

project 03

April 2, 2020

```
[47]: #Suja Basnet
      #project 03

      from sklearn.datasets import load_iris
      from sklearn.model_selection import KFold
      from sklearn import metrics
      from sklearn.linear_model import LogisticRegression
      from sklearn.metrics import classification_report
      from sklearn.metrics import accuracy_score
      from sklearn.model_selection import cross_val_score
      import numpy as np
      import pandas as pd
      from sklearn.model_selection import train_test_split
```

```
[48]: iris_data = load_iris()
```

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[49]: data_input = iris_data.data
      data_output = iris_data.target
      print(data_output)
```

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[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 2 2 2 2 2 2 2 2 2
 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
 2 2]
```

```
[50]: vec = np.arange(0,15)
      kf = KFold(n_splits =10, shuffle=True)
      print("Train Set                               Test Set ")
      for train_set,test_set in kf.split(vec):
          print(train_set, test_set)
```

Train Set	Test Set
[0 1 3 4 5 6 7 8 10 11 12 13 14]	[2 9]
[0 1 2 3 4 5 6 7 8 9 10 12 14]	[11 13]
[0 1 2 4 5 6 7 8 9 11 12 13 14]	[3 10]
[1 2 3 4 5 6 7 9 10 11 12 13 14]	[0 8]
[0 1 2 3 5 6 7 8 9 10 11 12 13]	[4 14]

```
[ 0  1  2  3  4  5  7  8  9 10 11 12 13 14] [6]
[ 0  1  2  3  4  6  7  8  9 10 11 12 13 14] [5]
[ 0  1  2  3  4  5  6  7  8  9 10 11 13 14] [12]
[ 0  1  2  3  4  5  6  8  9 10 11 12 13 14] [7]
[ 0  2  3  4  5  6  7  8  9 10 11 12 13 14] [1]
```

```
[51]: #Split the data into 80% training and 20% testing
x_train, x_test, y_train, y_test = train_test_split(data_input, data_output,
test_size=0.2, random_state=40)
```

```
[52]: #Train the model
model = LogisticRegression()
model.fit(x_train, y_train) #Training the model
#Test the model
predictions = model.predict(x_test)
print(predictions)# printing predictions
print()# Printing new line
#Check precision, recall, f1-score
print( classification_report(y_test, predictions) )
print( accuracy_score(y_test, predictions))
# import warnings filter
from warnings import simplefilter
# ignore all future warnings
simplefilter(action='ignore', category=FutureWarning)
import warnings
warnings.filterwarnings('ignore')
```

```
[0 1 2 2 1 2 1 1 1 0 1 0 0 2 1 2 2 2 1 1 2 2 1 0 1 0 0 2 0 1]
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	8
1	1.00	1.00	1.00	12
2	1.00	1.00	1.00	10
accuracy			1.00	30
macro avg	1.00	1.00	1.00	30
weighted avg	1.00	1.00	1.00	30

1.0

```
[53]: from sklearn.metrics import make_scorer, accuracy_score,
precision_score, recall_score, f1_score
2
import numpy as np
from sklearn.metrics import accuracy_score, precision_score, recall_score
from sklearn.model_selection import cross_validate
```

```

kf2 = KFold(n_splits=10, shuffle=True, random_state=52)
scoring = {'acc': 'accuracy',
'prec_macro': 'precision_macro',
'rec_micro': 'recall_macro',
}
scores = cross_validate(model, data_input, data_output, scoring=scoring, cv=kf2,
    ↪return_train_score=True)
print(scores['test_acc'])
print("Mean accuracy: ", cross_val_score(model, data_input, data_output, cv=kf2,
    ↪scoring='accuracy').mean())
print(scores['test_prec_macro'])
print("Mean precision: ", cross_val_score(model, data_input,
    ↪data_output, cv=kf2, scoring='precision_macro').mean())
print(scores['test_rec_micro'])
print("Mean Recall: ", cross_val_score(model, data_input, data_output,
    ↪cv=kf2, scoring='recall_macro').mean())
print("F1score:
    ↪", cross_val_score(model, data_input, data_output, scoring=make_scorer(f1_score, average='weight',
    ↪labels=[2]), cv=kf2))
#precision = cross_val_score(model, data_input, data_output, scoring='precision',
    ↪cv=kf2)
#print ("Precision: " + str(round(100*precision.mean(), 2)) + "%")
#print("Precision: ", cross_val_score(model, data_input,
    ↪data_output, cv=kf2, scoring='precision'))

```

```

[0.93333333 1.          0.93333333 1.          0.93333333 0.86666667
 1.          1.          0.93333333 1.          ]
Mean accuracy: 0.96
[0.96296296 1.          0.91666667 1.          0.88888889 0.88571429
 1.          1.          0.94444444 1.          ]
Mean precision: 0.9598677248677248
[0.88888889 1.          0.95833333 1.          0.95833333 0.88571429
 1.          1.          0.93333333 1.          ]
Mean Recall: 0.9624603174603175
F1score: [0.94117647 1.          0.85714286 1.          0.93333333 0.8
 1.          1.          0.88888889 1.          ]

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