

## Titanic Dataset – Exploratory Data Analysis (EDA)

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**Tools Used:** Python (Pandas, Matplotlib, Seaborn)

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### Objective

Perform exploratory data analysis on the Titanic dataset to extract patterns, detect anomalies, and uncover relationships using visual and statistical exploration.

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### Dataset Overview

- **Source:** Titanic Dataset (tested.csv)
  - **Rows:** 418
  - **Columns:** 12
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### Step 1: Data Preview

- Viewed top 5 rows
  - Data types, missing values
  - Basic statistics
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### Step 2: Univariate Analysis

#### Gender Distribution

```
sns.countplot(x='sex', data=df)
```

```
plt.title('Gender Distribution')
```

```
plt.show()
```

✅ **Observation:** The dataset contains more male passengers than female passengers.

#### Age Distribution

```
df['age'].hist(bins=30)
```

```
plt.title('Age Distribution')
```

```
plt.xlabel('Age')
```

```
plt.show()
```

✅ **Observation:** Most passengers are between 20–40 years. The distribution is right-skewed.

#### Age Boxplot

```
sns.boxplot(x=df['age'])  
  
plt.title('Age Boxplot')  
  
plt.show()
```

✅ **Observation:** There are several outliers, including infants and elderly passengers.

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### Step 3: Bivariate Analysis

#### Survival by Gender

```
sns.countplot(x='survived', hue='sex', data=df)  
  
plt.title('Survival Count by Gender')  
  
plt.show()
```

✅ **Observation:** Females had a significantly higher survival rate than males.

#### Survival by Passenger Class

```
sns.countplot(x='survived', hue='pclass', data=df)  
  
plt.title('Survival by Class')  
  
plt.show()
```

✅ **Observation:** 1st class passengers had a much better chance of survival than 3rd class passengers.

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### Step 4: Correlation Analysis

#### Correlation Heatmap

```
plt.figure(figsize=(10,6))  
  
sns.heatmap(df.corr(numeric_only=True), annot=True, cmap='coolwarm')  
  
plt.title('Correlation Matrix')  
  
plt.show()
```

✅ **Observation:** Fare and Pclass show moderate correlation. Survival is positively correlated with Fare and negatively with Pclass.

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### Step 5: Pairplot

```
sns.pairplot(df[['age', 'fare', 'pclass', 'survived']], hue='survived')  
  
plt.show()
```

✅ **Observation:** Higher fare and lower Pclass (1st class) passengers show better survival clustering.

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### Summary of Insights

-  **Gender:** Women had much higher survival rates.
-  **Class:** 1st class passengers were more likely to survive.
-  **Fare:** Higher-paying passengers had better survival chances.
-  **Age:** Younger passengers (especially children) were slightly more likely to survive.
-  **Outliers:** Found in Fare and Age.
-  **Missing Values:** Some nulls in Age, Embarked, etc.
-  **Correlations:** Survival weakly correlated with fare and class.

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### Deliverables

| File Name              | Description                    |
|------------------------|--------------------------------|
| Task_5_EDA.ipynb       | Jupyter Notebook with code     |
| titanic_eda_vaheed.pdf | Final EDA Report in PDF format |