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"import cv2\n",

"import numpy as np\n",

"from keras.datasets import mnist\n",

"from keras.layers import Dense, Flatten, MaxPooling2D, Dropout\n",

"from keras.layers.convolutional import Conv2D\n",

"from keras.models import Sequential\n",

"from tensorflow.keras.utils import to\_categorical\n",

"import matplotlib.pyplot as plt"

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"(X\_train, y\_train), (X\_test, y\_test) = mnist.load\_data()"

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"plt.imshow(X\_train[0], cmap=\"gray\")\n",

"plt.show()\n",

"print (y\_train[0])"

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"print (\"Shape of X\_train: {}\".format(X\_train.shape))\n",

"print (\"Shape of y\_train: {}\".format(y\_train.shape))\n",

"print (\"Shape of X\_test: {}\".format(X\_test.shape))\n",

"print (\"Shape of y\_test: {}\".format(y\_test.shape))"

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"Shape of y\_test: (10000,)\n"

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"# RESHAPING SO AS TO CONVERT IMAGES FOR OUR MODEL"

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"X\_train = X\_train.reshape(60000, 28, 28, 1)\n",

"X\_test = X\_test.reshape(10000, 28, 28, 1)"

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"y\_train = to\_categorical(y\_train)\n",

"y\_test = to\_categorical(y\_test)"

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"model = Sequential()\n",

"\n",

"## DECLARE THE LAYERS\n",

"layer\_1 = Conv2D(64, kernel\_size=3, activation='relu', input\_shape=(28, 28, 1))\n",

"layer\_2 = MaxPooling2D(pool\_size=2)\n",

"layer\_3 = Conv2D(32, kernel\_size=3, activation='relu')\n",

"layer\_4 = MaxPooling2D(pool\_size=2)\n",

"layer\_5 = Dropout(0.5)\n",

"layer\_6 = Flatten()\n",

"layer\_7 = Dense(128, activation=\"relu\")\n",

"layer\_8 = Dropout(0.5)\n",

"layer\_9 = Dense(10, activation='softmax')\n",

"\n",

"##ADD THE LAYERS TO THE MODEL\n",

"model.add(layer\_1)\n",

"model.add(layer\_2)\n",

"model.add(layer\_3)\n",

"model.add(layer\_4)\n",

"model.add(layer\_5)\n",

"model.add(layer\_6)\n",

"model.add(layer\_7)\n",

"model.add(layer\_8)\n",

"model.add(layer\_9)"

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"model.compile(optimizer='adam', loss='categorical\_crossentropy', metrics=['accuracy'])"

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"model.fit(X\_train, y\_train, validation\_data=(X\_test, y\_test), epochs=3)"

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"Epoch 2/3\n",

"1875/1875 [==============================] - 83s 44ms/step - loss: 0.2960 - accuracy: 0.9122 - val\_loss: 0.0714 - val\_accuracy: 0.9779\n",

"Epoch 3/3\n",

"1875/1875 [==============================] - 81s 43ms/step - loss: 0.2110 - accuracy: 0.9369 - val\_loss: 0.0559 - val\_accuracy: 0.9836\n"

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"example = X\_train[1]\n",

"prediction = model.predict(example.reshape(1, 28, 28, 1))\n",

"print (\"Prediction (Softmax) from the neural network:\\n\\n {}\".format(prediction))\n",

"hard\_maxed\_prediction = np.zeros(prediction.shape)\n",

"hard\_maxed\_prediction[0][np.argmax(prediction)] = 1\n",

"print (\"\\n\\nHard-maxed form of the prediction: \\n\\n {}\".format(hard\_maxed\_prediction))\n",

"\n",

"print (\"\\n\\n--------- Prediction --------- \\n\\n\")\n",

"plt.imshow(example.reshape(28, 28), cmap=\"gray\")\n",

"plt.show()\n",

"print(\"\\n\\nFinal Output: {}\".format(np.argmax(prediction)))"

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"1/1 [==============================] - 0s 114ms/step\n",

"Prediction (Softmax) from the neural network:\n",

"\n",

" [[9.9999928e-01 3.4740781e-13 1.6311596e-08 2.2861069e-11 1.0698914e-12\n",

" 1.6004112e-10 1.7066007e-07 7.1788651e-11 4.5166394e-07 2.4545757e-08]]\n",

"\n",

"\n",

"Hard-maxed form of the prediction: \n",

"\n",

" [[1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]]\n",

"\n",

"\n",

"--------- Prediction --------- \n",

"\n",

"\n"

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"\n",

"\n",

"Final Output: 0\n"

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"metrices=model.evaluate(X\_test,y\_test,verbose=0)\n",

"print(\"Metrices(test loss and Test Accuracy):\")\n",

"print(metrices)"

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"image = cv2.imread('test\_image.jpg')\n",

"image = np.full((100,80,3), 12, dtype = np.uint8)\n",

"grey = cv2.cvtColor(image.copy(), cv2.COLOR\_BGR2GRAY)\n",

"ret, thresh = cv2.threshold(grey.copy(), 75, 255, cv2.THRESH\_BINARY\_INV)\n",

"contours,hierarchy = cv2.findContours(thresh.copy(), cv2.RETR\_EXTERNAL, cv2.CHAIN\_APPROX\_SIMPLE)\n",

"preprocessed\_digits = []\n",

"\n",

"for c in contours:\n",

" x,y,w,h = cv2.boundingRect(c)\n",

" \n",

" # Creating a rectangle around the digit in the original image (for displaying the digits fetched via contours)\n",

" cv2.rectangle(image, (x,y), (x+w, y+h), color=(0, 255, 0), thickness=2)\n",

" \n",

" # Cropping out the digit from the image corresponding to the current contours in the for loop\n",

" digit = thresh[y:y+h, x:x+w]\n",

" \n",

" # Resizing that digit to (18, 18)\n",

" resized\_digit = cv2.resize(digit, (18,18))\n",

" \n",

" # Padding the digit with 5 pixels of black color (zeros) in each side to finally produce the image of (28, 28)\n",

" padded\_digit = np.pad(resized\_digit, ((5,5),(5,5)), \"constant\", constant\_values=0)\n",

" \n",

" # Adding the preprocessed digit to the list of preprocessed digits\n",

" preprocessed\_digits.append(padded\_digit)\n",

"\n",

"print(\"\\n\\n\\n----------------Contoured Image--------------------\")\n",

"import os, types\n",

"import pandas as pd\n",

"\n",

"def \_\_iter\_\_(self): return 0\n",

"\n",

"print=(\"\\n\\n\\n----------------Contoured Image--------------------\")\n",

"plt.imshow(image, cmap=\"gray\")\n",

"plt.show()\n",

" \n",

"inp = np.array(preprocessed\_digits)\n"

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"text": [

"\n",

"\n",

"\n",

"----------------Contoured Image--------------------\n"

]

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"image/png": "\n"

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