

# Assignment 1 temperature

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## 1 Assignment 1

### 1.1 The data set

The considered data set has been generated by the Korean Meteorological Administration over the city of Seoul (SK), in the time period between 2013 and 2017, uploaded on the UC Irvine Machine Learning Repository. The data consists in the LDAPS (Local Data Assimilation and Prediction System) model's forecasts, the present-day minimum and maximum temperatures and other auxiliary variables, while the two output variables are the next-day maximum and minimum air temperatures.

### 1.2 Applied goals in studying the data

The aim of studying such data is to perform a bias correction of the future minimum and maximum air temperature forecasts of the Korean Meteorological Administration LDAPS model. Precise forecasts of maximum and minimum air temperatures are crucial to anticipate and prevent the possible damage caused by extreme meteorological events, such as heat waves and cold spells. Bias correction is a process used to address bias errors in the predictions of a Numerical Weather Prediction (NWP) climate model, that present some discrepancies when compared to observed data. These errors usually arise with extreme, but unusually occurring, rainfall intensities or temperatures, that are underestimated by NWP models, due to a variety of factors (e.g. a large grid size - which leads to a less detailed representation of the weather, due to a reduced spacial resolution - and the simplification of thermodynamic processes). Therefore, long-term data sets with a model's output and historic data from the same period are used to correct extreme temperatures with linear regressions: the adjusted future time series is then constructed with the updated regression parameters.

### 1.3 Exploratory analysis

The study area is the metropolitan city of Seoul (SK), which is located in a geographical region surrounded by four mountains and split in two by the Han River. Precipitation is particularly abundant in summer, due to the East Asian monsoon, therefore the period is characterized by temperatures above 30°C.

The data set is composed of data collected from the 30th of June 2013 to the 30th of August 2017 and it consists of the following variables: the ones produced by the LDAPS model for the forecasts are the next-day maximum and minimum air temperatures (°C), next-day maximum and minimum relative humidity (%), next-day average wind speed (m/s), next-day average latent heat flux (W/m<sup>2</sup>), next-day average cloud cover (%), next-day average precipitation (%). Additionally, the data collected in-situ comprises of the observed

next-day maximum and minimum air temperatures ( $^{\circ}\text{C}$ ) - used as target variables -, as well as the present max and min air temperatures, obtained from a grid with 25 automatic weather stations scattered around the city. Other auxiliary variables are the latitude and longitude (in coordinates), elevation and slope (grouped as topographic variables) and daily incoming solar radiation ( $\text{wh/m}^2$ ).

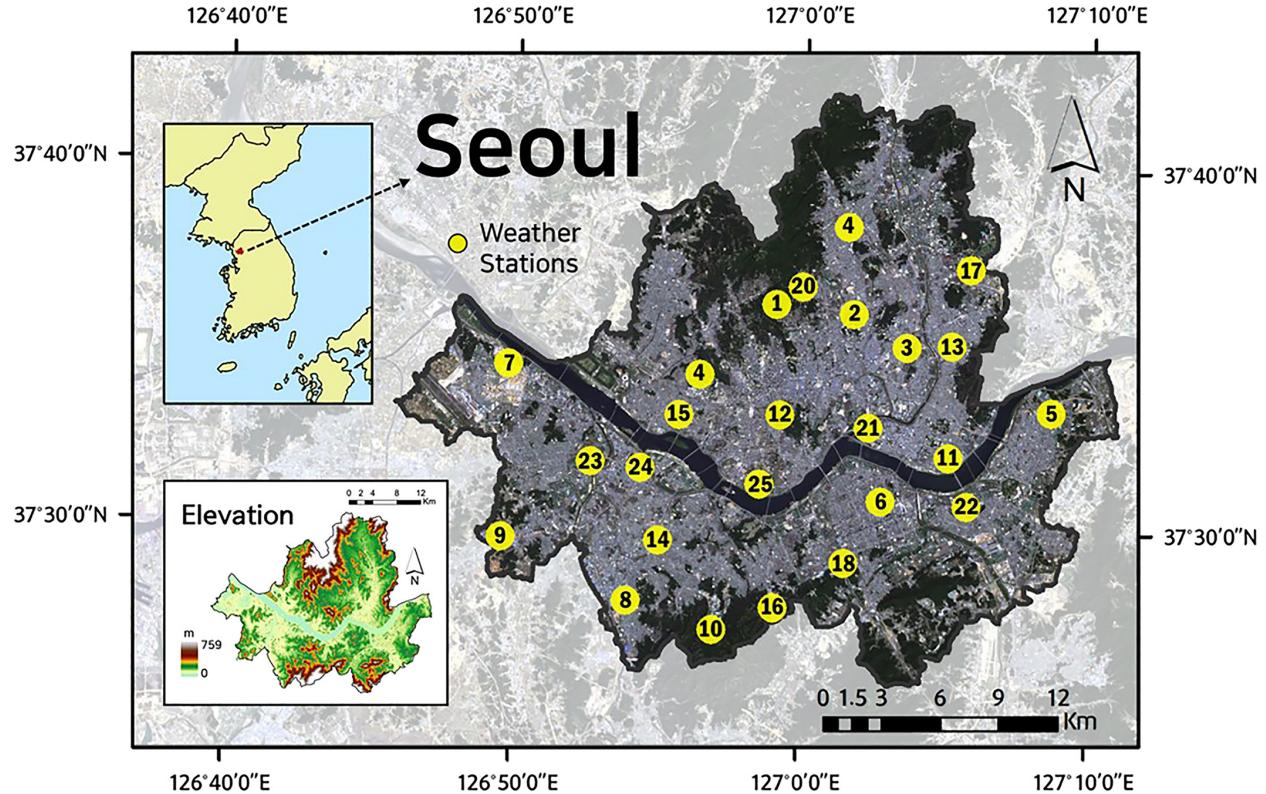


Figure 1: Location of automatic weather stations

We now read the data, remove the 1248 missing values and provide a summary of the data set. The number of data is 7752 and the regression variables are 25.

```
data = read.csv("Bias_correction.csv", header = T)
head(data)
```

```
##   station      Date Present_Tmax Present_Tmin     RHmin     RHmax Tmax_lapse
## 1       1 30/06/2013      28.7      21.4 58.25569 91.11636  28.07410
## 2       2 30/06/2013      31.9      21.6 52.26340 90.60472  29.85069
## 3       3 30/06/2013      31.6      23.3 48.69048 83.97359  30.09129
## 4       4 30/06/2013      32.0      23.4 58.23979 96.48369  29.70463
## 5       5 30/06/2013      31.4      21.9 56.17410 90.15513  29.11393
## 6       6 30/06/2013      31.9      23.5 52.43713 85.30725  29.21934
##   Tmin_lapse      WS      LH      CC1      CC2      CC3      CC4 PPT1
## 1 23.00694 6.818887 69.45181 0.2339475 0.2038957 0.1616969 0.1309282 0
## 2 24.03501 5.691890 51.93745 0.2255082 0.2517714 0.1594441 0.1277273 0
## 3 24.56563 6.138224 20.57305 0.2093437 0.2574694 0.2040915 0.1421253 0
## 4 23.32618 5.650050 65.72714 0.2163720 0.2260024 0.1611574 0.1342487 0
## 5 23.48648 5.735004 107.96554 0.1514069 0.2499953 0.1788925 0.1700210 0
## 6 23.82261 6.182295 50.23139 0.1852788 0.2808180 0.2328410 0.1463629 0
##   PPT2 PPT3 PPT4      lat      lon      DEM Slope Solar.radiation Next_Tmax
## 1     0     0     0 37.6046 126.991 212.3350 2.7850      5992.896     29.1
```

```

## 2 0 0 0 37.6046 127.032 44.7624 0.5141 5869.312 30.5
## 3 0 0 0 37.5776 127.058 33.3068 0.2661 5863.556 31.1
## 4 0 0 0 37.6450 127.022 45.7160 2.5348 5856.965 31.7
## 5 0 0 0 37.5507 127.135 35.0380 0.5055 5859.552 31.2
## 6 0 0 0 37.5102 127.042 54.6384 0.1457 5873.781 31.5
##   Next_Tmin
## 1 21.2
## 2 22.5
## 3 23.9
## 4 24.3
## 5 22.5
## 6 24.0
sum(is.na(data))

## [1] 1248
data = na.omit(data)
data$date = as.Date(data$date, format = "%d/%m/%Y")
as.factor(data$station)

## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
## [25] 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23
## [49] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [73] 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
## [97] 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## [121] 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
## [145] 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## [169] 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17
## [193] 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
## [217] 17 18 19 20 21 22 23 24 25 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
## [241] 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
## [265] 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
## [289] 16 17 18 19 20 21 22 23 24 25 2 3 4 5 6 7 8 9 10 11 12 13 14 15
## [313] 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14
## [337] 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13
## [361] 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12
## [385] 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11
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## [433] 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 2 3 4 5 6 7 8 9 10
## [457] 11 12 13 14 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10
## [481] 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9
## [505] 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8
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## [625] 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4
## [649] 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3
## [673] 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2
## [697] 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1
## [721] 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
## [745] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
## [769] 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23
## [793] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [817] 23 24 25 1 2 3 4 5 6 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22

```





```

## [3433] 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
## [3457] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
## [3481] 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23
## [3505] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [3529] 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
## [3553] 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## [3577] 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
## [3601] 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
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## [3649] 18 19 20 21 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## [3673] 20 21 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 16 17 18 19 20
## [3697] 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 16 17 18 19 20
## [3721] 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
## [3745] 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## [3769] 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17
## [3793] 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
## [3817] 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 13 14 15 16
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## [3889] 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14
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## [3937] 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12
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## [4081] 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7
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## [4129] 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5
## [4153] 7 8 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 7 8
## [4177] 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 7 8
## [4201] 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 7 8
## [4225] 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 7 8
## [4249] 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7
## [4273] 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6
## [4297] 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5
## [4321] 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4
## [4345] 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3
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## [4393] 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1
## [4417] 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
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## [4489] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
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## [4537] 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
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## [4585] 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## [4609] 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17
## [4633] 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
## [4657] 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
## [4681] 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 15
## [4705] 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14

```



```

## [6025] 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
## [6049] 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
## [6073] 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 16 17 18 22
## [6097] 23 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [6121] 23 24 25 1 2 3 5 6 7 8 9 10 11 12 13 14 15 16 17 18 20 21 22 23
## [6145] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 20 21 22 23
## [6169] 24 25 1 2 3 4 5 6 7 8 9 10 11 13 14 15 16 17 18 19 20 21 22 23
## [6193] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 17 18 19 20 21 22 23
## [6217] 24 25 1 2 3 4 5 6 7 9 10 12 13 14 15 16 17 18 19 20 21 22 23 25
## [6241] 1 2 3 4 5 6 7 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
## [6265] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
## [6289] 25 1 2 3 4 5 6 7 8 9 10 12 13 14 15 16 17 18 19 20 21 22 23 24
## [6313] 25 1 2 3 4 5 6 7 8 9 10 12 13 14 15 16 17 18 19 20 21 22 23 24
## [6337] 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23
## [6361] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [6385] 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
## [6409] 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## [6433] 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
## [6457] 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## [6481] 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17
## [6505] 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 14 15 16 17
## [6529] 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
## [6553] 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
## [6577] 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14
## [6601] 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13
## [6625] 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12
## [6649] 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11
## [6673] 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10
## [6697] 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8 9
## [6721] 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7 8
## [6745] 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7
## [6769] 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7
## [6793] 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6 7
## [6817] 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5 6
## [6841] 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4 5
## [6865] 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4
## [6889] 5 6 7 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3 4
## [6913] 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2 3
## [6937] 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1 2
## [6961] 3 4 5 6 7 8 9 10 11 12 14 15 16 17 18 19 20 21 22 23 24 25 1 2
## [6985] 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 1
## [7009] 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
## [7033] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
## [7057] 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23
## [7081] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [7105] 23 24 25 1 2 3 4 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [7129] 23 24 25 1 2 3 4 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [7153] 23 24 25 1 2 3 4 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [7177] 23 24 25 1 2 3 4 6 7 8 9 10 11 12 13 14 15 16 17 18 19 21 22 23
## [7201] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 21 22 23
## [7225] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [7249] 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
## [7273] 22 23 24 25 1 2 3 4 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
## [7297] 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20

```

```

## [7321] 21 22 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
## [7345] 20 21 22 23 24 25 1 2 3 4 6 7 8 9 10 11 12 13 14 15 16 17 18 19
## [7369] 20 21 22 23 24 25 1 2 3 4 6 7 8 9 10 11 12 13 14 15 16 17 18 19
## [7393] 20 21 22 23 24 25 1 2 3 4 6 7 8 9 10 11 12 13 15 16 17 18 19 20
## [7417] 21 23 24 25 1 2 3 4 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
## [7441] 23 24 25 1 2 3 4 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23
## [7465] 24 25 1 2 3 4 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23
## [7489] 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [7513] 23 24 25 1 2 3 4 5 6 7 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [7537] 23 24 25 1 2 3 4 5 6 7 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## [7561] 23 24 25 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
## [7585] 22 23 24 25
## 25 Levels: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 ... 25

```

```
summary(data)
```

```

##      station          Date        Present_Tmax    Present_Tmin
## Min.   : 1.00  Min.   :2013-06-30  Min.   :20.00  Min.   :11.3
## 1st Qu.: 7.00  1st Qu.:2014-07-15  1st Qu.:27.80  1st Qu.:21.6
## Median :13.00  Median :2015-07-29  Median :29.90  Median :23.4
## Mean   :13.01  Mean   :2015-07-27  Mean   :29.75  Mean   :23.2
## 3rd Qu.:19.00  3rd Qu.:2016-08-14  3rd Qu.:32.00  3rd Qu.:24.8
## Max.   :25.00  Max.   :2017-08-30  Max.   :37.60  Max.   :29.9
##      RHmin         RHmax        Tmax_lapse    Tmin_lapse
## Min.   :19.79  Min.   : 58.94  Min.   :17.62  Min.   :14.27
## 1st Qu.:45.96  1st Qu.: 84.20  1st Qu.:27.67  1st Qu.:22.09
## Median :55.02  Median : 89.78  Median :29.71  Median :23.76
## Mean   :56.72  Mean   : 88.36  Mean   :29.62  Mean   :23.51
## 3rd Qu.:67.12  3rd Qu.: 93.74  3rd Qu.:31.71  3rd Qu.:25.16
## Max.   :98.52  Max.   :100.00  Max.   :38.54  Max.   :29.62
##      WS            LH           CC1          CC2
## Min.   : 2.883  Min.   :-13.60  Min.   :0.00000  Min.   :0.00000
## 1st Qu.: 5.675  1st Qu.: 37.21  1st Qu.:0.1465  1st Qu.:0.1403
## Median : 6.548  Median : 56.90  Median :0.3157  Median :0.3117
## Mean   : 7.094  Mean   : 62.49  Mean   :0.3685  Mean   :0.3555
## 3rd Qu.: 8.029  3rd Qu.: 84.24  3rd Qu.:0.5742  3rd Qu.:0.5572
## Max.   :21.858  Max.   :213.41  Max.   :0.9673  Max.   :0.9684
##      CC3            CC4          PPT1          PPT2
## Min.   :0.00000  Min.   :0.00000  Min.   :0.00000  Min.   : 0.00000
## 1st Qu.:0.1009  1st Qu.:0.08149  1st Qu.: 0.00000  1st Qu.: 0.00000
## Median :0.2618  Median :0.22746  Median : 0.00000  Median : 0.00000
## Mean   :0.3175  Mean   :0.29827  Mean   : 0.58901  Mean   : 0.48074
## 3rd Qu.:0.4964  3rd Qu.:0.49813  3rd Qu.: 0.05259  3rd Qu.: 0.01774
## Max.   :0.9838  Max.   :0.97471  Max.   :23.70154  Max.   :21.62166
##      PPT3            PPT4          lat          lon
## Min.   : 0.000000  Min.   : 0.000000  Min.   :37.46  Min.   :126.8
## 1st Qu.: 0.000000  1st Qu.: 0.000000  1st Qu.:37.51  1st Qu.:126.9
## Median : 0.000000  Median : 0.000000  Median :37.55  Median :127.0
## Mean   : 0.275007  Mean   : 0.265373  Mean   :37.54  Mean   :127.0
## 3rd Qu.: 0.007855  3rd Qu.: 0.000017  3rd Qu.:37.58  3rd Qu.:127.0
## Max.   :15.841235  Max.   :16.655469  Max.   :37.65  Max.   :127.1
##      DEM            Slope        Solar.radiation  Next_Tmax
## Min.   :12.37  Min.   :0.0985  Min.   :4330  Min.   :17.40
## 1st Qu.:28.70  1st Qu.:0.2713  1st Qu.:5001  1st Qu.:28.20
## Median :45.72  Median :0.6180  Median :5442  Median :30.40

```

```

##  Mean    : 61.92   Mean    :1.2598   Mean    :5344    Mean    :30.24
##  3rd Qu.: 59.83   3rd Qu.:1.7678   3rd Qu.:5729    3rd Qu.:32.60
##  Max.   :212.34   Max.   :5.1782   Max.   :5993    Max.   :38.90
##  Next_Tmin
##  Min.   :11.30
##  1st Qu.:21.30
##  Median :23.10
##  Mean   :22.91
##  3rd Qu.:24.60
##  Max.   :29.80

str(data)

## 'data.frame': 7588 obs. of  25 variables:
## $ station      : int  1 2 3 4 5 6 7 8 9 10 ...
## $ Date         : Date, format: "2013-06-30" "2013-06-30" ...
## $ Present_Tmax : num  28.7 31.9 31.6 32 31.4 31.9 31.4 32.1 31.4 31.6 ...
## $ Present_Tmin : num  21.4 21.6 23.3 23.4 21.9 23.5 24.4 23.6 22 20.5 ...
## $ RHmin        : num  58.3 52.3 48.7 58.2 56.2 ...
## $ RHmax        : num  91.1 90.6 84 96.5 90.2 ...
## $ Tmax_lapse   : num  28.1 29.9 30.1 29.7 29.1 ...
## $ Tmin_lapse   : num  23 24 24.6 23.3 23.5 ...
## $ WS           : num  6.82 5.69 6.14 5.65 5.74 ...
## $ LH           : num  69.5 51.9 20.6 65.7 108 ...
## $ CC1          : num  0.234 0.226 0.209 0.216 0.151 ...
## $ CC2          : num  0.204 0.252 0.257 0.226 0.25 ...
## $ CC3          : num  0.162 0.159 0.204 0.161 0.179 ...
## $ CC4          : num  0.131 0.128 0.142 0.134 0.17 ...
## $ PPT1         : num  0 0 0 0 0 0 0 0 0 0 ...
## $ PPT2         : num  0 0 0 0 0 0 0 0 0 0 ...
## $ PPT3         : num  0 0 0 0 0 0 0 0 0 0 ...
## $ PPT4         : num  0 0 0 0 0 0 0 0 0 0 ...
## $ lat          : num  37.6 37.6 37.6 37.6 37.6 ...
## $ lon          : num  127 127 127 127 127 ...
## $ DEM          : num  212.3 44.8 33.3 45.7 35 ...
## $ Slope         : num  2.785 0.514 0.266 2.535 0.505 ...
## $ Solar.radiation: num  5993 5869 5864 5857 5860 ...
## $ Next_Tmax    : num  29.1 30.5 31.1 31.7 31.2 31.5 30.9 31.1 31.3 30.5 ...
## $ Next_Tmin    : num  21.2 22.5 23.9 24.3 22.5 24 23.4 22.9 21.6 21 ...
## - attr(*, "na.action")= 'omit' Named int [1:164] 226 272 301 451 465 628 832 857 882 914 ...
## ..- attr(*, "names")= chr [1:164] "226" "272" "301" "451" ...
selected_variables <- c("Present_Tmax", "Present_Tmin", "RHmin", "RHmax", "Tmax_lapse", "Tmin_lapse", "WS")
selected_data <- data[, selected_variables]
correlation_matrix <- cor(selected_data)
print(correlation_matrix)

##              Present_Tmax Present_Tmin       RHmin       RHmax   Tmax_lapse
## Present_Tmax 1.000000000 0.61520818 -0.20905865 -0.30694324 0.57377689
## Present_Tmin  0.615208176 1.00000000 0.12219987 -0.01857640 0.46790418
## RHmin        -0.209058645 0.12219987 1.00000000 0.57835786 -0.56957978
## RHmax        -0.306943243 -0.01857640 0.57835786 1.00000000 -0.37729261
## Tmax_lapse   0.573776885 0.46790418 -0.56957978 -0.37729261 1.00000000
## Tmin_lapse   0.628704554 0.77206103 0.08523684 -0.11896461 0.65342497
## WS          -0.125952332 -0.03877922 0.28942243 0.13126857 -0.31601080

```

```

## LH          0.134246486 -0.01281686 -0.07219513  0.23898592  0.04366273
## CC1        -0.316333001  0.08434794  0.61283130  0.43593663 -0.44087760
## CC2        -0.216976122  0.09003759  0.74477409  0.39062032 -0.52537969
## CC3        -0.146391153 -0.00471934  0.68806867  0.22473396 -0.54272826
## CC4        -0.143069385 -0.04669442  0.51393680  0.12724159 -0.42973910
## PPT1       -0.114916546  0.11157678  0.25903888  0.26801864 -0.11565947
## PPT2       -0.104779098  0.06620729  0.38957186  0.22758086 -0.24836865
## PPT3       -0.127247895 -0.05242821  0.23751206  0.13243636 -0.19374912
## PPT4       -0.107294512 -0.07039097  0.16731078  0.11810122 -0.17041549
## lat         -0.054133706 -0.08141654  0.08502748  0.19481965 -0.04405242
## lon         0.007694507 -0.04517409 -0.07758996  0.02556256  0.09156832
## DEM         -0.189438545 -0.25408359  0.10068442  0.17655730 -0.18143288
## Slope       -0.107079402 -0.14842304  0.12245405  0.21943629 -0.16389116
## Solar.radiation -0.022114328  0.05976502  0.24216837  0.14662206  0.04741393
## Next_Tmax   0.610356613  0.46352292 -0.44682411 -0.28978793  0.83572903
## Next_Tmin   0.621048410  0.79697478  0.09324949 -0.07641549  0.58996315
##           Tmin_lapse      WS      LH      CC1
## Present_Tmax 0.628704554 -0.125952332  0.134246486 -0.316333001
## Present_Tmin 0.772061034 -0.038779221 -0.012816863  0.084347945
## RHmin        0.085236844  0.289422426 -0.072195133  0.612831303
## RHmax        -0.118964608  0.131268569  0.238985920  0.435936625
## Tmax_lapse   0.653424970 -0.316010798  0.043662729 -0.440877597
## Tmin_lapse   1.000000000 -0.134651681 -0.139674730  0.008713542
## WS            -0.134651681  1.000000000  0.004426528  0.285348577
## LH            -0.139674730  0.004426528  1.000000000 -0.148314260
## CC1           0.008713542  0.285348577 -0.148314260  1.000000000
## CC2           0.046536557  0.257226197 -0.265381931  0.779510001
## CC3           -0.044998320  0.238154536 -0.246709265  0.516698171
## CC4           -0.079535786  0.218869842 -0.173156106  0.362670096
## PPT1          0.034956109  0.141502506 -0.018707932  0.446929869
## PPT2          -0.002743111  0.181633685 -0.084375307  0.389367961
## PPT3          -0.093116424  0.147199395  0.009495158  0.166763283
## PPT4          -0.103540878  0.134428080  0.012371860  0.088896522
## lat           -0.099451729  0.033025192  0.133309349 -0.010267692
## lon           -0.024702519 -0.063008633  0.024475724 -0.007962488
## DEM           -0.198395839  0.188539918  0.056361555 -0.017377763
## Slope         -0.187864337  0.169222835  0.087516874 -0.023892750
## Solar.radiation 0.157763236  0.120629731 -0.044955505  0.218285237
## Next_Tmax    0.592118965 -0.351106889  0.156779046 -0.459704617
## Next_Tmin    0.886512767 -0.102021255 -0.060392495 -0.012876541
##           CC2      CC3      CC4      PPT1
## Present_Tmax -0.216976122 -0.146391153 -0.143069385 -0.114916546
## Present_Tmin  0.090037594 -0.004719340 -0.046694424  0.111576782
## RHmin         0.744774092  0.688068671  0.513936801  0.259038884
## RHmax         0.390620323  0.224733962  0.127241586  0.268018642
## Tmax_lapse   -0.525379690 -0.542728256 -0.429739103 -0.115659470
## Tmin_lapse   0.046536557 -0.044998320 -0.079535786  0.034956109
## WS            0.257226197  0.238154536  0.218869842  0.141502506
## LH            -0.265381931 -0.246709265 -0.173156106 -0.018707932
## CC1           0.779510001  0.516698171  0.362670096  0.446929869
## CC2           1.000000000  0.724546065  0.527333079  0.304441603
## CC3           0.724546065  1.000000000  0.794463222  0.154952486
## CC4           0.527333079  0.794463222  1.000000000  0.115967791
## PPT1          0.304441603  0.154952486  0.115967791  1.000000000

```

```

## PPT2          0.479510983  0.336253550  0.286265945  0.367922364
## PPT3          0.238179029  0.337345684  0.265709306  0.012846918
## PPT4          0.157194397  0.303787107  0.380596989  0.004376576
## lat           -0.002642199  0.003826780  -0.010611835 -0.005840691
## lon           -0.004757250  0.013947591  -0.004455768 -0.004375157
## DEM           -0.016249366  -0.001249168  -0.010239794  0.005412335
## Slope         -0.021173255  -0.005903459  -0.017890555  0.010340376
## Solar.radiation 0.169346323  0.116691186  0.114267925  0.084803769
## Next_Tmax     -0.498294901  -0.523060771  -0.454038585  -0.125954218
## Next_Tmin     0.030357996  -0.056489533  -0.086197216  0.018031316
##                  PPT2          PPT3          PPT4          lat
## Present_Tmax -0.104779098 -0.127247895 -0.107294512 -0.054133706
## Present_Tmin  0.066207289 -0.052428208 -0.070390969 -0.081416543
## RHmin         0.389571855  0.237512064  0.167310782  0.085027482
## RHmax         0.227580863  0.132436360  0.118101224  0.194819648
## Tmax_lapse    -0.248368650 -0.193749119 -0.170415485 -0.044052417
## Tmin_lapse    -0.002743111 -0.093116424 -0.103540878 -0.099451729
## WS            0.181633685  0.147199395  0.134428080  0.033025192
## LH            -0.084375307  0.009495158  0.012371860  0.133309349
## CC1           0.389367961  0.166763283  0.088896522 -0.010267692
## CC2           0.479510983  0.238179029  0.157194397 -0.002642199
## CC3           0.336253550  0.337345684  0.303787107  0.003826780
## CC4           0.286265945  0.265709306  0.380596989 -0.010611835
## PPT1          0.367922364  0.012846918  0.004376576 -0.005840691
## PPT2          1.000000000  0.205285257  0.121936971  0.015198224
## PPT3          0.205285257  1.000000000  0.288093165  0.031821763
## PPT4          0.121936971  0.288093165  1.000000000  0.008178147
## lat           0.015198224  0.031821763  0.008178147  1.000000000
## lon           0.016957509  0.018093874  0.041095055  0.289999101
## DEM           0.002411003  0.011345072 -0.009829926  0.029343775
## Slope         0.003702524  0.018372179 -0.008689008  0.071446200
## Solar.radiation 0.107620412 -0.057430108  0.033248768  0.036514051
## Next_Tmax     -0.186722283 -0.203553786 -0.190720113 -0.055367134
## Next_Tmin     -0.012461816 -0.068663981 -0.114899690 -0.085135814
##                  lon          DEM          Slope          Solar.radiation
## Present_Tmax  0.007694507 -0.189438545 -0.107079402 -0.022114328
## Present_Tmin  -0.045174094 -0.254083591 -0.148423044  0.059765018
## RHmin         -0.077589959  0.100684418  0.122454049  0.242168370
## RHmax         0.025562558  0.176557304  0.219436293  0.146622063
## Tmax_lapse    0.091568315 -0.181432879 -0.163891155  0.047413934
## Tmin_lapse    -0.024702519 -0.198395839 -0.187864337  0.157763236
## WS            -0.063008633  0.188539918  0.169222835  0.120629731
## LH            0.024475724  0.056361555  0.087516874 -0.044955505
## CC1           -0.007962488 -0.017377763 -0.023892750  0.218285237
## CC2           -0.004757250 -0.016249366 -0.021173255  0.169346323
## CC3           0.013947591 -0.001249168 -0.005903459  0.116691186
## CC4           -0.004455768 -0.010239794 -0.017890555  0.114267925
## PPT1          -0.004375157  0.005412335  0.010340376  0.084803769
## PPT2          0.016957509  0.002411003  0.003702524  0.107620412
## PPT3          0.018093874  0.011345072  0.018372179 -0.057430108
## PPT4          0.041095055 -0.009829926 -0.008689008  0.033248768
## lat           0.289999101  0.029343775  0.071446200  0.036514051
## lon           1.000000000  0.006684501  0.037615582  0.005177649
## DEM           0.006684501  1.000000000  0.784179395  0.053240680

```

```

## Slope          0.037615582  0.784179395  1.000000000  0.025516575
## Solar.radiation 0.005177649  0.053240680  0.025516575  1.000000000
## Next_Tmax      0.003501996 -0.174577585 -0.104278815  0.016145245
## Next_Tmin      -0.045757947 -0.249303047 -0.148378803  0.125516826
##                      Next_Tmax   Next_Tmin
## Present_Tmax    0.610356613  0.62104841
## Present_Tmin    0.463522925  0.79697478
## RHmin           -0.446824108  0.09324949
## RHmax            -0.289787928 -0.07641549
## Tmax_lapse       0.835729028  0.58996315
## Tmin_lapse       0.592118965  0.88651277
## WS                -0.351106889 -0.10202126
## LH                 0.156779046 -0.06039250
## CC1              -0.459704617 -0.01287654
## CC2              -0.498294901  0.03035800
## CC3              -0.523060771 -0.05648953
## CC4              -0.454038585 -0.08619722
## PPT1             -0.125954218  0.01803132
## PPT2             -0.186722283 -0.01246182
## PPT3             -0.203553786 -0.06866398
## PPT4             -0.190720113 -0.11489969
## lat               -0.055367134 -0.08513581
## lon                0.003501996 -0.04575795
## DEM               -0.174577585 -0.24930305
## Slope            -0.104278815 -0.14837880
## Solar.radiation  0.016145245  0.12551683
## Next_Tmax         1.000000000  0.61662918
## Next_Tmin         0.616629184  1.000000000

```

In order to inspect the features of the variables, we produce exploratory plots. The following code shows the relationship between the model forecast of next-day average cloud cover (in %) and the forecasts for the minimum and maximum temperature levels:

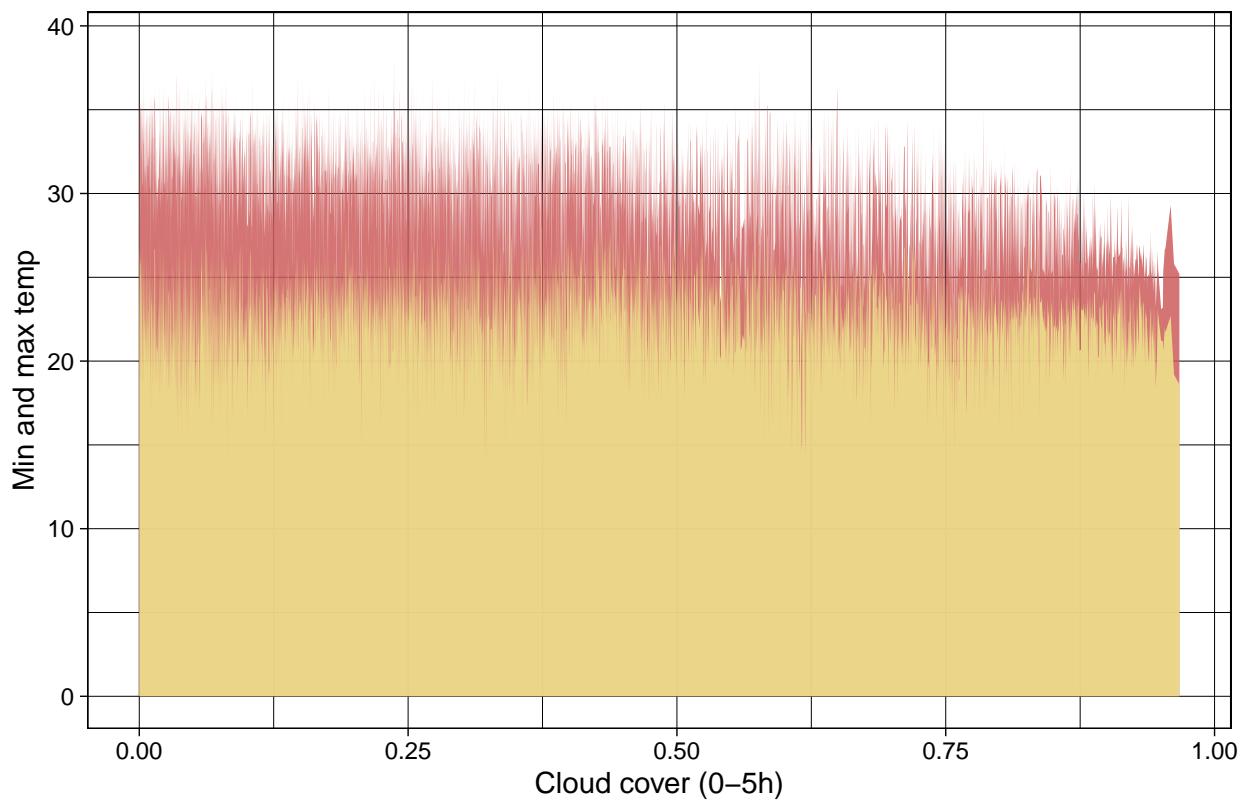
```

library(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.2.3
par(mfrow = c(1,4), mar = c(4,4,2,1), bty = "l")
ggplot(data) +
  geom_area(aes(x = data$CC1, y = data$Next_Tmax),
            fill = 'indianred', alpha = 0.85) +
  geom_area(aes(x = data$CC1, y = data$Next_Tmin),
            fill = 'khaki', alpha = 0.85) +
  labs(
    title = "Next-day maximum and minimum air temperatures based on cloud cover",
    x="Cloud cover (0-5h)",
    y="Min and max temp"
  ) +
  theme_linedraw()

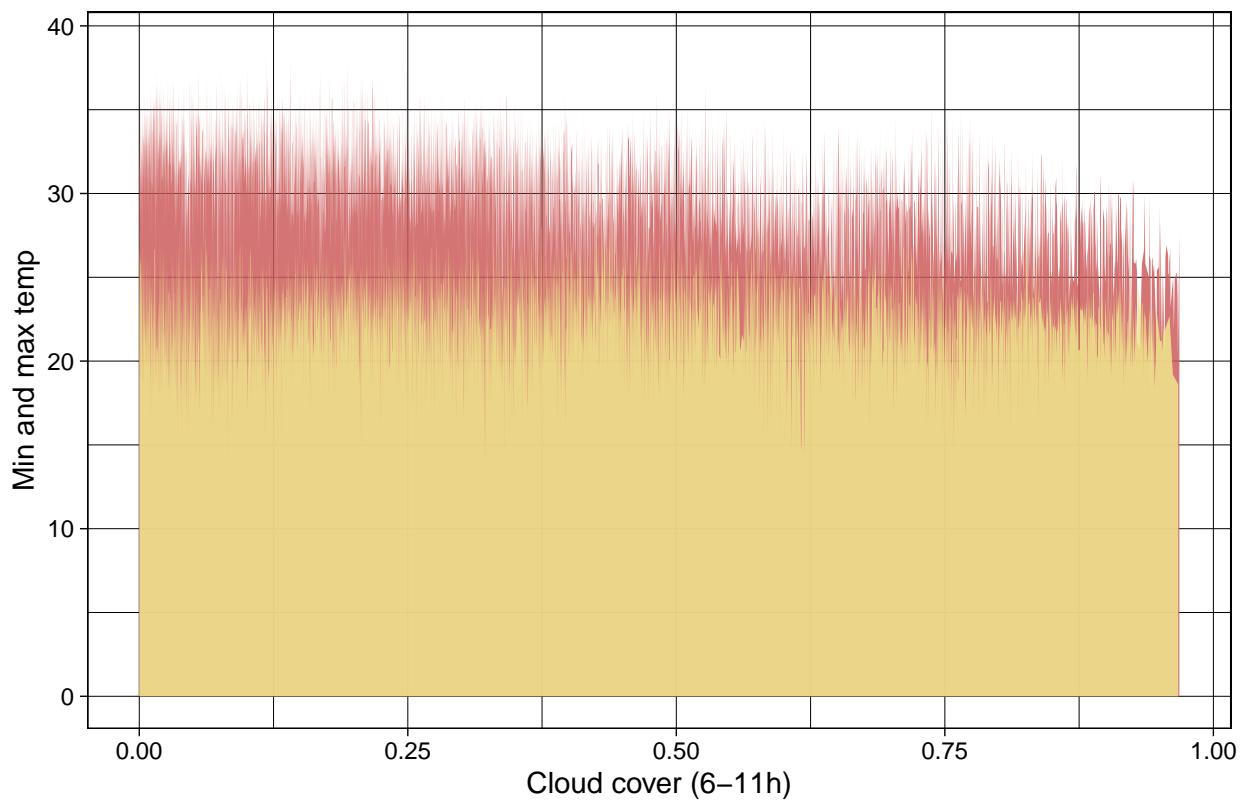
```

Next-day maximum and minimum air temperatures based on cloud cover



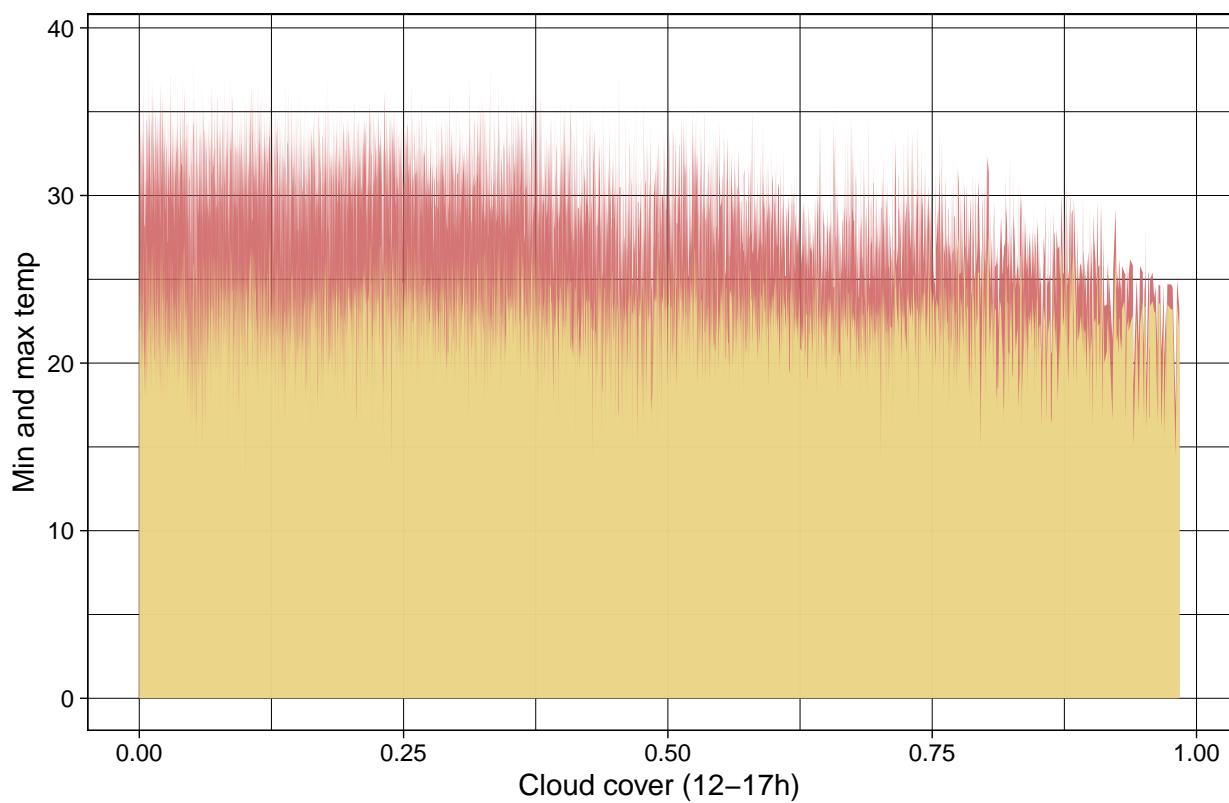
```
ggplot(data) +  
  geom_area(aes(x = data$CC2, y = data$Next_Tmax),  
            fill = 'indianred', alpha = 0.85)+  
  geom_area(aes(x = data$CC1, y = data$Next_Tmin),  
            fill = 'khaki', alpha = 0.85)+  
  labs(  
    title = "Next-day maximum and minimum air temperatures based on cloud cover",  
    x="Cloud cover (6-11h)",  
    y="Min and max temp"  
)  
  theme_linedraw()
```

Next-day maximum and minimum air temperatures based on cloud cover



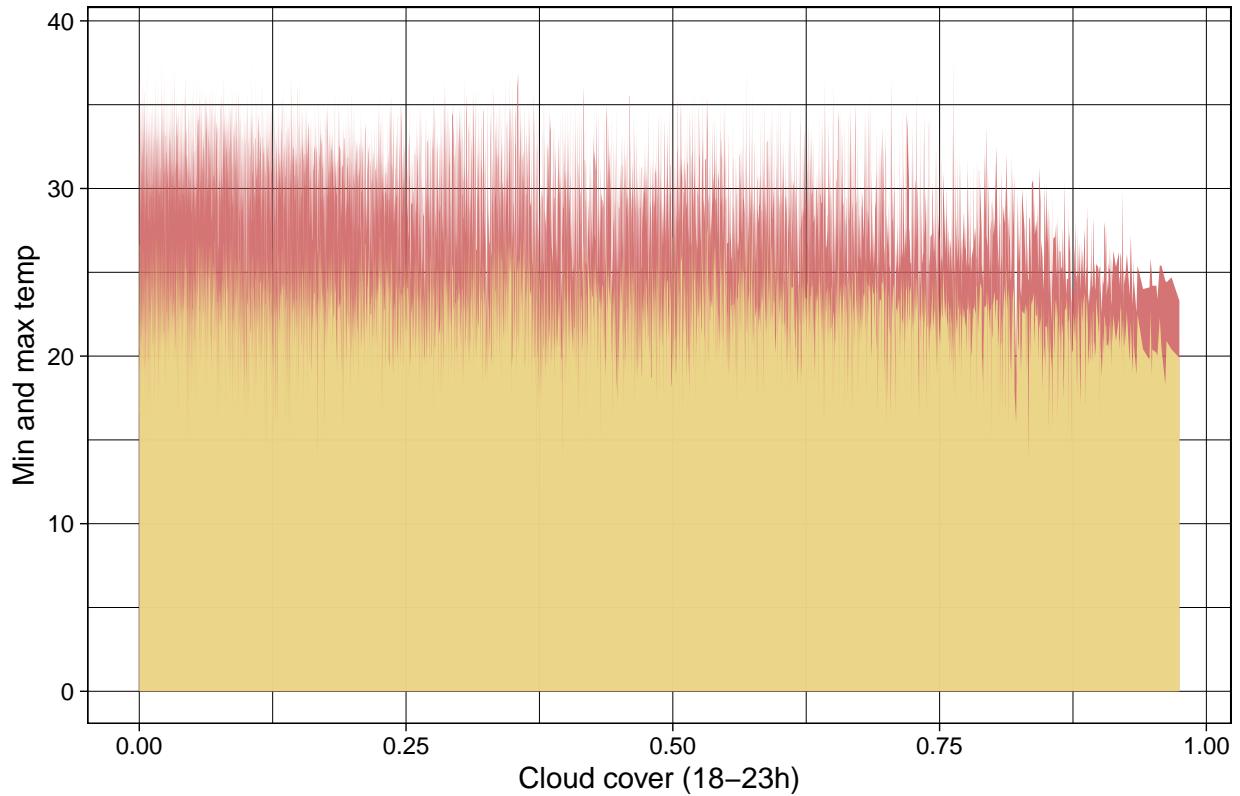
```
ggplot(data) +
  geom_area(aes(x = data$CC3, y = data$Next_Tmax),
            fill = 'indianred', alpha = 0.85) +
  geom_area(aes(x = data$CC3, y = data$Next_Tmin),
            fill = 'khaki', alpha = 0.85) +
  labs(
    title = "Next-day maximum and minimum air temperatures based on cloud cover",
    x="Cloud cover (12-17h)",
    y="Min and max temp"
  ) +
  theme_linedraw()
```

Next-day maximum and minimum air temperatures based on cloud cover



```
ggplot(data) +
  geom_area(aes(x = data$CC4, y = data$Next_Tmax),
            fill = 'indianred', alpha = 0.85) +
  geom_area(aes(x = data$CC4, y = data$Next_Tmin),
            fill = 'khaki', alpha = 0.85) +
  labs(
    title = "Next-day maximum and minimum air temperatures based on cloud cover",
    x="Cloud cover (18-23h)",
    y="Min and max temp"
  ) +
  theme_linedraw()
```

### Next-day maximum and minimum air temperatures based on cloud cover

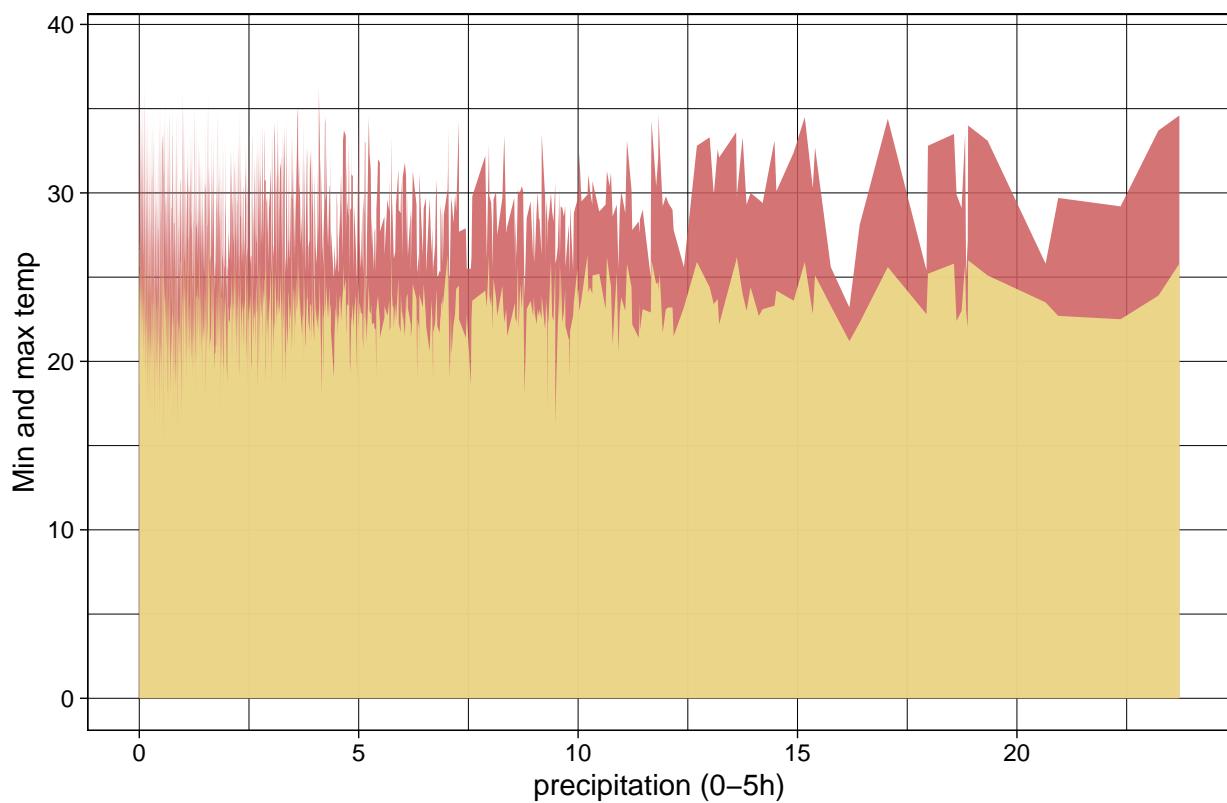


We can observe a slight negative correlation, in particular between the next-day maximum temperature forecast and the next-day forecast of the level of cloud cover.

The following plots show the relation between temperature forecasts and the model forecasts of next-day average precipitations (in %):

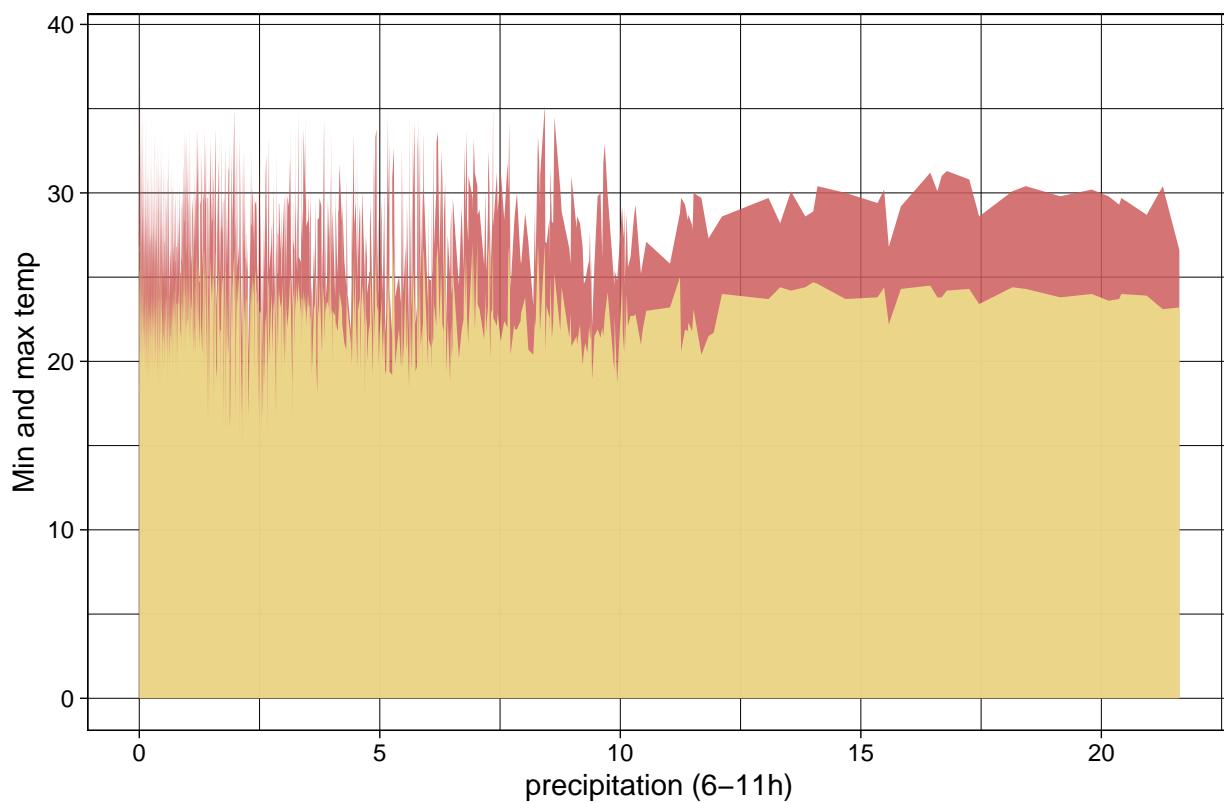
```
par(mfrow = c(1,4), mar = c(4,4,2,1), bty = "l")
ggplot(data) +
  geom_area(aes(x = data$PPT1, y = data$Next_Tmax),
            fill = 'indianred', alpha = 0.85) +
  geom_area(aes(x = data$PPT1, y = data$Next_Tmin),
            fill = 'khaki', alpha = 0.85) +
  labs(
    title = "Next-day maximum and minimum air temperatures based on precipitation",
    x="precipitation (0-5h)",
    y="Min and max temp"
  ) +
  theme_linedraw()
```

Next-day maximum and minimum air temperatures based on precipitation



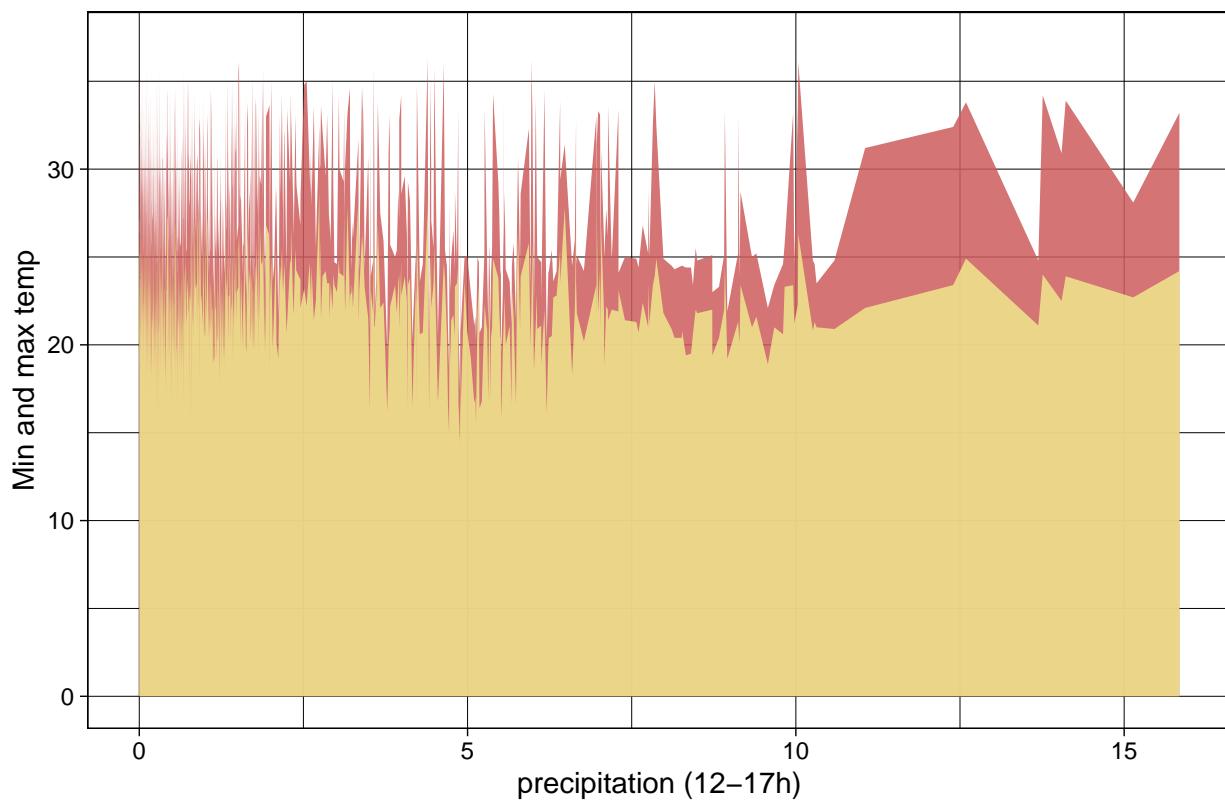
```
ggplot(data) +
  geom_area(aes(x = data$PPT2, y = data$Next_Tmax),
            fill = 'indianred', alpha = 0.85) +
  geom_area(aes(x = data$PPT2, y = data$Next_Tmin),
            fill = 'khaki', alpha = 0.85) +
  labs(
    title = "Next-day maximum and minimum air temperatures based on precipitation",
    x="precipitation (6-11h)",
    y="Min and max temp"
  ) +
  theme_linedraw()
```

### Next-day maximum and minimum air temperatures based on precipitation



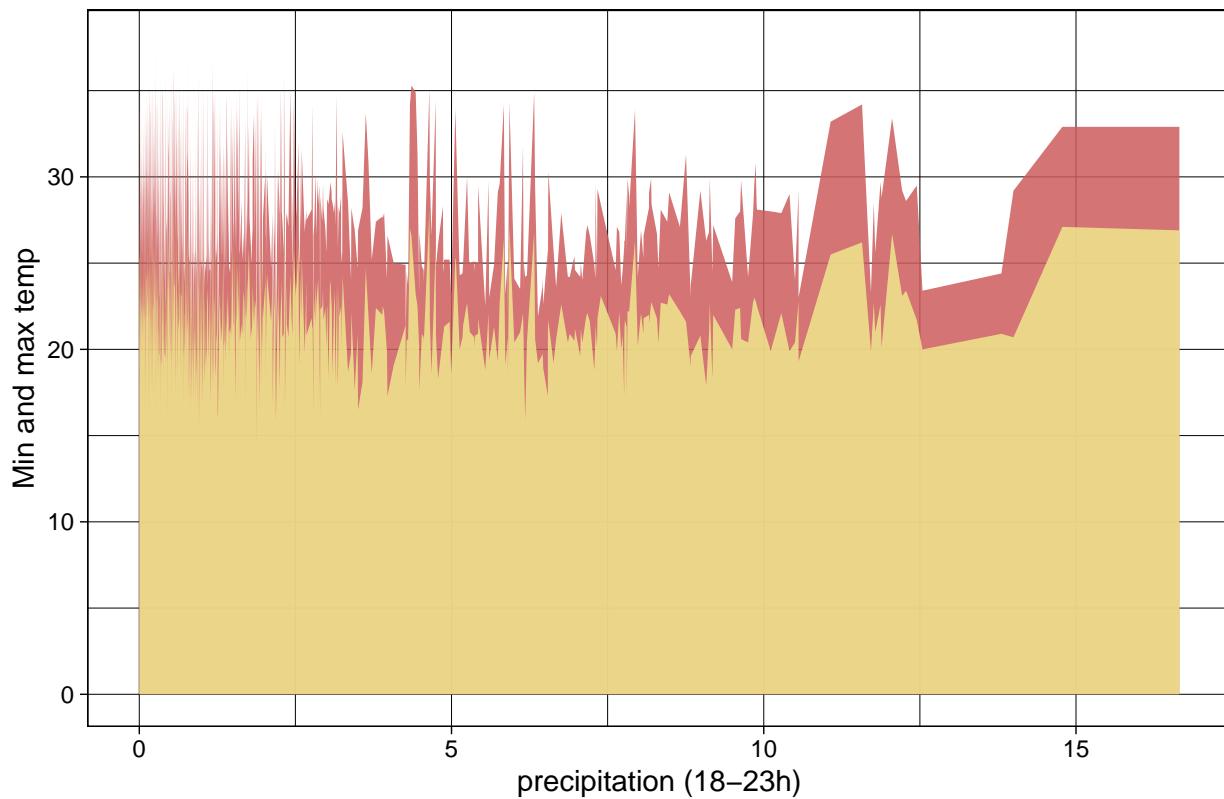
```
ggplot(data) +
  geom_area(aes(x = data$PPT3, y = data$Next_Tmax),
            fill = 'indianred', alpha = 0.85) +
  geom_area(aes(x = data$PPT3, y = data$Next_Tmin),
            fill = 'khaki', alpha = 0.85) +
  labs(
    title = "Next-day maximum and minimum air temperatures based on precipitation",
    x="precipitation (12-17h)",
    y="Min and max temp"
  ) +
  theme_linedraw()
```

### Next-day maximum and minimum air temperatures based on precipitation



```
ggplot(data) +
  geom_area(aes(x = data$PPT4, y = data$Next_Tmax),
            fill = 'indianred', alpha = 0.85) +
  geom_area(aes(x = data$PPT4, y = data$Next_Tmin),
            fill = 'khaki', alpha = 0.85) +
  labs(
    title = "Next-day maximum and minimum air temperatures based on precipitation",
    x="precipitation (18-23h)",
    y="Min and max temp"
  ) +
  theme_linedraw()
```

## Next-day maximum and minimum air temperatures based on precipitation

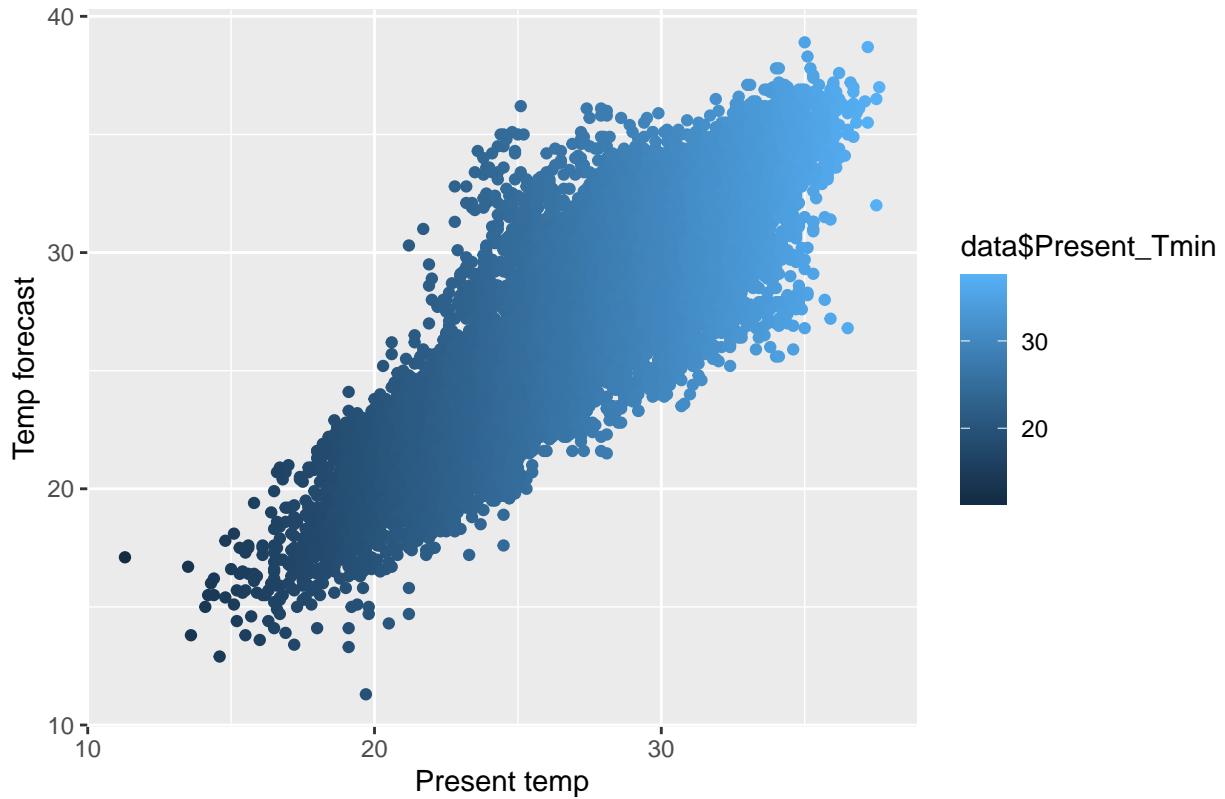


the results show a positive correlation between air temperature and precipitation. As the level of precipitation increases, the variability of the temperature forecasts also decreases.

The following plot shows the relation between the present minimum and maximum air temperature and the next-day max and min temperatures:

```
ggplot(data) +
  geom_point(aes(x = data$Present_Tmin, y = data$Next_Tmin, color=data$Present_Tmin))+
  geom_point(aes(x = data$Present_Tmax, y = data$Next_Tmax, color= data$Present_Tmax))+  
  labs(  
    title = "Next-day air temperatures based on present temperatures",  
    x="Present temp",  
    y="Temp forecast"  
)
```

## Next-day air temperatures based on present temperatures

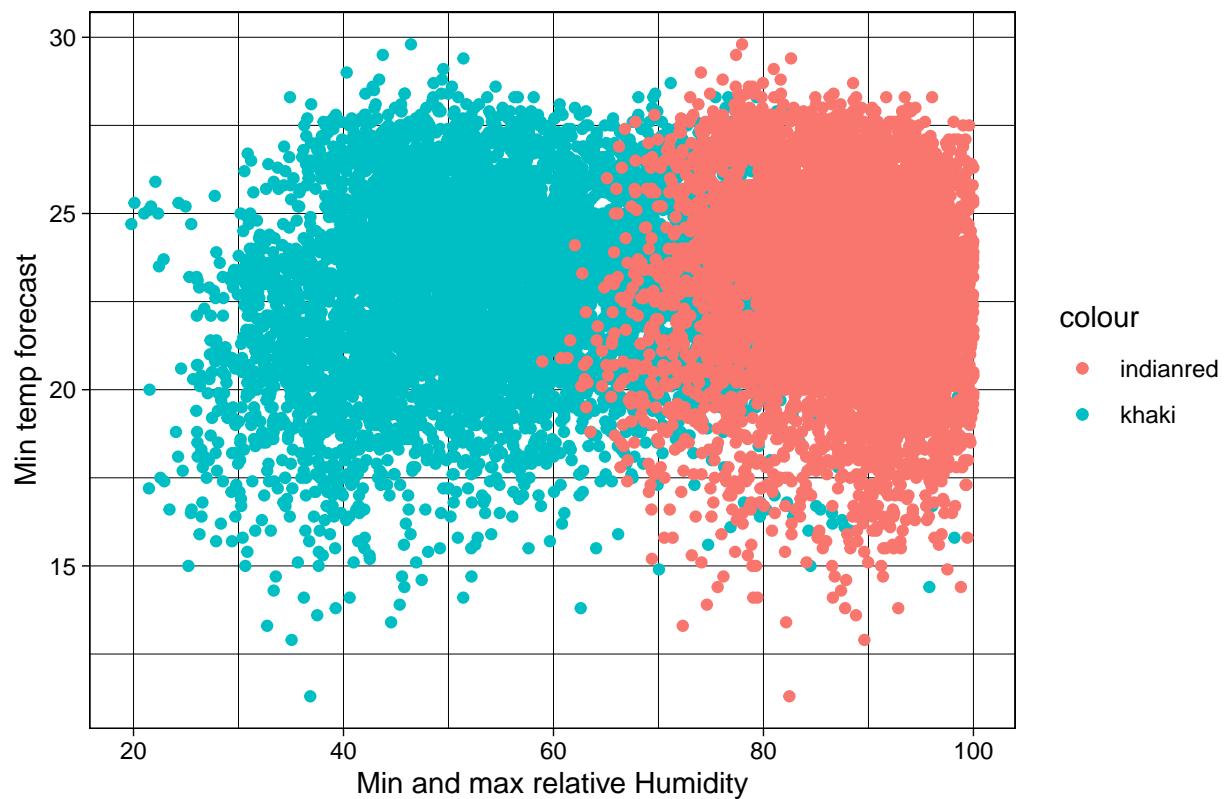


There is a clear positive correlation between present and future temperatures.

The following plot shows the relation between the model forecast of next-day minimum and maximum relative humidity and the next-day max and min temperatures:

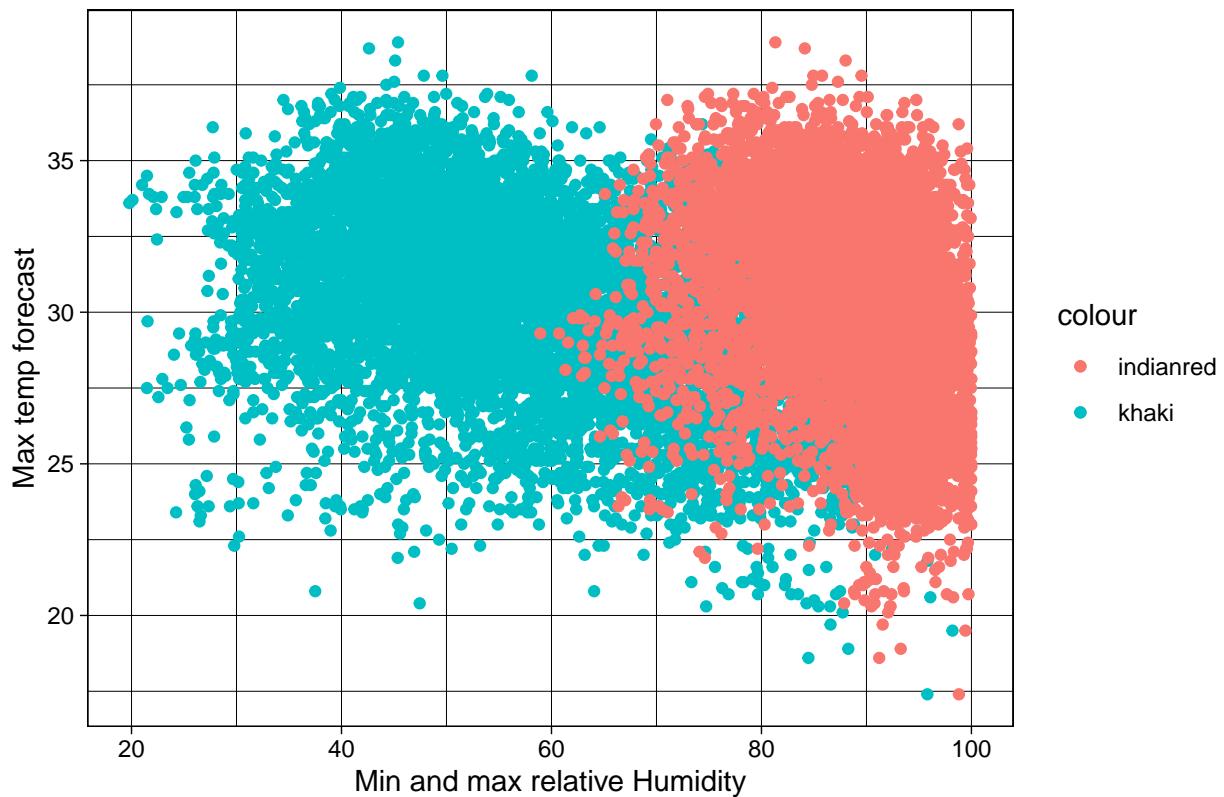
```
par(mfrow = c(1,2), mar = c(4,4,2,1), bty = "1")
ggplot(data) +
  geom_point(aes(x = data$RHmin, y = data$Next_Tmin, color="khaki"))+
  geom_point(aes(x = data$RHmax, y = data$Next_Tmin, color= "indianred"))+
  labs(
    title = "Next-day min air temperature based on relative humidity forecasts",
    x="Min and max relative Humidity",
    y="Min temp forecast"
  )+
  theme_linedraw()
```

### Next-day min air temperature based on relative humidity forecasts



```
ggplot(data) +  
  geom_point(aes(x = data$RHmin, y = data$Next_Tmax, color="khaki"))+  
  geom_point(aes(x = data$RHmax, y = data$Next_Tmax, color= "indianred"))+  
  labs(  
    title = "Next-day max air temperatures based on relative humidity forecasts",  
    x="Min and max relative Humidity",  
    y="Max temp forecast"  
) +  
  theme_linedraw()
```

## Next-day max air temperatures based on relative humidity forecasts

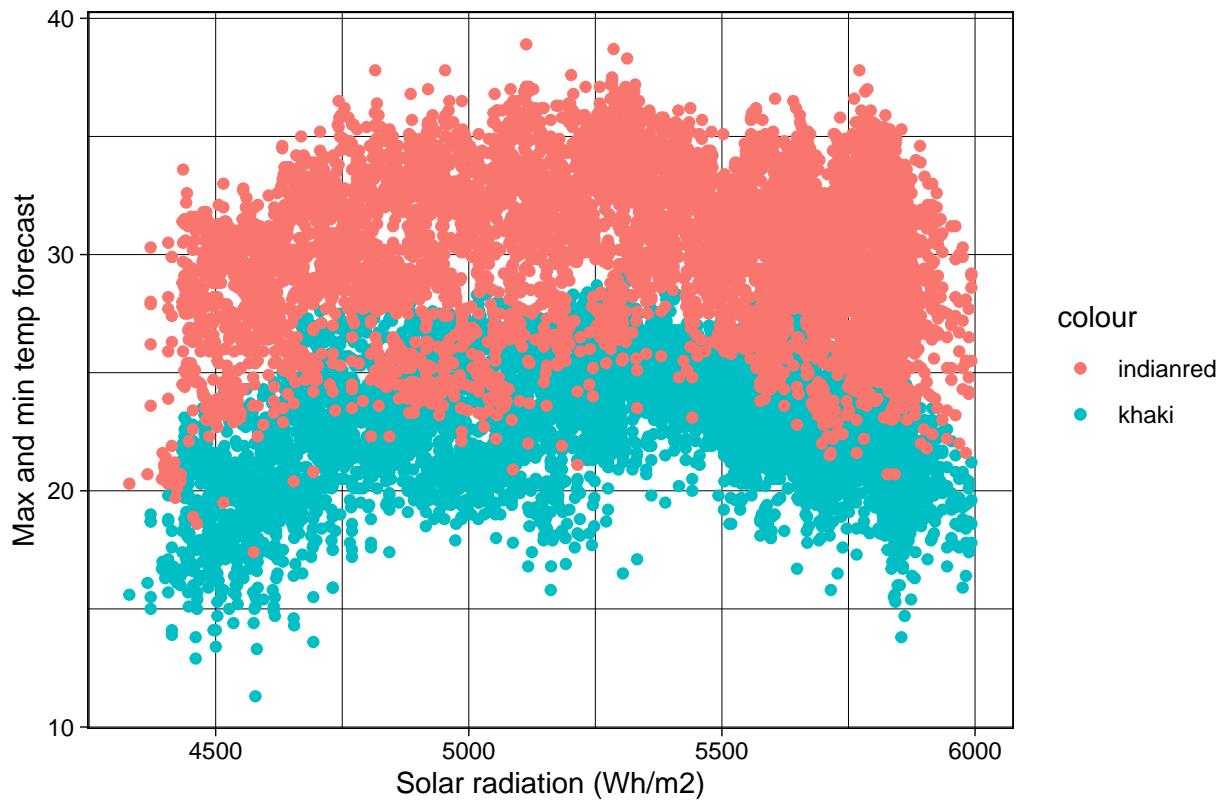


The correlation seems weak, but the plot between relative humidity and max temperature forecast appears to be negative.

Next, we can observe how incoming solar radiation ( $\text{Wh/m}^2$ ) affects the next day temperature forecasts:

```
ggplot(data) +
  geom_point(aes(x = data$Solar.radiation, y = data$Next_Tmin, color="khaki"))+
  geom_point(aes(x = data$Solar.radiation, y = data$Next_Tmax, color= "indianred"))+
  labs(
    title = "Next-day max air temperatures based on incoming solar radiation",
    x="Solar radiation (Wh/m2)",
    y="Max and min temp forecast"
  )+
  theme_linedraw()
```

Next-day max air temperatures based on incoming solar radiation



The relation between the two variables seems to be non-linear: temperature and solar radiation are positively correlated until a solar radiation between 5000 and 5500 Wh/m<sup>2</sup>, but then the data starts to show a negative correlation.

#### 1.4 Issues of validity, reliability and sample selection

The main issue of this data set is the number of missing data, which can affect the overall quality of the analysis. Other than this, the sampling has been done over a long period of time and in different locations, so to avoid bias selection.