

Energy Data Analysis with R

Reto Marek

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Preface

Welcome to the introduction of energy data analysis with R.

This documents gives you a short overview of the statistical software R and their capability of analyzing data sets in the context of building monitoring data and in general of time series.

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Chapter 1

Installing R and R Studio

- Before we can start the first analysis, we have to install “R” and “RStudio”.
- “R” is a programming language used for statistical computing while “RStudio” provides a graphical user interface.
- “R” may be used without “RStudio”, but “RStudio” may not be used without “R”. Both, “R” and “RStudio” are free of charge and there are no licence fees.

1.1 Download and install

Installation instructions according to (Grolemund and Wickham, 2015)

1.1.1 Windows

Installation according

1.1.2 Mac

1.1.3 Linux

Chapter 2

R Basics

2.1 Packages

2.1.1 Install from CRAN

- Close all projects in R Studio
- `install.packages("ggplot2")`

2.1.2 Install from github

```
install.packages("devtools")  
library(devtools)  
install_github("retomarek/redutils")
```

2.1.3 Loading

- `library(ggplot2)`

2.2 Importing data

2.2.1 csv file

```
df <- read.csv("datafile.csv")  
df <- read.csv("datafile.csv", header=FALSE, stringsAsFactors=FALSE)
```

```
df <- read.csv("https://github.com/retomarek/r/raw/master/datasets/buildingMonitoringT
```

Attention: By default, strings in the data are treated as factors. `read.csv()` is a convenience wrapper function around `read.table()`. If you need more control over the input, see `?read.table`

2.2.2 Excel File

```
# Only need to install once
install.packages("xlsx")

library(xlsx)

df <- read.xlsx("datafile.xlsx", 1)
df <- read.xlsx("datafile.xls", sheetIndex=2)
df <- read.xlsx("datafile.xls", sheetName="Revenues")
```

For reading older Excel files in the `.xls` format, the `gdata` package has the function `read.xls()`:

```
# Only need to install once
install.packages("gdata")

library(gdata)
# Read first sheet
df <- read.xls("datafile.xls")
df <- read.xls("datafile.xls", sheet=2)
```

Both the `xlsx` and `gdata` packages require other software to be installed on your computer. For `xlsx`, you need to install Java on your machine. For `gdata`, you need Perl, which comes as standard on Linux and Mac OS X, but not Windows. On Windows, you'll need ActiveState Perl. The Community Edition can be obtained for free.

2.3 Data manipulations

2.3.1 data frames

2.3.1.1 change row names of df

```
names(df) <- c("Column1", "Column2", "Column3")
```

2.3.2 wide to long

```
# wide format
head(df)
```

```
##           time WthStnPress WthStnHum WthStnRain WthStnSolRad
## 1 2018-09-30T22:00:00.000Z    1012.30      87.0        0.8          0
## 2 2018-09-30T23:00:00.000Z    1011.90      87.5        1.1          0
## 3 2018-10-01T00:00:00.000Z    1011.45      87.5        0.5          0
## 4 2018-10-01T01:00:00.000Z    1010.90      86.5        0.5          0
## 5 2018-10-01T02:00:00.000Z    1010.55      88.0        0.6          0
## 6 2018-10-01T03:00:00.000Z    1010.20      89.0        0.1          0
##   WthStnTemp WthStnWindDir WthStnWindSpd BldgEnergyHotwater BldgEnergyHeating
## 1      12.80        157.50          3.2              0              0
## 2      12.35         11.25          1.6             19              0
## 3      11.90        146.25          2.4              0              0
## 4      11.90        157.50          0.8              0              0
## 5      11.60        146.25          2.4              0              0
## 6      11.75         22.50          0.8              0              0
##   FlatHum FlatTemp FlatVolFlowColdwater FlatVolFlowHotwater
## 1      NA      NA              0.006              0
## 2      NA      NA              0.000              0
## 3      NA      NA              0.000              0
## 4      NA      NA              0.000              0
## 5      NA      NA              0.006              0
## 6      NA      NA              0.000              0
```

```
# convert wide to long format
df.long <- as.data.frame(tidyr::pivot_longer(df,
                                             cols = -time,
                                             names_to = "name",
                                             values_to = "value",
                                             values_drop_na = TRUE)
)

# long format
head(df.long)
```

```
##           time           name  value
## 1 2018-09-30T22:00:00.000Z WthStnPress 1012.3
```

```
## 2 2018-09-30T22:00:00.000Z    WthStnHum    87.0
## 3 2018-09-30T22:00:00.000Z    WthStnRain    0.8
## 4 2018-09-30T22:00:00.000Z    WthStnSolRad    0.0
## 5 2018-09-30T22:00:00.000Z    WthStnTemp    12.8
## 6 2018-09-30T22:00:00.000Z    WthStnWindDir    157.5
```

2.3.3 long to wide

```
# long format
head(df.long)
```

```
##           time           name  value
## 1 2018-09-30T22:00:00.000Z    WthStnPress 1012.3
## 2 2018-09-30T22:00:00.000Z    WthStnHum    87.0
## 3 2018-09-30T22:00:00.000Z    WthStnRain    0.8
## 4 2018-09-30T22:00:00.000Z    WthStnSolRad    0.0
## 5 2018-09-30T22:00:00.000Z    WthStnTemp    12.8
## 6 2018-09-30T22:00:00.000Z    WthStnWindDir    157.5
```

```
# convert long table into wide table
df.wide <- as.data.frame(tidyr::pivot_wider(df.long,
                                             names_from = "name",
                                             values_from = "value")
                        )
```

```
# wide format
head(df.wide)
```

```
##           time WthStnPress WthStnHum WthStnRain WthStnSolRad
## 1 2018-09-30T22:00:00.000Z    1012.30    87.0    0.8    0
## 2 2018-09-30T23:00:00.000Z    1011.90    87.5    1.1    0
## 3 2018-10-01T00:00:00.000Z    1011.45    87.5    0.5    0
## 4 2018-10-01T01:00:00.000Z    1010.90    86.5    0.5    0
## 5 2018-10-01T02:00:00.000Z    1010.55    88.0    0.6    0
## 6 2018-10-01T03:00:00.000Z    1010.20    89.0    0.1    0
##   WthStnTemp WthStnWindDir WthStnWindSpd BldgEnergyHotwater BldgEnergyHeating
## 1    12.80    157.50    3.2    0    0
## 2    12.35    11.25    1.6    19    0
## 3    11.90    146.25    2.4    0    0
## 4    11.90    157.50    0.8    0    0
## 5    11.60    146.25    2.4    0    0
## 6    11.75    22.50    0.8    0    0
##   FlatVolFlowColdwater FlatVolFlowHotwater FlatHum FlatTemp
```

## 1	0.006	0	NA	NA
## 2	0.000	0	NA	NA
## 3	0.000	0	NA	NA
## 4	0.000	0	NA	NA
## 5	0.006	0	NA	NA
## 6	0.000	0	NA	NA

Chapter 3

Explorative Data Analysis

3.1 Get overview

Get an overview of the whole data set and specific series of it

3.1.1 Load data

Load test data set in a data frame (e.g. from a csv-file)

```
df <- read.csv("https://github.com/retomarek/r/raw/master/datasets/buildingMonitoringTestDataSet.csv")
```

3.1.2 Names

show the column headers of the data frame

```
names(df)
```

```
## [1] "time" "WthStnPress" "WthStnHum"
## [4] "WthStnRain" "WthStnSolRad" "WthStnTemp"
## [7] "WthStnWindDir" "WthStnWindSpd" "BldgEnergyHotwater"
## [10] "BldgEnergyHeating" "FlatHum" "FlatTemp"
## [13] "FlatVolFlowColdwater" "FlatVolFlowHotwater"
```

3.1.3 Structure

show the structure of the data frame

```
str(df)
```

```
## 'data.frame':    16394 obs. of  14 variables:
## $ time           : chr  "2018-09-30T22:00:00.000Z" "2018-09-30T23:00:00.000Z"
## $ WthStnPress    : num  1012 1012 1011 1011 1011 ...
## $ WthStnHum      : num  87 87.5 87.5 86.5 88 89 86.5 81 78 80.5 ...
## $ WthStnRain     : num  0.8 1.1 0.5 0.5 0.6 0.1 0.2 0 0 0 ...
## $ WthStnSolRad   : num  0 0 0 0 0 0 0 3 24.5 ...
## $ WthStnTemp     : num  12.8 12.4 11.9 11.9 11.6 ...
## $ WthStnWindDir  : num  157.5 11.2 146.2 157.5 146.2 ...
## $ WthStnWindSpd  : num  3.2 1.6 2.4 0.8 2.4 0.8 0.8 3.2 4 3.2 ...
## $ BldgEnergyHotwater : num  0 19 0 0 0 ...
## $ BldgEnergyHeating : num  0 0 0 0 0 0 0 0 0 ...
## $ FlatHum        : num  NA NA NA NA NA NA NA NA NA NA ...
## $ FlatTemp       : num  NA NA NA NA NA NA NA NA NA NA ...
## $ FlatVolFlowColdwater: num  0.006 0 0 0 0.006 ...
## $ FlatVolFlowHotwater : num  0 0 0 0 0 ...
```

3.1.4 Head/Tail

```
head(df)
```

```
##           time WthStnPress WthStnHum WthStnRain WthStnSolRad
## 1 2018-09-30T22:00:00.000Z    1012.30      87.0        0.8          0
## 2 2018-09-30T23:00:00.000Z    1011.90      87.5        1.1          0
## 3 2018-10-01T00:00:00.000Z    1011.45      87.5        0.5          0
## 4 2018-10-01T01:00:00.000Z    1010.90      86.5        0.5          0
## 5 2018-10-01T02:00:00.000Z    1010.55      88.0        0.6          0
## 6 2018-10-01T03:00:00.000Z    1010.20      89.0        0.1          0
##   WthStnTemp WthStnWindDir WthStnWindSpd BldgEnergyHotwater BldgEnergyHeating
## 1      12.80      157.50          3.2          0              0
## 2      12.35       11.25          1.6          19              0
## 3      11.90      146.25          2.4          0              0
## 4      11.90      157.50          0.8          0              0
## 5      11.60      146.25          2.4          0              0
## 6      11.75       22.50          0.8          0              0
##   FlatHum FlatTemp FlatVolFlowColdwater FlatVolFlowHotwater
## 1      NA      NA          0.006          0
## 2      NA      NA          0.000          0
## 3      NA      NA          0.000          0
## 4      NA      NA          0.000          0
## 5      NA      NA          0.006          0
## 6      NA      NA          0.000          0
```



```
tail(df)
```

```
##                                time WthStnPress WthStnHum WthStnRain WthStnSolRad
## 16389 2020-08-13T18:00:00.000Z    1011.650     74.75    2.19964          9
## 16390 2020-08-13T19:00:00.000Z    1012.000     79.00    2.19964          0
## 16391 2020-08-13T20:00:00.000Z    1011.950     78.25    2.19964          0
## 16392 2020-08-13T21:00:00.000Z    1012.025     76.50    2.19964          0
## 16393 2020-08-13T22:00:00.000Z    1012.250     73.00    0.00000          0
## 16394 2020-08-13T23:00:00.000Z          NA          NA          NA          NA
##           WthStnTemp WthStnWindDir WthStnWindSpd BldgEnergyHotwater
## 16389      22.000      162.00      0.000000      NA
## 16390      20.175      124.25      1.609340      NA
## 16391      19.350      125.00      0.402335      NA
## 16392      19.900       93.00      1.609340      NA
## 16393      20.625      116.25      2.414010      NA
## 16394         NA         NA         NA      NA
##           BldgEnergyHeating FlatHum FlatTemp FlatVolFlowColdwater
## 16389              NA      NA      NA      NA
## 16390              NA      NA      NA      NA
## 16391              NA      NA      NA      NA
## 16392              NA      NA      NA      NA
## 16393              NA      NA      NA      NA
## 16394              NA      NA      NA      NA
##           FlatVolFlowHotwater
## 16389              NA
## 16390              NA
## 16391              NA
## 16392              NA
## 16393              NA
## 16394              NA
```

3.1.5 Five number summary

reveals details of a specific series

```
summary(df$WthStnTemp)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.    NA's
##    -5.25   5.50   11.25   11.99   17.35   40.30     12
```

3.2 Data Wrangling

3.2.1 season from date

```
# install redutils library
# devtools::install_github("retomarek/redutils", ref = "master")

# get season from a date
redutils::season(as.Date("2019-04-01"))

## [1] "Spring"

redutils::season(as.Date("2019-04-01"), c("Winter", "Frühling", "Sommer", "Herbst"))

## [1] "Frühling"

# apply it for a data frame
df.season <- dplyr::mutate(df, season = redutils::season(df$time))
head(df.season)
```

##		time	WthStnPress	WthStnHum	WthStnRain	WthStnSolRad
## 1	2018-09-30T22:00:00.000Z	1012.30	87.0	0.8	0	
## 2	2018-09-30T23:00:00.000Z	1011.90	87.5	1.1	0	
## 3	2018-10-01T00:00:00.000Z	1011.45	87.5	0.5	0	
## 4	2018-10-01T01:00:00.000Z	1010.90	86.5	0.5	0	
## 5	2018-10-01T02:00:00.000Z	1010.55	88.0	0.6	0	
## 6	2018-10-01T03:00:00.000Z	1010.20	89.0	0.1	0	
##		WthStnTemp	WthStnWindDir	WthStnWindSpd	BldgEnergyHotwater	BldgEnergyHeating
## 1		12.80	157.50	3.2	0	0
## 2		12.35	11.25	1.6	19	0
## 3		11.90	146.25	2.4	0	0
## 4		11.90	157.50	0.8	0	0
## 5		11.60	146.25	2.4	0	0
## 6		11.75	22.50	0.8	0	0
##		FlatHum	FlatTemp	FlatVolFlowColdwater	FlatVolFlowHotwater	season
## 1		NA	NA	0.006	0	Fall
## 2		NA	NA	0.000	0	Fall
## 3		NA	NA	0.000	0	Fall
## 4		NA	NA	0.000	0	Fall
## 5		NA	NA	0.006	0	Fall
## 6		NA	NA	0.000	0	Fall

Chapter 4

Data Visualizations

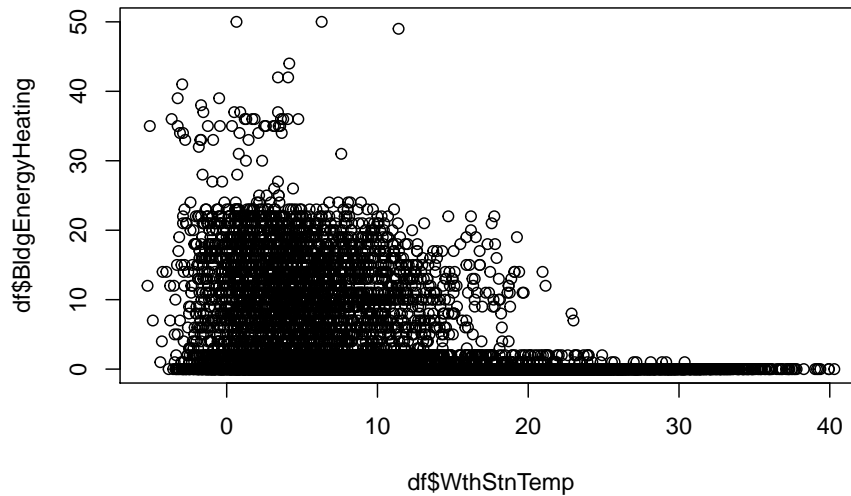
4.1 General Plots

4.1.1 Scatterplot

4.1.1.1 plot()

```
# load data set
df <- read.csv("https://github.com/retomarek/r/raw/master/datasets/buildingMonitoringTestDataSet.csv")

# crate simple scatterplot
plot(df$WthStnTemp, df$BldgEnergyHeating)
```



4.2 Building Energy Signature

4.2.1 static

```
library(ggplot2)
library(plotly)
library(dplyr)
library(reduutils)
library(lubridate)

# load data set
df <- read.csv("https://github.com/retomarek/r/raw/master/datasets/buildingMonitoringT
               stringsAsFactors=FALSE, sep = ",")

# select data and calculate season
data <- df %>%
  select(time, WthStnTemp, BldgEnergyHeating) %>%
  mutate(season = reduutils::season(df$time)) %>%
  na.omit()

# Aggregate data to daily values
data$time <- parse_date_time(data$time, "YmdHMS", tz = "Europe/Zurich")
```

```

data$year <- as.Date(cut(data$time, breaks = "year"))
data$month <- as.Date(cut(data$time, breaks = "month"))
data$day <- as.Date(cut(data$time, breaks = "day"))

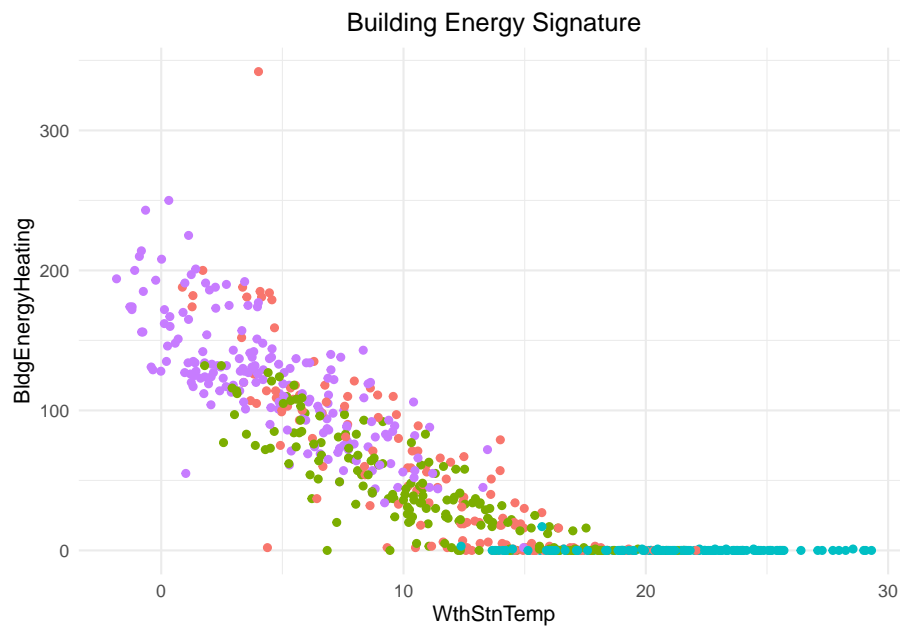
data <- data %>%
  select(day, WthStnTemp, BldgEnergyHeating, season) %>%
  group_by(day) %>%
  mutate(WthStnTemp = mean(WthStnTemp))

data <- data %>%
  group_by(day) %>%
  mutate(BldgEnergyHeating = sum(BldgEnergyHeating))

data <- data %>%
  unique()

# static chart with ggplot
p <- ggplot2::ggplot(data) +
  ggplot2::geom_point(aes(x = WthStnTemp,
                          y = BldgEnergyHeating, color=season,
                          text = paste("</br>Date: ", as.Date(data$day),
                                       "</br>Temp: ", round(data$WthStnTemp, digits = 1), "\u00B0C",
                                       "</br>Energy: ", round(data$BldgEnergyHeating, digits = 0), "kWh",
                                       "</br>Season: ", data$season)))
  ) +
  ggtitle("Building Energy Signature") +
  theme_minimal() +
  theme(
    legend.position="none",
    plot.title = element_text(hjust = 0.5)
  )
p

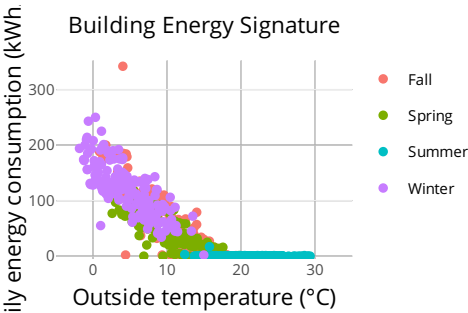
```



4.2.2 interactive

Make ggplot2 chart above interactive with plotly

```
# continuation from upper ggplot code section
plotly::ggplotly(p, tooltip = c("text")) %>%
  layout(xaxis = list(title = "Outside temperature (\u00B0C)",
    range = c(min(-5,min(data$WthStnTemp)), max(35,max(data$WthStnTemp))),
    yaxis = list(title = "Daily energy consumption (kWh/d)",
    range = c(-5, max(data$BldgEnergyHeating) + 10)),
    showlegend = TRUE
  ) %>%
  plotly::config(displayModeBar = FALSE, displaylogo = FALSE)
```



Bibliography

Grolemund, G. and Wickham, H. (2015). *Hands-on programming with R*.
O'Reilly, Sebastopol, second release edition.