

AI-Powered SQL for Real-World Data: From Questions to Insights

MSR20

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INTRODUCTION

Accessing and analyzing real-world data (RWD) is essential for healthcare decision-making. However, many professionals—including physicians—lack SQL expertise, and even for those proficient, running queries can be time-consuming. To address this, we developed an AI-powered system that transforms natural language questions into SQL queries, enabling quick and easy database exploration and early insight generation.

OBJECTIVE

To assist non-technical users in exploring databases, identifying trends, and conducting basic feasibility assessments. The system also serves as a supportive tool that enhances understanding of database content and promotes early-stage insight generation. Technically skilled users can benefit as well by quickly generating and testing simple queries, improving efficiency.

METHOD

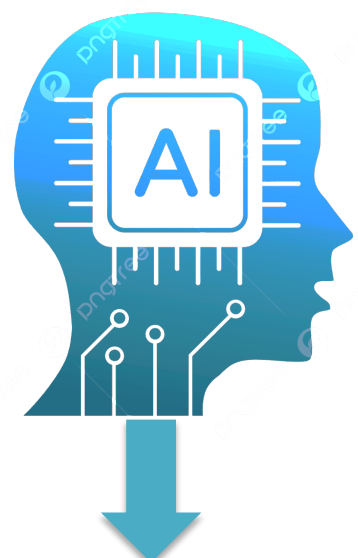
A Large Language Model (LLM) was fine-tuned for SQL generation, enabling it to interpret domain-specific natural language questions and produce accurate, executable SQL queries with visual outputs. The model was trained on curated SQL examples to ensure functionality, enhance its capabilities, and execute effectively on a synthetic database.

TECHNICAL SOLUTION

1. User enters a question in natural language



2. LLM translates this question to SQL query

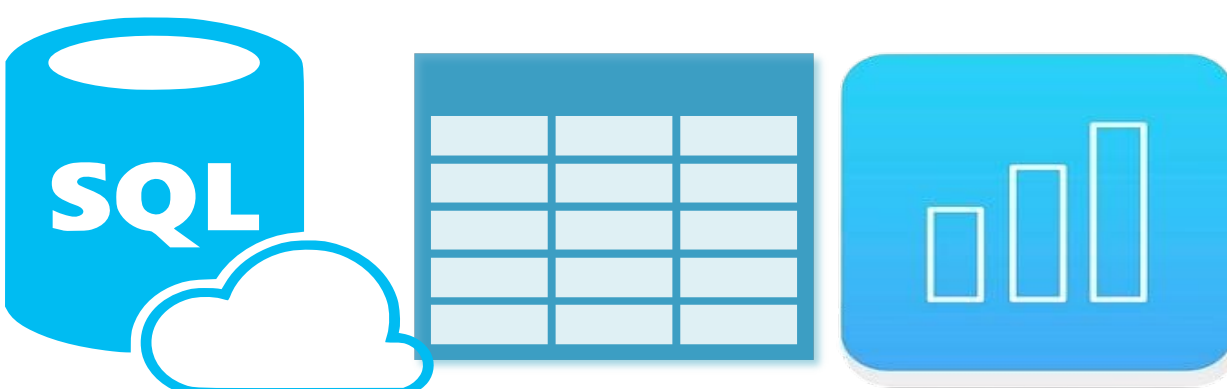


3. Execute the SQL query on the Database



4. The system executes the query and displays:

the SQL code tabular output graphs



USER INTERFACE DEMONSTRATION

Enter your question:

For all the patients with Anemia who are not currently dead I want you to give me the total claim ammount at daily level from 2024 until 2025

☒ Show Chart

Select chart type:

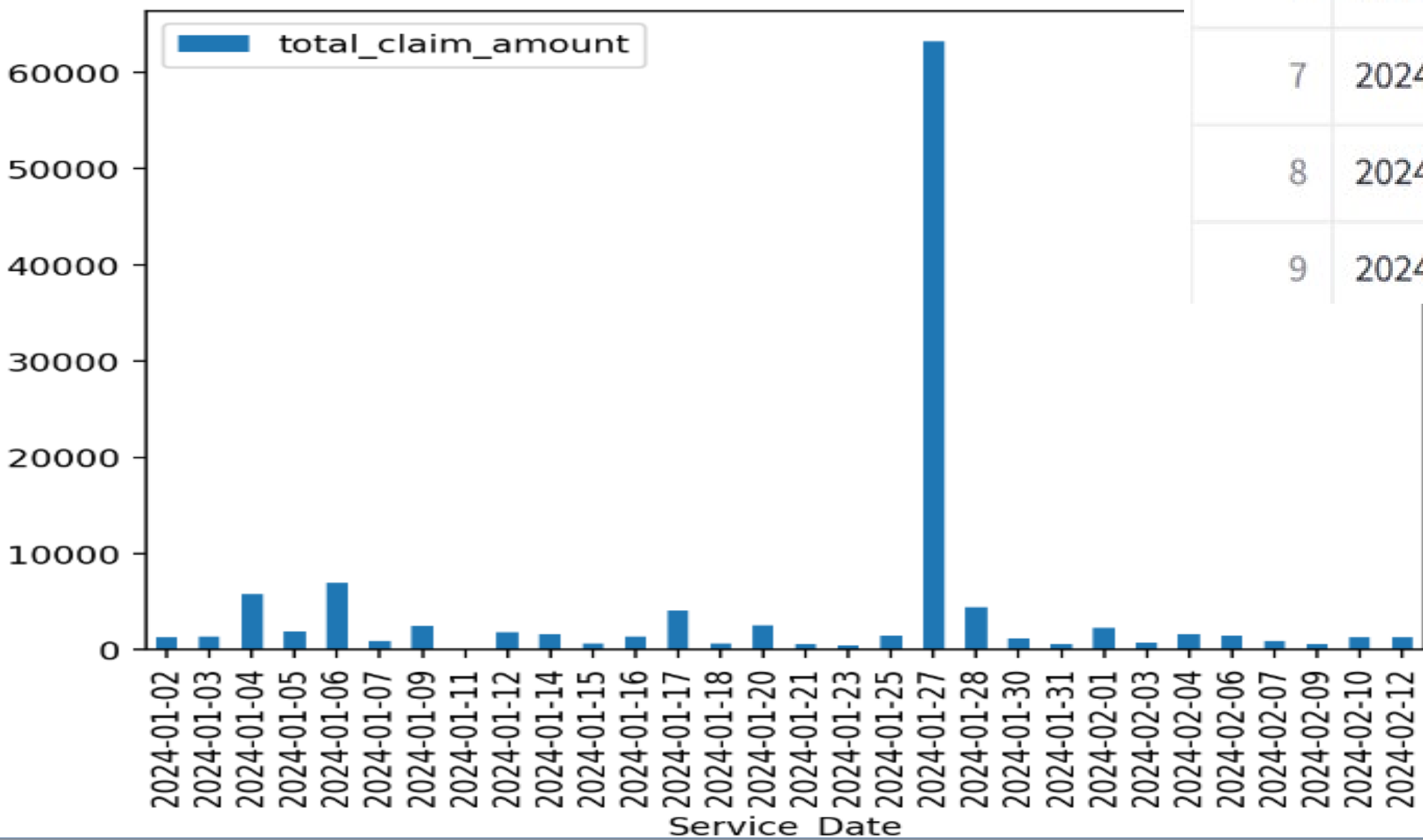
Bar Chart

OUTPUT DEMONSTRATION

```
SELECT      DATE(claims.SERVICEDATE) AS Service_Date
            ,SUM(claims_transactions.AMOUNT) AS total_claim_amount
FROM        patients
INNER JOIN  conditions
            ON patients.Id = conditions.PATIENT
            AND conditions.DESRIPTION LIKE '%Anemia%'
INNER JOIN  claims
            ON patients.Id = claims.PATIENTID
            AND claims.SERVICEDATE >= '2024-01-01'
            AND claims.SERVICEDATE <= '2025-12-31'
INNER JOIN  claims_transactions
            ON claims.Id = claims_transactions.CLAIMID
WHERE       patients.DEATHDATE IS NULL
GROUP BY    Service_Date
ORDER BY    Service_Date
LIMIT 30;
```

Result:

| | Service_Date | total_claim_amount |
|---|--------------|--------------------|
| 0 | 2024-01-02 | 1,266.65 |
| 1 | 2024-01-03 | 1,372.47 |
| 2 | 2024-01-04 | 5,797.14 |
| 3 | 2024-01-05 | 1,902.54 |
| 4 | 2024-01-06 | 6,961.86 |
| 5 | 2024-01-07 | 936.9 |
| 6 | 2024-01-09 | 2,443.07 |
| 7 | 2024-01-11 | 171.1 |
| 8 | 2024-01-12 | 1,820.24 |
| 9 | 2024-01-14 | 1,653.49 |



RESULTS

This system serves as a supportive tool that enhances understanding of database content, facilitates trend identification, and enables early-stage insight generation, thereby streamlining healthcare decision-making. By combining AI-driven query generation with intuitive visual outputs, it allows both non-technical and technical users to efficiently extract meaningful insights. The tool has been trained to handle and execute more complex queries, enabling technically skilled users to quickly generate and test advanced queries, further improving workflow efficiency.