

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
# Load the data
train_df = pd.read_csv('/content/train.csv')
test_df = pd.read_csv('/content/test.csv')
# Combine the training and testing data
df = pd.concat([train_df, test_df])
# Data Cleaning
print(df.isnull().sum())

```

```

PassengerId      0
Survived          418
Pclass            0
Name              0
Sex               0
Age              263
SibSp             0
Parch             0
Ticket           0
Fare              1
Cabin            1014
Embarked          2
dtype: int64

```

```

# Fill missing values in Age with median age
df['Age'].fillna(df['Age'].median(), inplace=True)

```

```

# Fill missing values in Embarked with most frequent value
df['Embarked'].fillna(df['Embarked'].mode()[0], inplace=True)

```

```

# Convert categorical variables to numerical variables
df['Sex'] = df['Sex'].map({'male': 0, 'female': 1})
df['Embarked'] = df['Embarked'].map({'S': 0, 'C': 1, 'Q': 2})

```

```

# Exploratory Data Analysis (EDA)
# Summary statistics
print(df.describe())

```

```

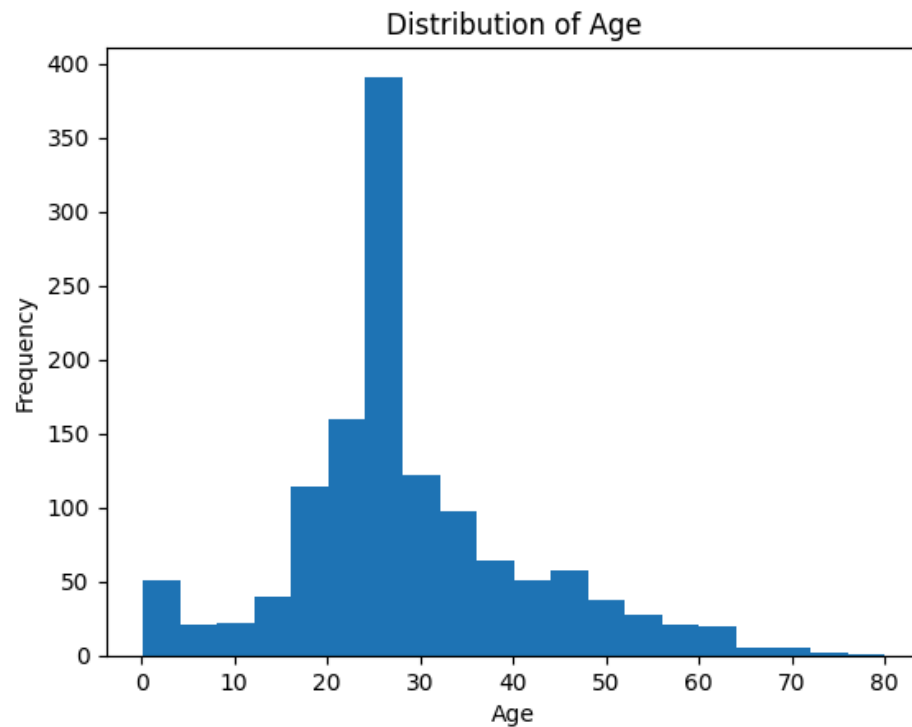
PassengerId      Survived      Pclass      Sex      Age \
count  1309.000000  891.000000  1309.000000  1309.000000  1309.000000
mean    655.000000    0.383838    2.294882    0.355997    29.503186
std     378.020061    0.486592    0.837836    0.478997    12.905241
min       1.000000    0.000000    1.000000    0.000000    0.170000

```

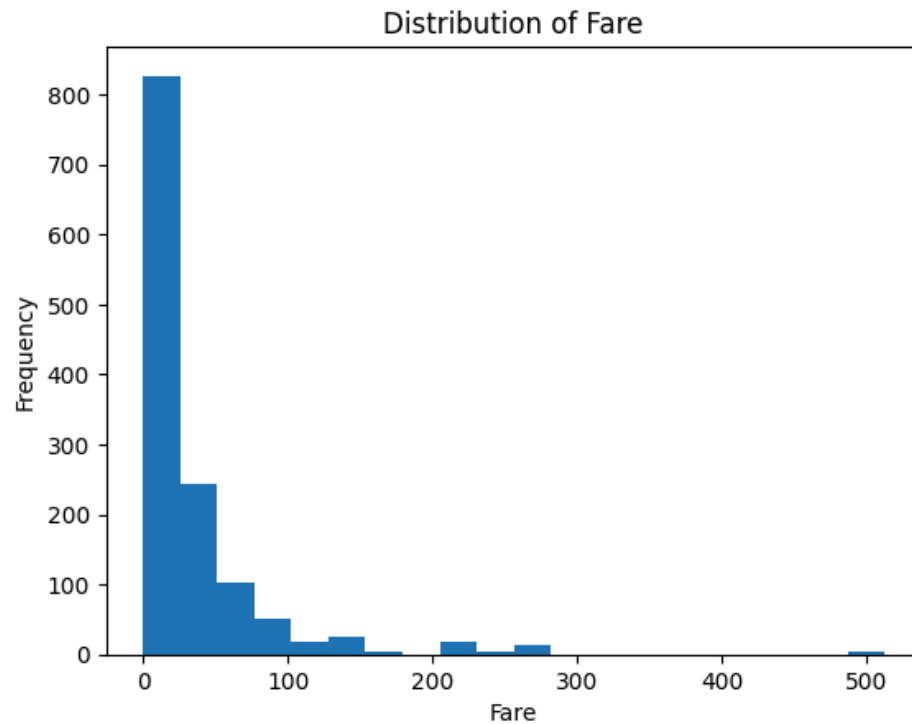
25%	328.000000	0.000000	2.000000	0.000000	22.000000
50%	655.000000	0.000000	3.000000	0.000000	28.000000
75%	982.000000	1.000000	3.000000	1.000000	35.000000
max	1309.000000	1.000000	3.000000	1.000000	80.000000

	SibSp	Parch	Fare	Embarked
count	1309.000000	1309.000000	1308.000000	1309.000000
mean	0.498854	0.385027	33.295479	0.394194
std	1.041658	0.865560	51.758668	0.653499
min	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	7.895800	0.000000
50%	0.000000	0.000000	14.454200	0.000000
75%	1.000000	0.000000	31.275000	1.000000
max	8.000000	9.000000	512.329200	2.000000

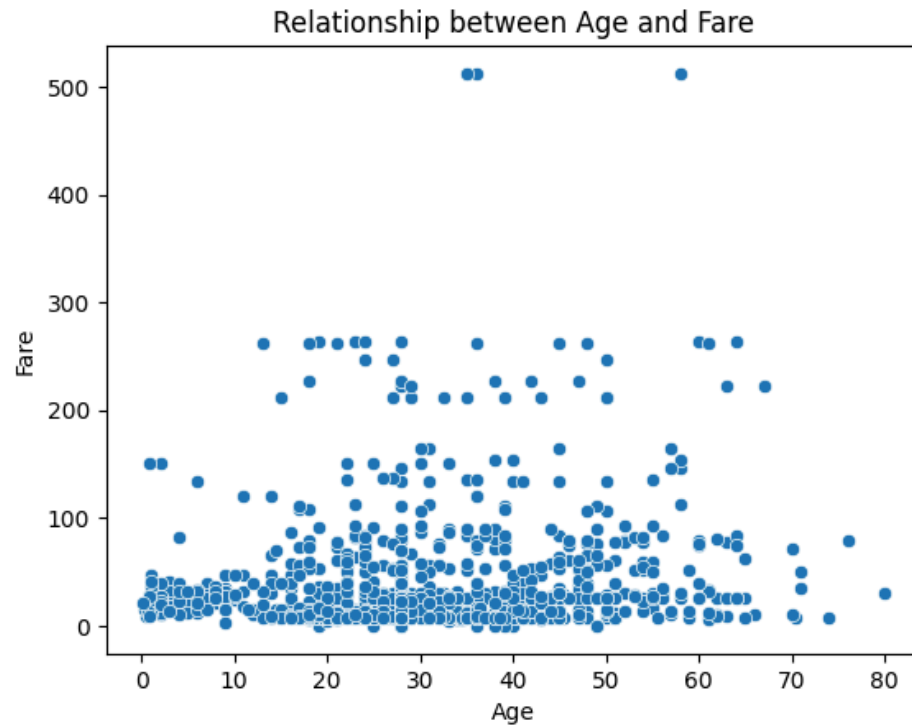
```
# Distribution of Age
plt.hist(df['Age'], bins=20)
plt.xlabel('Age')
plt.ylabel('Frequency')
plt.title('Distribution of Age')
plt.show()
```



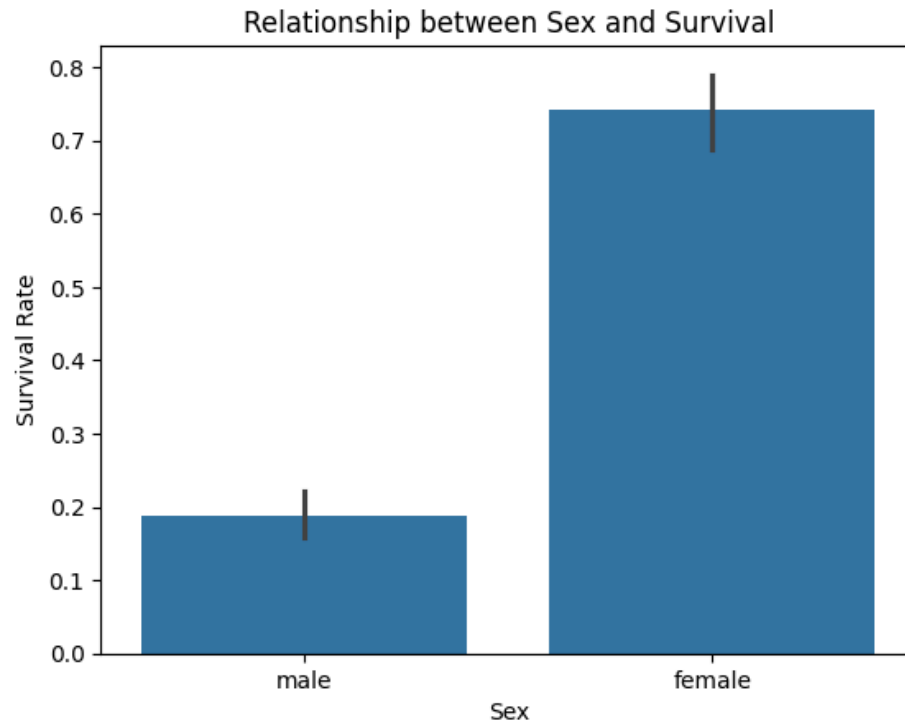
```
# Distribution of Fare
plt.hist(df['Fare'], bins=20)
plt.xlabel('Fare')
plt.ylabel('Frequency')
plt.title('Distribution of Fare')
plt.show()
```



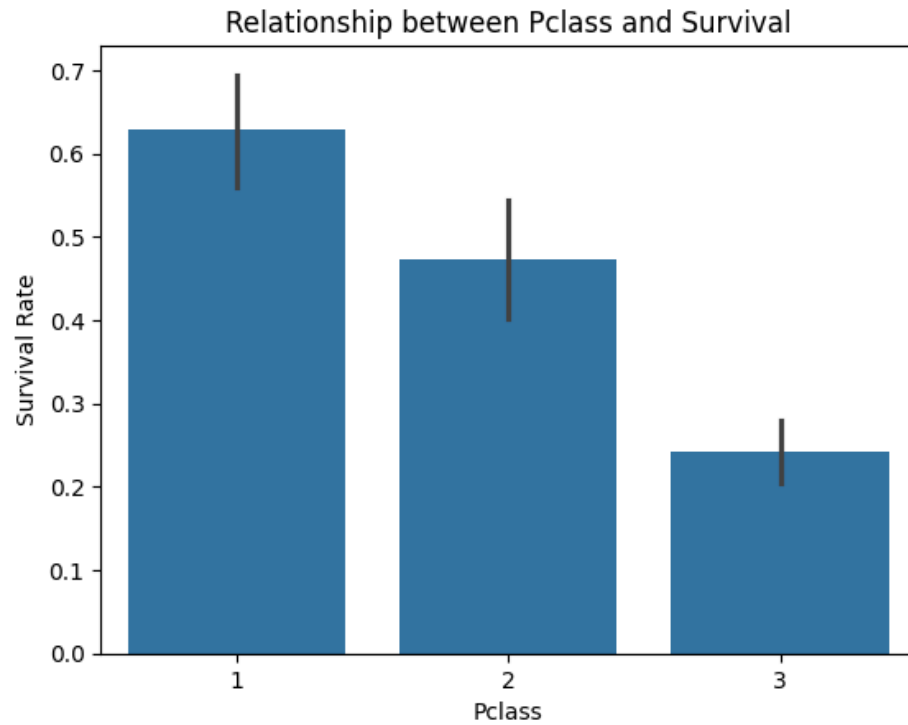
```
# Relationship between Age and Fare
df_reset = df.reset_index(drop=True)
sns.scatterplot(x='Age', y='Fare', data=df_reset)
plt.xlabel('Age')
plt.ylabel('Fare')
plt.title('Relationship between Age and Fare')
plt.show()
```



```
# Relationship between Sex and Survival
sns.barplot(x='Sex', y='Survived', data=train_df)
plt.xlabel('Sex')
plt.ylabel('Survival Rate')
plt.title('Relationship between Sex and Survival')
plt.show()
```




```
# Relationship between Pclass and Survival
sns.barplot(x='Pclass', y='Survived', data=train_df)
plt.xlabel('Pclass')
plt.ylabel('Survival Rate')
plt.title('Relationship between Pclass and Survival')
plt.show()
```



```
# Correlation matrix
numerical_df = df.select_dtypes(include=['number'])
corr_matrix = numerical_df.corr()
print(corr_matrix)

# Heatmap of correlation matrix
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', square=True)
plt.title('Correlation Matrix')
plt.show()
```



	PassengerId	Survived	Pclass	Sex	Age	SibSp	\
PassengerId	1.000000	-0.005007	-0.038354	-0.013406	0.025799	-0.055224	
Survived	-0.005007	1.000000	-0.338481	0.543351	-0.064910	-0.035322	
Pclass	-0.038354	-0.338481	1.000000	-0.124617	-0.377908	0.060832	
Sex	-0.013406	0.543351	-0.124617	1.000000	-0.053663	0.109609	
Age	0.025799	-0.064910	-0.377908	-0.053663	1.000000	-0.189972	
SibSp	-0.055224	-0.035322	0.060832	0.109609	-0.189972	1.000000	
Parch	0.008942	0.081629	0.018322	0.213125	-0.125851	0.373587	
Fare	0.031428	0.257307	-0.558629	0.185523	0.179256	0.160238	
Embarked	0.040143	0.106811	0.038875	0.120423	0.018654	-0.073461	

	Parch	Fare	Embarked
PassengerId	0.008942	0.031428	0.040143
Survived	0.081629	0.257307	0.106811
Pclass	0.018322	-0.558629	0.038875
Sex	0.213125	0.185523	0.120423
Age	-0.125851	0.179256	0.018654
SibSp	0.373587	0.160238	-0.073461
Parch	1.000000	0.221539	-0.095523
Fare	0.221539	1.000000	0.061126
Embarked	-0.095523	0.061126	1.000000

