## **EXPT NO: 5** A python program to implement Multi Layer

DATE: 20/09/2024 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:
```

## **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'),
])
```

#### **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.

```
model.fit(x_train, y_train, epochs=10,
    batch_size=2000,
    validation_split=0.2)
```

#### **OUTPUT:**

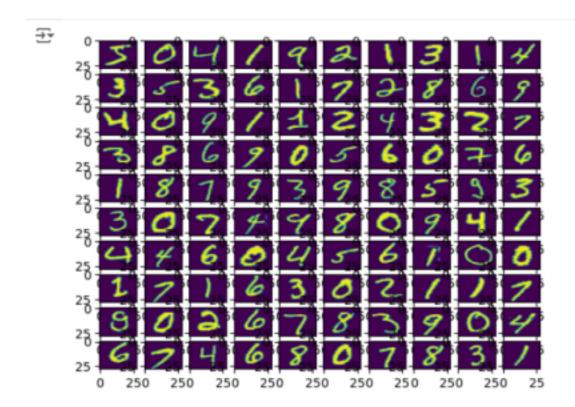
```
→ Epoch 1/10
    24/24 -
                              - 5s 115ms/step - accuracy: 0.3546 - loss: 2.1596 - val accuracy: 0.68
    Epoch 2/10
    24/24 -
                              4s 53ms/step - accuracy: 0.7116 - loss: 1.3743 - val_accuracy: 0.82€
    Epoch 3/10
                              - 1s 53ms/step - accuracy: 0.8221 - loss: 0.8221 - val_accuracy: 0.872
    24/24 -
    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
                             3s 102ms/step - accuracy: 0.8993 - loss: 0.3852 - val_accuracy: 0.91
    24/24 -
    Epoch 7/10
                             - 3s 104ms/step - accuracy: 0.9088 - loss: 0.3416 - val_accuracy: 0.91
    24/24 -
    Epoch 8/10
                              - 2s 92ms/step - accuracy: 0.9119 - loss: 0.3188 - val_accuracy: 0.922
    24/24 -
    Epoch 9/10
                              - 2s 92ms/step - accuracy: 0.9191 - loss: 0.2911 - val_accuracy: 0.92€
    24/24 -
    Epoch 10/10
    24/24 -
                             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.

#### **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.