# Analysing

R for Data Science Basel R Bootcamp









February 2019

### What is analysing?

### **Create Groups**

Group data by certain variables

- For all males (sex == "male")
- For all people in placebo condition (condition == "placebo")

### Calculate summaries

- Count number of cases
- Calculate mean of age (mean(age))
- Calculate number of events (sum(events))

### **Bonus: Statistical Analyses**

- Simple hypothesis tests (t-test, correlation test)
- Generalised linear model (regression, ANOVA)

Raw data (First 5 out of 1,000 rows)

id	sex	education	income	happiness
1	male	SEK_III	6300	5
2	male	obligatory_school	10900	7
3	female	SEK_III	5100	7
4	male	SEK_III	4200	7
5	male	SEK_III	4000	5

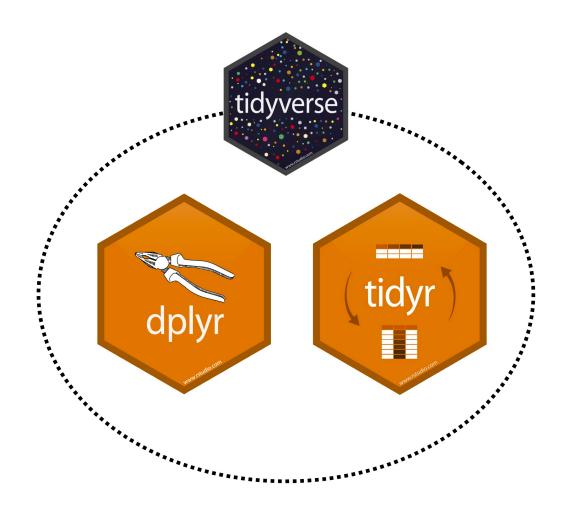
### Aggregated data

education	sex	N	Inc_mean	Hap_mean
apprenticeship	female	2168	7663.0	6.9
apprenticeship	male	1818	7388.9	6.9
obligatory_school	female	714	7746.1	6.9
obligatory_school	male	525	7293.7	6.8
SEK_II	female	469	7385.0	6.9
SEK_II	male	272	7254.7	6.9

# dplyr

To calculate grouped summary analyses, we will use dplyr (again!)

```
# Load packages individually
# install.packages('dplyr')
library(dplyr)
# Or just use the tidyverse!
# install.packages('tidyverse')
library(tidyverse)
```



### The Pipe! %>%

dplyr makes extensive use of a new operator called the "Pipe" %>%

Read the "Pipe" %>% as "And Then..."

```
# Start with data
data %>% # AND THEN...
DO_SOMETHING %>% # AND THEN...
DO_SOMETHING %>% # AND THEN...
DO_SOMETHING %>% # AND THEN...
```



This is not a pipe (but %>% is!)

# summarise()

Use summarise() to create new columns of summary statistics

```
df %>%
 summarise(
   NAME = SUMMARY_FUN(A),
   NAME = SUMMARY_FUN(B)
```

#### **Summary functions**

Function	Purpose
n()	Number of cases in each group
<pre>mean(), median(), max(), min() sum()</pre>	Summary stats

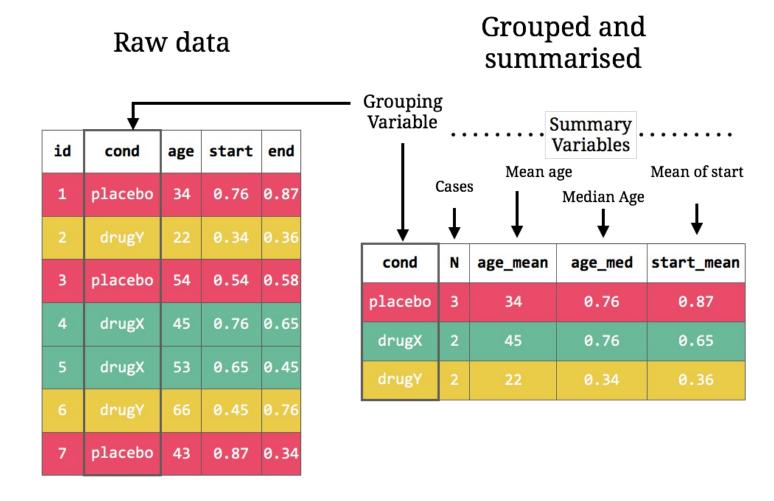
```
# Calculate summary statistics
baselers %>%
  summarise(
   N = n(),
    age_mean = mean(age),
    height_median = median(height),
    children_max = max(children, na.rm = TRUE)
```

```
## # A tibble: 1 x 4
        N age_mean height_median children_max
    <int>
              <dbl>
                            <dbl>
                                         <dbl>
              44.6
## 1 10000
                            171.
```

The result of summarise() will always be a tibble!

**Important** You can only include summary functions that return a single value (i.e.; can't use table())

### **Grouped Aggregation**



# group\_by(),summarise()

Use group\_by() to group data according to one or more columns

After grouping data, use summarise() to calculate summary statistics across groups of data

#### **Statistical functions**

Function	Purpose
n()	Number of cases in each group
<pre>mean(), median(), max(), min() sum()</pre>	Summary stats

```
# Group data by arm, and calculate many
# summary statistics
baselers %>%
  group_by(sex) %>%
  summarise(
   N = n()
   age_mean = mean(age),
   height_median = median(height),
   children_max = max(children)
```

```
## # A tibble: 2 x 5
               N age_mean height_median children_max
    <chr> <int>
                    <dbl>
                                  <dbl>
                                               <dbl>
## 1 female 5000
                     45.4
                                   164
                                                   6
                     43.8
## 2 male
            5000
                                   178.
                                                   6
```

### Combine wrangling with analysing

You can easily combine multiple wrangling (filtering, slicing, renaming) and analysing operations at once!

Just use the pipe %>%

```
baselers %>%
 filter(sex == "male" & children > 0) %>% # male parents only
  group_by(confession) %>%
  summarise(
   N = n(),
   age_mean = mean(age),
   income_median = median(income, na.rm = TRUE)
```

```
## # A tibble: 6 x 4
    confession
                             N age_mean income_median
    <chr>
                                   <dbl>
                          <int>
                                                 <dbl>
                                   43.5
## 1 <NA>
                           703
                                                  7000
## 2 catholic
                                   44.0
                                                  7100
                           1401
## 3 confessionless
                          1125
                                   43.8
                                                  7100
## 4 evangelical-reformed
                                   43.9
                                                  7200
## 5 muslim
                           155
                                   41.5
                                                  6800
## 6 other
                            247
                                    44.0
                                                  6900
```

### Here is part of the baselers dataframe

```
baselers %>%
 select(sex, fasnacht, age, income) %>%
 slice(1:5)
```

```
## # A tibble: 5 x 4
           fasnacht
    sex
                     age income
   <chr> <chr>
                   <dbl> <dbl>
                           6300
## 1 male
                      44
           no
## 2 male
                      65 10900
          no
## 3 female no
                      31
                           5100
                           4200
## 4 male
                           4000
## 5 male
```

```
## # A tibble: 2 x 4
    fasnacht
                  N age_mean income_mean
     <chr>
              <int>
                       <dbl>
                                   <dbl>
## 1 no
               9706
                        44.6
                                   7527.
                                   7692.
## 2 yes
                294
                        45.3
```

#### Here is part of the baselers dataframe

```
baselers %>%
 select(sex, fasnacht, age, income) %>%
 slice(1:5)
```

```
## # A tibble: 5 x 4
           fasnacht
                     age income
    sex
   <chr> <chr>
                   <dbl> <dbl>
                          6300
## 1 male
                      44
          no
                      65 10900
## 2 male
          no
## 3 female no
                      31 5100
## 4 male
                           4200
                           4000
## 5 male
```

```
baselers %>%
  group_by(fasnacht) %>%
  summarise(
    N = n(),
    age_mean = mean(age),
    income_mean = mean(income, na.rm = TRUE)
```

```
## # A tibble: 2 x 4
    fasnacht
                  N age_mean income_mean
     <chr>
              <int>
                       <dbl>
                                   <dbl>
## 1 no
               9706
                        44.6
                                   7527.
                                   7692.
## 2 yes
                294
                        45.3
```

### Here is part of the baselers dataframe

```
baselers %>%
 select(sex, fasnacht, age, income) %>%
  slice(1:5)
```

```
## # A tibble: 5 x 4
           fasnacht
                      age income
    sex
    <chr> <chr>
                    <dbl> <dbl>
                           6300
## 1 male
                       44
           no
## 2 male
                       65 10900
           no
## 3 female no
                       31
                           5100
## 4 male
                            4200
                            4000
## 5 male
```

```
## # A tibble: 4 x 5
              fasnacht [2]
## # Groups:
    fasnacht sex
                        N age_mean income_mean
    <chr>
              <chr> <int>
                              <dbl>
                                         <dbl>
## 1 no
                                         7646.
             female
                     4886
                              45.4
## 2 no
             male
                      4820
                              43.8
                                         7407.
## 3 yes
                                         7829.
             female
                      114
                              46.4
## 4 yes
             male
                      180
                              44.6
                                         7602
```

#### Here is part of the baselers dataframe

```
baselers %>%
 select(sex, fasnacht, age, income) %>%
  slice(1:5)
```

```
## # A tibble: 5 x 4
                     age income
           fasnacht
    sex
   <chr> <chr>
                    <dbl> <dbl>
## 1 male
                      44
                           6300
           no
                      65 10900
## 2 male
          no
## 3 female no
                      31
                           5100
## 4 male
                           4200
                           4000
## 5 male
```

```
baselers %>%
  group_by(fasnacht, sex) %>%
  summarise(
    N = n()
    age_mean = mean(age),
    income_mean = mean(income, na.rm = TRUE)
```

```
## # A tibble: 4 x 5
## # Groups: fasnacht [2]
    fasnacht sex
                        N age_mean income_mean
    <chr>
              <chr> <int>
                              <dbl>
                                          <dbl>
                                          7646.
## 1 no
             female
                     4886
                              45.4
## 2 no
                              43.8
                                         7407.
             male
                      4820
## 3 yes
                      114
                              46.4
                                         7829.
             female
                                          7602
## 4 yes
             male
                      180
                               44.6
```

### Here is part of the baselers dataframe

```
baselers %>%
 select(sex, fasnacht, age, income) %>%
 slice(1:5)
```

```
## # A tibble: 5 x 4
           fasnacht
                     age income
    sex
   <chr> <chr>
                    <dbl> <dbl>
                           6300
## 1 male
                      44
           no
## 2 male
                      65 10900
          no
## 3 female no
                      31
                           5100
## 4 male
                           4200
                           4000
## 5 male
```

```
## # A tibble: 2 x 5
## # Groups: fasnacht [2]
    fasnacht sex
                       N age_mean income_mean
    <chr>
             <chr> <int>
                            <dbl>
                                        <dbl>
## 1 no
                             43.8
                                        7407.
             male
                    4820
## 2 yes
             male
                     180
                             44.6
                                        7602
```

#### Here is part of the baselers dataframe

```
baselers %>%
 select(sex, fasnacht, age, income) %>%
 slice(1:5)
```

```
## # A tibble: 5 x 4
           fasnacht
                     age income
    sex
   <chr> <chr>
                   <dbl> <dbl>
                      44
                          6300
## 1 male
          no
                      65 10900
## 2 male
          no
## 3 female no
                      31 5100
## 4 male
                           4200
## 5 male
                           4000
```

```
baselers %>%
  filter(sex == "male") %>% # male patients only
  group_by(fasnacht, sex) %>%
  summarise(
    N = n(),
    age_mean = mean(age),
    income_mean = mean(income, na.rm = TRUE)
## # A tibble: 2 x 5
```

```
## # Groups: fasnacht [2]
    fasnacht sex
                       N age_mean income_mean
    <chr>
                            <dbl>
             <chr> <int>
                                        <dbl>
                             43.8
                                        7407.
## 1 no
             male
                    4820
## 2 yes
             male
                     180
                             44.6
                                        7602
```

### Here is part of the baselers dataframe

```
baselers %>%
 select(sex, fasnacht, age, income) %>%
 slice(1:5)
```

```
## # A tibble: 5 x 4
           fasnacht
                      age income
    sex
    <chr> <chr>
                    <dbl> <dbl>
                           6300
## 1 male
                       44
           no
## 2 male
                       65 10900
           no
## 3 female no
                       31
                           5100
## 4 male
                           4200
                           4000
## 5 male
```

```
## # A tibble: 4 x 3
    education
                           N income_mean
    <chr>
                       <int>
                                   <dbl>
                                   7555.
## 1 SEK_III
                        4034
                                   7551.
## 2 obligatory_school
                        1239
## 3 apprenticeship
                        3986
                                   7538.
## 4 SEK_II
                                   7338.
                         741
```

#### Here is part of the baselers dataframe

```
baselers %>%
 select(sex, fasnacht, age, income) %>%
 slice(1:5)
```

```
## # A tibble: 5 x 4
                     age income
           fasnacht
    sex
    <chr> <chr>
                    <dbl> <dbl>
                           6300
## 1 male
                      44
           no
                      65 10900
## 2 male
          no
## 3 female no
                      31
                           5100
## 4 male
                           4200
                           4000
## 5 male
```

```
baselers %>%
  group_by(education) %>%
  summarise(
    N = n(),
    income_mean = mean(income, na.rm = TRUE)
  ) %>%
  arrange(desc(income_mean))
```

```
## # A tibble: 4 x 3
    education
                           N income_mean
     <chr>
                       <int>
                                   <dbl>
## 1 SEK_III
                        4034
                                   7555.
                                   7551.
## 2 obligatory_school
                        1239
## 3 apprenticeship
                                   7538.
                        3986
## 4 SEK_II
                         741
                                   7338.
```

### What have we not covered yet? Statistics!

Statistical functions (almost) always require two key arguments

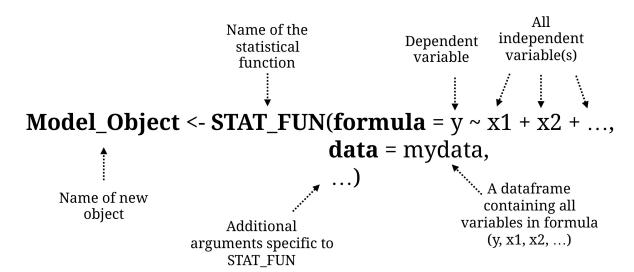
data	A dataframe
formula	A formula specifying variables in the model

A formula specifies a dependent variable (y) as a function of one or more **independent** variables (x1, x2, ...) in the form:

formula = 
$$y \sim x1 + x2 + ...$$

#### How to create a statistical object:

```
# Example: Create regression object (my_glm)
my_glm <- glm(formula = income ~ age + height,</pre>
               data = baselers)
```



### Simple hypothesis tests

All of the basic **one and two sample hypothesis tests** are included in the stats package.

These tests take either a **formula** for the argument formula, or individual vectors for the arguments x, and y

Hypothesis Test	R Function
t-test	t.test()
Correlation Test	cor.test()
Chi-Square Test	<pre>chisq.test()</pre>

### t-test with t.test()

```
# 2-sample t-test
t.test(formula = income ~ sex,
       data = baselers)
##
      Welch Two Sample t-test
## data: income by sex
## t = 4, df = 8500, p-value = 6e-05
## alternative hypothesis: true difference in means is not (
## 95 percent confidence interval:
## 120.6 352.2
## sample estimates:
## mean in group female
                         mean in group male
##
                   7650
                                        7414
```

# Regression with glm(), lm()

How to create a regression model predicting, e.g., how much money people spend on food as a function of income?

#### Part of the baselers dataframe:

food	income	happiness
610	6300	5
1550	10900	7
720	5100	7
680	4200	7
260	4000	5

### Generalized regression with glm()

```
# food (y) on income (x1) and happiness (x2)
food_glm <- glm(formula = food ~ income + happiness,</pre>
                 data = baselers)
# Print food_glm
food_glm
## Call: glm(formula = food ~ income + happiness, data = baselers)
## Coefficients:
## (Intercept)
                     income
                               happiness
                      0.101
                                  52.205
      -302.089
## Degrees of Freedom: 8509 Total (i.e. Null); 8507 Residual
    (1490 observations deleted due to missingness)
## Null Deviance:
                         1.27e+09
## Residual Deviance: 6.06e+08
                                   AIC: 119000
```

### Exploring statistical objects

Explore statistical objects using generic functions such as print(), summary(), predict() and plot().

**Generic** functions different things depending on the class label of the object.

```
# Create statistical object
obj <- STAT_FUN(formula = ...,</pre>
               data = ...)
names(obj)
                # Elements
print(obj) # Print
summary(obj)
                # Summary
plot(obj)
             # Plotting
predict(obj, ..) # Predict
```

```
# Create a glm object
my_glm <- glm(formula = income ~ happiness + age,
              data = baselers)
summary(my_glm)
## Call:
## glm(formula = income ~ happiness + age, data = baselers)
## Deviance Residuals:
              1Q Median
                              3Q
                                     Max
     Min
   -4045
            -835
                             814
                                    4899
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
                           94.363 16.70 < 2e-16 ***
## (Intercept) 1575.497
                           12.520 -8.02 1.2e-15 ***
## happiness -100.431
               149.312
                            0.815 183.31 < 2e-16 ***
## age
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' 201/23
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```

# tidy()

The tidy() function from the broom package **converts** the most important results of many statistical object like "glm" to a data frame.

```
# install and load broom
install.packages('broom')
library(broom)
```



my\_glm is a long list with many kinds of outputs and an ugly printout

```
# Print (raw glm)
my_glm
##
## Call: glm(formula = income ~ happiness + age, data = baselers
##
## Coefficients:
## (Intercept)
                  happiness
                                     age
##
          1575
                       -100
                                     149
## Degrees of Freedom: 8509 Total (i.e. Null); 8507 Residual
    (1490 observations deleted due to missingness)
## Null Deviance:
                         6.33e + 10
## Residual Deviance: 1.28e+10
                                   AIC: 145000
# Class (raw glm)
class(my_glm)
## [1] "glm" "lm"
```

# tidy()

The tidy() function from the broom package **converts** the most important results of many statistical object like "glm" to a data frame.

```
# install and load broom
install.packages('broom')
library(broom)
```



```
# Print (tidy glm)
tidy(my_glm)
## # A tibble: 3 x 5
                estimate std.error statistic p.value
    term
    <chr>
                             <dbl>
                                                <dbl>
                   <dbl>
                                       <dbl>
                                       16.7 1.33e-61
                   1575.
                            94.4
## 1 (Intercept)
                            12.5
                                       -8.02 1.18e-15
## 2 happiness
                   -100.
## 3 age
                    149.
                             0.815
                                      183. 0.
# Class (tidy glm)
class(tidy(my_glm))
## [1] "tbl_df"
                   "tbl"
                                "data.frame"
```

### **Practical**