

In [4]: # TASK 5: Exploratory Data Analysis - Titanic Dataset

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

# Setup display settings
pd.set_option('display.max_columns', 100)
sns.set(style='whitegrid')
plt.rcParams['figure.figsize'] = (10,6)

# Step 2: Load the Dataset
# Make sure the Titanic dataset (train.csv) is in the same folder
df = pd.read_csv('train.csv')
print("✅ Dataset loaded successfully!")
print("Shape of dataset:", df.shape)
print(df.head())

# Step 3: Basic Information
print("\n--- Dataset Info ---")
print(df.info())

print("\n--- Missing Values ---")
print(df.isnull().sum())

print("\n--- Summary Statistics ---")
print(df.describe(include='all').T)

# Step 4: Data Cleaning (Safe Version)
# Clean column names (remove spaces and make consistent capitalization)
df.columns = df.columns.str.strip().str.capitalize()

# Fill missing Age
if 'Age' in df.columns:
    df['Age'] = df['Age'].fillna(df['Age'].median())

# Fill missing Embarked
if 'Embarked' in df.columns:
    df['Embarked'] = df['Embarked'].fillna(df['Embarked'].mode()[0])

# Drop unwanted columns if they exist
for col in ['Cabin', 'Ticket']:
    if col in df.columns:
        df.drop(columns=[col], inplace=True)

print("\n✅ Missing values after cleaning:")
print(df.isnull().sum())

# Step 5: Feature Engineering
if all(x in df.columns for x in ['Sibsp', 'Parch']):
    df['Familysize'] = df['Sibsp'] + df['Parch'] + 1
    df['Isalone'] = (df['Familysize'] == 1).astype(int)

if 'Name' in df.columns:
    df['Title'] = df['Name'].str.extract('([A-Za-z]+)\.')
    df['Title'] = df['Title'].replace(['Mlle', 'Ms'], 'Miss')
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df['Title'] = df['Title'].replace(['Mme'], 'Mrs')

# Step 6: Univariate Analysis
if 'Sex' in df.columns:
    sns.countplot(x='Sex', data=df)
    plt.title('Gender Distribution')
    plt.show()

if 'Survived' in df.columns:
    sns.countplot(x='Survived', data=df)
    plt.title('Survival Distribution')
    plt.show()

if 'Age' in df.columns:
    sns.histplot(df['Age'], bins=30, kde=True)
    plt.title('Age Distribution')
    plt.show()

if 'Fare' in df.columns:
    sns.histplot(df['Fare'], bins=30, kde=True, color='green')
    plt.title('Fare Distribution')
    plt.show()

# Step 7: Bivariate Analysis
if all(x in df.columns for x in ['Sex', 'Survived']):
    sns.countplot(x='Sex', hue='Survived', data=df)
    plt.title('Survival by Gender')
    plt.show()

if all(x in df.columns for x in ['Pclass', 'Survived']):
    sns.countplot(x='Pclass', hue='Survived', data=df)
    plt.title('Survival by Passenger Class')
    plt.show()

if all(x in df.columns for x in ['Survived', 'Age']):
    sns.boxplot(x='Survived', y='Age', data=df)
    plt.title('Age vs Survival')
    plt.show()

if all(x in df.columns for x in ['Survived', 'Fare']):
    sns.boxplot(x='Survived', y='Fare', data=df)
    plt.title('Fare vs Survival')
    plt.show()

# Step 8: Multivariate Analysis
num_df = df.select_dtypes(include=[np.number])
if not num_df.empty:
    corr = num_df.corr()
    sns.heatmap(corr, annot=True, cmap='coolwarm', fmt=".2f")
    plt.title('Correlation Matrix')
    plt.show()

# Step 9: Summary of Observations
print("\n✅ SUMMARY OF FINDINGS:")
print("""
1. More females survived than males.
2. Passengers in 1st class had higher survival rates.
3. Younger passengers had slightly better chances of survival.
4. Higher fares were linked to higher survival probability.
5. Passengers traveling alone had lower survival rates.

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""")
```

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# Step 10: Export Cleaned Data
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df.to_csv('titanic_cleaned.csv', index=False)
```

```
print("\n✅ Cleaned dataset saved as 'titanic_cleaned.csv'")
```

✔ Dataset loaded successfully!

Shape of dataset: (1310, 14)

	pclass	survived		name	sex	\
0	1.0	1.0		Allen, Miss. Elisabeth Walton	female	
1	1.0	1.0		Allison, Master. Hudson Trevor	male	
2	1.0	0.0		Allison, Miss. Helen Loraine	female	
3	1.0	0.0		Allison, Mr. Hudson Joshua Creighton	male	
4	1.0	0.0	Allison, Mrs. Hudson J C (Bessie Waldo Daniels)		female	

	age	sibsp	parch	ticket	fare	cabin	embarked	boat	body	\
0	29.0000	0.0	0.0	24160	211.3375	B5	S	2	NaN	
1	0.9167	1.0	2.0	113781	151.5500	C22 C26	S	11	NaN	
2	2.0000	1.0	2.0	113781	151.5500	C22 C26	S	NaN	NaN	
3	30.0000	1.0	2.0	113781	151.5500	C22 C26	S	NaN	135.0	
4	25.0000	1.0	2.0	113781	151.5500	C22 C26	S	NaN	NaN	

	home.dest
0	St Louis, MO
1	Montreal, PQ / Chesterville, ON
2	Montreal, PQ / Chesterville, ON
3	Montreal, PQ / Chesterville, ON
4	Montreal, PQ / Chesterville, ON

--- Dataset Info ---

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

RangeIndex: 1310 entries, 0 to 1309

Data columns (total 14 columns):

#	Column	Non-Null Count	Dtype
0	pclass	1309 non-null	float64
1	survived	1309 non-null	float64
2	name	1309 non-null	object
3	sex	1309 non-null	object
4	age	1046 non-null	float64
5	sibsp	1309 non-null	float64
6	parch	1309 non-null	float64
7	ticket	1309 non-null	object
8	fare	1308 non-null	float64
9	cabin	295 non-null	object
10	embarked	1307 non-null	object
11	boat	486 non-null	object
12	body	121 non-null	float64
13	home.dest	745 non-null	object

dtypes: float64(7), object(7)

memory usage: 143.4+ KB

None

--- Missing Values ---

pclass	1
survived	1
name	1
sex	1
age	264
sibsp	1
parch	1
ticket	1
fare	2
cabin	1015
embarked	3
boat	824

```
body          1189
home.dest     565
dtype: int64
```

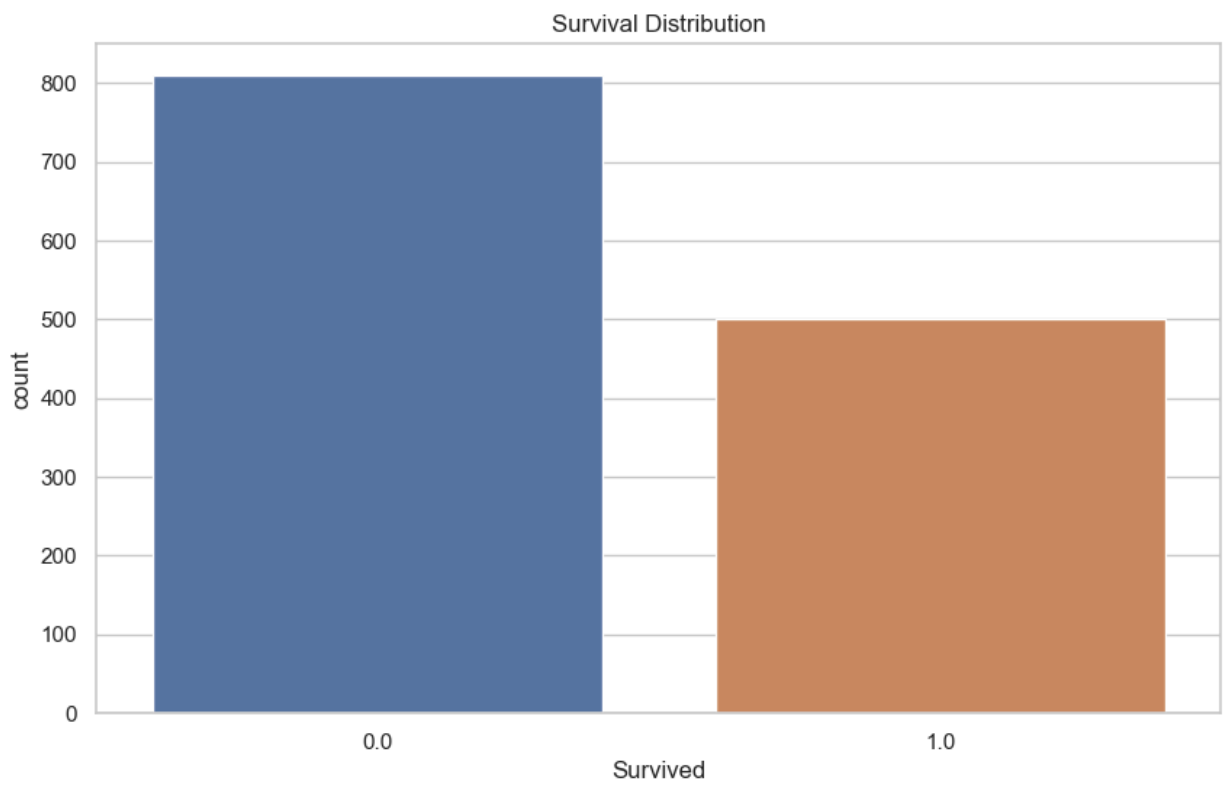
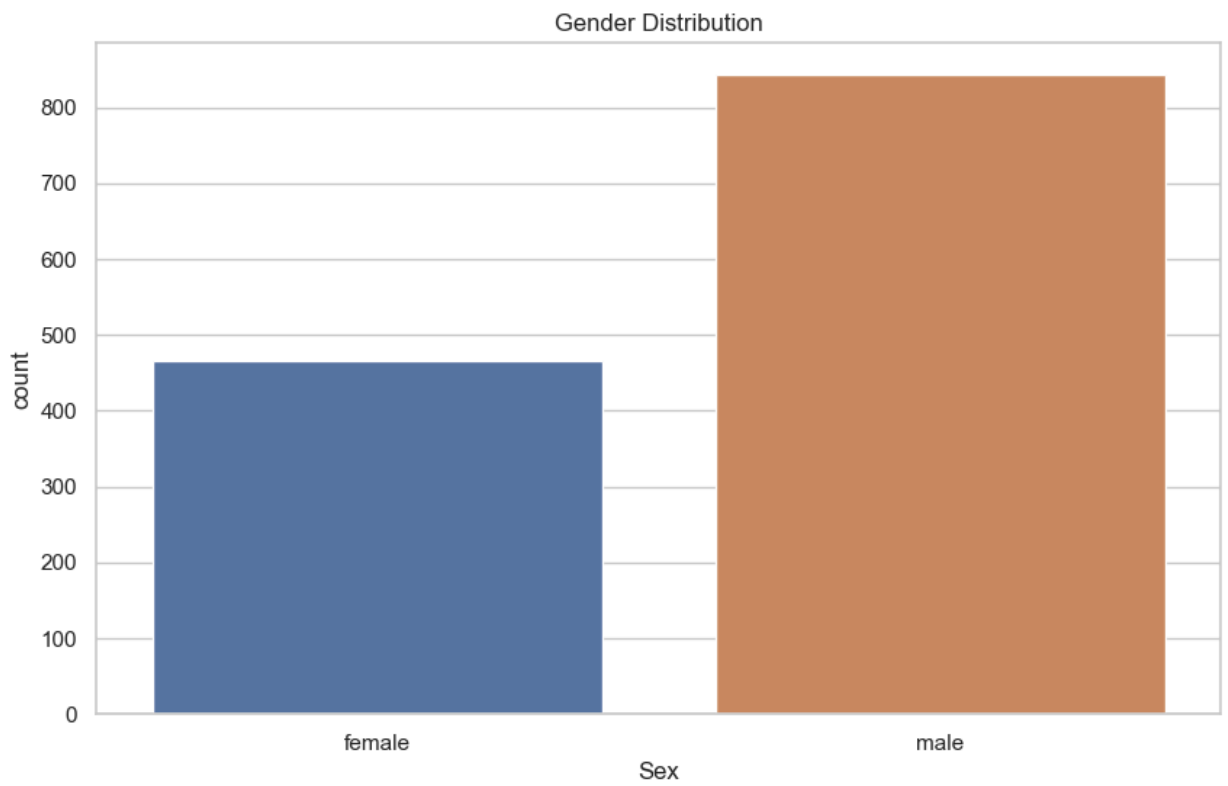
--- Summary Statistics ---

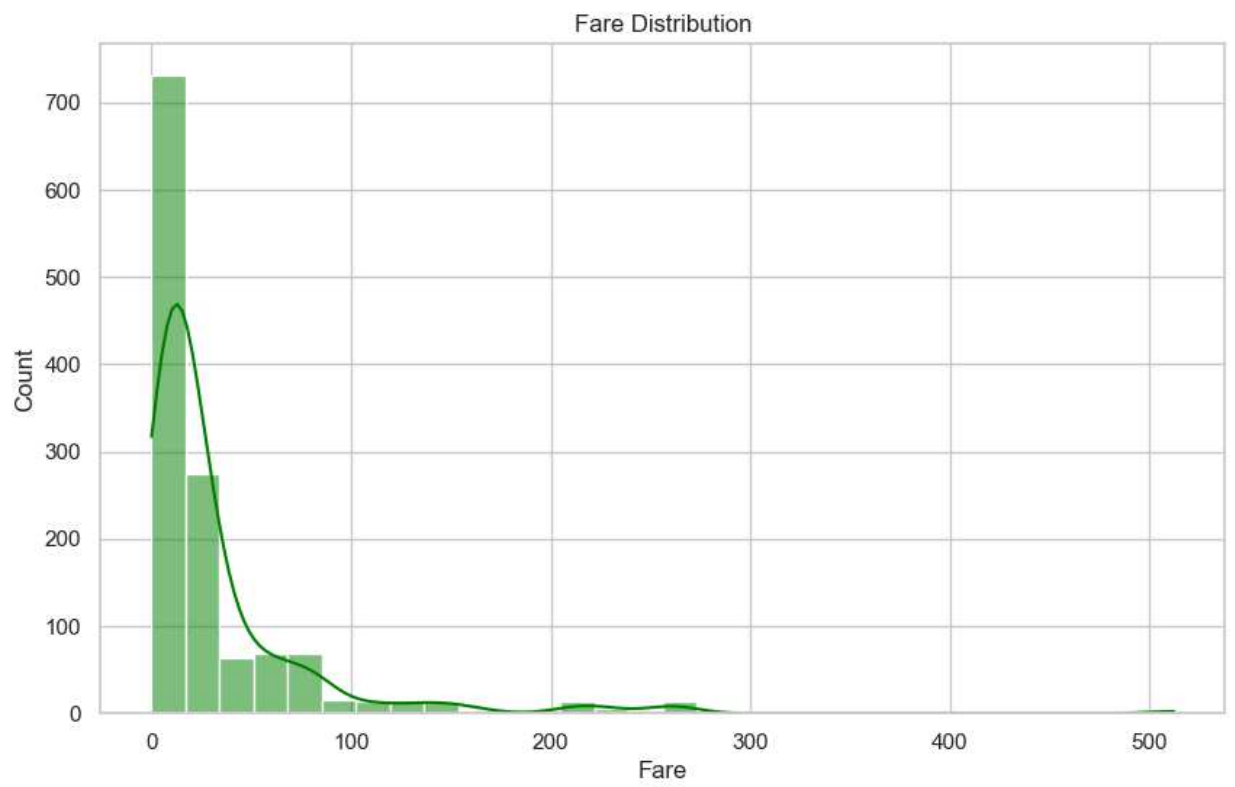
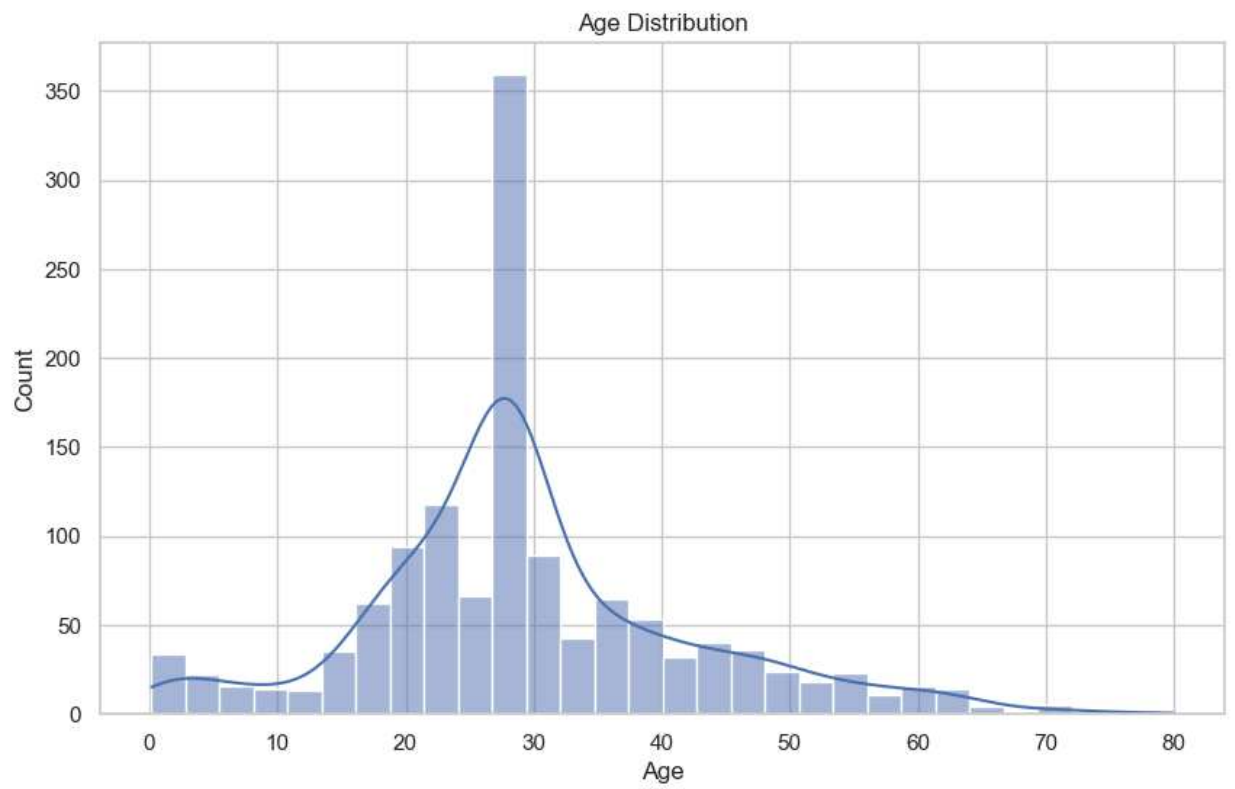
	count	unique	top	freq	mean	std	\
pclass	1309.0	NaN	NaN	NaN	2.294882	0.837836	
survived	1309.0	NaN	NaN	NaN	0.381971	0.486055	
name	1309	1307	Connolly, Miss. Kate	2	NaN	NaN	
sex	1309	2	male	843	NaN	NaN	
age	1046.0	NaN	NaN	NaN	29.881135	14.4135	
sibsp	1309.0	NaN	NaN	NaN	0.498854	1.041658	
parch	1309.0	NaN	NaN	NaN	0.385027	0.86556	
ticket	1309	929	CA. 2343	11	NaN	NaN	
fare	1308.0	NaN	NaN	NaN	33.295479	51.758668	
cabin	295	186	C23 C25 C27	6	NaN	NaN	
embarked	1307	3	S	914	NaN	NaN	
boat	486	27	13	39	NaN	NaN	
body	121.0	NaN	NaN	NaN	160.809917	97.696922	
home.dest	745	369	New York, NY	64	NaN	NaN	

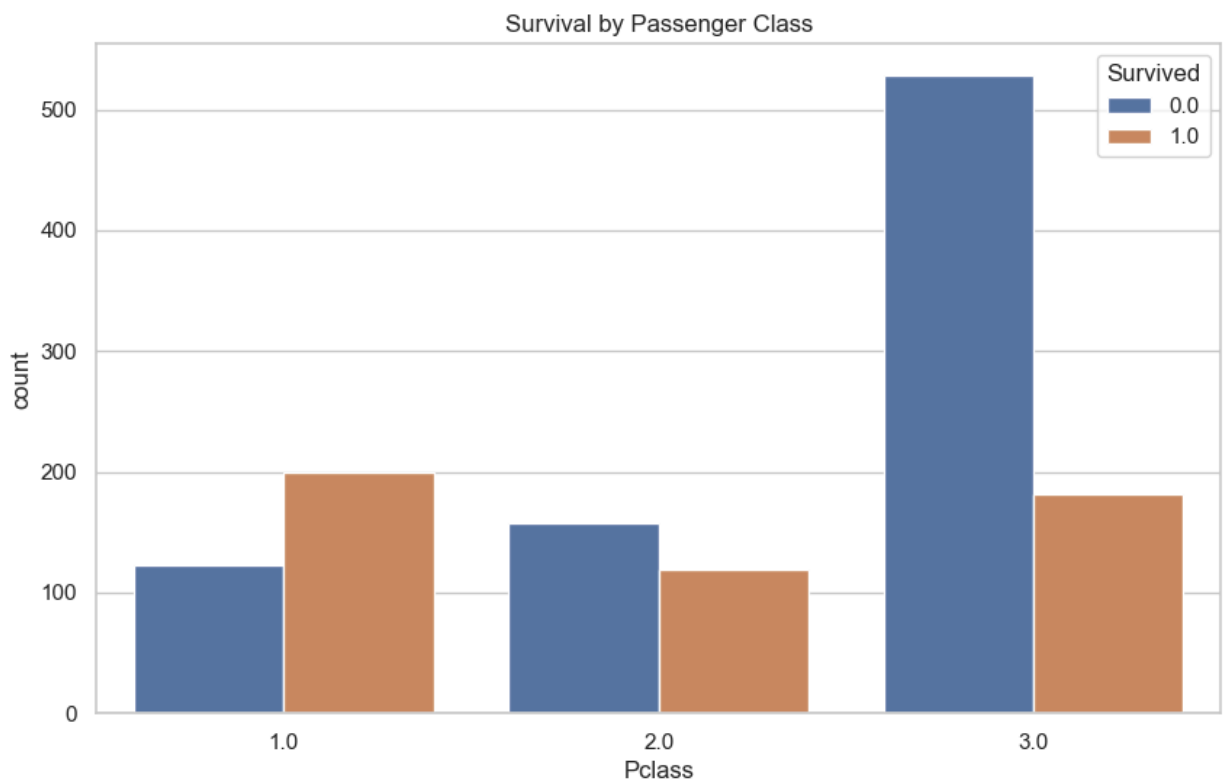
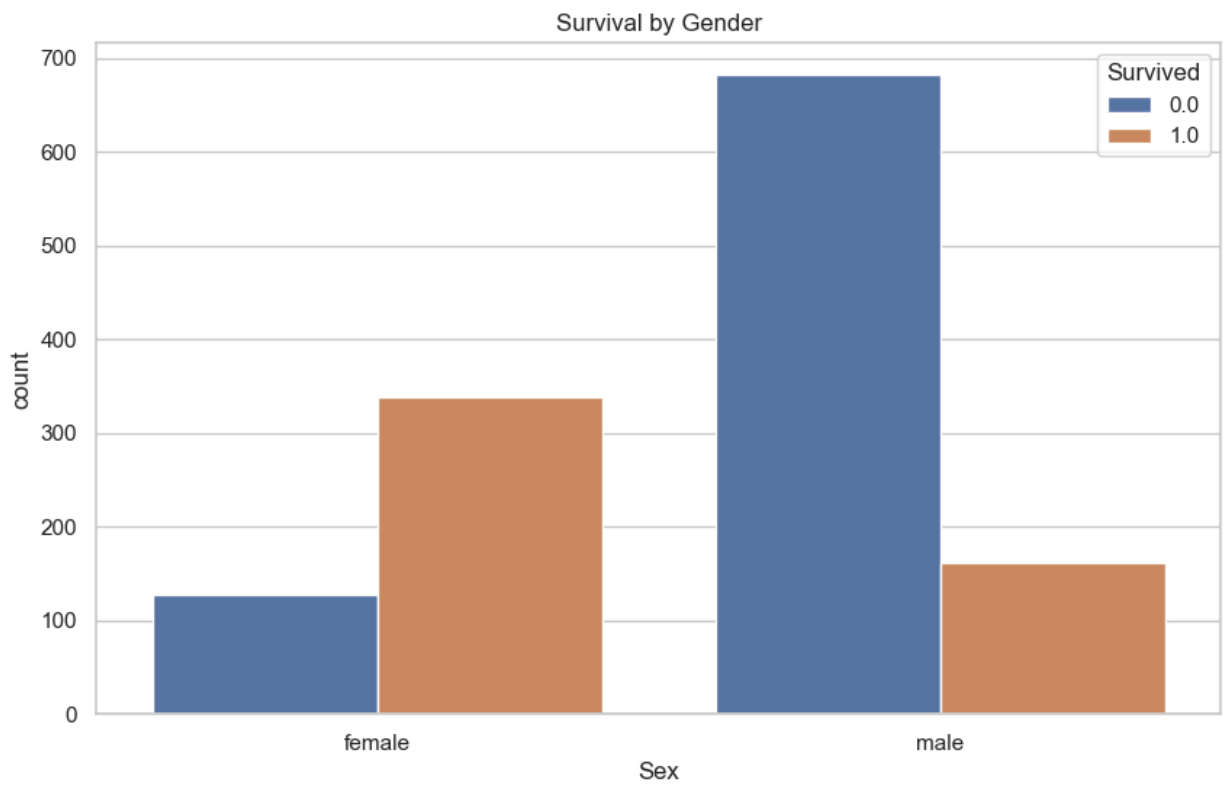
	min	25%	50%	75%	max
pclass	1.0	2.0	3.0	3.0	3.0
survived	0.0	0.0	0.0	1.0	1.0
name	NaN	NaN	NaN	NaN	NaN
sex	NaN	NaN	NaN	NaN	NaN
age	0.1667	21.0	28.0	39.0	80.0
sibsp	0.0	0.0	0.0	1.0	8.0
parch	0.0	0.0	0.0	0.0	9.0
ticket	NaN	NaN	NaN	NaN	NaN
fare	0.0	7.8958	14.4542	31.275	512.3292
cabin	NaN	NaN	NaN	NaN	NaN
embarked	NaN	NaN	NaN	NaN	NaN
boat	NaN	NaN	NaN	NaN	NaN
body	1.0	72.0	155.0	256.0	328.0
home.dest	NaN	NaN	NaN	NaN	NaN

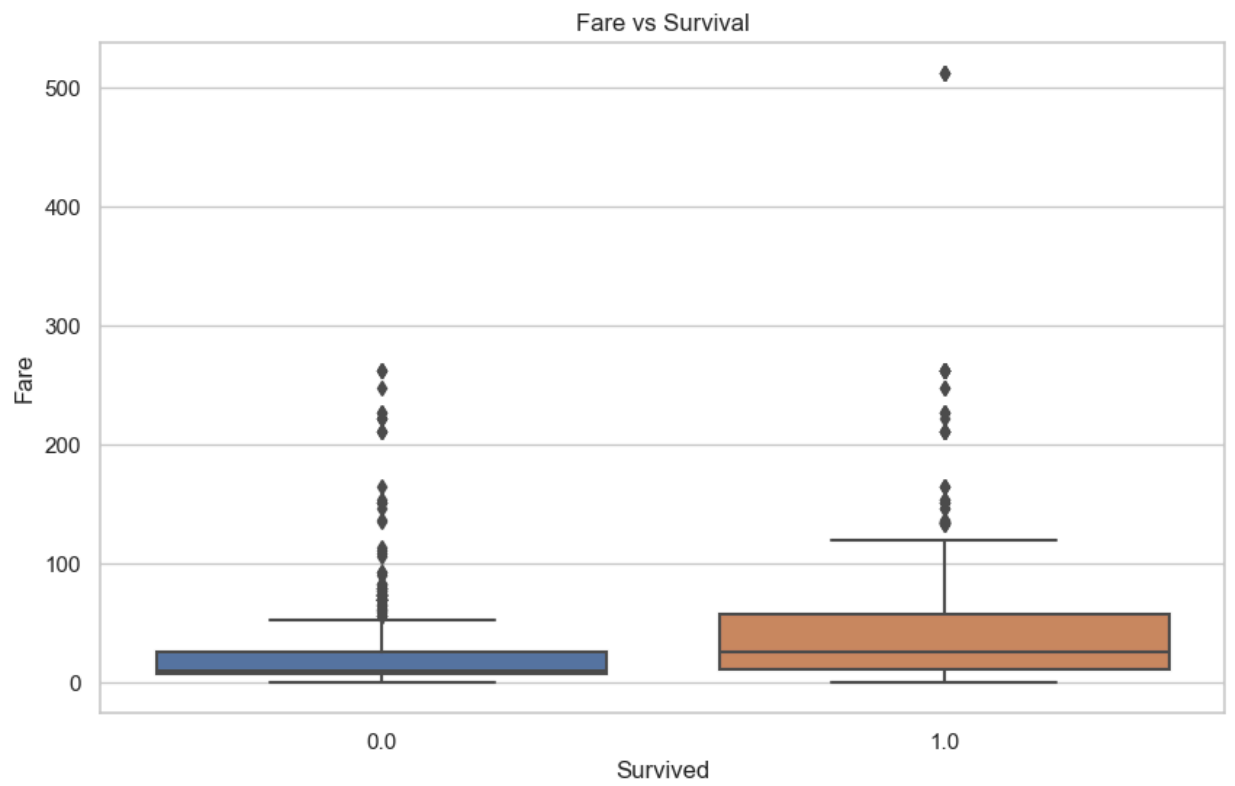
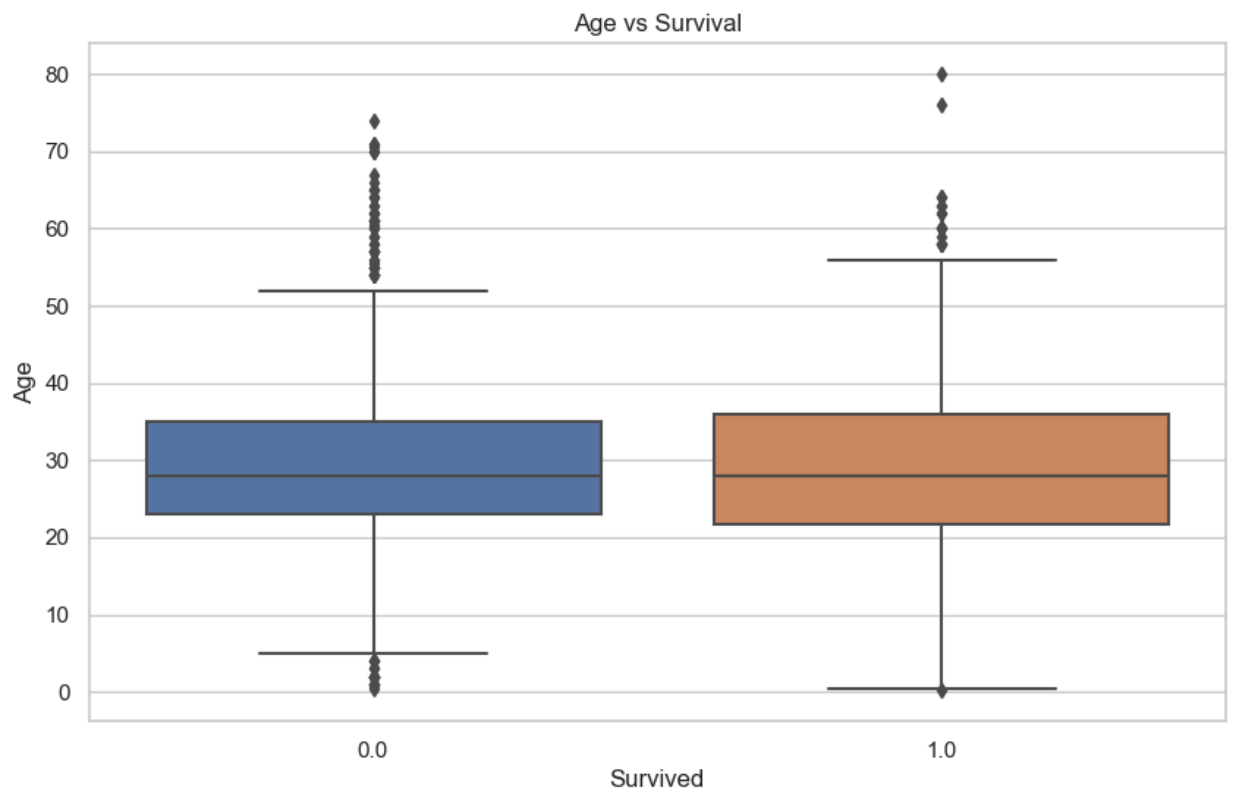
✅ Missing values after cleaning:

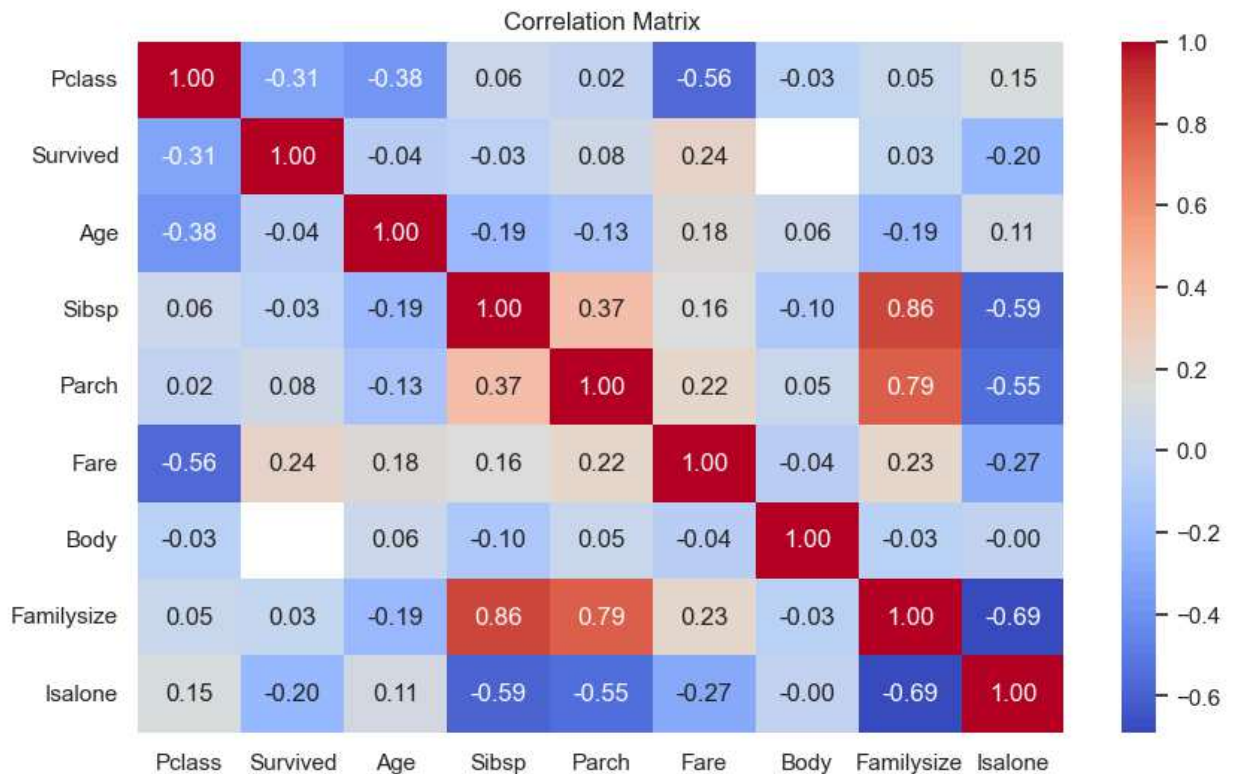
```
Pclass      1
Survived    1
Name        1
Sex         1
Age         0
Sibsp       1
Parch       1
Fare        2
Embarked     0
Boat        824
Body        1189
Home.dest    565
dtype: int64
```











✓ SUMMARY OF FINDINGS:

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✓ Cleaned dataset saved as 'titanic_cleaned.csv'

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