EXPT NO: 5 A python program to implement Multi Layer

DATE:20/9/24 Perceptron With Backpropagation

AIM:

To write a python program to implement Multilayer perceptron with backpropagation .

PROCEDURE:

Implementing Multilayer perceptron with backpropagation using the Keras dataset involve the following steps:

Step 1: Import Necessary Libraries

First, import the libraries that are essential for data manipulation, visualization, and model building.

```
# importing modules
import tensorflow as tf
import numpy as np
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Flatten
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import Activation
import matplotlib.pyplot as plt
```

Step 2: Load the Keras Dataset

The Keras dataset can be loaded.

```
(x_train, y_train), (x_test, y_test) = tf.keras.datasets.mnist.load_data()
OUTPUT:
```

231501073

Step 3: Data Preprocessing

Ensure the data is clean and ready for modeling. Since the Iris dataset is clean, minimal preprocessing is needed.

```
# Cast the records into float values
x train = x train.astype('float32')
x test = x test.astype('float32')
# normalize image pixel values by dividing
# by 255
gray scale = 255
x train /= gray scale
x_test /= gray_scale
print("Feature matrix:", x train.shape)
print("Target matrix:", x test.shape)
print("Feature matrix:", y train.shape)
print("Target matrix:", y test.shape)
```

OUTPUT:

```
Feature matrix: (60000, 28, 28)
Target matrix: (10000, 28, 28)
Feature matrix: (60000,)
Target matrix: (10000,)
```

Step 4 : Train a Model

```
model = Sequential([
  # reshape 28 row * 28 column data to 28*28 rows
  Flatten(input_shape=(28, 28)),
  # dense layer 1
  Dense(256, activation='sigmoid'),
  # dense layer 2
  Dense(128, activation='sigmoid'),
  # output layer
 Dense(10, activation='sigmoid'),
1)
```

OUTPUT:

```
/usr/local/lib/python3.10/dist-packages/keras/src/layers/reshaping/flatten.py:37: UserWarning: super().__init__(**kwargs)
```

Step 5 : Make Predictions

Use the model to make predictions based on the independent variable.

OUTPUT:

```
→ Epoch 1/10
                             - 5s 115ms/step - accuracy: 0.3546 - loss: 2.1596 - val_accuracy: 0.68
    24/24 -
    Epoch 2/10
    24/24 ---
                              - 4s 53ms/step - accuracy: 0.7116 - loss: 1.3743 - val_accuracy: 0.820
    Epoch 3/10
    24/24 -
                             - 1s 53ms/step - accuracy: 0.8221 - loss: 0.8221 - val_accuracy: 0.872
    Epoch 4/10
    24/24 -
                              - 3s 65ms/step - accuracy: 0.8720 - loss: 0.5676 - val_accuracy: 0.892
    Epoch 5/10
    24/24 -
                             - 2s 99ms/step - accuracy: 0.8907 - loss: 0.4444 - val_accuracy: 0.902
    Epoch 6/10
    24/24 -
                             3s 102ms/step - accuracy: 0.8993 - loss: 0.3852 - val_accuracy: 0.91
    Epoch 7/10
    24/24 ---
                             - 3s 104ms/step - accuracy: 0.9088 - loss: 0.3416 - val_accuracy: 0.91
    Epoch 8/10
    24/24 -
                             - 2s 92ms/step - accuracy: 0.9119 - loss: 0.3188 - val accuracy: 0.922
    Epoch 9/10
    24/24 -
                              - 2s 92ms/step - accuracy: 0.9191 - loss: 0.2911 - val_accuracy: 0.926
    Epoch 10/10
                              - 3s 99ms/step - accuracy: 0.9245 - loss: 0.2704 - val accuracy: 0.925
    <keras.src.callbacks.history.History at 0x7d9ca1406a40>
```

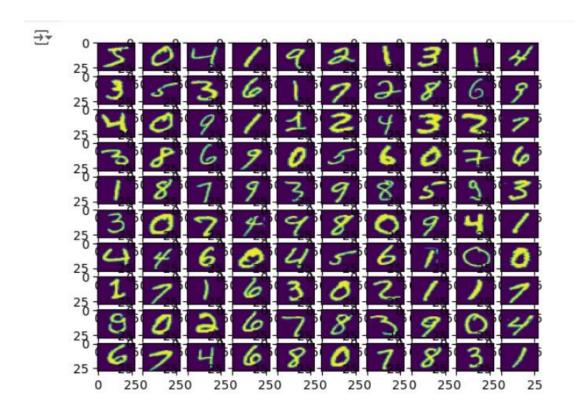
Step 6 : Evaluate the Model

Evaluate the model performance.

```
results = model.evaluate(x_test, y_test, verbose = 0)
print('test loss, test acc:', results)
```

OUTPUT:

→ test loss, test acc: [0.2589016258716583, 0.9277999997138977]



RESULT:

This step-by-step process will help us to implement MultiLayer Perceptron with Backpropagation models using the Keras dataset and analyze their performance.