

```
import pandas as pd
import numpy as np
import random
from sklearn import preprocessing
le = preprocessing.LabelEncoder()
```

```
from sklearn.neighbors import KNeighborsClassifier
```

```
dataFile = pd.read_csv('Titanic_train.csv')
```

```
dataFile.head()
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs)	female	38.0	1	0	PC 17599

```
dataFile = dataFile[["Survived", "Pclass", "Age", "Sex", "SibSp", "Parch", "Fare", "Embarked"]]
```

```
dataFile=dataFile.dropna()
```

```
dataFile.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 712 entries, 0 to 890
Data columns (total 8 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Survived    712 non-null    int64
1   Pclass      712 non-null    int64
2   Age         712 non-null    float64
3   Sex         712 non-null    object
4   SibSp       712 non-null    int64
5   Parch       712 non-null    int64
6   Fare        712 non-null    float64
7   Embarked    712 non-null    object
dtypes: float64(2), int64(4), object(2)
memory usage: 50.1+ KB
```

```
pclass = list(dataFile["Pclass"])
sex = list(dataFile["Sex"])
age = list(dataFile["Age"])
sibSp = list(dataFile["SibSp"])
parch = list(dataFile["Parch"])
fare = list(dataFile["Fare"])
#cabin = list(dataFile["Cabin"])
embarked = list(dataFile["Embarked"])
```

```
survived=list(dataFile["Survived"])
```

```
pclass_encoded=le.fit_transform(pclass)
sex_encoded=le.fit_transform(sex)
age_encoded=le.fit_transform(age)
sibSp_encoded=le.fit_transform(sibSp)
parch_encoded=le.fit_transform(parch)
fare_encoded=le.fit_transform(fare)
#cabin_encoded=le.fit_transform(cabin)
embarked_encoded=le.fit_transform(embarked)
survived_encoded=le.fit_transform(survived)
```

```
data=list(zip(pclass_encoded,sex_encoded,age_encoded,sibSp_encoded,parch_encoded,embarked_encoded,fare_encoded))#,cabin_encoded))
print(data)
```

```
[(2, 1, 28, 1, 0, 2, 16), (0, 0, 51, 1, 0, 0, 180), (2, 0, 34, 0, 0, 2, 32), (0, 0, 47, 1, 0, 2, 163), (2, 1, 47, 0, 0, 2, 34), (0,
```

```
label=le.fit_transform(survived_encoded)
print(label)
```

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

```
from sklearn.model_selection import train_test_split
```

```
x_train, x_test, y_train, y_test = train_test_split(data, label, test_size=570)
```

```
for i in range (3,38):
    neigh = KNeighborsClassifier(n_neighbors=i)
    neigh.fit(x_train,y_train)
    neigh.score(x_test,y_test)
    print(i, " ",neigh.score(x_test,y_test))
```

```
3  0.6614035087719298
4  0.656140350877193
5  0.6543859649122807
6  0.6736842105263158
7  0.6684210526315789
8  0.6578947368421053
9  0.6614035087719298
10 0.6719298245614035
11 0.6859649122807018
12 0.6684210526315789
13 0.6719298245614035
14 0.656140350877193
15 0.6701754385964912
16 0.6701754385964912
17 0.6736842105263158
18 0.6789473684210526
19 0.6684210526315789
20 0.6754385964912281
21 0.6754385964912281
22 0.6736842105263158
23 0.6736842105263158
24 0.6859649122807018
25 0.6807017543859649
26 0.6719298245614035
27 0.6701754385964912
28 0.6736842105263158
29 0.6701754385964912
30 0.6666666666666666
31 0.6631578947368421
32 0.6649122807017543
33 0.6649122807017543
34 0.6526315789473685
35 0.6543859649122807
36 0.6491228070175439
37 0.6473684210526316
```

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0.7901591895803184
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