

**Can the logistic  
regression predict  
which passenger will  
survive on the Titanic?**



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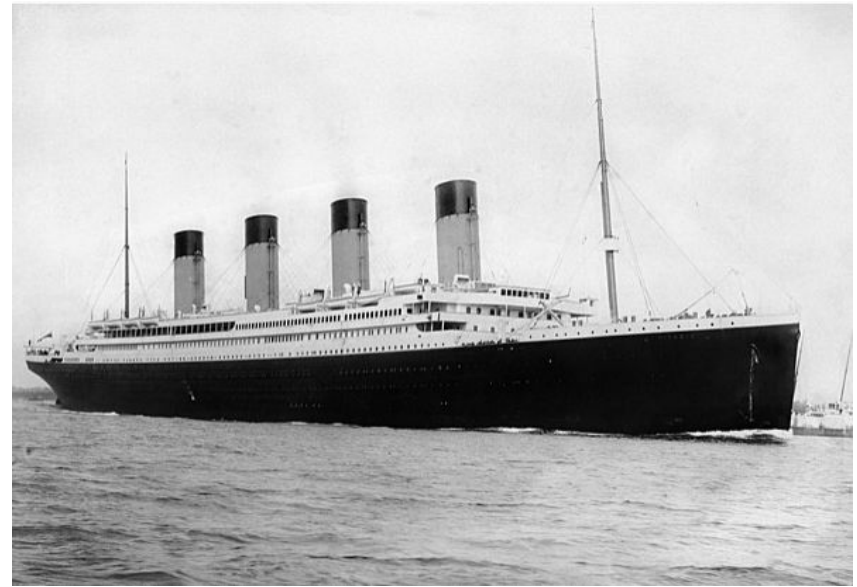


# Backstory of titanic



**RMS Titanic:** Luxurious, most advanced ship passenger at that time

- Left Southampton on 10th April 1912 to NY
- Hit North Atlantic Ocean iceberg at 11:40 PM
- 3 hours to sink on 14 April 1912



- 2,240 passengers on board incl 885 crew members
- Lifeboats for  $\frac{1}{3}$  passengers only
- 1500 died



# The dataset

Train data for exploratory analysis:

- 891 entries
- 11 features
- 1 target variable: Survived
- 5 categorical + 7 numerical

Cleaning:

- Missing values (%)
- Deleted 'Cabin'
- Replaced NaN age with mean
- Age range:0-80
- Embarked deleted 2 rows

=> only 889 entries + 11 columns

Cabin	77.10
Age	19.87
Embarked	0.22

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):
#   Column          Non-Null Count  Dtype
---  -
0   PassengerId     891 non-null    int64
1   Survived        891 non-null    int64
2   Pclass          891 non-null    int64
3   Name            891 non-null    object
4   Sex             891 non-null    object
5   Age             714 non-null    float64
6   SibSp           891 non-null    int64
7   Parch           891 non-null    int64
8   Ticket          891 non-null    object
9   Fare            891 non-null    float64
10  Cabin           204 non-null    object
11  Embarked        889 non-null    object
dtypes: float64(2), int64(5), object(5)
memory usage: 83.7+ KB
```

# Example

PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
680	1	1	Cardeza, Mr. Thomas Drake Martinez	male	36.0	0	1	PC 17755	512.3292	B51 B53 B55	C
738	1	1	Lesurer, Mr. Gustave J	male	35.0	0	0	PC 17755	512.3292	B101	C

Sibsp = Number of sibling/spouse aboard (0-5)

Parch = Number of parents/children aboard (0-6)

-> Majority were on the boat on their own

## SibSp

0	608
1	209
2	28
4	18
3	16
8	7
5	5

## Parch

0	678
1	118
2	80
5	5
3	5
4	4
6	1

## 2nd cleaning: Pre-processing

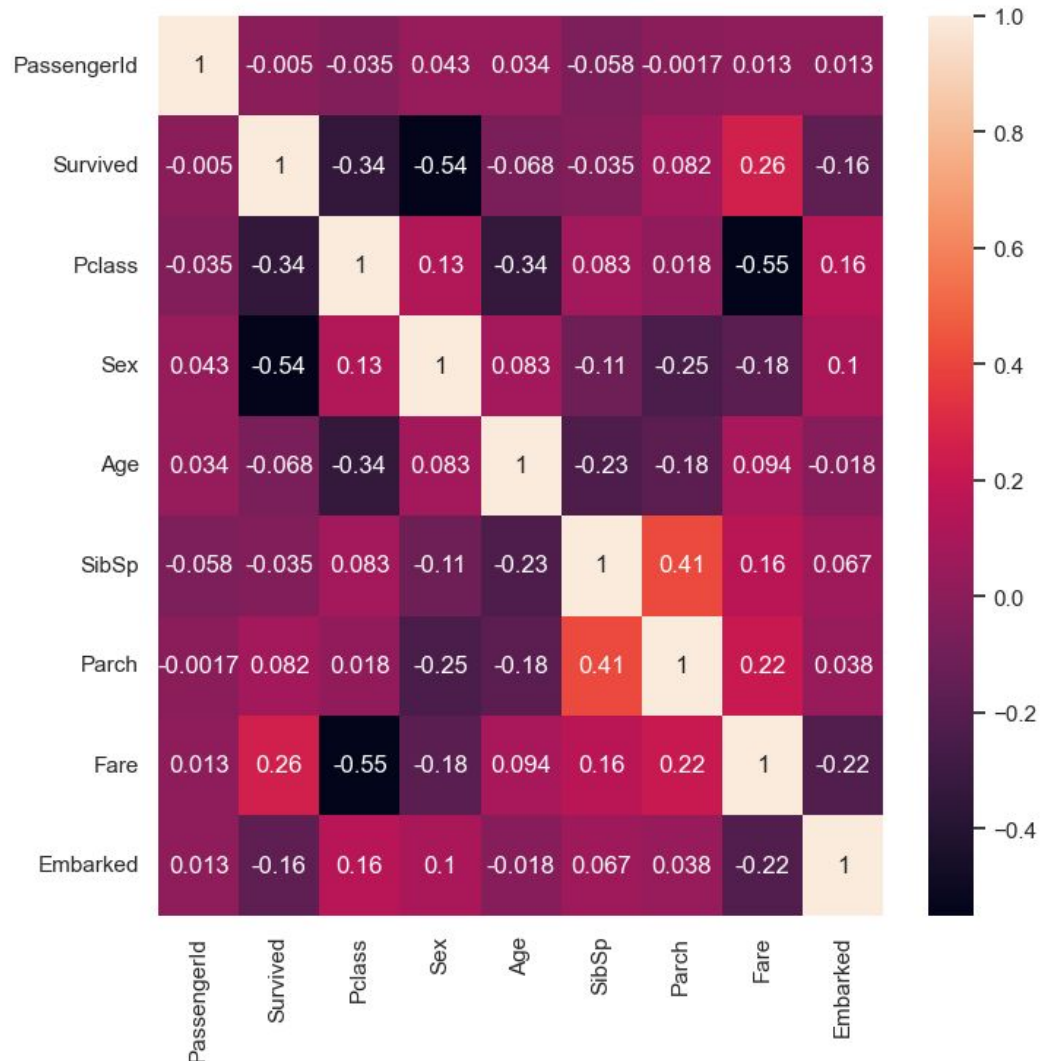
- Encode: 'Age' and 'Embark
- Standardise all

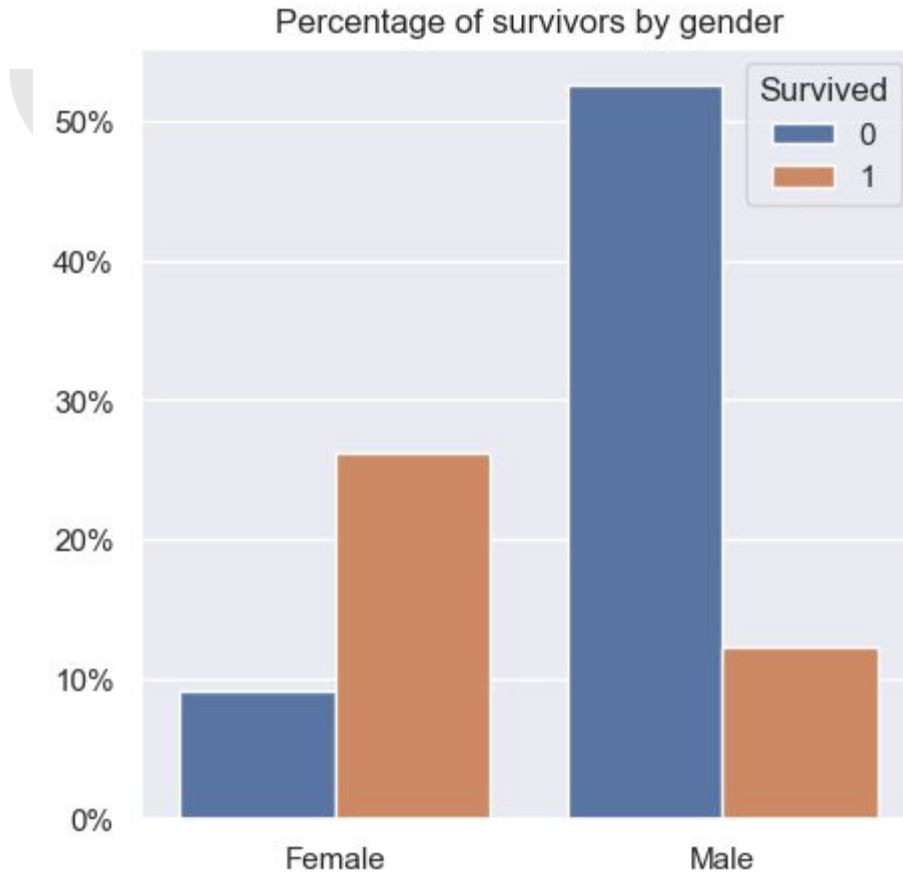


# Correlation Matrix

Top 4 correlated variables:

- Sex: -0.54
- Pclass: 0.26
- Fare: 0.26
- Embarked: -0.16
- Remaining: <0.1





1) **Sex:** gender male/female

Correlation: -0.54

- Female survival rate: 26.15%
- Male survival: 12.23%

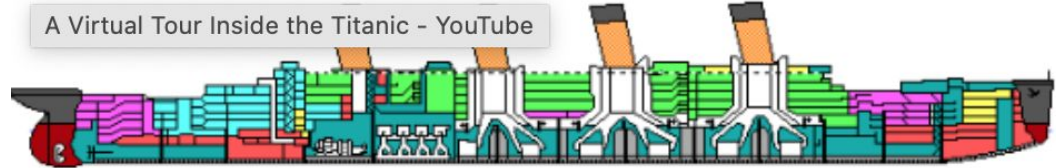


## 2) PClass = Proxy for socioeconomic classes , correlation at 0.26

Pclass

2	184
1	216
3	491

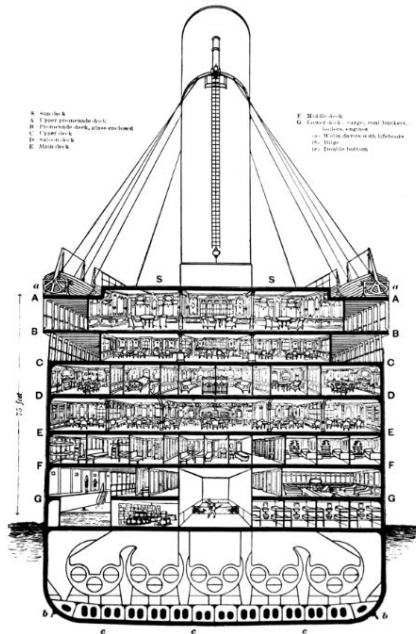
A Virtual Tour Inside the Titanic - YouTube



**R.M.S. Titanic**

### Legend

<span style="color: green;">■</span> First Class	<span style="color: yellow;">■</span> Crew - living
<span style="color: cyan;">■</span> Second Class	<span style="color: teal;">■</span> Crew - work
<span style="color: magenta;">■</span> Third Class	<span style="color: red;">■</span> Cargo & Stores



The Forward First Class Grand Staircase of *Titanic's* sister ship RMS *Olympic*. *Titanic's* staircase would have looked nearly identical. No known photos of *Titanic's* staircase exist.



The gymnasium on the boat deck, which was equipped with the latest exercise machines



The à la carte restaurant on B Deck (pictured here on sister ship RMS *Olympic*), run as a concession by Italian-born chef [Gaspere Gatti](#)



The First Class lounge of RMS *Olympic*, *Titanic's* sister ship



The First Class Turkish baths, located along the Starboard side of F-Deck

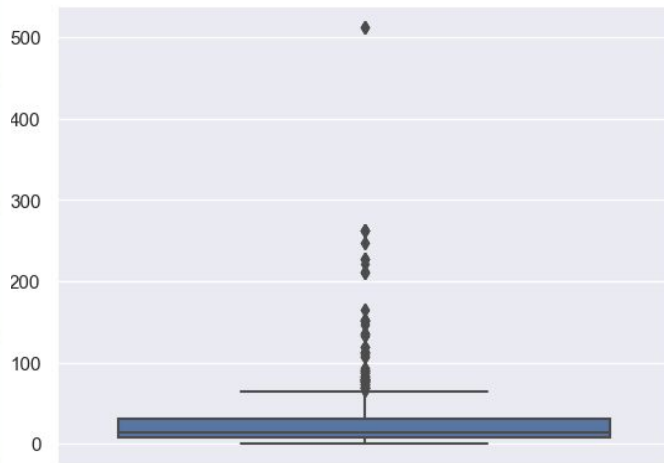
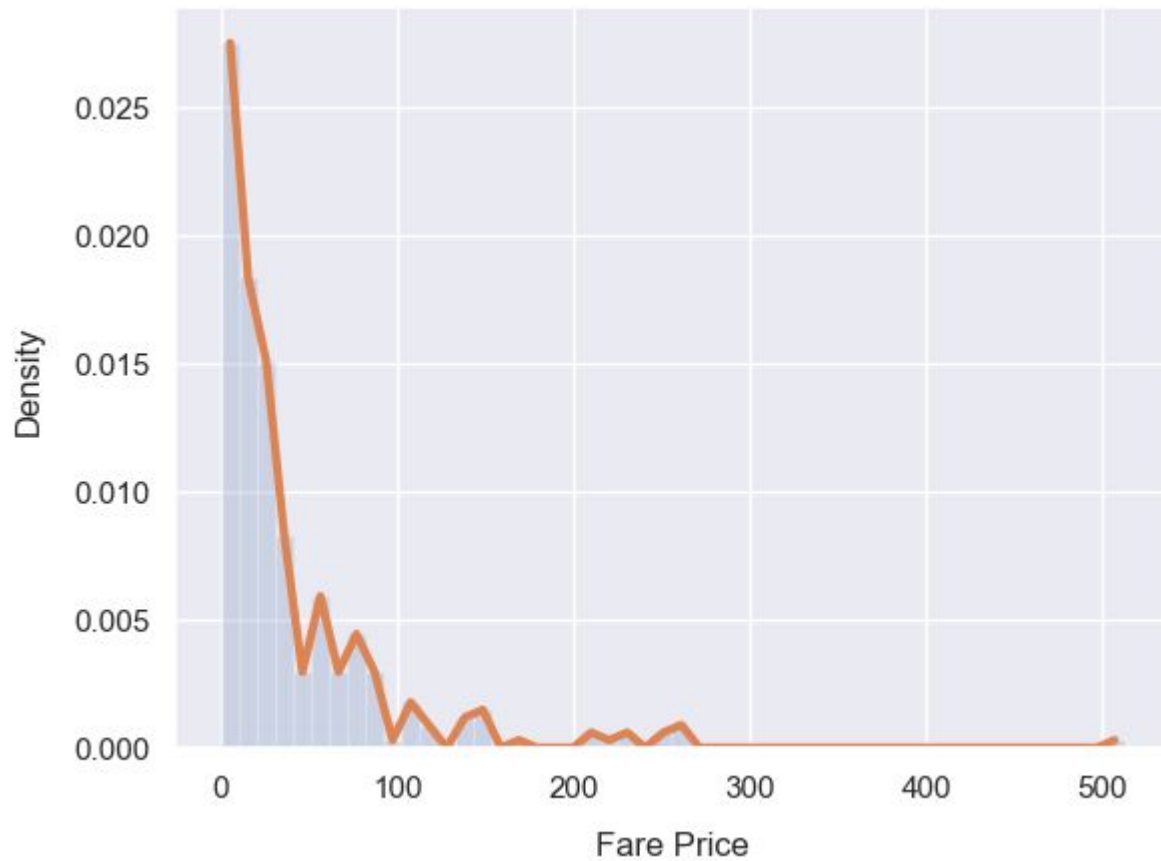


Percentage of survivors by social-economic class



- Highest chance of survival in First Class
- Lowest: Third Class

3) Fare: price tickets ranged from 0.0 to 513 shillings.



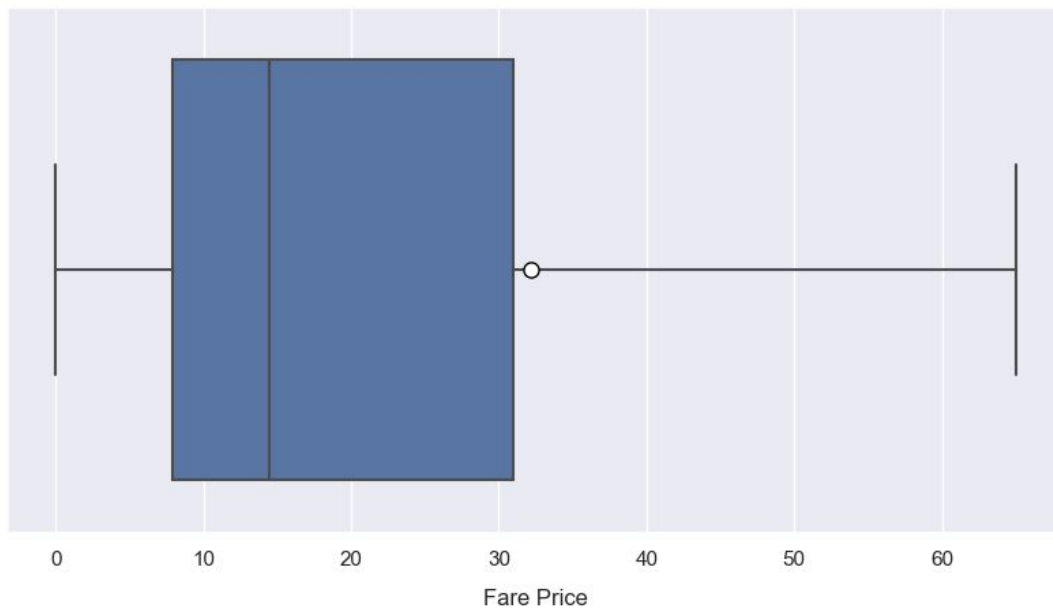
Pclass	Average Fare
1	84.154687
2	20.662183
3	13.675550

Mean: 32 (affected by outliers)

Median: 14

Range: 0 to 513

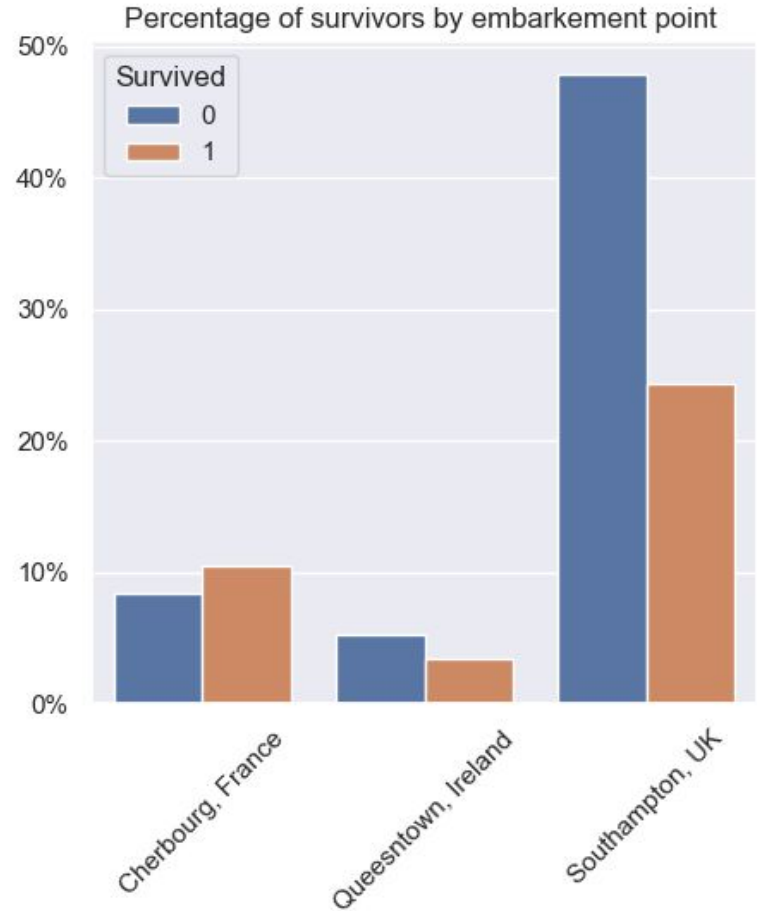
->separating into bins for histogram





#### 4) Embarked: 3 ports to embark from

- Highest survival:  
Boarding from France



# Building Logistic Regression Model

## -> cleaning

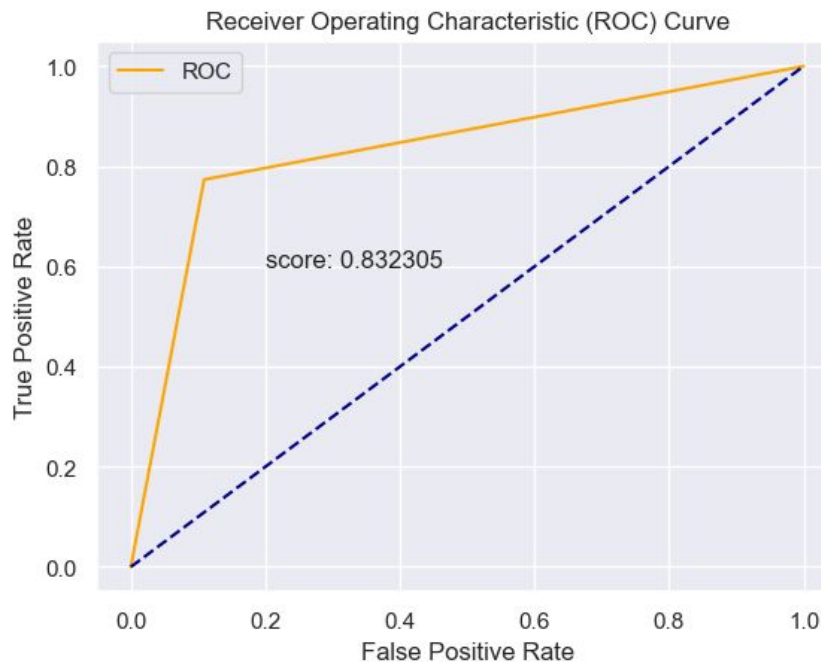
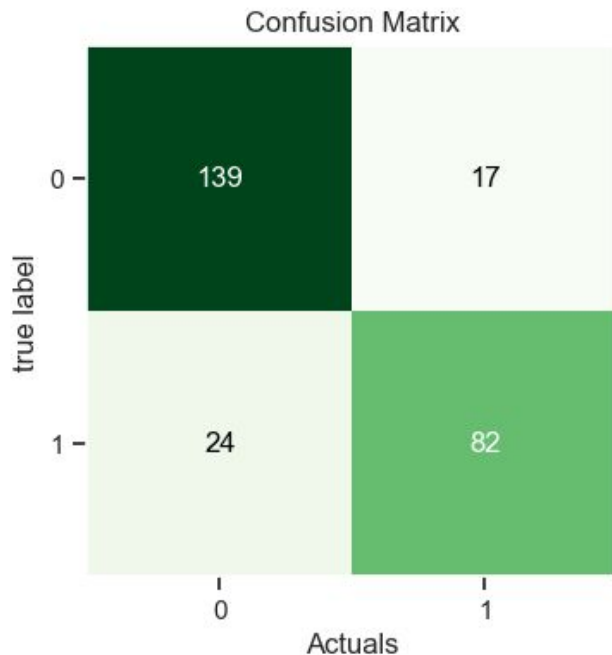
- Dropped 'Name' and 'Ticket' : words
- Dropped 'Cabin' as 77% Nan
- Replaced all missing values of 'Sex','Fare' with its median and 'Embarked' with its mode
- One-hot encoding for 'Sex' and 'Embarked'

Clean data:

	PassengerId	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	1	3	1	22	1	0	7.2500	2
1	2	1	0	38	1	0	71.2833	0
2	3	3	0	26	0	0	7.9250	2
3	4	1	0	35	1	0	53.1000	2
4	5	3	1	35	0	0	8.0500	2

# Evaluation of model

Accuracy score: 0.8435





# Conclusion

- Our logistic regression model's accuracy is at 84%.
- There are room for improvements therefore there could be further cleaning such as converting age into classes.





# Appendix

Github

Week 6

Date: 09/06/2023