healthcare-data-analysis

September 21, 2024

Data Source: https://www.kaggle.com/datasets/prasad22/healthcare-dataset

About Dataset Context: This synthetic healthcare dataset has been created to serve as a valuable resource for data science, machine learning, and data analysis enthusiasts. It is designed to mimic real-world healthcare data, enabling users to practice, develop, and showcase their data manipulation and analysis skills in the context of the healthcare industry.

Dataset Information: Each column provides specific information about the patient, their admission, and the healthcare services provided, making this dataset suitable for various data analysis and modeling tasks in the healthcare domain. Here's a brief explanation of each column in the dataset -

- **Name**: This column represents the name of the patient associated with the healthcare record.
- **Age**: The age of the patient at the time of admission, expressed in years.
- **Gender**: Indicates the gender of the patient, either "Male" or "Female."
- **Blood Type**: The patient's blood type, which can be one of the common blood types (e.g., "A+", "O-", etc.).
- **Medical Condition**: This column specifies the primary medical condition or diagnosis associated with the patient, such as "Diabetes," "Hypertension," "Asthma," and more.
- **Date of Admission**: The date on which the patient was admitted to the healthcare facility.
- **Doctor**: The name of the doctor responsible for the patient's care during their admission.
- **Hospital**: Identifies the healthcare facility or hospital where the patient was admitted.
- **Insurance Provider**: This column indicates the patient's insurance provider, which can be one of several options, including "Aetna," "Blue Cross," "Cigna," "UnitedHealthcare," and "Medicare."
- **Billing Amount**: The amount of money billed for the patient's healthcare services during their admission. This is expressed as a floating-point number.
- **Room Number**: The room number where the patient was accommodated during their admission.
- **Admission Type**: Specifies the type of admission, which can be "Emergency," "Elective," or "Urgent," reflecting the circumstances of the admission.
- **Discharge Date**: The date on which the patient was discharged from the healthcare facility, based on the admission date and a random number of days within a realistic range.
- **Medication**: Identifies a medication prescribed or administered to the patient during their admission. Examples include "Aspirin," "Ibuprofen," "Penicillin," "Paracetamol," and "Lipitor."
- **Test Results**: Describes the results of a medical test conducted during the patient's admission. Possible values include "Normal," "Abnormal," or "Inconclusive," indicating the outcome of the test.

[1]:
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

Reding the dataset and showing first 10 rows

[2]:
```
data=pd.read_csv("healthcare_dataset.csv")
data.head(10)
```

[2]:

	Name	Age	Gender	Blood Type	Medical Condition	Date of Admission	Doctor	Hospital	Insurance Provider	Billing Amount	Room Number	Admission Type
0	Bobby JacksOn	30	Male	B-	Cancer	2024-01-31	Matthew Smith	Sons and Miller	Blue Cross	18856.281306	328	Urgent
1	LesLie TErRy	62	Male	A+	Obesity	2019-08-20	Samantha Davies	Kim Inc	Medicare	33643.327287	265	Emergency
2	DaNnY sMitH	76	Female	A-	Obesity	2022-09-22	Tiffany Mitchell	Cook PLC	Aetna	27955.096079	205	Emergency
3	andrEw waTtS	28	Female	O+	Diabetes	2020-11-18	Kevin Wells	Hernandez Rogers and Vang,	Medicare	37909.782410	450	Elective
4	adrIENNE bEll	43	Female	AB+	Cancer	2022-09-19	Kathleen Hanna	White-White	Aetna	14238.317814	458	Urgent
5	EMILY JOHNSOn	36	Male	A+	Asthma	2023-12-20	Taylor Newton	Nunez-Humphrey	UnitedHealthcare	48145.110951	389	Urgent
6	edwArD EDWaRDs	21	Female	AB-	Diabetes	2020-11-03	Kelly Olson	Group Middleton	Medicare	19580.872345	389	Emergency
7	CHrisTInA MARtinez	20	Female	A+	Cancer	2021-12-28	Suzanne Thomas	Powell Robinson and Valdez,	Cigna	45820.462722	277	Emergency
8	JASmINe aGuIlaR	82	Male	AB+	Asthma	2020-07-01	Daniel Ferguson	Sons Rich and	Cigna	50119.222792	316	Elective
9	ChRISTopher BerG	58	Female	AB-	Cancer	2021-05-23	Heather Day	Padilla-Walker	UnitedHealthcare	19784.631062	249	Elective

```
  Discharge Date   Medication  Test Results
0     2024-02-02  Paracetamol        Normal
1     2019-08-26    Ibuprofen  Inconclusive
2     2022-10-07      Aspirin        Normal
3     2020-12-18    Ibuprofen      Abnormal
4     2022-10-09   Penicillin      Abnormal
5     2023-12-24    Ibuprofen        Normal
6     2020-11-15  Paracetamol  Inconclusive
7     2022-01-07  Paracetamol  Inconclusive
8     2020-07-14      Aspirin      Abnormal
9     2021-06-22  Paracetamol  Inconclusive
```

Size of the dataset

[3]:
```
data.shape
```

[3]:
```
(55500, 15)
```

Types of the data

[4]:
```
data.dtypes
```

[4]:
```
Name                   object
Age                     int64
Gender                 object
Blood Type             object
Medical Condition      object
Date of Admission      object
Doctor                 object
Hospital               object
Insurance Provider     object
Billing Amount        float64
Room Number             int64
Admission Type         object
Discharge Date         object
Medication             object
Test Results           object
dtype: object
```

Frequency distribution

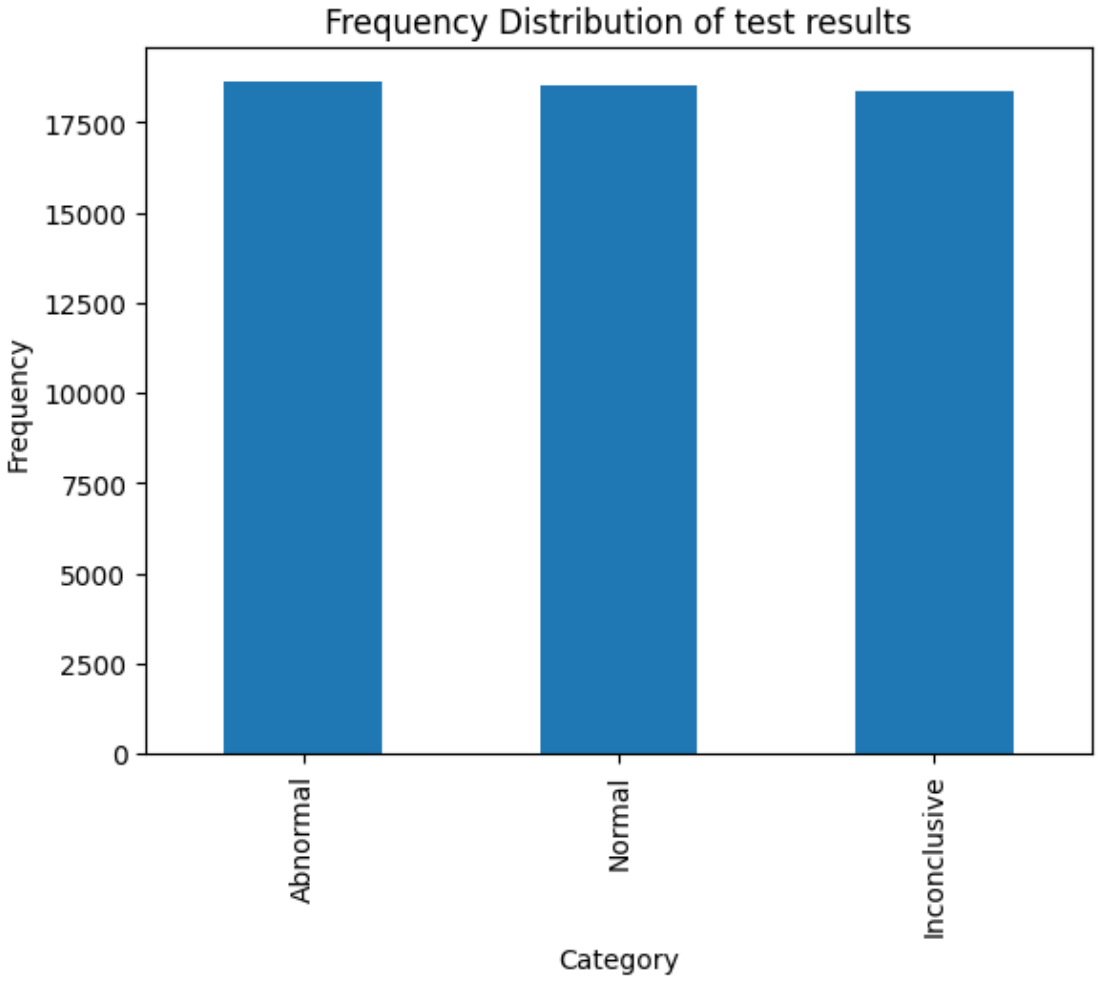
[5]:
```
## counting the frequency of test result
freq_count=data['Test Results'].value_counts()
freq_count
```

[5]:
```
Test Results
Abnormal        18627
Normal          18517
```

```
Inconclusive    18356
Name: count, dtype: int64
```

Bar diagram of the frequency distribution of the following

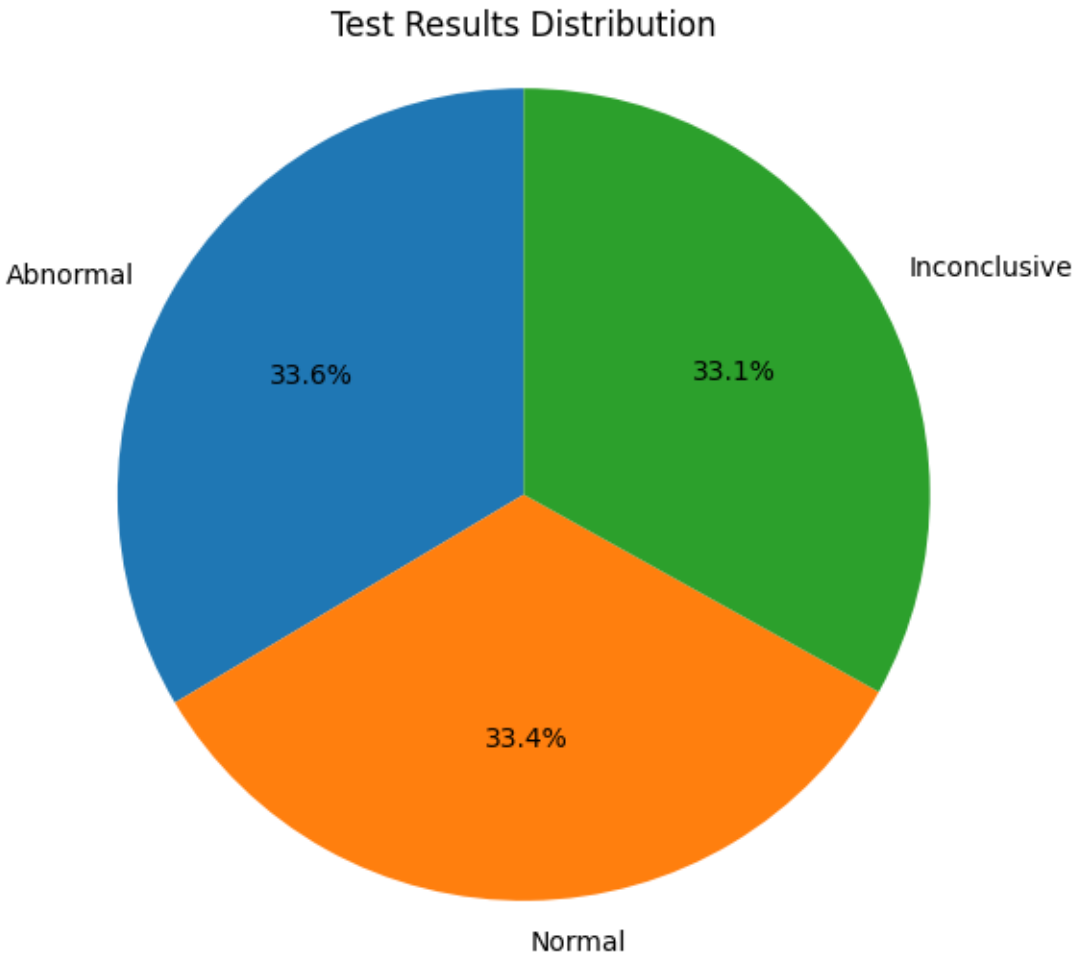
[6]:
```
freq_count.plot(kind='bar')
plt.title('Frequency Distribution of test results')
plt.xlabel('Category')
plt.ylabel('Frequency')
plt.show()
```



Pie chart of the frequency distribution of the following

[7]:
```
# Get the frequency of each unique value in 'Test Results'
test_results_counts =data['Test Results'].value_counts()

# Create a pie chart
```

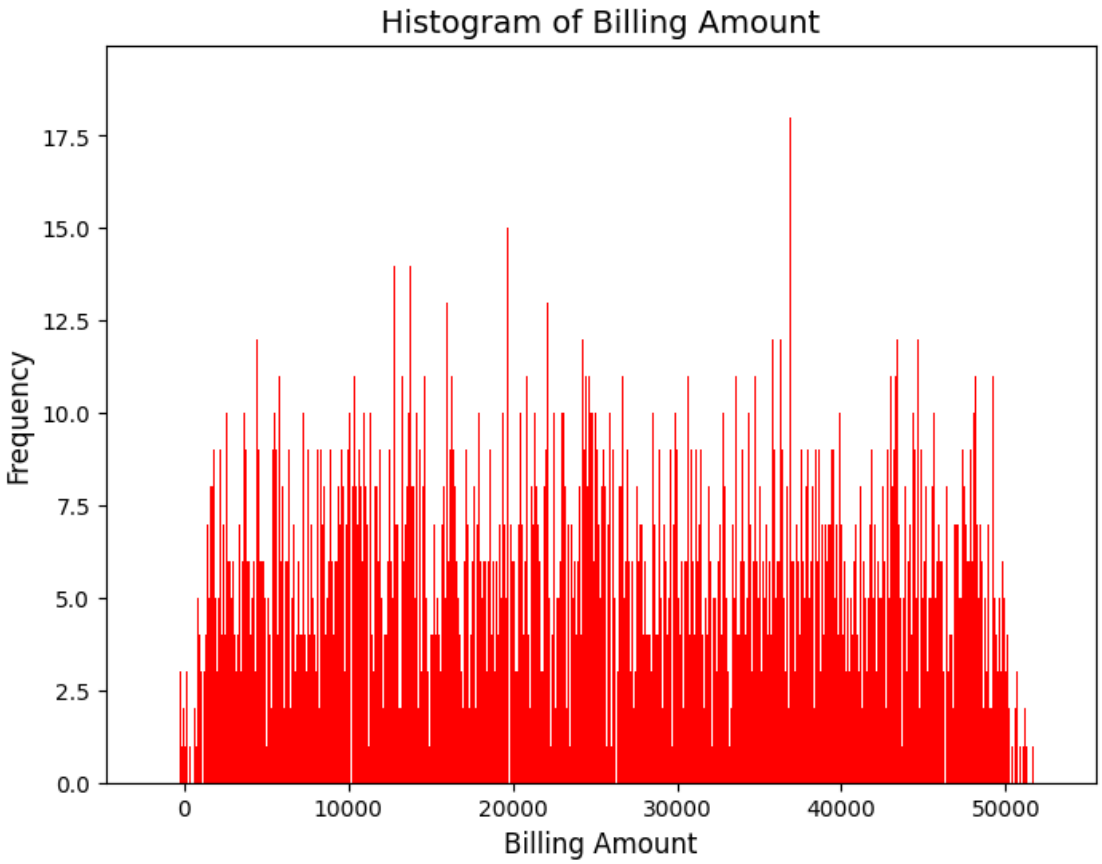
```
plt.figure(figsize=(6,6))
plt.pie(test_results_counts, labels=test_results_counts.index, autopct='%1.
↪1f%%', startangle=90)
plt.title('Test Results Distribution')
plt.axis('equal')  # Equal aspect ratio ensures the pie chart is circular.
plt.show()
```



Histogram of the frequency distribution of the following

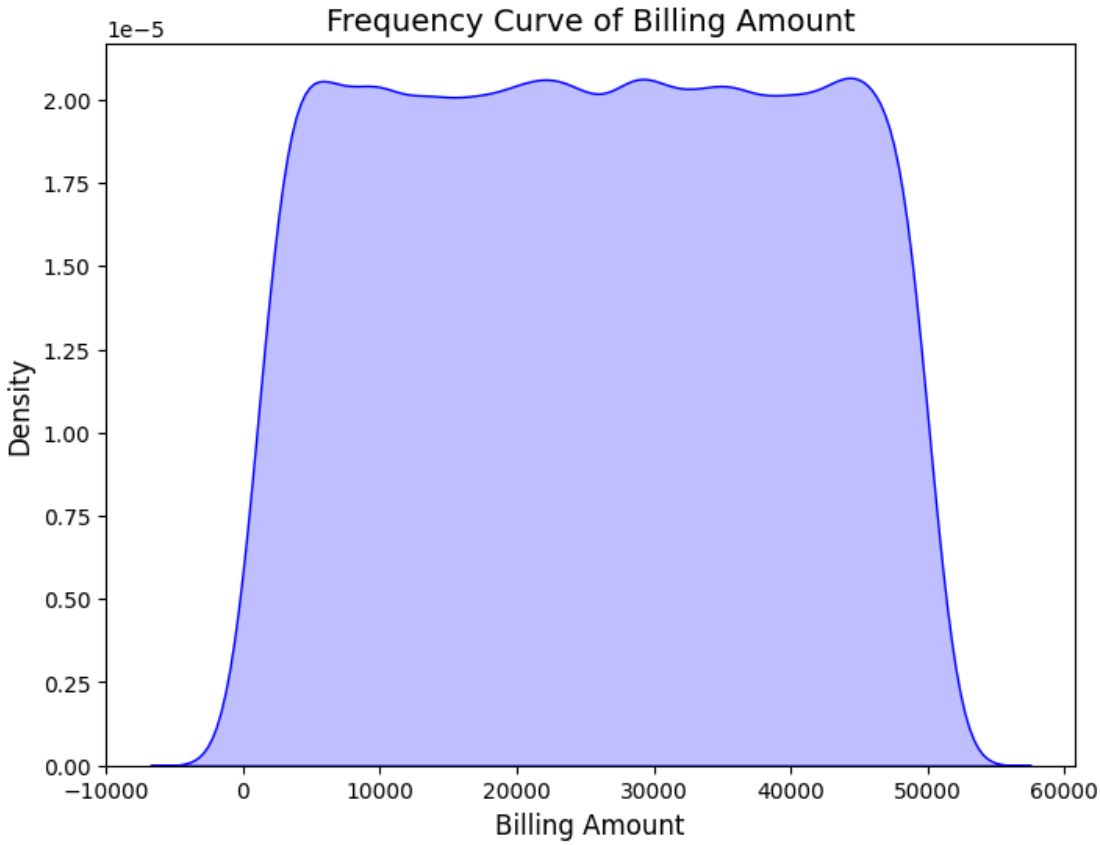
[8]:
```
plt.figure(figsize=(8, 6))
plt.hist(data['Billing Amount'], bins=10000, color='red')

# Add titles and labels
plt.title('Histogram of Billing Amount', fontsize=14)
plt.xlabel('Billing Amount', fontsize=12)
plt.ylabel('Frequency', fontsize=12)
```

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



frequency curve of the frequency distribution of the following

[9]:
```
plt.figure(figsize=(8, 6))
sns.kdeplot(data['Billing Amount'], color='blue', fill=True)

# Add titles and labels
plt.title('Frequency Curve of Billing Amount', fontsize=14)
plt.xlabel('Billing Amount', fontsize=12)
plt.ylabel('Density', fontsize=12)

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



[10]:
```
data.columns
```

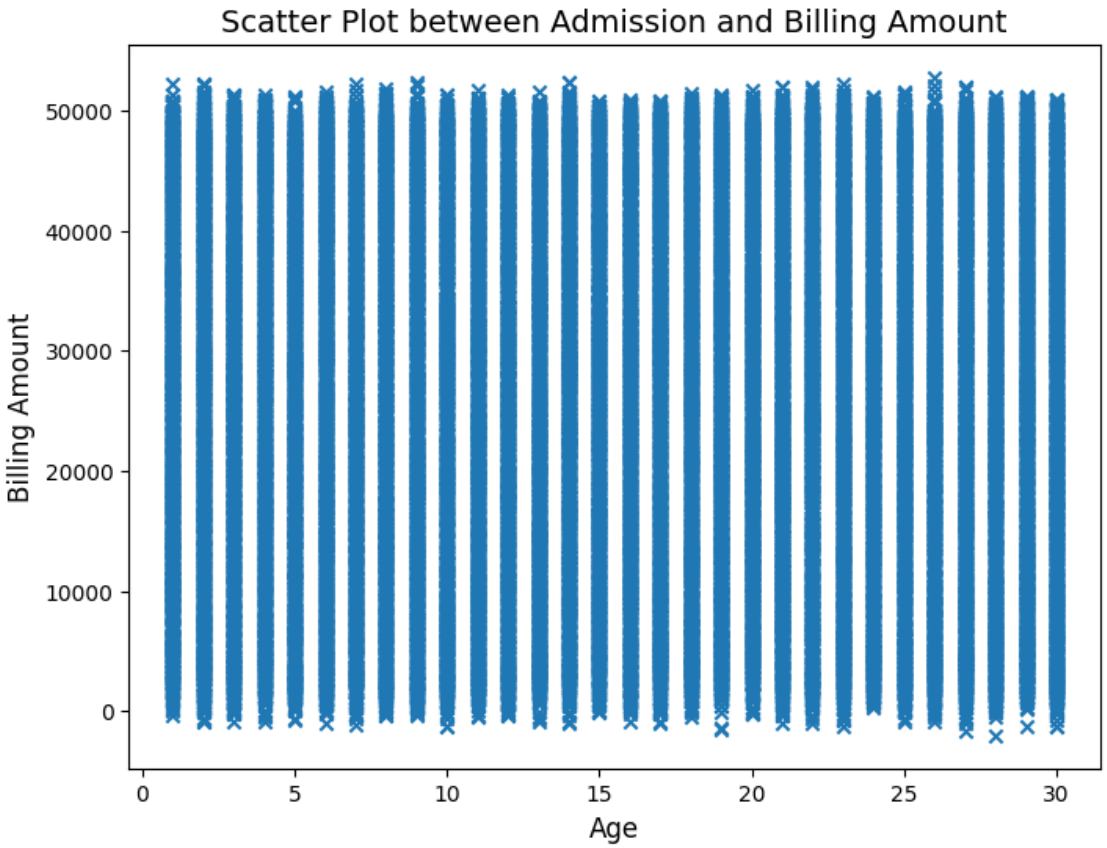
[10]:
```
Index(['Name', 'Age', 'Gender', 'Blood Type', 'Medical Condition',
       'Date of Admission', 'Doctor', 'Hospital', 'Insurance Provider',
       'Billing Amount', 'Room Number', 'Admission Type', 'Discharge Date',
       'Medication', 'Test Results'],
      dtype='object')
```

Scatter Plot between Age and Billing Amount

[11]:
```
# Convert the 'Date of Admission' and 'Discharge Date' columns to datetime
data['Date of Admission'] = pd.to_datetime(data['Date of Admission'])
data['Discharge Date'] = pd.to_datetime(data['Discharge Date'])
# Calculate the difference in days between the two dates
data['Admission Duration'] = (data['Discharge Date'] - data['Date of␣
 ↪Admission']).dt.days
# Display the DataFrame with the new column
```

Scatter Plot between Admission and Billing Amount

[12]:
```
plt.figure(figsize=(8, 6))
plt.scatter(data['Admission Duration'], data['Billing Amount'],marker='x')

# Add titles and labels
plt.title('Scatter Plot between Admission and Billing Amount', fontsize=14)
plt.xlabel('Age', fontsize=12)
plt.ylabel('Billing Amount', fontsize=12)

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

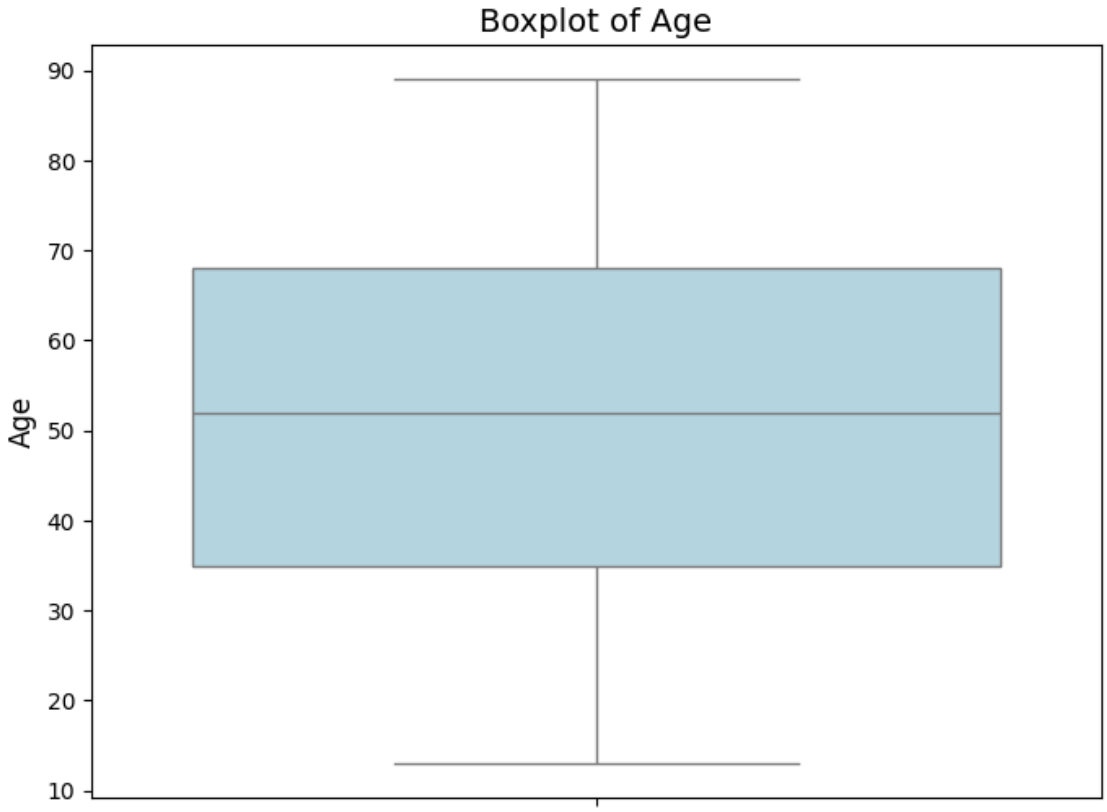


Box plot of age column

[13]:
```
plt.figure(figsize=(8, 6))
sns.boxplot(y=data['Age'], color='lightblue')

# Add titles and labels
plt.title('Boxplot of Age', fontsize=14)
plt.ylabel('Age', fontsize=12)
```

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

```
c:\Users\Acer\AppData\Local\Programs\Python\Python312\Lib\site-
packages\seaborn\categorical.py:640: FutureWarning: SeriesGroupBy.grouper is
deprecated and will be removed in a future version of pandas.
  positions = grouped.grouper.result_index.to_numpy(dtype=float)
```



calculating the quartiles of the age column

[14]:
```
Q1 = data['Age'].quantile(0.25)  # First quartile (25th percentile)
Q2 = data['Age'].quantile(0.50)  # Second quartile (50th percentile or median)
Q3 = data['Age'].quantile(0.75)  # Third quartile (75th percentile)

# Display the quartiles
print(f"First Quartile (Q1): {Q1}")
print(f"Second Quartile (Q2): {Q2}")
print(f"Third Quartile (Q3): {Q3}")
```

```
First Quartile (Q1): 35.0
Second Quartile (Q2): 52.0
Third Quartile (Q3): 68.0
```

MEAN, MEDIAN, SKEWNESS, RANGE, MEAN DEVIATION, VARIANCE, STANDARD DEVIATION and coefficient of variation (CV) for the "Age" and "Billing Amount" columnsT

[15]:
```
from scipy.stats import skew

# Sample data for 'Age' and 'Billing Amount'
copy_data = {'Age': data['Age'],
        'Billing Amount': data['Billing Amount']}
df = pd.DataFrame(copy_data)

# Function to calculate mean deviation
def mean_deviation(series):
    mean_value = series.mean()
    return (series - mean_value).abs().mean()
# Calculate statistics for 'Age' and 'Billing Amount'
for column in ['Age', 'Billing Amount']:
    print(f"Statistics for {column}:")
    print(f"Mean: {df[column].mean()}")
    print(f"Median: {df[column].median()}")
    print(f"Skewness: {skew(df[column])}")
    print(f"Range: {df[column].max() - df[column].min()}")
    print(f"Mean Deviation: {mean_deviation(df[column])}")
    print(f"Variance: {df[column].var()}")
    print(f"Standard Deviation: {df[column].std()}")
    print(f"Coefficient of Variation (CV): {df[column].std() / df[column].
 ↪mean()}")
    print()
```

```
Statistics for Age:
Mean: 51.53945945945946
Median: 52.0
Skewness: -0.005735115665703919
Range: 76
Mean Deviation: 16.948137092768444
Variance: 384.2561953149387
Standard Deviation: 19.602453808514348
Coefficient of Variation (CV): 0.38033875430791986

Statistics for Billing Amount:
Mean: 25539.316097211795
Median: 25538.069375965664
Skewness: -0.0009977690118698563
Range: 54772.76887632831
Mean Deviation: 12297.475837537435
Variance: 201965437.04053578
Standard Deviation: 14211.454430864414
Coefficient of Variation (CV): 0.5564539933947535
```

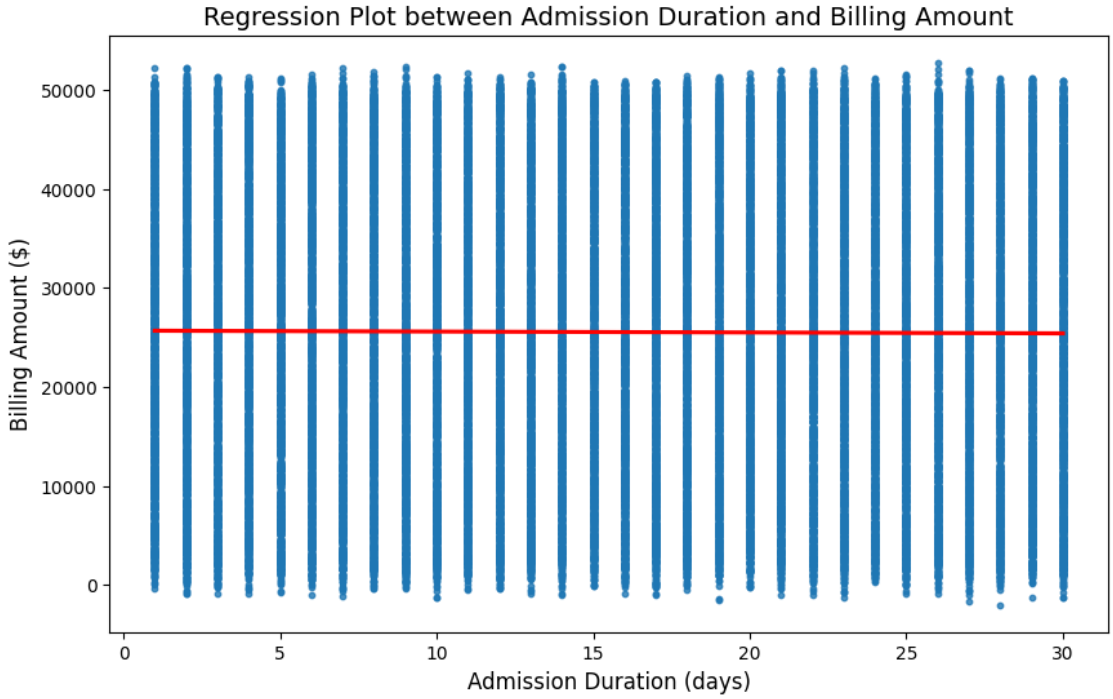
Correlation and correlation coefficient between the "Admission Duration" and "Billing Amount" columns

[16]:
```
import pandas as pd

# Sample data for 'Age' and 'Billing Amount'
copy_data = {'Admission Duration': data['Admission Duration'],
        'Billing Amount': data['Billing Amount']}
df = pd.DataFrame(copy_data)

# Calculate the correlation matrix
correlation_matrix = df.corr()

# Extract the correlation coefficient between 'Age' and 'Billing Amount'
corr_coefficient = df['Admission Duration'].corr(df['Billing Amount'])

# Display the correlation matrix and the specific correlation coefficient
print("Correlation Matrix:")
print(correlation_matrix)

print(f"\nCorrelation Coefficient between Age and Billing Amount:␣
 ↪{corr_coefficient:.4f}")
```

```
Correlation Matrix:
                    Admission Duration  Billing Amount
Admission Duration            1.000000       -0.005602
Billing Amount               -0.005602        1.000000

Correlation Coefficient between Age and Billing Amount: -0.0056
```

Regression Plot between Admission Duration and Billing Amount

[17]:
```
plt.figure(figsize=(10, 6))
sns.regplot(x='Admission Duration', y='Billing Amount', data=data,␣
 ↪scatter_kws={'s':10}, line_kws={'color':'red'})

# Add titles and labels
plt.title('Regression Plot between Admission Duration and Billing Amount',␣
 ↪fontsize=14)
plt.xlabel('Admission Duration (days)', fontsize=12)
plt.ylabel('Billing Amount ($)', fontsize=12)

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


Regression Plot between Admission Duration and Billing Amount
Billing Amount ($)
50000
40000
30000
20000
10000
0
0
5
10
15
20
25
30
Admission Duration (days)