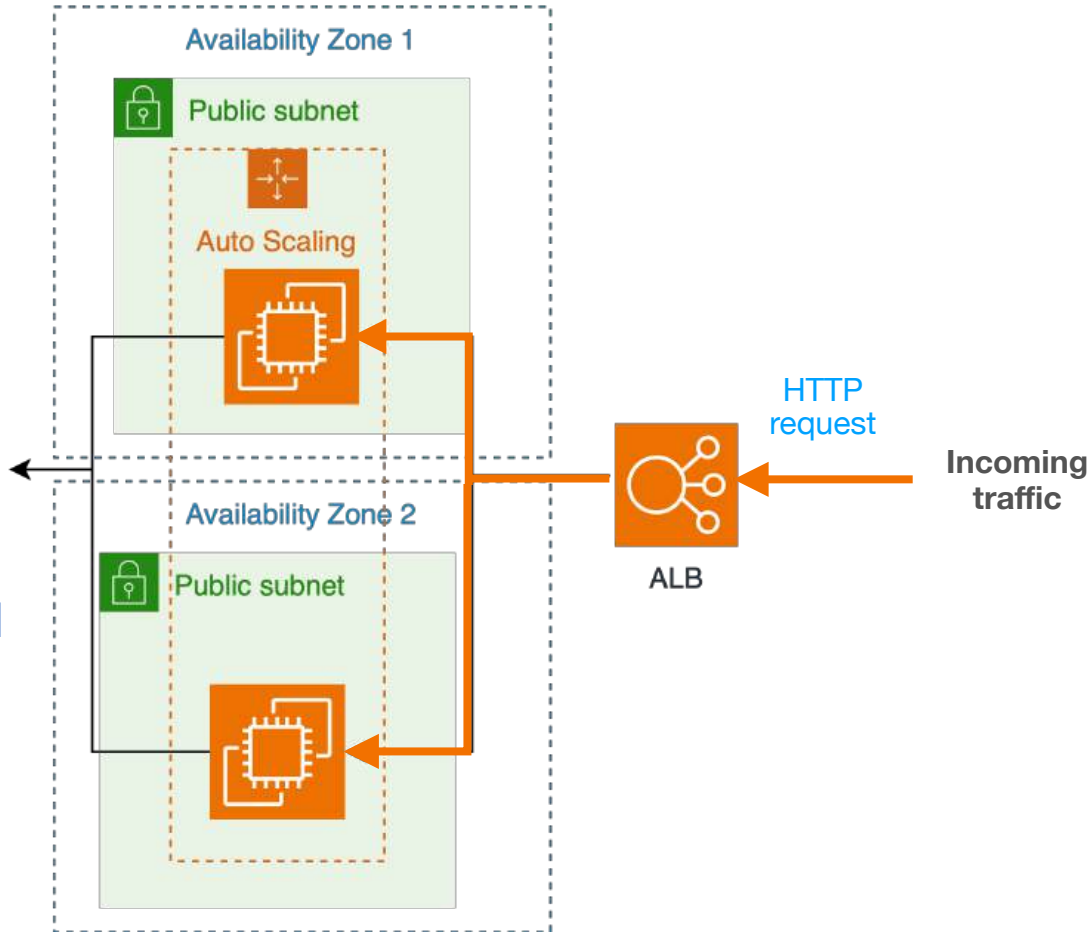


Monitoring the Web App

Monitor the incoming traffic and usage of computing resources.

**DataOps
Observability**

**Operational
Excellence**

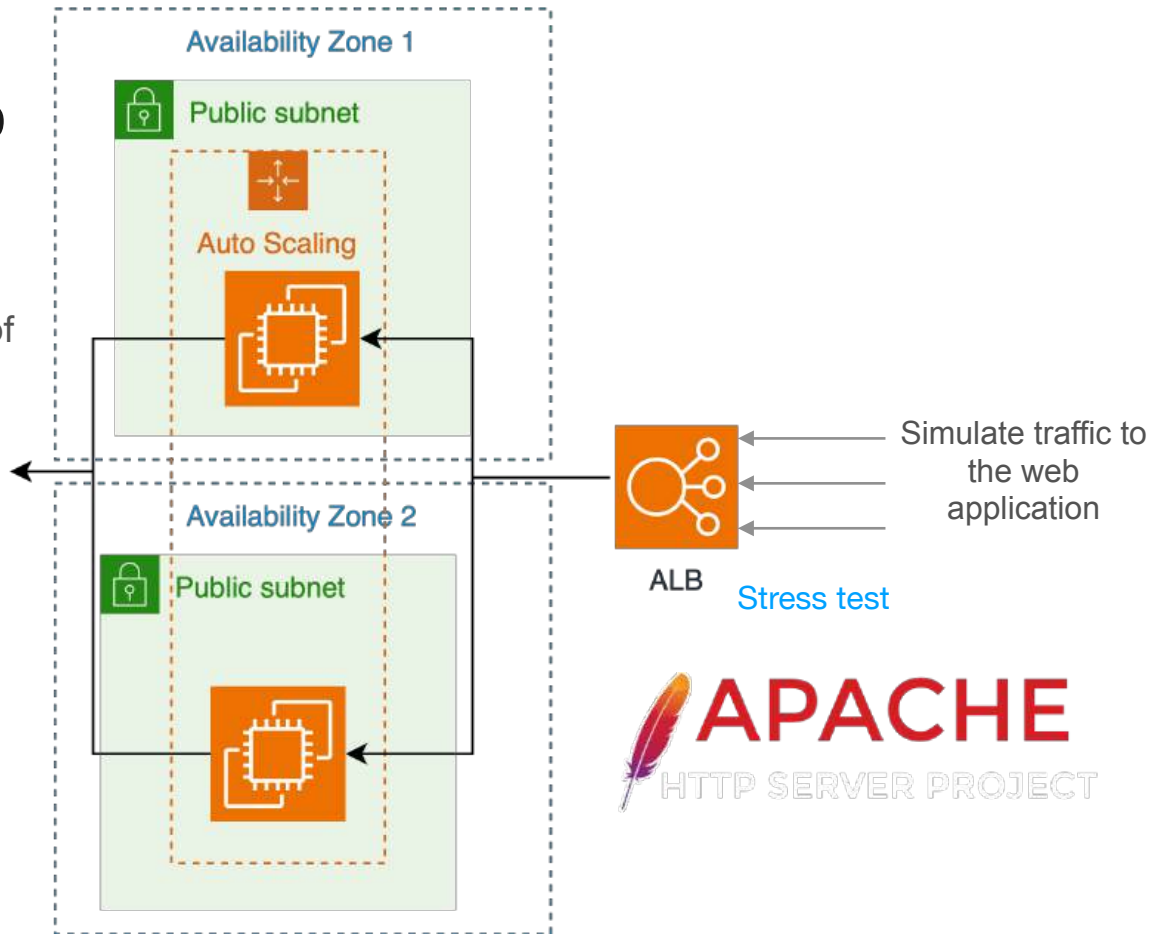


Monitoring the Web App

Monitor the incoming traffic and usage of computing resources.



Amazon CloudWatch





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Lab Walkthrough

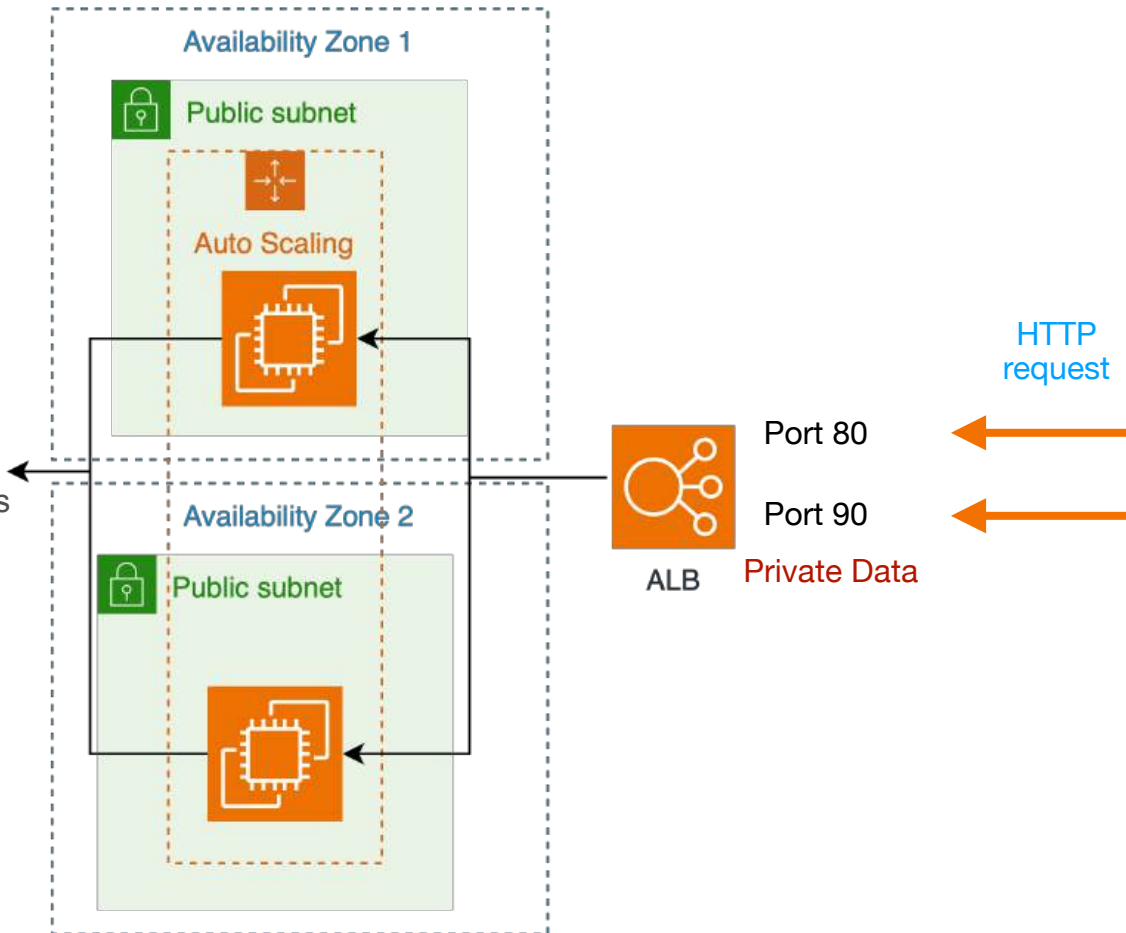
Applying the Principles of Good Data Architecture

Security

Prioritize Security

Security Pillar

- Configure the ALB to only receive certain types of requests.
- A request needs the address and the port number.
- A port number is a virtual identifier that applications use to differentiate between types of traffic

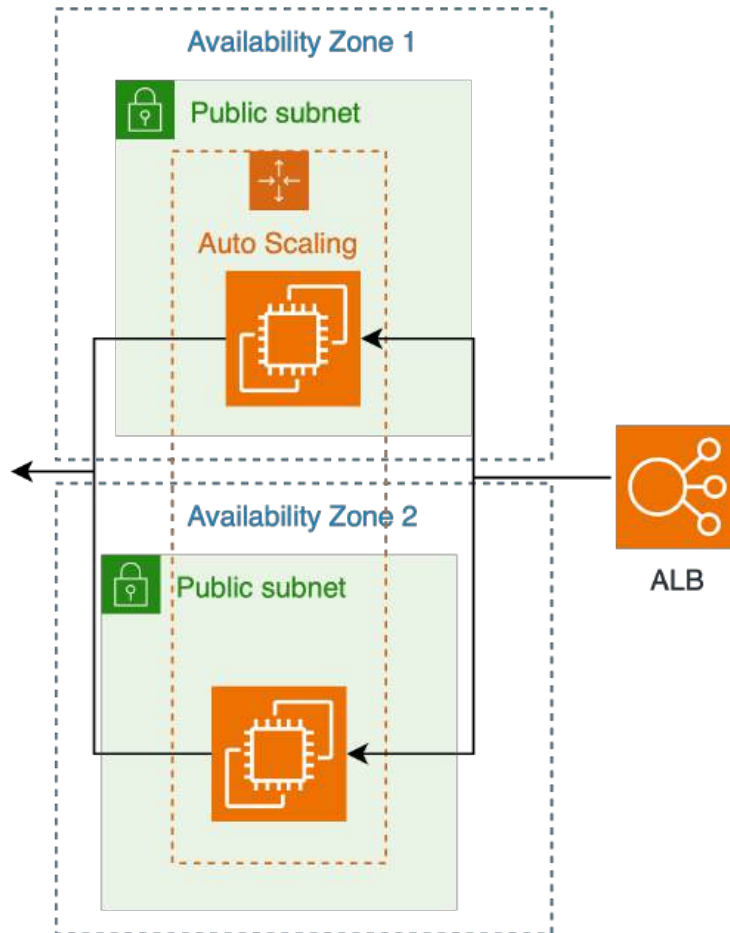


Reliability

Plan for failure

If something goes wrong with one of the availability zones, the requests can still be processed by the EC2 of the other zone.

Reliability Pillar

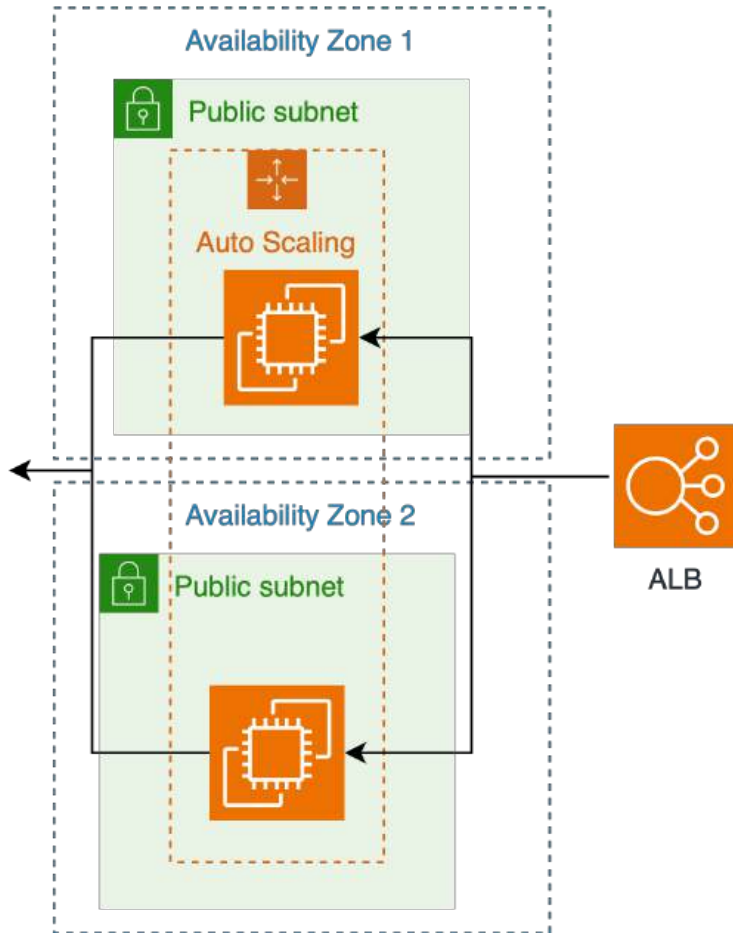


Cost Optimization

Embrace FinOps

Cost Optimization Pillar

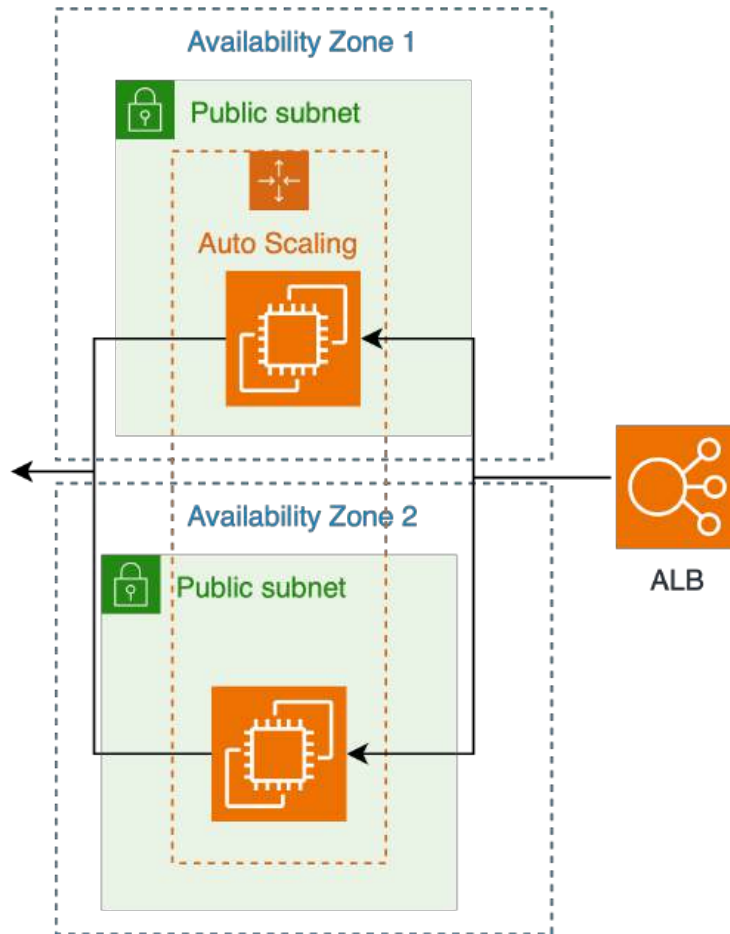
- Use the right resources efficiently.
- Currently, t3.micro instances are used.
- Switch to t3.nano.
- Modify the Launch template of the auto scaling group.



Scalability

Architecting for scalability

- With auto scaling groups, you only pay for what you use.
- Extra EC2 resources will be automatically removed when demand decreases.





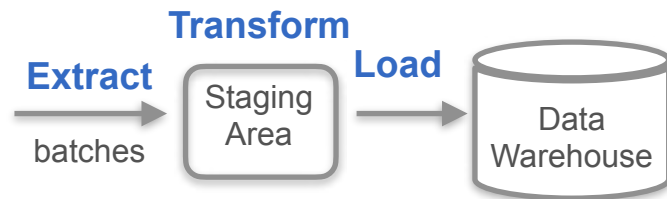
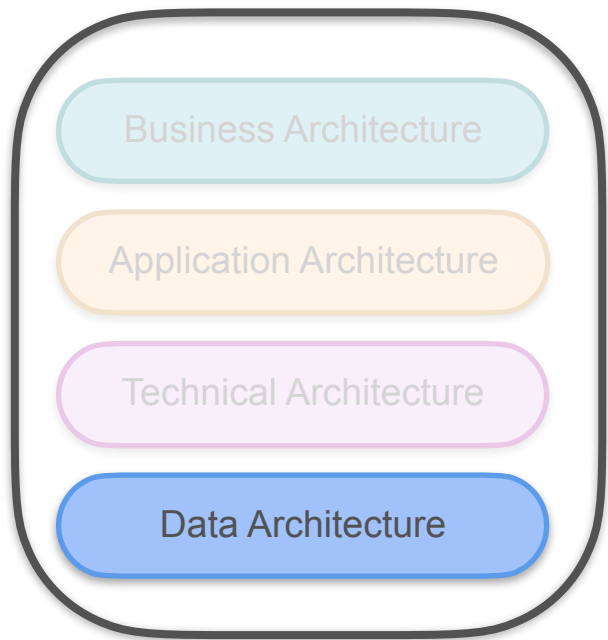
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Data Architecture

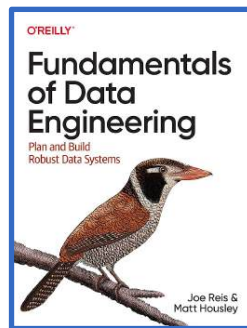
Week 3 Summary

Week 3 Summary

Enterprise Architecture



Principles of
good data
architecture



Hands-on Lab



AWS Well-Architected