

A Python Program to implement multilayer Perceptron with Back Propagation

Aim:-

To implement multilayer perceptron 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("banknote.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.describe()`)

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(columns = "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.Curt	Entropy	Class
0	3.02160	0.0061	-2.80730	-0.44690	0
1	4.54990	0.1074	-2.45660	-1.46210	0
2	3.06600	-2.6303	4.92420	0.10845	0
3	3.45680	0.6328	-4.01120	-3.59440	0
4	0.82024	-4.4682	4.57160	-0.06080	0
5	4.30040	0.6715	-3.06060	-2.16250	0
6	3.59120	0.0129	0.72084	0.50421	0
7	2.09230	-0.8100	8.46360	-0.60216	0
8	3.20320	5.7588	-0.75345	-0.01261	0
9	1.53560	0.1772	-2.27180	-0.73535	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 'MLP Classifier' from sklearn.neural-network.

Step 8: Initialize MLP Classifier

* Create an instance of 'MLP Classifier' with activation = 'relu'.

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

Step 9: Fit the Classifier

* Fit the model using clf.fit(x_train, y_train)

Step 10: Make prediction

* 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 MLP Classifier with `activation = 'logistic'`

- * Fit the model and make prediction

- * Display the confusion matrix and classification report

- * Repeat 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 MLPClassifier
from sklearn.metrics import classification
report, confusion_matrix
# Load dataset
bnotes = pd.read_csv('./input/banknote-data.csv')
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

```
m1p = MLPClassifier(max_iter=500,  
activation='logistic')
```

```
m1p.fit(x=train, y=train)
```

```
pred = m1p.predict(x=test)
```

```
print(pred)
```

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

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

activation function : tanh

```
m1p = MLPClassifier(max_iter=500, activation  
= 'tanh')
```

```
m1p.fit(x=train, y=train)
```

```
pred = m1p.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, y_test = train_test~~

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

Activation function : relu

```
m1p = MLPClassifier(max_iter=500,  
activation='logistic')
```

```
m1p.fit(x=train, y=train)
```

```
pred = m1p.predict(x=test)
```

```
print(m1p.predict(x=test))
```

```
print(pred)
```

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

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

```

# activation function: identify
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))

```

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).