

# Titanic Dataset - EDA and Feature Engineering

This report contains Exploratory Data Analysis (EDA) and Feature Engineering steps performed on the Titanic Dataset.

It includes Python code, output summaries, and visualizations to understand the factors affecting survival and to prepare the dataset for machine learning model building.

## Python Code Used

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import LabelEncoder

data = pd.read_csv("Titanic-Dataset.csv")

# Data Cleaning
data['Age'].fillna(data['Age'].median(), inplace=True)
data['Embarked'].fillna(data['Embarked'].mode()[0], inplace=True)
data.drop('Cabin', axis=1, inplace=True)

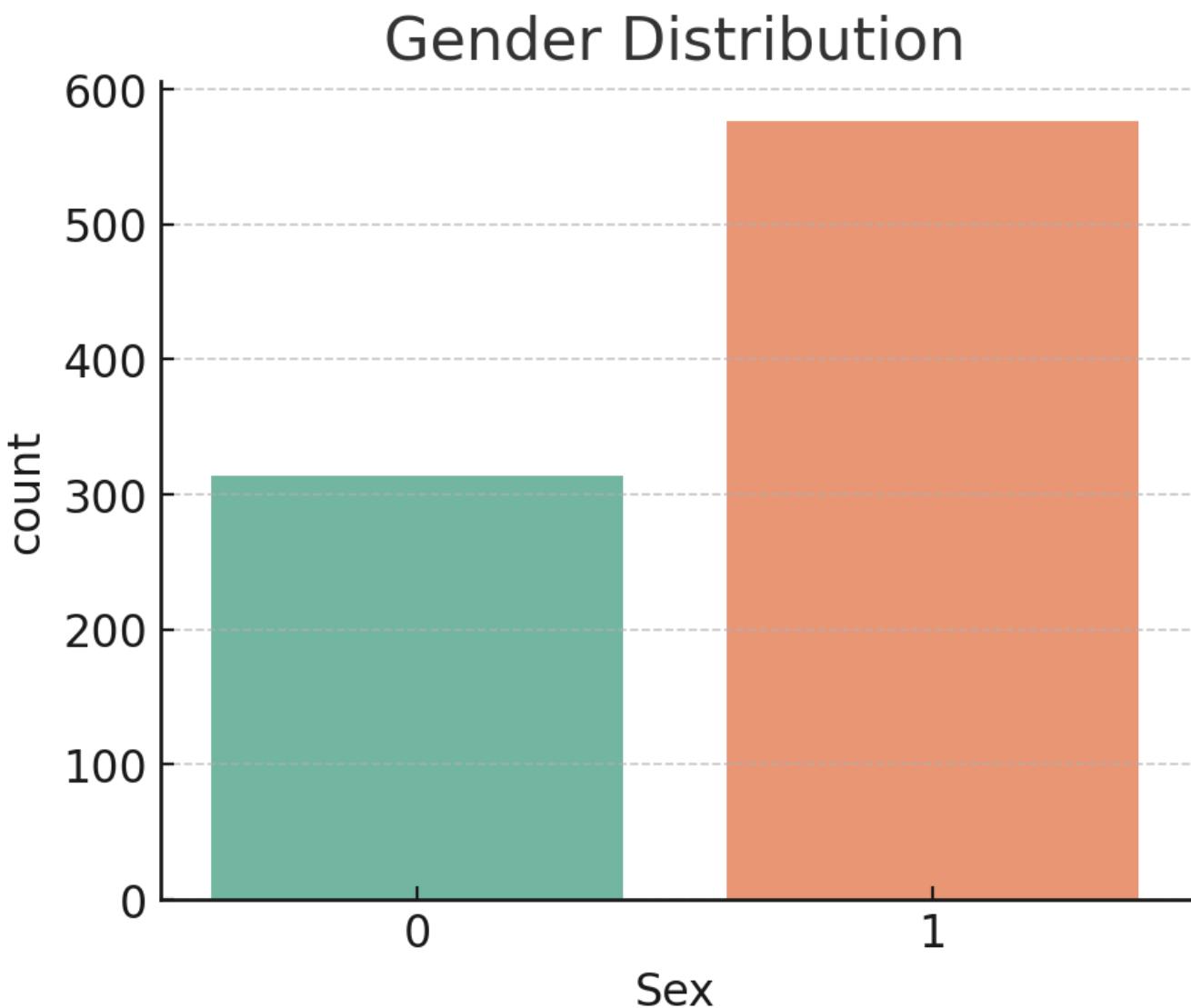
# Feature Engineering
data['FamilySize'] = data['SibSp'] + data['Parch'] + 1
data['IsAlone'] = 1
data.loc[data['FamilySize'] > 1, 'IsAlone'] = 0
data['Title'] = data['Name'].str.extract(' ([A-Za-z]+)\.', expand=False)
data['Title'] = data['Title'].replace(['Lady', 'Countess', 'Capt', 'Col', 'Don', 'Dr', 'Major', 'Rev', 'Sir', 'Jonkheer', 'Dona'], 'Rare')
data['Title'] = data['Title'].replace({'Mlle': 'Miss', 'Ms': 'Miss', 'Mme': 'Mrs'})

# Encoding
label = LabelEncoder()
data['Sex'] = label.fit_transform(data['Sex'])
data['Embarked'] = label.fit_transform(data['Embarked'])
data['Title'] = label.fit_transform(data['Title'])
data.drop(['Name', 'Ticket'], axis=1, inplace=True)
```

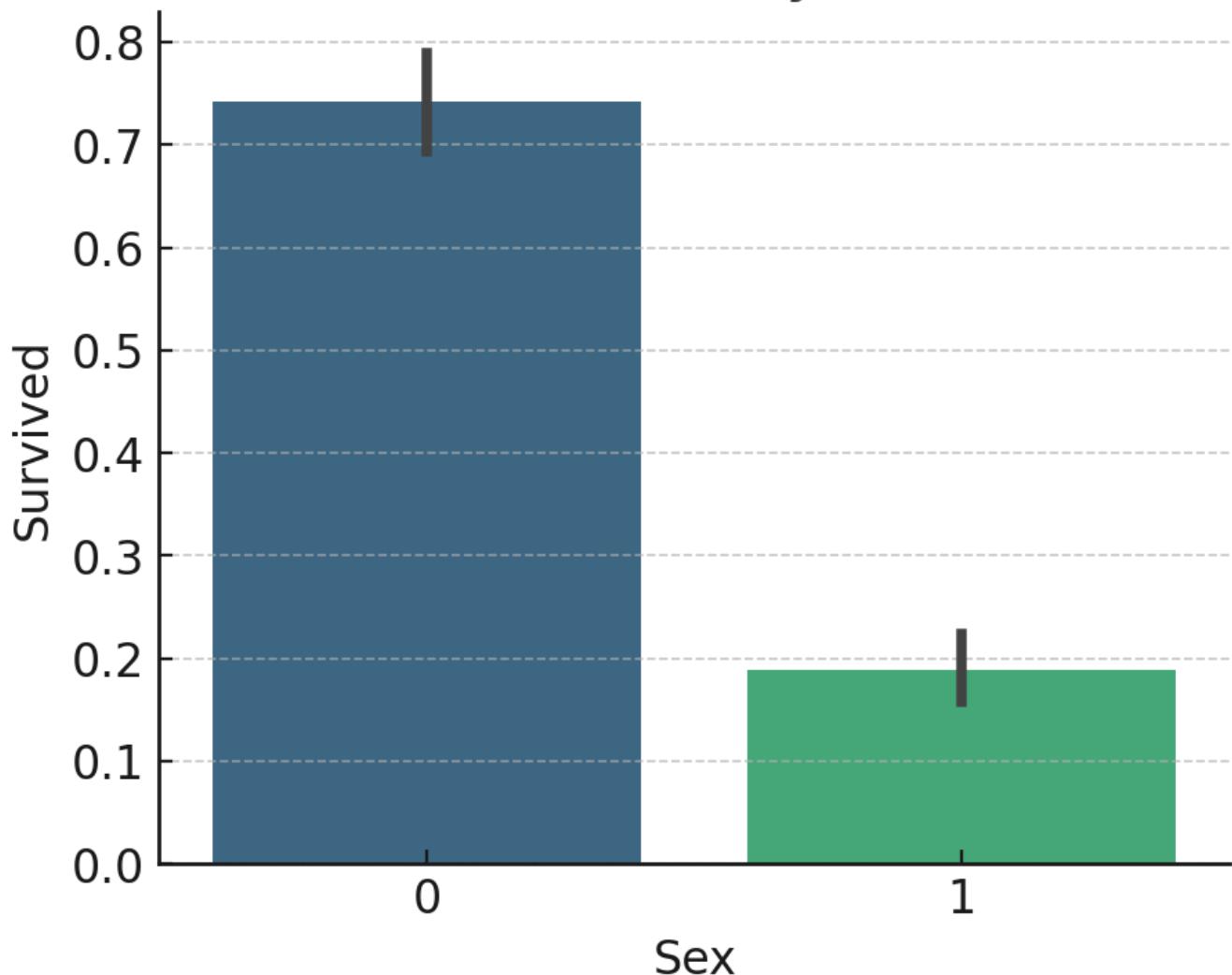
## Output Summary

- Dataset Shape: (891, 12)
- Missing Columns: Age, Cabin, Embarked
- Cleaned: Filled Age and Embarked, Dropped Cabin
- New Columns: FamilySize, IsAlone, Title
- Encoded: Sex, Embarked, Title
- Final Dataset Ready for Model Building

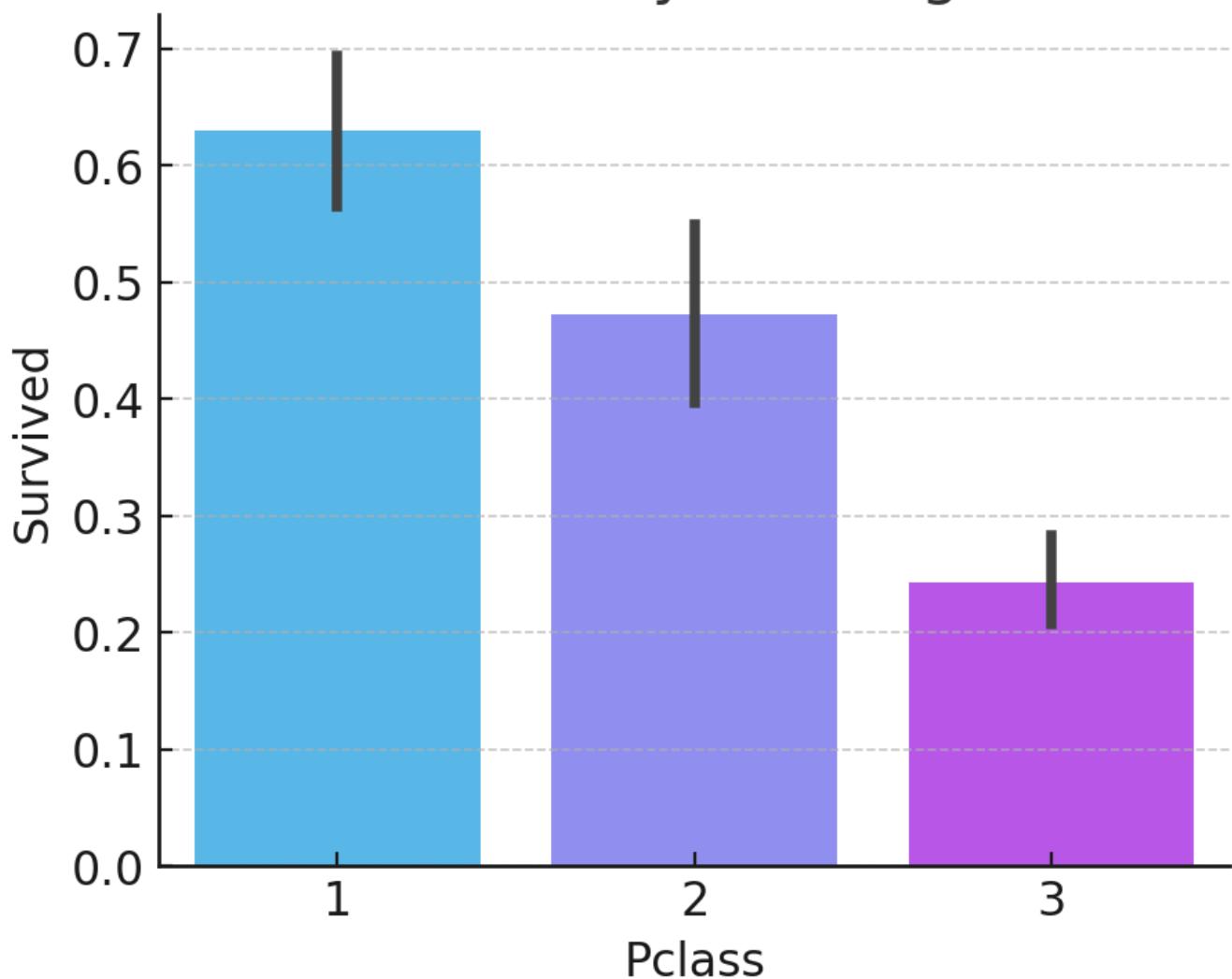
## Visualizations



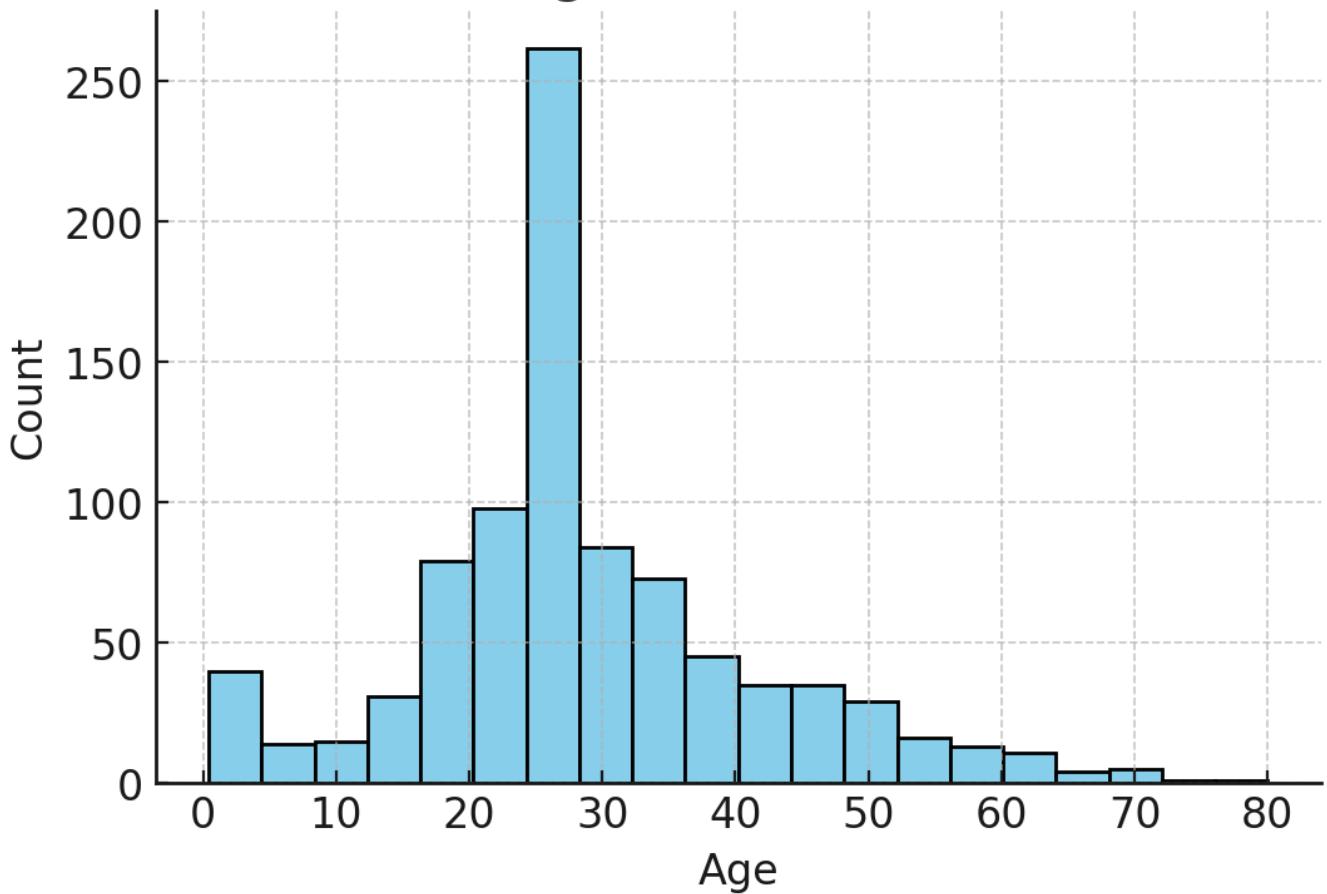
# Survival Rate by Gender



# Survival Rate by Passenger Class



# Age Distribution



Correlation Heatmap

