

# ASSIGNMENT-3

## WEEK-3

**NAME: N. SUSRUTHA HASINI**

**MORNING SESSION (10:00 – 12:00)**

### Perform Data preprocessing on Titanic dataset

#### Data Preprocessing

- o Import the Libraries.
- o Importing the dataset.
- o Checking for Null Values.
- o Data Visualization.
- o Outlier Detection
- o Splitting Dependent and Independent variables
- o Perform Encoding
- o Splitting Data into Train and Test
- o Feature Scaling.

#### o Import the Libraries.

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

#### o Importing the dataset.

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

```
In [3]:
df.head()
```

Out[3]:

	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	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence	female	38.0	1	0	PC 17599	71.2833	C85	C

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
				Briggs Th...								
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

In [4]:  
df.describe()

Out[4]:

	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

In [5]:  
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

```
In [6]:
df.corr()

/tmp/ipykernel_8747/1134722465.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()
```

Out[6]:

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
PassengerId	1.000000	-0.005007	-0.035144	0.036847	-0.057527	-0.001652	0.012658
Survived	-0.005007	1.000000	-0.338481	-0.077221	-0.035322	0.081629	0.257307
Pclass	-0.035144	-0.338481	1.000000	-0.369226	0.083081	0.018443	-0.549500
Age	0.036847	-0.077221	-0.369226	1.000000	-0.308247	-0.189119	0.096067
SibSp	-0.057527	-0.035322	0.083081	-0.308247	1.000000	0.414838	0.159651
Parch	-0.001652	0.081629	0.018443	-0.189119	0.414838	1.000000	0.216225
Fare	0.012658	0.257307	-0.549500	0.096067	0.159651	0.216225	1.000000

```
In [7]:
df.corr().Parch.sort_values(ascending=False)

/tmp/ipykernel_8747/1925834096.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().Parch.sort_values(ascending=False)
```

Out[7]:

```
Parch          1.000000
SibSp          0.414838
Fare           0.216225
Survived       0.081629
Pclass         0.018443
PassengerId   -0.001652
Age           -0.189119
Name: Parch, dtype: float64
```

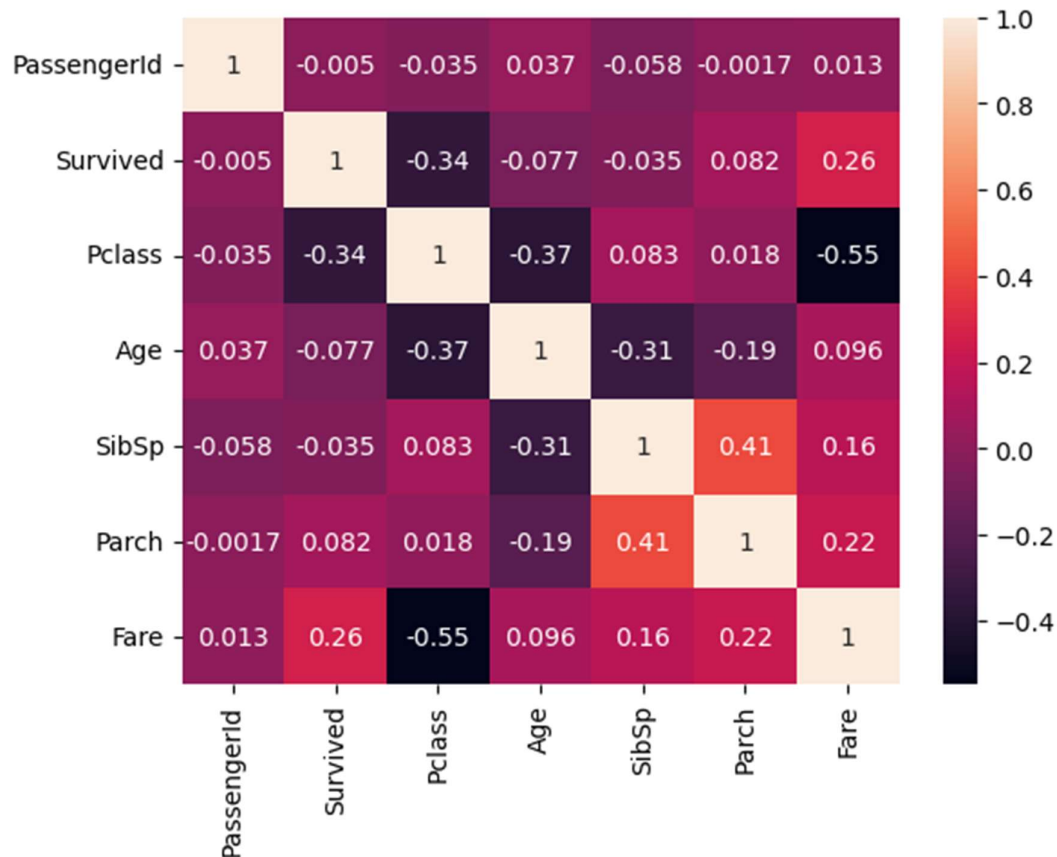
```
In [8]:
sns.heatmap(df.corr(),annot=True)

/tmp/ipykernel_8747/4277794465.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)
```

Out[8]:

```
<Axes: >
```



## o Checking for null values

```
In [9]:
df.isnull().sum()
```

```
Out[9]:
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 [10]:
df.drop(["Cabin"],axis=1,inplace=True)
```

```
In [11]:
mean_age = df['Age'].mean()
mean_age
```

```
Out[11]:
```

29.69911764705882

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

```
In [13]:
mode_embarked = df["Embarked"].mode()[0]
mode_embarked
```

```
Out[13]:
'S'
```

```
In [14]:
df["Embarked"].fillna(mode_embarked, inplace=True)
```

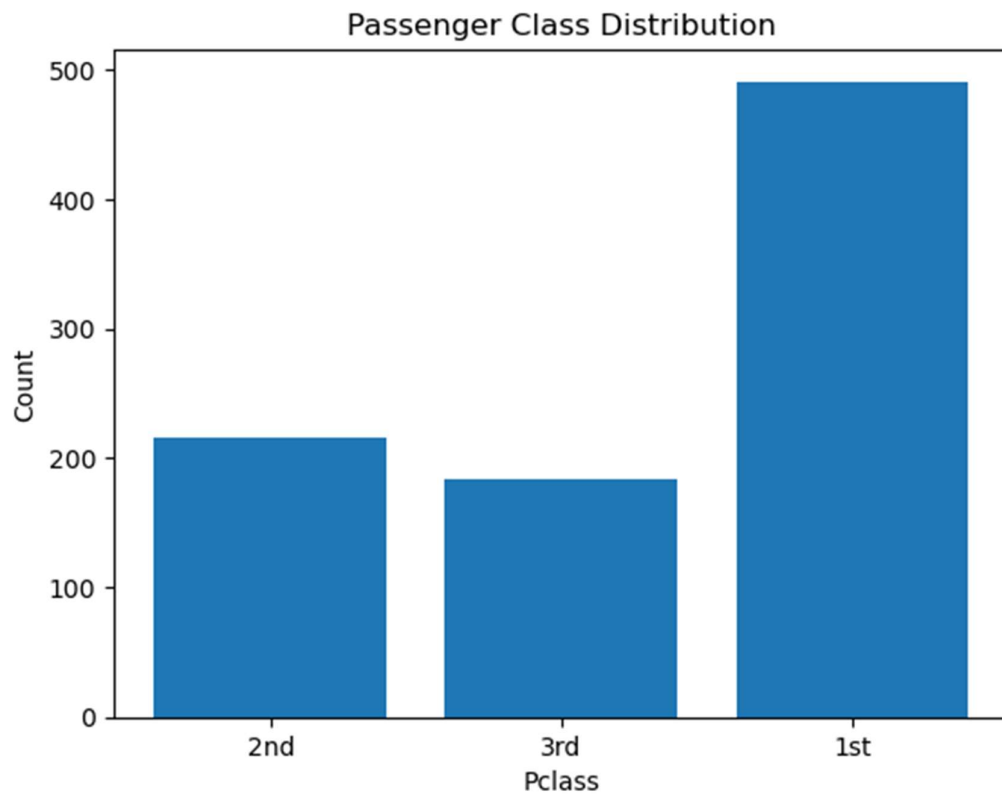
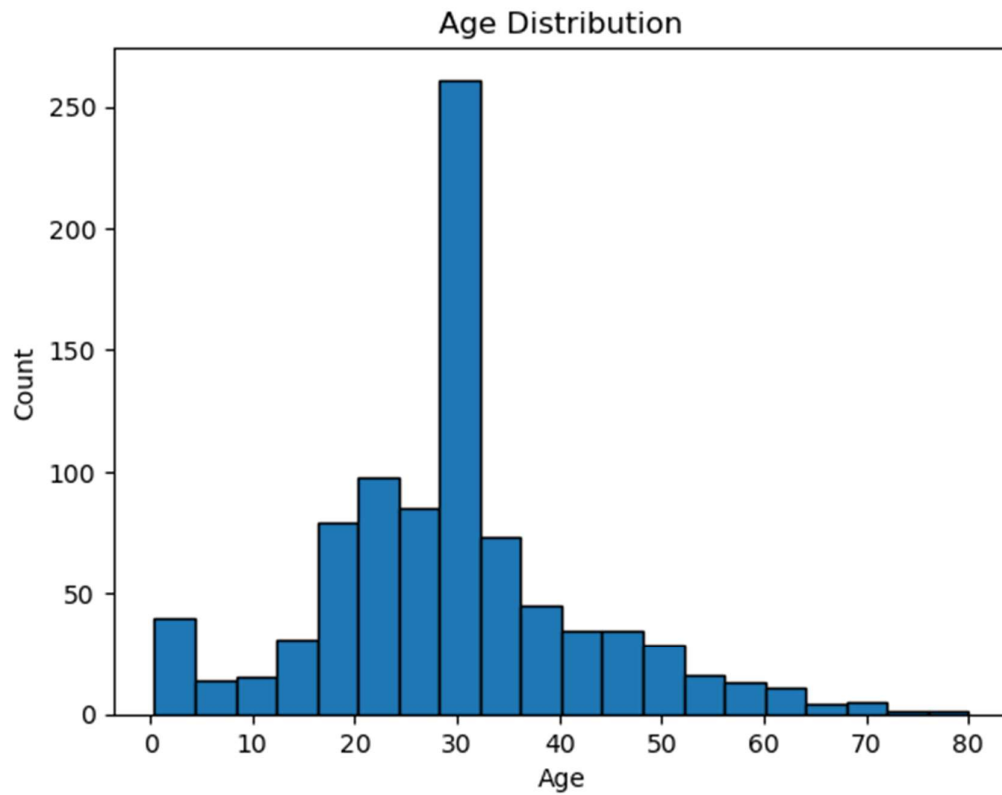
```
In [15]:
df.isnull().sum()
```

```
Out[15]:
PassengerId      0
Survived          0
Pclass           0
Name             0
Sex              0
Age              0
SibSp            0
Parch            0
Ticket           0
Fare             0
Embarked         0
dtype: int64
```

## o Data visualization

```
In [16]:
# Example 1: Create a histogram of the 'Age' column
plt.hist(df['Age'], bins=20, edgecolor='k')
plt.xlabel('Age')
plt.ylabel('Count')
plt.title('Age Distribution')
plt.show()
```

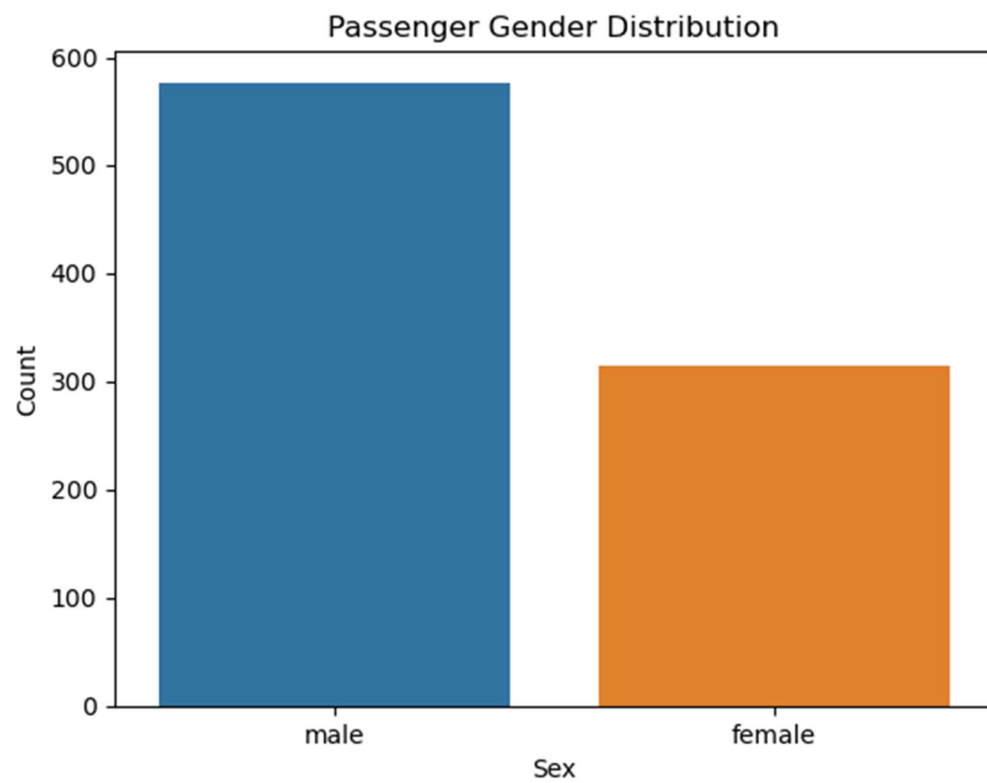
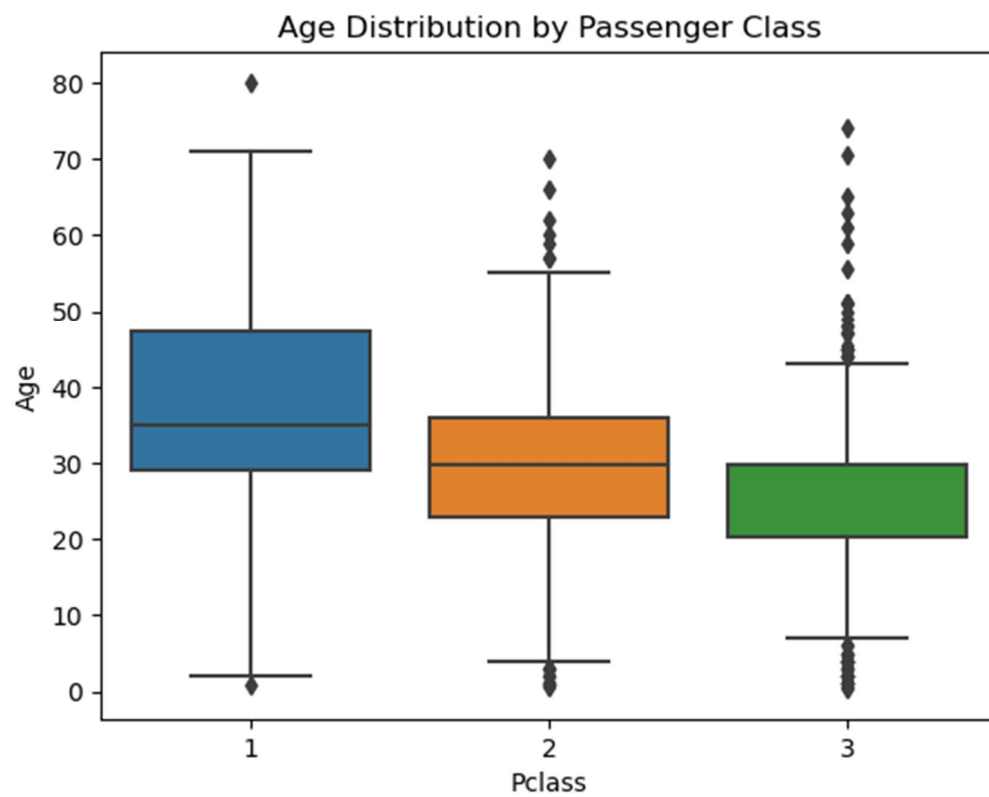
```
# Example 2: Create a bar chart for the 'Pclass' column
pclass_counts = df['Pclass'].value_counts()
plt.bar(pclass_counts.index, pclass_counts.values)
plt.xlabel('Pclass')
plt.ylabel('Count')
plt.title('Passenger Class Distribution')
plt.xticks(pclass_counts.index, labels=['1st', '2nd', '3rd'])
plt.show()
```



```
In [17]:  
# Example 1: Create a box plot of 'Age' by 'Pclass'  
sns.boxplot(x='Pclass', y='Age', data=df)
```

```
plt.xlabel('Pclass')  
plt.ylabel('Age')  
plt.title('Age Distribution by Passenger Class')  
plt.show()
```

```
# Example 2: Create a countplot of 'Sex'  
sns.countplot(x='Sex', data=df)  
plt.xlabel('Sex')  
plt.ylabel('Count')  
plt.title('Passenger Gender Distribution')  
plt.show()
```





```

In [18]:
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 11 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            891 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  Embarked       891 non-null    object
dtypes: float64(2), int64(5), object(4)
memory usage: 76.7+ KB

```

```

In [19]:
df.drop(["PassengerId", "Name", "Ticket"], axis=1, inplace=True)

```

```

In [20]:
df.head()

```

Out[20]:

	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	71.2833	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

## o Outlier Detection

```

In [21]:
# Select numerical columns from the dataset
numerical_attributes = df.select_dtypes(include=['int64', 'float64'])

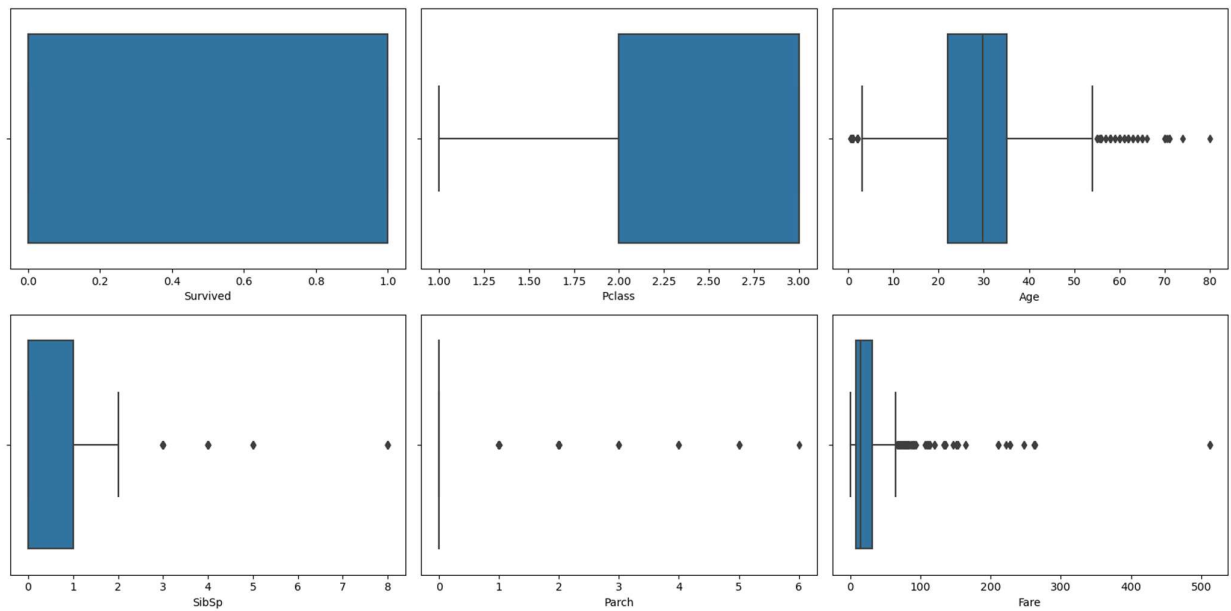
# Create box plots for each numerical attribute
plt.figure(figsize=(16, 8)) # Adjust the figure size for better
visualization

for i, column in enumerate(numerical_attributes.columns):
    plt.subplot(2, 3, i+1) # Create subplots in a 2x3 grid
    sns.boxplot(x=column, data=numerical_attributes, whis=1.5) # Adjust
    whis as needed

plt.tight_layout()

```

```
plt.show()
```



## o Removal of Outliers

In [22]:

```
# Load your dataset into a pandas DataFrame (assuming your dataset is loaded as 'df')
```

```
# Define the numerical attributes
```

```
numerical_attributes = ['Age', 'SibSp', 'Parch', 'Fare']
```

```
# Define a function to detect and potentially remove outliers
```

```
def detect_and_remove_outliers(df, column):
```

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

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

```
    IQR = Q3 - Q1
```

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

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

```
    outliers = df[(df[column] < lower_bound) | (df[column] > upper_bound)]
```

```
    return outliers
```

```
# Detect and potentially remove outliers for each numerical attribute
```

```
outliers_dict = {}
```

```
for attribute in numerical_attributes:
```

```
    outliers = detect_and_remove_outliers(df, attribute)
```

```
    outliers_dict[attribute] = outliers
```

```
# Print the detected outliers for each numerical attribute
```

```
for attribute, outliers in outliers_dict.items():
```

```
    print(f"Outliers in {attribute}:")
```

```
    print(outliers)
```

```
# Optionally, remove the outliers from the DataFrame
for attribute, outliers in outliers_dict.items():
    df = df[~df.index.isin(outliers.index)]
```

```
# Now, the DataFrame 'df' contains the data with potential outliers
removed.
```

```
Outliers in Age:
```

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
7	0	3	male	2.00	3	1	21.0750	S
11	1	1	female	58.00	0	0	26.5500	S
15	1	2	female	55.00	0	0	16.0000	S
16	0	3	male	2.00	4	1	29.1250	Q
33	0	2	male	66.00	0	0	10.5000	S
..	...	...	...	...	...	...	...	...
827	1	2	male	1.00	0	2	37.0042	C
829	1	1	female	62.00	0	0	80.0000	S
831	1	2	male	0.83	1	1	18.7500	S
851	0	3	male	74.00	0	0	7.7750	S
879	1	1	female	56.00	0	1	83.1583	C

```
[66 rows x 8 columns]
```

```
Outliers in SibSp:
```

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
7	0	3	male	2.000000	3	1	21.0750	S
16	0	3	male	2.000000	4	1	29.1250	Q
24	0	3	female	8.000000	3	1	21.0750	S
27	0	1	male	19.000000	3	2	263.0000	S
50	0	3	male	7.000000	4	1	39.6875	S
59	0	3	male	11.000000	5	2	46.9000	S
63	0	3	male	4.000000	3	2	27.9000	S
68	1	3	female	17.000000	4	2	7.9250	S
71	0	3	female	16.000000	5	2	46.9000	S
85	1	3	female	33.000000	3	0	15.8500	S
88	1	1	female	23.000000	3	2	263.0000	S
119	0	3	female	2.000000	4	2	31.2750	S
159	0	3	male	29.699118	8	2	69.5500	S
164	0	3	male	1.000000	4	1	39.6875	S
171	0	3	male	4.000000	4	1	29.1250	Q
176	0	3	male	29.699118	3	1	25.4667	S
180	0	3	female	29.699118	8	2	69.5500	S
182	0	3	male	9.000000	4	2	31.3875	S
201	0	3	male	29.699118	8	2	69.5500	S
229	0	3	female	29.699118	3	1	25.4667	S
233	1	3	female	5.000000	4	2	31.3875	S
261	1	3	male	3.000000	4	2	31.3875	S
266	0	3	male	16.000000	4	1	39.6875	S
278	0	3	male	7.000000	4	1	29.1250	Q
324	0	3	male	29.699118	8	2	69.5500	S
341	1	1	female	24.000000	3	2	263.0000	S

374	0	3	female	3.000000	3	1	21.0750	S
386	0	3	male	1.000000	5	2	46.9000	S
409	0	3	female	29.699118	3	1	25.4667	S
480	0	3	male	9.000000	5	2	46.9000	S
485	0	3	female	29.699118	3	1	25.4667	S
541	0	3	female	9.000000	4	2	31.2750	S
542	0	3	female	11.000000	4	2	31.2750	S
634	0	3	female	9.000000	3	2	27.9000	S
642	0	3	female	2.000000	3	2	27.9000	S
683	0	3	male	14.000000	5	2	46.9000	S
686	0	3	male	14.000000	4	1	39.6875	S
726	1	2	female	30.000000	3	0	21.0000	S
787	0	3	male	8.000000	4	1	29.1250	Q
792	0	3	female	29.699118	8	2	69.5500	S
813	0	3	female	6.000000	4	2	31.2750	S
819	0	3	male	10.000000	3	2	27.9000	S
824	0	3	male	2.000000	4	1	39.6875	S
846	0	3	male	29.699118	8	2	69.5500	S
850	0	3	male	4.000000	4	2	31.2750	S
863	0	3	female	29.699118	8	2	69.5500	S

Outliers in Parch:

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
7	0	3	male	2.000000	3	1	21.0750	S
8	1	3	female	27.000000	0	2	11.1333	S
10	1	3	female	4.000000	1	1	16.7000	S
13	0	3	male	39.000000	1	5	31.2750	S
16	0	3	male	2.000000	4	1	29.1250	Q
..	...	...	...	...	...	...	...	...
871	1	1	female	47.000000	1	1	52.5542	S
879	1	1	female	56.000000	0	1	83.1583	C
880	1	2	female	25.000000	0	1	26.0000	S
885	0	3	female	39.000000	0	5	29.1250	Q
888	0	3	female	29.699118	1	2	23.4500	S

[213 rows x 8 columns]

Outliers in Fare:

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
1	1	1	female	38.000000	1	0	71.2833	C
27	0	1	male	19.000000	3	2	263.0000	S
31	1	1	female	29.699118	1	0	146.5208	C
34	0	1	male	28.000000	1	0	82.1708	C
52	1	1	female	49.000000	1	0	76.7292	C
..	...	...	...	...	...	...	...	...
846	0	3	male	29.699118	8	2	69.5500	S
849	1	1	female	29.699118	1	0	89.1042	C
856	1	1	female	45.000000	1	1	164.8667	S
863	0	3	female	29.699118	8	2	69.5500	S
879	1	1	female	56.000000	0	1	83.1583	C

[116 rows x 8 columns]

## o Comparing the Before and After removing the outliers

In [23]:

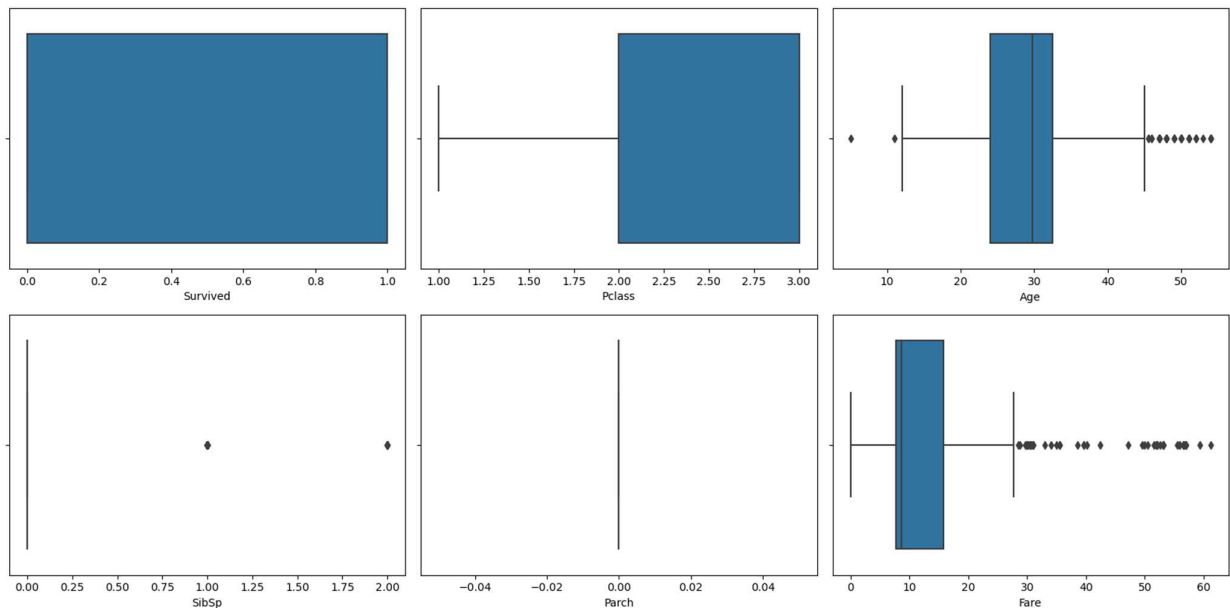
```
import matplotlib.pyplot as plt
import seaborn as sns

# Select numerical columns from the dataset
numerical_attributes = df.select_dtypes(include=['int64', 'float64'])

# Create box plots for each numerical attribute
plt.figure(figsize=(16, 8)) # Adjust the figure size for better
visualization

for i, column in enumerate(numerical_attributes.columns):
    plt.subplot(2, 3, i+1) # Create subplots in a 2x3 grid
    sns.boxplot(x=column, data=numerical_attributes, whis=1.5) # Adjust
    whis as needed

plt.tight_layout()
plt.show()
```



In [24]:

```
df.head()
```

Out[24]:

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

## o Splitting the Dependent and Independent variables

In [25]:

```
# Assuming 'df' is your DataFrame containing the dataset
X = df.drop('Survived', axis=1) # Independent variables
y = df['Survived'] # Dependent variable
```

In [26]:

X.head()

Out[26]:

	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

In [27]:

y.head()

Out[27]:

0 0

2 1

3 1

4 0

5 0

Name: Survived, dtype: int64

## o Perform Encoding

In [28]:

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

In [29]:

```
le=LabelEncoder()
```

In [30]:

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

In [31]:

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

In [32]:

Embarked

Out[32]:

	Q	S
0	0	1
2	0	1
3	0	1
4	0	1
5	1	0
...	...	...
884	0	1
886	0	1

	Q	S
887	0	1
889	0	0
890	1	0

577 rows × 2 columns

In [33]:

```
X=pd.concat([X,Embarked],axis=1)
```

In [34]:

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

## o Splitting Data into Train and Test

In [35]:

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=2)
```

In [36]:

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

Out[36]:

```
((461, 8), (116, 8), (461,), (116,))
```

## o Feature Scaling

In [37]:

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

In [38]:

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

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

In [39]:

```
x_train
```

Out[39]:

```
array([[ 0.6931394 , -1.76523977,  0.0155626 , ..., -0.04945919,
        -0.34879005, -1.74466606],
       [ 0.6931394 ,  0.56649528, -1.4470116 , ..., -0.5034821 ,
        -0.34879005,  0.57317559],
       [ 0.6931394 ,  0.56649528,  0.16538695, ..., -0.5612962 ,
        -0.34879005,  0.57317559],
       ...,
       [ 0.6931394 ,  0.56649528,  0.0155626 , ...,  0.7106139 ,
         2.86705424, -1.74466606],
       [-2.03794831,  0.56649528,  0.0155626 , ..., -1.18255274,
        -0.34879005,  0.57317559],
       [ 0.6931394 , -1.76523977,  0.0155626 , ..., -0.57501479,
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In [40]:

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x_test
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Out[40]:

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