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

df = pd.read_csv('Titanic-Dataset.csv')
df
```

	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]

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	\
count	891.000000	891.000000	891.000000	714.000000	891.000000	
mean	446.000000	0.383838	2.308642	29.699118	0.523008	
std	257.353842	0.486592	0.836071	14.526497	1.102743	
min	1.000000	0.000000	1.000000	0.420000	0.000000	
25%	223.500000	0.000000	2.000000	20.125000	0.000000	
50%	446.000000	0.000000	3.000000	28.000000	0.000000	
75%	668.500000	1.000000	3.000000	38.000000	1.000000	
max	891.000000	1.000000	3.000000	80.000000	8.000000	

Parch	Fare
-------	------

```
count    891.000000    891.000000
mean      0.381594    32.204208
std       0.806057    49.693429
min       0.000000     0.000000
25%       0.000000     7.910400
50%       0.000000    14.454200
75%       0.000000    31.000000
max       6.000000   512.329200
```

```
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
```

```
df.corr()
```

```
C:\Users\hp\AppData\Local\Temp\ipykernel_4088\1134722465.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.
```

```
df.corr()
```

	PassengerId	Survived	Pclass	Age	SibSp
Parch \					
PassengerId	1.000000	-0.005007	-0.035144	0.036847	-0.057527
Survived	-0.005007	1.000000	-0.338481	-0.077221	-0.035322
Pclass	-0.035144	-0.338481	1.000000	-0.369226	0.083081
Age	0.036847	-0.077221	-0.369226	1.000000	-0.308247
SibSp	-0.057527	-0.035322	0.083081	-0.308247	1.000000
Parch	-0.001652	0.081629	0.018443	-0.189119	0.414838
Fare	0.012658	0.257307	-0.549500	0.096067	0.159651

```

Fare
PassengerId 0.012658
Survived     0.257307
Pclass       -0.549500
Age          0.096067
SibSp        0.159651
Parch        0.216225
Fare         1.000000

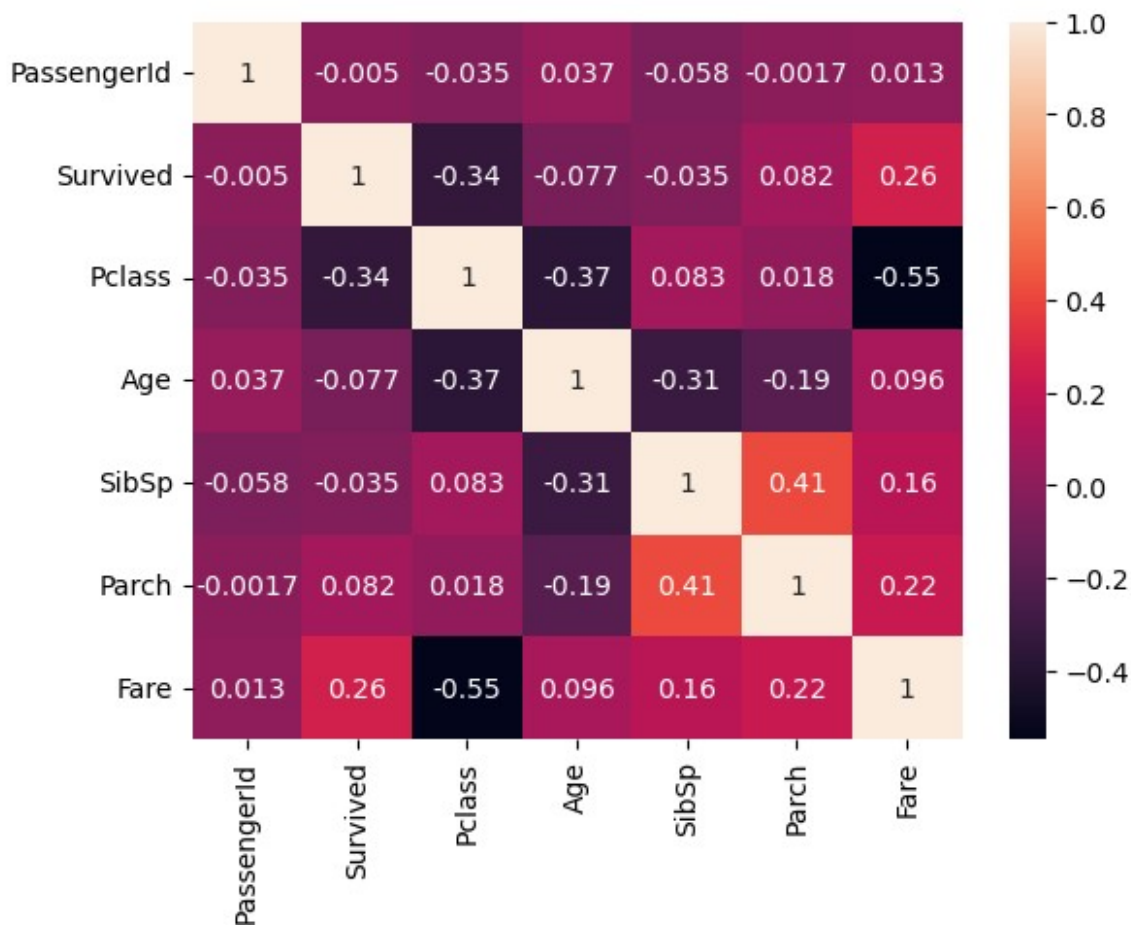
```

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

C:\Users\hp\AppData\Local\Temp\ipykernel_4088\4277794465.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.

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

<Axes: >



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

```
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
```

```
df['Age'].mean()
```

```
29.69911764705882
```

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

```
df
```

	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 \		
0	Braund, Mr. Owen Harris	male
22.000000		
1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female
38.000000		
2	Heikkinen, Miss. Laina	female
26.000000		
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female
35.000000		
4	Allen, Mr. William Henry	male
35.000000		
..
...		

```

886                                Montvila, Rev. Juozas    male
27.000000
887                                Graham, Miss. Margaret Edith  female
19.000000
888                Johnston, Miss. Catherine Helen "Carrie"  female
29.699118
889                                Behr, Mr. Karl Howell    male
26.000000
890                                Dooley, Mr. Patrick    male
32.000000

```

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

```
[891 rows x 12 columns]
```

```
df["Cabin"]=df["Cabin"].fillna(df["Cabin"].mode()[0])
```

```
df
```

	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 \		
0	Braund, Mr. Owen Harris	male
22.000000		
1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female
38.000000		
2	Heikkinen, Miss. Laina	female

```

26.000000
3      Futrelle, Mrs. Jacques Heath (Lily May Peel)  female
35.000000
4      Allen, Mr. William Henry  male
35.000000
..      ...
...
886      Montvila, Rev. Juozas  male
27.000000
887      Graham, Miss. Margaret Edith  female
19.000000
888      Johnston, Miss. Catherine Helen "Carrie"  female
29.699118
889      Behr, Mr. Karl Howell  male
26.000000
890      Dooley, Mr. Patrick  male
32.000000

```

	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	A/5 21171	7.2500	B96 B98	S
1	1	0	PC 17599	71.2833	C85	C
2	0	0	STON/O2. 3101282	7.9250	B96 B98	S
3	1	0	113803	53.1000	C123	S
4	0	0	373450	8.0500	B96 B98	S
..
886	0	0	211536	13.0000	B96 B98	S
887	0	0	112053	30.0000	B42	S
888	1	2	W./C. 6607	23.4500	B96 B98	S
889	0	0	111369	30.0000	C148	C
890	0	0	370376	7.7500	B96 B98	Q

[891 rows x 12 columns]

```
df["Embarked"]=df["Embarked"].fillna(df["Embarked"].mode()[0])
```

```
df
```

	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	

Age \	Name	Sex
0	Braund, Mr. Owen Harris	male
22.000000		
1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female
38.000000		
2	Heikkinen, Miss. Laina	female
26.000000		
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female
35.000000		
4	Allen, Mr. William Henry	male
35.000000		
..
...		
886	Montvila, Rev. Juozas	male
27.000000		
887	Graham, Miss. Margaret Edith	female
19.000000		
888	Johnston, Miss. Catherine Helen "Carrie"	female
29.699118		
889	Behr, Mr. Karl Howell	male
26.000000		
890	Dooley, Mr. Patrick	male
32.000000		

	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	A/5 21171	7.2500	B96 B98	S
1	1	0	PC 17599	71.2833	C85	C
2	0	0	STON/O2. 3101282	7.9250	B96 B98	S
3	1	0	113803	53.1000	C123	S
4	0	0	373450	8.0500	B96 B98	S
..
886	0	0	211536	13.0000	B96 B98	S
887	0	0	112053	30.0000	B42	S
888	1	2	W./C. 6607	23.4500	B96 B98	S
889	0	0	111369	30.0000	C148	C
890	0	0	370376	7.7500	B96 B98	Q

[891 rows x 12 columns]

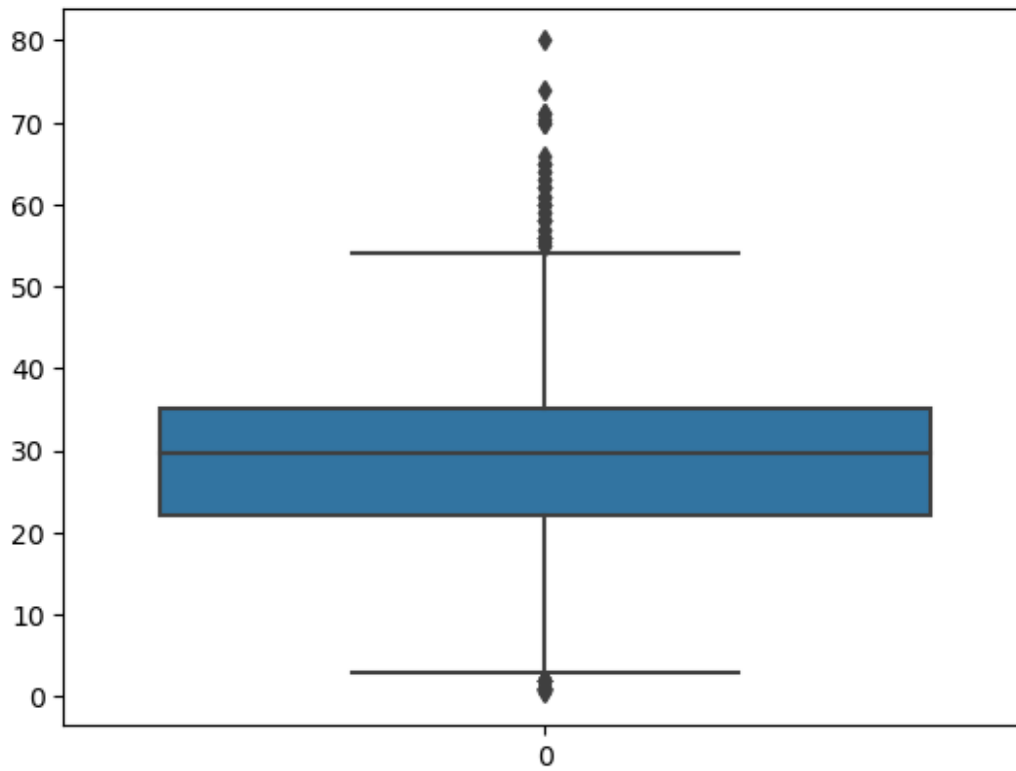
df.isnull().any()

PassengerId	False
Survived	False
Pclass	False
Name	False
Sex	False
Age	False
SibSp	False
Parch	False


```
Ticket      False
Fare        False
Cabin       False
Embarked    False
dtype: bool
```

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

```
<Axes: >
```



```
q1 = df.Age.quantile(0.25)
```

```
q3 = df.Age.quantile(0.75)
```

```
print(q1)
```

```
22.0
```

```
print(q3)
```

```
35.0
```

```
IQR = q3-q1
```

```
IQR
```

```
13.0
```

```
upper_limit = q3+1.5*IQR
upper_limit
```

```
54.5
```

```
lower_limit = q1-1.5*IQR
lower_limit
```

```
2.5
```

```
df.median()
```

```
C:\Users\hp\AppData\Local\Temp\ipykernel_4088\530051474.py:1:
FutureWarning: The default value of numeric_only in DataFrame.median
is deprecated. In a future version, it will default to False. In
addition, specifying 'numeric_only=None' is deprecated. Select only
valid columns or specify the value of numeric_only to silence this
warning.
```

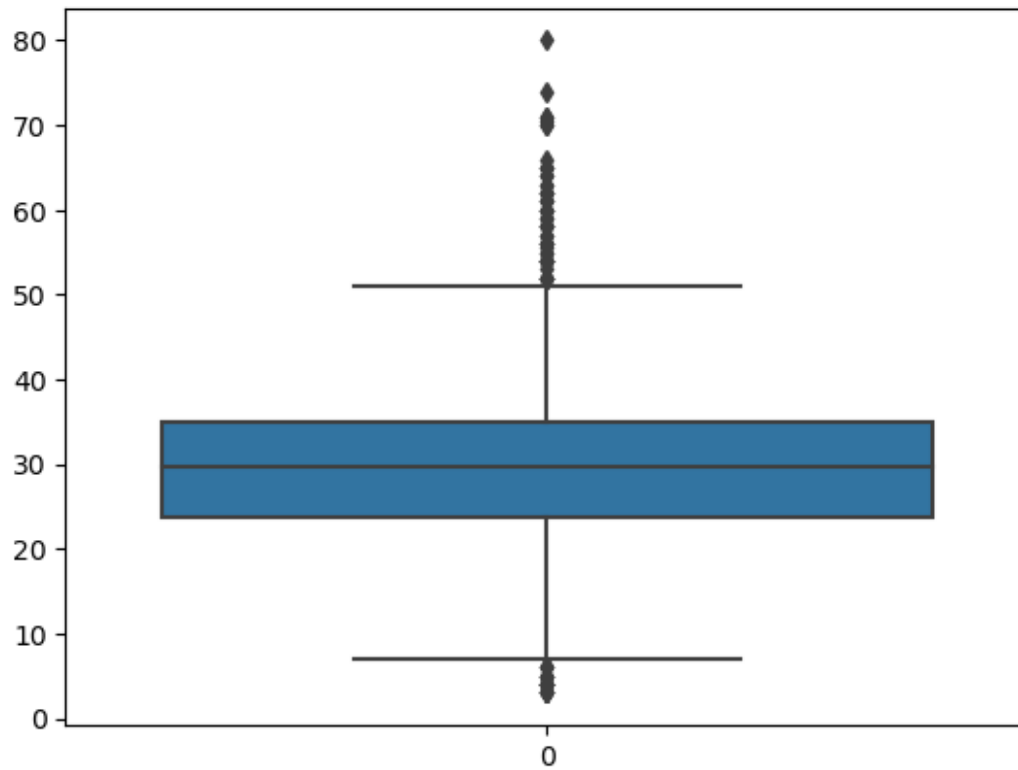
```
df.median()
```

```
PassengerId    446.000000
Survived        0.000000
Pclass         3.000000
Age            29.699118
SibSp          0.000000
Parch          0.000000
Fare           14.454200
dtype: float64
```

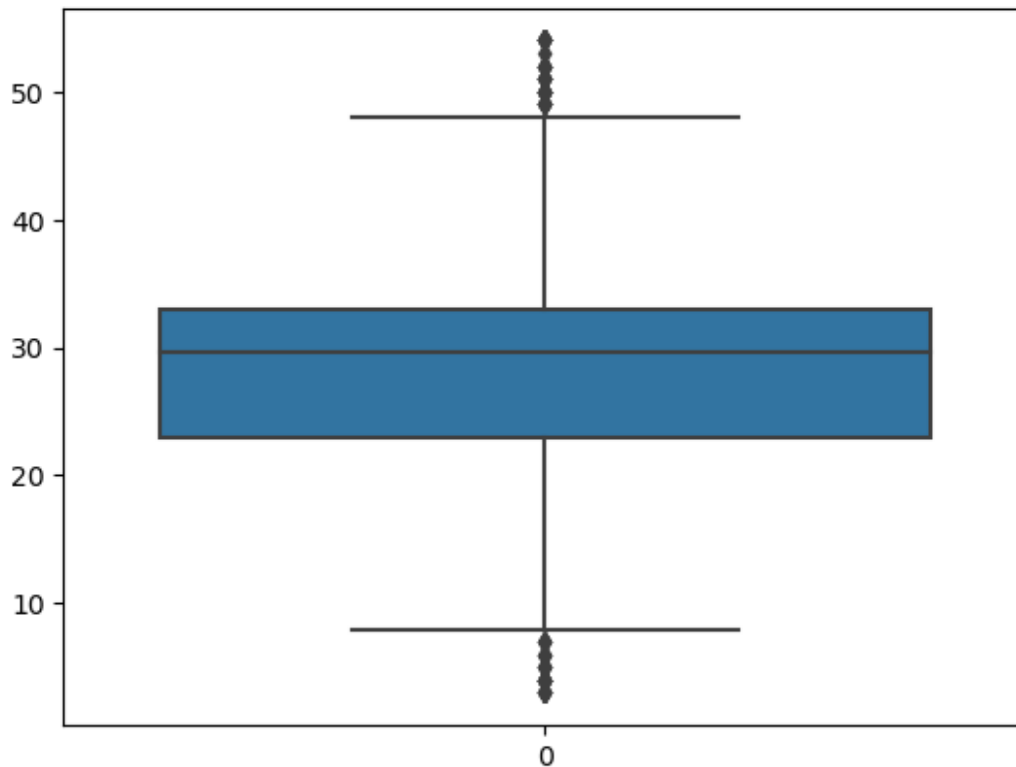
```
df['Age'] = np.where(df['Age']<lower_limit,29.69,df['Age'])
```

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

```
<Axes: >
```



```
df = df[df.Age<upper_limit]
sns.boxplot(df.Age)
<Axes: >
```



```
from scipy import stats
z_scores = np.abs(stats.zscore(df['Age']))
max_threshold=3
outliers = df['Age'][z_scores > max_threshold]

# Print and visualize the outliers
print("Outliers detected using Z-Score:")
print(outliers)

Outliers detected using Z-Score:
Series([], Name: Age, dtype: float64)

from scipy import stats
z_scores = np.abs(stats.zscore(df['Fare']))
max_threshold=3
outliers = df['Fare'][z_scores > max_threshold]

# Print and visualize the outliers
print("Outliers detected using Z-Score:")
print(outliers)

Outliers detected using Z-Score:
27      263.0000
88      263.0000
118     247.5208
258     512.3292
```

```

299    247.5208
311    262.3750
341    263.0000
377    211.5000
380    227.5250
527    221.7792
557    227.5250
679    512.3292
689    211.3375
700    227.5250
716    227.5250
730    211.3375
737    512.3292
742    262.3750
779    211.3375

```

Name: Fare, dtype: float64

```
column_name = 'Fare'
```

```
# Calculate the first quartile (Q1) and third quartile (Q3)
```

```
Q1 = df[column_name].quantile(0.25)
```

```
Q3 = df[column_name].quantile(0.75)
```

```
# Calculate the IQR
```

```
IQR = Q3 - Q1
```

```
# Define the lower and upper bounds for outliers
```

```
lower_bound = Q1 - 1.5 * IQR
```

```
upper_bound = Q3 + 1.5 * IQR
```

```
# Filter rows with values outside the IQR bounds
```

```
df_cleaned = df[(df[column_name] > lower_bound) & (df[column_name]
<upper_bound)]
```

```
# Display the original and cleaned DataFrame sizes
```

```
print(f"Original DataFrame size: {df.shape}")
```

```
print(f"Cleaned DataFrame size: {df_cleaned.shape}")
```

```
df_cleaned
```

Original DataFrame size: (849, 12)

Cleaned DataFrame size: (741, 12)

	PassengerId	Survived	Pclass	\
0	1	0	3	
2	3	1	3	
3	4	1	1	
4	5	0	3	
5	6	0	3	
...	
886	887	0	2	

887	888	1	1
888	889	0	3
889	890	1	1
890	891	0	3

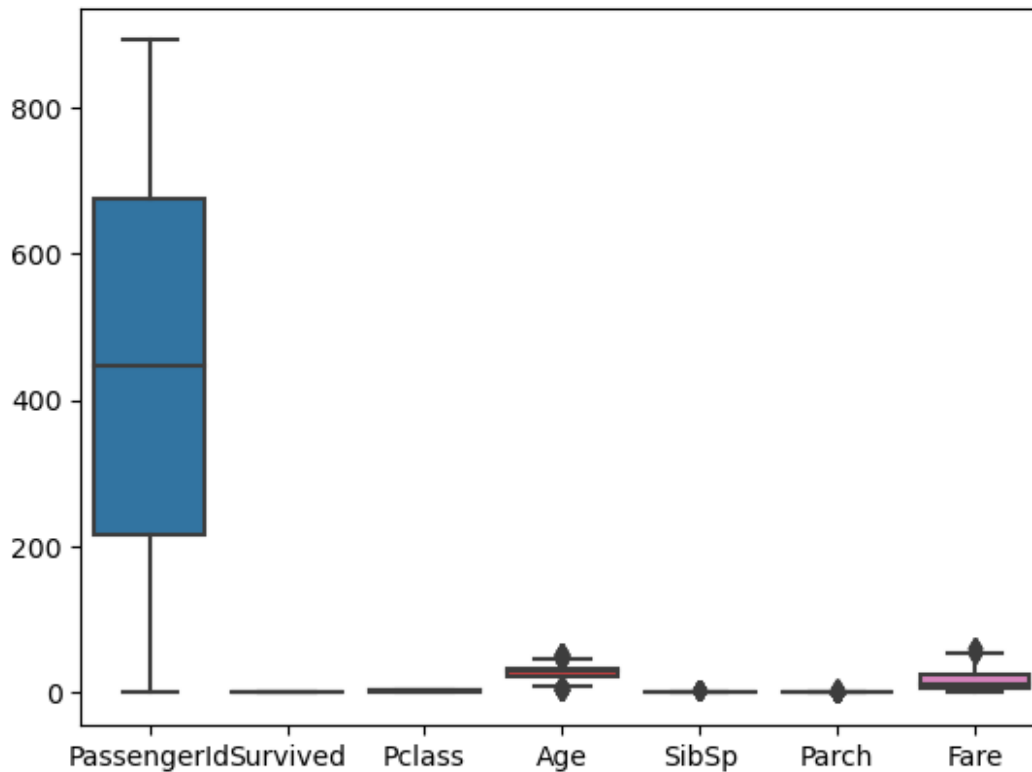
SibSp \	Name	Sex	Age
0	Braund, Mr. Owen Harris	male	22.000000
1			
2	Heikkinen, Miss. Laina	female	26.000000
0			
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.000000
1			
4	Allen, Mr. William Henry	male	35.000000
0			
5	Moran, Mr. James	male	29.699118
0			
..
...			
886	Montvila, Rev. Juozas	male	27.000000
0			
887	Graham, Miss. Margaret Edith	female	19.000000
0			
888	Johnston, Miss. Catherine Helen "Carrie"	female	29.699118
1			
889	Behr, Mr. Karl Howell	male	26.000000
0			
890	Dooley, Mr. Patrick	male	32.000000
0			

	Parch	Ticket	Fare	Cabin	Embarked
0	0	A/5 21171	7.2500	B96 B98	S
2	0	STON/O2. 3101282	7.9250	B96 B98	S
3	0	113803	53.1000	C123	S
4	0	373450	8.0500	B96 B98	S
5	0	330877	8.4583	B96 B98	Q
..
886	0	211536	13.0000	B96 B98	S
887	0	112053	30.0000	B42	S
888	2	W./C. 6607	23.4500	B96 B98	S
889	0	111369	30.0000	C148	C
890	0	370376	7.7500	B96 B98	Q

[741 rows x 12 columns]

sns.boxplot(df_cleaned)

<Axes: >



Separating dependent and independent variables

```
x = df.iloc[:, [2, 4, 5, 6, 7, 9, 11]]
```

```
y = df.iloc[:, 1]
```

```
x.head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	male	22.0	1	0	7.2500	S
1	1	female	38.0	1	0	71.2833	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

```
y.head()
```

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

```
Name: Survived, dtype: int64
```

```
df.shape
```

```
(849, 12)
```

```
x.shape
```

```
(849, 7)
```

```
y.shape
```

```
(849,)
```

One hot encoding on Embarked column

```
x.Embarked.value_counts()
```

```
S    618
```

```
C    157
```

```
Q     74
```

```
Name: Embarked, dtype: int64
```

```
embarked=pd.get_dummies(x["Embarked"],drop_first=True)
```

```
embarked
```

	Q	S
0	0	1
1	0	0
2	0	1
3	0	1
4	0	1
..
886	0	1
887	0	1
888	0	1
889	0	0
890	1	0

```
[849 rows x 2 columns]
```

```
x=pd.concat([x,embarked],axis=1)
```

```
x.head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked	Q	S
0	3	male	22.0	1	0	7.2500	S	0	1
1	1	female	38.0	1	0	71.2833	C	0	0
2	3	female	26.0	0	0	7.9250	S	0	1
3	1	female	35.0	1	0	53.1000	S	0	1
4	3	male	35.0	0	0	8.0500	S	0	1

```
x.drop(["Embarked"],axis=1,inplace=True)
```

```
x.head(10)
```


	Pclass	Sex	Age	SibSp	Parch	Fare	Q	S
0	3	male	22.000000	1	0	7.2500	0	1
1	1	female	38.000000	1	0	71.2833	0	0
2	3	female	26.000000	0	0	7.9250	0	1
3	1	female	35.000000	1	0	53.1000	0	1
4	3	male	35.000000	0	0	8.0500	0	1
5	3	male	29.699118	0	0	8.4583	1	0
6	1	male	54.000000	0	0	51.8625	0	1
7	3	male	29.690000	3	1	21.0750	0	1
8	3	female	27.000000	0	2	11.1333	0	1
9	2	female	14.000000	1	0	30.0708	0	0

```
x.shape
```

```
(849, 8)
```

Label Encoding on sex column

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

```
le=LabelEncoder()
```

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

```
x["Sex"]
```

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

```
..
```

```
886    1
887    0
888    0
889    1
890    1
```

```
Name: Sex, Length: 849, dtype: int32
```

```
x["Sex"].value_counts()
```

```
1    545
0    304
```

```
Name: Sex, dtype: int64
```

```
x["Sex"].nunique()
```

```
2
```

```
x.head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Q	S
0	3	1	22.0	1	0	7.2500	0	1
1	1	0	38.0	1	0	71.2833	0	0
2	3	0	26.0	0	0	7.9250	0	1
3	1	0	35.0	1	0	53.1000	0	1
4	3	1	35.0	0	0	8.0500	0	1

```
x.Sex.value_counts()
```

```
1    545
0    304
Name: Sex, dtype: int64
```

Splitting into training and testing set

```
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)
```

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

```
((594, 8), (255, 8), (594,), (255,))
```

Feature Scaling

```
from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
```

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

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

```
x_train
```

```
array([[ -0.41834557, -1.31330524,  1.17453394, ..., -0.24582194,
        -0.32262861,  0.62017367],
       [  0.79383622, -1.31330524,  0.0443624 , ..., -0.49093682,
        -0.32262861, -1.61245155],
       [  0.79383622,  0.76143761,  0.0443624 , ..., -0.48045329,
         3.09953914, -1.61245155],
       ...,
       [  0.79383622,  0.76143761,  0.67449708, ..., -0.48552931,
        -0.32262861,  0.62017367],
       [  0.79383622,  0.76143761, -0.72560612, ..., -0.4744627 ,
        -0.32262861,  0.62017367],
       [  0.79383622,  0.76143761, -1.22564298, ..., -0.49434746,
        -0.32262861,  0.62017367]])
```