import csv import numpy as np from matplotlib import pyplot as plt from matplotlib import patches as mpatches from sklearn import preprocessing #import tensorflow as tf from tensorflow import keras from tensorflow.keras.callbacks import TensorBoard import keras.backend as K WARNING:tensorflow:From C:\Anaconda\lib\site-packages\tensorflow\python\compat\v2_compat.py:96: disable_resource_variables (from tensorflow.python.ops.variable_scope) is deprecated and will be removed in a future version. Instructions for updating: non-resource variables are not supported in the long term In [4]: linkLength = 2 Q1 = []Q2 = []Q3 = []posX = []posY = []titaEnd = []samples = 2000**def** Xe (a,b,c): In [7]: # return the X,Y,Tita for a given 2 joint angles return linkLength*math.cos(a)+linkLength*math.cos(a+b)+linkLength*math.cos(a+b+c) **def** Ye (e,f,g): return linkLength*math.sin(e) + linkLength*math.sin(e + f) + linkLength*math.sin(e + f + g) def tita (h,i,j): return math.degrees(h)+math.degrees(j)+math.degrees(j) def customloss (yTrue, yPred): In [8]: return K.sum((yTrue - yPred)**2) def build_model(): # NN Model model = keras.Sequential() model.add(keras.layers.Dense(3)) #model.add(keras.layers.Dense(100, use_bias=True, activation='relu')) #model.add(keras.layers.Dense(100, use_bias=True, activation='relu')) model.add(keras.layers.Dense(100, use_bias=True, activation='relu')) model.add(keras.layers.Dense(50, use_bias=True, activation='relu')) model.add(keras.layers.Dense(25, use_bias=True, activation='relu')) model.add(keras.layers.Dense(3, use_bias=True, activation='linear')) model.compile(optimizer=tf.train.AdamOptimizer(0.01), loss=customloss, metrics=['accuracy']) # mean squared error return model file = open ("traing_data.csv", "w") # Data Set Creation for i in range (0, samples): q1= round(random.uniform(0, math.pi), 2) q2= round(random.uniform(-math.pi,0),2) q3= round(random.uniform(-math.pi/2, math.pi/2), 2) Q1.append(q1) file.write(str(q1)) file.write(",") Q2.append(q2) file.write(str(q2)) file.write(",") Q3.append(q3) file.write(str(q3)) file.write(",") X = Xe(q1, q2, q3)posX.append(X)file.write(str(round(X, 2))) file.write(",") Y = Ye(q1, q2, q3)posY.append(Y) file.write(str(round(Y, 2))) file.write(",") T = tita(q1,q2,q3)titaEnd.append(T) file.write(str(round(T, 2))) file.write("\n") file.close() for i in range(0,len(posX)): In [14]: plt.plot([posX[i],posX[i]+0.2*math.cos(math.radians(titaEnd[i]))], [posY[i],posY[i]+0.2*math.sin(math.radians(titaEnd[i]))], 'k-') #Plotting the data set plt.scatter(posX, posY) plt.xlabel("X Axis") plt.ylabel("Y Axis") plt.title("Data set of 2000 possible endeffector positions and orientations") Out[14]: Text(0.5, 1.0, 'Data set of 2000 possible endeffector positions and orientations') Data set of 2000 possible endeffector positions and orientations X Axis dataMat =np.c_[Q1,Q2,Q3,posX,posY,titaEnd] # Augmenting to the data marix In [10]: for i in range (0, samples): # Removing duplicated end effector positions check1 = dataMat[i,3]check2 = dataMat[i, 4]check3 = dataMat[i,5]for j in range (0, samples): **if** i != j: #print(i,j,'checking..') if $(dataMat[j,3] == check1 \ and \ dataMat[j,4] == check2 \ and \ dataMat[j,5] == check3)$: print(i, j, dataMat[j, 3], dataMat[j, 4], dataMat[j, 5]) 1291 1556 4.511551809575435 2.1076857876216626 -20.053522829578803 1556 1291 4.511551809575435 2.1076857876216626 -20.053522829578803 In [11]: data =dataMat[:,[3,4,5]] #X,Y,Tita output =dataMat[:,[0,1,2]] # Q1,Q2,Q3 train_input = data[0:int(0.7*samples),:] #Separate data set in to Train, Test And Validation train_output = output[0:int(0.7*samples),:] test_input = data[int(0.7*samples):int(0.85*samples),:] test_output = output[int(0.7*samples):int(0.85*samples),:] validate_input = data[int(0.85*samples):int(samples),:] validate_output = output[int(0.85*samples):int(samples),:] print("Train INPUT----") print(np.shape(train_input)) print(train_input) print("OUTPUT-----") print(np.shape(output)) print(output) Train INPUT-----(1400, 3) [[-0.9391886 4.35367827 107.71606548] 2.59982349 0.19221537 -56.14986392] 3.17188899 -0.64199706 -34.37746771] -1.14218422 -158.13635146] -1.62020541 -2.48693856 5.33926192 115.16451682¹ 4.85883363 -1.3829172 -6.30253575]] OUTPUT-----(2000, 3) [[2.61 -1.81 1.08] [1.83 -1.87 -0.94] [1.47 -2.32 0.25] [2.77 -1.61 -0.51] [1.64 -2.16 -0.44] [0.42 -1.5 1.5]] In [12]: x_scaler = preprocessing.MinMaxScaler(feature_range=(-1,1)) y_scaler = preprocessing.MinMaxScaler(feature_range=(-1,1)) x_scaler_test = preprocessing.MinMaxScaler(feature_range=(-1,1)) y_scaler_test = preprocessing.MinMaxScaler(feature_range=(-1,1)) x_scaler_eva = preprocessing.MinMaxScaler(feature_range=(-1,1)) y_scaler_eva = preprocessing.MinMaxScaler(feature_range=(-1,1)) dataX = x_scaler.fit_transform(train_input) dataY = y_scaler.fit_transform(train_output) dataX_test = x_scaler_test.fit_transform(test_input) dataY_test = y_scaler_test.fit_transform(test_output) dataX_eva = x_scaler_eva.fit_transform(validate_input) dataY_eva = y_scaler_eva.fit_transform(validate_output) In [25]: model = build_model() # Building the model history = model.fit(dataX, dataY, nb_epoch=200) #train the model [loss, mae] =model.evaluate(dataX_test, dataY_test, verbose=0) #evaluation print("Testing set Mean Abs Error: \${:7.2f}".format(mae)) WARNING:tensorflow:The `nb_epoch` argument in `fit` has been renamed `epochs`. Train on 1400 samples Epoch 1/200 Epoch 2/200 Epoch 3/200 Epoch 5/200 Epoch 6/200 Epoch 7/200 Epoch 9/200 Epoch 10/200 Epoch 11/200 Epoch 12/200 Epoch 13/200 Epoch 14/200 Epoch 15/200 Epoch 16/200 Epoch 17/200 Epoch 18/200 Epoch 19/200 Epoch 20/200 Epoch 21/200 Epoch 22/200 Epoch 23/200 Epoch 24/200 Epoch 25/200 Epoch 26/200 Epoch 27/200 Epoch 28/200 Epoch 29/200 Epoch 30/200 Epoch 31/200 Epoch 32/200 Epoch 33/200 Epoch 34/200 Epoch 35/200 Epoch 36/200 Epoch 37/200 Epoch 38/200 Epoch 39/200 Epoch 40/200 Epoch 41/200 Epoch 42/200 Epoch 43/200 Epoch 44/200 Epoch 45/200 Epoch 46/200 Epoch 47/200 Epoch 48/200 Epoch 49/200 Epoch 50/200 Epoch 51/200 Epoch 52/200 Epoch 53/200 Epoch 54/200 Epoch 55/200 Epoch 56/200 Epoch 57/200 Epoch 58/200 Epoch 59/200 Epoch 60/200 Epoch 61/200 Epoch 62/200 Epoch 63/200 Epoch 64/200 Epoch 65/200 Epoch 66/200 Epoch 67/200 Epoch 68/200 Epoch 69/200 Epoch 70/200 Epoch 71/200 Epoch 72/200 Epoch 73/200 Epoch 74/200 Epoch 75/200 Epoch 76/200 Epoch 77/200 Epoch 78/200 Epoch 79/200 Epoch 80/200 Epoch 81/200 Epoch 82/200 Epoch 83/200 Epoch 84/200 Epoch 85/200 Epoch 86/200 Epoch 87/200 Epoch 88/200 Epoch 89/200 Epoch 90/200 Epoch 91/200 Epoch 92/200 Epoch 93/200 Epoch 94/200 Epoch 95/200 Epoch 96/200 Epoch 97/200 Epoch 98/200 Epoch 99/200 Epoch 100/200 Epoch 101/200 Epoch 102/200 Epoch 103/200 Epoch 104/200 Epoch 105/200 Epoch 106/200 Epoch 107/200 Epoch 108/200 Epoch 109/200 Epoch 110/200 Epoch 111/200 Epoch 112/200 Epoch 113/200 Epoch 114/200 Epoch 115/200 Epoch 116/200 Epoch 117/200 Epoch 118/200 Epoch 119/200 Epoch 120/200 Epoch 121/200 Epoch 122/200 Epoch 123/200 Epoch 124/200 Epoch 125/200 Epoch 126/200 Epoch 127/200 Epoch 128/200 Epoch 129/200 Epoch 130/200 Epoch 131/200 Epoch 132/200 Epoch 133/200 Epoch 134/200 Epoch 135/200 Epoch 136/200 Epoch 137/200 Epoch 138/200 Epoch 139/200 Epoch 140/200 Epoch 141/200 Epoch 142/200 Epoch 143/200 Epoch 144/200 Epoch 145/200 Epoch 146/200 Epoch 147/200 Epoch 148/200 Epoch 149/200 Epoch 150/200 Epoch 151/200 Epoch 152/200 Epoch 153/200 Epoch 154/200 Epoch 155/200 Epoch 156/200 Epoch 157/200 Epoch 158/200 Epoch 159/200 Epoch 160/200 Epoch 161/200 Epoch 162/200 Epoch 163/200 Epoch 164/200 Epoch 165/200 Epoch 166/200 Epoch 167/200 Epoch 168/200 Epoch 169/200 Epoch 170/200 Epoch $171/\bar{2}00$ Epoch 172/200 Epoch 173/200 Epoch 174/200 Epoch 175/200 Epoch 176/200 Epoch 177/200 Epoch 178/200 Epoch 179/200 Epoch 180/200 Epoch 181/200 Epoch 182/200 Epoch 183/200 Epoch 184/200 Epoch 185/200 Epoch 186/200 Epoch 187/200 Epoch 188/200 Epoch 189/200 Epoch 190/200 Epoch 191/200 Epoch 192/200 Epoch 193/200 Epoch 194/200 Epoch 195/200 Epoch 196/200 Epoch 197/200 Epoch 198/200 Epoch 199/200 Testing set Mean Abs Error: \$ 0.83 dataX_input = x_scaler.transform(validate_input) In [20]: test_prediction = model.predict(dataX_input) #predict real_prediction =y_scaler.inverse_transform(test_prediction) plt.clf() plt.scatter(validate_output[:,0],real_prediction[:,0],c='b') # Plotting Actual angles(x: desired output(Joint angles used to genarate Xe, Ye nd Titae, y: output from pr plt.scatter(validate_output[:,1],real_prediction[:,1],c='g') plt.scatter(validate_output[:,2],real_prediction[:,2],c='r') plt.xlabel('True Values angles in rad') plt.ylabel('Predictions angles in rad') plt.title("True Value Vs Prediction") plt.legend("If all predicted values equal to the desired(true) value, this will be lie on 45 degree line") #plt.show(block = False)Out[20]: <matplotlib.legend.Legend at 0x26d492d5640> True Value Vs Prediction rad angles in 1 0 Predictions -1 -3 0 True Values angles in rad plt.savefig('True Value Vs Prediction.png') **#_Tensor Board** #tensorboard --logdir=logs/ print(validate_input[100,0]," ",validate_input[100,1]) print(Xe(real_prediction[100,0],real_prediction[100,1],real_prediction[100,2])," ",Ye(real_prediction[100,0],real_prediction[100,1],real_prediction[100,2]))

import tensorflow.compat.v1 as tf

tf.disable_v2_behavior()

import random import math

******** 2.7762071937488724 1.2390796503099544 2.743028565762362 1.1906972985494644 ******* <Figure size 432x288 with 0 Axes> In [31]: $single_data_1 = np.array([[5, 2, 60]])$ single_data = x_scaler.transform(single_data_1) single_prediction = model.predict(single_data) single_real_prediction = y_scaler.inverse_transform(single_prediction) print(single_data_1[0,0]," ", single_data_1[0,1]) print(Xe(single_real_prediction[0,0], single_real_prediction[0,1], single_real_prediction[0,2]), " ", Ye(single_real_prediction[0,0], single_real_prediction[0,1], single_real_prediction[0,2]), " ", Ye(single_real_prediction[0,0], single_real_prediction[0,1], single_real_prediction[0,2]), " ", Ye(single_real_prediction[0,0], single_real_prediction[0,1], single_re print("*****************************") 4.790735226042725 2.0810691858973818 model.evaluate(dataX_test,dataY_test) [3.3232549158732096, 0.82666665] Out[45]: