

```
In [23]: import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
td = pd.read_csv(r'C:\Users\B4B3EE\Downloads\archive\tested.csv')
td

Out[23]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	892	0	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	NaN	Q
1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	NaN	S
2	894	0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	NaN	Q
3	895	0	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	NaN	S
4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	NaN	S
...	...	...	...	...	...	...	...	...	...	...	...	...
413	1305	0	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.0500	NaN	S
414	1306	1	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	C105	C
415	1307	0	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	NaN	S
416	1308	0	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.0500	NaN	S
417	1309	0	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	NaN	C

418 rows x 12 columns

```
In [24]: td.shape
Out[24]: (418, 12)

In [25]: td.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 418 entries, 0 to 417
Data columns (total 12 columns):
# Column Non-Null Count Dtype
---
0 PassengerId 418 non-null int64
1 Survived 418 non-null int64
2 Pclass 418 non-null int64
3 Name 418 non-null object
4 Sex 418 non-null object
5 Age 332 non-null float64
6 SibSp 418 non-null int64
7 Parch 418 non-null int64
8 Ticket 418 non-null object
9 Fare 417 non-null float64
10 Cabin 91 non-null object
11 Embarked 418 non-null object
dtypes: float64(2), int64(5), object(5)
memory usage: 39.3+ KB

In [26]: td.isnull().sum()
Out[26]: PassengerId 0
Survived 0
Pclass 0
Name 0
Sex 0
Age 86
SibSp 0
Parch 0
Ticket 0
Fare 1
Cabin 327
Embarked 0
dtype: int64

In [27]: td=td.drop(columns='Cabin',axis=1)

In [29]: td['Age'].fillna(td['Age'].mean(),inplace=True)

In [30]: td['Embarked'].fillna(td['Embarked'].mode()[0],inplace=True)

In [31]: td['Fare'].fillna(td['Fare'].mode()[0],inplace=True)

In [32]: td.isnull().sum().sum()
Out[32]: 0

In [34]: td['Survived'].value_counts()
Out[34]: Survived
0 266
1 152
Name: count, dtype: int64

In [35]: td.describe()
Out[35]:
```

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
count	418.000000	418.000000	418.000000	418.000000	418.000000	418.000000	418.000000
mean	1100.500000	0.363636	2.265550	30.272590	0.447368	0.392344	35.560497
std	120.810458	0.481622	0.841838	12.634534	0.896760	0.981429	55.857145
min	892.000000	0.000000	1.000000	0.170000	0.000000	0.000000	0.000000
25%	996.250000	0.000000	1.000000	23.000000	0.000000	0.000000	7.895800
50%	1100.500000	0.000000	3.000000	30.272590	0.000000	0.000000	14.454200
75%	1204.750000	1.000000	3.000000	35.750000	1.000000	0.000000	31.471875
max	1309.000000	1.000000	3.000000	76.000000	8.000000	9.000000	512.329200

```
In [37]: import matplotlib.pyplot as plt
import seaborn as sns
Matplotlib is building the font cache; this may take a moment.

In [38]: sns.set()

In [39]: sns.countplot(x='Survived',data=td)
Out[39]: <Axes: xlabel='Survived', ylabel='count'>
```



```
In [41]: sns.countplot(x='Sex',data=td)
Out[41]: <Axes: xlabel='Sex', ylabel='count'>
```



```
In [42]: sns.countplot(x='Sex',hue='Survived',data=td)
Out[42]: <Axes: xlabel='Sex', ylabel='count'>
```



```
In [43]: sns.countplot(x='Pclass',data=td)
Out[43]: <Axes: xlabel='Pclass', ylabel='count'>
```



```
In [44]: sns.countplot(x='Pclass',hue='Survived',data=td)
Out[44]: <Axes: xlabel='Pclass', ylabel='count'>
```



```
In [45]: td['Sex'].value_counts()
Out[45]: Sex
male 266
female 152
Name: count, dtype: int64

In [46]: td['Embarked'].value_counts()
Out[46]: Embarked
S 276
C 192
Q 46
Name: count, dtype: int64

In [47]: td.replace({'Sex':{'male':0,'female':1},'Embarked':{'S':0,'C':1,'Q':2}},inplace=True)
td

Out[47]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Embarked
0	892	0	3	Kelly, Mr. James	0	34.50000	0	0	330911	7.8292	2
1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	1	47.00000	1	0	363272	7.0000	0
2	894	0	2	Myles, Mr. Thomas Francis	0	62.00000	0	0	240276	9.6875	2
3	895	0	3	Wirz, Mr. Albert	0	27.00000	0	0	315154	8.6625	0
4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	1	22.00000	1	1	3101298	12.2875	0
...	...	...	...	...	...	...	...	...	...	...	...
413	1305	0	3	Spector, Mr. Woolf	0	30.27259	0	0	A.5. 3236	8.0500	0
414	1306	1	1	Oliva y Ocana, Dona. Fermina	1	39.00000	0	0	PC 17758	108.9000	1
415	1307	0	3	Saether, Mr. Simon Sivertsen	0	38.50000	0	0	SOTON/O.Q. 3101262	7.2500	0
416	1308	0	3	Ware, Mr. Frederick	0	30.27259	0	0	359309	8.0500	0
417	1309	0	3	Peter, Master. Michael J	0	30.27259	1	1	2668	22.3583	1

418 rows x 11 columns

```
In [49]: X=td.drop(columns=['PassengerId','Name','Ticket'],axis=1)
Y=td['Survived']
print(X)
print(Y)

Survived Pclass Sex Age SibSp Parch Fare Embarked
0 0 3 0 34.50000 0 0 7.8292 2
1 1 3 1 47.00000 1 0 7.0000 0
2 0 2 0 62.00000 0 0 9.6875 2
3 0 3 0 27.00000 0 0 8.6625 0
4 1 3 1 22.00000 1 1 12.2875 0
... ..
413 0 3 0 30.27259 0 0 8.0500 0
414 1 1 1 39.00000 0 0 108.9000 1
415 0 3 0 38.50000 0 0 7.2500 0
416 0 3 0 30.27259 0 0 8.0500 0
417 0 3 0 30.27259 1 1 22.3583 1

[418 rows x 8 columns]
0 0
1 1
2 0
3 0
4 1
..
413 0
414 1
415 0
416 0
417 0
Name: Survived, Length: 418, dtype: int64

In [50]: X_train,X_test,Y_train,Y_test=train_test_split(X,Y,test_size=0.2,random_state=2)
print(X.shape,X_train.shape,X_test.shape)
(418, 8) (334, 8) (84, 8)

In [53]: from sklearn.linear_model import LogisticRegression
model=LogisticRegression()
model.fit(X_train,Y_train)

Out[53]: LogisticRegression()

In [54]: X_train_prediction=model.predict(X_train)
print(X_train_prediction)

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

In [56]: train_data_accuracy=accuracy_score(Y_train,X_train_prediction)
print('Accuracy Score of training data:',train_data_accuracy)
Accuracy Score of training data: 1.0

In [57]: X_test_prediction=model.predict(X_test)
print(X_test_prediction)

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

In [59]: test_data_accuracy=accuracy_score(Y_test,X_test_prediction)
print('Accuracy score of testing data:',test_data_accuracy)
Accuracy score of testing data: 1.0

In [ ]:
```