```
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 south.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 south.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'
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
7092/7092 [============================ ] - 4s 503us/sample - loss: 0.0072 - 1
Epoch 9/700
Epoch 10/700
Epoch 11/700
7092/7092 [=============== ] - 3s 492us/sample - loss: 0.0071 - 1
Epoch 12/700
Epoch 13/700
7092/7092 [=============== ] - 3s 485us/sample - loss: 0.0071 - 1
Epoch 14/700
7092/7092 [============== ] - 3s 480us/sample - loss: 0.0071 - 1
Epoch 15/700
Epoch 16/700
Epoch 17/700
7092/7092 [=============== ] - 3s 490us/sample - loss: 0.0070 - 1
Epoch 18/700
Epoch 19/700
Epoch 20/700
Epoch 21/700
Epoch 22/700
Epoch 23/700
7092/7092 [=============== ] - 3s 491us/sample - loss: 0.0069 - 1
Epoch 24/700
Epoch 25/700
7092/7092 [=============] - 3s 476us/sample - loss: 0.0070 - 1
Epoch 26/700
Epoch 27/700
Epoch 28/700
7092/7092 [=============] - 4s 501us/sample - loss: 0.0069 - 1
Enoch 29/700
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7092/7092 [============= ] - 3s 489us/sample - loss: 0.0069 - ]
Epoch 30/700
Epoch 31/700
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
7092/7092 [=============== ] - 4s 498us/sample - loss: 0.0066 - 1
Epoch 41/700
Epoch 42/700
7092/7092 [============== ] - 4s 516us/sample - loss: 0.0066 - ]
Epoch 43/700
Epoch 44/700
Epoch 45/700
Epoch 46/700
Epoch 47/700
Epoch 48/700
Epoch 49/700
Epoch 50/700
Epoch 51/700
Epoch 52/700
Epoch 53/700
Epoch 54/700
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
7092/7092 [=============== ] - 3s 491us/sample - loss: 0.0063 - 1
Epoch 70/700
7092/7092 [=============== ] - 3s 487us/sample - loss: 0.0062 - 1
Epoch 71/700
Epoch 72/700
Epoch 73/700
7092/7092 [=============== ] - 3s 476us/sample - loss: 0.0062 - 1
Epoch 74/700
Epoch 75/700
Epoch 76/700
7092/7092 [=============== ] - 3s 486us/sample - loss: 0.0062 - 1
Epoch 77/700
7092/7092 [=========================== ] - 4s 513us/sample - loss: 0.0062 - 1
Epoch 78/700
Epoch 79/700
7092/7092 [=============== ] - 3s 478us/sample - loss: 0.0062 - 1
Epoch 80/700
7092/7092 [=============== ] - 3s 476us/sample - loss: 0.0062 - 1
Epoch 81/700
7092/7092 [=============== ] - 3s 479us/sample - loss: 0.0062 - 1
Epoch 82/700
7092/7092 [=============== ] - 3s 485us/sample - loss: 0.0062 - 1
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 470us/sample - loss: 0.0062 - ]
Epoch 89/700
Epoch 90/700
Epoch 91/700
Epoch 92/700
Epoch 93/700
Epoch 94/700
7092/7092 [=============] - 4s 497us/sample - loss: 0.0061 - ]
Epoch 95/700
       7092/7092 [======
Epoch 96/700
Epoch 97/700
7092/7092 [=============] - 3s 481us/sample - loss: 0.0062 - ]
Epoch 98/700
(10, 1)
(10, 1)
(10,)
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