DATA MANAGEMENT IN R

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DISCLAIMER

I owe a debt of gratitude to many people as the thoughts and code in these slides are the process of years-long development cycles and discussions with my team, friends, colleagues and peers. When someone has contributed to the content of the slides, I have credited their authorship.

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• You must ensure that the content is not used for further training of the model



SLIDE MATERIALS AND SOURCE CODE

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Materials

- lecture slides on Moodle
- course page: www.gerkovink.com/sur
- source: github.com/gerkovink/sur



RECAP

Yesterday we learned

- 1. How to use R and RStudio
- 2. How to install packages
- 3. How to use simple data containers
- 4. How to do subsetting in base R with [] and \$
- 5. How to use logical operators to subset data
- 6. How to adhere to code conventions en style

TODAY

- Importeren en bestuderen van datasets
- Begrijpen en toepassen van verschillende datatypes en database formats
- Variabelen labelen en (her)coderen
- De blauwdruk van R: frames en environments
- Pipes
- Formules gebruiken in functies

NEW PACKAGES WE USE

```
library(tibble) # tibbles variation on data frames
library(dplyr) # data manipulation
library(haven) # in/exporting data
library(magrittr) # pipes
library(labelled) # labelled data manipulation
library(tidyr) # data tidying
library(broom) # tidying model outputs
```







IMPORTING DATA: STATA

```
stata data <- read dta("files/03-poverty-analysis-data-2022-rt001-housing-plus.dta")
 2 head(stata data)
# A tibble: 6 × 114
       hhid domain2
                        psu domain gp subdom district fortnight panel
                                                                        hhid16
                                                     <dbl> <dbl+1> <dbl>
       <dbl> <dbl+lbl> <dbl+lbl> <dbl+lbl> <chr>
1 1102500401 1.1
                      10250 1 [Gre... 1 [Param... Paramar... 1 1 [Pan... 6.02e6
                 10250 1 [Gre... 1 [Param... Paramar... 1 1 [Pan... 6.02e6
2 1102500501 1.1
                 10250 1 [Gre... 1 [Param... Paramar... 1 1 [Pan... 6.02e6
3 1102500502 1.1
                 10250 1 [Gre... 1 [Param... Paramar... 1 1 [Pan... 6.02e6
4 1102501202 1.1
                 10743 1 [Gre... 1 [Param... Paramar... 1 1 [Pan... 6.04e6
5 1107430501 1.1
6 1107430801 1.1
                  10743 1 [Gre... 1 [Param... Paramar... 1 1 [Pan... 6.04e6
# i 105 more variables: lat cen <dbl>, long cen <dbl>, result <dbl+lbl>,
    end date n <date>, Year s <dbl>, Month s <dbl>, Day <dbl>, stratum <dbl>,
   hhid text <chr>, HHsize <dbl>, HHsize2 <dbl>, interv <dbl>, end date <chr>,
   q17 02 <dbl+lbl>, q17 03a <dbl+lbl>, q17 03b <dbl+lbl>, q17 04 <dbl+lbl>,
   q12a <dbl+lbl>, q12 01a <dbl>, q12 01b <dbl>, q12 02a <dbl>, q12 02b <dbl>,
    q12 03a <dbl>, q12 03b <dbl>, q12 04a <dbl>, q12 04b <dbl>,
    q12 05 <dbl+lbl>, q13 01 <dbl+lbl>, q13 01 ot <chr>, q13 02 <dbl+lbl>, ...
```

IMPORTING DATA: SPSS

```
spss data <- read sav("files/SUR 2023 LAPOP AmericasBarometer v1.0 w orginal.sav",
  2
                          user na = TRUE)
    spss data2 <- read sav("files/SUR 2023 LAPOP AmericasBarometer v1.0 w orginal.sav")
    head(spss data)
# A tibble: 6 × 162
                          nationality
  idnum
            pais
                                          estratopri
                                                              estratosec strata
  <dbl+lbl> <dbl+lbl>
                        <dbl+lbl>
                                          <dbl+lbl>
                                                              <dbl+lbl>
                                                                           <dbl+>
            27 [Suriname] 27 [Surinamese] 2702 [Wanica / Par... 2 [Medium ... 2702
1 5581
2 5642
            27 [Suriname] 27 [Surinamese] 2701 [Paramaribo] 1 [Large (... 2701
            27 [Suriname] 27 [Surinamese] 2701 [Paramaribo] 1 [Large (... 2701
3 4622
            27 [Suriname] 27 [Surinamese] 2702 [Wanica / Par... 2 [Medium ... 2702
4 4034
            27 [Suriname] 27 [Surinamese] 2701 [Paramaribo] 1 [Large (... 2701
5 9206
6 2101
            27 [Suriname] 27 [Surinamese] 2702 [Wanica / Par... 2 [Medium ... 2702
# i 156 more variables: prov <dbl+lbl>, municipio <dbl+lbl>, upm <dbl+lbl>,
    ur <dbl+lbl>, cluster <dbl+lbl>, year <dbl+lbl>, wave <dbl+lbl>,
    wt <dbl+lbl>, g1tc r <dbl+lbl>, g2 <dbl+lbl>, a4n <dbl+lbl>,
    soct2 <dbl+lbl>, idio2 <dbl+lbl>, mesfut1 <dbl+lbl>, cp8 <dbl+lbl>,
    it1 <dbl+lbl>, jc10 <dbl+lbl>, jc13 <dbl+lbl>, jc15a <dbl+lbl>,
    jc16a <dbl+lbl>, vic1ext <dbl+lbl>, aoj11 <dbl+lbl>, aoj12 <dbl+lbl>,
    pese1 <dbl+lbl>, pese2 <dbl+lbl>, aoj17 <dbl+lbl>, ivol24 <dbl+lbl>, ...
```



QUICK INSPECTION: MISSINGNESS

```
1 sum(is.na(stata_data)) # total number of NAs
[1] 31355
1 sum(is.na(spss_data)) # total number of NAs
[1] 28174
1 sum(is.na(spss_data2)) # total number of NAs
[1] 28174
```

LET'S LOOK AT GENDER

```
1 head(spss dataq1tc r, n = 20)
<labelled spss<double>[20]>: Gender
 [1] 888888
[11]
Missing values: 888888, 988888, 999999
Labels:
 value
                                          label
                                       Man/male
                                   Woman/female
      3 Does not identify as either man or woman
 888888
                                             DK
 988888
                                             NR
                                            N/A
 999999
 1 head(spss data2$q1tc r, n = 20)
<labelled<double>[20]>: Gender
 [1] NA 2 1 2 1 2 2 1 1 2 1 2
Labels:
                                          label
 value
                                       Man/male
      1
                                   Woman/female
      3 Does not identify as either man or woman
 888888
                                             DK
 988888
                                             NR
                                            N/A
 999999
```

LET'S LOOK AT GENDER

743 746

CORRECTING THE MISSINGS

In base R, there is only one type of missing value: NA. In SPSS and Stata, there are multiple types of missing values. In R, we can use the haven package to convert these to NA.

```
1 spss_data <- zap_missing(spss_data) # set all special missing values to NA
```

LET'S LOOK AT GENDER AGAIN

```
1 table(spss_data$q1tc_r)

1   2   3
743 746   1

1 table(spss_data2$q1tc_r)

1   2   3
743 746   1
```



TAGGED MISSING VALUES

Alternatively, if you'd still like to use *special NAs*, you can use haven::tag_na()

```
1 x <- c(1:5, tagged_na("a"), tagged_na("z"), NA)
2
3 # Tagged NA's work identically to regular NAs
4 x

[1] 1 2 3 4 5 NA NA NA

1 is.na(x)</pre>
```

[1] FALSE FALSE FALSE FALSE TRUE TRUE TRUE

EXPLORING DATA SETS

1 glimpse(spss_data)

```
Rows: 1,539
Columns: 162
$ idnum
           <dbl+lbl> 5581, 5642, 4622, 4034, 9206, 2101, 3574, 709, 8666, ...
$ pais
           $ estratopri <dbl+lbl> 2702, 2701, 2701, 2702, 2701, 2702, 2701, 2701, 2701, ...
$ estratosec <dbl+lbl> 2, 1, 1, 2, 1, 2, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 2, 1, ...
           <dbl+lbl> 2702, 2701, 2701, 2702, 2701, 2702, 2701, 2701, 2701, ...
$ strata
$ prov
           <dbl+lbl> 2702, 2701, 2701, 2702, 2701, 2702, 2701, 2701, 2701, ...
          <dbl+lbl> 270214, 270109, 270109, 270214, 270109, 270214, 270109...
$ municipio
$ upm
           <dbl+lbl> 43, 21, 21, 43, 21, 43, 21, 21, 21, 21, 21, 21, 43, 21...
           <dbl+lbl> 2, 1, 1, 2, 1, 2, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 2, 1, ...
$ ur
$ cluster
           <dbl+1bl> 87, 67, 5, 234, 67, 234, 233, 233, 5, 92, 67,...
$ year
           <dbl+1bl> 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, ...
$ wave
           <dbl+1bl> 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, ...
$ wt
           <dbl+lbl> NA, 2, 1, 2, 1, 2, 2, 1, 1, 2, 1, 2...
$ q1tc r
           <dbl+lbl> 59, 61, 30, 36, 34, 53, 75, 27, 27, 50, 60, 34, 50, 30...
$ q2
$ a4n
           <dbl+lbl> 1, 1, 1, 1, 1, 1, 1, 77, 1, 1, 1, 77...
```



EXPLORING DATA SETS

1 glimpse(stata data)

```
Rows: 2,502
Columns: 114
$ hhid
            <dbl> 1102500401, 1102500501, 1102500502, 1102501202, 11074305...
$ domain2
            $ psu
            <dbl> 10250, 10250, 10250, 10250, 10743, 10743, 10743, 10743, ...
$ domain
            <dbl+lbl> 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2.
            <dbl+lbl> 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0...
$ qp subdom
$ district
            <chr> "Paramaribo", "Paramaribo", "Paramaribo", "Paramaribo", ...
$ fortnight
            $ panel
            $ hhid16
            <dbl> 6015041, 6015051, 6015052, 6015122, 6039051, 6039081, 60...
$ lat cen
            <dbl> 5.847621, 5.847621, 5.847621, 5.847621, 5.819147, 5.8191...
$ long cen
            <dbl> -55.17032, -55.17032, -55.17032, -55.17032, -55.21745, -...
$ result
            $ end date n
            <date> 2022-01-04, 2022-01-05, 2022-01-10, 2022-01-04, 2022-01...
$ Year s
            <dbl> 2022, 2022, 2022, 2022, 2022, 2022, 2022, 2022, 2022, 202.
$ Month s
            <dbl> 4, 5, 10, 4, 5, 5, 5, 5, 5, 7, 7, 7, 7, 7, 7, 15, 9, 14,...
$ Day
            <dbl> 2, 2, 2, 2, 5, 5, 5, 5, 5, 4, 4, 4, 4, 4, 4, 6, 6, 6, 6, ...
$ stratum
```



LABELS AND FACTORS

Currently, the q1tc_r variable is a numeric vector - even though the SPSS labels are still recorded. In R, we often use factors to represent categorical data. Factors are stored as integers with labels attached.

```
1 is.factor(spss_data$q1tc_r)
[1] FALSE
```

We can easily convert the q1tc_r variable to a factor using the haven package's as_factor() function, which will also preserve the labels.

```
1 spss_data <- as_factor(spss_data)
2 is.factor(spss_data$q1tc_r)</pre>
```

[1] TRUE

EXPLORING AGAIN - FACTORED

1 glimpse(as_factor(spss_data))

```
Rows: 1,539
Columns: 162
$ idnum
             <fct> 5581, 5642, 4622, 4034, 9206, 2101, 3574, 709, 8666, 1566,...
$ pais
             <fct> Suriname, Suriname, Suriname, Suriname, Suriname, Suriname...
$ nationality <fct> Surinamese, Surinamese, Surinamese, Surinamese, Surinamese...
$ estratopri <fct> Wanica / Para, Paramaribo, Paramaribo, Wanica / Para, Para...
$ estratosec <fct> "Medium (Between 3,000 and 10,000 inhabitants)", "Large (M...
             <fct> 2702, 2701, 2701, 2702, 2701, 2702, 2701, 2701, 2701, 2701...
$ strata
$ prov
             <fct> Wanica, Paramaribo, Paramaribo, Wanica, Paramaribo, Wanica...
            <fct> Saramacca Polder, Flora, Flora, Saramacca Polder, Flora, S...
$ municipio
$ upm
             <fct> 43, 21, 21, 43, 21, 43, 21, 21, 21, 21, 21, 21, 43, 21, 21...
             <fct> Rural, Urban, Urban, Rural, Urban, Rural, Urban, Urban, Ur...
$ ur
$ cluster
             <fct> 87, 67, 5, 234, 67, 234, 233, 233, 5, 92, 67, 233, 234, 5,...
$ year
             <fct> 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023...
$ wave
             <fct> 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023...
$ wt
             $ q1tc r
             <fct> NA, Woman/female, Man/male, Woman/female, Man/male, Woman/...
             <fct> 59, 61, 30, 36, 34, 53, 75, 27, 27, 50, 60, 34, 50, 30, 18...
$ q2
             <fct> "Economic issues", "Economic issues", "Economic issues", "...
$ a4n
```



EXPLORING AGAIN - FACTORED

1 glimpse(as_factor(stata_data))

```
Rows: 2,502
Columns: 114
$ hhid
              <dbl> 1102500401, 1102500501, 1102500502, 1102501202, 11074305...
$ domain2
              $ psu
              <dbl> 10250, 10250, 10250, 10250, 10743, 10743, 10743, 10743, ...
$ domain
              <fct> Great Paramaribo, Great Paramaribo, Great Paramaribo, Gr...
              <fct> Paramaribo, Paramaribo, Paramaribo, Paramaribo, Paramari...
$ qp subdom
              <chr> "Paramaribo", "Paramaribo", "Paramaribo", "Paramaribo", ...
$ district
$ fortnight
              <fct> Panel, Panel, Panel, Panel, Panel, Panel, Panel, Panel, ...
$ panel
$ hhid16
              <dbl> 6015041, 6015051, 6015052, 6015122, 6039051, 6039081, 60...
              <dbl> 5.847621, 5.847621, 5.847621, 5.847621, 5.819147, 5.8191...
$ lat cen
$ long cen
              <dbl> -55.17032, -55.17032, -55.17032, -55.17032, -55.21745, -...
              <fct> Interview finalized - Fully completed, Interview finaliz...
$ result
$ end date n
              <date> 2022-01-04, 2022-01-05, 2022-01-10, 2022-01-04, 2022-01...
$ Year s
              <dbl> 2022, 2022, 2022, 2022, 2022, 2022, 2022, 2022, 2022, 202.
$ Month s
              <dbl> 4, 5, 10, 4, 5, 5, 5, 5, 5, 7, 7, 7, 7, 7, 7, 15, 9, 14,...
$ Day
              <dbl> 2, 2, 2, 2, 5, 5, 5, 5, 5, 4, 4, 4, 4, 4, 4, 6, 6, 6, 6, ...
$ stratum
```



PIPES

Pipes are a way to chain together multiple operations in a more readable way. The pipe operator %>% takes the output of the left-hand side and passes it as the first argument to the function on the right-hand side. In R, there is now also the |> operator, which is a base R pipe.

Remember

\$ idnum

\$ pais

```
1 glimpse(as_factor(stata_data))
```

With a pipe this would be

```
1 spss_data %>%
2 as_factor() %>%
3 glimpse()
```

and with the base R pipe |> this would be

```
1 spss_data |>
2 as_factor() |>
3 glimpse()

Rows: 1,539
Columns: 162
```

<fct> 5581, 5642, 4622, 4034, 9206, 2101, 3574, 709, 8666, 1566,...

<fct> Suriname, Suriname, Suriname, Suriname, Suriname, Suriname...

\$ nationality <fct> Surinamese, Surinamese, Surinamese, Surinamese... \$ estratopri <fct> Wanica / Para, Paramaribo, Paramaribo, Wanica / Para, Paramaribo, Paramaribo, Wanica / Para, Paramaribo, Paramaribo, Wanica / Para, Paramaribo, Para



```
<fct> "Medium (Between 3,000 and 10,000 inhabitants)", "Large (M...
$ estratosec
$ strata
             <fct> 2702, 2701, 2701, 2702, 2701, 2702, 2701, 2701, 2701, 2701, 2701...
             <fct> Wanica, Paramaribo, Paramaribo, Wanica, Paramaribo, Wanica...
$ prov
$ municipio
             <fct> Saramacca Polder, Flora, Flora, Saramacca Polder, Flora, S...
$ upm
             <fct> 43, 21, 21, 43, 21, 43, 21, 21, 21, 21, 21, 21, 43, 21, 21...
$ ur
             <fct> Rural, Urban, Urban, Rural, Urban, Rural, Urban, Urban, Ur...
$ cluster
             <fct> 87, 67, 5, 234, 67, 234, 233, 233, 5, 92, 67, 233, 234, 5,...
$ year
             <fct> 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023...
$ wave
             <fct> 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023, 2023...
$ wt
             $ q1tc r
             <fct> NA, Woman/female, Man/male, Woman/female, Man/male, Woman/...
$ q2
             <fct> 59, 61, 30, 36, 34, 53, 75, 27, 27, 50, 60, 34, 50, 30, 18...
             <fct> "Economic issues", "Economic issues", "Economic issues", "...
$ a4n
```

PIPES IN DETAIL

- %>% is the pipe operator from the magrittr package
- |> is the base R pipe operator introduced in R 4.1.0

Both operators allow you to chain together multiple operations in a more readable way

A pipe is a way to pass the output of one function as the input to another function, without having to create intermediate variables.

A %>% B is equivalent to B(A), where A is the output of the left-hand side and B is the function on the right-hand side. A is expected to be the first argument in function B.

The next step in a pipe is always expected to be a function! If the next step is not a function, you need to be clever, otherwise you'll get an error.



WHAT IF THE NEXT STEP IS NOT THE FIRST ARGUMENT?

You can use the **placeholder** • (or _ with the native R pipe) to indicate where the output of the previous step should go in the next function.

For example,



WHAT IF THE NEXT STEP IS NOT THE FIRST ARGUMENT?

You can use the **placeholder** • (or _ with the native R pipe) to indicate where the output of the previous step should go in the next function.

For example,

OTHER PIPES

- %\$% is the exposition pipe from the magrittr package, which allows you to use the names of the variables in the data frame directly without having to use the \$ operator.
- %<>% is the assignment pipe from the magrittr package, which allows you to modify the data frame in place, without the need for calling assign() or <- again.

There are more pipes (like the %T>% pipe), but they can be very confusing and we therefore skip them in this course.

```
1 stata_data %>%
2    filter(q13_05 == 1 | q13_05 == 2) %$% # Note the exposition pipe
3    t.test(Year_s ~ q13_05)

Welch Two Sample t-test

data: Year_s by q13_05
t = -1, df = 1429, p-value = 0.3175
alternative hypothesis: true difference in means between group 1 and group 2 is not equal to 0
95 percent confidence interval:
    -0.0020710668    0.0006724654
sample estimates:
mean in group 1 mean in group 2
    2022.000    2022.001
```

RENAMING VARIABLES WITH A PIPE



RECODING WITH mutate() EN recode()

```
spss data <- spss data %>%
      mutate(gender rec = recode(gender,
  3
                                  "Man/male" = "male",
                                 "Woman/female" = "female"))
    spss data$gender %>% head()
                 Woman/female Man/male
                                           Woman/female Man/male
[1] <NA>
[6] Woman/female
6 Levels: Man/male Woman/female ... N/A
    spss data$gender_rec %>% head()
           female male female male
                                     female
[1] <NA>
Levels: male female Does not identify as either man or woman DK NR N/A
```



LABELING VARIABLES

With labeled variables, we can add an additional layer of description to variable, very similar to what SPSS and STATA do.

```
1 spss_data <- set_variable_labels(spss_data, gender_rec = "Gerecodeerd geslacht")
2 spss_data$gender_rec %>% glimpse()

Factor w/ 6 levels "male", "female",..: NA 2 1 2 1 2 2 1 1 2 ...
- attr(*, "label")= chr "Gerecodeerd geslacht"
```

SELECTING AND FILTERING

```
spss data %<>%
     rename(age = q2)
   spss data %>%
     filter(age > 18) %>%
     select(age, gender) %>%
     summary()
                                                   gender
     age
           Man/male
18
       :0
                                                      :0
19
           Woman/female
                                                      :0
            Does not identify as either man or woman:0
20
21
       :0
                                                      :0
            DK
22
       :0
            NR
                                                      : 0
23
       : 0
            N/A
                                                      : 0
(Other):0
```



CALCULATIONS AND SUMMARISING

age is also a factor, hence mean () is meaningless. We have to convert age to numeric:



JOINING DATA

INNER JOIN

With an inner join, we combine two data frames based on a common key. Only the rows with matching keys in both data frames are kept.



LEFT JOIN

With a left join, we keep all rows from the left data frame and only the matching rows from the right data frame. If there is no match, the result will contain NA for the columns from the right data frame.

```
1 band_members %>% left_join(band_instruments)

# A tibble: 3 × 3
  name band plays
  <chr>  <chr>  <chr>  <chr>  1 Mick Stones <NA>
2 John Beatles guitar
3 Paul Beatles bass
```

RIGHT JOIN

With a right join, we keep all rows from the right data frame and only the matching rows from the left data frame. If there is no match, the result will contain NA for the columns from the left data frame.



FULL JOIN

With a full join, we keep all rows from both data frames. If there is no match, the result will contain NA for the columns from the other data frame.



MODELING IN R

To model objects based on other objects, we use \sim (tilde)

For example, to model body mass index (BMI) on weight, we would type

```
1 BMI ~ weight
```

The \sim is used to separate the left- and right-hand sides in a model **formula**.

For functions (or models), within models we use \mathbf{I} () - For example, to model body mass index (BMI) on its deterministic function of weight and height, we would type

```
1 BMI ~ I(weight / height<sup>2</sup>)
```

MODELING CONTINUED

We already saw the use of the \sim operator in the t test() function, where we specified the outcome variable on the left-hand side and the grouping variable on the right-hand side.

```
1 stata_data %>%
2 filter(q13_05 == 1 | q13_05 == 2) %$%
3 t.test(Year_s ~ q13_05)
```



USING FORMULAS

```
1  # Use a formula in function lm() in a pipe
2  mtcars %>%
3  lm(mpg ~ wt + hp, data = .)

Call:
lm(formula = mpg ~ wt + hp, data = .)

Coefficients:
(Intercept)  wt  hp
37.22727  -3.87783  -0.03177
```



USING FORMULAS WITH broom

With the broom package, we can easily tidy up the output of models and other functions that return complex objects.

TABLES AND CONTINGENCY TABLES

0 213

0

0 165

0

Nickerie

Saramacca

Wanica

Sipaliwini

Paramaribo 150 147 112

Para

```
stata data %$%
      table(district)
district.
Brokopondo Commewijne
                         Coronie
                                  Marowijne
                                               Nickerie
                                                               Para Paramaribo
                               33
                                                    213
                                                                 27
                                                                           945
                  170
                                          75
 Saramacca Sipaliwini
                          Wanica
       233
                  118
                              681
    stata data %$%
      table(district, stratum)
            stratum
district
                                                   10
                                                       11 12
                                                                13
                                                                    14 15
                                                                           16
  Brokopondo
 Commewijne
                           0
                               0 46
                                                0 124
                                                                             0
 Coronie
                       0
                          33
                                                                             0
                            0
                                   47
 Marowijne
                                                                         0
                                                                            28
```

0

51

57

0

0 182

0 233

0 120 129 122

0

0 118

0 164 227

TABLES

```
stata data %$%
 2
      table(district, stratum) %>% # calculate table
      prop.table() %>% # convert to proportions
      round(3) # round to 3 decimals
 4
           stratum
district
                1
                                                                    10
                      2
                                  4
                                        5
                                                                          11
 Brokopondo 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000
 Commewijne 0.000 0.000 0.000 0.000 0.018 0.000 0.000 0.000 0.050 0.000
 Coronie
            0.000 0.000 0.000 0.013 0.000 0.000 0.000 0.000 0.000 0.000
 Marowijne 0.000 0.000 0.000 0.000 0.019 0.000 0.000 0.000 0.000 0.000
 Nickerie 0.000 0.000 0.000 0.085 0.000 0.000 0.000 0.000 0.000 0.000
            0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.009 0.000 0.000
 Para
 Paramaribo 0.060 0.059 0.045 0.000 0.066 0.000 0.000 0.000 0.000 0.000 0.048
  Saramacca 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.073 0.020 0.000 0.000
 Sipaliwini 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000
 Wanica
            0.000 0.000 0.000 0.000 0.000 0.000 0.093 0.000 0.023 0.000 0.000
           stratum
district
                     13
                           14
               12
                              15
                                      16
  Brokopondo 0.000 0.000 0.000 0.000 0.001
 Commewijne 0.000 0.000 0.000 0.000 0.000
 Coronie
            0.000 0.000 0.000 0.000 0.000
 Marowijne 0.000 0.000 0.000 0.001
 Nickerie
            0.000 0.000 0.000 0.000 0.000
```



TIBBLES VS DATA.FRAMES

Tibbles are a modern reimagining of data frames in R. They are part of the tidyverse and provide a more user-friendly interface for working with data.

```
is tibble(band members)
[1] TRUE
    band members %>%
      full join(band instruments)
# A tibble: 4 \times 3
 name band
                plays
  <chr> <chr>
                <chr>
1 Mick Stones <NA>
2 John Beatles guitar
3 Paul Beatles bass
4 Keith <NA>
                quitar
    band members %>%
      full join(band instruments) %>%
      as.data.frame()
           band plays
   name
1 Mick Stones
                  <NA>
  John Beatles guitar
                 bass
  Paul Beatles
4 Keith
          <NA> quitar
```

LAYERS IN R

There are several 'layers' in R. Some layers you are allowed to fiddle around in, some are forbidden. In general there is the following distinction:

- The global environment.
- User environments
- Functions
- Packages
- Namespaces

ENVIRONMENTS

The global environment can be seen as an olympic-size swimming pool. Everything you do has its place there.

If you'd like, you may create another, separate environment to work in.

• A user environment would by default not have access to other environments

FUNCTIONS

- If you create a function, it is positioned in the global environment.
- Everything that happens in a function, stays in a function. Unless you specifically tell the function to share the information with the global environment.
- See functions as a shampoo bottle in a swimming pool to which you add some water. If you'd like to see the color of the mixture, you'd have to squeeze the bottle for it to come out.

PACKAGE AND NAMESPACES

Namespaces are a way to organize functions and data in R. Every package has its own namespace, which means that functions and data in one package do not interfere with functions and data in another package.

- Everything needed to run the functions in a package is neatly contained within its own space
- See packages as separate (mini) pools that are connected to the main pool (the global environment)



%IN%

The %in% operator is used to check if elements of one vector are present in another vector. It returns a logical vector indicating whether each element of the first vector is found in the second vector.

```
1 x <- c(1, 2, 3, 4, 5)

2 y <- c(3, 4, 5, 6, 7)

3 x %in% y
```

[1] FALSE FALSE TRUE TRUE TRUE

grepl()

The grepl() function is used to search for a pattern in a character vector. It returns a logical vector indicating whether the pattern is found in each element of the character vector.

```
1 x <- c("apple", "banana", "cherry", "date")
2 pattern <- "a"
3 grepl(pattern, x)

[1] TRUE TRUE FALSE TRUE

1 grepl("^a", x) # starts with a

[1] TRUE FALSE FALSE FALSE

1 grepl("e$", x) # ends with e

[1] TRUE FALSE FALSE TRUE

1 grepl("cherry", x)

[1] FALSE FALSE TRUE FALSE</pre>
```



ANSCOMBE DATA

```
anscombe |>
      as tibble()
# A tibble: 11 × 8
      x1
            x2
                                          у3
                  x3
                        x4
                              у1
                                    y2
                                                у4
   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
                            8.04
      10
            10
                  10
                                  9.14
                                        7.46
                                              6.58
                            6.95
                                  8.14 6.77
                                              5.76
      13
                           7.58
                                  8.74 12.7
                                              7.71
            13
                  13
      9
             9
                            8.81
                                  8.77
                                       7.11
                                              8.84
                                        7.81 8.47
      11
                  11
                           8.33
                                  9.26
            11
      14
                            9.96
                                  8.1
                                        8.84 7.04
            14
                  14
                            7.24
                                  6.13
                                       6.08 5.25
                           4.26
                                  3.1
                                        5.39 12.5
      12
            12
                  12
                         8 10.8
                                  9.13
                                        8.15 5.56
10
                         8 4.82 7.26
                                        6.42 7.91
11
                         8 5.68
                                  4.74
                                        5.73 6.89
```



ANSCOMBE DATA PROPERTIES

```
anscombe |>
     cor() |>
     round(2)
     x1
           x2
                 x3
                      x4
                            у1
                                  y2
                                       у3
x1 1.00 1.00 1.00 -0.50 0.82 0.82 0.82 -0.31
   1.00 1.00 1.00 -0.50 0.82
                                0.82 \quad 0.82 \quad -0.31
   1.00 1.00 1.00 -0.50 0.82 0.82 0.82 -0.31
x4 -0.50 -0.50 -0.50 1.00 -0.53 -0.72 -0.34 0.82
   0.82 0.82 0.82 -0.53 1.00 0.75 0.47 -0.49
y2 0.82 0.82 0.82 -0.72 0.75 1.00 0.59 -0.48
  0.82 0.82 0.82 -0.34 0.47 0.59 1.00 -0.16
y4 -0.31 -0.31 -0.31 0.82 -0.49 -0.48 -0.16 1.00
```



ADDING SOME RANDOM NUMBERS



anscombe_new DATA PROPERTIES



PRACTICAL

