

nihal-21bcb7146-assg-3

September 20, 2023

ASSIGNMENT 3 NIHAL SHAIK 21BCB7146

1 DATA PREPROCESSING

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

IMPORTING 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	Cummings, 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.shape
```

```
[5]: (891, 12)
```

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

```
[6]:
```

	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

```
[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.corr()
```

```
C:\Users\lenovo\AppData\Local\Temp\ipykernel_11992\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()
```

```
[8]:
```

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

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

```
[9]: df.corr().Fare.sort_values(ascending=False)
```

C:\Users\lenovo\AppData\Local\Temp\ipykernel_11992\60082530.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().Fare.sort_values(ascending=False)
```

```
[9]: Fare          1.000000
Survived      0.257307
Parch         0.216225
SibSp         0.159651
Age           0.096067
PassengerId   0.012658
Pclass        -0.549500
Name: Fare, dtype: float64
```

CHECKING FOR NULL VALUES

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

```
[10]: 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
```

```
[11]: df = df.drop(['Cabin'], axis=1)
df
```

```
[11]:
```

	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	Embarked
0	0	A/5 21171	7.2500	S
1	0	PC 17599	71.2833	C
2	0	STON/O2. 3101282	7.9250	S
3	0	113803	53.1000	S
4	0	373450	8.0500	S
..
886	0	211536	13.0000	S
887	0	112053	30.0000	S
888	2	W./C. 6607	23.4500	S
889	0	111369	30.0000	C

```
890      0      370376    7.7500      Q
```

```
[891 rows x 11 columns]
```

We dropped cabin beacuse it has highest number of null values.

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

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

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

```
[14]: PassengerId      False
      Survived        False
      Pclass          False
      Name            False
      Sex             False
      Age             False
      SibSp           False
      Parch           False
      Ticket          False
      Fare            False
      Embarked        False
      dtype: bool
```

Finally,we can observe there are no null values in any attribute

```
[15]: df.Embarked.nunique()
```

```
[15]: 3
```

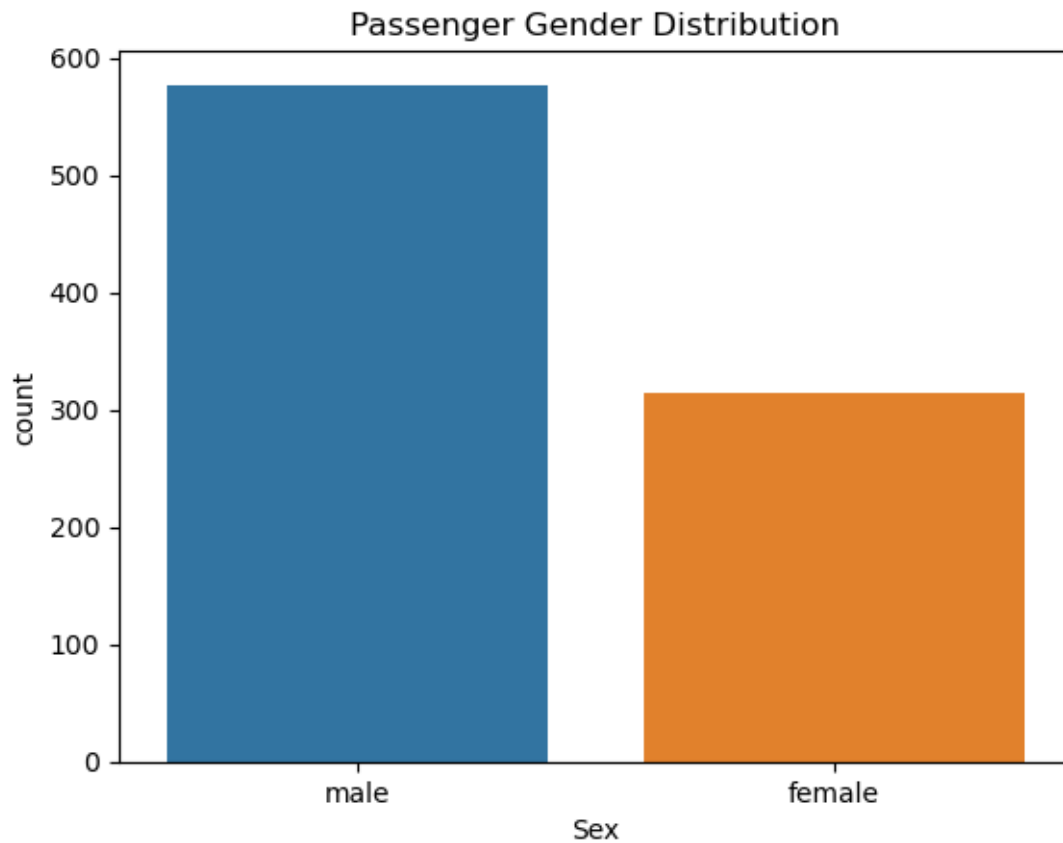
```
[16]: df.Embarked.unique()
```

```
[16]: array(['S', 'C', 'Q'], dtype=object)
```

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

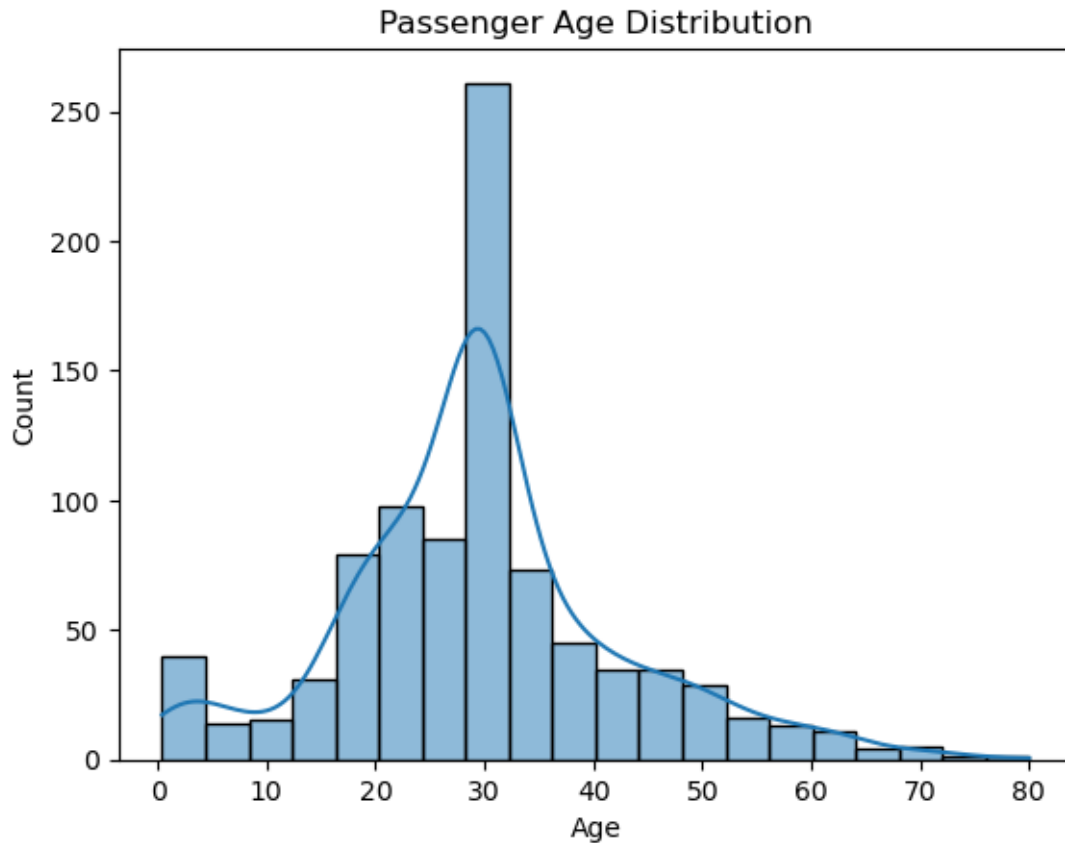
```
[17]: S      646
      C      168
      Q       77
      Name: Embarked, dtype: int64
```

```
[18]: sns.countplot(data=df, x='Sex')
      plt.title('Passenger Gender Distribution')
      plt.show()
```



INFERENCE: We can observe that there are more number of male passengers than female passengers

```
[19]: sns.histplot(data=df, x='Age', bins=20, kde=True)
plt.title('Passenger Age Distribution')
plt.xlabel('Age')
plt.ylabel('Count')
plt.show()
```



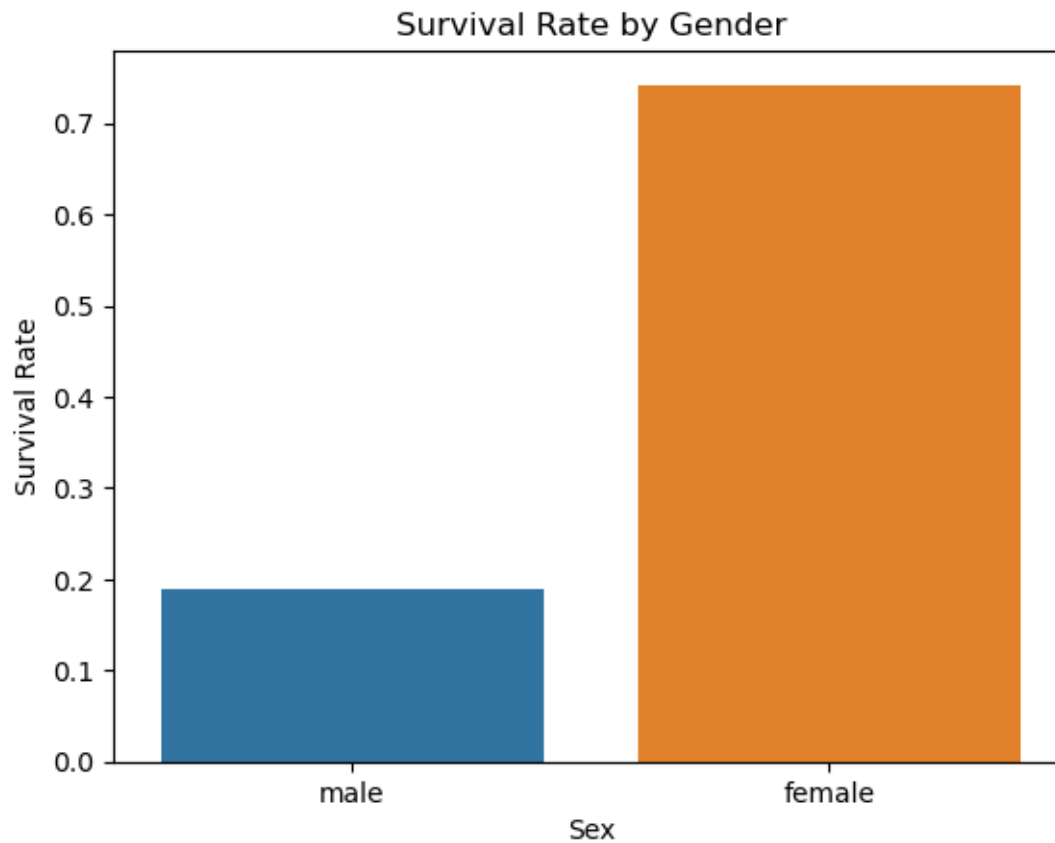
INFERENCE: The histogram of the 'Age' distribution provides insights into the age of Titanic passengers, showing that the majority were 30 to 40 aged adults, but there were also significant numbers of younger and older passengers.

```
[20]: sns.barplot(data=df, x='Sex', y='Survived', ci=None)
plt.title('Survival Rate by Gender')
plt.ylabel('Survival Rate')
plt.show()
```

C:\Users\lenovo\AppData\Local\Temp\ipykernel_11992\3687825708.py:1:
FutureWarning:

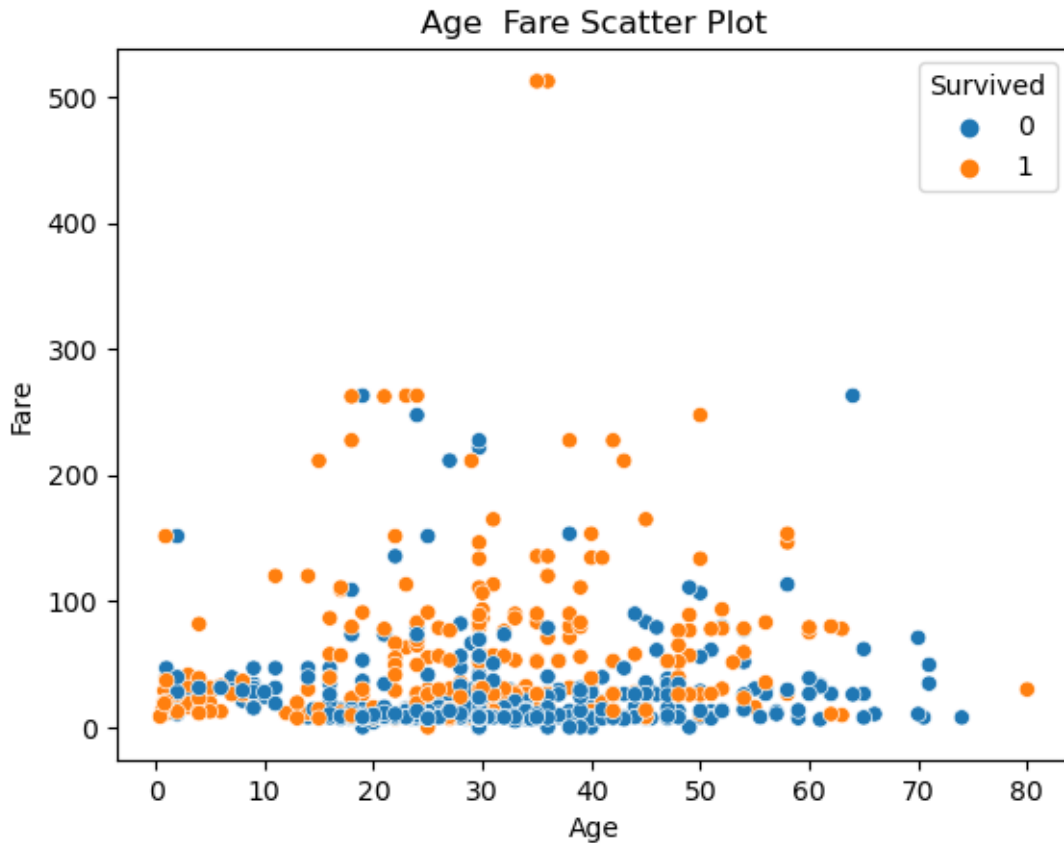
The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

```
sns.barplot(data=df, x='Sex', y='Survived', ci=None)
```

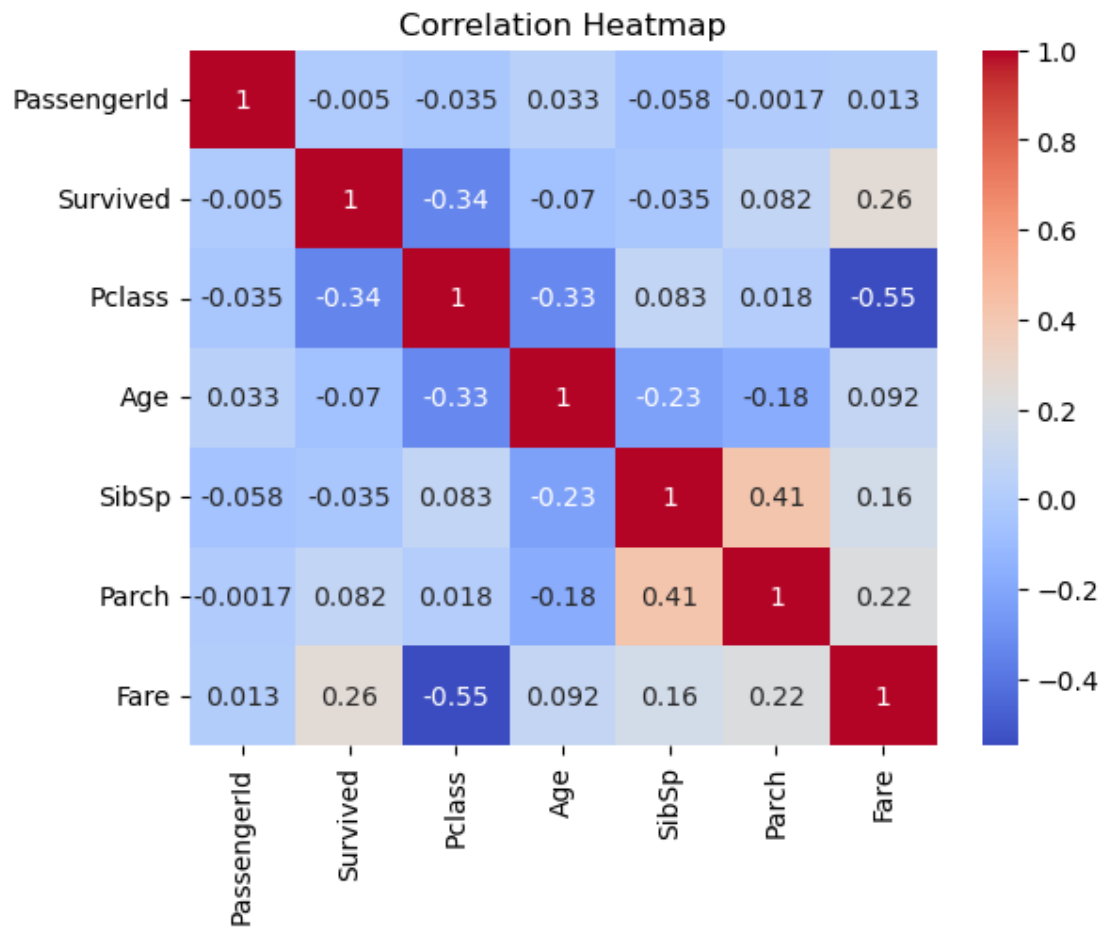
INFERENCE: we can observe that female passengers have high survival rate than male passengers

```
[21]: sns.scatterplot(data=df, x='Age', y='Fare', hue='Survived')
plt.title('Age Fare Scatter Plot')
plt.xlabel('Age')
plt.ylabel('Fare')
plt.show()
```



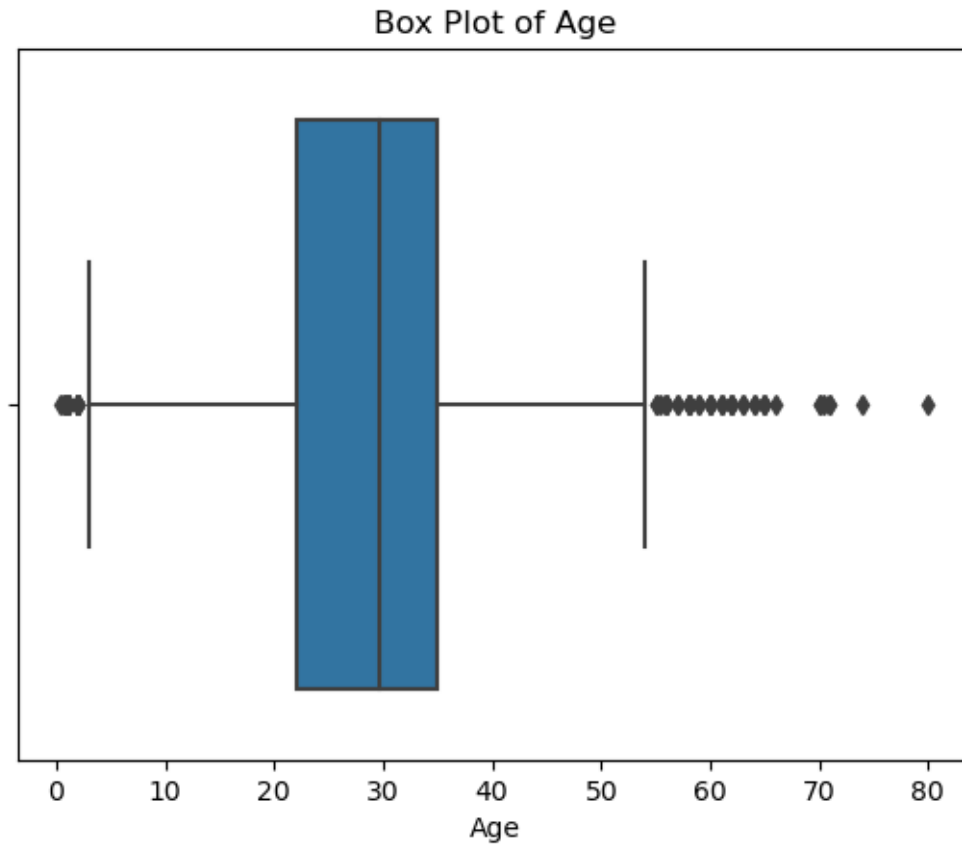
```
[22]: correlation_matrix = df.corr()  
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')  
plt.title('Correlation Heatmap')  
plt.show()
```

```
C:\Users\lenovo\AppData\Local\Temp\ipykernel_11992\2298098936.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.  
    correlation_matrix = df.corr()
```



OUTLIER DETECTION

```
[23]: sns.boxplot(data=df, x='Age')
plt.title('Box Plot of Age')
plt.show()
```



```
[24]: df.shape
```

```
[24]: (891, 11)
```

```
[25]: q1=df.Age.quantile(0.25)
      q3=df.Age.quantile(0.75)
      print(q1)
      print(q3)
```

```
22.0
35.0
```

```
[26]: IQR=q3-q1
      IQR
```

```
[26]: 13.0
```

```
[27]: upper_limit=q3+1.5*IQR
      upper_limit
```

[27]: 54.5

```
[28]: lower_limit=q3-1.5*IQR  
lower_limit
```

[28]: 15.5

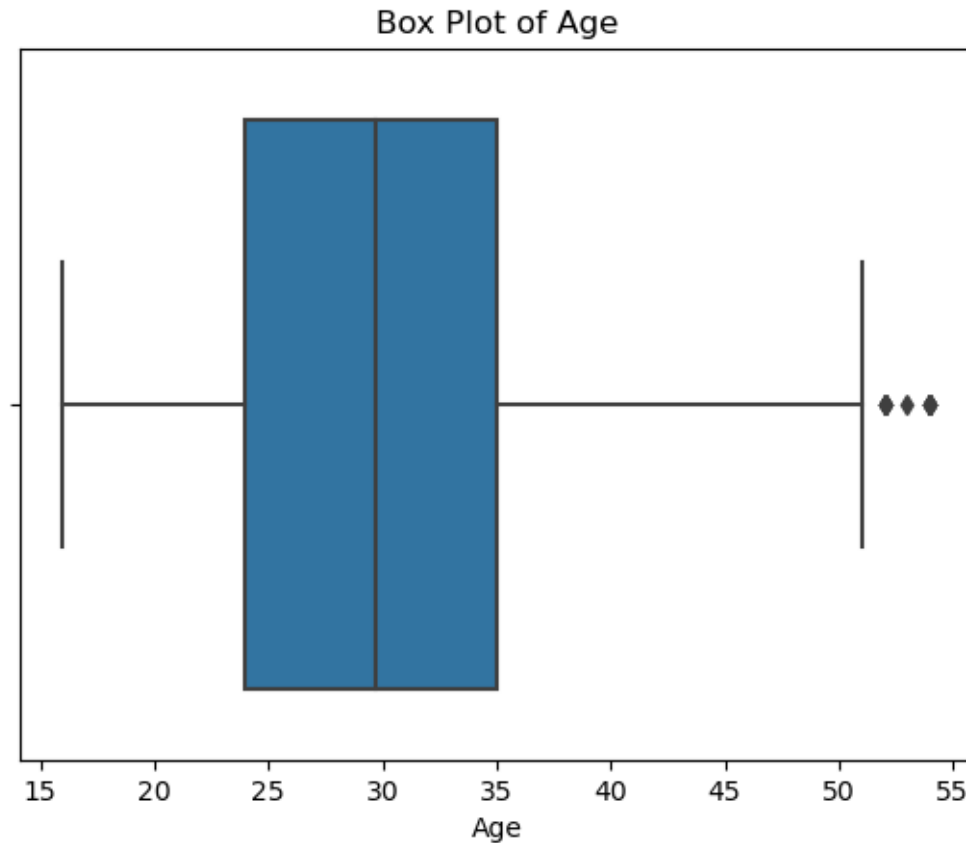
```
[29]: from scipy import stats  
  
z_scores = np.abs(stats.zscore(df['Age']))  
outliers = (z_scores > 3)  
z_scores
```

```
[29]: 0      0.592481  
1      0.638789  
2      0.284663  
3      0.407926  
4      0.407926  
...  
886    0.207709  
887    0.823344  
888    0.000000  
889    0.284663  
890    0.177063  
Name: Age, Length: 891, dtype: float64
```

```
[30]: df_no_outliers = df[(df['Age'] >= lower_limit) & (df['Age'] <= upper_limit)]  
print("Original dataset shape:", df.shape)  
print("Dataset shape after removing outliers:", df_no_outliers.shape)
```

```
Original dataset shape: (891, 11)  
Dataset shape after removing outliers: (766, 11)
```

```
[31]: sns.boxplot(data=df_no_outliers, x='Age')  
plt.title('Box Plot of Age')  
plt.show()
```



SPLITTING DEPENDENT AND INDEPENDENT VARIABLES

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

```
[32]:   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 Embarked
0      0   A/5 21171   7.2500         S
```

1	0	PC 17599	71.2833	C
2	0	STON/O2. 3101282	7.9250	S
3	0	113803	53.1000	S
4	0	373450	8.0500	S

```
[33]: X=df.drop (columns=["Fare"],axis=1)
X.head()
```

```
[33]: 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	Embarked
0	0	A/5 21171	S
1	0	PC 17599	C
2	0	STON/O2. 3101282	S
3	0	113803	S
4	0	373450	S

```
[34]: X.shape
```

```
[34]: (891, 10)
```

```
[35]: type(X)
```

```
[35]: pandas.core.frame.DataFrame
```

```
[36]: y=df["Fare"]
y.head()
```

```
[36]: 0    7.2500
1    71.2833
2     7.9250
3    53.1000
4     8.0500
Name: Fare, dtype: float64
```

ENCODING

```
[37]: X.head()
```

```
[37]:
```

	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	Embarked
0	0	A/5 21171	S
1	0	PC 17599	C
2	0	STON/O2. 3101282	S
3	0	113803	S
4	0	373450	S

```
[38]: from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
X["Sex"]=le.fit_transform(X["Sex"])
X.head()
```

```
[38]:
```

	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	Parch	\
0	Braund, Mr. Owen Harris	1	22.0	1	0	
1	Cumings, Mrs. John Bradley (Florence Briggs Th...	0	38.0	1	0	
2	Heikkinen, Miss. Laina	0	26.0	0	0	
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	35.0	1	0	
4	Allen, Mr. William Henry	1	35.0	0	0	

	Ticket	Embarked
0	A/5 21171	S
1	PC 17599	C
2	STON/O2. 3101282	S

3	113803	S
4	373450	S

```
[39]: X["Embarked"]=le.fit_transform(X["Embarked"])
X.head()
```

```
[39]: 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	Parch	\
0	Braund, Mr. Owen Harris	1	22.0	1	0	
1	Cumings, Mrs. John Bradley (Florence Briggs Th...	0	38.0	1	0	
2	Heikkinen, Miss. Laina	0	26.0	0	0	
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	35.0	1	0	
4	Allen, Mr. William Henry	1	35.0	0	0	

	Ticket	Embarked
0	A/5 21171	2
1	PC 17599	0
2	STON/O2. 3101282	2
3	113803	2
4	373450	2

```
[40]: print(le.classes_)
```

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

```
[41]: mapping=dict(zip(le.classes_,range(len(le.classes_))))
mapping
```

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

FEATURE SCALING

```
[49]: from sklearn.preprocessing import MinMaxScaler
ms=MinMaxScaler()
from sklearn.preprocessing import MinMaxScaler
numerical_features = ['Age', 'Fare']
data = df[numerical_features]
ms = MinMaxScaler()
scaled_data = ms.fit_transform(data)
df_scaled = pd.DataFrame(scaled_data, columns=numerical_features)
print(df_scaled.head())
```

	Age	Fare
0	0.271174	0.014151
1	0.472229	0.139136
2	0.321438	0.015469
3	0.434531	0.103644
4	0.434531	0.015713

SPLITTING DATA INTO TRAIN AND TEST

```
[50]: 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=0)
```

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

```
(712, 10) (179, 10) (712,) (179,)
```

```
[52]: X = df.drop('Survived', axis=1)
y = df['Survived']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
↳ random_state=42)
```

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

```
(712, 10) (179, 10) (712,) (179,)
```

```
[54]: df= df.drop(['PassengerId', 'Name', 'Ticket'], axis=1)
```

```
[55]: df
```

```
[55]:
```

	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	71.2833	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 x 8 columns]
```

```
[ ]:
```