A4 - Data Processing

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FIRST THINGS TO DO

Don't try to kiss your data on the first date; rather, you just want to get to know the data:

1. Import the data
2. Review the codebook
3. **Learn about the data**
4. Quick (visual) understanding of the data

LEARN ABOUT THE DATA

So what are the first things we want to know about our data?

- dimensions
- data types (i.e. character, integer, factor, etc.)
- missing values
- summary statistics

What are some functions to extract this information?

LEARN ABOUT THE DATA

- So what are the first things we want to know about our data?
- dimensions: `dim()`, `ncol()`, `nrow()`, `names()`
- data types: `str()`, `class()`, `is.`, `as.`
- missing values: `is.na()`, `sum(is.na())`, `colSums(is.na())`
- summary statistics: `summary()`, `quantile()`, `var()`, `sd()`, `table()`

Data Frames

Example data:

```
install.packages("AmesHousing")
```

```
ames_data <- AmesHousing::make_ames()
```

```
typeof(ames_data)
```

```
## [1] "list"
```

```
head(names(ames_data))
```

```
## [1] "Lot_Area"      "Street"        "Alley"         "Lot_Shape"
```

Transfer to data.frame

- Transfer data to a data.frame:

```
ames_df <- data.frame(ames_data)
```

Number of rows/columns

- Find out the number of rows/columns

```
nrow(ames_df) # rows
```

```
## [1] 2930
```

```
ncol(ames_df) # columns
```

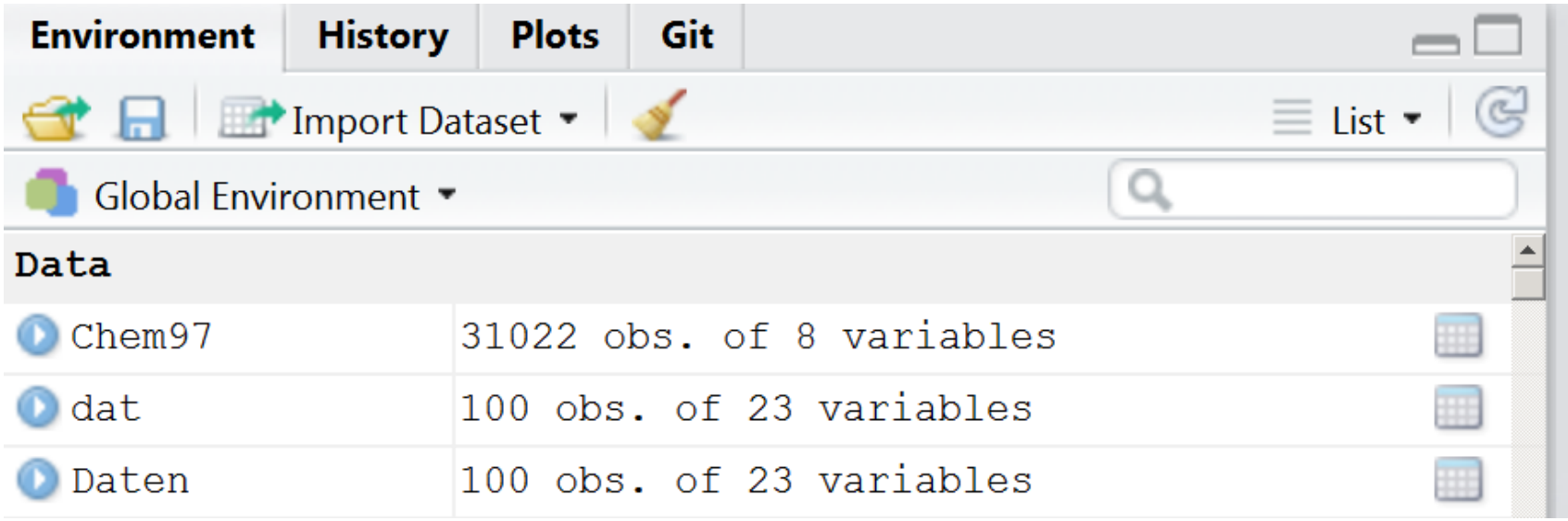
```
## [1] 81
```

View the data

- See some lines:

```
head(ames_df) # first lines
tail(ames_df) # last lines
```

- Overview with Rstudio:



Accessing Columns with the dollar sign

- The dollar sign can also be used to address individual columns

```
head(ames_df$Lot_Area)
```

```
## [1] 31770 11622 14267 11160 13830  9978
```

```
ames_df$Lot_Area[1:6]
```

```
## [1] 31770 11622 14267 11160 13830  9978
```

Accessing Columns with the number or the name

```
head(ames_df[,5])
head(ames_df[,"Street"]) # the same result
```

Subsetting dataset

```
Street <- ames_df$Street
table(Street)
```

```
## Street
## Grvl Pave
##   12 2918
```

```
ames_df[Street=="Grvl",]
# same result:
ames_df[Street!="Pave",]
```

Logical operations in indexing

```
head(ames_df[ames_df$Lot_Area>9000,1:4])
```

```
##                                MS_SubClass                 MS_Zonin
## 1 One_Story_1946_and_Newer_All_Styles  Residential_Low_Densit
## 2 One_Story_1946_and_Newer_All_Styles Residential_High_Densit
## 3 One_Story_1946_and_Newer_All_Styles  Residential_Low_Densit
## 4 One_Story_1946_and_Newer_All_Styles  Residential_Low_Densit
## 5           Two_Story_1946_and_Newer  Residential_Low_Densit
## 6           Two_Story_1946_and_Newer  Residential_Low_Densit
##   Lot_Area
## 1    31770
## 2    11622
## 3    14267
## 4    11160
## 5    13830
## 6     9978
```

Exercise: Vectors and Indexing

Assume that we have registered the height and weight for four people: Heights in cm are 180, 165, 160, 193; weights in kg are 87, 58, 65, 100. Make two vectors, height and weight, with the data. The bodymass index (BMI) is defined as

$$\frac{\text{weight in kg}}{(\text{height in m})^2}$$

Make a vector with the BMI values for the four people, and a vector with the natural logarithm to the BMI values. Finally make a vector with the weights for those people who have a BMI larger than 25.

The airquality data

```
data(airquality)
Ozone <- airquality$Ozone
```

airquality {datasets} R Documentation

New York Air Quality Measurements

Description

Daily air quality measurements in New York, May to September 1973.

Usage

```
airquality
```

Format

A data frame with 154 observations on 6 variables.

[,1] Ozone	numeric Ozone (ppb)
[,2] Solar.R	numeric Solar R (lang)
[,3] Wind	numeric Wind (mph)
[,4] Temp	numeric Temperature (degrees F)
[,5] Month	numeric Month (1–12)
[,6] Day	numeric Day of month (1–31)

Details

Daily readings of the following air quality values for May 1, 1973 (a Tuesday) to September 30, 1973.

Other important options

- save result to an object

```
subDat <- airquality[Ozone>30,]
```

- multiple conditions can be linked with &

```
airquality[Ozone>18 & airquality$Month==5,]
```

- the or argument - one of the two conditions must be fullfilled

```
airquality[Ozone>18 | airquality$Month==5,]
```

Missing values

- Missing values are defined as `NA` in R
- Math functions usually have a way to exclude missing values in their calculations.
- `mean()`, `median()`, `colSums()`, `var()`, `sd()`, `min()` and `max()` all take the `na.rm` argument.

```
mean(Ozone)
```

```
## [1] NA
```

```
mean(Ozone,na.rm=T)
```

```
## [1] 42.12931
```

Find the missing values:

```
head(is.na(Ozone))
```

```
## [1] FALSE FALSE FALSE FALSE  TRUE FALSE
```

```
which(is.na(Ozone))
```

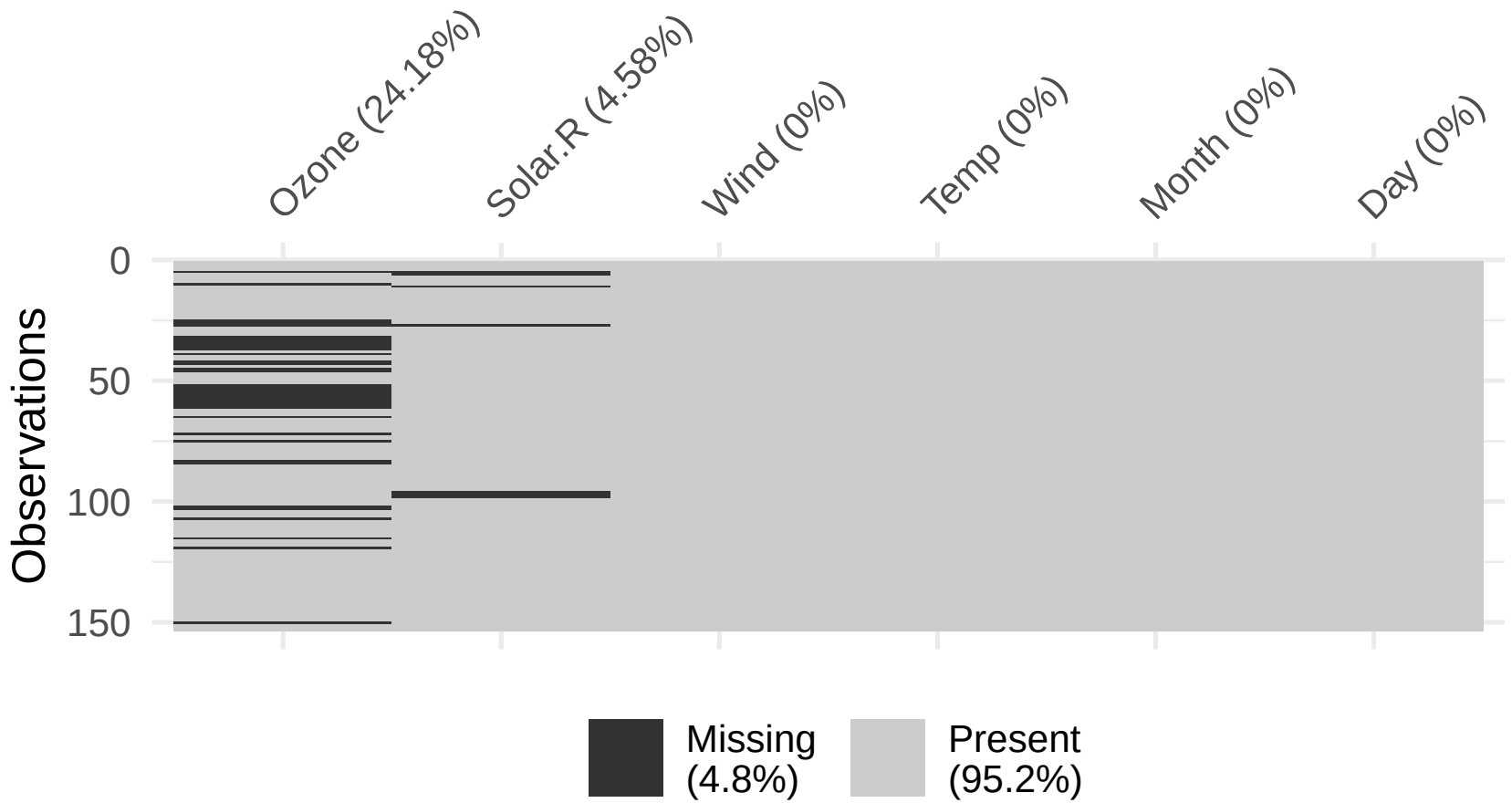
```
##   [1]   5  10  25  26  27  32  33  34  35  36  37  39  42  43
##  [20]  55  56  57  58  59  60  61  65  72  75  83  84 102 103
```

```
table(is.na(Ozone))
```

```
##
## FALSE  TRUE
##   116    37
```

Missing Data Visualisations

```
# Data Structures, Summaries, and Visualisations
# for Missing Data
library(naniar)
vis_miss(airquality)
```



The command complete.cases()

- returns a logical vector indicating which cases are complete.

```
nrow(airquality)
# list rows of data without missing values
airq_comp <- gpdat[complete.cases(airquality),]
nrow(airq_comp)
```

A shorthand alternative

- An shorthand alternative is to simply use `na.omit()` to omit all rows containing missing values.

```
airq_comp <- na.omit(airquality)
nrow(airq_comp)
```

```
## [1] 111
```

Very simple imputation

```
airq <- airquality
airq$Ozone[is.na(airq$Ozone)] <-mean(airq$Ozone,
                                     na.rm = T)
```

Comparing mean and variance

```
mean(airquality$Ozone, na.rm = T)
```

```
## [1] 42.12931
```

```
mean(airq$Ozone)
```

```
## [1] 42.12931
```

```
var(airquality$Ozone, na.rm = T)
```

```
## [1] 1088.201
```

```
var(airq$Ozone)
```

```
## [1] 823.3096
```

NAs per column

- For data frames, a convenient shortcut to compute the total missing values in each column is to use `colSums()`:

```
colSums(is.na(airquality))
```

```
##   Ozone Solar.R    Wind    Temp   Month     Day
##      37       7       0       0       0       0
```

```
colSums(is.na(airq))
```

```
##   Ozone Solar.R    Wind    Temp   Month     Day
##       0       7       0       0       0       0
```

CRAN Task View: Missing Data

CRAN Task View: Missing Data

Maintainer: Julie Josse, Nicholas Tierney and Nathalie Vialaneix (r-miss-tastic team)
Contact: r-miss-tastic at clementine.wf
Version: 2019-07-02
URL: https://CRAN.R-project.org/view=MissingData

Missing data are very frequently found in datasets. Base R provides a few options to handle them using computations that involve only observed data (`na.rm = TRUE` in functions `mean`, `var`, ... or `use = complete.obs|na.or.complete|pairwise.complete.obs` in functions `cov`, `cor`, ...). The base package stats also contains the generic function `na.action` that extracts information of the `NA` action used to create an object.

These basic options are complemented by many packages on CRAN, which we structure into main topics:

- Exploration of missing data
- Likelihood based approaches
- Single imputation
- Multiple imputation
- Weighting methods
- Specific types of data
- Specific application fields

Exercise: Missing values

1. How many missing values are in the built-in data set airquality?
2. Which variables are the missing values concentrated in?
3. How would you impute the mean or median for these values?
4. How would you omit all rows containing missing values?

Rename the column names

- With the command `colnames` you get the column names

```
colnames(airquality)
```

- We can rename the column names:

```
colnames(airquality)[1] <- "var1"
```

- The same applies to the row names

```
rownames(airquality)
```

The apply family

```
apply(airq,2,mean)
```

```
##      Ozone    Solar.R       Wind       Temp      Month        Day
## 42.129310         NA  9.957516 77.882353  6.993464 15.803922
```

```
# the following doesn`t make sense:
apply(airq,1,mean)
```

```
##   [1] 51.90000 40.16667 42.60000 68.91667       NA       NA 6
##   [9] 20.35000 54.78822       NA 61.28333 65.70000 64.31667 2
##  [17] 73.50000 30.40000 75.91667 25.28333 17.28333 74.60000 2
##  [25] 35.28822 68.67155       NA 24.66667 71.15000 76.28333 7
##  [33] 70.13822 62.70489 55.22155 61.12155 68.57155 43.45000 7
##  [41] 79.41667 70.33822 68.55489 46.66667 81.32155 79.27155 5
##  [49] 25.86667 40.25000 43.71667 50.40489 34.47155 40.45489 6
##  [57] 47.85489 34.23822 44.27155 33.33822 51.18822 83.35000 6
##  [65] 41.50489 56.43333 76.81667 76.68333 79.55000 80.45000 6
##  [73] 63.38333 52.98333 76.67155 28.55000 69.81667 70.88333 7
##  [81] 67.75000 21.98333 69.97155 76.77155 83.26667 76.00000 3
##  [89] 70.90000 75.73333 74.06667 73.53333 36.48333 22.96667 3
##  [97]       NA       NA 80.83333 72.38333 72.00000 63.78822 4
```

The command apply()

```
apply(airq,2,var)
```

```
##      Ozone    Solar.R       Wind       Temp      Month
## 823.309608         NA  12.411539  89.591331   2.006536  78.57
```

```
apply(airq,2,sd)
```

```
##     Ozone   Solar.R      Wind      Temp     Month       Day
## 28.693372        NA  3.523001  9.465270  1.416522  8.864520
```

```
apply(X = airq,MARGIN = 2,FUN = range)
```

```
##      Ozone Solar.R Wind Temp Month Day
## [1,]     1      NA  1.7   56     5   1
## [2,]   168      NA 20.7   97     9  31
```

The arguments of the command `apply()`

- If `MARGIN=1` the function `mean` is applied for rows,
- If `MARGIN=2` the function `mean` is applied for columns,
- Instead of `mean` you could also use `var`, `sd` or `length`.

Example command tapply()

```
tapply(airq$Wind, airq$Month,mean)
```

```
##         5         6         7         8         9
## 11.622581 10.266667  8.941935  8.793548 10.180000
```

- Other commands can also be used. . . .. also self-scripted commands

Exercise: using the tapply() command

- Calculate the average ozone value by month using the airquality dataset and the tapply command.

Links and resources

- **Tidy data** - the package `tidyr`
- Homepage for **the `tidyverse` collection**
- **Data wrangling with R and RStudio**
- Hadley Wickham - **Tidy Data**
- Hadley Wickham - **Advanced R**
- Colin Gillespie and Robin Lovelace **Efficient R programming**
- **Quick-R about missing values**
- **Recode missing values**