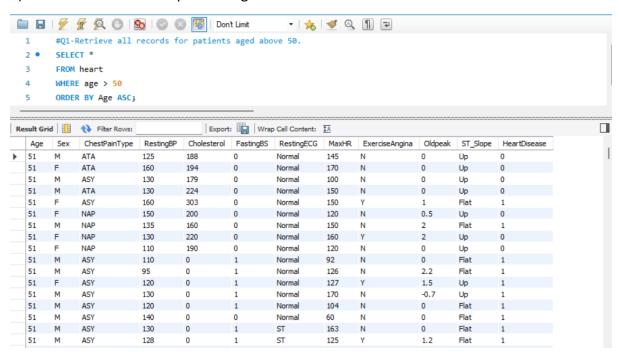
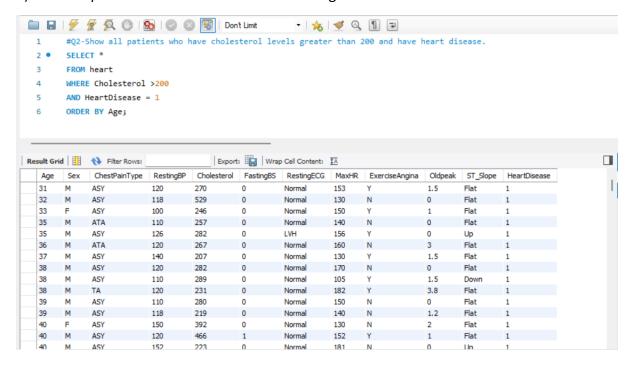
Heart Failure Prediction

SQL Queries:

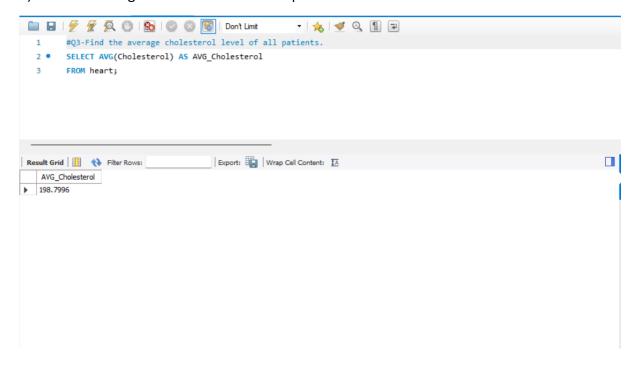
1) Retrieve all records for patients aged above 50.



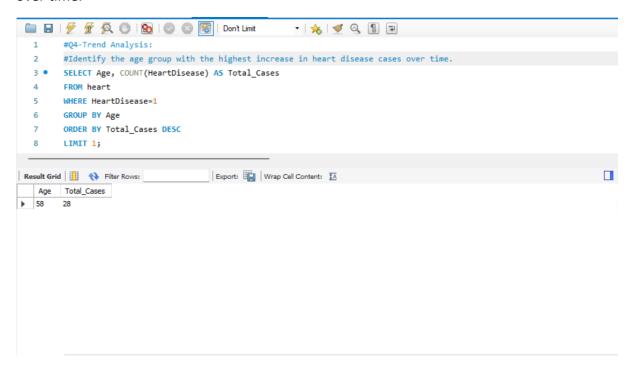
2) Show all patients who have cholesterol levels greater than 200 and have heart disease.



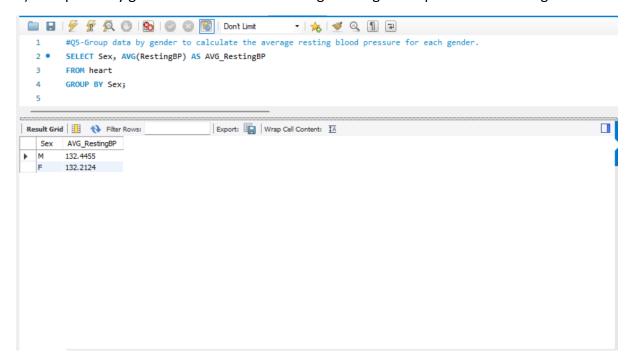
3) Find the average cholesterol level of all patients.



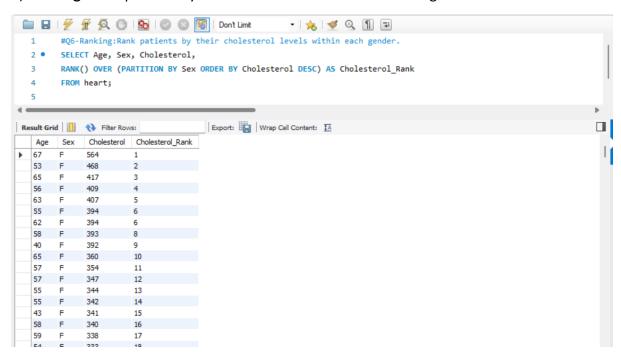
4) **Trend Analysis:** Identify the age group with the highest increase in heart disease cases over time.



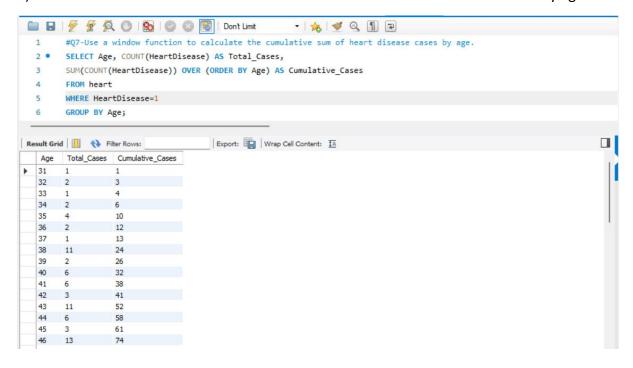
5) Group data by gender to calculate the average resting blood pressure for each gender.



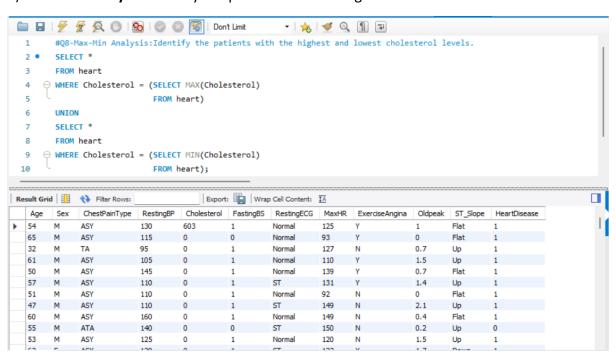
6) Ranking: Rank patients by their cholesterol levels within each gender.



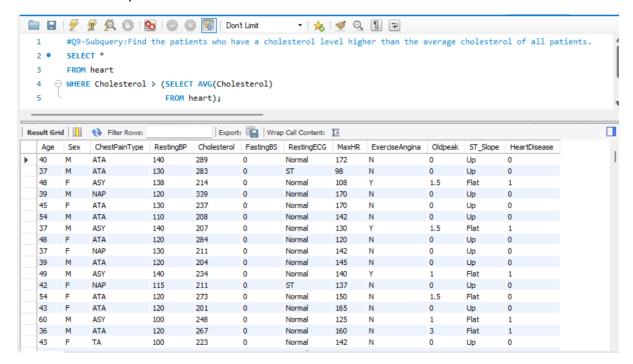
7) Use a window function to calculate the cumulative sum of heart disease cases by age.



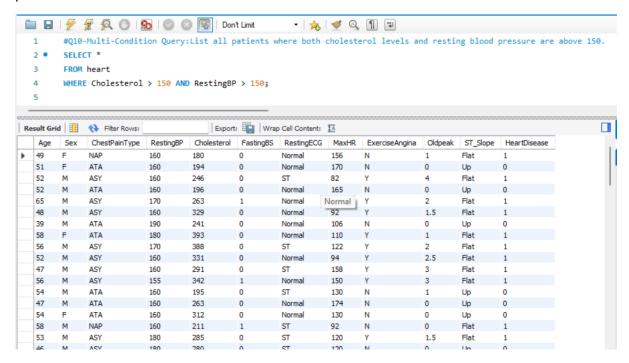
8) Max-Min Analysis: Identify the patients with the highest and lowest cholesterol levels.



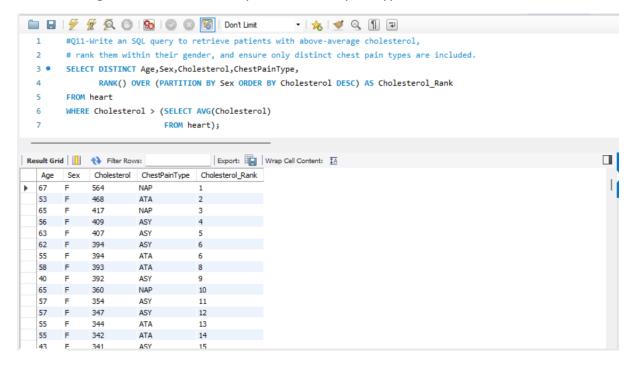
9) **Subquery:** Find the patients who have a cholesterol level higher than the average cholesterol of all patients.



10) **Multi-Condition Query:** List all patients where both cholesterol levels and resting blood pressure are above 150.



11) Write an SQL query to retrieve patients with above-average cholesterol, rank them within their gender, and ensure only distinct chest pain types are included.



Executive Summary

Objective

The primary goal of this project is to develop a predictive model for heart failure using Linear Regression, Logistic Regression and k-NN which includes data preprocessing, exploratory data analysis (EDA), model training, and evaluation.

Data Preprocessing

The dataset, stored as heart.csv, is loaded and examined. The preprocessing steps include:

- Checking data types of columns.
- Determining the number of records.
- Summary statistics (describe()) to understand variable distributions.
- Handling missing values (isnull().sum()).

Exploratory Data Analysis (EDA)

Several visualizations and insights are provided, including:

- Heart Disease by Gender: A countplot is used to analyze the distribution of heart disease across genders.
- Other potential exploratory insights such as bar plot, box plot (for detecting outliers) scatter plot, pie chart are included.

Model Training

The model development process includes:

- 1. Feature Engineering:
 - Data is split into training and testing sets using train_test_split().
 - Standardization and normalization is performed for numerical features.
- 2. Model Selection:
 - Linear Regression, Logistic Regression, k-NN is chosen as the predictive model.
 - o The models are trained on the dataset using from scikit-learn.

3. Performance Metrics:

- The performance is evaluated using:
 - Accuracy Score
 - Precision Score
 - Recall Score
 - Confusion Matrix
 - Regression Summary (dmba.regressionSummary)
 - Adjusted R² Score

Key Findings & Inferences

- The model attempts to predict heart failure based on various health-related features.
- Insights from **EDA** may help identify key risk factors for heart disease.
- Performance metrics indicate how well the models generalizes to unseen data.

Limitations & Future Scope

- Regression Models and k-NN may not be the best model for classification problems like heart disease prediction. A more suitable alternative could be used i.e machine learning algorithms (Random Forest, XGBoost, Neural Networks, etc.).
- Additional feature selection techniques and hyperparameter tuning could improve the model's predictive accuracy.
- Further data augmentation or external datasets could be used to enhance model robustness.