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import matplotlib as mpl
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
from sklearn.preprocessing import MinMaxScaler
from scipy import stats
import seaborn as sns
import tensorflow as tf
from tensorflow import keras
import sklearn
import sys
import io
n \text{ steps} = 50
forecast = 10
def plot_learning_curves(loss, val_loss):
    plt.figure()
    plt.plot(np.arange(len(loss)), loss, "b.-", label="Training loss")
   plt.plot(np.arange(len(val_loss)), val_loss, "r.-", label="Validation loss")
   plt.gca().xaxis.set_major_locator(mpl.ticker.MaxNLocator(integer=True))
    plt.legend(fontsize=14)
    plt.xlabel("Epochs")
    plt.ylabel("Loss")
   plt.grid(True)
#LOAD DATA
data west = pd.read csv('denoised data 1015 north.csv')
data west o = np.array(data west.FLOW)
# #EXRTEND DATA
array to concatinate = data west o[288:]
for iter in range (35):
    data west o = np.concatenate([data west o,array to concatinate])
# #SCALE AND RESHAPE DATA
scaler = MinMaxScaler()
array = data west o.reshape(-1, 1)
array_scaled = scaler.fit_transform(array)
flow reshaped = array scaled[:(len(array scaled) - (len(array scaled) % (n steps+forecast
print(flow reshaped.shape)
# #TRAIN SET, VALIDATION SET, TEST SET
test = int(0.7 * flow_reshaped.shape[0])
valid = int(0.9 * flow reshaped.shape[0])
X_train= flow_reshaped[:test, :n_steps] #first 50, last 10
X_valid = flow_reshaped[test:valid, :n_steps]
X test = flow reshaped[valid:, :n steps]
#prepare targets
Y = np.empty((flow reshaped.shape[0], n steps, forecast))
for step ahead in range(1, forecast + 1):
    Y[:, :, step ahead - 1] = flow reshaped[:, step ahead:step ahead + n steps, 0]
y train = Y[:test]
y valid = Y[test:valid]
y test = Y[valid:]
# # NEW MODEL TO FIT HYPERPARAMETERS
model = keras.models.Sequential([
   keras.layers.SimpleRNN(20, return_sequences=True, input_shape=[None, 1]),
    keras.layers.SimpleRNN(20, return_sequences=True),
    keras.layers.TimeDistributed(keras.layers.Dense(forecast))
])
def last time step mse(Y true, Y pred):
    return keras.metrics.mean squared error(Y true[:, -1], Y pred[:, -1])
model.compile(loss="mse", optimizer="adam", metrics=[last_time_step_mse])
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early stopping cb = keras.callbacks.EarlyStopping(patience=10, restore best weights=True)
history = model.fit(X_train, y_train, epochs=700,
                                             validation_data=(X_valid, y_valid), callbacks=[early_stopping_cb])
model.save("rnn 1014 norht.h5")
plot_learning_curves(history.history["loss"], history.history["val_loss"])
#50 minutes forecast
flow_unscaled = array[:(len(array) - (len(array) % (n_steps + forecast)))].reshape(-1, (n
y test_unscaled = flow_unscaled[valid:, n_steps:, 0]
y_real_rescaled = y_test_unscaled[-1, :].reshape(-1, 1)
print(y real rescaled.shape)
flow_not_reshaped = array[:(len(array) - (len(array) % (n_steps+forecast)))]
#flow prediction
y_pred = model.predict(X_test[-1, :].reshape(-1, n_steps, 1)) #shape (1, 50, 10)
y_{pred} = y_{pred}[-1,-1,:].reshape(-1,1)
y pred rescaled = scaler.inverse_transform(y_pred).reshape(-1, 1) #shape (10, 1)
print(y_pred_rescaled.shape)
#time
time_not_reshaped = np.array(data_west['TIME'][:(len(data_west['TIME']) - (len(data_west[
time_reshaped = np.array(data_west['TIME'][:(len(data_west['TIME']) - (len(data_west['TIME')) - (len(data_west['TIME'
         reshape(-1, (n steps+forecast), 1)
valid time = int(0.9 * time reshaped.shape[0])
y time test = time reshaped[valid time:, n steps:, 0]
print(y time test[-1, :].shape)
def plot prediction(y real resacled, y pred rescaled, flow not reshaped, time not reshape
         plt.figure()
         plt.title("50 minutes prediction", fontsize=14)
         plt.plot(time_not_reshaped[-300:-forecast], flow_not_reshaped[-300:-forecast], 'b-')
        plt.plot(y_time_test[-1, :], y_real_resacled, 'ro-', label = 'Real values')
plt.plot(y_time_test[-1, :], y_pred_rescaled, 'gx-', label = 'Predicted values')
         plt.legend(loc="upper left")
        plt.xlabel("Time")
plt.ylabel('Volume')
plot_prediction(y_real_rescaled, y_pred_rescaled, flow_not_reshaped, time_not_reshaped, y
plt.show()
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(10132, 60, 1)
Train on 7092 samples, validate on 2026 samples
Epoch 1/700
Epoch 2/700
Epoch 3/700
Epoch 4/700
Epoch 5/700
Epoch 6/700
Epoch 7/700
Epoch 8/700
Epoch 9/700
Epoch 10/700
Epoch 11/700
Epoch 12/700
Epoch 13/700
7092/7092 [============] - 3s 479us/sample - loss: 0.0106 - 1
Epoch 14/700
Epoch 15/700
Epoch 16/700
Epoch 17/700
Epoch 18/700
Epoch 19/700
Epoch 20/700
Epoch 21/700
Epoch 22/700
Epoch 23/700
Epoch 24/700
Epoch 25/700
Epoch 26/700
Epoch 27/700
Epoch 28/700
Enoch 29/700
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7092/7092 [============= ] - 3s 484us/sample - loss: 0.0103 - 1
Epoch 30/700
Epoch 31/700
7092/7092 [============================] - 3s 478us/sample - loss: 0.0102 -
Epoch 32/700
Epoch 33/700
Epoch 34/700
Epoch 35/700
Epoch 36/700
Epoch 37/700
Epoch 38/700
Epoch 39/700
Epoch 40/700
Epoch 41/700
Epoch 42/700
Epoch 43/700
Epoch 44/700
Epoch 45/700
Epoch 46/700
7092/7092 [============] - 3s 474us/sample - loss: 0.0100 - 1
Epoch 47/700
Epoch 48/700
Epoch 49/700
Epoch 50/700
Epoch 51/700
Epoch 52/700
Epoch 53/700
Epoch 54/700
7092/7092 [============] - 3s 480us/sample - loss: 0.0099 - 1
Epoch 55/700
Epoch 56/700
Epoch 57/700
Epoch 58/700
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Epoch 59/700
Epoch 60/700
Epoch 61/700
Epoch 62/700
Epoch 63/700
Epoch 64/700
Epoch 65/700
Epoch 66/700
Epoch 67/700
Epoch 68/700
Epoch 69/700
Epoch 70/700
Epoch 71/700
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Epoch 79/700
Epoch 80/700
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Epoch 82/700
Epoch 83/700
Epoch 84/700
Epoch 85/700
Epoch 86/700
Epoch 87/700
Epoch 88/700
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7092/7092 [==============] - 3s 469us/sample - loss: 0.0086 - 1
Epoch 89/700
Epoch 90/700
Epoch 91/700
Epoch 92/700
Epoch 93/700
Epoch 94/700
Epoch 95/700
Epoch 96/700
7092/7092 [=============] - 3s 469us/sample - loss: 0.0083 - 1
Epoch 97/700
7092/7092 [=============] - 3s 479us/sample - loss: 0.0084 - 1
Epoch 98/700
Epoch 99/700
Epoch 100/700
Epoch 101/700
Epoch 102/700
Epoch 103/700
Epoch 104/700
Epoch 105/700
7092/7092 [=============] - 3s 472us/sample - loss: 0.0081 - 1
Epoch 106/700
Epoch 107/700
7092/7092 [==============] - 3s 466us/sample - loss: 0.0081 - 1
Epoch 108/700
Epoch 109/700
Epoch 110/700
Epoch 111/700
Epoch 112/700
Epoch 113/700
Epoch 114/700
Epoch 115/700
Epoch 116/700
Epoch 117/700
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Epoch 118/700
Epoch 119/700
Epoch 120/700
Epoch 121/700
Epoch 122/700
Epoch 123/700
7092/7092 [============================] - 3s 492us/sample - loss: 0.0078 -
Epoch 124/700
Epoch 125/700
Epoch 126/700
Epoch 127/700
Epoch 128/700
Epoch 129/700
Epoch 130/700
Epoch 131/700
Epoch 132/700
7092/7092 [=============] - 4s 494us/sample - loss: 0.0076 - 1
Epoch 133/700
Epoch 134/700
Epoch 135/700
Epoch 136/700
Epoch 137/700
7092/7092 [============] - 3s 486us/sample - loss: 0.0076 - 1
Epoch 138/700
Epoch 139/700
Epoch 140/700
Epoch 141/700
Epoch 142/700
Epoch 143/700
Epoch 144/700
Epoch 145/700
Epoch 146/700
Epoch 147/700
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Epoch 162/700
Epoch 163/700
Epoch 164/700
7092/7092 [============] - 3s 468us/sample - loss: 0.0074 - 1
Epoch 165/700
Epoch 166/700
Epoch 167/700
Epoch 168/700
Epoch 169/700
Epoch 170/700
7092/7092 [============] - 3s 474us/sample - loss: 0.0073 - 1
Epoch 171/700
Epoch 172/700
Epoch 173/700
Epoch 174/700
Epoch 175/700
Epoch 176/700
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