

231501167

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EX NO: 1 Setting up the Python environment and libraries- Jupyter Notebook

Create a new notebook for Python

Write and execute Python code

Create new cells for code and Markdown

Demonstrate the application of Jupyter Widgets, Jupyter AI

```
import ipywidgets as widgets
from IPython.display import display

slider = widgets.IntSlider(description='Slider:', min=0, max=100,
value=25)

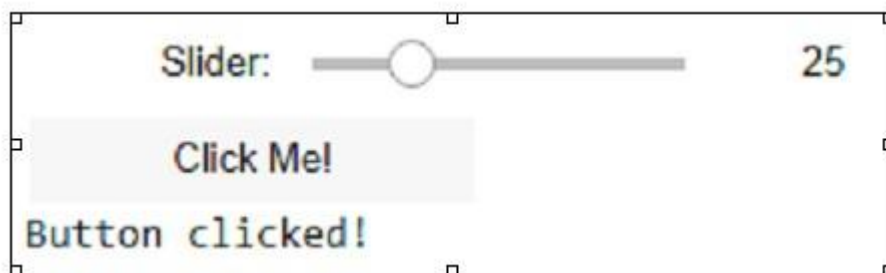
display(slider)

button = widgets.Button(description="Click Me!")

display(button)

def on_button_click(b):
    print("Button clicked!")

button.on_click(on_button_click)
```



EXP NO:2

EDA-Data Import and Export

Importing data from CSV, Excel, SQL databases, and web scraping

Handling different data formats

Export a DataFrame to an Excel file.

```
import pandas as pd
```

```
df_csv = pd.read_csv('/content/data.csv')
```

```
df_csv.head()
```

	Make	Model	Year	Engine Fuel Type	Engine HP	Engine Cylinders	Transmission Type	Driven_Wheels	Number of Doors	Market	Category	Vehicle Size	Vehicle Style	highwa MP
0	BMW	Series M	2011	premium unleaded (required)	335.0	6.0	MANUAL	rear wheel drive	2.0	Tuner,Luxury,High- Performance	Factory	Compact	Coupe	2
1	BMW	Series	2011	premium unleaded (required)	300.0	6.0	MANUAL	rear wheel drive	2.0	Luxury,Performance		Compact	Convertible	2
2	BMW	Series	2011	premium unleaded (required)	300.0	6.0	MANUAL	rear wheel drive	2.0	Luxury,High- Performance		Compact	Coupe	2
3	BMW	Series	2011	premium unleaded (required)	230.0	6.0	MANUAL	rear wheel drive	2.0	Luxury,Performance		Compact	Coupe	2
4	BMW	Series	2011	premium unleaded (required)	230.0	6.0	MANUAL	rear wheel drive	2.0	Luxury		Compact	Convertible	2

```
df_excel = pd.read_excel('/content/data.xlsx')
```

```
df_excel.head()
```

	Make	Model	Year	Engine Fuel Type	Engine HP	Engine Cylinders	Transmission Type	Driven_Wheels	Number of Doors	Market	Category	Vehicle Size	Vehicle Style	highwa MP
0	BMW	Series M	2011	premium unleaded (required)	335.0	6.0	MANUAL	rear wheel drive	2.0	Tuner,Luxury,High- Performance	Factory	Compact	Coupe	2
1	BMW	Series	2011	premium unleaded (required)	300.0	6.0	MANUAL	rear wheel drive	2.0	Luxury,Performance		Compact	Convertible	2
2	BMW	Series	2011	premium unleaded (required)	300.0	6.0	MANUAL	rear wheel drive	2.0	Luxury,High- Performance		Compact	Coupe	2
3	BMW	Series	2011	premium unleaded (required)	230.0	6.0	MANUAL	rear wheel drive	2.0	Luxury,Performance		Compact	Coupe	2
4	BMW	Series	2011	premium unleaded (required)	230.0	6.0	MANUAL	rear wheel drive	2.0	Luxury		Compact	Convertible	2

```
import sqlite3
```

```
conn = sqlite3.connect(':memory:')
```

```
df.to_sql('data_table', conn, index=False, if_exists='replace')
```

```
[9] import sqlite3
    # Create an in-memory SQLite database
    conn = sqlite3.connect(':memory:')

    # Save DataFrame as SQL table
    df.to_sql('data_table', conn, index=False, if_exists='replace')
```

11914

```
query = "SELECT * FROM data_table LIMIT 5;"
```

```
result = pd.read_sql_query(query, conn)
```

result

	Make	Model	Year	Engine Fuel Type	Engine HP	Engine Cylinders	Transmission Type	Driven_Wheels	Number of Doors	Market	Category	Vehicle Size	Vehicle Style	highwa MP
0	BMW	Series M	2011	premium unleaded (required)	335.0	6.0	MANUAL	rear wheel drive	2.0	Tuner,Luxury,High- Performance	Factory	Compact	Coupe	2
1	BMW	Series	2011	premium unleaded (required)	300.0	6.0	MANUAL	rear wheel drive	2.0	Luxury,Performance		Compact	Convertible	2
2	BMW	Series	2011	premium unleaded (required)	300.0	6.0	MANUAL	rear wheel drive	2.0	Luxury,High- Performance		Compact	Coupe	2
3	BMW	Series	2011	premium unleaded (required)	230.0	6.0	MANUAL	rear wheel drive	2.0	Luxury,Performance		Compact	Coupe	2
4	BMW	Series	2011	premium unleaded (required)	230.0	6.0	MANUAL	rear wheel drive	2.0	Luxury		Compact	Convertible	2

```
df.to_html('data.htm', index=False)
```

```
df_scraped = pd.read_html('data.htm')[0]
```

```
print(df_scraped.head())
```

	Make	Model	Year	Engine	Fuel Type	Engine HP	\
0	BMW	1 Series M	2011	premium	unleaded (required)	335.0	
1	BMW	1 Series	2011	premium	unleaded (required)	300.0	
2	BMW	1 Series	2011	premium	unleaded (required)	300.0	
3	BMW	1 Series	2011	premium	unleaded (required)	230.0	
4	BMW	1 Series	2011	premium	unleaded (required)	230.0	
	Engine	Cylinders	Transmission	Type	Driven_Wheels	Number of Doors	\
0		6.0		MANUAL	rear wheel drive	2.0	
1		6.0		MANUAL	rear wheel drive	2.0	
2		6.0		MANUAL	rear wheel drive	2.0	
3		6.0		MANUAL	rear wheel drive	2.0	
4		6.0		MANUAL	rear wheel drive	2.0	
	Market Category	Vehicle Size	Vehicle Style	\			
0	Factory Tuner,Luxury,High-Performance	Compact	Coupe				
1	Luxury,Performance	Compact	Convertible				
2	Luxury,High-Performance	Compact	Coupe				
3	Luxury,Performance	Compact	Coupe				
4	Luxury	Compact	Convertible				
	highway	MPG	city	mpg	Popularity	MSRP	
0		26		19	3916	46135	
1		28		19	3916	40650	
2		28		20	3916	36350	
3		28		18	3916	29450	
4		28		18	3916	34500	

EX NO: 3 EDA-Data Cleaning

- ☐ **Handling missing values: detection, filling, and dropping**
- ☐ **Removing duplicates and unnecessary data**
- ☐ **Data type conversion and ensuring consistency**
- ☐ **Normalize data (e.g., standardization, min-max scaling).**

```
import pandas as pd  
  
df=pd.read_csv('/content/data.csv')  
  
print(df.isnull().sum())  
  
print(df.info())
```

```

Make          0
Model         0
Year          0
Engine Fuel Type  3
Engine HP     69
Engine Cylinders 30
Transmission Type  0
Driven_Wheels  0
Number of Doors  6
Market Category 3742
Vehicle Size   0
Vehicle Style  0
highway MPG    0
city mpg       0
Popularity     0
MSRP           0
dtype: int64
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11914 entries, 0 to 11913
Data columns (total 16 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Make                  11914 non-null  object
1   Model                 11914 non-null  object
2   Year                  11914 non-null  int64
3   Engine Fuel Type      11911 non-null  object
4   Engine HP             11845 non-null  float64
5   Engine Cylinders      11884 non-null  float64
6   Transmission Type     11914 non-null  object
7   Driven_Wheels         11914 non-null  object
8   Number of Doors       11908 non-null  float64
9   Market Category       8172 non-null   object
10  Vehicle Size          11914 non-null  object
11  Vehicle Style         11914 non-null  object
12  highway MPG           11914 non-null  int64

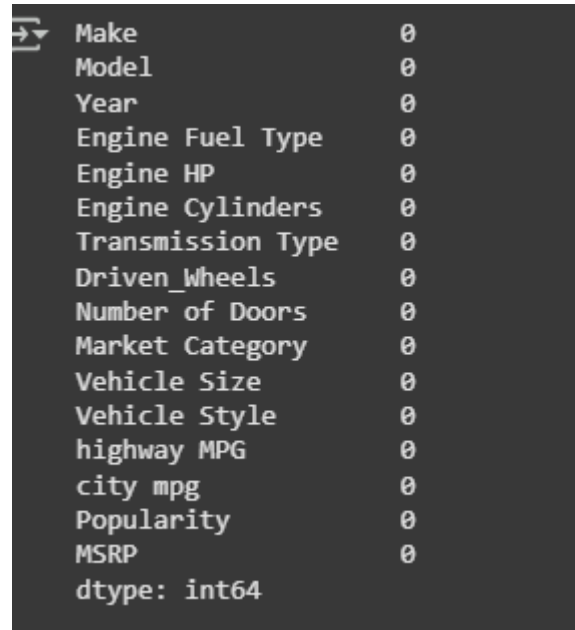
```



```

df.dropna(inplace=True)
print(df.isnull().sum())

```

A terminal window with a dark background showing a list of car features and their counts. The features are listed on the left, and the counts (all 0) are on the right. The last line shows the data type for the entire list.

Make	0
Model	0
Year	0
Engine Fuel Type	0
Engine HP	0
Engine Cylinders	0
Transmission Type	0
Driven_Wheels	0
Number of Doors	0
Market Category	0
Vehicle Size	0
Vehicle Style	0
highway MPG	0
city mpg	0
Popularity	0
MSRP	0
dtype: int64	

```
df.drop_duplicates(inplace=True)
```

```
df['Make'] = df['Make'].str.title()
```

```
df['Model'] = df['Model'].str.title()
```

```
df['Transmission Type'] = df['Transmission Type'].astype('category')
```

```
df['Driven_Wheels'] = df['Driven_Wheels'].astype('category')
```

```
df['Vehicle Size'] = df['Vehicle Size'].astype('category')
```

```
df['Vehicle Style'] = df['Vehicle Style'].astype('category')
```

```
print(df.dtypes)
```

```
Make          object
Model         object
Year          int64
Engine Fuel Type  object
Engine HP      float64
Engine Cylinders float64
Transmission Type category
Driven_Wheels  category
Number of Doors float64
Market Category object
Vehicle Size   category
Vehicle Style  category
highway MPG    int64
city mpg       int64
Popularity     int64
MSRP          int64
dtype: object
```

```
from sklearn.preprocessing import StandardScaler, MinMaxScaler
```

```
numeric_cols = ['Engine HP', 'Engine Cylinders', 'highway MPG', 'city mpg', 'Popularity',  
'MSRP']
```

```
scaler = StandardScaler()
```

```
df[numeric_cols] = scaler.fit_transform(df[numeric_cols])
```

```
# Min-Max Scaling (optional alternative)
```

```
# minmax = MinMaxScaler()
```

```
# df[numeric_cols] = minmax.fit_transform(df[numeric_cols])
```

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


EX NO: 4 EDA-Data Inspection and Analysis

- ❑ Viewing and inspecting DataFrames
- ❑ Filtering and subsetting data using conditions
- ❑ Descriptive statistics: measures of central tendency (mean, median, mode) and measures of dispersion (range, variance, standard deviation)

```
import pandas as pd
```

```
df=pd.read_csv("data.csv")
```

```
print(df.head())
```

```
print("Rows:", df.shape[0], "Columns:", df.shape[1])
```

```
print(df.dtypes)
```

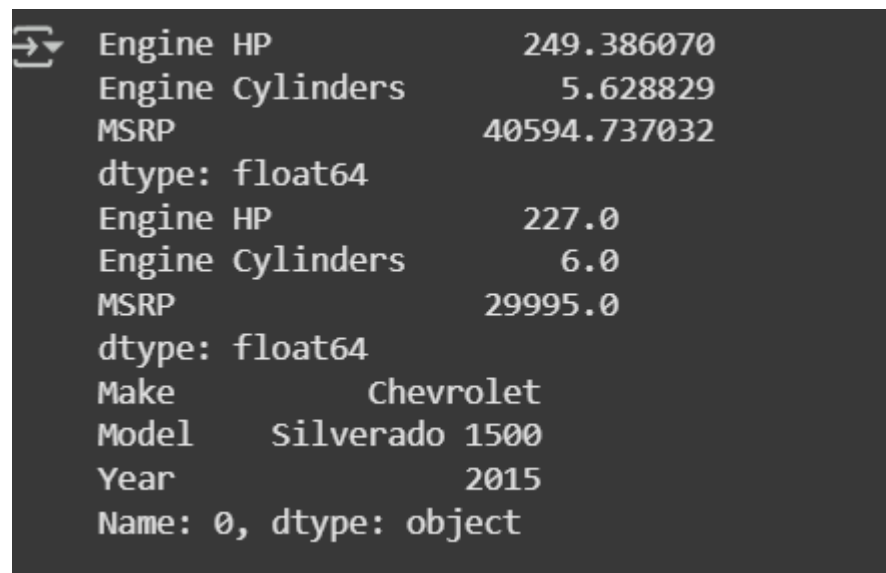
```
print(df.isnull().sum())
```

```
print(df.describe())
```

	Make	Model	Year	Engine	Fuel Type	Engine HP	\
0	BMW	1 Series M	2011	premium unleaded (required)		335.0	
1	BMW	1 Series	2011	premium unleaded (required)		300.0	
2	BMW	1 Series	2011	premium unleaded (required)		300.0	
3	BMW	1 Series	2011	premium unleaded (required)		230.0	
4	BMW	1 Series	2011	premium unleaded (required)		230.0	
	Engine	Cylinders	Transmission	Type	Driven_Wheels	Number of Doors	\
0		6.0	MANUAL		rear wheel drive	2.0	
1		6.0	MANUAL		rear wheel drive	2.0	
2		6.0	MANUAL		rear wheel drive	2.0	
3		6.0	MANUAL		rear wheel drive	2.0	
4		6.0	MANUAL		rear wheel drive	2.0	
	Market Category	Vehicle Size	Vehicle Style	\			
0	Factory Tuner,Luxury,High-Performance	Compact	Coupe				
1	Luxury,Performance	Compact	Convertible				
2	Luxury,High-Performance	Compact	Coupe				
3	Luxury,Performance	Compact	Coupe				
4	Luxury	Compact	Convertible				
	highway MPG	city mpg	Popularity	MSRP			
0	26	19	3916	46135			
1	28	19	3916	40650			
2	28	20	3916	36350			
3	28	18	3916	29450			
4	28	18	3916	34500			

```
car_after_2015 = df[df['Year'] > 2015]
high_hp_cars = df[df['Engine HP'] > 300]
selected_columns = df[['Make', 'Model', 'MSRP']]
luxury_cars = df[df['Market Category'].str.contains('Luxury', na=False)]
```

```
print(df[['Engine HP', 'Engine Cylinders', 'MSRP']].mean())
print(df[['Engine HP', 'Engine Cylinders', 'MSRP']].median())
print(df[['Make', 'Model', 'Year']].mode().iloc[0])
```



```
Engine HP      249.386070
Engine Cylinders  5.628829
MSRP           40594.737032
dtype: float64
Engine HP      227.0
Engine Cylinders  6.0
MSRP           29995.0
dtype: float64
Make      Chevrolet
Model    Silverado 1500
Year      2015
Name: 0, dtype: object
```

```
range_values = df[['Engine HP', 'Engine Cylinders', 'MSRP']].max() - df[['Engine HP', 'Engine Cylinders', 'MSRP']].min()
print("Range:\n", range_values)
print("Variance:\n", df[['Engine HP', 'Engine Cylinders', 'MSRP']].var())
print("Standard Deviation:\n", df[['Engine HP', 'Engine Cylinders', 'MSRP']].std())
```

```
Range:
  Engine HP          946.0
  Engine Cylinders    16.0
  MSRP              2063902.0
  dtype: float64

Variance:
  Engine HP          1.192286e+04
  Engine Cylinders    3.170392e+00
  MSRP              3.613104e+09
  dtype: float64

Standard Deviation:
  Engine HP          109.191870
  Engine Cylinders    1.780559
  MSRP              60109.103604
  dtype: float64
```

```
import matplotlib.pyplot as plt
```

```
# Histogram of Engine HP
```

```
df['Engine HP'].dropna().hist(bins=20)
```

```
plt.title("Distribution of Engine HP")
```

```
plt.xlabel("Engine HP")
```

```
plt.ylabel("Frequency")
```

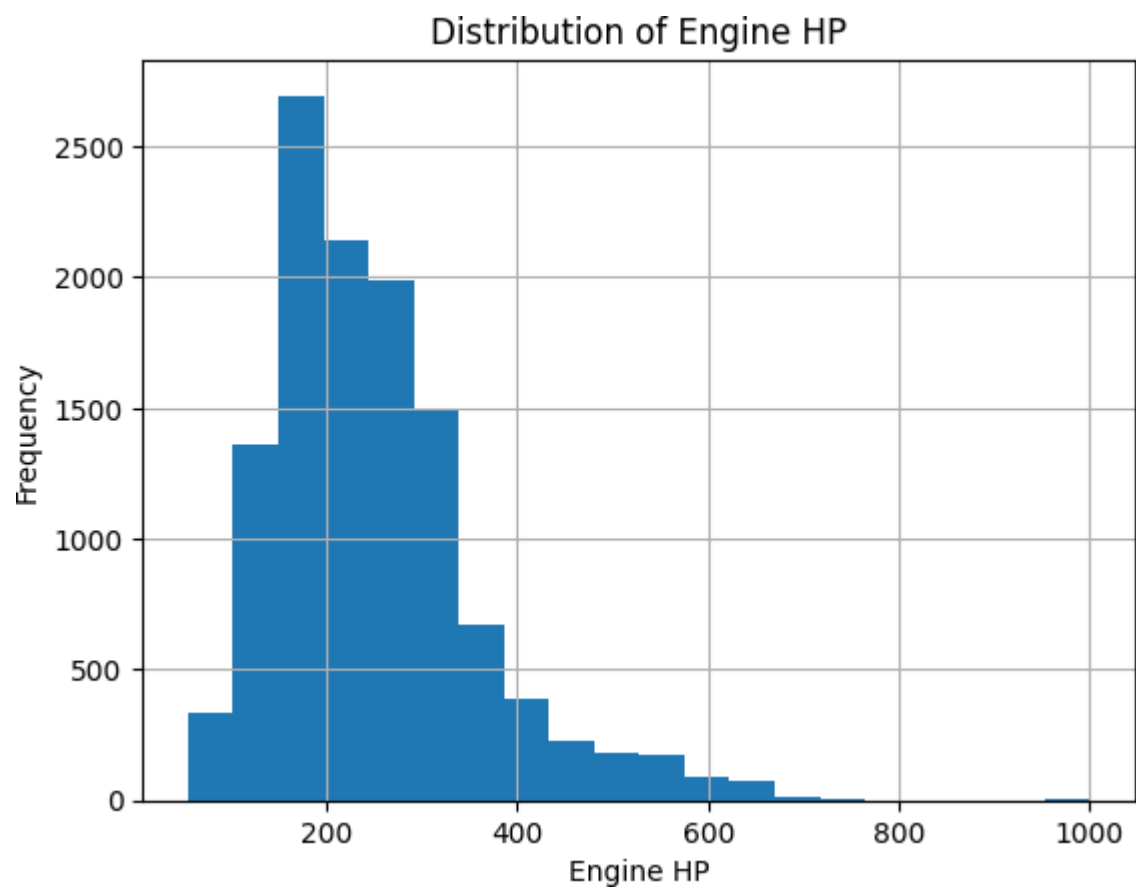
```
plt.show()
```

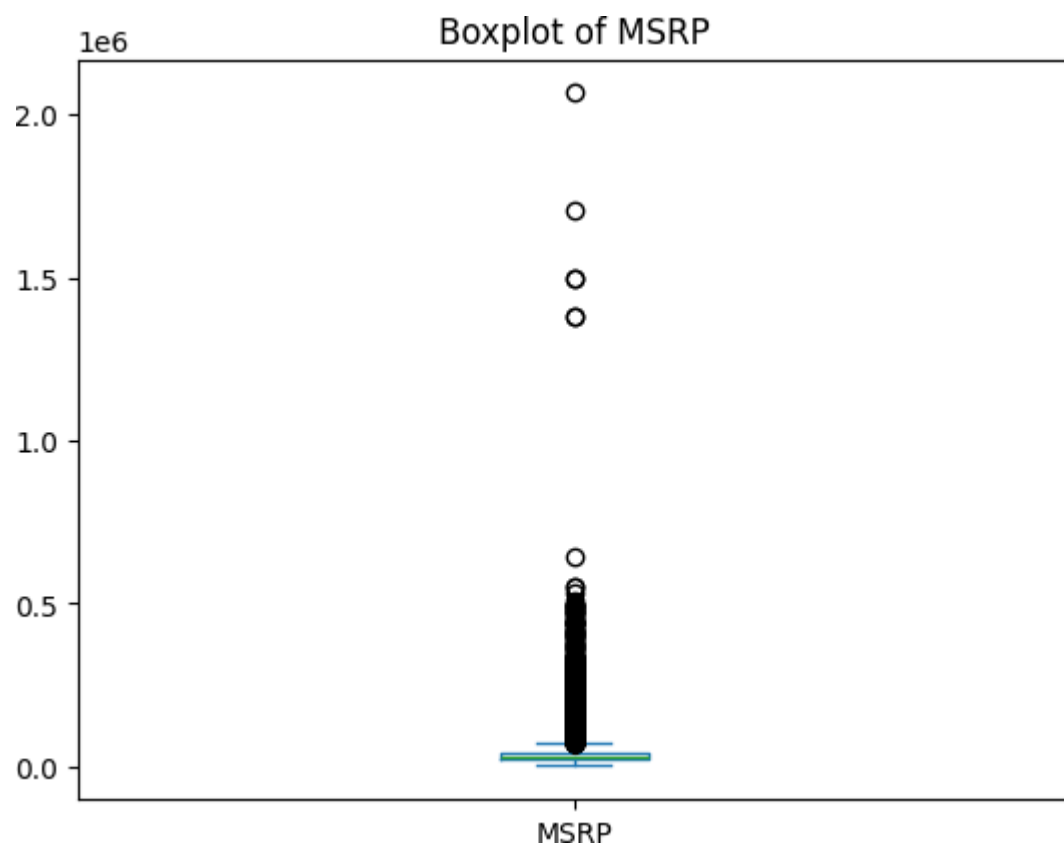
```
# Boxplot of MSRP
```

```
df['MSRP'].dropna().plot(kind='box')
```

```
plt.title("Boxplot of MSRP")
```

```
plt.show()
```





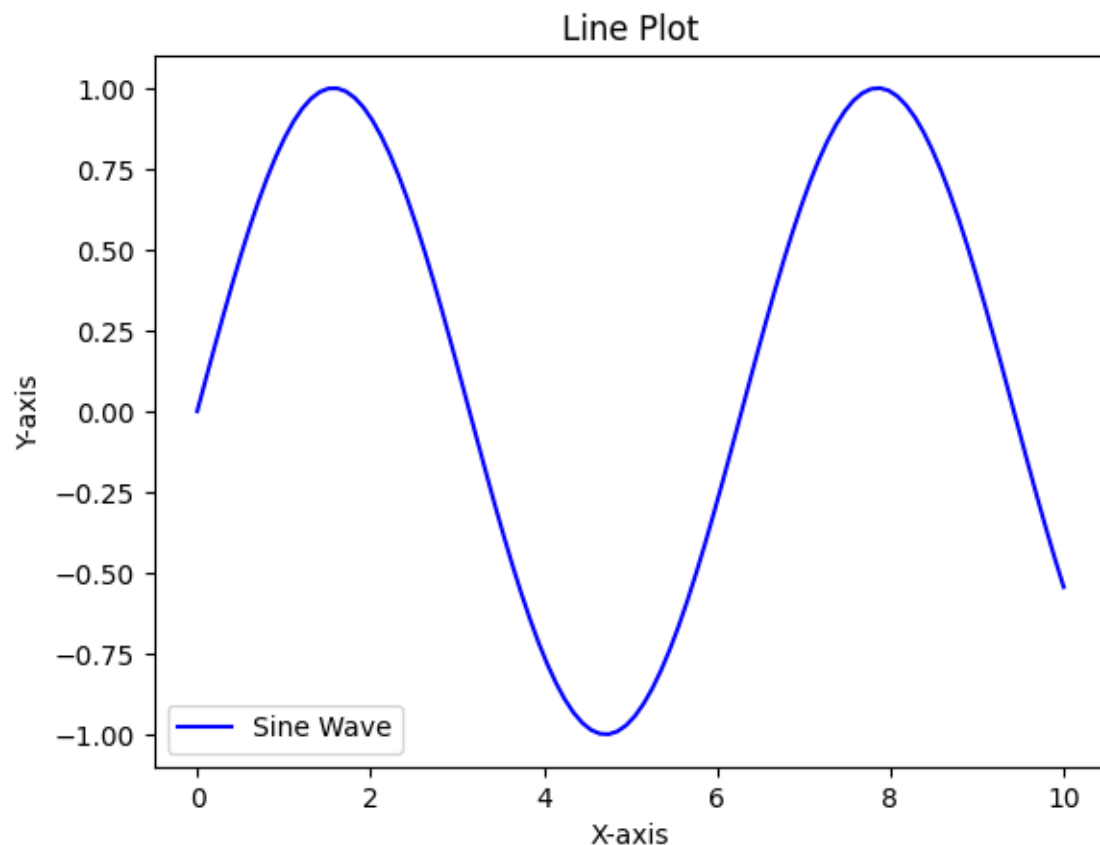
EX NO : 5 EDA – DATA VISUALIZATION

```
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
```

```
df = pd.read_csv("/content/data.csv") # Replace with your file
df.head()
```

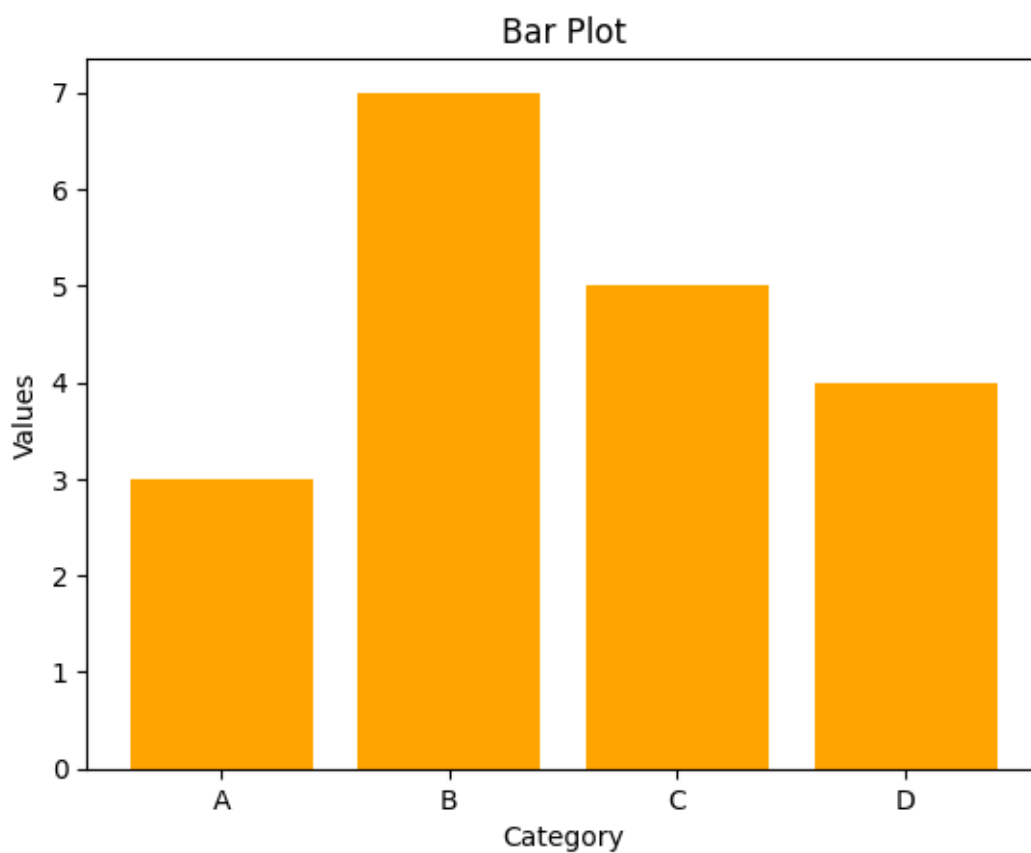
LINE CHART:

```
x = np.linspace(0, 10, 100)
y = np.sin(x)
plt.plot(x, y, color='blue', label='Sine Wave')
plt.xlabel('X-axis')
plt.ylabel('Y-axis')
plt.title('Line Plot')
plt.legend()
plt.show()
```



BAR CHART:

```
categories = ['A', 'B', 'C', 'D']  
values = [3, 7, 5, 4]  
plt.bar(categories, values, color='orange')  
plt.xlabel('Category')  
plt.ylabel('Values')  
plt.title('Bar Plot')  
plt.show()
```



HISTOGRAM:

```
data = np.random.randn(1000)  
plt.hist(data, bins=20, color='purple', edgecolor='black')  
plt.xlabel('Value')  
plt.ylabel('Frequency')  
plt.title('Histogram')  
plt.show()
```

