

```
In [ ]: #TITANIC SURVIVAL PREDICTION TASK (1)
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```
In [1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
```

```
In [2]: # Load data
df=pd.read_csv("Titanic-Dataset.csv")
```

```
In [3]: df
```

Out[3]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.25
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.28
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.92
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.10
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.05
...	...	...	...	...	...	...	...	...	...	...
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.00
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.00
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.45
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.00
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.73

891 rows × 12 columns



```
In [4]: df.head()
```

Out[4]:

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

```
In [5]: df.shape
```

Out[5]: (891, 12)

```
In [6]: df.info()
```

```
<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
```

```
In [7]: df.describe()
```

Out[7]:

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000
mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208
std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429
min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400
50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000
max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200

```
In [8]: df.duplicated().sum()
```

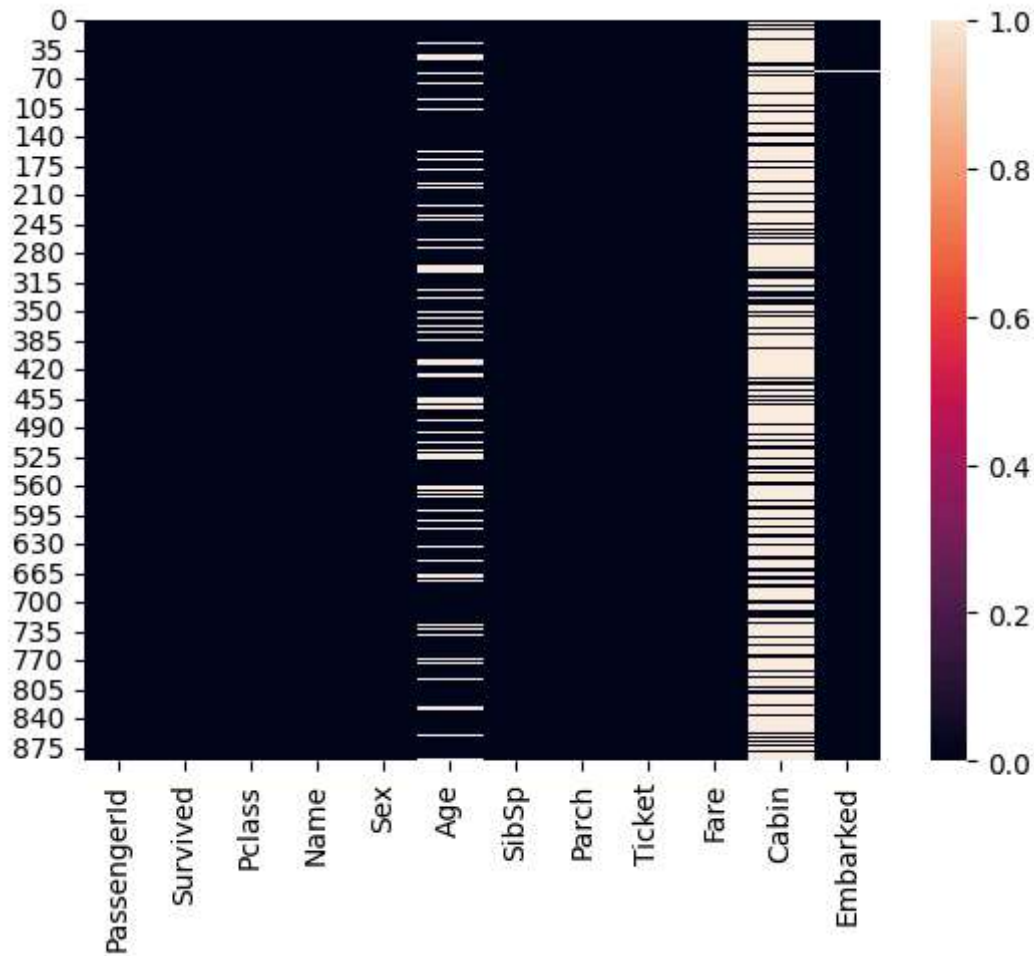
Out[8]: 0

```
In [9]: df['Survived'].value_counts()
```

Out[9]: Survived  
0 549  
1 342  
Name: count, dtype: int64

```
In [12]: sns.heatmap(df.isnull())
```

Out[12]: <Axes: >



```
In [13]: df['Sex']
```

```
Out[13]: 0      male
1      female
2      female
3      female
4      male
...
886     male
887     female
888     female
889     male
890     male
Name: Sex, Length: 891, dtype: object
```

```
In [15]: #Data Cleaning
df.drop(['PassengerId', 'Name', 'Ticket', 'Cabin', 'Embarked', 'Fare'], axis=1, inplace=True)
```

```
In [16]: df.Age=df.Age.fillna(df.Age.mean())
```

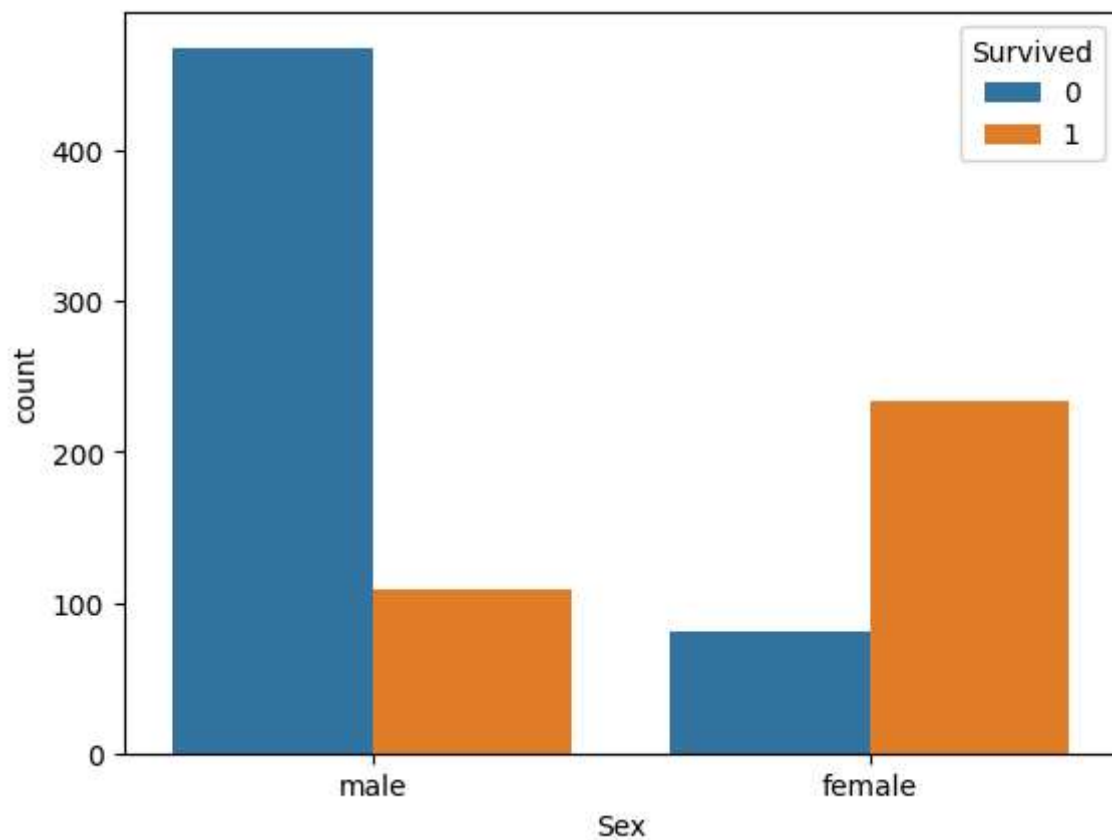
```
In [17]: df
```

Out[17]:

	Survived	Pclass	Sex	Age	SibSp	Parch
0	0	3	male	22.000000	1	0
1	1	1	female	38.000000	1	0
2	1	3	female	26.000000	0	0
3	1	1	female	35.000000	1	0
4	0	3	male	35.000000	0	0
...	...	...	...	...	...	...
886	0	2	male	27.000000	0	0
887	1	1	female	19.000000	0	0
888	0	3	female	29.699118	1	2
889	1	1	male	26.000000	0	0
890	0	3	male	32.000000	0	0

891 rows × 6 columns

```
In [21]: #Data Aanlysis
sns.countplot(x='Sex', hue='Survived', data=df)
plt.show()
```

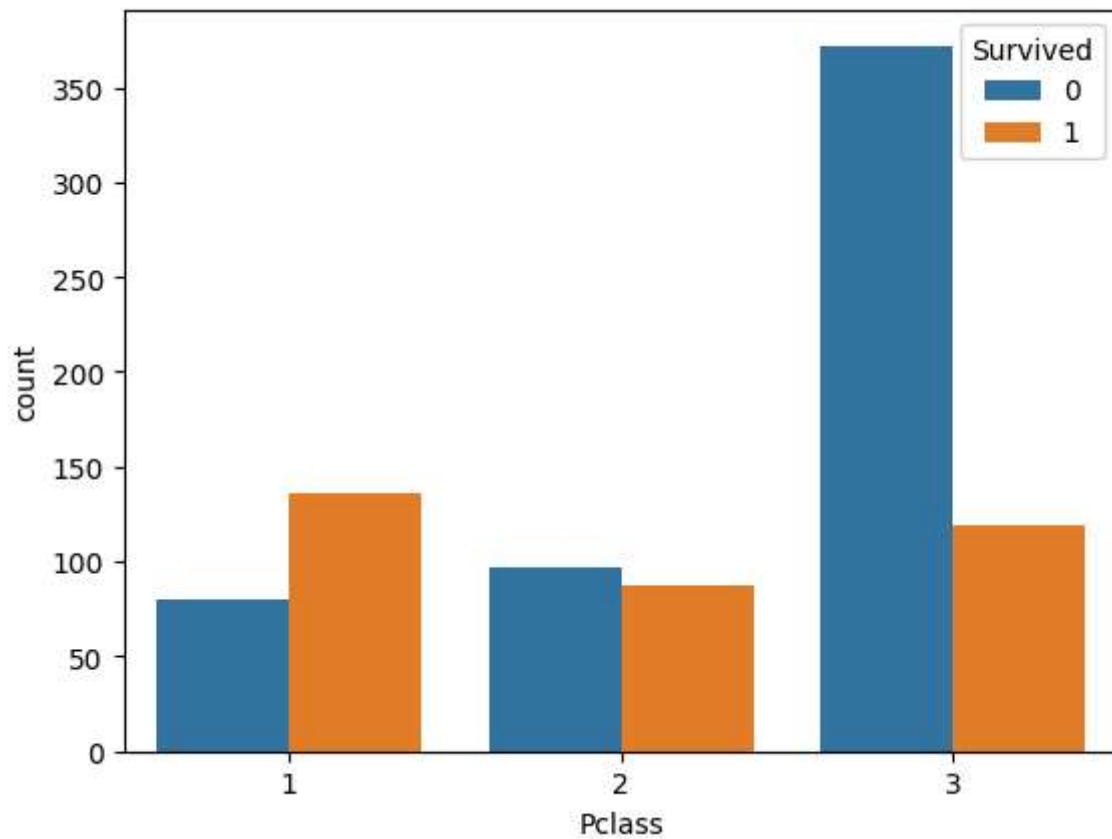


```
In [19]: df.groupby('Sex')[['Survived']].mean()
```

```
Out[19]:
```

Survived	
Sex	
female	0.742038
male	0.188908

```
In [20]: sns.countplot(x='Pclass', hue='Survived', data=df)  
plt.show()
```



```
In [22]: df['Sex'].unique()
```

```
Out[22]: array(['male', 'female'], dtype=object)
```

```
In [23]: age=df.Age.groupby(df.Survived).value_counts()  
age
```

```
Out[23]: Survived  Age
0      29.699118   125
        21.000000    19
        28.000000    18
        18.000000    17
        25.000000    17
        ...
1      43.000000     1
        47.000000     1
        53.000000     1
        55.000000     1
        80.000000     1
Name: count, Length: 144, dtype: int64
```

```
In [24]: df[df['Survived']==1 ] ['Age'].max()
```

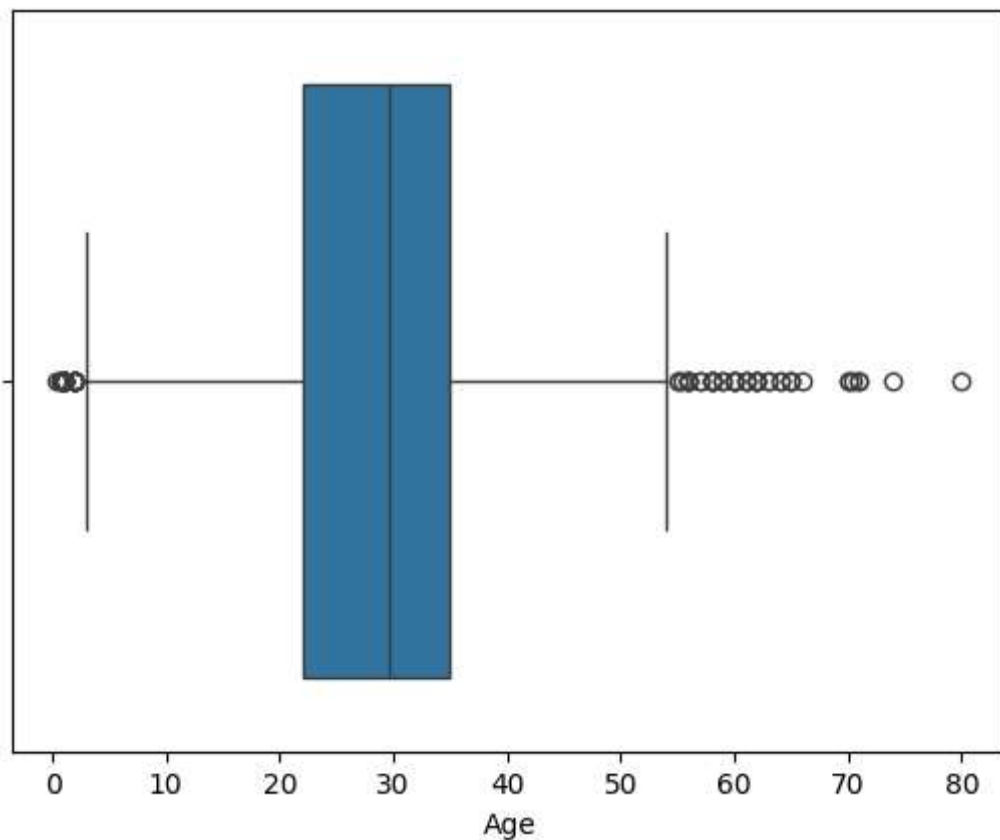
```
Out[24]: 80.0
```

```
In [25]: df[df['Survived']==0 ] ['Age'].mode()
```

```
Out[25]: 0      29.699118
Name: Age, dtype: float64
```

```
In [26]: sns.boxplot(x='Age',data=df)
```

```
Out[26]: <Axes: xlabel='Age'>
```



```
In [27]: df=df.drop(['Age'],axis=1)
```



```
In [40]: df
```

```
Out[40]:
```

	Survived	Pclass	Sex	SibSp	Parch
0	0	3	male	1	0
1	1	1	female	1	0
2	1	3	female	0	0
3	1	1	female	1	0
4	0	3	male	0	0
...	...	...	...	...	...
886	0	2	male	0	0
887	1	1	female	0	0
888	0	3	female	1	2
889	1	1	male	0	0
890	0	3	male	0	0

891 rows × 5 columns

```
In [41]: #Split features & target
x=df.drop('Survived',axis=1)
y=df.Survived
```

```
In [42]: x.shape,y.shape
```

```
Out[42]: ((891, 4), (891,))
```

```
In [52]: from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
```

```
In [56]: #Split train & test data
X = df.drop('Survived', axis=1)
y = df['Survived']
```

```
In [75]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
df= LogisticRegression(max_iter=1000)
```

```
In [79]: print(X_train.dtypes)
```

```
Pclass    int64
Sex        object
SibSp      int64
Parch      int64
dtype: object
```

```
In [80]: X_train = pd.get_dummies(X_train, drop_first=True)
```

```
X_test = pd.get_dummies(X_test, drop_first=True)
```

```
In [81]: X_test = X_test.reindex(columns=X_train.columns, fill_value=0)
```

```
In [82]: log = LogisticRegression(random_state=0)
log.fit(X_train, y_train)
```

```
Out[82]: LogisticRegression
LogisticRegression(random_state=0)
```

```
In [83]: pred=print(log.predict(X_test))
```

```
[0 0 0 1 1 1 0 1 1 0 0 0 0 0 1 0 1 0 0 0 0 1 0 0 0 0 0 0 1 0 1 0 1 0 0 0
 1 1 0 0 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 1 0 1 1 1 0 1 1 0 0 1 0 0 0 1 1 1 0 1
 0 0 1 1 1 1 0 1 1 0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 1 0 0 0 1 0 0 0 1
 0 1 0 0 0 0 0 1 1 0 1 1 0 0 0 1 0 1 0 1 0 0 1 0 1 1 0 0 1 0 1 0 0 1 1 0 0
 1 0 0 0 0 1 0 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 1 0 1 0 0 1 1 1]
```

```
In [84]: print(Y_test)
```

```
Survived
495      0
648      0
278      0
31       1
255      1
..      ...
780      1
837      0
215      1
833      0
372      0
```

```
[179 rows x 1 columns]
```

```
In [85]: data = {
    'PassengerId': [1, 2, 3, 4, 5],
    'Survived': [1, 0, 0, 1, 1]
}
df = pd.DataFrame(data)

def check_survival_status(survived):
    if survived == 1:
        return 'Survived'
    else:
        return 'Not Survived'
```

```
In [86]: df['Survival_Status'] = df['Survived'].apply(check_survival_status)
```

```
In [87]: df
```

Out[87]:

	PassengerId	Survived	Survival_Status
0	1	1	Survived
1	2	0	Not Survived
2	3	0	Not Survived
3	4	1	Survived
4	5	1	Survived

In [ ]: