

## 21522351\_Tuan3\_Cau4

April 11, 2024

```
[ ]: %matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import seaborn as sns

from sklearn.model_selection import train_test_split
from sklearn import tree
from sklearn.naive_bayes import GaussianNB
from sklearn import metrics
```

```
[ ]: df = pd.read_csv('Data/winequality-red.csv')
```

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

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 1599 entries, 0 to 1598

Data columns (total 12 columns):

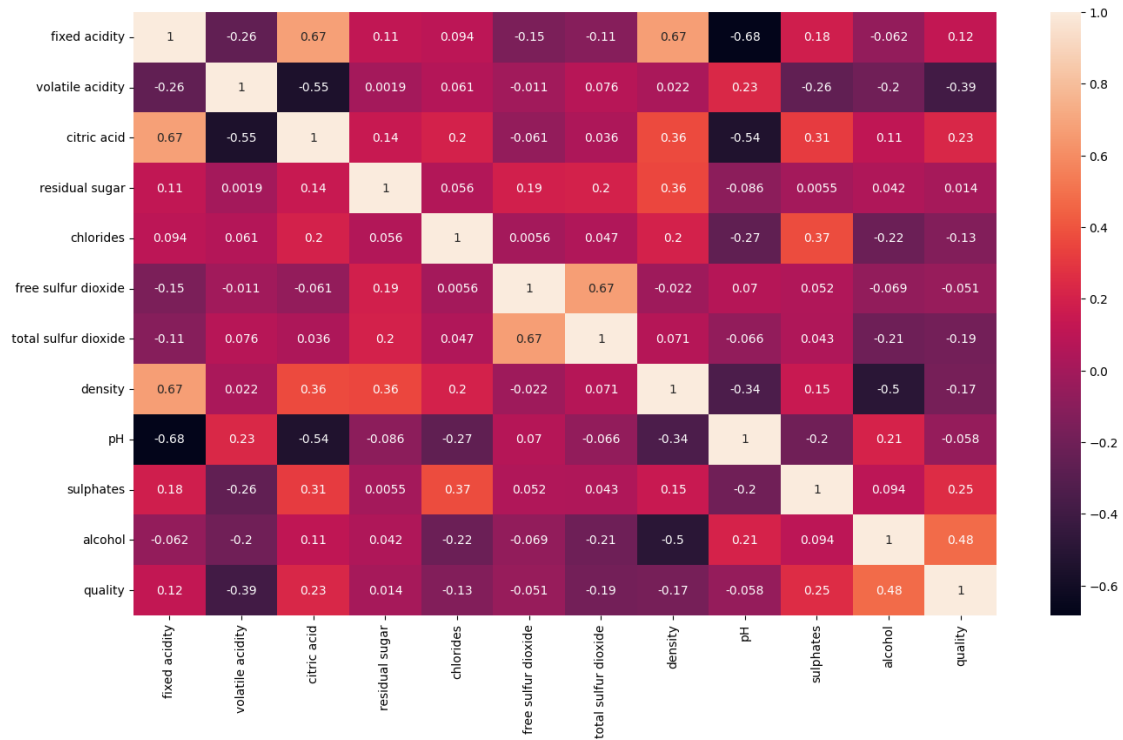
#	Column	Non-Null Count	Dtype
0	fixed acidity	1599 non-null	float64
1	volatile acidity	1599 non-null	float64
2	citric acid	1599 non-null	float64
3	residual sugar	1599 non-null	float64
4	chlorides	1599 non-null	float64
5	free sulfur dioxide	1599 non-null	float64
6	total sulfur dioxide	1599 non-null	float64
7	density	1599 non-null	float64
8	pH	1599 non-null	float64
9	sulphates	1599 non-null	float64
10	alcohol	1599 non-null	float64
11	quality	1599 non-null	int64

dtypes: float64(11), int64(1)

memory usage: 150.0 KB

```
[ ]: plt.figure(figsize=(16,9))
sns.heatmap(df.corr(method='pearson'),annot=True)
```

```
[ ]: <Axes: >
```



```
[ ]: feature = df.drop('quality',axis=1)
label = df['quality']
```

```
[ ]: feature.select_dtypes(exclude=['int64']).columns
```

```
[ ]: Index(['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',
          'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density',
          'pH', 'sulphates', 'alcohol'],
          dtype='object')
```

```
[ ]: feature_onehot = pd.get_dummies(feature,columns=feature.
    ↪select_dtypes(exclude=['int64']).columns)
feature_onehot
```

```
[ ]:
    fixed acidity_4.6  fixed acidity_4.7  fixed acidity_4.9 \
0                False                False                False
1                False                False                False
2                False                False                False
3                False                False                False
4                False                False                False
...                ...                ...                ...
```

1594	False	False	False
1595	False	False	False
1596	False	False	False
1597	False	False	False
1598	False	False	False

	fixed acidity_5.0	fixed acidity_5.1	fixed acidity_5.2	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	
...	...	...	...	
1594	False	False	False	
1595	False	False	False	
1596	False	False	False	
1597	False	False	False	
1598	False	False	False	

	fixed acidity_5.3	fixed acidity_5.4	fixed acidity_5.5	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	
...	...	...	...	
1594	False	False	False	
1595	False	False	False	
1596	False	False	False	
1597	False	False	False	
1598	False	False	False	

	fixed acidity_5.6	...	alcohol_13.0	alcohol_13.1	alcohol_13.2	\
0	False	...	False	False	False	
1	False	...	False	False	False	
2	False	...	False	False	False	
3	False	...	False	False	False	
4	False	...	False	False	False	
...	...	...	...	...	...	
1594	False	...	False	False	False	
1595	False	...	False	False	False	
1596	False	...	False	False	False	
1597	False	...	False	False	False	
1598	False	...	False	False	False	

	alcohol_13.3	alcohol_13.4	alcohol_13.5	alcohol_13.566666666666698	\
0	False	False	False	False	

1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
...	...	...	...	...
1594	False	False	False	False
1595	False	False	False	False
1596	False	False	False	False
1597	False	False	False	False
1598	False	False	False	False

	alcohol_13.6	alcohol_14.0	alcohol_14.9
0	False	False	False
1	False	False	False
2	False	False	False
3	False	False	False
4	False	False	False
...	...	...	...
1594	False	False	False
1595	False	False	False
1596	False	False	False
1597	False	False	False
1598	False	False	False

[1599 rows x 1453 columns]

```
[ ]: #câu 6
x_train,x_test,y_train,y_test = train_test_split(feature,label,test_size=0.
↪3,random_state=42)
```

```
[ ]: clf = tree.DecisionTreeClassifier(criterion="entropy",random_state=0)
clf.fit(x_train,y_train)
```

```
[ ]: DecisionTreeClassifier(criterion='entropy', random_state=0)
```

```
[ ]: tree_pred = clf.predict(x_test)

tree_score = metrics.accuracy_score(y_test,tree_pred)
print("Accruracy:",tree_score)
print("Report:",metrics.classification_report(y_test,tree_pred))
```

Accruracy: 0.5833333333333334

Report:                    precision      recall    f1-score    support

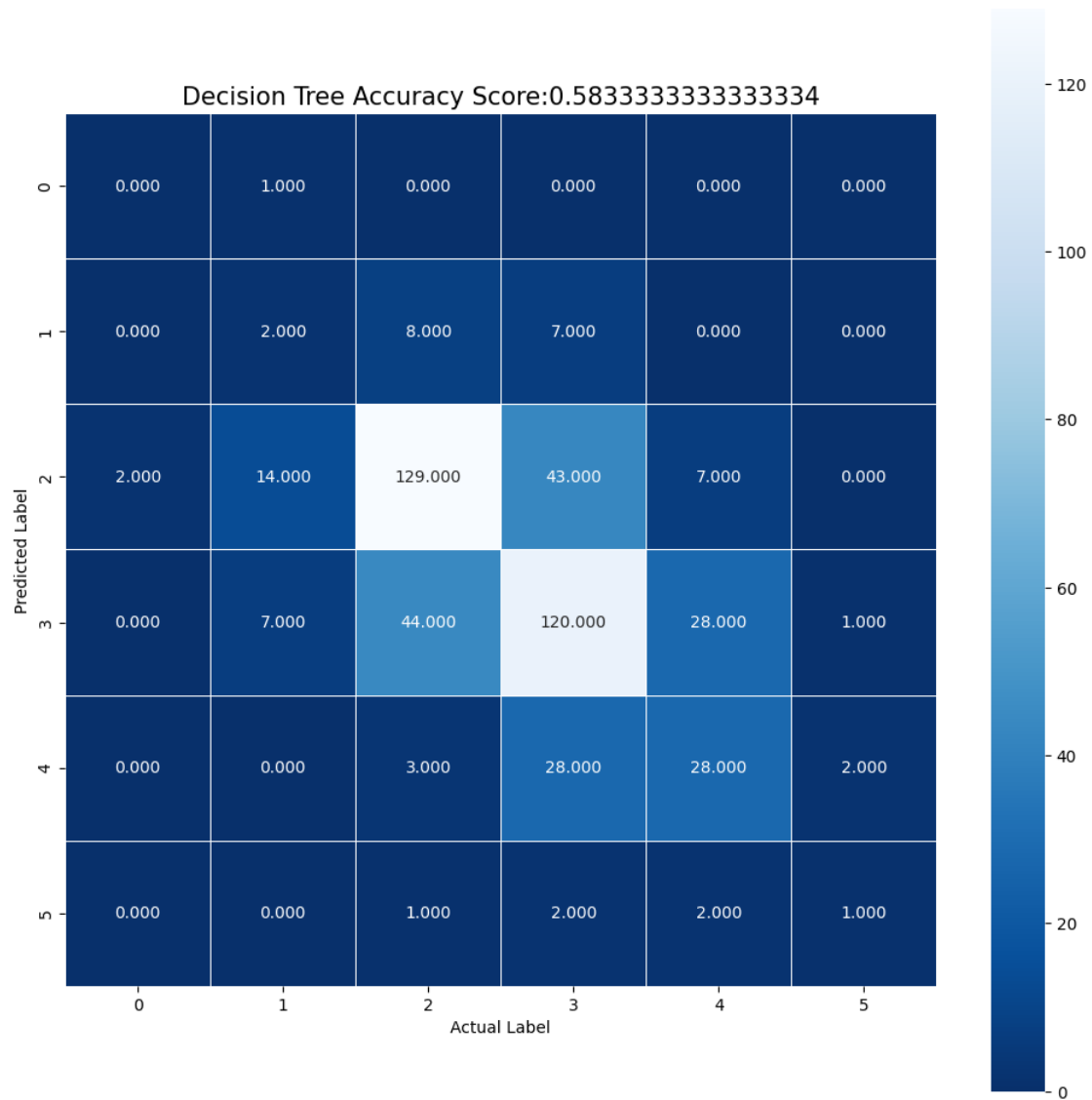
3	0.00	0.00	0.00	1
4	0.08	0.12	0.10	17
5	0.70	0.66	0.68	195
6	0.60	0.60	0.60	200

7	0.43	0.46	0.44	61
8	0.25	0.17	0.20	6
accuracy			0.58	480
macro avg	0.34	0.33	0.34	480
weighted avg	0.59	0.58	0.59	480

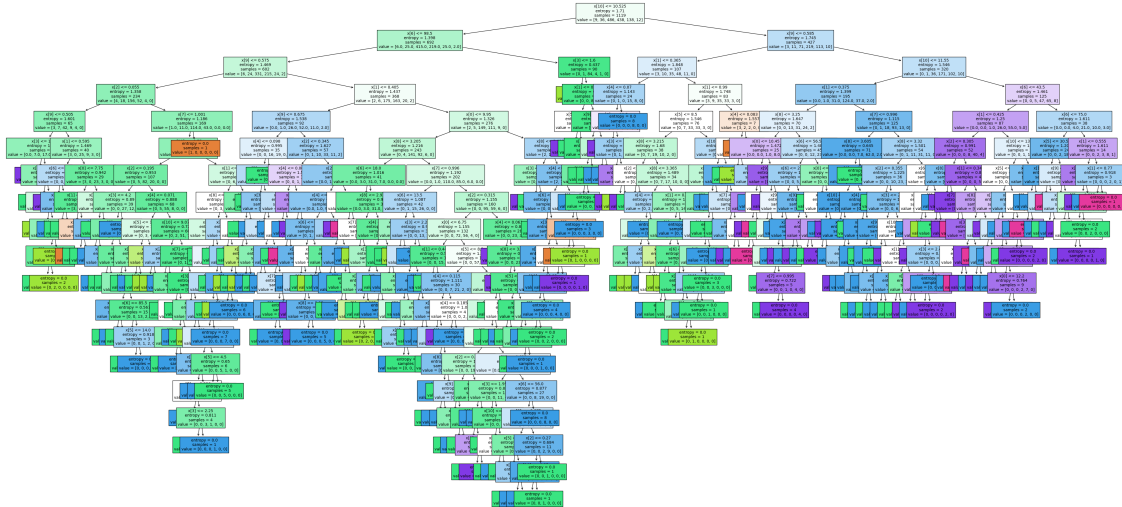
```
[ ]: tree_cm = metrics.confusion_matrix(y_test,tree_pred)
```

```
[ ]: plt.figure(figsize=(12,12))
sns.heatmap(tree_cm,annot=True, fmt=".3f",linewidth=.
↪5,square=True,cmap='Blues_r')
plt.xlabel('Actual Label')
plt.ylabel('Predicted Label')
title = 'Decision Tree Accuracy Score:{0}'.format(tree_score)
plt.title(title,size=15)
```

```
[ ]: Text(0.5, 1.0, 'Decision Tree Accuracy Score:0.5833333333333334')
```



```
[ ]: fig, ax = plt.subplots(figsize=(50,24))
tree.plot_tree(clf,filled=True,fontsize=10)
plt.savefig('decision_tree',dpi=100)
plt.show()
```



```
[ ]: #câu 8
clf = tree.DecisionTreeClassifier(criterion="gini",random_state=0)
clf.fit(x_train,y_train)
```

```
[ ]: DecisionTreeClassifier(random_state=0)
```

```
[ ]: tree_pred = clf.predict(x_test)
tree_score = metrics.accuracy_score(y_test,tree_pred)
print("Accuracy:",tree_score)
print("Report:",metrics.classification_report(y_test,tree_pred))
```

Accuracy: 0.5625

Report:		precision	recall	f1-score	support
	3	0.00	0.00	0.00	1
	4	0.07	0.06	0.06	17
	5	0.64	0.62	0.63	195
	6	0.57	0.58	0.57	200
	7	0.45	0.49	0.47	61
	8	0.33	0.33	0.33	6
	accuracy			0.56	480
	macro avg	0.34	0.35	0.35	480
	weighted avg	0.56	0.56	0.56	480

```
[ ]: tree_cm = metrics.confusion_matrix(y_test,tree_pred)
```

```
[ ]: plt.figure(figsize=(12,12))
```

```

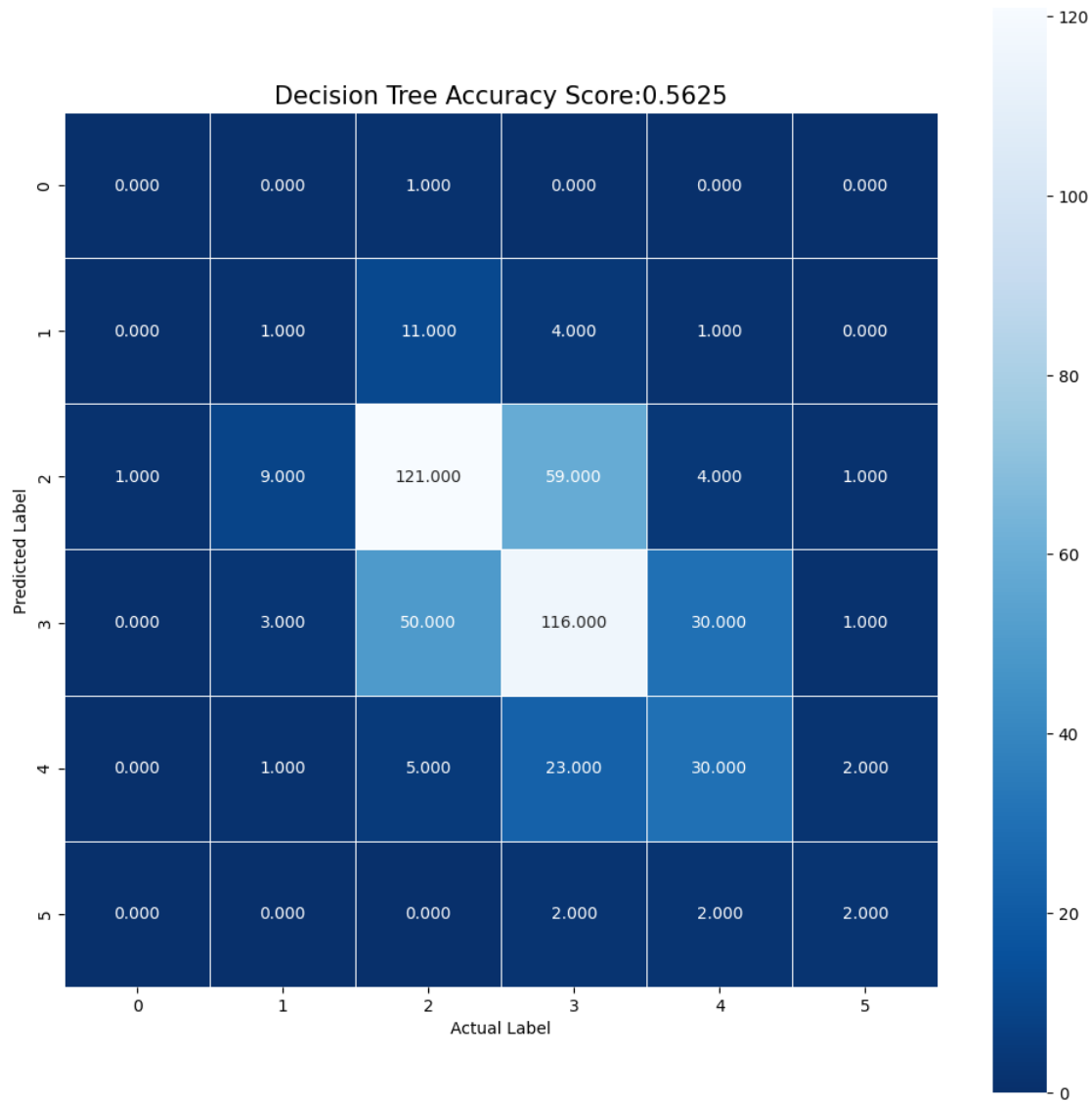
sns.heatmap(tree_cm,annot=True, fmt=".3f",linewidth=.
↪5,square=True,cmap='Blues_r')
plt.xlabel('Actual Label')
plt.ylabel('Predicted Label')
title = 'Decision Tree Accuracy Score:{0}'.format(tree_score)
plt.title(title,size=15)

```

```

[ ]: Text(0.5, 1.0, 'Decision Tree Accuracy Score:0.5625')

```



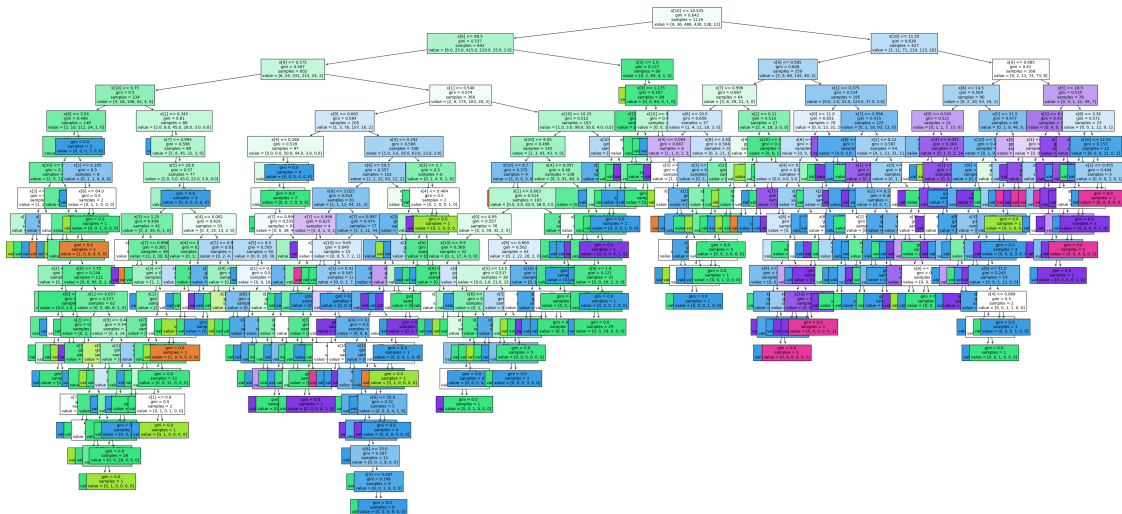
```

[ ]: fig, ax = plt.subplots(figsize=(50,24))
tree.plot_tree(clf,filled=True,fontsize=10)
plt.savefig('decision_tree',dpi=100)

```



```
plt.show()
```



```
[ ]: #câu 9
gnb = GaussianNB()
bayes_pred = gnb.fit(x_train, y_train).predict(x_test)
```

```
[ ]: bayes_score = metrics.accuracy_score(y_test, bayes_pred)
print("Accuracy: ", bayes_score)
print("Report: ", metrics.classification_report(y_test, bayes_pred))
```

Accuracy: 0.5416666666666666

Report:		precision	recall	f1-score	support
	3	0.00	0.00	0.00	1
	4	0.12	0.12	0.12	17
	5	0.68	0.62	0.65	195
	6	0.52	0.54	0.53	200
	7	0.40	0.49	0.44	61
	8	0.00	0.00	0.00	6
	accuracy			0.54	480
	macro avg	0.29	0.29	0.29	480
	weighted avg	0.55	0.54	0.54	480

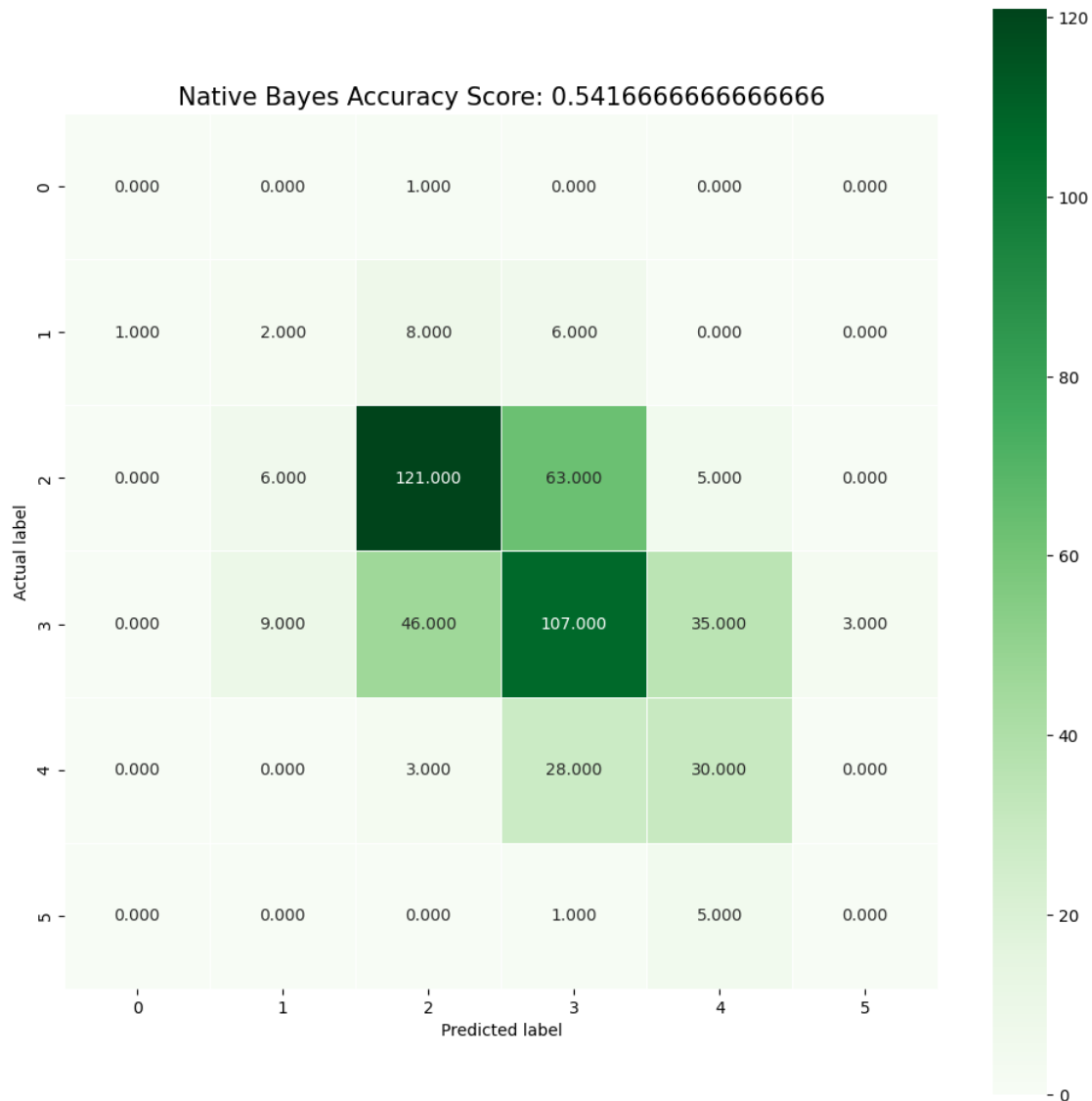
```
[ ]: bayes_cm = metrics.confusion_matrix(y_test, bayes_pred)
plt.figure(figsize=(12,12))
sns.heatmap(bayes_cm,annot=True, fmt=".3f",linewidth=.
```

↪5,square=True,cmap='Greens')

```
plt.ylabel('Actual label')
```

```
plt.xlabel('Predicted label')
title = 'Native Bayes Accuracy Score: {0}'.format(bayes_score)
plt.title(title, size=15)
```

```
[ ]: Text(0.5, 1.0, 'Native Bayes Accuracy Score: 0.5416666666666666')
```



```
[ ]: #Câu 10: So sánh kết quả các mô hình
      # Độ chính xác của các thuật toán
      # Thuật toán cây ID3: 58.34%
      # Thuật toán CART: 56.25%
      # Thuật toán cây Naive Bayes: 54.167%
      # => thuật toán ID3 cho độ chính xác tốt nhất
```