

```

import pandas as pd

titanic_data = pd.read_csv('/content/train.csv')

print(titanic_data.info())

print(titanic_data.head())

print("Missing Values:")
print(titanic_data.isnull().sum())

titanic_data.drop('Cabin', axis=1, inplace=True)

titanic_data['Age'].fillna(titanic_data['Age'].median(), inplace=True)

titanic_data['Embarked'].fillna(titanic_data['Embarked'].mode()[0], inplace=True)

print("Missing Values After Handling:")
print(titanic_data.isnull().sum())

titanic_data['Sex'] = titanic_data['Sex'].map({'male': 0, 'female': 1})

titanic_data = pd.get_dummies(titanic_data, columns=['Embarked'], drop_first=True)

titanic_data.drop(['PassengerId', 'Name', 'Ticket'], axis=1, inplace=True)

print("Cleaned Dataset:")
print(titanic_data.head())

```

```

<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
None
   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
Missing Values:
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
Missing Values After Handling:
PassengerId 0
Survived     0
Pclass       0
```

```
# Import necessary libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

# Load the Titanic dataset
titanic_data = pd.read_csv('/content/train.csv')

# Display basic information about the dataset
print(titanic_data.info())

# Display the first few rows of the dataset
print(titanic_data.head())

# Summary statistics for numerical features
print(titanic_data.describe())

# Visualize the distribution of 'Age' using a histogram
plt.figure(figsize=(8, 6))
sns.histplot(titanic_data['Age'], bins=30, kde=True)
plt.title('Distribution of Age')
plt.xlabel('Age')
plt.ylabel('Count')
plt.show()

# Survival distribution based on 'Pclass'
plt.figure(figsize=(8, 6))
sns.countplot(x='Pclass', hue='Survived', data=titanic_data)
plt.title('Survival Distribution by Passenger Class')
plt.xlabel('Passenger Class')
plt.ylabel('Count')
plt.show()

# Survival distribution based on 'Sex'
plt.figure(figsize=(8, 6))
sns.countplot(x='Sex', hue='Survived', data=titanic_data)
plt.title('Survival Distribution by Gender')
plt.xlabel('Gender')
plt.ylabel('Count')
plt.show()

# Survival distribution based on 'Embarked' port
plt.figure(figsize=(8, 6))
sns.countplot(x='Embarked', hue='Survived', data=titanic_data)
plt.title('Survival Distribution by Embarked Port')
plt.xlabel('Embarked Port')
plt.ylabel('Count')
plt.show()

# Correlation heatmap for numerical features
plt.figure(figsize=(10, 8))
correlation_matrix = titanic_data.corr()
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt=".2f")
plt.title('Correlation Heatmap')
plt.show()
```

```
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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
None
```

	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

	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

Distribution of Age

