EDA Titanic cleaning data process report

1. Title & Objective

Exploratory Data Analysis of Titanic Dataset

Objective: The goal is to explore the dataset to identify survival trends and relationships between passenger attributes and survival probability.

2. Dataset Overview

Source: train.csv (Titanic dataset)

Number of rows: 891

• Number of columns: 12

Key Features:

o PassengerId: Unique passenger identifier

Survived: Survival status (0 = No, 1 = Yes)

Pclass: Ticket class (1 = 1st, 2 = 2nd, 3 = 3rd)

o Name: Passenger's name

o Sex: Gender

Age: Age in years

SibSp: Number of siblings/spouses aboard

Parch: Number of parents/children aboard

Ticket: Ticket number

Fare: Passenger fare

o Cabin: Cabin number

o **Embarked**: Port of embarkation (C = Cherbourg, Q = Queenstown, S = Southampton)

3. Data Cleaning Summary

Missing Values:

o Age: 177 missing values

o Cabin: 687 missing values

Embarked: 2 missing values

- Fixed data types for analysis (Age → float, Fare → float)
- No duplicate rows found

4. Statistical Summary

From .describe() and .value_counts() analysis:

• Age: Mean = 29.7, Min = 0.42, Max = 80

• Fare: Mean = 32.2, Min = 0, Max = 512.3

• **Survival rate**: 38.38% survived (342/891)

• Gender count: 577 males, 314 females

• Class distribution: 1st (216), 2nd (184), 3rd (491)

5. Visual Insights

• **Bar Chart**: Higher survival rate for females than males

• Histogram (Age): Most passengers between 20–40 years old

• **Boxplot (Fare vs Class)**: Higher ticket class → Higher fare

• **Heatmap**: Survival is positively correlated with Fare and Pclass (inverse correlation)

• Pairplot: Clear separation between classes and survival probability

6. Key Findings

- Females had a much higher survival rate than males
- 1st class passengers had the highest survival probability
- Younger passengers, especially children, survived more often
- Higher fares were generally linked to better survival rates
- Port of embarkation "C" passengers had slightly higher survival rates than others

7. Conclusion & Next Steps

- Gender, class, and age are strong predictors of survival
- Missing age and cabin data could be imputed for better modeling
- Next step: Build a predictive model (e.g., logistic regression, decision tree) using these insights