

An abstract graphic on the left side of the slide, consisting of a complex network of white lines connecting small dots, resembling a molecular structure or a data network. The dots are slightly blurred, giving a sense of depth and movement. The lines form a dense, interconnected web that tapers off towards the right.A solid pink horizontal bar located in the top right corner of the slide.

Case Study: Titanic

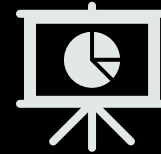
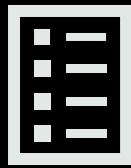
Shashank Maindola

Background

- The titanic causality is one of the worst sea causalities to happen in the history of mankind
- But there are lot of key information which can be evaluated
- We have a given data set to get any valuable information out of it
- Data Science tools were used for analyzing and filtering the given raw data set
- We have used Python based Anaconda distribution for evaluating and analysing the data



Methodology



Raw Data

Process &
Analysis

Result

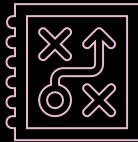
Processing Raw Data



Getting raw data



Finding missing values



Filtering/imputing the values accordingly

Getting the Data & Processing

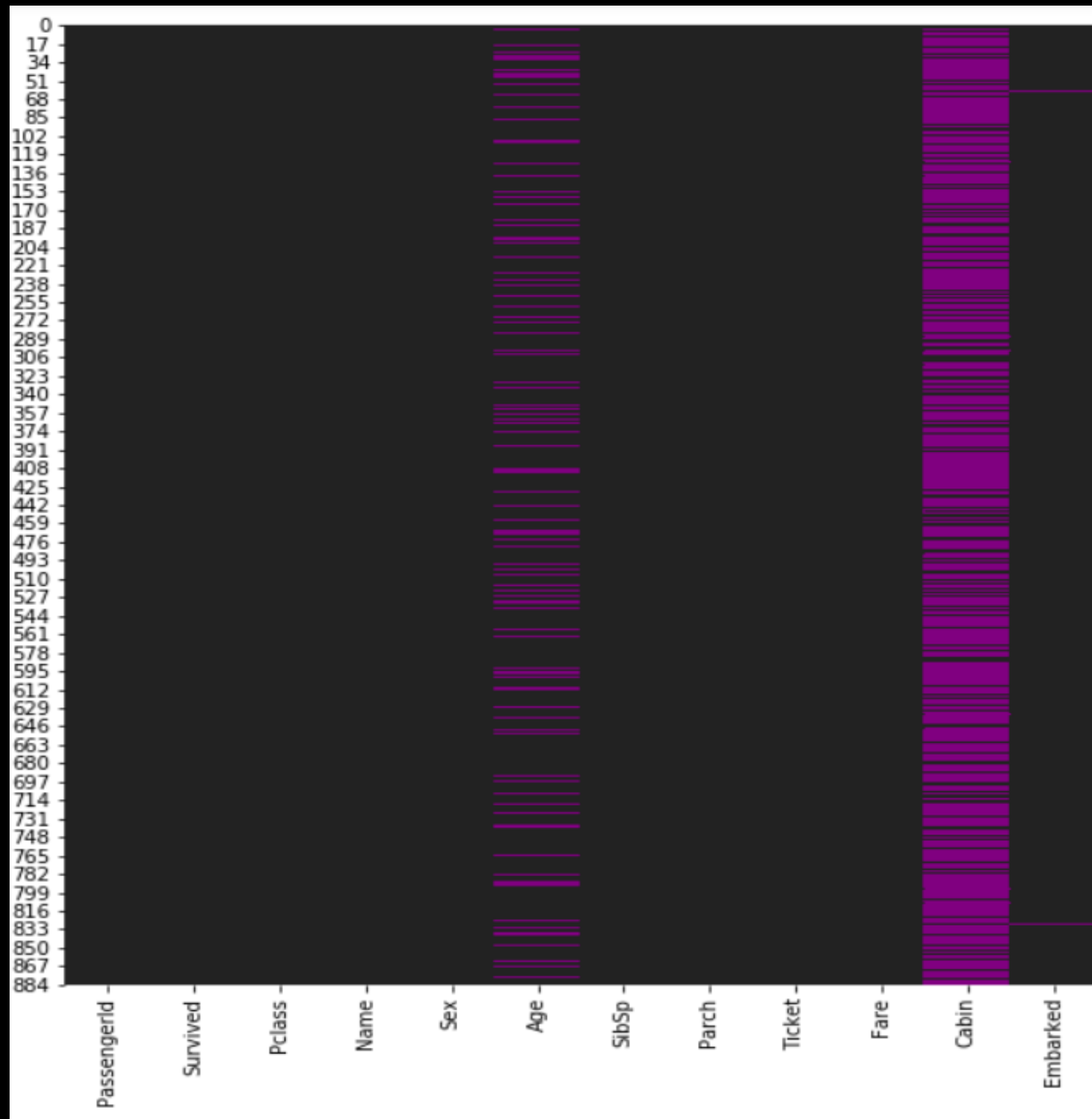
The data was imported from the 'kaggle' and was converted to a data frame

Passenger	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
1	0	3	Braund, Mr. Owen Harris	male	22	1	0	A/5 21171	7.25	null	S
2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Thayer)	female	38	1	0	PC 17599	71.2833	C85	C

There was a need to find out the missing data which was to be counted

Output:

Passenger	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	0	0	0	0	177	0	0	0	0	687	2



Missing values

Getting the Data & Processing(Cont.)

We renamed the column

```
miss_val.rename(columns={'index':'variables',0:'missing_count'},inplace=True)
```

Calculated the percentage of missing data

Sorted the missing data

The result:

variables	missing_count	missing_Per
Cabin	687	77.10437710437711
Age	177	19.86531986531986
Embarked	2	0.224466891133557

Treatment of missing values

Eliminated the values 'Cabin' as the 77% is missing

```
df.drop('Cabin',axis=1,inplace=True)
```

Almost 20% values related to age are missing so it needs to be treated as well, being a numeric value we imputed the missing values with the mean values

```
#impute missing val
df.fillna(value=df['Age'].mean(),inplace=True)
```

Finally, we see that embarked has 2 values missing we replaced them with

```
df['Embarked'].fillna(value='S',inplace=True)
```

The Result:

[illegible]

Analysis



General Analysis



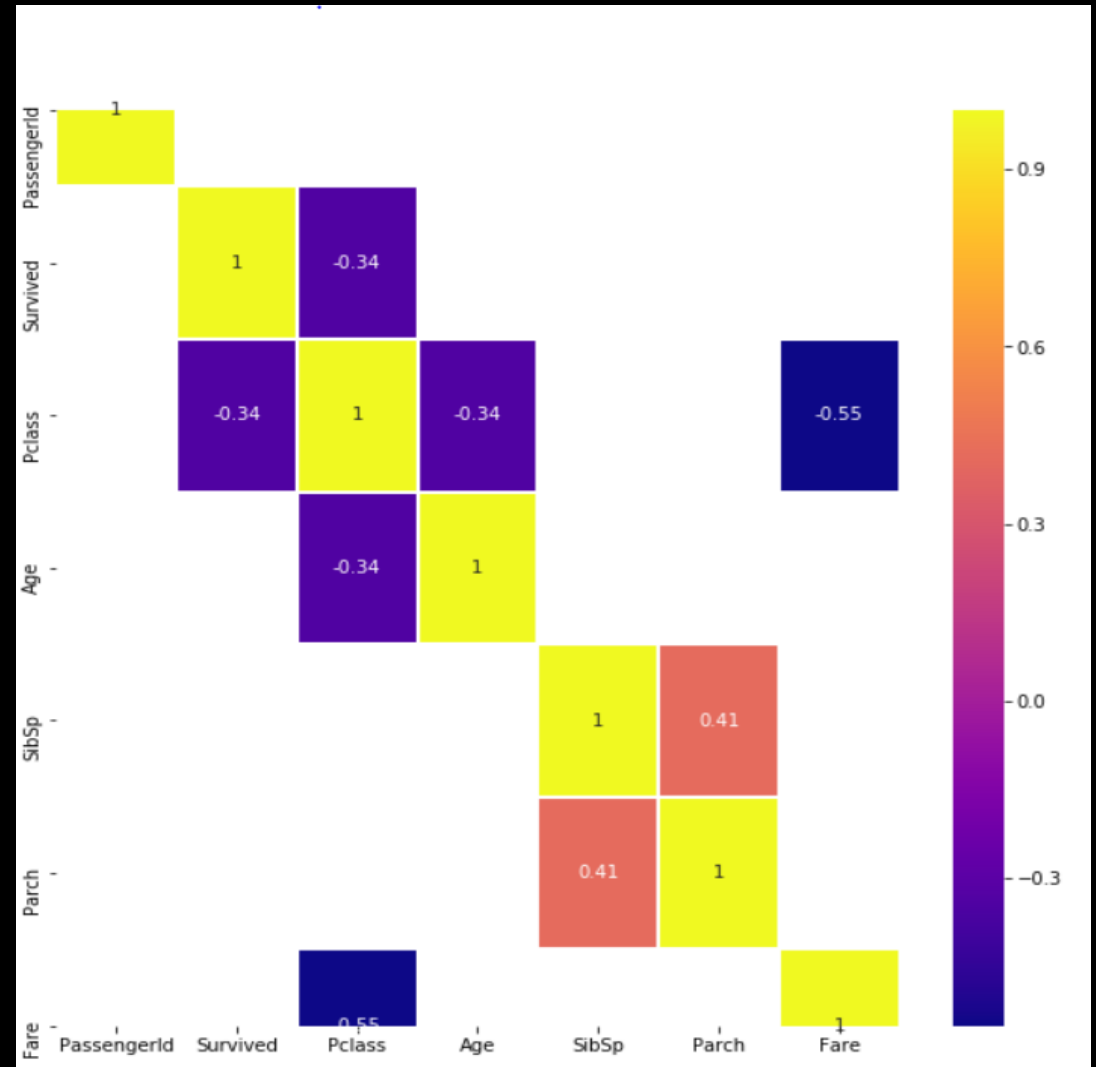
Business Analysis



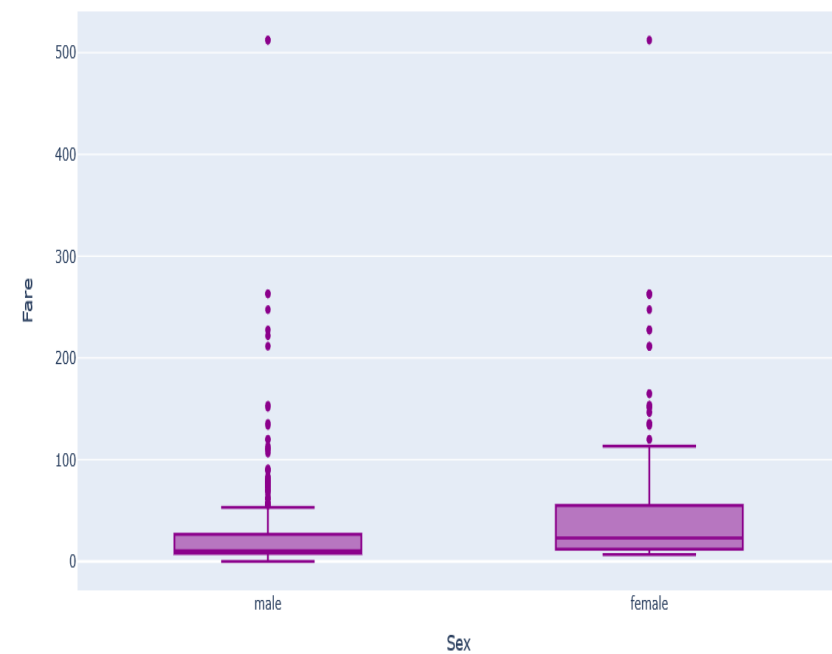
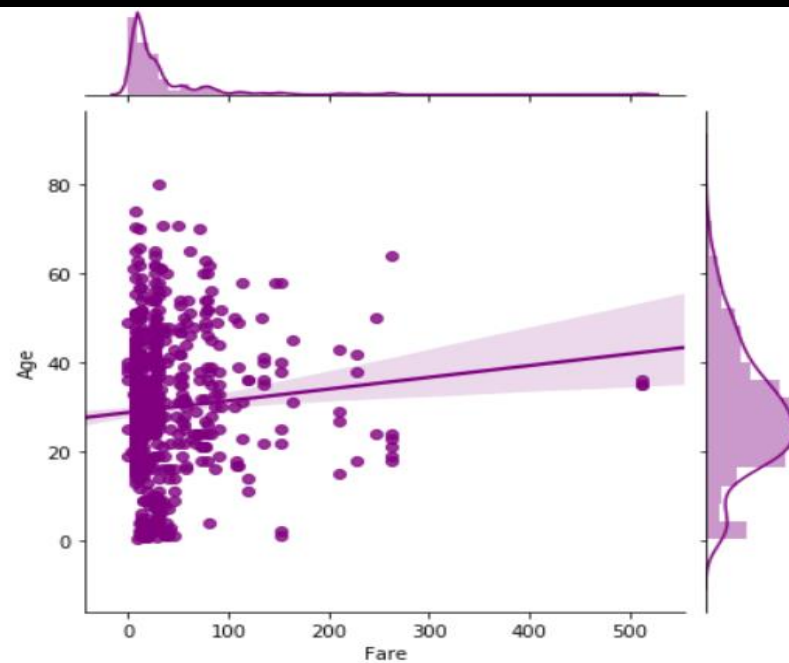
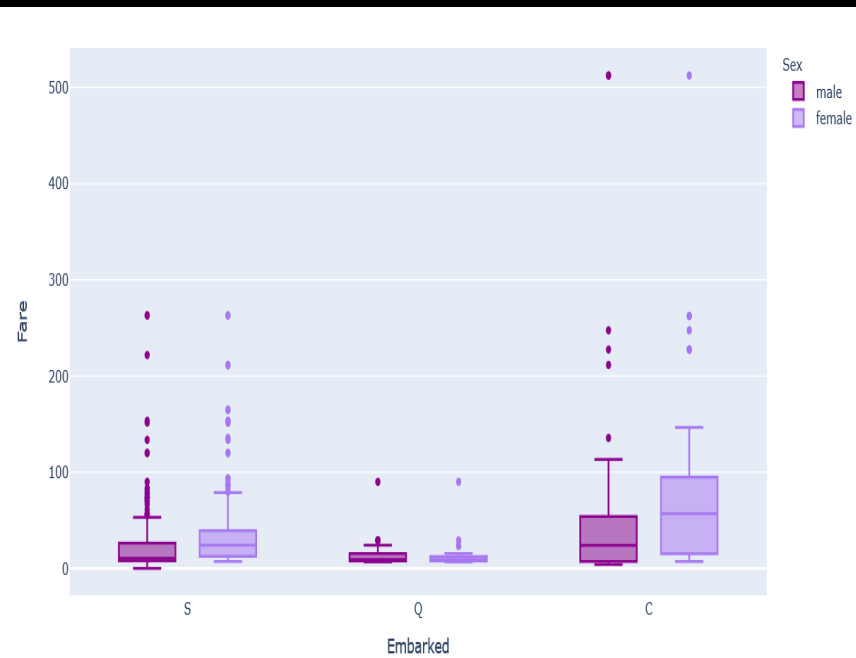
Survival Analysis

General Analysis

**Correlation
Plot**

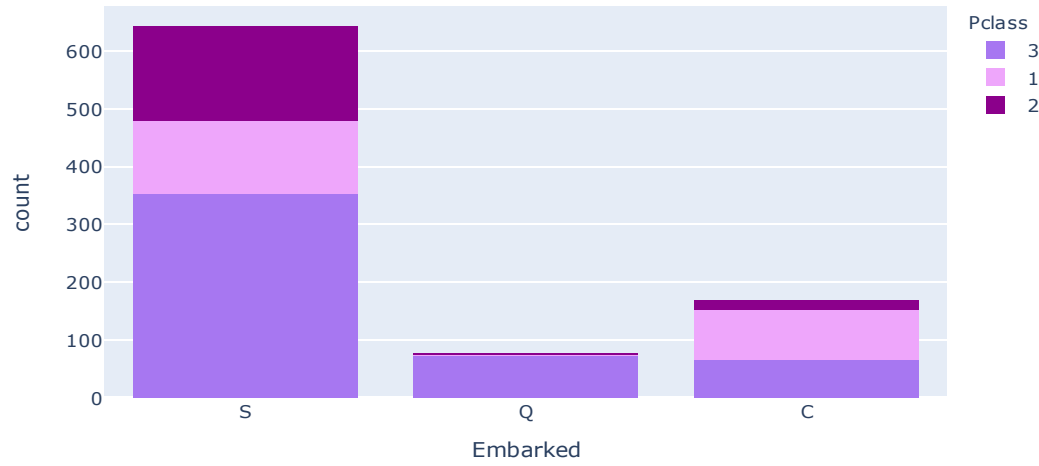


General Analysis(cont.)

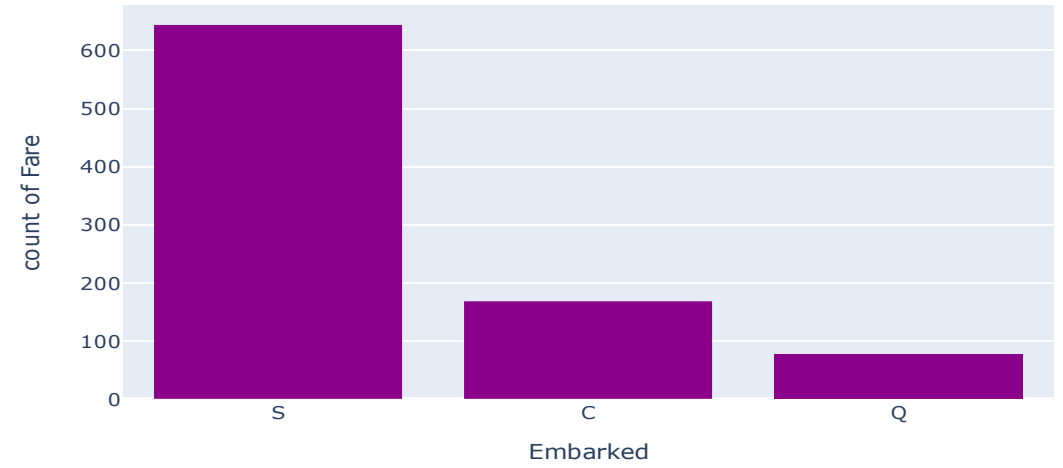


Business Analysis

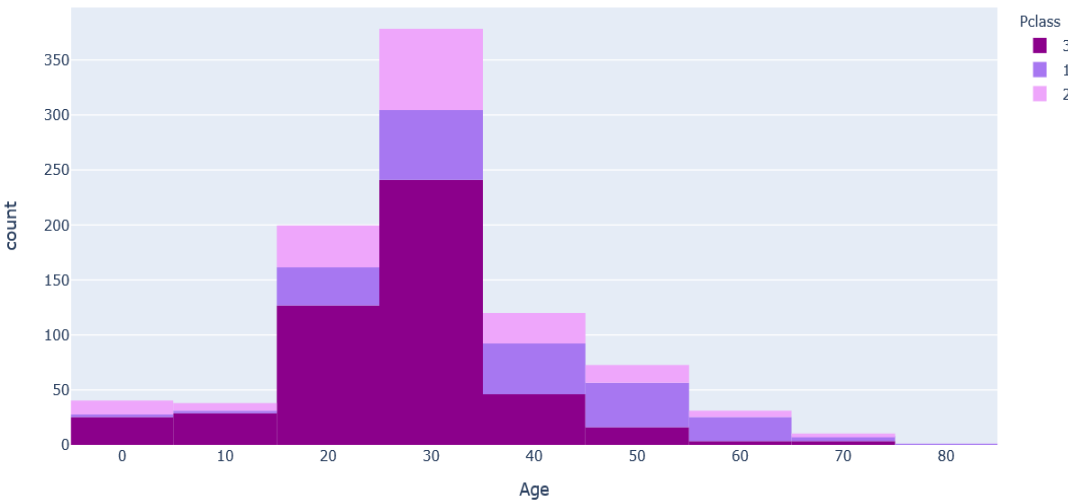
Passenger count of each port & their respective classes



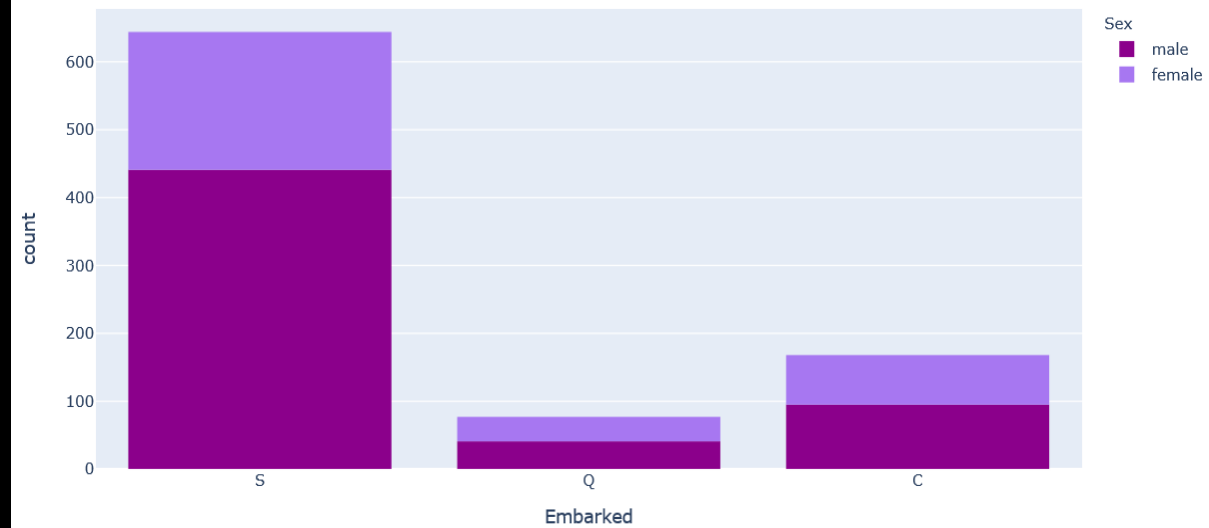
Total earning from each Port



Age wise passenger class

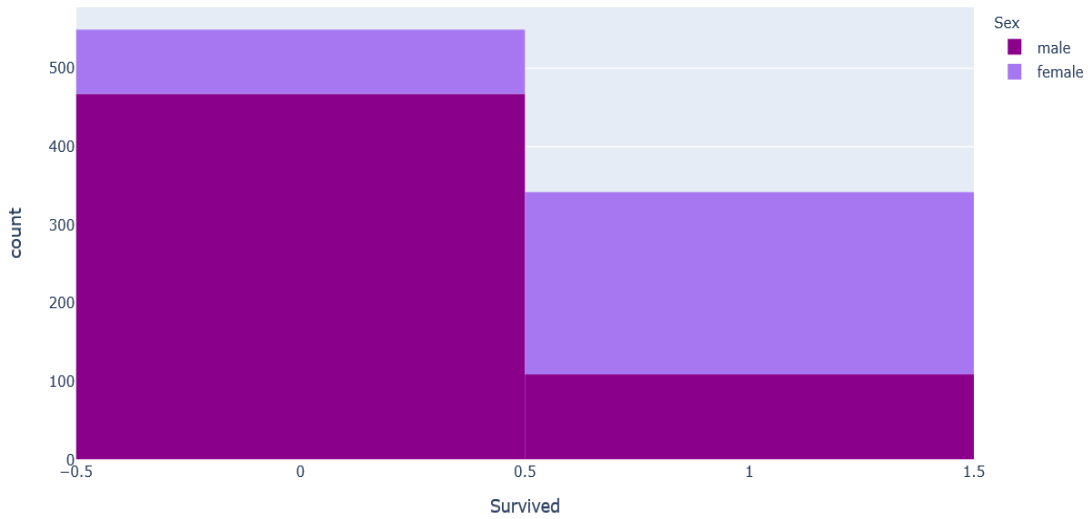


Passenger from each port

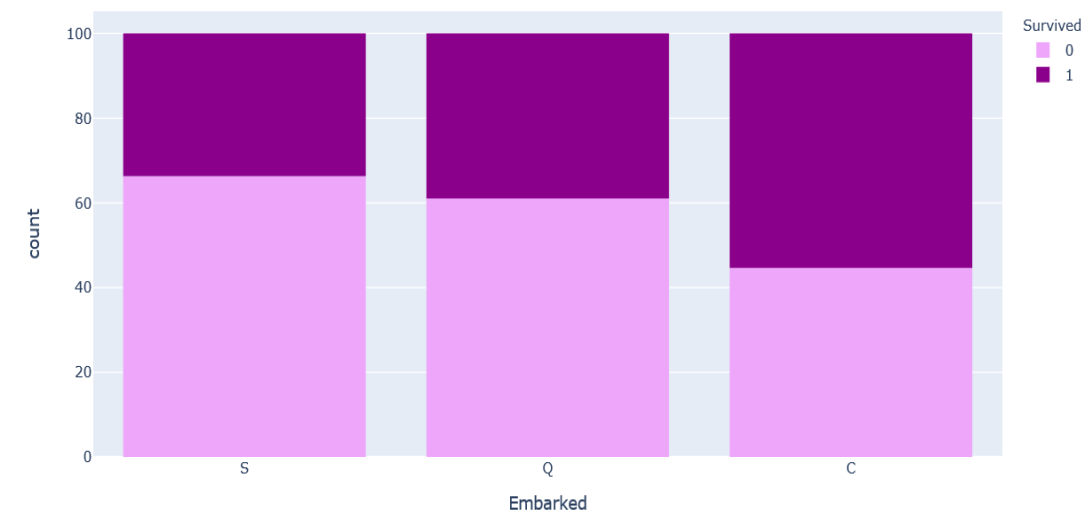


Survival Analysis

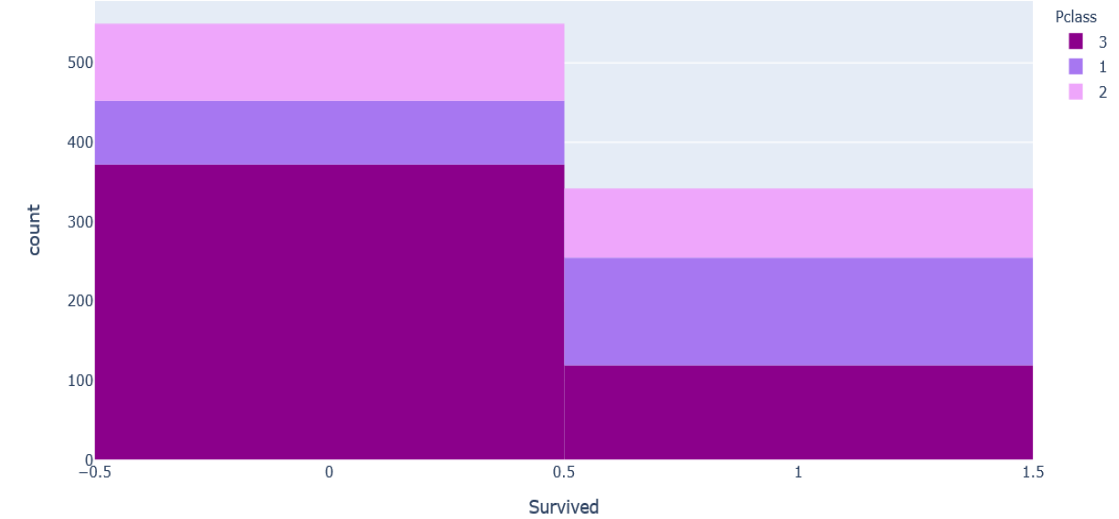
Survived



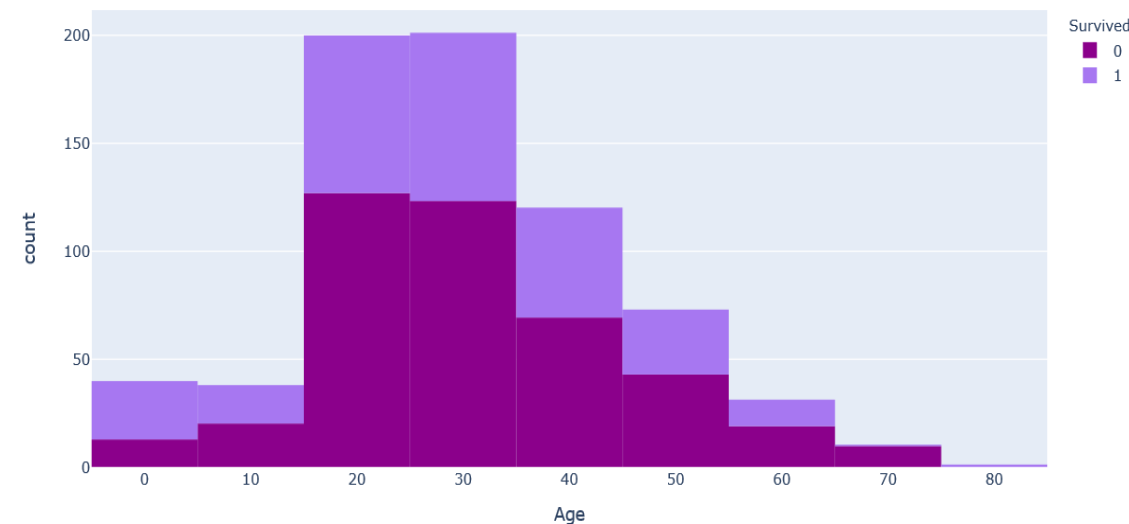
Survival Percent based on Port



Survival based on class



Survival based on age



Key Outcomes



- Generally females passengers paid more fare than males may be due to the fact the society was not open to the females so in order to travel they might had to pay more amount
- Chances of survival are directly related to the passenger class as the passengers were more privileged



- Most of the people boarded from Southampton and maximum paid for the 3rd class, so for starting a budget liner, Southampton is the best port
- Most of the passengers were youngsters around 30's and single so if someone wants to start a business then they should keep youngsters in mind
- If someone wants to start a luxury liner then Cherbourg is the best port



- Most of the surviving passengers were females as they were allowed to board the lifeboats first
- The survival percentage of people from Cherbourg was highest as the people paid more fare and had more privileges
- Survival percentage of the 1st class passengers was highest, as they were having more privileges
- Survival of youngsters was highest as they could survive for more time in icy cold conditions

Glossary

Variable	Definition	Key
Survival	Survival	0 = No, 1 = Yes
Pclass	Ticket class	1 = 1st, 2 = 2nd, 3 = 3rd
Sex	Sex	null
Age	Age in years	null
Sibsp	# of siblings / spouses aboard the Titanic	null
Parch	# of parents / children aboard the Titanic	null
Ticket	Ticket number	null
Fare	Passenger fare	null
Cabin	Cabin number	null
Embarked	Port of Embarkation	C = Cherbourg, Q = Queenstown, S = Southampton