Task 5 — Exploratory Data Analysis (EDA)

1. Introduction

This report demonstrates a complete Exploratory Data Analysis (EDA) workflow using Python. It contains data loading, cleaning, summary statistics, visualizations, and interpretations. A sample synthetic dataset (Titanic-like) is used for demonstration; replace it with your own

2. Dataset Overview

The synthetic dataset used: rows = 200, columns = 6. Columns: survived (0/1), class (First/Second/Third), sex, age, fare, embarked. Replace the data-loading cell in the notebook with: pd.read_csv('yourfile.csv') and rerun the

3. Data Cleaning & Preprocessing

Cleaning steps applied in the notebook:

- Inspect missing values and percent missing per column.
- Imputed 'age' with median for summary/statistics and plotting.
- Left other categorical variables as-is for value_counts; encoding guidance provided.
- Removed negative fares by taking absolute value when generating synthetic data; for real data check for errors.

4. Summary Statistics

Summary statistics (table truncated in PDF; full in notebook):

```
survived class sex
                       age
                              fare embarked
count 200.000000 200 200 180.000000 200.000000
                                             200
        NaN 3 2 NaN
                              NaN
uniaue
                                     3
       NaN Third male
                        NaN
                               NaN
                                       S
top
                               NaN 128
       NaN 95 119
                       NaN
freq
      0.365000 NaN NaN 27.652222 70.992900
std 0.482638 NaN NaN 14.627278 55.235444
     0.000000 NaN NaN -12.600000 1.950000
25%
      0.000000 NaN NaN 18.425000 23.402500
50%
      0.000000 NaN NaN 27.450000 57.575000
75%
      1.000000 NaN NaN 36.625000 102.112500
                                             NaN
```

5ax Visualizations & Key Findings NaN

Visualizations generated (see figures). Interpretations:

- Age Distribution: Shows central tendency and spread; missing ages were imputed with median in analysis.
- Fare Boxplot: Reveals skew and outliers consider log-transform for modeling.
- Survival Rate by Class: Differences indicate class-related survival disparities.
- Age vs Fare Scatter: Checks correlation; in this synthetic data there's no strong linear relation.

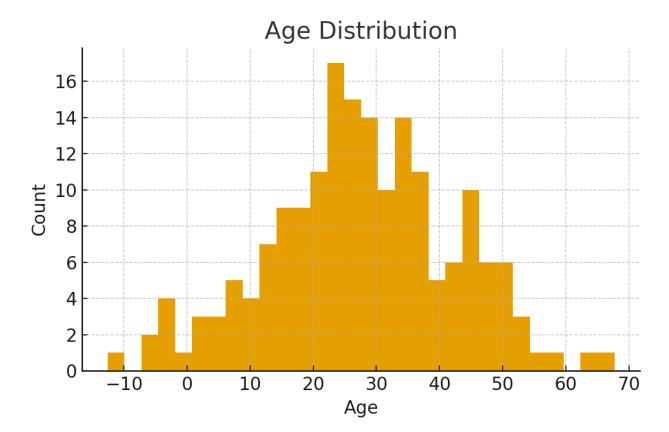


Figure 1 — Age Distribution

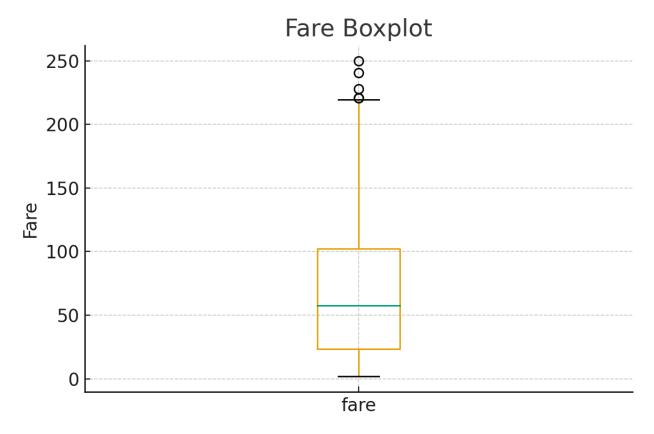


Figure 2 — Fare Boxplot

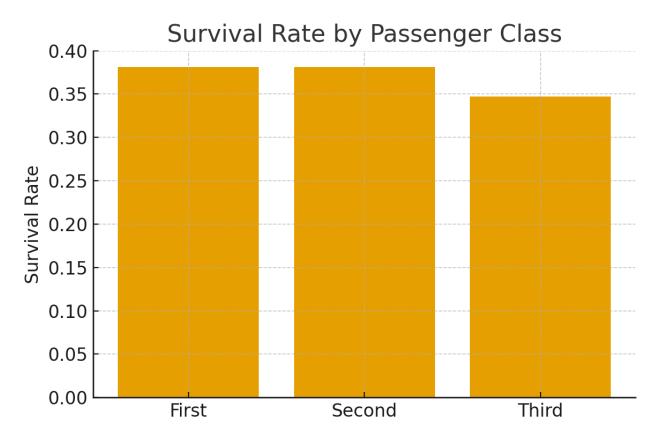


Figure 3 — Survival Rate by Class

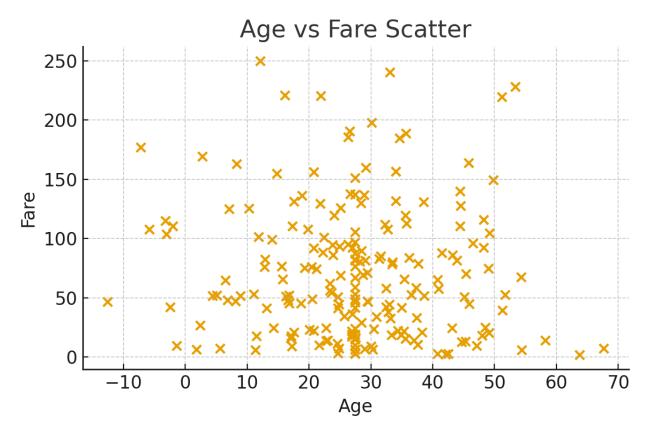


Figure 4 — Age vs Fare

6. Conclusions & Recommendations

Example conclusions from the synthetic dataset:

- Differences in survival rate by class suggest the model should include class as an important predictor.
- Missing age values should be handled carefully; consider models that can handle missingness or use informed imputation.