

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

df = pd.read_csv("/content/drive/MyDrive/DATASETS/Titanic-Dataset.csv")

from google.colab import drive
drive.mount('/content/drive')

Mounted at /content/drive

df.head()
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.25
1	2	1	1	Cummings, Mrs. John Bradley (Florence Briggs Th...)	female	38.0	1	0	PC 17599	71.28

```
df.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
```

```
df.describe()
```

	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

```
df.corr()
```

```
<ipython-input-10-2f6f6606aa2c>:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future ver
df.corr()
```

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
<b>PassengerId</b>	1.000000	-0.005007	-0.035144	0.036847	-0.057527	-0.001652	0.012658
<b>Survived</b>	-0.005007	1.000000	-0.338481	-0.077221	-0.035322	0.081629	0.257307

```
df.corr().Survived.sort_values(ascending = False)
```

```
<ipython-input-11-936bc0a2ea37>:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future ve
df.corr().Survived.sort_values(ascending = False)
```

```
Survived    1.000000
Fare         0.257307
Parch       0.081629
PassengerId -0.005007
SibSp       -0.035322
Age         -0.077221
Pclass      -0.338481
Name: Survived, dtype: float64
```

```
df.isnull().any()
```

```
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
```

```
sum(df.Cabin.isnull())
```

```
687
```

```
sum(df.Age.isnull())
```

```
177
```

```
df["Age"].fillna(df["Age"].mean(),inplace=True)
```

```
sum(df.Embarked.isnull())
```

```
2
```

```
df.describe()
```

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
<b>count</b>	891.000000	891.000000	891.000000	891.000000	891.000000	891.000000	891.000000
<b>mean</b>	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208
<b>std</b>	257.353842	0.486592	0.836071	13.002015	1.102743	0.806057	49.693429
<b>min</b>	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
<b>25%</b>	223.500000	0.000000	2.000000	22.000000	0.000000	0.000000	7.910400
<b>50%</b>	446.000000	0.000000	3.000000	29.699118	0.000000	0.000000	14.454200
<b>75%</b>	668.500000	1.000000	3.000000	35.000000	1.000000	0.000000	31.000000
<b>max</b>	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200

```
plt.scatter(df["Fare"],df["Survived"])
```

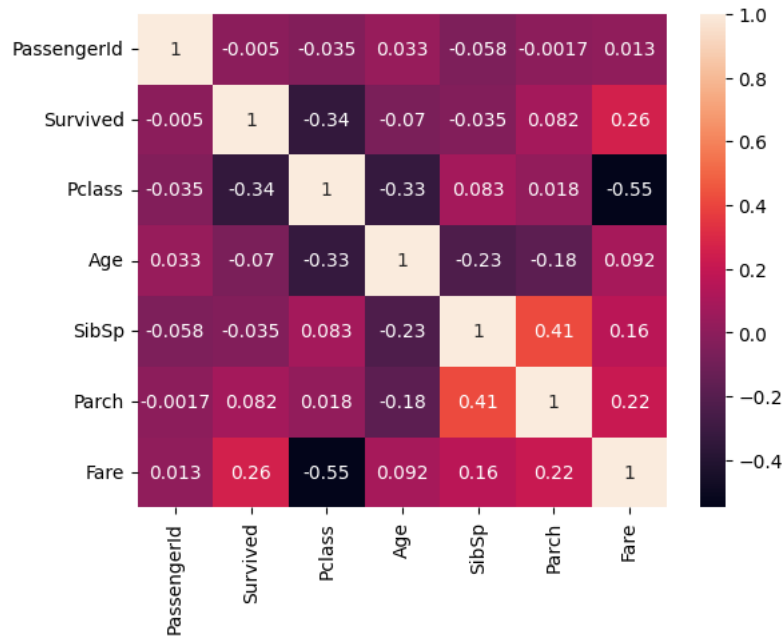
<matplotlib.collections.PathCollection at 0x7c90b8ff11b0>



```
sns.heatmap(df.corr(),annot=True)
```

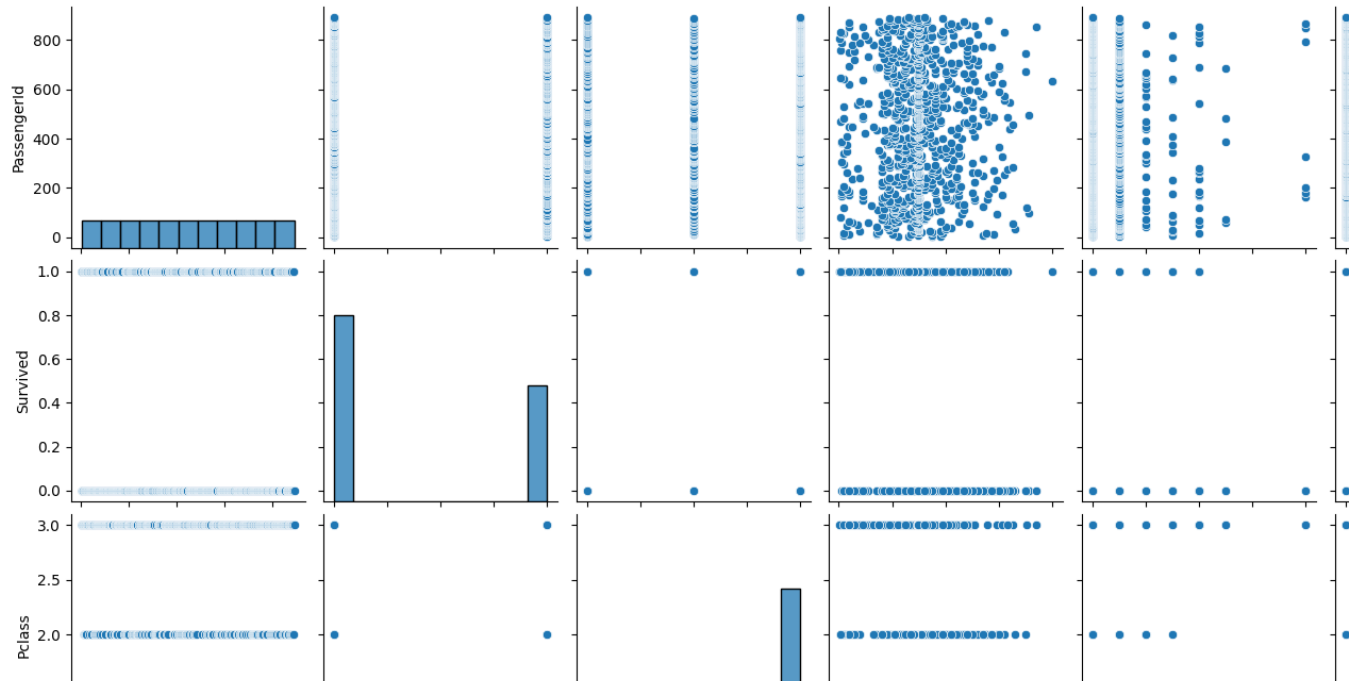
<ipython-input-21-8df7bcac526d>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future ver  
sns.heatmap(df.corr(),annot=True)

<Axes: >



```
sns.pairplot(df)
```

<seaborn.axisgrid.PairGrid at 0x7c90b6e2bac0>



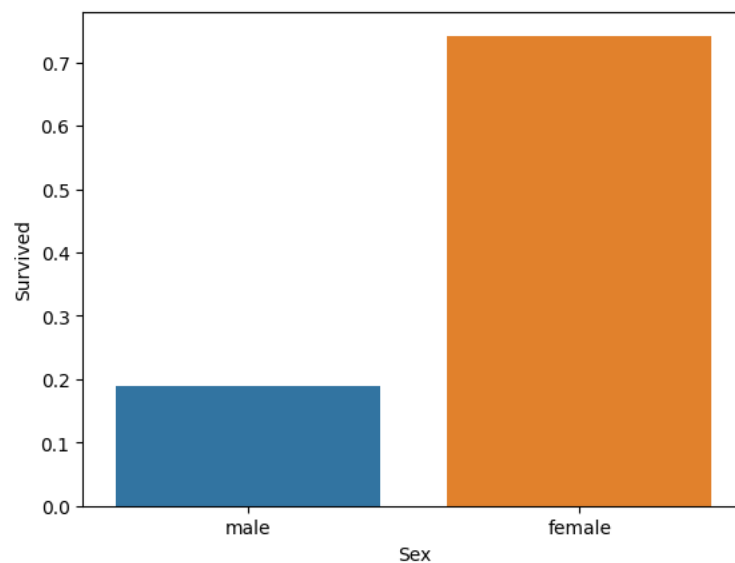
```
sns.barplot(x=df["Sex"],y=df["Survived"],ci=0)
```

<ipython-input-22-8ae461271d98>:1: FutureWarning:

The `ci` parameter is deprecated. Use `errorbar=('ci', 0)` for the same effect.

```
sns.barplot(x=df["Sex"],y=df["Survived"],ci=0)
```

<Axes: xlabel='Sex', ylabel='Survived'>



```
sns.barplot(x=df["Embarked"],y=df["Survived"],ci=0)
```

```
<ipython-input-23-d5b0276940a6>:1: FutureWarning:
```

The `ci` parameter is deprecated. Use `errorbar=('ci', 0)` for the same effect.

```
sns.barplot(x=df["Embarked"],y=df["Survived"],ci=0)
<Axes: xlabel='Embarked', ylabel='Survived'>
```

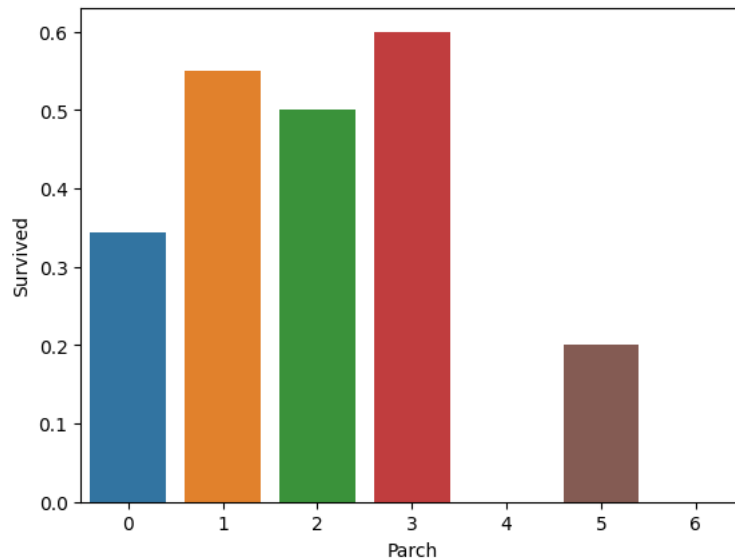


```
sns.barplot(x=df["Parch"],y=df["Survived"],ci=0)
```

```
<ipython-input-26-a1496fefeaf8>:1: FutureWarning:
```

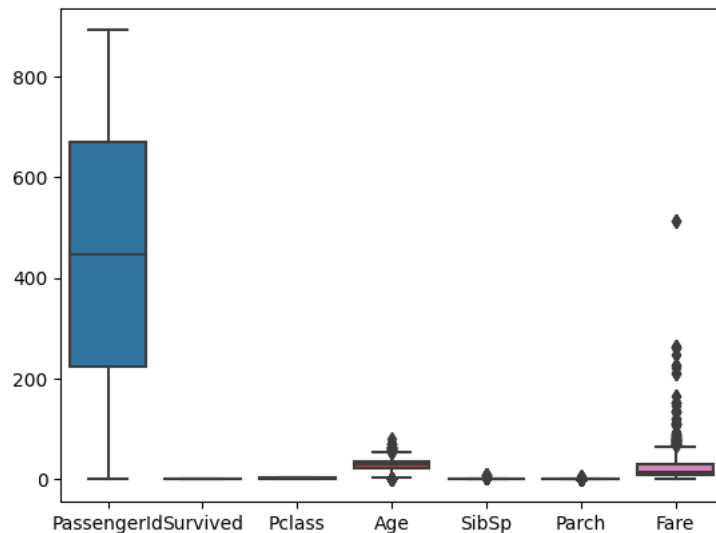
The `ci` parameter is deprecated. Use `errorbar=('ci', 0)` for the same effect.

```
sns.barplot(x=df["Parch"],y=df["Survived"],ci=0)
<Axes: xlabel='Parch', ylabel='Survived'>
```



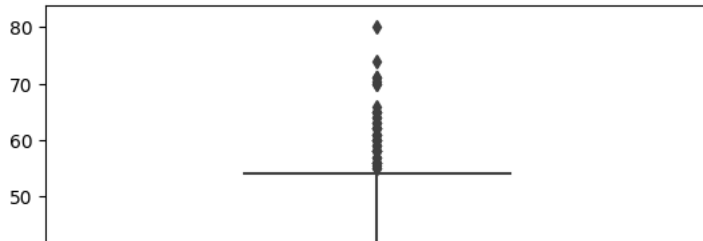
```
sns.boxplot(df)
```

```
<Axes: >
```



```
sns.boxplot(df.Age)
```

&lt;Axes: &gt;



```
Q1 = df['Age'].quantile(0.25)
Q3 = df['Age'].quantile(0.75)
```

```
IQR = Q3 - Q1
```

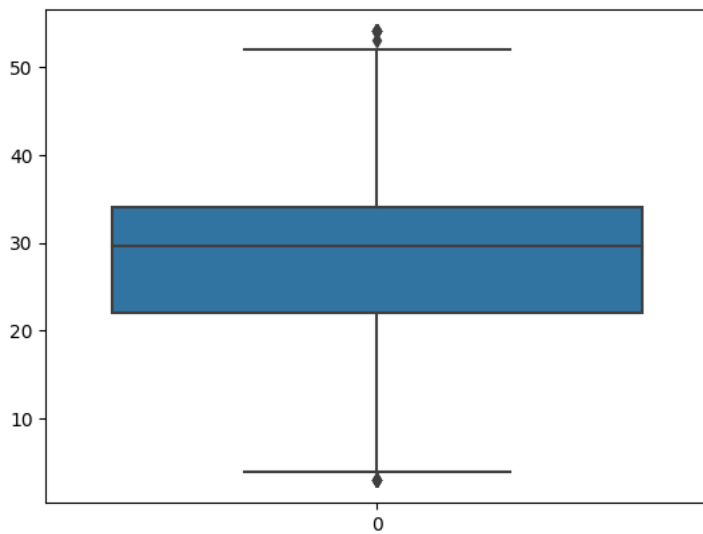
```
threshold = 1.5 * IQR
```

```
df = df[(df['Age'] >= Q1 - threshold) & (df['Age'] <= Q3 + threshold)]
```

```
0 ↓
```

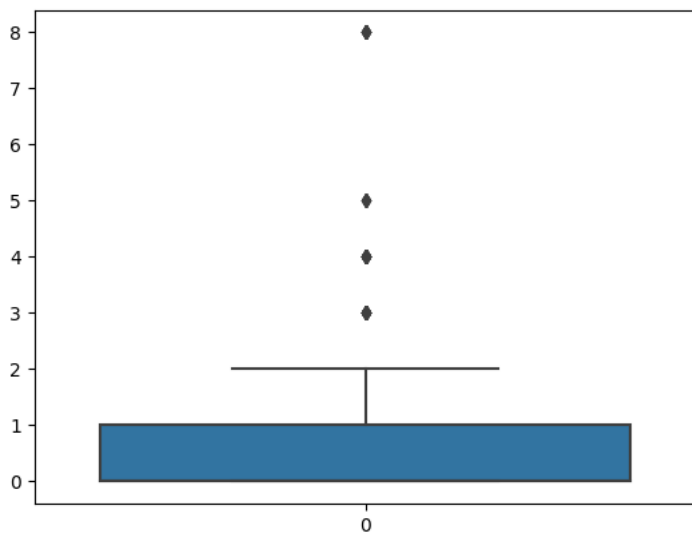
```
sns.boxplot(df.Age)
```

&lt;Axes: &gt;



```
sns.boxplot(df.SibSp)
```

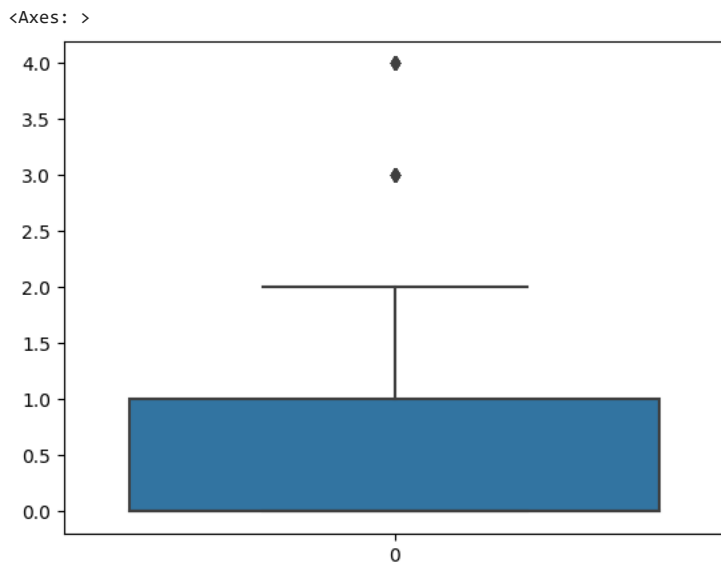
&lt;Axes: &gt;



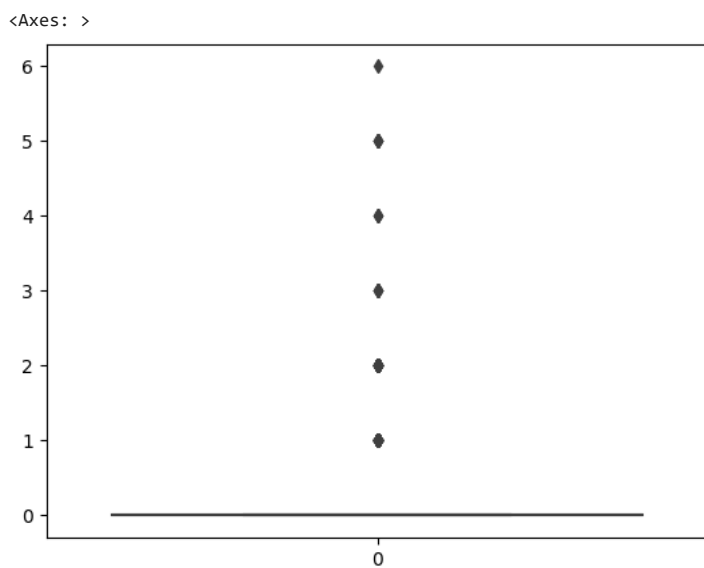
```
p99 = df.SibSp.quantile(0.99)
```

```
df = df[df.SibSp < p99]
```

```
sns.boxplot(df.SibSp)
```



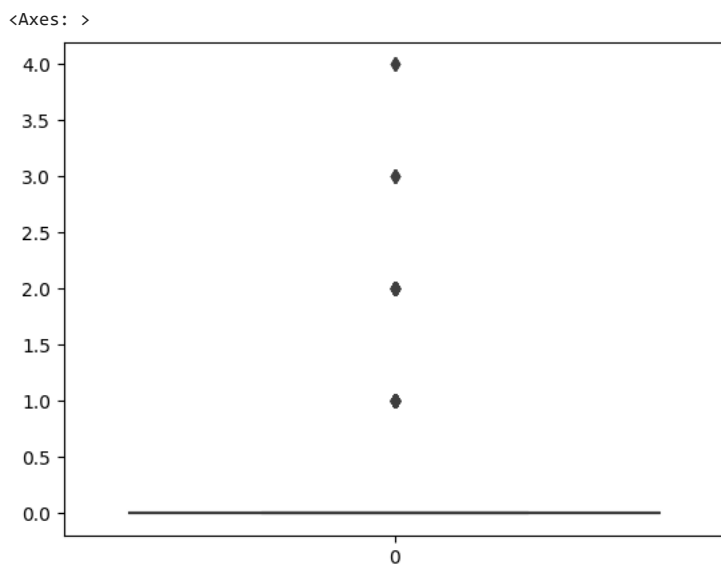
```
sns.boxplot(df.Parch)
```



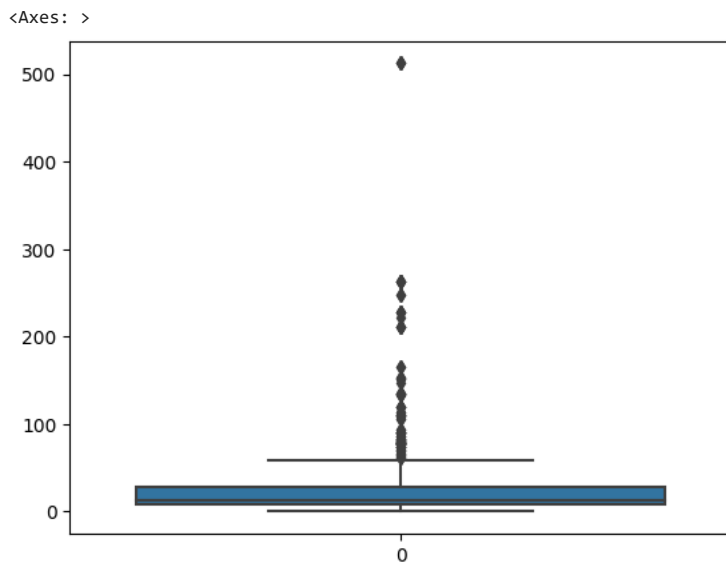
```
p99 = df.Parch.quantile(0.99)
```

```
df = df[df.Parch < p99]
```

```
sns.boxplot(df["Parch"])
```



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



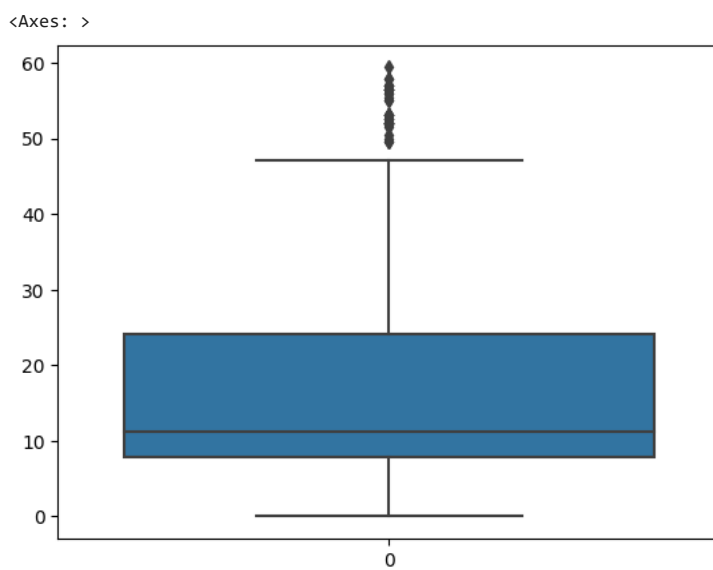
```
Q1 = df['Fare'].quantile(0.25)
Q3 = df['Fare'].quantile(0.75)
```

```
IQR = Q3 - Q1
```

```
threshold = 1.5 * IQR
```

```
df = df[(df['Fare'] >= Q1 - threshold) & (df['Fare'] <= Q3 + threshold)]
```

```
sns.boxplot(df.Fare)
```



```
x = df.drop(columns=["Survived", "PassengerId", "Name", "Ticket", "Cabin"], axis=1) # Independent variables should be in df or 2d array
```

```
x.head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked	
0	3	male	22.000000	1	0	7.2500	S	
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	
5	3	male	29.699118	0	0	8.4583	Q	

```
y = pd.Series(df["Survived"])
```

```
y.head()
```



```
0    0
2    1
3    1
4    0
5    0
Name: Survived, dtype: int64
```

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

```
le = LabelEncoder()
```

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

```
x.head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.000000	1	0	7.2500	S
2	3	0	26.000000	0	0	7.9250	S
3	1	0	35.000000	1	0	53.1000	S
4	3	1	35.000000	0	0	8.0500	S
5	3	1	29.699118	0	0	8.4583	Q

```
print(le.classes_)
```

```
['female' 'male']
```

```
mapping=dict(zip(le.classes_,range(len(le.classes_))))
```

```
mapping
```

```
{'female': 0, 'male': 1}
```

```
le1 = LabelEncoder()
```

```
x["Embarked"] = le1.fit_transform(x["Embarked"])
```

```
x.head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.000000	1	0	7.2500	2
2	3	0	26.000000	0	0	7.9250	2
3	1	0	35.000000	1	0	53.1000	2
4	3	1	35.000000	0	0	8.0500	2
5	3	1	29.699118	0	0	8.4583	1

```
print(le1.classes_)
```

```
['C' 'Q' 'S']
```

```
mapping1=dict(zip(le1.classes_,range(len(le1.classes_))))
```

```
mapping1
```



```
{'C': 0, 'Q': 1, 'S': 2}
```

```
from sklearn.preprocessing import MinMaxScaler
```

```
ms = MinMaxScaler()
```

```
x_Scaled = pd.DataFrame(ms.fit_transform(x),columns = x.columns)
```

```
x_Scaled.head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked	
0	1.0	1.0	0.372549	0.25	0.0	0.122054	1.0	
1	1.0	0.0	0.450980	0.00	0.0	0.133418	1.0	
2	0.0	0.0	0.627451	0.25	0.0	0.893939	1.0	
3	1.0	1.0	0.627451	0.00	0.0	0.135522	1.0	

```
from sklearn.model_selection import train_test_split

x_train,x_test,y_train,y_test = train_test_split(x_Scaled,y,test_size = 0.2,random_state =0)

print(x_train.shape,x_test.shape,y_train.shape,y_test.shape)

(564, 7) (142, 7) (564,) (142,)
```