#### **About Titanic Dataset**

The Titanic dataset is a famous dataset that contains information about passengers aboard the Titanic ship, which sank in 1912 after colliding with an iceberg. The dataset is often used in data science and machine learning education and competitions as a starting point for exploring data analysis and predictive modeling techniques.

The Titanic dataset contains information about **1309** passengers, including their age, gender, ticket class, cabin, port of embarkation, and whether they survived or not. The goal of many analyses and models built on the Titanic dataset is to predict whether a given passenger would have survived the disaster.

The variables in the Titanic dataset are as follows: PassengerId: Unique identifier for each passenger Survived: Whether the passenger survived (0 = No, 1 = Yes) Pclass: Ticket class (1 = 1st, 2 = 2nd, 3 = 3rd) Name: Passenger name Sex: Passenger gender Age: Passenger age SibSp: Number of siblings/spouses aboard the Titanic Parch: Number of parents/children aboard the Titanic Ticket: Ticket number Fare: Passenger fare Cabin: Cabin number Embarked: Port of embarkation (C = Cherbourg, Q = Queenstown, S = Southampton) As mentioned earlier, the main objective of many analyses and models built on the Titanic dataset is to predict whether a given passenger would have survived the disaster, based on their demographic and travel information. This is a binary classification problem, where the target variable is Survived and the predictors are the other variables in the dataset.

#### Importing Libraries

import pandas as pd

## Data Loading

data=pd.read csv('/content/titanic.csv')

data.head(5)

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fa
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.25
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.28
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.92
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.10
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.05



data.tail(5)

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fa
1304	1305	0	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.05
1305	1306	1	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.90
1306	1307	0	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.25
1307	1308	0	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.05
1308	1309	0	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.35



#### → Data Dimention: No. of Rows and Columns

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1309 entries, 0 to 1308
Data columns (total 12 columns):

		································	
#	Column	Non-Null Count	Dtype
0	PassengerId	1309 non-null	int64
1	Survived	1309 non-null	int64
2	Pclass	1309 non-null	int64
3	Name	1309 non-null	object
4	Sex	1309 non-null	object
5	Age	1046 non-null	float64
6	SibSp	1309 non-null	int64
7	Parch	1309 non-null	int64
8	Ticket	1309 non-null	object
9	Fare	1308 non-null	float64
10	Cabin	295 non-null	object
11	Embarked	1307 non-null	object
dtype	es: float64(2	), int64(5), obj	ect(5)

memory usage: 122.8+ KB

### → Get Overall Statistics About The Dataframe

data.describe(include='all')

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Tick
count	1309.000000	1309.000000	1309.000000	1309	1309	1046.000000	1309.000000	1309.000000	13
unique	NaN	NaN	NaN	1307	2	NaN	NaN	NaN	9
top	NaN	NaN	NaN	Connolly, Miss. Kate	male	NaN	NaN	NaN	C 23
freq	NaN	NaN	NaN	2	843	NaN	NaN	NaN	
mean	655.000000	0.377387	2.294882	NaN	NaN	29.881138	0.498854	0.385027	Nŧ
std	378.020061	0.484918	0.837836	NaN	NaN	14.413493	1.041658	0.865560	Na
min	1.000000	0.000000	1.000000	NaN	NaN	0.170000	0.000000	0.000000	Nŧ
25%	328.000000	0.000000	2.000000	NaN	NaN	21.000000	0.000000	0.000000	Na
50%	655.000000	0.000000	3.000000	NaN	NaN	28.000000	0.000000	0.000000	Na
75%	982.000000	1.000000	3.000000	NaN	NaN	39.000000	1.000000	0.000000	Na
max	1309.000000	1.000000	3.000000	NaN	NaN	80.000000	8.000000	9.000000	Ní



# Data Preprocessing & Data Cleaning

#### → Data Filtering

	Name	Age				
0	Braund, Mr. Owen Harris	22.0				
1	Cumings, Mrs. John Bradley (Florence Briggs Th	38.0				
2	Heikkinen, Miss. Laina	26.0				
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)					
4	Allen, Mr. William Henry	35.0				
sum(data['S	ex']=='male')					
843						
data[data['	Sex']=='male'].head()					

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Far€
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500
5	6	0	3	Moran, Mr. James	male	NaN	0	0	330877	8.4583
6	7	0	1	McCarthy, Mr. Timothy J	male	54.0	0	0	17463	51.8625
7	8	0	3	Palsson, Master. Gosta Leonard	male	2.0	3	1	349909	21.0750



494

sum(data['Survived']==1)

### ▼ Check Missing (Null) Values In The Dataset

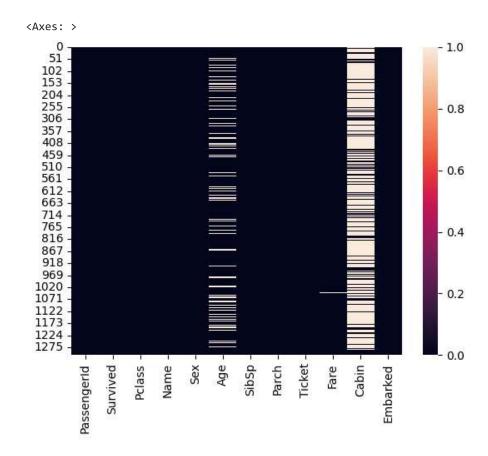
data.isnull().sum()

PassengerId Survived Pclass Name Sex 263 Age SibSp 0 Parch Ticket Fare Cabin 1014 Embarked dtype: int64

import seaborn as sns

import matplotlib.pyplot as plt

sns.heatmap(data.isnull())



per\_missing = data.isnull().sum() \* 100 / len(data)

### → Drop the Column

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

data.isnull().sum()

Passen	gerId	0
Surviv	ed	0
Pclass		0
Name		0
Sex		0
Age		263
SibSp		0
Parch		0
Ticket		0
Fare		1
Embark	ed	2
dtype:	int64	

# Handle Missing Values

data['Embarked'].mode()

```
Name: Embarked, dtype: object
data['Embarked'].fillna('S',inplace=True)
data.isnull().sum()
    PassengerId
    Survived
                     0
    Pclass
                     0
    Name
                     0
    Sex
    Age
                   263
    SibSp
                     0
    Parch
                     0
    Ticket
                     0
    Fare
                     1
    Embarked
    dtype: int64
data['Age']
    0
             22.0
    1
             38.0
     2
             26.0
    3
             35.0
    4
             35.0
    1304
             NaN
    1305
            39.0
            38.5
    1306
    1307
             NaN
    1308
             NaN
    Name: Age, Length: 1309, dtype: float64
data['Age'].fillna(data['Age'].mean(), inplace = True)
data.isnull().sum()
    PassengerId
    Survived
                   0
    Pclass
    Name
    Sex
    Age
    SibSp
    Parch
    Ticket
    Fare
                   1
    Embarked
    dtype: int64
data.head()
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticke <sup>.</sup>
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 2117
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2 310128;
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	11380;
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450

data['Sex'].unique()

```
array(['male', 'female'], dtype=object)
```

data['Gender']=data['Sex'].map({'male':1, 'female':0})

data.head(5)

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Far
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.250
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.283
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.925
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.100
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.050



# → Data Encoding

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Ge
0	1	0	3	Braund, Mr. Owen Harris	male	22.000000	1	0	A/5 21171	7.2500	
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.000000	1	0	PC 17599	71.2833	
2	3	1	3	Heikkinen, Miss. Laina	female	26.000000	0	0	STON/O2. 3101282	7.9250	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.000000	1	0	113803	53.1000	
4	5	0	3	Allen, Mr. William Henry	male	35.000000	0	0	373450	8.0500	
1304	1305	0	3	Spector, Mr. Woolf	male	29.881138	0	0	A.5. 3236	8.0500	
1305	1306	1	1	Oliva y Ocana, Dona. Fermina	female	39.000000	0	0	PC 17758	108.9000	
1306	1307	0	3	Saether, Mr. Simon Sivertsen	male	38.500000	0	0	SOTON/O.Q. 3101262	7.2500	
1307	1308	0	3	Ware, Mr. Frederick	male	29.881138	0	0	359309	8.0500	
1308	1309	0	3	Peter, Master. Michael J	male	29.881138	1	1	2668	22.3583	

1309 rows × 14 columns



data1=pd.get\_dummies(data,columns=['Embarked'],drop\_first=True)

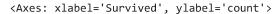
data1.head(1)

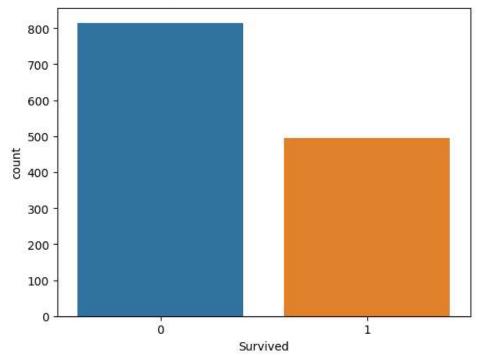
	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Gender	Emb
(	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.25	1	



# Visual Analysis

#### → How Many People Survived And How Many Died?





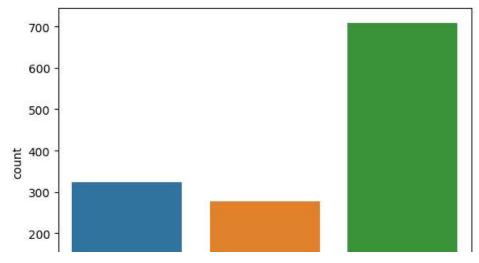
### → How Many Passengers Were In First Class, Second Class, and Third Class?

```
data['Pclass'].value_counts()

3    709
1    323
2    277
Name: Pclass, dtype: int64

sns.countplot(x='Pclass', data=data)
```





# Number of Male And Female Passengers

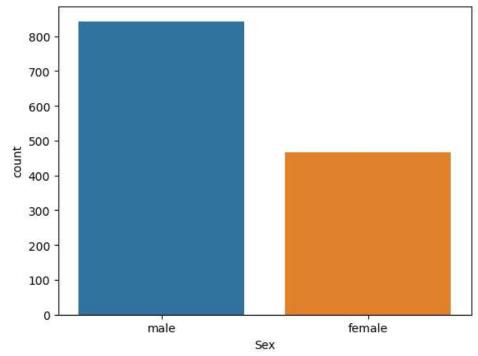
data['Sex'].value\_counts()

male 843 female 466

Name: Sex, dtype: int64

sns.countplot(x = 'Sex', data = data)

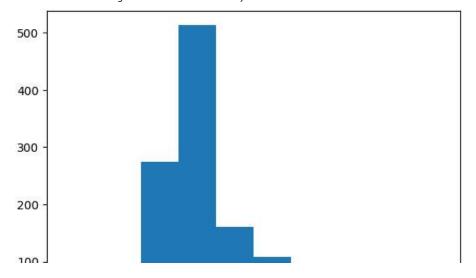
<Axes: xlabel='Sex', ylabel='count'>



plt.hist(data['Age'])

(array([ 72., 62., 274., 513., 161., 108., 65., 41., 10., 3.]), array([ 0.17 , 8.153, 16.136, 24.119, 32.102, 40.085, 48.068, 56.051, 64.034, 72.017, 80. ]),

<BarContainer object of 10 artists>)

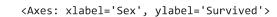


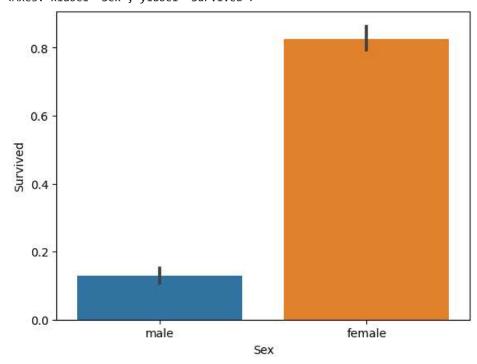
### ▼ 12. Bivariate Analysis

on

How Has Better Chance of Survival Male or Female?

sns.barplot(x='Sex',y='Survived',data=data)

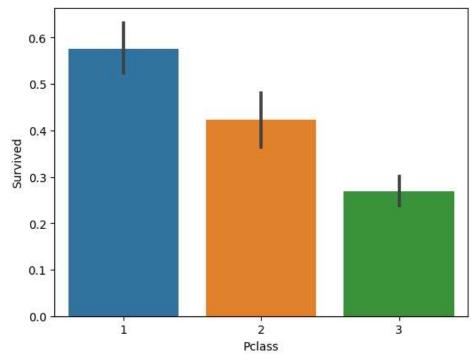




Which Passenger Class Has Better Chance of Surviva(First, Second, Or Third Class)?

sns.barplot(x="Pclass", y="Survived",data=data)

<Axes: xlabel='Pclass', ylabel='Survived'>



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