## 20250410 01

April 10, 2025

## 1 Titanic Dataset: Mini Analysis Report

In this mini project, we analyze a simplified Titanic dataset to explore potential relationships between fare, passenger class and survival status. Visualization techniques are applied to illustrate data distributions and group differences.

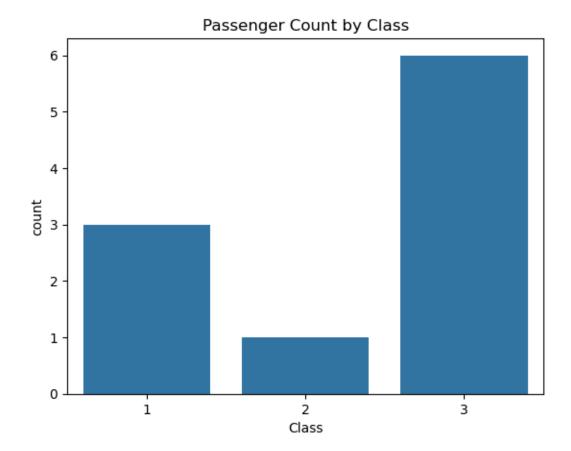
```
[5]: import pandas as pd
     df = pd.read_csv('cleaned_titanic.csv')
     df.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 10 entries, 0 to 9
    Data columns (total 7 columns):
                   Non-Null Count Dtype
         Column
         ----
                   _____
                                   ____
     0
         ID
                   10 non-null
                                   int64
         Survived 10 non-null
                                   int64
     1
     2
         Class
                   10 non-null
                                   int64
     3
         Name
                   10 non-null
                                   object
     4
         Gender
                   10 non-null
                                   object
     5
         Age
                   10 non-null
                                   float64
         Fare
                   10 non-null
                                   float64
    dtypes: float64(2), int64(3), object(2)
    memory usage: 692.0+ bytes
```

The dataset contains 10 records with no missing values.

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

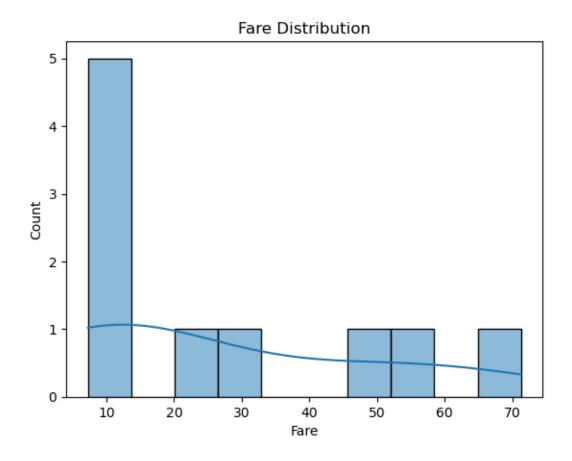
sns.countplot(x = 'Class', data = df)
plt.title('Passenger Count by Class')
plt.show
```

[7]: <function matplotlib.pyplot.show(close=None, block=None)>



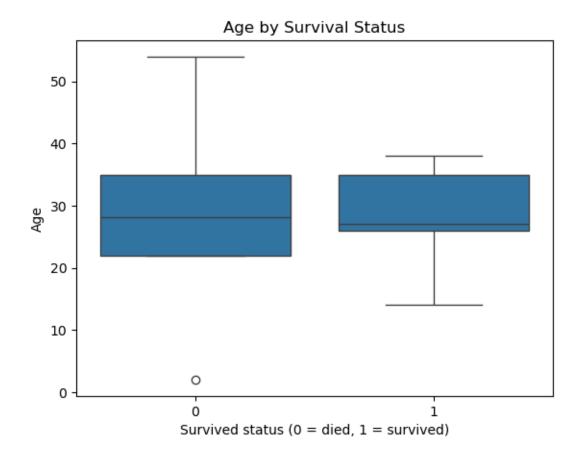
Most passengers were in class 3, suggesting tickets with lower price were sold most commonly.

```
[9]: sns.histplot(df['Fare'], bins = 10, kde = True)
plt.title('Fare Distribution')
plt.show()
```



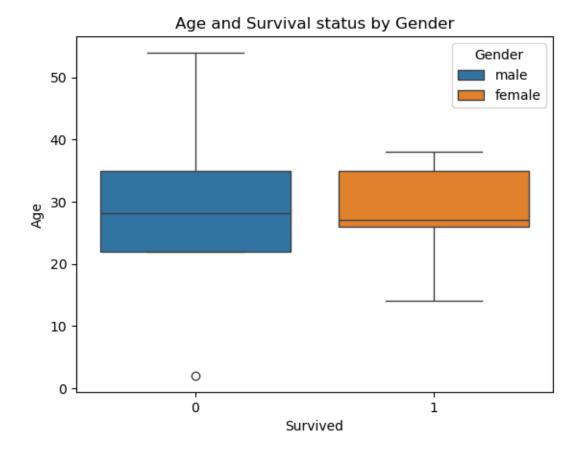
Fare distribution is right-skewed, with most tickets priced around 10 dollars. A few tickets were sold in higher price, which create visible outliers.

```
[21]: sns.boxplot(x = 'Survived', y = 'Age', data = df)
   plt.xlabel('Survived status (0 = died, 1 = survived)')
   plt.title('Age by Survival Status')
   plt.show()
```



The median ages of survivors and non-survivors are similar. However, the most extreme ages (both youngest and oldest) belong to non-survivors, indicating greater age deviance in said group.

```
[23]: sns.boxplot(x = 'Survived', y = 'Age', hue = 'Gender', data = df)
plt.title('Age and Survival status by Gender')
plt.show()
```



The chart shows that all survivors were female and all non-survivors were male in this sample.

## 1.0.1 Conclusion:

- Most Passengers in the dataset were from class 3 and the fare distribution is uneven, with the majority paying lower prices.
- There is no clear linear relationship between age and survival status, though non-survivors exhibit greater variance in age.
- Gender appears to be the strongest factor in survival. All female passengers survived, while all male passengers did not, in this sample.

[]: