

A Python Program to implement multilayered Perception with Back Propagation

Aim:-

To implement multilayered perception with back propagation using python

Algorithm :-

Step 1: Import the Necessary Libraries

- * Import pandas as pd

- * Import numpy as np

Step 2: Read and Display the Dataset

- * Use pd.read_csv('banknotes.csv') to read the dataset

- * Assign the result to a variable (eg. data)

- * Display the first ten rows using data.head(10)

Step 3: Display the Data Set Dimensions

- * Use the shape attribute on the dataset (eg. data.shape)

Step 4: Import Train-Test Split Module

- * Import train-test-split from 'sklearn model_selection'

Step 5: Split Dataset with 80-20 Ratio.

- * Assign the features to a variable (eg. x-data.drop(['target']))

- * Assign the target variable to another variable (eg. 'y-data['target'])

- * Use 'train-test-split' to split the dataset

	Image.Var	Image.Skew	Image.Gini	Entropy	Class
0	0.62160	0.6661	-0.80730	0.44690	0
1	4.64800	0.1074	-0.46660	1.46210	0
2	0.06600	-0.6303	1.03420	0.10845	0
3	0.45660	0.6228	-0.01120	0.69440	0
4	0.82024	-4.4682	4.67160	0.96680	0
5	4.36040	0.6718	-0.96000	3.16260	0
6	0.80120	0.0129	0.79080	0.66421	0
7	2.09220	-0.8100	0.46560	0.60216	0
8	3.20320	0.7688	-0.78348	0.61261	0
9	1.69560	0.1772	-0.27160	0.73595	0

* Use train-test-split to split the dataset into training and testing sets with a ratio of 0.2.

* Assign the result to 'x-train', 'x-test', 'y-train', and 'y-test'.

Step 7: Import MLPClassifier Module

* Import 'MLPClassifier' from 'sklearn.neural_network'.

Step 8: Initialize MLP classifier

* Create an instance of 'MLPClassifier' with activation = 'relu'.

* Assign the instance to a variable (eg. clf).

Step 9: Fit the classifier

* Fit the model using `clf.fit(x-training-train)`

Step 10: Make predictions

* Use the predict() function on 'x-test'

(eg, ~~pred = clf.predict(x-test)~~)

* ~~Display the prediction~~

Step 10: Import Metrics Module

* Import 'confusion_matrix' from 'sklearn.metrics'

* Import 'classification_report' from 'sklearn.metrics'.

Step 11: Display Confusion Matrix.

* Use `confusion_matrix(y-test, pred)` to

generate the confusion matrix

* Display the confusion matrix

Step 13: Display Classification Report

* Use `classification_report(y-test, pred)` to generate the classification report

* Display the classification report

Step 14: Repeat step 9-13 with different Activation Function

* Initialize MLPClassifier with activation = "logistic"

* Fit the model and make prediction

* Display the confusion matrix and classification report

* Report for activation = 'tanh'

* Repeat for activation = 'identity'.

Step 15: Repeat steps 7-14 with 70-30 Ratio.

* Use `'train_test_split'` to split the dataset into training and testing sets with a ratio of 0.3

* Assign the results to 'x-train', 'y-train' and 'y-test'.

* Repeat steps 7-14 with the new training and testing sets

PROGRAM:

```
import pandas as pd
import numpy as np.
from sklearn.model_selection import train_test_
    -split
from sklearn.neural_network import MLPClass
    -ifier
from sklearn.metrics import classification_
    report, confusion_matrix
# Load dataset
bnotes = pd.read_csv('input/banknote-dataset/
    bank-note-data.csv')
print(bnotes.head(10))
x = bnotes.drop(['Class', axis=1)
y = bnotes['Class']
print(x.head(2))
print(y.head(2))
x_train, x-test, y-train, y-test = train-test-split_
    (x, y) test_size = 0.2
# activation function: relu
mlp = MLPClassifier(max_iter = 500, activation
    = 'relu')
mlp.fit(x-train, y-train)
pred = mlp.predict(x-test)
print(pred)
print(confusion_matrix(y-test, pred))
print(classification_report(y-test, pred))
```

```
# activation function : logistic
```

```
mlp = MLPClassifier(max_iter=500,
```

```
activation='logistic')
```

```
mlp.fit(x_train, y_train)
```

```
pred = mlp.predict(x_test)
```

```
print(pred)
```

```
print(confusion_matrix(y_test, pred))
```

```
print(classification_report(y_test, pred))
```

```
# activation function : tanh
```

```
mlp = MLPClassifier(max_iter=500, activation
```

```
= 'identity')
```

```
mlp.fit(x_train, y_train)
```

```
pred = mlp.predict(x_test)
```

```
print(pred)
```

```
print(confusion_matrix(y_test, pred))
```

```
print(classification_report(y_test, pred))
```

```
# ----- train-test ratio = 0.3 - - -
```

~~x-train, x-test, y-train, = train-test,~~

~~split(x, y, test_size=0.3)~~

```
# Activation function: relu
```

```
mlp = MLPClassifier(max_iter=500,
```

```
activation='logistic')
```

```
mlp.fit(x_train, y_train)
```

```
pred = mlp.predict(x_test)
```

```
print(mlp.predict(x_test))
```

```
print(pred)
```

```
print(confusion_matrix(y_test, pred))
```

```
print(classification_report(y_test, pred))
```

```
# activation function: identity  
mlp = MLPClassifier(hidden_layer_sizes=[50],  
activation='identity')  
mlp.fit(x_train, y_train)  
pred = mlp.predict(x_test)  
print(pred)  
print(confusion_matrix(y_test, pred))  
print(classification_report(y_test, pred))
```

Steps followed in the program:

- Import libraries and modules.
- Load dataset from file.
- Separate dataset into train and test.
- Import the neural network module.
- Import the classifier by specifying it.
- Fit the model on training data.
- Print the predicted output.

Output obtained from the program:

Result:-

Thus the python program to implement multi-layer perceptron with back propagation on the given dataset (banknotes.csv) has been executed successfully and its results have been analysed successfully for different activation functions (relu, logistic, tanh, identity) with two different training-testing ratio(0.2 and 0.3).