21BCE8975 - Assignment-3

September 21, 2023

1 Assignment-3

Name: Chigulla Somu Vinay Raj

Registration number: 21BCE8975

Email: vinayraj.21bce8975@vitapstudent.ac.in

Mobile Number: 9491604860

2 1. IMPORT THE LIBRARIES

[1]:
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
```

3 2. IMPORT THE DATASET

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

[3]:
```
df
```

[3]:

	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]
```

[4]:
```
df.head()
```

[4]:
```
   PassengerId  Survived  Pclass  \
0            1         0       3
1            2         1       1
2            3         1       3
3            4         1       1
4            5         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

   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
```

[5]:
```
df.tail()
```

[5]:
```
     PassengerId  Survived  Pclass                                      Name  \
886          887         0       2                     Montvila, Rev. Juozas
887          888         1       1              Graham, Miss. Margaret Edith
888          889         0       3  Johnston, Miss. Catherine Helen "Carrie"
889          890         1       1                     Behr, Mr. Karl Howell
890          891         0       3                       Dooley, Mr. Patrick

        Sex   Age  SibSp  Parch      Ticket   Fare Cabin Embarked
886    male  27.0      0      0      211536  13.00   NaN        S
887  female  19.0      0      0      112053  30.00   B42        S
888  female   NaN      1      2  W./C. 6607  23.45   NaN        S
889    male  26.0      0      0      111369  30.00  C148        C
890    male  32.0      0      0      370376   7.75   NaN        Q
```

[6]:
```
df.shape
```

[6]:
```
(891, 12)
```

[7]:
```
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
```

[8]:
```
df.describe()
```

[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

[9]:
```
corr=df.corr()
corr
```

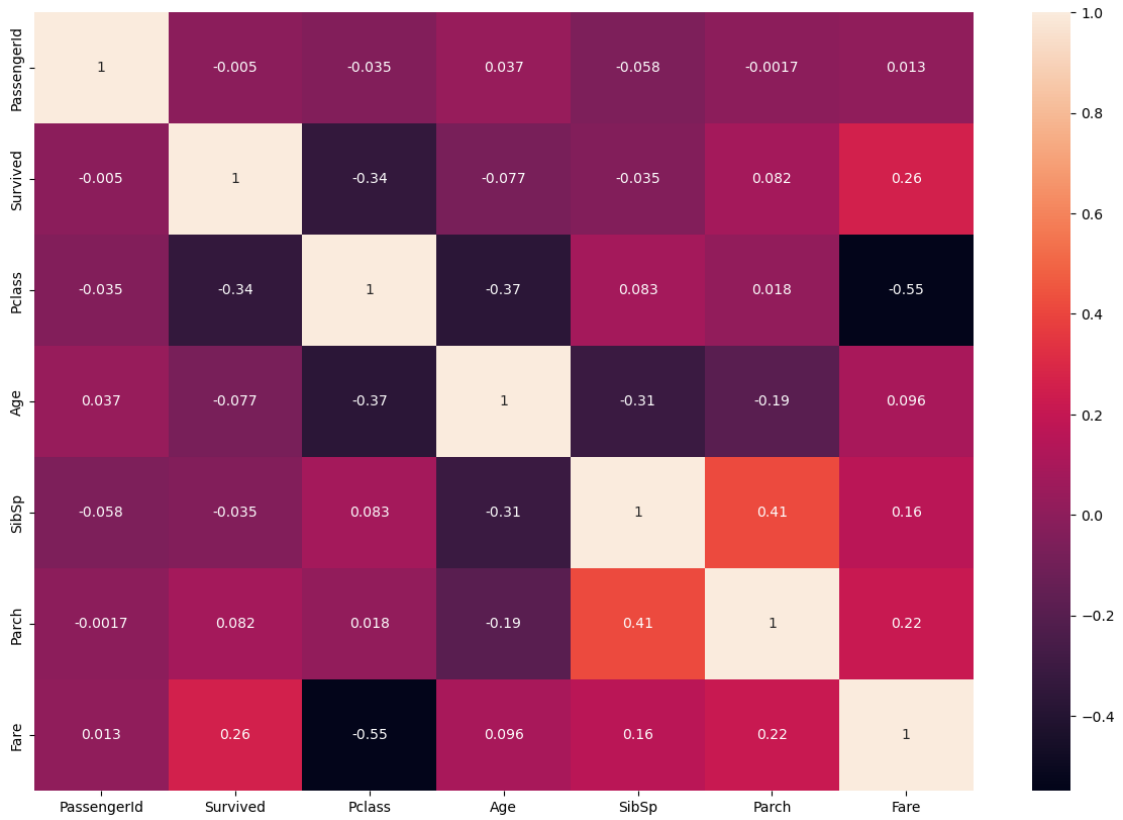
```
C:\Users\DELL\AppData\Local\Temp\ipykernel_6436\3182140910.py:1: FutureWarning:
The default value of numeric_only in DataFrame.corr is deprecated. In a future
version, it will default to False. Select only valid columns or specify the
value of numeric_only to silence this warning.
  corr=df.corr()
```

[9]:

	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

[10]:
```
plt.subplots(figsize=(15,10))
sns.heatmap(corr,annot=True)
```

[10]: <Axes: >



[11]:
```
df.Survived.value_counts()
```

[11]:
```
0    549
1    342
Name: Survived, dtype: int64
```

[12]:
```
df.Sex.value_counts()
```

[12]:
```
male      577
female    314
Name: Sex, dtype: int64
```

[13]:
```
df.Embarked.value_counts()
```

[13]:
```
S    644
C    168
```

```
Q     77
Name: Embarked, dtype: int64
```

4 3. CHECK FOR NULL VALUES

[14]:
```
df.isnull().any()
```

[14]:
```
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
```

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

[15]:
```
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
```

Fill null values in the 'Age' column with the mean age

[16]:
```
mean_age = df['Age'].mean()
df['Age'].fillna(mean_age, inplace=True)
```

Fill null values in the 'Embarked' column with the most common value

[17]:
```
most_common_embarked = df['Embarked'].mode()[0]
df['Embarked'].fillna(most_common_embarked, inplace=True)
```

[18]:
```
df.drop(['Cabin'],axis=1, inplace=True)
```

[19]:
```
df.drop(['Ticket'],axis=1, inplace=True)
```

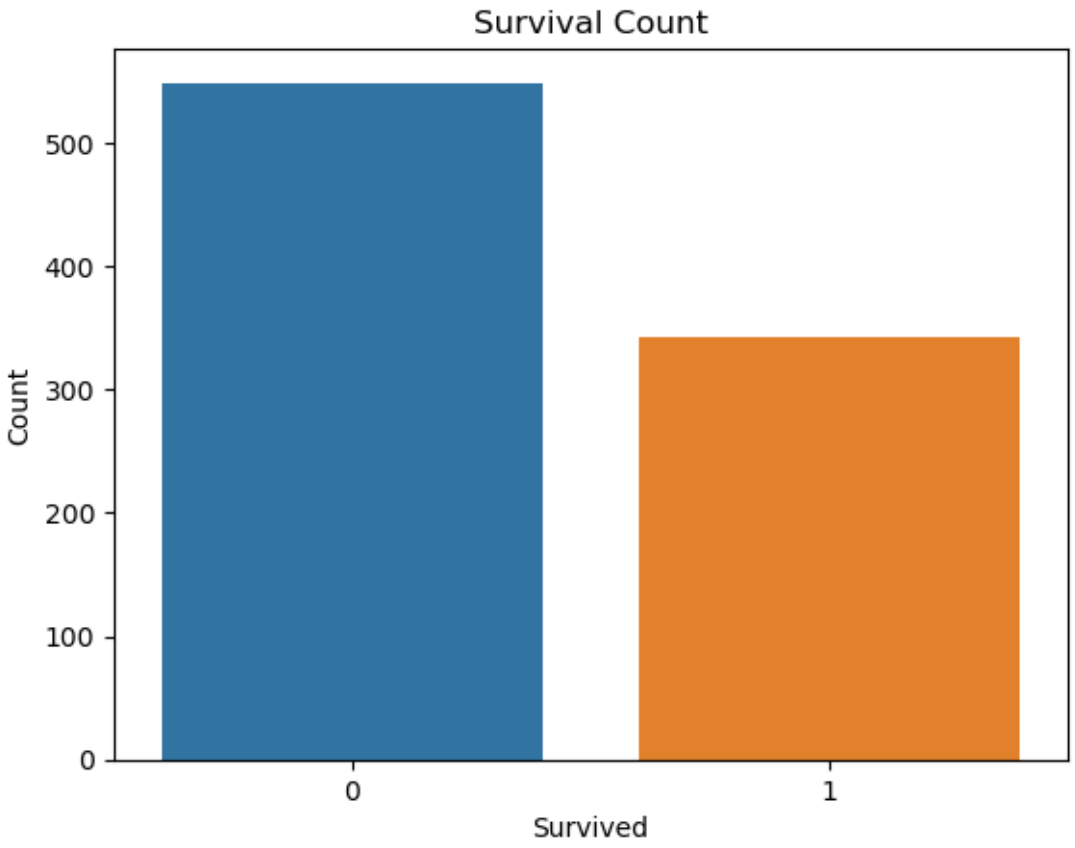
[20]:
```
df.drop(['Name'],axis=1,inplace=True)
```

[21]:
```
print(df.isnull().sum())
```

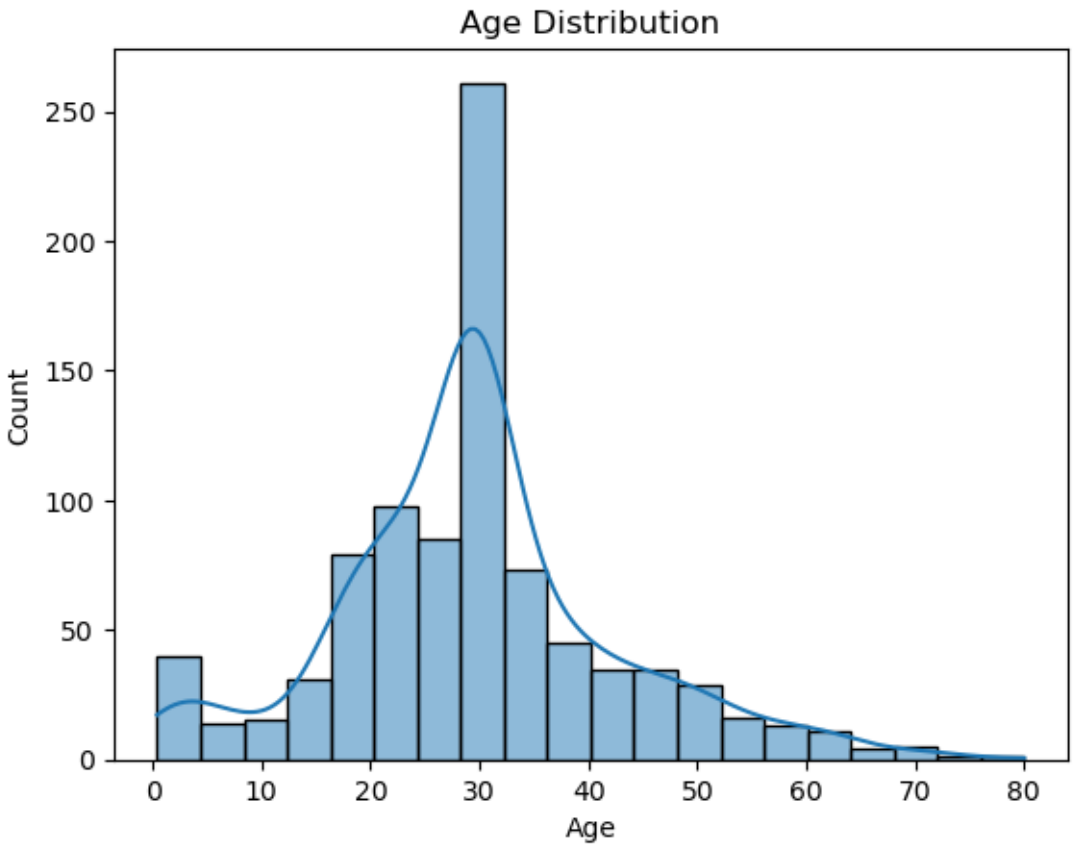
```
PassengerId    0
Survived       0
Pclass         0
Sex            0
Age            0
SibSp          0
Parch          0
Fare           0
Embarked       0
dtype: int64
```

5 4. Data Visualization

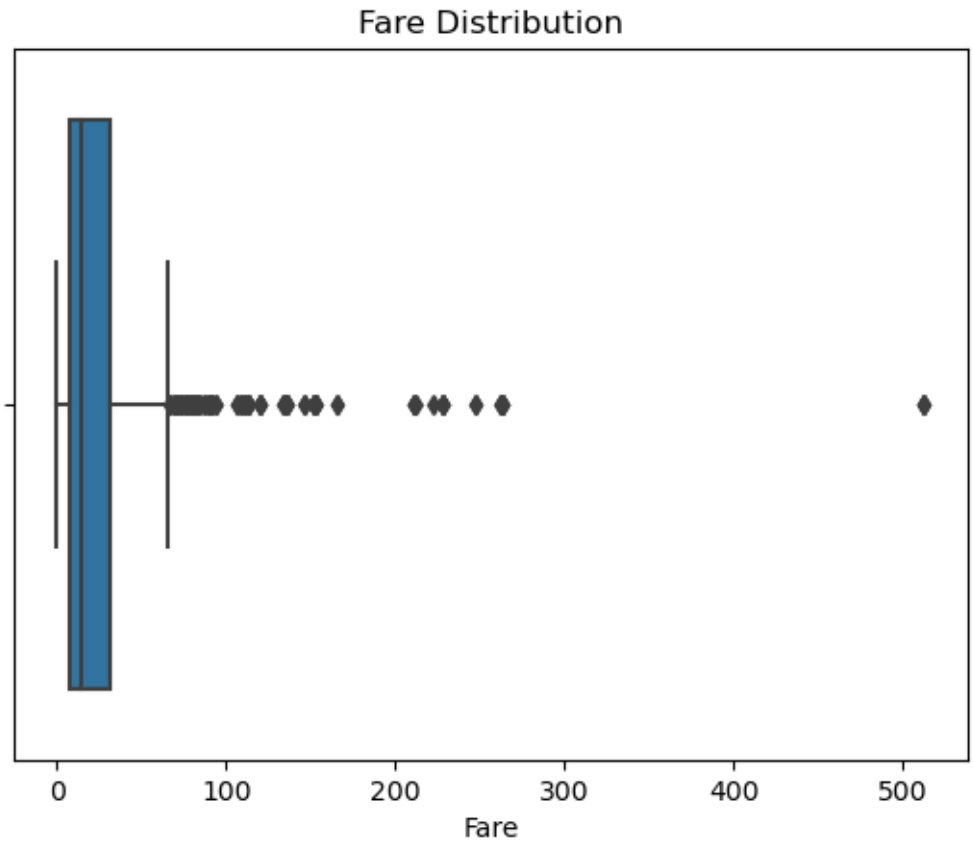
[22]:
```
# Visualize the distribution of the 'Survived' column (0 = Not Survived, 1 =␣
↪Survived)
sns.countplot(data=df, x='Survived')
plt.title('Survival Count')
plt.xlabel('Survived')
plt.ylabel('Count')
plt.show()
```



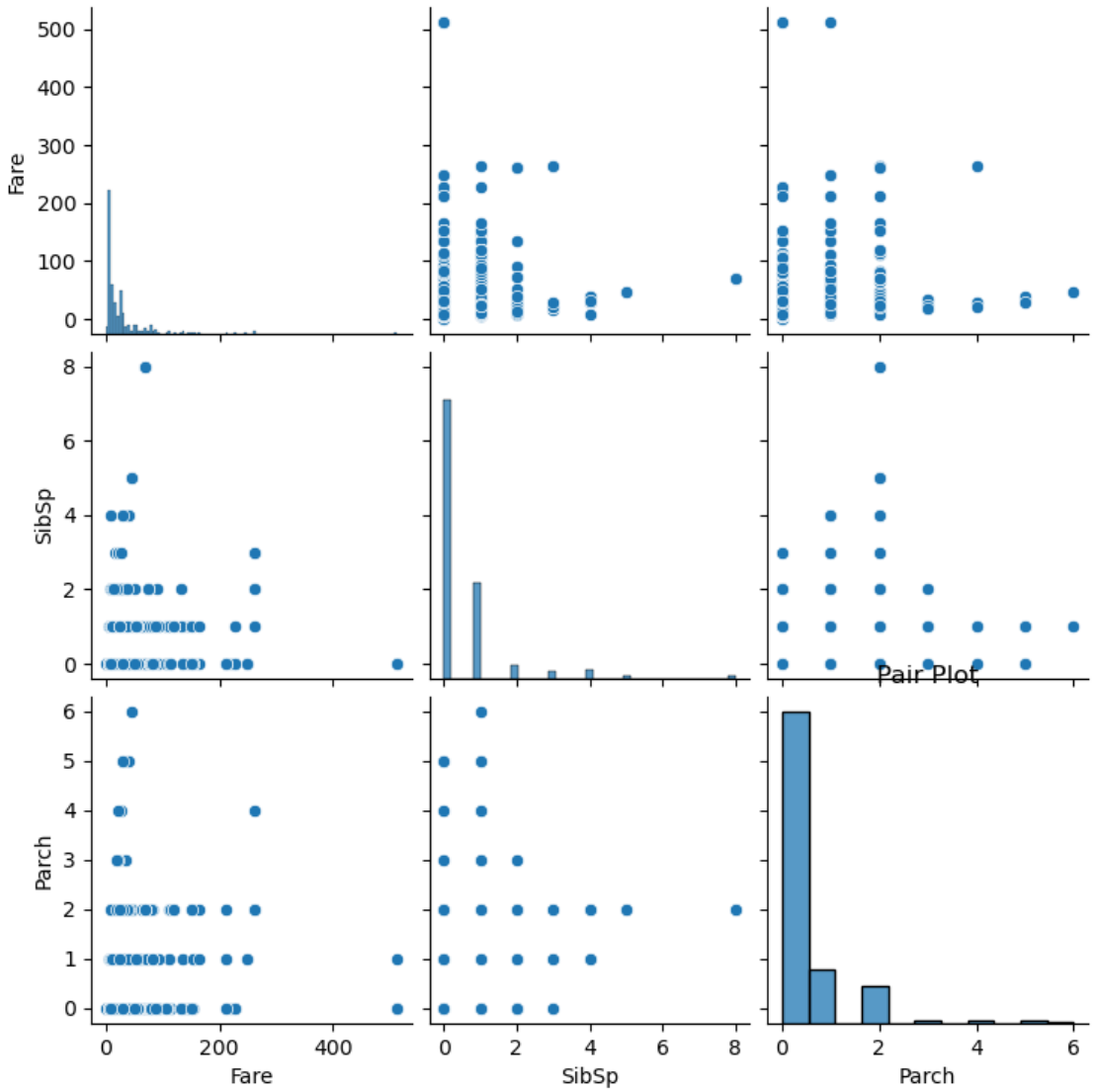
[23]:
```
#Visualize the distribution of the 'Age' column
sns.histplot(data=df, x='Age', bins=20, kde=True)
plt.title('Age Distribution')
plt.xlabel('Age')
plt.ylabel('Count')
plt.show()
```



[24]:
```
#Visualize the distribution of the 'Fare' column and detect outliers we will␣
 ↪handle outliers in the next step
sns.boxplot(data=df, x='Fare')
plt.title('Fare Distribution')
plt.xlabel('Fare')
plt.show()
```

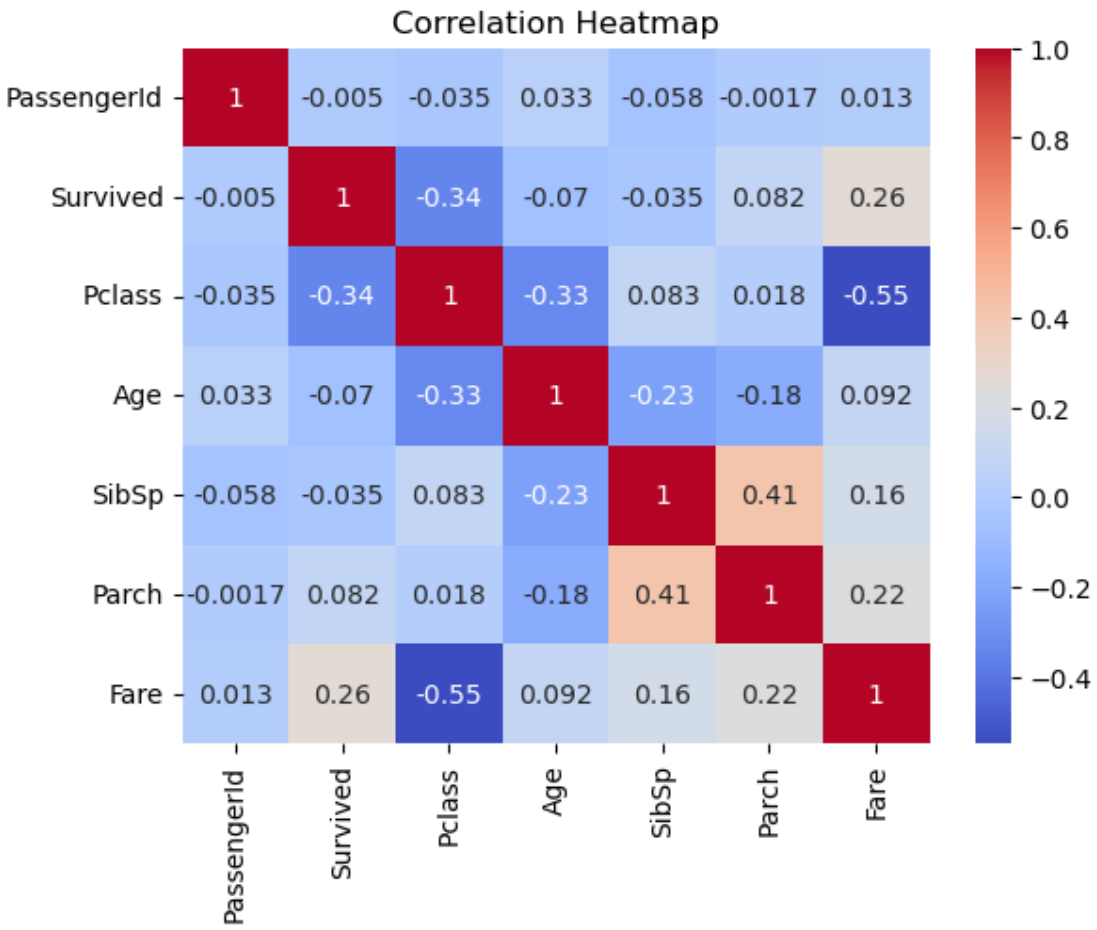


[25]:
```
#Pair plot for selected numerical columns
sns.pairplot(data=df[['Fare', 'SibSp', 'Parch']])
plt.title('Pair Plot')
plt.show()
```



[26]:
```
corr_matrix = df.corr()
sns.heatmap(corr_matrix, annot=True,cmap='coolwarm')
plt.title('Correlation Heatmap')
plt.show()
```

```
C:\Users\DELL\AppData\Local\Temp\ipykernel_6436\554220597.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_matrix = df.corr()
```



6 5. Detect and Handle Outliers

[27]:
```
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:
96     71.0
116    70.5
493    71.0
630    80.0
672    70.0
745    70.0
851    74.0
```

```
Name: Age, dtype: float64
```

[28]:
```
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
438    263.0000
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
```

[29]:
```
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: (891, 9)
Cleaned DataFrame size: (775, 9)
```

[29]:

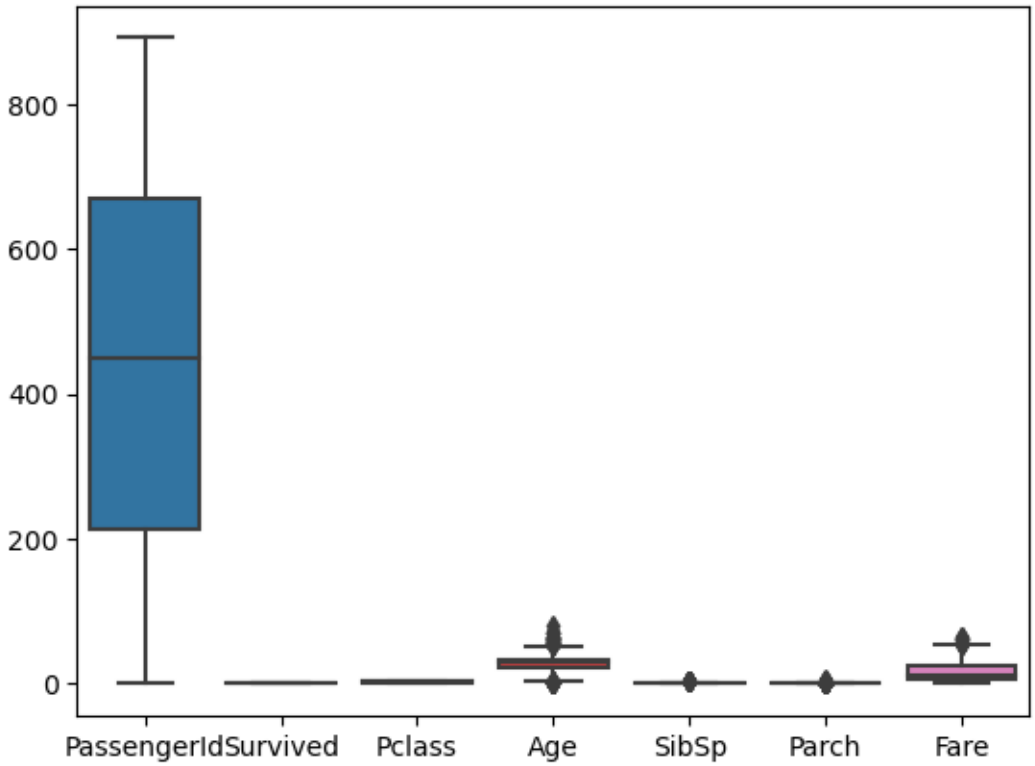
	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	\
0	1	0	3	male	22.000000	1	0	7.2500	
2	3	1	3	female	26.000000	0	0	7.9250	
3	4	1	1	female	35.000000	1	0	53.1000	
4	5	0	3	male	35.000000	0	0	8.0500	
5	6	0	3	male	29.699118	0	0	8.4583	
..	…	…	…	…	…	…	…	…	
886	887	0	2	male	27.000000	0	0	13.0000	
887	888	1	1	female	19.000000	0	0	30.0000	
888	889	0	3	female	29.699118	1	2	23.4500	
889	890	1	1	male	26.000000	0	0	30.0000	
890	891	0	3	male	32.000000	0	0	7.7500	

	Embarked
0	S
2	S
3	S
4	S
5	Q
..	…
886	S
887	S
888	S
889	C
890	Q

[775 rows x 9 columns]

[30]:
```
sns.boxplot(df_cleaned)
```

[30]:
```
<Axes: >
```



[31]:
```
df=df_cleaned
```

7 6.Splitting Dependent and Independent variables

[32]:
```
x=df.drop('Survived', axis=1)
y=df['Survived']
```

[33]:
```
x.head()
```

[33]:

	PassengerId	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	1	3	male	22.000000	1	0	7.2500	S
2	3	3	female	26.000000	0	0	7.9250	S
3	4	1	female	35.000000	1	0	53.1000	S
4	5	3	male	35.000000	0	0	8.0500	S
5	6	3	male	29.699118	0	0	8.4583	Q

[34]:
```
y.head()
```

[34]:
```
0    0
2    1
3    1
```

```
4    0
5    0
Name: Survived, dtype: int64
```

8 7. Perform Encoding

[35]:
```
en = LabelEncoder()
x['Sex'] = en.fit_transform(x['Sex'])
```

[36]:
```
x.head()
```

[36]:

	PassengerId	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	1	3	1	22.000000	1	0	7.2500	S
2	3	3	0	26.000000	0	0	7.9250	S
3	4	1	0	35.000000	1	0	53.1000	S
4	5	3	1	35.000000	0	0	8.0500	S
5	6	3	1	29.699118	0	0	8.4583	Q

[37]:
```
x = pd.get_dummies(x,columns=['Embarked'])
```

[38]:
```
x.head()
```

[38]:

	PassengerId	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked_C	Embarked_Q	Embarked_S
0	1	3	1	22.000000	1	0	7.2500	0	0	1
2	3	3	0	26.000000	0	0	7.9250	0	0	1
3	4	1	0	35.000000	1	0	53.1000	0	0	1
4	5	3	1	35.000000	0	0	8.0500	0	0	1
5	6	3	1	29.699118	0	0	8.4583	0	1	0

9 8. Feature Scaling

[39]:
```
scale = StandardScaler()
x[['Age', 'Fare']] = scale.fit_transform(x[['Age', 'Fare']])
```

[40]:
```
x.head()
```

[40]:

	PassengerId	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked_C	Embarked_Q	Embarked_S
0	1	3	1	-0.556219	1	0	-0.779117	0	0	1
2	3	3	0	-0.243027	0	0	-0.729373	0	0	1

```
3            4       1    0  0.461654      1      0  2.599828           0
4            5       3    1  0.461654      0      0 -0.720161           0
5            6       3    1  0.046606      0      0 -0.690071           0

   Embarked_Q  Embarked_S
0           0           1
2           0           1
3           0           1
4           0           1
5           1           0
```

10 9. Splitting the data into Train and Test

[41]:
```
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2,␣
↪random_state=42)
```

[42]:
```
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(620, 10)
(155, 10)
(620,)
(155,)
```