Introduction

weather_features.csv contains hourly weather data from 2015 to 2018 for 5 major cities in Spain.

```
weather_features <- read.csv("weather_features.csv")</pre>
df <- data.frame(</pre>
  Column = colnames(weather_features),
  Description = c("datetime index localized to CET",
                  "name of city (Barcelona, Bilbao, Madrid, Seville, Valencia)",
                  "temperature (K)",
                  "minimum temperature (K)",
                  "maximum temperature (K)",
                  "pressure (hPa)",
                  "humidity (%)",
                  "wind speed (m/s)",
                  "wind direction (degrees)",
                  "rain in last hour (mm)",
                  "rain in last 3 hours (mm)",
                  "snow in last 3 hours (mm)",
                  "cloud cover (%)",
                  "weather description - code",
                  "weather description - short",
                  "weather description - long",
                  "weather icon")
knitr::kable(df, format = "markdown")
```

Column	Description
$\overline{\mathrm{dt}}$ iso	datetime index localized to CET
city_name	name of city (Barcelona, Bilbao, Madrid, Seville, Valencia)
temp	temperature (K)
temp_min	minimum temperature (K)
temp_max	maximum temperature (K)
pressure	pressure (hPa)
humidity	humidity (%)
wind_speed	wind speed (m/s)
wind_deg	wind direction (degrees)
rain_1h	rain in last hour (mm)
rain_3h	rain in last 3 hours (mm)
$snow_3h$	snow in last 3 hours (mm)
$clouds_all$	cloud cover (%)
weather_id	weather description - code
weather_main	weather description - short
weather_description	weather description - long
weather_icon	weather icon

Missing Values

```
sapply(weather_features, function(col)
{ sum(sapply(col, function(x) {
   (is.na(x)) + (x == "")
})) })
```

##	dt iso	city_name	temp	temp_min
##	0	0	0	0
##	temp_max	pressure	humidity	wind_speed
##	0	0	0	0
##	wind_deg	rain_1h	rain_3h	snow_3h
##	0	0	0	0
##	clouds_all	${\tt weather_id}$	weather_main	weather_description
##	0	0	0	0
##	weather_icon			
##	0			

Data has no blank or NA cells in any of the columns. The last 4 columns are non-numerical, and will be dropped. The data for the city of Barcelona has city_name "Barcelona", the extra space will be removed. The datetime index will be switched to date format.

```
weather_features <- weather_features[1:13]
weather_features[weather_features$city_name == "Barcelona",]$city_name = "Barcelona"
weather_features$dt_iso <- as.POSIXct(weather_features$dt_iso)</pre>
```

Next, we check that there are no missing rows.

Since we have hourly data for 3 regular and 1 leap year, we expect there to be $24 \times 365 \times 4 + 24 = 35064$ unique datetime indices.

```
dates_occurences <- table(weather_features$dt_iso)
length(dates_occurences)</pre>
```

```
## [1] 1461
```

This matches our expectations.

Since we have 5 cities, we expect each datetime index to appear 5 times.

```
sum(sapply(dates_occurences, function(date) { date < 5 }))</pre>
```

```
## [1] 0
```

This indicates that we have no missing rows.

Partition Based on City

This means a total of $35064 \times 5 = 175320$ rows. However, our data has 178396 rows. This indicates the presence of duplicated rows.

To fix this, we first create separate data frames for each city, then remove duplicates.

```
library(dplyr)
## Warning: package 'dplyr' was built under R version 4.4.2
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
Barcelona <- weather_features[weather_features$city_name == "Barcelona", ] %>% distinct()
Bilbao <- weather_features[weather_features$city_name == "Bilbao", ] %>% distinct()
Madrid <- weather_features[weather_features$city_name == "Madrid", ] %>% distinct()
Seville <- weather_features[weather_features$city_name == "Seville", ] %>% distinct()
Valencia <- weather_features[weather_features$city_name == "Valencia", ] %>% distinct()
dim(Barcelona)
## [1] 33363
                13
dim(Bilbao)
## [1] 34177
                13
dim(Madrid)
## [1] 33221
                13
dim(Seville)
## [1] 32366
                13
dim(Valencia)
## [1] 33227
                13
```