

NEURAL NETWORK DEEP LEARNING

ICP 5

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GitHub:

Repository URL for the source code : <https://github.com/vxr22100/NNPL/tree/main/ICP5>

Zoom Recording:

<https://github.com/vxr22100/NNPL/blob/main/ICP5/ICP5%20Video.mp4>

Question 1

Implement Naïve Bayes method using scikit-learn library

Use dataset available with name glass

Use train_test_split to create training and testing part

Evaluate the model on test part using score and classification_report(y_true, y_pred)

Program & Explanation:

```
In [1]: #importing set of libraries
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import classification_report, accuracy_score
import warnings
warnings.filterwarnings("ignore")
from sklearn import metrics
```

```
In [5]: #importing the given dataset glass.csv
dst_Data = pd.read_csv("glass.csv")
dst_Data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 214 entries, 0 to 213
Data columns (total 10 columns):
#   Column  Non-Null Count  Dtype
---  -
0    RI      214 non-null      float64
1    Na      214 non-null      float64
2    Mg      214 non-null      float64
3    Al      214 non-null      float64
4    Si      214 non-null      float64
5    K       214 non-null      float64
6    Ca      214 non-null      float64
7    Ba      214 non-null      float64
8    Fe      214 non-null      float64
9    Type    214 non-null      int64
dtypes: float64(9), int64(1)
memory usage: 16.8 KB
```

- Importing the set of libraries and csv file and printing information about glass csv dataframe which includes index, columns, non-null values and memory usage.

```
#splitting the dataset which is excluding last columns
X = dst_Data.iloc[:, :-1]
y = dst_Data.iloc[:, -1]
#splitting the dataset into train and test datasets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
#creating a Gaussian Naïve Bayes model
gn = GaussianNB()
#fitting train data
gn.fit(X_train, y_train)
# predicting the test dataset
y_pred = gn.predict(X_test)
# evaluating the model on the test dataset
print("Accuracy: ", accuracy_score(y_test, y_pred)*100)
print("Classification Report: \n", classification_report(y_test, y_pred))
```

```
Accuracy: 37.2093023255814
Classification Report:
              precision    recall  f1-score   support

     1         0.19         0.44         0.27         9
     2         0.33         0.16         0.21        19
     3         0.33         0.20         0.25         5
     5         0.00         0.00         0.00         2
     6         0.67         1.00         0.80         2
     7         1.00         1.00         1.00         6

 accuracy          0.37         0.37         0.37        43
 macro avg         0.42         0.47         0.42        43
 weighted avg         0.40         0.37         0.36        43
```

- Splitting the dataset using iloc function into features(x) and target variable(y). Then the data is split into training and testing sets using the ‘train_test_split’ function.
- Creating a Gaussian Naïve Bayes classifier using the ‘GaussianNB’ class.
- ‘fit’ method is used to train the classifier and ‘predict’ method is used to make predictions on test data.
- Finally evaluating and printing the accuracy and classification report.

Question 2

Implement linear SVM method using scikit library

Use the same dataset above

Use train_test_split to create training and testing part

Evaluate the model on test part using score and classification_report(y_true, y_pred)

Program & Explanation:

```
#importing set of libraries
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import classification_report, accuracy_score
```

```
#loading the glass dataset
dst_Data = pd.read_csv("glass.csv")
dst_Data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 214 entries, 0 to 213
Data columns (total 10 columns):
 #   Column  Non-Null Count  Dtype
---  -
 0    RI      214 non-null    float64
 1    Na      214 non-null    float64
 2    Mg      214 non-null    float64
 3    Al      214 non-null    float64
 4    Si      214 non-null    float64
 5    K       214 non-null    float64
 6    Ca      214 non-null    float64
 7    Ba      214 non-null    float64
 8    Fe      214 non-null    float64
 9    Type    214 non-null    int64
dtypes: float64(9), int64(1)
memory usage: 16.8 KB
```

- Importing the set of libraries and csv file and printing information about glass csv dataframe which includes index, columns, non-null values and memory usage.

```

#splitting the dataset into training and testing datasets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
#creating a linear SVM model
svm = SVC(kernel='linear')
#fitting the training dataset
svm.fit(X_train, y_train)
#predicting the target values using the test dataset
y_pred = svm.predict(X_test)
#evaluating the model on the test dataset
print("Accuracy: ", accuracy_score(y_test, y_pred)*100)
print("Classification Report: \n", classification_report(y_test, y_pred))

```

Accuracy: 51.162790697674424

Classification Report:

	precision	recall	f1-score	support
1	0.36	0.89	0.52	9
2	0.58	0.37	0.45	19
3	0.00	0.00	0.00	5
5	0.50	0.50	0.50	2
6	0.00	0.00	0.00	2
7	0.86	1.00	0.92	6
accuracy			0.51	43
macro avg	0.38	0.46	0.40	43
weighted avg	0.48	0.51	0.46	43

- The data is split into training and testing sets using ‘**train_test_split**’ function.
- Linear SVM classifier is created using ‘svc’ class with parameter ‘kernel’ set to ‘linear’.
- ‘fit’ method is used to train classifier and ‘predict’ method is used to predict data.
- Finally evaluating and printing the accuracy and classification report.

The accuracy of GaussianNB is 37.2, whereas the accuracy of Linear SVM is 51.16, which makes the SVM algorithm an accurate algorithm. As our datasets are linearly separable, we can use Linear SVM. When the datasets are not linearly separable, we can use NB, etc. As our dataset is linear, we got more accuracy for Linear SVM Algorithm compared to Naive Bayes Algorithm.