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# 1. Import Libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

# Set styles
sns.set(style="whitegrid")
%matplotlib inline

# 2. Load Dataset
df = pd.read_csv("train.csv")

# 3. Data Overview
print("Dataset Shape:", df.shape)
display(df.head())
display(df.info())
display(df.describe())
print("Missing Values:\n", df.isnull().sum())

# 4. Univariate Analysis

# Plotting categorical features
categorical = ['Survived', 'Pclass', 'Sex', 'Embarked']
for col in categorical:
    sns.countplot(x=col, data=df)
    plt.title(f'Distribution of {col}')
    plt.show()

# Plotting numerical features
numerical = ['Age', 'Fare', 'SibSp', 'Parch']
for col in numerical:
    plt.figure(figsize=(12, 5))
    plt.subplot(1, 2, 1)
    sns.histplot(df[col].dropna(), kde=True)
    plt.title(f'Histogram of {col}')

    plt.subplot(1, 2, 2)
    sns.boxplot(x=col, data=df)
    plt.title(f'Boxplot of {col}')
    plt.show()

# 5. Bivariate Analysis

# Survival vs Sex
sns.countplot(x='Sex', hue='Survived', data=df)
plt.title("Survival Count by Sex")
plt.show()

# Survival vs Pclass

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sns.countplot(x='Pclass', hue='Survived', data=df)
plt.title("Survival Count by Pclass")
plt.show()

# Age vs Survival
sns.histplot(data=df, x='Age', hue='Survived', kde=True,
element="step")
plt.title("Age Distribution by Survival")
plt.show()

# 6. Correlation and Heatmap
plt.figure(figsize=(10, 6))
numeric_df = df.select_dtypes(include=[np.number])
sns.heatmap(numeric_df.corr(), annot=True, cmap='coolwarm', fmt=".2f")
plt.title("Correlation Heatmap")
plt.show()

# Pairplot (optional, heavy)
# sns.pairplot(df[['Survived', 'Pclass', 'Age', 'Fare', 'SibSp',
'Parch']], hue='Survived')

# 7. Handling Skewness
from scipy.stats import skew

print("Skewness of numerical features:")
for col in ['Age', 'Fare']:
    val = skew(df[col].dropna())
    print(f"{col}: {val:.2f}")

# Apply log1p to fix skewness in 'Fare'
df['Fare_log'] = np.log1p(df['Fare'])

# Before and after log transformation
sns.histplot(df['Fare'], kde=True, color='red', label='Original')
sns.histplot(df['Fare_log'], kde=True, color='blue', label='Log
Transformed')
plt.title("Fare Distribution Before and After Log Transformation")
plt.legend()
plt.show()

# 8. Summary of Findings
summary = """
- Most passengers did not survive (~62%).
- Females had a much higher survival rate than males.
- Passengers in 1st class had a better chance of survival.
- Age and Fare show right skew; Fare was log transformed.
- Missing values found in Age, Cabin, and Embarked.
- Correlation highest between Pclass & Fare, SibSp & Parch.
"""
print(summary)

```

Dataset Shape: (891, 12)

	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

<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	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

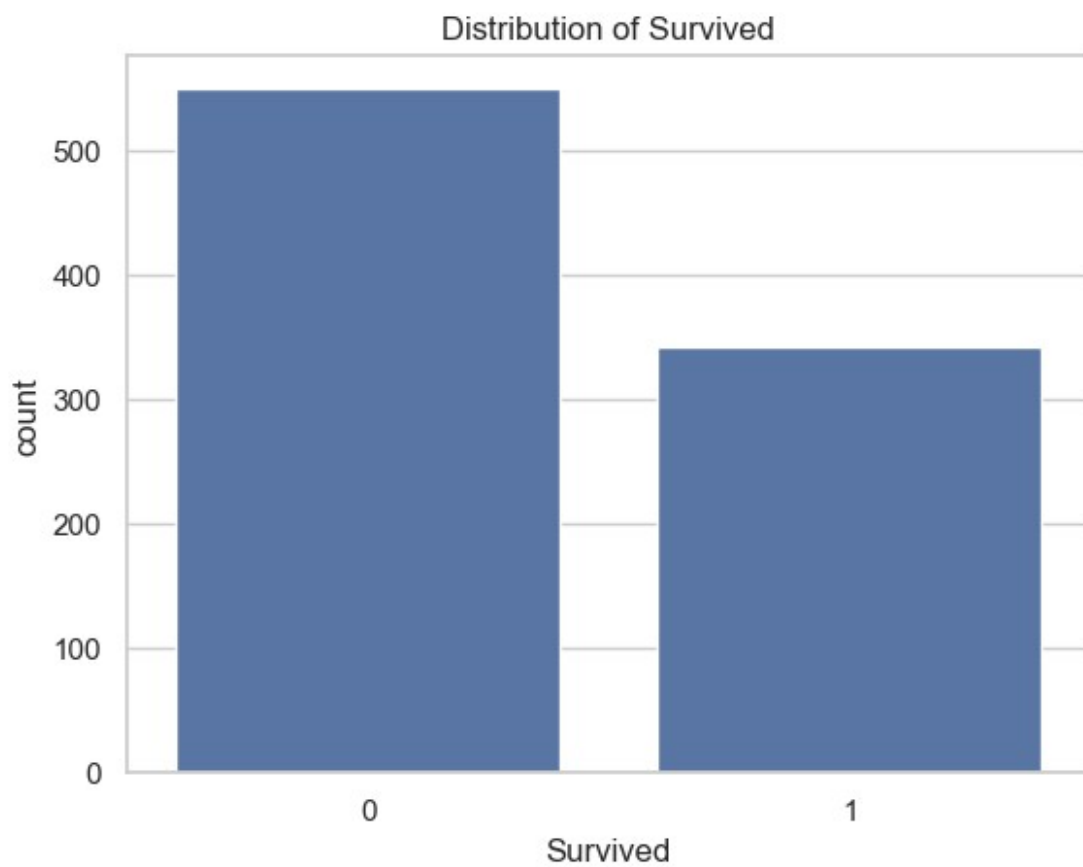
	Parch	Fare
count	891.000000	891.000000
mean	0.381594	32.204208
std	0.806057	49.693429
min	0.000000	0.000000
25%	0.000000	7.910400
50%	0.000000	14.454200
75%	0.000000	31.000000
max	6.000000	512.329200

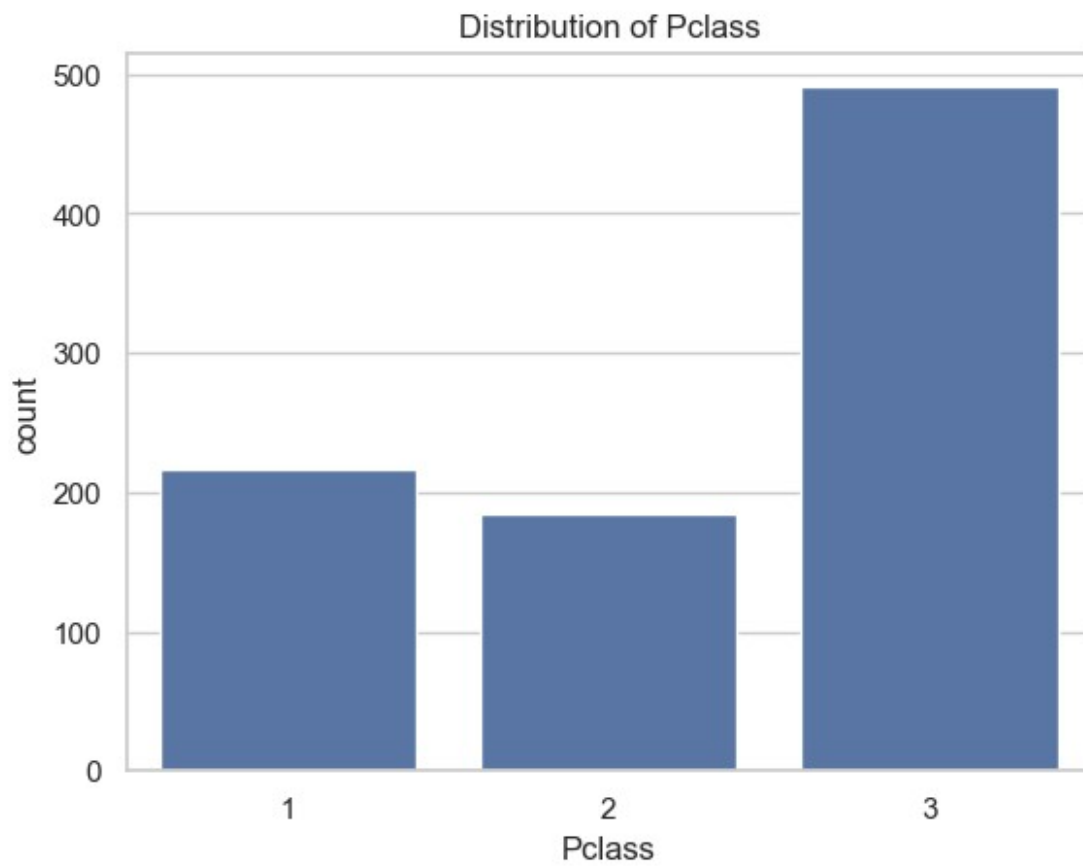
  

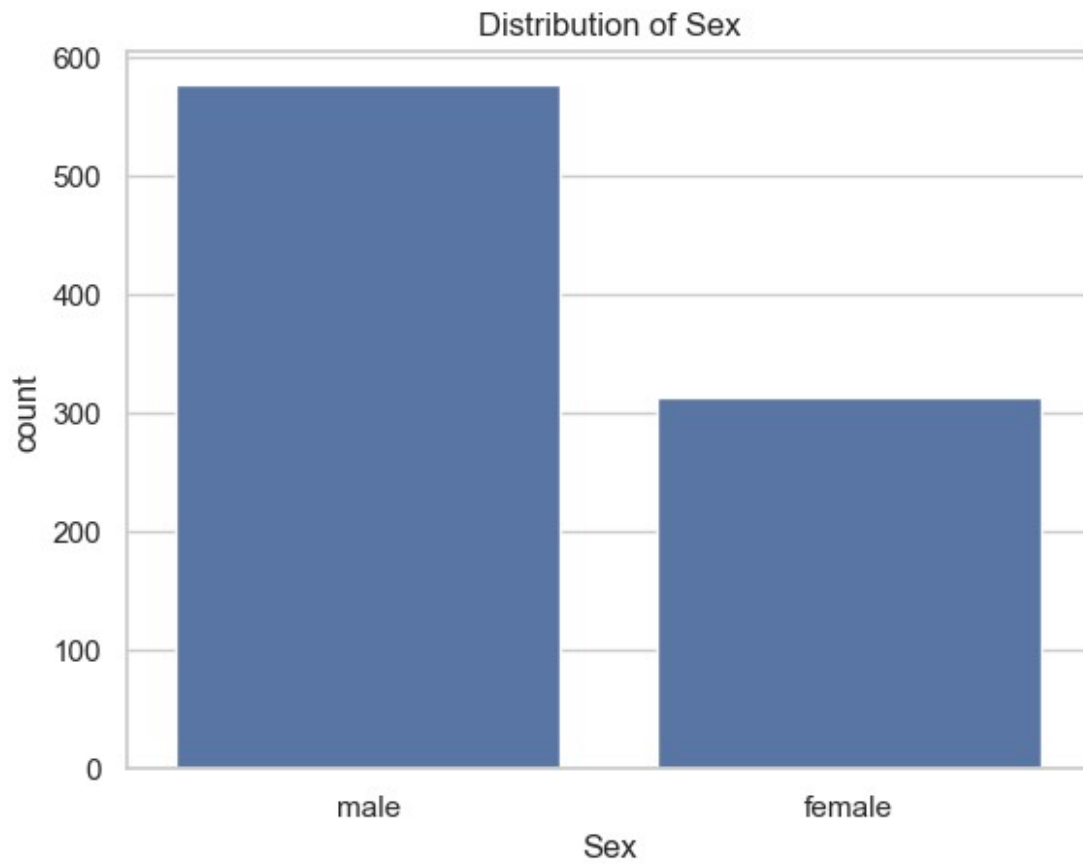
Missing Values:

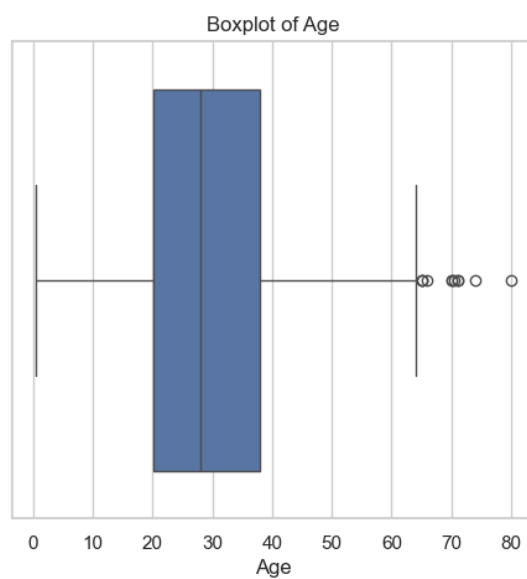
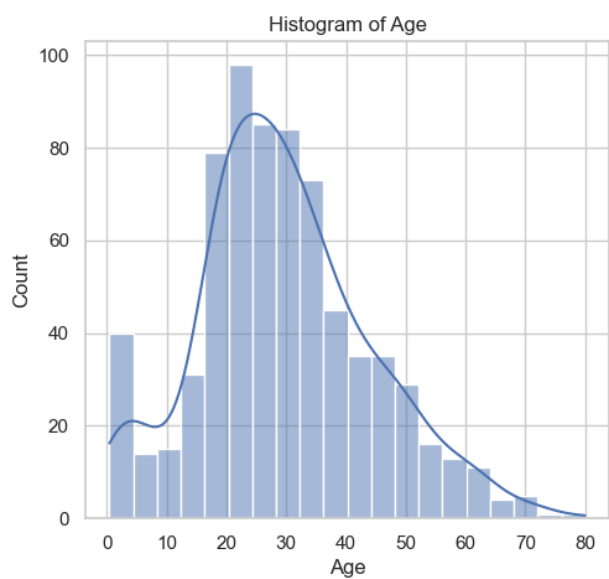
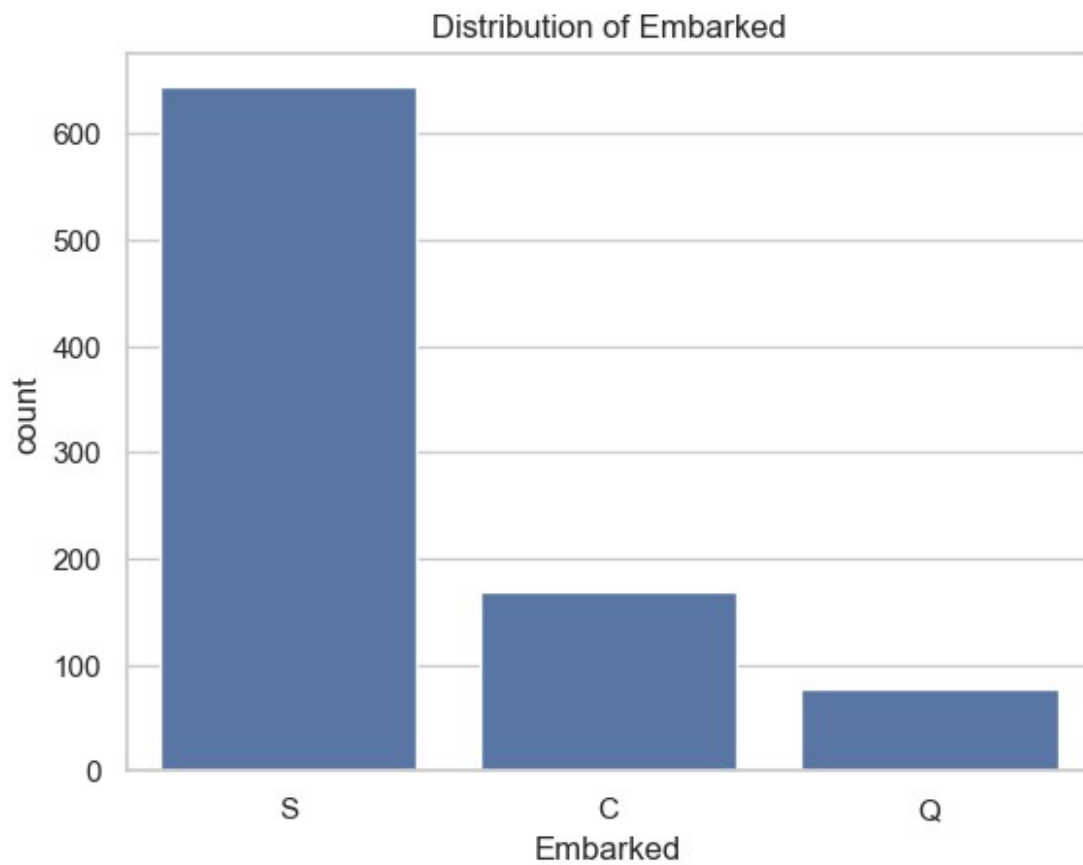
PassengerId	0
Survived	0
Pclass	0
Name	0
Sex	0
Age	177
SibSp	0
Parch	0
Ticket	0
Fare	0
Cabin	687
Embarked	2

dtype: int64



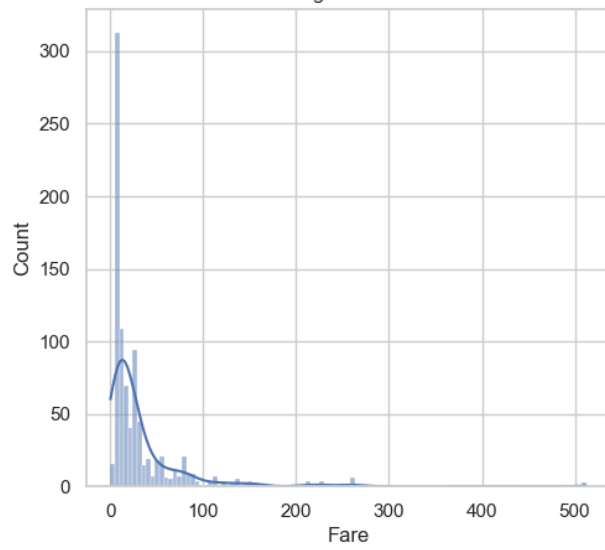




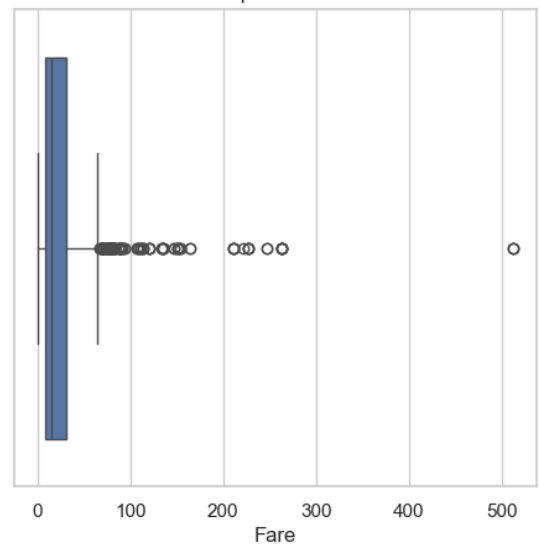




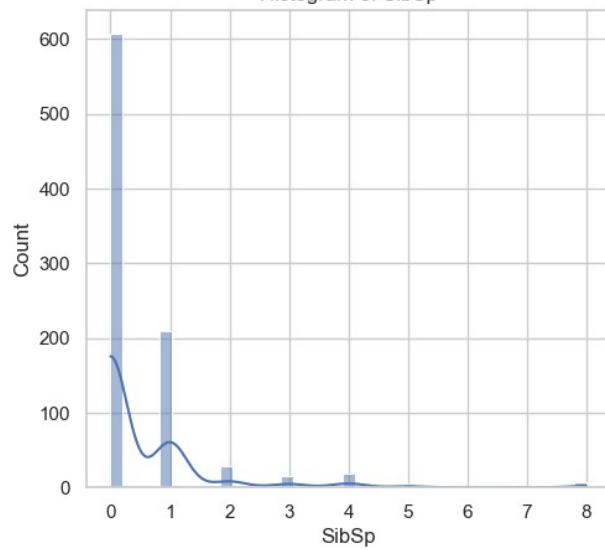
Histogram of Fare



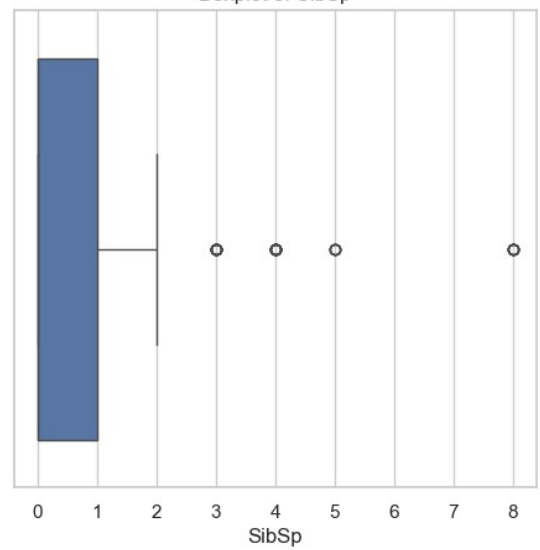
Boxplot of Fare

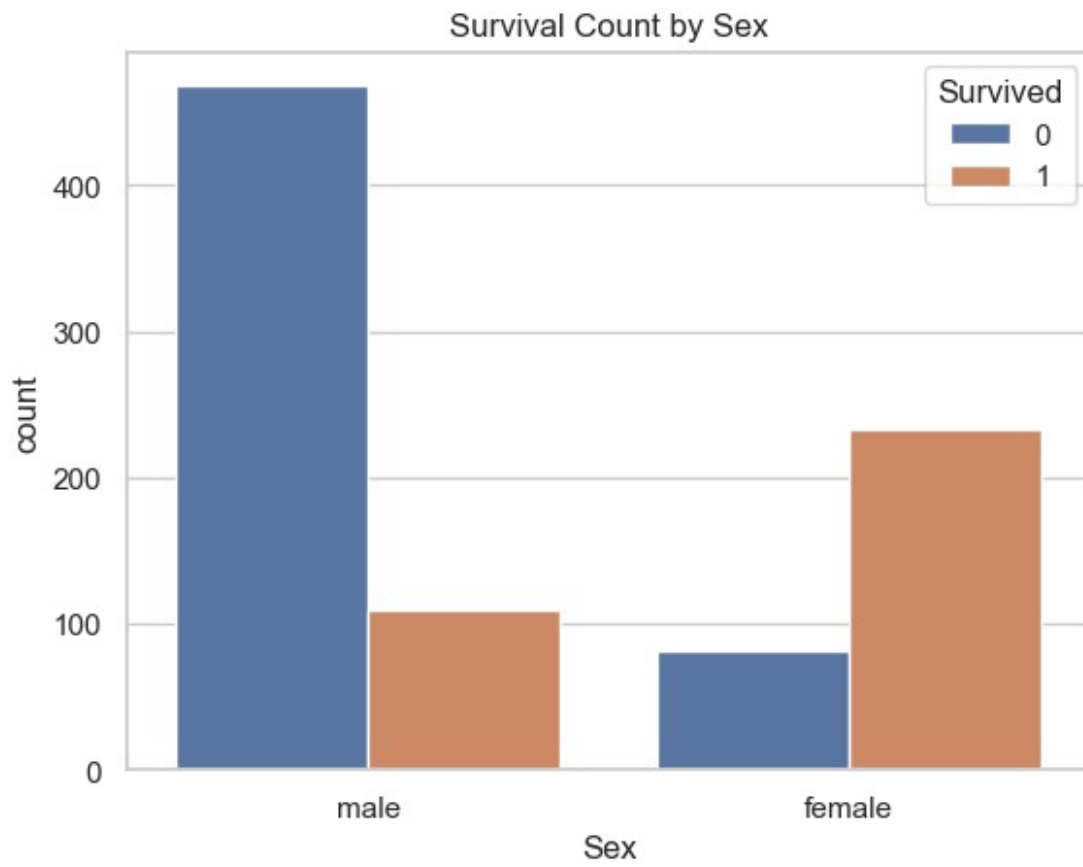
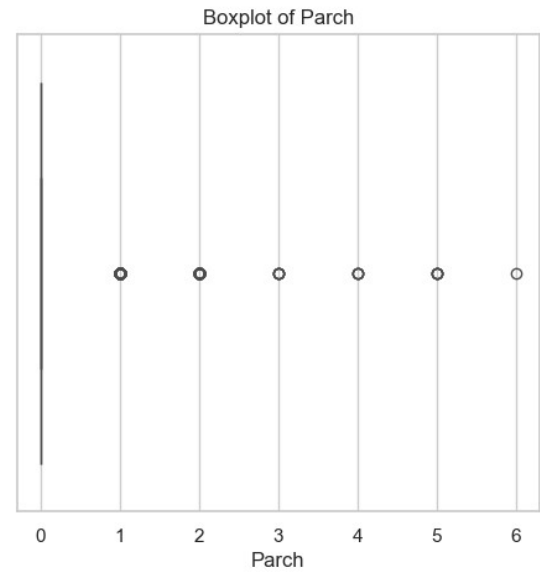
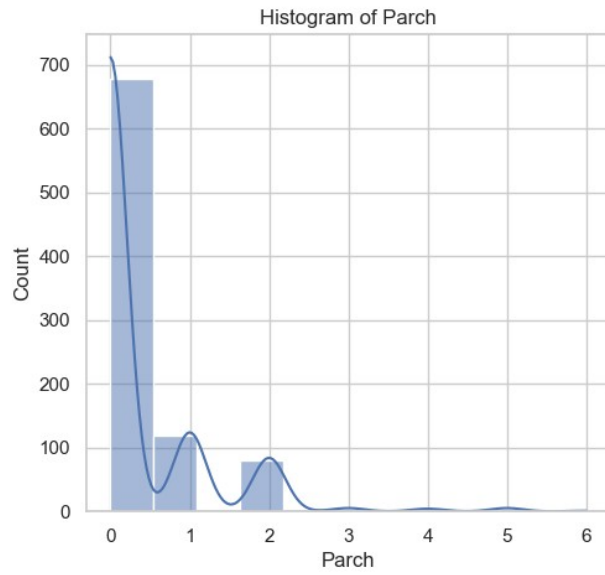


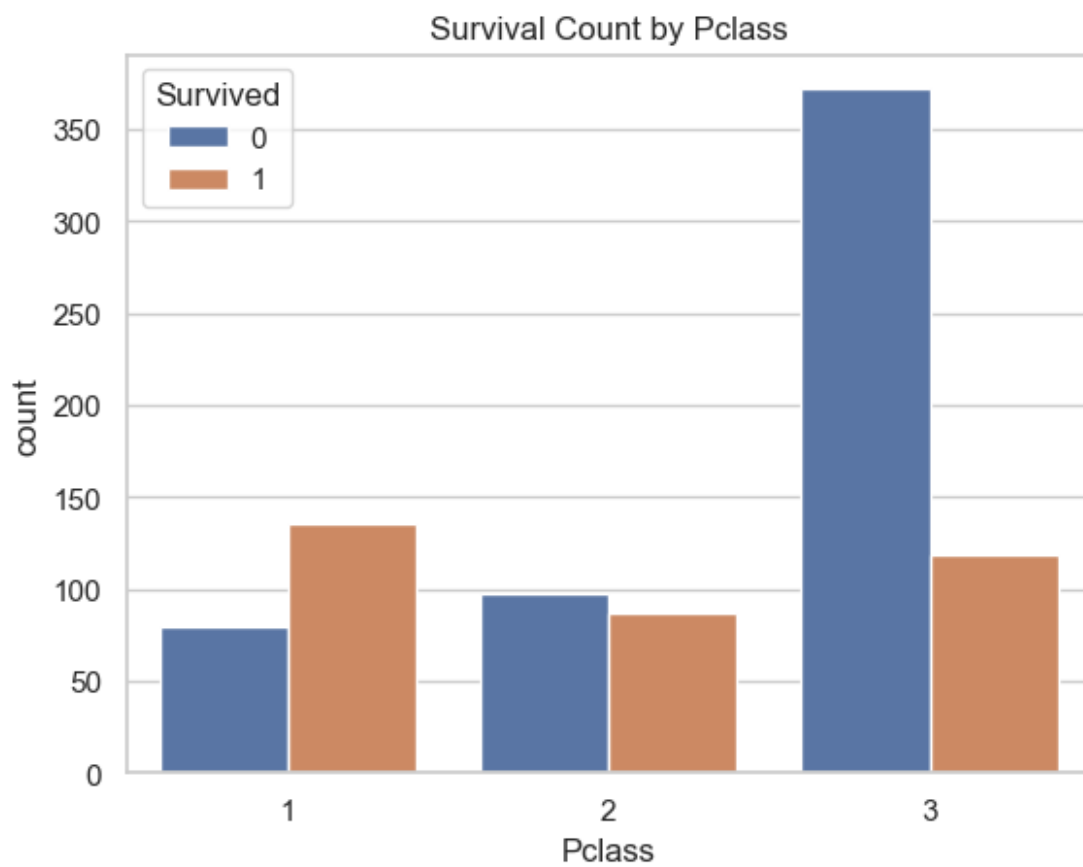
Histogram of SibSp

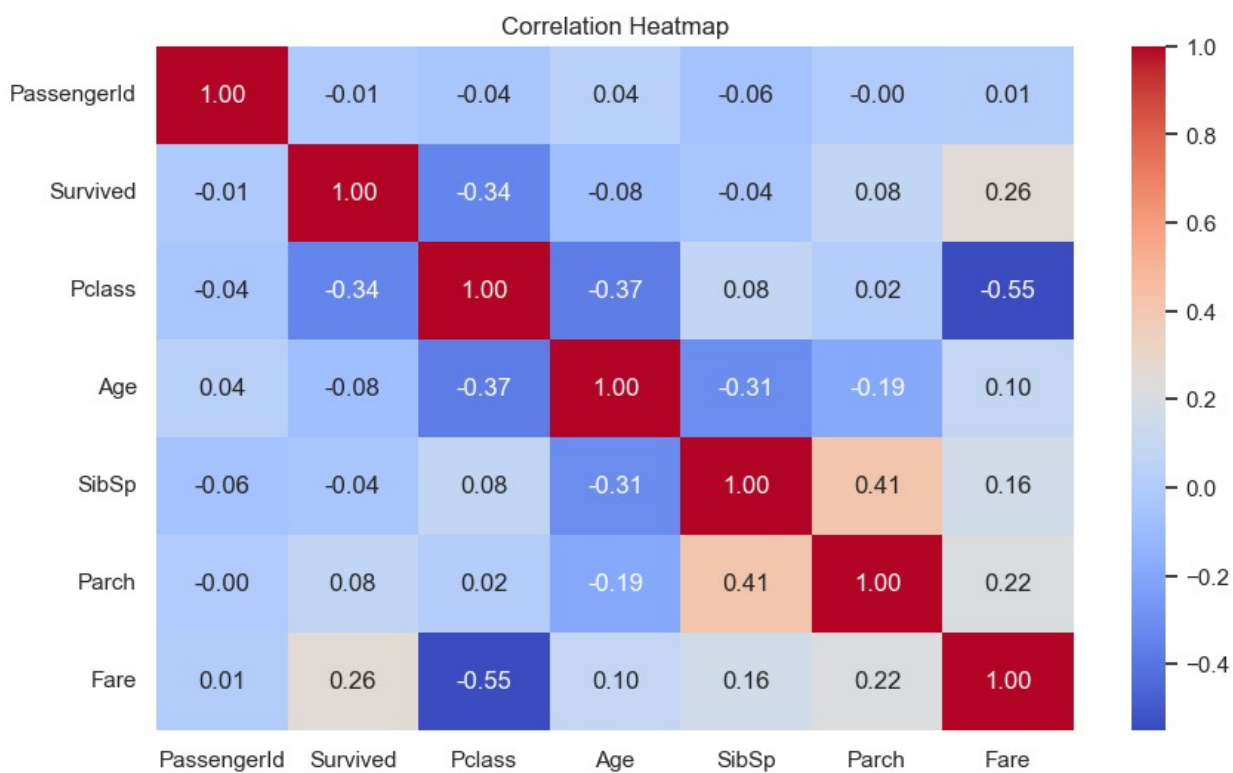
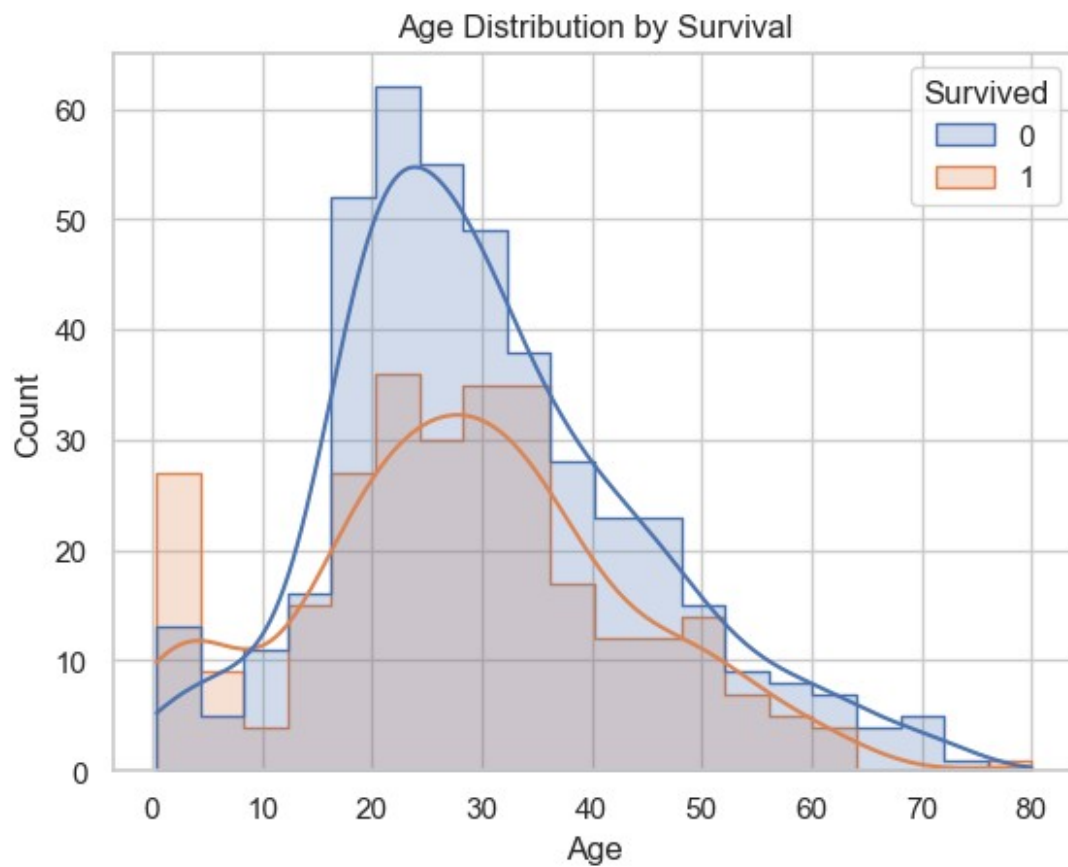


Boxplot of SibSp





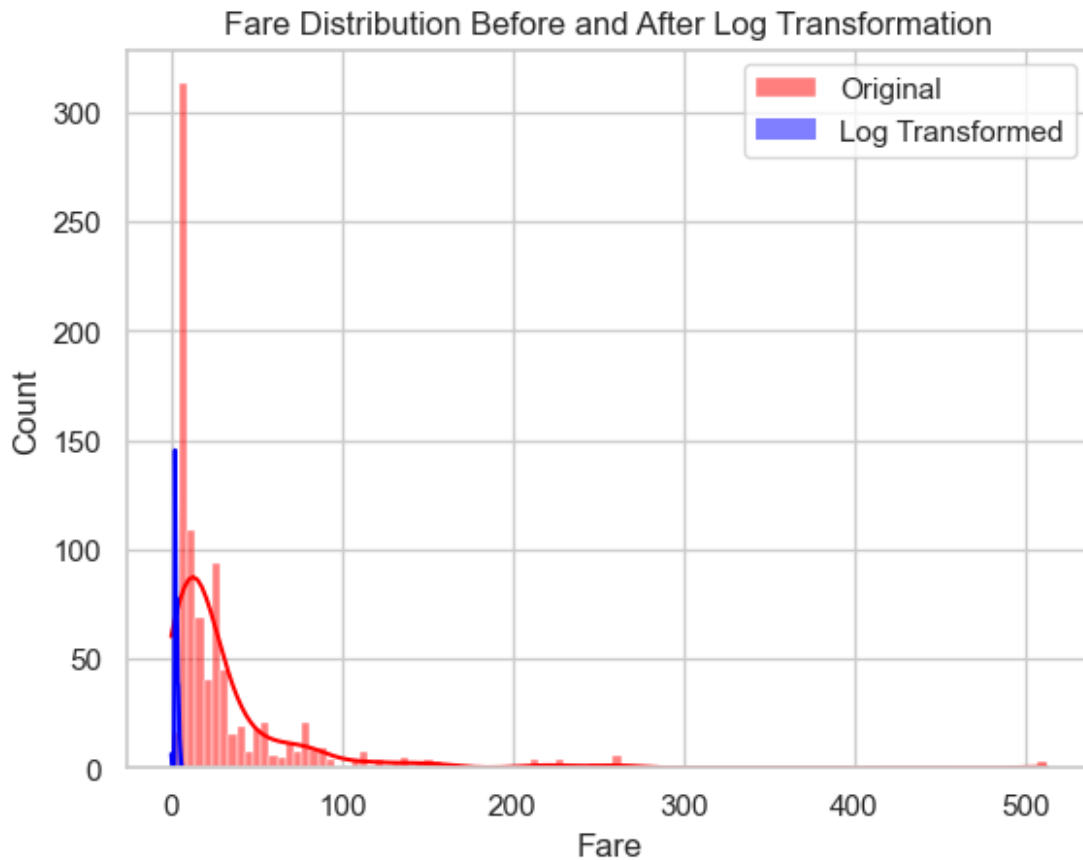




Skewness of numerical features:

Age: 0.39

Fare: 4.78



- Most passengers did not survive (~62%).
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