

# PS19

January 26, 2024

## 1 Mini-project #2 (RNN exercise)

### 1.1 Task: Weather prediction

### 1.2 Jena climate dataset

This dataset is prepared by Max Planck Institute for Biogeochemistry. It is measured in Jena, Germany, from Jan. 10 2009 to December 31 2016.

### 1.3 Loading Jena climate dataset

```
[1]: import pandas as pd
```

```
[2]: data = pd.read_csv('jena_climate_2009_2016.csv')
      data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 420551 entries, 0 to 420550
Data columns (total 15 columns):
 #   Column                Non-Null Count  Dtype  
---  -
 0   Date Time             420551 non-null object  
 1   p (mbar)              420551 non-null float64 
 2   T (degC)              420551 non-null float64 
 3   Tpot (K)              420551 non-null float64 
 4   Tdew (degC)          420551 non-null float64 
 5   rh (%)               420551 non-null float64 
 6   VPmax (mbar)         420551 non-null float64 
 7   VPact (mbar)         420551 non-null float64 
 8   VPdef (mbar)         420551 non-null float64 
 9   sh (g/kg)            420551 non-null float64 
10  H2OC (mmol/mol)      420551 non-null float64 
11  rho (g/m**3)         420551 non-null float64 
12  wv (m/s)             420551 non-null float64 
13  max. wv (m/s)        420551 non-null float64 
14  wd (deg)             420551 non-null float64 
dtypes: float64(14), object(1)
memory usage: 48.1+ MB
```

### 1.3.1 Temperatures

T (degC): Temperature in Celsius

Tpot (K): Temperature in Kelvin

### 1.3.2 Others

p (mbar): The pascal SI derived unit of pressure

Tdew (degC): Temperature in Celsius relative to humidity ( )

rh (%): Relative humidity is a measure of how saturated the air is with water vapor ( )

VPmax (mbar): Saturation vapor pressure ( )

VPact (mbar): Vapor pressure ( )

VPdef (mbar): Vapor pressure deficit

sh (g/kg): Specific humidity ( )

H2OC (mmol/mol): Water vapor concentration ( )

rho (g/m\*\*3): Airtight

wv (m/s): Wind speed

max. wv (m/s): Maximum wind speed

wd (deg): Wind direction in degrees

```
[3]: # Look up the first five examples
data.head(5)
```

```
[3]:      Date Time  p (mbar)  T (degC)  Tpot (K)  Tdew (degC)  rh (%)  \
0  01.01.2009 00:10:00   996.52    -8.02   265.40     -8.90   93.3
1  01.01.2009 00:20:00   996.57    -8.41   265.01     -9.28   93.4
2  01.01.2009 00:30:00   996.53    -8.51   264.91     -9.31   93.9
3  01.01.2009 00:40:00   996.51    -8.31   265.12     -9.07   94.2
4  01.01.2009 00:50:00   996.51    -8.27   265.15     -9.04   94.1

      VPmax (mbar)  VPact (mbar)  VPdef (mbar)  sh (g/kg)  H2OC (mmol/mol)  \
0           3.33         3.11         0.22         1.94           3.12
1           3.23         3.02         0.21         1.89           3.03
2           3.21         3.01         0.20         1.88           3.02
3           3.26         3.07         0.19         1.92           3.08
4           3.27         3.08         0.19         1.92           3.09

      rho (g/m**3)  wv (m/s)  max. wv (m/s)  wd (deg)
0          1307.75       1.03         1.75     152.3
1          1309.80       0.72         1.50     136.1
2          1310.24       0.19         0.63     171.6
3          1309.19       0.34         0.50     198.0
4          1309.00       0.32         0.63     214.3
```

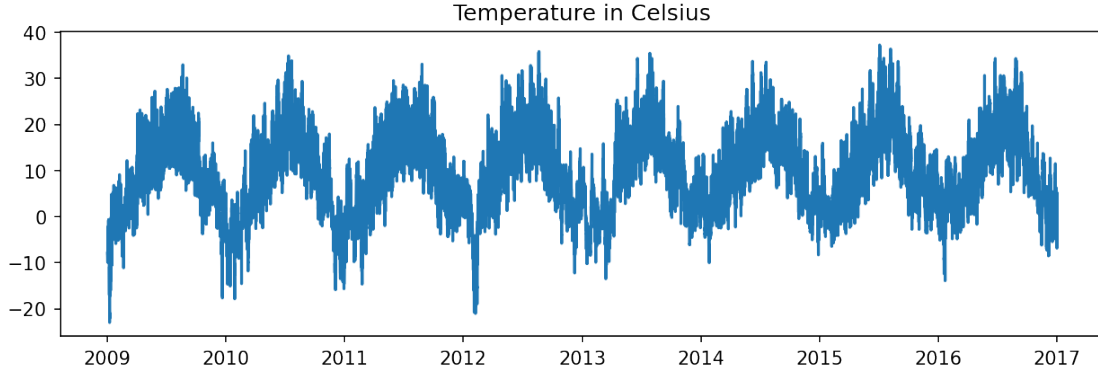
## 1.4 Data visualization: Temperature in celsius

```
[4]: T_data = data['T (degC)']  
date_time = pd.to_datetime(data['Date Time'],format='%d.%m.%Y %H:%M:%S')
```

```
[5]: print(T_data)  
print(T_data.shape)  
print(date_time)
```

```
0      -8.02  
1      -8.41  
2      -8.51  
3      -8.31  
4      -8.27  
...  
420546  -4.05  
420547  -3.35  
420548  -3.16  
420549  -4.23  
420550  -4.82  
Name: T (degC), Length: 420551, dtype: float64  
(420551,)  
0      2009-01-01 00:10:00  
1      2009-01-01 00:20:00  
2      2009-01-01 00:30:00  
3      2009-01-01 00:40:00  
4      2009-01-01 00:50:00  
...  
420546  2016-12-31 23:20:00  
420547  2016-12-31 23:30:00  
420548  2016-12-31 23:40:00  
420549  2016-12-31 23:50:00  
420550  2017-01-01 00:00:00  
Name: Date Time, Length: 420551, dtype: datetime64[ns]
```

```
[6]: import matplotlib.pyplot as plt  
  
plt.figure(figsize=(10,3), dpi=150)  
plt.plot(date_time, T_data)  
plt.title('Temperature in Celsius')  
plt.show()
```



## 1.5 Correlation analysis

We employ Pearson correlation defined as:

$$\rho_{ij} := \frac{\text{Cov}(x_i, x_j)}{\sqrt{\text{Var}(x_i)\text{Var}(x_j)}}.$$

It is in between -1 and 1.

```
[7]: data.corr()
```

```
[7]:
```

	p (mbar)	T (degC)	Tpot (K)	Tdew (degC)	rh (%)	\
p (mbar)	1.000000	-0.045375	-0.124718	-0.066755	-0.018352	
T (degC)	-0.045375	1.000000	0.996827	0.895708	-0.572416	
Tpot (K)	-0.124718	0.996827	1.000000	0.894911	-0.567127	
Tdew (degC)	-0.066755	0.895708	0.894911	1.000000	-0.156615	
rh (%)	-0.018352	-0.572416	-0.567127	-0.156615	1.000000	
VPmax (mbar)	-0.031546	0.951113	0.947293	0.799271	-0.615842	
VPact (mbar)	-0.054370	0.867673	0.866205	0.968344	-0.151494	
VPdef (mbar)	-0.003401	0.761744	0.756962	0.435752	-0.843835	
sh (g/kg)	-0.069762	0.866755	0.866533	0.967599	-0.150841	
H2OC (mmol/mol)	-0.069804	0.867177	0.866955	0.968044	-0.150969	
rho (g/m**3)	0.307640	-0.963410	-0.981345	-0.885232	0.514282	
wv (m/s)	-0.005701	-0.004689	-0.004195	-0.008718	-0.005020	
max. wv (m/s)	-0.007760	-0.002871	-0.002224	-0.009091	-0.009921	
wd (deg)	-0.063258	0.038732	0.043599	0.049877	-0.015912	

	VPmax (mbar)	VPact (mbar)	VPdef (mbar)	sh (g/kg)	\
p (mbar)	-0.031546	-0.054370	-0.003401	-0.069762	
T (degC)	0.951113	0.867673	0.761744	0.866755	
Tpot (K)	0.947293	0.866205	0.756962	0.866533	
Tdew (degC)	0.799271	0.968344	0.435752	0.967599	
rh (%)	-0.615842	-0.151494	-0.843835	-0.150841	
VPmax (mbar)	1.000000	0.824865	0.875588	0.824460	

VPact (mbar)	0.824865	1.000000	0.449154	0.999851
VPdef (mbar)	0.875588	0.449154	1.000000	0.448641
sh (g/kg)	0.824460	0.999851	0.448641	1.000000
H2OC (mmol/mol)	0.824493	0.999856	0.448689	0.999997
rho (g/m**3)	-0.901536	-0.850241	-0.698290	-0.853325
wv (m/s)	-0.004018	-0.009600	0.001852	-0.009479
max. wv (m/s)	-0.002213	-0.010316	0.005317	-0.010163
wd (deg)	-0.009583	0.018418	-0.030881	0.019376

	H2OC (mmol/mol)	rho (g/m**3)	wv (m/s)	max. wv (m/s)	\
p (mbar)	-0.069804	0.307640	-0.005701	-0.007760	
T (degC)	0.867177	-0.963410	-0.004689	-0.002871	
Tpot (K)	0.866955	-0.981345	-0.004195	-0.002224	
Tdew (degC)	0.968044	-0.885232	-0.008718	-0.009091	
rh (%)	-0.150969	0.514282	-0.005020	-0.009921	
VPmax (mbar)	0.824493	-0.901536	-0.004018	-0.002213	
VPact (mbar)	0.999856	-0.850241	-0.009600	-0.010316	
VPdef (mbar)	0.448689	-0.698290	0.001852	0.005317	
sh (g/kg)	0.999997	-0.853325	-0.009479	-0.010163	
H2OC (mmol/mol)	1.000000	-0.853769	-0.009477	-0.010158	
rho (g/m**3)	-0.853769	1.000000	0.003240	0.001086	
wv (m/s)	-0.009477	0.003240	1.000000	0.948477	
max. wv (m/s)	-0.010158	0.001086	0.948477	1.000000	
wd (deg)	0.019607	-0.058072	-0.015322	-0.014471	

	wd (deg)
p (mbar)	-0.063258
T (degC)	0.038732
Tpot (K)	0.043599
Tdew (degC)	0.049877
rh (%)	-0.015912
VPmax (mbar)	-0.009583
VPact (mbar)	0.018418
VPdef (mbar)	-0.030881
sh (g/kg)	0.019376
H2OC (mmol/mol)	0.019607
rho (g/m**3)	-0.058072
wv (m/s)	-0.015322
max. wv (m/s)	-0.014471
wd (deg)	1.000000

## 1.6 Statistics of data

```
[8]: data.describe()
```

```
[8]:
```

	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
min	913.600000	-23.010000	250.600000	-25.010000
25%	984.200000	3.360000	277.430000	0.240000
50%	989.580000	9.420000	283.470000	5.220000
75%	994.720000	15.470000	289.530000	10.070000
max	1015.350000	37.280000	311.340000	23.110000

	rh (%)	VPmax (mbar)	VPact (mbar)	VPdef (mbar)	\
count	420551.000000	420551.000000	420551.000000	420551.000000	
mean	76.008259	13.576251	9.533756	4.042412	
std	16.476175	7.739020	4.184164	4.896851	
min	12.950000	0.950000	0.790000	0.000000	
25%	65.210000	7.780000	6.210000	0.870000	
50%	79.300000	11.820000	8.860000	2.190000	
75%	89.400000	17.600000	12.350000	5.300000	
max	100.000000	63.770000	28.320000	46.010000	

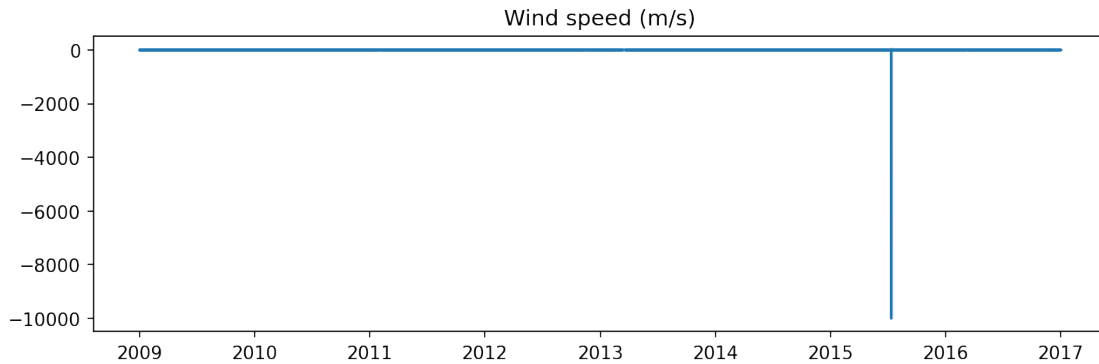
	sh (g/kg)	H2OC (mmol/mol)	rho (g/m**3)	wv (m/s)	\
count	420551.000000	420551.000000	420551.000000	420551.000000	
mean	6.022408	9.640223	1216.062748	1.702224	
std	2.656139	4.235395	39.975208	65.446714	
min	0.500000	0.800000	1059.450000	-9999.000000	
25%	3.920000	6.290000	1187.490000	0.990000	
50%	5.590000	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

## 1.7 Missing entries in wind speed (m/s)

```
[9]: import matplotlib.pyplot as plt

plt.figure(figsize=(10,3), dpi=150)
wv = data['wv (m/s)']
plt.plot(date_time, wv)
plt.title('Wind speed (m/s)')
plt.show()
```



## 1.8 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.

```
[10]: 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\_41920\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](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
wv[wv_missing_idx] = wv_mean
```

```
[11]: 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\_41920\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](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

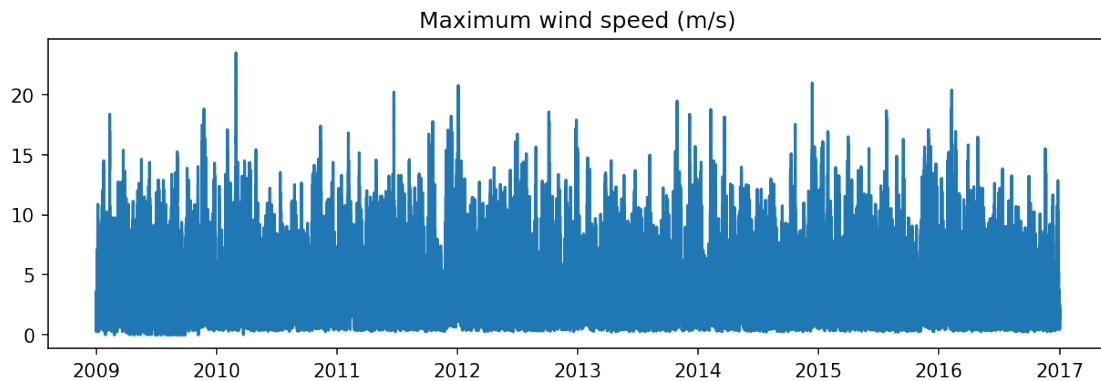
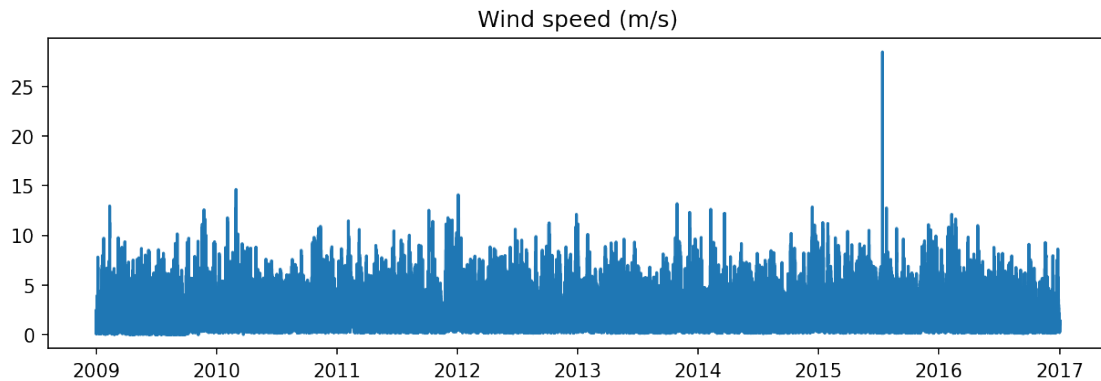
```
max_wv[missing_idx] = max_wv_mean
```

```
[12]: import matplotlib.pyplot as plt

plt.figure(figsize=(10,3), dpi=150)
plt.plot(date_time, wv)
```

```
plt.title('Wind speed (m/s)')
plt.show()

plt.figure(figsize=(10,3), dpi=150)
plt.plot(date_time, max_wv)
plt.title('Maximum wind speed (m/s)')
plt.show()
```



## 1.9 Check if missing entries are properly filled up

```
[13]: data.describe()
```

```
[13]:
```

	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	
min	913.600000	-23.010000	250.600000	-25.010000	
25%	984.200000	3.360000	277.430000	0.240000	
50%	989.580000	9.420000	283.470000	5.220000	



75%	994.720000	15.470000	289.530000	10.070000
max	1015.350000	37.280000	311.340000	23.110000

	rh (%)	VPmax (mbar)	VPact (mbar)	VPdef (mbar)	\
count	420551.000000	420551.000000	420551.000000	420551.000000	
mean	76.008259	13.576251	9.533756	4.042412	
std	16.476175	7.739020	4.184164	4.896851	
min	12.950000	0.950000	0.790000	0.000000	
25%	65.210000	7.780000	6.210000	0.870000	
50%	79.300000	11.820000	8.860000	2.190000	
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)	\
count	420551.000000	420551.000000	420551.000000	420551.000000	
mean	6.022408	9.640223	1216.062748	2.130282	
std	2.656139	4.235395	39.975208	1.542271	
min	0.500000	0.800000	1059.450000	0.000000	
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	3.532242	174.743738
std	2.340355	86.681693
min	0.000000	0.000000
25%	1.760000	124.900000
50%	2.960000	198.100000
75%	4.740000	234.100000
max	23.500000	360.000000

## 1.10 Remove date\_time column

```
[14]: data.pop('Date Time')
data
```

```
[14]:
```

	p (mbar)	T (degC)	Tpot (K)	Tdew (degC)	rh (%)	VPmax (mbar)	\
0	996.52	-8.02	265.40	-8.90	93.30	3.33	
1	996.57	-8.41	265.01	-9.28	93.40	3.23	
2	996.53	-8.51	264.91	-9.31	93.90	3.21	
3	996.51	-8.31	265.12	-9.07	94.20	3.26	
4	996.51	-8.27	265.15	-9.04	94.10	3.27	
...	...	...	...	...	...	...	
420546	1000.07	-4.05	269.10	-8.13	73.10	4.52	
420547	999.93	-3.35	269.81	-8.06	69.71	4.77	

420548	999.82	-3.16	270.01	-8.21	67.91	4.84
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)	sh (g/kg)	H2OC (mmol/mol)	rho (g/m**3)	\
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
420549	3.20	1.26	1.99		3.20	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
420548	1.08	2.00	215.2
420549	1.49	2.16	225.8
420550	1.23	1.96	184.9

[420551 rows x 14 columns]

[ ]:

[ ]:

[ ]: