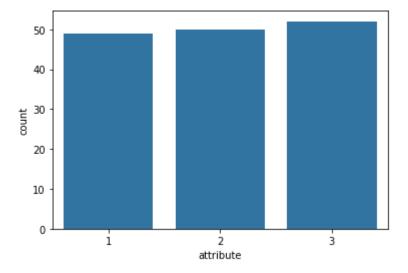
```
In [1]:
        import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
        df=pd.read_csv('tae.data',names=['speaker', 'instructor', 'course', 'semester', 'class
In [2]:
        df.head()
Out[2]:
           speaker instructor course semester class_size attribute
                                                             3
                 1
                         23
                                  3
                                                   19
         0
         1
                 2
                          15
                                  3
                                                   17
                                                             3
        2
                 1
                                           2
                                                   49
                                                             3
                         23
                                  3
        3
                 1
                                           2
                                                             3
                          5
                                  2
                                                   33
         4
                 2
                          7
                                           2
                                                   55
                                                             3
                                 11
        df.isna().sum()
In [3]:
                       0
        speaker
Out[3]:
        instructor
                       0
        course
                       0
        semester
                       0
        class_size
                       0
        attribute
        dtype: int64
        df.info()
In [4]:
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 151 entries, 0 to 150
        Data columns (total 6 columns):
                          Non-Null Count Dtype
         #
             Column
             ----
                          -----
                                          ----
         0
                                           int64
             speaker
                          151 non-null
             instructor 151 non-null
                                           int64
         1
         2
                          151 non-null
             course
                                           int64
         3
             semester
                          151 non-null
                                           int64
             class size 151 non-null
                                           int64
         5
             attribute
                          151 non-null
                                           int64
        dtypes: int64(6)
        memory usage: 7.2 KB
        df.describe()
In [5]:
```

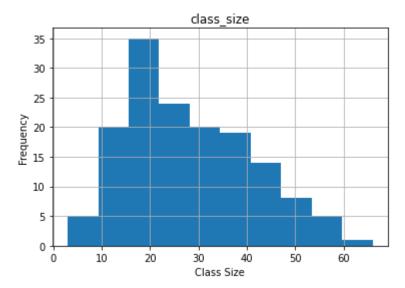
Out[5]:

	speaker	instructor	course	semester	class_size	attribute
count	151.000000	151.000000	151.000000	151.000000	151.000000	151.000000
mean	1.807947	13.642384	8.105960	1.847682	27.867550	2.019868
std	0.395225	6.825779	7.023914	0.360525	12.893758	0.820327
min	1.000000	1.000000	1.000000	1.000000	3.000000	1.000000
25%	2.000000	8.000000	3.000000	2.000000	19.000000	1.000000
50%	2.000000	13.000000	4.000000	2.000000	27.000000	2.000000
75%	2.000000	20.000000	15.000000	2.000000	37.000000	3.000000
max	2.000000	25.000000	26.000000	2.000000	66.000000	3.000000



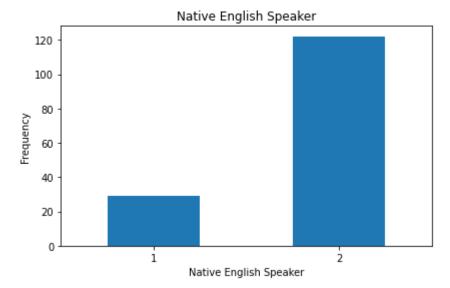
```
In [7]: numerical_columns = ['class_size']

for column in numerical_columns:
    df[column].hist()
    plt.title(column)
    plt.xlabel('Class Size')
    plt.ylabel('Frequency')
    plt.show()
```

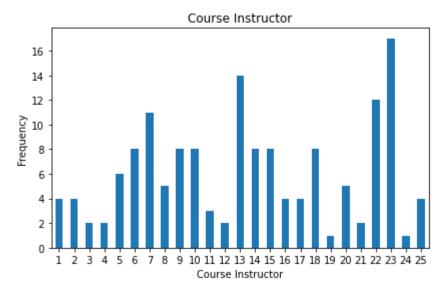


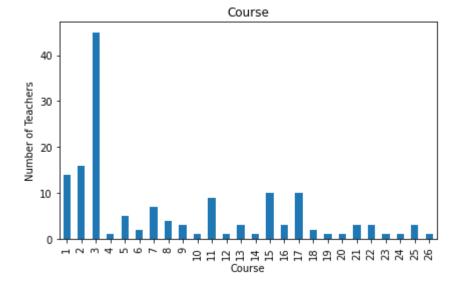
```
In [8]: instructor_counts = df['speaker'].value_counts().sort_index()

instructor_counts.plot(kind='bar')
plt.title('Native English Speaker')
plt.xlabel('Native English Speaker')
plt.ylabel('Frequency')
plt.xticks(rotation=0)
plt.tight_layout()
plt.show()
```

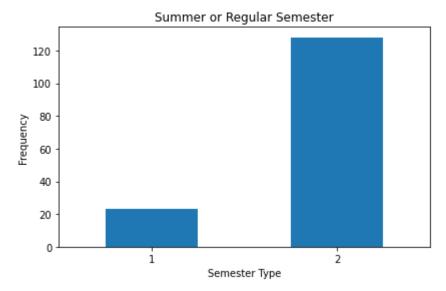


```
In [9]: instructor_counts = df['instructor'].value_counts().sort_index()
    instructor_counts.plot(kind='bar')
    plt.title('Course Instructor')
    plt.xlabel('Course Instructor')
    plt.ylabel('Frequency')
    plt.xticks(rotation=0)
    plt.tight_layout()
    plt.show()
```



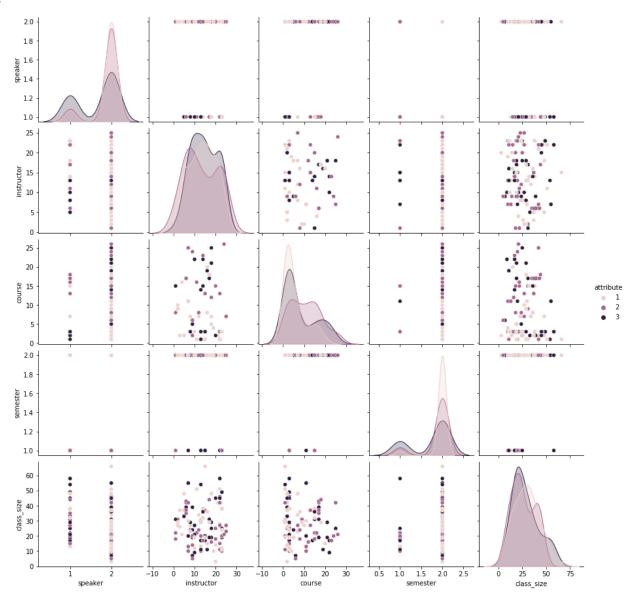


```
In [11]:
    semester_counts = df['semester'].value_counts().sort_index()
    semester_counts.plot(kind='bar')
    plt.title('Summer or Regular Semester')
    plt.xlabel('Semester Type')
    plt.ylabel('Frequency')
    plt.xticks(rotation=0)
    plt.tight_layout()
    plt.show()
```



In [12]: sns.pairplot(df,hue='attribute')

Out[12]: <seaborn.axisgrid.PairGrid at 0x244425aae20>



```
Task 3 – Data Modelling
          x=df.drop('attribute',axis=1)
In [13]:
          x.sample(5)
Out[13]:
              speaker instructor course semester class_size
                   2
                                             2
                                                      27
          46
                            10
                                    3
                   2
                            22
                                             2
                                                      46
          17
          57
                   2
                            13
                                    3
                                             1
                                                      10
          62
                   2
                                   15
                                                      22
          45
                   2
                             9
                                    5
                                             2
                                                      19
          y=df['attribute']
In [14]:
          y.sample(5)
                 2
Out[14]:
          115
                 1
          83
                 3
          17
                 2
          107
          Name: attribute, dtype: int64
          Split dataset to training and testing
In [15]:
         from sklearn.model_selection import train_test_split
```

```
In [15]: from sklearn.model_selection import train_test_split

from sklearn.metrics import accuracy_score,confusion_matrix
from sklearn.metrics import plot_confusion_matrix
from sklearn.metrics import classification_report

from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
```

Suite1: 50% for training and 50% for testing

```
In [16]: x_train, x_test, y_train, y_test =train_test_split(x,y,test_size=0.5)
```

Support Vector Machine(SVM)

```
In [18]: pred=model.predict(x_test)
    pred
```

```
Out[18]:
                 1, 2, 2, 2, 2, 2, 1, 2, 2, 1, 2, 2, 2, 2, 1, 2, 2, 1, 2, 2, 1,
                 2, 2, 2, 2, 2, 2, 2, 1, 1, 2, 1, 1, 2, 1, 1, 1, 1, 2, 1, 2, 2,
                 1, 2, 1, 2, 2, 1, 2, 2, 2], dtype=int64)
In [19]:
          y_test
                 3
          11
Out[19]:
          91
                 3
          2
                 3
          36
                 1
          125
                 3
                 3
          47
          132
                 2
          43
                 3
          96
                 2
          124
                 3
          Name: attribute, Length: 76, dtype: int64
          Classification accuracy
In [20]:
          accuracy_score(y_test,pred)
          0.42105263157894735
Out[20]:
          confusion_matrix(y_test,pred)
In [21]:
          array([[11, 14, 0],
Out[21]:
                 [ 2, 21, 0],
                 [12, 16, 0]], dtype=int64)
In [22]: s=sns.heatmap(confusion_matrix(y_test,pred), annot = True)
          s.set_title('Confusion Matrix')
          s.set xlabel('Predicted labels', fontsize=10)
          s.set ylabel('True labels', fontsize=10)
          Text(33.0, 0.5, 'True labels')
Out[22]:
                           Confusion Matrix
                                                           -20.0
                     11
                                                0
                                                           - 17.5
            0
                                                           - 15.0
                                                           - 12.5
          Frue labels
                     2
                                  21
                                                0
                                                           - 10.0
                                                           - 7.5
                                                           - 5.0
                                                0
            2
                                                            2.5
```

In [23]: report=pd.DataFrame(classification_report(y_test,pred,output_dict=True))
 report

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- 0.0

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Predicted labels

C:\Users\user\AppData\Roaming\Python\Python39\site-packages\sklearn\metrics_classifi cation.py:1334: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

C:\Users\user\AppData\Roaming\Python\Python39\site-packages\sklearn\metrics_classifi cation.py:1334: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

C:\Users\user\AppData\Roaming\Python\Python39\site-packages\sklearn\metrics_classifi cation.py:1334: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

76.000000

76.000000

warn prf(average, modifier, msg start, len(result))

Out[23]:		1	2	3	accuracy	macro avg	weighted avg
	precision	0.44	0.411765	0.0	0.421053	0.283922	0.269350
	recall	0.44	0.913043	0.0	0.421053	0.451014	0.421053
	f1-score	0.44	0.567568	0.0	0.421053	0.335856	0.316501

support 25.00 23.000000 28.0 0.421053

Decision tree

```
In [24]: model = DecisionTreeClassifier()
model.fit(x_train,y_train)
```

```
Out[24]: • DecisionTreeClassifier

DecisionTreeClassifier()
```

```
In [25]: pred=model.predict(x_test)
pred
```

```
Out[25]: array([1, 2, 3, 2, 3, 2, 3, 1, 1, 2, 2, 3, 2, 2, 2, 1, 2, 3, 2, 2, 3, 2, 2, 3, 1, 1, 1, 1, 1, 1, 1, 2, 1, 2, 3, 1, 2, 2, 3, 3, 1, 3, 3, 1, 3, 2, 3, 1, 2, 2, 1, 1, 3, 2, 2, 3, 1, 3, 2, 1], dtype=int64)
```

```
In [26]:
           y_test
                   3
Out[26]:
           91
                   3
           2
                   3
                   1
           36
           125
                   3
                  . .
           47
                   3
           132
                   2
           43
                   3
           96
                   2
```

Name: attribute, Length: 76, dtype: int64

124

Classification accuracy

```
accuracy_score(y_test,pred)
In [27]:
          0.6052631578947368
Out[27]:
In [28]:
          confusion_matrix(y_test,pred)
          array([[17, 6, 2],
Out[28]:
                 [3, 14, 6],
                 [ 5, 8, 15]], dtype=int64)
          s=sns.heatmap(confusion_matrix(y_test,pred), annot = True)
In [29]:
          s.set title('Confusion Matrix')
          s.set_xlabel('Predicted labels', fontsize=10)
          s.set_ylabel('True labels', fontsize=10)
          Text(33.0, 0.5, 'True labels')
Out[29]:
                           Confusion Matrix
                                                            - 16
                     17
                                   6
                                                2
                                                            - 14
                                                            - 12
          Frue labels
                                                            - 10
                     3
                                  14
                                                6
                                                            - 8
                                                             6
                                                15
                     ò
                                                ż
                                   1
                             Predicted labels
In [30]:
          report=pd.DataFrame(classification report(y test,pred,output dict=True))
          report
```

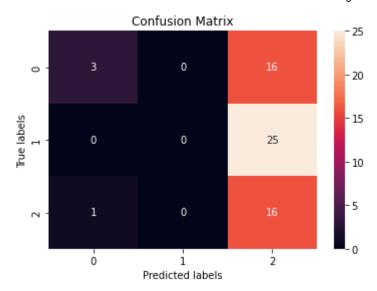
Out[30]:		1	2	3	accuracy	macro avg	weighted avg
	precision	0.68	0.500000	0.652174	0.605263	0.610725	0.615275
	recall	0.68	0.608696	0.535714	0.605263	0.608137	0.605263
	f1-score	0.68	0.549020	0.588235	0.605263	0.605752	0.606553
	support	25.00	23.000000	28.000000	0.605263	76.000000	76.000000

Suite2: 60% for training and 40% for testing

```
In [31]: x_train, x_test, y_train, y_test =train_test_split(x,y,test_size=0.4)
```

Support Vector Machine(SVM)

```
model=SVC()
In [32]:
        model.fit(x_train,y_train)
Out[32]:
        ▼ SVC
        SVC()
        pred=model.predict(x_test)
In [33]:
        pred
        Out[33]:
              1, 3, 3, 3, 3, 3, 3, 3, 3, 3, 1, 1], dtype=int64)
In [34]:
        y_test
        80
              3
Out[34]:
              2
        18
        113
             1
        119
             1
        69
             1
             1
        34
        46
             3
        75
             1
        76
             1
        126
              3
        Name: attribute, Length: 61, dtype: int64
        Classification accuracy
        accuracy_score(y_test,pred)
In [35]:
        0.3114754098360656
Out[35]:
        confusion matrix(y test,pred)
In [36]:
        array([[ 3, 0, 16],
Out[36]:
              [0, 0, 25],
              [ 1, 0, 16]], dtype=int64)
In [37]: s=sns.heatmap(confusion_matrix(y_test,pred), annot = True)
        s.set_title('Confusion Matrix')
        s.set xlabel('Predicted labels', fontsize=10)
        s.set_ylabel('True labels', fontsize=10)
        Text(33.0, 0.5, 'True labels')
Out[37]:
```



In [38]: report=pd.DataFrame(classification_report(y_test,pred,output_dict=True))
 report

C:\Users\user\AppData\Roaming\Python\Python39\site-packages\sklearn\metrics_classifi cation.py:1334: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

C:\Users\user\AppData\Roaming\Python\Python39\site-packages\sklearn\metrics_classifi cation.py:1334: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

C:\Users\user\AppData\Roaming\Python\Python39\site-packages\sklearn\metrics_classifi cation.py:1334: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

Out[38]:

	1	2	3	accuracy	macro avg	weighted avg
precision	0.750000	0.0	0.280702	0.311475	0.343567	0.311835
recall	0.157895	0.0	0.941176	0.311475	0.366357	0.311475
f1-score	0.260870	0.0	0.432432	0.311475	0.231101	0.201768
support	19.000000	25.0	17.000000	0.311475	61.000000	61.000000

Decision tree

In [39]: model = DecisionTreeClassifier()
model.fit(x_train,y_train)

Out[39]: • DecisionTreeClassifier

DecisionTreeClassifier()

In [40]: pred=model.predict(x_test)
pred

```
array([1, 2, 1, 1, 1, 2, 3, 1, 3, 2, 2, 1, 1, 2, 1, 1, 2, 2, 1, 3, 1, 1,
Out[40]:
                 1, 3, 2, 3, 1, 3, 3, 1, 3, 2, 2, 2, 3, 2, 1, 2, 3, 2, 1, 1, 2,
                 1, 2, 2, 3, 3, 2, 3, 2, 1, 2, 1, 3, 1, 3, 1, 1, 3], dtype=int64)
         y_test
In [41]:
                 3
          80
Out[41]:
          18
                 2
                 1
          113
          119
                 1
          69
                 1
          34
                 1
          46
                 3
          75
                 1
          76
                 1
          126
                 3
          Name: attribute, Length: 61, dtype: int64
          accuracy_score(y_test,pred)
In [42]:
          0.6229508196721312
Out[42]:
          confusion_matrix(y_test,pred)
In [43]:
          array([[14, 2, 3],
Out[43]:
                 [7, 14, 4],
                 [ 3, 4, 10]], dtype=int64)
In [44]: s=sns.heatmap(confusion_matrix(y_test,pred), annot = True)
          s.set_title('Confusion Matrix')
          s.set_xlabel('Predicted labels', fontsize=10)
          s.set_ylabel('True labels', fontsize=10)
          Text(33.0, 0.5, 'True labels')
Out[44]:
                           Confusion Matrix
                                                          - 14
                                  2
                                               3
                    14
                                                          - 12
            0
                                                          - 10
          Frue labels
                                 14
                                               4
                                                           - 8
                                                           - 6
                     3
                                  4
```

In [45]: report=pd.DataFrame(classification_report(y_test,pred,output_dict=True))
report

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Predicted labels

Out[45]:

		1	2	3	accuracy	macro avg	weighted avg
	precision	0.583333	0.700000	0.588235	0.622951	0.623856	0.632514
	recall	0.736842	0.560000	0.588235	0.622951	0.628359	0.622951
	f1-score	0.651163	0.622222	0.588235	0.622951	0.620540	0.621765
	support	19.000000	25.000000	17.000000	0.622951	61.000000	61.000000

Suite3: 80% for training and 20% for testing

```
In [46]: x_train, x_test, y_train, y_test =train_test_split(x,y,test_size=0.2)
```

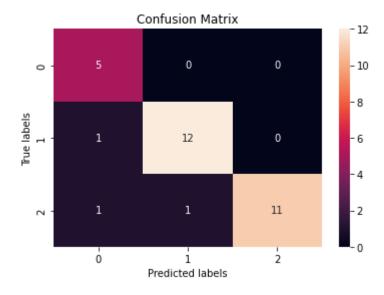
Support Vector Machine(SVM)

```
pred=model.predict(x_test)
In [47]:
          pred
          array([2, 2, 2, 1, 1, 2, 2, 2, 1, 2, 3, 3, 3, 2, 2, 3, 1, 2, 1, 2, 3, 3,
Out[47]:
                  1, 2, 2, 3, 3, 1, 3, 3, 3], dtype=int64)
In [48]:
          y_test
                  2
Out[48]:
                  2
          103
                  2
          134
          34
                  1
          114
                  1
          92
                  3
          99
                  2
          18
                  2
          148
                  1
          59
                  2
          45
                  3
          2
                  3
                  3
          41
          65
                  2
                  2
          97
          1
                  3
          29
                  1
          55
                  2
          120
                  1
          64
                  2
                  3
          8
          13
                  3
          88
                  3
          14
                  2
                  2
          54
          39
                  3
          93
                  3
          95
                  2
          44
                  3
          43
                  3
          52
          Name: attribute, dtype: int64
```

Classification accuracy

```
In [49]:
         accuracy_score(y_test,pred)
         0.9032258064516129
Out[49]:
In [50]:
         confusion_matrix(y_test,pred)
         array([[ 5, 0, 0],
Out[50]:
                [ 1, 12, 0],
                [ 1, 1, 11]], dtype=int64)
In [51]: s=sns.heatmap(confusion_matrix(y_test,pred), annot = True)
         s.set title('Confusion Matrix')
         s.set_xlabel('Predicted labels', fontsize=10)
         s.set_ylabel('True labels', fontsize=10)
         Text(33.0, 0.5, 'True labels')
```

Out[51]:



report=pd.DataFrame(classification_report(y_test,pred,output_dict=True)) In [52]: report

Out[52]:		1	2	3	accuracy	macro avg	weighted avg
	precision	0.714286	0.923077	1.000000	0.903226	0.879121	0.921659
	recall	1.000000	0.923077	0.846154	0.903226	0.923077	0.903226
	f1-score	0.833333	0.923077	0.916667	0.903226	0.891026	0.905914
	support	5.000000	13.000000	13.000000	0.903226	31.000000	31.000000

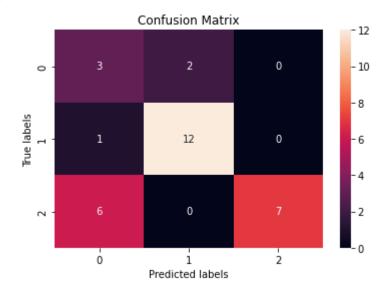
Decision tree

```
model = DecisionTreeClassifier()
In [53]:
         model.fit(x_train,y_train)
```

```
Out[53]:
          ▼ DecisionTreeClassifier
         DecisionTreeClassifier()
          pred=model.predict(x test)
In [54]:
          pred
         array([2, 2, 2, 1, 2, 3, 1, 2, 2, 2, 3, 1, 1, 2, 2, 3, 1, 2, 1, 2, 3, 1,
Out[54]:
                 1, 2, 2, 3, 1, 2, 3, 3, 1], dtype=int64)
In [55]:
         y_test
                 2
         129
Out[55]:
         103
                 2
                 2
         134
                 1
         34
         114
                 1
         92
                 3
                 2
         99
                 2
         18
         148
                 1
         59
                 2
         45
                 3
                 3
         2
                 3
         41
                 2
         65
         97
                 2
         1
                 3
                 1
         29
         55
                 2
         120
                 1
                 2
         64
         8
                 3
                 3
         13
                 3
         88
         14
                 2
                 2
          54
                 3
         39
         93
                 3
         95
                 2
         44
                 3
         43
                 3
         52
         Name: attribute, dtype: int64
         Classification accuracy
In [56]:
         accuracy_score(y_test,pred)
         0.7096774193548387
Out[56]:
          confusion_matrix(y_test,pred)
In [57]:
         array([[ 3, 2, 0],
Out[57]:
                 [ 1, 12, 0],
                 [ 6, 0, 7]], dtype=int64)
         s=sns.heatmap(confusion_matrix(y_test,pred), annot = True)
In [58]:
          s.set_title('Confusion Matrix')
```

```
s.set_xlabel('Predicted labels', fontsize=10)
s.set_ylabel('True labels', fontsize=10)
```

Out[58]: Text(33.0, 0.5, 'True labels')



In [59]: report=pd.DataFrame(classification_report(y_test,pred,output_dict=True))
 report

Out[59]:		1	2	3	accuracy	macro avg	weighted avg
	precision	0.3	0.857143	1.000000	0.709677	0.719048	0.827189
	recall	0.6	0.923077	0.538462	0.709677	0.687179	0.709677
	f1-score	0.4	0.888889	0.700000	0.709677	0.662963	0.730824
	support	5.0	13.000000	13.000000	0.709677	31.000000	31.000000

Tn [].