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# Check whether a simple NN having one neuron can learn a linear
function, e.g.,  $y = ax + b$ .
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#
# How to run:
# $ Python_Path ProgramFile_Path
# $ Tensorflow/bin/python DeepLearning/1D_LinearFunction_Learner.py

from tensorflow.keras.layers import Input, Dense
from tensorflow.keras import Model
import numpy as np
from tensorflow.keras.callbacks import EarlyStopping, History
import matplotlib.pyplot as plt

DIR = '/home/bibrity/DeepLearning/'

def main():
    # Build a model
    model = build_model()

    # Prepare data sets
    testX, testY, trainX, trainY = prepare_datasets()

    # Train the model
    callbackList = [EarlyStopping(monitor = 'val_loss', patience =
20), History()]
    history = model.fit(trainX, trainY, epochs = 300, batch_size =
32, callbacks = callbackList, validation_split = 0.2)
    plot_loss(history)

    # Test the performance
    predictedY = model.predict(testX)
    print(testY)
    print(predictedY)

    predictedA = model.layers[1].get_weights()[0][0][0]
    predictedB = model.layers[1].get_weights()[1][0]
    print('a: {}, b: {}'.format(predictedA, predictedB))

def prepare_datasets():
    testX = np.arange(100)
    testY = hidden_function(testX)
    trainX = np.arange(100, 65000)
    trainY = hidden_function(trainX)

    return testX, testY, trainX, trainY

def plot_loss(history):
    loss = history.history['loss']
    valLoss = history.history['val_loss']
    epochs = range(1, len(loss) + 1)

    plt.figure(figsize = (20, 20))
    plt.rcParams['font.size'] = '20'
    plt.plot(epochs, loss, 'bo-', label = 'Training loss')

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plt.plot(epochs, valLoss, 'k*-', label = 'Validation loss')
plt.title('Training Loss Vs. Validation Loss')
plt.legend()

figPath = DIR + 'TrainvsVal_Loss.png'
plt.savefig(figPath)
plt.close()

def hidden_function(x):
    a = 5; b = 3
    y = a * x + b

    return y

def build_model():
    # Layers of the model.
    inputs = Input(1,)
    outputs = Dense(1)(inputs)

    # Build the model.
    model = Model(inputs, outputs)

    # Configures the model for training.
    model.compile(loss = 'mse', optimizer = 'rmsprop')

    # Display the model architecture.
    model.summary()

    return model

if __name__ == '__main__':
    main()

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