Time series prediction with LSTM (student notebook)

Neural networks like Long Short-Term Memory (LSTM) recurrent neural networks are able to almost seamlessly model problems with multiple input variables.

This is a great benefit in time series forecasting, where classical linear methods can be difficult to adapt to multivariate or multiple input forecasting problems.

In this lab, you will discover how you can develop an LSTM model for multivariate time series forecasting with the Keras deep learning library.

Out[]: run previous cell, wait for 2 seconds

Imports

```
import ssl
ssl._create_default_https_context = ssl._create_unverified_context
```

```
# plottinh
         import matplotlib.pvplot as plt
         import plotly.express as px
In [ ]:
         # data
         import math
         import numpy as np
         import pandas as pd
In [ ]:
         # ML
         import tensorflow as tf
         from sklearn.compose import ColumnTransformer
         from sklearn.impute import SimpleImputer
         from sklearn.metrics import mean squared error
         from sklearn.pipeline import FeatureUnion, Pipeline
         from sklearn.preprocessing import MinMaxScaler, OneHotEncoder, StandardScaler
         from tensorflow.keras import utils
         from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint
         from tensorflow.keras.layers import (LSTM, AveragePooling1D, Bidirectional,
                                              Dense, Embedding, Flatten, Input,
                                              RepeatVector)
         from tensorflow.keras.models import Model, load model
```

LSTM network for uni-variate time series

LSTM can be used to model univariate time series forecasting problems.

These problems consist of a single set of observations and a model is needed to learn from the past set of observations in order to forecast the next value in the sequence.

We will demonstrate a number of variations of the LSTM model for univariate time series forecasting.

Be careful not to draw hasty conclusions about the relative performance of the models. The number of layers or neurons are highly variable between models.

Data preparation

A first example

Consider a given univariate sequence: [10, 20, 30, 40, 50, 60, 70, 80, 90]

We can divide the sequence into multiple input/output patterns called samples, where three time steps are used as input and one time step is used as output for the one-step prediction that is being learned.

```
X, y
10, 20, 30 40
20, 30, 40 50
30, 40, 50 60
```

The series_to_supervised() function below implements this behavior and will split a given univariate sequence into multiple samples where each sample has a specified number of time steps (n_in, by default 3) and the output has also a specified number of time steps (n out, by default 1).

By default, the data to predict is the last columns.

```
In [ ]:
         def series to supervised(data, n in=3, n out=1, output=None, dropnan=True):
              n vars = 1 if type(data) is list else data.shape[1]
              output = [data.columns[-1]] if output is None else output
              df = pd.DataFrame(data)
              cols, names = list(), list()
              # input sequence (t-n, ... t-1)
              for i in range(n in, 0, -1):
                  cols += [df.shift(i)]
                  names += [f"{col}(t-{i})" for col in data.columns]
              # forecast sequence (t, t+1, ... t+n)
              for i in range(0, n out):
                  cols += [df[output].shift(-i)]
                  if i == 0:
                      names += [f"{j}(t)" for j in output]
                  else:
                      names += [f''\{j\}(t+\{i\})'' \text{ for } j \text{ in } output]
              # put it all together
```

```
agg = pd.concat(cols, axis=1)
agg.columns = names

# drop rows with NaN values
if dropnan:
    agg.dropna(inplace=True)
return agg
```

TODO – Students

• create the time series mentioned in the first exemple.

```
In [ ]:
         my_time_series = pd.DataFrame(list(range(10,100,10)))
         my_time_series
Out[]:
        0 10
        1 20
        2 30
        3 40
        4 50
        5 60
        6 70
        7 80
        8 90
In [ ]:
         data = series_to_supervised(my_time_series, n_in=3, n_out=1)
         data.head()
```

```
Out[]:
             0(t-3) 0(t-2) 0(t-1) 0(t)
             10.0
                    20.0
                          30.0
                                40
          3
              20.0
                    30.0
                          40.0
                                50
              30.0
                    40.0
                          50.0
                                60
                    50.0
                          60.0
                                70
              40.0
                    60.0 70.0 80
              50.0
```

Head of the previous dataset

```
0(t-3) 0(t-2) 0(t-1) 0(t)
   0.0 10.0
                20.0
                        30
   10.0
            20.0
                    30.0
                            40
   20.0
           30.0
                    40.0
                            50
5
                            60
6
   30.0
           40.0
                    50.0
                            70 </font>
   40.0
           50.0
                    60.0
```

Do the same for a more sophisticated series

```
In []:
    SIZE = 250
    time_stamps = range(SIZE)

fct = lambda x: x*math.sin(x)
    time_series = pd.DataFrame(data={"data":[fct(x) for x in range(SIZE)]})

n_features = my_time_series.shape[1] # for univariate time series
time_series
```

```
Out[]:
                      data
            0
                 0.000000
            1
                 0.841471
            2
                 1.818595
            3
                 0.423360
                 -3.027210
          245
                -10.832078
               200.922960
          246
          247
               228.921375
          248
                45.818526
          249 -181.063635
         250 rows × 1 columns
```

TODO – Students

- Plot the time series generated thanks to $x \rightarrow x*\sin(x)$ with 250 timestamps
- label x and y axis

TODO - Students

- Using series_to_supervise split data into samples with n_in = 6 and n_out = 1
- Put the result in variable data

```
In []:
    """ FILL """
    n_in = 6
    n_out = 1
    data = series_to_supervised(data=time_series, n_in=n_in, n_out=n_out)
    data.head()
```

Out[]:		data(t-6)	data(t-5)	data(t-4)	data(t-3)	data(t-2)	data(t-1)	data(t)
	6	0.000000	0.841471	1.818595	0.423360	-3.027210	-4.794621	-1.676493
	7	0.841471	1.818595	0.423360	-3.027210	-4.794621	-1.676493	4.598906
	8	1.818595	0.423360	-3.027210	-4.794621	-1.676493	4.598906	7.914866
	9	0.423360	-3.027210	-4.794621	-1.676493	4.598906	7.914866	3.709066
	10	-3.027210	-4.794621	-1.676493	4.598906	7.914866	3.709066	-5.440211

Head of the previous dataset

	data(t-6)	data(t-5)	data(t-4)	data(t-3)	data(t-2)	data(t-1)	data(t)	
6	0.000000	0.841471	1.818595	0.423360	-3.027210	-4.794621	-1.676493	
7	0.841471	1.818595	0.423360	-3.027210	-4.794621	-1.676493	4.598906	
8	1.818595	0.423360	-3.027210	-4.794621	-1.676493	4.598906	7.914866	
9	0.423360	-3.027210	-4.794621	-1.676493	4.598906	7.914866	3.709066	
10	-3.027210	-4.794621	-1.676493	4.598906	7.914866	3.709066	-5.440211	

Contrary to the approaches used so far, we cannot separate the data into TRAIN, VALID and TEST in a random way, since we are dealing with time series where the order is important.

The TRAIN data will therefore necessarily be at the beginning, then we will find the VALIDATION data and finally the TEST data.

Here are the chosen indices.

```
In []:
    testAndValid = 0.1
    SPLIT = int(testAndValid*len(data))
    idx_train = len(data)-2*SPLIT
```

```
idx test = len(data)-SPLIT
         print(f"TRAIN=time series[:{idx train}]")
         print(f"VALID=time series[{idx train}:{idx test}]")
         print(f"TEST=time series[{idx test}:]")
         TRAIN=data[:idx train]
         VAL=data[idx_train:idx_test]
         TEST=data[idx test:]
         TRAIN=time series[:196]
         VALID=time series[196:220]
         TEST=time series[220:]
In [ ]:
         plt.figure(figsize=(10, 4))
         plt.plot(TRAIN["data(t)"], label="train")
         plt.plot(VAL["data(t)"], label="val")
         plt.plot(TEST["data(t)"], label="test")
         plt.legend()
         plt.xlabel("time stamps")
         plt.ylabel("time series")
         Text(0, 0.5, 'time series')
Out[ ]:
                     train
            200
                     test
            100
         time series
              0
           -100
           -200
                              50
                                            100
                                                          150
                                                                        200
                                                                                       250
```

time stamps

TODO - Students

complete the code for preprocessing your train/validation/test datasets.

```
In []: # split into input and outputs
    train_X, train_y = TRAIN.values[:, :-n_out], TRAIN.values[:, -n_out]
    val_X, val_y = VAL.values[:, :-n_out], VAL.values[:, -n_out]
    test_X, test_y = TEST.values[:, :-n_out], TEST.values[:, -n_out]

# reshape input to be 3D [samples, timesteps, features]
    train_X = train_X.reshape((-1, n_in, n_features))
    val_X = val_X.reshape((-1, n_in, n_features))
    test_X = test_X.reshape((-1, n_in, n_features))

    train_X.shape, train_y.shape

Out[]: ((196, 6, 1), (196,))
```

Build a first network using LSTM cells

TODO – Students

train y.shape = (196,)

- · Look carefully at the following cell
- What is the impact of the return_sequences parameter of the LSTM cell? (change the value: False or True and observe the shape of the output).

If return sequences=True, the LSTM and the Dense layer get an extra dimension, which has value n in.

```
In []:
    LSTM_SIZE = 16

inputs = Input(shape=(n_in, n_features))
    hidden = LSTM(LSTM_SIZE, return_sequences=False, activation="relu")(inputs)
# hidden = LSTM(LSTM_SIZE, return_sequences=True, activation='relu')(inputs) # check what changing `return_sequence`
    outputs = Dense(n_out, activation="linear")(hidden)
```

model = Model(inputs=inputs, outputs=outputs)
model.summary()

Model: "model"

Layer (type)	Output Shape	Param #		
input_1 (InputLayer)	[(None, 6, 1)]	0		
lstm (LSTM)	(None, 16)	1152		
dense (Dense)	(None, 1)	17		

Total params: 1,169 Trainable params: 1,169 Non-trainable params: 0

2022-02-03 13:20:15.360918: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:939] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-02-03 13:20:15.415097: W tensorflow/stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic library 'libcudnn.so.8'; dlerror: libcudnn.so.8: cannot open shared object file: No such file or directory 2022-02-03 13:20:15.415120: W tensorflow/core/common_runtime/gpu/gpu_device.cc:1850] Cannot dlopen some GPU libraries. Please make sure the missing libraries mentioned above are installed properly if you would like to use GPU. Follow the guide at https://www.tensorflow.org/install/gpu for how to download and setup the required libraries for your platform.

Skipping registering GPU devices...

2022-02-03 13:20:15.416141: I tensorflow/core/platform/cpu_feature_guard.cc:151] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX2 FMA

To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.

TODO - Students

- Complete the function build_and_fit used to train your RNN model.
- · compile: as usual
- fit: as usual but...
 - Be careful, you have to set the shuffle parameter to false in order to take the data in order.
 - Use the validation set to control the overfitting in the earlystopping callback

```
In [ ]: def build and fit(model, X_train, y_train, X_val, y_val, X_test, y_test, patience=150, epochs=200):
```

```
model.compile(
    optimizer="adam",
    loss="mse",
    metrics=["mae"],
es = EarlyStopping(
    monitor="val loss",
    patience=patience,
    verbose=1.
    restore best weights=True,
    mode="min",
history = model.fit(
    x=X train,
    y=y train,
    validation data=(X val, y val),
    epochs=epochs,
    callbacks=[es],
    use multiprocessing=True,
    workers=6,
) # epochs = 200
y pred = model.predict(X test)
# plot history
plt.figure(figsize=(20, 8))
plt.subplot(311)
plt.plot(history.history["loss"][3:], label="loss")
plt.plot(history.history["val loss"][3:], label="val loss")
plt.legend()
plt.subplot(312)
plt.plot(history.history["mae"][3:], label="mae")
plt.plot(history.history["val mae"][3:], label="val mae")
plt.legend()
plt.subplot(313)
plt.plot(range(len(y_train)), y_train, label="train")
plt.plot(range(len(y train), len(y train) + len(y val)), y val, label="valid")
plt.plot(
    range(len(y train) + len(y val), len(y train) + len(y val) + len(y pred)),
    y test,
    label="test",
```

```
plt.plot(
    range(len(y_train) + len(y_val), len(y_train) + len(y_val) + len(y_pred)),
    y_pred,
    label="predict",
)

plt.legend(loc="center left")
plt.show()

return model

history = build_and_fit(model, train_X, train_y, val_X, val_y, test_X, test_y)
```

```
Epoch 1/200
e: 142.4521
Epoch 2/200
141.4368
Epoch 3/200
e: 140.5119
Epoch 4/200
e: 139.6716
Epoch 5/200
138.8918
Epoch 6/200
e: 138.0099
Epoch 7/200
e: 137.0667
Epoch 8/200
e: 135.9030
Epoch 9/200
e: 133.4701
Epoch 10/200
e: 128.7614
Epoch 11/200
e: 121.7736
Epoch 12/200
e: 107.6457
Epoch 13/200
88.7976
Epoch 14/200
64.1259
Epoch 15/200
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51.1207
Epoch 16/200
33.1972
Epoch 17/200
6.5784
Epoch 18/200
0.5762
Epoch 19/200
6766
Epoch 20/200
4.0201
Epoch 21/200
938
Epoch 22/200
441
Epoch 23/200
596
Epoch 24/200
8854
Epoch 25/200
0683
Epoch 26/200
592
Epoch 27/200
8813
Epoch 28/200
1680
Epoch 29/200
Epoch 30/200
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90
Epoch 31/200
070
Epoch 32/200
033
Epoch 33/200
72
Epoch 34/200
975
Epoch 35/200
Epoch 36/200
Epoch 37/200
06
Epoch 38/200
Epoch 39/200
Epoch 40/200
59
Epoch 41/200
36
Epoch 42/200
26
Epoch 43/200
37
Epoch 44/200
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Epoch 45/200
Epoch 46/200
78
Epoch 47/200
34
Epoch 48/200
74
Epoch 49/200
Epoch 50/200
08
Epoch 51/200
Epoch 52/200
Epoch 53/200
Epoch 54/200
Epoch 55/200
81
Epoch 56/200
Epoch 57/200
27
Epoch 58/200
Epoch 59/200
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Epoch 60/200
Epoch 61/200
Epoch 62/200
Epoch 63/200
82
Epoch 64/200
Epoch 65/200
Epoch 66/200
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Epoch 75/200
Epoch 76/200
Epoch 77/200
28
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99
Epoch 79/200
Epoch 80/200
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Epoch 90/200
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Epoch 91/200
Epoch 92/200
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Epoch 108/200
Epoch 109/200
Epoch 110/200
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Epoch 111/200
Epoch 112/200
Epoch 113/200
Epoch 114/200
Epoch 115/200
Epoch 116/200
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Epoch 118/200
Epoch 119/200
Epoch 120/200
Epoch 121/200
Epoch 122/200
Epoch 123/200
Epoch 124/200
Epoch 125/200
Epoch 126/200
2
Epoch 127/200
Epoch 128/200
```

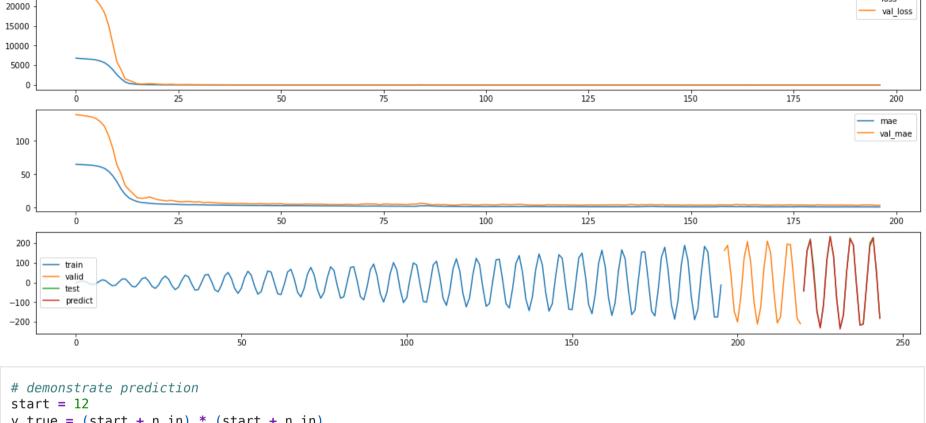
```
Epoch 129/200
Epoch 130/200
Epoch 131/200
Epoch 132/200
Epoch 133/200
Epoch 134/200
Epoch 135/200
Epoch 136/200
Epoch 137/200
Epoch 138/200
Epoch 139/200
Epoch 140/200
Epoch 141/200
Epoch 142/200
Epoch 143/200
Epoch 144/200
Epoch 145/200
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Epoch 146/200
Epoch 147/200
Epoch 148/200
Epoch 149/200
Epoch 150/200
Epoch 151/200
Epoch 152/200
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Epoch 155/200
7
Epoch 156/200
Epoch 157/200
Epoch 158/200
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Epoch 163/200
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Epoch 164/200
Epoch 165/200
Epoch 166/200
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Epoch 177/200
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Epoch 181/200
Epoch 182/200
Epoch 183/200
Epoch 184/200
Epoch 185/200
```

04-notebook-RNN-LIMONIER

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Epoch 186/200
Epoch 187/200
Epoch 188/200
Epoch 189/200
Epoch 190/200
Epoch 191/200
Epoch 192/200
Epoch 193/200
Epoch 194/200
Epoch 195/200
Epoch 196/200
Epoch 197/200
Epoch 198/200
Epoch 199/200
Epoch 200/200
```



```
In []: # demonstrate prediction
    start = 12
    y_true = (start + n_in) * (start + n_in)
    x_input = np.array([fct(x) for x in range(start, start + n_in)])
    x_input = x_input.reshape((1, n_in, n_features))
    yhat = model.predict(x_input, verbose=0)
    print(yhat, y_true)
```

[[-12.219949]] 324

Stacked Bi-LSTM

In order to improve the performance of the model, it's possible to:

- stack LSTM with return sequence=True for all levels except the last one where return sequence=False
- use Bi-LSTM

TODO – Students

loss

• Build a model stacking 3 BI-LSTM layers

```
In [ ]:
         inputs = Input(shape=(n in, n features))
         bi lstm1 = Bidirectional(
             layer=LSTM(
                 units=LSTM SIZE,
                 return sequences=True,
                 activation="relu",
         (inputs)
         bi lstm2 = Bidirectional(
             layer=LSTM(
                 units=LSTM SIZE,
                 return sequences=True,
                 activation="relu",
         )(bi lstm1)
         bi lstm3 = Bidirectional(
             layer=LSTM(
                 units=LSTM SIZE,
                 return sequences=False,
                 activation="relu",
         )(bi lstm2)
         outputs = Dense(n out, activation="linear")(bi lstm3)
         model = Model(inputs=inputs, outputs=outputs)
         model.summary()
         # Bidirectional(LSTM(LSTM SIZE, return sequences=True, activation='relu'))
```

Model: "model_1"

Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, 6, 1)]	0
bidirectional (Bidirectiona l)	(None, 6, 32)	2304
<pre>bidirectional_1 (Bidirectio nal)</pre>	(None, 6, 32)	6272
<pre>bidirectional_2 (Bidirectio nal)</pre>	(None, 32)	6272
dense_1 (Dense)	(None, 1)	33

Total params: 14,881 Trainable params: 14,881 Non-trainable params: 0

```
In [ ]:
```

model = build_and_fit(model, train_X, train_y, val_X, val_y, test_X, test_y)

```
Epoch 1/200
e: 138.8743
Epoch 2/200
e: 137.6566
Epoch 3/200
e: 135.2447
Epoch 4/200
e: 130.3098
Epoch 5/200
e: 120.2873
Epoch 6/200
e: 94.8881
Epoch 7/200
56.1267
Epoch 8/200
46.2576
Epoch 9/200
34.7541
Epoch 10/200
5.4910
Epoch 11/200
6.4811
Epoch 12/200
5.8728
Epoch 13/200
5.0106
Epoch 14/200
7394
Epoch 15/200
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1669
Epoch 16/200
5364
Epoch 17/200
4511
Epoch 18/200
899
Epoch 19/200
771
Epoch 20/200
203
Epoch 21/200
620
Epoch 22/200
082
Epoch 23/200
526
Epoch 24/200
09
Epoch 25/200
02
Epoch 26/200
55
Epoch 27/200
00
Epoch 28/200
1
Epoch 29/200
Epoch 30/200
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44
Epoch 31/200
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Epoch 102/200
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Epoch 118/200
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Epoch 127/200
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Epoch 129/200
Epoch 130/200
Epoch 131/200
2
Epoch 132/200
```

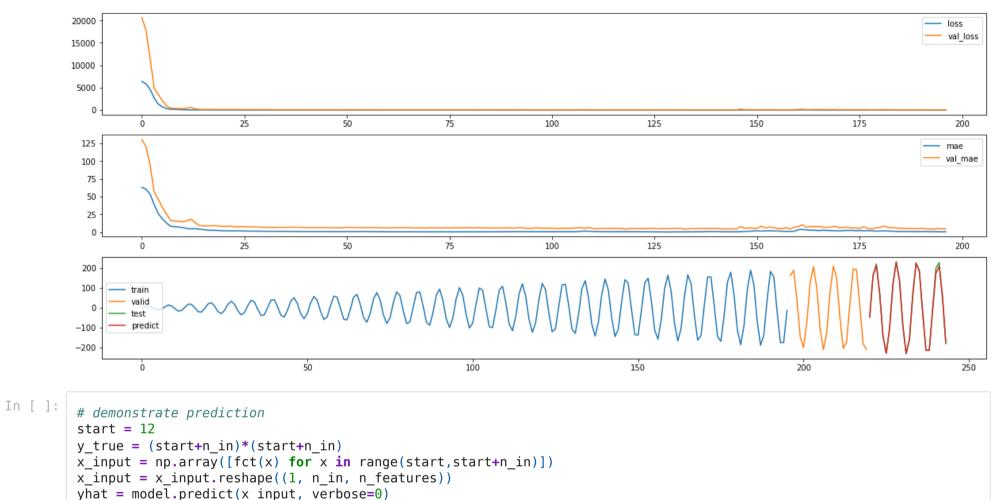
```
Epoch 133/200
Epoch 134/200
Epoch 135/200
Epoch 136/200
Epoch 137/200
Epoch 138/200
Epoch 139/200
Epoch 140/200
Epoch 141/200
Epoch 142/200
Epoch 143/200
Epoch 144/200
Epoch 145/200
Epoch 146/200
Epoch 147/200
```

```
Epoch 148/200
Epoch 149/200
Epoch 150/200
77
Epoch 151/200
Epoch 152/200
Epoch 153/200
Epoch 154/200
Epoch 155/200
Epoch 156/200
7
Epoch 157/200
Epoch 158/200
Epoch 159/200
Epoch 160/200
Epoch 161/200
Epoch 162/200
```

```
Epoch 163/200
Epoch 164/200
068
Epoch 165/200
5400
Epoch 166/200
88
Epoch 167/200
77
Epoch 168/200
768
Epoch 169/200
Epoch 170/200
456
Epoch 171/200
61
Epoch 172/200
Epoch 173/200
622
Epoch 174/200
Epoch 175/200
74
Epoch 176/200
09
```

```
Epoch 177/200
Epoch 178/200
Epoch 179/200
03
Epoch 180/200
Epoch 181/200
Epoch 182/200
Epoch 183/200
Epoch 184/200
Epoch 185/200
60
Epoch 186/200
Epoch 187/200
Epoch 188/200
Epoch 189/200
Epoch 190/200
Epoch 191/200
```

```
Epoch 192/200
Epoch 193/200
Epoch 194/200
Epoch 195/200
Epoch 196/200
Epoch 197/200
Epoch 198/200
Epoch 199/200
Epoch 200/200
7
```



[[-13.641433]] 324

print(yhat, y_true)

LSTM network for multi-variate time series

Multivariate time series data means data where there is more than one observation for each time step.

There are two main models we may need with multivariate time series data. These are the multiple input series or the multiple parallel series depending on whether we want to predict one or more of the variables.

In this notebook, we focus on the first case: as input, several time series and as output (the prediction), a single time series.

Prepare the data

We reuse the same series_to_supervise() function in order to build a dataset with:

- n in elements for each series
- n_out elements for each series to be predict

You have also to select one (Multiple Input Series) or many time series (Multiple Parallel Series) to predict.

```
In [ ]:
         # Get the time series
         fct2 = lambda x: 2*x
         time series1 = [fct(x) for x in range(SIZE)]
         time series2 = [fct2(x) \text{ for } x \text{ in } range(SIZE)]
         out seq = [time series1[i]+time series2[i] for i in range(SIZE)]
In [ ]:
         # Get the dataset
         dataset = pd.DataFrame(data={"f1":time_series1, "f2":time_series2, "output":out_seq}, index=range(SIZE))
         n features = dataset.shape[1] # for multivariate time series
         dataset.head()
                 f1 f2
Out[ ]:
                         output
         0 0.000000 0 0.000000
         1 0.841471 2 2.841471
         2 1.818595 4 5.818595
         3 0.423360 6 6.423360
         4 -3.027210 8 4.972790
```

As with the univariate time series, we must structure these data into samples with input and output elements. An LSTM model needs sufficient context to learn a mapping from an input sequence to an output value. LSTMs can support parallel input time series as separate variables or features. Therefore, we need to split the data into samples maintaining the order of observations across the two input sequences.

If we chose six input time steps for the three features, we have to transform the dataset in the following way.

```
In []: # As previously, prepare the dataset
''' In the followig example, we select
- n_in number of time steps (6)
- n_out number of time steps of output
- and one serie to predict : output
'''

n_features = dataset.shape[1] # for multivariate time series
n_in = 6
n_out = 1
output = ["output"]

data = series_to_supervised(dataset, n_in, n_out, output=output)
data.head()
```

```
Out[ ]:
                          f2(t-
                                output(t-
                                                      f2(t-
                                                            output(t-
                                                                                  f2(t-
                                                                                         output(t-
                                                                                                              f2(t-
                                                                                                                     output(t-
                                                                                                                                           f2(t-
                                                                                                                                                  output(t-
                   f1(t-6)
                                              f1(t-5)
                                                                          f1(t-4)
                                                                                                       f1(t-3)
                                                                                                                                   f1(t-2)
                                                                                                                                                               f1(t-
                                                        5)
                                                                                                                3)
                                                                                                                                             2)
                0.000000
                            0.0 0.000000
                                           0.841471
                                                       2.0 2.841471
                                                                       1.818595
                                                                                   4.0
                                                                                         5.818595
                                                                                                    0.423360
                                                                                                               6.0
                                                                                                                     6.423360
                                                                                                                                -3.027210
                                                                                                                                            8.0
                                                                                                                                                  4.972790
                                                                                                                                                            -4.79462
                0.841471
                           2.0 2.841471
                                           1.818595
                                                       4.0 5.818595
                                                                       0.423360
                                                                                  6.0
                                                                                         6.423360
                                                                                                   -3.027210
                                                                                                               8.0
                                                                                                                     4.972790
                                                                                                                                -4.794621
                                                                                                                                          10.0
                                                                                                                                                  5.205379
                                                                                                                                                            -1.67649
                            4.0 5.818595
                                           0.423360
                                                                      -3.027210
                                                                                         4.972790
                                                                                                                     5.205379
                                                                                                                                                             4.59890
                1.818595
                                                       6.0 6.423360
                                                                                  8.0
                                                                                                   -4.794621
                                                                                                             10.0
                                                                                                                               -1.676493
                                                                                                                                          12.0 10.323507
                0.423360
                           6.0 6.423360
                                          -3.027210
                                                       8.0
                                                            4.972790
                                                                      -4.794621
                                                                                 10.0
                                                                                         5.205379
                                                                                                  -1.676493
                                                                                                              12.0
                                                                                                                    10.323507
                                                                                                                                4.598906
                                                                                                                                                18.598906
                                                                                                                                                             7.91486
                                                                                                                                          14.0
           10 -3.027210
                           8.0 4.972790 -4.794621 10.0 5.205379 -1.676493 12.0 10.323507 4.598906
                                                                                                             14.0 18.598906
                                                                                                                                7.914866
                                                                                                                                          16.0 23.914866
                                                                                                                                                             3.70906
```

```
In []: # Split dataset into TRAIN, VAL and TEST
    testAndValid = 0.1

SPLIT = int(testAndValid * len(data))
    idx_train = len(data) - 2 * SPLIT
    idx_test = len(data) - SPLIT

print(f"TRAIN=time_series[:{idx_train}]")
    print(f"VALID=time_series[{idx_train}:{idx_test}]")
    print(f"TEST=time_series[{idx_test}:]")

TRAIN = data[:idx_train]
    VAL = data[idx_train:idx_test]
    TEST = data[idx_test:]
```

```
TRAIN=time_series[:196]
VALID=time_series[196:220]
TEST=time_series[220:]

TODO - Students
```

• build train X, val X, test X and train y, val y and test y as before. Then print the shapes of tensors

- train X is a 3D-tensor (196, 6, 3) for me
- train y is a 1D-tensor (196,)

```
In []: # split into input and outputs
    train_X, train_y = TRAIN.values[:, :-n_out], TRAIN.values[:, -n_out]
    val_X, val_y = VAL.values[:, :-n_out], VAL.values[:, -n_out]
    test_X, test_y = TEST.values[:, :-n_out], TEST.values[:, -n_out]

# reshape input to be 3D [samples, timesteps, features]
    train_X = train_X.reshape((-1, n_in, n_features))
    val_X = val_X.reshape((-1, n_in, n_features))
    test_X = test_X.reshape((-1, n_in, n_features)))
    train_X.shape, train_y.shape
Out[]: ((196, 6, 3), (196,))
```

Build a neuronal model

Any of the varieties of LSTMs in the previous section can be used, such as a Vanilla, Stacked, Bidirectional. It's also possible to use CNN or mixed CNN and LSTM.

We will use a Vanilla LSTM where the number of time steps and parallel series (features) are specified for the input layer via the input_shape argument.

```
In []: # Build model
inputs = Input(shape=(n_in, n_features))
hidden = LSTM(LSTM_SIZE, return_sequences=False, activation="relu")(inputs)
outputs = Dense(n_out, activation="linear")(hidden)
model = Model(inputs, outputs)
model.summary()
```

Model: "model_2"

Layer (type)	Output Shape	Param #
input_3 (InputLayer)	[(None, 6, 3)]	0
lstm_4 (LSTM)	(None, 16)	1280
dense_2 (Dense)	(None, 1)	17

Total params: 1,297 Trainable params: 1,297 Non-trainable params: 0

·

```
In [ ]:
```

model = build_and_fit(model, train_X, train_y, val_X, val_y, test_X, test_y)

```
Epoch 1/200
mae: 696.6531
Epoch 2/200
mae: 663.4233
Epoch 3/200
mae: 636.8749
Epoch 4/200
mae: 609.9120
Epoch 5/200
mae: 578.4326
Epoch 6/200
mae: 527.9244
Epoch 7/200
ae: 488.8546
Epoch 8/200
ae: 425.9018
Epoch 9/200
ae: 390.1928
Epoch 10/200
ae: 355.1621
Epoch 11/200
ae: 308.0824
Epoch 12/200
ae: 259.8584
Epoch 13/200
e: 230.0773
Epoch 14/200
e: 201.5160
Epoch 15/200
```

```
e: 174.3538
Epoch 16/200
e: 143.5351
Epoch 17/200
127.4922
Epoch 18/200
119.8643
Epoch 19/200
102.0894
Epoch 20/200
89.4920
Epoch 21/200
78,2692
Epoch 22/200
69.0479
Epoch 23/200
63.0500
Epoch 24/200
59.7849
Epoch 25/200
57.9682
Epoch 26/200
53.5880
Epoch 27/200
8.4082
Epoch 28/200
5.8267
Epoch 29/200
2.4257
Epoch 30/200
```

```
2.4464
Epoch 31/200
0.6197
Epoch 32/200
0.2490
Epoch 33/200
7.6561
Epoch 34/200
9.2536
Epoch 35/200
8.4021
Epoch 36/200
6.4717
Epoch 37/200
3.1745
Epoch 38/200
2.1756
Epoch 39/200
6.9275
Epoch 40/200
5.3632
Epoch 41/200
5.1058
Epoch 42/200
2.3610
Epoch 43/200
3.2980
Epoch 44/200
1.8396
```

```
Epoch 45/200
1.2955
Epoch 46/200
9.5361
Epoch 47/200
9.7878
Epoch 48/200
9.8330
Epoch 49/200
9.0601
Epoch 50/200
8.1741
Epoch 51/200
7.7834
Epoch 52/200
7.0558
Epoch 53/200
6.6753
Epoch 54/200
6.0750
Epoch 55/200
6.2544
Epoch 56/200
3.9693
Epoch 57/200
6.4673
Epoch 58/200
3.1303
Epoch 59/200
```

```
4.8810
Epoch 60/200
4.3012
Epoch 61/200
0.4045
Epoch 62/200
4.7867
Epoch 63/200
4.9515
Epoch 64/200
7.6362
Epoch 65/200
1.4327
Epoch 66/200
9.2959
Epoch 67/200
7.1463
Epoch 68/200
4.5360
Epoch 69/200
5.1490
Epoch 70/200
1.8144
Epoch 71/200
7.4736
Epoch 72/200
9.3891
Epoch 73/200
0.5494
Epoch 74/200
```

```
9.4629
Epoch 75/200
9.6737
Epoch 76/200
5.7442
Epoch 77/200
6.3531
Epoch 78/200
9.1692
Epoch 79/200
8.7802
Epoch 80/200
9.3884
Epoch 81/200
9.9211
Epoch 82/200
2.4998
Epoch 83/200
2.1765
Epoch 84/200
1.8011
Epoch 85/200
7.0053
Epoch 86/200
5.1690
Epoch 87/200
1.9538
Epoch 88/200
8.4207
```

```
Epoch 89/200
4.7021
Epoch 90/200
2.9489
Epoch 91/200
4984
Epoch 92/200
5409
Epoch 93/200
0.0158
Epoch 94/200
1045
Epoch 95/200
0305
Epoch 96/200
1483
Epoch 97/200
8902
Epoch 98/200
2.5584
Epoch 99/200
0313
Epoch 100/200
2663
Epoch 101/200
3.9586
Epoch 102/200
0062
Epoch 103/200
```

```
9.7555
Epoch 104/200
8290
Epoch 105/200
8.0474
Epoch 106/200
1.8281
Epoch 107/200
6.2668
Epoch 108/200
8.8583
Epoch 109/200
8.2668
Epoch 110/200
0921
Epoch 111/200
8057
Epoch 112/200
6407
Epoch 113/200
0845
Epoch 114/200
3074
Epoch 115/200
8420
Epoch 116/200
4127
Epoch 117/200
9724
Epoch 118/200
```

```
6266
Epoch 119/200
9909
Epoch 120/200
0362
Epoch 121/200
7674
Epoch 122/200
9324
Epoch 123/200
4826
Epoch 124/200
187
Epoch 125/200
063
Epoch 126/200
809
Epoch 127/200
8643
Epoch 128/200
2308
Epoch 129/200
0233
Epoch 130/200
9369
Epoch 131/200
427
Epoch 132/200
132
```

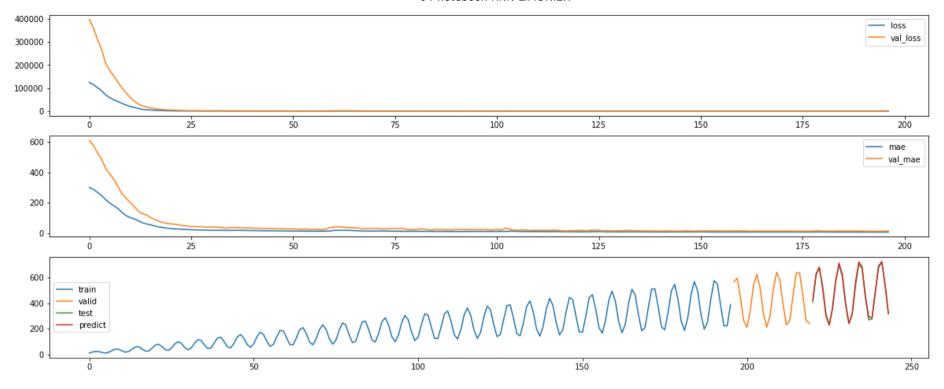
```
Epoch 133/200
652
Epoch 134/200
049
Epoch 135/200
792
Epoch 136/200
5104
Epoch 137/200
7366
Epoch 138/200
807
Epoch 139/200
836
Epoch 140/200
127
Epoch 141/200
405
Epoch 142/200
360
Epoch 143/200
256
Epoch 144/200
942
Epoch 145/200
954
Epoch 146/200
963
Epoch 147/200
```

```
3409
Epoch 148/200
450
Epoch 149/200
314
Epoch 150/200
231
Epoch 151/200
697
Epoch 152/200
732
Epoch 153/200
757
Epoch 154/200
508
Epoch 155/200
133
Epoch 156/200
285
Epoch 157/200
332
Epoch 158/200
506
Epoch 159/200
237
Epoch 160/200
481
Epoch 161/200
255
Epoch 162/200
```

```
119
Epoch 163/200
576
Epoch 164/200
640
Epoch 165/200
456
Epoch 166/200
914
Epoch 167/200
497
Epoch 168/200
056
Epoch 169/200
705
Epoch 170/200
532
Epoch 171/200
550
Epoch 172/200
521
Epoch 173/200
575
Epoch 174/200
066
Epoch 175/200
922
Epoch 176/200
777
```

```
Epoch 177/200
278
Epoch 178/200
644
Epoch 179/200
445
Epoch 180/200
036
Epoch 181/200
686
Epoch 182/200
577
Epoch 183/200
085
Epoch 184/200
619
Epoch 185/200
276
Epoch 186/200
059
Epoch 187/200
052
Epoch 188/200
746
Epoch 189/200
329
Epoch 190/200
897
Epoch 191/200
```

```
131
Epoch 192/200
561
Epoch 193/200
244
Epoch 194/200
357
Epoch 195/200
619
Epoch 196/200
205
Epoch 197/200
912
Epoch 198/200
544
Epoch 199/200
919
Epoch 200/200
261
```



Lab work: Air Pollution Forecasting

This is a dataset that reports on the weather and the level of pollution each hour for five years at the US embassy in Beijing, China.

The data includes the date-time, the pollution called PM2.5 concentration, and the weather information including dew point, temperature, pressure, wind direction, wind speed and the cumulative number of hours of snow and rain. The complete feature list in the raw data is as follows:

- 1. No: row number
- 2. year: year of data in this row
- 3. month: month of data in this row
- 4. day: day of data in this row
- 5. hour: hour of data in this row
- 6. pm2.5: PM2.5 concentration
- 7. DEWP: Dew Point
- 8. TEMP: Temperature

- 9. PRES: Pressure
- 10. cbwd: Combined wind direction
- 11. lws: Cumulated wind speed
- 12. Is: Cumulated hours of snow
- 13. Ir: Cumulated hours of rain

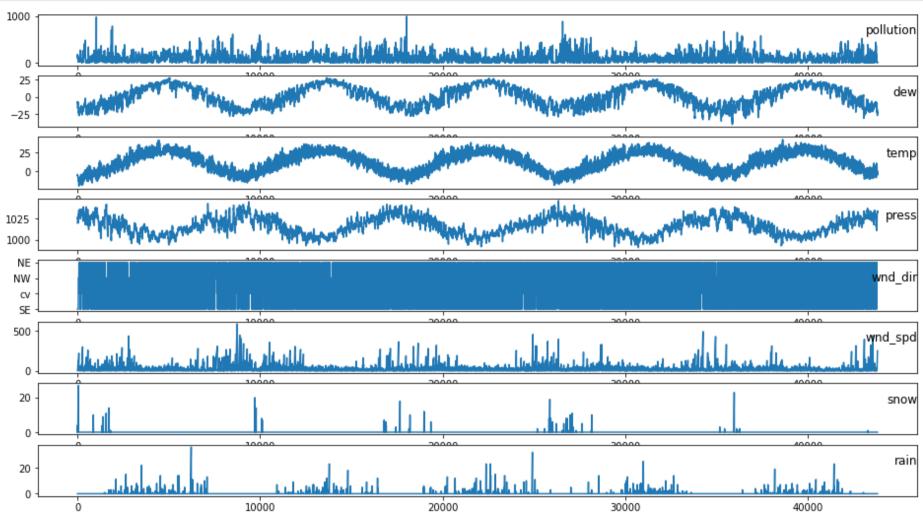
We can use this data and frame a forecasting problem where, given the weather conditions and pollution for prior hours, we forecast the pollution at the next hour.

This dataset can be used to frame other forecasting problems.

Load the data

```
In [ ]:
          DATAPATH = "https://www.i3s.unice.fr/~riveill/dataset/pollution.csv"
In [ ]:
          # Read the dataset
          data = pd.read csv(DATAPATH, sep=",", header=0, index col=0)
          data.head()
Out[ ]:
                                               press wnd_dir wnd_spd snow rain
                           pollution dew temp
                      date
         2010-01-02 00:00:00
                              129.0
                                     -16
                                          -4.0 1020.0
                                                          SE
                                                                  1.79
                                                                           0
         2010-01-02 01:00:00
                              148.0
                                     -15
                                          -4.0 1020.0
                                                          SE
                                                                  2.68
                                                                           0
         2010-01-02 02:00:00
                              159.0
                                     -11
                                          -5.0 1021.0
                                                                  3.57
                                                                           0
                                          -5.0 1022.0
         2010-01-02 03:00:00
                              181.0
                                                          SE
                                                                  5.36
                                                                                0
                                                                           1
         2010-01-02 04:00:00
                              138.0
                                      -7
                                          -5.0 1022.0
                                                          SE
                                                                  6.25
                                                                           2
                                                                                0
In [ ]:
          plt.figure(figsize=(16,9))
          for i, column in enumerate(data.columns):
              plt.subplot(len(data.columns), 1, i+1)
              plt.plot(data[column].to numpy())
```

plt.title(column, y=0.5, loc='right')
plt.show()



Construct the dataset

The first step is to prepare the pollution dataset for the LSTM.

This involves framing the dataset as a supervised learning problem and normalizing the input variables.

We will frame the supervised learning problem as predicting the pollution at the current hour (t) given the pollution measurement and weather conditions at the prior time step.

TODO – Students

• using series_to_supervised function build the dataset

```
In [ ]: dataset = series_to_supervised(data=data, n_in=n_in, n_out=n_out, output=output)
    dataset.head()
```

Out[]:		pollution(t- 6)	dew(t- 6)	temp(t- 6)	press(t- 6)	wnd_dir(t- 6)	wnd_spd(t- 6)	snow(t- 6)	rain(t- 6)	pollution(t- 5)	dew(t- 5)	 rain(t- 2)	pollution(t- 1)	dew(t- 1)	temp(t- 1)	р
	date															
	2010-01- 02 06:00:00	129.0	-16.0	-4.0	1020.0	SE	1.79	0.0	0.0	148.0	-15.0	 0.0	109.0	-7.0	-6.0	
	2010-01- 02 07:00:00	148.0	-15.0	-4.0	1020.0	SE	2.68	0.0	0.0	159.0	-11.0	 0.0	105.0	-7.0	-6.0	
	2010-01- 02 08:00:00	159.0	-11.0	-5.0	1021.0	SE	3.57	0.0	0.0	181.0	-7.0	 0.0	124.0	-7.0	-5.0	
	2010-01- 02 09:00:00	181.0	-7.0	-5.0	1022.0	SE	5.36	1.0	0.0	138.0	-7.0	 0.0	120.0	-8.0	-6.0	
	2010-01- 02 10:00:00	138.0	-7.0	-5.0	1022.0	SE	6.25	2.0	0.0	109.0	-7.0	 0.0	132.0	-7.0	-5.0	

5 rows × 49 columns

In []: dataset.dtypes

		nollution(+ 6)	floo+64
Out[]:	pollution(t-6)	float64 float64
		<pre>dew(t-6) temp(t-6)</pre>	float64
		press(t-6)	float64
		wnd_dir(t-6)	object
		wnd_spd(t-6)	float64
		snow(t-6)	float64
		rain(t-6)	float64
		pollution(t-5)	float64 float64
		dew(t-5)	
		temp(t-5)	float64 float64
		press(t-5)	
		wnd_dir(t-5)	object float64
		wnd_spd(t-5)	float64
		snow(t-5)	float64
		<pre>rain(t-5) pollution(t-4)</pre>	float64
		dew(t-4)	float64
		temp(t-4)	float64
		press(t-4)	float64
		wnd_dir(t-4)	object
		wnd_spd(t-4)	float64
		snow(t-4)	float64
		rain(t-4)	float64
		pollution(t-3)	float64
		dew(t-3)	float64
		temp(t-3)	float64
		press(t-3)	float64
		wnd_dir(t-3)	object
		wnd_spd(t-3)	float64
		snow(t-3)	float64
		rain(t-3)	float64
		pollution(t-2)	float64
		dew(t-2)	float64
		temp(t-2)	float64
		press(t-2)	float64
		wnd_dir(t-2)	object
		wnd_spd(t-2)	float64
		snow(t-2)	float64
		rain(t-2)	float64
		pollution(t-1)	float64
		dew(t-1)	float64
		temp(t-1)	float64
		press(t-1)	float64

wnd_dir(t-1)	object
wnd_spd(t-1)	float64
snow(t-1)	float64
rain(t-1)	float64
pollution(t)	float64
dtype: object	

dataset.dtypes gives the following result for me

pollution(t-6)	float64
dew(t-6)	float64
temp(t-6)	float64
press(t-6)	float64
wnd dir(t-6)	object
wnd_spd(t-6)	float64
snow(t-6)	float64
rain(t-6)	float64
pollution(t-5)	float64
dew(t-5)	float64
temp(t-5)	float64
press(t-5)	float64
wnd_dir(t-5)	object
wnd_spd(t-5)	float64
snow(t-5)	float64
rain(t-5)	float64
pollution(t-4)	float64
dew(t-4)	float64
temp(t-4)	float64
press(t-4)	float64
wnd_dir(t-4)	object
wnd_spd(t-4)	float64
snow(t-4)	float64
rain(t-4)	float64
pollution(t-3)	float64
dew(t-3)	float64
temp(t-3)	float64
press(t-3)	float64
wnd_dir(t-3)	object
wnd_spd(t-3)	float64

```
float64
snow(t-3)
rain(t-3)
                  float64
pollution(t-2)
                  float64
dew(t-2)
                  float64
temp(t-2)
                  float64
press(t-2)
                  float64
wnd dir(t-2)
                  obiect
wnd spd(t-2)
                  float64
snow(t-2)
                  float64
rain(t-2)
                  float64
pollution(t-1)
                  float64
dew(t-1)
                  float64
temp(t-1)
                  float64
press(t-1)
                  float64
wnd dir(t-1)
                  obiect
wnd spd(t-1)
                  float64
snow(t-1)
                  float64
rain(t-1)
                  float64
pollution(t)
                  float64
dtype: object
```


First, we must split the prepared dataset into train and test sets. To speed up the training of the model for this demonstration, we will only fit the model on the first year of data, then evaluate it on the remaining 4 years of data.

The example below splits the dataset into train and test sets, then splits the train and test sets into input and output variables. Finally, the inputs (X) are reshaped into the 3D format expected by LSTMs, namely [samples, timesteps, features].

```
In []: # get the values
    values = dataset.values

# split into train and test sets
    n_train_hours = 365 * 24
    train = values[:n_train_hours, :]
```

```
val = values[n_train_hours:2*n_train_hours, :]
test = values[2*n_train_hours:, :]

# split into input and outputs
train_X, train_y = train[:, :-n_out], np.array(train[:, -n_out], dtype="float64")
val_X, val_y = val[:, :-n_out], np.array(val[:, -n_out], dtype="float64")
test_X, test_y = test[:, :-n_out], np.array(test[:, -n_out], dtype="float64")

# reshape input to be 3D [samples, timesteps, features]
train_X = train_X.reshape((train_X.shape[0], n_in, n_features))
val_X = val_X.reshape((val_X.shape[0], n_in, n_features))
test_X = test_X.reshape((test_X.shape[0], n_in, n_features))
train_X.shape, train_y.shape
Out[]: ((8760, 6, 8), (8760,))
```

Encode and normalize dataset

Data encoding and normalization

- The wind direction feature is label encoded (integer encoded).
- All features are normalized

And then the dataset is transformed into a supervised learning problem. The weather variables for the hour to be predicted (t) are then removed.

```
numeric_features = [
    i for i, t in enumerate(dataset.dtypes[:-n_out]) if t in ["float64", "int32"]
]
numeric_transformer = Pipeline(
    steps=[("imputer", SimpleImputer(strategy="median")), ("scaler", StandardScaler())]
)

categorical_features = [
    i for i in range(len(dataset.columns) - n_out) if i not in numeric_features
]
categorical_transformer = OneHotEncoder(handle_unknown="ignore")

preprocessor = ColumnTransformer(
    transformers=[
        ("num", numeric_transformer, numeric_features),
        ("cat", categorical_transformer, categorical_features),
```

```
train_X_enc = preprocessor.fit_transform(train_X.reshape(len(train_X), -1)).reshape(
    len(train_X), n_in, -1
)
val_X_enc = preprocessor.fit_transform(val_X.reshape(len(val_X), -1)).reshape(
    len(val_X), n_in, -1
)
test_X_enc = preprocessor.transform(test_X.reshape(len(test_X), -1)).reshape(
    len(test_X), n_in, -1
)
n_features = train_X_enc.shape[2] # Change with oneHotEncode
n_features
```

Out[]: 11

Running the code below prepare the data. Executing the next cell, prints the first 5 rows of the transformed dataset. We can see the 8 input variables (input series) and the 1 output variable (pollution level at the current hour).

Build, Compile, Fit, Predict and Evaluate a model

TODO - Students

- · Build your model
 - Put the number of hidden layers you want. If possible more than one.

```
In []: LSTM_SIZE = 256

inputs = Input(shape=(n_in, n_features))
hidden = LSTM(LSTM_SIZE, return_sequences=False, activation="relu")(inputs)
outputs = Dense(n_out, activation="linear")(hidden)

model = Model(inputs=inputs, outputs=outputs)
model.summary()
```

Model: "model 3"

Layer (type)	Output Shape	Param #
input_4 (InputLayer)	[(None, 6, 11)]	0
lstm_5 (LSTM)	(None, 256)	274432
dense_3 (Dense)	(None, 1)	257

Total params: 274,689 Trainable params: 274,689 Non-trainable params: 0

TODO – Students

• Compile your model

TODO – Students

• Fit your model using EarlyStopping

TODO – Students

Plot learning curve

```
# compile, fit and plot
model = build_and_fit(model, train_X_enc, train_y, val_X_enc, val_y, test_X_enc, test_y, patience=20)
```

```
Epoch 1/200
mae: 31.8709
Epoch 2/200
mae: 25.8883
Epoch 3/200
ae: 20.5113
Epoch 4/200
ae: 20.4938
Epoch 5/200
ae: 17.6048
Epoch 6/200
e: 18.2300
Epoch 7/200
e: 18.2607
Epoch 8/200
e: 15.5426
Epoch 9/200
e: 15.9985
Epoch 10/200
e: 15.8439
Epoch 11/200
e: 14.4974
Epoch 12/200
e: 15.8724
Epoch 13/200
e: 14.7106
Epoch 14/200
e: 15.2156
Epoch 15/200
```

```
e: 14.8492
Epoch 16/200
e: 15.5546
Epoch 17/200
e: 16.9702
Epoch 18/200
e: 15.2084
Epoch 19/200
e: 16.0889
Epoch 20/200
e: 14.9737
Epoch 21/200
e: 17.0761
Epoch 22/200
e: 15.8118
Epoch 23/200
e: 14.9230
Epoch 24/200
e: 20.2004
Epoch 25/200
e: 16.8313
Epoch 26/200
e: 14.7679
Epoch 27/200
e: 15.6340
Epoch 28/200
e: 15.9832
Epoch 29/200
e: 19.6713
Epoch 30/200
```

```
e: 17.7752
Epoch 31/200
                      - ETA: Os - loss: 648.0208 - mae: 14.4144Restoring model weights from the en
d of the best epoch: 11.
e: 15.1601
Epoch 00031: early stopping
                                                            loss
                                                             val loss
1000
800
                         10
                                   15
                                              20
                                                        25
                                                             mae
20
                                                             val mae
18
16
                         10
                                   15
                                              20
                                                        25
    Ó
1000
800
600
400
200
                10000
                             20000
                                          30000
                                                       40000
```

TODO – Students

• Use your model to predict test set data

```
# make a prediction
y_test_pred = model.predict(test_X_enc)
y_test_pred
```

TODO - Students

Evaluate your model with RMSA

```
In [ ]: # calculate RMSE
rmse = np.sqrt(np.mean((y_test_pred.flatten() - test_y.flatten())**2))
print(f"Test RMSE: {rmse}")
```

Test RMSE: 26.458723208132696

Predict next day

Generally, what we are trying to predict is a pollution indicator for the day or per 12-hour period.

Modify the datasets to create a new column giving a pollution indicator per half day: little pollution, moderate pollution, heavy pollution.

We make a date id column of shape YYYYMMDD with padding zeros on the left to make sure that:

- · years have 4 characters
- months have 2 characters
- days have 2 characters

We also add a digit indicating whether the given hour is in the morning or in the afternoon.

```
In []:
    data.index = pd.to_datetime(data.index)
    idx = data.index

def date_to_id(date):
    """converts a date to its id"""
```

```
return (
       str(date.year).zfill(4)
       + str(date.month).zfill(2)
       + str(date.day).zfill(2)
       + str(int(date.hour >= 12)) # 0 if morning, else 1
# Convert dates to date id's
day id = pd.Series(idx).apply(date to id)
# Convert to list, otherwise all entries become nan when added to the df
data["day id"] = day id.values
# Group data by max of `day id` instead of something like mean
# because it is common in meteorology to consider whether a certain threshold is passed,
# rather than the mean over hour.
grouped data = data.groupby("day id").max()
print(f"\n---> Data grouped by day id\n{grouped data.head()}")
pollution halfday = grouped data.copy()["pollution"].to dict()
print(f"\n---> Dictionary conveting day id to max pollution that halfday:\n{pollution halfday}")
data["pollution"] = [pollution halfday[idx] for idx in data["day id"]]
data = data.drop(columns="day id")
data.head(20)
```

```
---> Data grouped by day id
          pollution dew temp
                                 press wnd dir wnd spd snow rain
day id
201001020
              181.0
                      -7 -4.0 1026.0
                                            SE
                                                 20.56
                                                                 0
201001021
              170.0
                      -7 -5.0 1028.0
                                            SE
                                                 55.43
                                                           3
201001030
                      -7 -6.0 1027.0
                                                102.80
               98.0
                                                          16
201001031
              107.0 -10 -9.0 1023.0
                                                127.84
                                                          27
                                                                 0
               79.0 -14 -9.0 1031.0
                                                108.61
                                                           0
                                                                 0
201001040
                                            NW
```

---> Dictionary conveting day id to max pollution that halfday: {'201001020': 181.0, '201001021': 170.0, '201001030': 98.0, '201001031': 107.0, '201001040': 79.0, '201001041': 33.0, '201001050': 36.0, '201001051': 106.0, '201001060': 77.0, '201001061': 132.0, '201001070': 130.0, '201001071': 198.0, '201001080': 275.0, '201001081': 250.0, '201001090': 196.0, '201001091': 66.0, '201001100': 83.0, '201001101': 88.0, '201001110': 27.0, '201001111': 28.0, '201001120': 37.0, '201001121': 31.0, '201001130': 36.0, '201001131': 96.0, '20 1001140': 257.0, '201001141': 94.0, '201001150': 37.0, '201001151': 102.0, '201001160': 271.0, '201001161': 269.0, '2 01001170': 263.0, '201001171': 317.0, '201001180': 407.0, '201001181': 435.0, '201001190': 383.0, '201001191': 485.0, '201001200': 389.0, '201001201': 29.0, '201001210': 26.0, '201001211': 72.0, '201001220': 35.0, '201001221': 49.0, '2 01001230': 49.0, '201001231': 22.0, '201001240': 0.0, '201001241': 0.0, '201001250': 0.0, '201001251': 0.0, '20100126 0': 0.0, '201001261': 340.0, '201001270': 300.0, '201001271': 143.0, '201001280': 34.0, '201001281': 25.0, '20100129 0': 25.0, '201001291': 52.0, '201001300': 70.0, '201001301': 94.0, '201001310': 85.0, '201001311': 145.0, '20100201 0': 78.0, '201002011': 72.0, '201002020': 104.0, '201002021': 88.0, '201002030': 161.0, '201002031': 114.0, '20100204 0': 104.0, '201002041': 91.0, '201002050': 126.0, '201002051': 100.0, '201002060': 85.0, '201002061': 189.0, '2010020 70': 92.0, '201002071': 164.0, '201002080': 197.0, '201002081': 258.0, '201002090': 273.0, '201002091': 182.0, '20100 2100': 22.0, '201002101': 18.0, '201002110': 12.0, '201002111': 24.0, '201002120': 26.0, '201002121': 69.0, '20100213 0': 108.0, '201002131': 267.0, '201002140': 980.0, '201002141': 67.0, '201002150': 45.0, '201002151': 116.0, '2010021 60': 146.0, '201002161': 198.0, '201002170': 247.0, '201002171': 49.0, '201002180': 107.0, '201002181': 296.0, '20100 2190': 191.0, '201002191': 165.0, '201002200': 205.0, '201002201': 282.0, '201002210': 261.0, '201002211': 154.0, '20 1002220': 154.0, '201002221': 100.0, '201002230': 150.0, '201002231': 282.0, '201002240': 302.0, '201002241': 368.0, '201002250': 266.0, '201002251': 248.0, '201002260': 82.0, '201002261': 103.0, '201002270': 138.0, '201002271': 124. 0, '201002280': 145.0, '201002281': 161.0, '201003010': 69.0, '201003011': 99.0, '201003020': 169.0, '201003021': 13 9.0, '201003030': 178.0, '201003031': 247.0, '201003040': 187.0, '201003041': 269.0, '201003050': 248.0, '201003051': 30.0, '201003060': 29.0, '201003061': 103.0, '201003070': 74.0, '201003071': 134.0, '201003080': 76.0, '201003081': 6 3.0, '201003090': 20.0, '201003091': 51.0, '201003100': 111.0, '201003101': 89.0, '201003110': 224.0, '201003111': 22 3.0, '201003120': 314.0, '201003121': 91.0, '201003130': 82.0, '201003131': 78.0, '201003140': 105.0, '201003141': 11 6.0, '201003150': 220.0, '201003151': 52.0, '201003160': 79.0, '201003161': 89.0, '201003170': 85.0, '201003171': 76. 0, '201003180': 166.0, '201003181': 239.0, '201003190': 249.0, '201003191': 250.0, '201003200': 700.0, '201003201': 1 56.0, '201003210': 60.0, '201003211': 121.0, '201003220': 784.0, '201003221': 226.0, '201003230': 86.0, '201003231': 69.0, '201003240': 76.0, '201003241': 134.0, '201003250': 69.0, '201003251': 59.0, '201003260': 72.0, '201003261': 14 2.0. '201003270': 134.0. '201003271': 71.0. '201003280': 99.0. '201003281': 120.0. '201003290': 218.0. '201003291': 3 66.0, '201003300': 288.0, '201003301': 195.0, '201003310': 129.0, '201003311': 200.0, '201004010': 49.0, '201004011': 54.0, '201004020': 52.0, '201004021': 77.0, '201004030': 138.0, '201004031': 168.0, '201004040': 206.0, '201004041': 173.0, '201004050': 199.0, '201004051': 186.0, '201004060': 33.0, '201004061': 34.0, '201004070': 135.0, '201004071': 140.0, '201004080': 107.0, '201004081': 158.0, '201004090': 150.0, '201004091': 121.0, '201004100': 102.0, '20100410 1': 58.0, '201004110': 91.0, '201004111': 87.0, '201004120': 116.0, '201004121': 88.0, '201004130': 19.0, '20100413

1': 69.0, '201004140': 100.0, '201004141': 96.0, '201004150': 206.0, '201004151': 237.0, '201004160': 266.0, '2010041 61': 289.0, '201004170': 165.0, '201004171': 223.0, '201004180': 212.0, '201004181': 351.0, '201004190': 389.0, '2010 04191': 198.0, '201004200': 110.0, '201004201': 95.0, '201004210': 52.0, '201004211': 51.0, '201004220': 53.0, '20100 4221': 47.0, '201004230': 65.0, '201004231': 39.0, '201004240': 122.0, '201004241': 117.0, '201004250': 142.0, '20100 4251': 135.0, '201004260': 20.0, '201004261': 26.0, '201004270': 46.0, '201004271': 50.0, '201004280': 27.0, '2010042 81': 38.0, '201004290': 21.0, '201004291': 27.0, '201004300': 40.0, '201004301': 54.0, '201005010': 97.0, '20100501 1': 80.0, '201005020': 151.0, '201005021': 65.0, '201005030': 151.0, '201005031': 160.0, '201005040': 133.0, '2010050 41': 228.0, '201005050': 283.0, '201005051': 260.0, '201005060': 20.0, '201005061': 41.0, '201005070': 165.0, '201005 071': 228.0, '201005080': 203.0, '201005081': 163.0, '201005090': 294.0, '201005091': 126.0, '201005100': 132.0. '201 005101': 63.0, '201005110': 13.0, '201005111': 54.0, '201005120': 55.0, '201005121': 54.0, '201005130': 83.0, '201005 131': 83.0, '201005140': 138.0, '201005141': 216.0, '201005150': 238.0, '201005151': 188.0, '201005160': 185.0, '2010 05161': 137.0. '201005170': 82.0, '201005171': 96.0, '201005180': 105.0, '201005181': 72.0, '201005190': 50.0, '20100 5191': 44.0, '201005200': 113.0, '201005201': 78.0, '201005210': 134.0, '201005211': 234.0, '201005220': 229.0, '2010 05221': 174.0, '201005230': 176.0, '201005231': 165.0, '201005240': 100.0, '201005241': 45.0, '201005250': 46.0, '201 005251': 57.0, '201005260': 139.0, '201005261': 80.0, '201005270': 108.0, '201005271': 96.0, '201005280': 157.0, '201 005281': 206.0, '201005290': 251.0, '201005291': 314.0, '201005300': 54.0, '201005301': 97.0, '201005310': 118.0, '20 1005311': 179.0, '201006010': 189.0, '201006011': 178.0, '201006020': 0.0, '201006021': 159.0, '201006030': 190.0, '2 01006031': 122.0, '201006040': 101.0, '201006041': 0.0, '201006050': 0.0, '201006051': 0.0, '201006060': 0.0, '201006 061': 0.0, '201006070': 0.0, '201006071': 172.0, '201006080': 191.0, '201006081': 0.0, '201006090': 0.0, '201006091': 70.0, '201006100': 63.0, '201006101': 80.0, '201006110': 97.0, '201006111': 100.0, '201006120': 160.0, '201006121': 1 29.0, '201006130': 154.0, '201006131': 198.0, '201006140': 82.0, '201006141': 142.0, '201006150': 229.0, '201006151': 183.0, '201006160': 155.0, '201006161': 139.0, '201006170': 51.0, '201006171': 50.0, '201006180': 93.0, '201006181': 100.0, '201006190': 120.0, '201006191': 117.0, '201006200': 240.0, '201006201': 71.0, '201006210': 86.0, '201006211': 132.0, '201006220': 140.0, '201006221': 0.0, '201006230': 136.0, '201006231': 69.0, '201006240': 120.0, '201006241': 98.0, '201006250': 231.0, '201006251': 227.0, '201006260': 252.0, '201006261': 160.0, '201006270': 163.0, '20100627 1': 131.0, '201006280': 166.0, '201006281': 200.0, '201006290': 249.0, '201006291': 196.0, '201006300': 205.0, '20100 6301': 235.0, '201007010': 241.0, '201007011': 124.0, '201007020': 46.0, '201007021': 41.0, '201007030': 94.0, '20100 7031': 100.0, '201007040': 168.0, '201007041': 87.0, '201007050': 132.0, '201007051': 44.0, '201007060': 62.0, '20100 7061': 59.0, '201007070': 57.0, '201007071': 80.0, '201007080': 77.0, '201007081': 73.0, '201007090': 96.0, '20100709 1': 123.0, '201007100': 108.0, '201007101': 117.0, '201007110': 181.0, '201007111': 231.0, '201007120': 164.0, '20100 7121': 211.0, '201007130': 206.0, '201007131': 121.0, '201007140': 143.0, '201007141': 174.0, '201007150': 216.0, '20 1007151': 286.0, '201007160': 310.0, '201007161': 283.0, '201007170': 196.0, '201007171': 186.0, '201007180': 267.0, '201007181': 177.0, '201007190': 275.0, '201007191': 286.0, '201007200': 63.0, '201007201': 49.0, '201007210': 109.0, '201007211': 79.0, '201007220': 227.0, '201007221': 206.0, '201007230': 255.0, '201007231': 201.0, '201007240': 125. 0, '201007241': 145.0, '201007250': 231.0, '201007251': 150.0, '201007260': 251.0, '201007261': 269.0, '201007270': 2 43.0, '201007271': 192.0, '201007280': 198.0, '201007281': 198.0, '201007290': 192.0, '201007291': 285.0, '20100730 0': 246.0, '201007301': 302.0, '201007310': 306.0, '201007311': 46.0, '201008010': 66.0, '201008011': 70.0, '20100802 0': 117.0, '201008021': 100.0, '201008030': 184.0, '201008031': 199.0, '201008040': 205.0, '201008041': 137.0, '20100 8050': 45.0, '201008051': 30.0, '201008060': 25.0, '201008061': 48.0, '201008070': 108.0, '201008071': 119.0, '201008 080': 71.0. 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 $\verb"Out[]: pollution dew temp press wnd_dir wnd_spd snow rain"$

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2010-01-02 02:00:00	181.0	-11	-5.0	1021.0	SE	3.57	0	0
2010-01-02 03:00:00	181.0	-7	-5.0	1022.0	SE	5.36	1	0
2010-01-02 04:00:00	181.0	-7	-5.0	1022.0	SE	6.25	2	0
2010-01-02 05:00:00	181.0	-7	-6.0	1022.0	SE	7.14	3	0
2010-01-02 06:00:00	181.0	-7	-6.0	1023.0	SE	8.93	4	0
2010-01-02 07:00:00	181.0	-7	-5.0	1024.0	SE	10.72	0	0
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2010-01-02 14:00:00	170.0	-9	-5.0	1025.0	SE	31.73	0	0
2010-01-02 15:00:00	170.0	-9	-5.0	1025.0	SE	35.75	0	0
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2010-01-02 17:00:00	170.0	-8	-5.0	1027.0	SE	39.33	0	0
2010-01-02 18:00:00	170.0	-8	-5.0	1027.0	SE	42.46	0	0
2010-01-02 19:00:00	170.0	-8	-5.0	1028.0	SE	44.25	0	0

In []: train_y

```
array([105., 124., 120., ..., 19., 18., 17.])
In [ ]:
         # get the values
         dataset = series to supervised(data=data, n in=n in, n out=n out, output=output)
         values = dataset.values
         # split into train and test sets
         n train hours = 365 * 24
         train = values[:n train hours, :]
         val = values[n train hours:2*n train hours, :]
         test = values[2*n train hours:, :]
         # redefine y values (X doesn't change)
         train y = np.array(train[:, -n out], dtype="float64")
         val y = np.array(val[:, -n out], dtype="float64")
         test y = np.array(test[:, -n out], dtype="float64")
In [ ]:
         train y.shape
        (8760,)
Out[ ]:
In [ ]:
         history = build and fit(model, train X enc, train_y, val_X_enc, val_y, test_X_enc, test_y, patience=30)
```

```
Epoch 1/200
mae: 36.6373
Epoch 2/200
mae: 38.6805
Epoch 3/200
mae: 33.4959
Epoch 4/200
mae: 35.2939
Epoch 5/200
mae: 37.0781
Epoch 6/200
mae: 34.9387
Epoch 7/200
mae: 33.9130
Epoch 8/200
mae: 36.5352
Epoch 9/200
mae: 32.7245
Epoch 10/200
mae: 37.0198
Epoch 11/200
mae: 38.1343
Epoch 12/200
mae: 38.9867
Epoch 13/200
mae: 43.3789
Epoch 14/200
mae: 35.2863
Epoch 15/200
```

```
mae: 33.9405
Epoch 16/200
mae: 37.0641
Epoch 17/200
mae: 36.7336
Epoch 18/200
mae: 34.5824
Epoch 19/200
mae: 33.8812
Epoch 20/200
mae: 38.5889
Epoch 21/200
mae: 38.5736
Epoch 22/200
mae: 35.1241
Epoch 23/200
mae: 34.2250
Epoch 24/200
mae: 36.7958
Epoch 25/200
mae: 39.0546
Epoch 26/200
mae: 39.8250
Epoch 27/200
mae: 37.4756
Epoch 28/200
mae: 37.4765
Epoch 29/200
mae: 36.8882
Epoch 30/200
```

```
mae: 39.5373
Epoch 31/200
mae: 38.5956
Epoch 32/200
mae: 39.0660
Epoch 33/200
mae: 36.7832
Epoch 34/200
mae: 34.9776
Epoch 35/200
mae: 35.4257
Epoch 36/200
nd of the best epoch: 6.
mae: 35.1554
Epoch 00036: early stopping
```

