


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


```
from google.colab import files
uploaded = files.upload()
```

 No file chosen Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.
Saving titanic.csv to titanic (1).csv

```
import io
```


```
df=pd.read_csv(io.BytesIO(uploaded['titanic (1).csv']))
```

```
df.head()
```




	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town	alive	alone
0	0	3	male	22.0	1	0	7.2500	S	Third	man	True	NaN	Southampton	no	False
1	1	1	female	38.0	1	0	71.2833	C	First	woman	False	C	Cherbourg	yes	False
2	1	3	female	26.0	0	0	7.9250	S	Third	woman	False	NaN	Southampton	yes	True
3	1	1	female	35.0	1	0	53.1000	S	First	woman	False	C	Southampton	yes	False
4	0	3	male	35.0	0	0	8.0500	S	Third	man	True	NaN	Southampton	no	True

```
print("\nDataset Info:")
print(df.info())
```

 Dataset Info:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 15 columns):
Column Non-Null Count Dtype

0 survived 891 non-null int64
1 pclass 891 non-null int64
2 sex 891 non-null object
3 age 714 non-null float64
4 sibsp 891 non-null int64
5 parch 891 non-null int64
6 fare 891 non-null float64
7 embarked 889 non-null object
8 class 891 non-null object
9 who 891 non-null object
10 adult_male 891 non-null bool
11 deck 203 non-null object
12 embark_town 889 non-null object
13 alive 891 non-null object
14 alone 891 non-null bool
dtypes: bool(2), float64(2), int64(4), object(7)
memory usage: 92.4+ KB
None

```
print("\nSummary Statistics:")
print(df.describe(include='all'))
```

 Summary Statistics:

	survived	pclass	sex	age	sibsp	parch	\
count	891.000000	891.000000	891	714.000000	891.000000	891.000000	
unique	NaN	NaN	2	NaN	NaN	NaN	
top	NaN	NaN	male	NaN	NaN	NaN	
freq	NaN	NaN	577	NaN	NaN	NaN	
mean	0.383838	2.308642	NaN	29.699118	0.523008	0.381594	
std	0.486592	0.836071	NaN	14.526497	1.102743	0.806057	
min	0.000000	1.000000	NaN	0.420000	0.000000	0.000000	
25%	0.000000	2.000000	NaN	20.125000	0.000000	0.000000	
50%	0.000000	3.000000	NaN	28.000000	0.000000	0.000000	
75%	1.000000	3.000000	NaN	38.000000	1.000000	0.000000	
max	1.000000	3.000000	NaN	80.000000	8.000000	6.000000	

	fare	embarked	class	who	adult_male	deck	embark_town	alive	\
count	891.000000	889	891	891	891	203	889	891	
unique	NaN	3	3	3	2	7	3	2	
top	NaN	S	Third	man	True	C	Southampton	no	
freq	NaN	644	491	537	537	59	644	549	

```

mean    32.204208    NaN    NaN    NaN    NaN    NaN    NaN    NaN
std     49.693429    NaN    NaN    NaN    NaN    NaN    NaN    NaN
min      0.000000    NaN    NaN    NaN    NaN    NaN    NaN    NaN
25%      7.910400    NaN    NaN    NaN    NaN    NaN    NaN    NaN
50%     14.454200    NaN    NaN    NaN    NaN    NaN    NaN    NaN
75%     31.000000    NaN    NaN    NaN    NaN    NaN    NaN    NaN
max     512.329200    NaN    NaN    NaN    NaN    NaN    NaN    NaN

```

```

alone
count    891
unique      2
top      True
freq     537
mean     NaN
std      NaN
min      NaN
25%      NaN
50%      NaN
75%      NaN
max      NaN

```

```

print("\nMissing Values:")
print(df.isnull().sum())

```



```

Missing Values:
survived      0
pclass        0
sex           0
age          177
sibsp         0
parch         0
fare          0
embarked      2
class         0
who           0
adult_male    0
deck         688
embark_town   2
alive         0
alone         0
dtype: int64

```

```

df = df.dropna(subset=['survived', 'sex'])
print(df)

```



```

survived  pclass    sex  age  sibsp  parch    fare  embarked  class \
0         0        3   male  22.0    1     0    7.2500         S   Third
1         1        1  female  38.0    1     0   71.2833         C   First
2         1        3  female  26.0    0     0    7.9250         S   Third
3         1        1  female  35.0    1     0   53.1000         S   First
4         0        3   male  35.0    0     0    8.0500         S   Third
..      ...      ...   ...   ...   ...   ...   ...   ...   ...
886      0        2   male  27.0    0     0   13.0000         S  Second
887      1        1  female  19.0    0     0   30.0000         S   First
888      0        3  female   NaN    1     2   23.4500         S   Third
889      1        1   male  26.0    0     0   30.0000         C   First
890      0        3   male  32.0    0     0    7.7500         Q   Third

```

```

who  adult_male  deck  embark_town  alive  alone
0    man        True  NaN  Southampton    no  False
1  woman       False   C   Cherbourg   yes  False
2  woman       False  NaN  Southampton   yes  True
3  woman       False   C   Southampton   yes  False
4    man        True  NaN  Southampton    no  True
..  ...      ...   ...   ...   ...   ...
886  man        True  NaN  Southampton    no  True
887  woman       False   B   Southampton   yes  True
888  woman       False  NaN  Southampton    no  False
889  man        True   C   Cherbourg   yes  True
890  man        True  NaN  Queenstown    no  True

```

```
[891 rows x 15 columns]
```

```

questions = [
    "What percentage of passengers survived?",
    "Does gender affect survival rate?",
    "Which class had higher survival rates?",
    "Were older passengers less likely to survive?",
    "Are there missing values that could affect analysis?"
]

print("\nBusiness Questions:")
for q in questions:
    print("-", q)

```



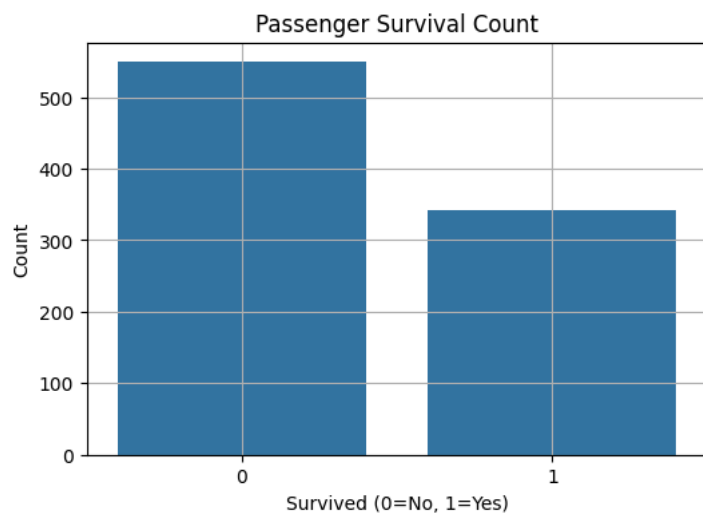
Business Questions:

- What percentage of passengers survived?
- Does gender affect survival rate?
- Which class had higher survival rates?
- Were older passengers less likely to survive?
- Are there missing values that could affect analysis?

```

plt.figure(figsize=(6,4))
sns.countplot(x='survived', data=df)
plt.title('Passenger Survival Count')
plt.xlabel('Survived (0=No, 1=Yes)')
plt.ylabel('Count')
plt.grid(True)
plt.show()

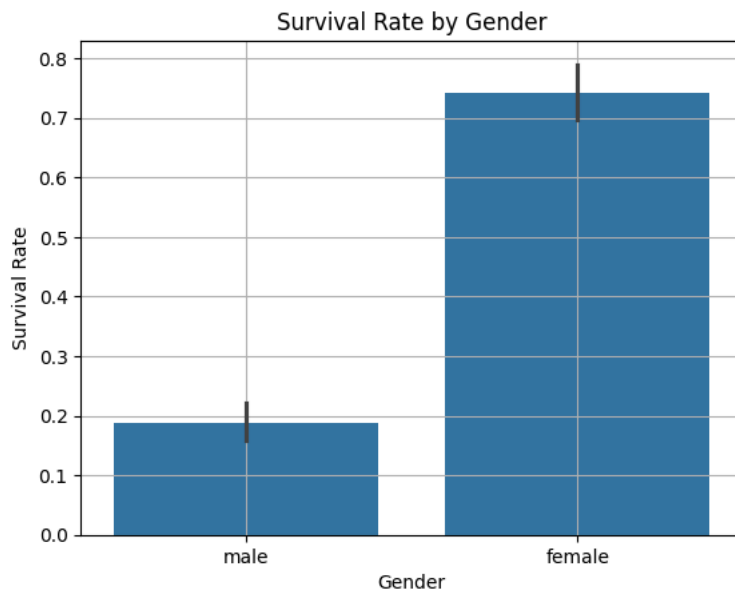
```



```

sns.barplot(x='sex', y='survived', data=df)
plt.title('Survival Rate by Gender')
plt.xlabel('Gender')
plt.ylabel('Survival Rate')
plt.grid(True)
plt.show()

```

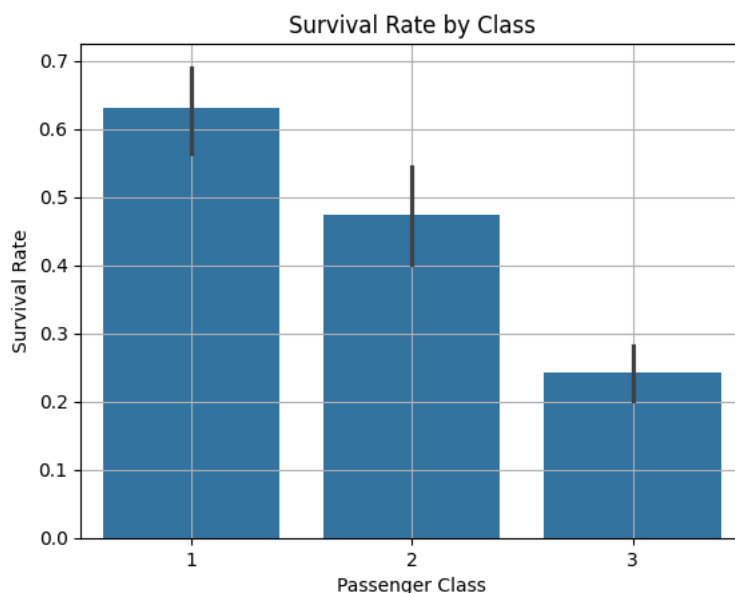


```
class_survival = df.groupby('pclass')['survived'].mean()
print("\nSurvival Rate by Passenger Class:")
print(class_survival)
```

```
sns.barplot(x='pclass', y='survived', data=df)
plt.title('Survival Rate by Class')
plt.xlabel('Passenger Class')
plt.ylabel('Survival Rate')
plt.grid(True)
plt.show()
```



```
Survival Rate by Passenger Class:
pclass
1    0.629630
2    0.472826
3    0.242363
Name: survived, dtype: float64
```



```
print("\nMissing Age values:", df['age'].isnull().sum())
print("Missing Embarked values:", df['embarked'].isnull().sum())
```



```
Missing Age values: 177
Missing Embarked values: 2
```

```
female_survival = df[df['sex'] == 'female']['survived'].mean()
male_survival = df[df['sex'] == 'male']['survived'].mean()

print("\nAverage Survival Rate by Gender:")
```

```
print("Female:", female_survival)
print("Male:", male_survival)
```



```
Average Survival Rate by Gender:
Female: 0.7420382165605095
Male: 0.18890814558058924
```

```
import scipy.stats as stats
```

```
t_stat, p_value = stats.ttest_ind(male_survival, female_survival)
```

```
print("\nTwo-Sample T-Test Result:")
print("t-statistic:", t_stat)
print("p-value:", p_value)
```

```
alpha = 0.05 # significance level
```

```
if p_value < alpha:
    print("\nResult: Reject Null Hypothesis (H0)")
    print("Conclusion: Survival rate differs significantly between males and females.")
else:
    print("\nResult: Fail to Reject Null Hypothesis (H0)")
    print("Conclusion: No significant difference in survival rates between genders.")
```



```
Two-Sample T-Test Result:
t-statistic: -19.297816550123354
```