Titanic EDA

June 11, 2025

```
[11]: import pandas as pd
      df = pd.read_csv("titanic.csv")
[15]: df.shape
      df.info()
      df.describe(include='all')
      df.isnull().sum()
      df['Sex'].value_counts()
      df['Embarked'].value_counts()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 891 entries, 0 to 890
     Data columns (total 12 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
                       714 non-null
                                        float64
          Age
      6
          SibSp
                        891 non-null
                                        int64
      7
          Parch
                        891 non-null
                                        int64
          Ticket
                       891 non-null
                                        object
          Fare
                        891 non-null
                                        float64
      10 Cabin
                        204 non-null
                                        object
      11 Embarked
                        889 non-null
                                        object
     dtypes: float64(2), int64(5), object(5)
     memory usage: 83.7+ KB
[15]: Embarked
      S
           644
      C
           168
            77
      Q
      Name: count, dtype: int64
```

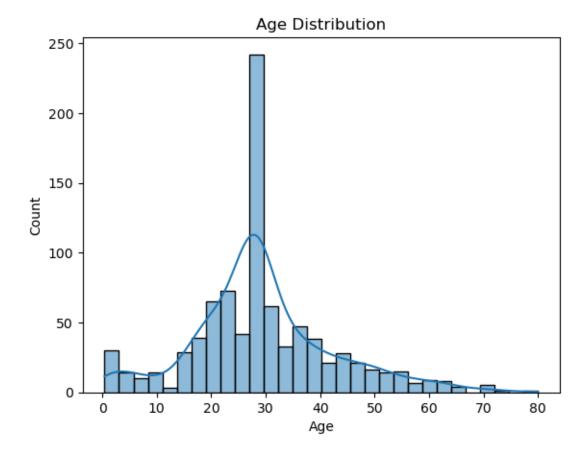
```
[17]: df['Age'].fillna(df['Age'].median(), inplace=True)
    df['Embarked'].fillna(df['Embarked'].mode()[0], inplace=True)
    df.drop('Cabin', axis=1, inplace=True)

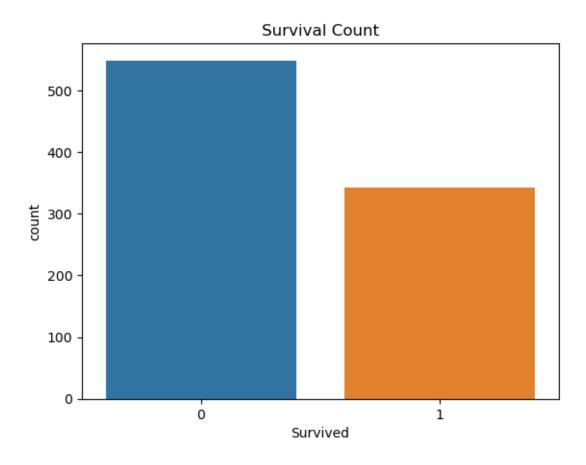
df.drop_duplicates(inplace=True)
```

```
[19]: import matplotlib.pyplot as plt
import seaborn as sns

sns.histplot(df['Age'], kde=True)
plt.title('Age Distribution')
plt.show()

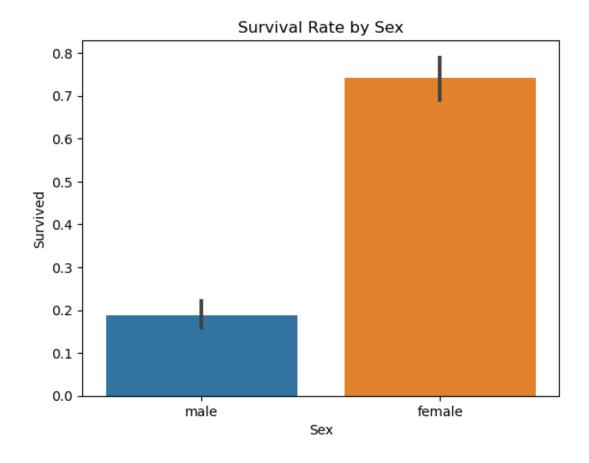
sns.countplot(x='Survived', data=df)
plt.title('Survival Count')
plt.show()
```

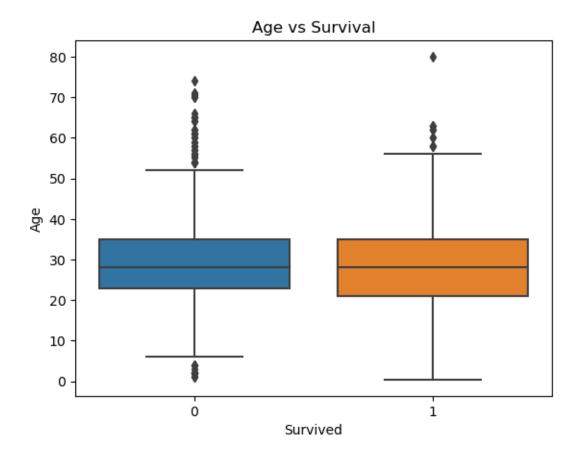




```
[21]: sns.barplot(x='Sex', y='Survived', data=df)
plt.title('Survival Rate by Sex')
plt.show()

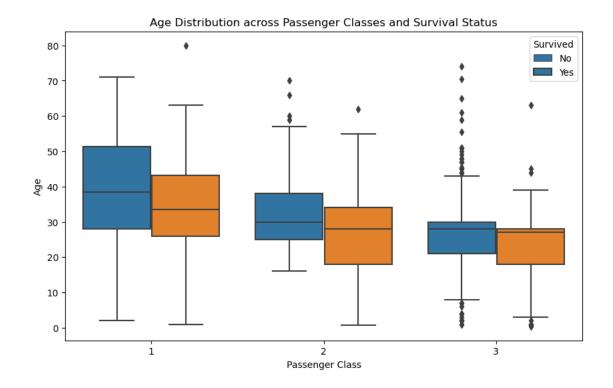
sns.boxplot(x='Survived', y='Age', data=df)
plt.title('Age vs Survival')
plt.show()
```



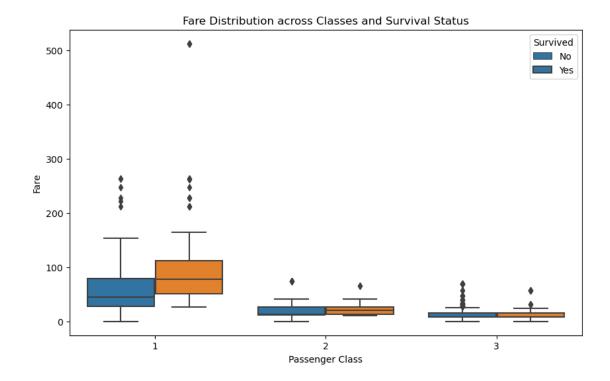


```
[31]: import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(10, 6))
    sns.boxplot(x='Pclass', y='Age', hue='Survived', data=df)
    plt.title('Age Distribution across Passenger Classes and Survival Status')
    plt.xlabel('Passenger Class')
    plt.ylabel('Age')
    plt.legend(title='Survived', labels=['No', 'Yes'])
    plt.show()
```



```
[33]: plt.figure(figsize=(10, 6))
    sns.boxplot(x='Pclass', y='Fare', hue='Survived', data=df)
    plt.title('Fare Distribution across Classes and Survival Status')
    plt.xlabel('Passenger Class')
    plt.ylabel('Fare')
    plt.legend(title='Survived', labels=['No', 'Yes'])
    plt.show()
```



[]: ## Summary of Insights:

- Females had a significantly higher survival rate than males.
- Passengers in 1st class had higher chances of survival.
- Younger passengers had slightly higher survival chances.
- Fare and class were positively correlated.
- Age was not strongly correlated with survival but showed some patterns.