In [1]: 1 import numpy as np, pandas as pd, matplotlib.pyplot as plt, seaborn as sns

1 Reading the Data Set

In [2]: 1 ds=pd.read\_csv("Titanic\_Dataset.csv")

In [3]:

1 ds

Out[3]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148
890	891	0	3	Dooley, Mr. Patrick	ma <b>l</b> e	32.0	0	0	370376	7.7500	NaN

891 rows × 12 columns

4

In [4]: 1 ds.head()

Out[4]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	En
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	
4												•

In [5]:

1 ds.describe()

## Out[5]:

	Passengerld	Survived	Pclass	Age	SibSp	Parch	Fare
count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000
mean	446.000000	0.383838	2.308642	29.699118	29.699118 0.523008 0.381		32.204208
std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429
min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400
50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000
max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200

In [6]:

1 ds.shape

Out[6]: (891, 12)

object

2 **Pclass** 891 non-null int64 891 non-null 3 Name object 4 Sex 891 non-null object 5 float64 Age 714 non-null 6 int64 891 non-null SibSp 7 Parch 891 non-null int64 8 Ticket 891 non-null object 9 float64 891 non-null Fare

11 Embarked 889 non-null object dtypes: float64(2), int64(5), object(5)

204 non-null

memory usage: 83.7+ KB

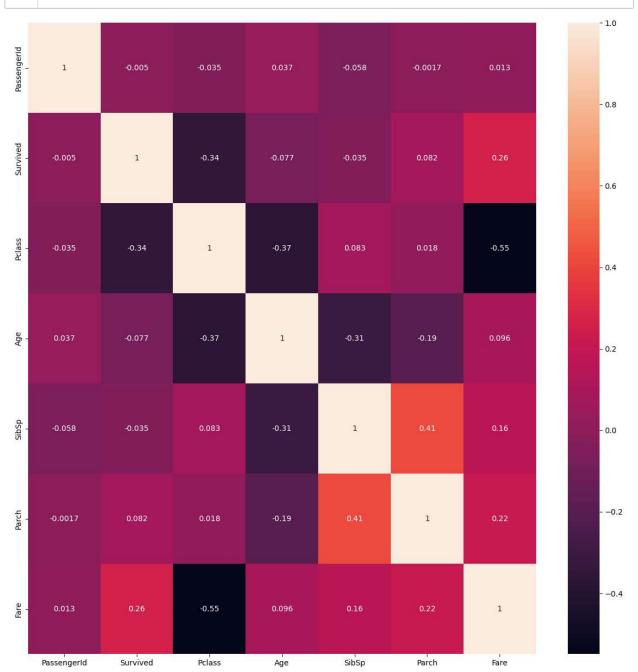
```
In [8]: 1 corr=ds.corr()
```

Cabin

10

In [25]:

```
plt.subplots(figsize=(15,15))
sns.heatmap(corr,annot=True);
```

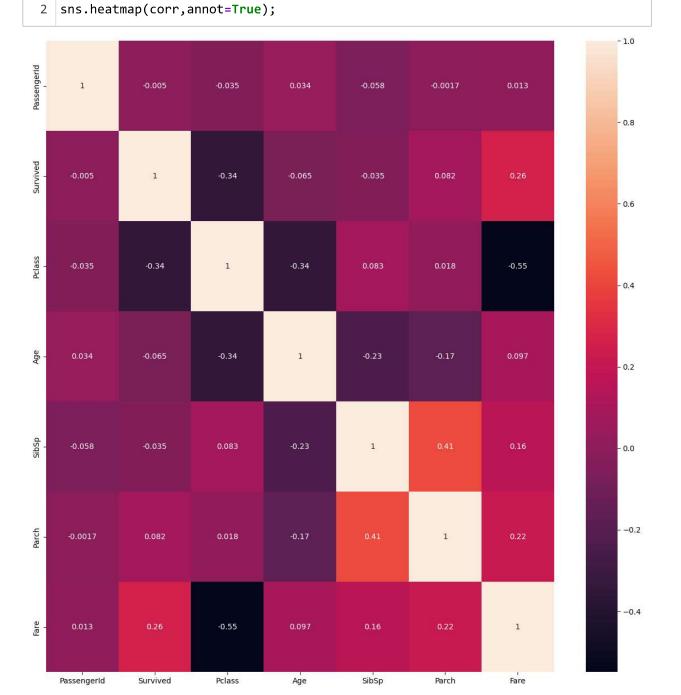


1 Checking the null values

```
1 ds.isnull().any()
In [10]:
Out[10]: PassengerId
                         False
          Survived
                         False
          Pclass
                         False
          Name
                         False
                         False
          Sex
          Age
                          True
                         False
          SibSp
                         False
          Parch
          Ticket
                         False
          Fare
                         False
          Cabin
                          True
                          True
          Embarked
          dtype: bool
In [11]:
              ds.isna().sum()
Out[11]: PassengerId
                           0
          Survived
                           0
          Pclass
                           0
          Name
                           0
          Sex
                           0
          Age
                         177
          SibSp
                           0
          Parch
                           0
          Ticket
                           0
          Fare
                           0
          Cabin
                         687
          Embarked
                           2
          dtype: int64
In [12]:
              ds.Embarked.value counts()
Out[12]: S
               644
          C
               168
                77
          Name: Embarked, dtype: int64
In [13]:
              ds.Age.mean()
Out[13]: 29.69911764705882
              ds.Age.median()
In [14]:
Out[14]: 28.0
In [15]:
              ds.Age.mode()
Out[15]: 0
               24.0
          Name: Age, dtype: float64
```

```
1 Filling the null values
In [16]:
              ds.Age.fillna(ds.Age.median(),inplace=True)
In [17]:
              ds.Age.isna().sum()
Out[17]: 0
In [18]:
              ds.Embarked.mode()
Out[18]: 0
               S
          Name: Embarked, dtype: object
In [19]:
              ds.Embarked.fillna(ds.Embarked.mode()[0],inplace=True)
In [20]:
              ds.Embarked.isna().sum()
Out[20]: 0
In [23]:
              ds.drop(['Cabin'],axis=1,inplace=True)
In [24]:
              ds.isnull().any()
Out[24]: PassengerId
                         False
          Survived
                         False
          Pclass
                         False
          Name
                         False
          Sex
                         False
          Age
                         False
          SibSp
                         False
          Parch
                         False
         Ticket
                         False
          Fare
                         False
          Embarked
                         False
          dtype: bool
 In [ ]:
              Data Visualization
In [26]:
              corr=ds.corr()
```

In [27]: 1 plt.subplots(figsize=(15,15))



1 Encoding

In [28]: 1 from sklearn.preprocessing import LabelEncoder

In [29]: 1 l=LabelEncoder()

In [30]: 1 ds["Sex"]=1.fit\_transform(ds["Sex"])

```
ds.Sex
In [33]:
Out[33]: 0
                 1
                 0
          2
                 0
          3
                 0
          4
                 1
          886
                 1
          887
                 0
          888
                 0
          889
                 1
          890
                 1
          Name: Sex, Length: 891, dtype: int32
In [34]:
              ds.Sex.value_counts()
Out[34]: 1
               577
               314
          Name: Sex, dtype: int64
In [37]:
              ds.shape
Out[37]: (891, 11)
              emb=pd.get_dummies(ds["Embarked"],drop_first=True)
In [39]:
            2
              emb
Out[39]:
               Q S
               0
            1
               0
                  0
            2
               0
                  1
             3
               0
                  1
           886
               0
                  1
           887
               0
                  1
           888
               0
                  1
           889
               0
                  0
           890
               1 0
          891 rows × 2 columns
```

localhost:8888/notebooks/AIML/21BCE9791\_Assignment\_3.ipynb#

Out[40]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Embarked (
0	1	0	3	Braund, Mr. Owen Harris	1	22.0	1	0	A/5 21171	7.2500	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	0	38.0	1	0	PC 17599	71.2833	С
2	3	1	3	Heikkinen, Miss. Laina	0	26.0	0	0	STON/O2. 3101282	7.9250	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	35.0	1	0	113803	53.1000	S
4	5	0	3	Allen, Mr. William Henry	1	35.0	0	0	373450	8.0500	S
4											

In [41]: 1 ds.drop(['Embarked'],axis=1,inplace=True)

In [42]: 1 ds.head()

Out[42]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Q	s
0	1	0	3	Braund, Mr. Owen Harris	1	22.0	1	0	A/5 21171	7.2500	0	1
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	0	38.0	1	0	PC 17599	71.2833	0	0
2	3	1	3	Heikkinen, Miss. Laina	0	26.0	0	0	STON/O2. 3101282	7.9250	0	1
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	35.0	1	0	113803	53.1000	0	1
4	5	0	3	Allen, Mr. William Henry	1	35.0	0	0	373450	8.0500	0	1

In [44]: 1 ds.shape

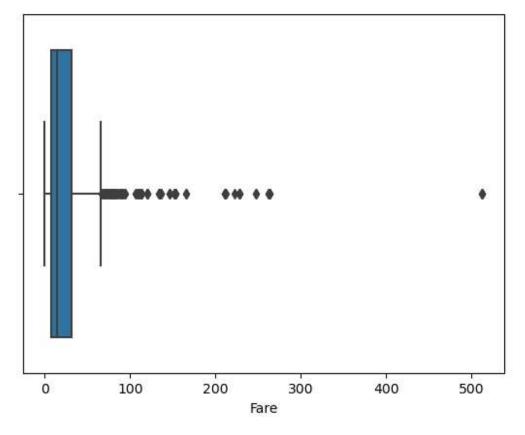
Out[44]: (891, 12)

1 Detecting and removing the outliers

```
In [47]: 1 sns.boxplot(ds['Fare'],data=ds);
2
```

C:\Users\M DIVYA\DataScience\lib\site-packages\seaborn\\_decorators.py:36: FutureWar ning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(



```
In [59]: 1 upper_limit =q3+1.5*IQR
2 upper_limit
```

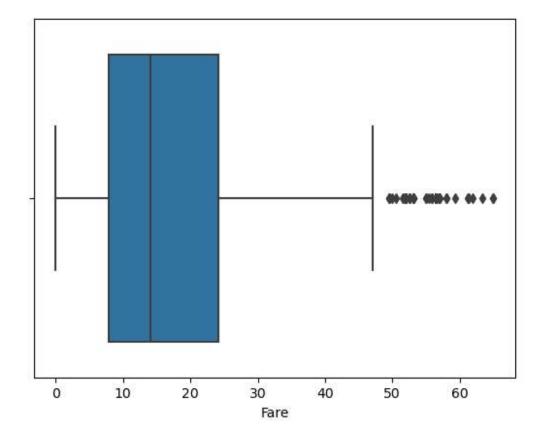
Out[59]: 65.6344

```
In [66]: 1 ds = ds[ds.Fare<upper_limit]</pre>
```

```
In [67]: 1 sns.boxplot(ds['Fare'],data=ds);
2
```

C:\Users\M DIVYA\DataScience\lib\site-packages\seaborn\\_decorators.py:36: FutureWar ning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an ex plicit keyword will result in an error or misinterpretation.

warnings.warn(



```
In [68]: 1 ds.shape
```

Out[68]: (891, 12)

In [69]: 1 ds.head()

Out[69]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Q	s
0	1	0	3	Braund, Mr. Owen Harris	1	22.0	1	0	A/5 21171	7.250	0	1
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	0	38.0	1	0	PC 17599	14.000	0	0
2	3	1	3	Heikkinen, Miss. Laina	0	26.0	0	0	STON/O2. 3101282	7.925	0	1
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	35.0	1	0	113803	53.100	0	1
4	5	0	3	Allen, Mr. William Henry	1	35.0	0	0	373450	8.050	0	1

1 Seperating Dependent and Independent Variables

```
In [71]: 1 x=ds.loc[:,['Age','SibSp','Parch','Fare']]
2 y=ds.iloc[:,1:2]
```

In [72]: 1 x.head()

Out[72]:

	Age	SibSp	Parch	Fare
0	22.0	1	0	7.250
1	38.0	1	0	14.000
2	26.0	0	0	7.925
3	35.0	1	0	53.100
4	35.0	0	0	8.050

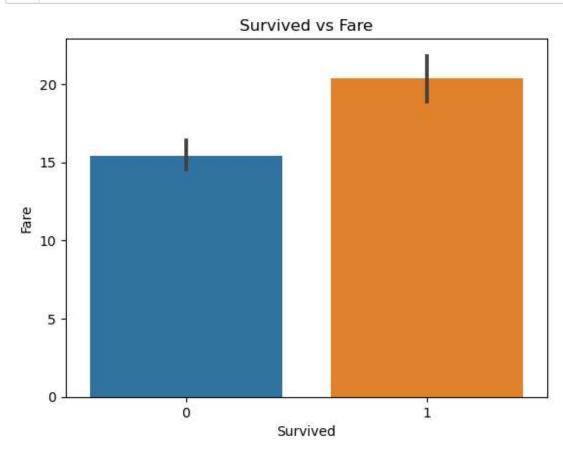
In [73]: 1 y.head()

Out[73]:

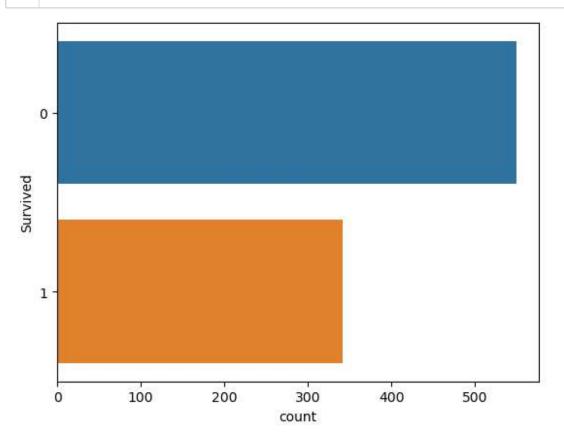
	Survived
0	0
1	1
2	1
3	1
4	0

1 Data Visualization

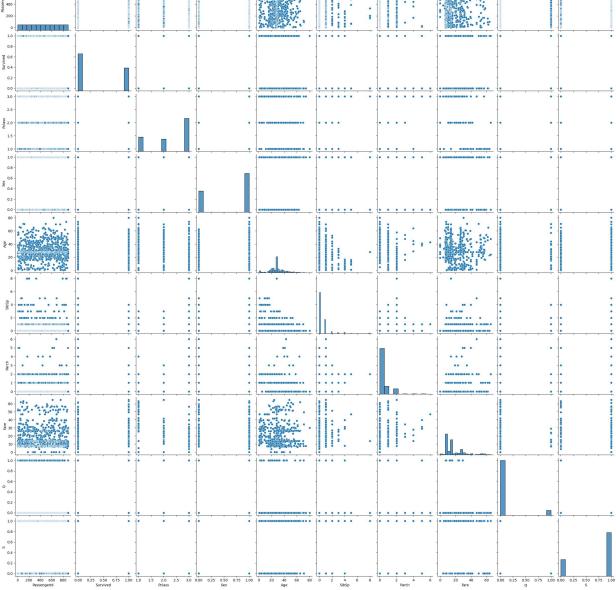
```
In [90]: 1 sns.barplot(data=ds,x='Survived',y='Fare')
2 plt.xlabel("Survived")
3 plt.ylabel("Fare")
4 plt.title("Survived vs Fare");
```



In [99]: 1 sns.countplot(data=ds,y="Survived",orient='h');



21BCE9791\_Assignment\_3 - Jupyter Notebook In [98]: sns.pairplot(data=ds);



- Splitting the data into training and testing
- In [74]: 1 from sklearn.model\_selection import train\_test\_split
  - x\_train,x\_test,y\_train,y\_test=train\_test\_split(x,y,test\_size=0.3,random\_state=0
- In [75]: x\_train.shape,x\_test.shape,y\_train.shape,y\_test.shape
- Out[75]: ((623, 4), (268, 4), (623, 1), (268, 1))
  - Feature Scaling

```
In [77]:
             from sklearn.preprocessing import StandardScaler
             ss=StandardScaler()
           2
In [80]:
             x_train=ss.fit_transform(x_train)
             x test=ss.transform(x test)
In [82]:
             x_train
Out[82]: array([[ 1.64654836, -0.457246 , -0.47299765, 0.68311366],
                [ 1.4930717 , 0.4033711 , -0.47299765, -0.29074647],
                [-2.19036814, 3.8458395, 1.93253327, 2.26224144],
                [-0.11843323, -0.457246, -0.47299765, -0.77703247],
                [0.49547341, 0.4033711, -0.47299765, -0.02691185],
                [ 2.33719333, 0.4033711 , 0.72976781, 1.64921395]])
In [83]:
           1 x test
Out[83]: array([[-0.0724674 , -0.53120385, -0.47809977, -0.15359735],
                [-0.0724674, -0.53120385, -0.47809977, -0.71667637],
                [-1.69302814, 3.68694819, 0.87064484, 1.04185031],
                [-0.14963696, 0.52333416, -0.47809977, -0.15393153],
                [-0.84416299, -0.53120385, -0.47809977, -0.72109409],
                [-0.0724674, -0.53120385, -0.47809977, 0.92739733]])
 In [ ]:
```