

# Popular streaming systems

STREAMING CONCEPTS



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# Streaming tools

- **Various tools available** depending on needs
- Allows designers to specify the **best tool for the job**
- **Common systems** include:
  - Celery
  - Kafka
  - Spark Streaming



# Celery

- **Distributed** task queue / **FIFO**
- Used primarily as a **job** or **task queue**
- Often used for **asynchronous tasks**
  - Sending password reset emails
  - Fulfilling digital orders
  - Resizing images
- Allows for **real-time processing** of significant quantity of messages
- Provides functionality for **management and scaling** (vertical & horizontal)



# Apache Kafka

- **Distributed** event streaming system
- Designed to send events between **producers** and **consumers**
  - **Producers** create events on a **topic**
  - **Topics** are basically **messages** of a specified form
  - **Consumers** receive new events
- Different consumers can **handle events as needed** (logging, transformations, relaying, etc)
- Handles **storing of events** as specified
- Extremely **powerful**, but can be **tricky** to **set up**



# Kafka applications

- Best used for **passing data between multiple systems**
  - **Single** source of truth
  - Change data **capture**
  - Data **backups**
  - Data system **migrations**

# Spark streaming

- Part of **Apache Spark**
- Designed to process **streaming** data
- Builds upon the capabilities of Spark to process data in **Scala, Python, SQL, and so forth**
- Useful for processing **large amounts** of data and in **machine learning** scenarios
- Able to **transition from batch to stream** processing fairly easily
- Not designed to store or log events, but primarily to **process or modify** the data



**Let's practice!**  
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# Real-world use case: streaming music service

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# Streaming music

- Consider the **scenario**
- **Not focusing on the actual music** being streamed
- More interested on the **user(s)**
  - Interactions
  - Music preferences
  - Other details

# Interactions

## Primary questions

- What?
  - When?
  - Where?
- Like / Don't play
  - Next / Previous / Skip
  - Select Channel / Playlist
  - Add / remove song from playlist

# How to store data

- Data is archived as a **log**
- Sent as **interactions** occur
- Number of interactions will **vary considerably** between users
- Logged data can be **analyzed later** on

# Analytics

- What about **preferences**?
  - Can be obtained from logged data
  - Favorite genres, bands, etc
  - Most popular times of day
- Other **details**?
  - Most popular app platform / version
  - Location data from stream

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# Real-world use case: sensor data

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# What is sensor data?

- **Automated devices** that monitor some aspect of interest
  - Temperature monitors
  - Electricity usage monitors
  - Vehicle presence detection
  - Many others
- Tend to communicate with **centralized services** for management and data reporting
- Can range from a **few sensors to millions or even billions**

# Connected doorbell

- Monitors primarily for **doorbell presses**
- Contains extra **video / audio capabilities**
- Allows live, remote **interaction**
- Can use camera / environmental sensors for additional **detection capabilities**



<sup>1</sup> C05731, CC BY-SA 4.0 [removed], via Wikimedia Commons



# What are we monitoring?

- Button **presses**
- **Movement** detection
- **Sound** detection
- Requires more **intensive interaction** than just logging of events

# Data handling

- How to **store data**
  - General event data (button press)
  - Sensor-based events (light sensor or audio pickup)
  - Raw data for later analysis
- Different services can have **different SLAs**, even in same product

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# Real-world use case: vaccination clinic

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# Data processing review

- **Batch**
  - Great for *large sets of data*
  - Potentially *poor latency*
- **Queue**
  - Awesome to *maintain order*
  - Can be *tricky to manage*
- **Stream**
  - *Fantastic for latency* / unknown data characteristics
  - *Scaling considerations*

# Complex systems

- **Not all processes fit** within a single processing type
- Many real-world scenarios may require **multiple components** to build the best processing model
- Concepts can be applied to **various components** as required

# Vaccination clinic

- **Multiple, simultaneous** moving pieces
- Vary based on **locale** and **requirements**
- Consider a **large self-contained** clinic
- Concepts apply to smaller pharmacies / doctor's offices as well

# Vaccination clinic areas

- **Arrival / entrance**
  - Entry and temperature check, with a single line
- **Registration**
  - Check-in & validation on info, multiple registrars
- **Vaccine administration**
  - Actual application of vaccine, multiple stations
- **Monitoring**
  - Patients checked for any post-application reactions, many seats
- **Departure**
  - Exit from clinic



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# Congratulations!

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# Next steps

- Learn more about **specific streaming platforms**
  - Apache Kafka ([Apache Kafka](#)) or ([Confluent](#))
  - Apache Spark ([Apache Spark](#))
- **Apply** current data implementations to stream processes
- Work with data consumers to better **determine best processing options for a given situation**

**Thank you!**  
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