```
import numpy as np
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout
from tensorflow.keras.datasets import mnist
from tensorflow.keras.utils import to_categorical
# Load MNIST dataset
(X_train, y_train), (X_test, y_test) = mnist.load_data()
# Normalize images
X_{train} = X_{train} / 255.0
X_{\text{test}} = X_{\text{test}} / 255.0
# Reshape to add channel dimension
X_train = X_train.reshape(-1, 28, 28, 1)
X_{\text{test}} = X_{\text{test.reshape}}(-1, 28, 28, 1)
# One-hot encode labels
y_train = to_categorical(y_train, 10)
y_test = to_categorical(y_test, 10)
     Downloading data from <a href="https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz">https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz</a>
     11490434/11490434
                                              0s Ous/step
# Build the CNN model
model = Sequential([
    Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)),
    MaxPooling2D((2, 2)),
    Dropout(0.3),
    Conv2D(64, (3, 3), activation='relu'),
    MaxPooling2D((2, 2)),
    Dropout(0.3),
    Flatten(),
    Dense(128, activation='relu'),
    Dropout(0.4).
    Dense(10, activation='softmax') # 10 classes for digits (0-9)
])
# Compile the model
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
🚁 /usr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_shape`/`inpu
       super().__init__(activity_regularizer=activity_regularizer, **kwargs)
# Train the model
history = model.fit(X_train, y_train, epochs=10, batch_size=32, validation_split=0.2)
# Evaluate the model
test_loss, test_acc = model.evaluate(X_test, y_test)
print(f"Test Accuracy: {test_acc:.2f}")
    Epoch 1/10
     1500/1500
                                    <mark>— 48s</mark> 31ms/step - accuracy: 0.8345 - loss: 0.5107 - val_accuracy: 0.9798 - val_loss: 0.0687
     Epoch 2/10
     1500/1500
                                    - 45s 30ms/step - accuracy: 0.9670 - loss: 0.1096 - val_accuracy: 0.9858 - val_loss: 0.0466
     Epoch 3/10
     1500/1500
                                    - 82s 30ms/step - accuracy: 0.9785 - loss: 0.0746 - val_accuracy: 0.9884 - val_loss: 0.0404
     Fnoch 4/10
     1500/1500
                                    – 82s 30ms/step - accuracy: 0.9806 - loss: 0.0641 - val_accuracy: 0.9898 - val_loss: 0.0342
     Epoch 5/10
     1500/1500
                                    - 81s 30ms/step - accuracy: 0.9819 - loss: 0.0554 - val_accuracy: 0.9896 - val_loss: 0.0342
     Epoch 6/10
     1500/1500
                                    - 48s 32ms/step - accuracy: 0.9838 - loss: 0.0518 - val_accuracy: 0.9908 - val_loss: 0.0301
     Epoch 7/10
     1500/1500
                                    - 81s 31ms/step - accuracy: 0.9851 - loss: 0.0484 - val_accuracy: 0.9903 - val_loss: 0.0336
     Epoch 8/10
     1500/1500
                                     - 80s 30ms/step - accuracy: 0.9856 - loss: 0.0452 - val_accuracy: 0.9898 - val_loss: 0.0328
     Epoch 9/10
     1500/1500
                                     - 85s 32ms/step - accuracy: 0.9870 - loss: 0.0405 - val_accuracy: 0.9913 - val_loss: 0.0328
     Epoch 10/10
     1500/1500
                                    - 80s 30ms/step - accuracy: 0.9887 - loss: 0.0372 - val_accuracy: 0.9930 - val_loss: 0.0281
```

```
# Save the trained model
model.save("handwritten_character_recognition.h5")

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is consi
```

handwritten digits prediction using deep learning

```
from sklearn.datasets import fetch openml
mnist=fetch_openml("mnist_784")
import numpy as np
import matplotlib.pyplot as plt
from sklearn.datasets import fetch_openml
from sklearn.model selection import train test split
from sklearn.neural_network import MLPClassifier
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
# Load the MNIST dataset
mnist = fetch_openml('mnist_784', version=1)
X, y = mnist.data, mnist.target
# Ensure that the data is in the correct format
X = X.astype(np.float32) # Convert pixel values to float32
                         # Convert target values to integers
y = y.astype(np.int8)
# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Initialize the MLP classifier
mlp = MLPClassifier(hidden_layer_sizes=(64, 64),
                    max_iter=50,
                    alpha=1e-4.
                    solver='adam'
                    random_state=42,
                    verbose=10.
                    tol=1e-4)
# Train the model
mlp.fit(X_train, y_train)
# Predict on the test data
y_pred = mlp.predict(X_test)
# Evaluate the model
print("Classification Report:")
print(classification_report(y_test, y_pred))
# Confusion matrix
print("Confusion Matrix:")
\verb|print(confusion_matrix(y_test, y_pred))|
accuracy = accuracy_score(y_test, y_pred)
print(f"Accuracy: {accuracy * 100:.2f}%")
# Visualize some predictions
fig, axes = plt.subplots(2, 5, figsize=(10, 5))
for ax, image, prediction, true_label in zip(axes.ravel(), X_test.to_numpy(), y_pred, y_test):
    ax.set_axis_off()
    image = image.reshape(28, 28)
    ax.imshow(image, cmap='gray')
    ax.set_title(f"Pred: {prediction}, True: {true_label}")
plt.tight_layout()
plt.show()
```

```
\rightarrow Iteration 1, loss = 2.10292369
    Iteration 2, loss = 0.63468546
    Iteration 3, loss = 0.41170029
    Iteration 4, loss = 0.32570292
    Iteration 5, loss = 0.27403755
    Iteration 6, loss = 0.24147559
    Iteration 7, loss = 0.21699058
    Iteration 8, loss = 0.19487723
    Iteration 9, loss = 0.17771913
    Iteration 10, loss = 0.16357822
    Iteration 11, loss = 0.15891865
    Iteration 12, loss = 0.14965729
    Iteration 13, loss = 0.12972488
    Iteration 14, loss = 0.13554853
    Iteration 15, loss = 0.13080247
    Iteration 16, loss = 0.13119637
    Iteration 17, loss = 0.11904129
    Iteration 18, loss = 0.11818511
    Iteration 19, loss = 0.11466374
    Iteration 20, loss = 0.11052036
    Iteration 21, loss = 0.10766985
    Iteration 22, loss = 0.10104356
    Iteration 23, loss = 0.09436340
    Iteration 24, loss = 0.08939378
    Iteration 25, loss = 0.09473639
    Iteration 26, loss = 0.08972806
    Iteration 27, loss = 0.08268881
    Iteration 28, loss = 0.08483314
    Iteration 29, loss = 0.08084884
    Iteration 30, loss = 0.07425225
    Iteration 31, loss = 0.06775623
    Iteration 32, loss = 0.06304130
    Iteration 33, loss = 0.07305904
    Iteration 34, loss = 0.06126501
    Iteration 35, loss = 0.05835775
    Iteration 36, loss = 0.06709495
    Iteration 37, loss = 0.06276928
    Iteration 38, loss = 0.05704420
    Iteration 39, loss = 0.05660724
    Iteration 40, loss = 0.04841852
    Iteration 41, loss = 0.05089388
    Iteration 42, loss = 0.05344457
    Iteration 43, loss = 0.05155276
    Iteration 44, loss = 0.04803985
    Iteration 45, loss = 0.04400076
    Iteration 46, loss = 0.04093383
    Iteration 47, loss = 0.03991760
    Iteration 48, loss = 0.03962692
    Iteration 49, loss = 0.04080373
    Iteration 50, loss = 0.04110693
    Classification Report:
                  precision
                                recall f1-score
                                                   support
                                            0.98
               0
                        0.98
                                  0.98
                                                       1343
               1
                       0.98
                                  0.98
                                            0.98
                                                      1600
               2
                        0.96
                                  0.94
                                            0.95
                                                       1380
               3
                       0.93
                                  0.95
                                            0.94
                                                       1433
               4
                       0.94
                                  0.97
                                            0.96
                                                      1295
               5
                       0.97
                                  0.93
                                            0.95
                                                       1273
                                                       1396
               6
                        0.98
                                  0.97
                                            0.98
               7
                       0.95
                                  0.97
                                            0.96
                                                      1503
               8
                       0.94
                                  0.93
                                            0.94
                                                      1357
               9
                        0.95
                                  0.94
                                            0.94
                                                      1420
                                            9.96
                                                      14000
        accuracy
       macro avg
                        0.96
                                  0.96
                                            0.96
                                                      14000
    weighted avg
                        0.96
                                  0.96
                                            0.96
                                                      14000
    Confusion Matrix:
    [[1314
                                                        31
         0 1565
                   3
                        3
                                        3
                                                 12
                                                       51
                             1
                                   1
         4
              8 1304
                       12
                              8
                                   3
                                        3
                                            24
                                                 11
                                                       31
         0
                  12 1364
                              2
                                  14
                                             9
                                                      18]
              1
                        0 1260
                                  0
                                        3
                                            10
                                                  3
                                                      14]
         3
              6
                   5
                        41
                             4 1178
                                       12
                                             4
                                                 15
                                                       5]
         6
              5
                   2
                        2
                            11
                                   8 1356
                                             1
                                                  3
                                                       2]
         4
              2
                   8
                        2
                             11
                                   1
                                        0 1458
                                                  2
                                                      15]
                                                      101
                  17
                             5
                                   9
                                             8 1266
         2
              9
                       29
                                        2
         8
                   3
                        9
                            36
                                   4
                                        1
                                            17
                                                  9 1331]]
    Accuracy: 95.69%
    /usr/local/lib/python3.11/dist-packages/sklearn/neural_network/_multilayer_perceptron.py:691: ConvergenceWarning: Stochastic Optimize
      warnings.warn(
         Pred: 8, True: 8
                                   Pred: 4, True: 4
                                                              Pred: 5, True: 8
                                                                                         Pred: 7, True: 7
                                                                                                                   Pred: 7, True: 7
```