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# Hands-on Lab: Generative AI for Querying Databases

**Estimated Effort: 30 mins** 

### Introduction

Processed data saved in a database table can be accessed, based on your requirements, using queries. Because queries are an essential part of any data professional's workflow, writing efficient queries is a necessary skillset. In this lab, you will learn how you can leverage Generative AI platforms to create optimized queries for your data, provided you can give the model enough context.

## Objective(s)

By the end of this lab, you'll be able to prompt a Generative AI model to create efficient queries for your data set.

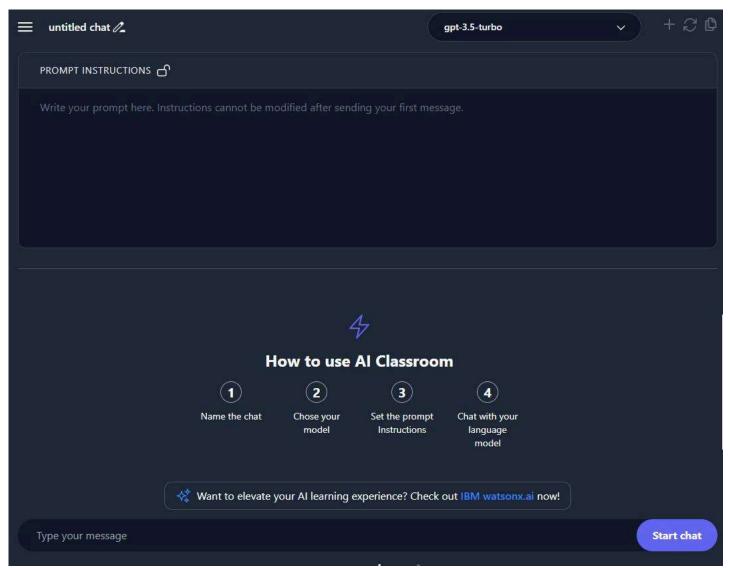
### The data set

For the purpose of this lab, you are making use of the Heart Disease Data set from the UCI ML library, available publically under the CCA 4.0 International license.

You can download the data set and run the queries generated in this lab using any SQL querying system.

# Giving the context

You might note that there is a section named Prompt Instructions on the Generative AI interface.



If you provide the model with the description of your data, you can generate efficient and readily usable queries for fetching the data based on your requirements.

Paste the following text in the prompt instructions to give the model the appropriate context for the data.

- 2. 2 3. 3 4. 4 5. 5

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```
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                                                                                                                                        about:blank
       9. 9
      10. 10
     12. 12
13. 13
     15. 15
16. 16
     18. 18
19. 19
     21. 21
22. 22
     24. 24
25. 25
      27. 27
       1. We have a Heart Disease prediction dataset with a single table which has the following attributes.

    age - age in years
    gender- gender (1 = male; 0 = female)

       4. 3. cp - chest pain type
5. -- Value 1: typical angina
6. -- Value 2: atypical angina
                         -- Value 3: non-anginal pain
                          -- Value 4: asymptomatic
       8.
       9. 4. trestbps - resting blood pressure (in mm Hg on admission to the hospital)
     10. 5. chol - serum cholestoral in mg/dl (1 = true; 0 = false)
11. 6. fbs - (fasting blood sugar > 120 mg/dl) (1 = true; 0 = false)
12. 7. restecg - resting electrocardiographic results
13. -- Value 0: normal
      14. -- Value 1: having ST-T wave abnormality (T wave inversions and/or ST elevation or depression of > 0.05 mV)
15. -- Value 2: showing probable or definite left ventricular hypertrophy by Estes' criteria
16. 8. thalach - maximum heart rate achieved
      17. 9. exang - exercise induced angina (1 = yes; \theta = no) 18. 10. oldpeak - ST depression induced by exercise relative to rest
      19. 11. slope - the slope of the peak exercise ST segment
20. -- Value 1: upsloping
21. -- Value 2: flat
                          -- Value 3: downsloping
     23. 12. ca - number of major vessels (0-3) colored by flourosopy
24. 13. thal - 3 = normal; 6 = fixed defect; 7 = reversable defect
      25. 14. num (the predicted attribute) - diagnosis of heart disease (angiographic disease status)
                          -- Value 0: < 50% diameter narrowing
-- Value 1: > 50% diameter narrowing
```

The model will now have enough context to generate SQL queries for your prompts.

Note: The prompt instructions are editable only until the first prompt is executed. Thereafter, you cannot change the instructions. If a mistake occurs and you want to start over, you can generate a fresh chat and add the correct prompt instructions before generating the SQL queries.

# **Prompts for Data Querying**

## Age Distribution

Consider the following prompt.

1.

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1. Write an SQL query to find the minimum, maximum, and average age of patients in the dataset.

Copied!

You can expect to the see the following query in response.

```
1. 1
2. 2
3. 3
4. 4
5. 5
6. 6

1. SELECT
2. MIN(age) AS min_age,
3. MAX(age) AS max_age,
4. AVG(age) AS avg_age
5. FROM
6. heart_disease_prediction_dataset;

Copied!
```

# **Gender Analysis**

Consider the following prompt.

1. 1

1. Write and SQL query to count the number of male and female patients in the dataset.

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You can expect to the see the following query in response.

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```
1. 1
2. 2
3. 3
4. 4
5. 5
6. 6
7. 7

1. SELECT
2. gender,
3. cOUNT(*) AS patient_count
4. FROM
5. heart_disease_prediction_dataset
6. GROUP BY
7. gender;

Copied!
```

# **Chest Pain Type Frequency**

Consider the following prompt.

1. 1

1. Write an SQL query to determine the frequency of each type of chest pain (typical angina, atypical angina, non-anginal pain, asymptomatic) among

```
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```

You can expect the following query to be generated.

```
1. 1
2. 2
3. 3
4. 4
5. 5
6. 6
7. 7

1. SELECT
2. cp,
3. cOUNT(*) AS pain_frequency
4. FROM
5. heart_disease_prediction_dataset
6. GROUP BY
7. cp;

Copied!
```

## Age Group Analysis and Target Variable

Consider the following prompt.

1. 1

1. Write an SQL query to investigate the distribution of the target variable (presence or absence of heart disease) within different age groups (e.

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You can expect the following output.

```
1. 1
2. 2
3. 3
  4. 4
5. 5
6. 6
7. 7
8. 8
9. 9
 10. 10
11. 11
 13. 13
14. 14
 16. 16
 17. 17
  1. SELECT
  2.
  3.
4.
                WHEN age BETWEEN 20 AND 30 THEN '20-30' WHEN age BETWEEN 31 AND 40 THEN '31-40'
  5.
                WHEN age BETWEEN 41 AND 50 THEN '41-50'
  6.
7.
8.
                WHEN age BETWEEN 51 AND 60 THEN '51-60'
                WHEN age BETWEEN 61 AND 70 THEN '61-70' ELSE 'Above 70'
  9.
            SUM(CASE WHEN num = 1 THEN 1 ELSE 0 END) AS heart_disease_count,
 10.
 11.
           SUM(CASE WHEN num = 0 THEN 1 ELSE 0 END) AS no_heart_disease_count
 12. FROM
           heart_disease_prediction_dataset
 13.
 14. GROUP BY
      age_group
ORDER BY
 15.
 16.
           age_group;
Copied!
```

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# **Practice Prompts**

Try to generate queries for the data set for the following prompts:

- 1. Cholesterol Range:
- 1 1
- 1. Find the range of cholesterol levels among patients (minimum, maximum).

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- 2. Age Range and Gender Analysis:
- 1. 1
- 1. Determine the age range (youngest and oldest) for male and female patients separately.

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- 3. Age Group Analysis and Target Variable:
- 1. 1
- 1. Investigate the distribution of the target variable (presence or absence of heart disease) within different age groups (e.g., 20-30, 30-40, etc.

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- 4. Maximum Heart Rate by Age Group:
- 1. 1
- 1. Find the maximum heart rate achieved during exercise for different age groups (e.g., 30-40, 40-50, etc.).

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- 5. Percentage of Patients with High Blood Sugar:
- 1. 1
- 1. Calculate the percentage of patients with fasting blood sugar greater than 120~mg/dl.

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- 6. Ratio of Patients with Resting Electrocardiographic Abnormality:
- 1. 3
- 1. Find the ratio of patients with abnormal resting electrocardiographic results to those with normal results.

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- 7. Number of Patients with Reversible Thalassemia:
- 1. 3
- 1. Count the number of patients with reversible thalassemia detected by thallium stress testing.

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- 8. Average Age of Patients with Chest Pain:
- 1. 1
- 1. Calculate the average age of patients who experienced chest pain during diagnosis.

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- 9. Distribution of Patients by Number of Major Vessels:
- 1. 1
- 1. Investigate the distribution of patients based on the number of major vessels colored by fluoroscopy  $(\theta-3)$ .

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# **Conclusion**

Congratulations on your successful completion of this lab!

You should now be able to use Generative AI to create efficient queries for retrieving relevant information from a database. Please note, that you need to provide the model with a detailed description of the attributes as context, to generate efficient and ready-to-use prompts.

### Author(s)

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