PS20

January 26, 2024

1 Mini-project #2 (RNN exercise)

- 1.1 Task: Weather prediction
- 1.2 Loading Jena climate dataset

```
[1]:
    import pandas as pd
[2]: data = pd.read_csv('jena_climate_2009_2016.csv')
     data.describe()
[2]:
                  p (mbar)
                                  T (degC)
                                                  Tpot (K)
                                                               Tdew (degC)
     count
            420551.000000
                             420551.000000
                                             420551.000000
                                                             420551.000000
     mean
                989.212776
                                  9.450147
                                                283.492743
                                                                   4.955854
     std
                  8.358481
                                  8.423365
                                                  8.504471
                                                                   6.730674
                913.600000
                                -23.010000
                                                250.600000
                                                                -25.010000
     min
     25%
                984.200000
                                  3.360000
                                                277.430000
                                                                   0.240000
                989.580000
     50%
                                                283.470000
                                  9.420000
                                                                   5.220000
     75%
                994.720000
                                 15.470000
                                                289.530000
                                                                  10.070000
               1015.350000
                                 37.280000
                                                311.340000
                                                                  23.110000
     max
                    rh (%)
                              VPmax (mbar)
                                              VPact (mbar)
                                                              VPdef (mbar)
            420551.000000
                             420551.000000
                                             420551.000000
                                                             420551.000000
     count
     mean
                 76.008259
                                 13.576251
                                                  9.533756
                                                                   4.042412
                 16.476175
                                  7.739020
                                                                   4.896851
     std
                                                  4.184164
                                                                   0.00000
     min
                 12.950000
                                  0.950000
                                                  0.790000
     25%
                 65.210000
                                  7.780000
                                                  6.210000
                                                                   0.870000
     50%
                 79.300000
                                 11.820000
                                                                   2.190000
                                                  8.860000
     75%
                 89.400000
                                 17.600000
                                                 12.350000
                                                                   5.300000
     max
                100.000000
                                 63.770000
                                                 28.320000
                                                                 46.010000
                 sh (g/kg)
                             H2OC (mmol/mol)
                                                rho (g/m**3)
                                                                     wv (m/s)
            420551.000000
                               420551.000000
                                               420551.000000
                                                                420551.000000
     count
                  6.022408
                                    9.640223
                                                 1216.062748
                                                                     1.702224
     mean
     std
                  2.656139
                                    4.235395
                                                    39.975208
                                                                    65.446714
                                                 1059.450000
                                                                -9999.000000
     min
                  0.500000
                                    0.800000
     25%
                  3.920000
                                    6.290000
                                                 1187.490000
                                                                     0.990000
     50%
                  5.590000
                                    8.960000
                                                 1213.790000
                                                                     1.760000
```

75%	7.800000	12.490000	1242.770000	2.860000
max	18.130000	28.820000	1393.540000	28.490000
	max. wv (m/s)	wd (deg)		
count	420551.000000	420551.000000		
mean	3.056555	174.743738		
std	69.016932	86.681693		
min	-9999.000000	0.000000		
25%	1.760000	124.900000		
50%	2.960000	198.100000		
75%	4.740000	234.100000		
max	23.500000	360.000000		

1.3 Data preprocessing

Wind speed (and maximum wind speed) is set to -9999.00 for missing entries. Let us fill up the missing entries with the mean.

```
[3]: wv = data['wv (m/s)']
    wv_missing_idx = (wv == -9999.00)
    wv_mean = wv[~wv_missing_idx].mean()
    wv[wv_missing_idx] = wv_mean
```

C:\Users\chsuh\AppData\Local\Temp\ipykernel_18376\832945953.py:4:
SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy wv[wv_missing_idx] = wv_mean

```
[4]: max_wv = data['max. wv (m/s)']
missing_idx = (max_wv == -9999.00)
max_wv_mean = max_wv[~missing_idx].mean()
max_wv[missing_idx] = max_wv_mean
```

C:\Users\chsuh\AppData\Local\Temp\ipykernel_18376\1423667913.py:4:
SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy max_wv[missing_idx] = max_wv_mean

1.4 Remove date_time column

```
[5]: data.pop('Date Time')
     data
[5]:
                        T (degC)
                                   Tpot (K)
                                              Tdew (degC)
                                                             rh (%)
                                                                     VPmax (mbar) \
              p (mbar)
                996.52
                            -8.02
                                      265.40
                                                     -8.90
                                                              93.30
                                                                              3.33
     0
     1
                996.57
                            -8.41
                                      265.01
                                                     -9.28
                                                              93.40
                                                                              3.23
     2
                996.53
                            -8.51
                                      264.91
                                                     -9.31
                                                                              3.21
                                                              93.90
     3
                996.51
                            -8.31
                                      265.12
                                                     -9.07
                                                              94.20
                                                                              3.26
     4
                996.51
                            -8.27
                                                     -9.04
                                                                              3.27
                                      265.15
                                                              94.10
     420546
               1000.07
                            -4.05
                                      269.10
                                                     -8.13
                                                              73.10
                                                                              4.52
                                      269.81
                                                                              4.77
     420547
                999.93
                            -3.35
                                                     -8.06
                                                              69.71
                                                                              4.84
     420548
                999.82
                            -3.16
                                      270.01
                                                     -8.21
                                                              67.91
     420549
                999.81
                            -4.23
                                      268.94
                                                     -8.53
                                                              71.80
                                                                              4.46
     420550
                999.82
                            -4.82
                                      268.36
                                                     -8.42
                                                              75.70
                                                                              4.27
              VPact (mbar)
                             VPdef (mbar)
                                                        H2OC (mmol/mol)
                                                                           rho (g/m**3)
                                            sh (g/kg)
     0
                      3.11
                                      0.22
                                                  1.94
                                                                    3.12
                                                                                 1307.75
     1
                      3.02
                                      0.21
                                                  1.89
                                                                    3.03
                                                                                 1309.80
     2
                      3.01
                                      0.20
                                                  1.88
                                                                    3.02
                                                                                 1310.24
     3
                      3.07
                                      0.19
                                                  1.92
                                                                    3.08
                                                                                 1309.19
     4
                      3.08
                                      0.19
                                                  1.92
                                                                    3.09
                                                                                 1309.00
     420546
                      3.30
                                      1.22
                                                  2.06
                                                                    3.30
                                                                                 1292.98
     420547
                      3.32
                                      1.44
                                                  2.07
                                                                    3.32
                                                                                 1289.44
     420548
                      3.28
                                      1.55
                                                  2.05
                                                                    3.28
                                                                                 1288.39
                                      1.26
                                                                    3.20
     420549
                      3.20
                                                  1.99
                                                                                 1293.56
     420550
                      3.23
                                      1.04
                                                  2.01
                                                                    3.23
                                                                                 1296.38
              wv (m/s)
                        max. wv (m/s)
                                         wd (deg)
     0
                  1.03
                                   1.75
                                            152.3
     1
                  0.72
                                   1.50
                                            136.1
     2
                  0.19
                                  0.63
                                            171.6
     3
                  0.34
                                  0.50
                                            198.0
     4
                  0.32
                                   0.63
                                            214.3
     420546
                  0.67
                                   1.52
                                            240.0
     420547
                  1.14
                                   1.92
                                            234.3
                  1.08
                                  2.00
                                            215.2
     420548
     420549
                  1.49
                                  2.16
                                            225.8
     420550
                  1.23
                                   1.96
                                            184.9
```

[420551 rows x 14 columns]

1.5 Downsampling

We sample every 60 minutes to contruct a dataset for training.

```
[6]: data = data[0::6] #data = data[::6] data
```

	uata									
[6]:		p (mbar)	T (degC)	Tpot (K)) Tdew (de	egC)	rh (%)	VPmax (m	bar) \	
	0	996.52	-8.02	265.40	3- C	3.90	93.30		3.33	
	6	996.50	-7.62	265.8	1 -8	3.30	94.80		3.44	
	12	996.63	-8.85	264.5	7 –9	70	93.50		3.12	
	18	996.87	-8.84	264.5	6 -9	.69	93.50		3.13	
	24	997.05	-9.23	264.1	5 -10	.25	92.20		3.03	
							•••			
	420522	1002.08	-1.40	271.59	9 -6	3.10	70.20		5.51	
	420528	1001.42	-2.15	270.90	O -7	7.08	68.77		5.21	
	420534	1001.05	-2.61	270.4	7 –6	5.97	71.80	,	5.04	
	420540	1000.51	-3.22	269.90		7.63	71.40	•	4.81	
	420546	1000.07	-4.05	269.10	3 – C	3.13	73.10	•	4.52	
					- ((-)			- > -		
	_				sh (g/kg)	H20C			_	\
	0		11	0.22	1.94			3.12	1307.75	
	6	3.		0.18	2.04			3.27	1305.68	
	12		92	0.20	1.82			2.93	1312.11	
	18		92	0.20	1.83			2.93	1312.37	
	24	2.	79	0.24	1.74			2.80	1314.62	
	 400E00	 2		1 61					1000 60	
	420522 420528		87 59	1.64 1.63	2.40 2.23			3.86 3.58	1282.68 1285.50	
	420526	3.		1.42	2.25			3.61	1287.20	
	420540	3.		1.42	2.23			3.44	1289.50	
	420546	3.		1.22	2.14			3.30	1292.98	
	120010	ο.	00	1.22	2.00				1232.30	
		wv (m/s)	max. wv (n	n/s) wd	(deg)					
	0	1.03		1.75	152.3					
	6	0.18	(0.63	166.5					
	12	0.16	(0.50	158.3					
	18	0.07	(0.25	129.3					
	24	0.10	(0.38	203.9					
			•••							
	420522	1.08	-	1.68	207.5					
	420528	0.79	-	1.24	184.3					
	420534	0.77	-	1.64	129.1					
	420540	0.85	-	1.54	207.8					
	420546	0.67	-	1.52	240.0					

[70092 rows x 14 columns]

1.6 Take temperature in celsius as label while taking everything as features

```
[7]: features = data
      labels = data[['T (degC)']]
      #labels = data['T (degC)']
 [8]: print(features.shape)
      print(labels.shape)
      print(type(features))
      print(type(labels))
     (70092, 14)
     (70092, 1)
     <class 'pandas.core.frame.DataFrame'>
     <class 'pandas.core.frame.DataFrame'>
     1.7 Normalization
 [9]: from sklearn.preprocessing import StandardScaler
      std_scaler = StandardScaler()
      features = std_scaler.fit_transform(features)
[10]: print(type(features))
      print(features)
      print(features.shape)
     <class 'numpy.ndarray'>
     [[ 0.87420457 -2.07391772 -2.12735513 ... -0.71190538 -0.76237653
       -0.2618485 ]
      [0.87181184 -2.02643323 -2.07914744 ... -1.26284569 -1.24217323
       -0.09825609]
      [0.8873646 -2.17244806 -2.2249463 ... -1.27580899 -1.29786392]
       -0.19272494]
      [ 1.41615816 -1.43168989 -1.53122591 ... -0.8804283 -0.80949942
       -0.52912624]
      [ 1.35155442 -1.50410375 -1.59824636 ... -0.82857509 -0.85233841
        0.37754438]
      [ 1.29891433 -1.60263408 -1.69231014 ... -0.9452448 -0.86090621
        0.74850745]]
     (70092, 14)
     1.8 Data split (train:val:test=7:2:1)
     Let us split chronologically, not randomly.
```

[11]: from sklearn.model_selection import train_test_split

```
[12]: print(X_train.shape)
print(X_val.shape)
print(X_test.shape)
(49063, 14)
```

(14019, 14) (7010, 14)

1.9 Set T_{window} and batch size

```
[13]: T = 24
batch_size = 16
```

1.10 Construct $\{(x_T^{(i)}, y_T^{(i)})\}_{i=1}^{m_T}$

```
[14]: from tensorflow.keras.preprocessing import timeseries_dataset_from_array
      # Train batch dataset
      dataset_train = timeseries_dataset_from_array(X_train[:-T],
                                                     y_train[T:],
                                                     sequence_length = T,
                                                     sequence_stride = 1,
                                                     batch_size = batch_size,
                                                     shuffle = True)
      # validation batch dataset
      dataset_val = timeseries_dataset_from_array(X_val[:-T],
                                                   y_val[T:],
                                                   sequence_length = T,
                                                   sequence_stride = 1,
                                                   batch_size=batch_size,
                                                   shuffle = False)
      # test batch dataset
      dataset_test = timeseries_dataset_from_array(X_test[:-T],
                                                    y_test[T:],
                                                    sequence_length = T,
                                                    sequence_stride = 1,
                                                    batch_size=batch_size,
```

```
shuffle = False)
[15]: print(type(dataset_train))
     <class 'tensorflow.python.data.ops.dataset_ops.BatchDataset'>
[16]: print(dataset_train.take)
     <bound method DatasetV2.take of <BatchDataset</pre>
     element_spec=(TensorSpec(shape=(None, None, 14), dtype=tf.float64, name=None),
     TensorSpec(shape=(None, 1), dtype=tf.float64, name=None))>>
[17]: print(len(dataset_train))
      print(len(X_train[:-T])//batch_size)
      print(len(dataset_val))
      print(len(X_val[:-T])//batch_size)
      print(len(dataset_test))
      print(len(X_test[:-T])//batch_size)
     3064
     3064
     874
     874
     436
     436
[18]: print(dataset_train.take)
      print(len(dataset_train))
      print(len(dataset_val))
      print(len(dataset_test))
     <bound method DatasetV2.take of <BatchDataset</pre>
     element_spec=(TensorSpec(shape=(None, None, 14), dtype=tf.float64, name=None),
     TensorSpec(shape=(None, 1), dtype=tf.float64, name=None))>>
     3064
     874
     436
[19]: for batch in dataset_train.take(5):
          inputs, labels = batch
          print(inputs.shape)
          print(labels.shape)
     (16, 24, 14)
     (16, 1)
     (16, 24, 14)
     (16, 1)
     (16, 24, 14)
     (16, 1)
     (16, 24, 14)
```

```
(16, 1)
     (16, 24, 14)
     (16, 1)
[20]: print(type(inputs))
     print(inputs)
    <class 'tensorflow.python.framework.ops.EagerTensor'>
    tf.Tensor(
     [[[-0.45256479 -0.261197 -0.22373935 ... -1.21099248 -1.28929612
       -1.60226942]
      [-0.4035138 \quad -0.34785621 \quad -0.31427574 \quad ... \quad -1.20451083 \quad -1.30643172
       -1.00607877
       [-0.37121193 -0.10568527 -0.07558889 ... -1.01654296 -1.01512658
       -1.42277859
      [-0.18816801 \quad 0.45107047 \quad 0.46292626 \dots \quad -0.89987325 \quad -0.98085539
       -1.87507705]
      [-0.17381162 \quad 0.4154571 \quad 0.42647654 \dots -0.69894208 \quad -0.86090621
        2.10217706]
       [-0.1295461 0.2825005 0.29008406 ... -0.24522653 -0.50105869
       -1.74523997]]
       \begin{bmatrix} 0.0857997 & -0.12824041 & -0.13437876 & \dots & -0.06374031 & 0.08155159 \end{bmatrix} 
       -0.10862462]
      [ 0.07503241  0.0154002  0.00906851 ... 1.31685129  1.16537806
        0.35219907]
       0.40288968]
      [-0.42265565 0.01183886 0.04551823 ... -0.26467148 -0.48392309
        0.18860666]
      0.21510403
      0.31648524]]
      [[-0.25037902 -0.18997026 -0.16965267 ... 0.31219543 0.86550513
        0.196671087
      [-0.20970259 \ -0.19471871 \ -0.17670746 \ ... \ 1.26499808 \ 1.26819164
        0.30035641]
      [-0.17261526 -0.17809913 -0.16377369 ... 3.50764922 3.56436156
        0.35680731]
      0.22547256]
      [ 0.32387643 -0.31580418 -0.34014328 ... 1.21962653 1.07113228
        0.3372223 ]
```

```
0.2807714 ]]
[[-0.11758244 0.70273831 0.70749209 ... -0.9452448 -1.03226218
  0.29690023]
[-0.10322606 0.6730605 0.67692136 ... -1.12673102 -1.27216052
  0.79113364]
0.23699315]
[ \ 0.19466896 \ \ 0.70748676 \ \ 0.68632774 \ ... \ -0.71838703 \ -0.73238924
  1.0042646 ]
 [ 0.2520945
           -1.77208295
 \hbox{ [ 0.21620354 \ 0.58996263 \ 0.56757222 \dots -1.02302461 \ -1.08366896 ] } 
 -1.4423636 ]]
[[-0.05537144 1.39838621 1.3929819 ... 0.68164952 0.80124664
  0.039991027
[-0.08408421 1.34496615 1.34242262 ... 0.92795225 0.64702627
  0.157501067
[-0.0948515
           1.20607399 1.20603013 ... 0.16959912 -0.14121116
  0.370632021
-1.46920658]
 0.45358029]
[-0.79472533 \quad 1.67973186 \quad 1.7363147 \quad ... \quad -0.05725866 \quad -0.02126199
 -0.8920249 ]]
[[-0.42624475 -1.37945694 -1.33839516 ... 0.24737892 -0.01269419
 -0.08327932]
 [-0.48367029 -1.41625742 -1.37014169 ... -0.24522653 -0.38539341
[-0.55425586 -1.39014095 -1.33957096 ... -0.34893294 -0.59958837
  0.490446187
[-1.5520247 \quad -0.54016843 \quad -0.41304271 \quad ... \quad -0.64708887 \quad -0.81378332
  0.9109478
[-1.61303935 \ -0.55797512 \ -0.42597648 \ ... \ -1.04895121 \ -0.97657149
 -0.23765525]
 [-1.66089397 -0.61851785 -0.48241475 ... -1.10728607 -1.08366896
 -1.38326296]]], shape=(16, 24, 14), dtype=float64)
```

```
[21]: print(type(labels))
     print(labels)
     <class 'tensorflow.python.framework.ops.EagerTensor'>
     tf.Tensor(
     [[ 10.34]
      [ 12.51]
      [6.46]
      [-4.75]
      [ 6.48]
      [0.77]
      [ 11.85]
      [ 27.94]
      [ 6.15]
      [ -1.23]
      [-15.31]
      [ 10.27]
      [-0.91]
      [ 14.12]
      [ 22.16]
      [ 4.54]], shape=(16, 1), dtype=float64)
 []:
 []:
```