PAREENITA A.SHIRSATH ROLL.NO: 57 B.E.A.I.&.D.S.

DL EXPERIMENT NO: 08

```
◆ Gemini
```

```
import matplotlib.pyplot as plt
import tensorflow as tf
import numpy as np
mnist = tf.keras.datasets.mnist
(x_train, y_train), (x_test, y_test) = mnist.load_data()
rows, cols = 28, 28
# Reshape the data into a 4D Array
x_train = x_train.reshape(x_train.shape[0], rows, cols, 1)
x_test = x_test.reshape(x_test.shape[0], rows, cols, 1)
input shape = (rows, cols, 1)
# Set type as float32
x_{train} = x_{train.astype('float32')} / 255.0
x_test = x_test.astype('float32') / 255.0
# Transform labels to one hot encoding
y_train = tf.keras.utils.to_categorical(y_train, 10)
def build_lenet(input_shape):
    # Define Sequential Model
   model = tf.keras.Sequential()
   # Convolution Layer
   model.add(tf.keras.layers.Conv2D(filters=6, strides=(1, 1), kernel_size=(5, 5),
activation='tanh', input_shape=input_shape))
    # SubSampling Layer
   model.add(tf.keras.layers.AveragePooling2D(pool_size=(2, 2), strides=(2, 2)))
   # Convolution Laver
    model.add(tf.keras.layers.Conv2D(filters=6, strides=(1, 1), kernel_size=(5, 5),
activation='tanh'))
    # SubSampling Layer
   model.add(tf.keras.layers.AveragePooling2D(pool_size=(2, 2), strides=(2, 2)))
    # Flatten the output (Important: Flatten before Dense layers)
    model.add(tf.keras.layers.Flatten())
    # Fully Connected Laver
    model.add(tf.keras.layers.Dense(units=120, activation='tanh'))
    # Dense Layers
    model.add(tf.keras.layers.Dense(units=84, activation='tanh'))
    # Output Layer
    model.add(tf.keras.layers.Dense(units=10, activation='softmax'))
    return model
# Build the model
lenet = build_lenet(input_shape)
# Compile the model
lenet.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
history = lenet.fit(x_train, y_train, epochs=epochs, batch_size=128, verbose=1)
# Check Accuracy of the Model
if len(y_test.shape) != 2 or y_test.shape[1] != 10:
   y_test = tf.keras.utils.to_categorical(y_test, 10)
loss, acc = lenet.evaluate(x_test, y_test)
print('Accuracy:', acc)
# Reshape for plotting
x_train_plot = x_train.reshape(x_train.shape[0], 28, 28)
x_test_plot = x_test.reshape(x_test.shape[0], 28, 28)
print('Training Data:', x_train_plot.shape, y_train.shape)
print('Test Data:', x_test_plot.shape, y_test.shape)
# Plot the Image
image_indices = [8888, 123, 4567]
image_indices = [8, 4567] # Changed indices to be within bounds
for idx in image_indices:
   # Plot the image
   \verb|plt.imshow(x_test_plot[idx], cmap='Greys')| \\
   plt.title(f"Image index: {idx}")
   plt.show()
    # Make prediction
    pred = lenet.predict(x_test[idx].reshape(1, rows, cols, 1))
   print(f'Predicted label for image {idx}:', pred.argmax())
```

