

Manipulating data

This guide is partly based on online material from Amy Willis, Kiirsti Owen and Amelia McNamara, and the book “R for Data Science” by Hadley Wickham and Garrett Grolemund. Thank you amazing R community!

Load packages

We will be using the readr, tidyr and dplyr packages from the Tidyverse family of packages. We will also load the “here” package that we will use to read in our data.

```
library(readr)
library(tidyr)
library(dplyr)
```

```
##
## 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
```

```
library(here)
```

```
## here() starts at C:/Users/Yuluo Chen/OneDrive - University of Strathclyde/Documents/Jing/JW_PU5058_A
```

First, let's practice with pivoting

We will start with a toy non-tidy dataset:

```
patient_ID <- c(1,2)
test_result_month1 <- c("a1" ,"a2")
test_result_month2 <- c("b1" ,"b2")
test_result_month3 <- c("c1" ,"c2")

patient_tests <- data.frame(patient_ID,test_result_month1,test_result_month2,test_result_month3)

patient_tests

##   patient_ID test_result_month1 test_result_month2 test_result_month3
## 1          1                a1                b1                c1
## 2          2                a2                b2                c2
```

The dataset is not tidy because each row contains three observations, one per month. A tidy dataset has one observation per row. To do this, we use `pivot_longer`.

- The first argument is the dataset to reshape, but as we are using the pipe (`%>%`) we are skipping the first argument.
- The next argument describes which columns need to be reshaped. In this case, it's every column apart from `patient_ID`.
- The `names_to` gives the name of the variable that will be created from the data stored in the column names, in this case the month.
- The `values_to` gives the name of the variable that will be created from the data stored in the cell value, in this case the test result.

```
tidy_patient_tests <- patient_tests %>%
  pivot_longer(
    c('test_result_month1', 'test_result_month2', 'test_result_month3'),
    names_to = 'month',
    values_to = 'test_result'
  )

tidy_patient_tests
```

```
## # A tibble: 6 x 3
##   patient_ID month      test_result
##       <dbl> <chr>      <chr>
## 1         1 test_result_month1 a1
## 2         1 test_result_month2 b1
## 3         1 test_result_month3 c1
## 4         2 test_result_month1 a2
## 5         2 test_result_month2 b2
## 6         2 test_result_month3 c2
```

As you can see, the data frame is now tidy (one observation per row), but it would be better if the “month” column just contained the month number (1,2,3). To do this we can add the arguments `names_prefix` to strip off the `test_result_month` prefix, and `names_transform` to convert month into an integer:

```
tidy_patient_tests <- patient_tests %>%
  pivot_longer(
    c('test_result_month1', 'test_result_month2', 'test_result_month3'),
    names_to = 'month',
    names_prefix = 'test_result_month',
    names_transform = list(month = as.integer),
    values_to = 'test_result'
  )

tidy_patient_tests
```

```
## # A tibble: 6 x 3
##   patient_ID month test_result
##       <dbl> <int> <chr>
## 1         1     1 a1
## 2         1     2 b1
## 3         1     3 c1
## 4         2     1 a2
## 5         2     2 b2
## 6         2     3 c2
```

Reading in the FEV data

We will use the same data as last week. So read in the data from file fev.csv and save it in an object called fev_data:

```
fev_data <- read_csv("fev.csv") # Add your code here!

## Rows: 654 Columns: 7
## -- Column specification -----
## Delimiter: ","
## dbl (7): seqnbr, subjid, age, fev, height, sex, smoke
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

Tip: If you got an error that “fev.csv” does not exist, check that you are working in the correct directory!

Operating on data: subsets

To select subsets of the data (not just columns with \$) use square brackets:

```
fev_data$fev[32] # 32nd element of the fev column
```

```
## [1] 3
```

```
fev_data[32,3] # 32nd element of the 3rd column
```

```
## # A tibble: 1 x 1
##   age
##   <dbl>
## 1     9
```

```
fev_data[32,"age"] # Same thing, but using the name of the 3rd column - better, as it is more readable
```

```
## # A tibble: 1 x 1
##   age
##   <dbl>
## 1     9
```

```
fev_data[32, ] # Everything in the 32nd row
```

```
## # A tibble: 1 x 7
##   seqnbr subjid age fev height sex smoke
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     32  7201     9     3   65.5     1     0
```

```
fev_data[32,1:3]
```

```
## # A tibble: 1 x 3
##   seqnbr subjid age
##   <dbl> <dbl> <dbl>
## 1     32  7201     9
```

```
fev_data[32,-5]
```

```
## # A tibble: 1 x 6
##   seqnbr subjid age fev sex smoke
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     32  7201     9     3     1     0
```

```
fev_data[32,-1:-2]
```

```
## # A tibble: 1 x 5
##   age fev height sex smoke
##   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     9     3  65.5     1     0
```

```
fev_data[32,c(1,3,5)] #c(1,3,5) is a vector of numbers (c means "combine")
```

```
## # A tibble: 1 x 3
##   seqnbr age height
##   <dbl> <dbl> <dbl>
## 1     32     9  65.5
```

```
c(1,3,5) %>%
  length
```

```
## [1] 3
```

-> How would you drop the 1st, 3rd and 5th column?

```
fev_data_no135 <- fev_data %>%
  select(-c(1,3,5))
```

Logicals

Besides numbers and strings of characters, R also stores logicals - TRUE and FALSE

Example: a new vector with elements that are TRUE if height is above 72 cm and FALSE otherwise:

```
is_tall <- fev_data$height > 72
```

Useful summary command:

```
table(is_tall)
```

```
## is_tall
## FALSE  TRUE
##   647    7
```

Which subjects in fev_data are tall?

```
is_tall_data <- fev_data[is_tall,]
```

Filtering (selecting rows)

```
fev_data %>%  
  filter(height > 72)
```

```
## # A tibble: 7 x 7  
##   seqnbr subjid   age   fev height   sex smoke  
##   <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>  
## 1    401  18841    14  4.27   72.5     1     0  
## 2    450  32741    13  4.22    74     1     0  
## 3    464  37241    13  4.88    73     1     0  
## 4    517  49541    13  5.08    74     1     0  
## 5    550  59941    14  4.27   72.5     1     0  
## 6    632  37441    17  5.63    73     1     0  
## 7    636  44241    16  3.64   73.5     1     0
```

```
fev_data %>%  
  filter(age == 6)
```

```
## # A tibble: 37 x 7  
##   seqnbr subjid   age   fev height   sex smoke  
##   <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>  
## 1      7   1752     6  1.92    58     0     0  
## 2      8   1753     6  1.42    56     0     0  
## 3     11   1952     6  1.60    53     0     0  
## 4     18   3551     6  1.88    53     0     0  
## 5     49  10841     6  1.65    55     1     0  
## 6     55  12241     6  1.63    54     1     0  
## 7     63  14251     6  1.48    51     0     0  
## 8     66  14541     6  1.75   57.5     1     0  
## 9     80  16151     6  1.72    53     0     0  
## 10    82  16252     6  1.70    53     0     0  
## # i 27 more rows
```

```
fev_data %>%  
  filter(age != 20)
```

```
## # A tibble: 654 x 7  
##   seqnbr subjid   age   fev height   sex smoke  
##   <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>  
## 1      1    301     9  1.71    57     0     0  
## 2      2    451     8  1.72   67.5     0     0  
## 3      3    501     7  1.72   54.5     0     0  
## 4      4    642     9  1.56    53     1     0  
## 5      5    901     9  1.90    57     1     0  
## 6      6   1701     8  2.34    61     0     0
```

```
## 7      7  1752      6  1.92  58      0      0
## 8      8  1753      6  1.42  56      0      0
## 9      9  1901      8  1.99  58.5    0      0
## 10     10  1951      9  1.94  60      0      0
## # i 644 more rows
```

```
fev_data %>%
  filter(age <= 20)
```

```
## # A tibble: 654 x 7
##   seqnbr subjid   age   fev height   sex smoke
##   <dbl>  <dbl> <dbl> <dbl>  <dbl> <dbl> <dbl>
## 1      1     301     9  1.71   57      0      0
## 2      2     451     8  1.72  67.5    0      0
## 3      3     501     7  1.72  54.5    0      0
## 4      4     642     9  1.56   53      1      0
## 5      5     901     9  1.90   57      1      0
## 6      6    1701     8  2.34   61      0      0
## 7      7    1752     6  1.92   58      0      0
## 8      8    1753     6  1.42   56      0      0
## 9      9    1901     8  1.99  58.5    0      0
## 10     10    1951     9  1.94   60      0      0
## # i 644 more rows
```

You can also filter by whether data are not a number (na):

```
fev_data %>%
  filter(is.na(age))  # opposite: !is.na(age)
```

```
## # A tibble: 0 x 7
## # i 7 variables: seqnbr <dbl>, subjid <dbl>, age <dbl>, fev <dbl>,
## #   height <dbl>, sex <dbl>, smoke <dbl>
```

You can combine multiple expressions with Boolean operators: & is “and”, | is “or”, and ! is “not”

```
fev_data %>%
  filter(age == 14 & smoke !=0)  # age is 14 AND smoker
```

```
## # A tibble: 7 x 7
##   seqnbr subjid   age   fev height   sex smoke
##   <dbl>  <dbl> <dbl> <dbl>  <dbl> <dbl> <dbl>
## 1    332   4952    14  2.24   66      0      1
## 2    358  10053    14  3.43   64      0      1
## 3    370  11642    14  3.96   72      1      1
## 4    384  15751    14  3.07   65      0      1
## 5    439  30042    14  4.31   69      1      1
## 6    556  61941    14  2.28   66      1      1
## 7    602  82743    14  4.76   68      1      1
```

```
fev_data %>%
  filter(age < 5 | height < 50) # younger than 5 OR shorter than 50 cm
```

```
## # A tibble: 18 x 7
##   seqnbr subjid   age   fev height   sex smoke
##   <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     21  4351     5 1.4     49     0     0
## 2     23  5152     4 0.839   48     0     0
## 3     26  5642     3 1.40   51.5    1     0
## 4     31  6851     5 1.28   49     0     0
## 5     59 13751     4 1.57   50     0     0
## 6     64 14252     4 1.58   49     0     0
## 7    104 23841     4 0.796   47     1     0
## 8    118 28551     5 1.20   46.5    0     0
## 9    157 38242     6 1.54   48     1     0
## 10   173 40541     4 1.79   52     1     0
## 11   181 43242     7 1.16   47     1     0
## 12   216 49551     4 1.10   48     0     0
## 13   222 50951     3 1.07   46     0     0
## 14   225 51341     6 1.42   49.5    1     0
## 15   233 54751     4 1.39   48     0     0
## 16   286 75951     4 1.42   49     0     0
## 17   299 80841     4 1.00   48     1     0
## 18   300 81241     6 1.43   49.5    1     0
```

Rules for filtering for categorical data: sex == "F" or sex != "F" sex %in% c("M", "F")

Selecting columns

```
fev_data %>%
  select(fev, height, age)
```

```
## # A tibble: 654 x 3
##   fev height   age
##   <dbl>  <dbl> <dbl>
## 1  1.71    57     9
## 2  1.72   67.5     8
## 3  1.72   54.5     7
## 4  1.56    53     9
## 5  1.90    57     9
## 6  2.34    61     8
## 7  1.92    58     6
## 8  1.42    56     6
## 9  1.99   58.5     8
## 10 1.94    60     9
## # i 644 more rows
```

```
fev_data %>%
  select(-seqnbr, -subjid)
```

```
## # A tibble: 654 x 5
##   age    fev height    sex smoke
##   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     9  1.71    57     0     0
## 2     8  1.72   67.5    0     0
## 3     7  1.72   54.5    0     0
## 4     9  1.56    53     1     0
## 5     9  1.90    57     1     0
## 6     8  2.34    61     0     0
## 7     6  1.92    58     0     0
## 8     6  1.42    56     0     0
## 9     8  1.99   58.5    0     0
## 10    9  1.94    60     0     0
## # i 644 more rows
```

Summarising data

```
fev_data %>%
  filter(age == 14 & smoke != 0) %>%
  summarise(mean(fev))
```

```
## # A tibble: 1 x 1
##   'mean(fev)'
##   <dbl>
## 1      3.43
```

You can name the summary variable:

```
fev_data %>%
  filter(age == 14 & smoke != 0) %>%
  summarise(my_mean = mean(fev))
```

```
## # A tibble: 1 x 1
##   my_mean
##   <dbl>
## 1      3.43
```

```
fev_data %>%
  filter(age == 14 & smoke != 0) %>%
  summarise(mean(fev), sd(fev))
```

```
## # A tibble: 1 x 2
##   'mean(fev)' 'sd(fev)'
##   <dbl>      <dbl>
## 1      3.43      0.976
```

To get the average FEV for both smokers and non-smokers we don't need to repeat for smoke==0. We can create a grouping variable:


```
fev_data %>%
  group_by(smoke)
```

```
## # A tibble: 654 x 7
## # Groups:   smoke [2]
##   seqnbr subjid   age   fev height   sex smoke
##   <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     1     301     9  1.71    57     0     0
## 2     2     451     8  1.72   67.5    0     0
## 3     3     501     7  1.72   54.5    0     0
## 4     4     642     9  1.56    53     1     0
## 5     5     901     9  1.90    57     1     0
## 6     6    1701     8  2.34    61     0     0
## 7     7    1752     6  1.92    58     0     0
## 8     8    1753     6  1.42    56     0     0
## 9     9    1901     8  1.99   58.5    0     0
## 10    10    1951     9  1.94    60     0     0
## # i 644 more rows
```

(Same exact data, it just prints the two groups)

```
fev_data %>%
  group_by(smoke) %>%
  summarise(mean(fev), sd(fev))
```

```
## # A tibble: 2 x 3
##   smoke 'mean(fev)' 'sd(fev)'
##   <dbl>      <dbl>      <dbl>
## 1     0         2.57      0.851
## 2     1         3.28      0.750
```

But what is the size of each group? `n()` gives us the number of observations in each group:

```
fev_data %>%
  group_by(smoke) %>%
  summarise(n = n(), mean = mean(fev), sd = sd(fev))
```

```
## # A tibble: 2 x 4
##   smoke    n mean   sd
##   <dbl> <int> <dbl> <dbl>
## 1     0  589  2.57 0.851
## 2     1   65  3.28 0.750
```

You can also group by your own variables:

```
fev_data %>%
  group_by(height < 60) %>%
  summarise(n(), mean(fev))
```

```
## # A tibble: 2 x 3
```

```
##   'height < 60' 'n()' 'mean(fev)'
##   <lgl>         <int>   <dbl>
## 1 FALSE         409     3.10
## 2 TRUE          245     1.86
```

A useful function: arrange

```
fev_data %>%
  group_by(age) %>%
  summarise(n(), mean(fev)) %>%
  arrange(age) # arrange by increasing age
```

```
## # A tibble: 17 x 3
##   age 'n()' 'mean(fev)'
##   <dbl> <int>   <dbl>
## 1     3     2     1.24
## 2     4     9     1.28
## 3     5    28     1.55
## 4     6    37     1.66
## 5     7    54     1.87
## 6     8    85     2.12
## 7     9    94     2.43
## 8    10    81     2.69
## 9    11    90     3.04
## 10   12    57     3.22
## 11   13    43     3.48
## 12   14    25     3.58
## 13   15    19     3.48
## 14   16    13     3.67
## 15   17     8     4.30
## 16   18     6     3.59
## 17   19     3     3.99
```

```
fev_data %>%
  group_by(age) %>%
  summarise(n(), mean(fev)) %>%
  arrange(desc(age)) # arrange by decreasing age
```

```
## # A tibble: 17 x 3
##   age 'n()' 'mean(fev)'
##   <dbl> <int>   <dbl>
## 1    19     3     3.99
## 2    18     6     3.59
## 3    17     8     4.30
## 4    16    13     3.67
## 5    15    19     3.48
## 6    14    25     3.58
## 7    13    43     3.48
## 8    12    57     3.22
## 9    11    90     3.04
## 10   10    81     2.69
## 11     9    94     2.43
```

```
## 12      8      85      2.12
## 13      7      54      1.87
## 14      6      37      1.66
## 15      5      28      1.55
## 16      4       9      1.28
## 17      3       2      1.24
```

Sorting columns

```
fev_data$age %>% sort #Sort a column
```

```
## [1] 3 3 4 4 4 4 4 4 4 4 4 4 5 5 5 5 5 5 5 5 5 5 5
## [26] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 6 6 6 6 6 6 6 6 6
## [51] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [76] 6 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7
## [101] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7
## [126] 7 7 7 7 7 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [151] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [176] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [201] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 9 9 9 9 9 9 9
## [226] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [251] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [276] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [301] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [326] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10
## [351] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10
## [376] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 11 11 11 11 11 11 11
## [401] 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11
## [426] 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11
## [451] 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11
## [476] 11 11 11 11 11 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12
## [501] 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12
## [526] 12 12 12 12 12 12 12 12 12 12 12 12 13 13 13 13 13 13 13 13 13 13 13
## [551] 13 13 13 13 13 13 13 13 13 13 13 13 13 13 13 13 13 13 13 13 13 13 13
## [576] 13 13 13 13 13 14 14 14 14 14 14 14 14 14 14 14 14 14 14 14 14 14 14
## [601] 14 14 14 14 14 15 15 15 15 15 15 15 15 15 15 15 15 15 15 15 15 15 16
## [626] 16 16 16 16 16 16 16 16 16 16 16 16 17 17 17 17 17 17 17 17 18 18 18
## [651] 18 19 19 19
```

```
fev_data$age %>% unique %>% sort # Sort unique values in a column
```

```
## [1] 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
```

table() gives you a count of a particular factor or combination of factor levels:

```
table(fev_data$age)
```

```
##
## 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
## 2 9 28 37 54 85 94 81 90 57 43 25 19 13 8 6 3
```

```
table(fev_data$age, fev_data$smoke)
```

```
##
##      0  1
##   3  2  0
##   4  9  0
##   5 28  0
##   6 37  0
##   7 54  0
##   8 85  0
##   9 93  1
##  10 76  5
##  11 81  9
##  12 50  7
##  13 30 13
##  14 18  7
##  15  9 10
##  16  6  7
##  17  6  2
##  18  4  2
##  19  1  2
```

→ **Problem 1:** Which subjects are male and which are female? (i.e. what does sex == 1 mean?)

```
fev_data_M <- fev_data %>%
  filter(sex==1) # Add your code here!
```

→ **Problem 2:** Why do smokers appear to have better lung function (higher forced expiratory volume - FEV)?

```
fev_data_fevPK <- fev_data %>%
  group_by(smoke) %>%
  summarise(N=n(), Mean_fev=mean(fev))
# Add your code here!
```

Useful function: rename

```
fev_data %>%
  rename(ID = subjid)
```

```
## # A tibble: 654 x 7
##   seqnbr    ID  age  fev height  sex smoke
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     1     301    9  1.71   57     0     0
## 2     2     451    8  1.72  67.5     0     0
## 3     3     501    7  1.72  54.5     0     0
## 4     4     642    9  1.56   53     1     0
## 5     5     901    9  1.90   57     1     0
## 6     6    1701    8  2.34   61     0     0
## 7     7    1752    6  1.92   58     0     0
## 8     8    1753    6  1.42   56     0     0
```

```
## 9      9 1901      8 1.99  58.5      0      0
## 10     10 1951      9 1.94   60      0      0
## # i 644 more rows
```

Mutate: compute new column

```
fev_data %>%
  mutate(heightdiff = height - mean(height))
```

```
## # A tibble: 654 x 8
##   seqnbr subjid  age  fev height  sex smoke heightdiff
##   <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>    <dbl>
## 1      1      301    9  1.71   57      0     0    -4.14
## 2      2      451    8  1.72  67.5      0     0     6.36
## 3      3      501    7  1.72  54.5      0     0    -6.64
## 4      4      642    9  1.56   53      1     0    -8.14
## 5      5      901    9  1.90   57      1     0    -4.14
## 6      6     1701    8  2.34   61      0     0   -0.144
## 7      7     1752    6  1.92   58      0     0    -3.14
## 8      8     1753    6  1.42   56      0     0    -5.14
## 9      9     1901    8  1.99  58.5      0     0    -2.64
## 10     10     1951    9  1.94   60      0     0    -1.14
## # i 644 more rows
```

Remember that to save these changes you need to assign to a new tibble:

```
new_fev_data <- fev_data %>%
  rename(id = subjid) %>%
  mutate(heightdiff = height - mean(height))
```