

Prodigy InfoTech Internship

Task 2

Perform data cleaning and exploratory data analysis (EDA) on a dataset of your choice, such as the Titanic dataset from Kaggle. Explore the relationships between variables and identify patterns and trends in the data.

Sample Dataset:- [Titanic Dataset](#)

```
[1] import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from google.colab import drive
drive.mount('/content/drive')
from google.colab import files

Mounted at /content/drive
```

➤ Understanding the shape of the Dataset:

```
[2] data=pd.read_csv('/content/drive/My Drive/titanic.csv', index_col='PassengerId')
```

```
[3] data.head()
```

Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 1309 entries, 1 to 1309
Data columns (total 11 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   Survived    1309 non-null   int64
1   Pclass      1309 non-null   int64
2   Name        1309 non-null   object
3   Sex         1309 non-null   object
4   Age         1046 non-null   float64
5   SibSp       1309 non-null   int64
6   Parch       1309 non-null   int64
7   Ticket      1309 non-null   object
8   Fare        1308 non-null   float64
9   Cabin       295 non-null    object
10  Embarked    1307 non-null   object
dtypes: float64(2), int64(4), object(5)
memory usage: 122.7+ KB
```

✓ 0s [5] data.describe()

	Survived	Pclass	Age	SibSp	Parch	Fare
count	1309.000000	1309.000000	1046.000000	1309.000000	1309.000000	1308.000000
mean	0.377387	2.294882	29.881138	0.498854	0.385027	33.295479
std	0.484918	0.837836	14.413493	1.041658	0.865560	51.758668
min	0.000000	1.000000	0.170000	0.000000	0.000000	0.000000
25%	0.000000	2.000000	21.000000	0.000000	0.000000	7.895800
50%	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
75%	1.000000	3.000000	39.000000	1.000000	0.000000	31.275000
max	1.000000	3.000000	80.000000	8.000000	9.000000	512.329200

➤ Data Cleaning:

✓ 0s data.drop(columns='Cabin').isna().mean()

Survived	0.000000
Pclass	0.000000
Name	0.000000
Sex	0.000000
Age	0.200917
SibSp	0.000000
Parch	0.000000
Ticket	0.000000
Fare	0.000764
Embarked	0.001528

dtype: float64

0s

data.dropna(subset=['Embarked'])

Survived

Pclass

Name

Sex

Age

SibSp

Parch

Ticket

Fare

Cabin

Embarked

PassengerId

1

0

3

Braund, Mr. Owen Harris

male

22.0

1

0

A/5 21171

7.2500

NaN

S

2

1

1

Cummings, Mrs. John Bradley (Florence Briggs Th...

female

38.0

1

0

PC 17599

71.2833

C85

C

3

1

3

Heikkinen, Miss. Laina

female

26.0

0

0

STON/O2. 3101282

7.9250

NaN

S

4

1

1

Futrelle, Mrs. Jacques Heath (Lily May Peel)

female

35.0

1

0

113803

53.1000

C123

S

5

0

3

Allen, Mr. William Henry

male

35.0

0

0

373450

8.0500

NaN

S

...

...

...

...

...

...

...

...

...

...

...

...

1305

0

3

Spector, Mr. Woolf

male

NaN

0

0

A.5. 3236

8.0500

NaN

S

1306

1

1

Oliva y Ocana, Dona. Fermina

female

39.0

0

0

PC 17758

108.9000

C105

C

1307

0

3

Saether, Mr. Simon Sivertsen

male

38.5

0

0

SOTON/O.Q. 3101262

7.2500

NaN

S

1308

0

3

Ware, Mr. Frederick

male

NaN

0

0

359309

8.0500

NaN

S

1309

0

3

Peter, Master. Michael J

male

NaN

1

1

2668

22.3583

NaN

C

1307 rows x 11 columns

```
[8] data['Age'] = data['Age'].round()
data['Pclass'] = data['Pclass'].map({1: 'Upper', 2: 'Middle', 3: 'Lower'})
data['Embarked'] = data['Embarked'].map({'C': 'Cherbourg', 'Q': 'Queenstown', 'S': 'Southampton'})
data['Survived'] = data['Survived'].map({0: 'Survived', 1: 'Not Survived'})
data['Sex'] = data['Sex'].str.title()
categorical_columns = ['Sex', 'Parch', 'SibSp', 'Pclass', 'Embarked', 'Survived']
data[categorical_columns] = data[categorical_columns].astype('category')
```

data

Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
PassengerId										
1	Survived	Lower	Braund, Mr. Owen Harris	Male	22.0	1	0	A/5 21171	7.2500	NaN Southampton
2	Not Survived	Upper	Cummings, Mrs. John Bradley (Florence Briggs Th...	Female	38.0	1	0	PC 17599	71.2833	C85 Cherbourg
3	Not Survived	Lower	Heikkinen, Miss. Laina	Female	26.0	0	0	STON/O2. 3101282	7.9250	NaN Southampton
4	Not Survived	Upper	Futrelle, Mrs. Jacques Heath (Lily May Peel)	Female	35.0	1	0	113803	53.1000	C123 Southampton
5	Survived	Lower	Allen, Mr. William Henry	Male	35.0	0	0	373450	8.0500	NaN Southampton
...
1305	Survived	Lower	Spector, Mr. Woolf	Male	NaN	0	0	A.5. 3236	8.0500	NaN Southampton
1306	Not Survived	Upper	Oliva y Ocana, Dona. Fermina	Female	39.0	0	0	PC 17758	108.9000	C105 Cherbourg
1307	Survived	Lower	Saether, Mr. Simon Sivertsen	Male	38.0	0	0	SOTON/O.Q. 3101262	7.2500	NaN Southampton
1308	Survived	Lower	Ware, Mr. Frederick	Male	NaN	0	0	359309	8.0500	NaN Southampton
1309	Survived	Lower	Peter, Master. Michael J	Male	NaN	1	1	2668	22.3583	NaN Cherbourg

1309 rows x 11 columns

```
[10] data.dropna(subset=['Age'], inplace=True)

[11] cols = [
    'Name', 'Sex', 'Age', 'Parch', 'SibSp',
    'Ticket', 'Pclass', 'Embarked', 'Fare',
    'Survived',
]
data = data[cols]

[12] data.to_csv('/content/drive/My Drive/titanic_new.csv')
```

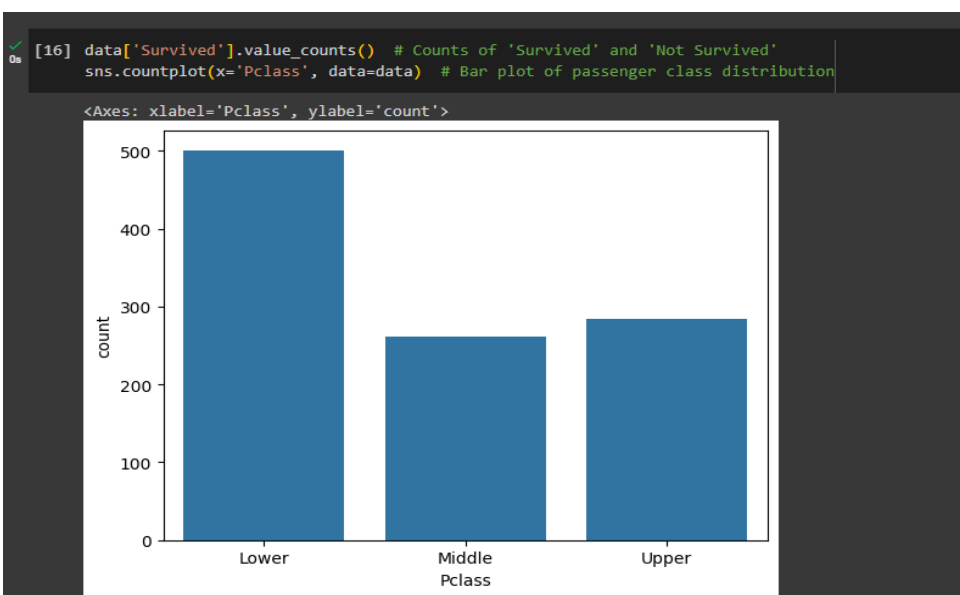
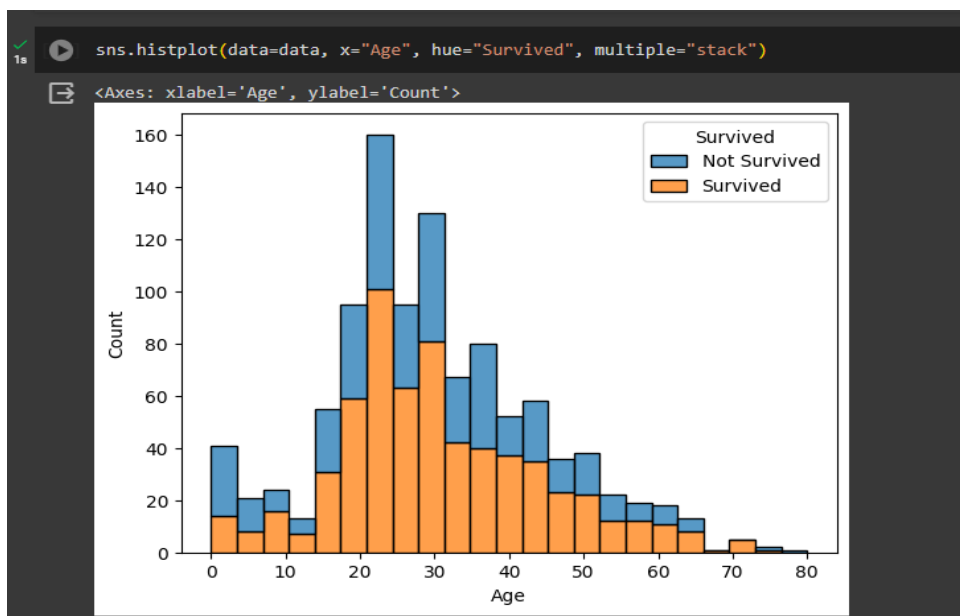
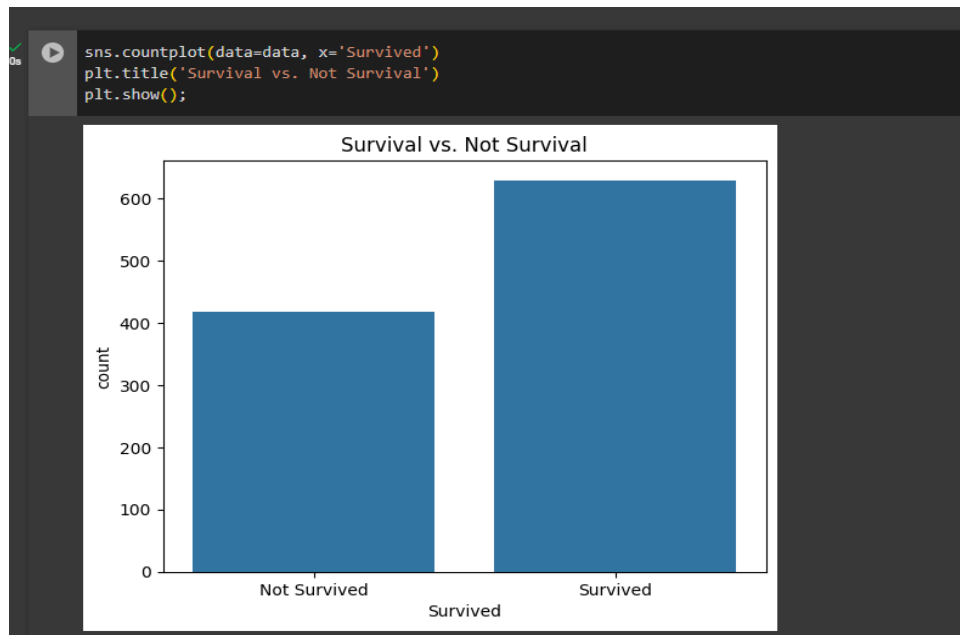
[13] data.head()

	Name	Sex	Age	Parch	SibSp	Ticket	Pclass	Embarked	Fare	Survived
PassengerId										
1	Braund, Mr. Owen Harris	Male	22.0	0	1	A/5 21171	Lower	Southampton	7.2500	Survived
2	Cummings, Mrs. John Bradley (Florence Briggs Th...	Female	38.0	0	1	PC 17599	Upper	Cherbourg	71.2833	Not Survived
3	Heikkinen, Miss. Laina	Female	26.0	0	0	STON/O2. 3101282	Lower	Southampton	7.9250	Not Survived
4	Futrelle, Mrs. Jacques Heath (Lily May Peel)	Female	35.0	0	1	113803	Upper	Southampton	53.1000	Not Survived
5	Allen, Mr. William Henry	Male	35.0	0	0	373450	Lower	Southampton	8.0500	Survived

```
[14] data.info()

<class 'pandas.core.frame.DataFrame'>
Index: 1046 entries, 1 to 1307
Data columns (total 10 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   Name         1046 non-null   object
1   Sex          1046 non-null   category
2   Age          1046 non-null   float64
3   Parch        1046 non-null   category
4   SibSp        1046 non-null   category
5   Ticket       1046 non-null   object
6   Pclass       1046 non-null   category
7   Embarked     1044 non-null   category
8   Fare         1045 non-null   float64
9   Survived     1046 non-null   category
dtypes: category(6), float64(2), object(2)
memory usage: 48.2+ KB
```

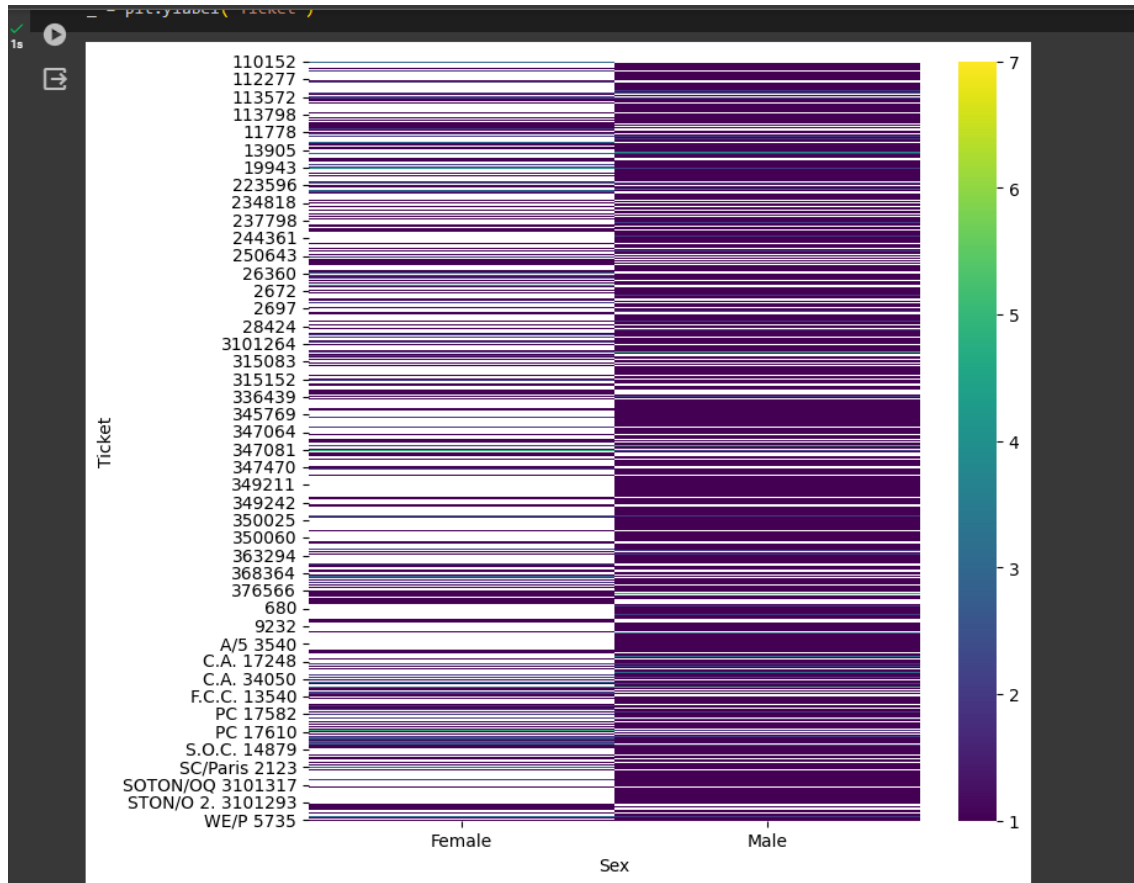
➤ Data Exploration :-



```

[20] plt.subplots(figsize=(8, 8))
      df_2dhist = pd.DataFrame({
          x_label: grp['Ticket'].value_counts()
          for x_label, grp in data.groupby('Sex')
      })
      sns.heatmap(df_2dhist, cmap='viridis')
      plt.xlabel('Sex')
      _ = plt.ylabel('Ticket')

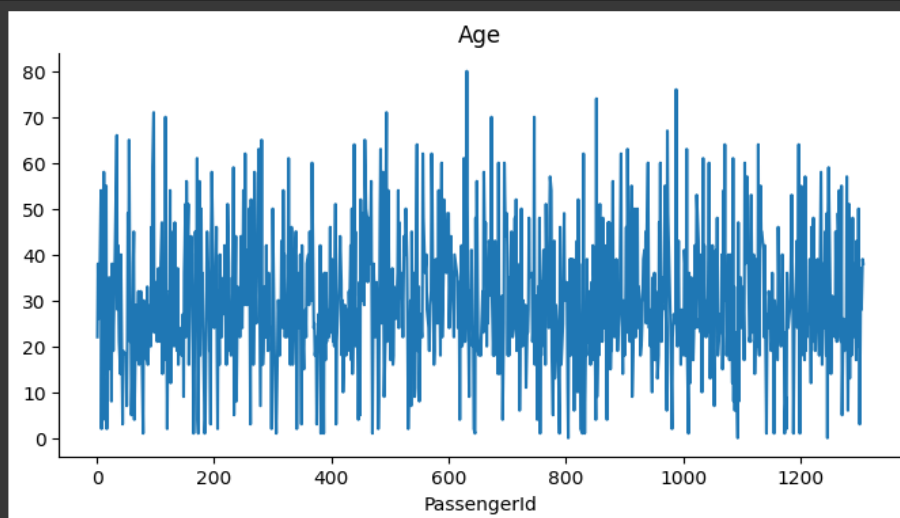
```

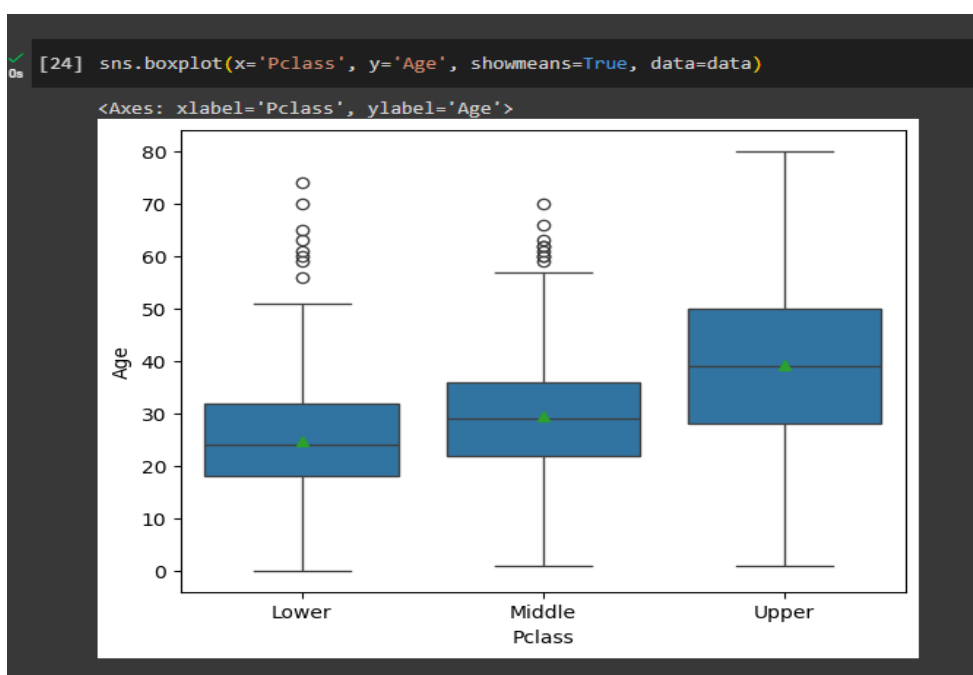
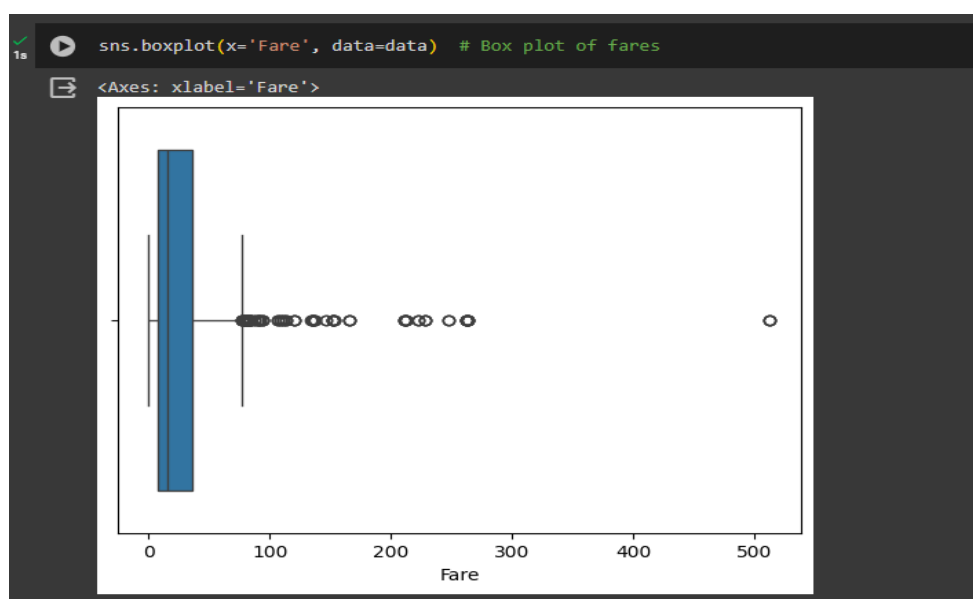
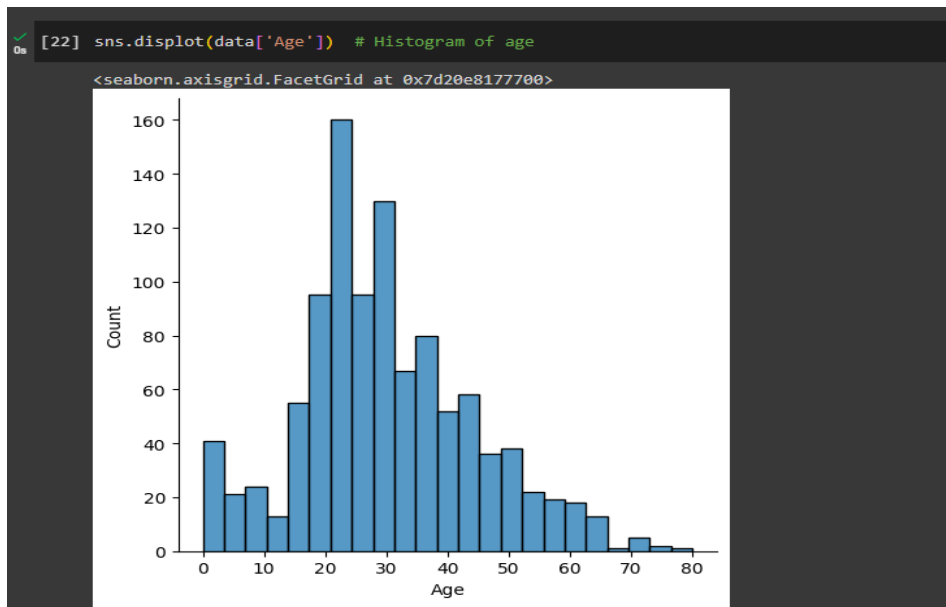


```

[21] data['Age'].plot(kind='line', figsize=(8, 4), title='Age')
      plt.gca().spines[['top', 'right']].set_visible(False)

```





➤ Data Visualization (Using Power BI) :

You can see the analysis report using this link : [Power BI Visualization \(Titanic\)](#)

