

Project Title: AI Agent for SQL Query Execution

Introduction

Organizations store vast amounts of data in relational databases. However, not all users are proficient in **SQL**. This project aims to build an **AI-powered SQL agent** that allows users to input queries in natural language, which are then converted into **SQL queries**, executed on a **PostgreSQL** database, and displayed in a structured format.

This project leverages **Google Gemini AI** for natural language processing (NLP), **Flask** for API development, **PostgreSQL** as the database, and **Streamlit** for the user interface.

Problem Statement

Many organizations require seamless interaction with databases for **data retrieval and insights**. However, non-technical users struggle with writing SQL queries. This AI agent will bridge the gap by enabling:

- Conversion of **natural language queries** to SQL
- Execution of queries on a **PostgreSQL** database
- Returning results in **JSON** or tabular format

Technologies & Libraries Used

1. Programming Language:

- **Python**

2. Database:

- **PostgreSQL (Primary)**
- **MySQL (Alternative)**

3. AI Model for SQL Conversion:

- **Google Gemini API** (via google-generativeai package)

4. Backend Framework:

- **Flask**

5. Frontend UI (Optional):

- **Streamlit**

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Dependencies & Libraries:

- Flask → For backend API
- psycopg2 → For PostgreSQL database connection
- mysql-connector-python → For MySQL database connection
- google-generativeai → For AI-based SQL conversion
- pandas → For data handling in Streamlit
- re → For cleaning SQL queries
- json → For formatting query results
- decimal → For handling numeric values
- datetime → For date handling

Installation Commands:

`pip install Flask psycopg2 mysql-connector-python google-generativeai pandas`

Project Architecture

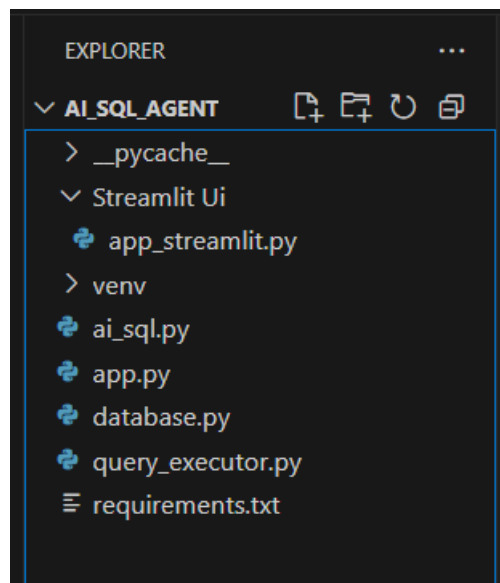
Backend (Flask API):

- database.py: Handles database connection and query execution.
- ai_sql.py: Converts natural language queries to SQL using Google Gemini API.
- query_executor.py: Processes AI-generated SQL and executes it.
- app.py: Main Flask application that serves the API.

Frontend (Streamlit UI):

- streamlit_app.py: Provides a simple UI for user input and displays query results.

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Code Explanation

1.database.py

This code connects to a PostgreSQL database using provided credentials. It returns a connection object on success, or None if the connection fails.

```
database.py ×
database.py > connect_to_db
1  import psycopg2
2
3  # Database connection details
4  DB_NAME = "company_data"
5  DB_USER = "postgres"
6  DB_PASSWORD = "newpassword"
7  DB_HOST = "localhost"
8  DB_PORT = "5432" # Default PostgreSQL port
9
10 def connect_to_db():
11     try:
12         conn = psycopg2.connect(
13             dbname=DB_NAME,
14             user=DB_USER,
15             password=DB_PASSWORD,
16             host=DB_HOST,
17             port=DB_PORT
18         )
19         print("✅ Connected to the database successfully!")
20         return conn
21     except Exception as e:
22         print(f"❌ Error connecting to the database: {e}")
23         return None
24
```

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2.AI Model (ai_sql.py)

- Converts user input to SQL using Google Gemini AI.
- Extracts the SQL query from the response.

```
ai_sql.py X
ai_sql.py > extract_sql
1 import google.generativeai as genai
2 import re
3
4 # Function to get API key from user input
5 API_KEY = input("Enter your Google API Key: ").strip()
6 genai.configure(api_key=API_KEY)
7
8 def extract_sql(text):
9     """
10     Extracts only the SQL query from AI-generated text and removes extra explanations.
11     """
12     sql_query = re.findall(r"SELECT.*?;", text, re.DOTALL) # Extract SQL query only
13     return sql_query[0] if sql_query else "Error: No valid SQL found."
14
15 def convert_to_sql(user_query):
16     """Function to convert natural language to SQL using Gemini AI."""
17     model = genai.GenerativeModel("gemini-1.5-pro") # Use correct model
18     response = model.generate_content(f"Convert this to SQL and return only the query:\n{user_query}")
19
20     if response and hasattr(response, "text"):
21         return extract_sql(response.text.strip()) # Extract only the SQL query
22     else:
23         return "Error: Failed to generate SQL."
24
```

3. Query Execution (query_executor.py)

```
query_executor.py X
query_executor.py > execute_query
1 from database import connect_to_db
2 from ai_sql import convert_to_sql
3 import re
4 import json
5 from decimal import Decimal
6 from datetime import date # Import date module
7
8 def clean_sql_query(ai_response):
9     """Cleans AI-generated SQL query by removing unnecessary characters."""
10     sql_query = re.sub(r"```sql|```", "", ai_response, flags=re.MULTILINE) # Remove
11     sql_query = re.sub(r"--.*", "", sql_query) # Remove inline comments
12     sql_query = sql_query.strip().split("\n\n")[0] # Keep only first SQL statement
13     return sql_query.strip()
14
15 def convert_to_serializable(data):
16     """Recursively converts Decimal and date values to JSON serializable types."""
17     if isinstance(data, list):
18         return [convert_to_serializable(item) for item in data]
19     elif isinstance(data, dict):
20         return {key: convert_to_serializable(value) for key, value in data.items()}
21     elif isinstance(data, Decimal):
22         return float(data) # Convert Decimal to float
23     elif isinstance(data, date):
24         return data.isoformat() # Convert date to string (YYYY-MM-DD)
25     return data
```

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```
query_executor.py X
query_executor.py > execute_query
27 def execute_query(user_query):
28     """Executes AI-generated SQL query and returns results in JSON format."""
29     conn = connect_to_db()
30     if conn is None:
31         return json.dumps({"error": "Database connection failed."})
32
33     ai_response = convert_to_sql(user_query) # AI generates SQL query
34     sql_query = clean_sql_query(ai_response) # Clean the SQL query
35
36     print(f"\n✅ Cleaned SQL Query:\n{sql_query}\n") # Debugging output
37
38     try:
39         cursor = conn.cursor()
40         cursor.execute(sql_query)
41         columns = [desc[0] for desc in cursor.description] # Get column names
42         results = cursor.fetchall()
43
44         conn.close()
45
46         # Convert results to JSON format
47         json_output = json.dumps(
48             convert_to_serializable([dict(zip(columns, row)) for row in results]),
49             indent=4
50         )
51         return json_output
52
```

- Executes AI-generated SQL queries on the database.
- Handles errors and returns structured JSON output.

4. User Interface (Streamlit UI)

Provides a simple UI for query input and execution.

```
app_streamlit.py X
Streamlit UI > app_streamlit.py > ...
1 import streamlit as st
2 from database import connect_to_db
3 from ai_sql import convert_to_sql
4 import pandas as pd
5
6 # Streamlit UI
7 st.title("AI-Powered SQL Agent")
8 st.write("Enter your natural language query below and get the SQL query along with results.")
9
10 # User input
11 user_query = st.text_input("Enter your query:")
12
13 if st.button("Generate SQL and Execute"):
14     if user_query:
15         # Convert to SQL using AI
16         sql_query = convert_to_sql(user_query)
17
18         st.subheader("Generated SQL Query")
19         st.code(sql_query, language="sql")
20
```

5. API Endpoint (app.py - Flask API)

```
app.py > ...
1  from flask import Flask, request, jsonify
2  from query_executor import execute_query # Import query execution function
3
4  app = Flask(__name__)
5
6
7  @app.route('/')
8  def home():
9      return "Welcome to the AI SQL Agent! Use the '/query' endpoint to interact."
10
11 @app.route('/query', methods=['POST'])
12 def query_database():
13     """Process natural language queries and return SQL execution results."""
14     data = request.get_json()
15     user_query = data.get("query", "")
16
17     if not user_query:
18         return jsonify({"error": "No query provided"}), 400
19
20     result = execute_query(user_query)
21     return jsonify({"result": result})
22
23 if __name__ == "__main__":
24     app.run(debug=True)
```

Flask API to handle user queries and return results.

6. Testing

a) Run the API:

Using CURL:

```
curl -X POST http://127.0.0.1:5000/query -H "Content-Type: application/json" -d '{"query": "Show all employees in Mumbai."}'
```

Using Postman:

1. Open Postman
2. Set request type to POST
3. URL: http://127.0.0.1:5000/query
4. Go to Body → raw → JSON
5. Input:

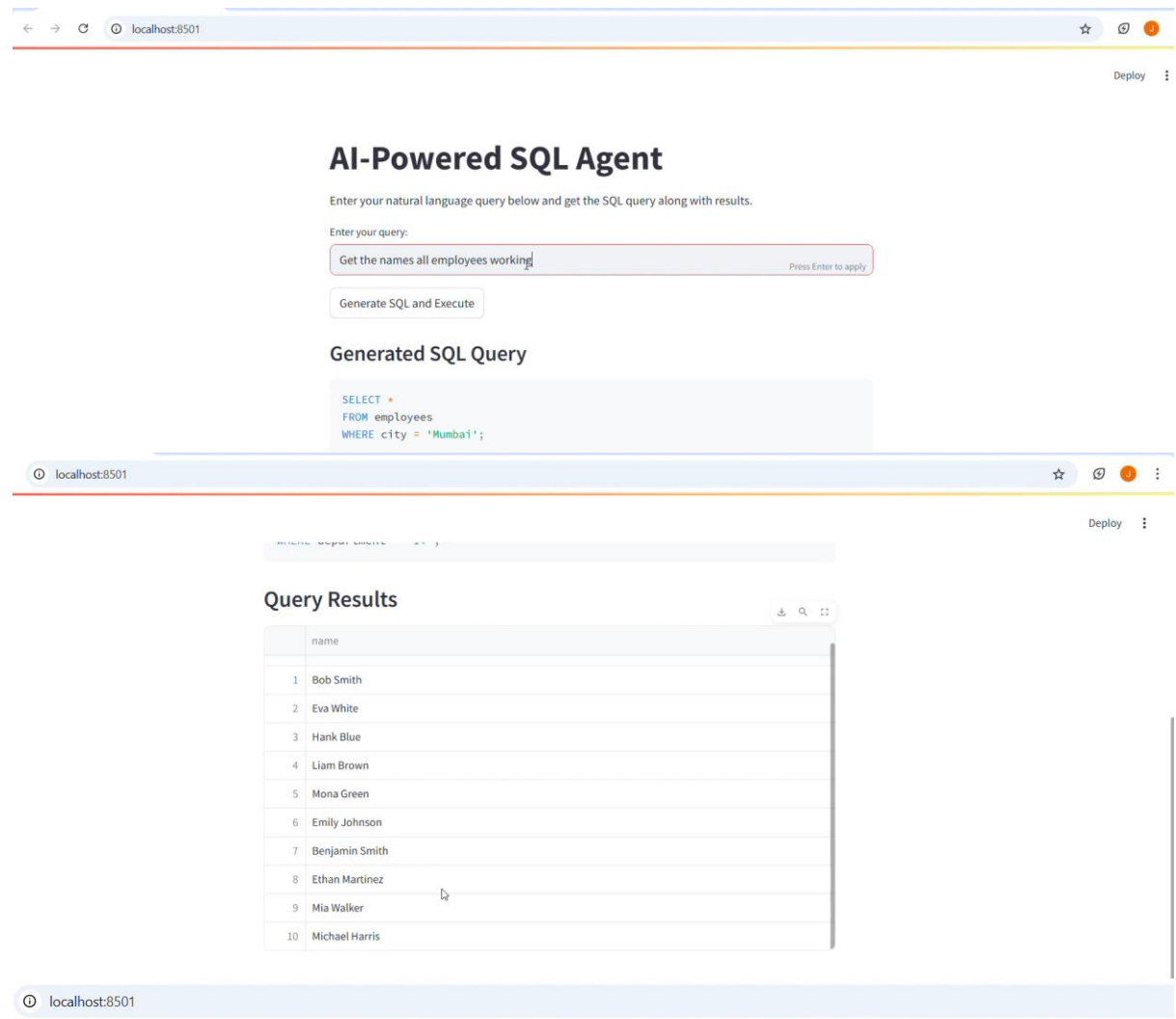
```
{
    "query": "List all employees in the IT department."}
```

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b) Run Streamlit UI:

streamlit run app_streamlit.py

Output:-



localhost:8501

Deploy

AI-Powered SQL Agent

Enter your natural language query below and get the SQL query along with results.

Enter your query:

Get the names all employees working Press Enter to apply

Generate SQL and Execute

Generated SQL Query

```
SELECT *
FROM employees
WHERE city = 'Mumbai';
```

localhost:8501

Deploy

Query Results

	name
1	Bob Smith
2	Eva White
3	Hank Blue
4	Liam Brown
5	Mona Green
6	Emily Johnson
7	Benjamin Smith
8	Ethan Martinez
9	Mia Walker
10	Michael Harris

localhost:8501

Enter your natural language query below and get the SQL query along with results.

Enter your query:

Find employees earning more than 80000

Generate SQL and Execute

Generated SQL Query

```
SELECT *
FROM employees
WHERE salary > 80000;
```

AI-Powered SQL Agent

Enter your natural language query below and get the SQL query along with results.

Enter your query:

Find employees earning more than 80000

Press Enter to apply

Generate SQL and Execute

Generated SQL Query

```
SELECT *
FROM employees
WHERE salary > 80000;
```

Query Results

⬇ 🔍 🗖

	id	name	position	department	salary	join_date	city
6	18	Emma White	Data Engineer	Data Science	98000	2023-03-20	San Francisco
7	21	Emily Johnson	Software Architect	IT	120000	2019-06-25	New York
8	22	Daniel Brown	Product Manager	Product	90000	2021-11-12	San Francisco
9	24	James Wilson	Finance Manager	Finance	87000	2020-04-17	Austin
10	25	Benjamin Smith	Cloud Engineer	IT	95000	2021-09-28	Mumbai

7. Conclusion

- Successfully developed an AI-powered SQL Agent.
- Allows users to interact with databases without SQL knowledge.
- Implements Google Gemini AI for natural language to SQL conversion.
- Provides Flask API for backend and Streamlit UI for user interaction.
- Ensures security with query validation and SQL injection protection.