Data training done using High Frq dataset In [2]: from dataPrep import DataPreparation from modelling import Model In [3]: drivingProfile = pd.read_csv('./parameter.csv') data = pd.read csv('./HighFrq/HighFrq input.csv') In [4]: dp = DataPreparation(drivingProfile) data.head() In [5]: # data.shape Out[5]: Time Drv_DeltaSteer Veh_Vx 0.000 0.0 19.444444 **1** 0.005 0.0 19.444285 **2** 0.010 0.0 19.443796 **3** 0.015 0.0 19.443005 **4** 0.020 0.0 19.441990 In [6]: Time = data['Time'].to numpy() deltaSteer = data['Drv_DeltaSteer'].to_numpy() Vx = data['Veh Vx'].to numpy() In [7]: print(len(Vx)) 4601 In [8]: [time, output] = dp.getOutput(Time, deltaSteer, Vx) [time, input] = dp.getInput(Time, deltaSteer, Vx) output = np.array(output) input = np.array(input) In [9]: # plt.plot(time, output[0]) plt.plot(time, input[0]) plt.show() # plt.plot(time, output[1]) plt.plot(time, input[1]) 4000 2000 0 -2000 -4000-600015 20 10 [<matplotlib.lines.Line2D at 0x1f983211a80>] Out[9]: 0.06 0.04 0.02 0.00 -0.02-0.04-0.065 10 15 20 modelObj = Model(input=input, output=output) In [10]: [sysId_noised, x_id_noised, y_id_noised, output_noised] = modelObj.model_withNoise(len(input[0]), [0.001, 0.01] In [11]: In [12]: print(f"For yaw rate :\nR2 Score: {r2_score(output_noised[0], y_id_noised[0])}, MSE: {mean squared error(print(f"For inertial acceleration :\nR2 Score: {r2_score(output_noised[1], y_id_noised[1])}, For yaw rate : R2 Score: 0.903151265747005, MSE: 0.0009516725797754343 For inertial acceleration : R2 Score: 0.9957360780214753, MSE: 0.010531821640179767 In [13]: plt.plot(time, output noised[0]) # plt.show() plt.plot(time, y_id_noised[0]) plt.legend(['output', 'y_id']) <matplotlib.legend.Legend at 0x1f983252590> Out[13]: output 0.3 0.2 0.1 0.0 -0.1-0.2-0.35 20 10 15 Model testing with other datasets Data prep and prediction In [14]: drivingProfile = pd.read_csv('./parameter.csv') data slc = pd.read csv('./SLC/SLC input.csv') data slc.head() time Drv DeltaSteer Out[14]: Veh_Vx 0.000 0.0 30.000000 0.0 29.999664 **1** 0.005 0.0 29.998813 **2** 0.010 **3** 0.015 0.0 29.997596 **4** 0.020 0.0 29.996121 In [15]: dp_slc = DataPreparation(drivingProfile) In [16]: Time_slc = data_slc['time'].to_numpy() deltaSteer_slc = data_slc['Drv_DeltaSteer'].to_numpy() Vx_slc = data_slc['Veh_Vx'].to_numpy() In [17]: [time_slc, output_slc] = dp_slc.getOutput(Time_slc, deltaSteer_slc, Vx_slc) [time_slc, input_slc] = dp_slc.getInput(Time_slc, deltaSteer_slc, Vx_slc) In [18]: output_slc = np.array(output_slc) input_slc = np.array(input_slc) output_noised_slc = output_slc + fset.white_noise_var(len(output_slc[0]), [0.001, 0.01]) In [19]: x_id_slc, y_id_slc = modelObj.predict(sysId_noised, input_slc) Metrics In [21]: print(f"For yaw rate :\nR2 Score: {r2_score(output_noised_slc[0], y_id_slc[0])}, MSE: {mean squared error plt.plot(time_slc, output_noised_slc[0]) plt.plot(time slc, y id slc[0]) plt.legend(['output', 'yId']) plt.show() print(f"For inertial acceleration :\nR2 Score: {r2_score(output_noised_slc[1], y_id_slc[1])}, MSE: {mean plt.plot(time_slc, output_noised_slc[1]) plt.plot(time_slc, y_id_slc[1]) plt.legend(['output', 'yId']) plt.show() For yaw rate : R2 Score: 0.8878084797663299, MSE: 0.0024295908750604396 0.6 output 0.4 0.2 0.0 -0.2-0.4-0.68 10 For inertial acceleration : R2 Score: 0.9926000070178109, MSE: 0.10819782297819984 output 10 yld 5 0 -5 -100 2 10 Circular **Data prep and Prediction** In [22]: data_circular = pd.read_csv('./Circular/circular_input.csv') data_circular.head() Out[22]: samplingTime Drv_DeltaSteer Veh Vx 0 0.000 0.0 30.000000 1 0.005 29.999664 2 0.010 29.998813 3 0.015 0.0 29.997596 4 0.020 0.0 29.996121 deltaSteer circular = data circular['Drv DeltaSteer'].to numpy() Vx circular = data circular['Veh Vx'].to numpy() In [24]: dp_circular = DataPreparation(drivingProfile) In [25]: [time_circular, output_circular] = dp.getOutput(Time circular, deltaSteer circular, Vx circular) [time circular, input circular] = dp.getInput(Time circular, deltaSteer circular, Vx circular) output circular = np.array(output circular) input circular = np.array(input circular) output_noised_cicular = output_circular + fset.white_noise_var(len(output_circular[0]), [0.001, 0.01]) In [26]: In [27]: x_id_circular, y_id_circular = modelObj.predict(sysId_noised, input_circular) **Metrics** In [28]: print(f"For yaw rate :\nR2 Score: {r2 score(output noised cicular[0], y id circular[0])}, plt.plot(time circular, output noised cicular[0]) plt.plot(time circular, y id circular[0]) plt.legend(['output', 'yId']) plt.show() print(f"For inertial acceleration :\nR2 Score: {r2 score(output noised cicular[1], y id circular[1])}, plt.plot(time circular, output noised cicular[1]) plt.plot(time circular, y id circular[1]) plt.legend(['output', 'yId']) plt.show() For yaw rate : R2 Score: 0.8769965478856241, MSE: 0.0020753748873728333 0.5 output yld 0.4 0.3 0.2 0.1 0.0 -0.16 10 For inertial acceleration : R2 Score: 0.9931842554211441, MSE: 0.07326340596399787 output yld 10 8 6 4 2 0 2 8 10 0 **Data prep and Prediction** In [29]: data_jturn = pd.read_csv('./JTurn/jTurn_input.csv') data_jturn.head() data_jturn.shape (2001, 3)Out[29]: In [30]: dp_jturn = DataPreparation(drivingProfile) In [31]: Time_jturn = data_jturn['samplingTime'].to_numpy() deltaSteer_jturn = data_jturn['Drv_DeltaSteer'].to_numpy() Vx_jturn = data_jturn['Veh_Vx'].to_numpy() In [32]: [time_jturn, output_jturn] = (dp.getOutput(Time_jturn, deltaSteer_jturn, Vx_jturn)) [time_jturn, input_jturn] = (dp.getInput(Time_jturn, deltaSteer_jturn, Vx_jturn)) output_jturn = np.array(output_jturn) output_noised_jturn = output_jturn + fset.white_noise_var(len(output_jturn[0]), [0.001, 0.01]) input_jturn = np.array(input_jturn) In [33]: x_id_jturn, y_id_jturn = modelObj.predict(sysId_noised, input jturn) **Metrics** In [34]: print(f"For yaw rate :\nR2 Score: {r2_score(output_noised_jturn[0], y_id_jturn[0])}, MSE: {mean_squared_e plt.plot(time_jturn, output_noised_jturn[0]) plt.plot(time_jturn, y_id_jturn[0]) plt.legend(['output', 'yId']) plt.show() print(f"For inertial acceleration :\nR2 Score: {r2_score(output_noised_jturn[1], y_id_jturn[1])}, MSE: {m plt.plot(time_jturn, output_noised_jturn[0]) plt.plot(time_jturn, y_id_jturn[0]) plt.legend(['output', 'yId']) plt.show() For yaw rate : R2 Score: 0.8729390757836556, MSE: 0.0011296475487978495 output yld 0.25 0.20 0.15 0.10 0.05 0.00 -0.054 0 2 6 8 10 For inertial acceleration : R2 Score: 0.9957406405322687, MSE: 0.02213787618698999 output yld 0.25 0.20 0.15 0.10 0.05 0.00 -0.050 2 4 6 8 10 **VDA Data Prep and Prediction** In [35]: data_vda = pd.read_csv('./VDA/VDA_input.csv') data_vda.head() data_vda.shape (1801, 3)Out[35]: In [36]: dp_vda = DataPreparation(drivingProfile) In [37]: Time_vda = data_vda['samplingTime'].to_numpy() deltaSteer_vda = data_vda['Drv_DeltaSteer'].to_numpy() Vx_vda = data_vda['Veh_Vx'].to_numpy() In [38]: [time_vda, output_vda] = (dp.getOutput(Time_vda, deltaSteer_vda, Vx_vda)) [time_vda, input_vda] = (dp.getInput(Time_vda, deltaSteer_vda, Vx_vda)) output_vda = np.array(output_vda) output_noised_vda = output_vda + fset.white_noise_var(len(output_vda[0]), [0.001, 0.01]) input_vda = np.array(input_vda) In [39]: x_id_vda, y_id_vda = modelObj.predict(sysId_noised, input vda) **Metrics** In [40]: print(f"For yaw rate :\nR2 Score: {r2_score(output_noised_vda[0], y_id_vda[0])}, MSE: {mean_squared_error plt.plot(time_vda, output_noised_vda[0]) plt.plot(time_vda, y_id_vda[0]) plt.legend(['output', 'yId']) plt.show() print(f"For inertial acceleration :\nR2 Score: {r2_score(output_noised_vda[1], y_id_vda[1])}, MSE: {mean_ plt.plot(time_vda, output_noised_vda[1]) plt.plot(time_vda, y_id_vda[1]) plt.legend(['output', 'yId']) plt.show() For yaw rate : R2 Score: 0.9855472184623414, MSE: 0.0018872564075950035 output 0.75 yld 0.50 0.25 0.00 -0.25-0.50-0.75-1.00-1.250 2 4 6 8 For inertial acceleration : R2 Score: 0.9991242219624897, MSE: 0.050139821516996835 20 output yld 10 0 -10-202 6 8

In [1]: | from sklearn.metrics import r2 score

import pandas as pd
import numpy as np

from sklearn.metrics import mean squared error

from sippy import functionset as fset

import matplotlib.pyplot as plt