!pip install pandas matplotlib seaborn markdown2 weasyprint

```
Requirement already satisfied: pandas in /usr/local/lib/python3.11/dist-packages (2.2.2)
    Requirement already satisfied: matplotlib in /usr/local/lib/python3.11/dist-packages (3.10.0)
    Requirement already satisfied: seaborn in /usr/local/lib/python3.11/dist-packages (0.13.2)
    Collecting markdown2
       Downloading markdown2-2.5.3-py3-none-any.whl.metadata (2.1 kB)
    Collecting weasyprint
      Downloading weasyprint-64.1-py3-none-any.whl.metadata (3.7 kB)
    Requirement already satisfied: numpy>=1.23.2 in /usr/local/lib/python3.11/dist-packages (from pandas) (1.26.4)
    Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas) (2.8.2)
    Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-packages (from pandas) (2025.1)
    Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas) (2025.1)
    Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (1.3.1)
    Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (0.12.1)
    Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (4.56.0)
    Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (1.4.8)
    Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (24.2)
    Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (11.1.0)
    Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (3.2.1)
    Collecting pydyf>=0.11.0 (from weasyprint)
      Downloading pydyf-0.11.0-py3-none-any.whl.metadata (2.5 kB)
     Requirement already satisfied: cffi>=0.6 in /usr/local/lib/python3.11/dist-packages (from weasyprint) (1.17.1)
    Collecting tinyhtml5>=2.0.0b1 (from weasyprint)
       Downloading tinyhtml5-2.0.0-py3-none-any.whl.metadata (2.9 kB)
     Requirement already satisfied: tinycss2>=1.4.0 in /usr/local/lib/python3.11/dist-packages (from weasyprint) (1.4.0)
    Collecting cssselect2>=0.1 (from weasyprint)
       Downloading cssselect2-0.7.0-py3-none-any.whl.metadata (2.9 kB)
    Collecting Pyphen>=0.9.1 (from weasyprint)
      Downloading pyphen-0.17.2-py3-none-any.whl.metadata (3.2 kB)
    Requirement already satisfied: pycparser in /usr/local/lib/python3.11/dist-packages (from cffi>=0.6->weasyprint) (2.22)
    Requirement already satisfied: webencodings in /usr/local/lib/python3.11/dist-packages (from cssselect2>=0.1->weasyprint) (0.5.1)
    Collecting brotli>=1.0.1 (from fonttools[woff]>=4.0.0->weasyprint)
       Downloading Brotli-1.1.0-cp311-cp311-manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata (5.5 kB)
    Collecting zopfli>=0.1.4 (from fonttools[woff]>=4.0.0->weasyprint)
       Downloading zopfli-0.2.3.post1-cp311-cp311-manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata (2.9 kB)
     Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.8.2->pandas) (1.17.0)
    Downloading markdown2-2.5.3-py3-none-any.whl (48 kB)
                                                48.5/48.5 kB 2.3 MB/s eta 0:00:00
    Downloading weasyprint-64.1-py3-none-any.whl (302 kB)
                                                 302.0/302.0 kB 11.6 MB/s eta 0:00:00
    Downloading cssselect2-0.7.0-py3-none-any.whl (15 kB)
    Downloading pydyf-0.11.0-py3-none-any.whl (8.1 kB)
    Downloading pyphen-0.17.2-py3-none-any.whl (2.1 MB)
                                                 2.1/2.1 MB 23.4 MB/s eta 0:00:00
    Downloading tinyhtml5-2.0.0-py3-none-any.whl (39 kB)
    Downloading Brotli-1.1.0-cp311-cp311-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (2.9 MB)
                                                · 2.9/2.9 MB 22.5 MB/s eta 0:00:00
    Downloading zopfli-0.2.3.post1-cp311-cp311-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (850 kB)
                                                 850.6/850.6 kB 36.5 MB/s eta 0:00:00
    Installing collected packages: brotli, zopfli, tinyhtml5, Pyphen, pydyf, markdown2, cssselect2, weasyprint
    Successfully installed Pyphen-0.17.2 brotli-1.1.0 cssselect2-0.7.0 markdown2-2.5.3 pydyf-0.11.0 tinyhtml5-2.0.0 weasyprint-64.1 zopfli-0
    4
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import markdown2
import weasyprint
from google.colab import files
uploaded = files.upload()
    Choose Files No file chosen
                                      Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to
     anahla
#import pandas as pd
df = pd.read_csv("diabetes_prediction_dataset.csv")
df.info()
df.head()
```

```
→ <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 100000 entries, 0 to 99999
    Data columns (total 9 columns):
    # Column
                            Non-Null Count
                                             Dtype
        -----
    0
        gender
                             100000 non-null object
    1
        age
                             100000 non-null float64
    2
        hypertension
                             100000 non-null int64
                             100000 non-null
        heart_disease
                                             int64
                             100000 non-null object
        smoking_history
     5
        bmi
                             100000 non-null
                                             float64
        HbA1c_level
                             100000 non-null
                                             float64
        blood glucose level 100000 non-null int64
        diabetes
                             100000 non-null int64
    dtypes: float64(3), int64(4), object(2)
    memory usage: 6.9+ MB
```

	gender	age	hypertension	heart_disease	smoking_history	bmi	HbA1c_level	blood_glucose_level	diabetes
0	Female	80.0	0	1	never	25.19	6.6	140	0
1	Female	54.0	0	0	No Info	27.32	6.6	80	0
2	Male	28.0	0	0	never	27.32	5.7	158	0
3	Female	36.0	0	0	current	23.45	5.0	155	0
4	Male	76.0	1	1	current	20 14	4.8	155	0

print(df.isnull().sum())

_	gender	0		
	age	0		
	hypertension	0		
	heart_disease	0		
	smoking_history	0		
	bmi	0		
	HbA1c_level	0		
	blood_glucose_level			
	diabetes	0		
	dtype: int64			

display(df.describe())

₹

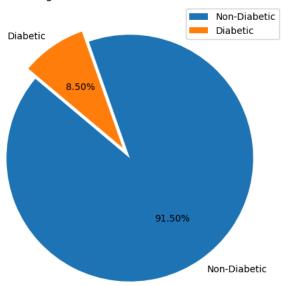
```
age hypertension heart_disease
                                                              bmi
                                                                    HbA1c_level blood_glucose_level
                                                                                                             diabetes
                                                                                         100000.000000 100000.000000
count 100000.000000
                      100000.00000 100000.000000 100000.000000 100000.000000
           41.885856
                           0.07485
                                          0.039420
                                                        27.320767
                                                                        5 527507
                                                                                            138.058060
                                                                                                             0.085000
mean
                                                                                                             0.278883
 std
           22.516840
                           0.26315
                                          0.194593
                                                         6.636783
                                                                        1.070672
                                                                                             40.708136
            0.080000
                           0.00000
                                          0.000000
                                                        10.010000
                                                                        3.500000
                                                                                             80.000000
                                                                                                             0.000000
min
                           0.00000
                                                                                            100.000000
25%
           24.000000
                                          0.000000
                                                        23.630000
                                                                        4.800000
                                                                                                             0.000000
50%
           43.000000
                           0.00000
                                          0.000000
                                                        27.320000
                                                                        5.800000
                                                                                            140.000000
                                                                                                             0.000000
75%
           60.000000
                           0.00000
                                          0.000000
                                                        29.580000
                                                                        6.200000
                                                                                            159.000000
                                                                                                             0.000000
           80.000000
                           1.00000
                                          1.000000
                                                        95.690000
                                                                        9.000000
                                                                                            300.000000
                                                                                                             1.000000
max
```

```
# Count occurrences of diabetic vs non-diabetic cases
diabetes_counts = df['diabetes'].value_counts().to_dict()  # Convert to dictionary to avoid KeyError

# Ensure both labels exist in the correct order
sizes = [diabetes_counts.get(0, 0), diabetes_counts.get(1, 0)]  # Handles cases where values might be missing
labels = ['Non-Diabetic', 'Diabetic']
colors = ['#1f77b4', '#ff7f0e']  # Custom colors for clarity
explode = (0, 0.1)  # Slightly separate "Diabetic" slice for emphasis

# Create the pie chart
plt.figure(figsize=(6, 6))
plt.pie(sizes, labels=labels, colors=colors, autopct='%1.2f%%', startangle=140, explode=explode)
plt.title("Percentage of Diabetic vs Non-Diabetic Patients")
plt.legend(labels, loc="upper right")
plt.show()
```

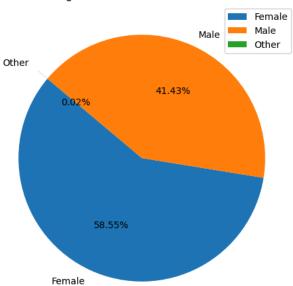

Percentage of Diabetic vs Non-Diabetic Patients



```
categorical_columns = ['gender', 'smoking_history'] # Adjust as needed
for col in categorical_columns:
    print(f"Unique values in {col}: {df[col].unique()}")
→ Unique values in gender: ['Female' 'Male' 'Other']
     Unique values in smoking_history: ['never' 'No Info' 'current' 'former' 'ever' 'not current']
# Count occurrences of each gender category
gender_counts = df['gender'].value_counts()
print(gender_counts)
     gender
     Female
                58552
                41430
     Male
     Other
                   18
     Name: count, dtype: int64
# Count occurrences of each gender category
gender_counts = df['gender'].value_counts()
# Data for the pie chart
labels = gender_counts.index.tolist() # Extract unique gender labels
sizes = gender_counts.values # Extract corresponding counts
colors = ['#1f77b4', '#ff7f0e', '#2ca02c'] # Assign colors
explode = (0, 0, 0.1) if 'Other' in labels else (0, 0) # Slightly separate 'Other' if it exists
# Create the pie chart
plt.figure(figsize=(6, 6))
plt.pie(sizes, labels=labels, colors=colors, autopct='%1.2f%%', startangle=140, explode=explode)
plt.title("Percentage of Gender Distribution in Dataset")
plt.legend(labels, loc="upper right")
plt.show()
```



Percentage of Gender Distribution in Dataset



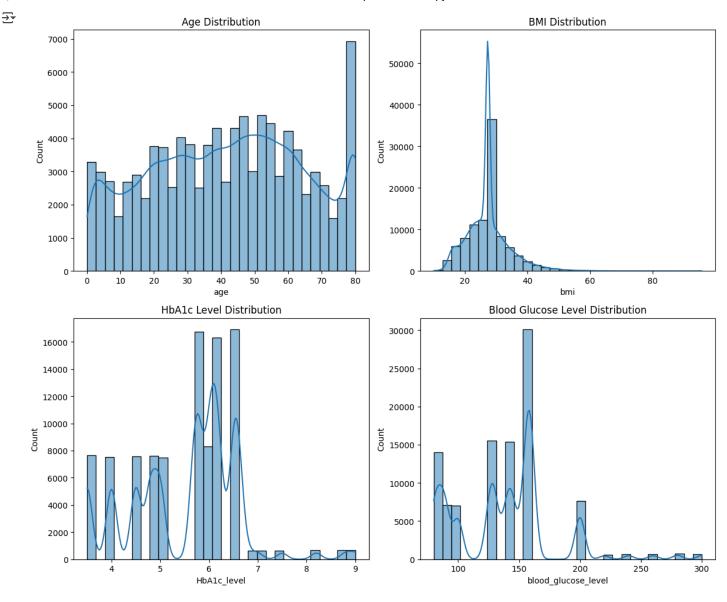
```
# Plot distributions
fig, axes = plt.subplots(2, 2, figsize=(12, 10))
sns.histplot(df['age'], bins=30, kde=True, ax=axes[0, 0])
axes[0, 0].set_title('Age Distribution')

sns.histplot(df['bmi'], bins=30, kde=True, ax=axes[0, 1])
axes[0, 1].set_title('BMI Distribution')

sns.histplot(df['HbA1c_level'], bins=30, kde=True, ax=axes[1, 0])
axes[1, 0].set_title('HbA1c Level Distribution')

sns.histplot(df['blood_glucose_level'], bins=30, kde=True, ax=axes[1, 1])
axes[1, 1].set_title('Blood Glucose Level Distribution')

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



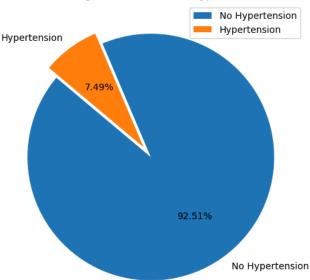
```
# Count occurrences of each hypertension category
hypertension_counts = df['hypertension'].value_counts()

# Data for the pie chart
labels = ['No Hypertension', 'Hypertension']
sizes = [hypertension_counts[0], hypertension_counts[1]] # Extract counts
colors = ['#1f77b4', '#ff7f0e'] # Custom colors for clarity
explode = (0, 0.1) # Slightly separate "Hypertension" slice for emphasis

# Create the pie chart
plt.figure(figsize=(6, 6))
plt.pie(sizes, labels=labels, colors=colors, autopct='%1.2f%%', startangle=140, explode=explode)
plt.title("Percentage of Patients with Hypertension")
plt.legend(labels, loc="upper right")
plt.show()
```



Percentage of Patients with Hypertension



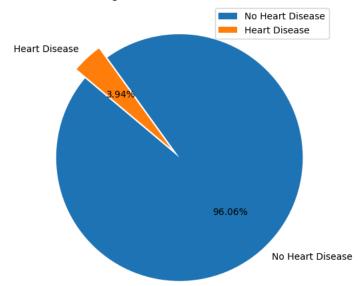
```
# Count occurrences of each heart disease category
heart_disease_counts = df['heart_disease'].value_counts()

# Data for the pie chart
labels = ['No Heart Disease', 'Heart Disease']
sizes = [heart_disease_counts[0], heart_disease_counts[1]] # Extract counts
colors = ['#1f77b4', '#ff7f0e'] # Custom colors for clarity
explode = (0, 0.1) # Slightly separate "Heart Disease" slice for emphasis

# Create the pie chart
plt.figure(figsize=(6, 6))
plt.pie(sizes, labels=labels, colors=colors, autopct='%1.2f%%', startangle=140, explode=explode)
plt.title("Percentage of Patients with Heart Disease")
plt.legend(labels, loc="upper right")
plt.show()
```



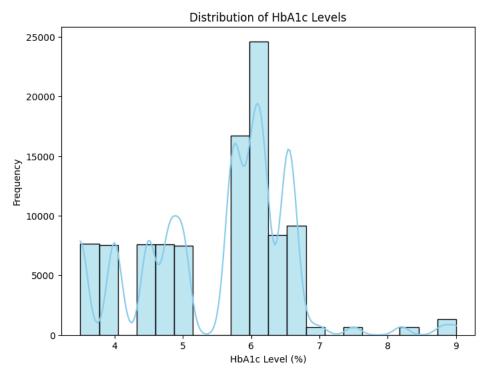
Percentage of Patients with Heart Disease



Count occurrences of each smoking history category
smoking_counts = df['smoking_history'].value_counts().reset_index()

```
smoking_counts.columns = ['Smoking History', 'Count']
# Calculate frequency (percentage)
smoking_counts['Frequency (%)'] = (smoking_counts['Count'] / df.shape[0]) * 100
# Display the table
print(smoking_counts)
<del>_</del>
       Smoking History Count Frequency (%)
               No Info 35816
                                       35.816
                                       35.095
                 never 35095
     1
                                        9.352
                         9352
                former
     2
     3
               current
                         9286
                                        9.286
           not current 6447
                                        6.447
                        4004
                                        4.004
                  ever
# Get the range of HbA1c levels
hbA1c_min = df['HbA1c_level'].min()
hbA1c_max = df['HbA1c_level'].max()
# Define HbA1c categories
hbA1c_bins = [0, 5.7, 6.4, float('inf')]
hbA1c\_labels = [' \le 5.7\% (Non-Diabetic)', '5.7\% - 6.4\% (Prediabetic)', ' \ge 6.5\% (Diabetic)']
# Categorize patients based on HbA1c levels
df['HbA1c_Category'] = pd.cut(df['HbA1c_level'], bins=hbA1c_bins, labels=hbA1c_labels, include_lowest=True)
# Count occurrences in each category
hbA1c_counts = df['HbA1c_Category'].value_counts().reset_index()
hbA1c_counts.columns = ['HbA1c Category', 'Count']
# Calculate percentage
\label{eq:hbA1c_counts['Counts['Count'] / df.shape[0]) * 100} \\ \text{hbA1c\_counts['Count'] / df.shape[0]) * 100} \\
# Display the results
print(f"HbA1c levels range from {hbA1c_min}% to {hbA1c_max}%.\n")
print(hbA1c_counts)
→ HbA1c levels range from 3.5% to 9.0%.
                   HbA1c Category Count Percentage (%)
            ≤ 5.7% (Non-Diabetic) 46270
                                                    46.270
     1 5.7% - 6.4% (Prediabetic) 32933
                                                    32.933
                                                    20.797
                ≥ 6.5% (Diabetic) 20797
# prompt: histogram for HbA1C
import matplotlib.pyplot as plt
# Plotting the histogram for HbA1c levels with a density curve
plt.figure(figsize=(8, 6))
sns.histplot(df['HbA1c_level'], bins=20, kde=True, color='skyblue') # Increased bins for better granularity
plt.title('Distribution of HbA1c Levels')
plt.xlabel('HbA1c Level (%)')
plt.ylabel('Frequency')
plt.show()
```

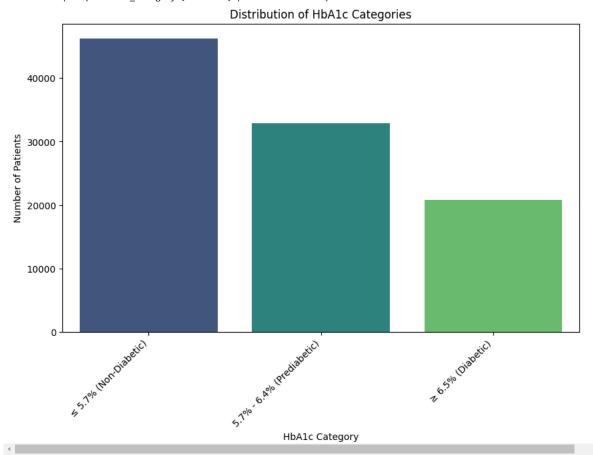




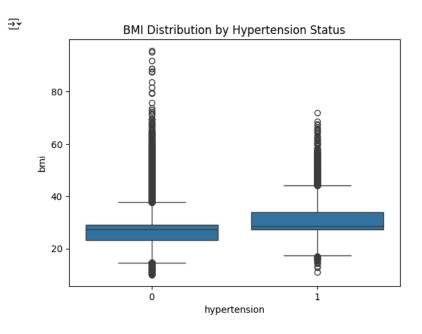
prompt: historgram for this About ≤ 5.7% being non-diabetic, 5.7% - 6.4% being prediabetic, and ≥ 6.5% being diabetic.
import matplotlib.pyplot as plt
Plotting the histogram for HbA1c categories
plt.figure(figsize=(10, 6))
sns.countplot(x='HbA1c_Category', data=df, palette='viridis')
plt.title('Distribution of HbA1c Categories')
plt.xlabel('HbA1c Category')
plt.ylabel('Number of Patients')
plt.xticks(rotation=45, ha='right') # Rotate x-axis labels for better readability
plt.show()

<ipython-input-32-6feb8259eba6>:6: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend sns.countplot(x='HbA1c_Category', data=df, palette='viridis')



import seaborn as sns
import matplotlib.pyplot as plt
sns.boxplot(x=df['hypertension'], y=df['bmi'])
plt.title("BMI Distribution by Hypertension Status")
plt.show()

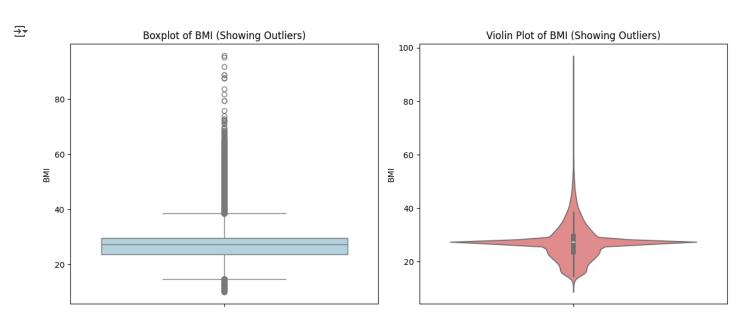


Set figure size
plt.figure(figsize=(12, 5))

```
# Create a boxplot for BMI to show outliers
plt.subplot(1, 2, 1)  # 1 row, 2 columns, position 1
sns.boxplot(y=df['bmi'], color='lightblue')
plt.title("Boxplot of BMI (Showing Outliers)")
plt.ylabel("BMI")

# Create a violin plot for BMI to show distribution and outliers
plt.subplot(1, 2, 2)  # 1 row, 2 columns, position 2
sns.violinplot(y=df['bmi'], color='lightcoral')
plt.title("Violin Plot of BMI (Showing Outliers)")
plt.ylabel("BMI")

# Show the plots
plt.tight_layout()
plt.show()
```

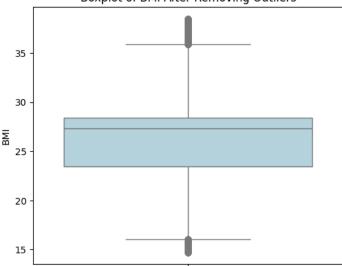


```
# Calculate IQR (Interquartile Range) for BMI
Q1 = df['bmi'].quantile(0.25) # 25th percentile
Q3 = df['bmi'].quantile(0.75) # 75th percentile
IQR = Q3 - Q1 # Interquartile Range
# Define lower and upper bounds for outliers
lower\_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR
# Remove outliers
df_cleaned = df[(df['bmi'] >= lower_bound) & (df['bmi'] <= upper_bound)]</pre>
# Print before and after removing outliers
print(f"Original dataset size: {df.shape[0]} rows")
print(f"Dataset size after removing outliers: {df_cleaned.shape[0]} rows")
# Save the cleaned dataset (optional)
df_cleaned.to_csv("cleaned_diabetes_dataset.csv", index=False)
    Original dataset size: 100000 rows
     Dataset size after removing outliers: 92914 rows
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
# Load dataset
file path = "diabetes prediction dataset.csv" # Update the path if needed
df = pd.read_csv(file_path)
```

```
# Calculate IQR (Interquartile Range) for BMI
Q1 = df['bmi'].quantile(0.25) # 25th percentile
Q3 = df['bmi'].quantile(0.75) # 75th percentile
IQR = Q3 - Q1 # Interquartile Range
# Define lower and upper bounds for outliers
lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR
# Remove outliers
df_cleaned = df[(df['bmi'] >= lower_bound) & (df['bmi'] <= upper_bound)]</pre>
# Plot the boxplot for BMI after removing outliers
plt.figure(figsize=(6, 5))
sns.boxplot(y=df_cleaned['bmi'], color='lightblue')
plt.title("Boxplot of BMI After Removing Outliers")
plt.ylabel("BMI")
# Show the plot
plt.show()
```

₹

Boxplot of BMI After Removing Outliers

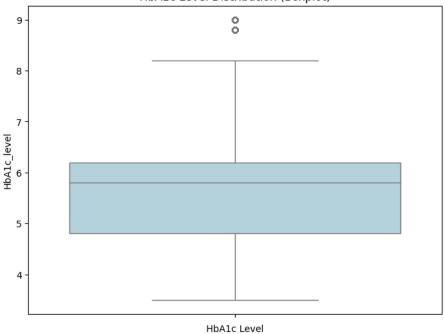


Start coding or generate with AI.

```
# Plot a boxplot for HbA1c level distribution
plt.figure(figsize=(8, 6))
sns.boxplot(y=df['HbA1c_level'], color="lightblue")
plt.title("HbA1c Level Distribution (Boxplot)")
plt.xlabel("HbA1c Level")
plt.show()
```



HbA1c Level Distribution (Boxplot)

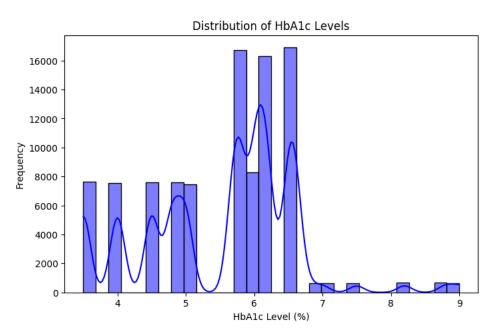


Start coding or generate with AI.

Start coding or $\underline{\text{generate}}$ with AI.

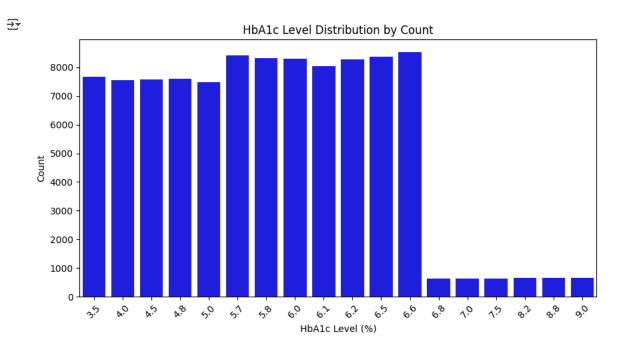
```
# Plot the distribution of HbA1c levels
plt.figure(figsize=(8, 5))
sns.histplot(df['HbA1c_level'], bins=30, kde=True, color='blue')
plt.title("Distribution of HbA1c Levels")
plt.xlabel("HbA1c Level (%)")
plt.ylabel("Frequency")
plt.show()
```





```
# Plot the distribution of HbA1c levels by count
plt.figure(figsize=(10, 5))
sns.barplot(x=hbA1c_distribution['HbA1c Level'], y=hbA1c_distribution['Count'], color='blue')
plt.title("HbA1c Level Distribution by Count")
plt.xlabel("HbA1c Level (%)")
```

```
plt.ylabel("Count")
plt.xticks(rotation=45)
plt.show()
```

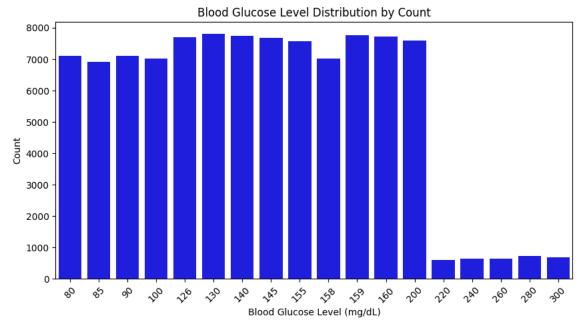


```
# Count occurrences of each Blood Glucose Level
blood_glucose_distribution = df['blood_glucose_level'].value_counts().reset_index()
blood_glucose_distribution.columns = ['Blood Glucose Level', 'Count']
```

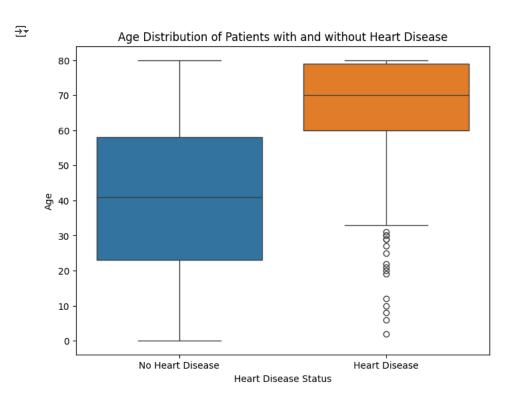
Display the distribution table
print(blood_glucose_distribution)

₹		Blood	Glucose	Level	Count
	0			130	7794
	1			159	7759
	2			140	7732
	3			160	7712
	4			126	7702
	5			145	7679
	6			200	7600
	7			155	7575
	8			90	7112
	9			80	7106
	10			158	7026
	11			100	7025
	12			85	6901
	13			280	729
	14			300	674
	15			240	636
	16			260	635
	17			220	603



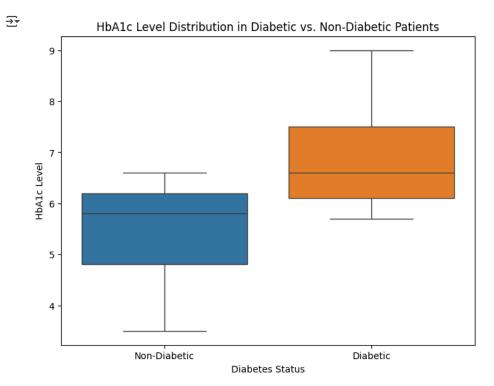


```
# Comparing the age distribution of patients with and without heart disease.
# Create a boxplot to compare age distribution between patients with and without heart disease
plt.figure(figsize=(8, 6))
sns.boxplot(x="heart_disease", y="age", data=df, hue="heart_disease", palette={0: "#1f77b4", 1: "#ff7f0e"}, legend=False)
plt.xticks([0, 1], ['No Heart Disease', 'Heart Disease'])
plt.title("Age Distribution of Patients with and without Heart Disease")
plt.xlabel("Heart Disease Status")
plt.ylabel("Age")
plt.show()
```

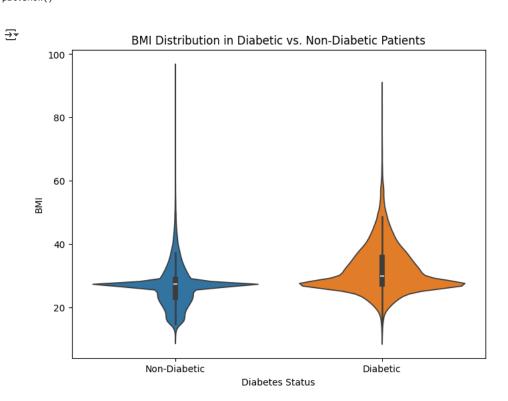


```
plt.figure(figsize=(8, 6))
sns.boxplot(x="diabetes", y="HbA1c_level", data=df, hue="diabetes", palette={0: "#1f77b4", 1: "#ff7f0e"}, legend=False)
plt.xticks([0, 1], ['Non-Diabetic', 'Diabetic'])
plt.title("HbA1c Level Distribution in Diabetic vs. Non-Diabetic Patients")
plt.xlabel("Diabetes Status")
plt.ylabel("HbA1c Level")
```

plt.show()



```
plt.figure(figsize=(8, 6))
sns.violinplot(x="diabetes", y="bmi", data=df, hue="diabetes", palette={0: "#1f77b4", 1: "#ff7f0e"}, legend=False)
plt.xticks([0, 1], ['Non-Diabetic', 'Diabetic'])
plt.title("BMI Distribution in Diabetic vs. Non-Diabetic Patients")
plt.xlabel("Diabetes Status")
plt.ylabel("BMI")
plt.show()
```



```
import pandas as pd

# Load the dataset
file_path = "diabetes_prediction_dataset.csv"  # Update if needed
```

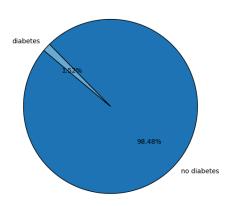
```
df_new = pd.read_csv(file_path)
# Define HbA1c categories for initial diagnosis
hbA1c_bins = [0, 5.7, 6.4, float('inf')]
hbA1c_labels = ['Normal', 'Prediabetes', 'Diabetes']
\label{lower} $$ df_new['HbA1c_Category'] = pd.cut(df_new['HbA1c_level'], bins=hbA1c_bins, labels=hbA1c_labels, include_lowest=True) $$ df_new['HbA1c_level'], bins=hbA1c_bins, labels=hbA1c_labels, include_lowest=True) $$ df_new['HbA1c_level'], bins=hbA1c_labels, include_lowest=True) $$ df_new['HbA1c_level'], bins=hbA1c_labels,
# Calculate the percentage of actual diabetes cases within each HbA1c category
hbA1c_diabetes_distribution = df_new.groupby('HbA1c_Category', observed=False)['diabetes'].mean() * 100
# Convert to DataFrame for better visualization
hbA1c_diabetes_summary = hbA1c_diabetes_distribution.reset_index()
hbA1c_diabetes_summary.columns = ['HbA1c Level', 'Percentage with Diabetes']
# Display the summary
print(hbA1c_diabetes_summary)
 ₹
                HbA1c Level Percentage with Diabetes
                                                                                 1.521504
                           Normal
                                                                                 7,909999
           1 Prediabetes
                        Diabetes
                                                                                24.960331
# Define diabetes status labels
labels = ['no diabetes', 'diabetes']
colors = ['#1f77b4', '#6baed6'] # Different shades of blue
# Create subplots for each HbA1c category
fig, axes = plt.subplots(1, 3, figsize=(15, 5))
# Loop through each category and plot a pie chart
for i, category in enumerate(hbA1c_diabetes_summary['HbA1c Level']):
         diabetes_percentage = hbA1c_diabetes_summary.loc[hbA1c_diabetes_summary['HbA1c Level'] == category, 'Percentage with Diabetes'].values[@
         no_diabetes_percentage = 100 - diabetes_percentage
         axes[i].pie([no_diabetes_percentage, diabetes_percentage],
                                   labels=labels,
                                   autopct='%1.2f%%',
                                   colors=colors.
                                   startangle=140,
                                   wedgeprops={'edgecolor': 'black'})
         axes[i].set_title(f"Initial Diagnosis = {category}")
# Adjust layout and show the plots
plt.tight_layout()
plt.show()
```

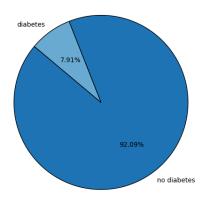


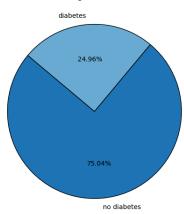










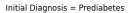


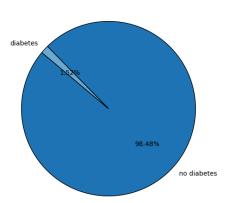
Double-click (or enter) to edit

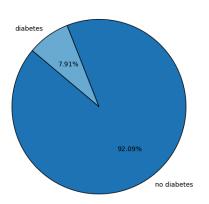
```
# prompt: add the title # Define diabetes status labels
# labels = ['no diabetes', 'diabetes']
# colors = ['#1f77b4', '#6baed6'] # Different shades of blue
# # Create subplots for each HbA1c category
# fig, axes = plt.subplots(1, 3, figsize=(15, 5))
# # Loop through each category and plot a pie chart
# for i, category in enumerate(hbA1c_diabetes_summary['HbA1c Level']):
      diabetes_percentage = hbA1c_diabetes_summary.loc[hbA1c_diabetes_summary['HbA1c Level'] == category, 'Percentage with Diabetes'].values
      no_diabetes_percentage = 100 - diabetes_percentage
#
#
      axes[i].pie([no_diabetes_percentage, diabetes_percentage],
#
                  labels=labels,
#
                  autopct='%1.2f%%',
#
                  colors=colors,
                  startangle=140,
                  wedgeprops={'edgecolor': 'black'})
      axes[i].set_title(f"Initial Diagnosis = {category}")
#
# # Adjust layout and show the plots
# plt.tight_layout()
# plt.show()
import matplotlib.pyplot as plt
# Define diabetes status labels
labels = ['no diabetes', 'diabetes']
colors = ['#1f77b4', '#6baed6'] # Different shades of blue
# Create subplots for each HbA1c category
fig, axes = plt.subplots(1, 3, figsize=(15, 5))
# Loop through each category and plot a pie chart
for i, category in enumerate(hbA1c_diabetes_summary['HbA1c Level']):
    diabetes_percentage = hbA1c_diabetes_summary.loc[hbA1c_diabetes_summary['HbA1c Level'] == category, 'Percentage with Diabetes'].values[@
    no_diabetes_percentage = 100 - diabetes_percentage
    axes[i].pie([no_diabetes_percentage, diabetes_percentage],
                labels=labels,
                autopct='%1.2f%%',
                colors=colors,
                startangle=140,
                wedgeprops={'edgecolor': 'black'})
    axes[i].set_title(f"Initial Diagnosis = {category}")
# Adjust layout and show the plots
plt.tight_layout()
plt.show()
```

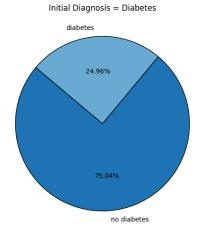


Initial Diagnosis = Normal



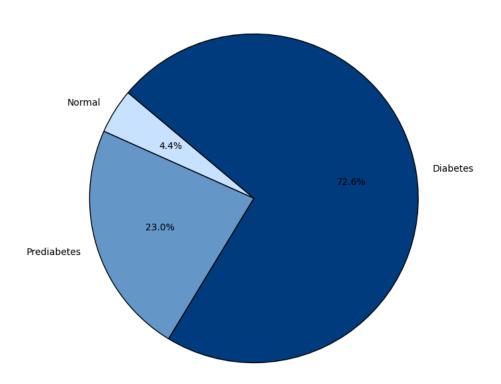




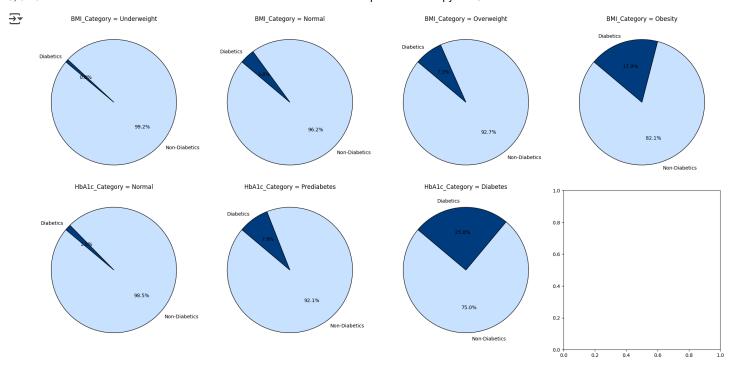




Percentage of Diabetes Cases by HbA1c Level



```
import matplotlib.pyplot as plt
# Define diabetes status labels and shades of blue
labels = ['Non-Diabetics', 'Diabetics']
colors = ['#cce5ff', '#004080'] # Light to dark blue shades
# Define ordinal categories
ordinal_categories = ['BMI_Category', 'HbA1c_Category']
# Create subplots for each ordinal category
fig, axes = plt.subplots(len(ordinal_categories), len(df_new['BMI_Category'].unique()), figsize=(20, 10))
# Loop through each ordinal category and plot a pie chart
for row, category in enumerate(ordinal_categories):
   for col, cat in enumerate(category_distribution.index):
       diabetes_percentage = category_distribution.loc[cat, 1] if 1 in category_distribution.columns else 0
       non_diabetes_percentage = category_distribution.loc[cat, 0] if 0 in category_distribution.columns else 100
       axes[row, col].pie([non_diabetes_percentage, diabetes_percentage],
                        labels=labels,
                        autopct='%1.1f%%',
                        colors=colors,
                        startangle=140,
                        wedgeprops={'edgecolor': 'black'})
       axes[row, col].set_title(f"{category} = {cat}")
# Adjust layout and show the plots
plt.tight_layout()
plt.show()
```



```
import matplotlib.pyplot as plt
# Define diabetes status labels and shades of blue
labels = ['Non-Diabetics', 'Diabetics']
colors = ['#cce5ff', '#004080'] # Light to dark blue shades
# Create subplots for BMI categories
fig, axes = plt.subplots(1, len(df_new['BMI_Category'].unique()), figsize=(20, 5))
# Group data by BMI category and diabetes status
bmi_distribution = df_new.groupby('BMI_Category', observed=False)['diabetes'].value_counts(normalize=True).unstack() * 100
# Loop through each BMI category and plot a pie chart
for col, category in enumerate(bmi_distribution.index):
    \label{eq:diabetes_percentage} \mbox{ = bmi\_distribution.loc[category, 1] if 1 in bmi\_distribution.columns else 0 }
    non\_diabetes\_percentage = bmi\_distribution.loc[category, 0] if 0 in bmi\_distribution.columns else 100
    axes[col].pie([non_diabetes_percentage, diabetes_percentage],
                  labels=labels,
                  autopct='%1.1f%%',
                  colors=colors,
                  startangle=140,
                  wedgeprops={'edgecolor': 'black'})
```

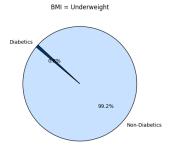
```
axes[col].set_title(f"BMI = {category}")

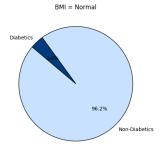
# Add overall title
plt.suptitle("Diabetes and Non-Diabetes Distribution Across BMI Categories", fontsize=16, fontweight='bold')

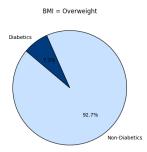
# Adjust layout and show the plots
plt.tight_layout(rect=[0, 0.03, 1, 0.95])
plt.show()
```

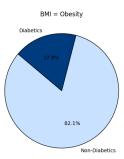


Diabetes and Non-Diabetes Distribution Across BMI Categories





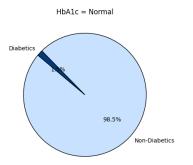


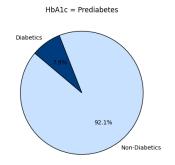


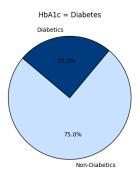
```
import matplotlib.pyplot as plt
# Define diabetes status labels and shades of blue
labels = ['Non-Diabetics', 'Diabetics']
colors = ['#cce5ff', '#004080'] # Light to dark blue shades
# Create subplots for HbA1c categories
fig, axes = plt.subplots(1, len(df_new['HbA1c_Category'].unique()), figsize=(20, 5))
# Group data by HbA1c category and diabetes status
hba1c_distribution = df_new.groupby('HbA1c_Category', observed=False)['diabetes'].value_counts(normalize=True).unstack() * 100
# Loop through each HbA1c category and plot a pie chart
for col, category in enumerate(hba1c_distribution.index):
    diabetes_percentage = hba1c_distribution.loc[category, 1] if 1 in hba1c_distribution.columns else 0
    non_diabetes_percentage = hba1c_distribution.loc[category, 0] if 0 in hba1c_distribution.columns else 100
    axes[col].pie([non_diabetes_percentage, diabetes_percentage],
                  labels=labels,
                  autopct='%1.1f%%',
                  colors=colors,
                  startangle=140,
                  wedgeprops={'edgecolor': 'black'})
    axes[col].set_title(f"HbA1c = {category}")
# Add overall title
plt.suptitle("Diabetes and Non-Diabetes Distribution Across HbA1c Categories", fontsize=16, fontweight='bold')
# Adjust layout and show the plots
plt.tight_layout(rect=[0, 0.03, 1, 0.95])
plt.show()
```



Diabetes and Non-Diabetes Distribution Across HbA1c Categories







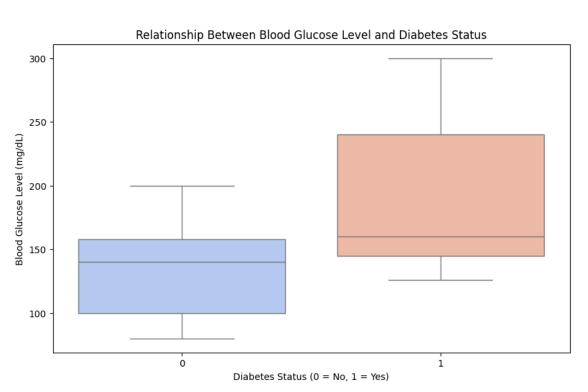
```
import matplotlib.pyplot as plt
import seaborn as sns

# Plot the distribution of Blood Glucose Levels vs. Diabetes Status
plt.figure(figsize=(10, 6))
sns.boxplot(x=df_new['diabetes'], y=df_new['blood_glucose_level'], hue=df_new['diabetes'], palette="coolwarm", legend=False)

# Set plot labels and title
plt.title("Relationship Between Blood Glucose Level and Diabetes Status")
plt.xlabel("Diabetes Status (0 = No, 1 = Yes)")
plt.ylabel("Blood Glucose Level (mg/dL)")

# Show the plot
plt.show()
```



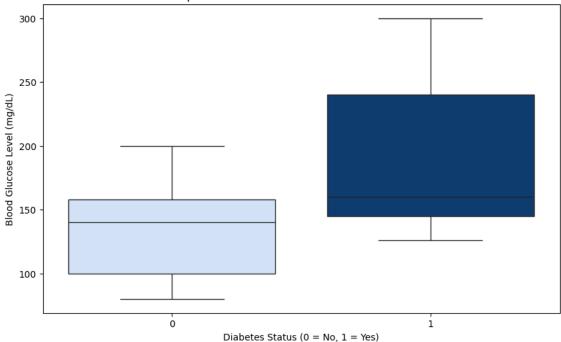


import matplotlib.pyplot as plt
import seaborn as sns

```
# Plot the distribution of Blood Glucose Levels vs. Diabetes Status using shades of blue
plt.figure(figsize=(10, 6))
sns.boxplot(x=df_new['diabetes'], y=df_new['blood_glucose_level'], hue=df_new['diabetes'],
           palette=['#cce5ff', '#004080'], legend=False) # Light to dark blue shades
# Set plot labels and title
plt.title("Relationship Between Blood Glucose Level and Diabetes Status")
plt.xlabel("Diabetes Status (0 = No, 1 = Yes)")
plt.ylabel("Blood Glucose Level (mg/dL)")
# Show the plot
plt.show()
```



Relationship Between Blood Glucose Level and Diabetes Status



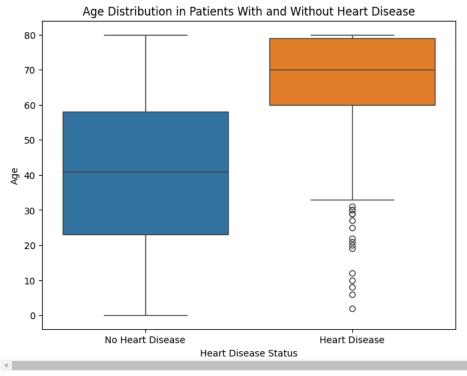
Start coding or generate with AI. Start coding or $\underline{\text{generate}}$ with AI. Start coding or generate with AI. plt.figure(figsize=(8, 6)) sns.boxplot(x=df['heart_disease'], y=df['age'], palette=["#1f77b4", "#ff7f0e"]) plt.xticks([0, 1], ['No Heart Disease', 'Heart Disease']) plt.title("Age Distribution in Patients With and Without Heart Disease") plt.xlabel("Heart Disease Status")

Start coding or generate with AI.

plt.ylabel("Age") plt.show()

```
<ipython-input-39-32f56af5aab9>:2: FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend sns.boxplot(x=df['heart_disease'], y=df['age'], palette=["#1f77b4", "#ff7f0e"])



```
# Define HbA1c level categories based on diabetes classification
hba1c_bins = [0, 5.7, 6.4, float('inf')]
hba1c_labels = ['Normal', 'Prediabetes', 'Diabetes']
df['HbA1c_Category'] = pd.cut(df['HbA1c_level'], bins=hba1c_bins, labels=hba1c_labels)
# Calculate percentage of diabetics in each HbA1c category
hba1c_diabetes_counts = df.groupby('HbA1c_Category', observed=True)['diabetes'].mean() * 100
# Create a structured table
hba1c_table = pd.DataFrame({
    'HbA1c Level Category': hba1c_labels,
    'Prediction': [f"{hba1c_diabetes_counts.get(category, 0):.2f}% have diabetes" for category in hba1c_labels]
})
# Display the table
print(hba1c_table)
      HbA1c Level Category
                                       Prediction
\overline{2}
                     Normal
                              1.52% have diabetes
                Prediabetes 7.91% have diabetes
     1
     2
                   Diabetes 24.96% have diabetes
# Define HbA1c level categories
hba1c_bins = [0, 5.7, 6.4, float('inf')]
hba1c_labels = ['< 5.7', '5.7 - 6.4', '>= 6.5']
df['HbA1c_Category'] = pd.cut(df['HbA1c_level'], bins=hba1c_bins, labels=hba1c_labels, include_lowest=True)
# Calculate percentage of diabetics in each HbA1c category
hba1c_diabetes_counts = df.groupby('HbA1c_Category', observed=True)['diabetes'].mean() * 100
# Create a structured table
hba1c_table = pd.DataFrame({
    'HbA1c Level': hba1c_labels,
    'Prediction': [f"{hbalc_diabetes_counts.get(category, 0):.2f}% have diabetes" for category in hbalc_labels]
})
# Display the table
print(hba1c_table)
```

```
HbA1c Level
                            Prediction
    0 < 5.7 1.52% have diabetes
    1 5.7 - 6.4 7.91% have diabetes
2 >= 6.5 24.96% have diabetes
# Define BMI categories
bmi_bins = [0, 18.5, 24.9, 29.9, float('inf')]
bmi_labels = ['Underweight', 'Normal', 'Overweight', 'Obesity']
df['BMI_Category'] = pd.cut(df['bmi'], bins=bmi_bins, labels=bmi_labels, include_lowest=True)
# Calculate percentage of diabetics in each BMI category
bmi_diabetes_counts = df.groupby('BMI_Category', observed=True)['diabetes'].mean() * 100
# Create a structured table
bmi_table = pd.DataFrame({
    'BMI Category': bmi_labels,
    'Prediction': [f"{bmi_diabetes_counts.get(category, 0):.2f}% have diabetes" for category in bmi_labels]
})
# Display the table
print(bmi_table)
      BMI Category
                             Prediction
    0 Underweight
                    0.75% have diabetes
          Normal 3.84% have diabetes
    1
    2 Overweight 7.25% have diabetes
    3
           Obesity 17.91% have diabetes
# Define Blood Glucose Level categories
glucose_bins = [0, 99, 125, float('inf')]
glucose_labels = ['≤ 99', '100 - 125', '≥ 126']
df['Glucose_Category'] = pd.cut(df['blood_glucose_level'], bins=glucose_bins, labels=glucose_labels, include_lowest=True)
# Calculate percentage of diabetics in each Blood Glucose Level category
glucose_diabetes_counts = df.groupby('Glucose_Category', observed=True)['diabetes'].mean() * 100
# Create a structured table
glucose_table = pd.DataFrame({
    'Blood Glucose Level': glucose_labels,
    'Prediction': [f"{glucose_diabetes_counts.get(category, 0):.2f}% have diabetes" for category in glucose_labels]
})
# Display the table
print(glucose_table)
₹
      Blood Glucose Level
                                   Prediction
                   ≤ 99
                          0.00% have diabetes
               100 - 125 0.00% have diabetes
    1
                   ≥ 126 11.83% have diabetes
Double-click (or enter) to edit
# Define Blood Glucose Level categories
glucose_bins = [0, 99, 125, float('inf')]
glucose_labels = ['≤ 99', '100 - 125', '≥ 126']
# Plot a bar chart for blood glucose categories (Fixing FutureWarning)
plt.figure(figsize=(8, 6))
sns.countplot(x="Glucose_Category", data=df, hue="Glucose_Category", palette="coolwarm", legend=False)
```