

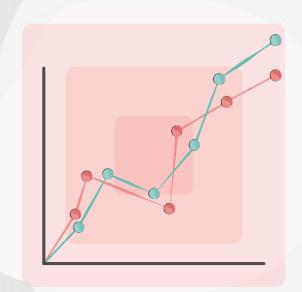
## **Data engineering**

Module: Data pipelines

**Topic:** From data pipelines to data

architectures

Welcome to today's webinar.



## Ice breaker

Scale it or fail it!

"Imagine you're hosting a dinner party for 4 people. Suddenly, 40 guests show up. What's your first move?"

#### Your options:

- A. Order 10 pizzas and hope for the best
- B. Start cooking in batches and delegate tasks
- C. Set up a buffet line and let people serve themselves
- D. Panic and pretend it was a prank





Submit your responses to the

chat or turn on your microphone

## Real-world case study

Zalando's scalable data platform on AWS

#### Challenge

- Billions of daily events from web, mobile, and logistics
- Real-time personalization and inventory updates
- Elastic scaling for peak events like Black Friday

#### Solution

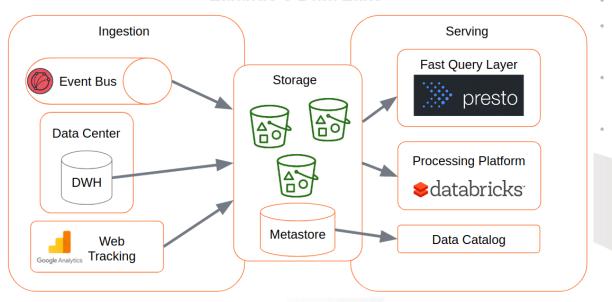
- Kafka + Kinesis for real-time streaming
- Serverless compute with AWS Lambda
- Modular microservices and Airflow orchestration
- Scalable storage with S3 and Redshift



Zalando delivers real-time recommendations, accurate inventory, rapid feature rollouts, and seamless performance during traffic spikes.



#### Zalando's Data Lake



Zalando's Real-Time Retail Engine: Scaling with AWS, Image source: <u>AWS</u>

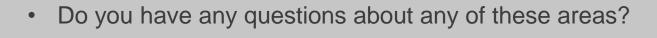


## e-learning recap

Reflecting on your learning...

The e-learning for this topic, covered the following areas:

- 1. Scalability means designing systems that grow with data and demand
- 2. Pipelines scale through horizontal scaling, stream processing, and decoupling
- 3. Performance and reliability require monitoring, fault tolerance, and testing
- 4. Scalable architectures are modular, loosely coupled, and event-driven
- 5. Tools like Spark, Kafka, and Airflow enable distributed, resilient workflows
- 6. Smart trade-offs balance speed, cost, and complexity



Did everything in the e-learning make sense?











# **Webinar Agenda**

#### Today, we will cover the following:

- 1. Core concept recap
- 2. Practical lab
- 3. Summary
- 4. Q&A







## Session aim and objectives

By the end of this session, you should be able to:

- Understand what scalability means in data pipelines and architectures
- Explore strategies and tools for building scalable systems
- Learn from real-world examples of scalable data design









# Recap of core concepts



## What is scalability?

#### From prototype to production

- Scalability is the ability of a system to grow with demand
- It applies to both data pipelines and overall architecture
- Critical for performance, reliability, and costefficiency

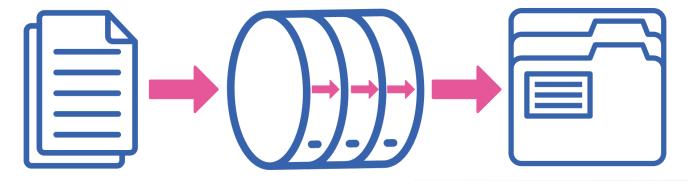
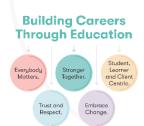


Figure: An abstraction of a data pipeline



Scalability is the ability of your data solution to grow as demand increases.





## Scaling data pipelines

Strategies for growth

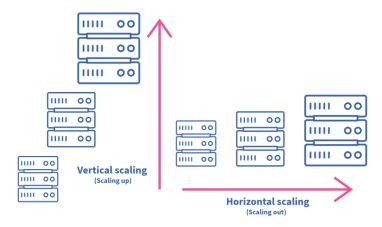


Figure: Vertical vs. Horizontal Scaling

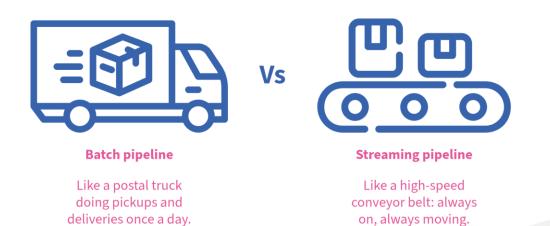


Figure: Difference between Batch and Streaming via an analogy.

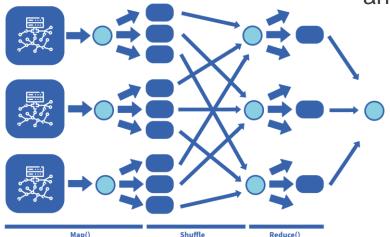


Figure: Diagram of how MapReduce works



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## Tools for scalable pipelines

The Engineer's Toolkit



**Apache Spark:** distributed processing for large-scale data





Apache Kafka: real-time messaging and decoupling



Airflow & Prefect: orchestrate complex workflows





## **Ensuring performance**

Monitoring and optimisation

- Track throughput, latency, and resource usage
- Optimise transformations to reduce compute load
- Choose storage formats that match your workload



Scalable systems depend on visibility monitoring and optimising performance is essential to ensure your pipelines stay fast, efficient, and reliable as they grow.





**Ensuring Performance:** Monitoring and optimisation strategies for scalable data pipelines, Image source: Grafana.com

## **Building for reliability**

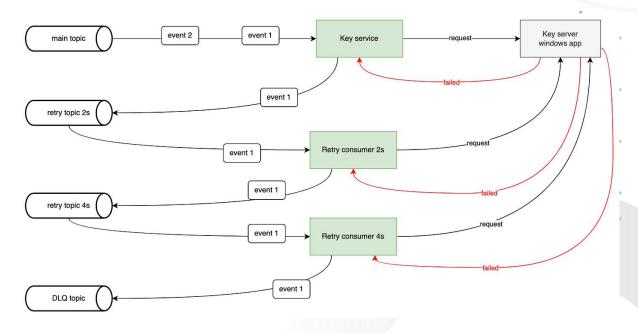
Fault tolerance and resilience

- Use retries, checkpoints, and idempotent design
- Decouple components with queues like Kafka
- Validate data early and test beyond unit tests



Reliability isn't about avoiding failure - it's about designing systems that recover gracefully, validate early, and keep data trustworthy at scale.





Resilient Event Processing: How Retry Logic and Dead Letter Queues Keep Pipelines Reliable, image source:

Medium.com



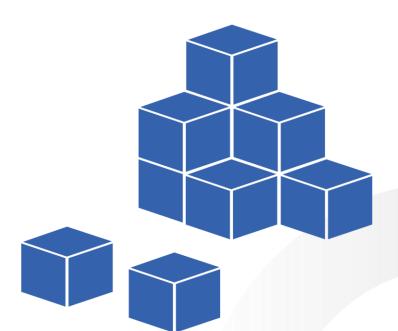
## Designing scalable architectures

#### Thinking beyond pipelines

- Modular, loosely coupled components scale better
- Event-driven design enables real-time responsiveness
- Use the right storage for the right data type



Scalable architectures are built, not bolted on - modularity, event-driven design, and smart storage choices are the foundation for systems that grow and evolve.



Scalable by Design: Modular,
Architectures That Grow with Your Data,
image source: <a href="mailto:freepic.com">freepic.com</a>





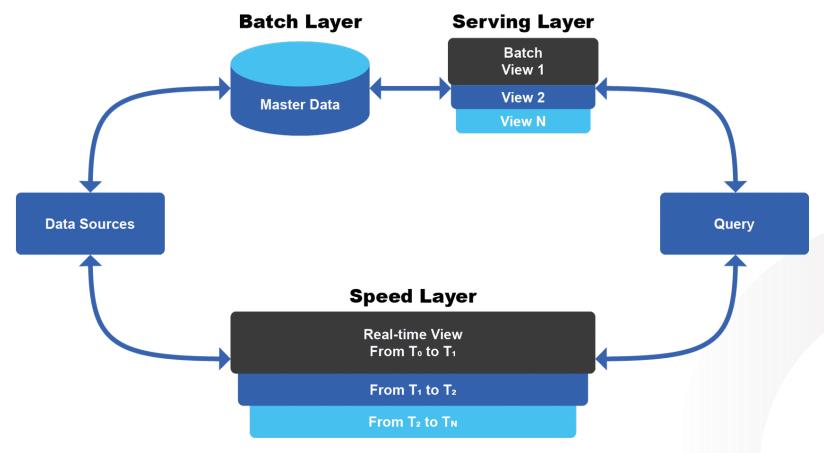


Trust and Respect.

Embrace Change.

## **Architecture patterns**

Lambda



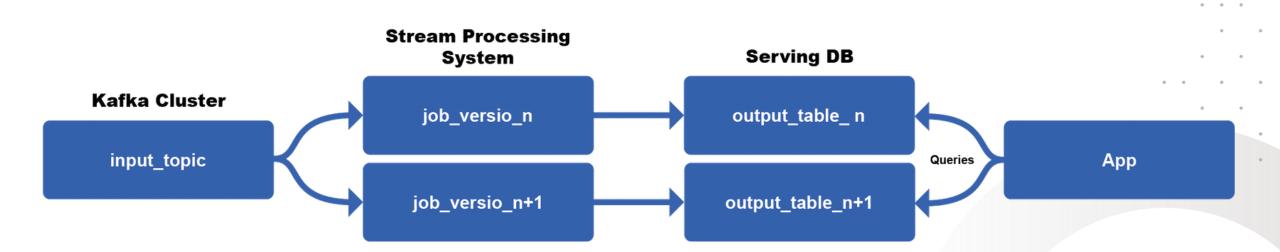
Lambda Architecture combining batching and streaming





## **Architecture patterns**

Kappa



Kappa Architecture for streaming based process

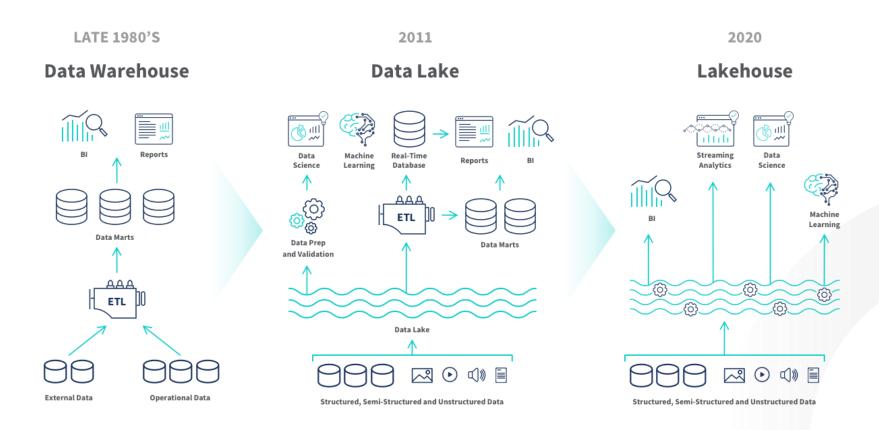


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## **Architecture patterns**

Lakehouse



Kappa Architecture, which is a purely streaming based process, image source: <u>HAZELCAST</u>



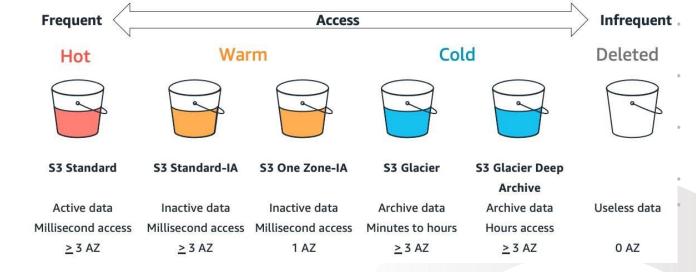


## **Balancing cost and performance**

# Building Careers Through Education Stronger Matters. Stronger Together. Student, Learner Learner Centric.

#### **Smart scaling**

- Store cold data cheaply; query hot data efficiently
- Use batch jobs to reduce always-on compute costs
- Auto-scale with guardrails to avoid surprise bills



Smart Scaling in the Cloud: Matching Storage and Compute to Your Workload Needs



Scalability isn't just about speed - it's about making smart, cost-effective choices that align with your workload and business goals.



### Real – World case studies

Lessons from the field...





Ride-sharing: Kafka + Flink for real-time GPS data

Image source: ottocar.co.uk



**Retail:** Lakehouse model for unified analytics

#### Image source:

Webandcrafts.com



**Media:** Auto-scaling clusters for burst traffic

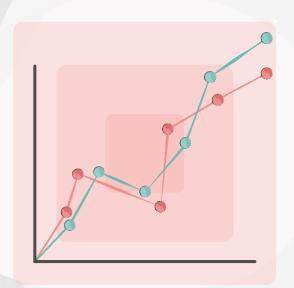
#### Image source:

BBC.co.uk





## **Practical lab**



## **Exercise part 4**

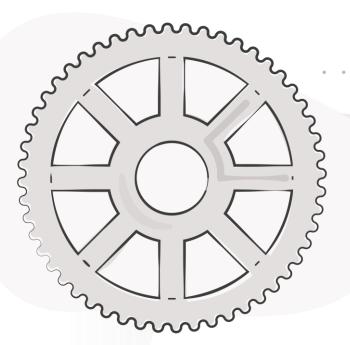
Next phase of design based on Scottish data

- Update the Database Schema: Update your database schema to accommodate data from the Scottish version of the software application.
- Re-implement the Data Pipeline: Re-implement the pipeline to clean, validate, and load the four data sets into your revised database schema.
- **Utilise Additional Test Data:** Use additional test data provided by the Scottish application team, including data for ten Scottish users and the login timestamps.
- Stakeholder Support: Your instructor will act as the stakeholder SMI and answer any questions about the new data.
- **Documentation Requirements:** Create full documentation for both the schema and the pipeline for implementation by other teams.

#### Files Provided in the Hub:

- The 10 sample records are in the "SC User Data.csv" file
- The login audit file is "SC-User-LoginTS.csv."









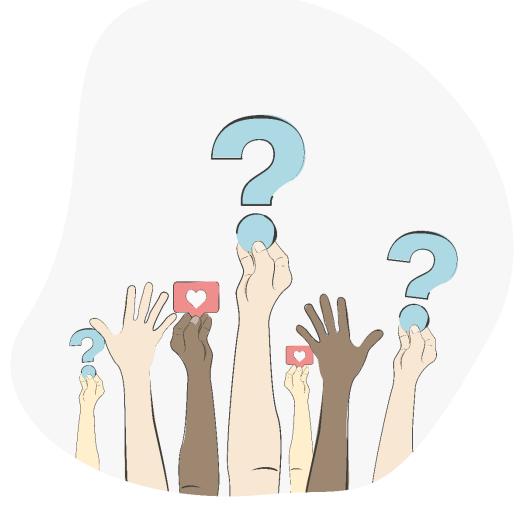
# **Key Learning Summary**

- Scalability Fundamentals Systems must handle growing data volume, complexity, and user demand without performance loss—across both pipelines and architectures.
- Scaling Data Pipelines Use horizontal scaling, stream processing, partitioning, and decoupling (e.g., Kafka) to improve throughput and resilience.
- **Key Tools for Scale** Leverage Apache Spark for distributed processing, Kafka for real-time streaming, and Airflow/Prefect for orchestration.
- **Performance & Reliability -** Monitor metrics like latency and resource usage, optimise transformations, and build in retries, validation, and fault tolerance.
- Architectural Design Principles Design modular, loosely coupled systems using eventdriven flows and polyglot persistence for flexibility and maintainability.
- Architecture Patterns Choose from Lambda (batch + stream), Kappa (stream-only), or Lakehouse (unified analytics) based on your data needs.
- Cost-Performance Trade-offs Store cold data cheaply, use batch jobs to reduce compute costs, and auto-scale with guardrails to avoid surprise bills.





# Any questions or feedback?



















# Thank you

