# Internship Task Report – Titanic Dataset (EDA)

#### 1. Introduction

For this task, I worked on the famous **Titanic dataset**, which is available on Kaggle. The goal of this project was to perform **Data Cleaning** and carry out **Exploratory Data Analysis (EDA)** to uncover patterns and insights hidden within the data.

# 2. Understanding the Dataset

The dataset is part of the Kaggle **Titanic Challenge**. It comes with two main files:

- $train.csv \rightarrow used for analysis (contains the survival information).$
- $test.csv \rightarrow used$  for prediction tasks (not used in this EDA).

Key columns in the dataset:

- Survived: Target column (0 = Not Survived, 1 = Survived)
- **Pclass**: Passenger class (1st, 2nd, 3rd)
- **Sex**: Gender of the passenger
- **Age**: Age of the passenger
- SibSp & Parch: Family members onboard
- Fare: Ticket price
- Embarked: Port of embarkation
- Cabin, Ticket, Name: Contain too much noise or are less useful for EDA

# 3. Data Cleaning Process

Before performing analysis, I cleaned the dataset to handle missing or irrelevant values:

- Age: Filled missing values with the median age.
- Embarked: Filled missing values with the most common port (mode).
- Cabin: Dropped this column because most values were missing.
- **HasCabin**: Added a new column (1 = passenger had a cabin, <math>0 = no cabin).
- Converted some categorical features like **Sex**, **Pclass**, **Embarked** into categorical data types for better analysis.

# 4. Exploratory Data Analysis (EDA)

### (A) Univariate Analysis

- Age: Most passengers were between 20–40 years old.
- Fare: The distribution was right-skewed, showing a few passengers paid very high fares.
- **Pclass**: Most passengers belonged to 3rd class.

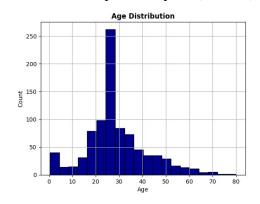
#### (B) Bivariate Analysis (with Survival)

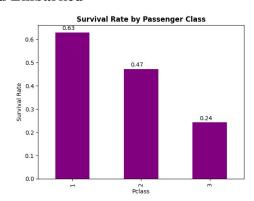
- Sex: Survival rate was much higher for women ( $\sim$ 74%) compared to men ( $\sim$ 19%).
- **Pclass**: 1st class passengers had the highest survival rate ( $\sim$ 63%), followed by 2nd class ( $\sim$ 47%), and 3rd class ( $\sim$ 24%).
- **Embarked**: Passengers who boarded at Cherbourg (Port C) had slightly better chances of survival.
- Age: Children under 12 had a higher survival rate compared to adults.

# 5. Visual Insights

I created multiple charts in Python (using Matplotlib) to visualize the findings:

- Histogram of Age distribution
- Bar charts for survival count
- Survival comparison by Sex, Pclass, and Embarked





# 6. Key Findings

- Women and children had significantly higher survival chances.
- Passengers from **wealthier classes (1st class)** survived more compared to those in 3rd class.
- Passengers boarding from **Port** C had better outcomes than others.
- Traveling with family improved chances, while being alone reduced survival chances.

# 7. Conclusion

The Titanic dataset required initial cleaning to handle missing values and irrelevant features. After data cleaning and EDA, I was able to identify strong patterns: survival was closely related to **gender**, **class**, **and age**.

These insights not only show the social dynamics during the Titanic tragedy but also lay the foundation for building predictive models in the future.