

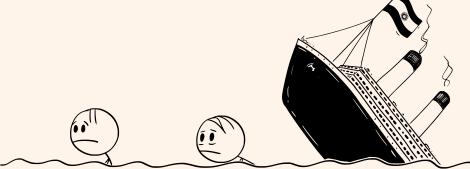
# **Titanic Data** Analysis: Solving Queries and Plotting Graphs

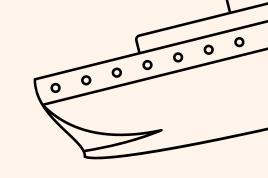




## Abstract

- The Titanic dataset is a popular dataset used for data analysis and machine learning.
- In this project, we will explore the dataset and perform various analyses to gain insights into the passengers on board in Titanic.
- Using Python and its data analysis libraries, we will clean and preprocess the data, visualize the data through plots and charts, and draw conclusions.
- Overall, this project will give us a better understanding of the passengers on board the Titanic and their survival compared with various other factors.



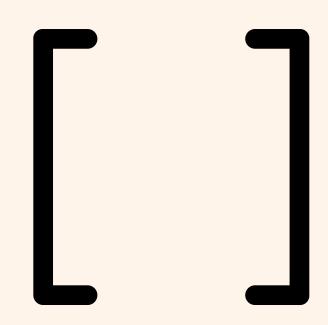


## Introduction

- Titanic dataset is taken, imported it to notebook/python environment using panda's library.
- the imported data set is observed and cleaned/preprocessed to obtain meaningful data.
- This data is used to analyze relations between different columns/ features.
- This is also arrange, pick and solve user queries.
- Different Python modules like matplotlib, seaborn, scikitlearn along with numpy and pandas are used.



#### Libraries Used:



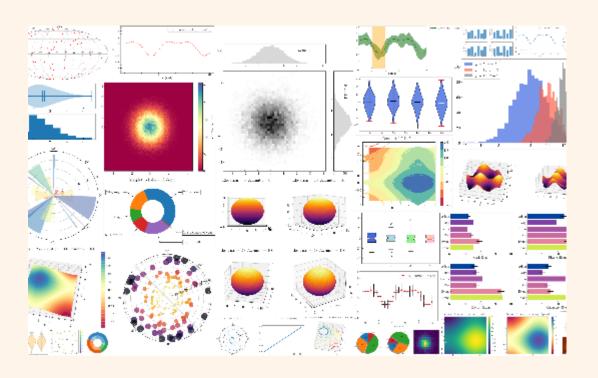
#### **Pandas**

A Python library for data manipulation and analysis. Used for reading and Doing operations easily on Dataset.



#### ScikitLearn

More Advanced Version of
Matplotlib. plots relations
between features easily, auto
scales the data.



#### Matplotlib

A Python library for creating visualizations. We will use it to plot data and explore the dataset.



## Queries

- Distributions of features
- Query to Replace something in data
- Searching a record
- Features vs Features (plots)
- which ages survied most?
- Records of people of age greater than given input
- Youngest, Oldest, Max, Min
- Total and mean
- Age Distributions?
- Names starting with given input
- Correlation matrix

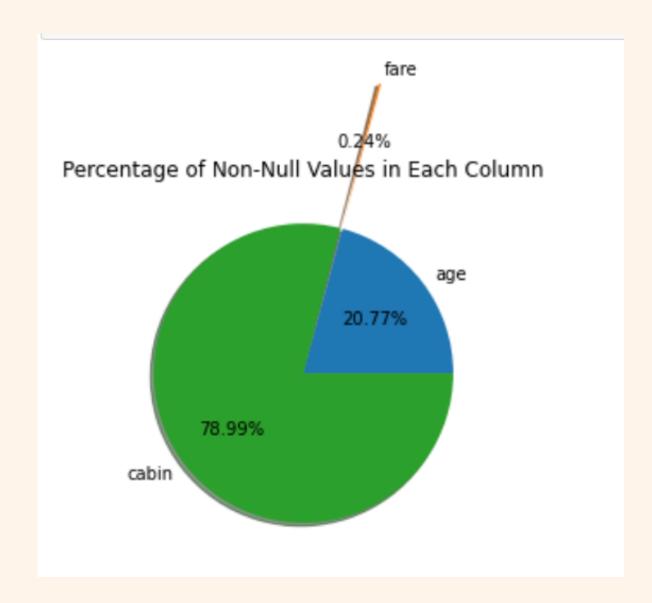
#### **DataSet**

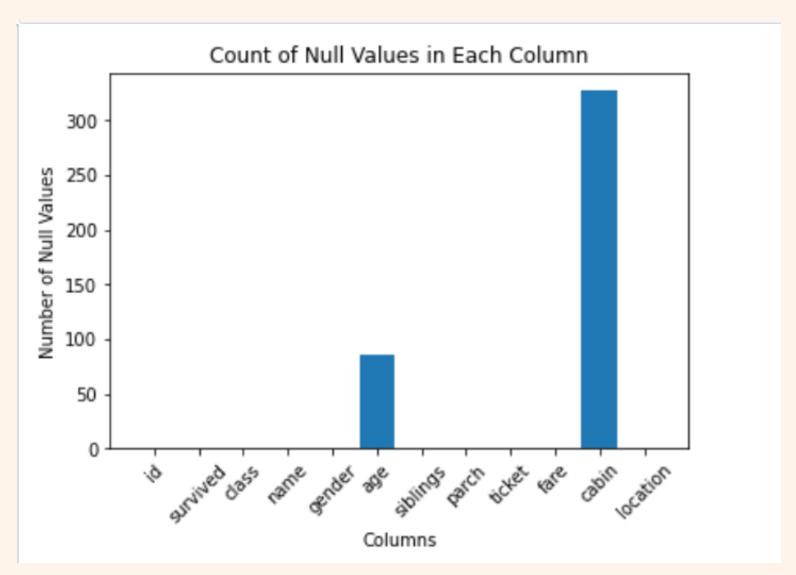
	id	survived	class	name	gender	age	siblings	parch	ticket	fare	cabin	location
0	892	1	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	NaN	Q
1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	NaN	S
2	894	0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	NaN	Q
3	895	0	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	NaN	S
4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	NaN	S
413	1305	0	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.0500	NaN	S
414	1306	1	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	C105	С
415	1307	0	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	NaN	S
416	1308	0	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.0500	NaN	S
417	1309	1	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	NaN	С

418 rows × 12 columns

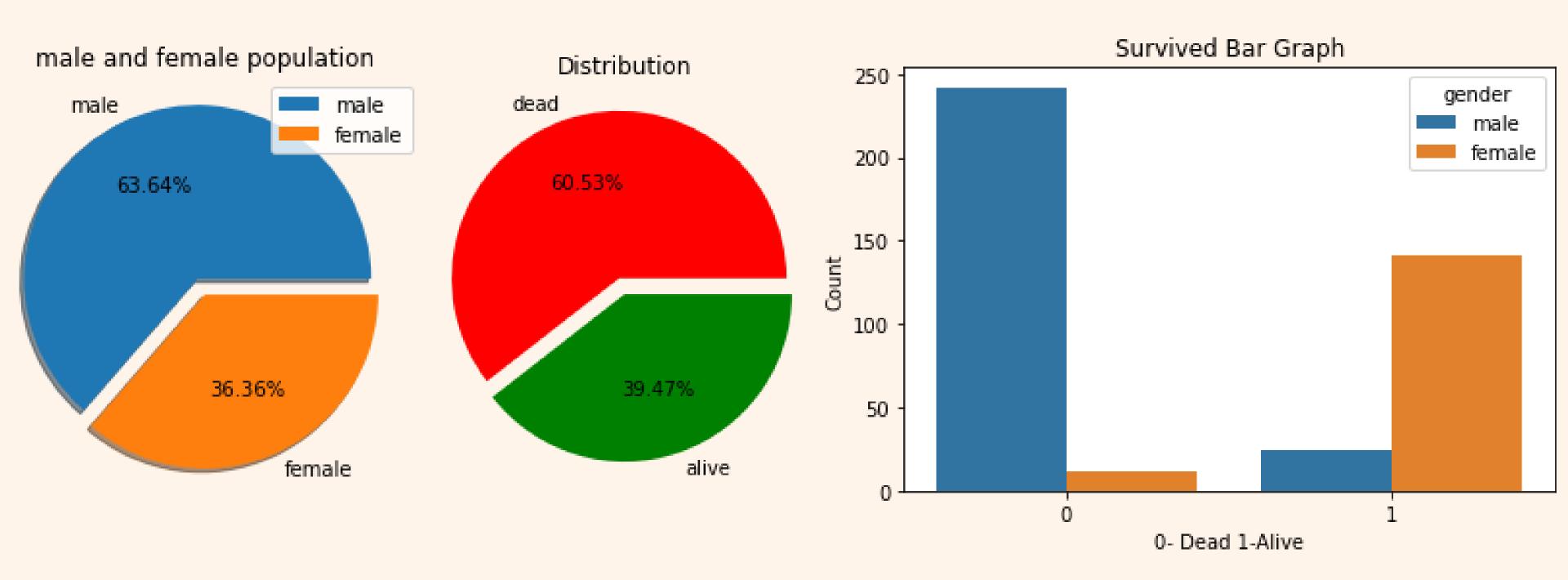
### **Data Cleaning**

df.isnull(	).sum()	#
id	0	
survived	0	
class	0	
name	0	
gender	0	
age	86	
siblings	0	
parch	0	
ticket	0	
fare	1	
cabin	327	
location	0	
dtype: int	:64	

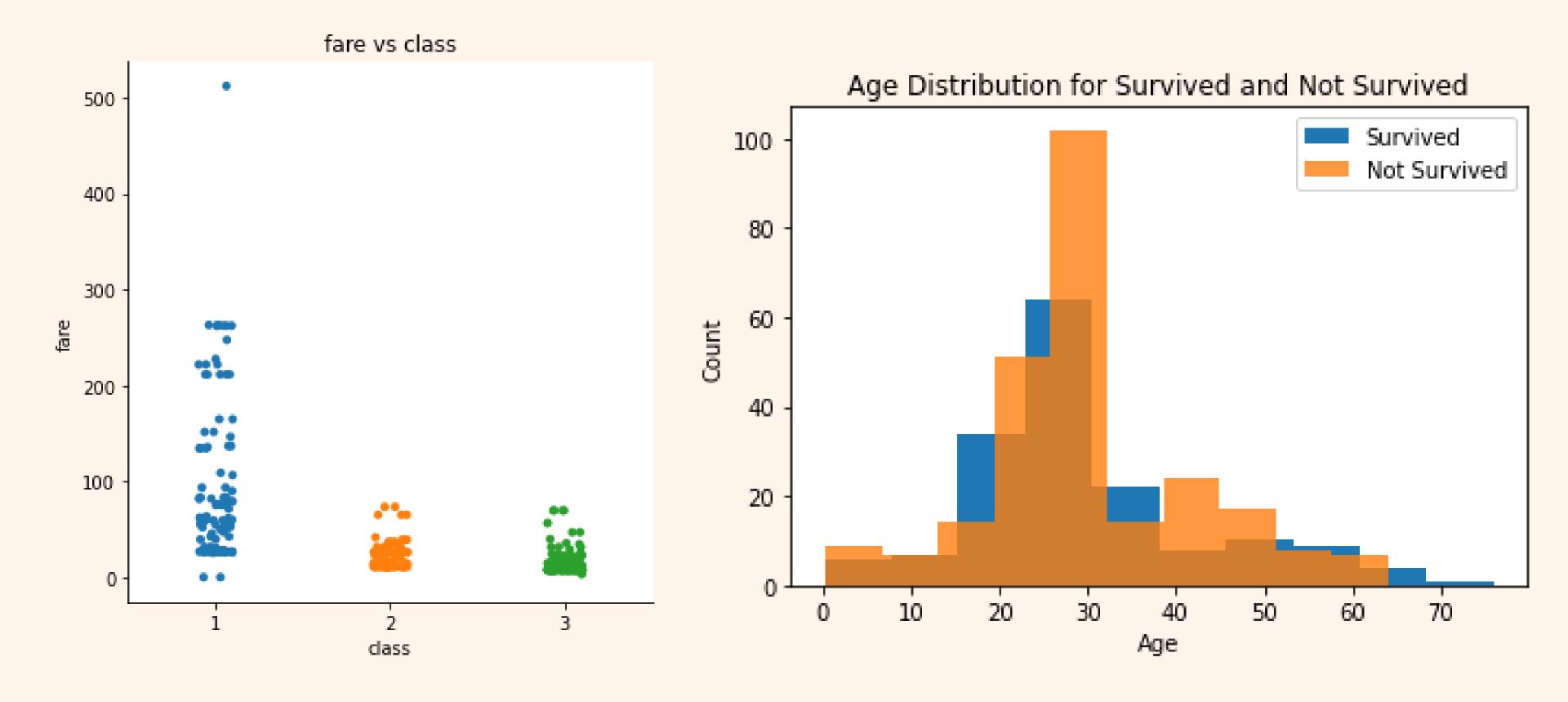




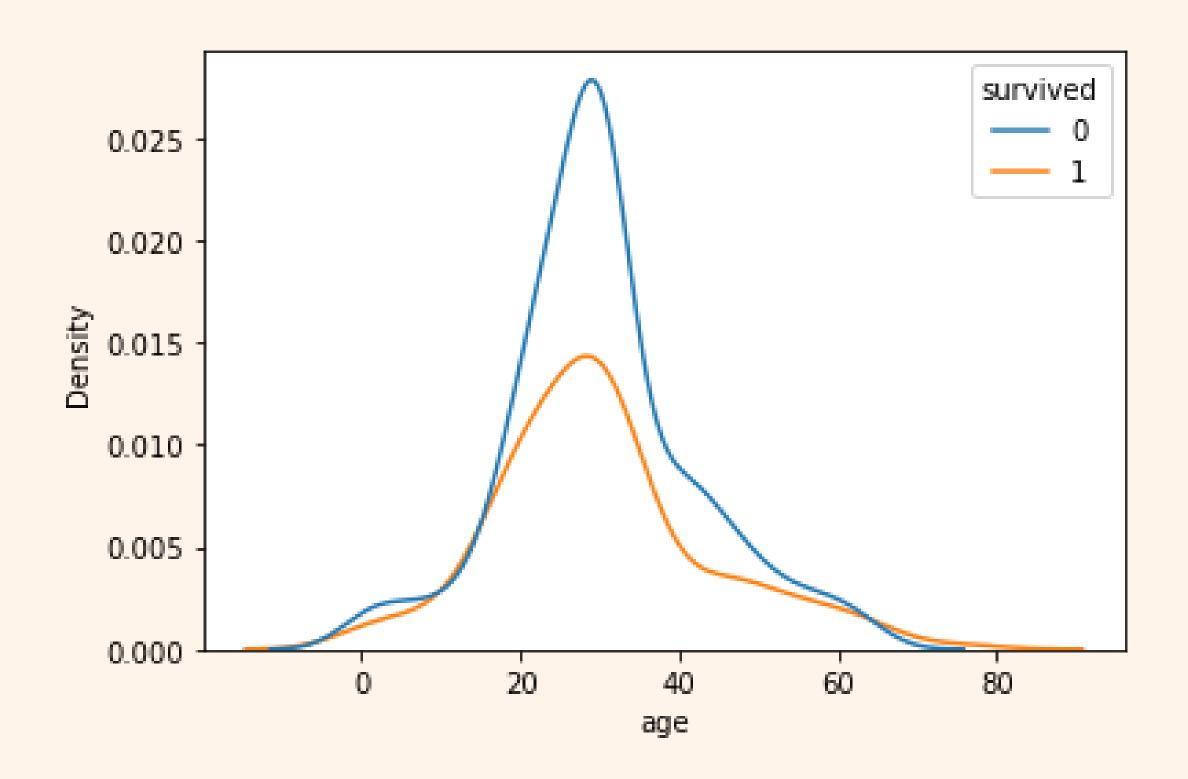
### **Distribution Graphs**



### **Corelation Graphs**



#### Same age vs Survived but using sns



#### \* Among males how many survived, among females how many survived?

```
g=df.groupby("gender")["survived"].value_counts()
  g
 gender survived
  female 1
                      141
                     11
  male
                      242
                       24
  Name: survived, dtype: int64
: f=len(df[df['gender']=='female'])
 m=len(df[df['gender']=='male'])
  print(f"Percentage of female survivors = {g[0]/f*100}")
  print(f"Percentage of female death = {g[1]/f*100}")
  print(f"Percentage of male survivors = \{g[-1]/m*100\}")
  print(f"Percentage of male death = {g[2]/m*100}")
  Percentage of female survivors = 92.76315789473685
  Percentage of female death = 7.236842105263158
  Percentage of male survivors = 9.022556390977442
  Percentage of male death = 90.97744360902256
```

## Advantages

- Visualize the Titanic dataset to discover patterns, correlations.
- Can Apply Similar procedures to many other Datasets
- Learning Libraries to implement on real data sets.
- Can also be used later for predictions/machine learning.



## Conclusion

- The survival rate for women and children was significantly higher than for men.
- Passengers in first, second class had a better chance of survival.
- Above conclusions were solely drawn due to visualization instead of studying 892 rows



#### References

https://www.kaggle.com/datasets/brendan45774/test-file

# Thank you!