

```
!pip install faker
```

```
Collecting faker
  Downloading Faker-23.1.0-py3-none-any.whl (1.7 MB)
    ━━━━━━━━━━━━━━━━━━━━━━━━━━━━ 1.7/1.7 MB 1.9 MB/s eta 0:00:00
Requirement already satisfied: python-dateutil>=2.4 in /usr/local/lib/python3.10/dist-packages (from faker) (2.8.2)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.4->faker) (1.16.0)
Installing collected packages: faker
Successfully installed faker-23.1.0
```

```

import random
import pandas as pd
from faker import Faker

fake = Faker()

# Generate data for 1000 patients
data = []
for _ in range(1000):
    name = fake.name()
    age = random.randint(1, 90) if random.random() < 0.95 else None # Introduce missing values in age
    if age is not None:
        if age <= 18:
            age_group = 'Child' if age <= 12 else 'Teenager'
        else:
            age_group = 'Adult'
    else:
        age_group = 'Unknown'

    gender = random.choice(['Male', 'Female'])
    race = random.choice(['Asian', 'Black', 'Hispanic', 'White', 'Other'])
    blood_group = random.choice(['A+', 'A-', 'B+', 'B-', 'AB+', 'AB-', 'O+', 'O-'])
    temperature = round(random.uniform(97.0, 99.0), 1)
    blood_pressure_sys = random.randint(90, 180) # Allow for outliers
    blood_pressure_dia = random.randint(60, 120) # Allow for outliers
    height = round(random.uniform(120, 200), 2) # in centimeters
    weight = round(random.uniform(40, 150), 2) # in kilograms

    # Simulate common health concerns based on age and gender
    common_health_concerns = []
    if age_group == 'Child':
        common_health_concerns.append('Vaccination')
        common_health_concerns.append('Growth and development')
    elif age_group == 'Teenager':
        common_health_concerns.append('Acne and skin issues')
        common_health_concerns.append('Mental health challenges')
        common_health_concerns.append('Sexual health education')
    elif age_group == 'Adult':
        common_health_concerns.append('Chronic diseases management')
        common_health_concerns.append('Heart health monitoring')
        common_health_concerns.append('Diet and exercise counseling')

    # Simulate additional health concerns
    additional_health_concerns = [fake.random.choice(['Asthma', 'Diabetes', 'Hypertension', 'Obesity', 'Allergies', 'Anxiety', 'Depression'])]

    # Combine common and additional health concerns
    health_concerns = common_health_concerns + additional_health_concerns

    # Determine if the patient needs medical attention based on health concerns and vital signs
    if any(condition in health_concerns for condition in ['Hypertension', 'Diabetes', 'Asthma', 'Allergies', 'Depression']):
        needs_medical_attention = 1
    elif blood_pressure_sys > 140 or blood_pressure_dia > 90 or weight > 100:
        needs_medical_attention = 1
    else:
        needs_medical_attention = 0

    # Simulate patients visiting for general checkups
    if random.random() < 0.2: # 20% chance of visiting for a general checkup
        health_concerns = 'General Checkup'
        needs_medical_attention = 0

    data.append({
        'Name': name,
        'Age': age,
        'Age Group': age_group,
        'Gender': gender,
        'Race': race,
        'Blood Group': blood_group,
        'Temperature': temperature,
        'Blood Pressure (Systolic)': blood_pressure_sys,
        'Blood Pressure (Diastolic)': blood_pressure_dia,
        'Height': height,
        'Weight': weight,
        'Health Concerns': ', '.join(health_concerns) if not isinstance(health_concerns, str) else health_concerns, # Convert list to
        'Needs Medical Attention': needs_medical_attention
    })
}

# Create a DataFrame
df = pd.DataFrame(data)

# Save the dataset to a CSV file

```

```
df.to_csv('medical_data_v3.csv', index=False)

from google.colab import drive
drive.mount('/content/drive')

Mounted at /content/drive

%cd Healthcare

[Errno 2] No such file or directory: 'Healthcare'
/content

%ls

drive/ medical_data_v2.csv medical_data_v3.csv newmedical_data.csv sample_data/
```

```
import pandas as pd

# Assuming your dataset is stored in a CSV file named 'medical_data.csv'
# Replace 'medical_data.csv' with the actual file path if it's located in a different directory
file_path = 'medical_data_v3.csv'

# Read the CSV file into a DataFrame
df = pd.read_csv(file_path)

# Display the first few rows of the DataFrame to verify the import
print("First few rows of the dataset:")
print(df.head())
```

```
First few rows of the dataset:
      Name   Age  Age Group Gender     Race Blood Group \
0    Kevin Wong  29.0   Adult Female  Black    AB+
1  Mr. David Meyers  84.0   Adult   Male  White     A-
2  Robert Sanchez DDS  26.0   Adult   Male Hispanic    A+
3    Brent Moore  42.0   Adult   Male  Black     B+
4  Alicia Kennedy  24.0   Adult   Male  Black     A+

      Temperature  Blood Pressure (Systolic)  Blood Pressure (Diastolic)  Height \
0            97.9                  178                      84        183.19
1            98.1                  130                      88        170.96
2            97.9                  121                     101        146.33
3            98.1                  169                     112        197.98
4            98.0                  100                      61        149.24

      Weight          Health Concerns \
0   73.15  Chronic diseases management, Heart health moni...
1   92.36  Chronic diseases management, Heart health moni...
2   77.41  Chronic diseases management, Heart health moni...
3  113.40  Chronic diseases management, Heart health moni...
4   79.90  Chronic diseases management, Heart health moni...

      Needs Medical Attention
0                           1
1                           1
2                           1
3                           1
4                           1
```

```
df
```

	Name	Age	Age Group	Gender	Race	Blood Group	Temperature	Blood Pressure (Systolic)	Blood Pressure (Diastolic)
0	Kevin Wong	29.0	Adult	Female	Black	AB+	97.9	178	
1	Mr. David Meyers	84.0	Adult	Male	White	A-	98.1	130	
2	Robert Sanchez DDS	26.0	Adult	Male	Hispanic	A+	97.9	121	
3	Brent Moore	42.0	Adult	Male	Black	B+	98.1	169	
4	Alicia Kennedy	24.0	Adult	Male	Black	A+	98.0	100	
...
995	Stacy Gallegos	36.0	Adult	Female	Black	A+	97.9	159	
996	Natalie Curtis	15.0	Teenager	Female	Asian	AB+	98.0	97	
997	Taylor Meyer	21.0	Adult	Male	Hispanic	A+	97.6	119	

```

import pandas as pd

breakdown_data = {
    'Column': ['Name', 'Age', 'Age Group', 'Gender', 'Race', 'Blood Group', 'Temperature', 'Blood Pressure (Systolic)', 'Blood Pressure (Diastolic)', 'Height', 'Weight', 'Health Concerns', 'Needs Medical Attention'],
    'Data Type': ['Nominal', 'Quantitative', 'Discrete', 'Nominal', 'Nominal', 'Nominal', 'Continuous', 'Continuous', 'Continuous', 'Continuous', 'Continuous', 'Nominal', 'Binary']
}

breakdown_df = pd.DataFrame(breakdown_data)

# Display the breakdown DataFrame
print("Breakdown of Dataset:")
print(breakdown_df)

```

	Breakdown of Dataset:		Data Type
0	Name		Nominal
1	Age	Quantitative, Discrete	
2	Age Group		Nominal
3	Gender		Nominal
4	Race		Nominal
5	Blood Group		Nominal
6	Temperature		Continuous
7	Blood Pressure (Systolic)		Continuous
8	Blood Pressure (Diastolic)		Continuous
9	Height		Continuous
10	Weight		Continuous
11	Health Concerns		Nominal
12	Needs Medical Attention		Binary

```
print (df)
```

	Name	Age	Age Group	Gender	Race	Blood Group	\
0	Kevin Wong	29.0	Adult	Female	Black	AB+	

1	Mr. David Meyers	84.0	Adult	Male	White	A-
2	Robert Sanchez DDS	26.0	Adult	Male	Hispanic	A+
3	Brent Moore	42.0	Adult	Male	Black	B+
4	Alicia Kennedy	24.0	Adult	Male	Black	A+
..
995	Stacy Gallegos	36.0	Adult	Female	Black	A+
996	Natalie Curtis	15.0	Teenager	Female	Asian	AB+
997	Taylor Meyer	21.0	Adult	Male	Hispanic	A+
998	Erin Morales	89.0	Adult	Male	Hispanic	A+
999	Gary Harper	1.0	Child	Male	Asian	B+

	Temperature	Blood Pressure (Systolic)	Blood Pressure (Diastolic)	\
0	97.9	178	84	
1	98.1	130	88	
2	97.9	121	101	
3	98.1	169	112	
4	98.0	100	61	
..	
995	97.9	159	68	
996	98.0	97	69	
997	97.6	119	112	
998	98.3	93	84	
999	98.0	128	81	

	Height	Weight	Health Concerns	\
0	183.19	73.15	Chronic diseases management, Heart health moni...	
1	170.96	92.36	Chronic diseases management, Heart health moni...	
2	146.33	77.41	Chronic diseases management, Heart health moni...	
3	197.98	113.40	Chronic diseases management, Heart health moni...	
4	149.24	79.90	Chronic diseases management, Heart health moni...	
..	
995	162.34	136.16	Chronic diseases management, Heart health moni...	
996	182.38	55.07	Acne and skin issues, Mental health challenges...	
997	185.71	117.14	General Checkup	
998	144.21	109.79	General Checkup	
999	194.56	77.49	General Checkup	

	Needs Medical Attention
0	1
1	1
2	1
3	1
4	1
..	...
995	1
996	0
997	0
998	0
999	0

[1000 rows x 13 columns]

```

import matplotlib.pyplot as plt
import seaborn as sns

# Initial Visualization
# Histograms for numerical variables
df.hist(figsize=(10, 8))
plt.tight_layout()
plt.show()

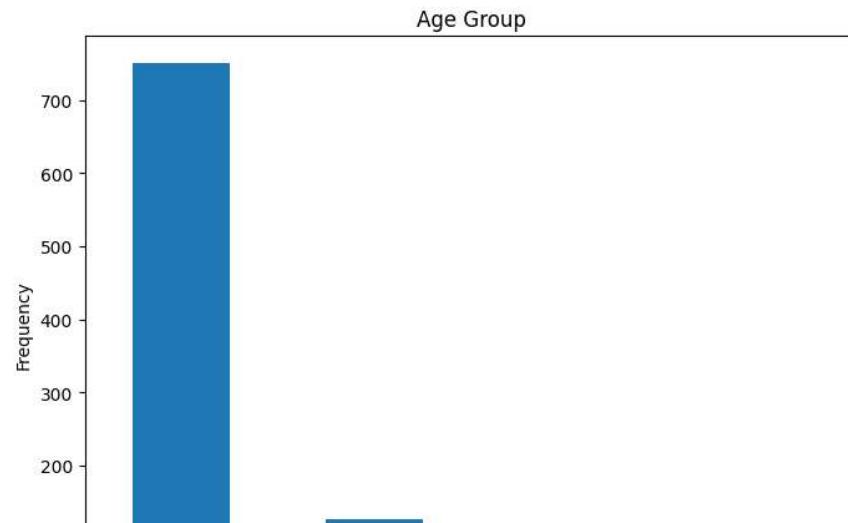
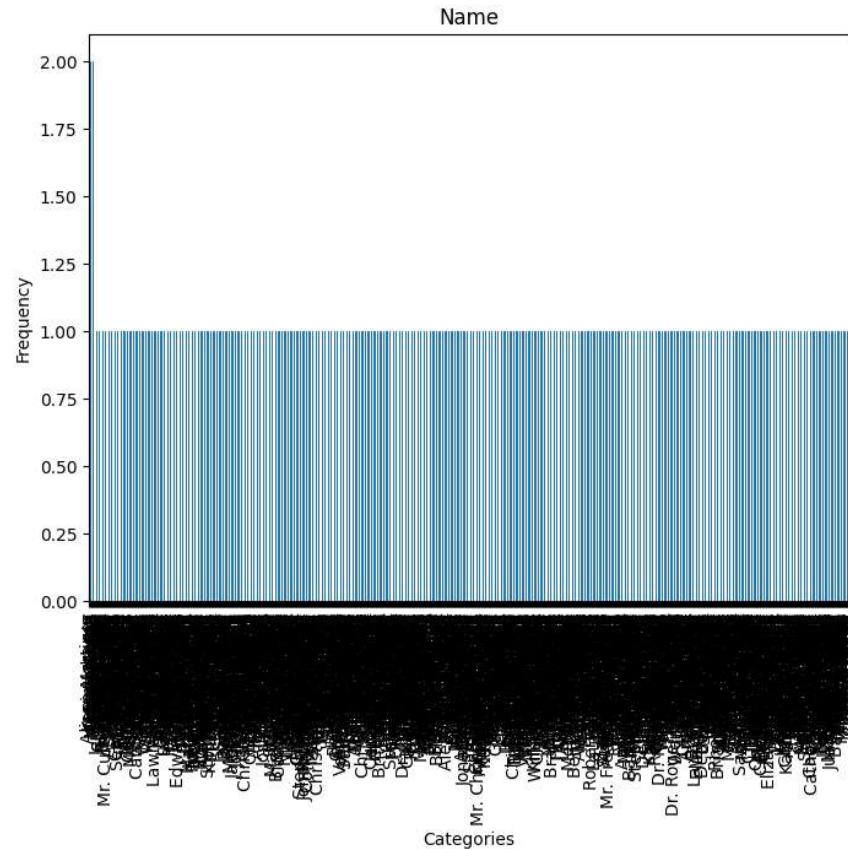
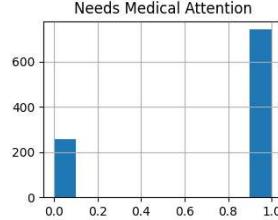
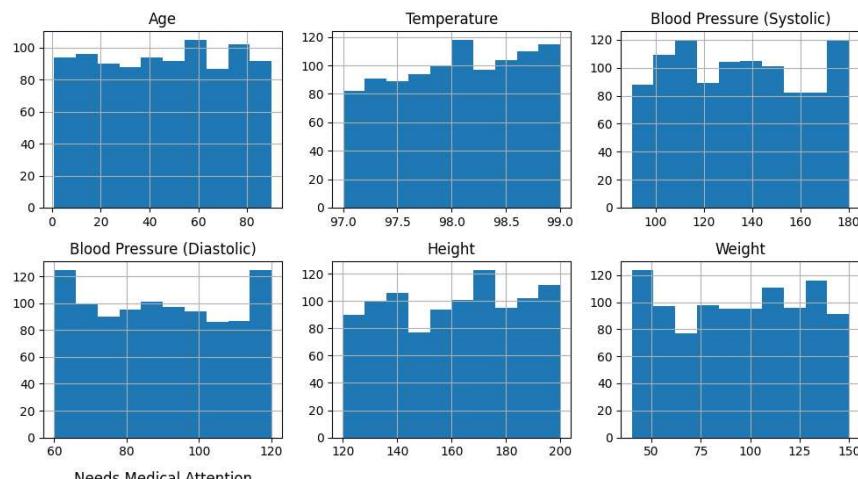
# Bar plots for categorical variables
categorical_cols = df.select_dtypes(include=['object']).columns
for col in categorical_cols:
    df[col].value_counts().plot(kind='bar', figsize=(8, 6))
    plt.title(col)
    plt.xlabel('Categories')
    plt.ylabel('Frequency')
    plt.show()

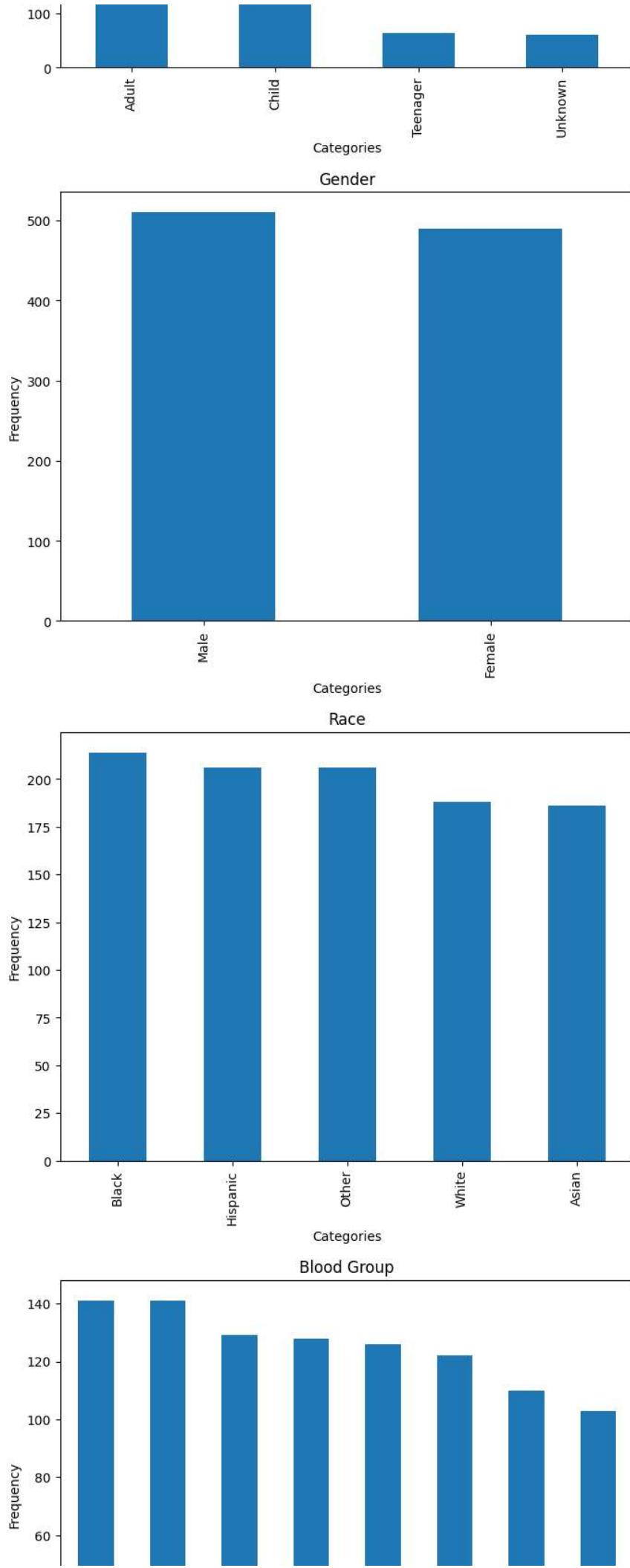
# Box plots for numerical variables
df.boxplot(figsize=(10, 6))
plt.show()

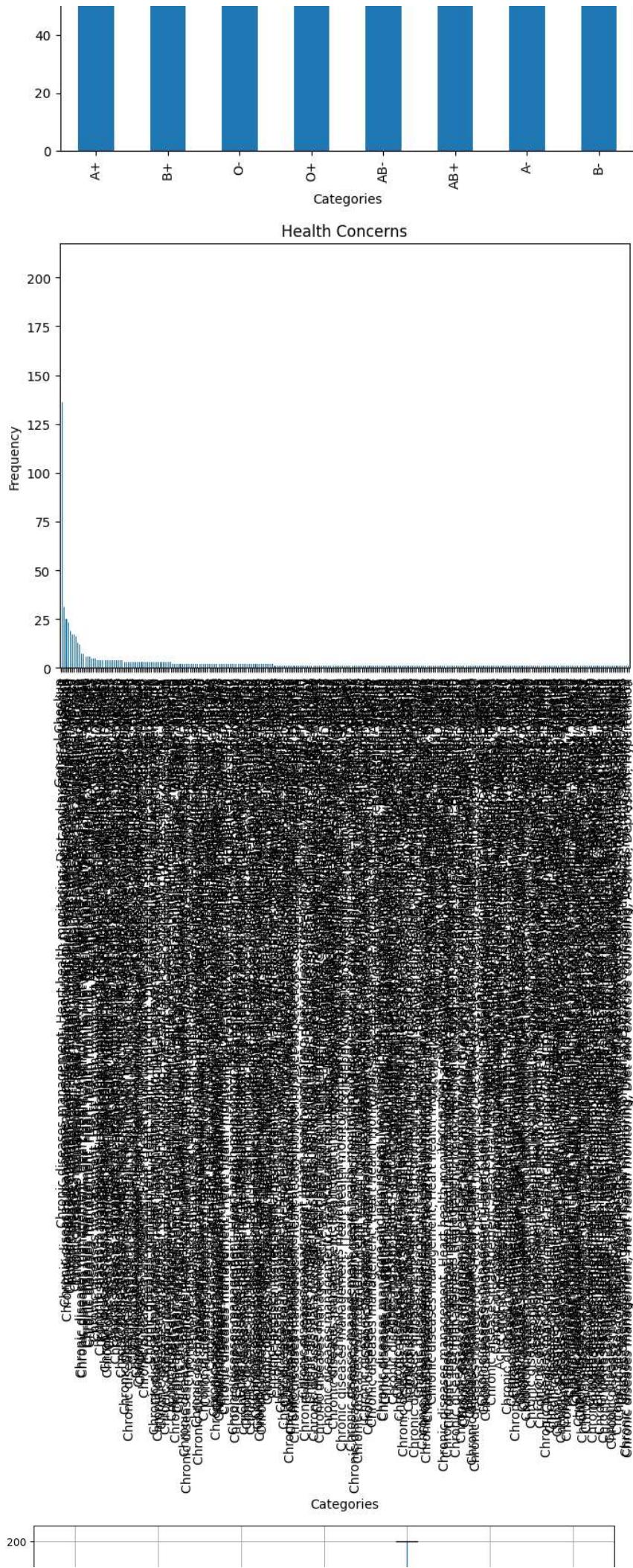
# Exploratory Data Analysis (EDA)
# Scatter plots for pairwise relationships
sns.pairplot(df)
plt.show()

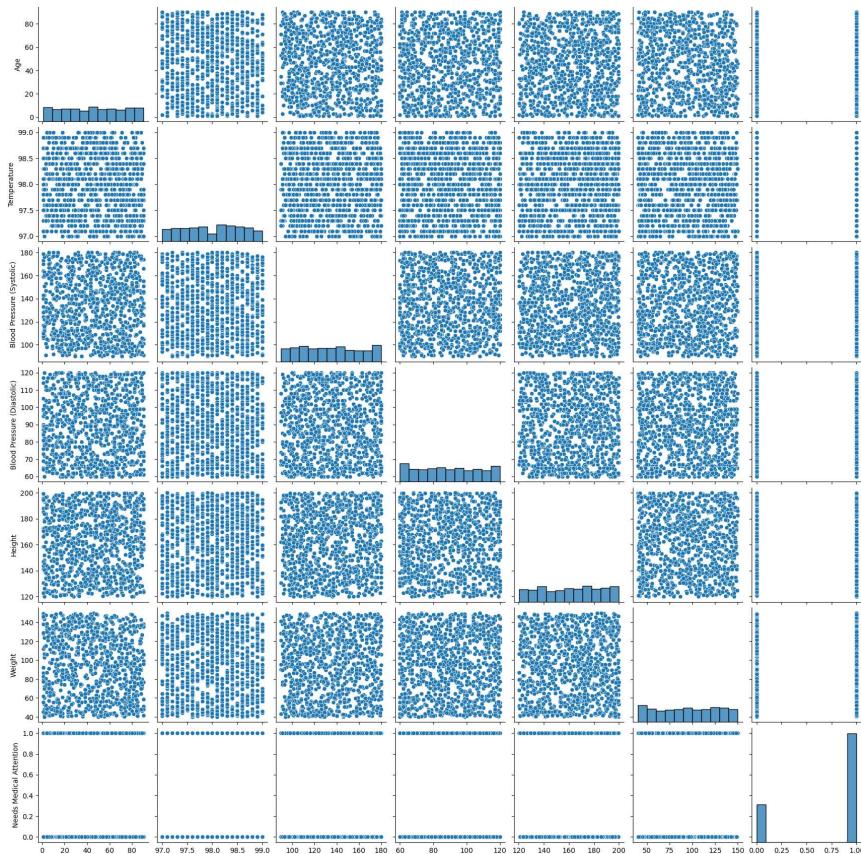
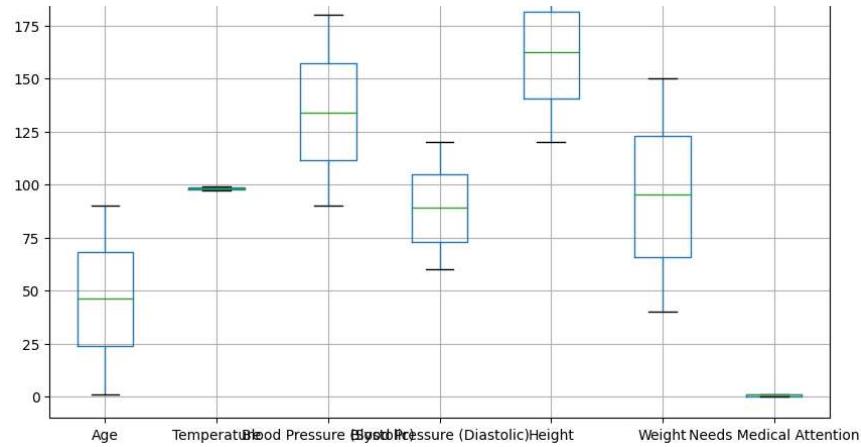
# Correlation matrix
corr_matrix = df.corr()
plt.figure(figsize=(10, 8))
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm')
plt.title('Correlation Matrix')
plt.show()

```

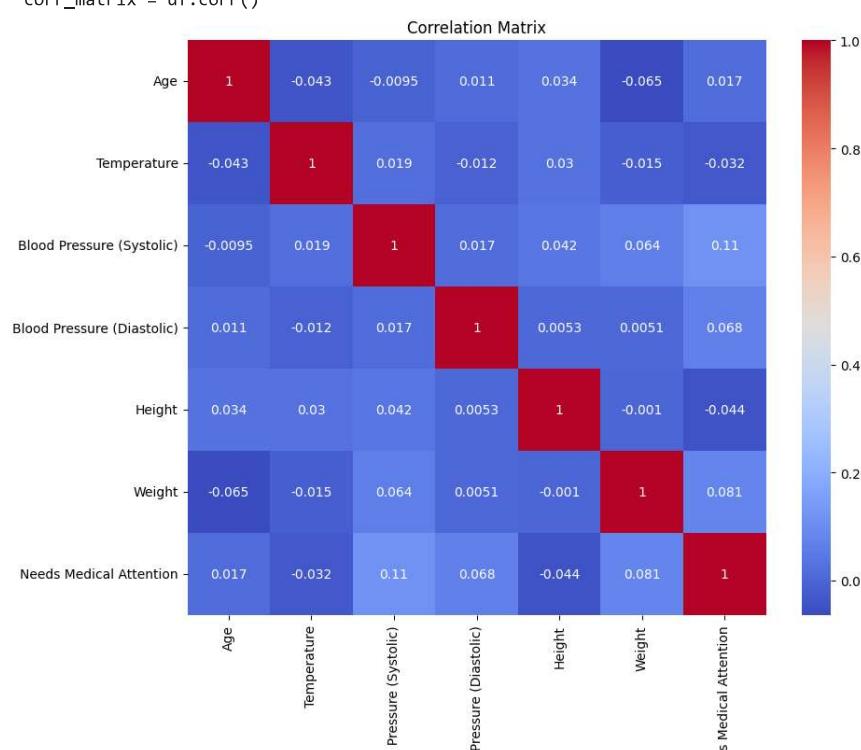








```
<ipython-input-54-6c5d4b929b0f>:29: FutureWarning: The default value of numeric_only
corr_matrix = df.corr()
```



Blood

Need

```
import pandas as pd  
  
# Selecting numerical columns for skewness calculation  
numerical_columns = df.select_dtypes(include=['int64', 'float64'])
```

```
# Calculating skewness for each numerical column
skewness = numerical_columns.skew()

# Displaying skewness for each column
print("Skewness for numerical columns:")
print(skewness)
```

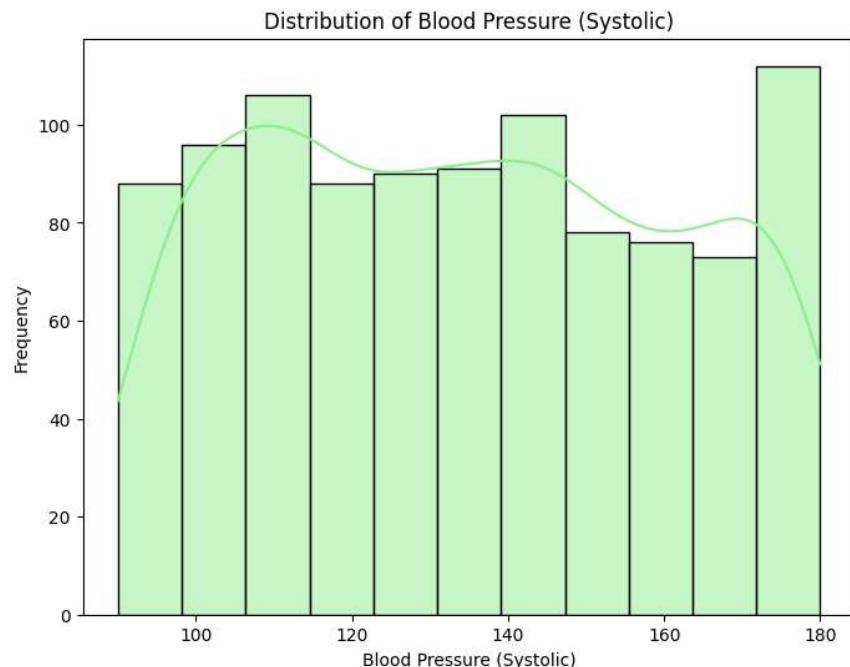
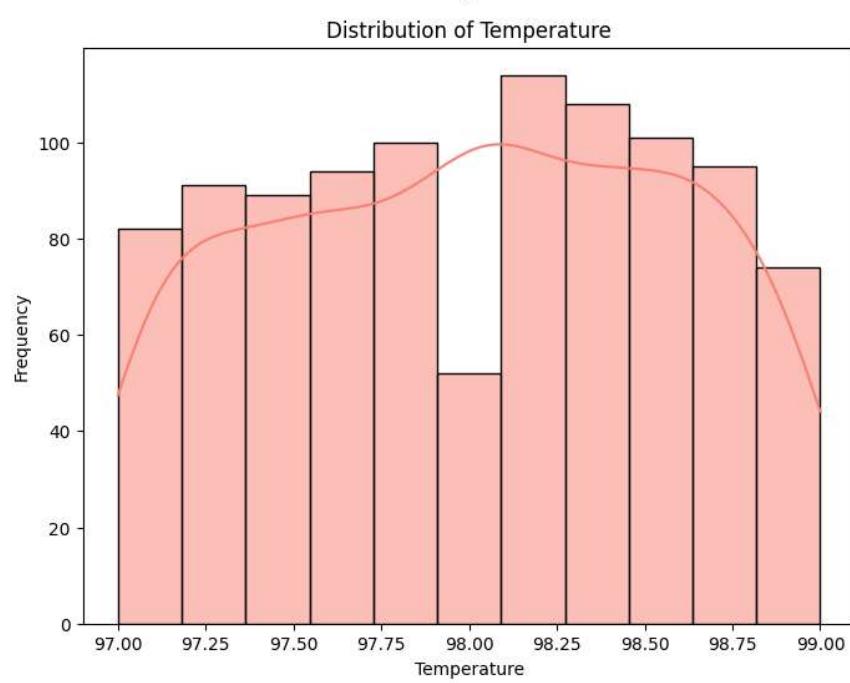
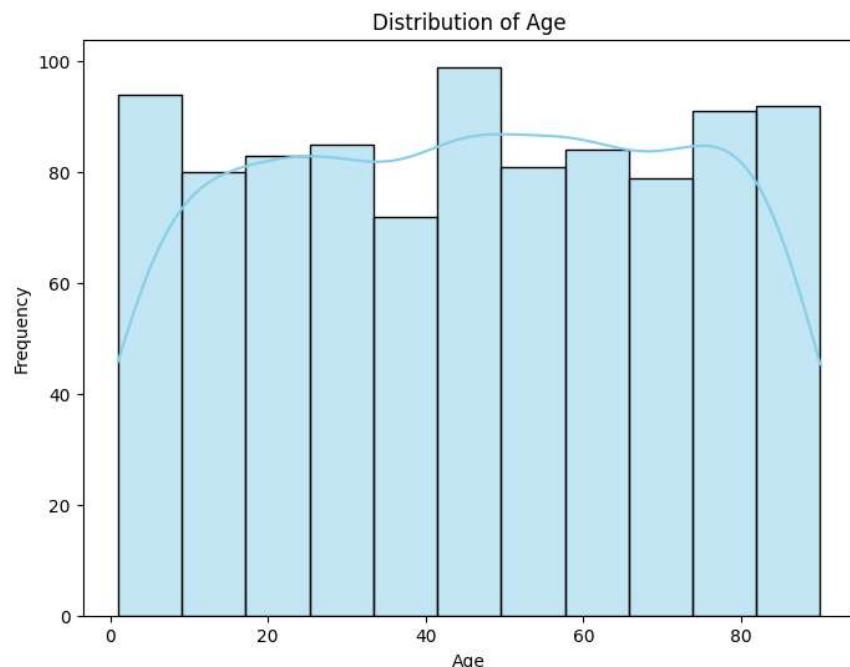
```
Skewness for numerical columns:
Age           -0.029920
Temperature   -0.060163
Blood Pressure (Systolic)  0.092493
Blood Pressure (Diastolic) 0.054590
Height        -0.068798
Weight         -0.048330
Needs Medical Attention -1.113852
dtype: float64
```

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

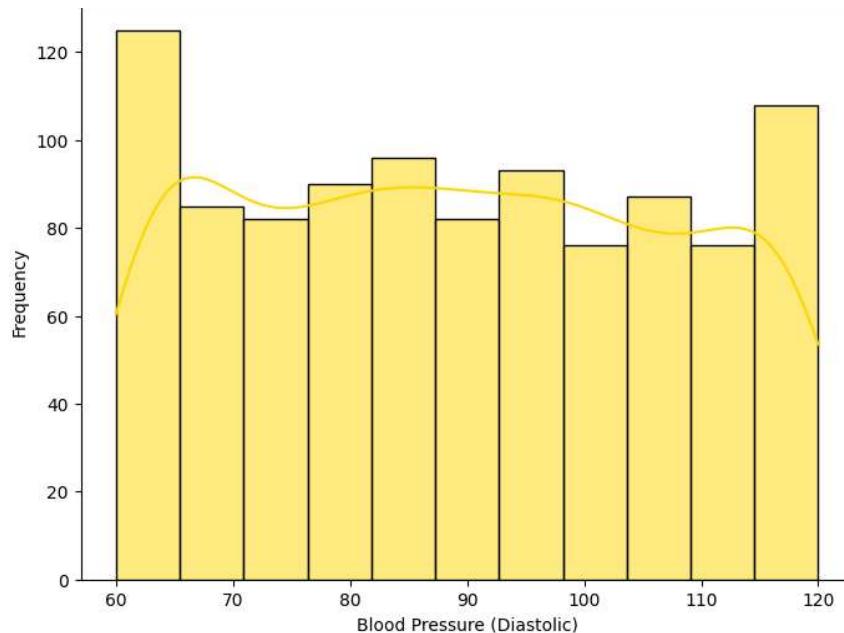
# Select numerical columns
numerical_columns = ['Age', 'Temperature', 'Blood Pressure (Systolic)', 'Blood Pressure (Diastolic)', 'Height', 'Weight', 'Needs Medical Attention']

# Define bright colors
bright_colors = ['skyblue', 'salmon', 'lightgreen', 'gold', 'lightcoral', 'lightblue', 'lightsalmon']

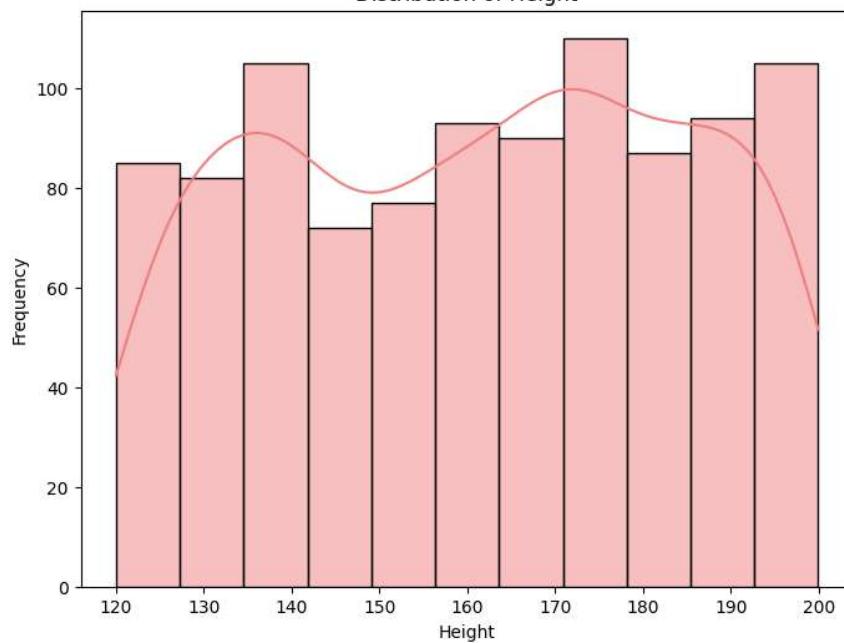
# Plot histograms for numerical columns with bright colors
for i, column in enumerate(numerical_columns):
    plt.figure(figsize=(8, 6))
    sns.histplot(df[column], kde=True, color=bright_colors[i])
    plt.title(f'Distribution of {column}')
    plt.xlabel(column)
    plt.ylabel('Frequency')
    plt.show()
```



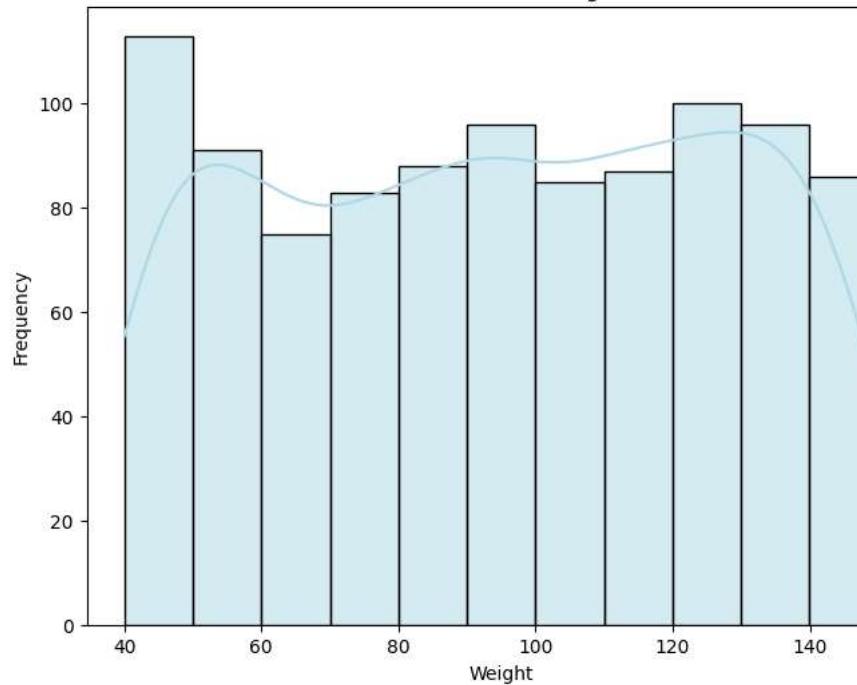
Distribution of Blood Pressure (Diastolic)



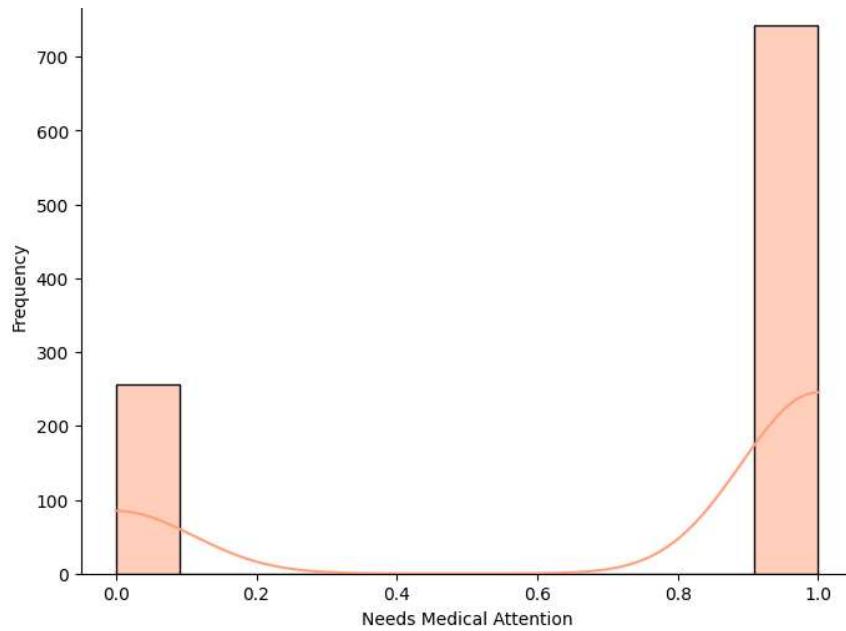
Distribution of Height



Distribution of Weight



Distribution of Needs Medical Attention



```
# Select rows where Age column has NaN values
missing_age_rows = df[df['Age'].isnull()]

# Display rows with missing Age values
print("Rows with missing Age values:")
print(missing_age_rows)
```

```
129      1
731      0
744      1
765      1
783      1
807      1
808      1
822      1
833      0
847      1
852      0
862      1
895      0
951      1
956      1
959      1
965      1
992      0
```

```
# Import necessary libraries
import pandas as pd

# Calculate the mean of the Age column
age_mean = df['Age'].mean()

# Impute missing values in the Age column with the mean
df['Age'].fillna(age_mean, inplace=True)
```

```
# Displaying only the 'Age' column
print(df['Age'])
```

```
0      29.0
1      84.0
2      26.0
3      42.0
4      24.0
...
995     36.0
996     15.0
997     21.0
998     89.0
999     1.0
Name: Age, Length: 1000, dtype: float64
```

```
# Define a dictionary mapping age ranges to numeric categories
age_category_dict_numeric = {
    (0, 25): 0,
    (26, 40): 1,
    (41, 60): 2,
    (61, 100): 3
}

# Function to map age to numeric category using the dictionary
def categorize_age_numeric(age):
    for age_range, category in age_category_dict_numeric.items():
        if age_range[0] <= age <= age_range[1]:
            return category
    return -1 # Handle unknown values

# Apply the function to create a new column 'Numeric Age Group'
df['Numeric Age Group'] = df['Age'].apply(categorize_age_numeric)
```

```
df
```

	Name	Age	Age Group	Gender	Race	Blood Group	Temperature	Blood Pressure (Systolic)	Blood Pressure (Diastolic)	Other
0	Kevin Wong	29.0	Adult	Female	Black	AB+	97.9	178	100	Normal
1	Mr. David Meyers	84.0	Adult	Male	White	A-	98.1	130	100	Normal
2	Robert Sanchez DDS	26.0	Adult	Male	Hispanic	A+	97.9	121	100	Normal
3	Brent Moore	42.0	Adult	Male	Black	B+	98.1	169	100	Normal
4	Alicia Kennedy	24.0	Adult	Male	Black	A+	98.0	100	100	Normal
...
995	Stacy Gallegos	36.0	Adult	Female	Black	A+	97.9	159	100	Normal
996	Natalie Curtis	15.0	Teenager	Female	Asian	AB+	98.0	97	100	Normal
997	Taylor Meyer	21.0	Adult	Male	Hispanic	A+	97.6	119	100	Normal
998	Erin Morales	89.0	Adult	Male	Hispanic	A+	98.3	93	100	Normal
999	Gary Harper	1.0	Child	Male	Asian	B+	98.0	128	100	Normal

1000 rows × 14 columns



```
# Create a dictionary to map gender categories to numeric values
gender_mapping = {'Male': 0, 'Female': 1}
```

```
# Map the 'Gender' column using the dictionary
df['Gender'] = df['Gender'].map(gender_mapping)
df
```

	Name	Age	Age Group	Gender	Race	Blood Group	Temperature	Blood Pressure (Systolic)	Blood Pressure (Diastolic)
0	Kevin Wong	29.0	Adult	1	Black	AB+	97.9	178	
1	Mr. David Meyers	84.0	Adult	0	White	A-	98.1	130	
2	Robert Sanchez DDS	26.0	Adult	0	Hispanic	A+	97.9	121	
3	Brent Moore	42.0	Adult	0	Black	B+	98.1	169	
4	Alicia Kennedy	24.0	Adult	0	Black	A+	98.0	100	
...
995	Stacy Gallegos	36.0	Adult	1	Black	A+	97.9	159	
996	Natalie Curtis	15.0	Teenager	1	Asian	AB+	98.0	97	
997	Taylor Meyer	21.0	Adult	0	Hispanic	A+	97.6	119	
998	Erin Morales	89.0	Adult	0	Hispanic	A+	98.3	93	
999	Gary Harper	1.0	Child	0	Asian	B+	98.0	128	

1000 rows × 14 columns

```
# Get unique values in the 'Race' column
unique_races = df['Race'].unique()
print(unique_races)
```

```
['Black' 'White' 'Hispanic' 'Asian' 'Other']
```

```
# Define a dictionary to map race values to numerical categories
race_mapping = {'Black': 0, 'Hispanic': 1, 'Asian': 2, 'Other': 3, 'White': 4}

# Map the 'Race' column using the defined dictionary
df['Race'] = df['Race'].map(race_mapping)

# Display the updated DataFrame
print(df.head())
```

	Name	Age	Age Group	Gender	Race	Blood Group	Temperature	Height	Weight
0	Kevin Wong	29.0	Adult	1	0	AB+	97.9	183.19	73.15
1	Mr. David Meyers	84.0	Adult	0	4	A-	98.1	170.96	92.36
2	Robert Sanchez DDS	26.0	Adult	0	1	A+	97.9	146.33	77.41
3	Brent Moore	42.0	Adult	0	0	B+	98.1	197.98	113.40
4	Alicia Kennedy	24.0	Adult	0	0	A+	98.0	149.24	79.90
					Blood Pressure (Systolic)	Blood Pressure (Diastolic)			
0					178	84			
1					130	88			
2					121	101			
3					169	112			
4					100	61			

```
Health Concerns  Needs Medical Attention \
0 Chronic diseases management, Heart health moni... 1
1 Chronic diseases management, Heart health moni... 1
2 Chronic diseases management, Heart health moni... 1
3 Chronic diseases management, Heart health moni... 1
4 Chronic diseases management, Heart health moni... 1
```

Numeric Age Group

	Numeric Age Group
0	1
1	3
2	1
3	2
4	0

```
# Find unique values in the 'Blood Group' column
unique_blood_groups = df['Blood Group'].unique()
```

```
# Display the unique values
print(unique_blood_groups)
```

```
['AB+' 'A-' 'A+' 'B+' 'B-' 'AB-' 'O+' 'O-' ]
```

```
# Define the mapping dictionary for blood group
```

```
blood_group_mapping = {'AB-': 0, 'A-': 1, 'O+'. 2, 'B+'. 3, 'A+'. 4, 'O-'. 5, 'AB+'. 6, 'B-'. 7}
```

```
# Map the values in the "Blood Group" column to numeric values
```

```
df['Blood Group'] = df['Blood Group'].map(blood_group_mapping)
df
```

	Name	Age	Age Group	Gender	Race	Blood Group	Temperature	Blood Pressure (Systolic)	Blood Pressure (Diastolic)
0	Kevin Wong	29.0	Adult	1	0	6	97.9	178	8
1	Mr. David Meyers	84.0	Adult	0	4	1	98.1	130	8
2	Robert Sanchez DDS	26.0	Adult	0	1	4	97.9	121	10
3	Brent Moore	42.0	Adult	0	0	3	98.1	169	11
4	Alicia Kennedy	24.0	Adult	0	0	4	98.0	100	6
...
995	Stacy Gallegos	36.0	Adult	1	0	4	97.9	159	6
996	Natalie Curtis	15.0	Teenager	1	2	6	98.0	97	6
997	Taylor Meyer	21.0	Adult	0	1	4	97.6	119	11
998	Erin Morales	89.0	Adult	0	1	4	98.3	93	8
999	Gary Harper	1.0	Child	0	2	3	98.0	128	8

1000 rows × 14 columns

```
import pandas as pd

# Fill blank rows with 'Unknown'
df['Health Concerns'].fillna('Unknown', inplace=True)

# Convert the 'Health Concerns' column to a list of health concerns
df['Health Concerns'] = df['Health Concerns'].str.split(',')

# Display the first few rows to verify the conversion
print(df.head())

df
```

	Name	Age	Age Group	Gender	Race	Blood Group	Temperature	\
0	Kevin Wong	29.0	Adult	1	0	6	97.9	
1	Mr. David Meyers	84.0	Adult	0	4	1	98.1	
2	Robert Sanchez DDS	26.0	Adult	0	1	4	97.9	
3	Brent Moore	42.0	Adult	0	0	3	98.1	
4	Alicia Kennedy	24.0	Adult	0	0	4	98.0	

	Blood Pressure (Systolic)	Blood Pressure (Diastolic)	Height	Weight	\
0	178	84	183.19	73.15	
1	130	88	170.96	92.36	
2	121	101	146.33	77.41	
3	169	112	197.98	113.40	
4	100	61	149.24	79.90	

	Health Concerns	Needs Medical Attention	\
0	[Chronic diseases management, Heart health management]	1	
1	[Chronic diseases management, Heart health management]	1	
2	[Chronic diseases management, Heart health management]	1	
3	[Chronic diseases management, Heart health management]	1	
4	[Chronic diseases management, Heart health management]	1	

	Numeric Age Group
0	1
1	3
2	1
3	2
4	0

	Name	Age	Age Group	Gender	Race	Blood Group	Temperature	Blood Pressure (Systolic)	Blood Pressure (Diastolic)	\
0	Kevin Wong	29.0	Adult	1	0	6	97.9	178	84	
1	Mr. David Meyers	84.0	Adult	0	4	1	98.1	130	88	
2	Robert Sanchez DDS	26.0	Adult	0	1	4	97.9	121	101	
3	Brent Moore	42.0	Adult	0	0	3	98.1	169	112	
4	Alicia Kennedy	24.0	Adult	0	0	4	98.0	100	61	
...	
995	Stacy Gallegos	36.0	Adult	1	0	4	97.9	159	119	
996	Natalie Curtis	15.0	Teenager	1	2	6	98.0	97	93	
997	Taylor Meyer	21.0	Adult	0	1	4	97.6	119	111	
998	Erin Morales	89.0	Adult	0	1	4	98.3	93	88	
999	Gary Harper	1.0	Child	0	2	3	98.0	128	106	

1000 rows x 14 columns

df['Health Concerns']

	Health Concerns
0	[Chronic diseases management, Heart health management]
1	[Chronic diseases management, Heart health management]
2	[Chronic diseases management, Heart health management]

```
3 [Chronic diseases management, Heart health mo...
4 [Chronic diseases management, Heart health mo...
...
995 [Chronic diseases management, Heart health mo...
996 [Acne and skin issues, Mental health challeng...
997 [General Checkup]
998 [General Checkup]
999 [General Checkup]
Name: Health Concerns, Length: 1000, dtype: object
```

```
import pandas as pd
# Fill blank cells with "Unknown"
df['Health Concerns'].fillna('Unknown', inplace=True)

# Split rows with multiple health conditions by commas and create a new column
df['Health Concern'] = df['Health Concerns'].str.split(',')

# Iterate over each row in the original DataFrame and handle missing values
new_rows = []

for index, row in df.iterrows():
    health_concerns = row['Health Concern']
    if isinstance(health_concerns, list):
        for health_condition in health_concerns:
            new_row = row.copy()
            new_row['Health Concern'] = health_condition.strip() if isinstance(health_condition, str) else 'Unknown'
            new_rows.append(new_row)
    else:
        # If there's no multiple health condition, keep the row with 'Health Concern' as is
        new_row = row.copy()
        new_row['Health Concern'] = health_concerns.strip() if isinstance(health_concerns, str) else 'Unknown'
        new_rows.append(new_row)

# Create a new DataFrame from the list of duplicated rows
new_df = pd.DataFrame(new_rows)

# Counting the number of health concerns in each row
new_df['Number of Health Concerns'] = new_df['Health Concerns'].apply(len)

# Exploring unique health concerns
unique_health_concerns = set()
for health_concerns in new_df['Health Concerns']:
    unique_health_concerns.update(health_concerns)

# Print unique health concerns
print("Unique Health Concerns:", unique_health_concerns)

# Creating a numerical representation using one-hot encoding
for concern in unique_health_concerns:
    new_df[concern] = new_df['Health Concerns'].apply(lambda x: 1 if concern in x else 0)

Unique Health Concerns: {'Asthma', 'Allergies', 'Depression', 'Diabetes', 'Sexual health education', 'Heart health monitoring'}
```

	Name	Age	Age Group	Gender	Race	Blood Group	Temperature	Blood Pressure (Systolic)	Blood Pressure (Diastolic)
0	Kevin Wong	29.0	Adult	1	0	6	97.9	178	8
1	Mr. David Meyers	84.0	Adult	0	4	1	98.1	130	8
2	Robert Sanchez DDS	26.0	Adult	0	1	4	97.9	121	10
3	Brent Moore	42.0	Adult	0	0	3	98.1	169	11
4	Alicia Kennedy	24.0	Adult	0	0	4	98.0	100	6
...
995	Stacy Gallegos	36.0	Adult	1	0	4	97.9	159	6
996	Natalie Curtis	15.0	Teenager	1	2	6	98.0	97	6
997	Taylor Meyer	21.0	Adult	0	1	4	97.6	119	11
998	Erin Morales	89.0	Adult	0	1	4	98.3	93	8
999	Gary Harper	1.0	Child	0	2	3	98.0	128	8

1000 rows × 15 columns

df

	Name	Age	Age Group	Gender	Race	Blood Group	Temperature	Blood Pressure (Systolic)	Blood Pressure (Diastolic)
0	Kevin Wong	29.0	Adult	1	0	6	97.9	178	8
1	Mr. David Meyers	84.0	Adult	0	4	1	98.1	130	8
2	Robert Sanchez DDS	26.0	Adult	0	1	4	97.9	121	10
3	Brent Moore	42.0	Adult	0	0	3	98.1	169	11
4	Alicia Kennedy	24.0	Adult	0	0	4	98.0	100	6
...
995	Stacy Gallegos	36.0	Adult	1	0	4	97.9	159	6
996	Natalie Curtis	15.0	Teenager	1	2	6	98.0	97	6
997	Taylor Meyer	21.0	Adult	0	1	4	97.6	119	11
998	Erin Morales	89.0	Adult	0	1	4	98.3	93	8
999	Gary Harper	1.0	Child	0	2	3	98.0	128	8

1000 rows × 15 columns

```
# Counting the number of health concerns in each row
df['Number of Health Concerns'] = df['Health Concerns'].apply(len)

# Exploring unique health concerns
unique_health_concerns = set()
for health_concerns in df['Health Concerns']:
    unique_health_concerns.update(health_concerns)

# Print unique health concerns
print("Unique Health Concerns:", unique_health_concerns)

# Creating a numerical representation using one-hot encoding
for concern in unique_health_concerns:
    df[concern] = df['Health Concerns'].apply(lambda x: 1 if concern in x else 0)
```

Unique Health Concerns: {'Asthma', 'Allergies', 'Depression', 'Diabetes', 'Sexual health education', 'Heart health monitoring',

df ['Health Concern']

0	NaN
1	NaN
2	NaN
3	NaN
4	NaN
..	
995	NaN
996	NaN

```
997    NaN  
998    NaN  
999    NaN  
Name: Health Concern, Length: 1000, dtype: float64
```

```
import pandas as pd  
# Drop the 'Health Concern' column  
df.drop(columns=['Health Concern'], inplace=True)
```



```
column_names = df.columns  
print(column_names)
```



```
Index(['Name', 'Age', 'Age Group', 'Gender', 'Race', 'Blood Group',  
       'Temperature', 'Blood Pressure (Systolic)',  
       'Blood Pressure (Diastolic)', 'Height', 'Weight', 'Health Concerns',  
       'Needs Medical Attention', 'Numeric Age Group',  
       'Number of Health Concerns', 'Asthma', 'Allergies', 'Depression',  
       'Diabetes', 'Sexual health education', 'Heart health monitoring',  
       'Chronic diseases management', 'Anxiety', 'Diabetes', 'Obesity',  
       'Depression', 'Mental health challenges', 'Growth and development',  
       'Obesity', 'Anxiety', 'Hypertension', 'Allergies', 'Vaccination',  
       'Unknown', 'Diet and exercise counseling', 'Acne and skin issues',  
       'General Checkup', 'Hypertension', 'Asthma'],  
      dtype='object')
```

```
# Select only numeric columns from the DataFrame  
numeric_df = df.select_dtypes(include='number')
```



```
# Display the selected numeric columns  
print(numeric_df)  
numeric_df.corr()
```

	Age	Gender	Race	Blood Group	Temperature	Blood Pressure (Systolic)	\
0	29.0	1	0	6	97.9	178	
1	84.0	0	4	1	98.1	130	
2	26.0	0	1	4	97.9	121	
3	42.0	0	0	3	98.1	169	
4	24.0	0	0	4	98.0	100	
..	
995	36.0	1	0	4	97.9	159	
996	15.0	1	2	6	98.0	97	
997	21.0	0	1	4	97.6	119	
998	89.0	0	1	4	98.3	93	
999	1.0	0	2	3	98.0	128	
	Blood Pressure (Diastolic)	Height	Weight	Needs Medical Attention	...	\	
0		84	183.19	73.15	1	...	
1		88	170.96	92.36	1	...	
2		101	146.33	77.41	1	...	
3		112	197.98	113.40	1	...	
4		61	149.24	79.90	1	...	
..	
995		68	162.34	136.16	1	...	
996		69	182.38	55.07	0	...	
997		112	185.71	117.14	0	...	
998		84	144.21	109.79	0	...	
999		81	194.56	77.49	0	...	
	Anxiety	Hypertension	Allergies	Vaccination	Unknown	\	
0	0	0	0	0	0	0	
1	0	0	0	0	0	0	
2	1	0	0	0	0	0	
3	0	0	0	0	0	0	
4	0	0	0	0	0	0	
..	
995	0	0	0	0	0	0	
996	0	0	0	0	0	0	
997	0	0	0	0	0	0	
998	0	0	0	0	0	0	
999	0	0	0	0	0	0	
	Diet and exercise counseling	Acne and skin issues	General Checkup	\			
0		1	0		0	0	
1		1	0		0	0	
2		1	0		0	0	
3		1	0		0	0	
4		1	0		0	0	
..	
995		1	0		0	0	
996		0	1		0	0	
997		0	0		0	1	
998		0	0		0	1	
999		0	0		0	1	
	Hypertension	Asthma					
0	0	0					
1	0	1					
2	0	0					
3	0	0					
4	0	0					
..					
995	0	0					
996	0	0					
997	0	0					
998	0	0					
999	0	0					

[1000 rows x 36 columns]

	Age	Gender	Race	Blood Group	Temperature	Blood Pressure (Systolic)	(D
Age	1.000000e+00	0.021437	-0.015645	-0.042293	-0.041872	-0.009177	
Gender	2.143728e-02	1.000000	0.018985	0.052414	-0.011855	-0.048921	
Race	-1.564549e-02	0.018985	1.000000	0.037033	-0.006760	0.057787	
Blood Group	-4.229306e-02	0.052414	0.037033	1.000000	0.021326	-0.012643	
Temperature	-4.187165e-02	-0.011855	-0.006760	0.021326	1.000000	0.019296	
Blood Pressure (Systolic)	-9.177439e-03	-0.048921	0.057787	-0.012643	0.019296	1.000000	
Blood Pressure (Diastolic)	1.062430e-02	0.012449	-0.016144	-0.029204	-0.012202	0.016760	

Height	3.325761e-02	0.047283	-0.008441	0.028555	0.029523	0.042426
Weight	-6.301085e-02	0.022803	-0.003821	-0.040494	-0.015252	0.063763
Needs Medical Attention	1.622124e-02	0.008835	0.020359	-0.006298	-0.031977	0.110760
Numeric Age Group	9.564945e-01	0.014556	-0.010882	-0.054028	-0.036830	0.013946
Number of Health Concerns	1.366514e-01	0.012977	0.007749	0.042663	-0.019376	0.044720
Asthma	-4.233181e-17	-0.008684	0.063130	-0.038544	0.033460	-0.020945
Allergies	1.271284e-17	-0.062117	0.002320	-0.020065	-0.006524	-0.038536
Depression	3.880559e-03	0.033578	0.019373	0.004919	-0.031516	-0.014142
Diabetes	3.544219e-17	0.000896	-0.014120	-0.009119	0.045991	-0.047740
Sexual health education	-2.616739e-01	-0.056995	-0.015157	0.050800	-0.003553	-0.049230
Heart health monitoring	4.541202e-01	0.054440	-0.016273	0.018692	-0.019152	0.011802
Chronic diseases management	4.541202e-01	0.054440	-0.016273	0.018692	-0.019152	0.011802
Anxiety	-5.910761e-17	0.015599	-0.037333	-0.052863	-0.030719	0.005900
Diabetes	3.286694e-02	-0.005674	0.021251	0.021311	-0.013531	0.005607
Obesity	2.707932e-18	0.015599	0.052505	-0.008037	-0.006065	0.005900
Depression	2.219259e-17	-0.012762	-0.017369	0.023982	0.030916	-0.059081
Mental health challenges	-2.616739e-01	-0.056995	-0.015157	0.050800	-0.003553	-0.049230
Growth and development	-5.342369e-01	-0.012844	0.042457	-0.025012	0.021323	0.023675
Obesity	7.392585e-02	0.014150	-0.011165	-0.006511	0.035502	0.009684
Anxiety	-3.643369e-02	0.030603	-0.003444	0.000919	0.004035	0.032617
Hypertension	8.153629e-02	0.005562	0.007411	0.001016	-0.008159	0.019658
Allergies	3.449431e-02	-0.034372	-0.024915	0.002059	0.015142	0.046206
Vaccination	-5.342369e-01	-0.012844	0.042457	-0.025012	0.021323	0.023675
Unknown	1.959636e-17	-0.013392	-0.034610	0.013621	-0.046388	0.087443
Diet and exercise counseling	4.541202e-01	0.054440	-0.016273	0.018692	-0.019152	0.011802
Acne and skin issues	-2.616739e-01	-0.056995	-0.015157	0.050800	-0.003553	-0.049230
General Checkup	-1.227239e-02	-0.011998	-0.002148	-0.012661	0.015432	-0.020565
Hypertension	-1.316623e-18	-0.010317	-0.022261	-0.011686	-0.013858	0.055456
Asthma	-1.285699e-02	-0.023955	0.041015	0.033228	-0.054645	0.028242

36 rows × 36 columns

```
column_names1 = numeric_df.columns
print(column_names1)

Index(['Age', 'Gender', 'Race', 'Blood Group', 'Temperature',
       'Blood Pressure (Systolic)', 'Blood Pressure (Diastolic)', 'Height',
       'Weight', 'Needs Medical Attention', 'Numeric Age Group',
       'Number of Health Concerns', 'Mental health challenges',
       'Growth and development', 'Vaccination', 'Obesity', 'Diabetes',
       'Depression', 'Anxiety', 'Hypertension',
       'Diet and exercise counseling', 'Allergies',
       'Sexual health education', 'Acne and skin issues',
       'Chronic diseases management', 'Heart health monitoring', 'Asthma'],
      dtype='object')
```

```
# Separate the input variables (features) from the target variable
X = numeric_df.drop(columns=['Needs Medical Attention']) # Input variables
y = numeric_df['Needs Medical Attention'] # Target variable
```

```
# Display the input variables and target variable
```

```
print("Input variables (features):")
```

```
print(X)
```

```
print("\nTarget variable:")
```

```
print(y)
```

		112	185.71	117.14	0	...
997		84	144.21	109.79	3	...
998		81	194.56	77.49	0	...
999						

	Anxiety	Hypertension	Allergies	Vaccination	Unknown	\
0	0	0	0	0	0	
1	0	0	0	0	0	
2	1	0	0	0	0	
3	0	0	0	0	0	
4	0	0	0	0	0	
..	
995	0	0	0	0	0	
996	0	0	0	0	0	
997	0	0	0	0	0	
998	0	0	0	0	0	
999	0	0	0	0	0	

	Diet and exercise counseling	Acne and skin issues	General Checkup	\
0	1	0	0	
1	1	0	0	
2	1	0	0	
3	1	0	0	
4	1	0	0	
..	
995	1	0	0	
996	0	1	0	
997	0	0	0	1
998	0	0	0	1
999	0	0	0	1

	Hypertension	Asthma
0	0	0
1	0	1
2	0	0
3	0	0
4	0	0
..
995	0	0
996	0	0
997	0	0
998	0	0
999	0	0

[1000 rows x 35 columns]

Target variable:

0	1
1	1
2	1
3	1
4	1
..	..
995	1
996	0
997	0
998	0
999	0

Name: Needs Medical Attention, Length: 1000, dtype: int64

```
from sklearn.model_selection import train_test_split
# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.20, random_state=0)
# Display the shapes of the training and testing sets
print("Training set - X shape:". X_train.shape)
```

```
print("Training set - X shape:", X_train.shape)
print("Training set - y shape:", y_train.shape)
print("Testing set - X shape:", X_test.shape)
print("Testing set - y shape:", y_test.shape)
```

```
Training set - X shape: (800, 35)
Training set - y shape: (800,)
Testing set - X shape: (200, 35)
Testing set - y shape: (200,)
```

```
from sklearn.linear_model import LogisticRegression
# Create the model
logistic_regression_model = LogisticRegression(solver='saga', max_iter=1000)

# Train the model
logistic_regression_model.fit(X_train, y_train)
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/linear_model/_sag.py:350: ConvergenceWarning:
  warnings.warn(
    LogisticRegression
LogisticRegression(max_iter=1000, solver='saga')
```

```
from sklearn.tree import DecisionTreeClassifier

# Create the model
decision_tree_model = DecisionTreeClassifier()

# Train the model
decision_tree_model.fit(X_train, y_train)
```

```
DecisionTreeClassifier()
DecisionTreeClassifier()
```

```
from sklearn.ensemble import RandomForestClassifier

# Create the model
random_forest_model = RandomForestClassifier()

# Train the model
random_forest_model.fit(X_train, y_train)
```

```
RandomForestClassifier()
RandomForestClassifier()
```

```
from sklearn.neighbors import KNeighborsClassifier

# Create the model
knn_model = KNeighborsClassifier()

# Train the model
knn_model.fit(X_train, y_train)
```

```
KNeighborsClassifier()
KNeighborsClassifier()
```

```
# Import necessary libraries
from sklearn.metrics import accuracy_score, classification_report

# Define a function to make predictions and evaluate the model
def evaluate_model(model, X_test, y_test):
    # Make predictions on the testing data
    y_pred = model.predict(X_test)

    # Calculate accuracy
    accuracy = accuracy_score(y_test, y_pred)
    print("Accuracy:", accuracy)

    # Display classification report
    print("\nClassification Report:")
    print(classification_report(y_test, y_pred))

    # Return the predictions
    return y_pred

# Evaluate each trained model
models = {
    "Logistic Regression": logistic_regression_model,
    "Decision Tree": decision_tree_model,
    "Random Forest": random_forest_model,
    "K-Nearest Neighbors": knn_model
}
```

```
for model_name, model in models.items():
    print("Model:", model_name)
    print("Predictions and Evaluation on Testing Data:")
    y_pred = evaluate_model(model, X_test, y_test)
    print("-----\n")
```

0	0.89	0.89	0.89	46
1	0.97	0.97	0.97	154

accuracy		0.95	200	
macro avg	0.93	0.93	0.93	200
weighted avg	0.95	0.95	0.95	200

```
-----  
Model: Decision Tree  
Predictions and Evaluation on Testing Data:  
Accuracy: 0.965
```

```
Classification Report:  
precision recall f1-score support  
  
0 0.88 0.98 0.93 46  
1 0.99 0.96 0.98 154  
  
accuracy 0.96 200  
macro avg 0.94 0.97 0.95 200  
weighted avg 0.97 0.96 0.97 200
```

```
-----  
Model: Random Forest  
Predictions and Evaluation on Testing Data:  
Accuracy: 0.965
```

```
Classification Report:  
precision recall f1-score support  
  
0 1.00 0.85 0.92 46  
1 0.96 1.00 0.98 154  
  
accuracy 0.96 200  
macro avg 0.98 0.92 0.95 200  
weighted avg 0.97 0.96 0.96 200
```

```
-----  
Model: K-Nearest Neighbors  
Predictions and Evaluation on Testing Data:  
Accuracy: 0.69
```

```
Classification Report:  
precision recall f1-score support
```

```
# Filter patients needing medical attention
df_medical_attention = numeric_df[numeric_df['Needs Medical Attention'] == 1]
```

```
# Display the subset of patients needing medical attention
print("Patients needing medical attention:")
print(df_medical_attention)
```

991	80.0	1	1	4	97.7	159
993	58.0	0	4	1	98.3	155
994	28.0	1	3	5	97.6	124
995	36.0	1	0	4	97.9	159
	Blood Pressure (Diastolic)	Height	Weight	Needs Medical Attention	...	\
0	84	183.19	73.15	1	...	
1	88	170.96	92.36	1	...	
2	101	146.33	77.41	1	...	
3	112	197.98	113.40	1	...	
4	61	149.24	79.90	1	...	
..	
990	105	162.68	52.39	1	...	
991	83	132.29	148.18	1	...	
993	74	163.94	103.98	1	...	
994	62	165.12	83.81	1	...	
995	68	162.34	136.16	1	...	
	Anxiety	Hypertension	Allergies	Vaccination	Unknown	\
0	0	0	0	0	0	
1	0	0	0	0	0	
2	1	0	0	0	0	
3	0	0	0	0	0	
4	0	0	0	0	0	
..	
990	0	0	0	0	0	
991	1	0	0	0	0	
993	0	0	0	0	0	
994	0	1	0	0	0	
995	0	0	0	0	0	
	Diet and exercise counseling	Acne and skin issues	General Checkup	...		\
0	1	0	0	...		
1	1	0	0	...		
2	1	0	0	...		
3	1	0	0	...		
4	1	0	0	...		
..		
990	1	0	0	...		
991	1	0	0	...		
993	1	0	0	...		
994	1	0	0	...		
995	1	0	0	...		
	Hypertension	Asthma				
0	0	0				
1	0	1				
2	0	0				
3	0	0				
4	0	0				
..				
990	0	0				
991	0	0				
993	0	1				
994	0	1				
995	0	0				

[743 rows x 36 columns]

```

from sklearn.cluster import KMeans

# Specify the number of clusters
num_clusters = 5 #can be adjusted based on the dataset and requirements

# Selecting only numeric columns for clustering
numeric_columns = ['Age', 'Temperature', 'Blood Pressure (Systolic)', 'Blood Pressure (Diastolic)', 'Height', 'Weight']

# Extracting numeric data for clustering
X = df_medical_attention[numeric_columns]

# Initialize the KMeans model
kmeans = KMeans(n_clusters=num_clusters, random_state=0)

# Fit the KMeans model to the data
kmeans.fit(X)

# Get the cluster labels
cluster_labels = kmeans.labels_

# Add the cluster labels to the subset dataframe
df_medical_attention['Cluster'] = cluster_labels

# Print the counts of patients in each cluster
print("Number of patients in each cluster:")
print(df_medical_attention['Cluster'].value_counts())

```

Number of patients in each cluster:

2	186
3	175
4	134
0	129
1	119

Name: Cluster, dtype: int64
/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change fr
warnings.warn(
<ipython-input-88-357cc1a7d179>:22: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus
df_medical_attention['Cluster'] = cluster_labels

```

# Analyze the clusters created by KMeans

# Calculate the centroids of each cluster
centroids = kmeans.cluster_centers_

# Assign cluster labels to the original dataframe
df_medical_attention['Cluster'] = kmeans.labels_

# Group the dataframe by cluster and calculate summary statistics for each cluster
cluster_summary = df_medical_attention.groupby('Cluster')[numeric_columns].describe()

# Display the cluster centroids
print("Centroids of Each Cluster:")
print(pd.DataFrame(centroids, columns=numeric_columns))

# Display the summary statistics for each cluster
print("\nSummary Statistics for Each Cluster:")
print(cluster_summary)

```

<ipython-input-90-2909b773166d>:7: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus
df_medical_attention['Cluster'] = kmeans.labels_

Centroids of Each Cluster:

	Age	Temperature	Blood Pressure (Systolic)	\
0	19.419957	98.004651	142.116279	
1	62.587699	97.939496	108.941176	
2	38.530805	98.023656	113.301075	
3	45.397277	97.998286	159.445714	
4	68.205399	98.009701	155.731343	

	Blood Pressure (Diastolic)	Height	Weight
0	91.124031	158.920078	65.420000
1	92.571429	156.928992	65.326723
2	87.360215	162.565054	123.083226
3	89.965714	158.791829	127.367486
4	90.022388	165.422985	75.639179

Summary Statistics for Each Cluster:

	Age	count	mean	std	min	25%	50%	75%	max	\
Cluster	0	129.0	19.419957	13.604296	1.0	6.00	17.0	30.0	47.0	
	1	119.0	62.587699	17.466522	24.0	47.00	62.0	77.0	90.0	
	2	186.0	38.530805	23.925644	1.0	18.00	37.5	56.0	89.0	
	3	175.0	45.397277	22.407144	1.0	29.00	45.0	61.5	90.0	
	4	134.0	68.205399	14.912736	39.0	56.25	70.5	81.0	90.0	
	Temperature	count	mean	...	Height	...	Weight	...	\\	
Cluster	0	129.0	98.004651	...	180.5400	199.48	129.0	65.420000		
	1	119.0	97.939496	...	177.3150	199.61	119.0	65.326723		
	2	186.0	98.023656	...	181.3825	199.54	186.0	123.083226		
	3	175.0	97.998286	...	183.3100	199.67	175.0	127.367486		
	4	134.0	98.009701	...	182.9050	199.84	134.0	75.639179		
	std	min	25%	50%	75%	max				
Cluster	0	17.719571	40.81	49.0100	62.340	80.3300	106.39			
	1	17.685420	40.26	50.0800	63.030	79.0650	102.96			
	2	15.287500	84.47	110.4325	124.115	135.5900	149.59			
	3	14.542984	94.73	116.1650	128.880	140.0600	149.93			
	4	18.855708	40.17	62.2200	77.000	90.8725	111.36			

[5 rows x 48 columns]

```
column_labels = df_medical_attention.columns
print(column_labels)
```

```
Index(['Age', 'Gender', 'Race', 'Blood Group', 'Temperature',
       'Blood Pressure (Systolic)', 'Blood Pressure (Diastolic)', 'Height',
       'Weight', 'Needs Medical Attention', 'Numeric Age Group',
       'Number of Health Concerns', 'Asthma', 'Allergies', 'Depression',
       'Diabetes', 'Sexual health education', 'Heart health monitoring',
       'Chronic diseases management', 'Anxiety', 'Diabetes', 'Obesity',
       'Depression', 'Mental health challenges', 'Growth and development',
       'Obesity', 'Anxiety', 'Hypertension', 'Allergies', 'Vaccination',
       'Unknown', 'Diet and exercise counseling', 'Acne and skin issues',
       'General Checkup', 'Hypertension', 'Asthma', 'Cluster'],
      dtype='object')
```

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

# Select relevant columns (health concerns and other features)
selected_columns = ['Asthma', 'Allergies', 'Depression', 'Diabetes', 'Hypertension', 'Obesity', 'Anxiety', 'Cluster']

# Create a subset of the DataFrame with selected columns
cluster_data = df_medical_attention[selected_columns]

# Group the data by cluster and calculate the mean of each feature
cluster_means = cluster_data.groupby('Cluster').mean().reset_index()

# Plot the distribution of health concerns and other features across clusters
plt.figure(figsize=(12, 8))
sns.barplot(data=cluster_means, x='Cluster', y='Asthma', palette='viridis')
plt.title('Average Asthma Prevalence by Cluster')
plt.xlabel('Cluster')
plt.ylabel('Average Asthma Prevalence')
plt.show()

# List of health concerns and other features
health_concerns = ['Asthma', 'Allergies', 'Depression', 'Diabetes', 'Hypertension', 'Obesity', 'Anxiety']

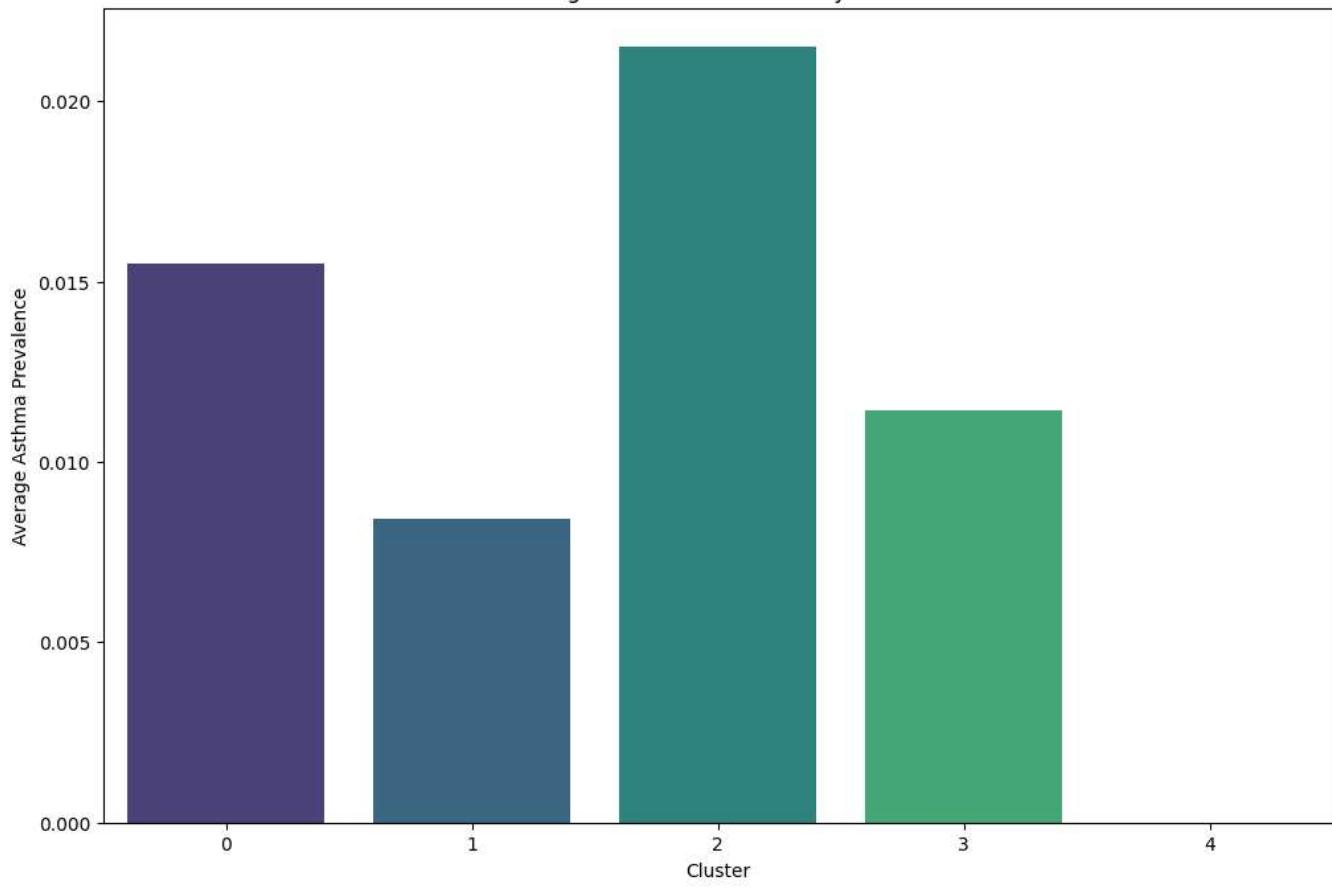
# Loop through each health concern and feature to create bar plots
for concern in health_concerns:
    plt.figure(figsize=(12, 8))
    sns.barplot(data=cluster_means, x='Cluster', y=concern, palette='viridis')
    plt.title(f'Average {concern} Prevalence by Cluster')
    plt.xlabel('Cluster')
    plt.ylabel(f'Average {concern} Prevalence')
    plt.show()
```

```
<ipython-input-99-2d66d7925f27>:15: FutureWarning:
```

```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `1
```

```
sns.barplot(data=cluster_means, x='Cluster', y='Asthma', palette='viridis')
```

Average Asthma Prevalence by Cluster

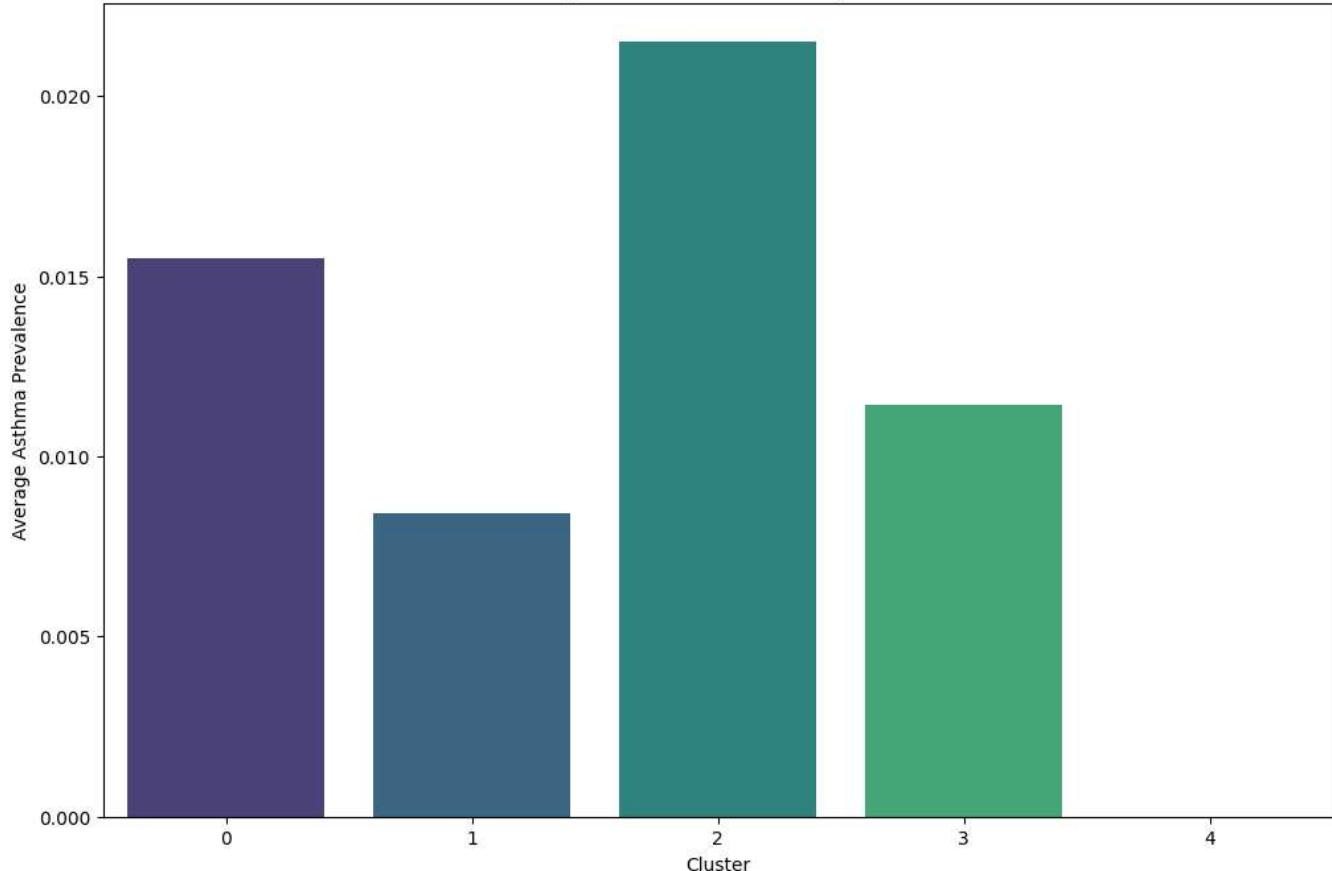


```
<ipython-input-99-2d66d7925f27>:27: FutureWarning:
```

```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `1
```

```
sns.barplot(data=cluster_means, x='Cluster', y='concern', palette='viridis')
```

Average Asthma Prevalence by Cluster

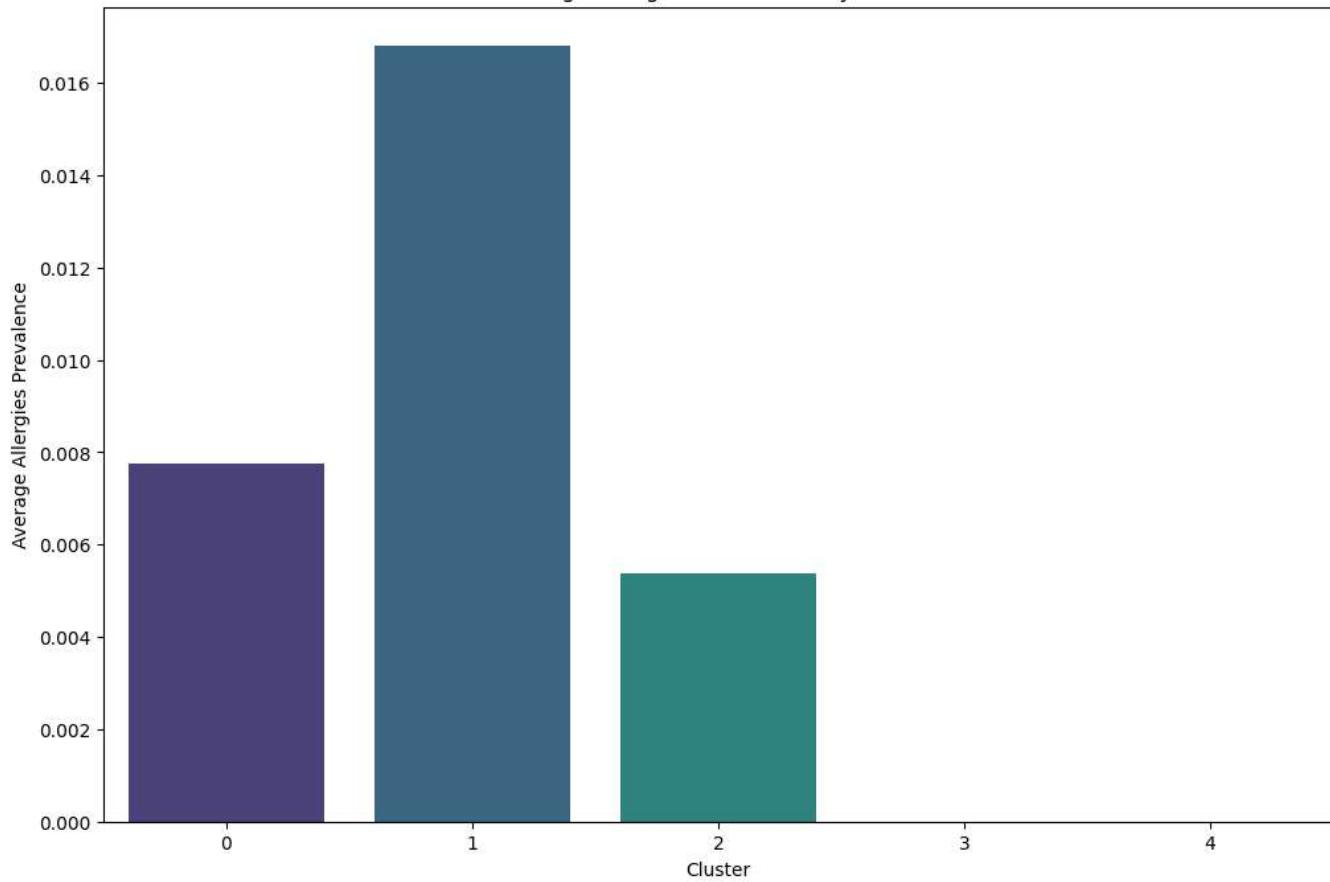


```
<ipython-input-99-2d66d7925f27>:27: FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `1

```
sns.barplot(data=cluster_means, x='Cluster', y=concern, palette='viridis')
```

Average Allergies Prevalence by Cluster

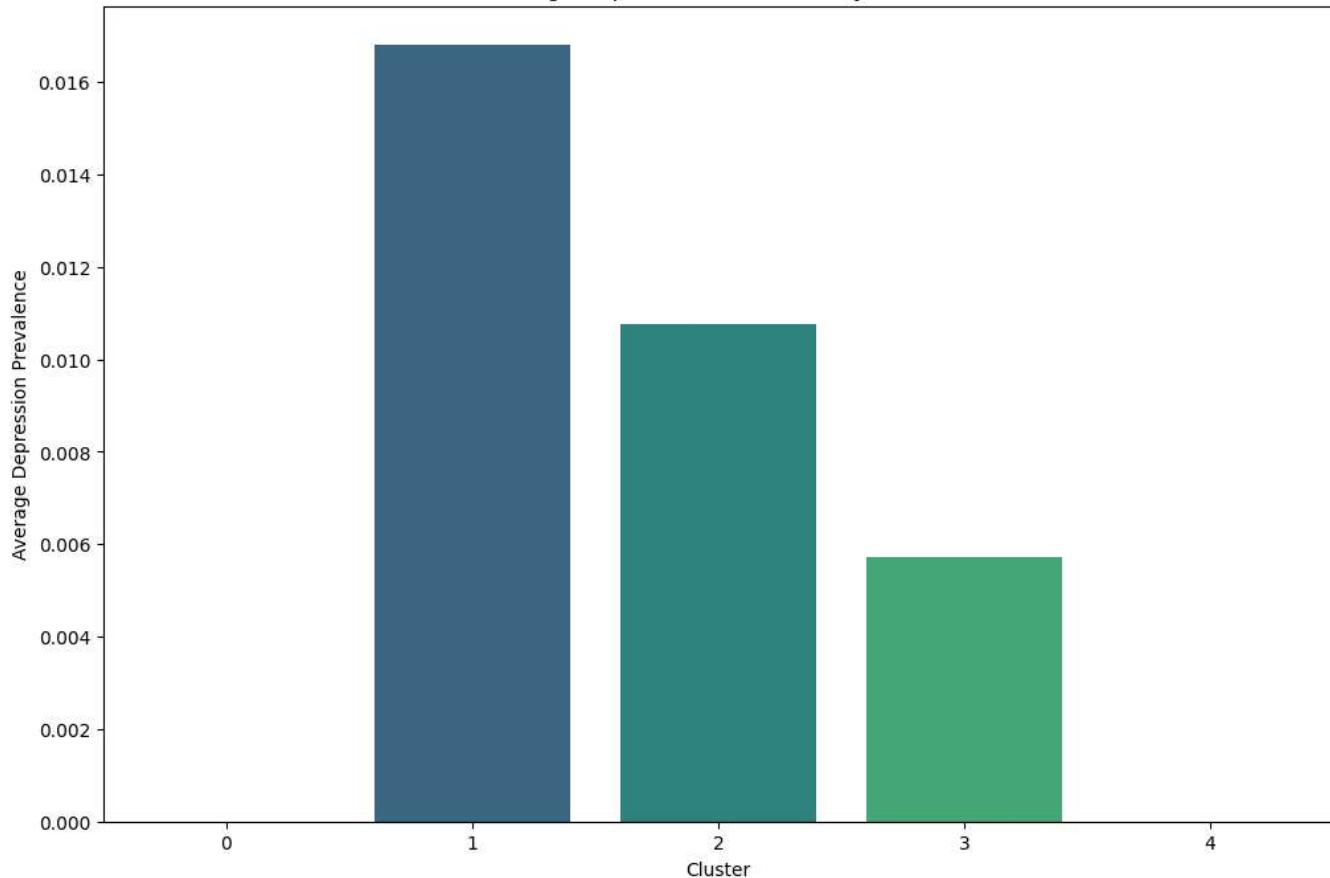


```
<ipython-input-99-2d66d7925f27>:27: FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `1

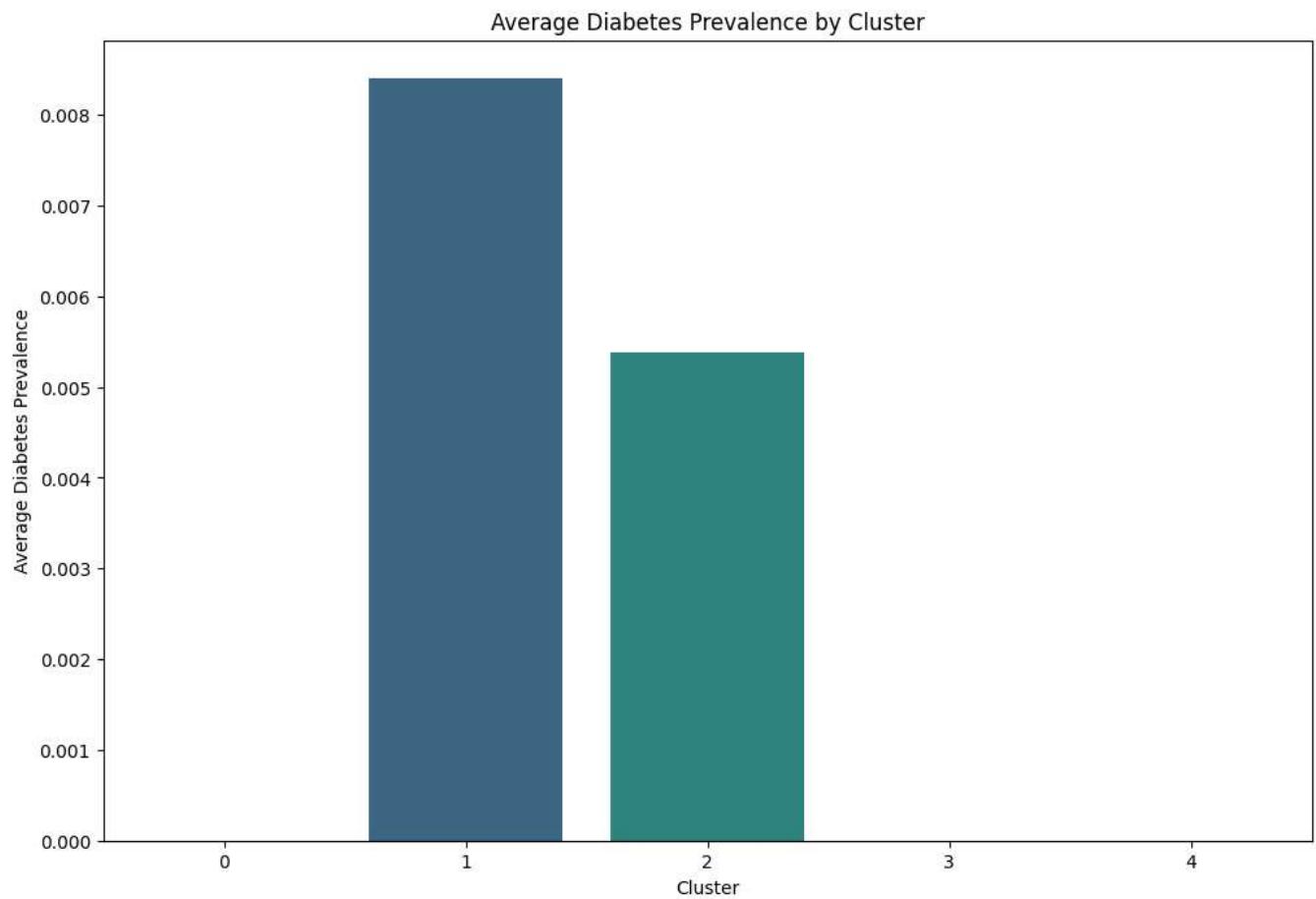
```
sns.barplot(data=cluster_means, x='Cluster', y=concern, palette='viridis')
```

Average Depression Prevalence by Cluster



```
<ipython-input-99-2d66d7925f27>:27: FutureWarning:
```

```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `l
sns.barplot(data=cluster_means, x='Cluster', y=concern, palette='viridis')
```



```
<ipython-input-99-2d66d7925f27>:27: FutureWarning:
```

```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `l
sns.barplot(data=cluster_means, x='Cluster', y=concern, palette='viridis')
```

Average Hypertension Prevalence by Cluster