

```
import pandas as pd
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]

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
%matplotlib inline

train =pd. read_csv('Titanic-Dataset.csv')
test= pd. read_csv('Titanic-Dataset.csv')
train. head ()
```

	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

```
train. 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

```
train.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 891 entries, 0 to 890
```

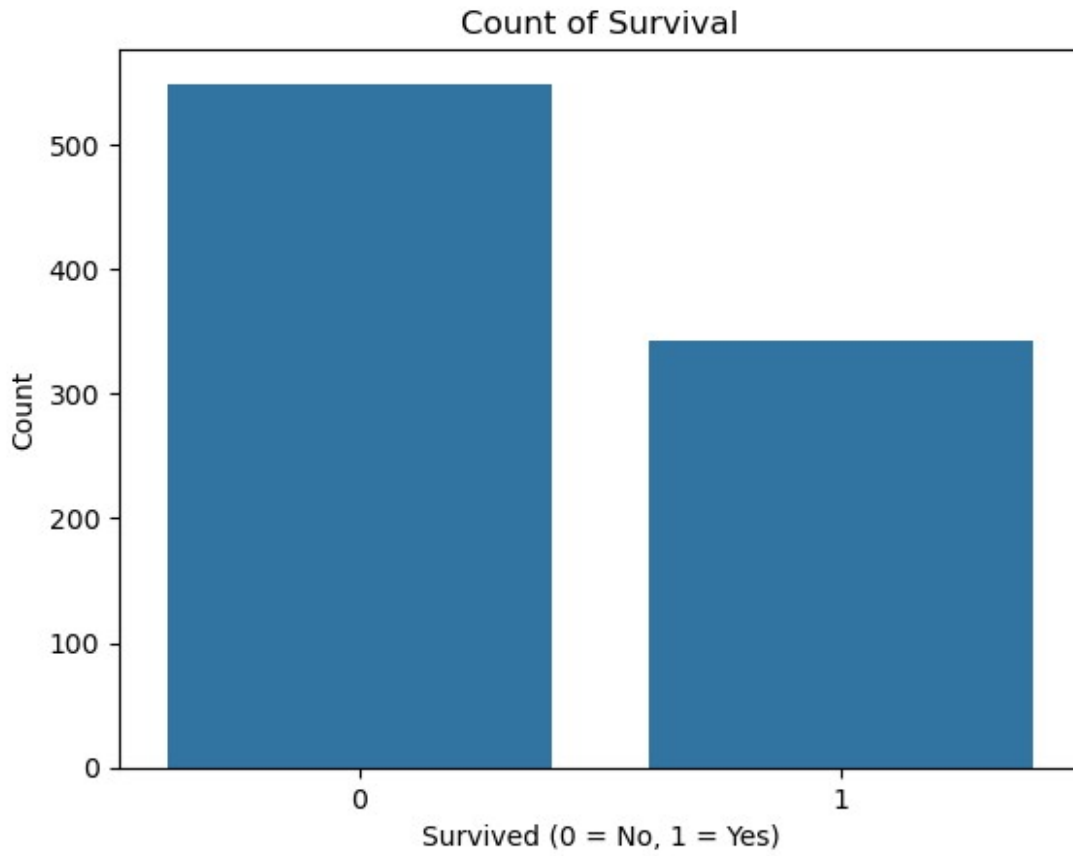
```
Data columns (total 12 columns):
```

#	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=pd.read_csv('Titanic-Dataset.csv')
train= pd.read_csv('Titanic-Dataset.csv')
sns.countplot(x='Survived', data=train)
plt.title("Count of Survival")
plt.xlabel("Survived (0 = No, 1 = Yes)")
plt.ylabel("Count")
plt.show()
```



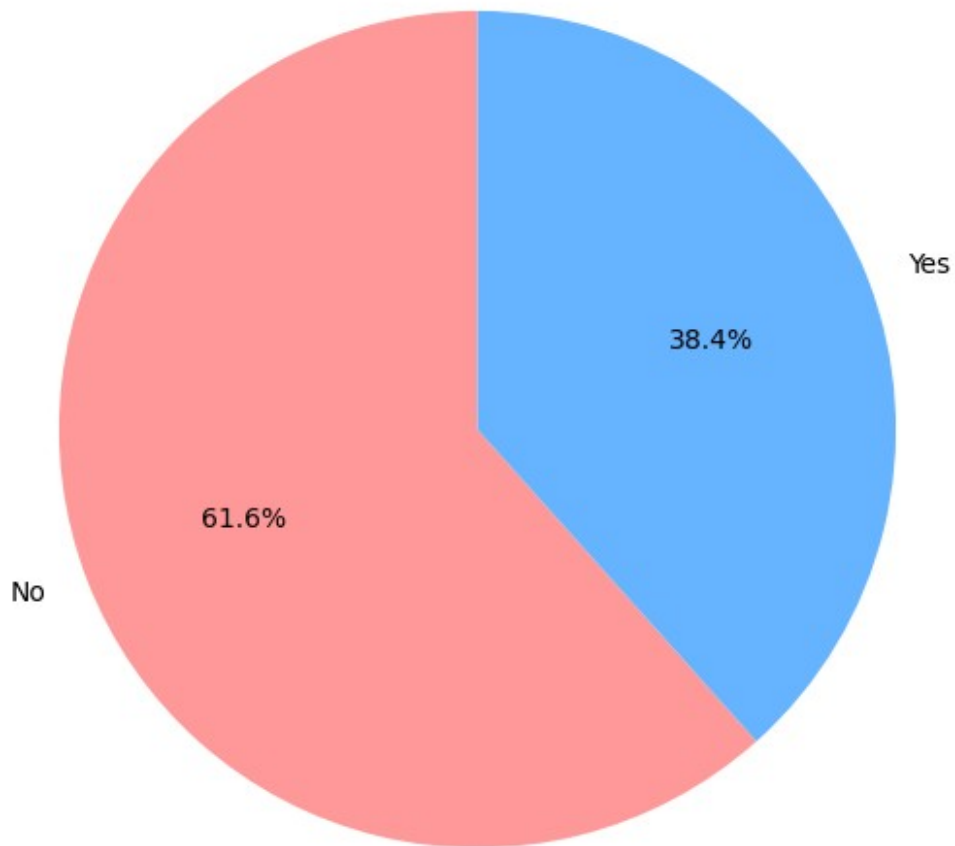
```
import pandas as pd
import matplotlib.pyplot as plt

# Load the dataset
train = pd.read_csv('Titanic-Dataset.csv')

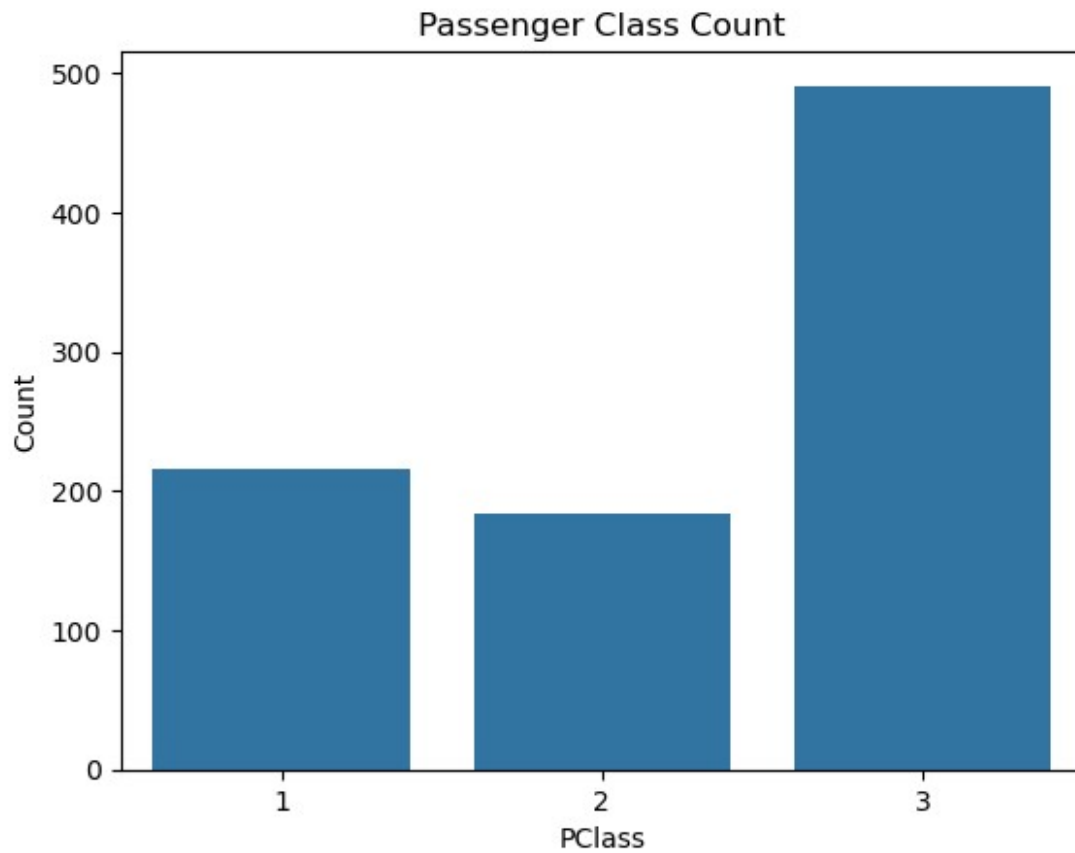
# Count the occurrences of 'Survived' values (0 and 1)
survival_counts = train['Survived'].value_counts()

# Plot pie chart
plt.figure(figsize=(7, 7))
plt.pie(survival_counts, labels=['No', 'Yes'], autopct='%1.1f%%',
startangle=90, colors=['#FF9999', '#66B3FF'])
plt.title("Survival Distribution")
plt.show()
```

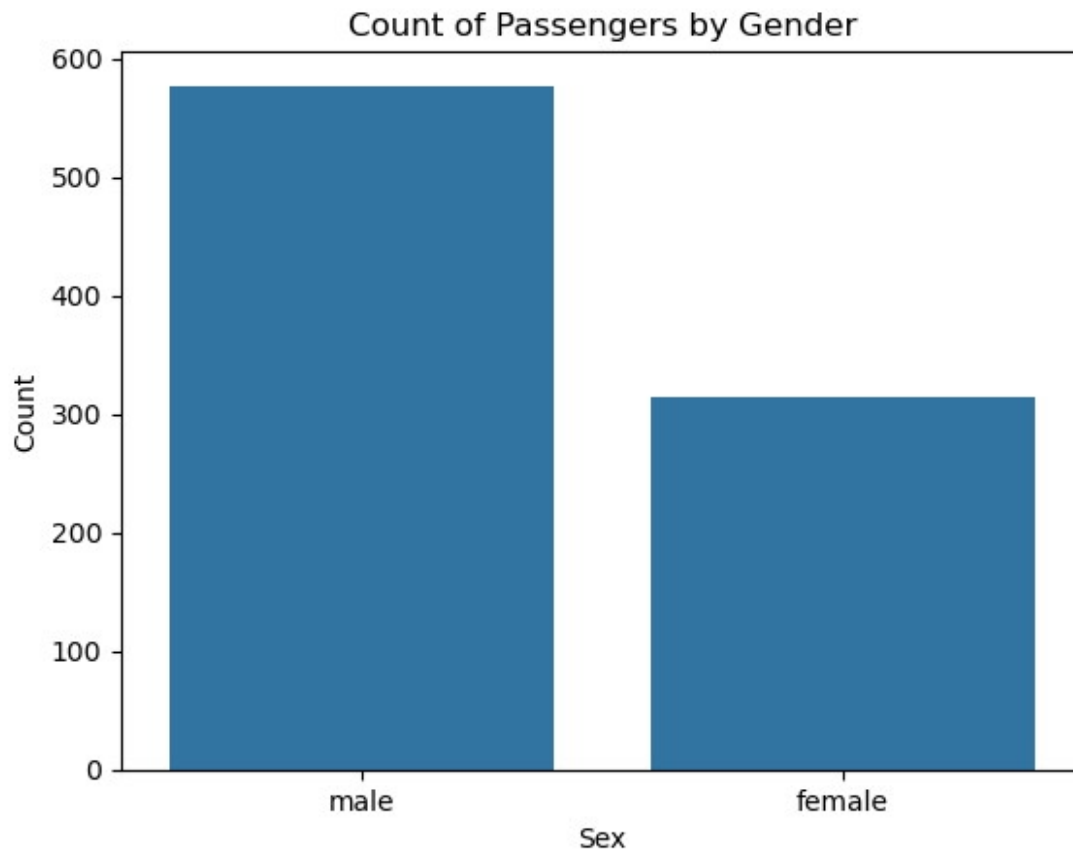
Survival Distribution



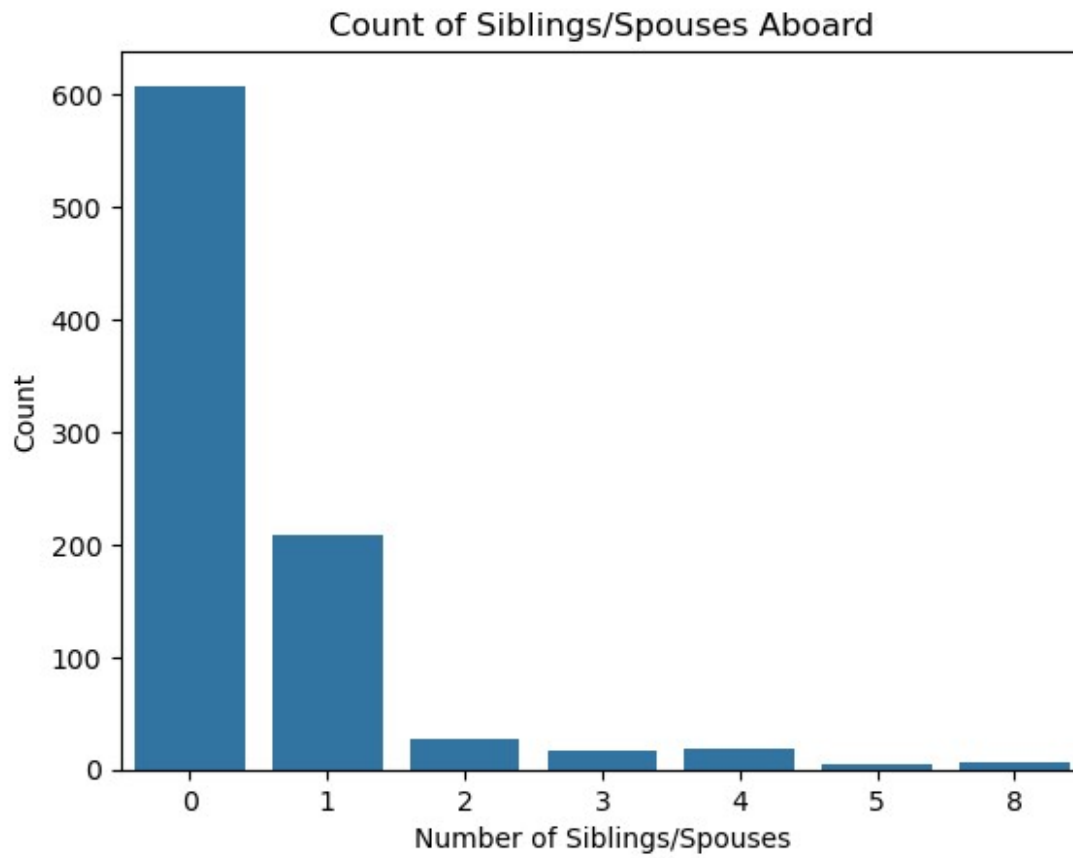
```
df=pd.read_csv('Titanic-Dataset.csv')
train= pd.read_csv('Titanic-Dataset.csv')
sns.countplot(x='Pclass', data=train)
plt.title("Passenger Class Count")
plt.xlabel("PClass")
plt.ylabel("Count")
plt.show()
```



```
train = pd.read_csv('Titanic-Dataset.csv')
sns.countplot(x='Sex', data=train)
plt.title("Count of Passengers by Gender")
plt.xlabel("Sex")
plt.ylabel("Count")
plt.show()
```

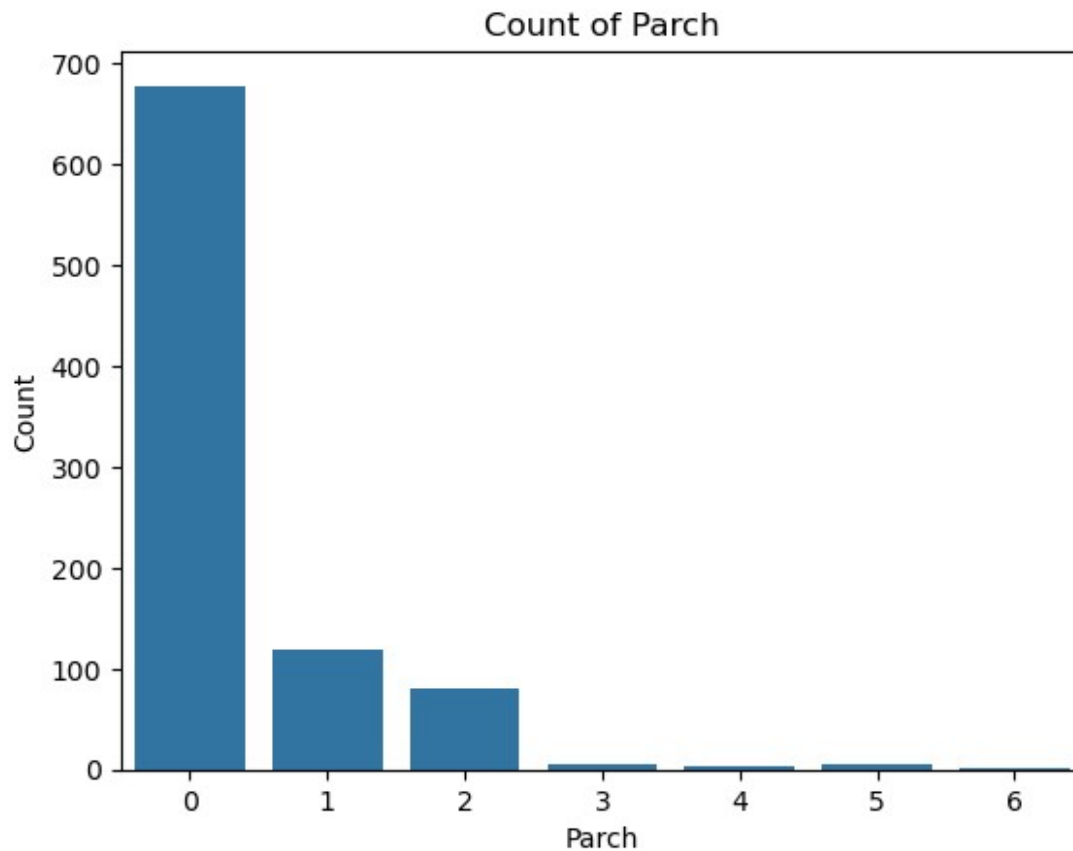


```
train = pd.read_csv('Titanic-Dataset.csv')
sns.countplot(x='SibSp', data=train)
plt.title("Count of Siblings/Spouses Aboard")
plt.xlabel("Number of Siblings/Spouses")
plt.ylabel("Count")
plt.show()
```

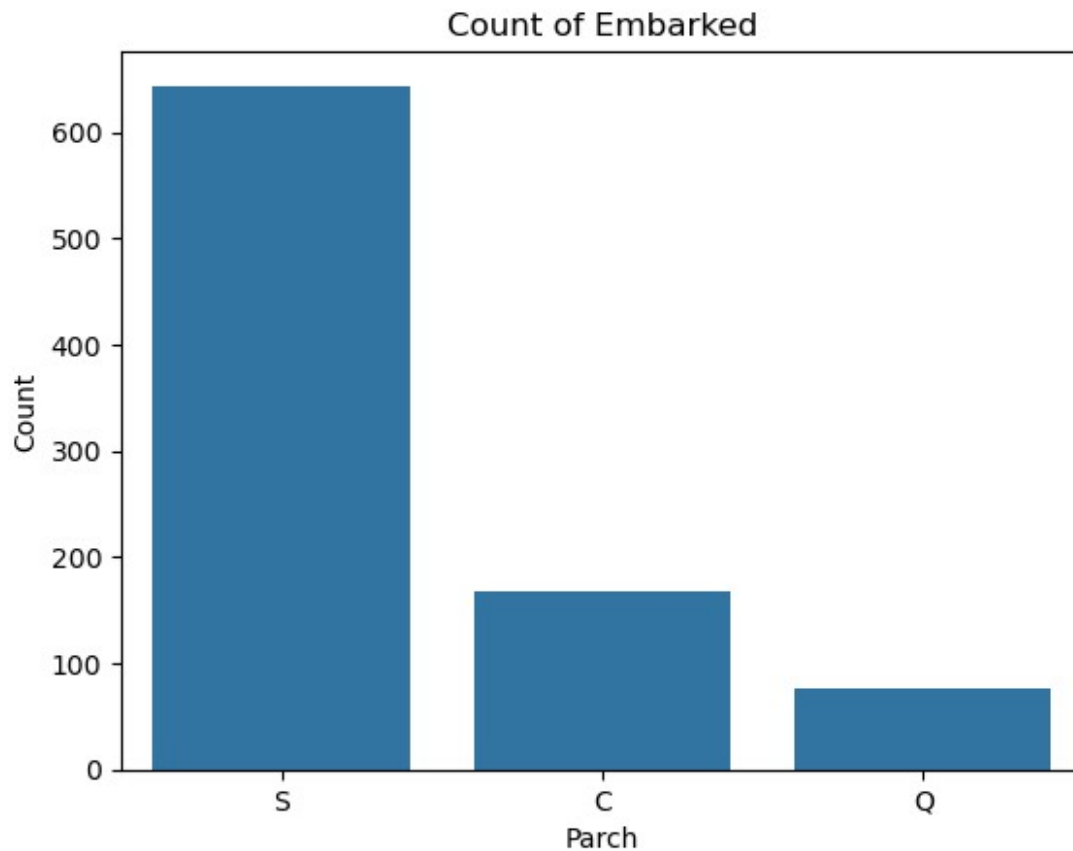


```
train = pd.read_csv('Titanic-Dataset.csv')
sns.countplot(x='Parch', data=train)
plt.title("Count of Parch")
plt.xlabel("Parch")
plt.ylabel("Count")
plt.show()
```

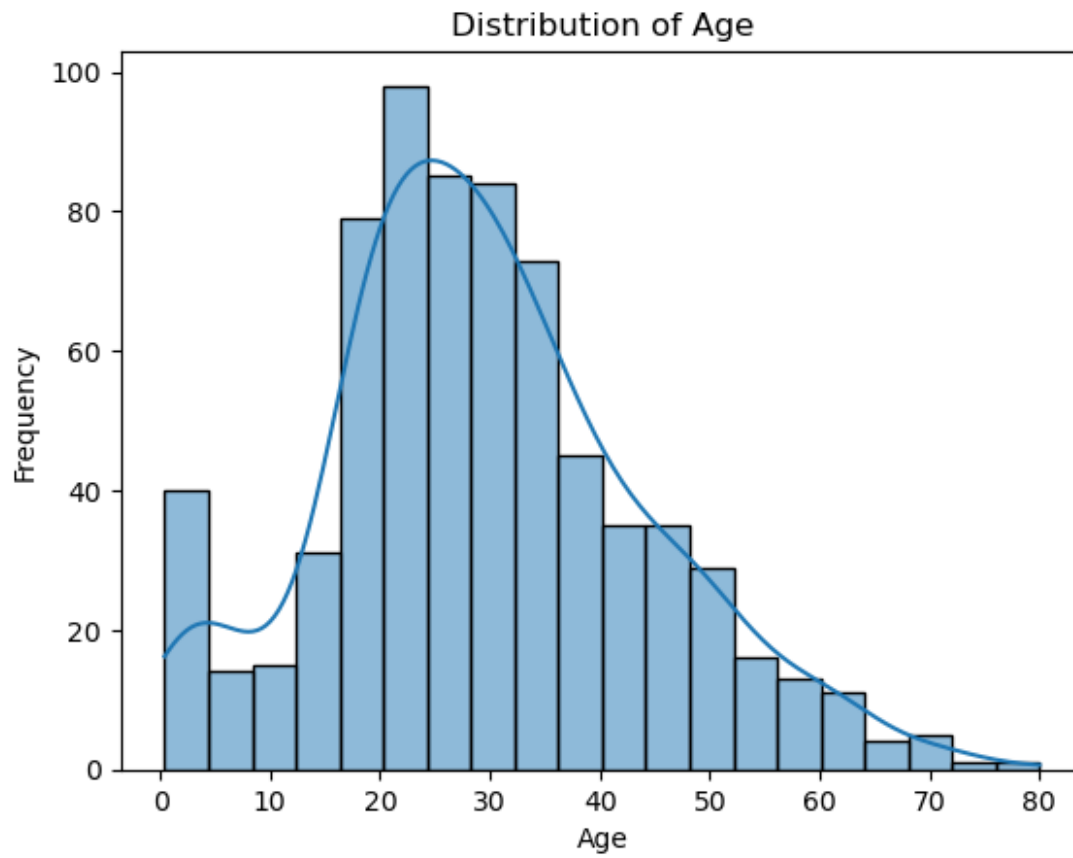




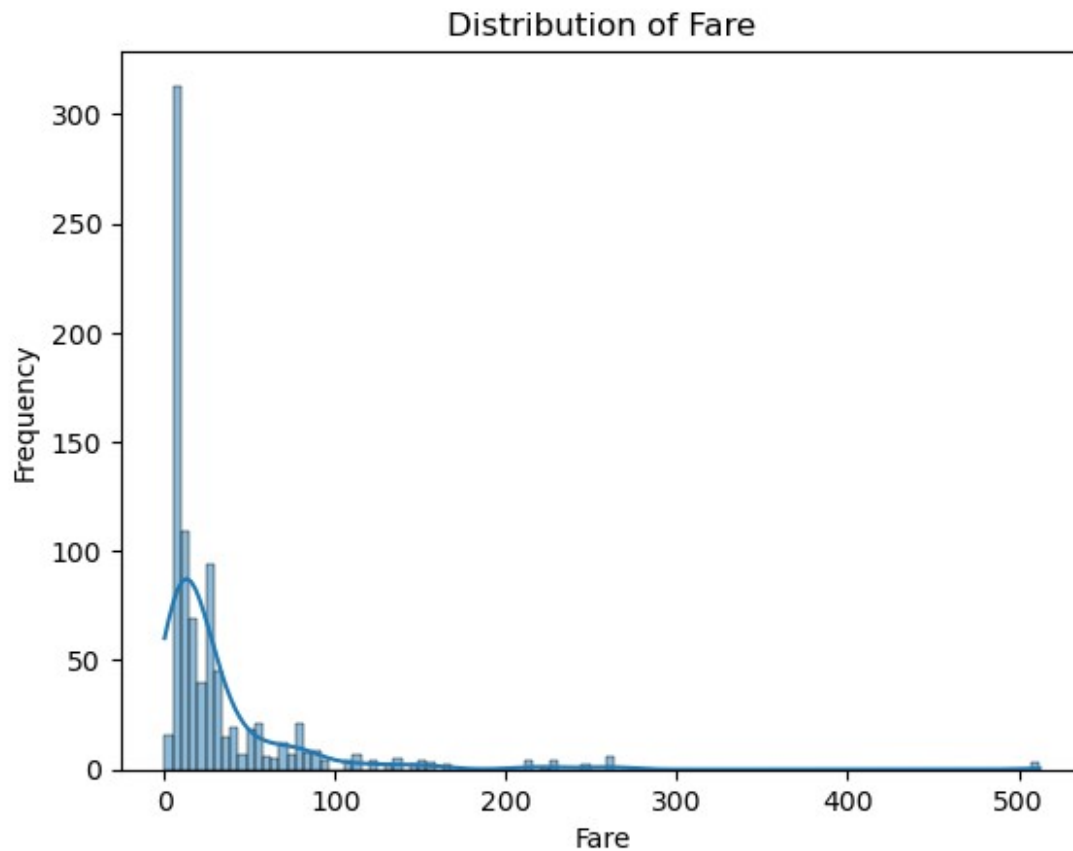
```
train = pd.read_csv('Titanic-Dataset.csv')
sns.countplot(x='Embarked', data=train)
plt.title("Count of Embarked")
plt.xlabel("Parch")
plt.ylabel("Count")
plt.show()
```



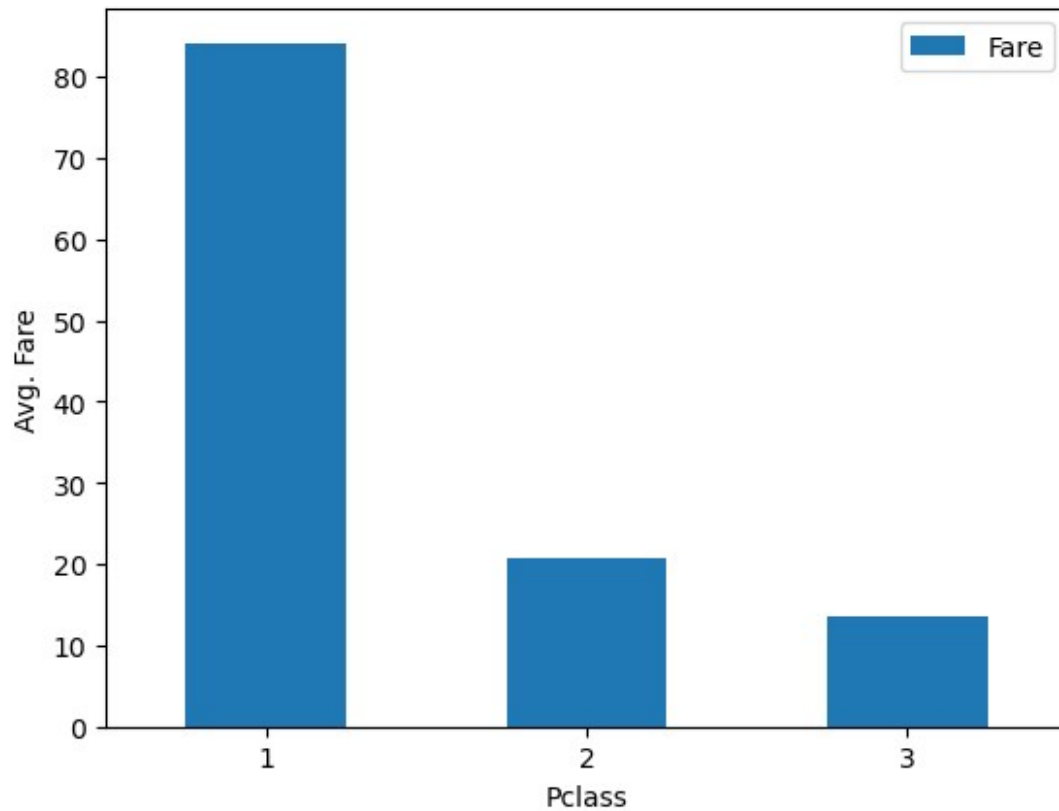
```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
train = pd.read_csv('Titanic-Dataset.csv')
sns.histplot(train['Age'], kde=True)
plt.title('Distribution of Age')
plt.xlabel('Age')
plt.ylabel('Frequency')
plt.show()
```



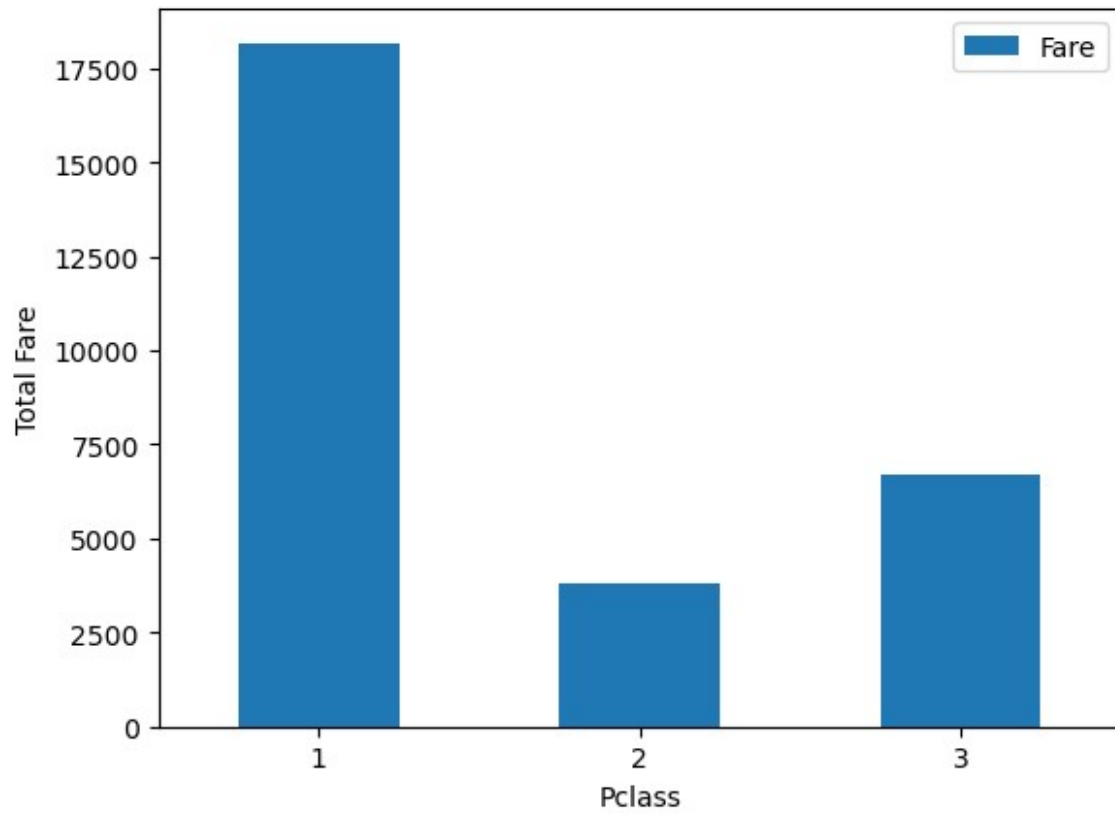
```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
train = pd.read_csv('Titanic-Dataset.csv')
sns.histplot(train['Fare'], kde=True)
plt.title('Distribution of Fare')
plt.xlabel('Fare')
plt.ylabel('Frequency')
plt.show()
```



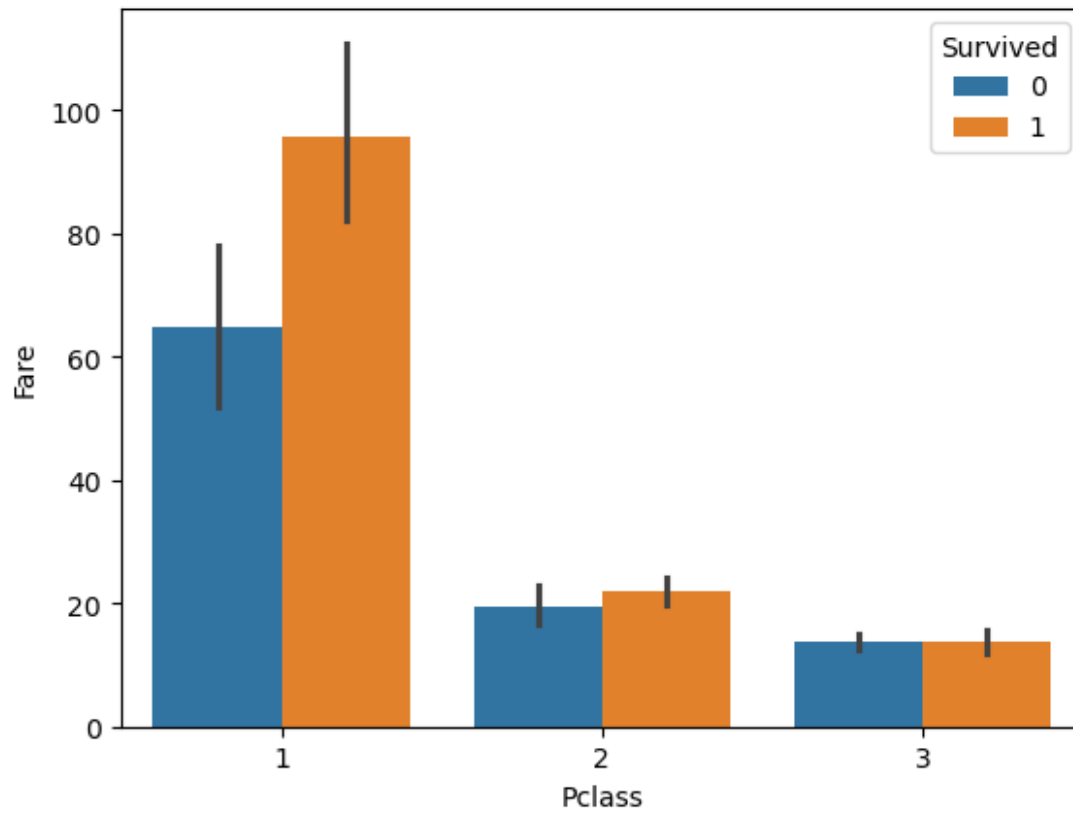
```
class_fare = train.pivot_table(index='Pclass', values='Fare')
class_fare.plot(kind='bar')
plt.xlabel('Pclass')
plt.ylabel('Avg. Fare')
plt.xticks(rotation=0)
plt.show()
```



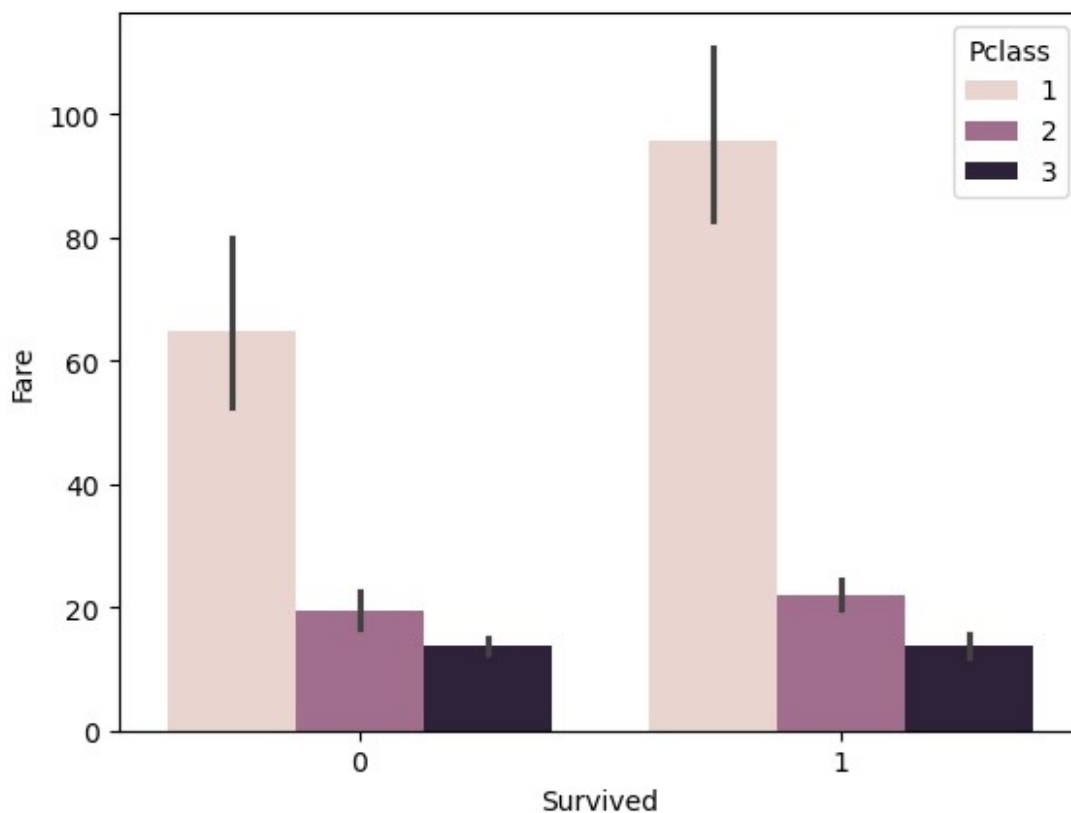
```
class_fare = train.pivot_table(index='Pclass', values='Fare',  
                                aggfunc="sum")  
class_fare.plot(kind='bar')  
plt.xlabel('Pclass')  
plt.ylabel('Total Fare')  
plt.xticks(rotation=0)  
plt.show()
```



```
sns. barplot(data=train, x='Pclass', y='Fare', hue='Survived')  
<Axes: xlabel='Pclass', ylabel='Fare'>
```



```
sns. barplot(data=train, x='Survived', y='Fare', hue='Pclass' )  
<Axes: xlabel='Survived', ylabel='Fare'>
```



```
import pandas as pd
train = pd.read_csv("Titanic-Dataset.csv")
test = pd.read_csv("Titanic-Dataset.csv")
train_len = len(train)
df = pd.concat([train, test], axis=0)
print(df.head())
```

	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

```
df.tail()
```

	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

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

```
PassengerId    0
Survived        0
Pclass          0
Name            0
Sex             0
Age            354
SibSp           0
Parch           0
Ticket          0
Fare            0
Cabin          1374
Embarked        4
dtype: int64
```

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

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

29.69911764705882

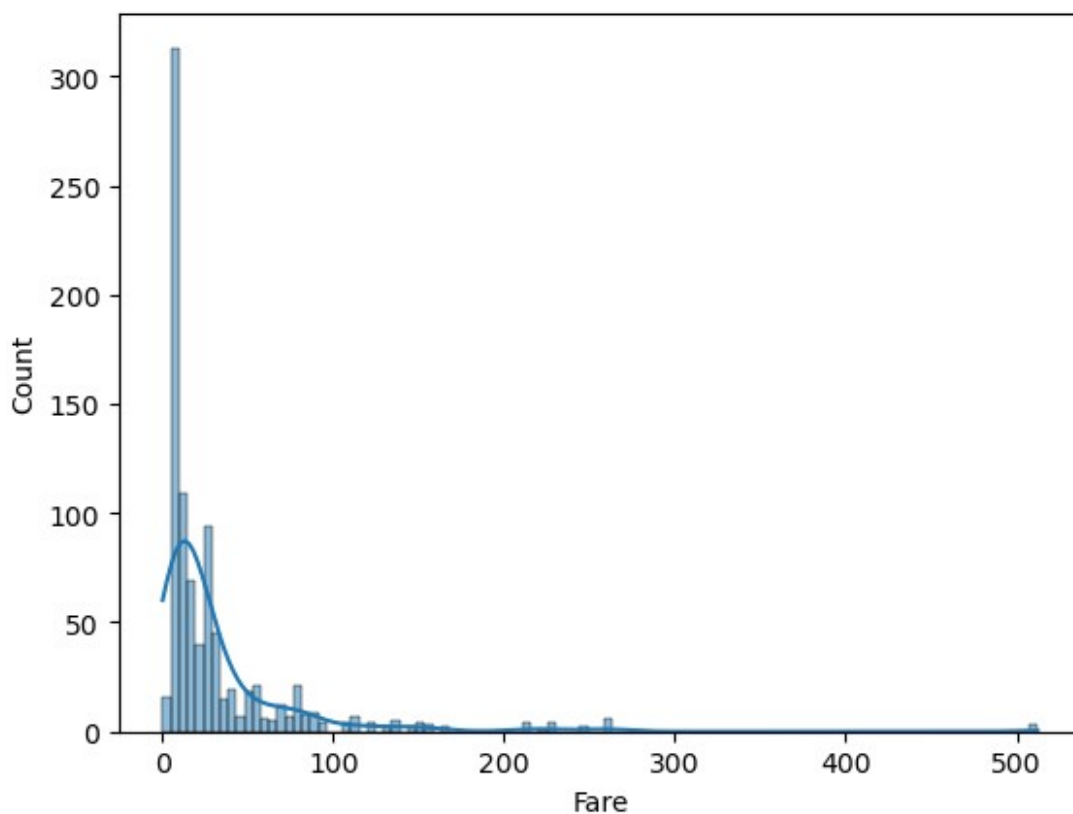
```
df['Age'] = df['Age'].fillna(df['Age'].mean())
df['Fare'] = df['Fare'].fillna(df['Fare'].mean())

df['Embarked'].mode()[0]

'S'

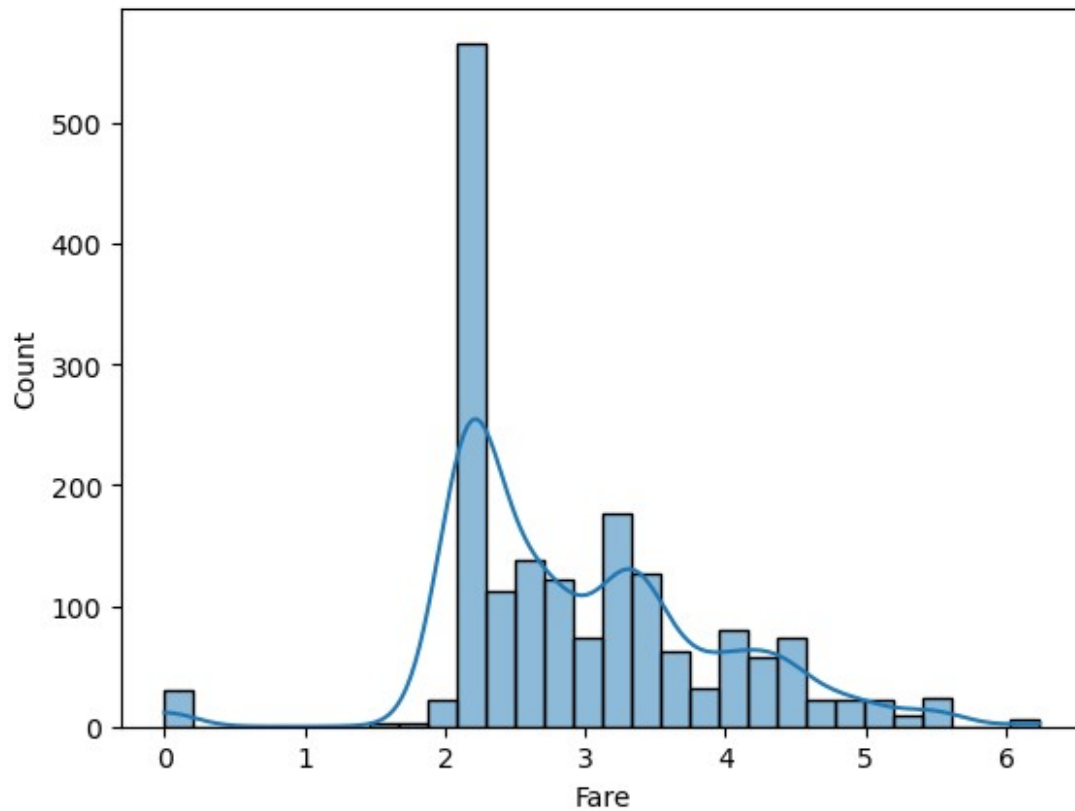
df['Embarked'] = df['Embarked'].fillna(df['Embarked'].mode()[0])

sns.histplot(train['Fare'], kde=True)
plt.show()
```

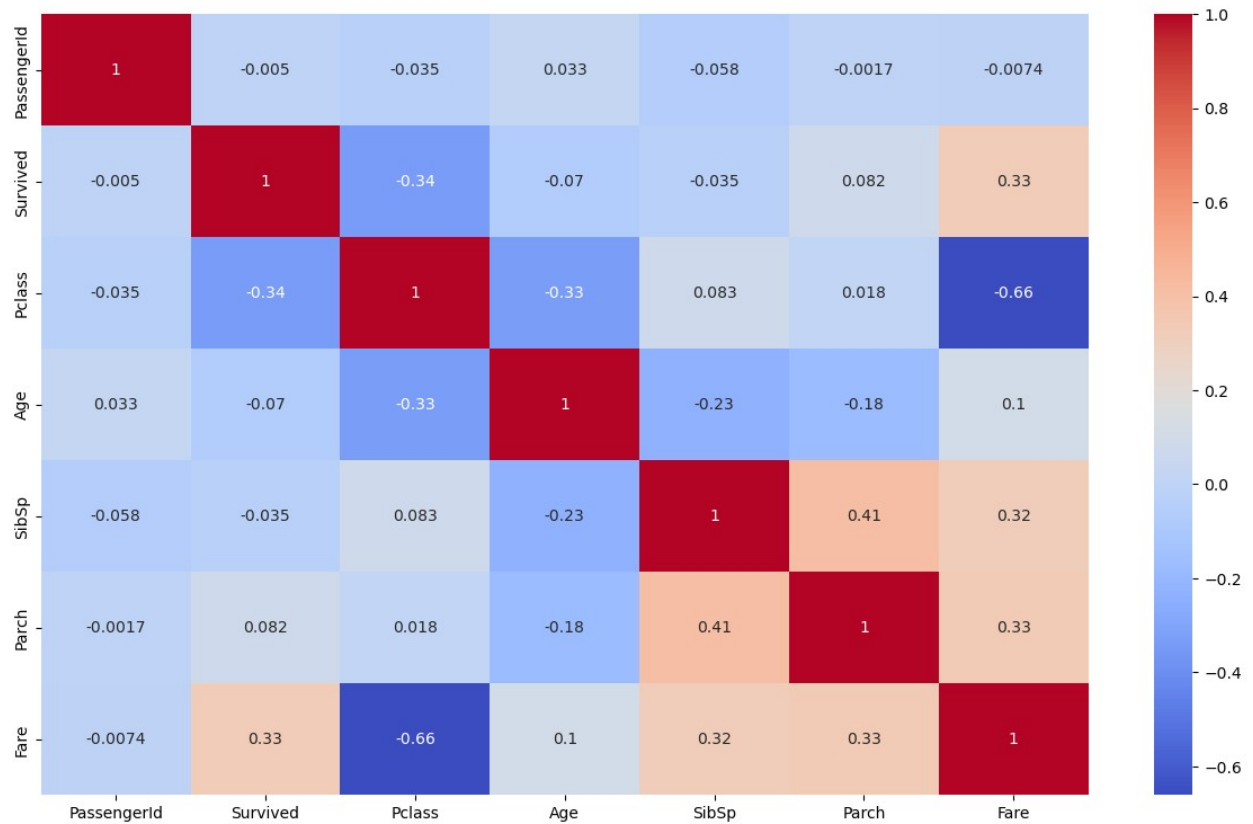


```
df['Fare'] = np. log(df ['Fare' ]+1)

sns.histplot(df['Fare'], kde=True)
plt.show()
```



```
import seaborn as sns
import matplotlib.pyplot as plt
numeric_df = df.select_dtypes(include=[float, int])
corr = numeric_df.corr()
plt.figure(figsize=(15, 9))
sns.heatmap(corr, annot=True, cmap='coolwarm')
plt.show()
```



```
df.head()
```

	PassengerId	Survived	Pclass	\
0	1	0	3	
1	2	1	1	
2	3	1	3	
3	4	1	1	
4	5	0	3	

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

	Parch	Ticket	Fare	Embarked
0	0	A/5 21171	2.110213	S
1	0	PC 17599	4.280593	C

2	0	STON/O2.	3101282	2.188856	S
3	0		113803	3.990834	S
4	0		373450	2.202765	S

```
df = df.drop(columns=[col for col in ['Name', 'Ticket'] if col in
df.columns], axis=1)
df.head()
```

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare
0	1	0	3	male	22.0	1	0	2.110213
1	2	1	1	female	38.0	1	0	4.280593
2	3	1	3	female	26.0	0	0	2.188856
3	4	1	1	female	35.0	1	0	3.990834
4	5	0	3	male	35.0	0	0	2.202765

	Embarked
0	S
1	C
2	S
3	S
4	S

```
from sklearn.preprocessing import LabelEncoder
cols = ['Sex', 'Embarked']
```

```
le = LabelEncoder()
for col in cols:
    df[col] = le.fit_transform(df[col])
df.head()
```

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	1	0	3	1	22.0	1	0	2.110213	
2									
1	2	1	1	0	38.0	1	0	4.280593	
0									
2	3	1	3	0	26.0	0	0	2.188856	
2									
3	4	1	1	0	35.0	1	0	3.990834	
2									
4	5	0	3	1	35.0	0	0	2.202765	
2									

```
train = df. iloc[ :train_len,: ]
test = df. iloc[train_len:,: ]
```

```
test.head()
```

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	1	0	3	1	22.0	1	0	2.110213	2
1	2	1	1	0	38.0	1	0	4.280593	0
2	3	1	3	0	26.0	0	0	2.188856	2
3	4	1	1	0	35.0	1	0	3.990834	2
4	5	0	3	1	35.0	0	0	2.202765	2

```
X = train.drop(columns=['PassengerId', 'Survived'], axis=1)
y = train['Survived']
```

```
X.head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.0	1	0	2.110213	2
1	1	0	38.0	1	0	4.280593	0
2	3	0	26.0	0	0	2.188856	2
3	1	0	35.0	1	0	3.990834	2
4	3	1	35.0	0	0	2.202765	2

```
from sklearn.model_selection import train_test_split, cross_val_score
import numpy as np
def classify(model, X, y):
    X_train, X_test, Y_train, Y_test = train_test_split(X, y,
test_size=0.25, random_state=42)
    model.fit(X_train, Y_train)
    print('Accuracy:', model.score(X_test, Y_test))
    score = cross_val_score(model, X, y, cv=5)
    print('CV Score:', np.mean(score))
```

```
from sklearn.linear_model import LogisticRegression
model = LogisticRegression()
classify(model, X, y)
```

```
Accuracy: 0.8071748878923767
CV Score: 0.7833971502102819
```

```
from sklearn.tree import DecisionTreeClassifier
model = DecisionTreeClassifier()
classify(model, X, y)
```

```
Accuracy: 0.726457399103139
CV Score: 0.7699516665620488
```

```
from sklearn.ensemble import RandomForestClassifier
model = RandomForestClassifier()
classify(model, X, y)
```

Accuracy: 0.8116591928251121  
CV Score: 0.8114933149205952

```
from sklearn.ensemble import ExtraTreesClassifier
model = ExtraTreesClassifier()
classify(model, X, y)
```

Accuracy: 0.7892376681614349  
CV Score: 0.7923921913250893

```
pip install lightgbm
```

Requirement already satisfied: lightgbm in c:\users\bhara\anaconda3\lib\site-packages (4.5.0)

Requirement already satisfied: numpy>=1.17.0 in c:\users\bhara\anaconda3\lib\site-packages (from lightgbm) (1.26.4)

```
Requirement already satisfied: scipy in c:\users\bhara\anaconda3\lib\
site-packages (from lightgbm) (1.13.1)
```

Note: you may need to restart the kernel to use updated packages.

```
from lightgbm import LGBMClassifier
model = LGBMClassifier ()
classify(model, X, y)
```

```
[LightGBM] [Info] Number of positive: 253, number of negative: 415
[LightGBM] [Info] Auto-choosing row-wise multi-threading, the overhead
of testing was 0.000581 seconds.
```

You can set `force row wise=true` to remove the overhead.

And if memory is not enough, you can set `force col wise=true`.

```
[LightGBM] [Info] Total Bins 191
```

```
[LightGBM] [Info] Number of data points in the train set: 668, number
of used features: 7
```

```
[LightGBM] [Info] [binary:BoostFromScore]: pavg=0.378743 ->
initscore=-0.494889
```

```
[LightGBM] [Info] Start training from score -0.494889
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

[illegible]



[illegible]

[illegible]

```
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
Accuracy: 0.8116591928251121  
[LightGBM] [Info] Number of positive: 273, number of negative: 439  
[LightGBM] [Info] Auto-choosing row-wise multi-threading, the overhead  
of testing was 0.000183 seconds.  
You can set `force_row_wise=true` to remove the overhead.  
And if memory is not enough, you can set `force_col_wise=true`.  
[LightGBM] [Info] Total Bins 203  
[LightGBM] [Info] Number of data points in the train set: 712, number
```



[illegible]

[illegible]

```
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Info] Number of positive: 274, number of negative: 439  
[LightGBM] [Info] Auto-choosing col-wise multi-threading, the overhead
```

[illegible]



[illegible]

[illegible]

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Info] Number of positive: 274, number of negative: 439
[LightGBM] [Info] Auto-choosing row-wise multi-threading, the overhead
of testing was 0.000428 seconds.
You can set `force_row_wise=true` to remove the overhead.
And if memory is not enough, you can set `force_col_wise=true`.
[LightGBM] [Info] Total Bins 204
[LightGBM] [Info] Number of data points in the train set: 713, number
```

of used features: 7

```
[LightGBM] [Info] [binary:BoostFromScore]: pavg=0.384292 ->
initscore=-0.471371
```

```
[LightGBM] [Info] Start training from score -0.471371
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
```

```
[LightGBM] [Warning] No further splits with positive gain, best gain:
```

[illegible]

[illegible]

```
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Warning] No further splits with positive gain, best gain:  
-inf  
[LightGBM] [Info] Number of positive: 274, number of negative: 439  
[LightGBM] [Info] Auto-choosing row-wise multi-threading, the overhead
```

[illegible]



[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

```
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
[LightGBM] [Warning] No further splits with positive gain, best gain:
-inf
CV Score: 0.8238277572029377
```

```
pip install catboost
```

```
Requirement already satisfied: catboost in c:\users\bhara\anaconda3\
lib\site-packages (1.2.7)
Requirement already satisfied: graphviz in c:\users\bhara\anaconda3\
lib\site-packages (from catboost) (0.20.3)
Requirement already satisfied: matplotlib in c:\users\bhara\anaconda3\
lib\site-packages (from catboost) (3.9.2)
Requirement already satisfied: numpy<2.0,>=1.16.0 in c:\users\bhara\
anaconda3\lib\site-packages (from catboost) (1.26.4)
Requirement already satisfied: pandas>=0.24 in c:\users\bhara\
anaconda3\lib\site-packages (from catboost) (2.2.2)
Requirement already satisfied: scipy in c:\users\bhara\anaconda3\lib\
site-packages (from catboost) (1.13.1)
Requirement already satisfied: plotly in c:\users\bhara\anaconda3\lib\
site-packages (from catboost) (5.24.1)
Requirement already satisfied: six in c:\users\bhara\appdata\roaming\
python\python312\site-packages (from catboost) (1.17.0)
Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\
bhara\appdata\roaming\python\python312\site-packages (from
pandas>=0.24->catboost) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in c:\users\bhara\
anaconda3\lib\site-packages (from pandas>=0.24->catboost) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in c:\users\bhara\
anaconda3\lib\site-packages (from pandas>=0.24->catboost) (2023.3)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\bhara\
anaconda3\lib\site-packages (from matplotlib->catboost) (1.2.0)
```



Requirement already satisfied: cycloer>=0.10 in c:\users\bhara\anaconda3\lib\site-packages (from matplotlib->catboost) (0.11.0)  
 Requirement already satisfied: fonttools>=4.22.0 in c:\users\bhara\anaconda3\lib\site-packages (from matplotlib->catboost) (4.51.0)  
 Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\bhara\anaconda3\lib\site-packages (from matplotlib->catboost) (1.4.4)  
 Requirement already satisfied: packaging>=20.0 in c:\users\bhara\appdata\roaming\python\python312\site-packages (from matplotlib->catboost) (24.2)  
 Requirement already satisfied: pillow>=8 in c:\users\bhara\anaconda3\lib\site-packages (from matplotlib->catboost) (10.4.0)  
 Requirement already satisfied: pyparsing>=2.3.1 in c:\users\bhara\anaconda3\lib\site-packages (from matplotlib->catboost) (3.1.2)  
 Requirement already satisfied: tenacity>=6.2.0 in c:\users\bhara\anaconda3\lib\site-packages (from plotly->catboost) (8.2.3)  
 Note: you may need to restart the kernel to use updated packages.

```
from catboost import CatBoostClassifier
model = CatBoostClassifier(verbose=0)
classify(model, X, y)
```

Accuracy: 0.8295964125560538  
 CV Score: 0.8226790534178645

```
model = LGBMClassifier()
model.fit(X, y)
```

[LightGBM] [Info] Number of positive: 342, number of negative: 549  
 [LightGBM] [Info] Auto-choosing col-wise multi-threading, the overhead of testing was 0.000552 seconds.  
 You can set `force\_col\_wise=true` to remove the overhead.  
 [LightGBM] [Info] Total Bins 222  
 [LightGBM] [Info] Number of data points in the train set: 891, number of used features: 7  
 [LightGBM] [Info] [binary:BoostFromScore]: pavg=0.383838 -> initscore=-0.473288  
 [LightGBM] [Info] Start training from score -0.473288  
 [LightGBM] [Warning] No further splits with positive gain, best gain: -inf

```
LGBMClassifier()
```

```
test.head()
```

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare
0	1	0	3	1	22.0	1	0	2.110213
2	2	1	1	0	38.0	1	0	4.280593
0	3	1	3	0	26.0	0	0	2.188856

```
2
3          4          1          1          0  35.0          1          0  3.990834
2
4          5          0          3          1  35.0          0          0  2.202765
2
```

```
X_test = train.drop(columns=[ 'PassengerId', 'Survived'], axis=1)
```

```
X_test. head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.0	1	0	2.110213	2
1	1	0	38.0	1	0	4.280593	0
2	3	0	26.0	0	0	2.188856	2
3	1	0	35.0	1	0	3.990834	2
4	3	1	35.0	0	0	2.202765	2

```
pred = model.predict(X_test)
```

```
pred
```

```
array([0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0,
0,
      1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
1,
      1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0,
1,
      1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 0,
0,
      1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
1,
      0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0,
0,
      0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0,
0,
      0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0,
0,
      0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0,
0,
      1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1,
0,
      0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0,
1,
      0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1,
0,
      1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0,
0,
      0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1,
1,
      0, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 1, 1,
1,
```

0,	1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0,
0,	0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 0,
0,	0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1,
0,	0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1,
1,	0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0,
0,	1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1,
0,	0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0,
1,	1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1,
0,	1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1,
0,	0, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0,
1,	1, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
1,	1, 1, 0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0,
0,	0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0,
1,	0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1,
0,	0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0,
0,	1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0,
1,	0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0,
0,	0, 0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1,
0,	1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0,
1,	0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 0,
0,	0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 0,
0,	0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0,
0,	0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0,
1,	0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1, 1,
1,	1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0,

```
1,
      1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0], dtype=int64)
```

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

	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

```
sub.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
```

```
sub[ 'Survived' ] = pred
```

```
sub. head()
```

	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

```
sub.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

```

```
sub.head()
```

	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

```
sub.to_csv('submission.csv', index=False)
```

```
import matplotlib.pyplot as plt
```

```
# Calculate the counts for each category in 'Sex'
```

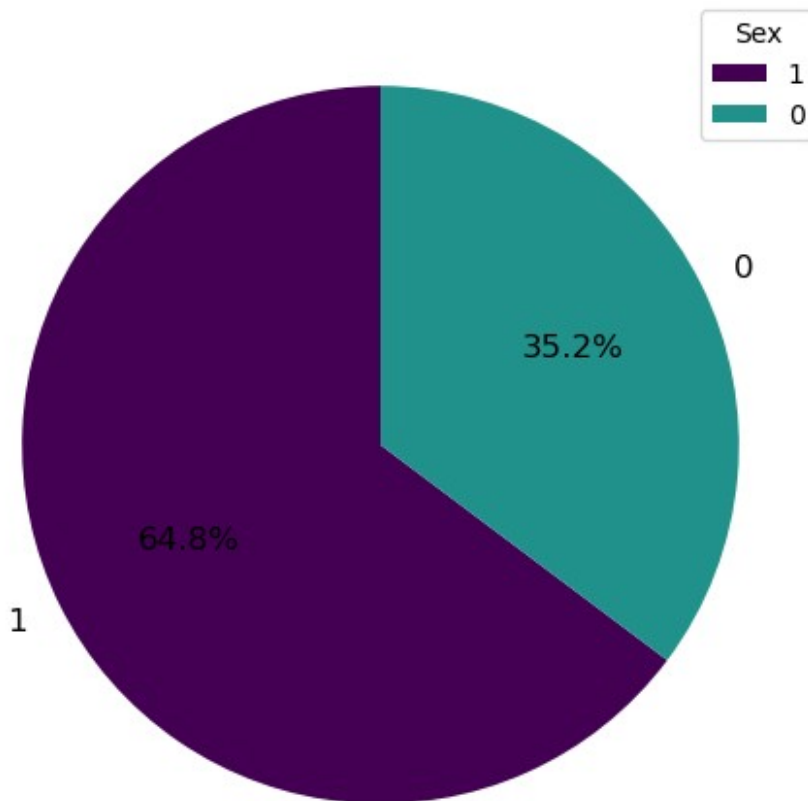
```
sex_counts = train['Sex'].value_counts()
```

```
# Create a pie chart
```

```
plt.figure(figsize=(8, 6))
```

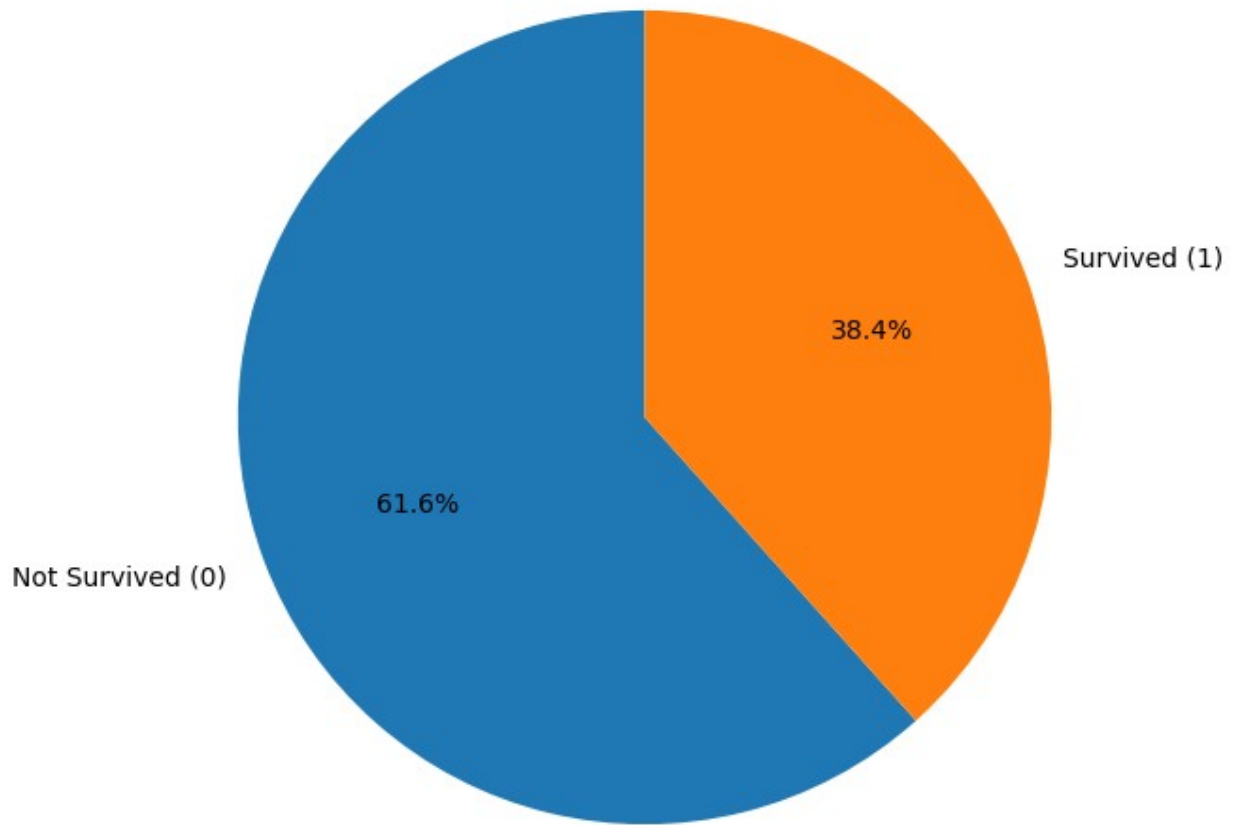
```
plt.pie(sex_counts, labels=sex_counts.index,
        autopct='%1.1f%%', startangle=90,
        colors=['#440154', '#21918c'], textprops={'fontsize': 12})
plt.title("Exploratory Data Analysis: Sex Distribution", fontsize=14)
plt.legend(title="Sex", loc="upper right", fontsize=10)
plt.show()
```

Exploratory Data Analysis: Sex Distribution



```
survival_counts = df['Survived'].value_counts()
plt.figure(figsize=(7, 7))
plt.pie(survival_counts, labels=['Not Survived (0)', 'Survived (1)'],
        autopct='%1.1f%%', startangle=90)
plt.title("Survival Distribution")
plt.ylabel("")
plt.show()
```

Survival Distribution



```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Index: 1782 entries, 0 to 890
```

```
Data columns (total 9 columns):
```

#	Column	Non-Null Count	Dtype
0	PassengerId	1782 non-null	int64
1	Survived	1782 non-null	int64
2	Pclass	1782 non-null	int64
3	Sex	1782 non-null	int32
4	Age	1782 non-null	float64
5	SibSp	1782 non-null	int64
6	Parch	1782 non-null	int64
7	Fare	1782 non-null	float64
8	Embarked	1782 non-null	int32



```
dtypes: float64(2), int32(2), int64(5)
memory usage: 125.3 KB
```

```
import pandas as pd
f=pd.read_csv("train.csv")
f
```

	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]

f.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

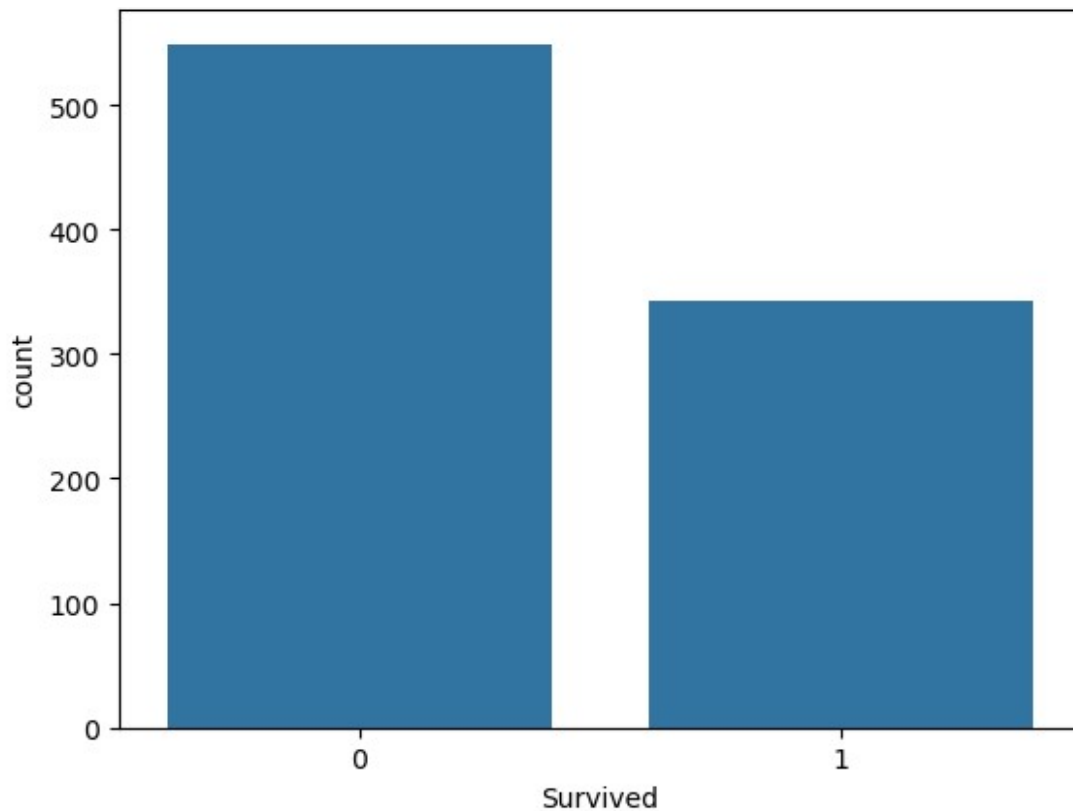
f.isnull().sum()

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

```
import seaborn as sns
sns.countplot(x="Survived", data=f)
```

```
<Axes: xlabel='Survived', ylabel='count'>
```



```
survival_counts = f.groupby(["Pclass", "Survived"]).size().unstack()  
print(survival_counts)
```

```
Survived    0    1  
Pclass  
1           80  136  
2           97   87  
3          372  119
```

```
# Check missing values in Class 1 grouped by Sex
```

```
missing_counts = f[f["Pclass"] == 1].isnull().sum()
```

```
# Count missing values separately for male and female passengers
```

```
missing_by_sex = f[f["Pclass"] == 1].groupby("Sex").apply(lambda x:  
x.isnull().sum())
```

```
print("Total Missing Values in Class 1:\n", missing_counts)
```

```
print("\nMissing Values in Class 1 by Sex:\n", missing_by_sex)
```

Total Missing Values in Class 1:

PassengerId	0
Survived	0
Pclass	0
Name	0
Sex	0
Age	30
SibSp	0
Parch	0
Ticket	0
Fare	0
Cabin	40
Embarked	2

dtype: int64

Missing Values in Class 1 by Sex:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch
Ticket \								
Sex								

female	0	0	0	0	0	9	0	0
male	0	0	0	0	0	21	0	0

	Fare	Cabin	Embarked
Sex			
female	0	13	2
male	0	27	0

```
surviving_men_class1 = f[(f["Pclass"] == 1) & (f["Sex"] == "male") & (f["Survived"] == 1)]
```

```
count_surviving_men_class1 = surviving_men_class1.shape[0]
```

```
print("Number of surviving men in Class 1:", count_surviving_men_class1)
```

Number of surviving men in Class 1: 45

```
surviving = f[(f["Survived"] == 1) & (f["Age"] == 47)]  
count_surviving = surviving.shape[0]  
print("Number of surviving :", count_surviving)
```

Number of surviving : 1

```
surviving_men_class2 = f[(f["Pclass"] == 2) & (f["Survived"] == 1)]
```

```
print("Number of surviving men in Class 2:",  
count_surviving_men_class)
```

Number of surviving men in Class 2: 136

```
a = df[df["Survived"] == 1]
```

```
b = a[a["Pclass"] == 1]
```

```
b.Age.value_counts()
```

Age	
29.699118	28
35.000000	18
36.000000	14
48.000000	10
24.000000	10
30.000000	10
38.000000	8
39.000000	8
49.000000	8
22.000000	8
17.000000	6
27.000000	6
52.000000	6
33.000000	6
42.000000	6
18.000000	6
16.000000	6
19.000000	6
58.000000	6
31.000000	6
40.000000	6
23.000000	6
45.000000	4
51.000000	4
56.000000	4
21.000000	4
28.000000	4
54.000000	4
50.000000	4
26.000000	4
44.000000	4
25.000000	4
60.000000	4
32.000000	4
14.000000	2

```
15.000000    2
62.000000    2
11.000000    2
43.000000    2
0.920000     2
29.000000    2
37.000000    2
63.000000    2
80.000000    2
4.000000     2
53.000000    2
41.000000    2
34.000000    2
47.000000    2
```

```
Name: count, dtype: int64
```

```
a = df[df["Survived"] == 0]
```

```
b = a[a["Pclass"] == 3]
```

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

```
Sex
```

```
1    600
0    144
```

```
Name: count, dtype: int64
```

```
# Filter for passengers who did not survive (died)
```

```
a = df[df["Survived"] == 0]
```

```
# Further filter for passengers in third class (Pclass = 3)
```

```
b = a[a["Pclass"] == 3]
```

```
# Count the number of females
```

```
num_female_died = b[b["Sex"] == "female"].shape[0]
```

```
print(f"Number of female passengers who died in third class:
{num_female_died}")
```

```
Number of female passengers who died in third class: 0
```