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hw3.pv
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ECE472 Deep Learning
Professor Curro
MNIST Classification
import os
import numpy as np
from matplotlib import pyplot as plt
import matplotlib.image as mpimg
from tensorflow.keras.utils import to categorical
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout, Conv2D, MaxPooling2D
from tensorflow.keras.optimizers import SGD
import pandas as pd
import tensorflow as tf
from keras.layers.core import Flatten
from sklearn.model_selection import train_test_split
script_path = os.path.dirname(os.path.realpath(__file__))
function for loading csv and adjusting attributes for model use
def load_data(path):
    df = pd.read_csv(path)
    df = df.values # load values
    np.random.shuffle(df) # shuffle dataset
    x = df[:, 1:].reshape(-1, 28, 28, 1) # reshape for model
    y = df[:, 0].astype(np.int32)
    v = tf.keras.utils.to categorical(
        y, 10
    ) # convert y_train to categorical by one-hot-encoding
    return x, y
X train, Y train = load data(script path + "/mnist train.csv") # loading
X test, Y test = load data(script path + "/mnist test.csv")
X_train, X_valid, Y_train, Y_valid = train_test_split(
    X_train, Y_train, test_size=0.1
) # splitting validation from train
X = X \text{ test}
X_train = X_train.astype("float32") # cast/convert type of array
X_test = X_test.astype("float32")
X_valid = X_valid.astype("float32")
X train /= 255
X test /= 255
X valid /= 255
# https://www.kaggle.com/code/cdeotte/how-to-choose-cnn-architecture-mnist/noteb
# informed architecture choices
model = Sequential()
model.add(
    Conv2D(
        32,
        (3, 3),
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         activation="relu",
        kernel initializer="he uniform".
        padding="same",
        input shape=(28, 28, 1),
model.add(
    Conv2D(
        64,
         (3.3).
        activation="relu",
         kernel initializer="he uniform",
        padding="same",
        input_shape=(28, 28, 1),
model.add(MaxPooling2D((2, 2)))
model.add(
    Conv2D(
        64, (3, 3), activation="relu", kernel_initializer="he_uniform", padding="sa
me"
model.add(
    Conv2D (
        64, (3, 3), activation="relu", kernel_initializer="he_uniform", padding="sa
me"
model.add(MaxPooling2D((3, 3)))
model.add(
    Conv2D(
        128, (3, 3), activation="relu", kernel_initializer="he_uniform", padding="sa
me"
model.add(
    Conv2D (
        128, (3, 3), activation="relu", kernel_initializer="he_uniform", padding="sa
me"
model.add(MaxPooling2D((3, 3)))
# add flatten
model.add(Flatten())
model.add(Dense(128, activation="relu", kernel_initializer="he_uniform"))
model.add(Dense(10, activation="softmax"))
model.compile(
    optimizer=SGD (learning_rate=0.01, momentum=0.9), loss="mse", metrics=["accurac
v"1
) # L2
# https://www.tensorflow.org/datasets/keras_example
  admit to changing hyperparameters based on performance
 mainly b/c i decided on using a more robust network, so
 i lowered the number of epochs for runtime
history = model.fit(
    X_train,
    Y train,
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    epochs=2,
batch_size=64,
    validation_data=(X_valid, Y_valid),
    verbose=1,
score = model.evaluate(X_test, Y_test, verbose=1)
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Epoch 1/2 844/844 [==== Epoch 2/2