

# Practical - 1

Name - Sarthak Abhaykumar Nahar

Batch - B4

Roll no. - 55

```
In [2]: import pandas as pd
```

```
df = pd.read_csv('titanic.csv')
```

```
In [3]: # Get the first five data
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

```
In [4]: # Get bottom 5 record
print(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

```
In [5]: #Get the Statistical Information about the dataset
print(df.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

```
In [7]: #Get the number of rows and columns
```

```
print(df.shape)
```

```
(891, 12)
```

```
In [10]: #Check for missing values
```

```
print(df.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
```

```
In [13]: df['Age'].fillna(df['Age'].median(), inplace=True)
df['Embarked'].fillna(df['Embarked'].mode()[0], inplace=True)
df.drop(columns=['Cabin'],inplace=True)
```

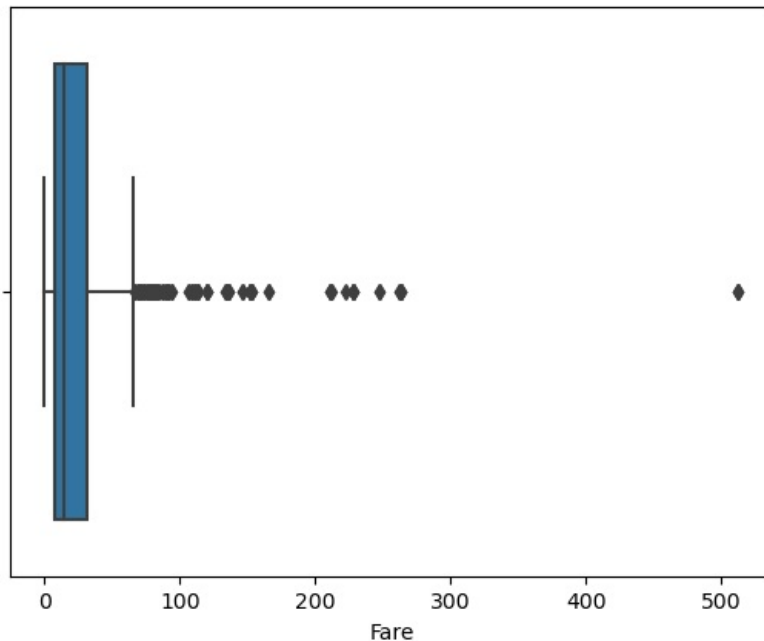
```
In [19]: duplicates = df.duplicated().sum()
print(f"Number of duplicates rows: {duplicates}")
df.drop_duplicates(inplace=True)
```

```
Number of duplicates rows: 0
```

```
In [20]: import matplotlib.pyplot as plt
```

```
In [21]: import seaborn as sns
```

```
In [22]: sns.boxplot(x=df['Fare'])
plt.show()
```

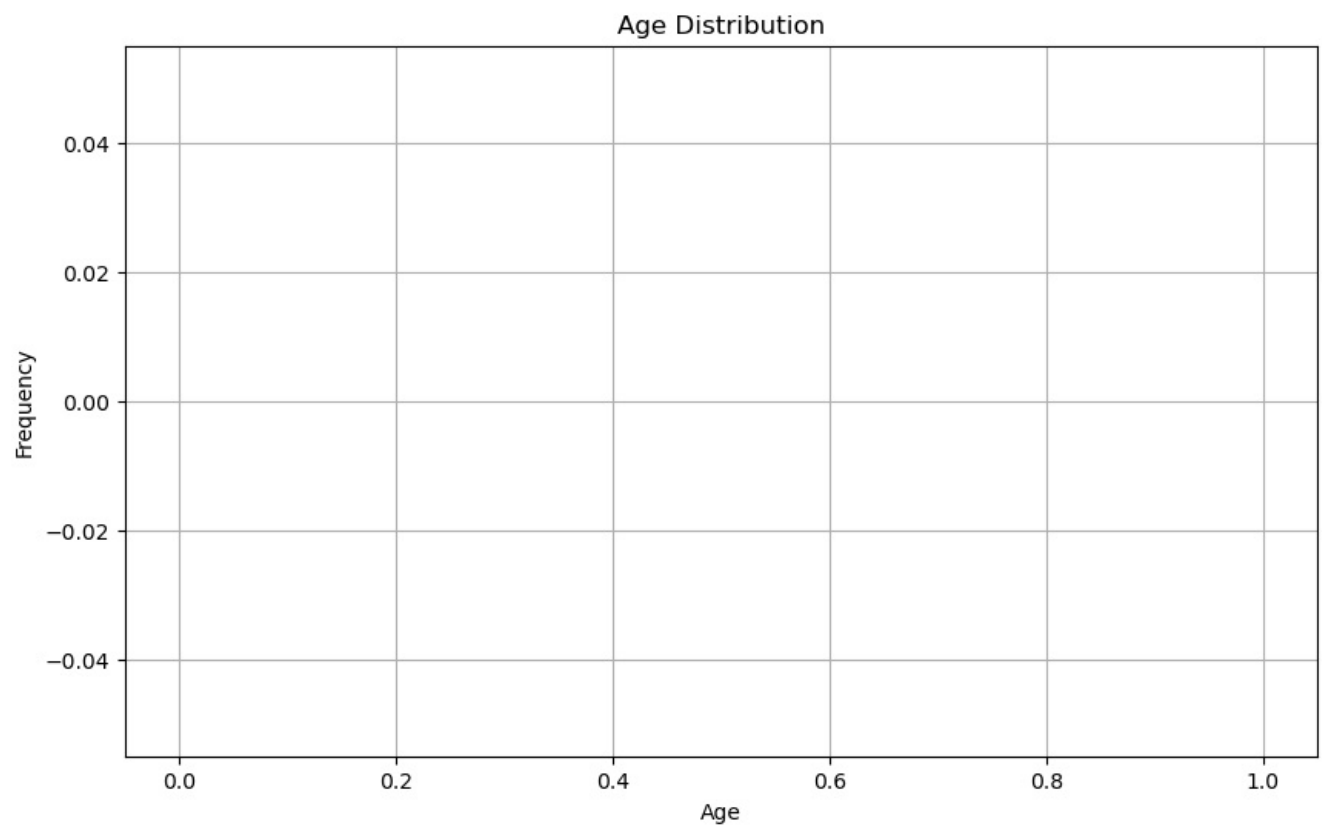


```
In [27]: Q1 = df['Fare'].quantile(0.25)
Q3 = df['Fare'].quantile(0.75)
IQR = Q3-Q1
df = df[~((df['Fare'] < (Q1 - 1.5*IQR)) | (df['Fare'] > (Q3 - 1.5*IQR)))]
```

```
In [28]: # Apply one-hot encoding to the 'Sex' column
df = pd.get_dummies(df, columns=['Sex', 'Embarked'], drop_first=True)
```

```
In [36]: import matplotlib.pyplot as plt
```

```
# Univariate analysis for Age
plt.figure(figsize=(10, 6))
df['Age'].hist(bins=30)
plt.title('Age Distribution')
plt.xlabel('Age')
plt.ylabel('Frequency')
plt.show()
```



```
In [38]: print(df.head())
print(df[['Pclass', 'Survived']].isnull().sum())
```

Empty DataFrame  
Columns: [PassengerId, Survived, Pclass, Name, Age, SibSp, Parch, Ticket, Fare]  
Index: []  
Pclass 0.0  
Survived 0.0  
dtype: float64

```
In [40]: import pandas as pd

# Load the dataset
df = pd.read_csv('titanic.csv')

# Check the initial data
print(df.head())
print(df.columns)
```

	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

Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp',  
'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'],  
dtype='object')

```
In [41]: # tells the no of columns
print(df.columns)
```

Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp',  
'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'],  
dtype='object')

```
In [42]: # Example: Check for dropped columns or filtering conditions
print(df.isnull().sum())

# Check any transformations applied to 'Pclass' or 'Survived'
```

```
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
```

```
In [43]: # Reload the dataset
df = pd.read_csv('titanic.csv')

# Reapply preprocessing, ensuring not to drop or alter essential columns
df['Age'].fillna(df['Age'].median(), inplace=True)
df['Embarked'].fillna(df['Embarked'].mode()[0], inplace=True)
df.drop(columns=['Cabin'], inplace=True)

# Confirm the presence of data
print(df['Pclass'].unique())
print(df['Survived'].unique())

[3 1 2]
[0 1]
```

```
In [44]: # to check the numeric value
print(df.dtypes)
```

```
PassengerId      int64
Survived          int64
Pclass           int64
Name             object
Sex              object
Age             float64
SibSp            int64
Parch            int64
Ticket           object
Fare             float64
Embarked         object
dtype: object
```

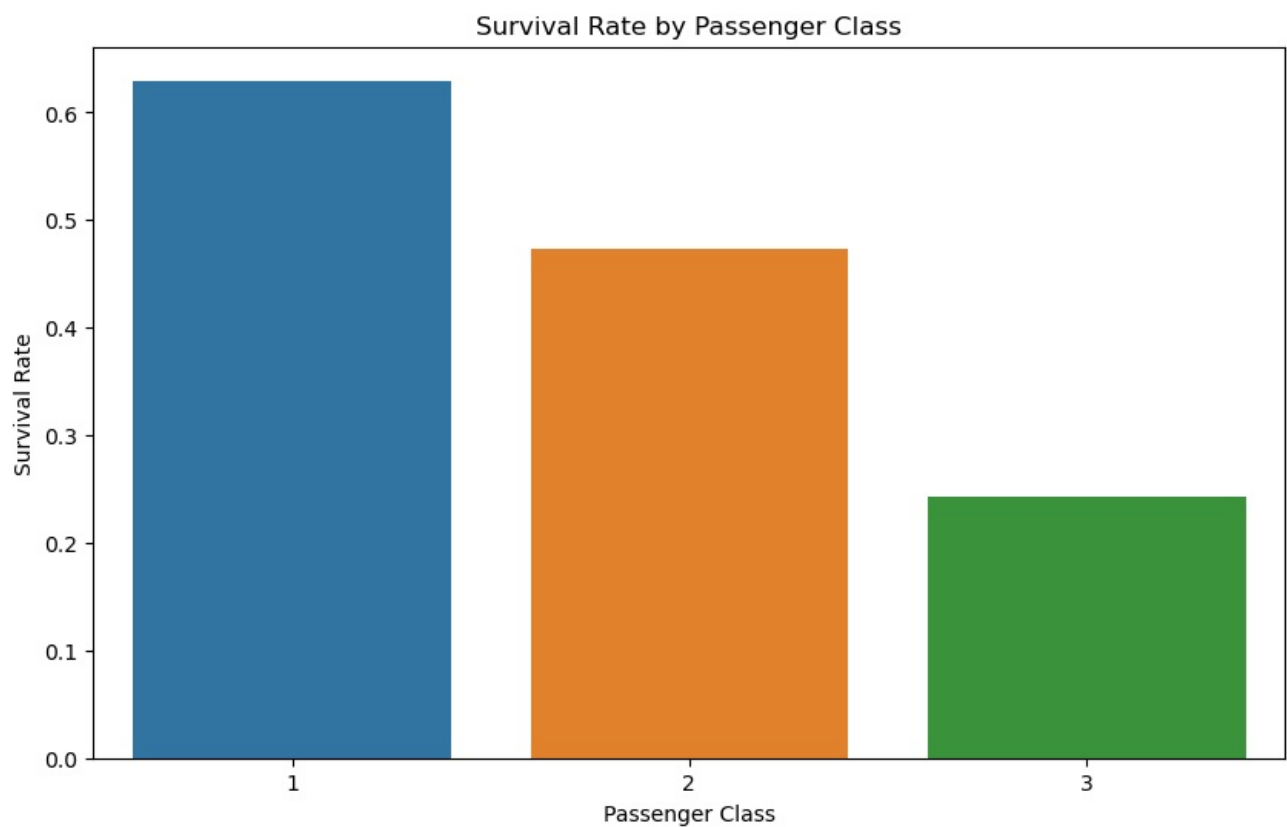
```
In [45]: import seaborn as sns
import matplotlib.pyplot as plt

# Ensure the DataFrame is not empty and the columns exist
if not df.empty and 'Pclass' in df.columns and 'Survived' in df.columns:
    plt.figure(figsize=(10, 6))
    sns.barplot(x='Pclass', y='Survived', data=df, ci=None)
    plt.title('Survival Rate by Passenger Class')
    plt.xlabel('Passenger Class')
    plt.ylabel('Survival Rate')
    plt.show()
else:
    print("Data is not available for plotting.")
```

C:\Users\cse\AppData\Local\Temp\ipykernel\_13196\257418103.py:7: FutureWarning:

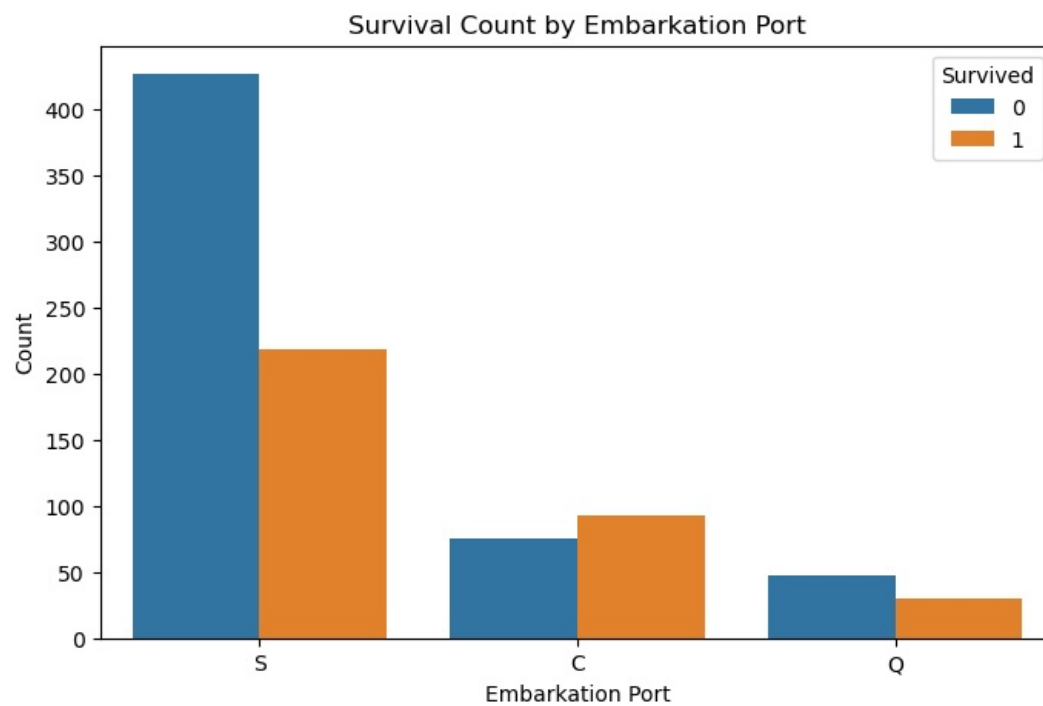
The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

```
sns.barplot(x='Pclass', y='Survived', data=df, ci=None)
```

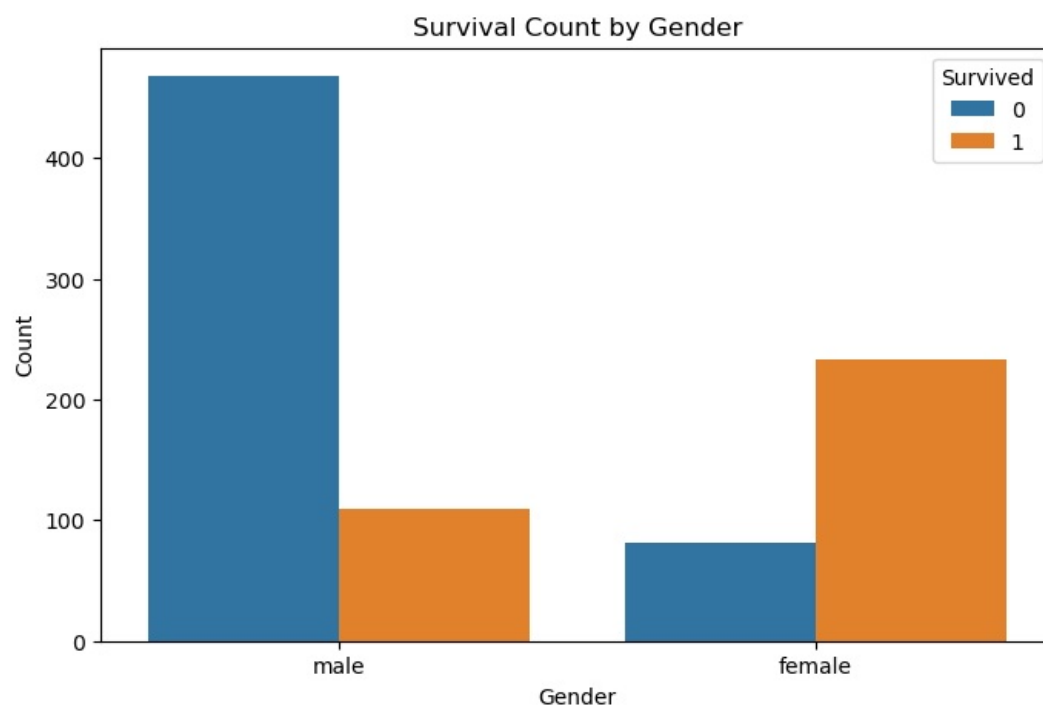


```
In [47]: import seaborn as sns
import matplotlib.pyplot as plt

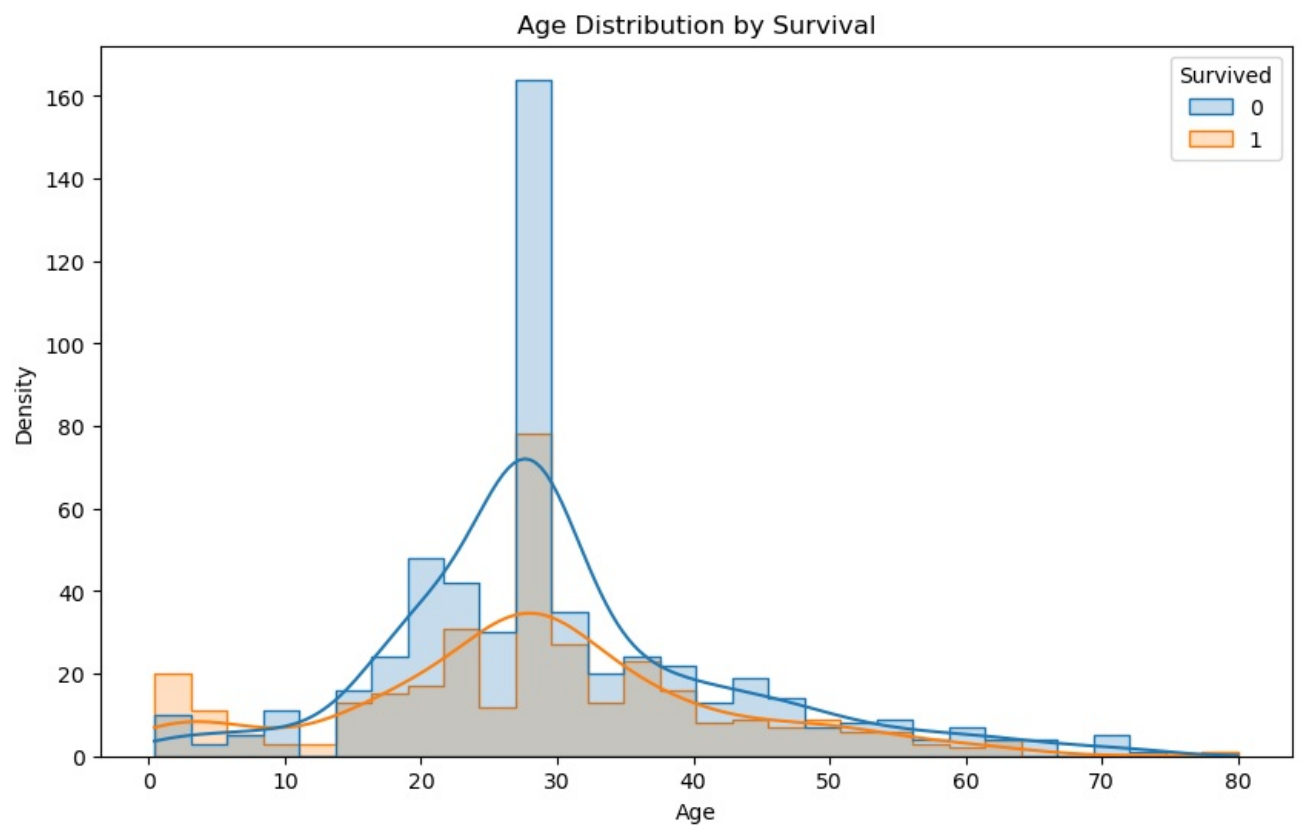
# Bivariate analysis: Survived vs. Embarked
plt.figure(figsize=(8, 5))
sns.countplot(x='Embarked', hue='Survived', data=df)
plt.title('Survival Count by Embarkation Port')
plt.xlabel('Embarkation Port')
plt.ylabel('Count')
plt.legend(title='Survived', loc='upper right')
plt.show()
```



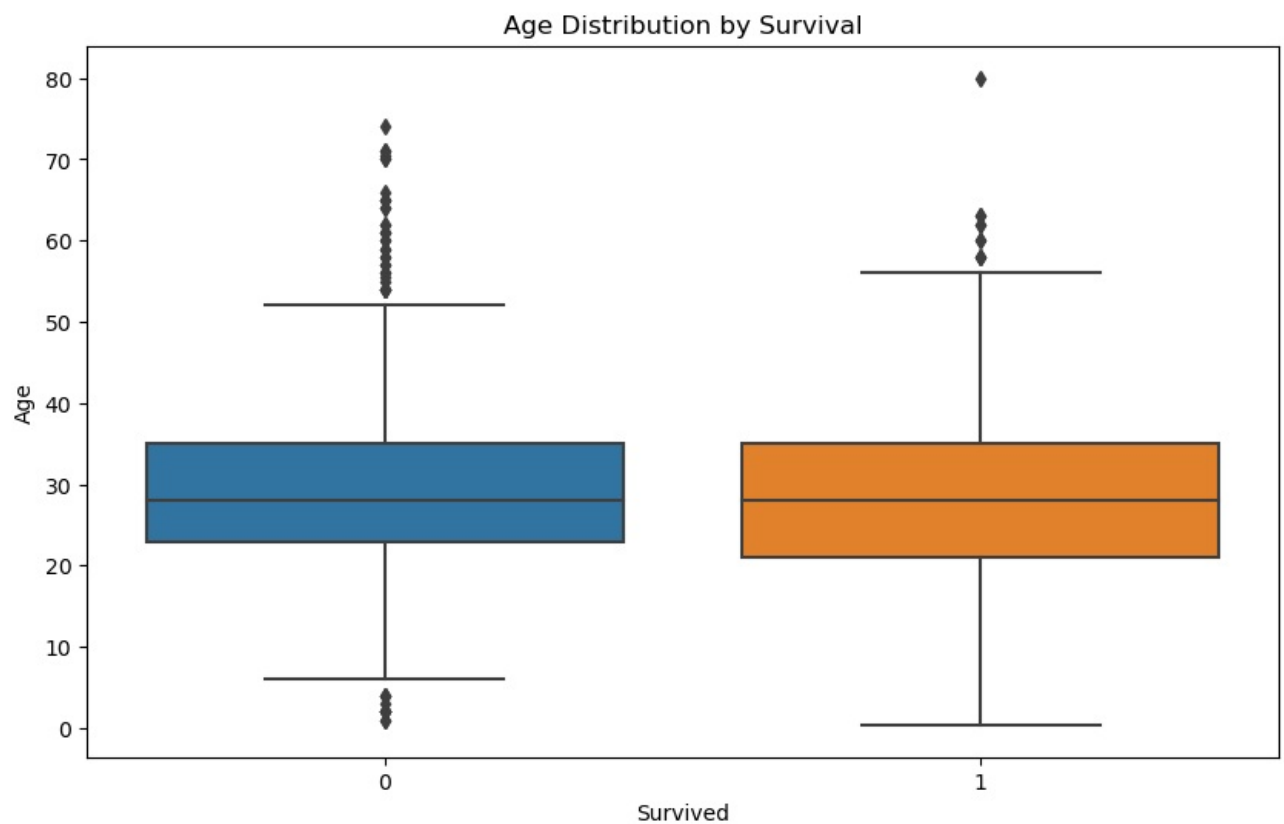
```
In [48]: # Bivariate analysis: Survived vs. Sex
plt.figure(figsize=(8, 5))
sns.countplot(x='Sex', hue='Survived', data=df)
plt.title('Survival Count by Gender')
plt.xlabel('Gender')
plt.ylabel('Count')
plt.legend(title='Survived', loc='upper right')
plt.show()
```



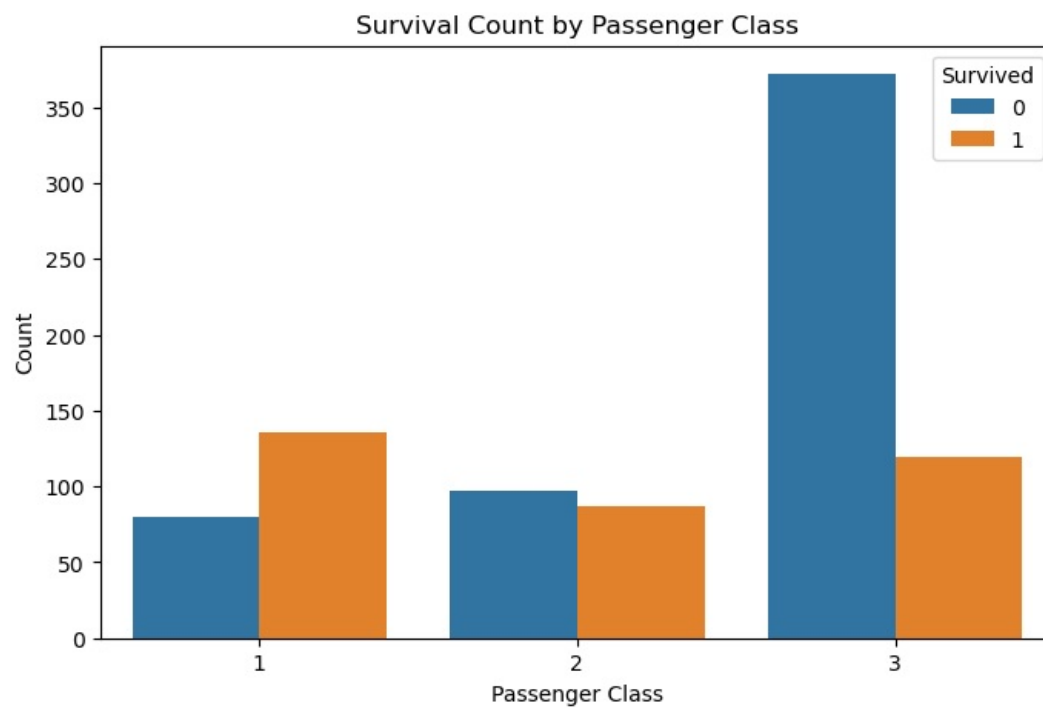
```
In [49]: # Bivariate analysis: Survived vs. Age
plt.figure(figsize=(10, 6))
sns.histplot(data=df, x='Age', hue='Survived', kde=True, bins=30, element='step')
plt.title('Age Distribution by Survival')
plt.xlabel('Age')
plt.ylabel('Density')
plt.show()
```



```
In [50]: # Boxplot: Age distribution by Survival
plt.figure(figsize=(10, 6))
sns.boxplot(x='Survived', y='Age', data=df)
plt.title('Age Distribution by Survival')
plt.xlabel('Survived')
plt.ylabel('Age')
plt.show()
```



```
In [51]: # Bivariate analysis: Survived vs. Pclass
plt.figure(figsize=(8, 5))
sns.countplot(x='Pclass', hue='Survived', data=df)
plt.title('Survival Count by Passenger Class')
plt.xlabel('Passenger Class')
plt.ylabel('Count')
plt.legend(title='Survived', loc='upper right')
plt.show()
```



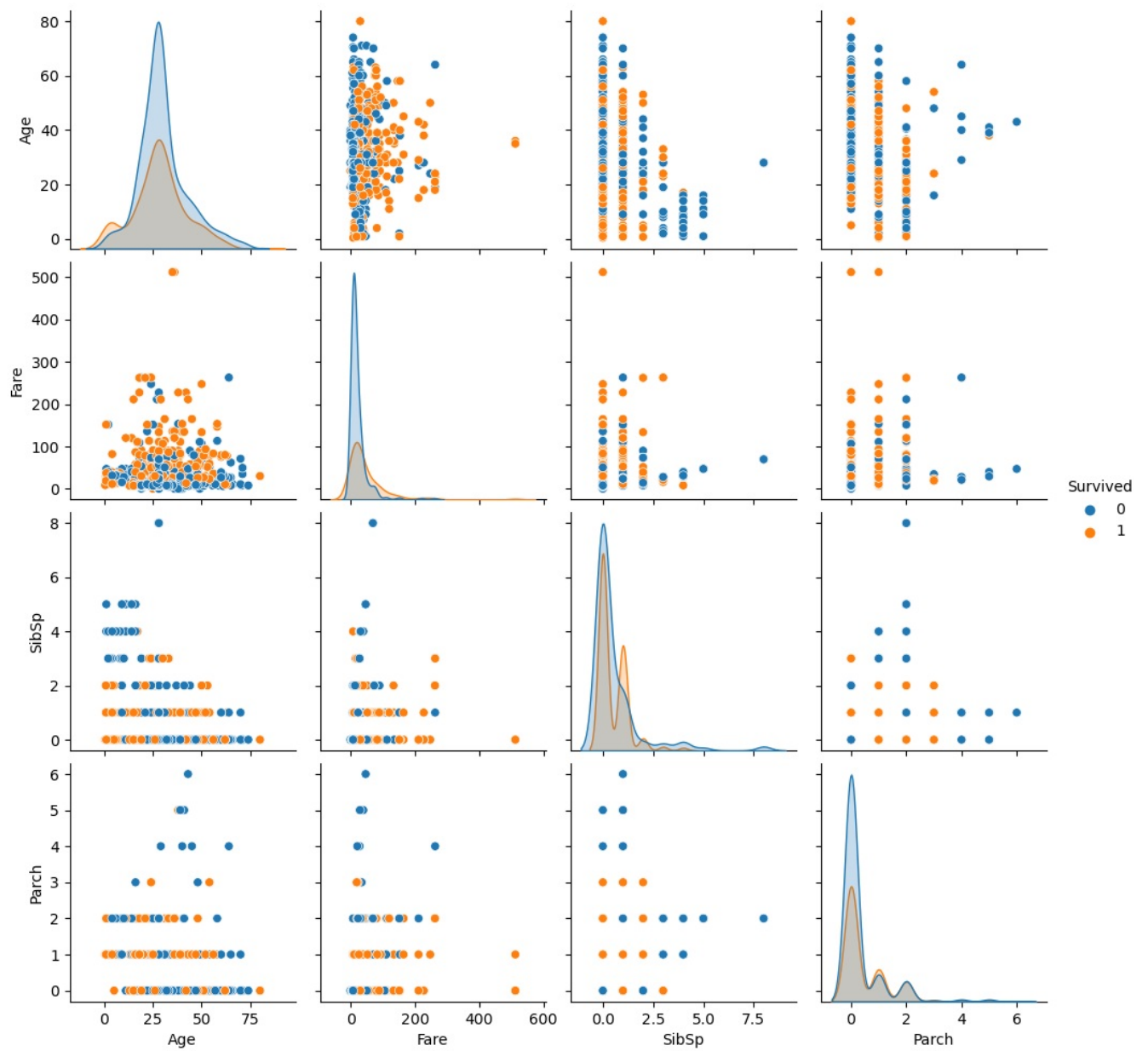
```
In [52]: import seaborn as sns
import matplotlib.pyplot as plt

# Pair plot for selected features
#Pair Plot: This is useful for seeing all pairwise relationships and distributions in one plot.
sns.pairplot(df, hue='Survived', vars=['Age', 'Fare', 'SibSp', 'Parch'], diag_kind='kde')
plt.suptitle('Pair Plot of Features', y=1.02)
plt.show()
```

C:\Users\cse\anaconda3\Lib\site-packages\seaborn\axisgrid.py:118: UserWarning: The figure layout has changed to tight  
self.\_figure.tight\_layout(\*args, \*\*kwargs)

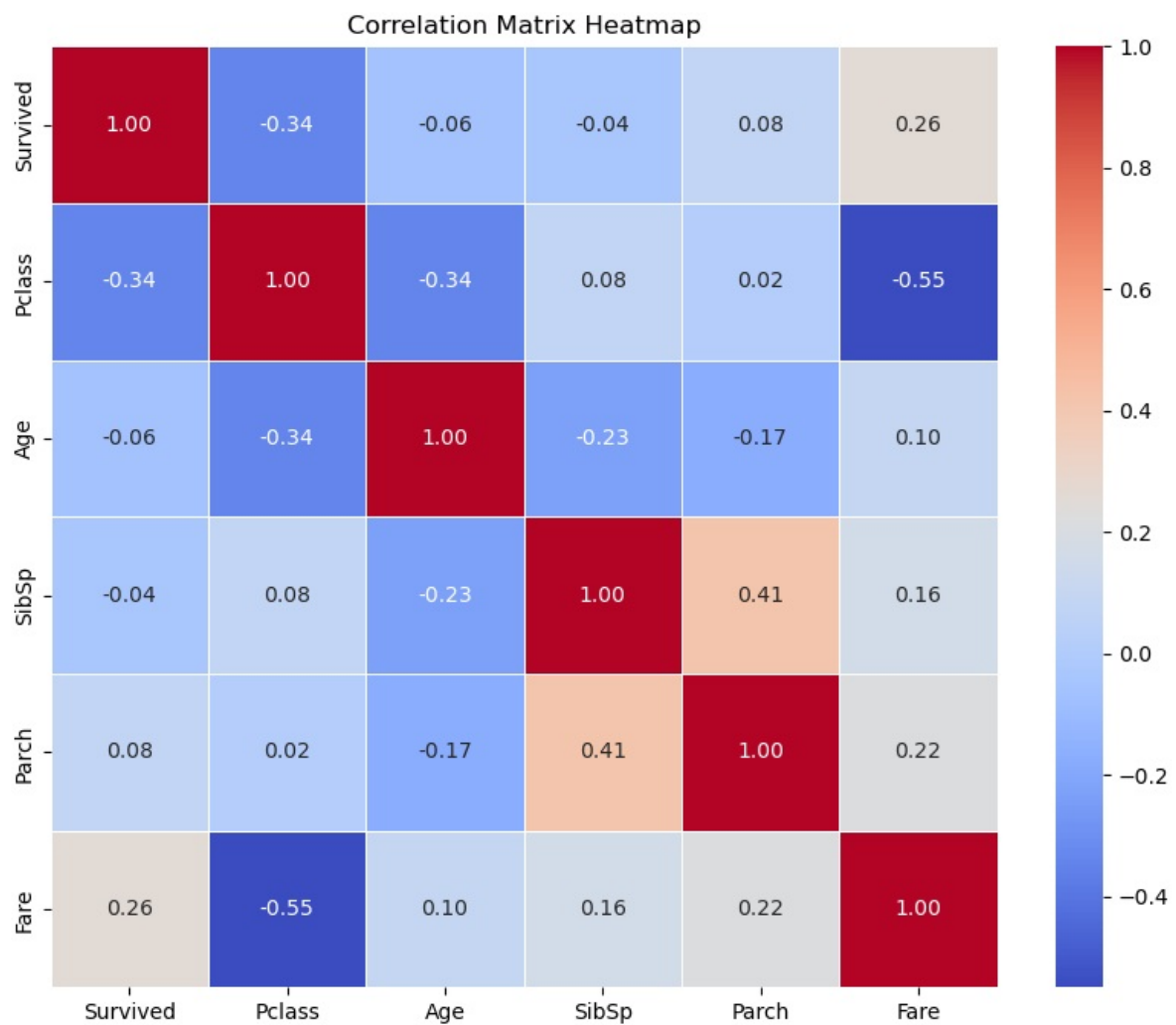


Pair Plot of Features



```
In [53]: # Compute correlation matrix
corr = df[['Survived', 'Pclass', 'Age', 'SibSp', 'Parch', 'Fare']].corr()

# Plot heatmap
#Correlation Heatmap: Provides a visual summary of relationships between numerical variables.
plt.figure(figsize=(10, 8))
sns.heatmap(corr, annot=True, cmap='coolwarm', fmt='.2f', linewidths=0.5)
plt.title('Correlation Matrix Heatmap')
plt.show()
```

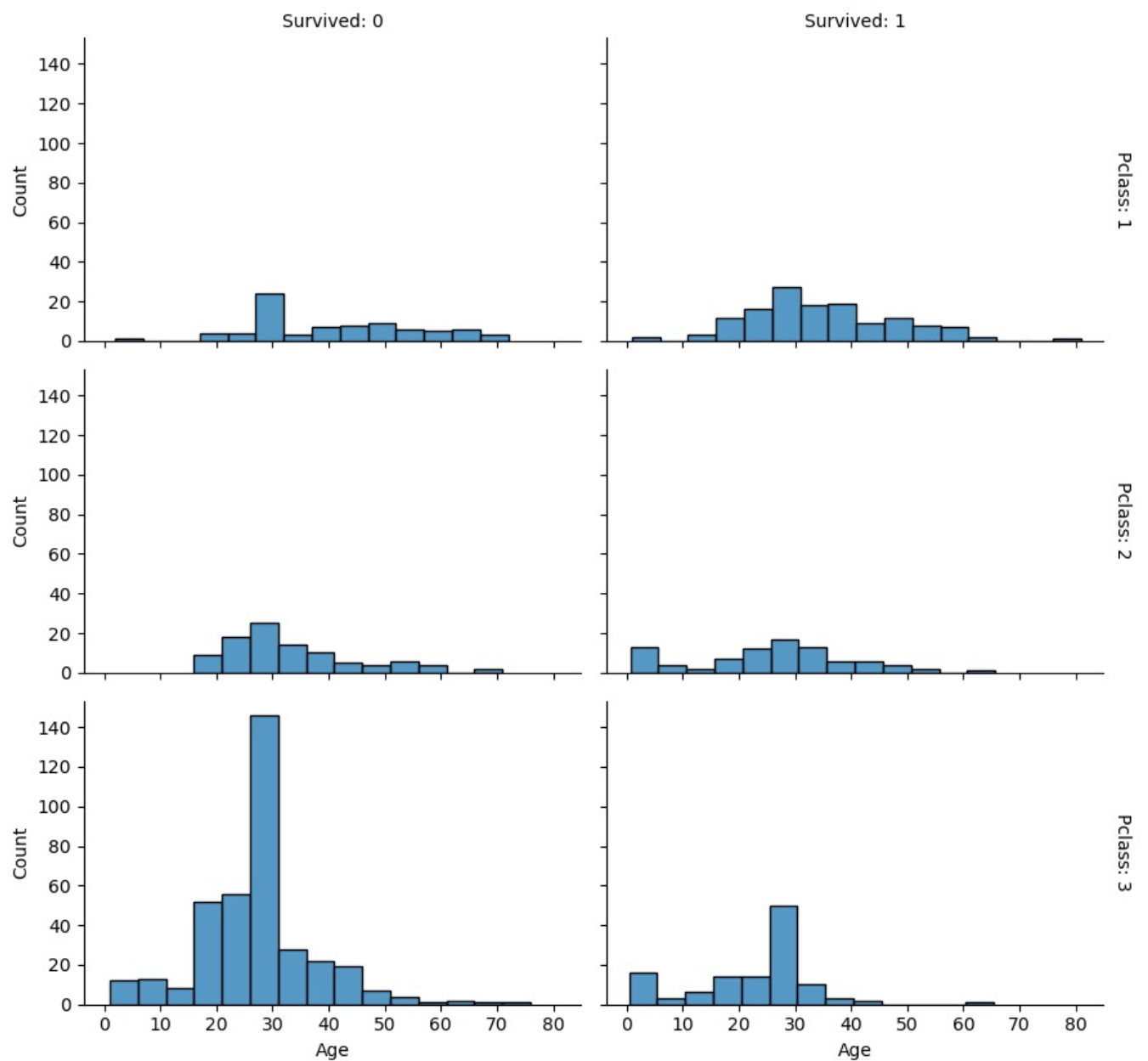


```
In [54]: # Facet Grid for Survived vs Age and Pclass
# Facet Grid: Allows you to compare distributions across subsets of data, which is great for exploring interact
g = sns.FacetGrid(df, col='Survived', row='Pclass', margin_titles=True, height=3, aspect=1.5)
g.map_dataframe(sns.histplot, x='Age', binwidth=5)
g.set_axis_labels('Age', 'Count')
g.set_titles(col_template='Survived: {col_name}', row_template='Pclass: {row_name}')
g.fig.subplots_adjust(top=0.9)
g.fig.suptitle('Distribution of Age by Survival and Passenger Class')
plt.show()
```

C:\Users\cse\anaconda3\Lib\site-packages\seaborn\axisgrid.py:118: UserWarning: The figure layout has changed to tight

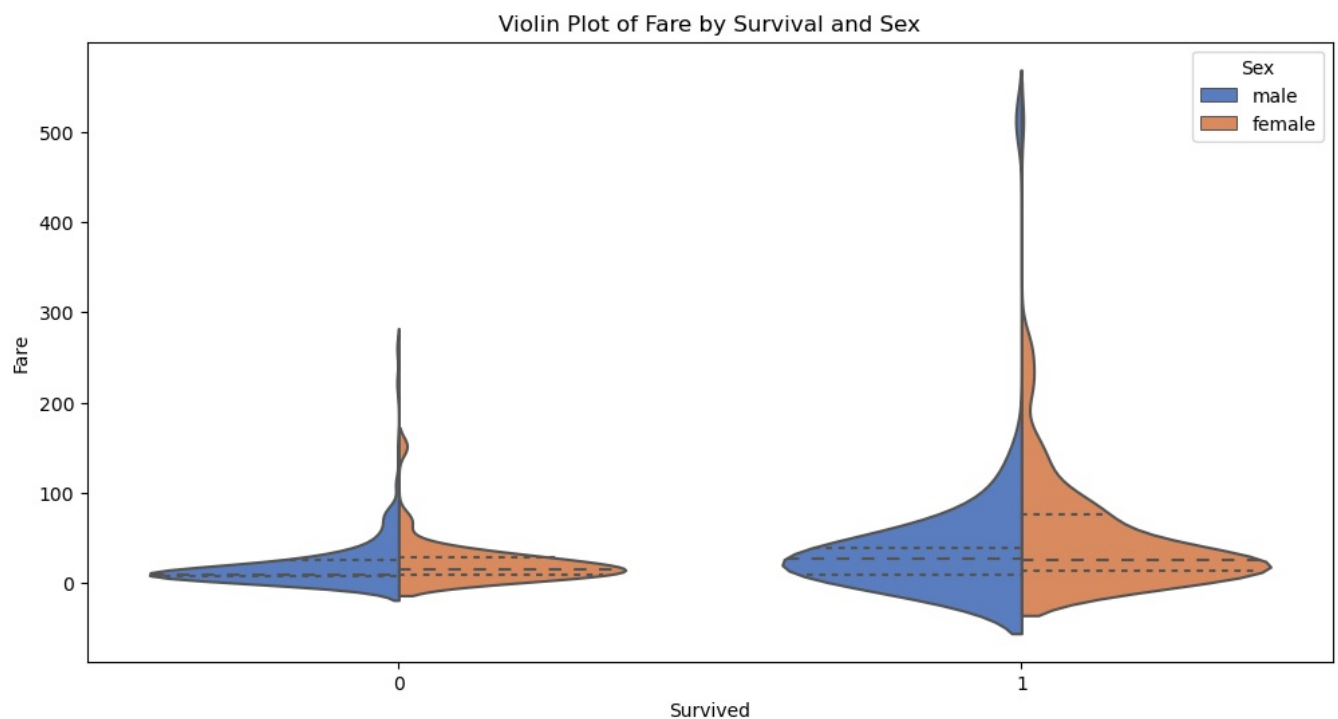
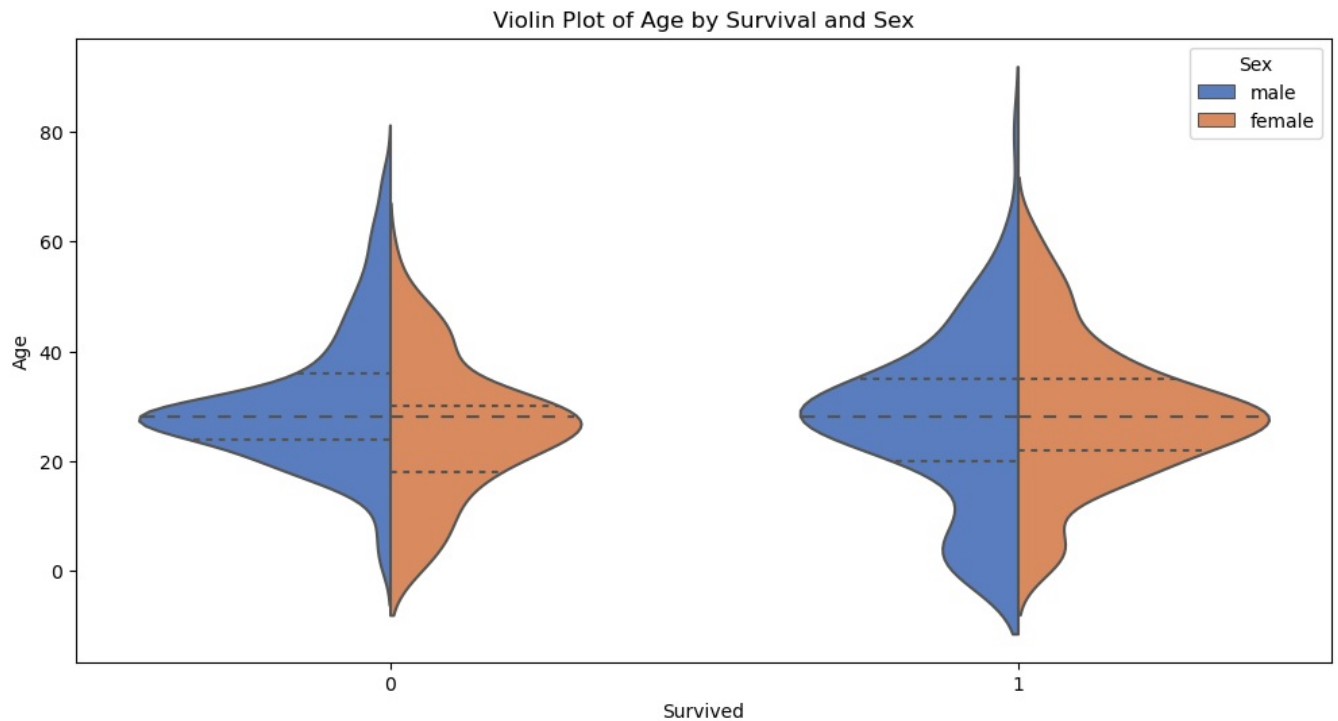
```
self._figure.tight_layout(*args, **kwargs)
```

Distribution of Age by Survival and Passenger Class



```
# Violin Plot: Displays the distribution of data points, showing the probability density of the data at different
plt.figure(figsize=(12, 6))
sns.violinplot(x='Survived', y='Age', hue='Sex', data=df, split=True, inner='quart', palette='muted')
plt.title('Violin Plot of Age by Survival and Sex')
plt.xlabel('Survived')
plt.ylabel('Age')
plt.show()

plt.figure(figsize=(12, 6))
sns.violinplot(x='Survived', y='Fare', hue='Sex', data=df, split=True, inner='quart', palette='muted')
plt.title('Violin Plot of Fare by Survival and Sex')
plt.xlabel('Survived')
plt.ylabel('Fare')
plt.show()
```



In [ ]:

Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js