



CHRIST
(DEEMED TO BE UNIVERSITY)
DELHI - NCR, INDIA

AI/ML Programming

MCA-475

CIA – 01

BY

Himanshu Heda (24225013)

SUBMITTED TO

Dr. Manjula Shannhog

SCHOOL OF SCIENCES

2024-25

Importing the Libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from datetime import date
```

✓ 13.3s

Python

Importing the CSV file to Perform some analysis

```
df = pd.read_csv('./Dataset/train.csv')
```

✓ 0.1s

Python

Performing few commands like head, tail, info, describe, shape, columns, dtypes

```
df.head()
```

✓ 0.0s

Python

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

```
df.tail()
```

✓ 0.0s

Python

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.00	NaN	S
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.00	B42	S
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.45	NaN	S
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.00	C148	C
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.75	NaN	Q

```
df.info()
```

✓ 0.1s

Python

```
<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          891 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        295 non-null    object
11  Embarked     891 non-null    object
```

```
df.info()
[5] ✓ 0.1s Python
...
<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
```

```
df.dtypes
[7] ✓ 0.0s Python
...
PassengerId    int64
Survived        int64
Pclass          int64
Name            object
Sex             object
Age            float64
SibSp           int64
Parch           int64
Ticket          object
Fare            float64
Cabin           object
Embarked        object
dtype: object
```

Is use to Replace the Existing Values in the DataFrame

```
df.rename(columns={'PassengerId': 'P_ID', 'Pclass': 'P_Class'}, inplace=True)
df.head()
[8] ✓ 0.0s Python
```

	P_ID	Survived	P_Class	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

[Generate](#) [+ Code](#) [+ Markdown](#)

```
df.sample()
✓ 0.0s Python
```

P_ID	Survived	P_Class	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked	
371	372	0	3	Wiklund, Mr. Jakob Alfred	male	18.0	1	0	3101267	6.4958	NaN	S

```
df.shape
✓ 0.0s Python
```

(891, 12)

```
df.columns
✓ 0.0s Python
```

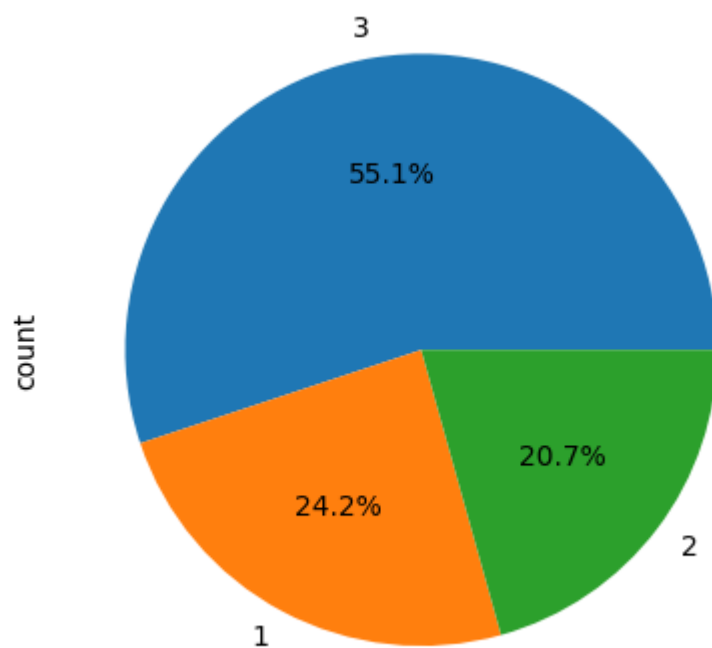
Index(['P_ID', 'Survived', 'P_Class', 'Name', 'Sex', 'Age', 'SibSp', 'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'], dtype='object')

```
df.isnull().sum()
✓ 0.0s Python
```

P_ID	0
Survived	0
P_Class	0
Name	0
Sex	0
Age	177
SibSp	0
Parch	0
Ticket	0
Fare	0
Cabin	687
Embarked	2

dtype: int64

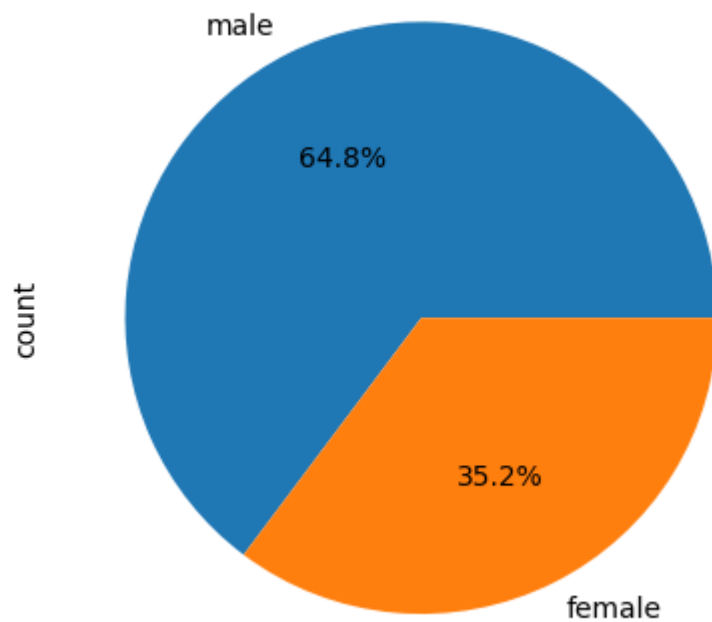
```
df['P_Class'].value_counts().plot(kind='pie', autopct='%1.1f%%')
```



```
[34] df.nunique()
✓ 0.0s Python
... P_ID      891
    Survived  2
    P_Class   3
    Name      891
    Sex        2
    Age       88
    SibSp       7
    Parch       7
    Ticket    681
    Fare     248
    Cabin     147
    Embarked   3
    dtype: int64

[25] df['Sex'].value_counts()
✓ 0.0s Python
... Sex
    male      577
    female    314
    Name: count, dtype: int64
```

```
df['Sex'].value_counts().plot(kind='pie', autopct='%1.1f%%')
```



```
df.sort_values('Age', ascending=False).head()
```

	P_ID	Survived	P_Class	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
630	631	1	1	Barkworth, Mr. Algernon Henry Wilson	male	80.0	0	0	27042	30.0000	A23	S
851	852	0	3	Svensson, Mr. Johan	male	74.0	0	0	347060	7.7750	NaN	S
493	494	0	1	Artagaveytia, Mr. Ramon	male	71.0	0	0	PC 17609	49.5042	NaN	C
96	97	0	1	Goldschmidt, Mr. George B	male	71.0	0	0	PC 17754	34.6542	A5	C
116	117	0	3	Connors, Mr. Patrick	male	70.5	0	0	370369	7.7500	NaN	Q

```
df.describe(include='all')
```

	P_ID	Survived	P_Class	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
count	891.000000	891.000000	891.000000	891	891	714.000000	891.000000	891.000000	891	891.000000	204	889
unique	NaN	NaN	NaN	891	2	NaN	NaN	NaN	681	NaN	147	3
top	NaN	NaN	NaN	Braund, Mr. Owen Harris	male	NaN	NaN	NaN	347082	NaN	B96 B98	S
freq	NaN	NaN	NaN	1	577	NaN	NaN	NaN	7	NaN	4	644
mean	446.000000	0.383838	2.308642	NaN	NaN	29.699118	0.523008	0.381594	NaN	32.204208	NaN	NaN
std	257.353842	0.486592	0.836071	NaN	NaN	14.526497	1.102743	0.806057	NaN	49.693429	NaN	NaN
min	1.000000	0.000000	1.000000	NaN	NaN	0.420000	0.000000	0.000000	NaN	0.000000	NaN	NaN
25%	223.500000	0.000000	2.000000	NaN	NaN	20.125000	0.000000	0.000000	NaN	7.910400	NaN	NaN
50%	446.000000	0.000000	3.000000	NaN	NaN	28.000000	0.000000	0.000000	NaN	14.454200	NaN	NaN
75%	668.500000	1.000000	3.000000	NaN	NaN	38.000000	1.000000	0.000000	NaN	31.000000	NaN	NaN
max	891.000000	1.000000	3.000000	NaN	NaN	80.000000	8.000000	6.000000	NaN	512.329200	NaN	NaN

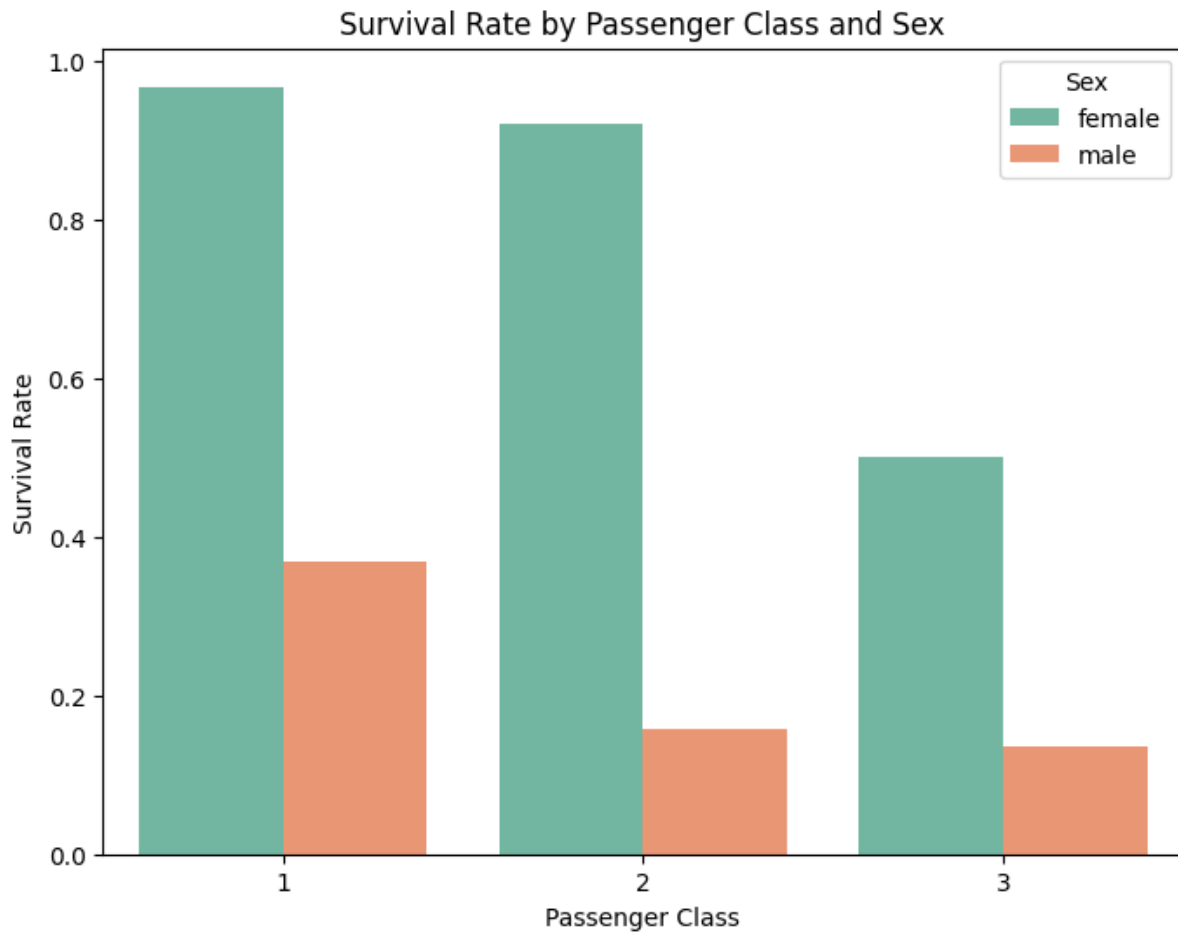
Perform the analysis on the basis of which gender is sitting in which class

```
# Groupby analysis: Survival rate by P_Class and Sex
grouped = df.groupby(['P_Class', 'Sex'])['Survived'].mean().reset_index()
print('Survival rate by P_Class and Sex:')
print(grouped)
```

	P_Class	Sex	Survived
0	1	female	0.968085
1	1	male	0.368852
2	2	female	0.921053
3	2	male	0.157407
4	3	female	0.500000
5	3	male	0.135447

```
# Plotting the survival rate by P_Class and Sex

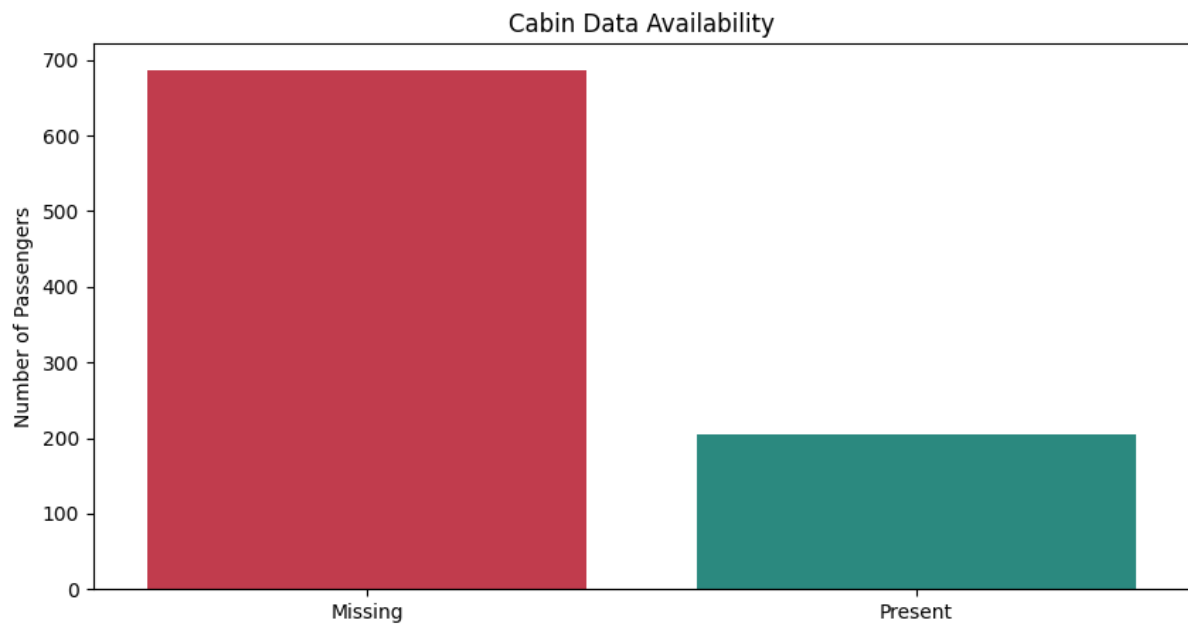
plt.figure(figsize=(8,6))
sns.barplot(data=grouped, x='P_Class', y='Survived', hue='Sex',
palette='Set2')
plt.title('Survival Rate by Passenger Class and Sex')
plt.xlabel('Passenger Class')
plt.ylabel('Survival Rate')
plt.legend(title='Sex')
plt.show()
```



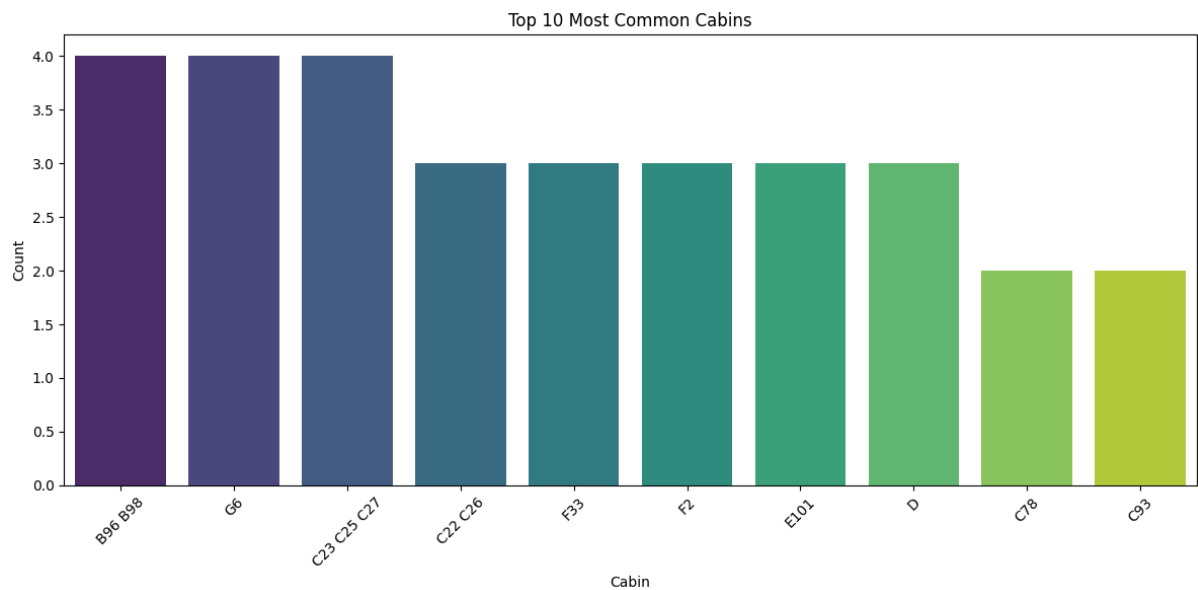
Perform the analysis on the basis of Cabin and Passenger

```
df['Cabin'].unique()
array([nan, 'C85', 'C123', 'E46', 'G6', 'C103', 'D56', 'A6',
       'C23 C25 C27', 'B78', 'D33', 'B30', 'C52', 'B28', 'C83', 'F33',
       'F G73', 'E31', 'A5', 'D10 D12', 'D26', 'C110', 'B58 B60', 'E101',
       'F E69', 'D47', 'B86', 'F2', 'C2', 'E33', 'B19', 'A7', 'C49', 'F4',
       'A32', 'B4', 'B80', 'A31', 'D36', 'D15', 'C93', 'C78', 'D35',
       'C87', 'B77', 'E67', 'B94', 'C125', 'C99', 'C118', 'D7', 'A19',
       'B49', 'D', 'C22 C26', 'C106', 'C65', 'E36', 'C54',
       'B57 B59 B63 B66', 'C7', 'E34', 'C32', 'B18', 'C124', 'C91', 'E40',
       'T', 'C128', 'D37', 'B35', 'E50', 'C82', 'B96 B98', 'E10', 'E44',
       'A34', 'C104', 'C111', 'C92', 'E38', 'D21', 'E12', 'E63', 'A14',
       'B37', 'C30', 'D20', 'B79', 'E25', 'D46', 'B73', 'C95', 'B38',
       'B39', 'B22', 'C86', 'C70', 'A16', 'C101', 'C68', 'A10', 'E68',
       'B41', 'A20', 'D19', 'D50', 'D9', 'A23', 'B50', 'A26', 'D48',
       'E58', 'C126', 'B71', 'B51 B53 B55', 'D49', 'B5', 'B20', 'F G63',
       'C62 C64', 'E24', 'C90', 'C45', 'E8', 'B101', 'D45', 'C46', 'D30',
       'E121', 'D11', 'E77', 'F38', 'B3', 'D6', 'B82 B84', 'D17', 'A36',
       'B102', 'B69', 'E49', 'C47', 'D28', 'E17', 'A24', 'C50', 'B42',
       'C148'], dtype=object)
```

```
# Graphical presentation of Cabin data
plt.figure(figsize=(10,5))
cabin_missing = df['Cabin'].isnull().sum()
cabin_present = df['Cabin'].notnull().sum()
sns.barplot(x=['Missing', 'Present'], y=[cabin_missing, cabin_present],
palette=['#d7263d', '#1b998b'])
plt.title('Cabin Data Availability')
plt.ylabel('Number of Passengers')
plt.show()
```



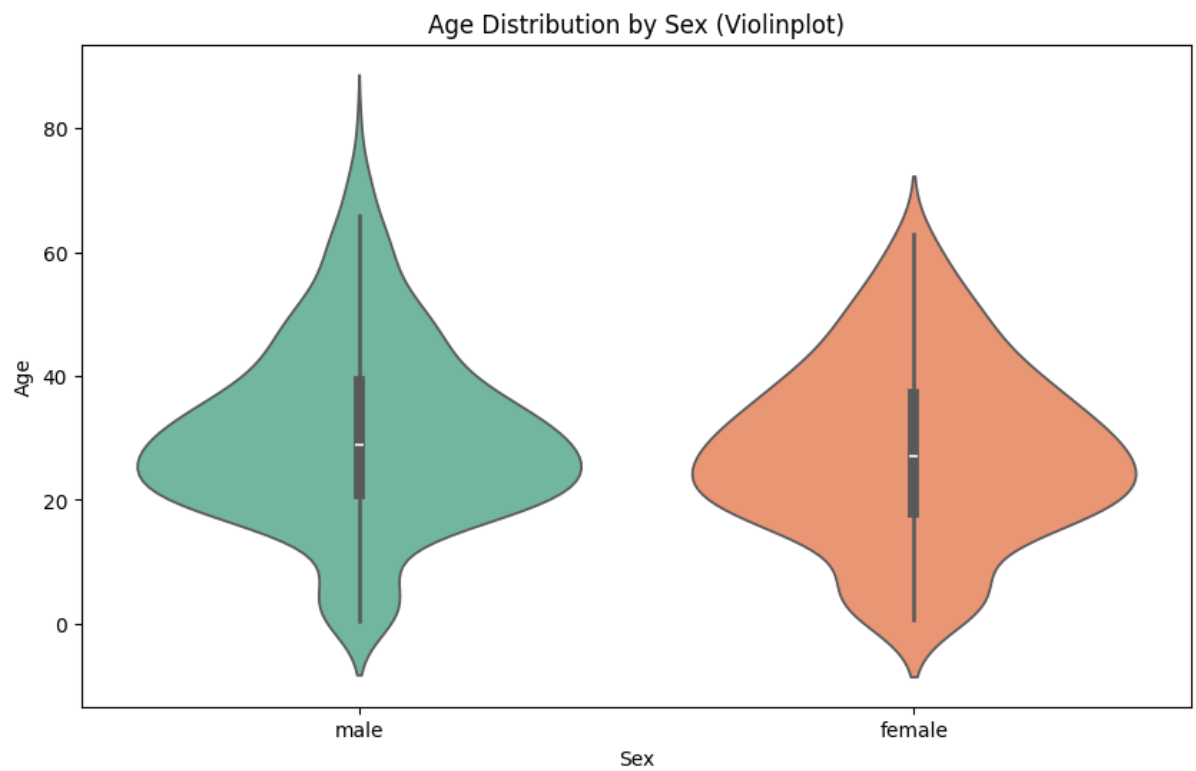
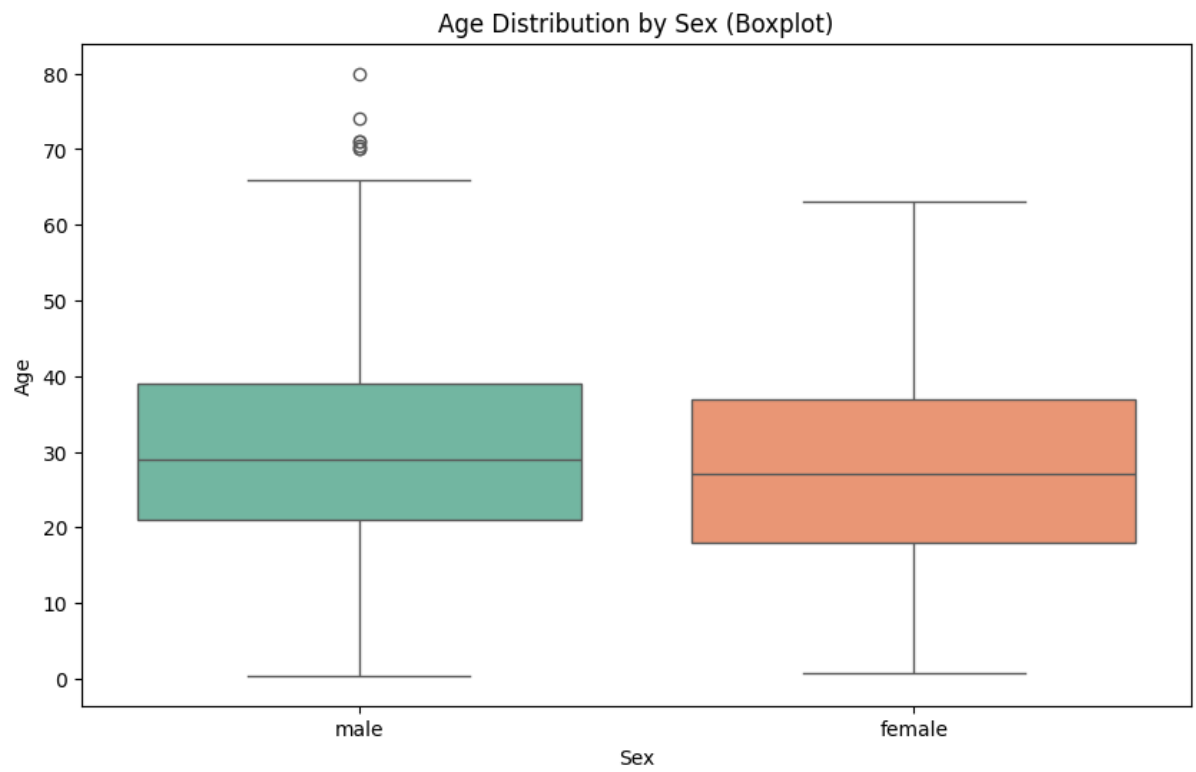
```
# Most common cabins (excluding missing)
cabin_counts = df['Cabin'].value_counts().head(10)
plt.figure(figsize=(12,6))
sns.barplot(x=cabin_counts.index, y=cabin_counts.values, palette='viridis')
plt.title('Top 10 Most Common Cabins')
plt.xlabel('Cabin')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

Perform the analysis on the basis of Age and Sex using Boxplot and Violinplot

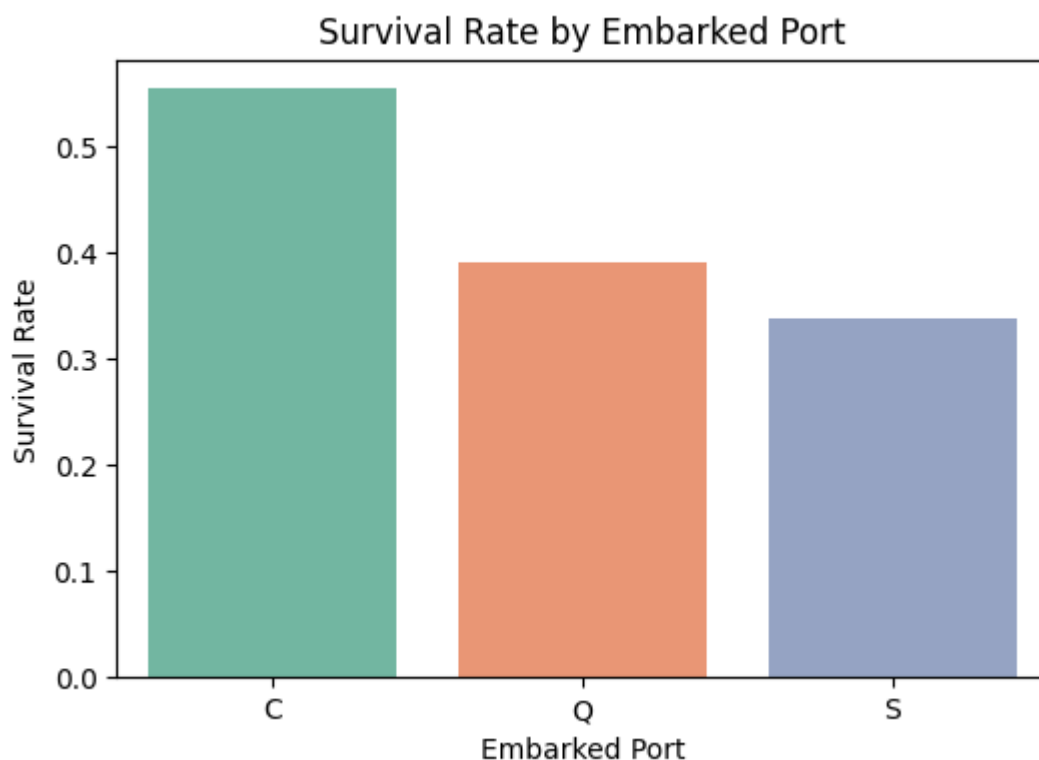
```
# Age distribution by Sex: Boxplot and Violinplot
plt.figure(figsize=(10,6))
sns.boxplot(data=df, x='Sex', y='Age', palette='Set2')
plt.title('Age Distribution by Sex (Boxplot)')
plt.xlabel('Sex')
plt.ylabel('Age')
plt.show()

plt.figure(figsize=(10,6))
sns.violinplot(data=df, x='Sex', y='Age', palette='Set2')
plt.title('Age Distribution by Sex (Violinplot)')
plt.xlabel('Sex')
plt.ylabel('Age')
plt.show()
```



Perform the analysis on the basis of how much Survivals on the Embarked Port using Barplot

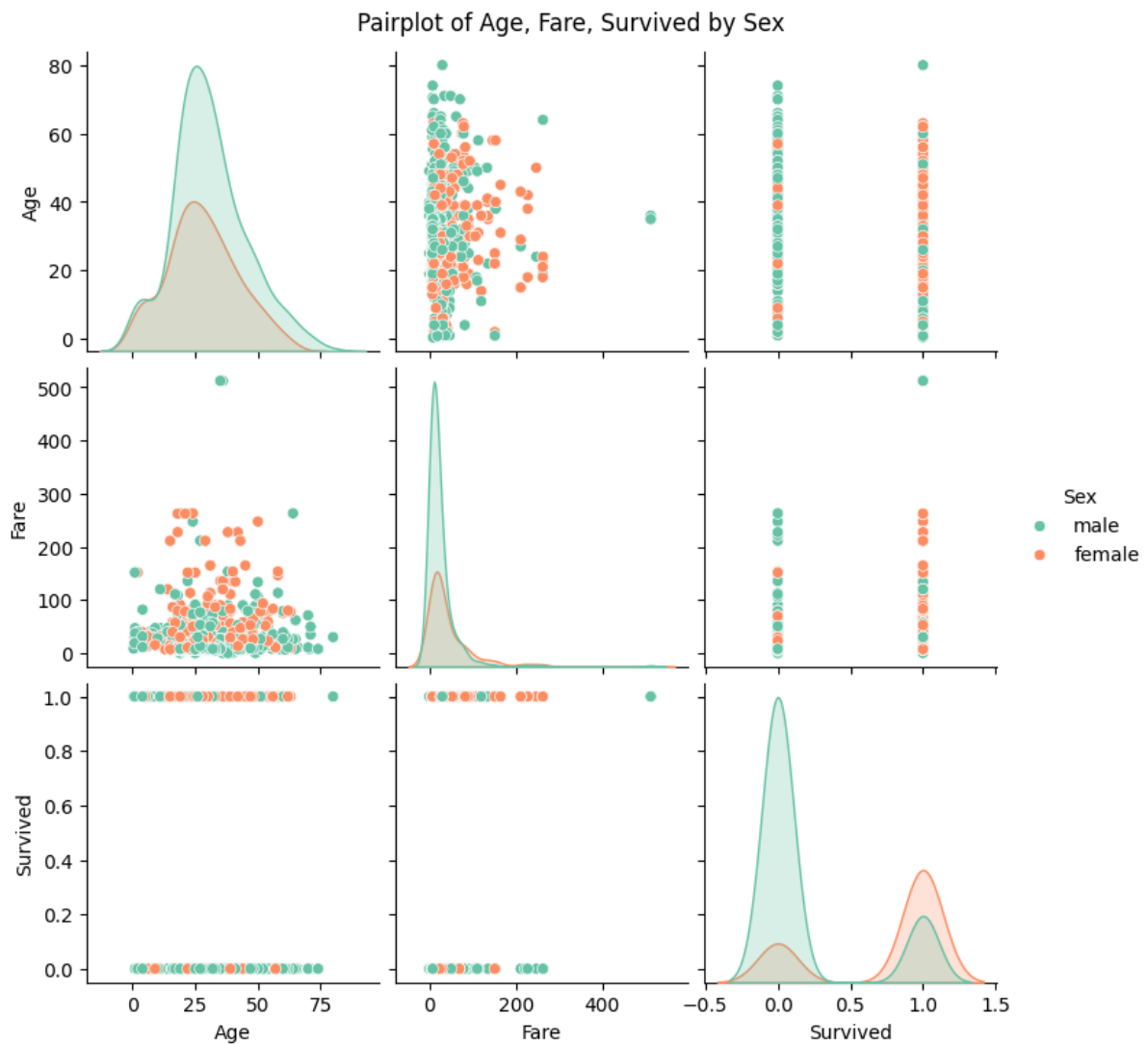
```
# Survival rate by Embarked port
embarked_survival = df.groupby('Embarked')['Survived'].mean().reset_index()
print('Survival rate by Embarked port:')
print(embarked_survival)
plt.figure(figsize=(6,4))
sns.barplot(data=embarked_survival, x='Embarked', y='Survived',
palette='Set2')
plt.title('Survival Rate by Embarked Port')
plt.ylabel('Survival Rate')
plt.xlabel('Embarked Port')
plt.show()
```



Perform the analysis on the basis of which Age group , what's the Fare and How much Survived using Pairplot

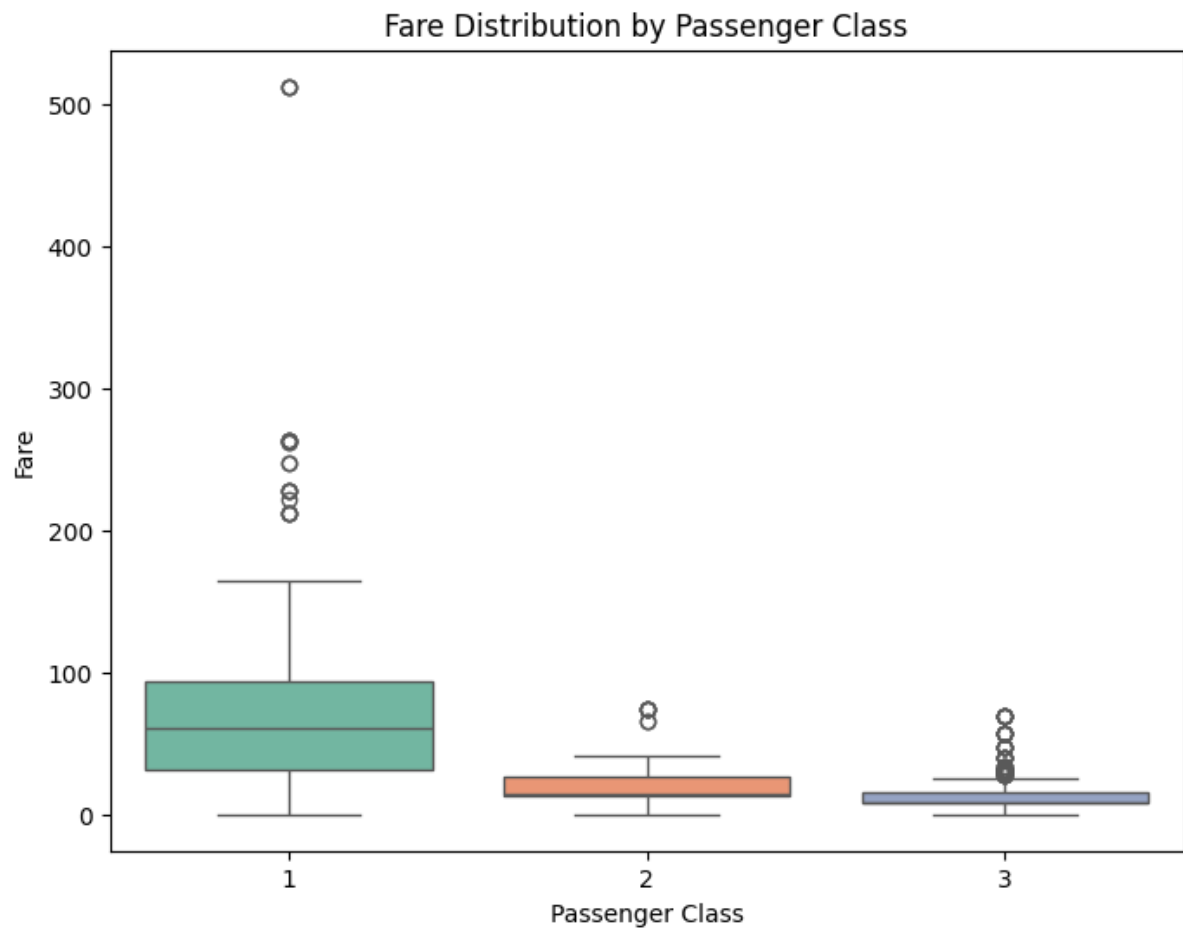
[Generate](#) [+ Code](#) [+ Markdown](#)

```
# Pairplot for selected features
sns.pairplot(df, vars=['Age', 'Fare', 'Survived'], hue='Sex', palette='Set2')
plt.suptitle('Pairplot of Age, Fare, Survived by Sex', y=1.02)
plt.show()
```



Perform the analysis on the basis of Passenger Class and how much Fare it is paying using Boxplot

```
# Distribution of Fare by Passenger Class
plt.figure(figsize=(8,6))
sns.boxplot(data=df, x='P_Class', y='Fare', palette='Set2')
plt.title('Fare Distribution by Passenger Class')
plt.xlabel('Passenger Class')
plt.ylabel('Fare')
plt.show()
```



Perform the analysis on the basis of Counting the SibSp(Siblings/Spouses) using Countplot

```
# Countplot for number of siblings/spouses aboard (SibSp)
plt.figure(figsize=(7,5))
sns.countplot(data=df, x='SibSp', palette='Set2')
plt.title('Number of Siblings/Spouses Aboard (SibSp)')
plt.xlabel('SibSp')
plt.ylabel('Count')
plt.show()
```

