

Final Project

Using SQL and Python

Dataset

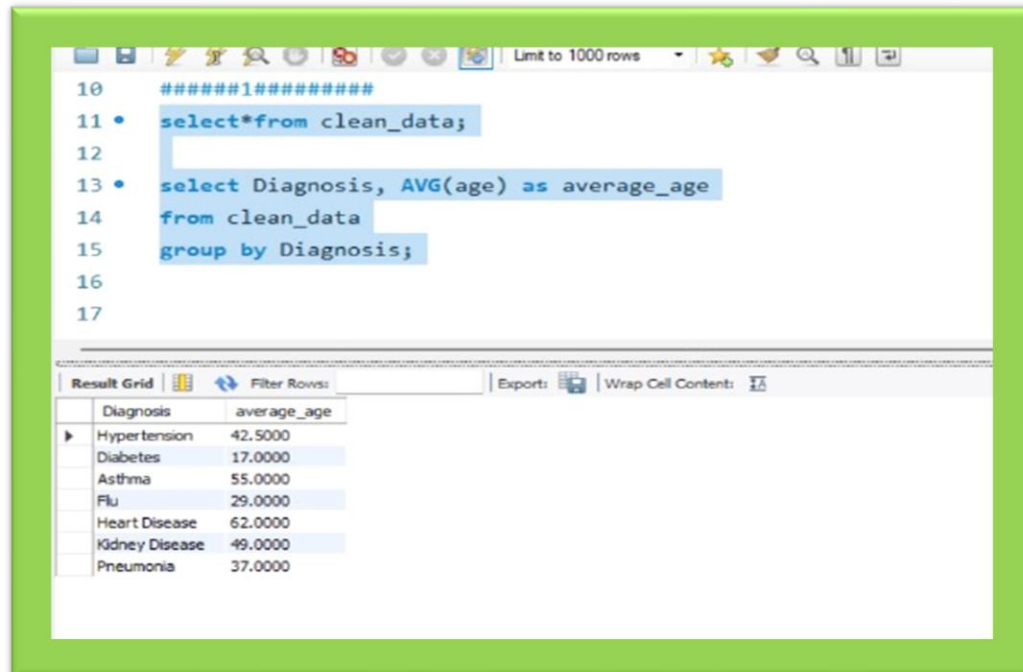
The dataset (hospital_patient_records.csv) contains the following columns:

- **PatientID:** Unique identifier for each patient.
- **Name:** Name of the patient.
- **Age:** Patient's age.
- **Gender:** Male or Female.
- **Diagnosis:** Primary diagnosis of the patient.
- **Medication:** Prescribed medication.
- **AdmissionDate:** Date of patient admission.
- **DischargeDate:** Date of discharge.
- **Doctor:** Assigned doctor's name.
- **Department:** Hospital department (e.g., Cardiology, Orthopedics).
- **Status:** Patient status (e.g., Admitted, Discharged, Under Observation).

Data Exploration and Analysis (SQL and Python)

1. SQL Queries

- What is the **average age** of patients for each diagnosis?



```

10 #####1#####
11 • select*from clean_data;
12
13 • select Diagnosis, AVG(age) as average_age
14 from clean_data
15 group by Diagnosis;
16
17

```

Diagnosis	average_age
Hypertension	42.5000
Diabetes	17.0000
Asthma	55.0000
Flu	29.0000
Heart Disease	62.0000
Kidney Disease	49.0000
Pneumonia	37.0000

Comment : The result average age of

'Hypertension' is '42.5000'

'Diabetes' is '17.0000"

Asthma' is '55.0000'

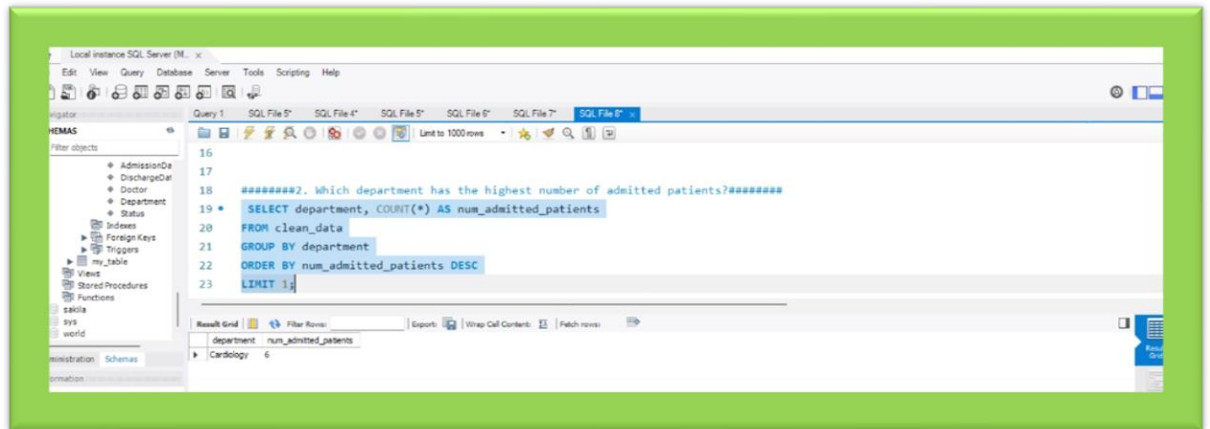
'Flu' is '29.0000'

'Heart Disease' is '62.0000'

'Kidney Disease' is '49.0000'

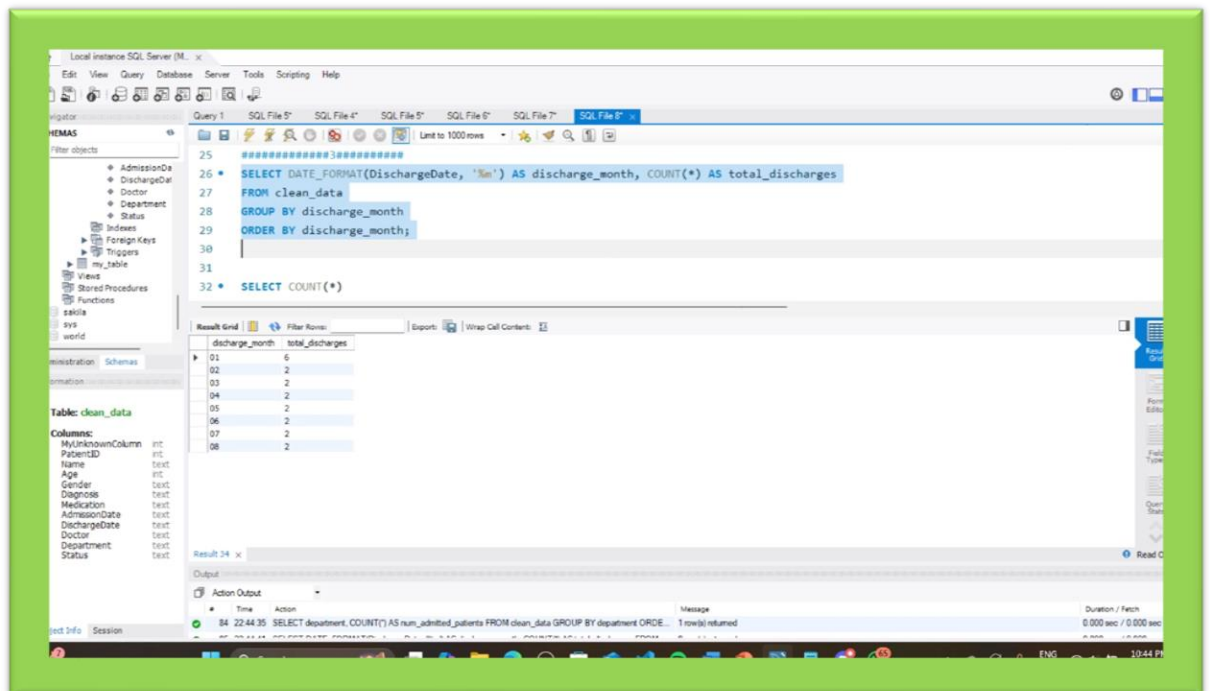
'Pneumonia' is '37.0000'

- Which department has the **highest number of admitted patients**?

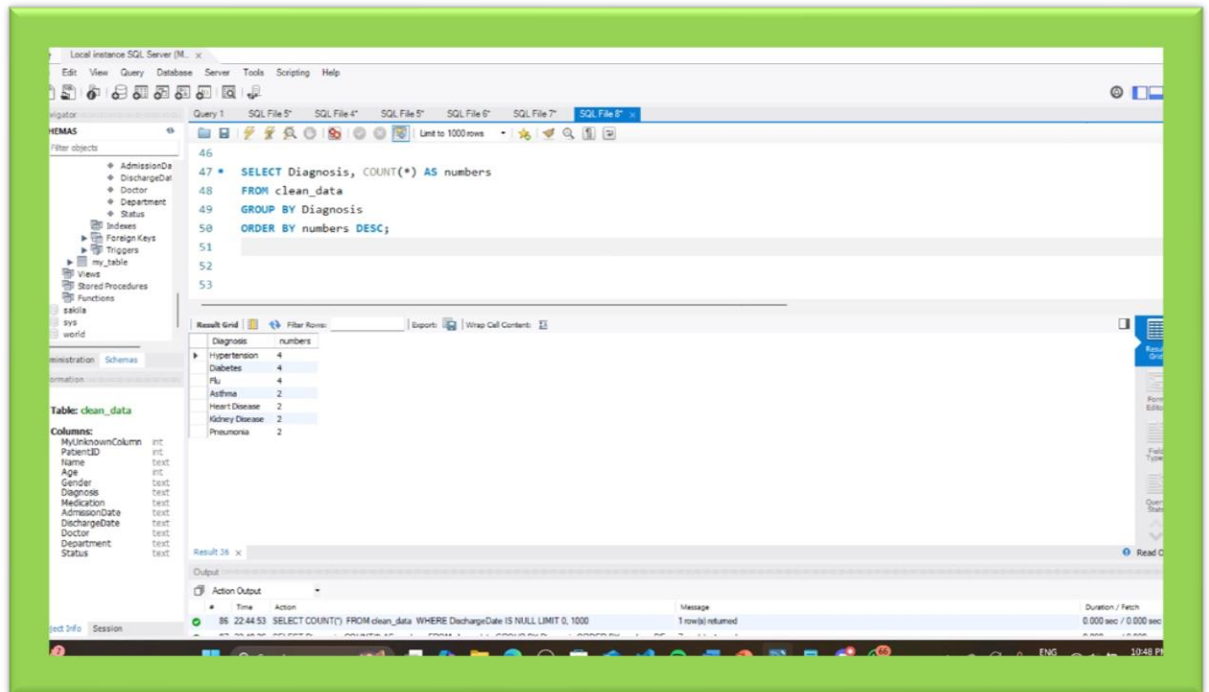


Comment: department has the **highest number of admitted patients** :Cardiology

- How many patients have been **discharged** per month?

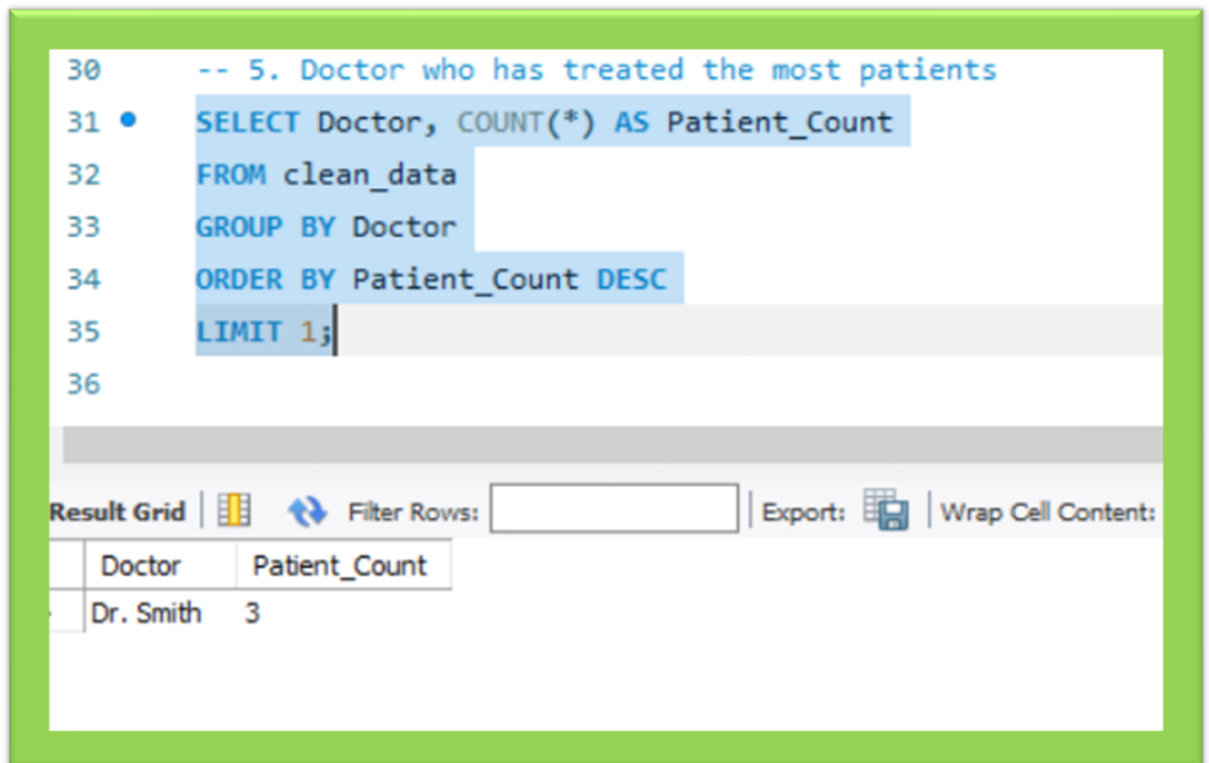


- What is the **most common diagnosis** among patients?



Comment: **most common diagnosis** among patients 'Hypertension'

- Which doctor has treated the **most patients**?



Comment: doctor has treated the **most patients 'Dr. Smith'**

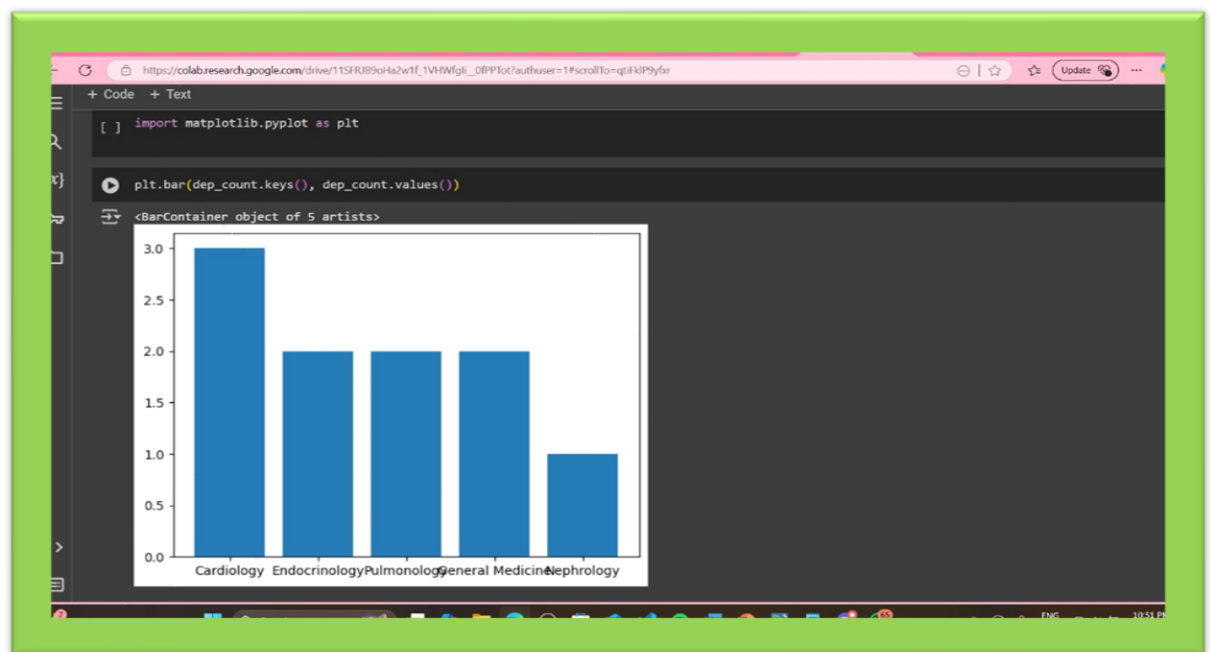
2. Python Analysis

- **Visualize the number of patients per department** using a bar chart.

```
Next steps: Generate code with df View recommended plots New interactive sheet
```

```
data = pd.read_csv('Clean_data.csv')
dep_count={}
for x in data['Department']:
    if x not in dep_count:
        dep_count[x]=1
    else:
        dep_count[x]+=1
dep_count
```

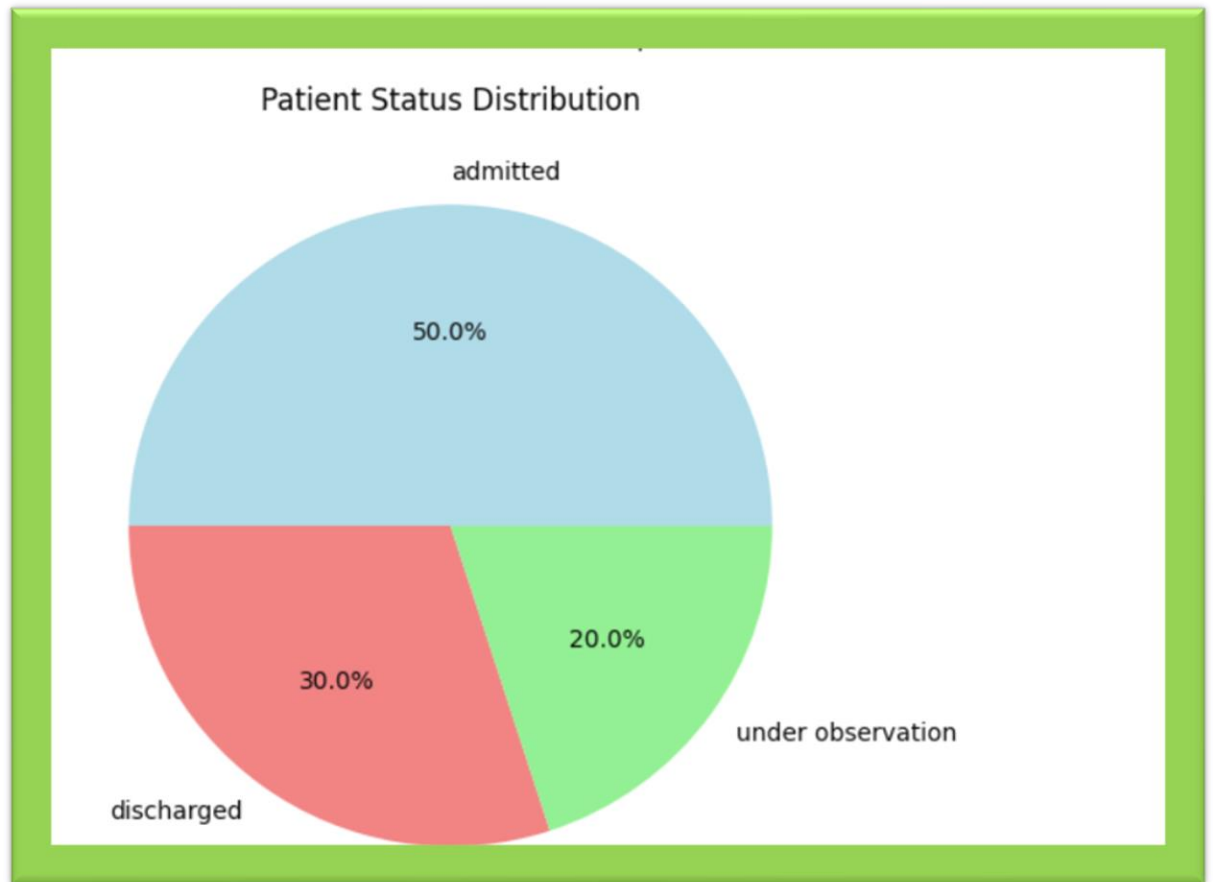
```
{'Cardiology': 3,
 'Endocrinology': 2,
 'Pulmonology': 2,
 'General Medicine': 2,
 'Nephrology': 1}
```



Comment: Cardiology has the highest number of patients (3), followed by Endocrinology, Pulmonology, and General Medicine (each with 2 patients).

Nephrology has the least number of patients (1)

- **Create a pie chart** showing the distribution of patient statuses (Admitted, Discharged, Under Observation).

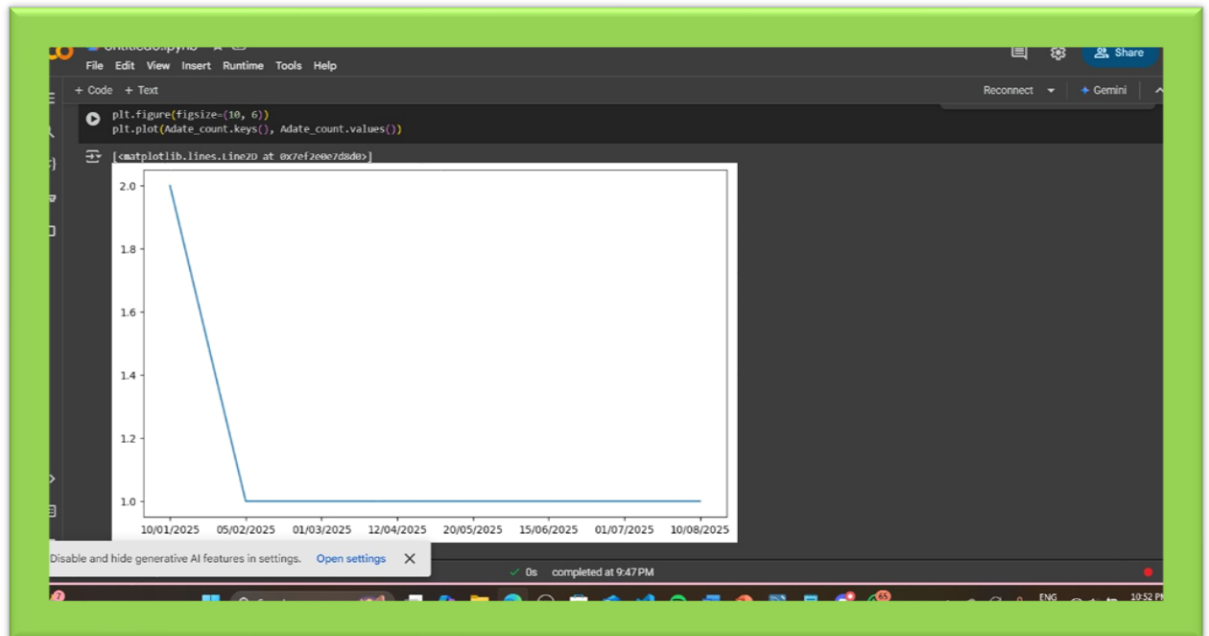


50% of patients are admitted, indicating a high hospitalization rate.

30% of patients are discharged

20% of patients are under observation

Generate a line chart showing **monthly hospital admissions trends**.



Theoretical Questions

1. Data Cleaning

- What are the common issues you might encounter in a messy dataset?
 1. Missing values
 2. Duplicate records
 3. Inconsistent data types
 4. Outliers
 5. Formatting errors
- How would you handle missing values in a dataset?
 1. Remove rows or columns with too many missing values
 2. Fill missing values with the mean, median, or mode
- What is the importance of data type consistency in data analysis?
 1. Ensuring accurate calculations
 2. Preventing errors during analysis and modeling
 3. Improving performance and efficient memory usage

2. SQL Queries

- What is the difference between INNER JOIN and LEFT JOIN?
 1. INNER JOIN returns only the matching records from both tables.
 2. LEFT JOIN returns all records from the left table and matching records from the right table
- How would you use the GROUP BY clause to aggregate data?
 1. The GROUP BY clause groups data by specific columns
- What is the purpose of the HAVING clause in SQL?
 1. The HAVING clause to apply conditions to the aggregated data

Python Analysis

- How would you use Pandas to clean a dataset with mixed data types?
 1. Convert columns with wrong data types to consistent types
 2. Handle invalid values
 3. Fill missing values or handle errors
- What are the benefits of using visualizations in data analysis?
 1. Making patterns, trends, and outliers easy to identify
 2. Providing a clearer understanding