

```
import pymysql
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import classification_report, accuracy_score

# Connect to MySQL using PyMySQL
db_connection = pymysql.connect(
    host="127.0.0.1",
    user="root",
    password="AnIsH@123#", # Replace with your MySQL password
    database="healthcare"
)

cursor = db_connection.cursor()
```

```
# Fetch patient details and health metrics
query = """
SELECT p.name, p.age, p.gender, p.weight, p.height, h.date_of_record, h.blood_pressure, h.heart_rate, h.cholesterol
FROM patients p
JOIN health_metrics h ON p.patient_id = h.patient_id;
"""
cursor.execute(query)
data = cursor.fetchall()

print(data)
```

↻ ((('John Doe', 45, 'Male', Decimal('78.50'), Decimal('175.00'), datetime.date(2024, 10, 1), '130/85', 78, Decimal('5.60')), ('Jane S...

```
# Convert the fetched data into a Pandas DataFrame
df = pd.DataFrame(data, columns=['name', 'age', 'gender', 'weight', 'height', 'date_of_record', 'blood_pressure', 'heart_rate', 'cholest
print(df)
# Close the database connection
cursor.close()
db_connection.close()
```

Data Preprocessing and Visualization (Assumed done before)

↻

16	Patient 17	58	Male	76.91	172.78	2024-10-14	130/85
17	Patient 18	45	Male	69.68	153.37	2024-10-15	120/78
18	Patient 19	55	Female	88.85	152.01	2024-10-16	140/88
19	Patient 20	35	Female	90.67	175.96	2024-10-17	115/75
20	Patient 21	30	Female	83.85	161.29	2024-10-18	130/84
21	Patient 22	32	Male	76.01	179.83	2024-10-19	135/89
22	Patient 23	53	Female	61.12	166.11	2024-10-20	120/80
23	Patient 24	48	Male	67.05	170.78	2024-10-21	145/93
24	Patient 25	58	Female	66.73	151.74	2024-10-22	130/87
25	Patient 26	53	Female	76.35	173.54	2024-10-23	120/76
26	Patient 27	50	Male	82.88	163.85	2024-10-24	140/90
27	Patient 28	47	Female	64.49	174.82	2024-10-25	110/70
28	Patient 29	53	Female	67.12	160.33	2024-10-26	150/95
29	Patient 30	35	Female	98.92	154.29	2024-10-27	125/80
30	Patient 31	53	Female	73.88	152.89	2024-10-28	135/88
31	Patient 32	43	Female	69.27	162.85	2024-10-29	140/89
32	Patient 33	43	Female	75.60	153.38	2024-10-30	120/77
33	Patient 34	41	Female	99.16	179.93	2024-10-31	145/92
34	Patient 35	31	Male	66.91	152.02	2024-11-01	130/85
35	Patient 36	54	Female	99.93	159.56	2024-11-02	150/96

	heart_rate	cholesterol
0	78	5.60

```

18      75      5.50
19      65      5.40
20      78      6.30
21      80      5.80
22      70      5.20
23      84      6.40
24      77      6.10
25      73      5.60
26      81      6.50
27      67      4.70
28      85      6.90
29      74      5.90
30      79      6.30
31      81      6.60
32      70      5.50
33      82      6.80
34      75      5.40
35      88      7.10

```

```
df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 36 entries, 0 to 35
Data columns (total 9 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   name                  36 non-null    object
1   age                   36 non-null    int64
2   gender                36 non-null    object
3   weight                36 non-null    object
4   height                36 non-null    object
5   date_of_record        36 non-null    object
6   blood_pressure        36 non-null    object
7   heart_rate            36 non-null    int64
8   cholesterol           36 non-null    object
dtypes: int64(2), object(7)
memory usage: 2.7+ KB

```

```
df['date_of_record'] = pd.to_datetime(df['date_of_record'])
df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 36 entries, 0 to 35
Data columns (total 9 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   name                  36 non-null    object
1   age                   36 non-null    int64
2   gender                36 non-null    object
3   weight                36 non-null    object
4   height                36 non-null    object
5   date_of_record        36 non-null    datetime64[ns]
6   blood_pressure        36 non-null    object
7   heart_rate            36 non-null    int64
8   cholesterol           36 non-null    object
dtypes: datetime64[ns](1), int64(2), object(6)
memory usage: 2.7+ KB

```

```

# Convert relevant columns to float
df['weight'] = df['weight'].astype(float)
df['height'] = df['height'].astype(float)
df['cholesterol'] = df['cholesterol'].astype(float)

# Calculate BMI for each patient
df['bmi'] = df['weight'] / ((df['height'] / 100) ** 2)
print(df.bmi)

```

```

0    25.632653
1    23.948577
2    28.242187
3    37.043735
4    29.081080
5    21.851449
6    24.809263
7    29.208367
8    32.491885
9    22.909343
10   22.935522
11   29.403128
12   24.519338
13   25.738354
14   33.645813
15   33.616488
16   25.762967
17   29.622882
18   38.451485

```

```

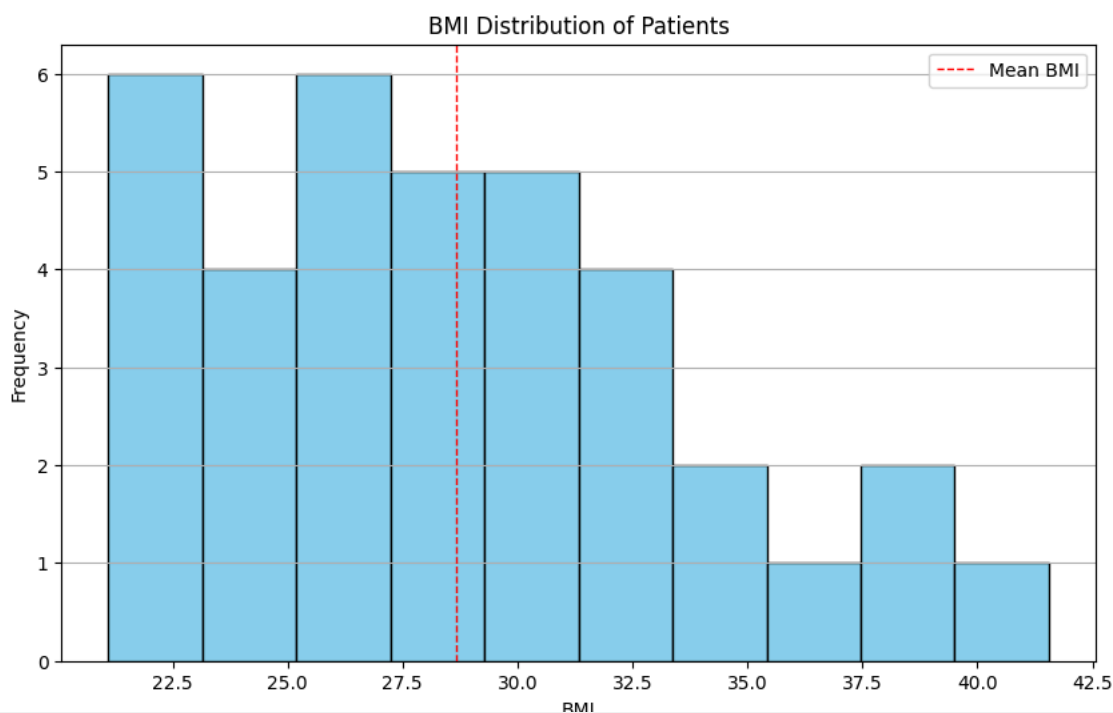
19 29.284358
20 32.232069
21 23.504252
22 22.150921
23 22.989248
24 28.981507
25 25.351862
26 30.871437
27 21.101345
28 26.110931
29 41.553590
30 31.605941
31 26.119789
32 32.135446
33 30.628756
34 28.952734
35 39.250738
Name: bmi, dtype: float64

```

```

# Plotting the BMI distribution
plt.figure(figsize=(10, 6))
plt.hist(df['bmi'], bins=10, color='skyblue', edgecolor='black')
plt.title('BMI Distribution of Patients')
plt.xlabel('BMI')
plt.ylabel('Frequency')
plt.grid(axis='y')
plt.axvline(df['bmi'].mean(), color='red', linestyle='dashed', linewidth=1, label='Mean BMI')
plt.legend()
plt.show()

```



```

# Calculate BMI
df['bmi'] = df['weight'] / ((df['height'] / 100) ** 2)

# Print and plot BMI
print("BMI for each patient:\n", df[['name', 'weight', 'height', 'bmi']])

```



```

BMI for each patient:
   name  weight  height      bmi
0  John Doe   78.50   175.00  25.63265306122448979591836735
1  Jane Smith  65.20   165.00  23.94857667584940312213039486
2  Alice Brown  72.30   160.00  28.2421875
3  Patient 4   84.05   150.63  37.04373473232336113790666688
4  Patient 5   72.92   158.35  29.08107956129923272331328891
5  Patient 6   64.36   171.62  21.85144896088890770880957394
6  Patient 7   69.93   167.89  24.80926339934639706434572489
7  Patient 8   71.20   156.13  29.20836682364536589220613662
8  Patient 9   97.20   172.96  32.49188531878096222619101613
9  Patient 10  72.79   178.25  22.90934341000336369841591442
10 Patient 11  73.94   179.55  22.93552191689347444231730972
11 Patient 12  83.73   168.75  29.40312757201646090534979424
12 Patient 13  69.36   168.19  24.51933816268376543122655999
13 Patient 14  67.04   161.39  25.73835356489883811566518934
14 Patient 15  77.49   151.76  33.64581286824992656884924147

```

```

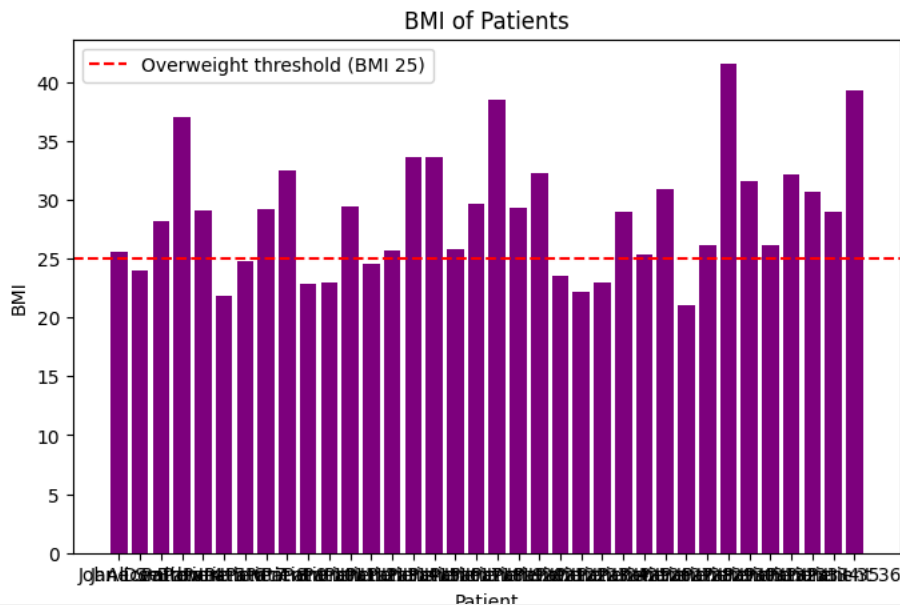
15 Patient 16 89.02 162.73 33.61648811176990624870508831
16 Patient 17 76.91 172.78 25.76296669106672965456882950
17 Patient 18 69.68 153.37 29.62288188051427788683879718
18 Patient 19 88.85 152.01 38.45148474901378649531144407
19 Patient 20 90.67 175.96 29.28435811296673524294435265
20 Patient 21 83.85 161.29 32.23206892814678431142465856
21 Patient 22 76.01 179.83 23.50425249938472570971795457
22 Patient 23 61.12 166.11 22.15092104577093162102364647
23 Patient 24 67.05 170.78 22.98924791674898337465592073
24 Patient 25 66.73 151.74 28.98150706234115437064666103
25 Patient 26 76.35 173.54 25.35186159167932444550747016
26 Patient 27 82.88 163.85 30.87143739263743409485424206
27 Patient 28 64.49 174.82 21.10134534115604170356157335
28 Patient 29 67.12 160.33 26.11093133478999289531361650
29 Patient 30 98.92 154.29 41.55358992624704068770670438
30 Patient 31 73.88 152.89 31.60594102879844962848709346
31 Patient 32 69.27 162.85 26.11978885089991571494437856
32 Patient 33 75.60 153.38 32.13544576904635990328829094
33 Patient 34 99.16 179.93 30.62875600526012745176419581
34 Patient 35 66.91 152.02 28.95273354393003323346291777
35 Patient 36 99.93 159.56 39.25073847791881421716187301

```

```

plt.figure(figsize=(8, 5))
plt.bar(df['name'], df['bmi'], color='purple')
plt.title('BMI of Patients')
plt.xlabel('Patient')
plt.ylabel('BMI')
plt.axhline(y=25, color='r', linestyle='--', label='Overweight threshold (BMI 25)')
plt.legend()
plt.show()

```



```

import mysql.connector
import pandas as pd
import matplotlib.pyplot as plt

# Step 1: Connect to the MySQL database
db_connection = pymysql.connect(
    host="127.0.0.1",
    user="root",
    password="AnIsH@123#", # Replace with your MySQL password
    database="healthcare"
)

cursor = db_connection.cursor()

# Step 2: Fetch data from MySQL
query = """
SELECT p.patient_id, p.name, p.weight, p.height
FROM patients p;
"""

cursor.execute(query)
data = cursor.fetchall()

# Check the data fetched
print("Data Fetched from Database:\n", data)

```

```
# Step 3: Convert the data into a Pandas DataFrame
# Ensure that we only create a DataFrame with the columns we are fetching
df = pd.DataFrame(data, columns=['patient_id', 'name', 'weight', 'height'])
cursor.close()
db_connection.close()
# Check if the DataFrame has the expected columns
print("DataFrame Columns: ", df.columns)

# Calculate BMI for each patient
df['bmi'] = df['weight'] / ((df['height'] / 100) ** 2)

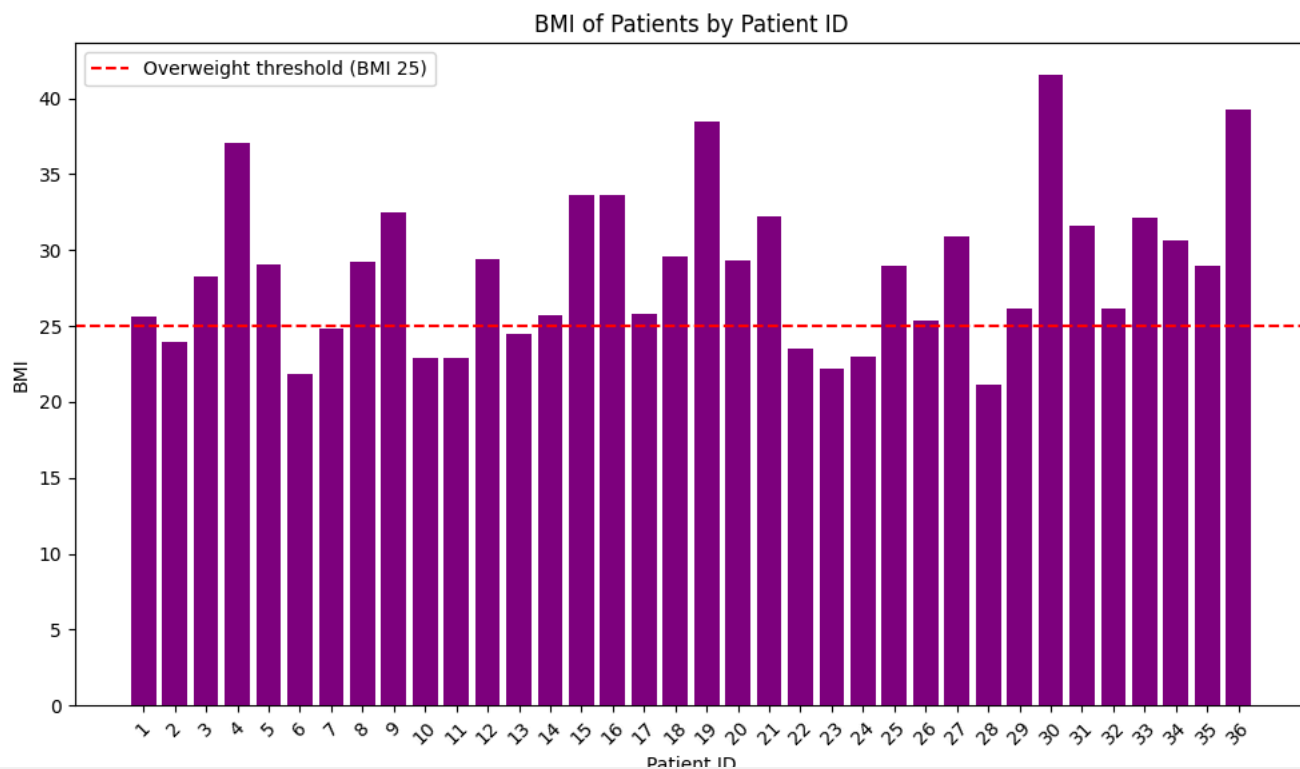
# Print BMI for each patient
print("BMI for each patient:\n", df[['patient_id', 'name', 'weight', 'height', 'bmi']])
```

➡ Data Fetched from Database:
 ((1, 'John Doe', Decimal('78.50'), Decimal('175.00')), (2, 'Jane Smith', Decimal('65.20'), Decimal('165.00')), (3, 'Alice Brown', Decimal('72.30'), Decimal('160.00')), (4, 'Patient 4', Decimal('84.05'), Decimal('150.63')), (5, 'Patient 5', Decimal('72.92'), Decimal('158.35')), (6, 'Patient 6', Decimal('64.36'), Decimal('171.62')), (7, 'Patient 7', Decimal('69.93'), Decimal('167.89')), (8, 'Patient 8', Decimal('71.20'), Decimal('156.13')), (9, 'Patient 9', Decimal('97.20'), Decimal('172.96')), (10, 'Patient 10', Decimal('72.79'), Decimal('178.25')), (11, 'Patient 11', Decimal('73.94'), Decimal('179.55')), (12, 'Patient 12', Decimal('83.73'), Decimal('168.75')), (13, 'Patient 13', Decimal('69.36'), Decimal('168.19')), (14, 'Patient 14', Decimal('67.04'), Decimal('161.39')), (15, 'Patient 15', Decimal('77.49'), Decimal('151.76')), (16, 'Patient 16', Decimal('89.02'), Decimal('162.73')), (17, 'Patient 17', Decimal('76.91'), Decimal('172.78')), (18, 'Patient 18', Decimal('69.68'), Decimal('153.37')), (19, 'Patient 19', Decimal('88.85'), Decimal('152.01')), (20, 'Patient 20', Decimal('90.67'), Decimal('175.96')), (21, 'Patient 21', Decimal('83.85'), Decimal('161.29')), (22, 'Patient 22', Decimal('76.01'), Decimal('179.83')), (23, 'Patient 23', Decimal('61.12'), Decimal('166.11')), (24, 'Patient 24', Decimal('67.05'), Decimal('170.78')), (25, 'Patient 25', Decimal('66.73'), Decimal('151.74')), (26, 'Patient 26', Decimal('76.35'), Decimal('173.54')), (27, 'Patient 27', Decimal('82.88'), Decimal('163.85')), (28, 'Patient 28', Decimal('64.49'), Decimal('174.82')), (29, 'Patient 29', Decimal('67.12'), Decimal('160.33')), (30, 'Patient 30', Decimal('98.92'), Decimal('154.29')), (31, 'Patient 31', Decimal('73.88'), Decimal('152.89')), (32, 'Patient 32', Decimal('69.27'), Decimal('162.85')), (33, 'Patient 33', Decimal('75.60'), Decimal('153.38')), (34, 'Patient 34', Decimal('99.16'), Decimal('179.93')), (35, 'Patient 35', Decimal('66.91'), Decimal('152.02')), (36, 'Patient 36', Decimal('99.93'), Decimal('159.56')))

BMI for each patient:

	patient_id	name	weight	height	bmi
0	1	John Doe	78.50	175.00	25.63265306122448979591836735
1	2	Jane Smith	65.20	165.00	23.94857667584940312213039486
2	3	Alice Brown	72.30	160.00	28.2421875
3	4	Patient 4	84.05	150.63	37.04373473232336113790666688
4	5	Patient 5	72.92	158.35	29.08107956129923272331328891
5	6	Patient 6	64.36	171.62	21.85144896088890770880957394
6	7	Patient 7	69.93	167.89	24.80926339934639706434572489
7	8	Patient 8	71.20	156.13	29.20836682364536589220613662
8	9	Patient 9	97.20	172.96	32.49188531878096222619101613
9	10	Patient 10	72.79	178.25	22.90934341000336369841591442
10	11	Patient 11	73.94	179.55	22.93552191689347444231730972
11	12	Patient 12	83.73	168.75	29.40312757201646090534979424
12	13	Patient 13	69.36	168.19	24.51933816268376543122655999
13	14	Patient 14	67.04	161.39	25.73835356489883811566518934
14	15	Patient 15	77.49	151.76	33.64581286824992656884924147
15	16	Patient 16	89.02	162.73	33.61648811176990624870508831
16	17	Patient 17	76.91	172.78	25.76296669106672965456882950
17	18	Patient 18	69.68	153.37	29.62288188051427788683879718
18	19	Patient 19	88.85	152.01	38.45148474901378649531144407
19	20	Patient 20	90.67	175.96	29.28435811296673524294435265
20	21	Patient 21	83.85	161.29	32.23206892814678431142465856
21	22	Patient 22	76.01	179.83	23.50425249938472570971795457
22	23	Patient 23	61.12	166.11	22.15092104577093162102364647
23	24	Patient 24	67.05	170.78	22.98924791674898337465592073
24	25	Patient 25	66.73	151.74	28.98150706234115437064666103
25	26	Patient 26	76.35	173.54	25.35186159167932444550747016
26	27	Patient 27	82.88	163.85	30.87143739263743409485424206
27	28	Patient 28	64.49	174.82	21.10134534115604170356157335
28	29	Patient 29	67.12	160.33	26.11093133478999289531361650
29	30	Patient 30	98.92	154.29	41.55358992624704068770670438
30	31	Patient 31	73.88	152.89	31.60594102879844962848709346
31	32	Patient 32	69.27	162.85	26.11978885089991571494437856
32	33	Patient 33	75.60	153.38	32.13544576904635990328829094
33	34	Patient 34	99.16	179.93	30.62875600526012745176419581
34	35	Patient 35	66.91	152.02	28.95273354393003323346291777
35	36	Patient 36	99.93	159.56	39.25073847791881421716187301

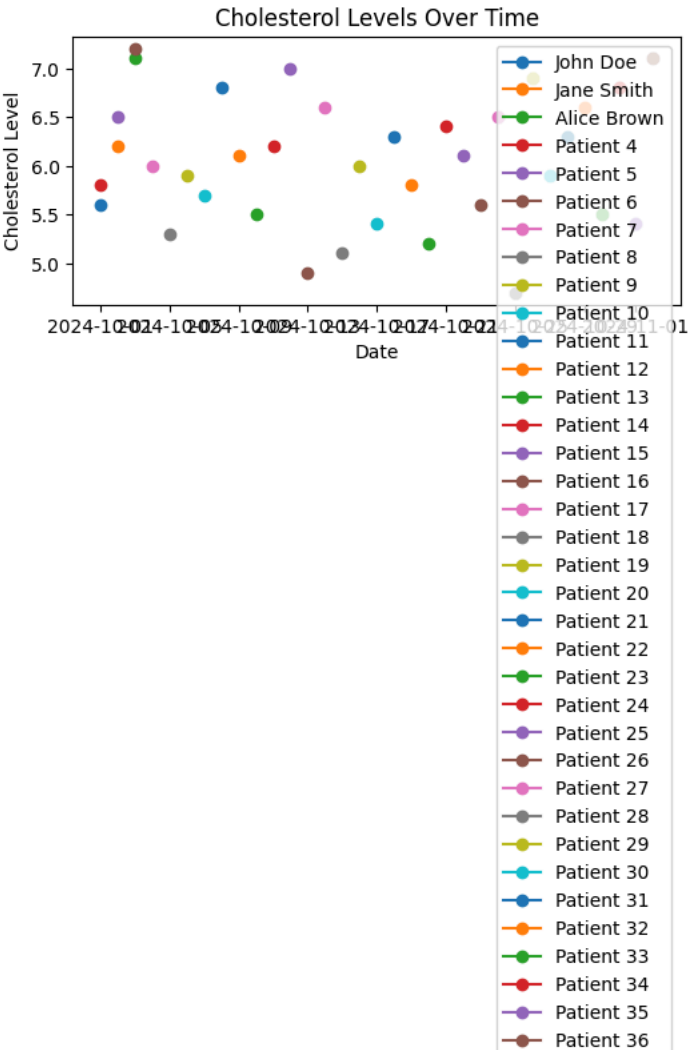
```
# Plotting BMI with patient_id on the x-axis
plt.figure(figsize=(10, 6))
plt.bar(df['patient_id'].astype(str), df['bmi'], color='purple') # Convert patient_id to string for plotting
plt.title('BMI of Patients by Patient ID')
plt.xlabel('Patient ID')
plt.ylabel('BMI')
plt.axhline(y=25, color='r', linestyle='--', label='Overweight threshold (BMI 25)')
plt.xticks(rotation=45) # Rotate x-axis labels for better readability
plt.legend()
plt.tight_layout()
plt.show()
```



```
plt.figure(figsize=(20, 12))
plt.subplot(4, 3, 2)
for name in df['name'].unique():
    patient_data = df[df['name'] == name]
    plt.plot(patient_data['date_of_record'], patient_data['cholesterol'], marker='o', label=name)

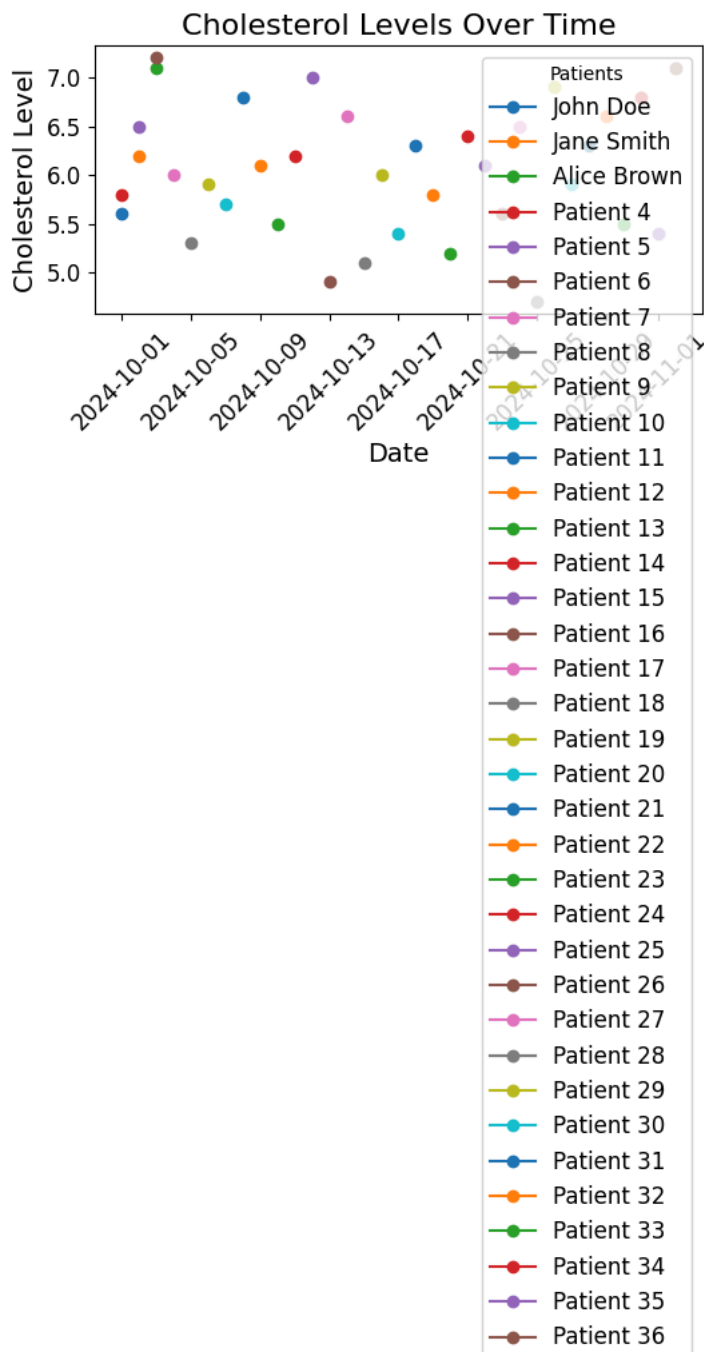
plt.title('Cholesterol Levels Over Time')
plt.xlabel('Date')
plt.ylabel('Cholesterol Level')
plt.legend()
```

 <matplotlib.legend.Legend at 0x1d539d58aa0>



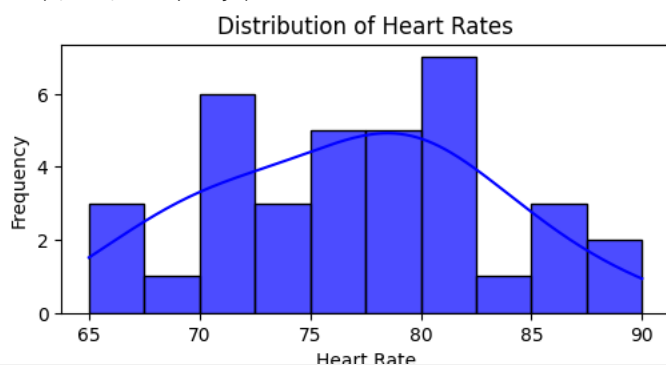
Start coding or [generate](#) with AI.

C:\Users\anish_rane\AppData\Local\Temp\ipykernel_16872\4146644537.py:25: UserWarning: Tight layout not applied. tight_layout can
plt.tight_layout() # Automatically adjust subplot parameters for a clean layout



```
plt.figure(figsize=(20, 12))
plt.subplot(4, 3, 3)
sns.histplot(df['heart_rate'], bins=10, kde=True, color='blue', alpha=0.7)
plt.title('Distribution of Heart Rates')
plt.xlabel('Heart Rate')
plt.ylabel('Frequency')
```

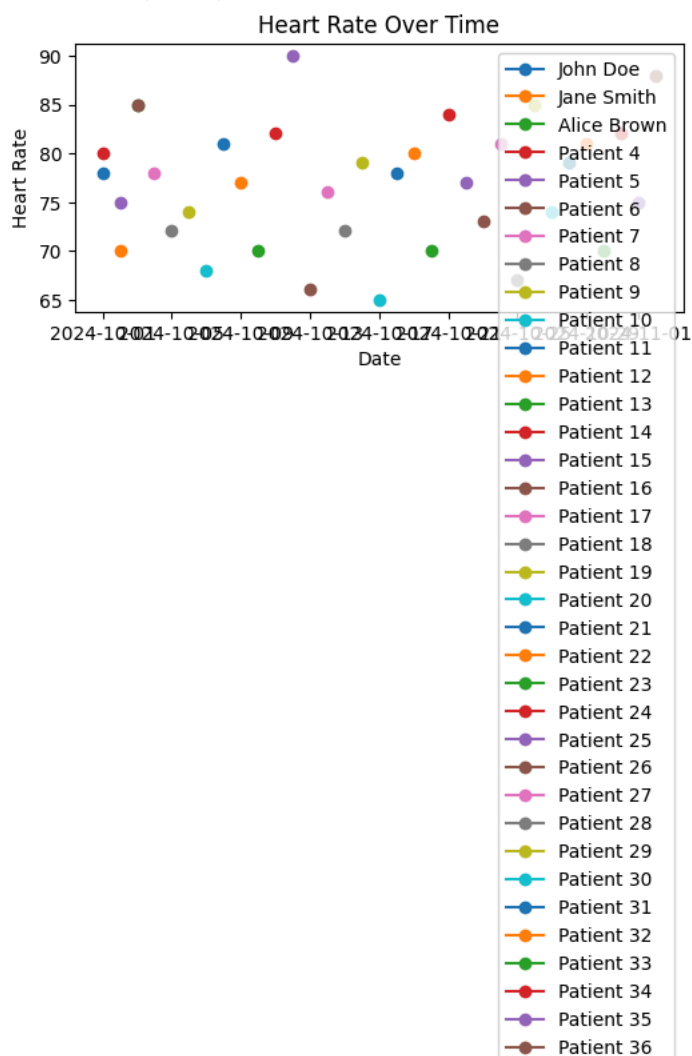
Text(0, 0.5, 'Frequency')



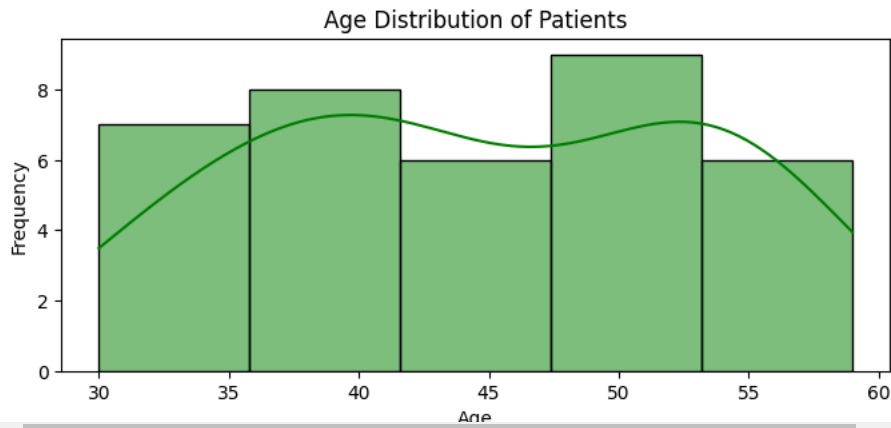

```
# Visualization 4: Heart Rate Over Time
plt.figure(figsize=(20, 12))
plt.subplot(4, 3, 4)
for name in df['name'].unique():
    patient_data = df[df['name'] == name]
    plt.plot(patient_data['date_of_record'], patient_data['heart_rate'], marker='o', label=name)

plt.title('Heart Rate Over Time')
plt.xlabel('Date')
plt.ylabel('Heart Rate')
plt.legend()
```

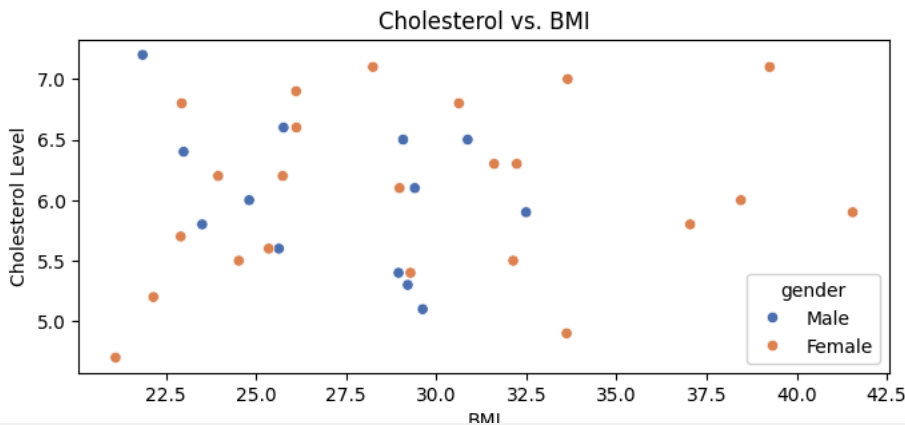
 <matplotlib.legend.Legend at 0x1d53aba7500>



```
# Visualization 5: Age Distribution
plt.figure(figsize=(20, 12))
plt.subplot(4, 3, 5)
sns.histplot(df['age'], bins=5, color='green', kde=True)
plt.title('Age Distribution of Patients')
plt.xlabel('Age')
plt.ylabel('Frequency')
plt.tight_layout()
plt.show()
```



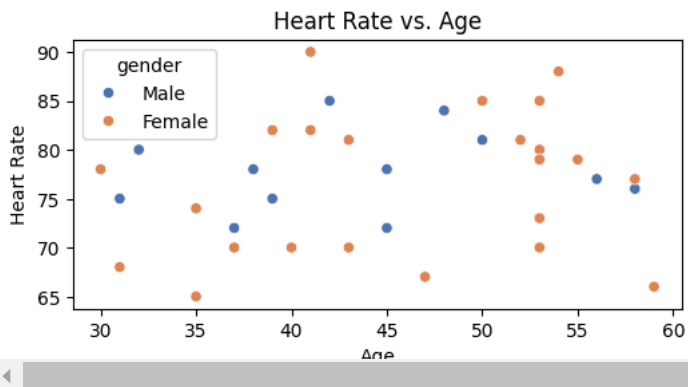
```
# Visualization 6: Cholesterol vs. BMI
plt.figure(figsize=(20, 12))
plt.subplot(4, 3, 6)
sns.scatterplot(x='bmi', y='cholesterol', data=df, hue='gender', palette='deep')
plt.title('Cholesterol vs. BMI')
plt.xlabel('BMI')
plt.ylabel('Cholesterol Level')
plt.tight_layout()
plt.show()
```



```
# Visualization 7: Heart Rate vs. Age
plt.figure(figsize=(20, 12))
plt.subplot(4, 3, 7)
sns.scatterplot(x='age', y='heart_rate', data=df, hue='gender', palette='deep')
plt.title('Heart Rate vs. Age')
plt.xlabel('Age')
plt.ylabel('Heart Rate')
```



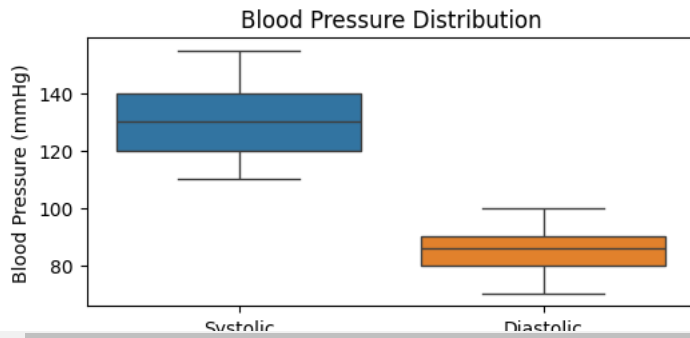
Text(0, 0.5, 'Heart Rate')



```
# Visualization 8: Blood Pressure Analysis
# Convert blood pressure to separate systolic and diastolic columns for analysis
plt.figure(figsize=(20, 12))
df[['systolic', 'diastolic']] = df['blood_pressure'].str.split('/', expand=True).astype(int)
plt.subplot(4, 3, 8)
sns.boxplot(data=df[['systolic', 'diastolic']])
plt.title('Blood Pressure Distribution')
```

```
plt.ylabel('Blood Pressure (mmHg)')
plt.xticks([0, 1], ['Systolic', 'Diastolic'])
```

```
[[<matplotlib.axis.XTick at 0x1d53c4258b0>,
<matplotlib.axis.XTick at 0x1d53c3d5c40>],
[Text(0, 0, 'Systolic'), Text(1, 0, 'Diastolic')]]
```

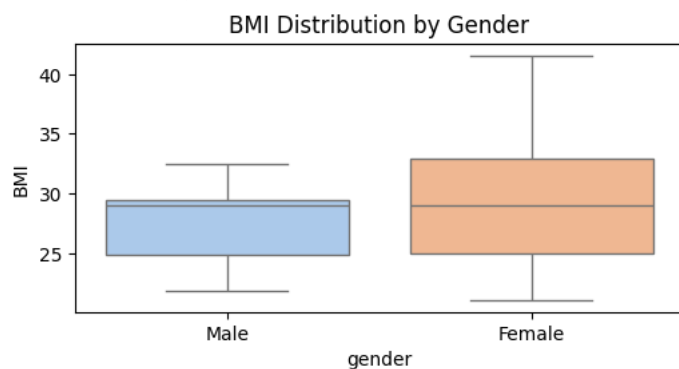


```
# Visualization 9: BMI Distribution by Gender
plt.figure(figsize=(20, 12))
plt.subplot(4, 3, 9)
sns.boxplot(x='gender', y='bmi', data=df, palette='pastel')
plt.title('BMI Distribution by Gender')
plt.ylabel('BMI')
```

```
C:\Users\anish rane\AppData\Local\Temp\ipykernel_16872\961257994.py:4: FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `le`

```
sns.boxplot(x='gender', y='bmi', data=df, palette='pastel')
Text(0, 0.5, 'BMI')
```



```
# Visualization 10: Cholesterol Levels by Gender
plt.figure(figsize=(20, 12))
plt.subplot(4, 3, 10)
sns.boxplot(x='gender', y='cholesterol', data=df, palette='pastel')
plt.title('Cholesterol Levels by Gender')
plt.ylabel('Cholesterol Level')

plt.tight_layout()
plt.show()
```

C:\Users\anish rane\AppData\Local\Temp\ipykernel_16872\1925240386.py:4: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `le

```
sns.boxplot(x='gender', y='cholesterol', data=df, palette='pastel')
```



```
import mysql.connector
import pandas as pd
import matplotlib.pyplot as plt

# Assuming you have already fetched your DataFrame `df` from the database as shown earlier

# 1. Count the number of males and females
gender_counts = df['gender'].value_counts()

# 2. Calculate average cholesterol and blood pressure by gender
# Convert blood pressure from string to two separate columns for systolic and diastolic
df[['systolic', 'diastolic']] = df['blood_pressure'].str.split('/', expand=True).astype(int)

average_cholesterol = df.groupby('gender')['cholesterol'].mean()
average_blood_pressure = df.groupby('gender')[['systolic', 'diastolic']].mean()

# 3. Plotting
plt.figure(figsize=(18, 12))

# Pie chart for Gender Distribution
plt.subplot(3, 2, 1)
plt.pie(gender_counts, labels=gender_counts.index, autopct='%1.1f%%', startangle=90, colors=['#ff9999', '#66b3ff'])
plt.title('Gender Distribution')

# Pie chart for Average Cholesterol Levels
plt.subplot(3, 2, 2)
plt.pie(average_cholesterol, labels=average_cholesterol.index, autopct='%1.1f%%', startangle=90, colors=['#ffcc99', '#99ff99'])
plt.title('Average Cholesterol Levels by Gender')

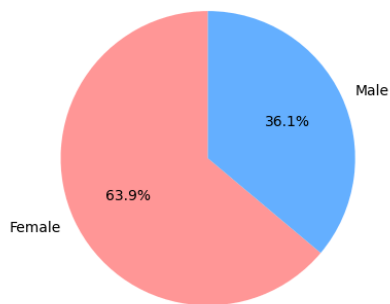
# Pie chart for Average Blood Pressure (Systolic)
plt.subplot(3, 2, 3)
plt.pie(average_blood_pressure['systolic'], labels=average_blood_pressure.index, autopct='%1.1f%%', startangle=90, colors=['#ffccff', '#99ffcc'])
plt.title('Average Systolic Blood Pressure by Gender')

# Pie chart for Average Blood Pressure (Diastolic)
plt.subplot(3, 2, 4)
plt.pie(average_blood_pressure['diastolic'], labels=average_blood_pressure.index, autopct='%1.1f%%', startangle=90, colors=['#fffb3e', '#99ff99'])
plt.title('Average Diastolic Blood Pressure by Gender')

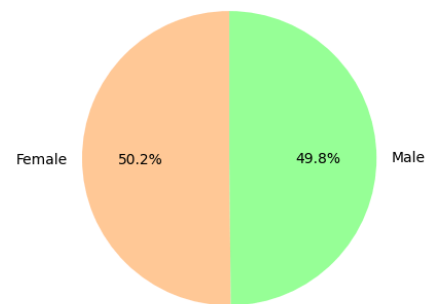
plt.tight_layout() # Automatically adjust subplot parameters for a clean layout
plt.show()
```



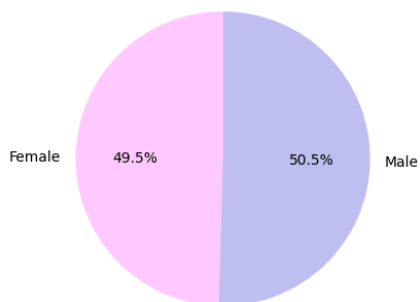
Gender Distribution



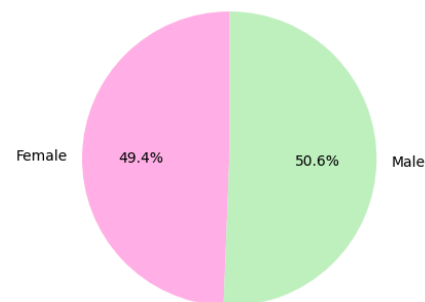
Average Cholesterol Levels by Gender



Average Systolic Blood Pressure by Gender



Average Diastolic Blood Pressure by Gender



```

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report
from sklearn.preprocessing import LabelEncoder

# Convert categorical gender column to numeric (0 for Female, 1 for Male)
le = LabelEncoder()
df['gender'] = le.fit_transform(df['gender']) # 0 for Female, 1 for Male

# Step 1: Categorize cholesterol levels
# Example: Normal < 200, Borderline 200-239, High >= 240
def categorize_cholesterol(value):
    if value < 200:
        return 'Normal'
    elif 200 <= value < 240:
        return 'Borderline'
    else:
        return 'High'

df['cholesterol_category'] = df['cholesterol'].apply(categorize_cholesterol)

# Convert cholesterol categories to numerical labels
le = LabelEncoder()
df['cholesterol_category'] = le.fit_transform(df['cholesterol_category'])

# Step 2: Define the features (X) and target (y)
X = df[['age', 'bmi', 'gender', 'heart_rate', 'systolic', 'diastolic']]
y = df['cholesterol_category']

# Step 3: Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Step 4: Standardize the features
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)

# Step 5: Train the model (RandomForestClassifier for classification)
model = RandomForestClassifier(random_state=42)
model.fit(X_train, y_train)

# Step 6: Make predictions on the test data

```

```

y_pred = model.predict(X_test)

# Step 7: Calculate accuracy
accuracy = accuracy_score(y_test, y_pred) * 100
print(f'Accuracy: {accuracy:.2f}%')

# Step 8: Classification report for detailed metrics
print(classification_report(y_test, y_pred, target_names=le.classes_))

```

```

↗ Accuracy: 100.00%
      precision    recall  f1-score   support

   Normal         1.00      1.00      1.00         8

 accuracy         1.00      1.00      1.00         8
 macro avg         1.00      1.00      1.00         8
 weighted avg         1.00      1.00      1.00         8

```

```

# Step 1: Find Alice's data in the DataFrame
alice_data = df[df['name'] == 'Alice Brown']

if alice_data.empty:
    print("Alice's data not found in the dataset.")
else:
    # Step 2: Select relevant features for prediction (same as used for training)
    X_alice = alice_data[['age', 'bmi', 'gender', 'heart_rate', 'systolic', 'diastolic']]

    # Step 3: Apply the same preprocessing (scaling)
    X_alice_scaled = scaler.transform(X_alice)

    # Step 4: Make the prediction
    alice_prediction = model.predict(X_alice_scaled)

    # Step 5: Convert the numeric prediction back to the category (Normal, Borderline, High)
    predicted_category = le.inverse_transform(alice_prediction)

    print(f"Alice's predicted cholesterol category: {predicted_category[0]}")

```

```

↗ Alice's predicted cholesterol category: Normal

```

```

import mysql.connector
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
import seaborn as sns

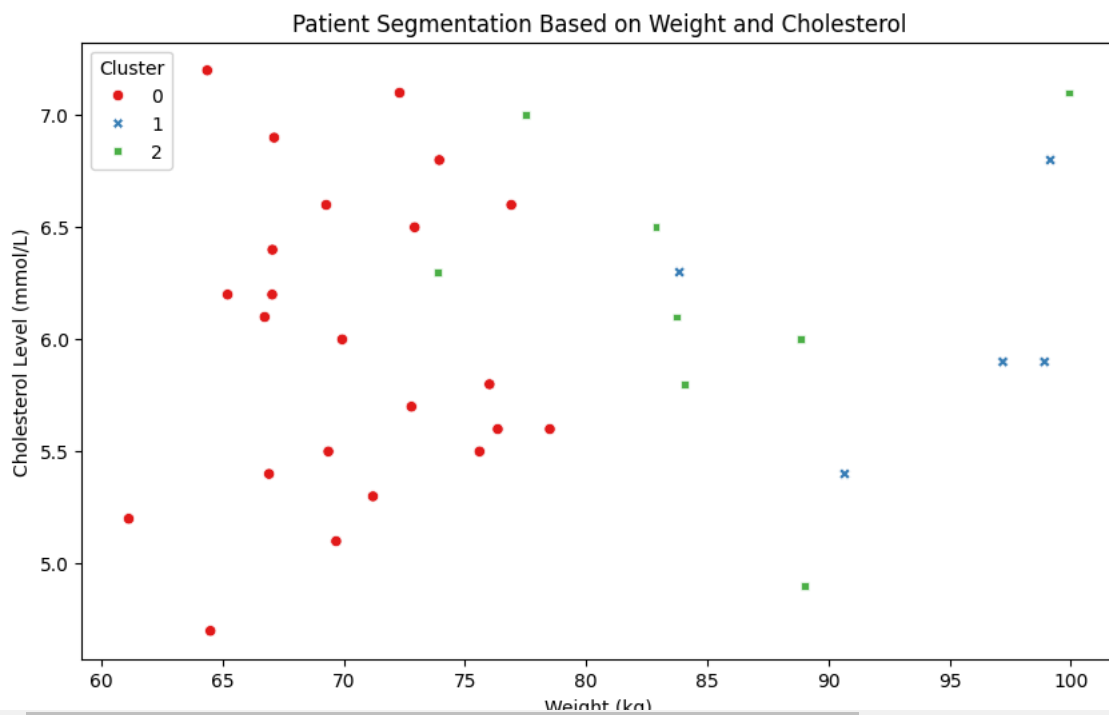
# Calculate BMI for each patient
df['bmi'] = df['weight'] / ((df['height'] / 100) ** 2)

# Step 4: Prepare data for clustering
features = df[['age', 'weight', 'cholesterol', 'bmi']]
scaler = StandardScaler()
scaled_features = scaler.fit_transform(features)

# Step 5: Apply K-Means Clustering
kmeans = KMeans(n_clusters=3, random_state=42)
df['cluster'] = kmeans.fit_predict(scaled_features)

# Step 6: Visualize Segmentation
plt.figure(figsize=(10, 6))
sns.scatterplot(x='weight', y='cholesterol', hue='cluster', data=df, palette='Set1', style='cluster')
plt.title('Patient Segmentation Based on Weight and Cholesterol')
plt.xlabel('Weight (kg)')
plt.ylabel('Cholesterol Level (mmol/L)')
plt.legend(title='Cluster')
plt.show()

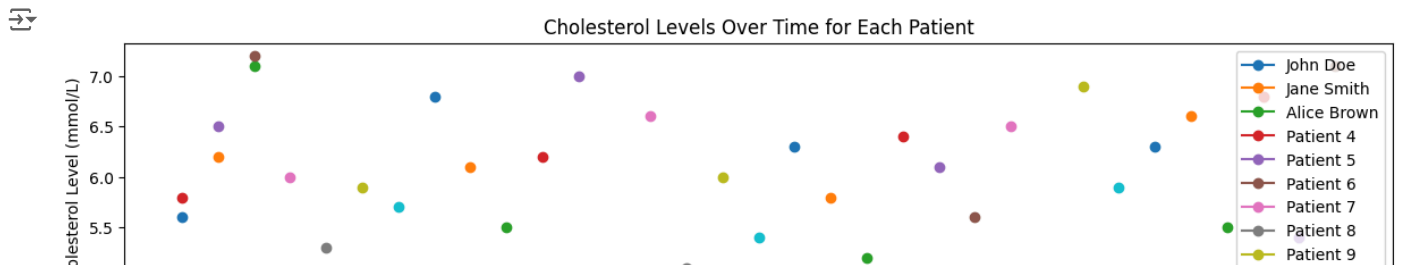
```



```
# Convert 'date_of_record' to datetime
df['date_of_record'] = pd.to_datetime(df['date_of_record'])

# Step 4: Plot Cholesterol Levels Over Time for Each Patient
plt.figure(figsize=(12, 6))
for name in df['name'].unique():
    patient_data = df[df['name'] == name]
    plt.plot(patient_data['date_of_record'], patient_data['cholesterol'], marker='o', label=name)

plt.title('Cholesterol Levels Over Time for Each Patient')
plt.xlabel('Date')
plt.ylabel('Cholesterol Level (mmol/L)')
plt.legend()
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



```
import numpy as np
import pandas as pd
import mysql.connector
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, classification_report

# Assuming you have already fetched your DataFrame `df` from the database

# Define conditions based on thresholds
def label_conditions(row):
    conditions = []
    if row['systolic'] > 140 or row['diastolic'] > 90:
        conditions.append('Hypertension')
    if row['cholesterol'] > 6.2:
        conditions.append('Hyperlipidemia')
    if row['bmi'] > 30:
        conditions.append('Obesity')
    return ', '.join(conditions) if conditions else 'Healthy'

# Apply function to label conditions
df['conditions'] = df.apply(label_conditions, axis=1)

# Visualize conditions
plt.figure(figsize=(10, 6))
sns.countplot(data=df, x='conditions', palette='viridis')
```