

In [1]:

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

In [38]:

```
data=pd.read_csv("Titanic-Dataset.csv")
```

In [39]:

```
data.head()
```

Out[39]:

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



In [40]:

```
data.tail()
```

Out[40]:

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

In [41]:

```
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 [42]:

```
data.describe()
```

Out[42]:

	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 [43]:

```
data.isnull().any()
```

Out[43]:

```
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 [44]:

```
data.isnull().sum()
```

Out[44]:

```
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 [45]:

```
mean=data["Age"].mean()
```

Filling the null values in Age column with Mean

In [46]:

```
data["Age"]=data["Age"].fillna(mean)
```

In [47]:

```
data["Age"].tail()
```

Out[47]:

```
886    27.000000
887    19.000000
888    29.699118
889    26.000000
890    32.000000
Name: Age, dtype: float64
```

In [48]:

```
data["Age"].isnull().sum()
```

Out[48]:

```
0
```

Filling the Null values in Embarked with mode

In [49]:

```
Em_mode=data["Embarked"].mode()
```

In [50]:

```
data["Embarked"]=data["Embarked"].fillna(Em_mode[0])
```

In [51]:

```
data["Embarked"].isnull().sum()
```

Out[51]:

0

In []:

Filling the null values in Cabin with mode

In [52]:

```
Cabin_mode=data["Cabin"].mode()
```

In [53]:

```
data["Cabin"]
```

Out[53]:

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

In [54]:

```
Cabin_mode
```

Out[54]:

```
0      B96 B98
1    C23 C25 C27
2           G6
Name: Cabin, dtype: object
```

In [55]:

```
data["Cabin"] = data["Cabin"].fillna(Cabin_mode[2])
```

In [56]:

```
data["Cabin"].isnull().sum()
```

Out[56]:

```
0
```

In [57]:

```
data["Cabin"]
```

Out[57]:

```
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
```

In [58]:

```
data.isnull().sum()
```

Out[58]:

```
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 [59]:

```
cor=data.corr()
```

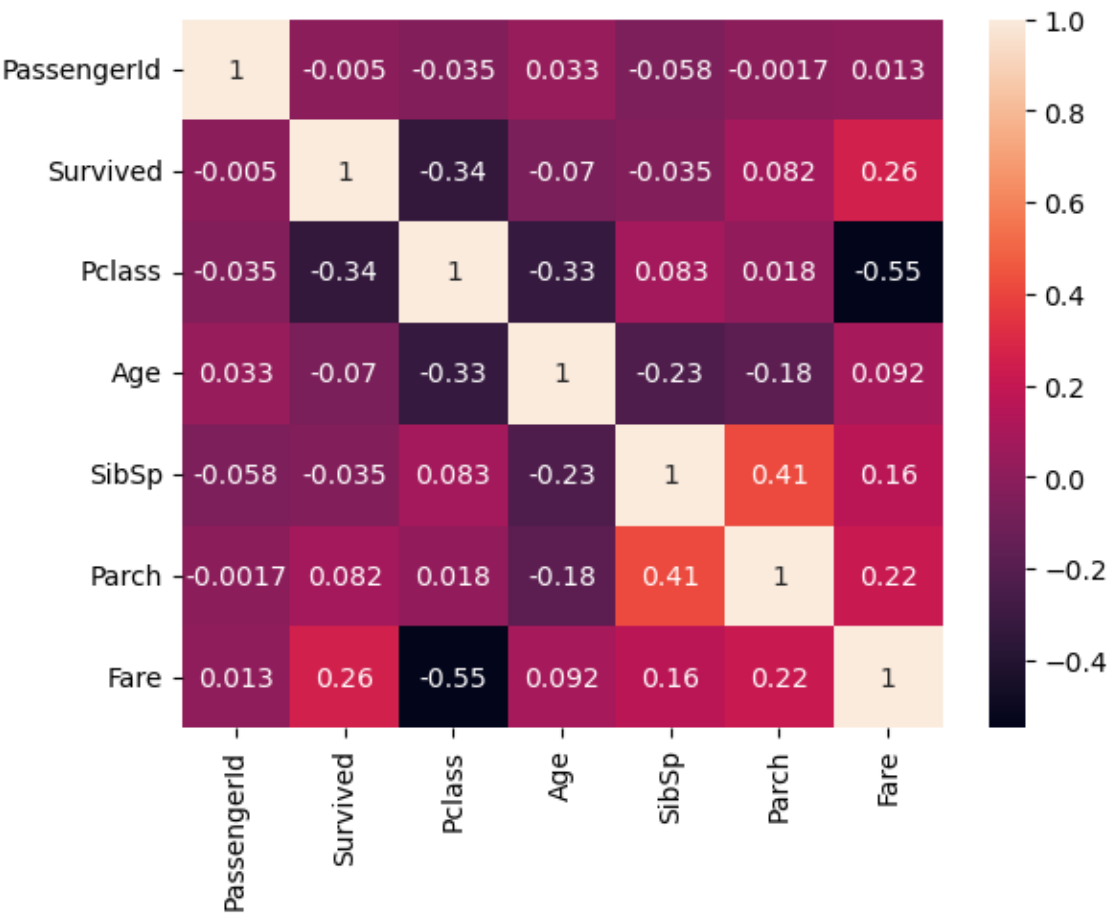
```
C:\Users\pichi\AppData\Local\Temp\ipykernel_25480\1426905697.py:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.
  cor=data.corr()
```

In [60]:

```
sns.heatmap(cor,annot=True)
```

Out[60]:

<Axes: >



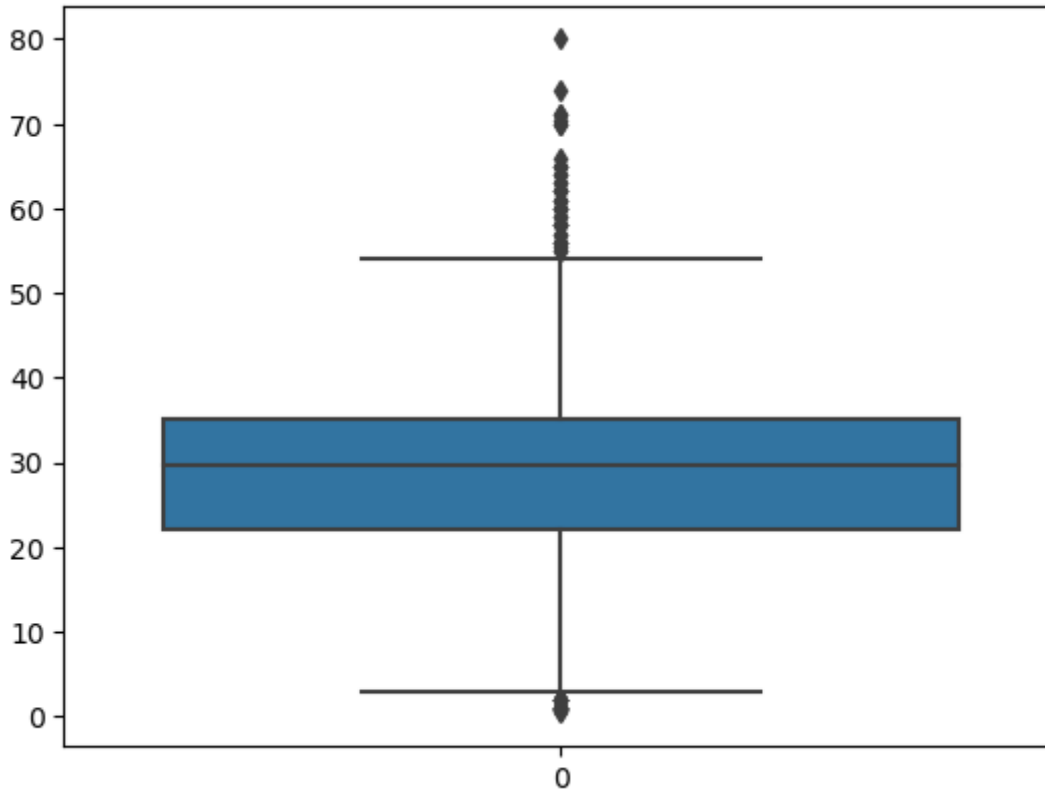
Handling the outliers

In [61]:

```
sns.boxplot(data["Age"])
```

Out[61]:

<Axes: >



Outliers

In [62]:

```
Age_q1 = data.Age.quantile(0.25)
Age_q3 = data.Age.quantile(0.75)
print(Age_q1)
print(Age_q3)
```

22.0

35.0

In [63]:

```
IQR_Age=Age_q3-Age_q1
IQR_Age
```

Out[63]:

13.0

In [64]:

```
upperlimit_Age=Age_q3+1.5*IQR_Age  
upperlimit_Age
```

Out[64]:

54.5

In [65]:

```
lower_limit_Age = Age_q1-1.5*IQR_Age  
lower_limit_Age
```

Out[65]:

2.5

In [66]:

```
median_Age=data["Age"].median()  
median_Age
```

Out[66]:

29.69911764705882

In [67]:

```
data["Age"]=np.where(data["Age"]>upperlimit_Age,median_Age,data["Age"])
```

In [68]:

```
(data["Age"]>54.5).sum()
```

Out[68]:

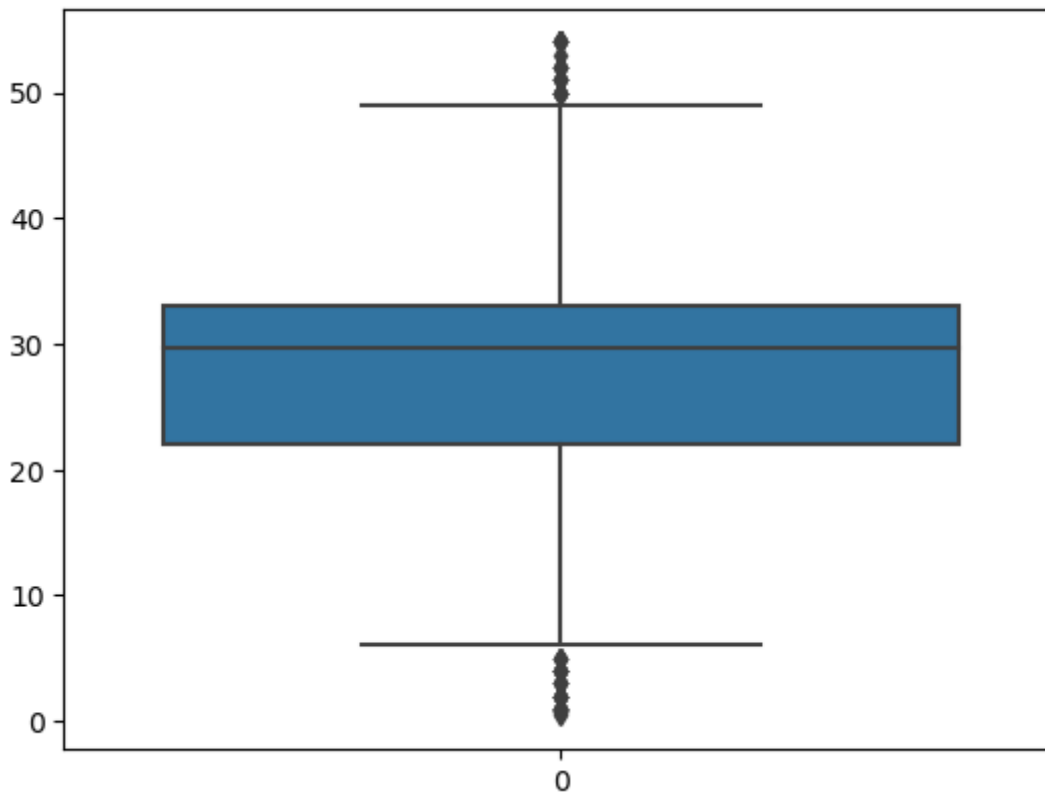
0

In [69]:

```
sns.boxplot(data["Age"])
```

Out[69]:

<Axes: >



In [70]:

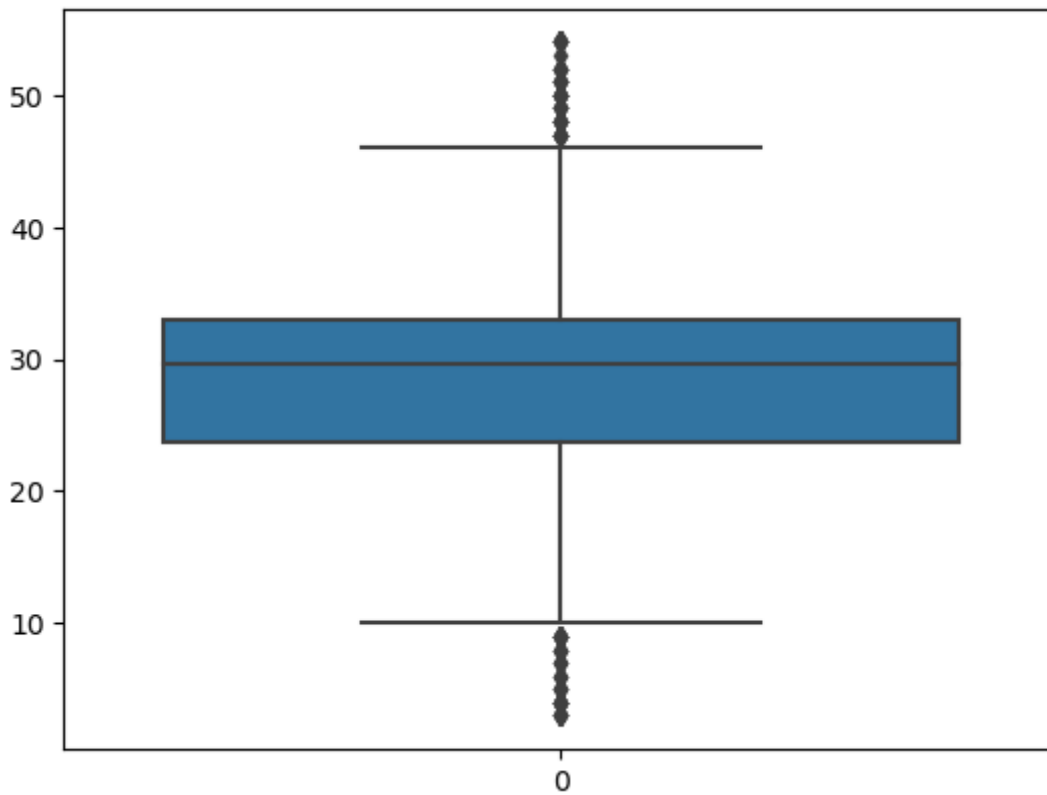
```
data["Age"] = np.where(data["Age"] < lower_limit_Age, median_Age, data["Age"])
```

In [71]:

```
sns.boxplot(data["Age"])
```

Out[71]:

<Axes: >

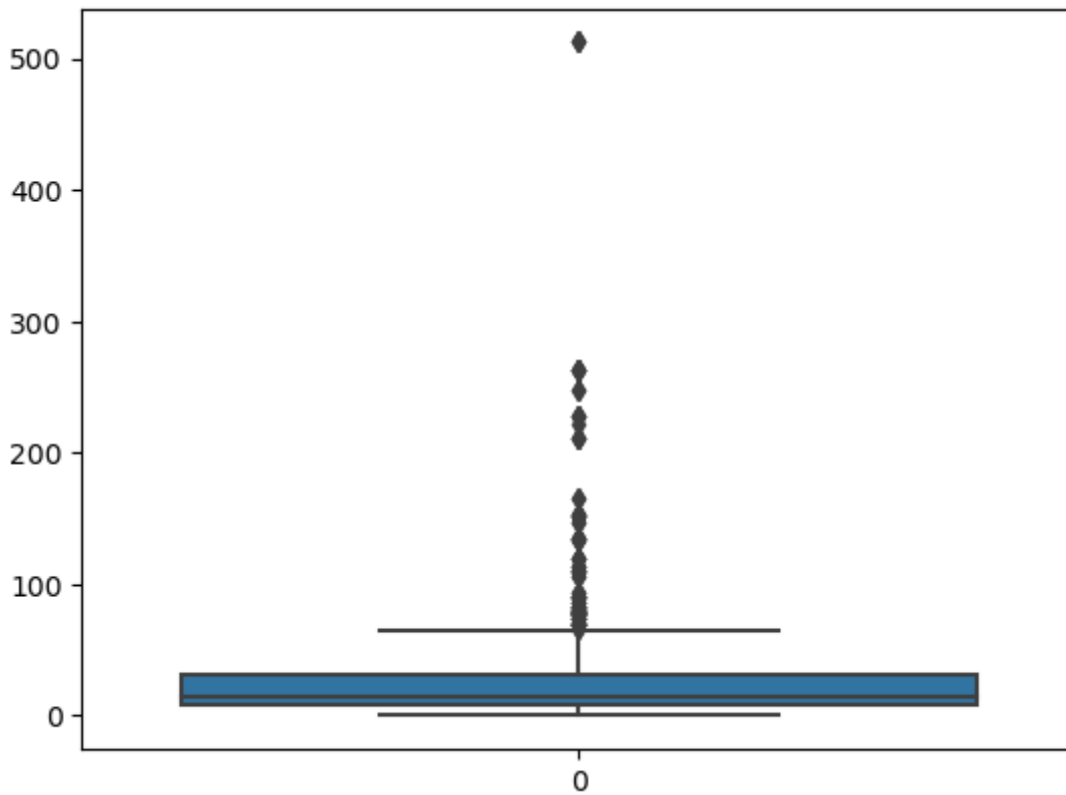


In [72]:

```
sns.boxplot(data["Fare"])
```

Out[72]:

<Axes: >



In [73]:

```
Fare_q1 = data.Fare.quantile(0.25)
Fare_q3 = data.Fare.quantile(0.75)
print(Fare_q1)
print(Fare_q3)
```

7.9104

31.0

In [74]:

```
IQR_Fare=Fare_q3-Fare_q1
IQR_Fare
```

Out[74]:

23.0896

In [75]:

```
upperlimit_Fare=Fare_q3+1.5*IQR_Fare
upperlimit_Fare
```

Out[75]:

65.6344

In [76]:

```
lower_limit_Fare = Fare_q1-1.5*IQR_Fare  
lower_limit_Fare
```

Out[76]:

-26.724

In [77]:

```
median_Fare=data["Fare"].median()  
median_Fare
```

Out[77]:

14.4542

In [78]:

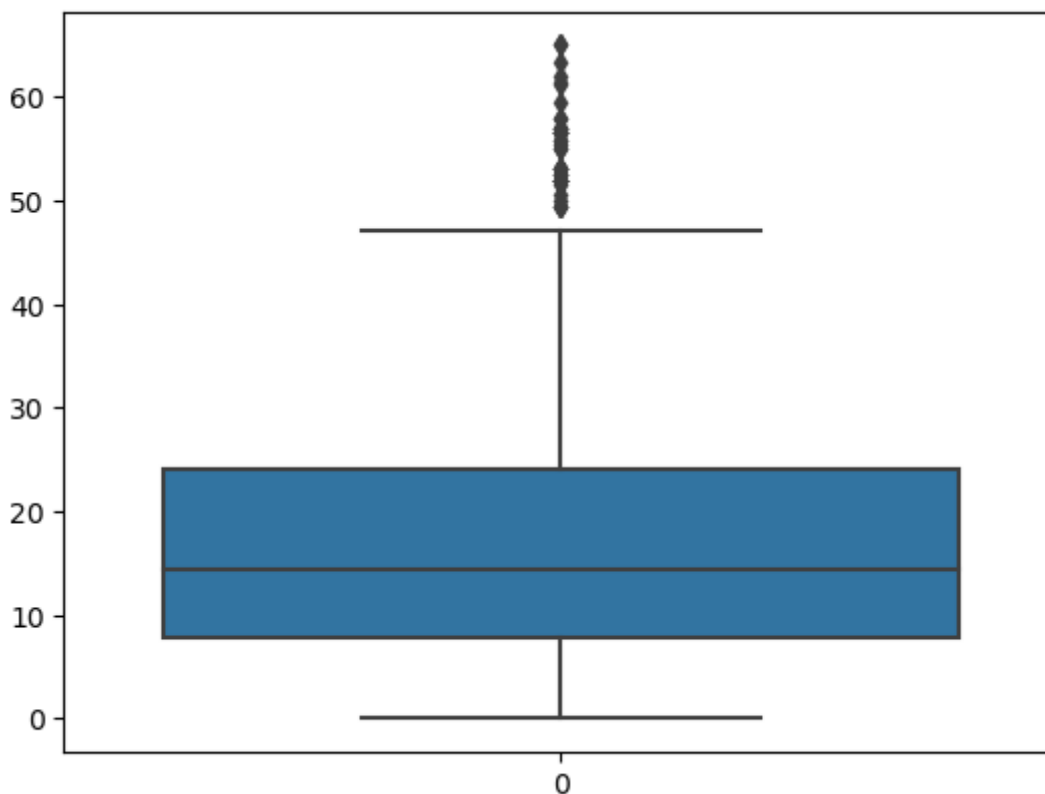
```
data['Fare'] = np.where(  
    (data['Fare'] > upperlimit_Fare),  
    median_Fare,  
    data['Fare']  
)
```

In [79]:

```
sns.boxplot(data["Fare"])
```

Out[79]:

<Axes: >



In [80]:

```
(data["Fare"]>65).sum()
```

Out[80]:

0

dropping the variables

In [84]:

```
data.drop(['Name'],axis=1,inplace=True)
```



```

-----
-
KeyError                                Traceback (most recent call las
t)
Cell In[84], line 1
----> 1 data.drop(['Name'],axis=1,inplace=True)
      2 data

File C:\ProgramData\anaconda3\lib\site-packages\pandas\util\_decorators.p
y:331, in deprecate_nonkeyword_arguments.<locals>.decorate.<locals>.wrappe
r(*args, **kwargs)
    325 if len(args) > num_allow_args:
    326     warnings.warn(
    327         msg.format(arguments=_format_argument_list(allow_args)),
    328         FutureWarning,
    329         stacklevel=find_stack_level(),
    330     )
--> 331 return func(*args, **kwargs)

File C:\ProgramData\anaconda3\lib\site-packages\pandas\core\frame.py:5399,
in DataFrame.drop(self, labels, axis, index, columns, level, inplace, erro
rs)
    5251 @deprecate_nonkeyword_arguments(version=None, allowed_args=["sel
f", "labels"])
    5252 def drop( # type: ignore[override]
    5253     self,
    5254     (...)
    5260     errors: IgnoreRaise = "raise",
    5261 ) -> DataFrame | None:
    5262     """
    5263     Drop specified labels from rows or columns.
    5264
    5265     (...)
    5397         weight  1.0      0.8
    5398     """
-> 5399     return super().drop(
    5400         labels=labels,
    5401         axis=axis,
    5402         index=index,
    5403         columns=columns,
    5404         level=level,
    5405         inplace=inplace,
    5406         errors=errors,
    5407     )

File C:\ProgramData\anaconda3\lib\site-packages\pandas\util\_decorators.p
y:331, in deprecate_nonkeyword_arguments.<locals>.decorate.<locals>.wrappe
r(*args, **kwargs)
    325 if len(args) > num_allow_args:
    326     warnings.warn(
    327         msg.format(arguments=_format_argument_list(allow_args)),
    328         FutureWarning,
    329         stacklevel=find_stack_level(),
    330     )
--> 331 return func(*args, **kwargs)

File C:\ProgramData\anaconda3\lib\site-packages\pandas\core\generic.py:450
5, in NDFrame.drop(self, labels, axis, index, columns, level, inplace, err
ors)
    4503 for axis, labels in axes.items():
    4504     if labels is not None:

```

```
-> 4505         obj = obj._drop_axis(labels, axis, level=level, errors=err
ors)
```

```
4507 if inplace:
4508     self._update_inplace(obj)
```

File C:\ProgramData\anaconda3\lib\site-packages\pandas\core\generic.py:454

```
6, in NDFrame._drop_axis(self, labels, axis, level, errors, only_slice)
4544     new_axis = axis.drop(labels, level=level, errors=errors)
4545     else:
-> 4546         new_axis = axis.drop(labels, errors=errors)
4547     indexer = axis.get_indexer(new_axis)
4549 # Case for non-unique axis
4550 else:
```

File C:\ProgramData\anaconda3\lib\site-packages\pandas\core\indexes\base.py:6934, in Index.drop(self, labels, errors)

```
6932 if mask.any():
6933     if errors != "ignore":
-> 6934         raise KeyError(f"{list(labels[mask])} not found in axis")
6935     indexer = indexer[~mask]
6936 return self.delete(indexer)
```

KeyError: "['Name'] not found in axis"

In [85]:

data

Out[85]:

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
0	1	0	3	male	22.000000	1	0	A/5 21171	7.2500	
1	2	1	1	female	38.000000	1	0	PC 17599	14.4542	
2	3	1	3	female	26.000000	0	0	STON/O2. 3101282	7.9250	
3	4	1	1	female	35.000000	1	0	113803	53.1000	C
4	5	0	3	male	35.000000	0	0	373450	8.0500	
...
886	887	0	2	male	27.000000	0	0	211536	13.0000	
887	888	1	1	female	19.000000	0	0	112053	30.0000	
888	889	0	3	female	29.699118	1	2	W./C. 6607	23.4500	
889	890	1	1	male	26.000000	0	0	111369	30.0000	C
890	891	0	3	male	32.000000	0	0	370376	7.7500	

891 rows × 11 columns



In [86]:

```
data.drop(['Ticket'],axis=1,inplace=True)
```

In [87]:

```
data
```

Out[87]:

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Cabin	Emba
0	1	0	3	male	22.000000	1	0	7.2500	G6	
1	2	1	1	female	38.000000	1	0	14.4542	C85	
2	3	1	3	female	26.000000	0	0	7.9250	G6	
3	4	1	1	female	35.000000	1	0	53.1000	C123	
4	5	0	3	male	35.000000	0	0	8.0500	G6	
...
886	887	0	2	male	27.000000	0	0	13.0000	G6	
887	888	1	1	female	19.000000	0	0	30.0000	B42	
888	889	0	3	female	29.699118	1	2	23.4500	G6	
889	890	1	1	male	26.000000	0	0	30.0000	C148	
890	891	0	3	male	32.000000	0	0	7.7500	G6	

891 rows × 10 columns



In [88]:

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

In [89]:

```
data
```

Out[89]:

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Cabin	Embarked
0	0	3	male	22.000000	1	0	7.2500	G6	S
1	1	1	female	38.000000	1	0	14.4542	C85	C
2	1	3	female	26.000000	0	0	7.9250	G6	S
3	1	1	female	35.000000	1	0	53.1000	C123	S
4	0	3	male	35.000000	0	0	8.0500	G6	S
...
886	0	2	male	27.000000	0	0	13.0000	G6	S
887	1	1	female	19.000000	0	0	30.0000	B42	S
888	0	3	female	29.699118	1	2	23.4500	G6	S
889	1	1	male	26.000000	0	0	30.0000	C148	C
890	0	3	male	32.000000	0	0	7.7500	G6	Q

891 rows × 9 columns

In [90]:

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

In [91]:

```
data
```

Out[91]:

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	0	3	male	22.000000	1	0	7.2500	S
1	1	1	female	38.000000	1	0	14.4542	C
2	1	3	female	26.000000	0	0	7.9250	S
3	1	1	female	35.000000	1	0	53.1000	S
4	0	3	male	35.000000	0	0	8.0500	S
...
886	0	2	male	27.000000	0	0	13.0000	S
887	1	1	female	19.000000	0	0	30.0000	S
888	0	3	female	29.699118	1	2	23.4500	S
889	1	1	male	26.000000	0	0	30.0000	C
890	0	3	male	32.000000	0	0	7.7500	Q

891 rows × 8 columns

Splitting the data

In [92]:

```
y=data["Survived"]
```

In [93]:

```
y.head()
```

Out[93]:

```
0    0
1    1
2    1
3    1
4    0
```

Name: Survived, dtype: int64

In [100]:

```
data
```

Out[100]:

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	male	22.000000	1	0	7.2500	S
1	1	female	38.000000	1	0	14.4542	C
2	3	female	26.000000	0	0	7.9250	S
3	1	female	35.000000	1	0	53.1000	S
4	3	male	35.000000	0	0	8.0500	S
...
886	2	male	27.000000	0	0	13.0000	S
887	1	female	19.000000	0	0	30.0000	S
888	3	female	29.699118	1	2	23.4500	S
889	1	male	26.000000	0	0	30.0000	C
890	3	male	32.000000	0	0	7.7500	Q

891 rows × 7 columns

Encoding

In [101]:

```
from sklearn.preprocessing import LabelEncoder
```

In [102]:

```
le=LabelEncoder()
```

In [103]:

```
data["Sex"]=le.fit_transform(data["Sex"])
```

In [104]:

```
data["Sex"]
```

Out[104]:

```
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 [105]:

```
data.head()
```

Out[105]:

	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 [106]:

```
data["Embarked"]=le.fit_transform(data["Embarked"])
```

In [107]:

```
data.head()
```

Out[107]:

	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 [108]:

```
data["Pclass"].nunique()
```

Out[108]:

```
3
```

In [109]:

```
data["Pclass"].unique()
```

Out[109]:

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

In [112]:

```
data["Sex"].unique()
```

Out[112]:

```
array([1, 0])
```

In [113]:

```
data["Embarked"].unique()
```

Out[113]:

```
array([2, 0, 1])
```

Splitting the train and test data

In [114]:

```
from sklearn.model_selection import train_test_split  
x_train,x_test,y_train,y_test=train_test_split(data,y,test_size=0.3,random_state=0)
```

In [119]:

```
x_train.shape,x_test.shape,y_train.shape,y_test.shape
```

Out[119]:

```
((623, 7), (268, 7), (623,), (268,))
```

Feature Scaling

In [123]:

```
from sklearn.preprocessing import StandardScaler
```

In [124]:

```
sc=StandardScaler()
```

In [125]:

```
x_train=sc.fit_transform(x_train)
```

In [126]:

```
x_train
```

Out[126]:

```
array([[ -1.5325562,   0.72592065,   2.25384386, ...,  -0.47299765,
         0.67925137,   0.56710989],
       [ -1.5325562,  -1.37756104,   2.04763778, ...,  -0.47299765,
        -0.26059483,  -2.03075381],
       [  0.84844757,   0.72592065,   0.05765817, ...,   1.93253327,
         2.26045064,   0.56710989],
       ...,
       [  0.84844757,   0.72592065,   0.05765817, ...,  -0.47299765,
        -0.78281017,  -0.73182196],
       [  0.84844757,  -1.37756104,   0.70729828, ...,  -0.47299765,
        -0.03170555,   0.56710989],
       [-0.34205431,   0.72592065,   0.05765817, ...,   0.72976781,
         1.64661898,   0.56710989]])
```

In [127]:

```
x_test=sc.fit_transform(x_test)
```


In [128]:

```
x_test
```

Out[128]:

```
array([[ 0.77963055,  0.76537495,  0.12227065, ..., -0.47809977,
        -0.15813988, -1.76531134],
       [ 0.77963055,  0.76537495,  0.12227065, ..., -0.47809977,
        -0.72165412,  0.63014911],
       [ 0.77963055,  0.76537495, -2.14803266, ...,  0.87064484,
        1.03823178, -0.56758111],
       ...,
       [ 0.77963055,  0.76537495, -0.14768767, ..., -0.47809977,
        -0.15847431, -1.76531134],
       [ 0.77963055, -1.30654916, -1.04784292, ..., -0.47809977,
        -0.72607524,  0.63014911],
       [-1.64991582,  0.76537495,  0.12227065, ..., -0.47809977,
        0.92369033, -1.76531134]])
```

In []:

In []: