

# Chanakya University



**CHANAKYA**  
UNIVERSITY

## **School of Mathematics and Natural Science**

**TOPIC: Real-Time Patient Health Monitoring System**

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**SUBJECT:**

Real Time Analytics (MAT503)

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## **Contribution Statement**

The project work was carried out collaboratively by all group members, with responsibilities divided to ensure smooth execution and timely completion.

- **Faizan Ansari:** Involved in end-to-end implementation of the project including dataset understanding, Kafka Producer development, Kafka Consumer integration with PostgreSQL, database setup, Apache Superset configuration, chart creation, and dashboard design. Also responsible for compiling the final report and documentation.
- **Umesh Naik:** Assisted in understanding project requirements, reviewing Kafka concepts, helping with PostgreSQL table verification, and validating data flow from Kafka to the database.
- **Dhanush Kumar V:** Supported testing, verification of outputs, taking screenshots for documentation, and reviewing Superset visualizations for correctness.

All members contributed through discussions, validation, and review of the final outputs.

## **STEP 1. Introduction**

In modern healthcare systems, continuous monitoring of patient vital signs is extremely important. Real-time data processing helps doctors and healthcare staff quickly identify abnormal conditions and take timely action. Traditional batch-based systems are not suitable for such real-time requirements.

This mini project implements a **Real-Time Healthcare Monitoring System** using:

- **Apache Kafka** for real-time data streaming
- **Java** for producer and consumer applications
- **PostgreSQL** for structured data storage
- **Apache Superset** for interactive data visualization

The system simulates patient vital signs such as heart rate, SpO<sub>2</sub>, blood pressure, and body temperature, streams them in real time, stores them in a database, and visualizes them using dashboards.

## **STEP 2. Problem Statement**

The objective of this project is to:

- Stream patient vital data in real time
- Store the streamed data reliably
- Visualize patient health trends and alerts

### **Key Requirements:**

- Kafka topic: patient\_vitals
- PostgreSQL database: HealthcareMonitoring
- Table: patient\_vitals
- Composite Primary Key: (patient\_id, recorded\_time)
- Dashboard with multiple analytical charts

### **STEP 3. System Architecture**

The overall architecture of the system is as follows:

1. CSV dataset acts as the data source
2. Kafka Producer reads CSV and publishes JSON messages to Kafka topic
3. Kafka Broker manages real-time streaming
4. Kafka Consumer subscribes to topic and consumes messages
5. Consumer inserts data into PostgreSQL using prepared statements
6. Apache Superset connects to PostgreSQL
7. Dashboards visualize real-time patient vitals

## **STEP 4. Dataset Description**

The dataset patient\_vitals.csv contains simulated healthcare data. Each row represents one recorded vital measurement of a patient.

### **Dataset Columns:**

- patient\_id – Unique identifier for each patient
- recorded\_time – Timestamp of measurement
- heart\_rate – Heart rate in BPM
- spo2 – Oxygen saturation level (%)
- systolic\_bp – Systolic blood pressure
- diastolic\_bp – Diastolic blood pressure
- body\_temperature – Body temperature in °C
- alert\_flag – Alert level (NORMAL / WARNING / CRITICAL)

	A	B	C	D	E	F	G	H
1	patient_id	recorded_time	heart_rate	spo2	systolic_bp	diastolic_bp	body_temperature	alert_flag
2	PAT001	2025-12-15T09:	72	98	120	78	36.6	NORMAL
3	PAT002	2025-12-15T09:	88	97	124	80	36.8	NORMAL
4	PAT003	2025-12-15T09:	96	95	132	84	37.2	WARNING
5	PAT004	2025-12-15T09:	110	93	140	90	38.1	CRITICAL
6	PAT005	2025-12-15T09:	78	98	118	76	36.5	NORMAL
7	PAT006	2025-12-15T09:	85	96	126	82	36.9	NORMAL
8	PAT007	2025-12-15T09:	102	94	138	88	37.8	WARNING
9	PAT008	2025-12-15T09:	115	92	145	92	38.5	CRITICAL
10	PAT009	2025-12-15T09:	74	99	116	74	36.4	NORMAL
11	PAT010	2025-12-15T09:	90	97	128	82	37	NORMAL
12	PAT011	2025-12-15T09:	98	95	134	86	37.4	WARNING
13	PAT012	2025-12-15T09:	112	93	142	90	38	CRITICAL
14	PAT013	2025-12-15T09:	76	98	120	78	36.6	NORMAL
15	PAT014	2025-12-15T09:	84	96	124	80	36.9	NORMAL
16	PAT015	2025-12-15T09:	100	94	136	86	37.6	WARNING
17	PAT016	2025-12-15T09:	118	91	148	94	38.7	CRITICAL
18	PAT017	2025-12-15T09:	79	99	118	76	36.5	NORMAL
19	PAT018	2025-12-15T09:	92	97	130	84	37.1	NORMAL
20	PAT019	2025-12-15T09:	104	94	140	88	37.9	WARNING
21	PAT020	2025-12-15T09:	120	90	150	96	39	CRITICAL

## **STEP 5. Kafka Topic Creation**

### **Objective**

The objective of this step is to create a Kafka topic that will act as a communication channel between the **Kafka Producer** and **Kafka Consumer**.

All patient vital records produced by the producer application will be published to this topic, and the consumer will subscribe to it for further processing and storage in PostgreSQL.

---

### **Topic Name**

#### **patient\_vitals**

This topic is used to stream patient health data such as:

- Patient ID
  - Recorded time
  - Heart rate
  - SpO<sub>2</sub>
  - Blood pressure
  - Body temperature
  - Alert flag
- 

### **Kafka Environment Assumption**

Before creating the topic, the following Kafka services were already running in the background:

- ZooKeeper service



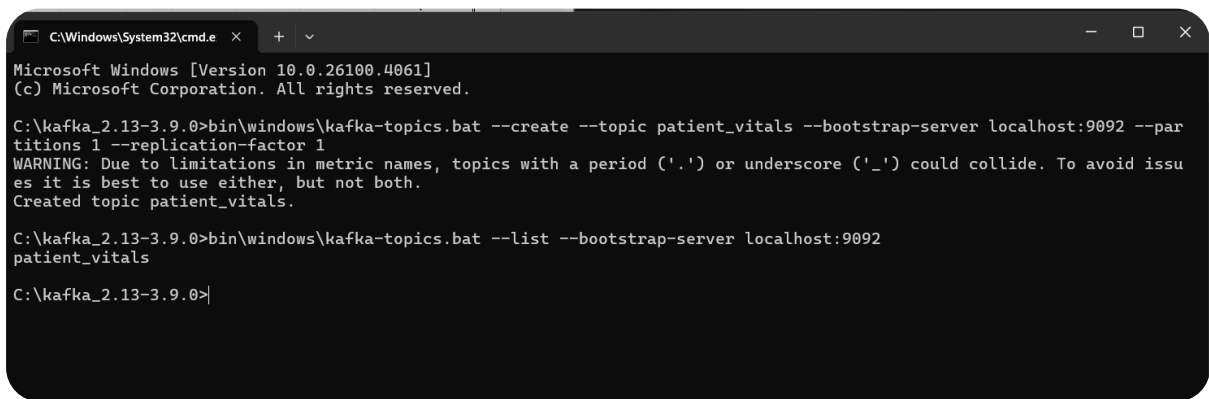
- Kafka broker service

These services are required for Kafka operations but do not generate direct outputs, hence their screenshots are not included.

---

## Command Used to Create Kafka Topic

The following command was executed using the Kafka command-line interface on Windows:



```
C:\Windows\System32\cmd.e  x  +  v
Microsoft Windows [Version 10.0.26100.4061]
(c) Microsoft Corporation. All rights reserved.

C:\kafka_2.13-3.9.0>bin\windows\kafka-topics.bat --create --topic patient_vitals --bootstrap-server localhost:9092 --part
tions 1 --replication-factor 1
WARNING: Due to limitations in metric names, topics with a period ('.') or underscore ('_') could collide. To avoid issu
es it is best to use either, but not both.
Created topic patient_vitals.

C:\kafka_2.13-3.9.0>bin\windows\kafka-topics.bat --list --bootstrap-server localhost:9092
patient_vitals

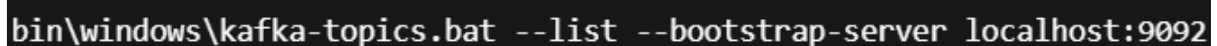
C:\kafka_2.13-3.9.0>|
```

## Explanation of Parameters

- `--topic patient_vitals` → Name of the Kafka topic
  - `--bootstrap-server localhost:9092` → Kafka broker address
  - `--partitions 1` → Single partition (sufficient for this project)
  - `--replication-factor 1` → Single replica (local setup)
- 

## Verification of Topic Creation

After creating the topic, the following command was executed to verify that the topic was successfully created:



```
bin\windows\kafka-topics.bat --list --bootstrap-server localhost:9092
```

## **STEP 6: Kafka Producer Implementation (Patient Vitals Producer)**

### **Objective**

The objective of this step is to develop a **Kafka Producer application in Java** that reads patient health data from a CSV file, converts each record into JSON format, and publishes it to the Kafka topic **patient\_vitals** in real time.

This producer simulates a real-world healthcare data stream where patient vitals are continuously generated and sent for downstream processing.

---

### **Technology Stack Used**

- **Programming Language:** Java
  - **IDE:** Eclipse
  - **Build Tool:** Maven
  - **Messaging System:** Apache Kafka
  - **Data Format:** CSV → JSON
- 

### **Dataset Used**

- File name: patient\_vitals.csv
  - Location: src/main/resources
  - Contains multiple rows of patient health records including:
    - patient\_id
    - recorded\_time
    - heart\_rate
    - spo2
    - systolic\_bp
    - diastolic\_bp
    - body\_temperature
    - alert\_flag
- 

### **Producer Design Flow**

1. Read the CSV file line by line.
  2. Skip the header row.
  3. Parse each row into individual fields.
  4. Convert the row into a structured JSON object.
  5. Send the JSON message to Kafka topic patient\_vitals.
  6. Introduce a small delay to simulate real-time streaming.
- 

### **Kafka Producer Configuration**

The producer is configured using the following properties:

- Kafka broker address: localhost:9092
- Key serializer: StringSerializer
- Value serializer: StringSerializer

These settings allow the producer to send string-based JSON messages to Kafka.

## Producer Code Implementation

The screenshot shows the Eclipse IDE with the `PatientVitalsProducer.java` file open. The code is a Java class that reads a CSV file and publishes data to a Kafka topic. The console output shows the Kafka messages being sent, which are in JSON format.

```

1 package com.healthcare.kafka;
2
3 import org.apache.kafka.clients.producer.KafkaProducer;
4 import org.apache.kafka.clients.producer.Producer;
5 import org.apache.kafka.clients.producer.ProducerConfig;
6 import org.apache.kafka.clients.producer.ProducerRecord;
7 import org.apache.kafka.common.serialization.StringSerializer;
8
9 import java.io.BufferedReader;
10 import java.io.InputStreamReader;
11 import java.util.Properties;
12 import java.util.concurrent.TimeUnit;
13
14 public class PatientVitalsProducer {
15
16     public static void main(String[] args) {
17
18         try {
19             // Kafka producer configuration
20             Properties props = new Properties();
21             props.put(ProducerConfig.BOOTSTRAP_SERVERS_CONFIG, "localhost:9092");
22             props.put(ProducerConfig.KEY_SERIALIZER_CLASS_CONFIG, StringSerializer.class.getName());
23             props.put(ProducerConfig.VALUE_SERIALIZER_CLASS_CONFIG, StringSerializer.class.getName());
24
25             Producer<String, String> producer = new KafkaProducer<>(props);
26
27             // Read CSV from src/main/resources
28             BufferedReader br = new BufferedReader(
29                 new InputStreamReader(
30                     PatientVitalsProducer.class
31                         .getClassLoader()
32                         .getResourceAsStream("patient_vitals.csv")
33                 )
34             );
35
36             String line;
37             br.readLine(); // skip header row
38
39             while ((line = br.readLine()) != null) {
40
41                 String[] d = line.split(",");
42
43                 // JSON output EXACTLY like your console output
44                 String json = "{"
45                     + "\"patient_id\":\"" + d[0] + "\""

```

The console output shows the following messages being sent to the Kafka topic:

```

Sent + {"patient_id":"PAT001","recorded_time":"2025-12-15T09:00:00","heart_rate":72,"spo2":98,"
Sent + {"patient_id":"PAT002","recorded_time":"2025-12-15T09:01:00","heart_rate":88,"spo2":97,"
Sent + {"patient_id":"PAT003","recorded_time":"2025-12-15T09:02:00","heart_rate":96,"spo2":95,"
Sent + {"patient_id":"PAT004","recorded_time":"2025-12-15T09:03:00","heart_rate":110,"spo2":93,"
Sent + {"patient_id":"PAT005","recorded_time":"2025-12-15T09:04:00","heart_rate":78,"spo2":98,"
Sent + {"patient_id":"PAT006","recorded_time":"2025-12-15T09:05:00","heart_rate":85,"spo2":96,"
Sent + {"patient_id":"PAT007","recorded_time":"2025-12-15T09:06:00","heart_rate":102,"spo2":94,"
Sent + {"patient_id":"PAT008","recorded_time":"2025-12-15T09:07:00","heart_rate":115,"spo2":92,"
Sent + {"patient_id":"PAT009","recorded_time":"2025-12-15T09:08:00","heart_rate":74,"spo2":99,"
Sent + {"patient_id":"PAT010","recorded_time":"2025-12-15T09:09:00","heart_rate":90,"spo2":97,"
Sent + {"patient_id":"PAT011","recorded_time":"2025-12-15T09:10:00","heart_rate":98,"spo2":95,"
Sent + {"patient_id":"PAT012","recorded_time":"2025-12-15T09:11:00","heart_rate":112,"spo2":93,"
Sent + {"patient_id":"PAT013","recorded_time":"2025-12-15T09:12:00","heart_rate":76,"spo2":98,"
Sent + {"patient_id":"PAT014","recorded_time":"2025-12-15T09:13:00","heart_rate":84,"spo2":96,"
Sent + {"patient_id":"PAT015","recorded_time":"2025-12-15T09:14:00","heart_rate":100,"spo2":94,"
Sent + {"patient_id":"PAT016","recorded_time":"2025-12-15T09:15:00","heart_rate":118,"spo2":91,"
Sent + {"patient_id":"PAT017","recorded_time":"2025-12-15T09:16:00","heart_rate":79,"spo2":99,"
Sent + {"patient_id":"PAT018","recorded_time":"2025-12-15T09:17:00","heart_rate":92,"spo2":97,"
Sent + {"patient_id":"PAT019","recorded_time":"2025-12-15T09:18:00","heart_rate":104,"spo2":94,"
Sent + {"patient_id":"PAT020","recorded_time":"2025-12-15T09:19:00","heart_rate":120,"spo2":90,"

```

## Execution Process

1. The Kafka producer application is run from Eclipse.
2. The CSV file is read successfully from the resources folder.
3. Each record is converted into JSON format.
4. Messages are continuously published to the Kafka topic.
5. Console output confirms message transmission.

## Outcome

- Patient vitals data is successfully streamed from CSV to Kafka.
- Data is published in structured JSON format.
- The Kafka topic `patient_vitals` now contains real-time patient health data.
- This data is ready to be consumed by the Kafka Consumer for database insertion.

## STEP 7: Kafka Consumer Implementation (Kafka → PostgreSQL)

### Objective

The objective of this step is to develop a **Kafka Consumer application in Java** that continuously reads patient health data from the Kafka topic **patient\_vitals**, parses the incoming JSON messages, and inserts the records into a **PostgreSQL database** using **prepared statements**.

This step completes the real-time data pipeline by persisting streaming data into a relational database for further analysis and visualization.

### Technology Stack Used

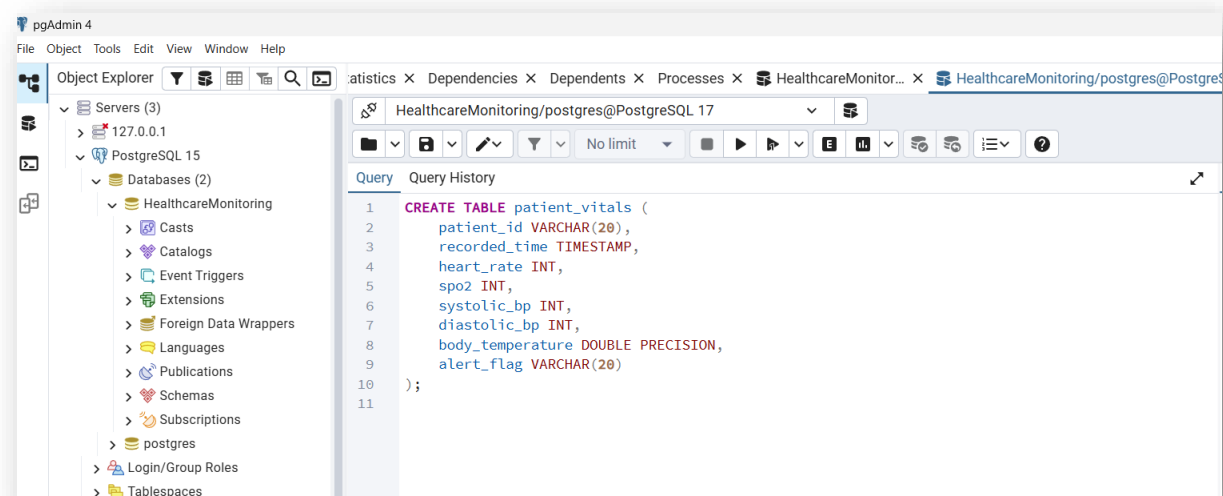
- **Programming Language:** Java
- **IDE:** Eclipse
- **Build Tool:** Maven
- **Messaging System:** Apache Kafka
- **Database:** PostgreSQL
- **Data Format:** JSON
- **JDBC Driver:** PostgreSQL JDBC

### Database Configuration

#### Database Details

- **Database Name:** HealthcareMonitoring
- **Table Name:** patient\_vitals
- **Primary Key:** (patient\_id, recorded\_time)

#### Table Structure



## Consumer Design Flow

1. Subscribe to Kafka topic patient\_vitals
2. Poll messages continuously from Kafka
3. Parse incoming JSON records
4. Extract individual health parameters
5. Insert records into PostgreSQL using prepared statements
6. Display confirmation messages in the console

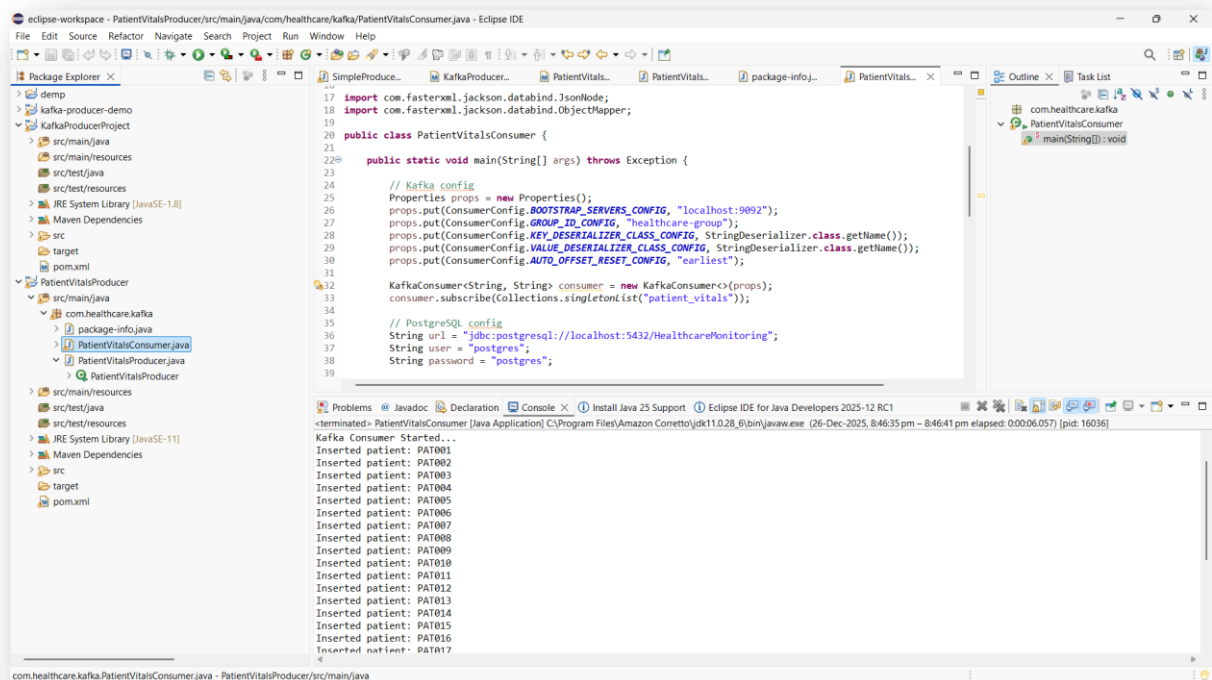
## Kafka Consumer Configuration

The consumer is configured with the following properties:

- **Bootstrap Server:** localhost:9092
- **Consumer Group ID:** healthcare-group
- **Key Deserializer:** StringDeserializer
- **Value Deserializer:** StringDeserializer
- **Offset Reset Policy:** earliest

These settings ensure reliable consumption of all messages from the beginning of the topic.

## Kafka Consumer Code Implementation



```
17 import com.fasterxml.jackson.databind.JsonNode;
18 import com.fasterxml.jackson.databind.ObjectMapper;
19
20 public class PatientVitalsConsumer {
21
22     public static void main(String[] args) throws Exception {
23
24         // Kafka config
25         Properties props = new Properties();
26         props.put(ConsumerConfig.BOOTSTRAP_SERVERS_CONFIG, "localhost:9092");
27         props.put(ConsumerConfig.GROUP_ID_CONFIG, "healthcare-group");
28         props.put(ConsumerConfig.KEY_DESERIALIZER_CLASS_CONFIG, StringDeserializer.class.getName());
29         props.put(ConsumerConfig.VALUE_DESERIALIZER_CLASS_CONFIG, StringDeserializer.class.getName());
30         props.put(ConsumerConfig.AUTO_OFFSET_RESET_CONFIG, "earliest");
31
32         KafkaConsumer<String, String> consumer = new KafkaConsumer<>(props);
33         consumer.subscribe(Collections.singletonList("patient_vitals"));
34
35         // PostgreSQL config
36         String url = "jdbc:postgresql://localhost:5432/HealthcareMonitoring";
37         String user = "postgres";
38         String password = "postgres";
39
40     }
41 }
```

com.healthcare.kafka.PatientVitalsConsumer - PatientVitalsProducer/src/main/java

Kafka Consumer Started...

Inserted patient: PAT001

Inserted patient: PAT002

Inserted patient: PAT003

Inserted patient: PAT004

Inserted patient: PAT005

Inserted patient: PAT006

Inserted patient: PAT007

Inserted patient: PAT008

Inserted patient: PAT009

Inserted patient: PAT010

Inserted patient: PAT011

Inserted patient: PAT012

Inserted patient: PAT013

Inserted patient: PAT014

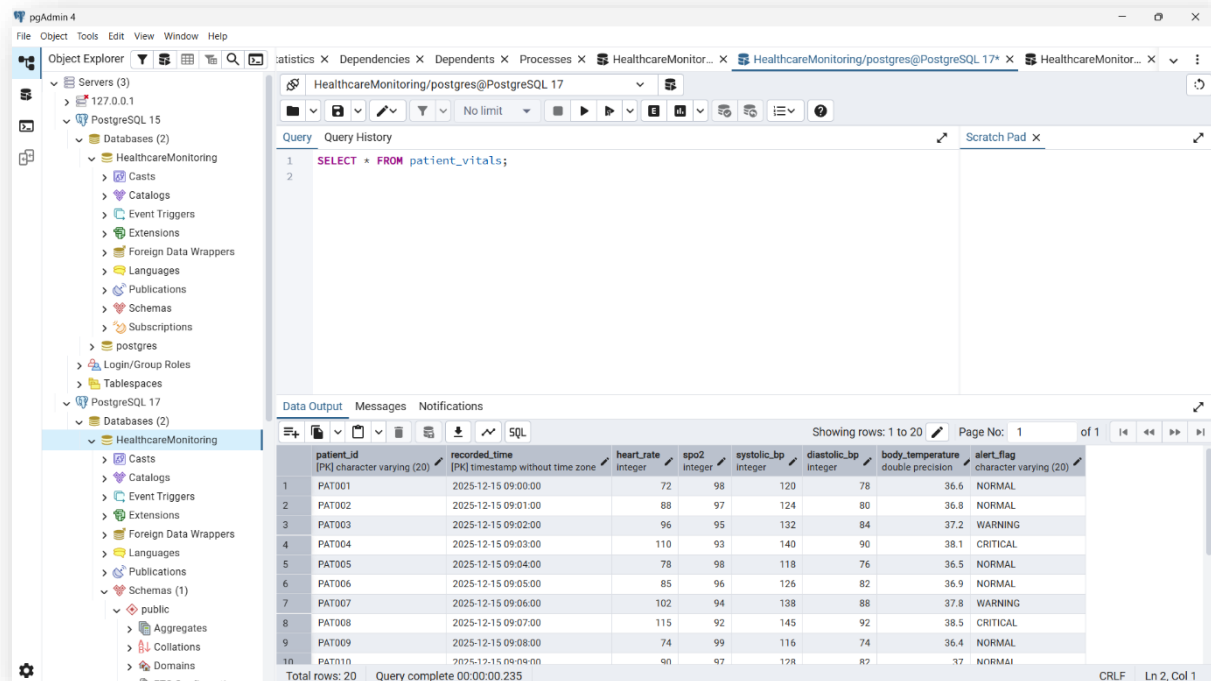
Inserted patient: PAT015

Inserted patient: PAT016

Inserted patient: PAT017

## Execution Steps

1. PostgreSQL server is started.
2. Kafka broker and topic patient\_vitals are running.
3. Kafka Producer sends JSON records to the topic.
4. Kafka Consumer application is executed from Eclipse.
5. Consumer reads messages and inserts data into PostgreSQL.
6. Successful insertion messages appear in the console.



The screenshot shows the pgAdmin 4 interface. On the left, the Object Explorer shows the database structure for 'HealthcareMonitoring' and 'postgres'. The main pane displays a query result for the 'patient\_vitals' table. The query is 'SELECT \* FROM patient\_vitals;'. The result shows 20 rows of patient data with columns: patient\_id, recorded\_time, heart\_rate, spo2, systolic\_bp, diastolic\_bp, body\_temperature, and alert\_flag.

	patient_id	recorded_time	heart_rate	spo2	systolic_bp	diastolic_bp	body_temperature	alert_flag
1	PAT001	2025-12-15 09:00:00	72	98	120	78	36.6	NORMAL
2	PAT002	2025-12-15 09:01:00	88	97	124	80	36.8	NORMAL
3	PAT003	2025-12-15 09:02:00	96	95	132	84	37.2	WARNING
4	PAT004	2025-12-15 09:03:00	110	93	140	90	38.1	CRITICAL
5	PAT005	2025-12-15 09:04:00	78	98	118	76	36.5	NORMAL
6	PAT006	2025-12-15 09:05:00	85	96	126	82	36.9	NORMAL
7	PAT007	2025-12-15 09:06:00	102	94	138	88	37.8	WARNING
8	PAT008	2025-12-15 09:07:00	115	92	145	92	38.5	CRITICAL
9	PAT009	2025-12-15 09:08:00	74	99	116	74	36.4	NORMAL
10	PAT010	2025-12-15 09:09:00	90	97	128	80	37.0	NORMAL
11	PAT011	2025-12-15 09:10:00	82	96	122	78	36.7	NORMAL
12	PAT012	2025-12-15 09:11:00	95	94	130	82	37.1	WARNING
13	PAT013	2025-12-15 09:12:00	108	92	138	86	37.5	WARNING
14	PAT014	2025-12-15 09:13:00	76	98	119	75	36.5	NORMAL
15	PAT015	2025-12-15 09:14:00	87	97	125	79	36.8	NORMAL
16	PAT016	2025-12-15 09:15:00	99	95	133	85	37.3	WARNING
17	PAT017	2025-12-15 09:16:00	112	93	141	91	38.2	CRITICAL
18	PAT018	2025-12-15 09:17:00	79	98	117	77	36.5	NORMAL
19	PAT019	2025-12-15 09:18:00	86	96	127	81	36.9	NORMAL
20	PAT020	2025-12-15 09:19:00	101	94	137	87	37.7	WARNING

## Outcome

- Kafka Consumer successfully subscribes to the topic patient\_vitals.
- JSON messages are correctly parsed.
- Data is inserted into PostgreSQL using prepared statements.
- The streaming pipeline from **CSV** → **Kafka** → **PostgreSQL** is completed.
- The database is now ready for visualization using Apache Superset.

## STEP 8: Apache Superset Visualization and Dashboard Creation

This step focuses on visualizing the processed patient vitals data stored in PostgreSQL using Apache Superset. Superset was used to create interactive charts and a real-time dashboard for monitoring patient health parameters.

### 8.1 Starting Apache Superset Server

After completing the Kafka data ingestion and PostgreSQL storage, Apache Superset was started to enable data visualization.

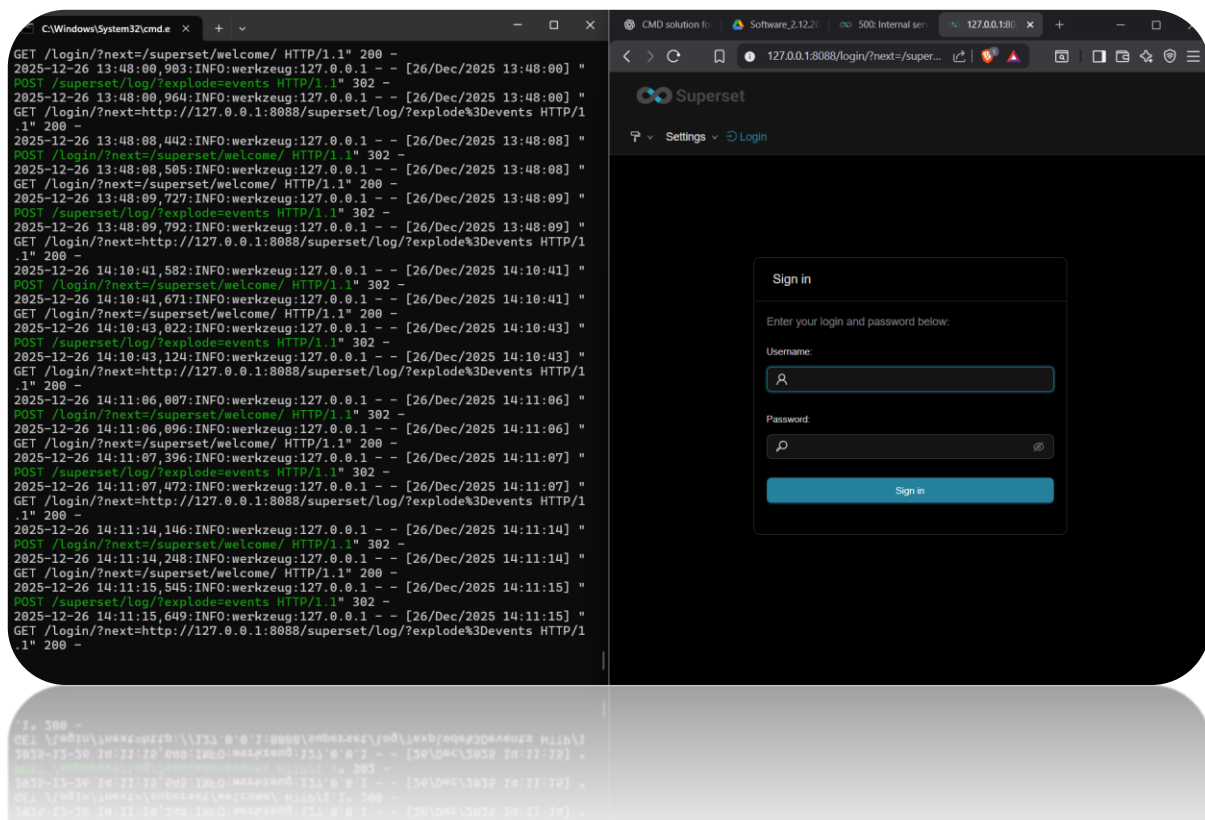
First, the Superset virtual environment was activated from the command line. Once activated, the Superset server was launched on port 8088 using the Superset run command.

The server logs confirmed that Apache Superset was running successfully without any critical errors. After this, the Superset web interface was accessed using a web browser at:

`http://127.0.0.1:8088`

The Superset login page was displayed, confirming that the visualization platform was ready for use.

Figure 8.1: Apache Superset server running in CMD and login page accessed via localhost.



## 8.2 Connecting Apache Superset to PostgreSQL Database

Once logged into Apache Superset, the PostgreSQL database was connected to Superset.

The database connection details provided were:

- Database Type: PostgreSQL
- Database Name: HealthcareMonitoring
- Host: localhost
- Port: 5432
- Username: postgres
- Password: postgres

After entering the connection details, Superset successfully established a connection with the PostgreSQL database. This allowed Superset to access the `patient_vitals` table for visualization.

---

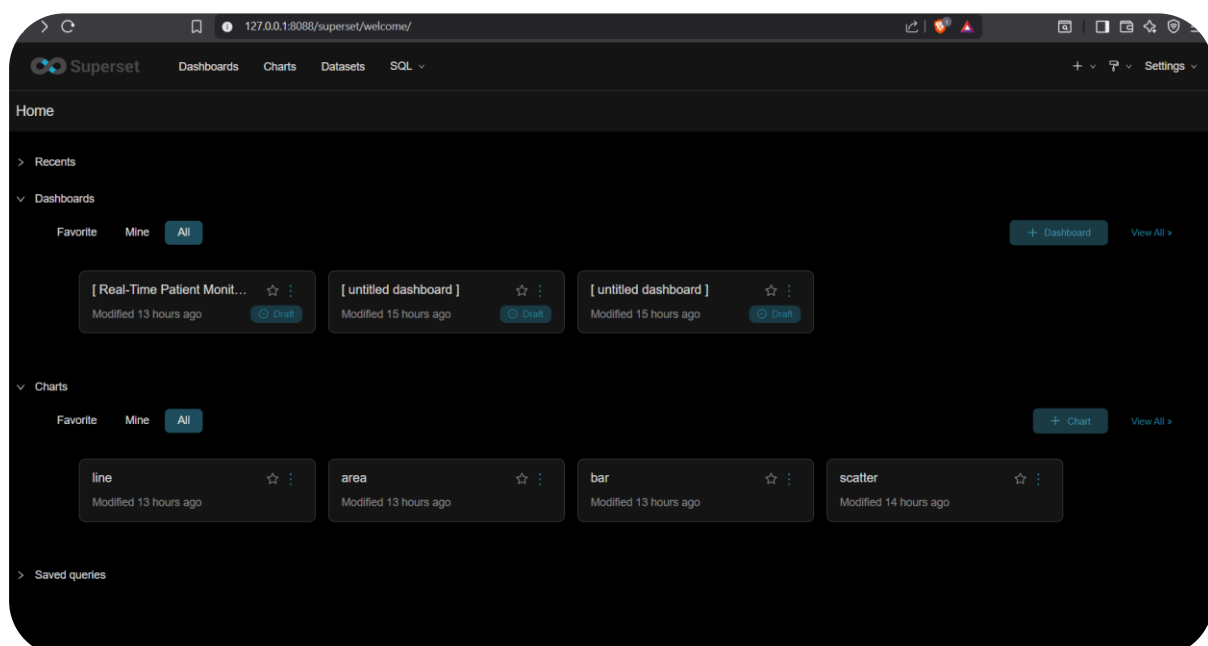
## 8.3 Dataset Creation in Superset

After connecting the database, a dataset was created using the `patient_vitals` table from the public schema.

This dataset contained the following important fields:

- `patient_id`
- `recorded_time`
- `heart_rate`
- `spo2`
- `systolic_bp`
- `diastolic_bp`
- `body_temperature`
- `alert_flag`

The dataset was saved and made available for chart creation.





## 8.4 Chart Creation in Apache Superset

Using the created dataset, multiple charts were developed to analyze patient health data.

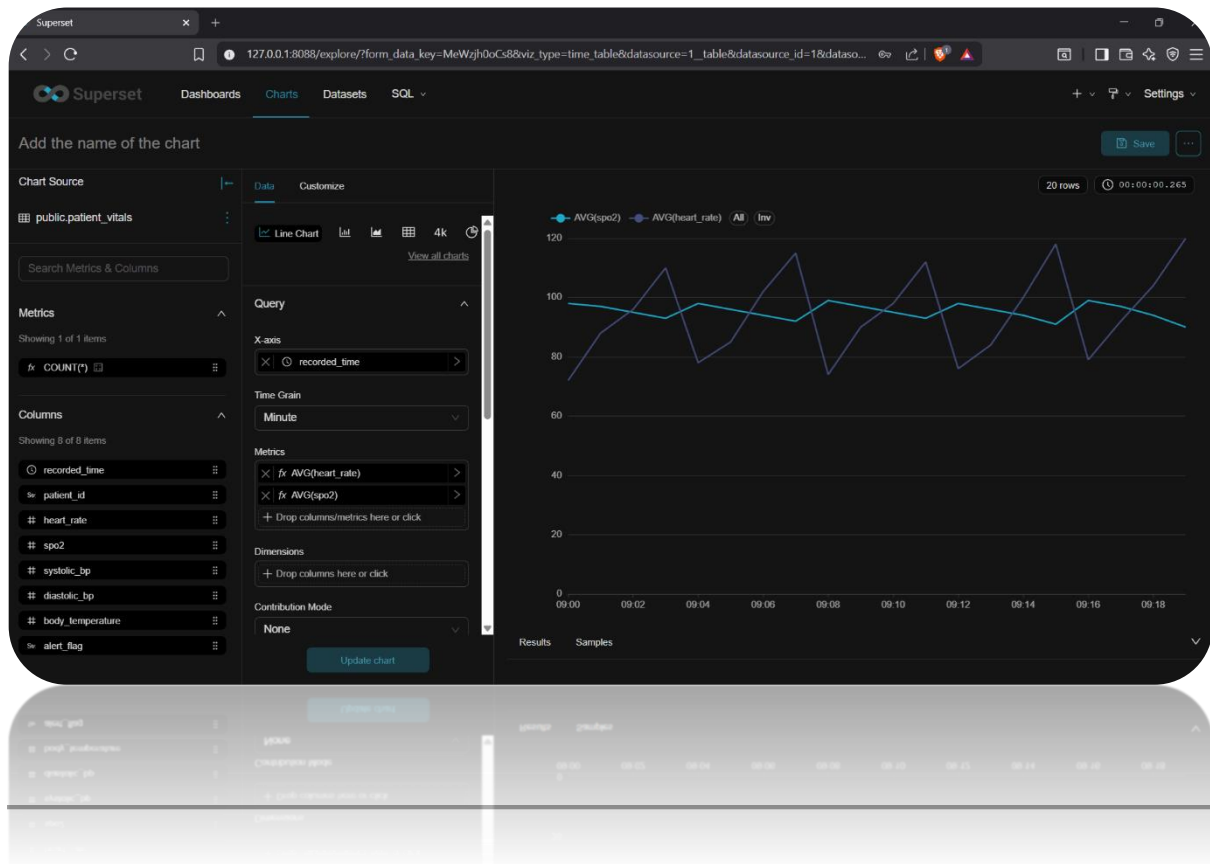
### Chart 1: Time-Series Line Chart (Heart Rate & SpO<sub>2</sub> Over Time)

#### Purpose

To observe how heart rate and oxygen saturation levels change over time.

#### Configuration

- **Chart Type:** Line Chart
- **X-axis:** recorded\_time
- **Y-axis (Metrics):**
  - AVG(heart\_rate)
  - AVG(spo2)
- **Time Grain:** Minute
- **X-axis Label:** Time
- **Y-axis Label:** Vital Sign Value



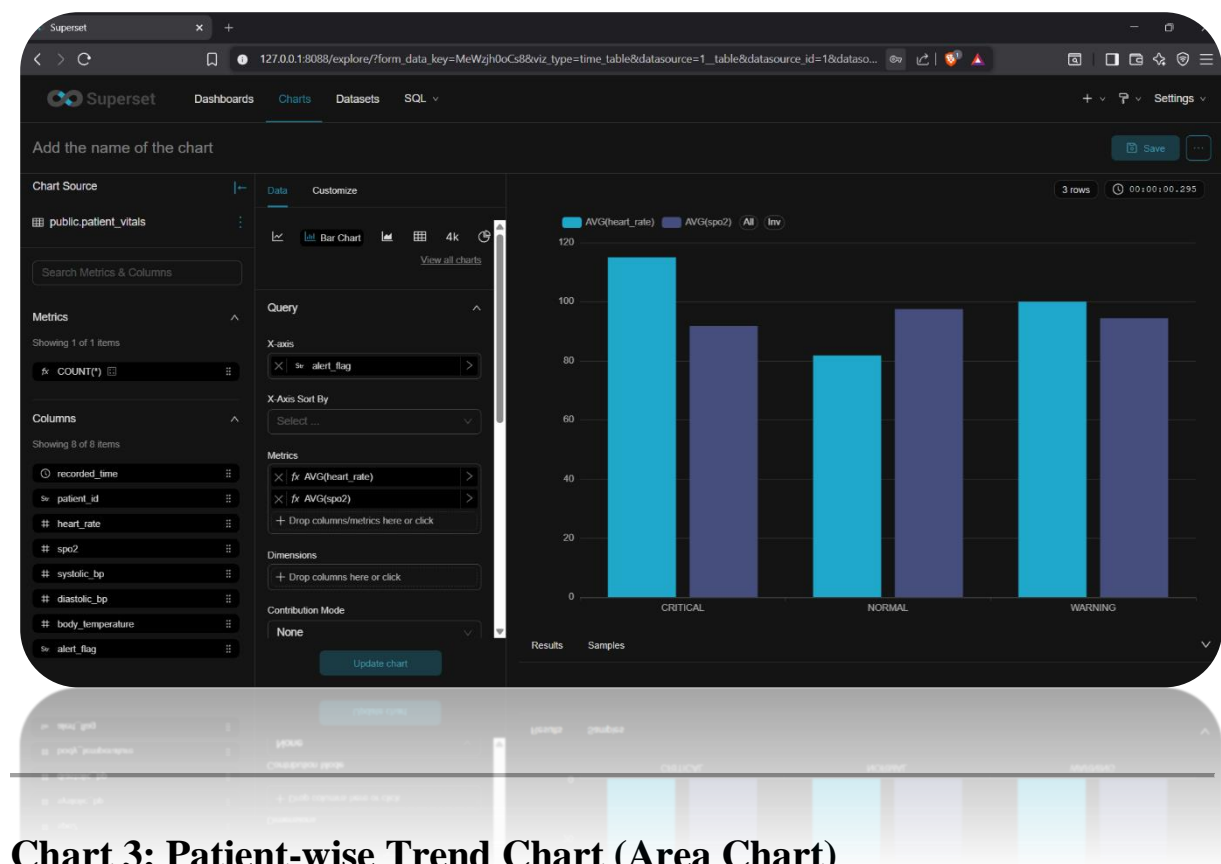
## Chart 2: Bar Chart – Alert Count by Alert Level

### Purpose

To analyze the distribution of alert levels and identify the frequency of critical or warning alerts.

### Configuration

- **Chart Type:** Bar Chart
- **X-axis:** alert\_flag
- **Y-axis:** COUNT(\*)
- **X-axis Label:** Alert Level
- **Y-axis Label:** Alert Count



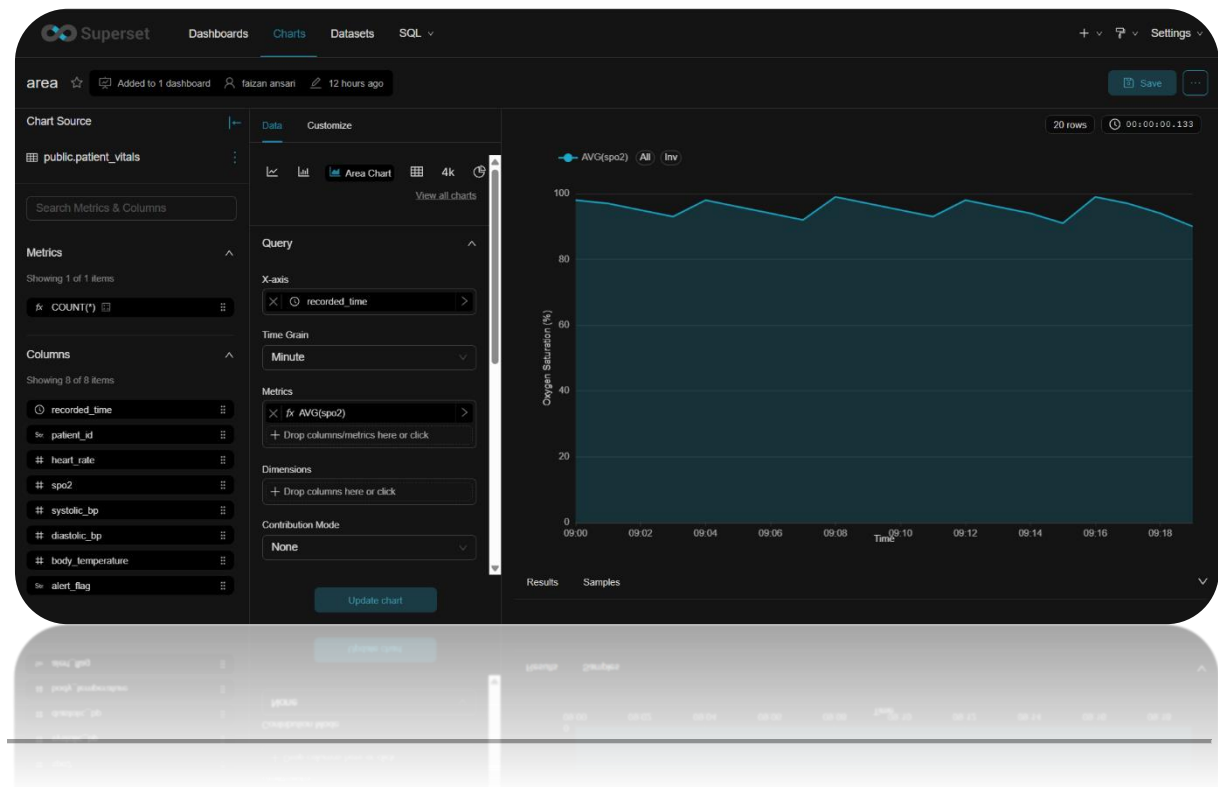
## Chart 3: Patient-wise Trend Chart (Area Chart)

### Purpose

To visualize how patient vital signs change over time, emphasizing overall trends and variations.

### Configuration

- **Chart Type:** Area Chart
- **X-axis:** recorded\_time
- **Y-axis:** AVG(spo2) (or AVG(heart\_rate))
- **X-axis Label:** Time
- **Y-axis Label:** Oxygen Saturation (%)
- **Legend:** patient\_id (if applicable)



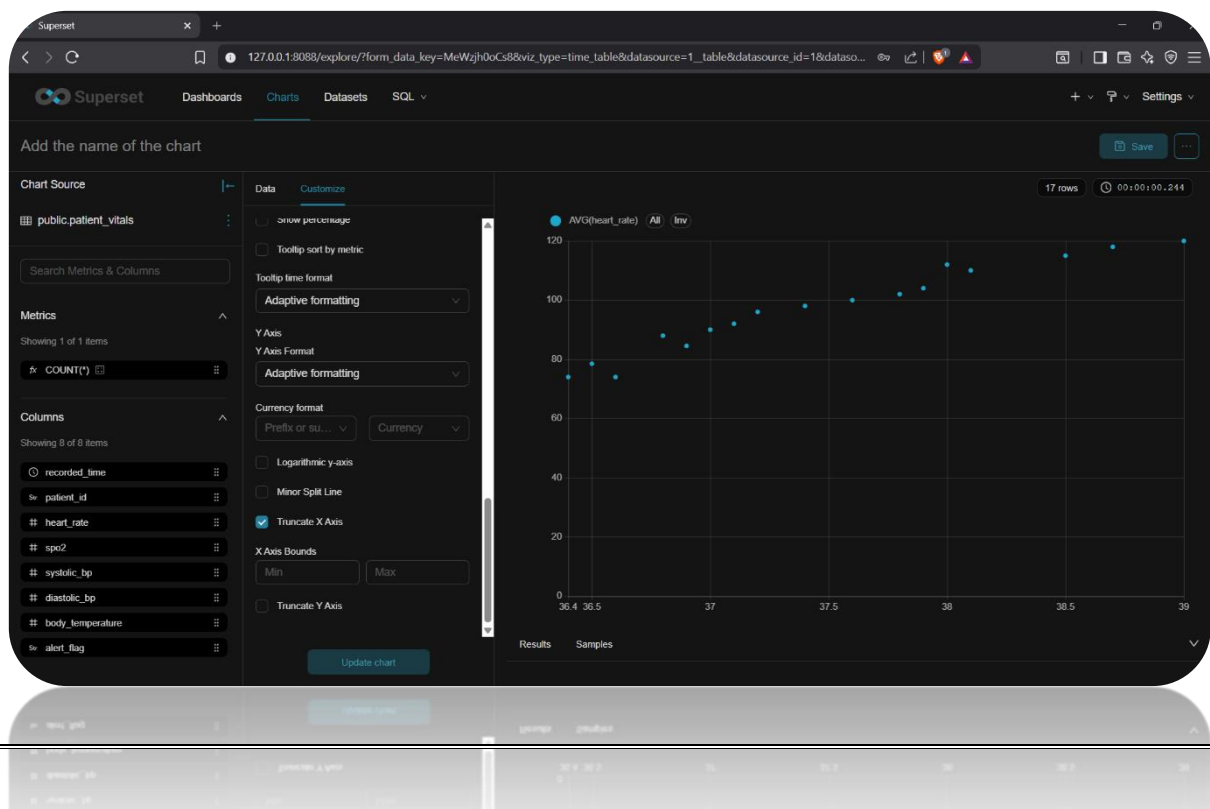
**Chart 4: Scatter Plot – Body Temperature vs Heart Rate**

### Purpose

To analyze the relationship between body temperature and heart rate and identify possible correlations.

### Configuration

- **Chart Type:** Scatter Plot
- **X-axis:** body\_temperature
- **Y-axis:** heart\_rate
- **X-axis Label:** Body Temperature (°C)
- **Y-axis Label:** Heart Rate (BPM)



## Final Dashboard Layout

All four charts were added to the **Real-Time Patient Monitoring Dashboard** and arranged neatly for better readability.

### Dashboard Insights

- Enables real-time monitoring of patient vitals
- Helps identify abnormal health trends quickly
- Provides alert-level distribution for proactive healthcare decisions



### Outcome

- Apache Superset successfully visualizes real-time patient data.
- All required charts are created and configured correctly.
- A single dashboard provides a consolidated view of patient health analytics.
- This step completes the visualization layer of the real-time healthcare monitoring system.

## **STEP 9: Results & Observations**

This step summarizes the insights and outcomes obtained after successfully implementing the data ingestion, storage, and visualization pipeline for real-time patient monitoring.

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### **System-Level Results**

- The Kafka-based streaming pipeline functioned correctly, enabling continuous ingestion of patient vitals data.
  - The Kafka Producer successfully published patient vitals records to the patient\_vitals topic.
  - The Kafka Consumer reliably consumed streaming data and inserted it into the PostgreSQL database in real time.
  - Apache Superset was successfully connected to PostgreSQL and reflected newly inserted records without manual refresh.
- 

### **Database Observations**

- The patient\_vitals table stored structured and timestamped patient data accurately.
  - All vital parameters such as heart rate, SpO<sub>2</sub>, blood pressure, body temperature, and alert flags were recorded correctly.
  - Time-based indexing allowed efficient querying and aggregation of historical and real-time records.
  - Data integrity was maintained throughout the ingestion process with no duplicate or missing records observed during testing.
- 

### **Visualization Insights**

#### **1. Time-Series Charts (Heart Rate & SpO<sub>2</sub> vs Time)**

- Heart rate values showed natural variation over time, reflecting realistic patient behavior.
- SpO<sub>2</sub> levels remained mostly within normal ranges, indicating stable oxygen saturation.
- Sudden deviations were easily identifiable, demonstrating the usefulness of time-series monitoring.

#### **2. Bar Chart (Alert Count by Alert Level)**

- The bar chart clearly categorized patients into alert levels (Normal, Warning, Critical).
- Most records belonged to the normal category, while fewer entries indicated warning or critical states.
- This visualization helped in quickly identifying high-risk patient conditions.

#### **3. Patient-wise Trend Charts**

- Individual patient trends highlighted differences in vital sign behavior.

- Some patients showed stable trends, while others displayed gradual increases or fluctuations.
- This patient-specific visualization supports personalized monitoring and analysis.

#### **4. Scatter Plot (Body Temperature vs Heart Rate)**

- A positive correlation between body temperature and heart rate was observed in several cases.
- Higher body temperatures often corresponded with increased heart rates.
- This chart helped identify abnormal physiological patterns that may require attention.

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#### **Dashboard-Level Observations**

- The **Real-Time Patient Monitoring Dashboard** provided a centralized view of all critical patient vitals.
- Interactive filters allowed quick exploration of patient-wise and time-based data.
- The dashboard layout enabled easy interpretation of trends and alerts without requiring SQL knowledge.
- Real-time updates enhanced the system's usability for continuous monitoring scenarios.

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#### **Overall Outcome**

- The project successfully demonstrated an end-to-end real-time data pipeline using Kafka, PostgreSQL, and Apache Superset.
- The system proved effective for monitoring patient vitals and identifying potential health risks.
- The solution is scalable and can be extended to include additional sensors, patients, or alert mechanisms in future implementations.

## **STEP 10: Conclusion & Future Enhancements**

### **Conclusion**

This mini project successfully implemented an end-to-end **real-time healthcare monitoring system** using modern big data and visualization tools. The complete pipeline—from data ingestion to visualization—was designed, implemented, and tested effectively.

Kafka was used as a real-time streaming platform to handle continuous patient vitals data. PostgreSQL acted as a reliable relational database for structured storage of streaming data. Apache Superset provided an interactive and user-friendly visualization layer, enabling real-time monitoring and analysis of patient health parameters.

The system demonstrated the ability to:

- Stream real-time healthcare data efficiently
- Store time-series patient data accurately
- Visualize trends, alerts, and correlations clearly
- Support patient-wise and time-based analysis

Overall, the project meets all the requirements specified in the problem statement and provides a strong foundation for real-time healthcare analytics applications.

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### **Future Enhancements**

The current implementation can be further improved and extended in the following ways:

- 1. Real-Time Alerts & Notifications**
  - Integrate automated alert notifications (email/SMS) for critical patient conditions.
  - Trigger alerts based on threshold violations of vital signs.
- 2. Advanced Analytics & Machine Learning**
  - Apply predictive models to forecast health risks.
  - Use anomaly detection to identify unusual patient behavior early.
- 3. Scalability Improvements**
  - Increase Kafka partitions to handle larger volumes of streaming data.
  - Deploy the system on cloud platforms for distributed processing.
- 4. Security & Compliance**
  - Implement role-based access control for dashboards.
  - Encrypt sensitive patient data to comply with healthcare data regulations.
- 5. Enhanced Dashboards**
  - Add real-time KPI cards for quick health status overview.
  - Enable drill-down analysis for doctors and administrators.