PHASE - 1

RECOGNIZING HANDWRITTEN DIGITS WITH DEEP LEARNING FOR SMARTER AI APPLICATIONS

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Problem Statement :

In an increasingly digital world, enabling machines to understand handwritten input is vital for applications such as digital forms, automated postal systems, and note-taking applications. Despite the widespread use of printed text, handwritten digits are still common in forms, invoices, and classroom work. Manually processing such data is time-consuming and error-prone. This project aims to solve the problem of recognizing handwritten digits using deep learning techniques, which can

significantly improve the efficiency and accuracy of data entry and verification processes.

☐ Objectives of the Project :

- Develop a deep learning model capable of accurately recognizing handwritten digits.
- Train the model on a publicly available dataset and evaluate its performance.
- Optimize the model for accuracy and performance.
- Visualize the results and decision-making process of the model.
- Optionally, deploy the model as an interactive web app or notebook-based tool for user testing.

☐ Scope of the Project :

Features to be built/analyzed:

- Image preprocessing pipeline for handwritten digit images.
- Convolutional Neural Network (CNN) architecture for digit recognition.
- Training, validation, and testing workflow with performance metrics.
- Visualization of misclassified digits and model insights.

Limitations and Constraints:

· Project will use the MNIST dataset only.

 Model will be trained and tested within a notebook environment (e.g., Google Colab).

 Deployment (if implemented) will be via basic tools like Streamlit or Gradio.

• Focused only on digits (0–9), not alphabets or symbols.

□ Data Sources :

Dataset: MNIST Handwritten Digits Dataset

Source: Public dataset from Kaggle or from tensorflow.keras.datasets.

Type: Public and static dataset.

Format: 28x28 grayscale images, labeled 0–9.

☐ High-Level Methodology:

 Data Collection: Dataset will be downloaded from Keras datasets or Kaggle.

 Data Cleaning: Normalize pixel values, convert labels to categorical form, check for shape consistency.

Exploratory Data Analysis (EDA): Visualize sample digits,
 class distribution, and pixel intensity histograms.

- Feature Engineering: Normalize images, reshape for CNN input, one-hot encoding of labels.
- Model Building: Use CNN (e.g., Conv2D, MaxPooling2D, Dense layers); test different architectures.
- Model Evaluation: Use accuracy, confusion matrix, and validation loss/accuracy plots.
- Visualization & Interpretation: Visualize filters, activation maps, confusion matrix, misclassifications.
- Deployment (Optional): Build a web interface using Streamlit or Gradio for real-time predictions.

☐ Tools and Technologies:

- Programming Language: Python
- Notebook/IDE: Google Colab or Jupyter Notebook
- · Libraries:
 - Data Processing: numpy, pandas
 - Visualization: matplotlib, seaborn
 - Modeling: tensorflow, keras, scikit-learn
- Optional Tools for Deployment: Streamlit, Gradio.

☐ Team Members and Roles:

Name	Role and Responsibilities
S.SANGDEENA	Project Lead – Oversees the entire project, coordinates tasks, and works on model training and optimization.
P.VAISHNAVI	Data Engineer – Responsible for data loading, preprocessing, and handling data cleaning and augmentation.
S.JOTHIGA	Exploratory Data Analyst – Performs EDA, visualizes patterns, class distribution, and contributes to insights.

Name	Role and Responsibilities
R.PRADHISH	Model Architect – Designs the CNN model architecture, experiments with hyperparameters, and evaluates models.
T.AGILESH	Deployment & Documentation Specialist – Builds the deployment interface (Streamlit/Gradio) and prepares the final report and presentations.