

Import Necessary Libraries

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

Import DataSet

```
In [2]: ▶ data=pd.read_csv("Titanic-Dataset.csv")
```

```
In [3]: ▶ data.head()
```

Out[3]:

	PassengerId	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	
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...)	female	38.0	1	0	PC 17599	71.2833	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	



In [4]: `data.tail()`

Out[4]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.00	Na
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.00	B4
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.45	Na
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.00	C14
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.75	Na



In [5]: `data.columns`

Out[5]: Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp', 'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'], dtype='object')

In [6]: data.info

```

Out[6]: <bound method DataFrame.info of      PassengerId  Survived  Pclass  \
0               1         0       3
1               2         1       1
2               3         1       3
3               4         1       1
4               5         0       3
..          ...      ...      ...
886            887         0       2
887            888         1       1
888            889         0       3
889            890         1       1
890            891         0       3

      Name      Sex  Age  SibSp
\
0      Braund, Mr. Owen Harris    male  22.0     1
1  Cumings, Mrs. John Bradley (Florence Briggs Th...  female  38.0     1
2      Heikkinen, Miss. Laina    female  26.0     0
3  Futrelle, Mrs. Jacques Heath (Lily May Peel)    female  35.0     1
4      Allen, Mr. William Henry    male  35.0     0
..          ...      ...      ...      ...
886      Montvila, Rev. Juozas    male  27.0     0
887      Graham, Miss. Margaret Edith    female  19.0     0
888  Johnston, Miss. Catherine Helen "Carrie"    female   NaN     1
889      Behr, Mr. Karl Howell    male  26.0     0
890      Dooley, Mr. Patrick    male  32.0     0

      Parch      Ticket    Fare Cabin Embarked
0         0      A/5 21171    7.2500   NaN      S
1         0      PC 17599   71.2833   C85      C
2         0  STON/O2. 3101282    7.9250   NaN      S
3         0      113803   53.1000  C123      S
4         0      373450    8.0500   NaN      S
..          ...      ...      ...      ...
886         0      211536   13.0000   NaN      S
887         0      112053   30.0000   B42      S
888         2      W./C. 6607   23.4500   NaN      S
889         0      111369   30.0000  C148      C
890         0      370376    7.7500   NaN      Q

[891 rows x 12 columns]>

```

In [7]: `data.info()`

```

<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

```

In [8]: `data.describe()`

Out[8]:

	PassengerId	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	0.523008	0.381594	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

Handling Null Values

```
In [9]: data.isnull().any()
```

```
Out[9]: PassengerId    False
Survived              False
Pclass                False
Name                  False
Sex                   False
Age                   True
SibSp                 False
Parch                 False
Ticket                False
Fare                  False
Cabin                 True
Embarked              True
dtype: bool
```

```
In [10]: data.isnull().sum()
```

```
Out[10]: 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
```

Filling the null values in Age column with mean

```
In [11]: mean_age=data["Age"].mean()
```

```
In [12]: data["Age"].fillna(mean_age,inplace=True)
```

In [13]: `data.tail()`

Out[13]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
886	887	0	2	Montvila, Rev. Juozas	male	27.000000	0	0	211536	13.00
887	888	1	1	Graham, Miss. Margaret Edith	female	19.000000	0	0	112053	30.00
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	29.699118	1	2	W./C. 6607	23.45
889	890	1	1	Behr, Mr. Karl Howell	male	26.000000	0	0	111369	30.00
890	891	0	3	Dooley, Mr. Patrick	male	32.000000	0	0	370376	7.75



In [14]: `data["Age"].isnull().sum()`

Out[14]: 0

Filling the null values in Cabin with mode

In [15]: `mode_cabin=data["Cabin"].mode()`

In [16]: `mode_cabin`

Out[16]: 0 B96 B98
1 C23 C25 C27
2 G6
Name: Cabin, dtype: object

In [17]: `data["Cabin"].fillna(mode_cabin[2],inplace=True)`

In [18]: `data["Cabin"].isnull().sum()`

Out[18]: 0

```
In [19]: data["Cabin"]
```

```
Out[19]: 0      G6
          1     C85
          2      G6
          3    C123
          4      G6
          ...
          886    G6
          887    B42
          888    G6
          889   C148
          890    G6
          Name: Cabin, Length: 891, dtype: object
```

Filling the Null values in Embarked with mode

```
In [20]: mode_emb=data["Embarked"].mode()
          mode_emb
```

```
Out[20]: 0    S
          Name: Embarked, dtype: object
```

```
In [21]: data["Embarked"].fillna(mode_emb[0],inplace=True)
```

```
In [22]: data["Embarked"].isnull().sum()
```

```
Out[22]: 0
```

```
In [23]: data.isnull().sum()
```

```
Out[23]: PassengerId    0
          Survived      0
          Pclass        0
          Name          0
          Sex           0
          Age           0
          SibSp         0
          Parch         0
          Ticket        0
          Fare          0
          Cabin         0
          Embarked      0
          dtype: int64
```

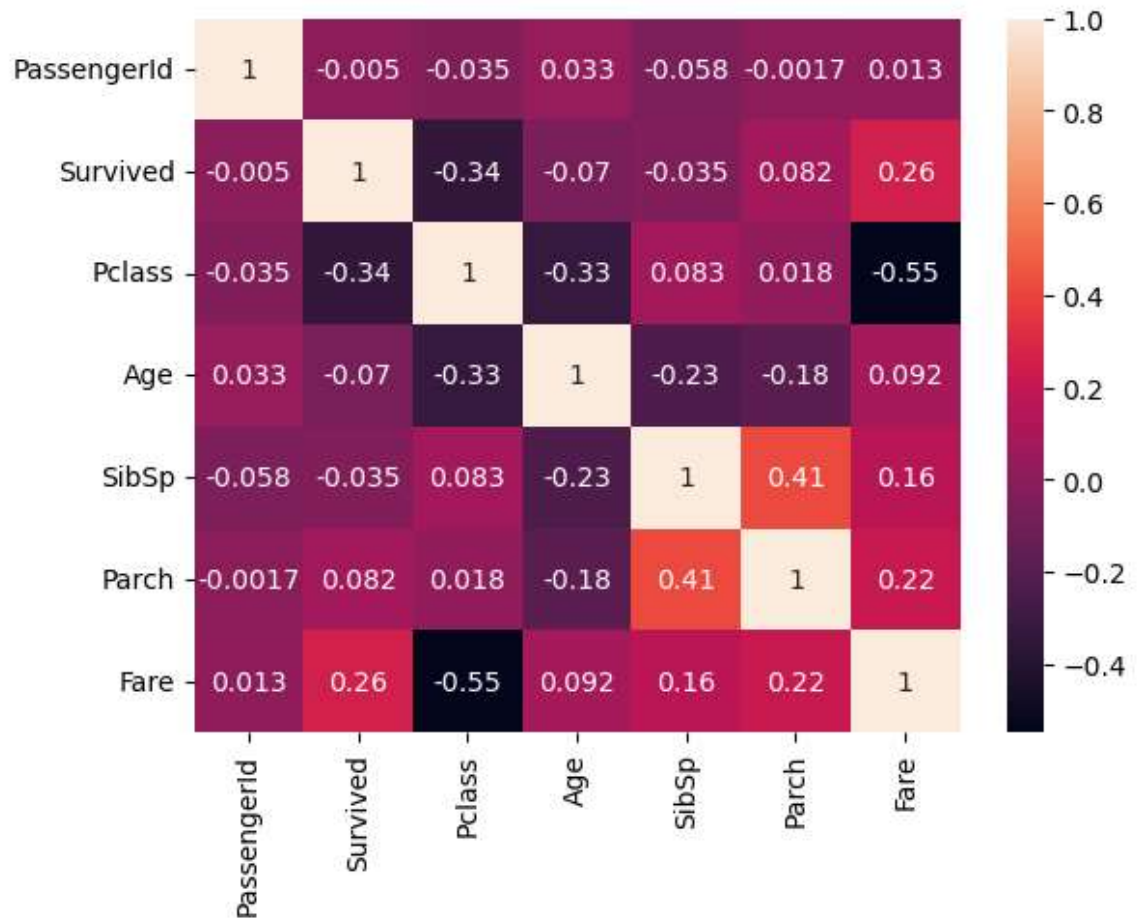
Data Visualisation

In [24]: `corr=data.corr()`

C:\Users\dines\AppData\Local\Temp\ipykernel_8724\2057684327.py:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.
`corr=data.corr()`

In [25]: `sns.heatmap(corr,annot=True)`

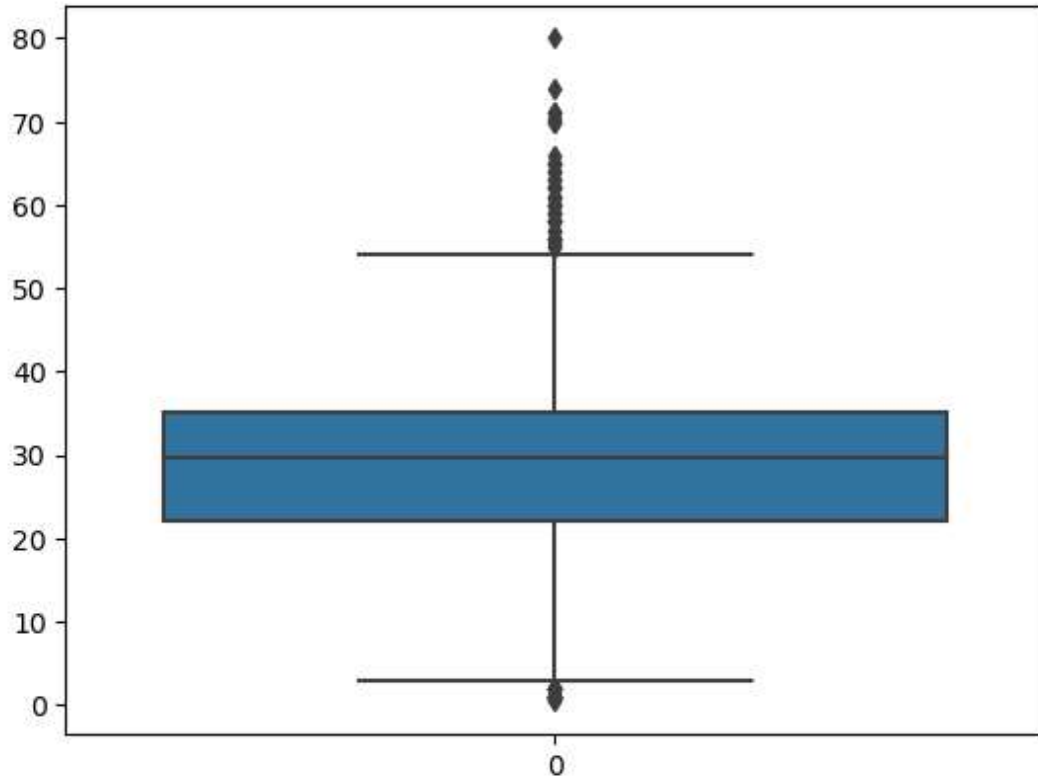
Out[25]: <Axes: >



Handling the outliers

```
In [26]: ▶ sns.boxplot(data["Age"])
```

```
Out[26]: <Axes: >
```



Inference: There are outliers in Age column

```
In [27]: ▶ age_q1=data.Age.quantile(0.25)
age_q2=data.Age.quantile(0.5)
age_q3=data.Age.quantile(0.75)
print(age_q1)
print(age_q2)
print(age_q3)
```

```
22.0
29.69911764705882
35.0
```

```
In [28]: ▶ IQR_AGE=age_q3-age_q1
IQR_AGE
```

```
Out[28]: 13.0
```

```
In [29]: upperlimit_age=age_q3+1.5*IQR_AGE  
upperlimit_age
```

Out[29]: 54.5

```
In [30]: lowerlimit_age=age_q1-1.5*IQR_AGE  
lowerlimit_age
```

Out[30]: 2.5

```
In [31]: median_age=data["Age"].median()  
median_age
```

Out[31]: 29.69911764705882

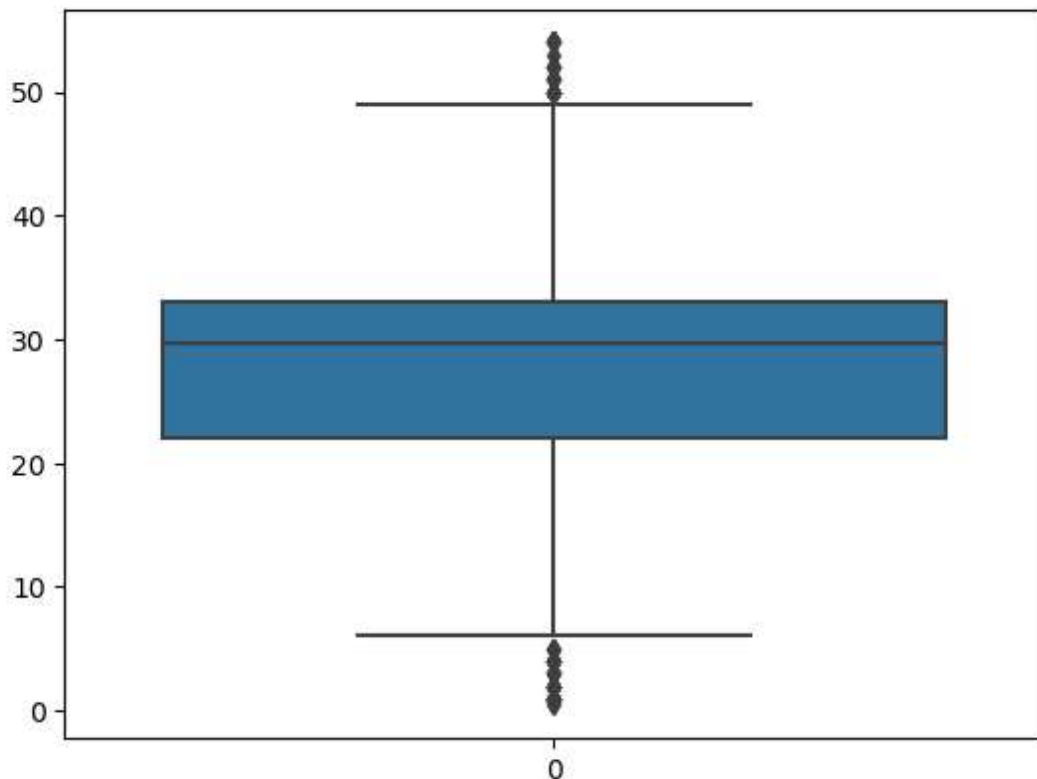
```
In [32]: data["Age"]=np.where(data["Age"]>upperlimit_age,median_age,data["Age"])
```

```
In [33]: (data["Age"]>upperlimit_age).sum()
```

Out[33]: 0

```
In [34]: sns.boxplot(data["Age"])
```

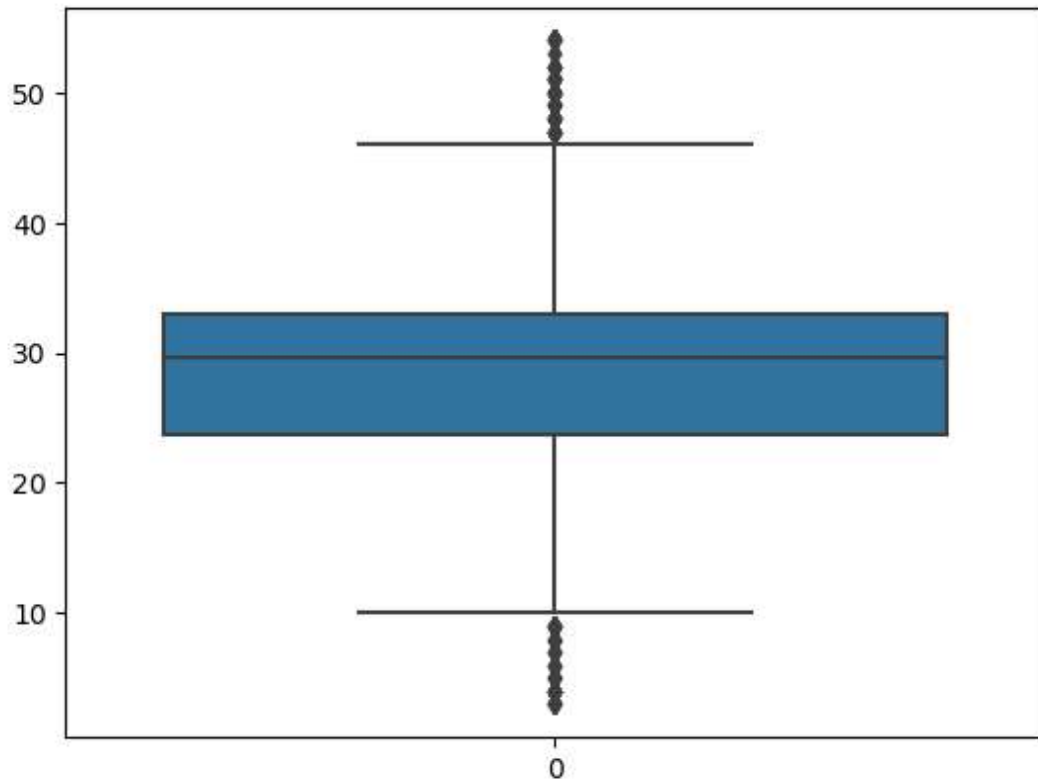
Out[34]: <Axes: >



```
In [35]: data["Age"] = np.where(data["Age"] < lowerlimit_age, median_age, data["Age"])
```

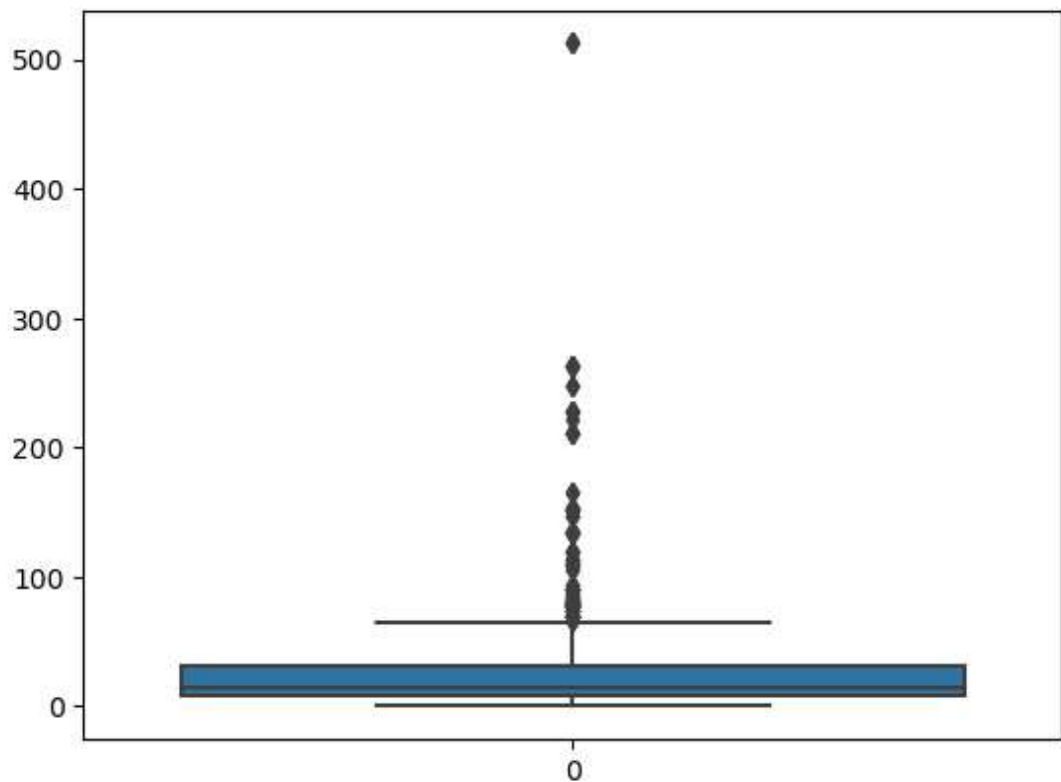
```
In [36]: sns.boxplot(data["Age"])
```

Out[36]: <Axes: >



```
In [37]: ▶ sns.boxplot(data["Fare"])
```

```
Out[37]: <Axes: >
```



```
In [38]: ▶ fare_q1=data.Fare.quantile(0.25)
fare_q2=data.Fare.quantile(0.5)
fare_q3=data.Fare.quantile(0.75)
print(fare_q1)
print(fare_q2)
print(fare_q3)
```

```
7.9104
14.4542
31.0
```

```
In [39]: ▶ IQR_FARE=fare_q3-fare_q1
IQR_FARE
```

```
Out[39]: 23.0896
```

```
In [40]: ▶ upperlimit_fare=fare_q3+1.5*IQR_FARE
upperlimit_fare
```

```
Out[40]: 65.6344
```

```
In [41]: lowerlimit_fare=fare_q1-1.5*IQR_FARE  
lowerlimit_fare
```

Out[41]: -26.724

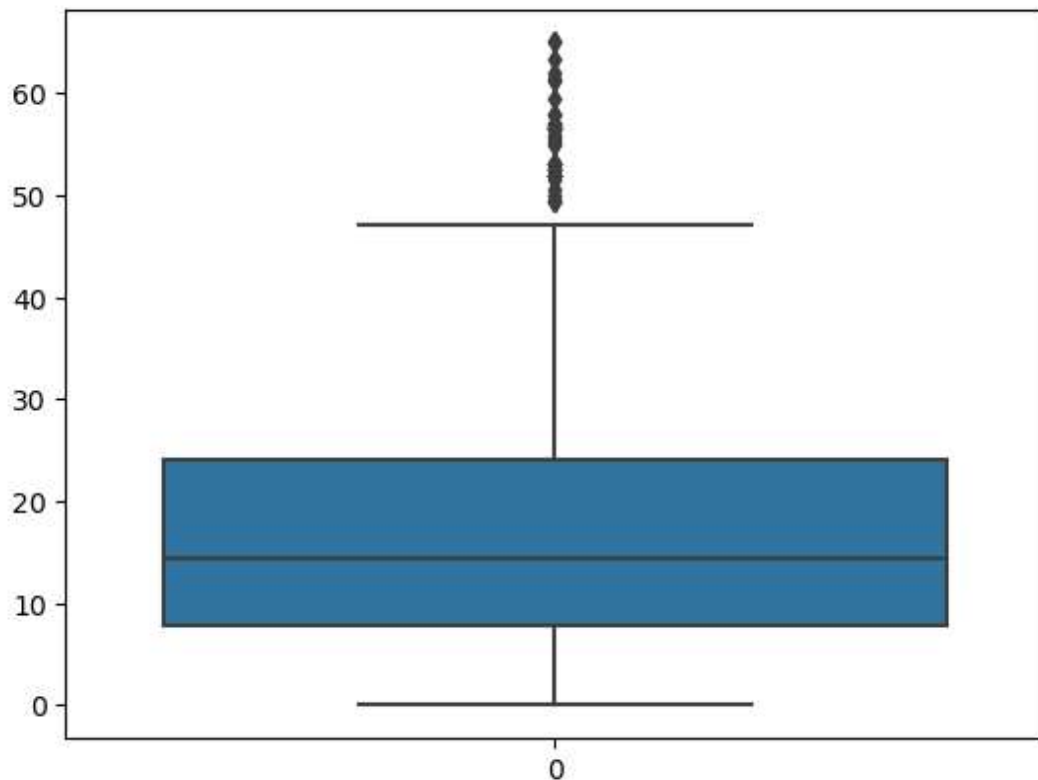
```
In [42]: median_fare=data["Fare"].median()  
median_fare
```

Out[42]: 14.4542

```
In [43]: data["Fare"]=np.where(data["Fare"]>upperlimit_fare,median_fare,data["Fare"])
```

```
In [44]: sns.boxplot(data["Fare"])
```

Out[44]: <Axes: >



```
In [45]: (data["Fare"]>upperlimit_fare).sum()
```

Out[45]: 0

Dropping Variables

In [46]: `data.head()`

Out[46]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500		
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...)	female	38.0	1	0	PC 17599	14.4542		
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250		
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000		
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500		

In [48]: `data.drop("Name",axis=1,inplace=True)`

In [49]: `data.head()`

Out[49]:

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	male	22.0	1	0	A/5 21171	7.2500	G6	
1	2	1	1	female	38.0	1	0	PC 17599	14.4542	C85	
2	3	1	3	female	26.0	0	0	STON/O2. 3101282	7.9250	G6	
3	4	1	1	female	35.0	1	0	113803	53.1000	C123	
4	5	0	3	male	35.0	0	0	373450	8.0500	G6	

In [50]: `data.drop("Ticket",axis=1,inplace=True)`

In [51]: `data.head()`

Out[51]:

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Cabin	Embarked
0	1	0	3	male	22.0	1	0	7.2500	G6	S
1	2	1	1	female	38.0	1	0	14.4542	C85	C
2	3	1	3	female	26.0	0	0	7.9250	G6	S
3	4	1	1	female	35.0	1	0	53.1000	C123	S
4	5	0	3	male	35.0	0	0	8.0500	G6	S

In [52]: `data.drop("PassengerId",axis=1,inplace=True)`

In [53]: `data.head()`

Out[53]:

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Cabin	Embarked
0	0	3	male	22.0	1	0	7.2500	G6	S
1	1	1	female	38.0	1	0	14.4542	C85	C
2	1	3	female	26.0	0	0	7.9250	G6	S
3	1	1	female	35.0	1	0	53.1000	C123	S
4	0	3	male	35.0	0	0	8.0500	G6	S

In [54]: `data.drop("Cabin",axis=1,inplace=True)`

In [55]: `data.head()`

Out[55]:

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	0	3	male	22.0	1	0	7.2500	S
1	1	1	female	38.0	1	0	14.4542	C
2	1	3	female	26.0	0	0	7.9250	S
3	1	1	female	35.0	1	0	53.1000	S
4	0	3	male	35.0	0	0	8.0500	S

Splitting Dependent and Independent variables

In [56]: `#independent variable
x=data.drop("Survived",axis=1)`

In [57]: `x.head()`

Out[57]:

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	male	22.0	1	0	7.2500	S
1	1	female	38.0	1	0	14.4542	C
2	3	female	26.0	0	0	7.9250	S
3	1	female	35.0	1	0	53.1000	S
4	3	male	35.0	0	0	8.0500	S

In [58]: `#dependent variable
y=data["Survived"]`

In [59]: `y.head()`

Out[59]:

0	0
1	1
2	1
3	1
4	0

Name: Survived, dtype: int64

Encoding

In [60]: `from sklearn.preprocessing import LabelEncoder`

In [61]: `le=LabelEncoder()`

In [62]: `x["Sex"]=le.fit_transform(x["Sex"])`

In [63]: `x["Sex"]`

Out[63]:

0	1
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 [64]: x.head()
```

```
Out[64]:
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.0	1	0	7.2500	S
1	1	0	38.0	1	0	14.4542	C
2	3	0	26.0	0	0	7.9250	S
3	1	0	35.0	1	0	53.1000	S
4	3	1	35.0	0	0	8.0500	S

```
In [65]: x["Embarked"]=le.fit_transform(data["Embarked"])
```

```
In [66]: x.head()
```

```
Out[66]:
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.0	1	0	7.2500	2
1	1	0	38.0	1	0	14.4542	0
2	3	0	26.0	0	0	7.9250	2
3	1	0	35.0	1	0	53.1000	2
4	3	1	35.0	0	0	8.0500	2

```
In [67]: x["Pclass"].unique()
```

```
Out[67]: array([3, 1, 2], dtype=int64)
```

```
In [68]: x["Pclass"].nunique()
```

```
Out[68]: 3
```

```
In [69]: x["Sex"].unique()
```

```
Out[69]: array([1, 0])
```

```
In [70]: x["Embarked"].unique()
```

```
Out[70]: array([2, 0, 1])
```

Splitting Data into Training data and Testing data

```
In [71]: from sklearn.model_selection import train_test_split
```

```
In [72]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state
```

```
In [73]: x_train.shape,x_test.shape,y_train.shape,y_test.shape
```

```
Out[73]: ((712, 7), (179, 7), (712,), (179,))
```

Feature Scaling

```
In [74]: from sklearn.preprocessing import StandardScaler
```

```
In [75]: sc=StandardScaler()
```

```
In [76]: x_train=sc.fit_transform(x_train)
```

```
In [77]: x_train
```

```
Out[77]: array([[ 0.81925059, -1.37207547,  0.07840244, ...,  1.95926403,
                  -0.17726299, -1.98156574],
                [-0.38096838,  0.72882288,  0.21232456, ..., -0.47741019,
                  -0.54667438,  0.5790056 ],
                [-0.38096838,  0.72882288,  0.21232456, ...,  0.74092692,
                  1.51640316, -1.98156574],
                ...,
                [ 0.81925059,  0.72882288,  0.07840244, ..., -0.47741019,
                  -0.76203333, -0.70128007],
                [ 0.81925059, -1.37207547,  0.72706025, ..., -0.47741019,
                  -0.00958083,  0.5790056 ],
                [-0.38096838,  0.72882288,  0.07840244, ...,  0.74092692,
                  1.67175552,  0.5790056 ]])
```

```
In [78]: x_test=sc.fit_transform(x_test)
```

```
In [79]: x_test
```

```
Out[79]: array([[ 0.86022947,  0.77344314,  0.0739225 , ..., -0.46006628,
                  -0.19571051, -1.80134224],
                [ 0.86022947,  0.77344314,  0.0739225 , ..., -0.46006628,
                  -0.76604362,  0.61394061],
                [ 0.86022947,  0.77344314, -2.17117908, ...,  0.88996427,
                  1.01513799, -0.59370081],
                ...,
                [-1.50871015, -1.29291987,  0.20258887, ..., -0.46006628,
                  -0.19604899, -1.80134224],
                [ 0.86022947,  0.77344314, -0.58866712, ..., -0.46006628,
                  -0.74092958,  0.61394061],
                [ 0.86022947,  0.77344314, -0.98429511, ..., -0.46006628,
                  -0.72476479,  0.61394061]])
```

In []: ▶