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[1]: !pip install pandas numpy matplotlib seaborn scipy statsmodels
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Requirement already satisfied: pandas in c:\users\dhharshini k s\anaconda3\lib\site-packages (2.2.3)  
Requirement already satisfied: numpy in c:\users\dhharshini k s\anaconda3\lib\site-packages (1.24.3)
```

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

```
[6]: df=pd.read_csv('C:/Users/Dharshini K S/Downloads/titanic.csv')
```

```
[8]: # Show the first few rows of the DataFrame  
df.head()  
  
# Get information about the data types and missing values  
df.info()  
  
# Get statistical summary of numerical columns  
df.describe()  
  
# Check for unique values in categorical columns  
df['Sex'].value_counts()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 891 entries, 0 to 890  
Data columns (total 11 columns):  
#   Column      Non-Null Count  Dtype  
---  ---  
0   PassengerId  891 non-null    int64  
1   Survived     891 non-null    int64  
2   Pclass       891 non-null    int64  
3   Name         891 non-null    object  
4   Sex          891 non-null    object  
5   Age          891 non-null    float64  
6   SibSp        891 non-null    int64  
7   Parch        891 non-null    int64  
8   Ticket       891 non-null    object  
9   Fare         891 non-null    float64  
10  Embarked     889 non-null    object  
dtypes: float64(2), int64(5), object(4)  
memory usage: 76.7+ KB
```

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[8]: Sex  
male      577  
female    314  
Name: count, dtype: int64
```

```
[9]: # Plot a histogram for a numerical column  
sns.histplot(data=df, x='Age', kde=True)  
  
# Plot a countplot for a categorical column  
sns.countplot(data=df, x='Sex')  
  
# Plot a boxplot to check for outliers in a numerical column  
sns.boxplot(data=df, y='Fare')
```

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[9]: <Axes: xlabel='Age', ylabel='Count'>
```

```
[10]: # Create a heatmap to visualize correlations between numerical variables  
corr_matrix = df.corr(numeric_only=True)  
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm')  
  
# Create a pairplot for a quick overview of relationships  
sns.pairplot(df[['Age', 'Fare', 'Survived']])  
  
# Use a violinplot to compare a numerical variable across categories  
sns.violinplot(x='Sex', y='Age', data=df)
```

```
[10]: <Axes: xlabel='Survived', ylabel='Count'>
```

```
[11]: # Example: Scatter plot of Age vs. Fare, colored by survival  
sns.scatterplot(x='Age', y='Fare', hue='Survived', data=df)
```

```
[11]: <Axes: xlabel='Survived', ylabel='Count'>
```

This histogram visualizes the distribution of ages among Titanic passengers. We can see that the majority of passengers were young adults, mainly between 20 and 40 years old. There's also a smaller peak for very young children. The majority of passengers did not survive. Female passengers and children had a higher survival rate. Passengers in first class had a much higher chance of survival than those in third class.