

Real-Time Wikimedia Streaming Analytics

Overview

This project is an **end-to-end real-time streaming analytics platform** that ingests live Wikipedia edit events, processes them in real time, and stores aggregated metrics for low-latency analytical querying.

The system is designed to mirror **production-grade streaming architectures** used at companies like TikTok, Netflix, and Uber.

Architecture

```
Wikimedia EventStreams (Live SSE API)
  ↓
Kafka (Real-time Event Log)
  ↓
Apache Flink (PyFlink Stream Processing)
  ↓
ClickHouse (Real-time Analytics Database)
  ↓
SQL / Dashboards
```

Technologies Used

- **Apache Kafka** – real-time data ingestion and buffering
 - **Apache Flink (PyFlink)** – event-time stream processing
 - **ClickHouse** – low-latency analytical database
 - **Docker & Docker Compose** – containerized deployment
 - **Python** – streaming logic
 - **SQL** – analytics and validation
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Data Flow

Data Ingestion

- Live Wikipedia edit events are streamed from **Wikimedia EventStreams**
- Events include wiki name, timestamp, and bot flag
- Events are continuously published into a Kafka topic (`wikimedia_events`)

Kafka

- Acts as a durable, scalable event log
- Decouples producers and consumers
- Enables fault tolerance and replayability

Apache Flink (PyFlink)

- Consumes events from Kafka
- Applies **event-time processing and watermarks**
- Filters bot edits
- Aggregates edit counts using **1-minute tumbling windows**
- Processes data continuously in memory

ClickHouse

- Stores aggregated results via JDBC sink
- Enables millisecond-level analytical SQL queries
- Serves as the real-time analytics store

ClickHouse Schema

```
CREATE TABLE analytics.wiki_edits (  
    wiki String,  
    window_start DateTime,  
    edit_count UInt64  
)  
ENGINE = MergeTree()  
ORDER BY (wiki, window_start);
```

How to Run

Start All Services

```
docker-compose up -d
```

Services started: - Kafka - Zookeeper - Flink JobManager & TaskManager - ClickHouse

Run PyFlink Job

```
docker exec -it flink-jobmanager  
/opt/flink/bin/flink run -py /opt/flink/wiki_stream.py
```

Check Flink UI:

```
http://localhost:8081
```

Verify Data in ClickHouse

```
docker exec -it clickhouse clickhouse-client
```

```
SELECT *  
FROM analytics.wiki_edits  
ORDER BY window_start DESC  
LIMIT 10;
```

New rows appear **every minute**.

Sample Analytics Queries

Top Wikis (Last 5 Minutes)

```
SELECT wiki, SUM(edit_count) AS total_edits  
FROM analytics.wiki_edits  
WHERE window_start > now() - INTERVAL 5 MINUTE  
GROUP BY wiki  
ORDER BY total_edits DESC;
```

Edit Volume Over Time

```
SELECT window_start, SUM(edit_count)  
FROM analytics.wiki_edits  
GROUP BY window_start  
ORDER BY window_start;
```



Validation & Testing

- Verified Kafka topic creation and consumption
 - Confirmed Flink job execution via Flink UI
 - Manually tested ClickHouse inserts
 - Validated real-time data ingestion using SQL queries
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Key Engineering Concepts Demonstrated

- Real-time streaming architectures
 - Kafka topic and broker management
 - Event-time vs processing-time semantics
 - Windowed aggregations
 - Stream-to-analytics database integration
 - Docker-based deployment
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Why This Project Is Strong

- Uses **production-grade technologies**
 - Demonstrates **real-time data engineering skills**
 - End-to-end pipeline with validation
 - SQL-accessible real-time analytics
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Resume Summary

Built a real-time streaming analytics platform using Kafka, PyFlink, and ClickHouse to process live Wikimedia events with event-time windowing and low-latency analytical queries.



Future Enhancements

- Grafana dashboards on ClickHouse
 - Exactly-once processing with Flink checkpoints
 - Alerting on edit spikes
 - Flink SQL implementation
 - Historical analysis with Hive / Presto
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Author

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