```
In [1]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report, accuracy_score
In [3]: data=pd.read_csv(r"C:\Users\Praveen T\Downloads\titanic.csv")
In [4]: data
```



Out[4]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
	0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85
	2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN
	3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123
	4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN
	•••											
	886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN
	887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42
	888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN
	889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148
	890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500	NaN

891 rows × 12 columns

In [5]: # number of columns present in the dataset.
len(data.columns)

Out[5]: 12

In [6]: # number of rows present in the dataset.
len(data)



In [7]: # Descriptive statistics
data.describe()

In [17]: data.isna().sum()

Out[7]: **PassengerId** Survived **Pclass** Age SibSp **Parch Fare** 891.000000 891.000000 891.000000 714.000000 891.000000 891.000000 891.000000 count 446.000000 0.383838 2.308642 29.699118 0.523008 0.381594 32.204208 mean 257.353842 0.486592 14.526497 49.693429 0.836071 1.102743 0.806057 std min 1.000000 0.000000 1.000000 0.420000 0.000000 0.000000 0.000000 0.000000 25% 223.500000 0.000000 2.000000 20.125000 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 891.000000 1.000000 3.000000 80.000000 8.000000 6.000000 512.329200 max

```
# finding number of missing values
In [8]:
         data.isna().sum()
         PassengerId
                          0
Out[8]:
         Survived
                          0
         Pclass
                          0
         Name
                          0
         Sex
                          0
                        177
         Age
         SibSp
                          0
         Parch
                          0
         Ticket
                          0
         Fare
                          0
         Cabin
                        687
         Embarked
                          2
         dtype: int64
In [9]: # solving the missing values -> 1 : Export Product Share (%) contains 100 in all the
         data['Age'].fillna(data['Age'].mean(),inplace=True)
In [12]:
         # Fill missing Cabin values with the most common Cabin value
         most_common_cabin = data['Cabin'].mode()[0]
         data['Cabin'].fillna(most_common_cabin, inplace=True)
         # Fill missing Embarked values using backward fill method
In [16]:
         data['Embarked'].fillna(method='bfill', inplace=True)
```



PassengerId 0 Out[17]: Survived 0 Pclass 0 Name 0 Sex 0 0 Age 0 0 0 0 SibSp Parch Ticket 0 Fare Cabin 0 Embarked dtype: int64

In [18]: data



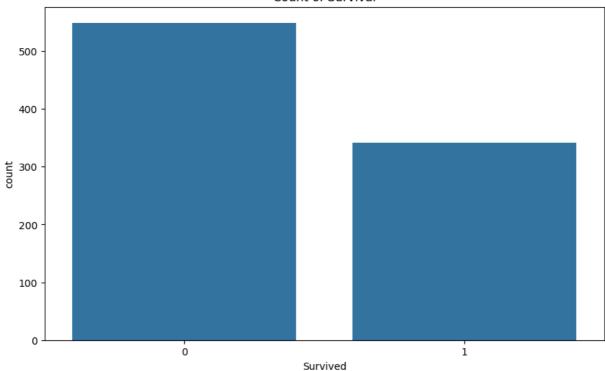
Out[18]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
	0	1	0	3	Braund, Mr. Owen Harris	male	22.000000	1	0	A/5 21171	7.2500
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.000000	1	0	PC 17599	71.2833
	2	3	1	3	Heikkinen, Miss. Laina	female	26.000000	0	0	STON/O2. 3101282	7.9250
	3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.000000	1	0	113803	53.1000
	4	5	0	3	Allen, Mr. William Henry	male	35.000000	0	0	373450	8.0500
	•••					•••					
	886	887	0	2	Montvila, Rev. Juozas	male	27.000000	0	0	211536	13.0000
	887	888	1	1	Graham, Miss. Margaret Edith	female	19.000000	0	0	112053	30.0000
	888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	29.699118	1	2	W./C. 6607	23.4500
	889	890	1	1	Behr, Mr. Karl Howell	male	26.000000	0	0	111369	30.0000
	890	891	0	3	Dooley, Mr. Patrick	male	32.000000	0	0	370376	7.7500

891 rows × 12 columns

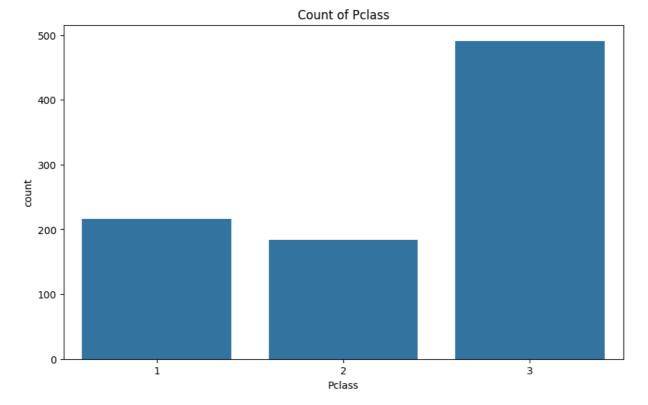
In [21]: plt.figure(figsize=(10, 6))
 sns.countplot(data=data, x='Survived')
 plt.title('Count of Survival')
 plt.show()



## Count of Survival

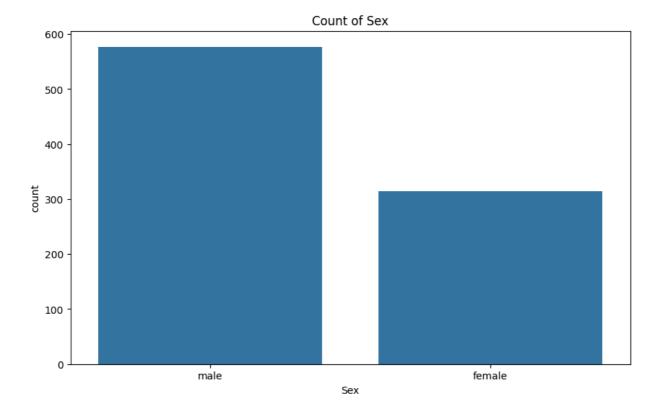


```
In [22]: # Countplot of Pclass
plt.figure(figsize=(10, 6))
sns.countplot(data, x='Pclass')
plt.title('Count of Pclass')
plt.show()
```

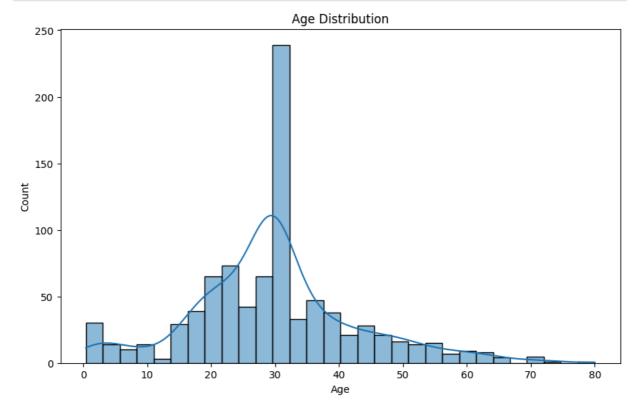


```
In [30]: # Countplot of Sex
plt.figure(figsize=(10, 6))
sns.countplot(data, x='Sex')
plt.title('Count of Sex')
plt.show()
```



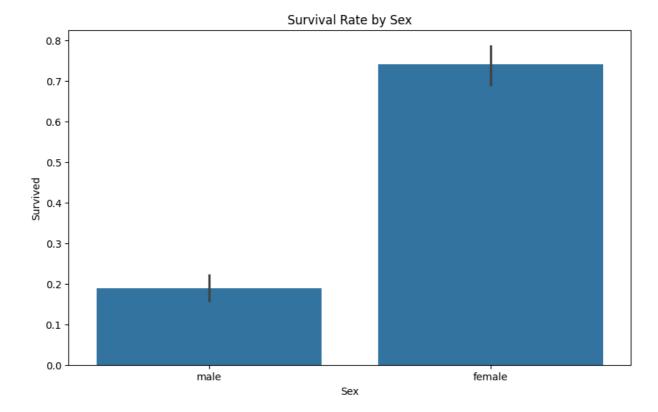


```
In [23]: # Age distribution
  plt.figure(figsize=(10, 6))
  sns.histplot(data, x='Age', kde=True)
  plt.title('Age Distribution')
  plt.show()
```



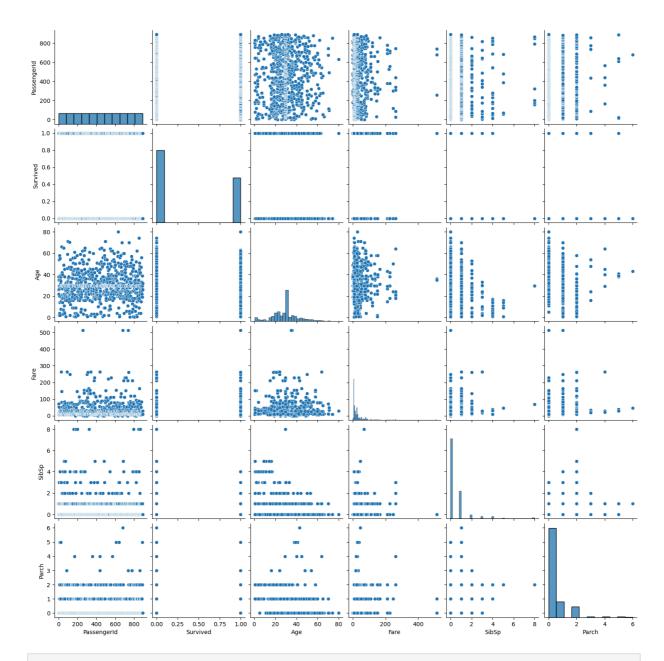
```
In [24]: # Survival rate by Sex
plt.figure(figsize=(10, 6))
sns.barplot(data, x='Sex', y='Survived')
plt.title('Survival Rate by Sex')
plt.show()
```





```
In [29]: # Pairplot for numerical features
sns.pairplot(data[['PassengerId','Survived', 'Age', 'Fare', 'SibSp', 'Parch']])
plt.show()
```

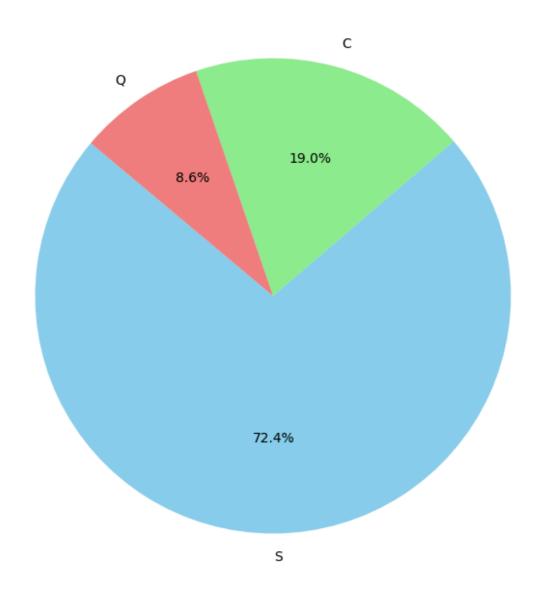




```
In [31]: # Create a pie chart for the 'Embarked' column
    embarked_counts = data['Embarked'].value_counts()
    plt.figure(figsize=(8, 8))
    plt.pie(embarked_counts, labels=embarked_counts.index, autopct='%1.1f%%', startangle=
    plt.title('Distribution of Embarked')
    plt.show()
```



## Distribution of Embarked



In [ ]:

