

Titanic Dataset – Exploratory Data Analysis (EDA) Report

Objective:

To analyze the Titanic dataset to identify patterns and insights that influenced passenger survival during the Titanic disaster.

Dataset Description:

- **Source:** `train.csv` from the Titanic Machine Learning dataset
 - **Features analyzed:**
 - `Survived` (target)
 - `Age`, `Sex`, `Pclass`, `Fare`, `Embarked`, `SibSp`, `Parch`
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Steps Performed:

1. Data Loading & Inspection

- Loaded using `pandas`
- Initial inspection with:
 - `.info()` for data types & null values
 - `.describe()` for summary statistics
 - `.isnull().sum()` to check missing data

2. Handling Missing Values

- No imputation steps were shown, but missing data was quantified, especially in columns like **Age** and **Cabin**.
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Visual Analysis:

Plot 1: Survival Count

- Used `seaborn.countplot()` on the **Survived** column
- Insight: More people died than survived (class imbalance).

Plot 2: Age Distribution

- Histogram with KDE for **Age**
- Insight: Most passengers were in the 20–40 age group.

Plot 3: Survival by Gender

- Count plot comparing **Sex** vs **Survived**
- Insight: Women had a significantly higher survival rate than men.

Plot 4: Survival by Passenger Class (Pclass)

- Count plot of **Pclass** vs **Survived**
 - Insight: First-class passengers had higher survival rates, indicating class-based disparity.
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Potential Further Steps (not in notebook but recommended):

- Impute missing values (e.g., median age).
 - Correlation heatmaps for numeric features.
 - Feature engineering (e.g., creating **FamilySize**, encoding **Sex** and **Embarked**).
 - Predictive modeling using logistic regression or decision trees.
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Conclusion:

This EDA provided a foundational understanding of factors affecting survival aboard the Titanic. Notably:

- Gender and class were strong indicators of survival.
- Age distribution offers demographic insights into passengers.