Loading & Cleaning Data in R

I know the file exists, why doesn't R?

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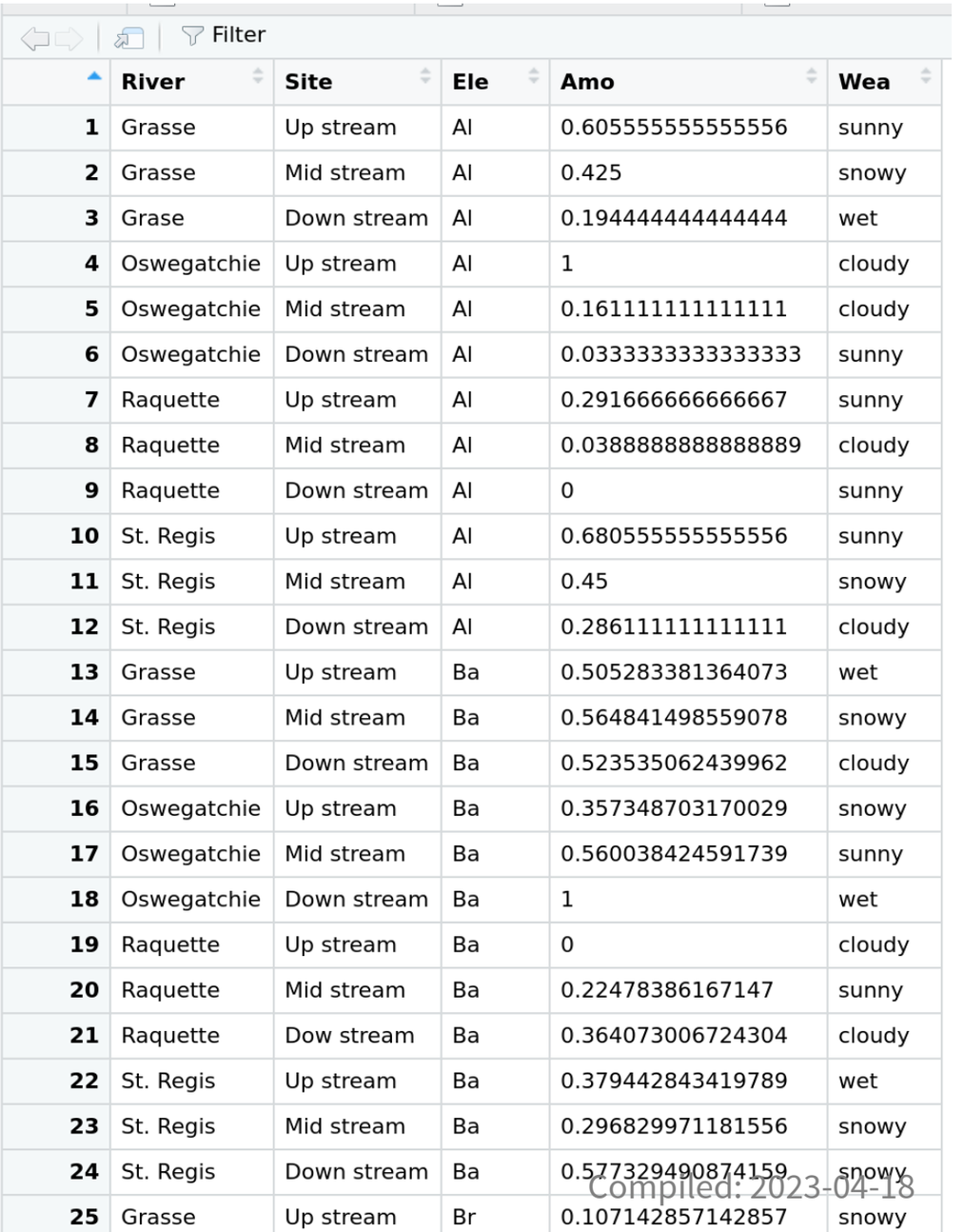

Filter

	River	Site	Ele	Amo	Wea
1	Grasse	Up stream	Al	0.605555555555556	sunny
2	Grasse	Mid stream	Al	0.425	snowy
3	Grase	Down stream	Al	0.194444444444444	wet
4	Oswegatchie	Up stream	Al	1	cloudy
5	Oswegatchie	Mid stream	Al	0.161111111111111	cloudy
6	Oswegatchie	Down stream	Al	0.0333333333333333	sunny
7	Raquette	Up stream	Al	0.291666666666667	sunny
8	Raquette	Mid stream	Al	0.0388888888888889	cloudy
9	Raquette	Down stream	Al	0	sunny
10	St. Regis	Up stream	Al	0.680555555555556	sunny
11	St. Regis	Mid stream	Al	0.45	snowy
12	St. Regis	Down stream	Al	0.286111111111111	cloudy
13	Grasse	Up stream	Ba	0.505283381364073	wet
14	Grasse	Mid stream	Ba	0.564841498559078	snowy
15	Grasse	Down stream	Ba	0.523535062439962	cloudy
16	Oswegatchie	Up stream	Ba	0.357348703170029	snowy
17	Oswegatchie	Mid stream	Ba	0.560038424591739	sunny
18	Oswegatchie	Down stream	Ba	1	wet
19	Raquette	Up stream	Ba	0	cloudy
20	Raquette	Mid stream	Ba	0.22478386167147	sunny
21	Raquette	Dow stream	Ba	0.364073006724304	cloudy
22	St. Regis	Up stream	Ba	0.379442843419789	wet
23	St. Regis	Mid stream	Ba	0.296829971181556	snowy
24	St. Regis	Down stream	Ba	0.577329490874159	snowy
25	Grasse	Up stream	Br	0.107142857142857	snowy

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First things first

- 💾 Save previous script
- 📂 Open New File

  (make sure you're in the RStudio Project)
- ✏️ Write `library(tidyverse)` at the top
- 💾 Save this new script

  (consider names like `cleaning.R` or `3_loading_and_cleaning.R`)

Side Note

R base vs. `tidyverse`

R base vs. `tidyverse`

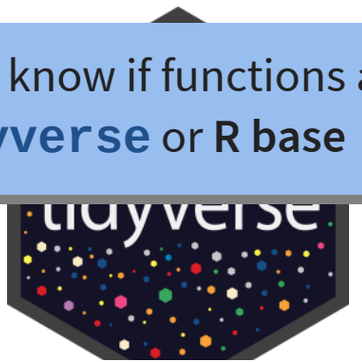
R base

- Basic R
- Packages are installed and loaded by default
- Base pipe `|>` *

`tidyverse`

- Collection of 'new' packages developed by a team closely affiliated with RStudio
  - e.g., `ggplot2`, `dplyr`, `tidyr`, `readr`
  - Packages designed to work well together
- Use a slightly different syntax
- tidyverse pipe `%>%` or base pipe `|>` *

Useful to know if functions are
`tidyverse` or **R base**



Dealing with data

1. Loading data

- Get your data into R

2. Looking for problems

- Typos
- Incorrectly loaded data

3. Fixing problems

- Corrections
- Renaming

4. Setting formats

- Dates
- Numbers
- Factors

5. Saving your data

Loading Data

Data types: What kind of data do you have?

Specific program files

Type	Extension	R Package	R function
Excel	.xls, .xlsx	`readxl`	`read_excel()`
Open Document	.ods	`readODS`	`read_ods()`
SPSS	.sav, .zsav, .por	`haven`	`read_spss()`
SAS	.sas7bdat	`haven`	`read_sas()`
Stata	.dta	`haven`	`read_dta()`
Database Files	.dbf	`foreign`	`read.dbf()`

Convenient but…

- Can be unreliable
- Can take longer

For files that don't change, better to save as a `*.csv` (Comma-separated-variables file)

Data types: What kind of data do you have?

General text files

Type	R base	`readr` package (`tidyverse`)
Comma separated	`read.csv()`	`read_csv()`, `read_csv2()`
Tab separated	`read.delim()`	`read_tsv()`
Space separated	`read.table()`	`read_table()`
Fixed-width	`read.fwf()`	`read_fwf()`

- **`readr` package** especially useful for big data sets (fast!)

We'll focus on

- `readxl` package → `read_excel()`
- `readr` package → `read_csv()`, `read_tsv()`

Where is my data?

Common error

```
1 my_data <- read_csv("weather.csv")
```

```
Error: 'weather.csv' does not exist in current working directory ('/home/steffi/Projects/Workshops/Two-day R
Workshop/qmd').
```

With no folder (just file name) R expects file to be in **Working directory**

Working directory is:

- Where your RStudio project is
- Your home directory (My Documents, etc.) [If not using RStudio Projects]
- Where you've set it (using `setwd()` or RStudio's Session > Set Working Directory)

Do use Projects in RStudio
Don't use `setwd()`

Where is my data?

A note on file paths (file locations)

```
/home
```

- folders separated by `/`
- `home` is a folder

Where is my data?

A note on file paths (file locations)

```
/home/steffi/
```

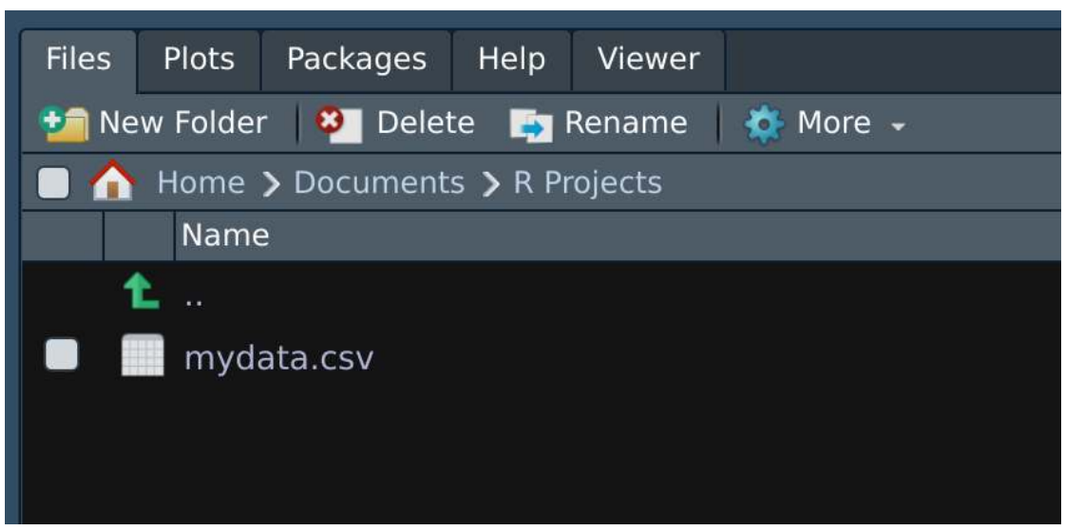
- folders separated by `/`
- `home` and `steffi` are folders
- `steffi` is a folder inside of `home`

Where is my data?

A note on file paths (file locations)

```
/home/steffi/Documents/R Projects/mydata.csv
```

- folders separated by `/`
- `home`, `steffi`, `Documents`, `R Projects` are folders
- `steffi` is inside of `home`, `Documents` is inside of `steffi`, etc.
- `mydata.csv` is a data file inside `R Projects` folder



RStudio Files Pane

Where is my data?

Absolute Paths

OS	Path
LINUX	/home/steffi/Documents/R Projects/mydata.csv
WINDOWS	C:/Users/steffi/My Documents/R Projects/mydata.csv
MAC	/users/steffi/Documents/R Projects/mydata.csv

Full location, folders and filename

Relative Paths

Path	Where to look
mydata.csv	Here (current directory)
../mydata.csv	Go up one directory (../)
data/mydata.csv	Stay here, go into “data” folder (data/)
../data/mydata.csv	Go up one directory (../), then into “data” folder (d

Only *relative* info
Use relative symbols (e.g., `../`)

With RStudio ‘Projects’ only need to use **relative** paths

Keep yourself organized

For simple projects

- Create an ‘RStudio Project’ for each Project (Chapter, Thesis, etc.)
- Create a specific “Data” folder within each project (one per project)

```
- Prospect Lake Quality                 # Project Folder
    - prospect_analysis.R
    - data                              # Data Folder
        - prospect_data_2017-01-01.csv
        - prospect_data_2017-02-01.csv
```

- Use **relative** paths to refer to this folder

```
d <- read_csv("data/prospect_data_2017-01-01.csv")
```

Let’s Load Some Data!

Your turn: Load some data

1. Create a '**data**' folder in your RStudio project
2. Put `rivers_correct.xlsx` file in the "**data**" folder
3. Load the package

```
1  library(readxl)
```

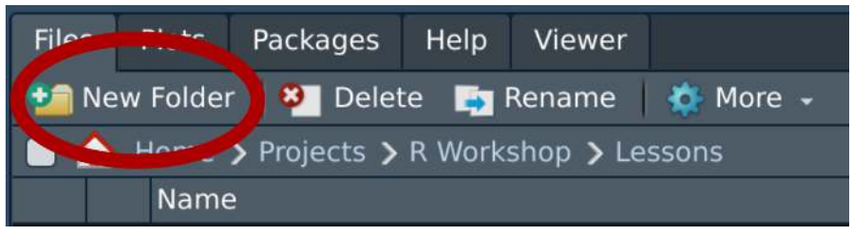
4. Read in the Excel file and assign to object `rivers`

```
1  rivers <- read_excel("data/rivers_correct.xlsx")
```

5. Use `head()` and `tail()` functions to look at the data

   e.g., `head(rivers)` and `tail(rivers)`

6. Click on the `rivers` object in your "Environment" pane to look at the whole data set



Click on "New Folder"

Use the *'tab'* key in RStudio when typing in the file name for auto-complete

How do I know which function to use?

Look at the file extension:

- `rivers_correct.csv`
- `.csv` → Comma-separated-variables → `read_csv()`

But not always obvious…

How do I know which function to use?

Look at the file: `master_moch.txt`

- Put this file in your **`data`** folder
- In lower right-hand pane, click on **Files**
  - Click on **data** folder
  - Click on **master_moch.txt**
  - Click “View File” (if asked)

```
ID  region  hab freq    freq.sd p.notes
MCB02   kam 0.5266879074    3.9806600009    3.9806600009    0.4592592593
MCB03   kam -0.9707703735   4.1090031783    4.1090031783    0.5
MCB04   kam -0.9707703735   4.2463067674    4.2463067674    0.5151515152
```

This **does not** read the file into R, but only shows you the contents as text.

Hmm, not comma-separated, maybe tab-separated?

How do I know what to use?

Peak:

- Pick a read function with your best guess (`read_csv()` is a good start)
- Use `n_max` to read only first few rows

```
  1  read_csv("data/master_moch.txt", n_max = 3)
```

```
# A tibble: 3 × 1
  `ID\tregion\thab\tfreq\tfreq.sd\tp.notes`
  <chr>
1 "MCB02\tkam\t0.5266879074\t3.9806600009\t3.9806600009\t0.4592592593"
2 "MCB03\tkam\t-0.9707703735\t4.1090031783\t4.1090031783\t0.5"
3 "MCB04\tkam\t-0.9707703735\t4.2463067674\t4.2463067674\t0.5151515152"
```

`\t` means tab, so this is tab-separated data

How do I know what to use?

Peak:

- Try again with `read_tsv()`

```
  1  read_tsv("data/master_moch.txt", n_max = 3)  # note change in function!
```

```
# A tibble: 3 × 6
  ID    region    hab  freq freq.sd p.notes
  <chr> <chr>   <dbl> <dbl>   <dbl>   <dbl>
1 MCB02 kam     0.527  3.98    3.98   0.459
2 MCB03 kam    -0.971  4.11    4.11   0.5
3 MCB04 kam    -0.971  4.25    4.25   0.515
```

Excellent!

Specifics of loading functions

col_names

- Geolocator data

```
1 my_data <- read_csv("data/geolocators.csv")
2 my_data
```

```
# A tibble: 20 × 2
   `02/05/11 22:29:59`  `64`
   <chr>               <dbl>
 1 02/05/11 22:31:59      64
 2 02/05/11 22:33:59      38
 3 02/05/11 22:35:59      38
 4 02/05/11 22:37:59      34
 5 02/05/11 22:39:59      30
 6 02/05/11 22:41:59      34
 7 02/05/11 22:43:59      40
 8 02/05/11 22:45:59      46
 9 02/05/11 22:47:59      48
10 02/05/11 22:49:59      46
# ℹ 10 more rows
```

Oops?

- `read_csv`, `read_tsv`, etc. assume that the first row contains the column names
- This file doesn't have headers

col_names

- Geolocator data

Declare no headings

```
my_data <- read_csv("data/geolocators.csv",
                    col_names = FALSE)
my_data
```

```
# A tibble: 21 × 2
   X1                  X2
   <chr>            <dbl>
 1 02/05/11 22:29:59    64
 2 02/05/11 22:31:59    64
 3 02/05/11 22:33:59    38
 4 02/05/11 22:35:59    38
 5 02/05/11 22:37:59    34
 6 02/05/11 22:39:59    30
 7 02/05/11 22:41:59    34
 8 02/05/11 22:43:59    40
 9 02/05/11 22:45:59    46
10 02/05/11 22:47:59    48
# ℹ 11 more rows
```

Name headings

```
my_data <- read_csv("data/geolocators.csv",
                    col_names = c("date", "light"))
my_data
```

```
# A tibble: 21 × 2
   date                light
   <chr>               <dbl>
 1 02/05/11 22:29:59    64
 2 02/05/11 22:31:59    64
 3 02/05/11 22:33:59    38
 4 02/05/11 22:35:59    38
 5 02/05/11 22:37:59    34
 6 02/05/11 22:39:59    30
 7 02/05/11 22:41:59    34
 8 02/05/11 22:43:59    40
 9 02/05/11 22:45:59    46
10 02/05/11 22:47:59    48
# ℹ 11 more rows
```

`skip` info rows before data

- Grain size data

```
my_data <- read_tsv("data/grain_size.txt")
my_data
```

```
# A tibble: 36 × 7
   `DATA DOWNLOAD: 2015-09-23` ...2  ...3        ...4     ...5  ...6  ...7
   <chr>                       <chr> <chr>       <chr>    <chr> <chr> <chr>
 1 SYSTEM 001                  <NA>  <NA>        <NA>     <NA>  <NA>  <NA>
 2 LOGGER X                    <NA>  <NA>        <NA>     <NA>  <NA>  <NA>
 3 lab_num                     CSP   sample_num  depth_lb csa   msa   fsa
 4 3177                        CSP01 CSP01-P-1-1 4        13.04 17.37 8.19
 5 3178                        CSP01 CSP01-P-1-2 12       10.74 16.9  7.92
 6 3179                        CSP01 CSP01-P-1-3 35       12.11 17.75 6.99
 7 3180                        CSP01 CSP01-P-1-4 53       17.61 18.16 6.29
 8 3181                        CSP01 CSP01-P-1-5 83       21.05 18.38 6.26
 9 3182                        CSP01 CSP01-P-1-6 105      19.02 18.43 6.28
10 3183                        CSP08 CSP08-P-1-1 10       11.6  17.14 8.18
# ℹ 26 more rows
```

`skip` info rows before data

- Grain size data

```
1 my_data <- read_tsv("data/grain_size.txt")
2 my_data
```

Look at the file:

- Click on **Files** tab
- Click on **data** folder
- Click on **grain_size.txt**
- Click **"View file"** (if asked)

```
DATA DOWNLOAD: 2015-09-23
SYSTEM 001
LOGGER X
lab_num CSP sample_num  depth_lb    csa msa fsa
3177    CSP01   CSP01-P-1-1 4   13.04   17.37   8.19
3178    CSP01   CSP01-P-1-2 12  10.74   16.9    7.92
3179    CSP01   CSP01-P-1-3 35  12.11   17.75   6.99
3180    CSP01   CSP01-P-1-4 53  17.61   18.16   6.29
3181    CSP01   CSP01-P-1-5 83  21.05   18.38   6.26
```

Ah ha!
Metadata was stored at the top of the file

`skip` info rows before data

- Grain size data
- Add `skip = 3` to skip the first three rows

```
my_data <- read_tsv("data/grain_size.txt", skip = 3)
my_data
```

```
# A tibble: 33 × 7
   lab_num CSP   sample_num  depth_lb   csa   msa   fsa
     <dbl> <chr> <chr>          <dbl> <dbl> <dbl> <dbl>
 1    3177 CSP01 CSP01-P-1-1        4 13.0  17.4   8.19
 2    3178 CSP01 CSP01-P-1-2       12 10.7  16.9   7.92
 3    3179 CSP01 CSP01-P-1-3       35 12.1  17.8   6.99
 4    3180 CSP01 CSP01-P-1-4       53 17.6  18.2   6.29
 5    3181 CSP01 CSP01-P-1-5       83 21.0  18.4   6.26
 6    3182 CSP01 CSP01-P-1-6      105 19.0
 7    3183 CSP08 CSP08-P-1-1       10 11.6
 8    3184 CSP08 CSP08-P-1-2       27 15.4
 9    3185 CSP08 CSP08-P-1-3       90 14.9  15.8   7.12
10    3186 CSP02 CSP02-P-1-1        5  8.75  8.64  3.41
# ℹ 23 more rows
```

Much better!

Your turn: Load this data set

Load the telemetry data set: `Sta A Data 2006-11-07.dmp`

1. Look at the file
2. Decide which R function to use based on delimiter (comma, space, or tab?)
3. Any other options need to be specified?

It should look like this:

```
# A tibble: 19 × 7
  StartDate Time     Frequency `Rate/Temp`   Pwr Ant      SD
      <dbl> <time>       <dbl>       <dbl> <dbl> <chr> <dbl>
1     39022 17:15:36      150.        34.8   175 M0        0
2     39022 17:19:14      148.        19.2    72 M0        0
3     39022 17:19:25      148.        19.7   194 M1        0
4     39022 17:20:04      149.        33.8   104 M0        0
5     39022 17:20:17      149.        33.7   152 M1        0
6     39022 17:20:57      150.        34.2   188 M0        0
7     39022 17:22:50      148.         9.8   188 M0        0
# ℹ 12 more rows
```

Too Easy?

Load some of your own tricky data

Looking for problems

Look at the data

- Make sure columns as expected (correctly assigned file format)
- Make sure no extra lines above the data (should we have used a skip?)
- Make sure column names look appropriate

```
1 library(palmerpenguins)
2 penguins
```

```
# A tibble: 344 × 8
   species island    bill_length_mm bill_depth_mm flipper_length_mm body_mass_g sex     year
   <fct>   <fct>              <dbl>         <dbl>             <int>       <int> <fct>  <int>
 1 Adelie  Torgersen           39.1          18.7               181        3750 male    2007
 2 Adelie  Torgersen           39.5          17.4               186        3800 female  2007
 3 Adelie  Torgersen           40.3          18                 195        3250 female  2007
 4 Adelie  Torgersen           NA            NA                  NA          NA <NA>    2007
 5 Adelie  Torgersen           36.7          19.3               193        3450 female  2007
 6 Adelie  Torgersen           39.3          20.6               190        3650 male    2007
 7 Adelie  Torgersen           38.9          17.8               181        3625 female  2007
 8 Adelie  Torgersen           39.2          19.6               195        4675 male    2007
 9 Adelie  Torgersen           34.1          18.1               193        3475 <NA>    2007
10 Adelie  Torgersen           42            20.2               190        4250 <NA>    2007
# ℹ 334 more rows
```

Look at the data

- Did the whole data set load?
- Are there extra blank lines at the end of the data?

```
1 tail(penguins)
```

```
# A tibble: 6 × 8
  species   island bill_length_mm bill_depth_mm flipper_length_mm body_mass_g sex     year
  <fct>     <fct>           <dbl>         <dbl>             <int>       <int> <fct>  <int>
1 Chinstrap Dream            45.7          17                 195        3650 female  2009
2 Chinstrap Dream            55.8          19.8               207        4000 male    2009
3 Chinstrap Dream            43.5          18.1               202        3400 female  2009
4 Chinstrap Dream            49.6          18.2               193        3775 male    2009
5 Chinstrap Dream            50.8          19                 210        4100 male    2009
6 Chinstrap Dream            50.2          18.7               198        3775 female  2009
```

skim()

- Are the formats correct?
  - numbers (`numeric`),
  - text (`character`)
  - date (`date`, `POSIXct`, `datetime`)
  - categories (`factor`)
- Are values appropriate?
  - Should there be `NA`s?
- Are there any typos?
- Number of rows expected?

```
1 library(skimr)
2 skim(penguins)
```

```
── Data Summary ────────────────────────
                           Values
Name                       penguins
Number of rows             344
Number of columns          8
_________________________
Column type frequency:
  factor                   3
  numeric                  5
_________________________
Group variables            None

── Variable type: factor ─────────────────────────────────────────────────────────────────────────────────────────────────────────────
  skim_variable n_missing complete_rate ordered n_unique top_counts
1 species               0         1     FALSE          3 Ade: 152, Gen: 124, Chi: 68
2 island                0         1     FALSE          3 Bis: 168, Dre: 124, Tor: 52
3 sex                  11         0.968 FALSE          2 mal: 168, fem: 165

── Variable type: numeric ────────────────────────────────────────────────────────────────────────────────────────────────────────────
  skim_variable     n_missing complete_rate   mean      sd     p0    p25    p50    p75   p100 hist
1 bill_length_mm            2         0.994   43.9    5.46    32.1   39.2   44.4   48.5   59.6 ▃▇▇▆▁
2 bill_depth_mm             2         0.994   17.2    1.97    13.1   15.6   17.3   18.7   21.5 ▅▇▇▆▂
3 flipper_length_mm         2         0.994  201.   14.1    172    190    197    213    231    ▂▇▃▅▂
4 body_mass_g               2         0.994 4202.  802.    2700   3550   4050   4750   6300   ▃▇▆▅▂
5 year                      0         1     2008.    0.818 2007   2007   2008   2009   2009   ▇▁▇▁▇
```

count()

- Check for sample sizes and potential typos in categorical columns

```
  1  count(penguins, species)
```

```
# A tibble: 3 × 2
  species       n
  <fct>     <int>
1 Adelie      152
2 Chinstrap    68
3 Gentoo      124
```

```
  1  count(penguins, island)
```

```
# A tibble: 3 × 2
  island        n
  <fct>     <int>
1 Biscoe      168
2 Dream       124
3 Torgersen    52
```

Example of problematic data

```
rivers <- read_csv("data/rivers_correct.csv")
rivers
```

```
# A tibble: 300 × 5
   `River Name` Site        Ele   Amo               Wea
   <chr>        <chr>       <chr> <chr>             <chr>
 1 Grasse       Up stream   Al    0.605555555555556 sunny
 2 Grasse       Mid stream  Al    0.425             cloudy
 3 Grase        Down stream Al    0.194444444444444 sunny
 4 Oswegatchie  Up stream   Al    1                 cloudy
 5 Oswegatchie  Mid stream  Al    0.161111111111111 snowy
 6 Oswegatchie  Down stream Al    0.0333333333333333 cloudy
 7 Raquette     Up stream   Al    0.291666666666667 cloudy
 8 Raquette     Mid stream  Al    0.0388888888888889 sunny
 9 Raquette     Down stream Al    0                 snowy
10 St. Regis    Up stream   Al    0.680555555555556 wet
# ℹ 290 more rows
```

- Column names are not great (`River Name` not R-friendly) or obvious (what is `Ele`?)
- `Amo` - should be numeric but isn't
- At least one typo in River (`Grase` should be `Grasse`)

Example of problematic data

```
1 skim(rivers)
```

```
── Variable type: character ──────────────────────────────────────────────────────────────────────────────────────────────────────────────
  skim_variable n_missing complete_rate min max empty n_unique whitespace
1 River Name            0          1      5  11     0        7          0
2 Site                  0          1      9  11     0        3          0
3 Ele                   0          1      1   2     0       25          0
4 Amo                  12          0.96   1  19     0      198          0
5 Wea                   0          1      3   6     0        4          0
```

- Not much additional info here

Example of problematic data

```
1 count(rivers, `River Name`)
```

```
# A tibble: 7 × 2
  `River Name`     n
  <chr>        <int>
1 Grase            1
2 Grasse          73
3 Oswegatchie     75
4 Raquette        74
5 St. Regis       75
6 grasse           1
7 raquette         1
```

```
1 count(rivers, Site)
```

```
# A tibble: 3 × 2
  Site            n
  <chr>       <int>
1 Down stream   100
2 Mid stream    100
3 Up stream     100
```

Typos in `River Name`

Fixing problems

Cleaning column names

`clean_names()` from the `janitor` package

```
1 library(janitor)
2 rivers <- clean_names(rivers)
3 rivers
```

```
# A tibble: 300 × 5
   river_name  site        ele   amo                wea
   <chr>       <chr>       <chr> <chr>              <chr>
 1 Grasse      Up stream   Al    0.605555555555556  sunny
 2 Grasse      Mid stream  Al    0.425              cloudy
 3 Grase       Down stream Al    0.194444444444444  sunny
 4 Oswegatchie Up stream   Al    1                  cloudy
 5 Oswegatchie Mid stream  Al    0.161111111111111  snowy
 6 Oswegatchie Down stream Al    0.0333333333333333 cloudy
 7 Raquette    Up stream   Al    0.291666666666667  cloudy
 8 Raquette    Mid stream  Al    0.0388888888888889 sunny
 9 Raquette    Down stream Al    0                  snowy
10 St. Regis   Up stream   Al    0.680555555555556  wet
# ℹ 290 more rows
```

Cleaning column names

`rename()` columns

```
1 rivers <- rename(rivers, element = ele, amount = amo)
2 rivers
```

```
# A tibble: 300 × 5
   river_name  site        element amount             wea
   <chr>       <chr>       <chr>   <chr>              <chr>
 1 Grasse      Up stream   Al      0.605555555555556  sunny
 2 Grasse      Mid stream  Al      0.425              cloudy
 3 Grase       Down stream Al      0.194444444444444  sunny
 4 Oswegatchie Up stream   Al      1                  cloudy
 5 Oswegatchie Mid stream  Al      0.161111111111111  snowy
 6 Oswegatchie Down stream Al      0.0333333333333333 cloudy
 7 Raquette    Up stream   Al      0.291666666666667  cloudy
 8 Raquette    Mid stream  Al      0.0388888888888889 sunny
 9 Raquette    Down stream Al      0                  snowy
10 St. Regis   Up stream   Al      0.680555555555556  wet
# ℹ 290 more rows
```

Subsetting columns

`select()`* columns you do want

```
1 rivers <- select(rivers, river_name, site, element, amount)
```

OR, un`select()` columns you don't want

```
1 rivers <- select(rivers, -wea)
2 rivers
```

```
# A tibble: 300 × 4
   river_name  site        element amount
   <chr>       <chr>       <chr>   <chr>
 1 Grasse      Up stream   Al      0.605555555555556
 2 Grasse      Mid stream  Al      0.425
 3 Grase       Down stream Al      0.194444444444444
 4 Oswegatchie Up stream   Al      1
 5 Oswegatchie Mid stream  Al      0.161111111111111
 6 Oswegatchie Down stream Al      0.0333333333333333
 7 Raquette    Up stream   Al      0.291666666666667
 8 Raquette    Mid stream  Al      0.0388888888888889
 9 Raquette    Down stream Al      0
10 St. Regis   Up stream   Al      0.680555555555556
# ℹ 290 more rows
```

Cleaning columns

Put it all together

```
rivers <- read_csv("data/rivers_correct.csv")
rivers <- clean_names(rivers)
rivers <- rename(rivers, element = ele, amount = amo)
rivers <- select(rivers, -wea)
rivers
```

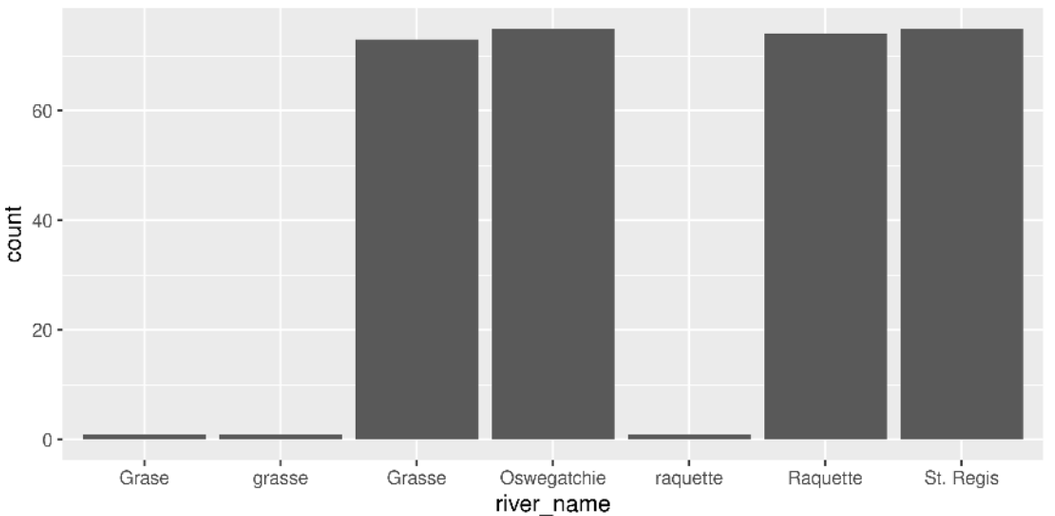
```
# A tibble: 300 × 4
   river_name  site        element amount
   <chr>       <chr>       <chr>   <chr>
 1 Grasse      Up stream   Al      0.605555555555556
 2 Grasse      Mid stream  Al      0.425
 3 Grase       Down stream Al      0.194444444444444
 4 Oswegatchie Up stream   Al      1
 5 Oswegatchie Mid stream  Al      0.161111111111111
 6 Oswegatchie Down stream Al      0.0333333333333333
 7 Raquette    Up stream   Al      0.291666666666667
 8 Raquette    Mid stream  Al      0.0388888888888889
 9 Raquette    Down stream Al      0
10 St. Regis   Up stream   Al      0.680555555555556
# ℹ 290 more rows
```

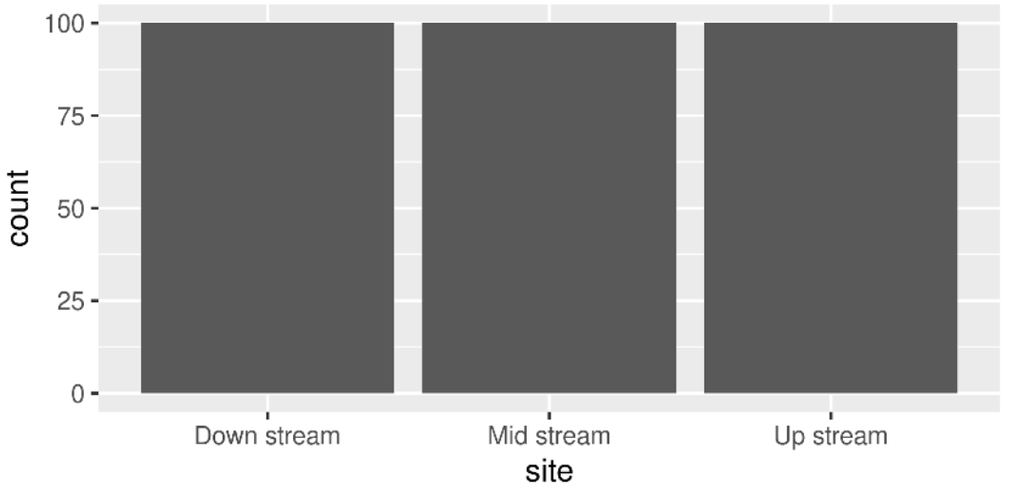
Note how code repeats data frame `rivers`...

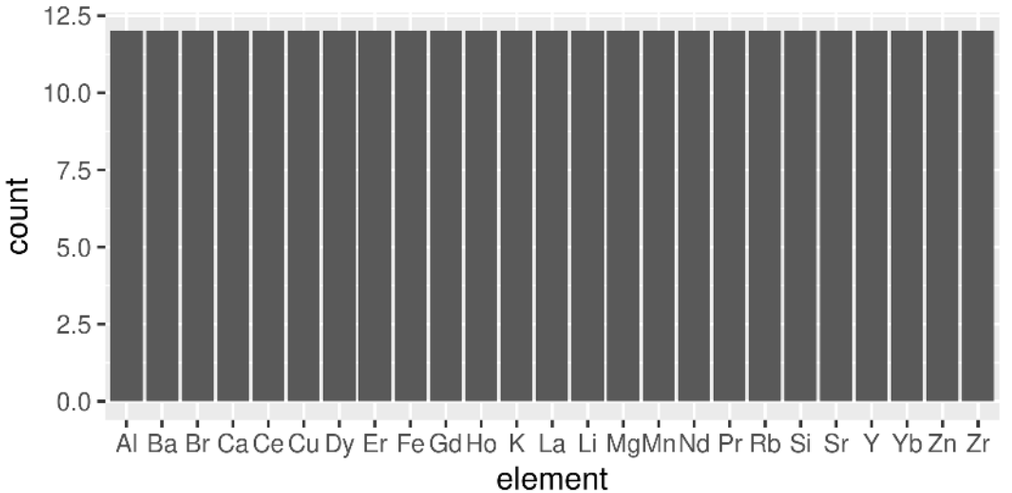
Fixing typos

Look for typos (Visually)

```
1  ggplot(data = rivers, aes(x = river_name)) + geom_bar()
2  ggplot(data = rivers, aes(x = site)) + geom_bar()
3  ggplot(data = rivers, aes(x = element)) + geom_bar()
```




100
75
50
25
0
count
Down stream
Mid stream
Up stream
site


12.5
10.0
7.5
5.0
2.5
0.0
count
Al Ba Br Ca Ce Cu Dy Er Fe Gd Ho K La Li Mg Mn Nd Pr Rb Si Sr Y Yb Zn Zr
element

Fixing typos

Look for typos with `count()`*

```
  1  count(rivers, river_name)
```

```
# A tibble: 7 × 2
  river_name      n
  <chr>       <int>
1 Grase           1
2 Grasse         73
3 Oswegatchie    75
4 Raquette       74
5 St. Regis      75
6 grasse          1
7 raquette        1
```

`filter()`* the data to highlight them

```
  1  filter(rivers, river_name == "Grase")
```

```
# A tibble: 1 × 4
  river_name site        element amount
  <chr>      <chr>       <chr>   <chr>
1 Grase      Down stream Al      0.194444444444444
```

Fixing typos

Replace typos

Combine the `if_else()` / `case_when()` functions with `mutate()` function

`mutate()` creates or changes columns in a data frame:

```
mutate(dataframe, column = new_values)
```

`if_else()` tests for a condition, and returns one value if FALSE and another if TRUE

```
if_else(condition, value_if_true, value_if_false)
```

`case_when()` tests for multiple conditions, and returns different values depending

```
case_when(condition1 ~ value_if_true1,
          condition2 ~ value_if_true2,
          condition3 ~ value_if_true3,
          TRUE ~ default_value)
```

Fixing typos

Replace typos

Combine the `if_else` function with the `mutate()` function

```
rivers <- mutate(rivers, river_name = if_else(river_name == "Grase", "Grasse", river_name))
```

Check that it's gone:

```
filter(rivers, river_name == "Grase")
```

```
# A tibble: 0 × 4
# ℹ 4 variables: river_name <chr>, site <chr>, element <chr>, amount <chr>
```

Iterative process

- Make some corrections
- Check the data
- Make some more corrections (either add to or modify existing code)

Your Turn: Fix another “Grasse” typo

1. Check the data with `count()`
2. Use `mutate()` and `if_else()` to fix the typo

```
rivers <- read_csv("data/rivers_correct.csv")
rivers <- clean_names(rivers)
rivers <- rename(rivers, element = ele, amount = amo)
rivers <- select(rivers, -wea)
rivers <- mutate(rivers, river_name = if_else(river_name == "Grase", "Grasse", river_name))

rivers <- mutate(???, ??? = ???)
```

Too Easy?
Examine and fix problems in your own data

Fixing typos

To be more efficient, fix all typos at once

```
rivers <- read_csv("data/rivers_correct.csv")
rivers <- clean_names(rivers)
rivers <- rename(rivers, element = ele, amount = amo)
rivers <- select(rivers, -wea)
rivers <- mutate(rivers,
                 river_name = if_else(river_name %in% c("Grase", "grasse"), "Grasse", river_name))
```

`==` compares one item to one other
`%in%` compares one item to many different ones

Tangent: `tidyverse` functions

`tidyverse` functions

`rename()`, `select()`, `mutate()`

- `tidyverse` functions always start with the **data**, followed by other arguments
- you can reference any **column** from '**data**'

```
rivers <- read_csv("data/rivers_correct.csv")
rivers <- clean_names(rivers)
rivers <- rename(rivers, element = ele, amount = amo)
rivers <- select(rivers, -wea)
rivers <- mutate(rivers,
                 if_else(river_name %in% c("Grase", "grasse"), "Grasse", river_name))
```

- `rename()` changes column names
- `select()` chooses columns to keep or to remove (with `-`)
- `mutate()` changes column contents

Why use `tidyverse` functions?

Pipes! `|>`* Allow you to string commands together

Instead of:

```
rivers <- read_csv("data/rivers_correct.csv")
rivers <- clean_names(rivers)
rivers <- rename(rivers, element = ele, amount = amo)
rivers <- select(rivers, -wea)
rivers <- mutate(rivers,
                 if_else(river_name %in% c("Grase", "grasse"), "Grasse", river_name))
```

We have:

```
rivers <- read_csv("data/rivers_correct.csv") |>
  clean_names() |>
  rename(element = ele, amount = amo) |>
  select(-wea) |>
  mutate(river_name = if_else(river_name %in% c("Grase", "grasse"), "Grasse", river_name))
```

Play around

Take a moment to play with this code in your console

Convert this:

```
rivers <- read_csv("data/rivers_correct.csv")
rivers <- clean_names(rivers)
rivers <- rename(rivers, element = ele, amount = amo)
rivers <- select(rivers, -wea)
rivers <- mutate(rivers,
                 river_name = if_else(river_name %in% c("Grase", "grasse"), "Grasse", river_name))
```

To this:

```
rivers <- read_csv("data/rivers_correct.csv") |>
  clean_names() |>
  rename(element = ele, amount = amo) |>
  select(-wea) |>
  mutate(river_name = if_else(river_name %in% c("Grase", "grasse"), "Grasse", river_name))
```

Back to the program!

Your turn: Fix the remaining typo

- Remember this is an iterative process (you may find your self reloading the data often)
- Find the typo (expect `river_name`: **Grasse**, **Oswegatchie**, **Raquette**, **St.Regis**)
- Add fix to code:

```
rivers <- read_csv("data/rivers_correct.csv") |>
  clean_names() |>
  rename(element = ele, amount = amo) |>
  select(-wea) |>
  mutate(river_name = if_else(river_name %in% c("Grase", "grasse"), "Grasse", river_name))
```

Remember…

Comparing single items

```
A == "hello"
A %in% "hello"
```

Comparing multiple items

```
A %in% c("hello", "bye"))
# NOT A == c("hello", "bye")
```

Too Easy?

Examine and fix problems in your own data

Fixing formats

Typos that affect classes (formats)

Look for problems

```
1 rivers
```

```
# A tibble: 300 × 4
   river_name  site        element amount
   <chr>       <chr>       <chr>   <chr>
 1 Grasse      Up stream   Al      0.605555555555556
 2 Grasse      Mid stream  Al      0.425
 3 Grasse      Down stream Al      0.194444444444444
 4 Oswegatchie Up stream   Al      1
 5 Oswegatchie Mid stream  Al      0.161111111111111
 6 Oswegatchie Down stream Al      0.0333333333333333
 7 Raquette    Up stream   Al      0.291666666666667
 8 Raquette    Mid stream  Al      0.0388888888888889
 9 Raquette    Down stream Al      0
10 St. Regis   Up stream   Al      0.680555555555556
# ℹ 290 more rows
```

Why all character (chr)?

Changing classes

Function	Input	Output
`as.character()`	Any vector	Text (Characters)
`as.numeric()`	Any vector (but returns NAs if not numbers)	Numbers
`as.logical()`	TRUE, FALSE, T, F, 0 (FALSE), any other number (all TRUE)	TRUE or FALSE
`as.factor()`	Any vector	Categories

For example…

```
1 a <- c(1, 2, 10)
2 as.character(a)
```

```
[1] "1"  "2"  "10"
```

```
1 as.numeric(a)
```

```
[1]  1  2 10
```

```
1 b <- c("hello", "bye", 1)
2 as.character(b)
```

```
[1] "hello" "bye"   "1"
```

```
1 as.numeric(b)
```

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

We'll deal with dates and times later…

Fixing numerical typos

Find the problem (when we don't know what they are)

- Make a new column and convert `amount` to numbers

```
1  rivers <- mutate(rivers, amount2 = as.numeric(amount))
```

```
Warning: There was 1 warning in `mutate()`.
ℹ In argument: `amount2 = as.numeric(amount)`.
Caused by warning:
! NAs introduced by coercion
```

`NAs introduced by coercion`
means the function was forced to create NAs.
This warning tells us that some values didn't convert to numbers

Fixing numerical typos

Find the problem (when we don't know what it is)

- Make a new column and convert `amount` to numbers
- Find out where the conversion didn't work

```
  1  filter(rivers, !is.na(amount), is.na(amount2))
```

```
# A tibble: 1 × 5
  river_name site       element amount amount2
  <chr>      <chr>      <chr>   <chr>    <dbl>
1 Raquette   Mid stream Ca      <0.1        NA
```

- `is.na()` is `TRUE` when the value is missing (`NA`)
- `!` turns a `TRUE` into a `FALSE` (and vice versa)
- This asks, which values are **not missing to begin with** (`!is.na(amount)`) but ***are* missing after** the conversion (`is.na(amount2)`)

Fixing numerical typos

Find the problem (when we know what it is):

```
1 filter(rivers, amount == "<0.1")
```

```
# A tibble: 1 × 5
  river_name site        element amount amount2
  <chr>      <chr>       <chr>   <chr>    <dbl>
1 Raquette   Mid stream Ca      <0.1        NA
```

Fix problem

```
1 rivers <- mutate(rivers, amount = if_else(amount == "<0.1", "0", amount))
```

Correct the class

```
1 rivers <- mutate(rivers, amount = as.numeric(amount))
```

Fixing numerical typos

Last, but not least, check…

```
1 rivers
```

```
# A tibble: 300 × 5
   river_name  site        element amount amount2
   <chr>       <chr>       <chr>    <dbl>   <dbl>
 1 Grasse      Up stream   Al       0.606   0.606
 2 Grasse      Mid stream  Al       0.425   0.425
 3 Grasse      Down stream Al       0.194   0.194
 4 Oswegatchie Up stream   Al       1       1
 5 Oswegatchie Mid stream  Al       0.161   0.161
 6 Oswegatchie Down stream Al       0.0333  0.0333
 7 Raquette    Up stream   Al       0.292   0.292
 8 Raquette    Mid stream  Al       0.0389  0.0389
 9 Raquette    Down stream Al       0       0
10 St. Regis   Up stream   Al       0.681   0.681
# ℹ 290 more rows
```

Put it together...

```
rivers <- read_csv("data/rivers_correct.csv") |>
  clean_names() |>
  rename(element = ele, amount = amo) |>
  select(-wea) |>
  mutate(river_name = case_when(river_name %in% c("Grase", "grasse") ~ "Grasse",
                                river_name == "raquette" ~ "Raquette",
                                TRUE ~ river_name),
         amount = if_else(amount == "<0.1", "0", amount),
         amount = as.numeric(amount))
```

And you have a clean, corrected data frame ready to use

- You have not changed the original data
- You have a **reproducible** record of all corrections
- You can alter these corrections at any time
- You have formatted your data for use in R
- Read these steps line by line to remind yourself what you did

Dates and Times

(Or why does R hate me?)

Dates and Times

- Date/times aren't always recognized as date/times

```
1 geolocators <- read_csv("data/geolocators.csv", col_names = c("time", "light"))
2 geolocators
```

```
# A tibble: 21 × 2
  time                light
  <chr>               <dbl>
1 02/05/11 22:29:59      64
2 02/05/11 22:31:59      64
3 02/05/11 22:33:59      38
4 02/05/11 22:35:59      38
5 02/05/11 22:37:59      34
6 02/05/11 22:39:59      30
# ℹ 15 more rows
```

Here `time` column is considered `chr` (character/text)


LUBRIDATE: wrangle times + dates!
YEARS
MONTHS
DAYS
Horst '18

`lubridate` package

- Part of `tidyverse`, but needs to be loaded separately
- Great for converting date/time formats

```
1 library(lubridate)
2 geolocators <- mutate(geolocators, time_formatted = dmy_hms(time))
3 geolocators
```

```
# A tibble: 21 × 3
  time              light time_formatted
  <chr>             <dbl> <dttm>
1 02/05/11 22:29:59    64 2011-05-02 22:29:59
2 02/05/11 22:31:59    64 2011-05-02 22:31:59
3 02/05/11 22:33:59    38 2011-05-02 22:33:59
4 02/05/11 22:35:59    38 2011-05-02 22:35:59
5 02/05/11 22:37:59    34 2011-05-02 22:37:59
6 02/05/11 22:39:59    30 2011-05-02 22:39:59
# ℹ 15 more rows
```

Now `time_formatted` column is considered `dttm` (Date/Time)

`lubridate` package

Generally, only the order of the **y**ear, **m**onth, **d**ay, **h**our, **m**inute, or **s**econd matters.

date/time	function	class
2018-01-01 13:09:11	`ymd_hms()`	dttm (POSIXct/POSIXt)
12/20/2019 10:00 PM	`mdy_hm()`	dttm (POSIXct/POSIXt)
31/01/2000 10 AM	`dmy_h()`	dttm (POSIXct/POSIXt)
31-01/2000	`dmy()`	Date

`lubridate` is smart enough to detect AMs and PMs

Saving data

(For the love of all that is good don’t *lose* that data!!!)

Saving data

Keep yourself organized

- Keep your R-created data in a **different** folder from your 'raw' data *
- If you have a lot going on, split your work into several scripts, and number the both the scripts AND the data sets produced:
- `1_cleaned.csv`
- `2_summarized.csv`
- `3_graphing.csv`

Save your data to file:

```
1 write_csv(rivers, "datasets/rivers_cleaned.csv")
```

Dealing with data

1. Loading data

- Get your data into R

2. Looking for problems

- Typos
- Incorrectly loaded data

3. Fixing problems

- Corrections
- Renaming

4. Setting formats

- Dates
- Numbers
- Factors

5. Saving your data

Wrapping up: Common mistakes

Assuming your data is in one format when it's not

- Print your data to the console and use `skim()` to explore the format of your data
- Use `skim()`, `count()`, `filter()`, `select()`, `ggplot()` to explore the content of your data

Wrapping up: Common mistakes

Confusing pipes with function arguments

- Pipes (`|>` or `%>%`) pass the *output* from one function as *input* to the next function:

```
my_data <- my_data |>                  # Pass my_data
  filter(my_column > 5) |>             # Pass my_data, filtered
  select(my_column, my_second_column)
```

- Arguments may be on different lines, but all part of *one* function

```
my_data <- my_data |>                   # Pass my_data
  mutate(my_column1 = if_else(...),     # No passing (no pipes!)
         my_column2 = if_else(...),     # Instead, give 3 arguments to mutate:
         my_column3 = if_else(...))     # Arguments separated by ",", and surrounded by ( )
```

Wrapping up: Further reading

- R for Data Science
  - Chapter 5: Transforming data
  - Chapter 8: RStudio Projects
  - Chapter 14: Strings
  - Chapter 15: Factors
  - Chapter 18: Pipes