Experiment 1

Aim :- To train a neural network to implement a digit classifier using the MNIST dataset by only using the numpy library.

Theory:-

- 1. **Dataset:** The MNIST dataset consists of 60,000 training images and 10,000 test images of handwritten digits (0-9), each of size 28x28 pixels.
- 2. Libraries Used: NumPy
- 3. **Network Architecture:** A simple feedforward neural network with the following layers:
 - a. Input Layer: 784 neurons (flattened 28x28 image pixels)
 - b. Hidden Layer: 128 neurons (ReLU activation)
 - c. Output Layer: 10 neurons (Softmax activation)

4. Training Method:

- a. Data Preprocessing: Normalize pixel values to range [0,1]
- b. Forward Propagation: Compute activations for each layer
- c. Loss Function: Categorical Cross-Entropy
- d. Backpropagation: Compute gradients using the chain rule
- e. Optimization: Stochastic Gradient Descent (SGD) with a learning rate of 0.01

Procedure:

1. Data Preprocessing:

- a. Load the MNIST dataset.
- b. Normalize the pixel values by dividing them by 255.
- c. Convert labels to one-hot encoding.

2. Model Initialization:

a. Initialize weights and biases randomly.

3. Training Loop:

a. For each epoch:

- i. Perform forward propagation to compute predictions.
- ii. Calculate the loss using categorical cross-entropy.
- iii. Perform backpropagation to compute gradients.
- iv. Update weights and biases using gradient descent.

4. Testing the Model:

- a. Perform forward propagation on test images.
- b. Compute accuracy by comparing predictions with actual labels.

Results:

- The model achieved an accuracy of approximately 85-90% on the test set after training for multiple epochs.
- Training time depended on computational resources but typically required several minutes per epoch.
- The performance of the model was observed to improve with increased hidden layer neurons and fine-tuning of learning rates.

Conclusion: The experiment successfully demonstrated that a neural network could be implemented and trained using only NumPy for digit classification on the MNIST dataset. The accuracy obtained is reasonable for a basic feedforward network without advanced optimizations. Further improvements, such as adding more hidden layers, using better weight initialization techniques, and implementing advanced optimizers like Adam, could enhance model performance.