EDA

IMPORTING LIBRARIES

```
In [1]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   import seaborn as sns
   import sklearn
   import warnings
   warnings.filterwarnings('ignore')
```

IMPORTING DATASET

In [2]: df=pd.read_csv("C:/Users/Yug/Downloads/titanic/train.csv")

DATA PREPROCESSING AND VISUALIZATION

In [3]: df.head(6) Out[3]: Passengerld Survived Pclass **Ticket** Age Name Sex SibSp Parch Fare Cabi Braund. male 22.0 0 1 0 3 Mr. Owen 0 A/5 21171 7.2500 Na Harris Cumings, Mrs. John Bradley female 38.0 2 0 PC 17599 71.2833 С٤ 1 1 1 (Florence **Briggs** Th... Heikkinen, STON/O2. 2 female 26.0 7.9250 Miss. Na 3101282 Laina Futrelle, Mrs. Jacques 3 4 female 35.0 0 113803 53.1000 C12 Heath (Lily May Peel) Allen, Mr. 5 0 3 William male 35.0 0 0 373450 8.0500 Na Henry Moran, 330877 8.4583 male NaN Na Mr. James In [4]: df.shape Out[4]: (891, 12) In [5]: df.size Out[5]: 10692

In [6]: | df.info()

```
<class 'pandas.core.frame.DataFrame'>
          RangeIndex: 891 entries, 0 to 890
          Data columns (total 12 columns):
           #
                              Non-Null Count Dtype
                Column
           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
                              891 non-null
                                                object
                Sex
           5
                                                float64
                Age
                              714 non-null
           6
                SibSp
                              891 non-null
                                                int64
           7
                Parch
                              891 non-null
                                                int64
                                                object
           8
                              891 non-null
                Ticket
           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
 In [7]: df.dtypes
 Out[7]: PassengerId
                             int64
          Survived
                             int64
          Pclass
                             int64
          Name
                            object
          Sex
                            object
          Age
                           float64
          SibSp
                             int64
                             int64
          Parch
          Ticket
                            object
                           float64
          Fare
          Cabin
                            object
          Embarked
                            object
          dtype: object
 In [8]: df.columns
 Out[8]: Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp',
                  'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'],
                 dtype='object')
 In [9]: df.select_dtypes(include='object').columns #RETURNS CATEGORICAL COLUMNS
 Out[9]: Index(['Name', 'Sex', 'Ticket', 'Cabin', 'Embarked'], dtype='object')
In [10]: sns.heatmap(df.corr(),annot=True)
Out[10]: <AxesSubplot:>
                                                                 -1.0
           Passengerld - 1
                           -0.005 -0.035 0.037 -0.058 -0.0017 0.013
                                                                 - 0.8
              Survived - -0.005
                             1
                                  -0.34
                                       -0.077 -0.035 0.082 0.26
                                                                 - 0.6
                Pclass - -0.035 -0.34
                                       -0.37
                                            0.083 0.018
                                                        -0.55
                                                                 - 0.4
                 Age - 0.037 -0.077
                                 -0.37
                                        1
                                            -0.31
                                                  -0.19
                                                       0.096
                                                                - 0.2
                SibSp - -0.058 -0.035 0.083
                                       -0.31
                                                        0.16
                                                                 - 0.0
                Parch -0.0017 0.082 0.018
                                                   1
                                                                  -0.2
                                       -0.19
                                                        0.22
                 Fare - 0.013
                                       0.096
                                             0.16
                                                  0.22
                                                         Fare
                                                   Parch
                             Survived
```

#COUNTS DISTINCT VALUES PRESENT IN A PA

In [11]: df.value_counts()

Out[11]: PassengerId Survived Pclass Name Parch Cabin Embarked Sex Age SibSp Ticket Fare 890 1 Behr, Mr. Karl Howell male 26.0 0 0 111369 30.0000 C148 C 1 337 Pears, Mr. Thomas Clinton 0 1 29.0 0 113776 66.6000 1 male 1 C2 Partner, Mr. Austen 332 0 1 male 45.5 0 113043 28.5000 C124 1 330 1 1 Hippach, Miss. Jean Gertrude 57.9792 1 female 16.0 1 B18 C 111361 328 2 Ball, Mrs. (Ada E Hall) 1 36.0 0 13.0000 female 28551 Ross, Mr. John Hugo 584 0 1 A10 male 36.0 0 0 40.1250 C 1 13049 582 Thayer, Mrs. John Borland (Marian Longstreth Mor 1 1 ris) female 39.0 1 17421 110.8833 C68 C Silvey, Mrs. William Baird (Alice Munger) 578 1 1 39.0 0 55.9000 E44 female 13507 S 1 573 Flynn, Mr. John Irwin ("Irving") 1 1 36.0 PC 17474 26.3875 E25 S male 2 1 1 Cumings, Mrs. John Bradley (Florence Briggs Thay female 38.0 1 PC 17599 71.2833 C85 er) C

In [12]: df['Pclass'].value_counts(sort=False)

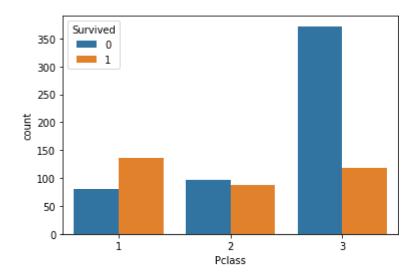
Length: 183, dtype: int64

Out[12]: 1 216 2 184 3 491

Name: Pclass, dtype: int64

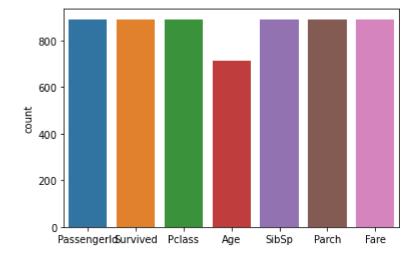
In [13]: sns.countplot(x='Pclass',hue='Survived',data=df)

Out[13]: <AxesSubplot:xlabel='Pclass', ylabel='count'>



In [14]: sns.countplot(data=df)

Out[14]: <AxesSubplot:ylabel='count'>



MISSING VALUES

In [15]: df.isna() #CHECKING NULL VALUES (FALSE= NOT NULL)

Out[15]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Emb
0	False	False	False	False	False	False	False	False	False	False	True	
1	False	False	False	False	False	False	False	False	False	False	False	
2	False	False	False	False	False	False	False	False	False	False	True	
3	False	False	False	False	False	False	False	False	False	False	False	
4	False	False	False	False	False	False	False	False	False	False	True	
886	False	False	False	False	False	False	False	False	False	False	True	
887	False	False	False	False	False	False	False	False	False	False	False	
888	False	False	False	False	False	True	False	False	False	False	True	
889	False	False	False	False	False	False	False	False	False	False	False	
890	False	False	False	False	False	False	False	False	False	False	True	

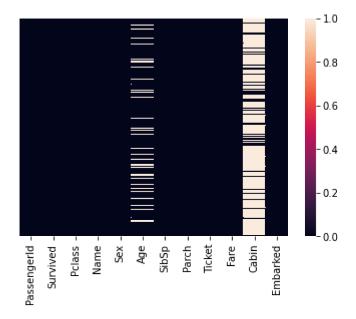
891 rows × 12 columns

In [16]: df.isna().sum()

Out[16]: PassengerId 0 Survived 0 Pclass 0 Name Sex 0 177 Age SibSp 0 Parch 0 Ticket 0 Fare 0 Cabin 687 Embarked 2 dtype: int64

In [17]: sns.heatmap(df.isnull(),yticklabels=False) #(single color in a column represer

Out[17]: <AxesSubplot:>

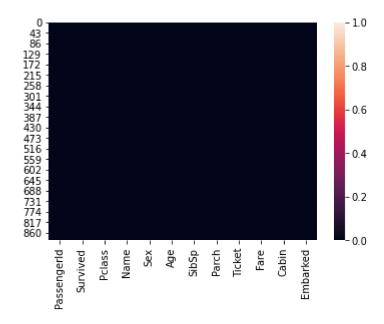


```
In [18]: df=df.fillna(
      'Embarked':df['Embarked'].replace(np.NaN,'S')
      })
                                             #treating the null values
In [19]: df.isna().sum()
```

```
Out[19]: PassengerId
                         0
         Survived
                         0
         Pclass
                         0
         Name
                         0
                         0
         Sex
                         0
         Age
         SibSp
         Parch
                         0
                         0
         Ticket
         Fare
                         0
         Cabin
                         1
         Embarked
         dtype: int64
```

```
In [20]: sns.heatmap(df.isna()) #no null values now
```

Out[20]: <AxesSubplot:>



DUPLICATE VALUES

```
In [21]: df.duplicated()
Out[21]: 0
                 False
                 False
         1
                 False
         2
         3
                 False
         4
                 False
         886
                 False
         887
                 False
         888
                 False
         889
                 False
         890
                 False
         Length: 891, dtype: bool
In [22]: df.duplicated().sum() #no duplicate value present
Out[22]: 0
In [23]: df.describe(include='object')
                                            #basic statistics
Out[23]:
                           Name
                                  Sex
                                       Ticket Cabin
                                                   Embarked
           count
                             891
                                  891
                                         891
                                               890
                                                         891
```

2

577

male 347082

891

1

top Murdlin, Mr. Joseph

681

7

147

C78

33

3

S

646

freq

unique

```
In [24]: def piechart(x):
    return x.value_counts().plot(kind='pie',autopct='%2f')

def countplot(x1):
    return sns.countplot(x=x1,data=df)

def histogram(x):
    return plt.hist(x,bins=10)

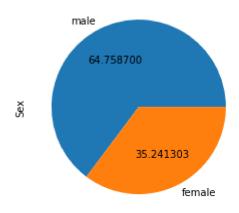
def distplot(x):
    return sns.distplot(x)

def boxplot(x):
    return sns.boxplot(x)
```

1. SEX

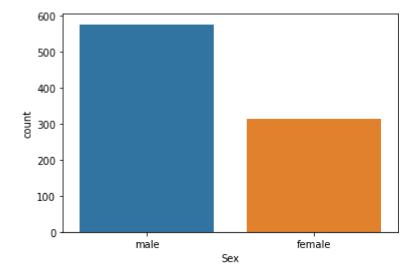
```
In [25]: piechart(df['Sex'])
```



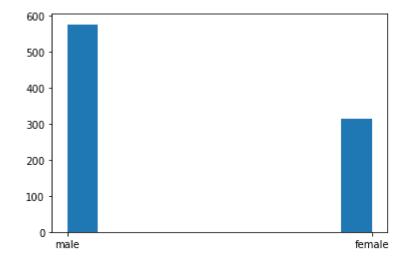




Out[26]: <AxesSubplot:xlabel='Sex', ylabel='count'>



```
In [27]: histogram(df['Sex'])
```

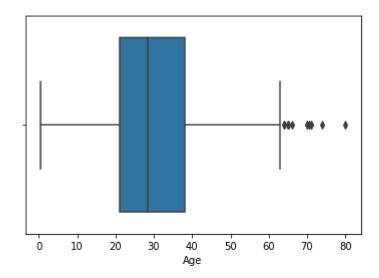


OBSERVATION: MALES > FEMALES (ALMOST DOUBLE)

2. AGE

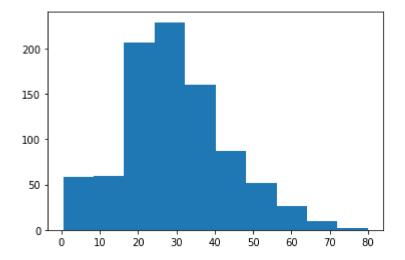
```
In [29]: boxplot(df['Age'])
```

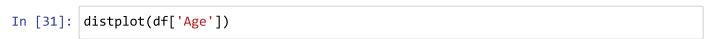
Out[29]: <AxesSubplot:xlabel='Age'>



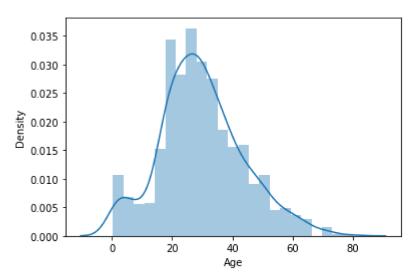
```
In [30]: histogram(df['Age'])
```

```
Out[30]: (array([ 58., 60., 207., 229., 160., 87., 52., 26., 10., 2.]),
array([ 0.42 , 8.378, 16.336, 24.294, 32.252, 40.21 , 48.168, 56.126,
64.084, 72.042, 80. ]),
<BarContainer object of 10 artists>)
```





Out[31]: <AxesSubplot:xlabel='Age', ylabel='Density'>

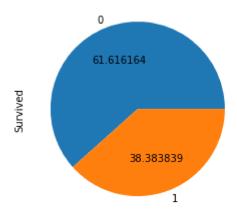


OBSERVATION: In titanic most of the people were in the range 20-40 and very few people were above 65. so age>65 is considered as outliers

3. Survived

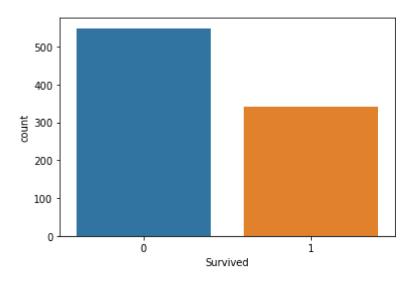
```
In [32]: piechart(df['Survived'])
```

Out[32]: <AxesSubplot:ylabel='Survived'>

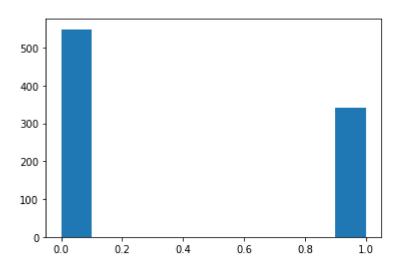


In [33]: countplot(df['Survived'])

Out[33]: <AxesSubplot:xlabel='Survived', ylabel='count'>

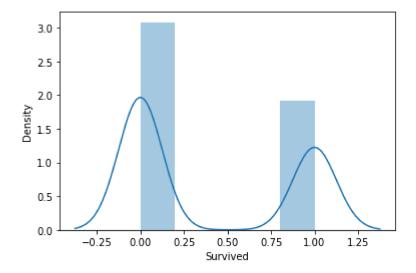


In [34]: histogram(df['Survived'])



```
In [35]: distplot(df['Survived'])
```

Out[35]: <AxesSubplot:xlabel='Survived', ylabel='Density'>

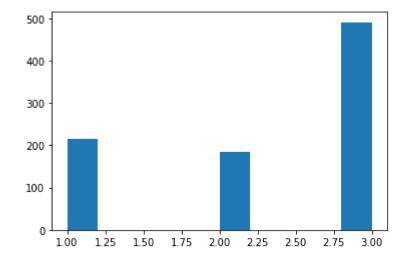


OBSERVATION: More people died and very few ratio of people were able to save themselves.

4. Pclass

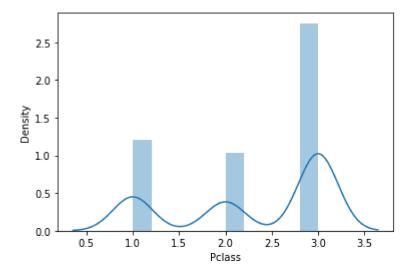
```
In [36]: histogram(df['Pclass'])
```

Out[36]: (array([216., 0., 0., 0., 0., 184., 0., 0., 0., 491.]), array([1., 1.2, 1.4, 1.6, 1.8, 2., 2.2, 2.4, 2.6, 2.8, 3.]), <BarContainer object of 10 artists>)



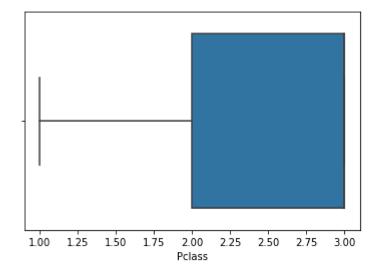
```
In [37]: distplot(df['Pclass'])
```

Out[37]: <AxesSubplot:xlabel='Pclass', ylabel='Density'>



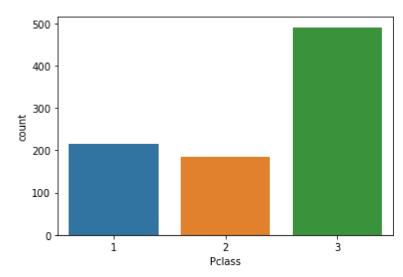
```
In [38]: boxplot(df['Pclass'])
```

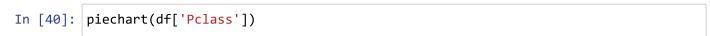
Out[38]: <AxesSubplot:xlabel='Pclass'>



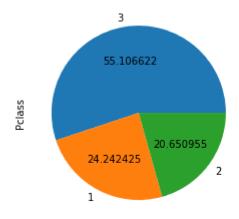


Out[39]: <AxesSubplot:xlabel='Pclass', ylabel='count'>





Out[40]: <AxesSubplot:ylabel='Pclass'>

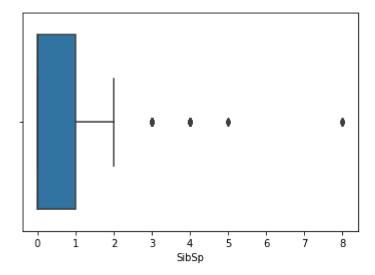


OBSERVATION: MOST OF THE PEOPLE BELONGS TO 3RD CLASS IN TITANIC.

4. SibSp

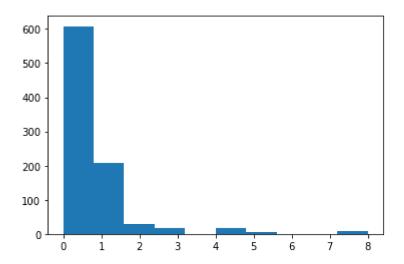
```
In [41]: boxplot(df['SibSp'])
```

Out[41]: <AxesSubplot:xlabel='SibSp'>



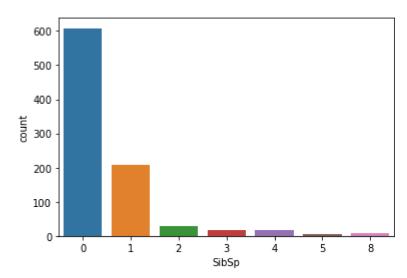
```
In [42]: histogram(df['SibSp'])
```

Out[42]: (array([608., 209., 28., 16., 0., 18., 5., 0., 0., 7.]), array([0., 0.8, 1.6, 2.4, 3.2, 4., 4.8, 5.6, 6.4, 7.2, 8.]), <BarContainer object of 10 artists>)



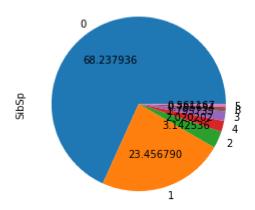
In [43]: countplot(df['SibSp'])

Out[43]: <AxesSubplot:xlabel='SibSp', ylabel='count'>



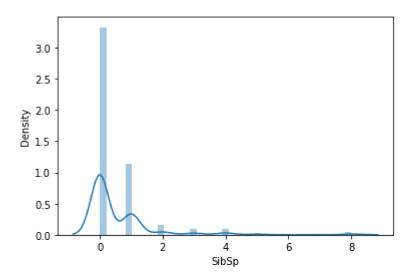
```
In [44]: piechart(df['SibSp'])
```

Out[44]: <AxesSubplot:ylabel='SibSp'>





Out[45]: <AxesSubplot:xlabel='SibSp', ylabel='Density'>

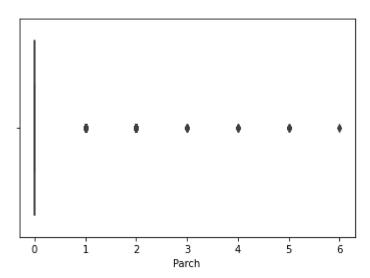


OBSERVATION: Most of the people who were in titanic were either alone or have 1 sibling along with them.

4. Parch

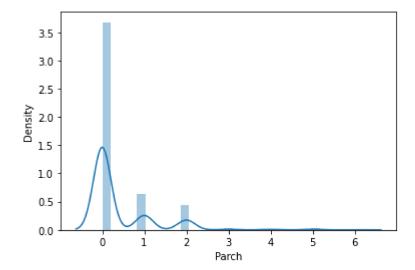


Out[46]: <AxesSubplot:xlabel='Parch'>

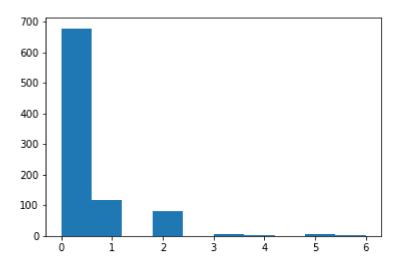


```
In [47]: distplot(df['Parch'])
```

Out[47]: <AxesSubplot:xlabel='Parch', ylabel='Density'>

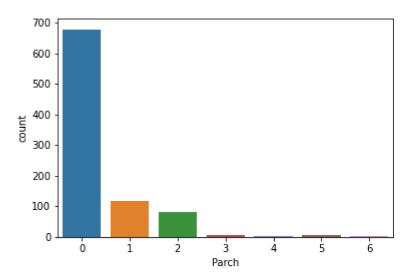


```
In [48]: histogram(df['Parch'])
```



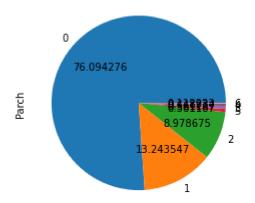
In [49]: countplot(df['Parch'])

Out[49]: <AxesSubplot:xlabel='Parch', ylabel='count'>



```
In [50]: piechart(df['Parch'])
```

Out[50]: <AxesSubplot:ylabel='Parch'>

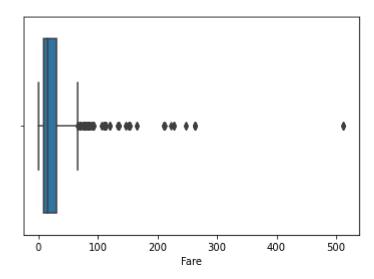


OBSERVATION: Only 24% of the people were there with their children in titanic .

5. 'Fare'

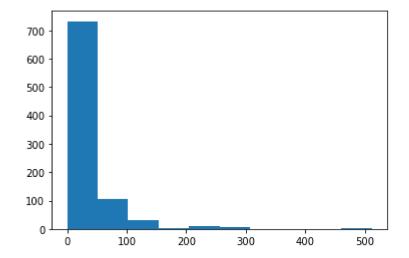
```
In [51]: boxplot(df['Fare'])
```

Out[51]: <AxesSubplot:xlabel='Fare'>



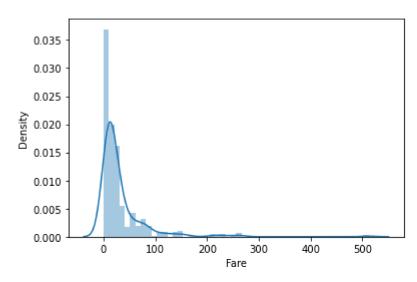
```
In [52]: histogram(df['Fare'])
```

Out[52]: (array([732., 106., 31., 2., 11., 6., 0., 0., 0., 0., 3.]), array([0. , 51.23292, 102.46584, 153.69876, 204.93168, 256.1646, 307.39752, 358.63044, 409.86336, 461.09628, 512.3292]), <BarContainer object of 10 artists>)



```
In [53]: distplot(df['Fare'])
```

Out[53]: <AxesSubplot:xlabel='Fare', ylabel='Density'>



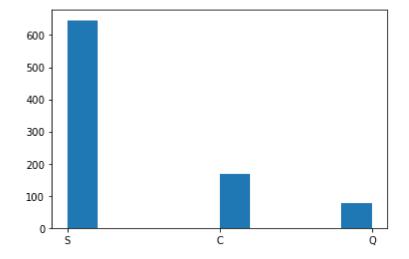
```
In [54]: df['Fare'].max()
```

Out[54]: 512.3292

OBSERVATION : FARE OF TITANIC MAINLY LIES IN THE RANGE 0-100 .FARE ABOVE 100 WAS CONSIDERED AS OUTLIER (MAX FARE- 512.3)

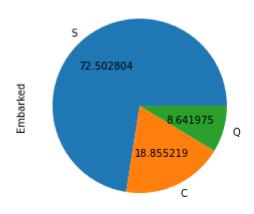
6. 'Embarked'

```
In [55]: histogram(df['Embarked'])
```



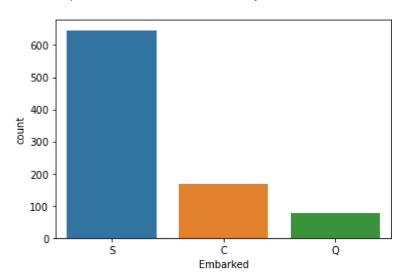
```
In [56]: piechart(df['Embarked'])
```

Out[56]: <AxesSubplot:ylabel='Embarked'>





Out[57]: <AxesSubplot:xlabel='Embarked', ylabel='count'>



OBSERVATION: S>C (ALMOST 4 TIMES) AND S>Q (ALMOST 9 TIMES)

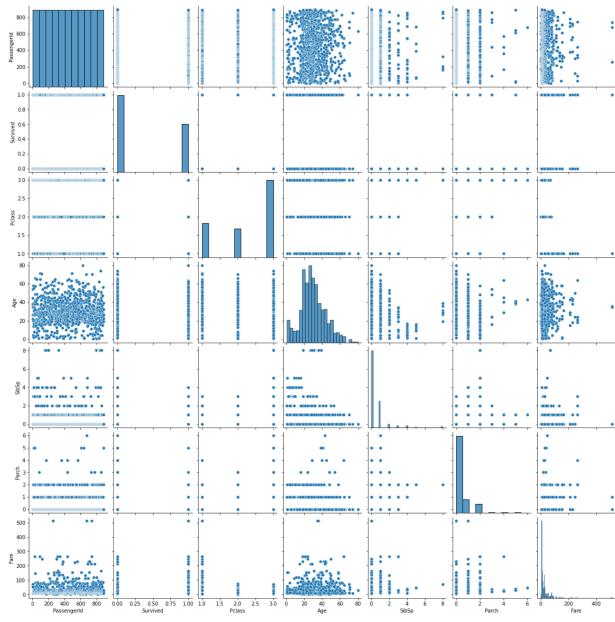
```
In [58]: df.hist(figsize=(15,15))
Out[58]: array([[<AxesSubplot:title={'center':'PassengerId'}>,
                         <AxesSubplot:title={'center':'Survived'}>,
                         <AxesSubplot:title={'center':'Pclass'}>],
                        [<AxesSubplot:title={'center':'Age'}>,
                        <AxesSubplot:title={'center':'SibSp'}>,
  <AxesSubplot:title={'center':'Parch'}>],
[<AxesSubplot:title={'center':'Fare'}>, <AxesSubplot:>,
                         <AxesSubplot:>]], dtype=object)
                            PassengerId
                                                                      Survived
                                                                                                               Pclass
                                                                                               500
                                                       500
                                                                                               400
                                                       400
               60
                                                                                               300
                                                       300
                                                                                               200
                                                       200
               20
                                                                                               100
                                                       100
                        200
                               400
                                     600
                                                                     0.4
                                                                           0.6
                                                                                 0.8
                                                                                                         1.5
                                                                                                                       2.5
                                                                       SibSp
                                                                                               700
                                                       600
                                                                                               600
               200
                                                       500
                                                                                               500
              150
                                                       400
                                                                                               400
                                                       300
              100
                                                                                               300
                                                                                               200
               50
                                                       100
                                Fare
               700
               600
               500
               400
               300
              100
```

BIVARIATE / MULTIVARIATE ANALYSIS

```
In [59]: def scatterplot(x,y):
             return sns.scatterplot(x,y)
                                            #num-num
         def barplot(x,y):
             return sns.barplot(x,y)
                                            #num-cat
         def distplot(x,y):
             return sns.distplot(x,y)
                                           #num-cat
         def heatmap(x,y):
             return sns.heatmap(x,y,annot=True)
                                                     #cat-cat
         def lineplot(x,y):
             return sns.lineplot(x,y)
         def clustermap(x,y):
             return sns.clustermap(pd.crosstab(x,y),annot=True) #cat-cat
         def boxplot(x,y):
             return sns.boxplot(x,y)
```

```
In [60]: sns.pairplot(df)
```

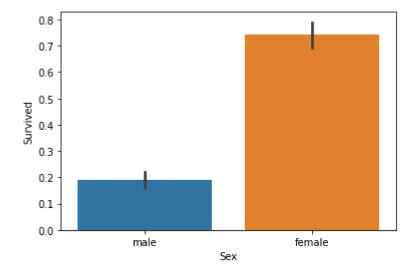
Out[60]: <seaborn.axisgrid.PairGrid at 0x1b4e818c880>



```
In [61]: df.select_dtypes(include='object').columns
Out[61]: Index(['Name', 'Sex', 'Ticket', 'Cabin', 'Embarked'], dtype='object')
In [62]: df.select_dtypes(include=['int64','float64']).columns
Out[62]: Index(['PassengerId', 'Survived', 'Pclass', 'Age', 'SibSp', 'Parch', 'Fare'], d type='object')
```

```
In [63]: barplot(df['Sex'],df['Survived'])
```

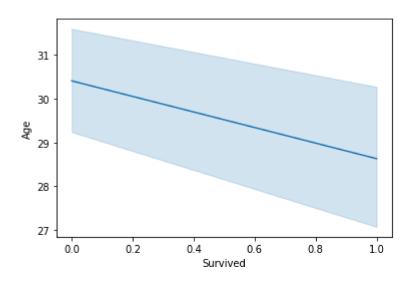
Out[63]: <AxesSubplot:xlabel='Sex', ylabel='Survived'>



OBSERVATION: SURVIVAL RATE OF FEMALE IS MORE THAN MALE.



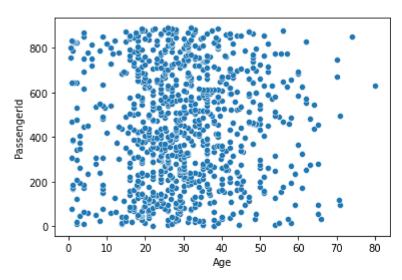
Out[64]: <AxesSubplot:xlabel='Survived', ylabel='Age'>



OBSERVATION: SURVIVAL RATE OF YOUNGER PEOPLE IS MORE THAN OLDER CROWD.



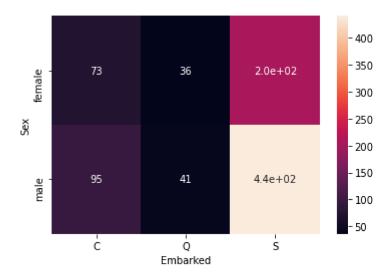
Out[65]: <AxesSubplot:xlabel='Age', ylabel='PassengerId'>



OBSERVATION: people of age 20-50 were travelling more.

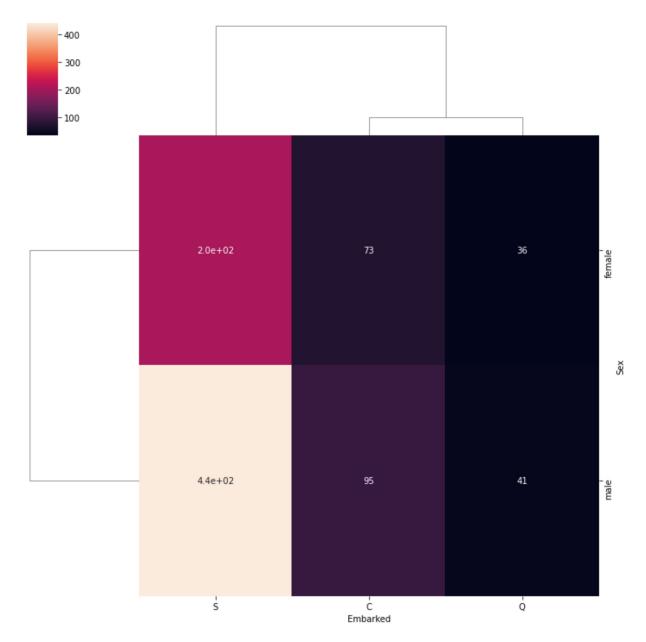
In [66]: sns.heatmap(pd.crosstab(df['Sex'],df['Embarked']),annot=True)

Out[66]: <AxesSubplot:xlabel='Embarked', ylabel='Sex'>



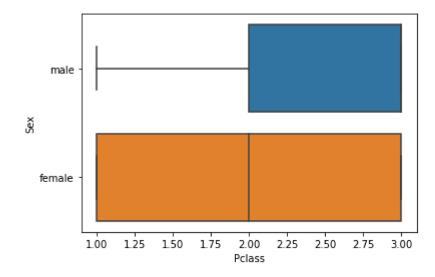
In [67]: clustermap(df['Sex'],df['Embarked'])

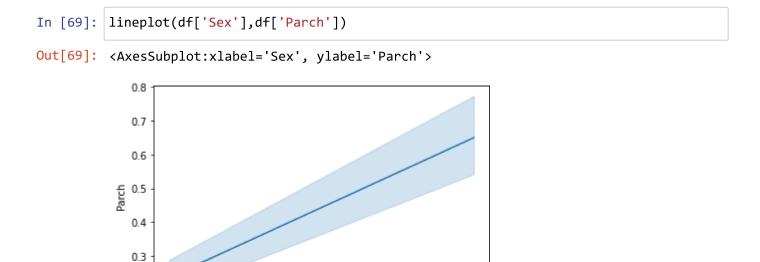
Out[67]: <seaborn.matrix.ClusterGrid at 0x1b4eb3c8fa0>



```
In [68]: boxplot(df['Pclass'],df['Sex'])
```

Out[68]: <AxesSubplot:xlabel='Pclass', ylabel='Sex'>





Observation: on an average, there are more females than males along with their child .

Sex

female

0.2

male