

ASSIGNMENT -3

November 5, 2023

Time Series

```
[1]: !rm jena_climate_2009_2016.csv.zip
```

rm: cannot remove 'jena_climate_2009_2016.csv.zip': No such file or directory

```
[2]: !wget https://s3.amazonaws.com/keras-datasets/jena_climate_2009_2016.csv.zip
!unzip jena_climate_2009_2016.csv.zip
```

```
--2023-11-05 19:45:34-- https://s3.amazonaws.com/keras-
datasets/jena_climate_2009_2016.csv.zip
Resolving s3.amazonaws.com (s3.amazonaws.com)... 52.216.76.94, 16.182.73.104,
16.182.65.24, ...
Connecting to s3.amazonaws.com (s3.amazonaws.com)|52.216.76.94|:443...
connected.
HTTP request sent, awaiting response... 200 OK
Length: 13565642 (13M) [application/zip]
Saving to: 'jena_climate_2009_2016.csv.zip'
```

```
100%[=====>] 13,565,642 10.0MB/s in 1.3s
```

```
2023-11-05 19:45:35 (10.0 MB/s) - 'jena_climate_2009_2016.csv.zip' saved
[13565642/13565642]
```

```
Archive: jena_climate_2009_2016.csv.zip
  inflating: jena_climate_2009_2016.csv
  inflating: __MACOSX/._jena_climate_2009_2016.csv
```

PROCESSING THE DATA:

```
[2]: import os
fname = os.path.join("jena_climate_2009_2016.csv")

with open(fname) as f:
    data = f.read()

lines = data.split("\n")
header = lines[0].split(",")
lines = lines[1:]
```

```
print(header)
print(len(lines))
```

```
['Date Time', 'p (mbar)', 'T (degC)', 'Tpot (K)', 'Tdew (degC)', 'rh (%)', 'VPmax (mbar)', 'VPact (mbar)', 'VPdef (mbar)', 'sh (g/kg)', 'H2OC (mmol/mol)', 'rho (g/m**3)', 'wv (m/s)', 'max. wv (m/s)', 'wd (deg)']
420451
```

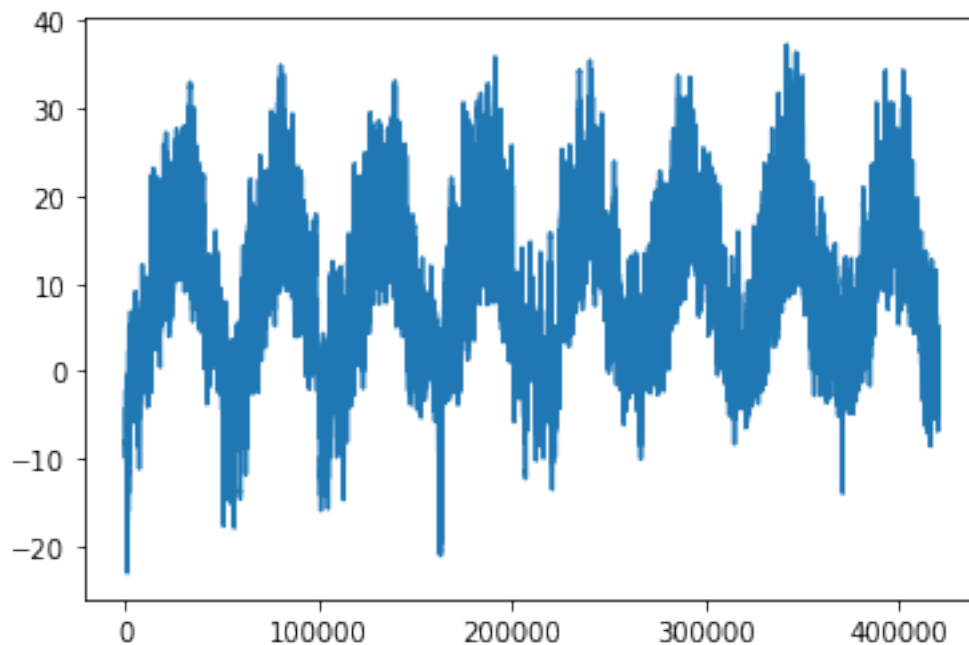
PLOTTING THE TEMPERATURE SERIES:

```
[4]: import numpy as np
temperature = np.zeros((len(lines),))
raw_data = np.zeros((len(lines), len(header) - 1))
for i, line in enumerate(lines):
    values = [float(x) for x in line.split(",")[1:]]
    temperature[i] = values[1]
    raw_data[i, :] = values[:]
```

PLOTTING THE FIRST 10 SERIES:

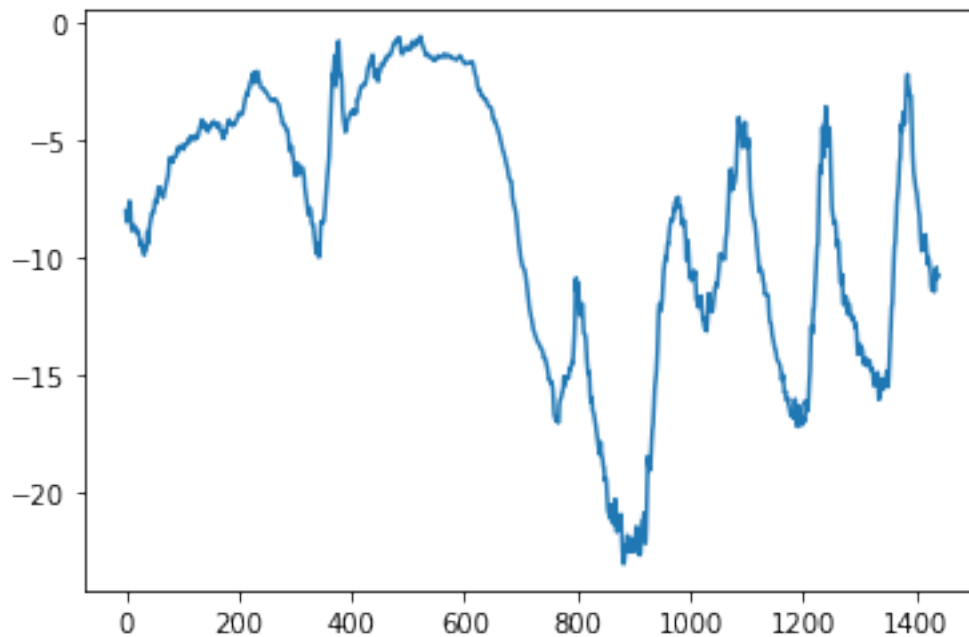
```
[5]: from matplotlib import pyplot as plt
plt.plot(range(len(temperature)), temperature)
```

```
[5]: [matplotlib.lines.Line2D at 0x7fcd67de5a58]
```



```
[6]: plt.plot(range(1440), temperature[:1440])
```

[6]: [<matplotlib.lines.Line2D at 0x7fcd5fce1198>]



COMPUTING THE SAMPLES USED FOR EACH DATA SPLIT:

```
[7]: num_train_samples = int(0.5 * len(raw_data))
num_val_samples = int(0.25 * len(raw_data))
num_test_samples = len(raw_data) - num_train_samples - num_val_samples
print("num_train_samples:", num_train_samples)
print("num_val_samples:", num_val_samples)
print("num_test_samples:", num_test_samples)
```

num_train_samples: 210225

num_val_samples: 105112

num_test_samples: 105114

NORMALISING THE DATA:

```
[8]: mean = raw_data[:num_train_samples].mean(axis=0)
raw_data -= mean
std = raw_data[:num_train_samples].std(axis=0)
raw_data /= std
```

```
[9]: import numpy as np
from tensorflow import keras
int_sequence = np.arange(10)
dummy_dataset = keras.utils.timeseries_dataset_from_array(
    data=int_sequence[:-3],
```

```

        targets=int_sequence[3:],
        sequence_length=3,
        batch_size=2,
    )

    for inputs, targets in dummy_dataset:
        for i in range(inputs.shape[0]):
            print([int(x) for x in inputs[i]], int(targets[i]))

```

```

[0, 1, 2] 3
[1, 2, 3] 4
[2, 3, 4] 5
[3, 4, 5] 6
[4, 5, 6] 7

```

TRAINING,TEST AND VALIDATION DATA SET:

```

[10]: sampling_rate = 6
        sequence_length = 120
        delay = sampling_rate * (sequence_length + 24 - 1)
        batch_size = 256

        train_dataset = keras.utils.timeseries_dataset_from_array(
            raw_data[:-delay],
            targets=temperature[delay:],
            sampling_rate=sampling_rate,
            sequence_length=sequence_length,
            shuffle=True,
            batch_size=batch_size,
            start_index=0,
            end_index=num_train_samples)

        val_dataset = keras.utils.timeseries_dataset_from_array(
            raw_data[:-delay],
            targets=temperature[delay:],
            sampling_rate=sampling_rate,
            sequence_length=sequence_length,
            shuffle=True,
            batch_size=batch_size,
            start_index=num_train_samples,
            end_index=num_train_samples + num_val_samples)

        test_dataset = keras.utils.timeseries_dataset_from_array(
            raw_data[:-delay],
            targets=temperature[delay:],
            sampling_rate=sampling_rate,
            sequence_length=sequence_length,
            shuffle=True,

```

```
batch_size=batch_size,  
start_index=num_train_samples + num_val_samples)
```

[]: OUTPUT:

```
[11]: for samples, targets in train_dataset:  
      print("samples shape:", samples.shape)  
      print("targets shape:", targets.shape)  
      break
```

samples shape: (256, 120, 14)

targets shape: (256,)

COMPUTING THE BASELINE MAE: NON MACHINE LEARNING BASELINE

```
[12]: def evaluate_naive_method(dataset):  
      total_abs_err = 0.  
      samples_seen = 0  
      for samples, targets in dataset:  
          preds = samples[:, -1, 1] * std[1] + mean[1]  
          total_abs_err += np.sum(np.abs(preds - targets))  
          samples_seen += samples.shape[0]  
      return total_abs_err / samples_seen  
  
      print(f"Validation MAE: {evaluate_naive_method(val_dataset):.2f}")  
      print(f"Test MAE: {evaluate_naive_method(test_dataset):.2f}")
```

Validation MAE: 2.44

Test MAE: 2.62

TRAINING & EVALUATING BASIC MACHINE LEARNING MODEL: DENSE LAYER

```
[14]: from tensorflow import keras  
      from tensorflow.keras import layers  
  
      inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))  
      x = layers.Flatten()(inputs)  
      x = layers.Dense(16, activation="relu")(x)  
      outputs = layers.Dense(1)(x)  
      model = keras.Model(inputs, outputs)  
  
      callbacks = [  
          keras.callbacks.ModelCheckpoint("jena_dense.keras",  
                                          save_best_only=True)  
      ]  
      model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])  
      history = model.fit(train_dataset,  
                          epochs=7,  
                          validation_data=val_dataset,
```

```

callbacks=callbacks)

model = keras.models.load_model("jena_dense.keras")
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")

```

```

Epoch 1/7
819/819 [=====] - 27s 31ms/step - loss: 13.2995 - mae:
2.8032 - val_loss: 13.0390 - val_mae: 2.8677
Epoch 2/7
819/819 [=====] - 26s 31ms/step - loss: 9.2294 - mae:
2.3870 - val_loss: 12.6537 - val_mae: 2.8345
Epoch 3/7
819/819 [=====] - 26s 31ms/step - loss: 8.5313 - mae:
2.2947 - val_loss: 11.6588 - val_mae: 2.6997
Epoch 4/7
819/819 [=====] - 24s 29ms/step - loss: 8.0392 - mae:
2.2305 - val_loss: 10.8673 - val_mae: 2.6056
Epoch 5/7
819/819 [=====] - 25s 30ms/step - loss: 7.6449 - mae:
2.1748 - val_loss: 10.3317 - val_mae: 2.5345
Epoch 6/7
819/819 [=====] - 24s 30ms/step - loss: 7.4097 - mae:
2.1405 - val_loss: 10.2984 - val_mae: 2.5230
Epoch 7/7
819/819 [=====] - 24s 29ms/step - loss: 7.1609 - mae:
2.1038 - val_loss: 10.8197 - val_mae: 2.5883
405/405 [=====] - 8s 19ms/step - loss: 11.4269 - mae:
2.6639
Test MAE: 2.66

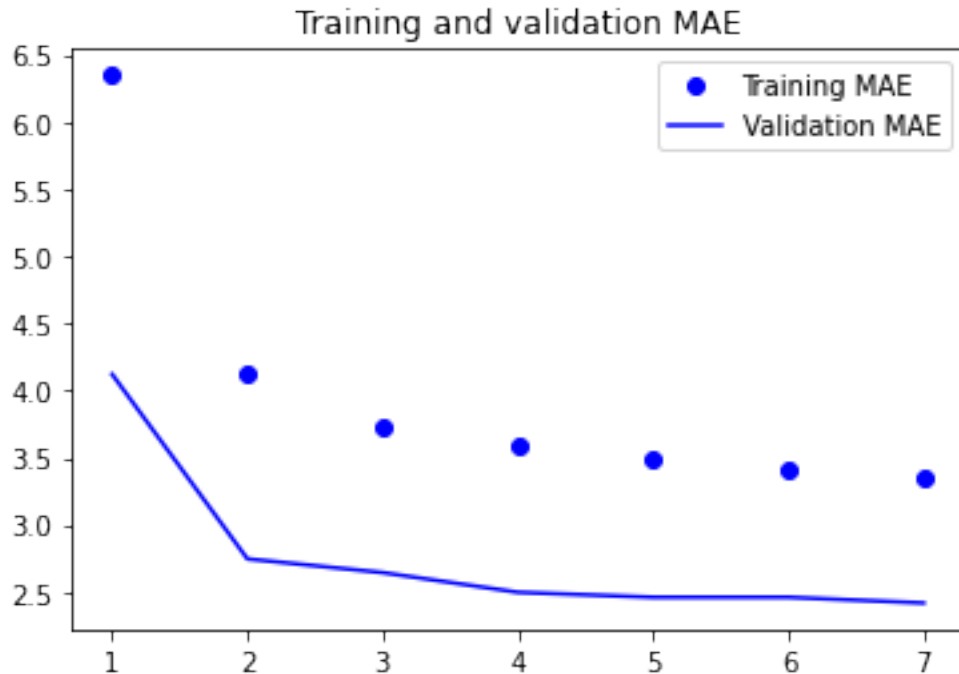
```

PLOTTING THE RESULT:

```

[20]: import matplotlib.pyplot as plt
loss = history.history["mae"]
val_loss = history.history["val_mae"]
epochs = range(1, len(loss) + 1)
plt.figure()
plt.plot(epochs, loss, "bo", label="Training MAE")
plt.plot(epochs, val_loss, "b", label="Validation MAE")
plt.title("Training and validation MAE")
plt.legend()
plt.show()

```



1D CONVOLUTION MODEL:

```
[25]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))
x = layers.Conv1D(8, 24, activation="relu")(inputs)
x = layers.MaxPooling1D(2)(x)
x = layers.Conv1D(8, 12, activation="relu")(x)
x = layers.MaxPooling1D(2)(x)
x = layers.Conv1D(8, 6, activation="relu")(x)
x = layers.GlobalAveragePooling1D()(x)
outputs = layers.Dense(1)(x)
model = keras.Model(inputs, outputs)

callbacks = [
    keras.callbacks.ModelCheckpoint("jena_conv.keras",
                                    save_best_only=True)
]
model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
                    callbacks=callbacks)

model = keras.models.load_model("jena_conv.keras")
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")
```

```

Epoch 1/7
819/819 [=====] - 60s 72ms/step - loss: 23.6116 - mae:
3.8005 - val_loss: 16.3333 - val_mae: 3.2228
Epoch 2/7
819/819 [=====] - 59s 72ms/step - loss: 16.3726 - mae:
3.2044 - val_loss: 14.7552 - val_mae: 3.0325
Epoch 3/7
819/819 [=====] - 58s 71ms/step - loss: 14.7005 - mae:
3.0401 - val_loss: 14.9092 - val_mae: 3.0411
Epoch 4/7
819/819 [=====] - 58s 71ms/step - loss: 13.7397 - mae:
2.9408 - val_loss: 14.6412 - val_mae: 3.0347
Epoch 5/7
819/819 [=====] - 59s 72ms/step - loss: 12.9683 - mae:
2.8536 - val_loss: 16.2347 - val_mae: 3.1999
Epoch 6/7
819/819 [=====] - 58s 71ms/step - loss: 12.3403 - mae:
2.7831 - val_loss: 16.9989 - val_mae: 3.2390
Epoch 7/7
819/819 [=====] - 58s 70ms/step - loss: 11.7086 - mae:
2.7085 - val_loss: 15.7274 - val_mae: 3.1138
405/405 [=====] - 12s 29ms/step - loss: 16.3334 - mae:
3.1985
Test MAE: 3.20

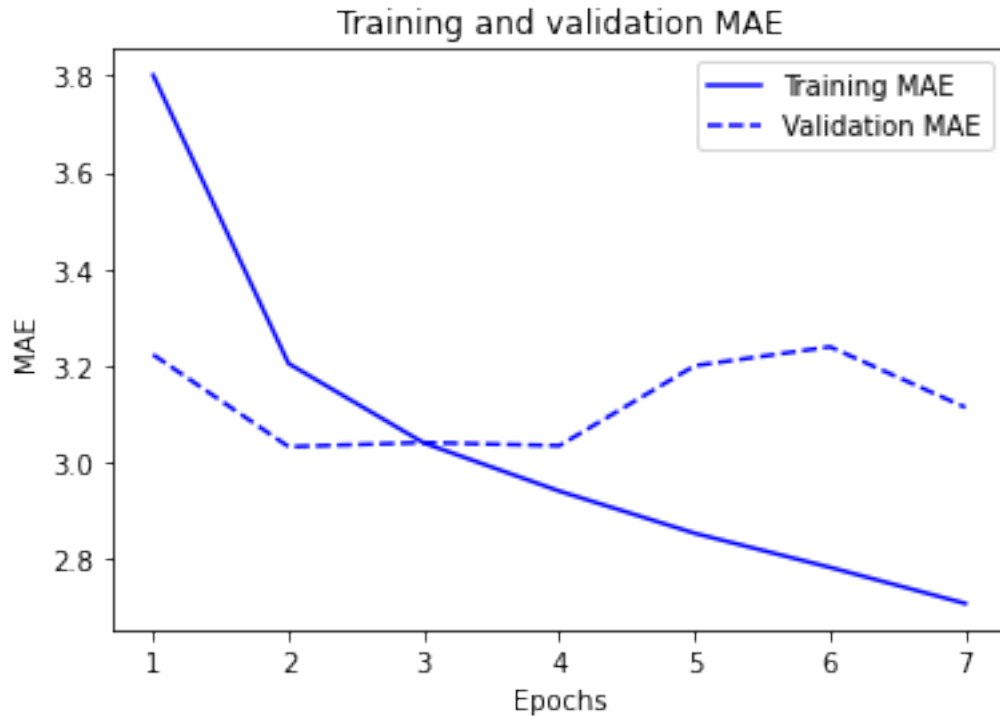
```

```

[26]: import matplotlib.pyplot as plt
      loss = history.history["mae"]
      val_loss = history.history["val_mae"]

      epochs = range(1, len(loss) + 1)
      plt.figure()
      plt.plot(epochs, loss, color="blue", linestyle="solid", label="Training MAE")
      plt.plot(epochs, val_loss, color="blue", linestyle="dashed", label="Validation_
↪MAE")
      plt.title("Training and validation MAE")
      plt.xlabel("Epochs")
      plt.ylabel("MAE")
      plt.legend()
      plt.show()

```

LSTM BASED MODEL: FIRST RECURRENT BASELINE

```
[27]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))
x = layers.LSTM(16)(inputs)
outputs = layers.Dense(1)(x)
model = keras.Model(inputs, outputs)

callbacks = [
    keras.callbacks.ModelCheckpoint("jena_lstm.keras",
                                    save_best_only=True)
]
model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
                    callbacks=callbacks)

model = keras.models.load_model("jena_lstm.keras")
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")
```

Epoch 1/7

819/819 [=====] - 68s 80ms/step - loss: 39.6437 - mae: 4.5914 - val_loss: 12.1645 - val_mae: 2.6560

Epoch 2/7

```

819/819 [=====] - 64s 78ms/step - loss: 10.9491 - mae:
2.5706 - val_loss: 9.9731 - val_mae: 2.4607
Epoch 3/7
819/819 [=====] - 62s 75ms/step - loss: 9.7183 - mae:
2.4295 - val_loss: 9.7276 - val_mae: 2.4247
Epoch 4/7
819/819 [=====] - 65s 80ms/step - loss: 9.2310 - mae:
2.3651 - val_loss: 9.7745 - val_mae: 2.4259
Epoch 5/7
819/819 [=====] - 65s 80ms/step - loss: 8.9661 - mae:
2.3311 - val_loss: 9.7748 - val_mae: 2.4305
Epoch 6/7
819/819 [=====] - 65s 80ms/step - loss: 8.6958 - mae:
2.2943 - val_loss: 9.7569 - val_mae: 2.4362
Epoch 7/7
819/819 [=====] - 65s 80ms/step - loss: 8.4829 - mae:
2.2664 - val_loss: 9.7781 - val_mae: 2.4370
405/405 [=====] - 13s 32ms/step - loss: 10.5993 - mae:
2.5655
Test MAE: 2.57

```

```
[ ]: SIMPLE RNN MODEL:
```

```

[28]: num_features = 14
inputs = keras.Input(shape=(None, num_features))
outputs = layers.SimpleRNN(16)(inputs)

model = keras.Model(inputs, outputs)

callbacks = [
    keras.callbacks.ModelCheckpoint("jena_SimRNN.keras",
                                    save_best_only=True)
]
model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
                    callbacks=callbacks)

model = keras.models.load_model("jena_SimRNN.keras")
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")

```

```

Epoch 1/7
819/819 [=====] - 43s 52ms/step - loss: 139.0193 - mae:
9.7127 - val_loss: 144.4249 - val_mae: 9.9398
Epoch 2/7
819/819 [=====] - 43s 53ms/step - loss: 136.4810 - mae:
9.5684 - val_loss: 143.7353 - val_mae: 9.8730

```

```

Epoch 3/7
819/819 [=====] - 43s 53ms/step - loss: 136.2950 - mae:
9.5525 - val_loss: 143.6657 - val_mae: 9.8670
Epoch 4/7
819/819 [=====] - 43s 52ms/step - loss: 136.2354 - mae:
9.5472 - val_loss: 143.6259 - val_mae: 9.8634
Epoch 5/7
819/819 [=====] - 43s 52ms/step - loss: 136.2116 - mae:
9.5427 - val_loss: 143.5809 - val_mae: 9.8537
Epoch 6/7
819/819 [=====] - 42s 52ms/step - loss: 136.1713 - mae:
9.5382 - val_loss: 143.5684 - val_mae: 9.8526
Epoch 7/7
819/819 [=====] - 41s 50ms/step - loss: 136.1624 - mae:
9.5378 - val_loss: 143.5494 - val_mae: 9.8509
405/405 [=====] - 10s 25ms/step - loss: 151.2899 - mae:
9.9169
Test MAE: 9.92

```

STACKING RNN LAYERS:

```

[29]: num_features = 14
      inputs = keras.Input(shape=(None, num_features))
      outputs = layers.SimpleRNN(16)(inputs)
      num_features = 14
      steps = 120
      inputs = keras.Input(shape=(steps, num_features))
      outputs = layers.SimpleRNN(16, return_sequences=False)(inputs)
      print(outputs.shape)
      num_features = 14
      steps = 120
      inputs = keras.Input(shape=(steps, num_features))
      outputs = layers.SimpleRNN(16, return_sequences=True)(inputs)
      print(outputs.shape)
      inputs = keras.Input(shape=(steps, num_features))
      x = layers.SimpleRNN(16, return_sequences=True)(inputs)
      x = layers.SimpleRNN(16, return_sequences=True)(x)
      outputs = layers.SimpleRNN(16)(x)

```

```

(None, 16)
(None, 120, 16)

```

LSTM - SIMPLE

```

[15]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))
      x = layers.LSTM(16)(inputs)
      outputs = layers.Dense(1)(x)
      model = keras.Model(inputs, outputs)

```

```

callbacks = [
    keras.callbacks.ModelCheckpoint("jena_lstm.keras",
                                    save_best_only=True)
]
model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
                    callbacks=callbacks)

model = keras.models.load_model("jena_lstm.keras")
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")

```

```

Epoch 1/7
819/819 [=====] - 56s 66ms/step - loss: 47.0226 - mae:
5.0047 - val_loss: 13.5562 - val_mae: 2.7893
Epoch 2/7
819/819 [=====] - 53s 65ms/step - loss: 11.0616 - mae:
2.5803 - val_loss: 9.7256 - val_mae: 2.4234
Epoch 3/7
819/819 [=====] - 54s 65ms/step - loss: 9.4351 - mae:
2.3981 - val_loss: 9.9041 - val_mae: 2.4479
Epoch 4/7
819/819 [=====] - 51s 63ms/step - loss: 8.8802 - mae:
2.3278 - val_loss: 9.9444 - val_mae: 2.4627
Epoch 5/7
819/819 [=====] - 51s 62ms/step - loss: 8.4791 - mae:
2.2742 - val_loss: 9.9536 - val_mae: 2.4617
Epoch 6/7
819/819 [=====] - 52s 63ms/step - loss: 8.2030 - mae:
2.2396 - val_loss: 10.0402 - val_mae: 2.4709
Epoch 7/7
819/819 [=====] - 51s 62ms/step - loss: 7.9682 - mae:
2.2083 - val_loss: 10.0586 - val_mae: 2.469868 - mae:
405/405 [=====] - 11s 27ms/step - loss: 10.9719 - mae:
2.5921
Test MAE: 2.59

```

PLOT THE LSTM:

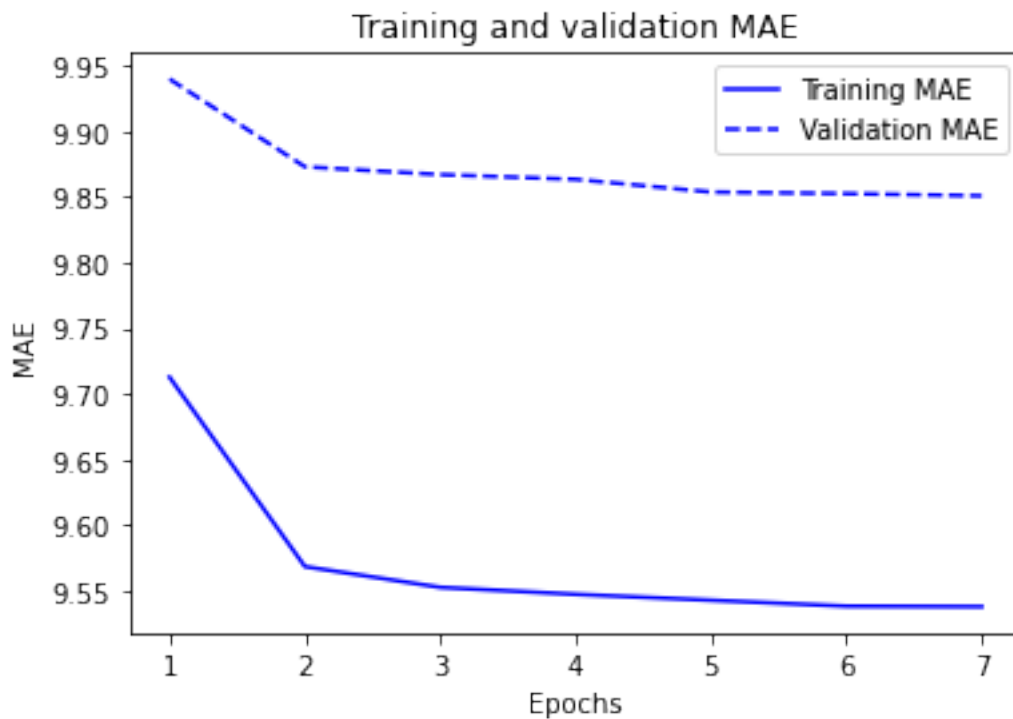
```

[30]: import matplotlib.pyplot as plt
      loss = history.history["mae"]
      val_loss = history.history["val_mae"]

      epochs = range(1, len(loss) + 1)
      plt.figure()
      plt.plot(epochs, loss, color="blue", linestyle="solid", label="Training MAE")

```

```
plt.plot(epochs, val_loss, color="blue", linestyle="dashed",label="Validation_↵
↵MAE")
plt.title("Training and validation MAE")
plt.xlabel("Epochs")
plt.ylabel("MAE")
plt.legend()
plt.show()
```



LSTM- DROPOUT REGULARIZATION

```
[ ]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))
x = layers.LSTM(16, recurrent_dropout=0.25)(inputs)
x = layers.Dropout(0.5)(x)
outputs = layers.Dense(1)(x)
model = keras.Model(inputs, outputs)

callbacks = [
    keras.callbacks.ModelCheckpoint("jena_lstm_dropout.keras",
                                    save_best_only=True)
]
model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
```

```
callbacks=callbacks)
```

```
model = keras.models.load_model("jena_lstm_dropout.keras")  
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")
```

Epoch 1/7

819/819 [=====] - 146s 176ms/step - loss: 45.1965 -
mae: 5.0064 - val_loss: 13.0267 - val_mae: 2.7319

Epoch 2/7

730/819 [=====>...] - ETA: 14s - loss: 19.9908 - mae:
3.4306

```
[ ]: import matplotlib.pyplot as plt  
loss = history.history["mae"]  
val_loss = history.history["val_mae"]  
  
epochs = range(1, len(loss) + 1)  
plt.figure()  
plt.plot(epochs, loss, color="black", linestyle="dashdot", label="Training MAE")  
plt.plot(epochs, val_loss, color="blue", linestyle="dashdot", label="Validation_↵  
↵MAE")  
plt.title("Training and validation MAE")  
plt.xlabel("Epochs")  
plt.ylabel("MAE")  
plt.legend()  
plt.show()
```

LSTM- 8 LAYERS

```
[16]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))  
x = layers.LSTM(8, return_sequences=True)(inputs)  
x = layers.LSTM(8)(x)  
outputs = layers.Dense(1)(x)  
model = keras.Model(inputs, outputs)  
  
callbacks = [  
    keras.callbacks.ModelCheckpoint("jena_LSTM_stacked1.keras",  
                                    save_best_only=True)  
]  
model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])  
history = model.fit(train_dataset,  
                    epochs=7,  
                    validation_data=val_dataset,  
                    callbacks=callbacks)  
model = keras.models.load_model("jena_LSTM_stacked1.keras")  
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")
```

Epoch 1/7

```

819/819 [=====] - 73s 86ms/step - loss: 66.5505 - mae:
6.1977 - val_loss: 32.5827 - val_mae: 4.2437
Epoch 2/7
819/819 [=====] - 73s 89ms/step - loss: 19.6884 - mae:
3.2833 - val_loss: 13.0442 - val_mae: 2.7470
Epoch 3/7
819/819 [=====] - 68s 83ms/step - loss: 10.8644 - mae:
2.5553 - val_loss: 10.7928 - val_mae: 2.5670
Epoch 4/7
819/819 [=====] - 75s 91ms/step - loss: 9.7278 - mae:
2.4324 - val_loss: 10.1817 - val_mae: 2.4860
Epoch 5/7
819/819 [=====] - 78s 95ms/step - loss: 9.4322 - mae:
2.3940 - val_loss: 9.9864 - val_mae: 2.4638
Epoch 6/7
819/819 [=====] - 82s 100ms/step - loss: 9.2045 - mae:
2.3646 - val_loss: 10.0388 - val_mae: 2.4600
Epoch 7/7
819/819 [=====] - 74s 90ms/step - loss: 9.0564 - mae:
2.3464 - val_loss: 9.8563 - val_mae: 2.4532
405/405 [=====] - 10s 24ms/step - loss: 10.8690 - mae:
2.5984
Test MAE: 2.60

```

```

[ ]: import matplotlib.pyplot as plt
loss = history.history["mae"]
val_loss = history.history["val_mae"]

epochs = range(1, len(loss) + 1)
plt.figure()
plt.plot(epochs, loss, color="blue", linestyle="solid", label="Training MAE")
plt.plot(epochs, val_loss, color="blue", linestyle="dashed", label="Validation_
↪MAE")
plt.title("Training and validation MAE")
plt.xlabel("Epochs")
plt.ylabel("MAE")
plt.legend()
plt.show()

```

LSTM - 16 LAYERS

```

[17]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))
x = layers.LSTM(16, return_sequences=True)(inputs)
x = layers.LSTM(16)(x)
outputs = layers.Dense(1)(x)
model = keras.Model(inputs, outputs)

callbacks = [

```

```

        keras.callbacks.ModelCheckpoint("jena_LSTM_stacked1.keras",
                                        save_best_only=True)
]
model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
                    callbacks=callbacks)
model = keras.models.load_model("jena_LSTM_stacked1.keras")
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")

```

```

Epoch 1/7
819/819 [=====] - 95s 112ms/step - loss: 40.4396 - mae:
4.5995 - val_loss: 13.0213 - val_mae: 2.7410
Epoch 2/7
819/819 [=====] - 93s 113ms/step - loss: 10.1922 - mae:
2.4755 - val_loss: 9.5310 - val_mae: 2.3967
Epoch 3/7
819/819 [=====] - 93s 113ms/step - loss: 8.8859 - mae:
2.3198 - val_loss: 9.4881 - val_mae: 2.4008
Epoch 4/7
819/819 [=====] - 93s 114ms/step - loss: 8.3576 - mae:
2.2506 - val_loss: 10.1544 - val_mae: 2.4799
Epoch 5/7
819/819 [=====] - 92s 112ms/step - loss: 7.9404 - mae:
2.1958 - val_loss: 10.2832 - val_mae: 2.5040
Epoch 6/7
819/819 [=====] - 85s 104ms/step - loss: 7.6751 - mae:
2.1572 - val_loss: 10.0266 - val_mae: 2.4688
Epoch 7/7
819/819 [=====] - 91s 111ms/step - loss: 7.5017 - mae:
2.1343 - val_loss: 9.8490 - val_mae: 2.4500
405/405 [=====] - 15s 36ms/step - loss: 11.5874 - mae:
2.6785
Test MAE: 2.68

```

```

[ ]: import matplotlib.pyplot as plt
loss = history.history["mae"]
val_loss = history.history["val_mae"]

epochs = range(1, len(loss) + 1)
plt.figure()
plt.plot(epochs, loss, color="blue", linestyle="solid", label="Training MAE")
plt.plot(epochs, val_loss, color="blue", linestyle="dashed", label="Validation_
→MAE")
plt.title("Training and validation MAE")
plt.xlabel("Epochs")

```



```
plt.ylabel("MAE")
plt.legend()
plt.show()
```

LSTM - 32 LAYERS

```
[18]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))
x = layers.LSTM(32, return_sequences=True)(inputs)
x = layers.LSTM(32)(x)
outputs = layers.Dense(1)(x)
model = keras.Model(inputs, outputs)

callbacks = [
    keras.callbacks.ModelCheckpoint("jena_LSTM_stacked2.keras",
                                    save_best_only=True)
]
model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
                    callbacks=callbacks)
model = keras.models.load_model("jena_LSTM_stacked2.keras")
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")
```

Epoch 1/7

819/819 [=====] - 123s 147ms/step - loss: 19.0733 -
mae: 3.1502 - val_loss: 10.5564 - val_mae: 2.5522

Epoch 2/7

819/819 [=====] - 121s 148ms/step - loss: 7.9323 - mae:
2.2019 - val_loss: 10.7405 - val_mae: 2.5648

Epoch 3/7

819/819 [=====] - 120s 146ms/step - loss: 6.3938 - mae:
1.9750 - val_loss: 11.7384 - val_mae: 2.6851

Epoch 4/7

819/819 [=====] - 116s 141ms/step - loss: 5.2708 - mae:
1.7918 - val_loss: 12.1749 - val_mae: 2.7385

Epoch 5/7

819/819 [=====] - 118s 144ms/step - loss: 4.3911 - mae:
1.6336 - val_loss: 13.1748 - val_mae: 2.8172

Epoch 6/7

819/819 [=====] - 119s 145ms/step - loss: 3.7401 - mae:
1.5048 - val_loss: 14.1781 - val_mae: 2.9448

Epoch 7/7

819/819 [=====] - 121s 147ms/step - loss: 3.2272 - mae:
1.3953 - val_loss: 14.2355 - val_mae: 2.9600

405/405 [=====] - 20s 47ms/step - loss: 11.2179 - mae:
2.6228

Test MAE: 2.62

LSTM- DROUPOUT, REGULARISED AND STACK MODEL

```
[16]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))
x = layers.LSTM(8, recurrent_dropout=0.5, return_sequences=True)(inputs)
x = layers.LSTM(8, recurrent_dropout=0.5)(x)
x = layers.Dropout(0.5)(x)
outputs = layers.Dense(1)(x)
model = keras.Model(inputs, outputs)

callbacks = [
    keras.callbacks.ModelCheckpoint("jena_stacked_LSTM_dropout.keras",
                                    save_best_only=True)
]
model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
                    callbacks=callbacks)
model = keras.models.load_model("jena_stacked_LSTM_dropout.keras")
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")
```

Epoch 1/7

819/819 [=====] - 232s 279ms/step - loss: 68.9645 -
mae: 6.3491 - val_loss: 31.1883 - val_mae: 4.1217

Epoch 2/7

819/819 [=====] - 231s 282ms/step - loss: 30.4259 -
mae: 4.1343 - val_loss: 13.4328 - val_mae: 2.7466

Epoch 3/7

819/819 [=====] - 230s 281ms/step - loss: 24.3147 -
mae: 3.7318 - val_loss: 11.7295 - val_mae: 2.6444

Epoch 4/7

819/819 [=====] - 229s 279ms/step - loss: 22.4774 -
mae: 3.5878 - val_loss: 10.4721 - val_mae: 2.4977

Epoch 5/7

819/819 [=====] - 229s 280ms/step - loss: 21.1514 -
mae: 3.4877 - val_loss: 10.0972 - val_mae: 2.4605

Epoch 6/7

819/819 [=====] - 229s 279ms/step - loss: 20.2269 -
mae: 3.4134 - val_loss: 10.0748 - val_mae: 2.4606

Epoch 7/7

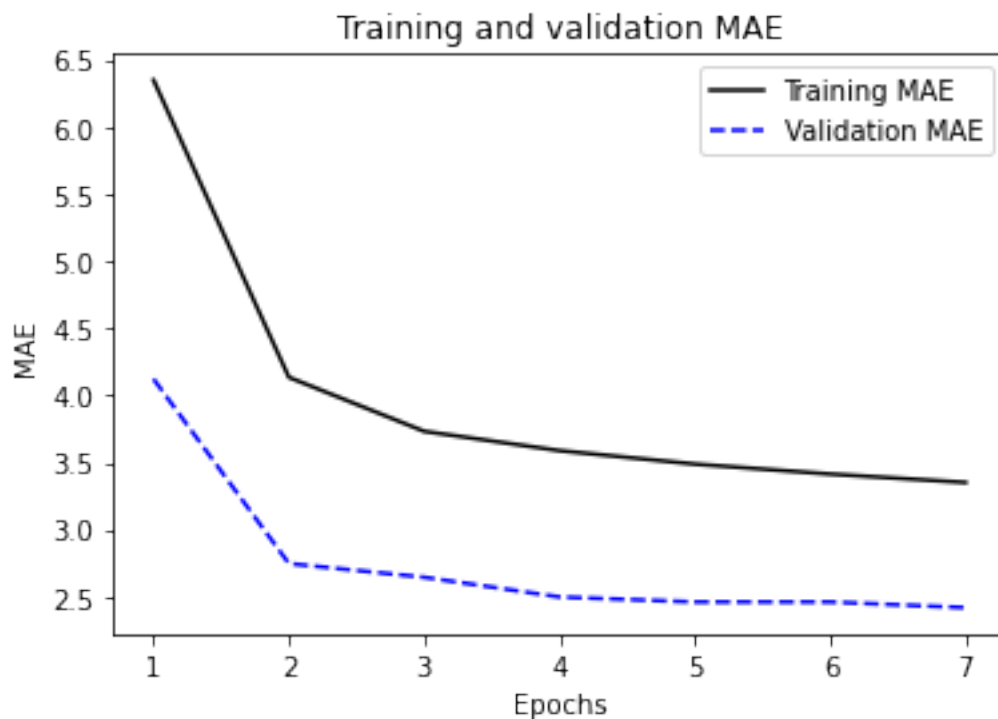
819/819 [=====] - 224s 274ms/step - loss: 19.3759 -
mae: 3.3508 - val_loss: 9.7040 - val_mae: 2.4182

405/405 [=====] - 20s 48ms/step - loss: 11.1056 - mae:
2.5970

Test MAE: 2.60

```
[19]: import matplotlib.pyplot as plt
loss = history.history["mae"]
val_loss = history.history["val_mae"]

epochs = range(1, len(loss) + 1)
plt.figure()
plt.plot(epochs, loss, color="black", linestyle="solid", label="Training MAE")
plt.plot(epochs, val_loss, color="blue", linestyle="dashed", label="Validation_
    MAE")
plt.title("Training and validation MAE")
plt.xlabel("Epochs")
plt.ylabel("MAE")
plt.legend()
plt.show()
```



BIDIRECTIONAL LSTM

```
[21]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))
x = layers.Bidirectional(layers.LSTM(16))(inputs)
outputs = layers.Dense(1)(x)
model = keras.Model(inputs, outputs)

callbacks = [
    keras.callbacks.ModelCheckpoint("jena_bidirec_LSTM.keras",
```

```

save_best_only=True)

]

model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
                    callbacks=callbacks)

model = keras.models.load_model("jena_bidirec_LSTM.keras")
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")

```

```

Epoch 1/7
819/819 [=====] - 115s 137ms/step - loss: 26.0296 -
mae: 3.6744 - val_loss: 10.1256 - val_mae: 2.4739
Epoch 2/7
819/819 [=====] - 115s 140ms/step - loss: 9.6192 - mae:
2.4264 - val_loss: 10.0175 - val_mae: 2.4432
Epoch 3/7
819/819 [=====] - 115s 140ms/step - loss: 8.4948 - mae:
2.2761 - val_loss: 9.7029 - val_mae: 2.4051
Epoch 4/7
819/819 [=====] - 115s 140ms/step - loss: 7.8801 - mae:
2.1936 - val_loss: 9.9285 - val_mae: 2.4333
Epoch 5/7
819/819 [=====] - 115s 141ms/step - loss: 7.5119 - mae:
2.1421 - val_loss: 10.0455 - val_mae: 2.4487
Epoch 6/7
819/819 [=====] - 115s 140ms/step - loss: 7.2317 - mae:
2.1007 - val_loss: 10.0826 - val_mae: 2.4536
Epoch 7/7
819/819 [=====] - 115s 140ms/step - loss: 6.9835 - mae:
2.0610 - val_loss: 10.4525 - val_mae: 2.4956
405/405 [=====] - 19s 46ms/step - loss: 11.4603 - mae:
2.6055
Test MAE: 2.61

```

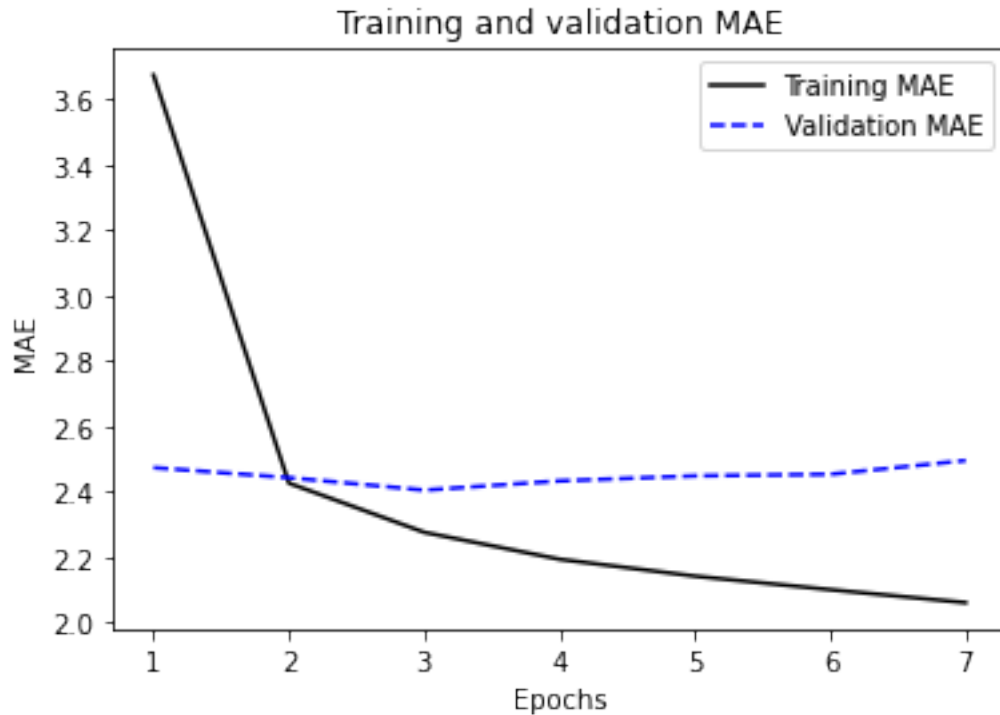
```

[22]: import matplotlib.pyplot as plt
loss = history.history["mae"]
val_loss = history.history["val_mae"]

epochs = range(1, len(loss) + 1)
plt.figure()
plt.plot(epochs, loss, color="black", linestyle="solid", label="Training MAE")
plt.plot(epochs, val_loss, color="blue", linestyle="dashed", label="Validation_
↪MAE")
plt.title("Training and validation MAE")

```

```
plt.xlabel("Epochs")
plt.ylabel("MAE")
plt.legend()
plt.show()
```



COMBINE 1 D CONVNET AND LSTM:

```
[23]: inputs = keras.Input(shape=(sequence_length, raw_data.shape[-1]))
x = layers.Bidirectional(layers.LSTM(16))(inputs)
outputs = layers.Dense(1)(x)
model = keras.Model(inputs, outputs)

callbacks = [
    keras.callbacks.ModelCheckpoint("jena_bidirec_LSTM.keras",
                                    save_best_only=True)
]

model.compile(optimizer="rmsprop", loss="mse", metrics=["mae"])
history = model.fit(train_dataset,
                    epochs=7,
                    validation_data=val_dataset,
                    callbacks=callbacks)

model = keras.models.load_model("jena_bidirec_LSTM.keras")
```

```
print(f"Test MAE: {model.evaluate(test_dataset)[1]:.2f}")
```

Epoch 1/7

819/819 [=====] - 118s 141ms/step - loss: 26.2874 - mae: 3.6911 - val_loss: 10.5428 - val_mae: 2.5172

Epoch 2/7

819/819 [=====] - 115s 141ms/step - loss: 9.5115 - mae: 2.4100 - val_loss: 9.9082 - val_mae: 2.4252

Epoch 3/7

819/819 [=====] - 115s 141ms/step - loss: 8.5254 - mae: 2.2753 - val_loss: 9.9273 - val_mae: 2.4359

Epoch 4/7

819/819 [=====] - 115s 140ms/step - loss: 7.9265 - mae: 2.1944 - val_loss: 10.3498 - val_mae: 2.4872

Epoch 5/7

819/819 [=====] - 115s 141ms/step - loss: 7.5152 - mae: 2.1347 - val_loss: 10.8683 - val_mae: 2.5475

Epoch 6/7

819/819 [=====] - 115s 140ms/step - loss: 7.1626 - mae: 2.0819 - val_loss: 11.0660 - val_mae: 2.5724

Epoch 7/7

819/819 [=====] - 115s 140ms/step - loss: 6.8987 - mae: 2.0417 - val_loss: 10.5593 - val_mae: 2.5235

405/405 [=====] - 19s 46ms/step - loss: 10.9262 - mae: 2.61541s

Test MAE: 2.62

```
[24]: import matplotlib.pyplot as plt
loss = history.history["mae"]
val_loss = history.history["val_mae"]

epochs = range(1, len(loss) + 1)
plt.figure()
plt.plot(epochs, loss, color="black", linestyle="solid", label="Training MAE")
plt.plot(epochs, val_loss, color="blue", linestyle="dashed", label="Validation_
↪MAE")
plt.title("Training and validation MAE")
plt.xlabel("Epochs")
plt.ylabel("MAE")
plt.legend()
plt.show()
```

