

Handwritten Digit Recognition using TensorFlow

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1 Introduction

In this project, we will walk through the process of building a handwritten digit recognition model using the MNIST dataset and TensorFlow. We will load and preprocess the data, create a neural network, train the model, and evaluate its performance.

2 Data Loading and Preprocessing

We start by loading the MNIST dataset using TensorFlow's Keras API. The dataset contains 60,000 training images and 10,000 test images.

```
(X_train, y_train), (X_test, y_test) = keras.datasets.mnist.load_data()
```

Next, we normalize the pixel values of the images to the range $[0, 1]$ by dividing by 255.

```
X_train = X_train / 255  
X_test = X_test / 255
```

We also flatten the images from 28x28 to 784 pixels.

```
X_train_flattened = X_train.reshape(len(X_train), 28*28)  
X_test_flattened = X_test.reshape(len(X_test), 28*28)
```

3 Neural Network Model

We create a simple feedforward neural network with two dense layers. The first layer has 150 neurons with ReLU activation, and the output layer has 10 neurons with a softmax activation.

```
model = keras.Sequential([  
    keras.layers.Flatten(input_shape=(28, 28)),  
    keras.layers.Dense(150, activation='relu'),  
    keras.layers.Dense(10, activation='softmax')  
)
```

We compile the model with the Adam optimizer and sparse categorical cross-entropy loss.

```
model.compile(optimizer='adam',  
              loss='sparse_categorical_crossentropy',  
              metrics=['accuracy'])
```

4 Model Training

We train the model on the training data for 40 epochs.

```
model.fit(X_train, y_train, epochs=40)
```

5 Model Evaluation

We evaluate the model on the test data.

```
model.evaluate(X_test, y_test)
```

6 Confusion Matrix

We create a confusion matrix to visualize the model's performance.

```
y_pred = model.predict(X_test)
y_pred_labels = np.argmax(y_pred, axis=1)
conf_matrix = confusion_matrix(y_test, y_pred_labels)
```

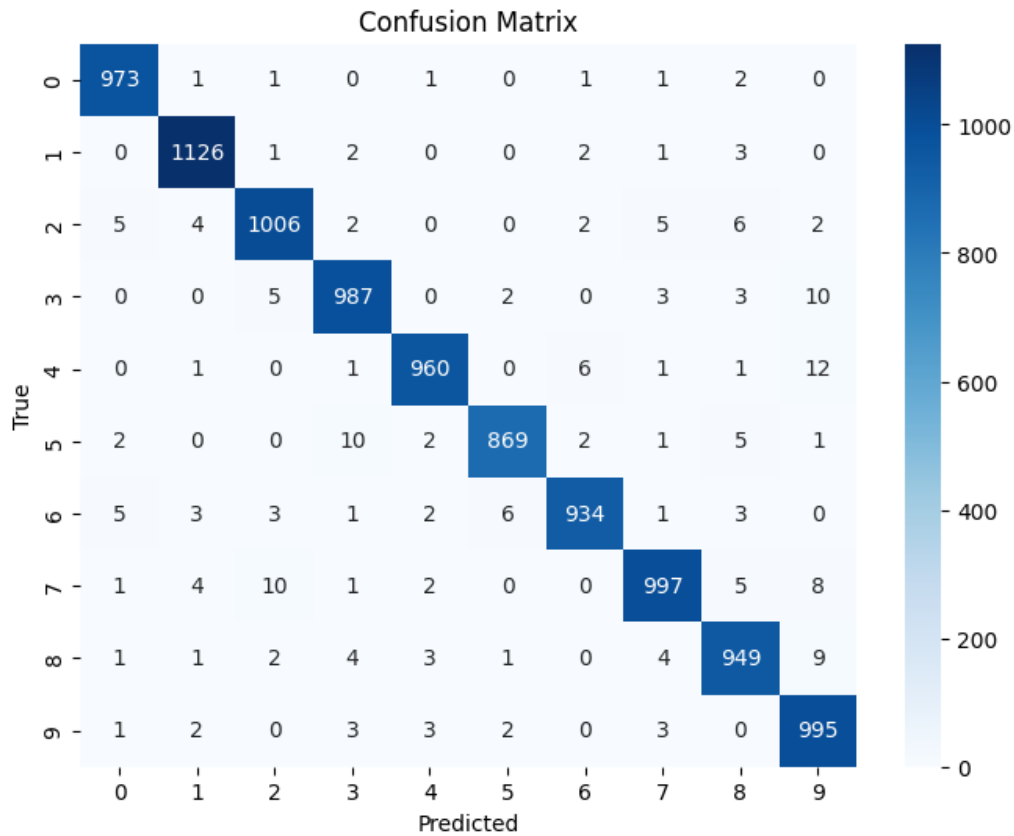


Figure 1: Confusion Matrix

7 Conclusion

In this project, we successfully built and trained a neural network model for handwritten digit recognition. The model achieved an accuracy of 98% on the MNIST dataset.