## TITANIC DATA CLEANING AND ANALYSIS

import csv import pan

import pandas as pd
import statistics as stat
from scipy.stats import norm
import seaborn as sea

import seaborn as sns

 $\textbf{import} \ \texttt{matplotlib.pyplot} \ \textbf{as} \ \texttt{plt}$ 

%matplotlib inline

In [2]: data=pd.read\_csv('titanic1.csv')

In [5]:

data.head(8)

**Parch** Out[5]: Passengerld Pclass Name Sex Age SibSp **Ticket Fare** Cabin **Embarked** 0 3 0 330911 7.8292 892 Kelly, Mr. James 34.5 0 NaN Q male Wilkes, Mrs. James 1 3 893 female 47.0 1 363272 7.0000 NaN S (Ellen Needs) Myles, Mr. Thomas 2 2 0 240276 894 male 62.0 0 9.6875 NaN Q Francis 3 895 3 Wirz, Mr. Albert 27.0 0 0 315154 8.6625 S male NaN Hirvonen, Mrs. 3 4 896 Alexander (Helga E female 22.0 1 3101298 12.2875 NaN S Lindqvist) Svensson, Mr. Johan 5 897 3 0 0 7538 9.2250 S male 14.0 NaN Cervin 6 898 3 0 0 330972 7.6292 Connolly, Miss. Kate female 30.0 NaN Q Caldwell, Mr. Albert male 26.0 7 899 2 1 248738 29.0000 NaN S

Francis

In [6]:

data.tail(5)

Passengerld **Pclass** SibSp Parch **Ticket** Cabin **Embarked** Out[6]: Name Sex Age Fare Spector, Mr. 413 1305 3 male NaN 0 0 A.5. 3236 8.0500 S NaN Woolf Oliva y Ocana, 0 0 С 414 1306 1 female 39.0 PC 17758 108.9000 C105 Dona. Fermina Saether, Mr. SOTON/O.Q. 415 1307 3 Simon male 38.5 0 7.2500 NaN S 3101262 Sivertsen Ware, Mr. 416 1308 3 male NaN 0 0 359309 8.0500 NaN S Frederick Peter, 3 С 417 1309 1 1 2668 22.3583 Master. male NaN NaN Michael J

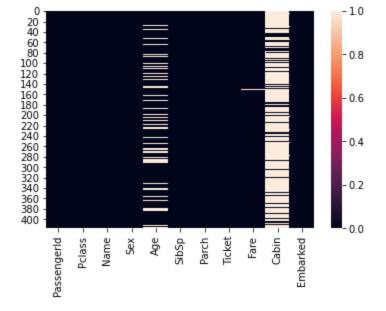
In [7]:

data.shape

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```
Out[7]: (418, 11)
 In [8]:
          print("Number of Columns", data.shape[1])
         print("Number of rows", data.shape[0])
         Number of Columns 11
         Number of rows 418
 In [9]:
         data.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 418 entries, 0 to 417
         Data columns (total 11 columns):
              Column
                           Non-Null Count Dtype
          0
              PassengerId 418 non-null
                                          int64
          1
              Pclass
                           418 non-null
                                          int64
          2
              Name
                           418 non-null
                                           object
          3
              Sex
                           418 non-null
                                           object
          4
              Age
                           332 non-null
                                          float64
          5
              SibSp
                          418 non-null
                                          int64
          6
              Parch
                           418 non-null
                                          int64
          7
              Ticket
                           418 non-null
                                          object
          8
                           417 non-null
                                          float64
              Fare
          9
              Cabin
                           91 non-null
                                           object
          10 Embarked
                          418 non-null
                                           object
         dtypes: float64(2), int64(4), object(5)
         memory usage: 36.0+ KB
In [11]:
          print("missing Value? ", data.isnull().values.any())
         missing Value? True
In [12]:
          data.isnull().sum()
Out[12]:
         PassengerId
         Pclass
                          0
         Name
                          0
         Sex
                          0
         Age
                         86
         SibSp
                          0
         Parch
                          0
         Ticket
                         0
         Fare
                          1
         Cabin
                        327
         Embarked
                          0
         dtype: int64
In [13]:
          sns.heatmap(data.isnull())
```

Out[13]: <AxesSubplot:>



In [14]: per=data.isnull().sum()\*100/len(data)

In [15]: print(per)

PassengerId 0.00000 Pclass 0.00000 0.00000 Name Sex 0.00000 20.574163 Age SibSp 0.00000 Parch 0.00000 Ticket 0.00000 Fare 0.239234 Cabin 78.229665 Embarked 0.000000

dtype: float64

In [16]: data.dropna(axis=0)

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Out[16]:		Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
	12	904	1	Snyder, Mrs. John Pillsbury (Nelle Stevenson)	female	23.0	1	0	21228	82.2667	B45	S
	14	906	1	Chaffee, Mrs. Herbert Fuller (Carrie Constance	female	47.0	1	0	W.E.P. 5734	61.1750	E31	S
	24	916	1	Ryerson, Mrs. Arthur Larned (Emily Maria Borie)	female	48.0	1	3	PC 17608	262.3750	B57 B59 B63 B66	С
	26	918	1	Ostby, Miss. Helene Ragnhild	female	22.0	0	1	113509	61.9792	B36	С
	28	920	1	Brady, Mr. John Bertram	male	41.0	0	0	113054	30.5000	A21	S
	404	1296	1	Frauenthal, Mr.	male	43.0	1	0	17765	27.7208	D40	С

Isaac Gerald

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
405	1297	2	Nourney, Mr. Alfred (Baron von Drachstedt")"	male	20.0	0	0	SC/PARIS 2166	13.8625	D38	С
407	1299	1	Widener, Mr. George Dunton	male	50.0	1	1	113503	211.5000	C80	С
411	1303	1	Minahan, Mrs. William Edward (Lillian E Thorpe)	female	37.0	1	0	19928	90.0000	C78	Q
414	1306	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	C105	С

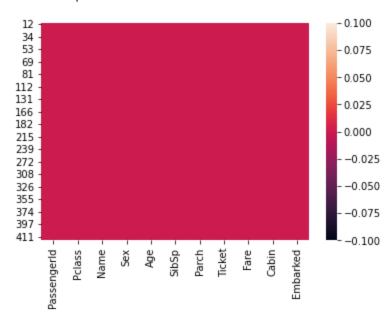
87 rows × 11 columns

```
In [17]:
    data.dropna(inplace=True)
    data.isnull().sum()
```

PassengerId 0 Out[17]: **Pclass** 0 Name 0 Sex 0 0 Age 0 SibSp Parch 0 Ticket Fare 0 Cabin 0 Embarked 0 dtype: int64

```
In [18]: sns.heatmap(data.isnull())
```

## Out[18]: <AxesSubplot:>



```
In [19]: dup=data.duplicated().any()
    print(dup)
```

False

```
PassengerId
                                   Pclass
                                                Age
                                                          SibSp
                                                                     Parch
                                                                                  Fare
Out[20]:
            count
                     87.000000
                                87.000000
                                           87.000000
                                                      87.000000
                                                                 87.000000
                                                                             87.000000
                                           39.247126
                                                       0.597701
                                                                  0.482759
            mean
                   1102.712644
                                 1.137931
                                                                             98.109198
                    126.751901
                                 0.435954
                                           15.218730
                                                       0.637214
                                                                  0.860801
                                                                             88.177319
              std
             min
                    904.000000
                                 1.000000
                                            1.000000
                                                       0.000000
                                                                  0.000000
                                                                              0.000000
             25%
                    986.000000
                                 1.000000
                                           27.000000
                                                       0.000000
                                                                  0.000000
                                                                             35.339600
             50%
                   1094.000000
                                 1.000000
                                           39.000000
                                                       1.000000
                                                                  0.000000
                                                                             71.283300
             75%
                   1216.000000
                                 1.000000
                                           50.000000
                                                       1.000000
                                                                  1.000000
                                                                            135.066650
             max
                   1306.000000
                                 3.000000
                                           76.000000
                                                       3.000000
                                                                  4.000000
                                                                            512.329200
In [21]:
            data.describe(include='all')
                    PassengerId
                                    Pclass
                                                Name
                                                         Sex
                                                                    Age
                                                                             SibSp
                                                                                        Parch
                                                                                               Ticket
                                                                                                             Fare
                                                                                                                   Cabin
                                                                                                                          Er
Out[21]:
            count
                      87.000000
                                 87.000000
                                                   87
                                                               87.000000
                                                                          87.000000
                                                                                     87.000000
                                                                                                   87
                                                                                                        87.000000
                                                                                                                      87
                                      NaN
                                                   87
                                                            2
                                                                                                                       72
           unique
                           NaN
                                                                    NaN
                                                                               NaN
                                                                                          NaN
                                                                                                   66
                                                                                                             NaN
                                            Earnshaw,
                                                                                                                     B57
                                                 Mrs.
                                                                                                   PC
                                                                                                                     B59
               top
                           NaN
                                      NaN
                                               Boulton
                                                       female
                                                                    NaN
                                                                               NaN
                                                                                          NaN
                                                                                                             NaN
                                                                                                17608
                                                                                                                     B63
                                                (Olive
                                                                                                                     B66
                                               Potter)
                           NaN
                                      NaN
                                                    1
                                                          44
                                                                    NaN
                                                                               NaN
                                                                                          NaN
                                                                                                    4
                                                                                                              NaN
                                                                                                                        3
              freq
                    1102.712644
                                  1.137931
                                                 NaN
                                                               39.247126
                                                                           0.597701
                                                                                      0.482759
                                                                                                  NaN
                                                                                                        98.109198
                                                                                                                     NaN
             mean
                                                         NaN
               std
                     126.751901
                                  0.435954
                                                 NaN
                                                         NaN
                                                               15.218730
                                                                           0.637214
                                                                                      0.860801
                                                                                                  NaN
                                                                                                        88.177319
                                                                                                                     NaN
              min
                     904.000000
                                  1.000000
                                                 NaN
                                                         NaN
                                                                1.000000
                                                                           0.000000
                                                                                      0.000000
                                                                                                 NaN
                                                                                                         0.000000
                                                                                                                     NaN
              25%
                     986.000000
                                  1.000000
                                                 NaN
                                                         NaN
                                                               27.000000
                                                                           0.000000
                                                                                      0.000000
                                                                                                  NaN
                                                                                                        35.339600
                                                                                                                     NaN
                    1094.000000
                                  1.000000
              50%
                                                 NaN
                                                         NaN
                                                               39.000000
                                                                           1.000000
                                                                                      0.000000
                                                                                                  NaN
                                                                                                        71.283300
                                                                                                                     NaN
              75%
                    1216.000000
                                                               50.000000
                                                                           1.000000
                                                                                      1.000000
                                                                                                       135.066650
                                  1.000000
                                                 NaN
                                                         NaN
                                                                                                  NaN
                                                                                                                     NaN
                    1306.000000
                                  3.000000
                                                              76.000000
                                                                           3.000000
                                                                                      4.000000
                                                                                                       512.329200
              max
                                                 NaN
                                                         NaN
                                                                                                 NaN
                                                                                                                     NaN
In [22]:
            data.columns
            data.groupby('Age')['Fare'].mean()
           Age
Out[22]:
           1.0
                      16.700000
           6.0
                     134.500000
           12.0
                      39.000000
           13.0
                     262.375000
           18.0
                      56.550000
           18.5
                      13.000000
           20.0
                      13.862500
           22.0
                       36.239600
           23.0
                       86.308333
           24.0
                      71.133350
           25.0
                       23.440300
           26.0
                       74.889600
           27.0
                     145.433333
```

In [20]:

data.describe()

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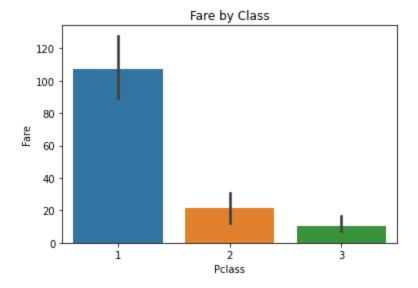
```
28.5
                     27.720800
            29.0
                    221.779200
            30.0
                    100.041675
            31.0
                     81.518750
            32.5
                    211.500000
            33.0
                     27.720800
            35.0
                    134.625000
            36.0
                    102.073975
            37.0
                     86.579150
            39.0
                     69.961100
            41.0
                     41.181250
            42.0
                     34.525000
            43.0
                     41.581250
            45.0
                     70.028125
            46.0
                     75.241700
            47.0
                    144.350000
            48.0
                    124.623950
            49.0
                       0.000000
            50.0
                    149.666667
            51.0
                     39.400000
            53.0
                     55.179150
            54.0
                     68.650000
            55.0
                     76.208325
            57.0
                    146.520800
            58.0
                    512.329200
            59.0
                     51.479200
            60.0
                    169.645850
            61.0
                    262.375000
            63.0
                    221.779200
            64.0
                     61.652767
            67.0
                    221.779200
            76.0
                      78.850000
            Name: Fare, dtype: float64
 In [23]:
            data.groupby('Age')['Fare'].mean().sort_values(ascending=False)
 Out[23]: Age
            58.0
                    512.329200
            28.0
                    263.000000
            13.0
                    262.375000
            61.0
                    262.375000
            29.0
                    221,779200
            67.0
                    221.779200
            63.0
                    221.779200
            32.5
                    211.500000
                    169.645850
            60.0
            50.0
                    149.666667
            57.0
                    146.520800
            27.0
                    145.433333
            47.0
                    144.350000
                    134.625000
            35.0
            6.0
                    134.500000
            48.0
                    124.623950
            36.0
                    102.073975
            30.0
                    100.041675
            37.0
                     86.579150
            23.0
                     86.308333
            31.0
                     81.518750
            76.0
                     78.850000
            55.0
                     76.208325
            46.0
                     75.241700
            26.0
                     74.889600
            24.0
                     71.133350
            45.0
                      70.028125
            39.0
                     69.961100
            54.0
                      68.650000
            64.0
                     61.652767
            18.0
                     56.550000
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```

```
59.0
         51.479200
43.0
         41.581250
41.0
         41.181250
51.0
         39.400000
12.0
         39.000000
22.0
         36.239600
42.0
         34.525000
33.0
         27.720800
28.5
         27.720800
25.0
         23.440300
1.0
         16.700000
20.0
         13.862500
18.5
         13.000000
          0.000000
49.0
```

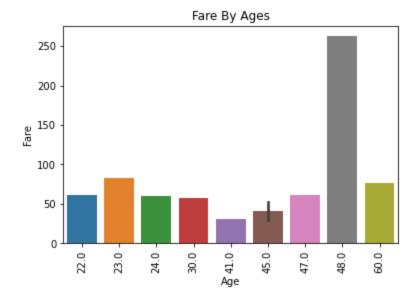
Name: Fare, dtype: float64

```
sns.barplot(x='Pclass',y='Fare',data=data)
plt.title("Fare by Class")
plt.show
```

Out[24]: <function matplotlib.pyplot.show(close=None, block=None)>



```
sns.barplot(x='Age', y='Fare', data=data.head(10))
plt.title("Fare By Ages")
plt.xticks(rotation=90)
plt.show()
```

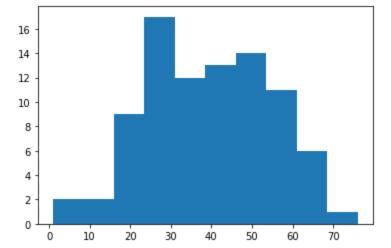


```
In [43]:
          data.hist(figsize=(10,10),color='orange')
         array([[<AxesSubplot:title={'center':'PassengerId'}>,
                  <AxesSubplot:title={'center':'Pclass'}>],
                 [<AxesSubplot:title={'center':'Age'}>,
                  <AxesSubplot:title={'center':'SibSp'}>],
                 PassengerId
                                                     80
          12
          10
                                                     60
           8
                                                     40
           6
           4
                                                     20
           2
           0
                                                     0
                           1100
                                  1200
            900
                   1000
                                          1300
                                                        1.0
                                                               1.5
                                                                       2.0
                                                                              2.5
                                                                                      3.0
                           Age
                                                                      SibSp
                                                     40
          15
                                                     30
          10
                                                     20
           5
                                                     10
           0
                                                     0
                     20
                             40
                                     60
                                                                            2.0
                                                        0.0
                                                             0.5
                                                                  1.0
                                                                       1.5
                                                                                 2.5
                                                                                      3.0
                           Parch
                                                                      Fare
          60
                                                     30
          50
                                                     25
          40
                                                     20
          30
                                                     15
          20
                                                     10
          10
                                                      5
                                                     0
           0
                                                             100
                                                                   200
                                                                         300
                                                                               400
                                                                                     500
In [27]:
          age=data['Age']
          fare=data['Fare']
          pc=data['Pclass']
          plt.hist(age)
```

(array([ 2., 2., 9., 17., 12., 13., 14., 11., 6., 1.]),

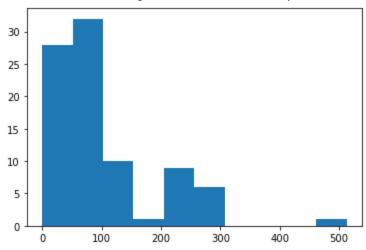
array([ 1. , 8.5, 16. , 23.5, 31. , 38.5, 46. , 53.5, 61. , 68.5, 76. ]), <BarContainer object of 10 artists>)

Out[27]:



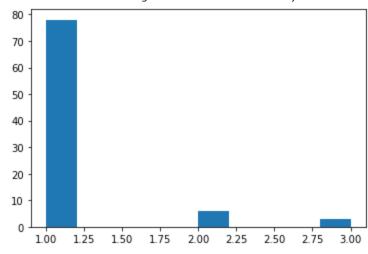
```
In [28]: plt.hist(fare)
```

Out[28]: (array([28., 32., 10., 1., 9., 6., 0., 0., 0., 1.]), 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 [30]: plt.hist(pc)
```

Out[30]: (array([78., 0., 0., 0., 0., 6., 0., 0., 0., 3.]), 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>)

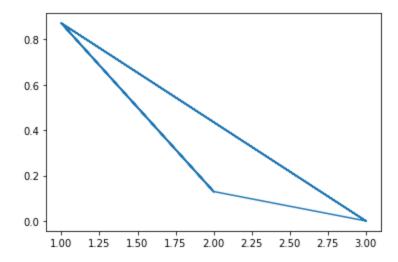


```
In [31]: age_mean=stat.mean(age)

Loading [MathJax]/extensions/Safe.js | t.mean(fare)
```

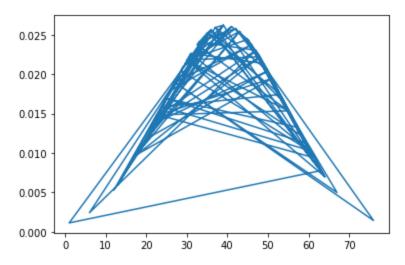
```
class_mean=stat.mean(pc)
age_st=stat.stdev(age)
fare_st=stat.stdev(fare)
class_st=stat.stdev(pc)
plt.plot(pc,norm.pdf(pc,class_mean,class_st))
```

Out[31]: [<matplotlib.lines.Line2D at 0x248279c9160>]



In [32]: plt.plot(age,norm.pdf(age,age\_mean,age\_st))

Out[32]: [<matplotlib.lines.Line2D at 0x24827a1cbb0>]



```
In [33]: plt.plot(fare, norm.pdf(fare, fare_mean, fare_st))
```

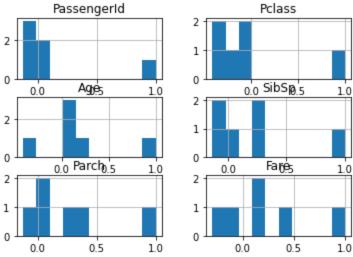
```
0.004 - 0.003 - 0.002 - 0.001 - 0.000 - 0 100 200 300 400 500
```

```
In [34]:
          data_cor=data.corr()
          print(data_cor)
                       PassengerId
                                      Pclass
                                                    Age
                                                            SibSp
                                                                      Parch
         PassengerId
                          1.000000
                                    0.004934
                                              0.055488 -0.087828 -0.122551 -0.097346
         Pclass
                          0.004934
                                    1.000000 -0.410924 -0.132790
                                                                   0.006411 -0.298186
                                              1.000000 0.062530
         Age
                          0.055488 -0.410924
                                                                   0.051144
                                                                             0.180567
         SibSp
                         -0.087828 -0.132790
                                              0.062530
                                                        1.000000
                                                                   0.252194
                                                                             0.213014
         Parch
                         -0.122551 0.006411
                                              0.051144 0.252194
                                                                   1.000000
                                                                             0.395685
         Fare
                         -0.097346 -0.298186
                                              0.180567
                                                        0.213014
                                                                   0.395685
                                                                             1.000000
In [35]:
          plt.hist(data_cor)
Out[35]: (array([[0., 0., 4., 1., 0., 0., 0., 0., 0., 1.],
                  [2., 1., 2., 0., 0., 0., 0., 0., 0., 1.],
                  [1., 0., 0., 3., 1., 0., 0., 0., 0., 1.],
                  [0., 1., 1., 1., 2., 0., 0., 0., 0., 1.],
                  [0., 0., 2., 1., 1., 1., 0., 0., 0., 1.],
                  [1., 0., 1., 0., 2., 1., 0., 0., 0., 1.]]),
          array([-0.41092369, -0.26983132, -0.12873895, 0.01235341,
                                                                        0.15344578,
                   0.29453815, 0.43563052, 0.57672289, 0.71781526, 0.85890763,
                   1.
          <a list of 6 BarContainer objects>)
          4.0
          3.5
          3.0
          2.5
          2.0
          1.5
          1.0
          0.5
          0.0
             -0.4
                                           0.6
                                                 0.8
```

In [36]:

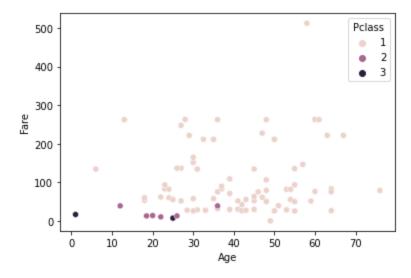
data\_cor.hist()

[<AxesSubplot:title={'center':'Parch'}>,
 <AxesSubplot:title={'center':'Fare'}>]], dtype=object)



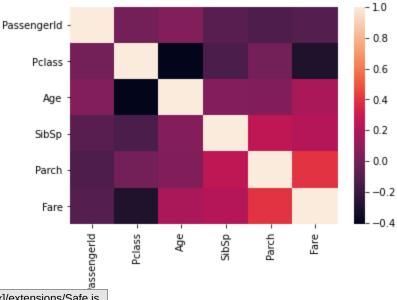
In [37]: sns.scatterplot(x='Age', y='Fare', data=data, hue="Pclass")

Out[37]: <AxesSubplot:xlabel='Age', ylabel='Fare'>



In [38]: sns.heatmap(data\_cor)

Out[38]: <AxesSubplot:>



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In [ ]:			