What is Apache Kafka?

Apache Kafka is an open-source, **distributed streaming platform** built for **real-time data pipelines** and **stream processing**. Originally developed by LinkedIn in 2011, Kafka is now used to **process trillions of messages per day**.

Real-Life Example:

Let's say you use a food delivery app like Swiggy or Zomato.

Here's how Kafka could help:

- 1. You place an order The order info is sent to Kafka.
- 2. Kafka stores it in an "Orders" box (called a *Topic*).
- 3. The delivery system checks that box and gets the order (as a *Consumer*).
- 4. At the same time, the kitchen system also reads the same order!
- 5. The payment service reads it too and starts processing.

One event (your order) is shared with many services at the same time — all thanks to Kafka.

What Makes Kafka Special?

• Handles a lot of data very fast 🚀

- Can store data safely even if systems crash 💾
- Sends the same data to many apps at the same time
- Helps systems talk to each other without waiting \(\tilde{\Z} \)

Summary:

Kafka is like a smart delivery boy for data.

It takes messages from one app, stores them safely, and delivers them to one or more other apps quickly and reliably.

Why is Kafka so Popular?

Kafka is the **de facto standard** for stream processing used by over **80% of Fortune 100 companies**. Here's why:

- **High Throughput** Millions of messages per second
- Low Latency As low as 2ms
- **High Scalability** Thousands of brokers, petabytes of data
- **Durable Storage** Distributed commit log ensures reliability
- **High Availability** Fault-tolerant across zones and regions

Kafka Architecture: How it Works

Kafka is like a **central hub** where data flows in real-time between different systems. It works with the following parts:

- **Producers** Apps or systems that **send data** into Kafka (e.g., website activity, logs, sensor data).
- Topics Named "folders" inside Kafka that store the data.
- **Consumers** Apps that **read data** from Kafka topics (e.g., dashboards, databases).
- **Brokers** Kafka servers that **manage and store the data** and help Producers and Consumers talk to Kafka.
- **ZooKeeper** / **KRaft** This helps **coordinate** all the Kafka servers and keep the system stable and in sync.

Kafka separates **compute** (processing data) from **storage**, so it can handle real-time data, store it safely, and scale easily.

Key Kafka Capabilities

- Stream Processing Kafka Streams lets you filter, join, and summarize data while it's still moving.
- **Pub/Sub Messaging** Producers publish data, and multiple consumers can read it at the same time.
- **Durability & Reliability** Data is **stored safely** and can be **replayed** if needed.
- Real-Time Analytics You can connect Kafka to tools like Druid or Elasticsearch to get live insights.
- Microservices Communication Kafka helps microservices talk to each other using fast, event-based messages.

Kafka Use Cases

- **Data Pipelines** Move data between systems in real time
- Stream Processing Real-time filtering, transformation, enrichment
- Streaming ETL Extract, transform, and load data continuously
- Event-Driven Microservices Reliable inter-service messaging
- Real-Time Analytics Insights and decisions on live data

Who Uses Kafka?

Major companies like **Uber, LinkedIn, Netflix, British Gas**, and many more rely on Kafka for mission-critical streaming infrastructure.

✓ Apache Kafka Installation Steps (on Ubuntu/Linux)

- # 1. Update system and install Java sudo apt update sudo apt install openidk-11-idk -y
- # 2. Verify Java installation java -version
- # 3. Download Kafka (example: version 3.7.0, Scala 2.13) wget https://downloads.apache.org/kafka/3.7.0/kafka_2.13-3.7.0.tgz
- #4. Extract Kafka archive

5. Navigate to Kafka directory cd kafka 2.13-3.7.0

Microservices eCommerce Project using Flask, Kafka & MongoDB

Q Project Overview

This project demonstrates a beginner-friendly **microservices architecture** that simulates the full lifecycle of an online order — from placing an order to shipping it.

- Flask is used to build lightweight, RESTful microservices.
- Kafka facilitates event-driven communication between services.
- MongoDB is used to persist data at each stage of the order.

Why Microservices?

Microservices split large applications into smaller, loosely coupled services that:

- Are easier to maintain and test.
- Can scale independently.

• Improve fault isolation.

Microservices Overview

Each service has a single responsibility and communicates via **Kafka topics**.

- order service: Accepts new orders.
- inventory service: Verifies inventory.
- payment service: Handles payment.
- shipping_service: Ships the order.
- All order data is stored and updated in MongoDB.

Project Structure

ecommerce-kafka-project/
 docker-compose.yml # Kafka, Zookeeper, MongoDB, etc.
 requirements.txt # Dependencies for all services
 order_service/app.py # Handles new order placements
 inventory_service/app.py # Verifies inventory after order
 payment_service/app.py # Processes payment after inventory

shipping_service/app.py # Marks order as shipped

X Step 1: Docker Setup (docker-compose.yml)

★ Introduction

This file sets up Kafka, Zookeeper, and MongoDB using Docker containers for a consistent and reproducible environment.

yaml

version: '3.8'

services:

zookeeper:

image: confluentinc/cp-zookeeper:latest
environment:

ZOOKEEPER_CLIENT_PORT: 2181

kafka:

image: confluentinc/cp-kafka:latest

ports:

- "9092:9092"

environment:
KAFKA_BROKER_ID: 1
KAFKA_ZOOKEEPER_CONNECT: zookeeper:2181
KAFKA_ADVERTISED_LISTENERS: PLAINTEXT://localhost:9092
KAFKA_OFFSETS_TOPIC_REPLICATION_FACTOR: 1
depends_on:
- zookeeper
mongo:
image: mongo
ports:
- "27017:27017"
volumes:
mongo_data:

Step 2: Install Python Dependencies

Introduction

Install the necessary libraries for Flask, Kafka, and MongoDB to interact within microservices.

requirements.txt:

nginx

flask

kafka-python

pymongo

Install using:

pip install -r requirements.txt



***** Introduction

This service receives incoming orders through an API, saves them in MongoDB, and publishes the event to Kafka.

order_service/app.py:

python

from flask import Flask, request, jsonify

```
from kafka import KafkaProducer
from pymongo import MongoClient
import json
app = Flask( name )
producer = KafkaProducer(
  bootstrap servers='localhost:9092',
  value_serializer=lambda v: json.dumps(v).encode('utf-8')
)
mongo = MongoClient("mongodb://localhost:27017/")
orders = mongo["ecommerce"]["orders"]
@app.route('/order', methods=['POST'])
def create order():
  data = request.json
  data['status'] = 'Order Placed'
```

```
orders.insert_one(data)

producer.send('order-created', data)

return jsonify({'message': 'Order placed successfully'}), 200

if __name__ == '__main__':

app.run(port=5000)
```

Step 4: Inventory Service

Introduction

This service listens to the order-created topic. It checks inventory and then sends an event to the inventory-checked topic.

inventory_service/app.py:

python

from kafka import KafkaConsumer, KafkaProducer

from pymongo import MongoClient

import json

consumer = KafkaConsumer(

```
'order-created',
  bootstrap servers='localhost:9092',
  group id='inventory-group',
  value_deserializer=lambda x: json.loads(x.decode('utf-8'))
)
producer = KafkaProducer(
  bootstrap_servers='localhost:9092',
  value serializer=lambda x: json.dumps(x).encode('utf-8')
)
mongo = MongoClient("mongodb://localhost:27017/")
orders = mongo["ecommerce"]["orders"]
for msg in consumer:
  order = msg.value
  print(f"[Inventory] Order received: {order}")
  order['status'] = 'Inventory Checked'
```

```
orders.update_one({'order_id': order['order_id']}, {'$set': {'status': 'Inventory
Checked'}})
producer.send('inventory-checked', order)
```

Step 5: Payment Service

A Introduction

This service listens to inventory-checked, processes the payment, and sends an event to the payment-processed topic.

payment_service/app.py:

python

from kafka import KafkaConsumer, KafkaProducer

from pymongo import MongoClient

import json

```
consumer = KafkaConsumer(
```

'inventory-checked',

bootstrap_servers='localhost:9092',

group_id='payment-group',

value_deserializer=lambda x: json.loads(x.decode('utf-8'))

```
)
producer = KafkaProducer(
  bootstrap_servers='localhost:9092',
  value serializer=lambda x: json.dumps(x).encode('utf-8')
)
mongo = MongoClient("mongodb://localhost:27017/")
orders = mongo["ecommerce"]["orders"]
for msg in consumer:
  order = msg.value
  print(f"[Payment] Processing: {order}")
  order['status'] = 'Payment Done'
  orders.update_one({'order_id': order['order_id']}, {'$set': {'status': 'Payment
Done'}})
  producer.send('payment-processed', order)
```

Step 6: Shipping Service

***** Introduction

The final service listens to payment-processed and marks the order as shipped in MongoDB.

```
shipping_service/app.py:

python
```

```
from kafka import KafkaConsumer from pymongo import MongoClient import json
```

```
consumer = KafkaConsumer(
   'payment-processed',
   bootstrap_servers='localhost:9092',
   group_id='shipping-group',
   value_deserializer=lambda x: json.loads(x.decode('utf-8'))
)
```

```
mongo = MongoClient("mongodb://localhost:27017/")
orders = mongo["ecommerce"]["orders"]

for msg in consumer:
    order = msg.value
    print(f''[Shipping] Shipped: {order}")
    orders.update_one({'order_id': order['order_id']}, {'$set': {'status': 'Shipped'}})
```

Step 7: Running the Project

Introduction

You'll now bring everything together: start services, run microservices, and test the end-to-end flow.

✓ Start Docker Services:

docker-compose up -d

W Run Python Services (Each in a separate terminal):

python order_service/app.py

python inventory service/app.py

```
python payment_service/app.py
python shipping service/app.py
```

Step 8: Test the Workflow



Simulate a real user placing an order and watch it move through all services.

Place an Order via curl:

```
curl -X POST http://localhost:5000/order \
   -H 'Content-Type: application/json' \
   -d '{"order id": "101", "user": "Alice", "item": "Laptop"}'
```

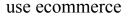
Step 9: View Orders in MongoDB

Introduction

Verify all updates at each microservice stage from within the MongoDB container.

docker exec -it \$(docker ps -qf "name=mongo") mongosh

Inside the shell:



db.orders.find().pretty()

© Final Notes

- This project teaches you how to build a **decoupled**, event-driven system.
- Services only communicate through Kafka, ensuring flexibility and scalability.
- MongoDB provides **centralized state** and traceability of the order.

NevOps Project: Real-Time Monitoring & Alerting System with Kafka

All setups, configuration files, and introductions are included. You can directly run this with **Docker Compose**, visualize logs with **Kibana**, view metrics in Grafana, and get alerts when something goes wrong.



Project Name:

Kafka-Powered Monitoring Pipeline

Goal:

To build a real-time monitoring system that:

- Collects logs and metrics from microservices
- Streams data using Kafka
- Indexes logs with Elasticsearch
- Visualizes logs (Kibana) and metrics (Grafana)
- Triggers alerts on issues (like CPU > 80%) via Slack/Email

X Tools Used:

- Kafka (message broker)
- Zookeeper (Kafka dependency)
- Filebeat (log shipper)
- Logstash (log processor)
- Elasticsearch (log storage & search)
- **Kibana** (log visualization)
- Prometheus (metrics collection)
- Grafana (dashboard)

- Alertmanager (for alerts)
- **Docker + Docker Compose** (container orchestration)

*** How the System Works:**

Step 1: Spin Up the Core Services with Docker Compose

Introduction:

We use Docker Compose to set up all services: Kafka, Zookeeper, Prometheus, Grafana, Elasticsearch, Kibana, Filebeat, and Logstash. This allows a complete monitoring system on your local machine.

yaml

```
# docker-compose.yml version: '3.7'
```

services:

```
zookeeper:
image: confluentinc/cp-zookeeper:7.2.1
environment:
ZOOKEEPER_CLIENT_PORT: 2181
ports:
- "2181:2181"
```

kafka:

```
image: confluentinc/cp-kafka:7.2.1
depends_on:
    - zookeeper
ports:
    - "9092:9092"
```

```
environment:
  KAFKA BROKER ID: 1
  KAFKA ZOOKEEPER CONNECT: zookeeper:2181
  KAFKA ADVERTISED LISTENERS: PLAINTEXT://kafka:9092
  KAFKA OFFSETS TOPIC REPLICATION FACTOR: 1
elasticsearch:
 image: docker.elastic.co/elasticsearch/elasticsearch:8.6.2
 environment:
  - discovery.type=single-node
  - xpack.security.enabled=false
  - bootstrap.memory lock=true
  - ES_JAVA_OPTS=-Xms512m -Xmx512m
 ports:
  - "9200:9200"
 ulimits:
  memlock:
   soft: -1
   hard: -1
 volumes:
  - esdata:/usr/share/elasticsearch/data
kibana:
 image: docker.elastic.co/kibana/kibana:8.6.2
 environment:
  - ELASTICSEARCH HOSTS=http://elasticsearch:9200
 ports:
  - "5601:5601"
```

prometheus:

depends on:

- elasticsearch

image: prom/prometheus

volumes:

 $\hbox{-./prometheus.yml:/etc/prometheus/prometheus.yml}\\$

```
ports:
   - "9090:9090"
 grafana:
  image: grafana/grafana
  ports:
   - "3000:3000"
 logstash:
  image: docker.elastic.co/logstash/logstash:8.6.2
  volumes:
   - ./logstash.conf:/usr/share/logstash/pipeline/logstash.conf
  depends on:
   - kafka
   - elasticsearch
  ports:
   - "5044:5044"
 filebeat:
  image: docker.elastic.co/beats/filebeat:8.6.2
  volumes:
   - ./filebeat.yml:/usr/share/filebeat/filebeat.yml
   - ./logs:/logs
  depends on:
   - kafka
volumes:
 esdata:
```

Step 2: Configure Filebeat to Ship Logs to Kafka

Q Introduction:

Filebeat monitors log files (like /logs/app.log) and sends them to Kafka.

```
yaml
```

```
# filebeat.yml
filebeat.inputs:
- type: log
    enabled: true
    paths:
    - /logs/*.log

output.kafka:
hosts: ["kafka:9092"]
topic: "logs"
codec.json:
    pretty: false
```

• Step 3: Configure Logstash to Process Logs from Kafka and Send to Elasticsearch

Introduction:

Logstash reads logs from Kafka and forwards them to Elasticsearch for indexing and Kibana visualization.

```
conf
```

```
# logstash.conf
input {
  kafka {
    bootstrap_servers => "kafka:9092"
  topics => ["logs"]
  codec => "json"
  }
}
filter {
```

```
# Add transformations or filters here
}

output {
  elasticsearch {
   hosts => ["http://elasticsearch:9200"]
   index => "app-logs-%{+YYYY.MM.dd}"
  }
  stdout { codec => rubydebug }
}
```

Step 4: Create Prometheus Configuration

Q Introduction:

yaml

Prometheus scrapes metrics from your services (e.g., Node Exporter, apps) and stores time-series data.

```
# prometheus.yml
global:
    scrape_interval: 15s

scrape_configs:
    - job_name: 'node'
    static_configs:
    - targets: ['localhost:9100']
```

Step 5: Launch Grafana and Add Prometheus Data Source

Introduction:

Grafana connects to Prometheus and visualizes your metrics with dashboards.

- Login at http://localhost:3000
- Default credentials: admin / admin
- Add Prometheus as a data source (http://prometheus:9090)
- Import dashboards or create your own

Step 6: Add Alerting Rules

Introduction:

Alertmanager integrated with Prometheus can send alerts when thresholds are breached.

yaml

```
# alerts/alert.rules
groups:
- name: system-alerts
rules:
- alert: HighCPUUsage
    expr: node_cpu_seconds_total > 80
    for: 1m
    labels:
        severity: warning
    annotations:
        summary: "High CPU usage detected"
        description: "CPU usage is over 80% for 1 minute"
```

You can plug this into Prometheus config and use Alertmanager to notify via Slack/Email.

Step 7: Create Kafka Topics (Logs, Metrics)

Introduction:

Kafka needs topics where logs and metrics will be published.

docker exec -it kafka kafka-topics --create --topic logs --bootstrap-server kafka:9092 --replication-factor 1 --partitions 1

Repeat this for other topics if needed.

• Step 8: View Logs in Kibana

Introduction:

Kibana provides a UI to view, filter, and visualize logs coming from Elasticsearch.

- Access: http://localhost:5601
- Go to "Stack Management" → "Index Patterns" → Add app-logs-*



Apache Kafka Commands

Start Zookeeper Server

bin/zookeeper-server-start.sh config/zookeeper.properties

Start Kafka Broker Server

bin/kafka-server-start.sh config/server.properties

Create a Kafka Topic

bin/kafka-topics.sh --create --topic test-topic --bootstrap-server localhost:9092 --partitions 1 --replication-factor 1

List Kafka Topics

bin/kafka-topics.sh --list --bootstrap-server localhost:9092

Start Kafka Producer (send messages)

bin/kafka-console-producer.sh --topic test-topic --bootstrap-server localhost:9092

Start Kafka Consumer (read messages)

bin/kafka-console-consumer.sh --topic test-topic --from-beginning --bootstrap-server localhost:9092

RabbitMQ and Kafka are message brokers, but they are used for different use-cases.

VS RabbitMQ vs Kafka: Key Differences

Feature	RabbitMQ	Apache Kafka
Туре	Message Queue (traditional broker)	Distributed Event Streaming Platform
Message Pattern	Push-based (broker pushes to consumers)	Pull-based (consumers pull from Kafka)

Ordering	Not guaranteed	Guaranteed per partition
Speed	Slower for high throughput	Designed for high throughput (millions/sec)
Storage	Transient (short-lived messages)	Durable (messages stored for days/weeks)
Use-case	Job queues, async processing	Real-time analytics, event streaming
Replay	Not possible after ACK	Yes, you can replay old messages
Built-in Features	Routing, priority queues, TTL, retries	Partitioning, replication, stream replay
Ease of Use	Easier for beginners	Needs deeper understanding (like partitions, offsets)
Dependencies	No Zookeeper	Needs Zookeeper (or KRaft mode)

& Summary: Which is best?

For	Best Choice
Simple task queues	✓ RabbitMQ
Complex routing	✓ RabbitMQ
Real-time streaming	Kafka
High data volume	Kafka
Replay old	Kafka

messages

Beginners

