

# Task 5 – Exploratory Data Analysis (EDA) on Titanic Dataset

## 1 Objective

Perform Exploratory Data Analysis (EDA) on the Titanic Dataset to discover patterns and relationships that affected passenger survival.

The aim is to demonstrate practical skills in data exploration, cleaning, and visual interpretation using Python.

## 2 Tools & Libraries Used

Python 3	- Programming language
Pandas	- Data loading & manipulation
NumPy	- Numeric operations
Matplotlib	- Basic visualizations
Seaborn	- Statistical plots & heatmaps
Google Colab / Jupyter Notebook -	Interactive coding environment

## 3 Dataset Overview

**Source:** Kaggle – Titanic Dataset

**File used:** train.csv

**Rows:** 891      **Columns:** 12

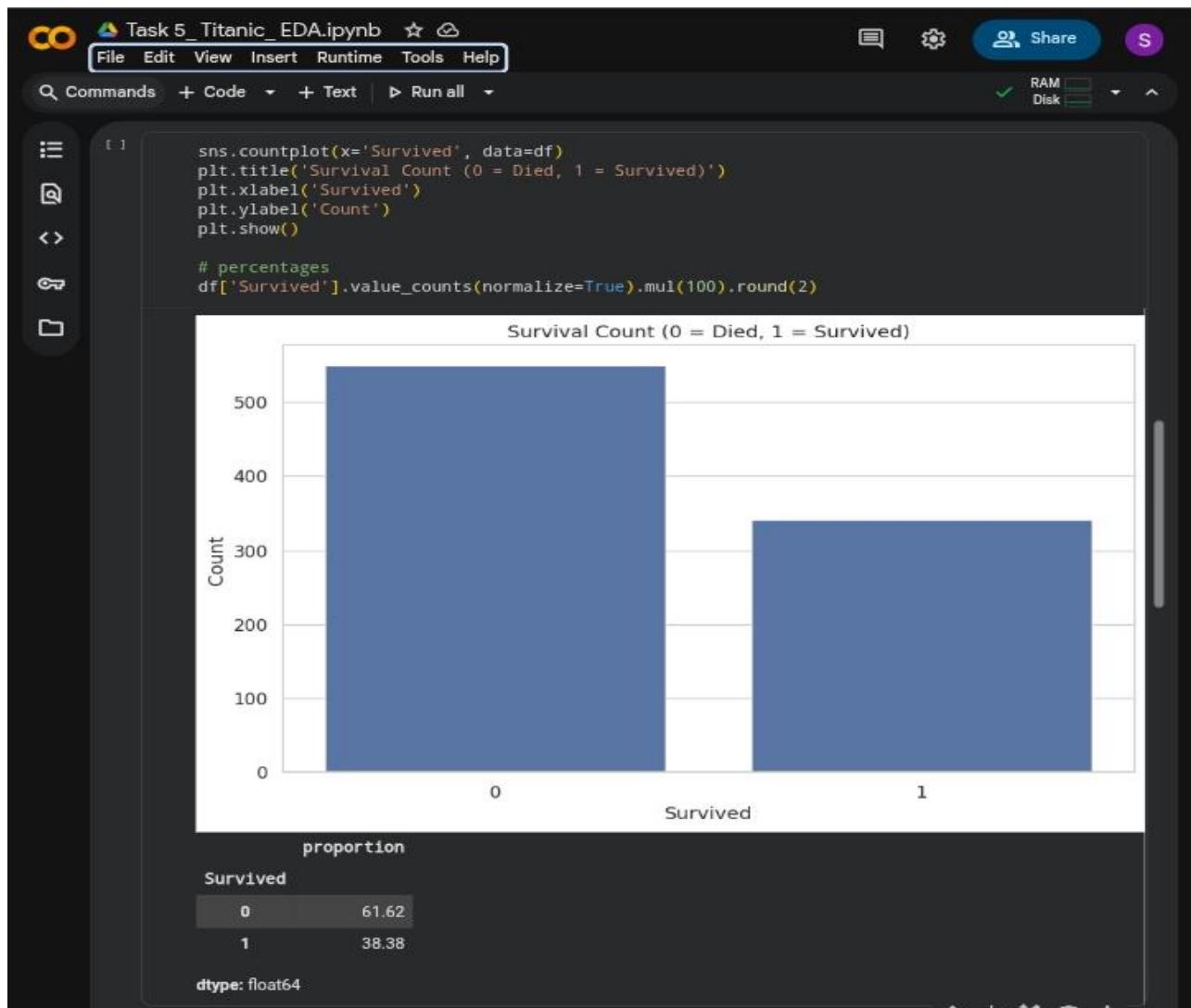
**Main Columns:**

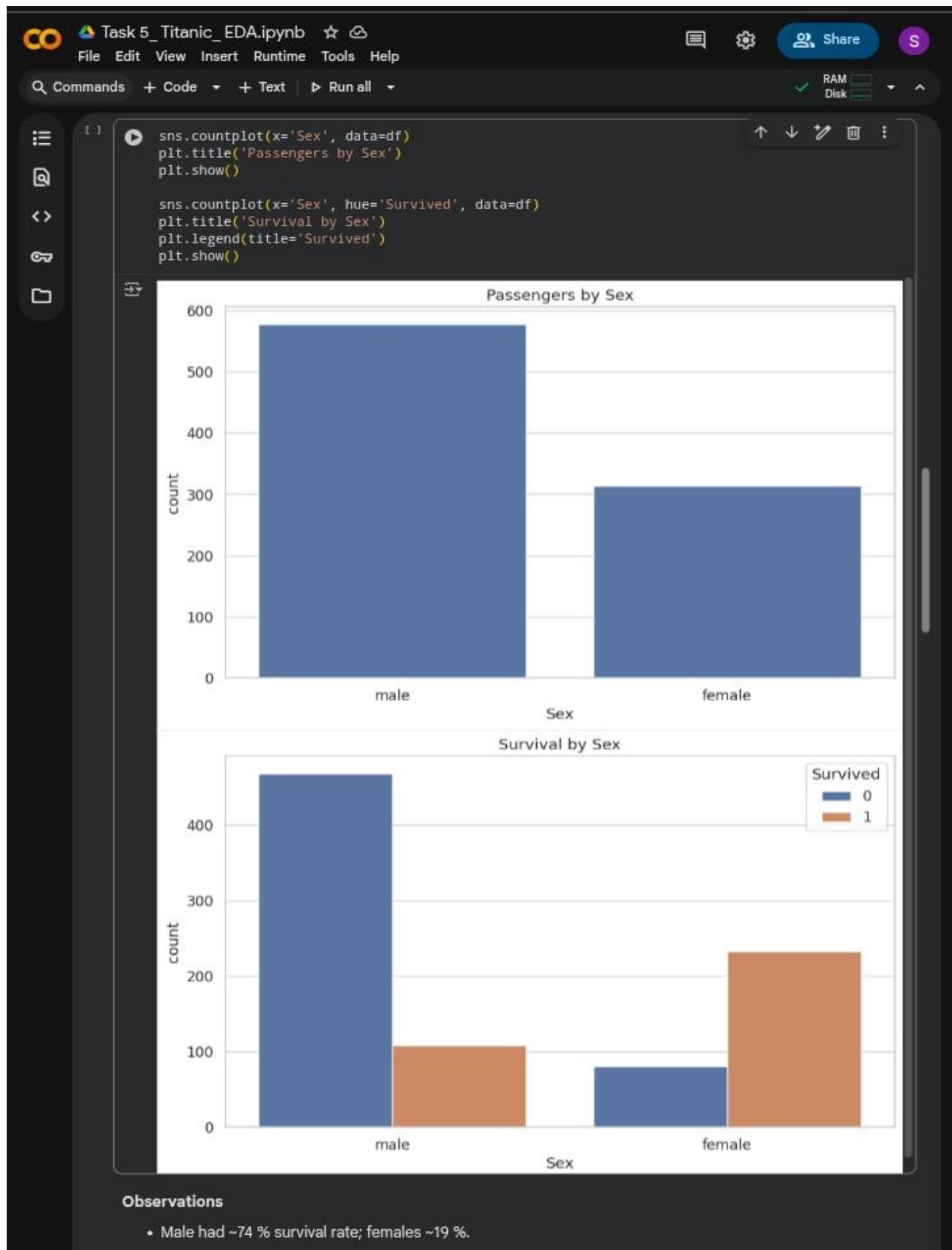
Survived → Target (1 = Survived, 0 = Did not)

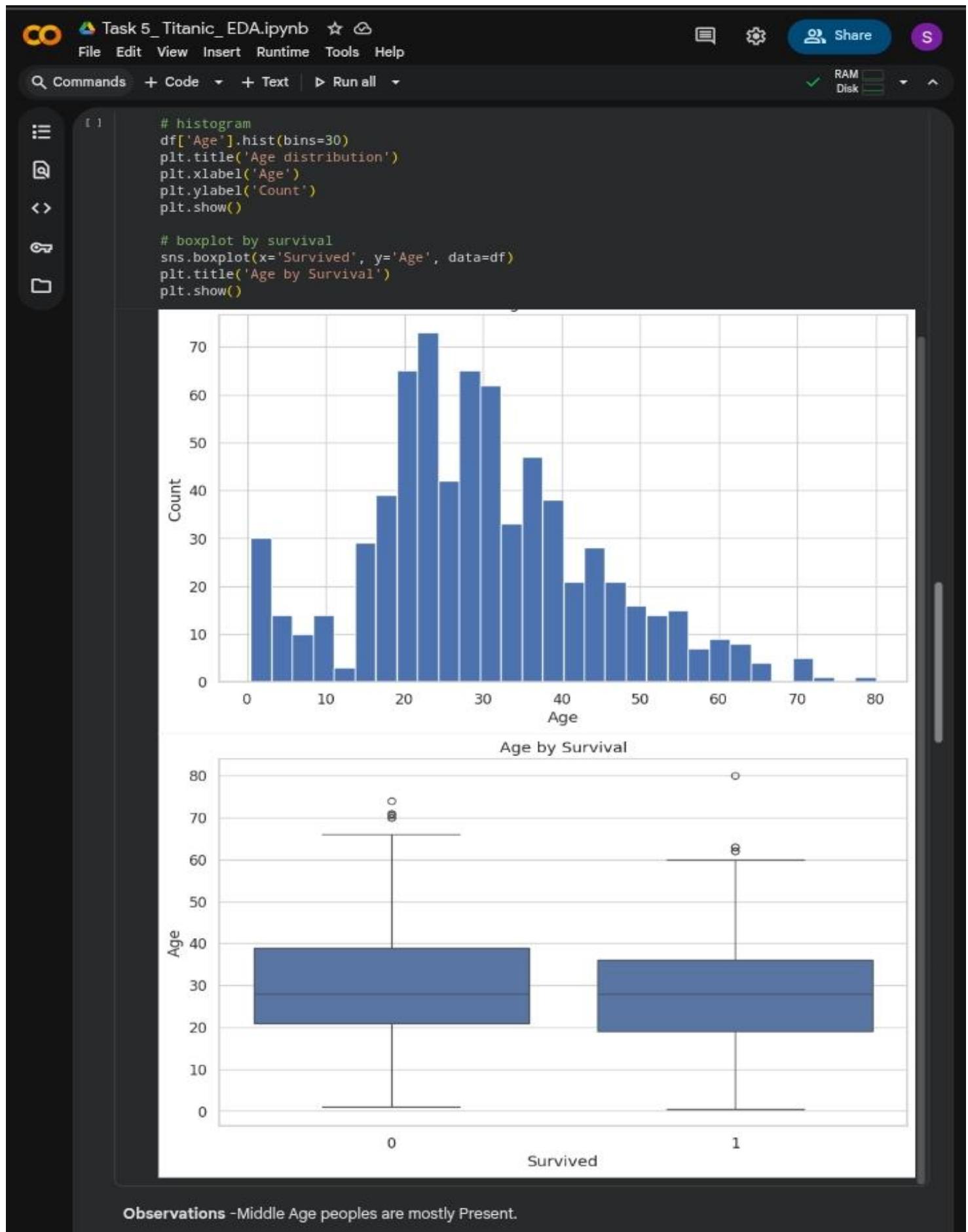
Pclass → Ticket class (1 = Upper, 2 = Middle, 3 = Lower)

Sex, Age, SibSp, Parch, Fare, Embarked, Cabin

## 4 Visual Highlights





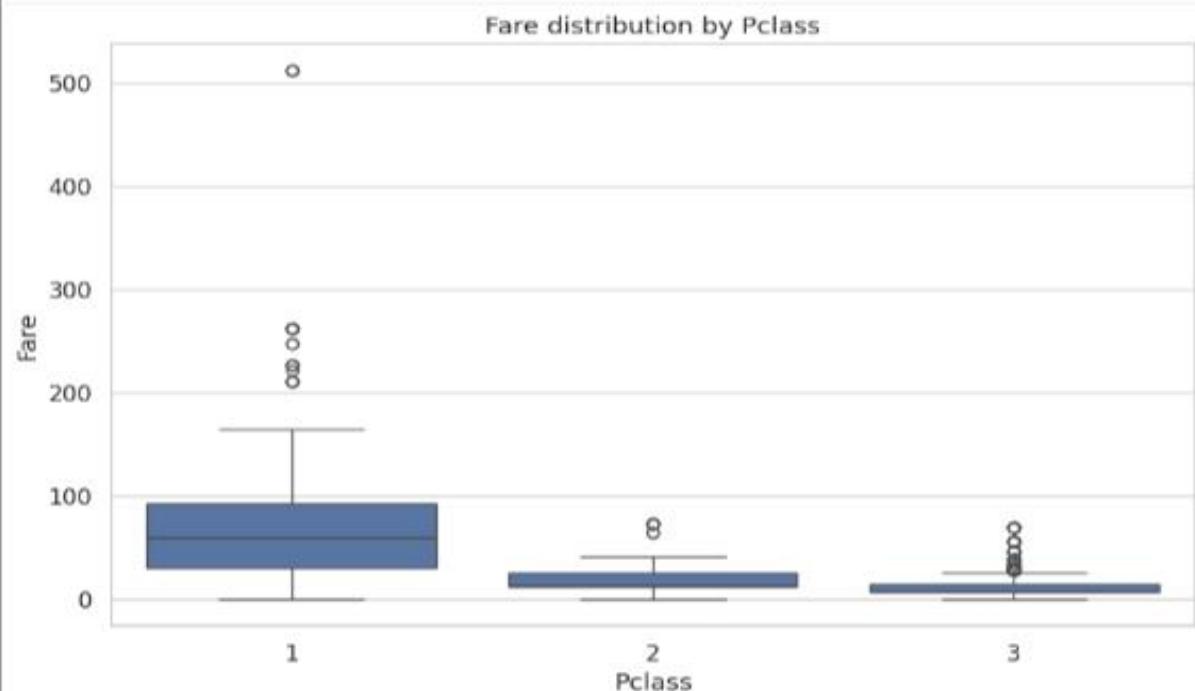
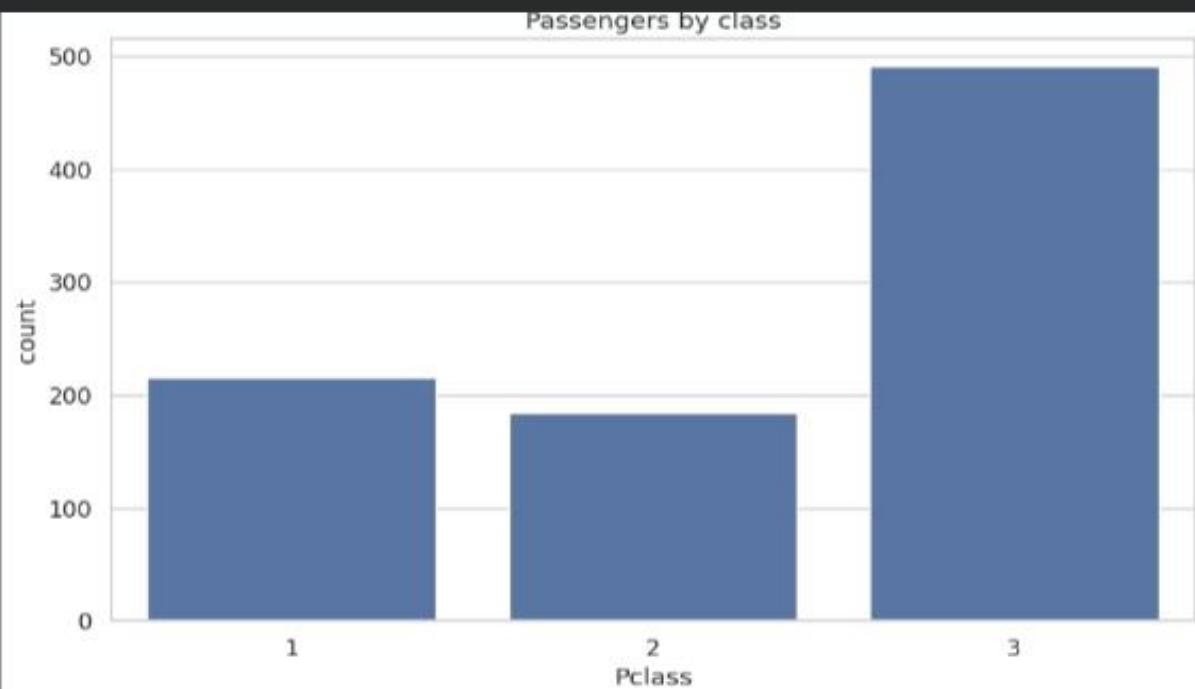




Commands

```
sns.countplot(x='Pclass', data=df)
plt.title('Passengers by class')
plt.show()

sns.boxplot(x='Pclass', y='Fare', data=df)
plt.title('Fare distribution by Pclass')
plt.show()
```



Observations -No of Passengers count in first class is More than the third class.



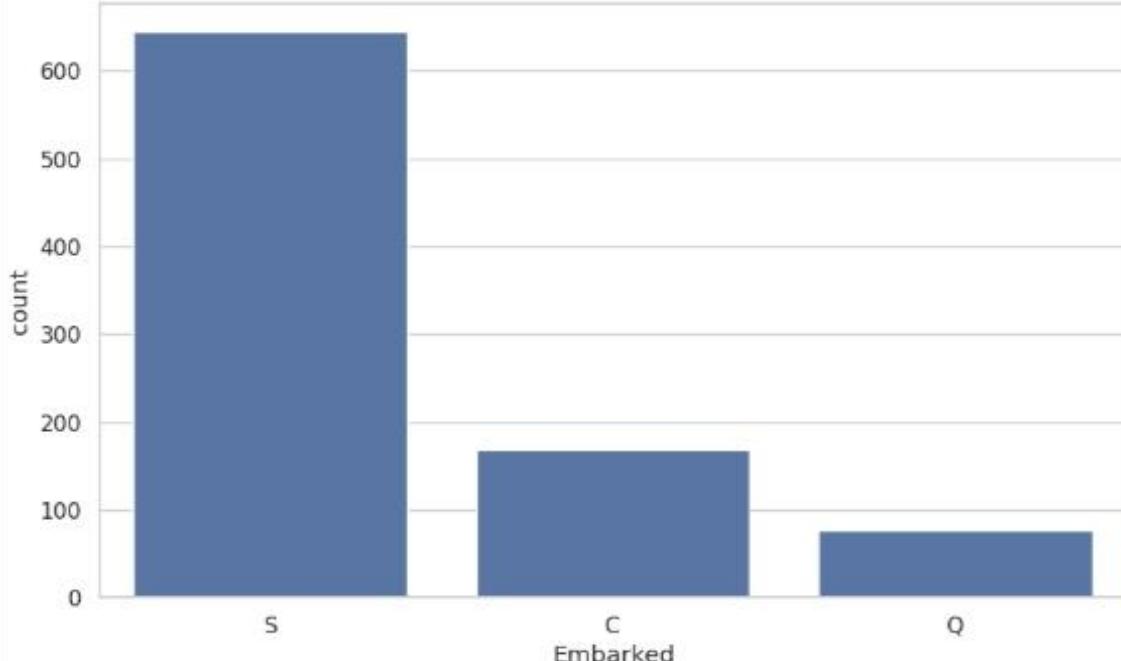
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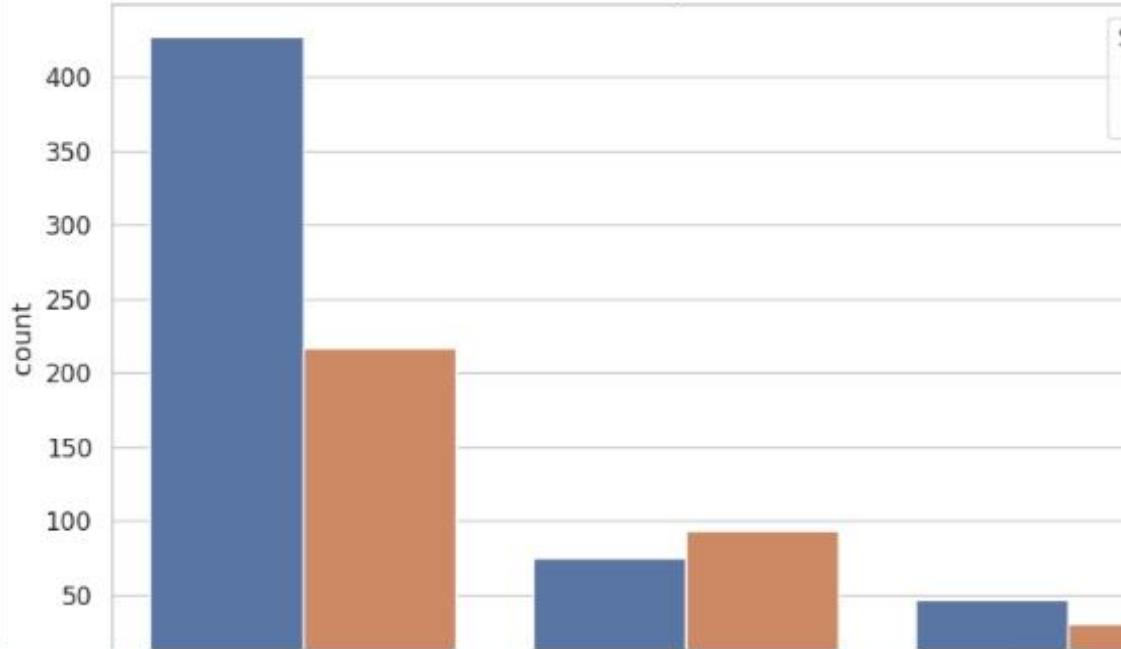
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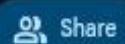
```
sns.countplot(x='Embarked', data=df)
plt.title('Passengers by Embarkation point')
plt.show()
sns.countplot(x='Embarked', hue='Survived', data=df)
plt.title('Survival by Embarked')
plt.show()
```

Passengers by Embarkation point



Survival by Embarked





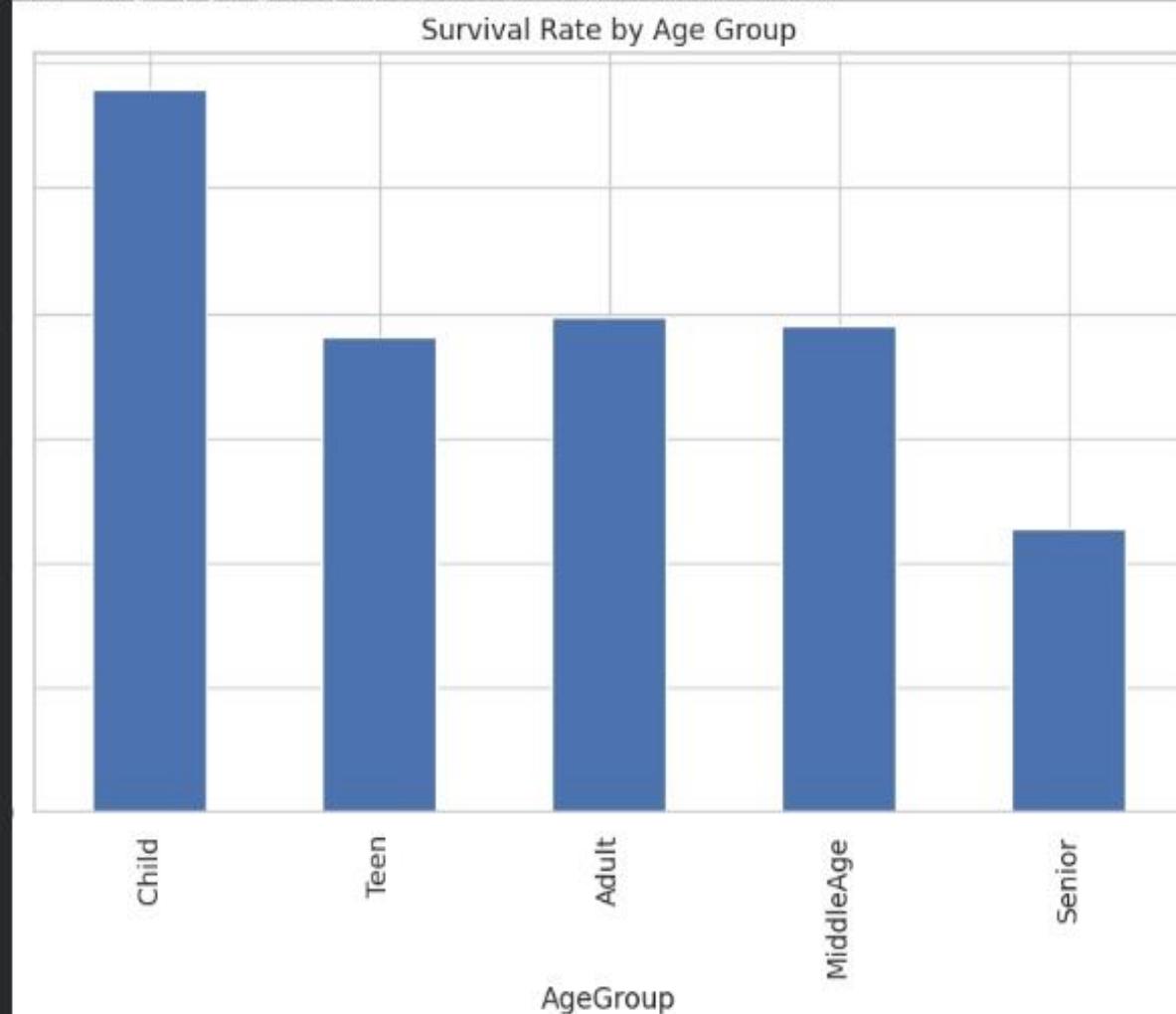
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RAM Disk

```
# create age bins
bins = [0,12,20,40,60,200]
labels = ['Child','Teen','Adult','MiddleAge','Senior']
df['AgeGroup'] = pd.cut(df['Age'], bins=bins, labels=labels)

# survival rate by age group
age_surv = df.groupby('AgeGroup')['Survived'].mean().mul(100).round(2)
age_surv
age_surv.plot(kind='bar')
plt.ylabel('Survival Rate (%)')
plt.title('Survival Rate by Age Group')
plt.show()
```

python-input-2794545954.py:7: FutureWarning: The default of observed=False is deprecated and w.  
surv = df.groupby('AgeGroup')['Survived'].mean().mul(100).round(2)



### Observations

- Younger passengers had better survival chances.



Task 5\_Titanic\_EDA.ipynb

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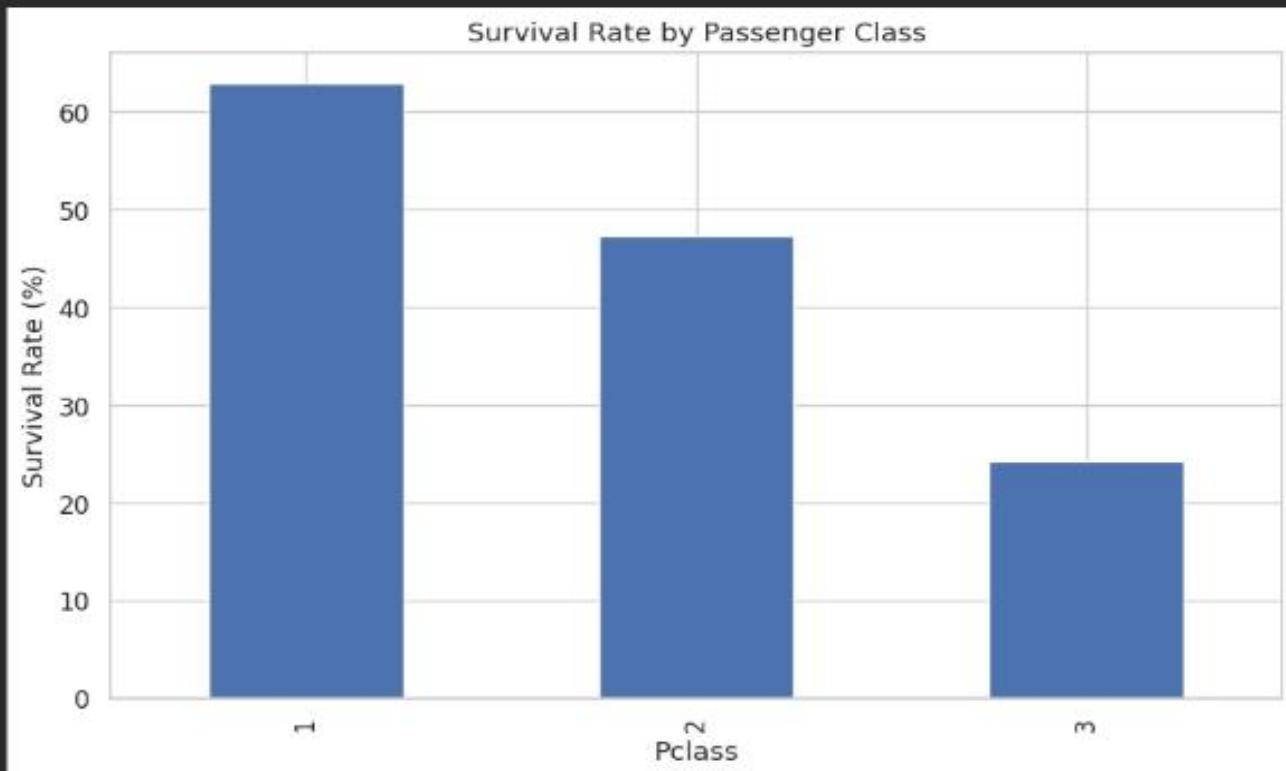


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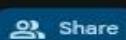


```
[1]: pcclass_surv = df.groupby('Pclass')['Survived'].mean().mul(100).round(2)
pcclass_surv.plot(kind='bar')
plt.ylabel('Survival Rate (%)')
plt.title('Survival Rate by Passenger Class')
plt.show()
```



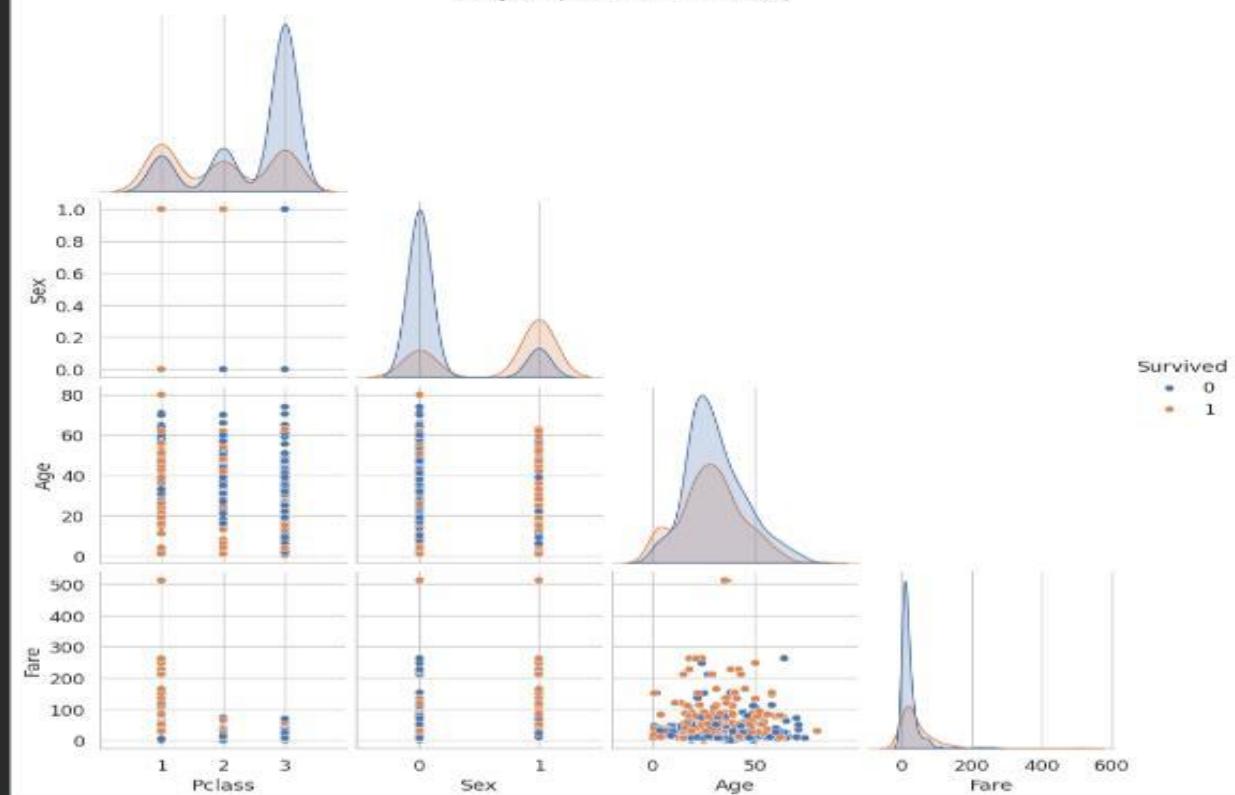
### Observations

- 1st class passengers survived more often than 3rd class.



```
sns.pairplot(corr_df[['Survived', 'Pclass', 'Sex', 'Age', 'Fare']], hue='Survived', corner=True)
plt.suptitle('Pairplot (selected features)', y=1.02)
plt.show()
```

Pairplot (selected features)



```
# family size
df['FamilySize'] = df['SibSp'] + df['Parch'] + 1

# title extraction
df['Title'] = df['Name'].str.extract(r',\s*(\w+\.\w+)', expand=False)
df['Title'] = df['Title'].replace(['Lady', 'Countess', 'Capt', 'Col', 'Don', 'Dr', 'Major', 'Rev', 'Sir'], 'Other')
df['Title'] = df['Title'].replace({'Mlle': 'Miss', 'Ms': 'Miss', 'Mme': 'Mrs'})

df[['Title', 'Survived']].groupby('Title').mean().sort_values(by='Survived', ascending=False)
```

Survived	
Title	Survived
the Countess	1.000000
Mrs	0.793651
Miss	0.702703
Master	0.575000
Rare	0.318182
Mr	0.156673

### Summary of Findings

- Sex and class strongly affect survival.
- Fare correlates positively with survival.
- Missing Cabin data could be replaced by "HasCabin" flag.

## 5 Key Findings & Insights

Women and children had significantly higher survival rates.

1<sup>st</sup>-class passengers survived far more than lower classes.

Higher fare = better survival chance.

Age and family size show moderate influence.

Missing Cabin information reduces data completeness.

## 6 Conclusion

EDA on the Titanic dataset revealed that gender, passenger class, and fare amount were the most important factors influencing survival.

This analysis demonstrates the importance of data visualization and cleaning for discovering patterns before machine-learning modelling.