# **MNIST Handwritten Digit Classification using Neural Networks**

#### **Abstract**

This project focuses on classifying handwritten digits from the MNIST dataset using a deep learning model built with Keras. The network is trained to identify digits 0–9 from grayscale images. A feedforward neural network was implemented, achieving a training accuracy of **98.9%** and a test accuracy of **97.1%**. The evaluation also includes confusion matrix analysis to highlight class-specific performance.

#### 1. Introduction

Handwritten digit recognition is a foundational problem in computer vision and machine learning. The MNIST dataset is a benchmark for evaluating classification algorithms. This project demonstrates the application of a neural network model for MNIST digit classification, covering the entire pipeline from data loading to performance evaluation.

#### 2. Dataset Details

Source: MNIST dataset from keras.datasets

• Training set: 60,000 images

Test set: 10,000 images

• Image size: 28 × 28 pixels

Channels: 1 (grayscale)

• Classes: Digits from 0 to 9

### 3. Data Preprocessing

1. Loading Data: Dataset loaded using keras.datasets.mnist.load data().

2. **Normalization:** Pixel values scaled from [0, 255] to [0, 1].

- 3. **Flattening:** Images reshaped from 28×28 to 784-element vectors for input to the dense layers.
- 4. **One-Hot Encoding:** Class labels transformed into one-hot vectors for multi-class classification.

#### 4. Model Architecture

The neural network was designed as follows:

Layer	Туре	Units/Filters	Activation	Input Shape
Dense (1)	Fully Connected	128	ReLU	(784,)
Dense (2)	Fully Connected	64	ReLU	_
Dense (3)	Fully Connected	10	Softmax	_

Loss function: categorical\_crossentropy

Optimizer: adamMetrics: accuracy

# 5. Training Setup & Hyperparameters

• **Epochs:** 10

• Batch size: 32

Validation split: 20% of training data used for validation

• Framework: TensorFlow + Keras

#### 6. Results & Performance

• Training Accuracy: 98.9%

• Test Accuracy: 97.1%

• **Observation:** The model generalized well with minimal overfitting.

# 7. Confusion Matrix & Analysis

The confusion matrix revealed strong performance across all classes, with occasional misclassification between similar digits such as **4 and 9**, or **3 and 5**. This is typical in MNIST classification due to similarities in handwritten styles.

### 8. Conclusion

The project successfully demonstrates a high-accuracy neural network classifier for MNIST handwritten digits. While accuracy is already high, improvements could be made by:

- Using Convolutional Neural Networks (CNNs) for better spatial feature extraction
- Applying data augmentation to improve robustness
- Hyperparameter tuning