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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

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)

n_steps = 50
forecast = 10

#EXTRACT FLOW, Z SCORE, OUTLIERS
data_west = pd.read_csv('denoised_data_1015_east.csv')
data_west_o = np.array(data_west.FLOW)

#EXRTEND DATA
array_to_concatinate = data_west_o
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_resaped = array_scaled[:len(array_scaled) - (len(array_scaled) % (n_steps+forecast))]

#TRAIN SET, VALIDATION SET, TEST SET
test = int(0.7 * flow_resaped.shape[0])
valid = int(0.9 * flow_resaped.shape[0])

X_train = flow_resaped[:test, :n_steps]
X_valid = flow_resaped[test:valid, :n_steps]
X_test = flow_resaped[valid:, :n_steps]
print(X_test.shape)
print(X_test[-1:].shape)

#prepare targets
Y = np.empty((flow_resaped.shape[0], n_steps, forecast))
for step_ahead in range(1, forecast + 1):
    Y[:, :, step_ahead - 1] = flow_resaped[:, step_ahead:step_ahead + n_steps, 0]

y_train = Y[:test]
y_valid = Y[test:valid]
y_test = Y[valid:]

#DEFINE AND COMPILE MODEL
model = keras.models.Sequential([
    keras.layers.GRU(20, return_sequences=True, input_shape=[None, 1]),
    keras.layers.GRU(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])

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model.compile(loss="mse", optimizer="adam", metrics=[last_time_step_mse])

early_stopping_cb = keras.callbacks.EarlyStopping(patience=15, 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("gru_1016_east.h5")
plot_learning_curves(history.history["loss"], history.history["val_loss"])

#50 MIN FORECAST
#flow
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_rescaled = 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_rescaled = np.array(data_test['TIME'][:,len(data_test['TIME']) - (len(data_test[
time_rescaled = np.array(data_test['TIME'][:,len(data_test['TIME']) - (len(data_test['TIM
    reshaped(-1, (n_steps+forecast), 1))

valid_time = int(0.9 * time_rescaled.shape[0])
y_time_test = time_rescaled[valid_time:, n_steps:, 0]
print(y_time_test[-1, :].shape)

def plot_prediction(y_real_rescaled, y_pred_rescaled, flow_not_rescaled, time_not_rescaled):
    plt.figure()
    plt.title("50 minutes prediction", fontsize=14)
    plt.plot(time_not_rescaled[-300:-forecast], flow_not_rescaled[-300:-forecast], 'b-')
    plt.plot(y_time_test[-1, :], y_real_rescaled, '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 (in 5 minutes intervals)")
    plt.ylabel('Volume (veh/h)')

plot_prediction(y_real_rescaled, y_pred_rescaled, flow_not_rescaled, time_not_rescaled, y
plt.show()

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7499/7499 [=====] - 11s 2ms/sample - loss: 0.0047 - 1a
Epoch 326/700
7499/7499 [=====] - 11s 2ms/sample - loss: 0.0047 - 1a
Epoch 327/700
7499/7499 [=====] - 11s 2ms/sample - loss: 0.0047 - 1a
Epoch 328/700
7499/7499 [=====] - 11s 1ms/sample - loss: 0.0047 - 1a
Epoch 329/700
7499/7499 [=====] - 11s 1ms/sample - loss: 0.0047 - 1a
Epoch 330/700
7499/7499 [=====] - 11s 1ms/sample - loss: 0.0048 - 1a
Epoch 331/700
7499/7499 [=====] - 11s 2ms/sample - loss: 0.0049 - 1a
Epoch 332/700
7499/7499 [=====] - 11s 1ms/sample - loss: 0.0047 - 1a
Epoch 333/700
7499/7499 [=====] - 11s 2ms/sample - loss: 0.0047 - 1a
Epoch 334/700
7499/7499 [=====] - 11s 1ms/sample - loss: 0.0047 - 1a
Epoch 335/700
7499/7499 [=====] - 12s 2ms/sample - loss: 0.0047 - 1a
Epoch 336/700
7499/7499 [=====] - 11s 1ms/sample - loss: 0.0047 - 1a
(10, 1)
(10, 1)
(10,)

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