

```

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_1016_east.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_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:]

#MODEL
def last_time_step_mse(Y_true, Y_pred):
    return keras.metrics.mean_squared_error(Y_true[:, -1], Y_pred[:, -1])

model = keras.models.Sequential()
model.add(keras.layers.InputLayer(input_shape=[None, 1]))
for rate in (1, 2, 4, 8) * 2:
    model.add(keras.layers.Conv1D(filters=20, kernel_size=2, padding="causal",
                                   activation="relu", dilation_rate=rate))

```

```

model.add(keras.layers.Conv1D(filters=10, kernel_size=1))
model.compile(loss="mse", optimizer="adam", metrics=[last_time_step_mse])

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("vawenet_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_steps + forecast))
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)))].reshape(-1, (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_west['TIME'][:, (len(data_west['TIME']) - (len(data_west['TIME']) % (n_steps + forecast)))])
time_rescaled = np.array(data_west['TIME'][:, (len(data_west['TIME']) - (len(data_west['TIME']) % (n_steps + forecast)))])
time_rescaled.reshape(-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, y_time_test):
    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")
    plt.ylabel('Volume')

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

```



```

Epoch 91/700
7381/7381 [=====] - 5s 675us/sample - loss: 0.0036 - 
Epoch 92/700
7381/7381 [=====] - 5s 677us/sample - loss: 0.0036 - 
Epoch 93/700
7381/7381 [=====] - 5s 671us/sample - loss: 0.0036 - 
Epoch 94/700
7381/7381 [=====] - 5s 679us/sample - loss: 0.0036 - 
Epoch 95/700
7381/7381 [=====] - 5s 679us/sample - loss: 0.0036 - 
Epoch 96/700
7381/7381 [=====] - 5s 721us/sample - loss: 0.0036 - 
Epoch 97/700
7381/7381 [=====] - 5s 682us/sample - loss: 0.0036 - 
Epoch 98/700
7381/7381 [=====] - 5s 668us/sample - loss: 0.0036 - 
Epoch 99/700
7381/7381 [=====] - 5s 673us/sample - loss: 0.0036 - 
Epoch 100/700
7381/7381 [=====] - 5s 684us/sample - loss: 0.0035 - 
Epoch 101/700
7381/7381 [=====] - 5s 674us/sample - loss: 0.0035 - 
Epoch 102/700
7381/7381 [=====] - 5s 683us/sample - loss: 0.0035 - 
(10, 1)
(10, 1)
(10,)

```



