#### In [1]:

```
import os
import tensorflow as tf
import numpy as np
import pandas as pd

from sklearn.model_selection import train_test_split
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import LSTM, Dense
from tensorflow.keras.losses import mse

from tensorflow.keras.metrics import RootMeanSquaredError, mean_squared_error
from tensorflow.keras.optimizers import Adam
import matplotlib.pyplot as plt
```

#### In [2]:

```
zip_path = tf.keras.utils.get_file(
    origin = "http://storage.googleapis.com/tensorflow/tf-keras-datasets/jena_climate_2009_2016.csv.
    fname = "jena_climate_2009_2016.csv.zip",
    extract = True)
```

#### In [3]:

```
csv_path, csv_extension = os.path.splitext(zip_path)
print("파일 경로 : ", csv_path, "파일 확장자명 : ", csv_extension)
```

파일 경로 : C:₩Users\wogml\.keras\datasets\jena\_climate\_2009\_2016.csv 파일 확장자명 : .zip

#### In [4]:

```
import numpy as np
import pandas as pd
import tensorflow as tf

import os

from sklearn.model_selection import train_test_split
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import LSTM, Dense
from tensorflow.keras.losses import mse

from tensorflow.keras.metrics import RootMeanSquaredError, mean_squared_error
from tensorflow.keras.optimizers import Adam
import matplotlib.pyplot as plt
```

### In [5]:

```
weather_data = pd.read_csv(csv_path)
weather_data
```

### Out[5]:

	Date Time	p (mbar)	T (degC)	Tpot (K)	Tdew (degC)	rh (%)	VPmax (mbar)	VPact (mbar)	VPdef (mbar)	sh (g/kg)	(mm
0	01.01.2009 00:10:00	996.52	-8.02	265.40	-8.90	93.30	3.33	3.11	0.22	1.94	
1	01.01.2009 00:20:00	996.57	-8.41	265.01	-9.28	93.40	3.23	3.02	0.21	1.89	
2	01.01.2009 00:30:00	996.53	-8.51	264.91	-9.31	93.90	3.21	3.01	0.20	1.88	
3	01.01.2009 00:40:00	996.51	-8.31	265.12	-9.07	94.20	3.26	3.07	0.19	1.92	
4	01.01.2009 00:50:00	996.51	-8.27	265.15	-9.04	94.10	3.27	3.08	0.19	1.92	
420546	31.12.2016 23:20:00	1000.07	-4.05	269.10	-8.13	73.10	4.52	3.30	1.22	2.06	
420547	31.12.2016 23:30:00	999.93	-3.35	269.81	-8.06	69.71	4.77	3.32	1.44	2.07	
420548	31.12.2016 23:40:00	999.82	-3.16	270.01	-8.21	67.91	4.84	3.28	1.55	2.05	
420549	31.12.2016 23:50:00	999.81	-4.23	268.94	-8.53	71.80	4.46	3.20	1.26	1.99	
420550	01.01.2017 00:00:00	999.82	-4.82	268.36	-8.42	75.70	4.27	3.23	1.04	2.01	

420551 rows × 15 columns

### In [6]:

```
print("Date Time 컬럼 데이터의 type : ", type(weather_data["Date Time"][0]))
```

Date Time 컬럼 데이터의 type : <class 'str'>

#### In [7]:

```
for i in range(len(weather_data)):
    if weather_data["Date Time"][i][6:10] == '2016':
        print("2016년 데이터는 ", str(i), "번째 행부터 시작입니다.")
        break
```

2016년 데이터는 368291 번째 행부터 시작입니다.

## In [8]:

```
weather_data = weather_data.iloc[368291:]
weather_data = weather_data.reset_index(drop = True)
weather_data
```

## Out[8]:

	Date Time	p (mbar)	T (degC)	Tpot (K)	Tdew (degC)	rh (%)	VPmax (mbar)	VPact (mbar)	VPdef (mbar)	sh (g/kg)	(mmo
0	01.01.2016 00:00:00	999.08	-0.01	273.22	-0.44	96.90	6.10	5.91	0.19	3.69	
1	01.01.2016 00:10:00	999.03	0.01	273.25	-0.41	97.00	6.11	5.93	0.18	3.70	
2	01.01.2016 00:20:00	999.07	0.06	273.29	-0.36	97.00	6.13	5.95	0.18	3.71	
3	01.01.2016 00:30:00	999.09	0.07	273.30	-0.36	96.90	6.14	5.95	0.19	3.71	
4	01.01.2016 00:40:00	999.09	-0.05	273.18	-0.50	96.80	6.09	5.89	0.19	3.68	
52255	31.12.2016 23:20:00	1000.07	-4.05	269.10	-8.13	73.10	4.52	3.30	1.22	2.06	
52256	31.12.2016 23:30:00	999.93	-3.35	269.81	-8.06	69.71	4.77	3.32	1.44	2.07	
52257	31.12.2016 23:40:00	999.82	-3.16	270.01	-8.21	67.91	4.84	3.28	1.55	2.05	
52258	31.12.2016 23:50:00	999.81	-4.23	268.94	-8.53	71.80	4.46	3.20	1.26	1.99	
52259	01.01.2017 00:00:00	999.82	-4.82	268.36	-8.42	75.70	4.27	3.23	1.04	2.01	

52260 rows × 15 columns

# In [9]:

```
weather_data = weather_data.iloc[:, 1:]
weather_data
```

## Out[9]:

	p (mbar)	T (degC)	Tpot (K)	Tdew (degC)	rh (%)	VPmax (mbar)	VPact (mbar)	VPdef (mbar)	sh (g/kg)	H2OC (mmol/mol)	ı (g/m³
0	999.08	-0.01	273.22	-0.44	96.90	6.10	5.91	0.19	3.69	5.92	1271
1	999.03	0.01	273.25	-0.41	97.00	6.11	5.93	0.18	3.70	5.94	1271
2	999.07	0.06	273.29	-0.36	97.00	6.13	5.95	0.18	3.71	5.96	1270
3	999.09	0.07	273.30	-0.36	96.90	6.14	5.95	0.19	3.71	5.96	1270
4	999.09	-0.05	273.18	-0.50	96.80	6.09	5.89	0.19	3.68	5.90	1271
52255	1000.07	-4.05	269.10	-8.13	73.10	4.52	3.30	1.22	2.06	3.30	1292
52256	999.93	-3.35	269.81	-8.06	69.71	4.77	3.32	1.44	2.07	3.32	1289
52257	999.82	-3.16	270.01	-8.21	67.91	4.84	3.28	1.55	2.05	3.28	1288
52258	999.81	-4.23	268.94	-8.53	71.80	4.46	3.20	1.26	1.99	3.20	1293
52259	999.82	-4.82	268.36	-8.42	75.70	4.27	3.23	1.04	2.01	3.23	1296

52260 rows × 14 columns

### In [10]:

```
def extract_inputoutput(dataframe, lookback_time = 2, predict_time = 1):
   dfx = pd.DataFrame()
   dfy = pd.DataFrame()
   for i in range(len(dataframe) - (lookback_time - 1) - (predict_time)):
       if i \% 10000 == 0:
           print(i)
       rowx = []
       for timestep in range(lookback_time):
           dfRename = dataframe.iloc[[i + timestep]]
           dfRename.index = [i]
           rowx.append(dfRename)
       rowx = pd.concat(rowx, axis = 1)
       dfx = pd.concat([dfx, rowx])
       rowx = []
       rowy = pd.DataFrame([dataframe["T (degC)"][i + lookback_time]])
       dfy = pd.concat([dfy, rowy], ignore_index = True)
   print("X, Y 데이터 분류 완료!")
   return dfx, dfy
```

#### In [11]:

```
x, t = extract_inputoutput(weather_data)
0
10000
```

30000 40000 50000

20000

X, Y 데이터 분류 완료!

## In [12]:

Х

## Out[12]:

	p (mbar)	T (degC)	Tpot (K)	Tdew (degC)	rh (%)	VPmax (mbar)	VPact (mbar)	VPdef (mbar)	sh (g/kg)	H2OC (mmol/mol)	
0	999.08	-0.01	273.22	-0.44	96.90	6.10	5.91	0.19	3.69	5.92	 9
1	999.03	0.01	273.25	-0.41	97.00	6.11	5.93	0.18	3.70	5.94	 9
2	999.07	0.06	273.29	-0.36	97.00	6.13	5.95	0.18	3.71	5.96	 9
3	999.09	0.07	273.30	-0.36	96.90	6.14	5.95	0.19	3.71	5.96	 9
4	999.09	-0.05	273.18	-0.50	96.80	6.09	5.89	0.19	3.68	5.90	 9
52253	1000.21	-3.76	269.39	-7.95	72.50	4.62	3.35	1.27	2.09	3.35	 7
52254	1000.11	-3.93	269.23	-8.09	72.60	4.56	3.31	1.25	2.06	3.31	 7
52255	1000.07	-4.05	269.10	-8.13	73.10	4.52	3.30	1.22	2.06	3.30	 6
52256	999.93	-3.35	269.81	-8.06	69.71	4.77	3.32	1.44	2.07	3.32	 6
52257	999.82	-3.16	270.01	-8.21	67.91	4.84	3.28	1.55	2.05	3.28	 7

52258 rows × 28 columns

## In [13]:

t

## Out[13]:

	0
0	0.06
1	0.07
2	-0.05
3	0.07
4	-0.05
52253	-4.05
52254	-3.35
52255	-3.16
52256	-4.23

**52257** -4.82

52258 rows × 1 columns

#### In [14]:

```
x_train, x_test, t_train, t_test = train_test_split(x, t, test_size=0.2, shuffle = False)
print("x_train shape : ", x_train.shape)
print("t_train shape : ", t_train.shape)
print("x_test shape : ", x_test.shape)
print("t_test shape : ", t_test.shape)
```

x\_train shape : (41806, 28) t\_train shape : (41806, 1) x\_test shape : (10452, 28) t\_test shape : (10452, 1)

#### In [15]:

```
timesteps = 2
feature = 14

x_train = np.array(x_train)
x_train = x_train.reshape(x_train.shape[0], timesteps, feature)

x_test = np.array(x_test)
x_test = x_test.reshape(x_test.shape[0], timesteps, feature)

t_train = np.array(t_train)
t_test = np.array(t_test)

print("reshape \( \frac{\pi}{2} \) x_train.shape)
print("t_train shape : ", t_train.shape)
print("reshape \( \frac{\pi}{2} \) x_test.shape : ", x_test.shape)
print("t_test shape : ", t_test.shape)
```

```
reshape 亭 x_train shape : (41806, 2, 14)
t_train shape : (41806, 1)
reshape 亭 x_test shape : (10452, 2, 14)
t_test shape : (10452, 1)
```

### In [16]:

```
cell_size = 128
timesteps = 2
feature = 14

model = Sequential(name = "Temp_LSTM")

model.add(LSTM(cell_size, input_shape = (timesteps, feature), return_sequences = True))
model.add(LSTM(cell_size))

model.add(Dense(1))

model.compile(loss = mse, optimizer = Adam(learning_rate=0.001), metrics = [RootMeanSquaredError()]
model.summary()
```

Model: "Temp\_LSTM"

Layer (type)	Output Shape	Param #
Istm (LSTM)	(None, 2, 128)	73216
lstm_1 (LSTM)	(None, 128)	131584
dense (Dense)	(None, 1)	129

Total params: 204,929 Trainable params: 204,929 Non-trainable params: 0

# In [17]:

model.fit(x\_train, t\_train, epochs=20, batch\_size = 32)

```
Epoch 1/20
1307/1307 [===========] - 24s 13ms/step - loss: 10.8442 - root_me
an_squared_error: 3.2930
Epoch 2/20
1307/1307 [=======
                       =============] - 16s 13ms/step - loss: 1.1526 - root_mea
n_squared_error: 1.0736
Epoch 3/20
1307/1307 [======
                             =======] - 17s 13ms/step - loss: 1.0211 - root_mea
n_squared_error: 1.0105
Epoch 4/20
1307/1307 [=======
                            =======] - 17s 13ms/step - loss: 0.8664 - root_mea
n_squared_error: 0.9308
Epoch 5/20
1307/1307 [=======
                            ========] - 17s 13ms/step - loss: 0.7837 - root_mea
n_squared_error: 0.8853
Epoch 6/20
                           ======] - 17s 13ms/step - loss: 0.7308 - root_mea
1307/1307 [======
n_squared_error: 0.8549
Epoch 7/20
1307/1307 [========
                         ========] - 17s 13ms/step - loss: 0.6293 - root_mea
n_squared_error: 0.7933
Epoch 8/20
1307/1307 [============] - 17s 13ms/step - loss: 0.6107 - root_mea
n_squared_error: 0.7815
Epoch 9/20
1307/1307 [=====
                            =======] - 17s 13ms/step - loss: 0.6499 - root_mea
n_squared_error: 0.8062
Epoch 10/20
1307/1307 [======
                            =======] - 17s 13ms/step - loss: 0.5708 - root_mea
n_squared_error: 0.7555
Epoch 11/20
1307/1307 [=========
                          ========] - 17s 13ms/step - loss: 0.5601 - root_mea
n_squared_error: 0.7484
Epoch 12/20
1307/1307 [============= ] - 17s 13ms/step - loss: 0.5073 - root_mea
n_squared_error: 0.7122
Epoch 13/20
1307/1307 [=====
                          ========] - 17s 13ms/step - loss: 0.4964 - root_mea
n_squared_error: 0.7046
Epoch 14/20
1307/1307 [===========] - 18s 14ms/step - loss: 0.5218 - root_mea
n_squared_error: 0.7224
Epoch 15/20
1307/1307 [============= ] - 19s 14ms/step - loss: 0.4651 - root_mea
n_squared_error: 0.6820
Epoch 16/20
1307/1307 [=======
                            ========] - 19s 14ms/step - loss: 0.5303 - root_mea
n_squared_error: 0.7282
Epoch 17/20
1307/1307 [========
                       ==========] - 19s 14ms/step - loss: 0.5475 - root_mea
n_squared_error: 0.7399
Epoch 18/20
1307/1307 [============= ] - 19s 14ms/step - loss: 0.5037 - root_mea
n_squared_error: 0.7097
Epoch 19/20
1307/1307 [======
                          ========] - 18s 14ms/step - loss: 0.4766 - root_mea
n_squared_error: 0.6904
Epoch 20/20
1307/1307 [=======
                         =========] - 18s 14ms/step - loss: 0.4760 - root_mea
n_squared_error: 0.6899
```

### Out[17]:

<keras.callbacks.History at 0x21584bdb3d0>

#### In [18]:

test RMSE: 0.43

```
pred = model.predict(x_test)
for i in range(1, 10):
   print("온도(T(deg C)) 예측 : ", round(pred[i][0], 2), "/ 정답 : ", round(t_test[i][0], 2))
327/327 [=========
                          ======] - 4s 7ms/step
온도(T(deg C)) 예측 : 6.14 / 정답 : 5.94
온도(T(deg C)) 예측 : 6.17 / 정답 : 6.16
온도(T(deg C)) 예측 : 6.22 / 정답 : 6.43
온도(T(deg C)) 예측 : 6.36 / 정답 : 6.58
온도(T(deg C)) 예측 : 6.47 / 정답 : 6.93
온도(T(deg C)) 예측 : 6.62 / 정답 : 7.21
온도(T(deg C)) 예측 : 6.83 / 정답 : 7.38
온도(T(deg C)) 예측 : 7.01 / 정답 : 7.56
온도(T(deg C)) 예측 : 7.12 / 정답 : 8.14
In [19]:
loss, rmse = model.evaluate(x_test, t_test, verbose = 1)
print("test loss (MSE) : ", round(loss, 6))
print("test RMSE : ", round(rmse, 2))
                   ========] - 4s 7ms/step - loss: 0.1818 - root_mean_sq
327/327 [=======
uared_error: 0.4263
test loss (MSE): 0.181767
```