

# Loading & Cleaning Data in R

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

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	River	Site	Ele	Amo	Wea
1	Grasse	Up stream	Al	0.6055555555555556	sunny
2	Grasse	Mid stream	Al	0.425	snowy
3	Grasse	Down stream	Al	0.1944444444444444	wet
4	Oswegatchie	Up stream	Al	1	cloudy
5	Oswegatchie	Mid stream	Al	0.1611111111111111	cloudy
6	Oswegatchie	Down stream	Al	0.0333333333333333	sunny
7	Raquette	Up stream	Al	0.2916666666666667	sunny
8	Raquette	Mid stream	Al	0.0388888888888889	cloudy
9	Raquette	Down stream	Al	0	sunny
10	St. Regis	Up stream	Al	0.6805555555555556	sunny
11	St. Regis	Mid stream	Al	0.45	snowy
12	St. Regis	Down stream	Al	0.2861111111111111	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	<code>readxl*</code>	<code>read_excel()</code>
Open Document	.ods	<code>readODS</code>	<code>read_ods()</code>
SPSS	.sav, .zsav, .por	<code>haven</code>	<code>read_spss()</code>
SAS	.sas7bdat	<code>haven</code>	<code>read_sas()</code>
Stata	.dta	<code>haven</code>	<code>read_dta()</code>
Database Files	.dbf	<code>foreign</code>	<code>read.dbf()</code>



## 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	<b>readr</b> package *
Comma separated	<code>read.csv()</code>	<code>read_csv()</code> , <code>read_csv2()</code>
Tab separated	<code>read.delim()</code>	<code>read_tsv()</code>
Space separated	<code>read.table()</code>	<code>read_table()</code>
Fixed-width	<code>read.fwf()</code>	<code>read_fwf()</code>



- **readr** package especially useful for big data sets (fast!)
- Error/warnings from **readr** are a bit more helpful

• 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/workshop-dealing-with-data').
```

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)

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

# Where is my data?

## A note on file paths (file locations)

```
1 /home
```

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

# Where is my data?

## A note on file paths (file locations)

```
1 /home/steffi/
```

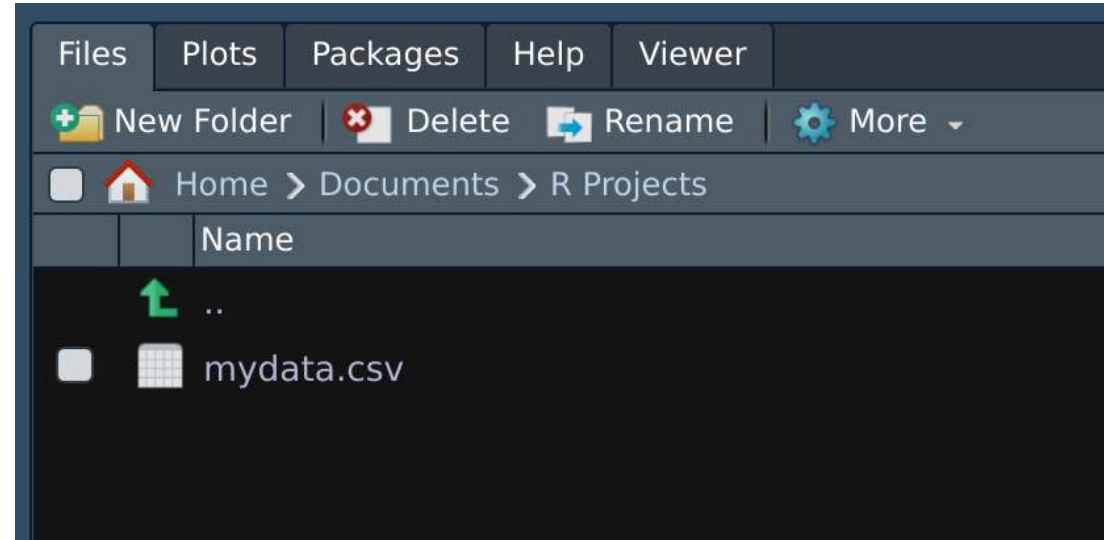
- folders separated by /
- `home` and `steffi` are folders
- `steffi` is a folder inside of `home`

# Where is my data?

## A note on file paths (file locations)

```
1 /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 (data/)

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

# 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)

```
1 - Prospect Lake Quality          # Project Folder
2   - prospect_analysis.R
3   - data                        # Data Folder
4     - prospect_data_2017-01-01.csv
5     - prospect_data_2017-02-01.csv
```

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

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

**Let's Load Some Data!**

# Your turn: Load some data

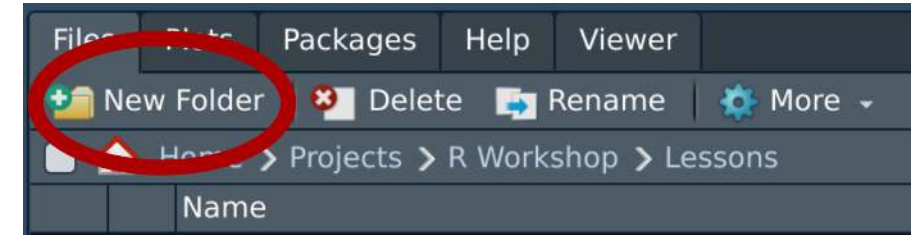
1. Create a 'data' folder in your RStudio project
2. Put `rivers_correct.xlsx` file in the "data" folder
3. Load the package

```
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



# Your turn: Load some data

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

```
1 head(rivers)
```

```
# A tibble: 6 × 7
  `River Name` Site      Ele      Amo `Temperature C°` Year Wea
  <chr>         <chr>    <chr>   <dbl>   <dbl> <dbl> <chr>
1 Grasse       Up stream  Al      0.606    10.9  2019 cloudy
2 Grasse       Mid stream Al      0.425     8.68  2020 cloudy
3 Grasse       Down stream Al      0.194     8.75  2021 snowy
4 Oswegatchie Up stream  Al       1      0.791  2022 sunny
5 Oswegatchie Mid stream Al      0.161     9.32  2023 cloudy
6 Oswegatchie Down stream Al      0.0333    10.6  2019 cloudy
```

```
1 tail(rivers)
```

```
# A tibble: 6 × 7
  `River Name` Site      Ele      Amo `Temperature C°` Year Wea
  <chr>         <chr>    <chr>   <dbl>   <dbl> <dbl> <chr>
1 Raquette     Up stream  Zr      0.333    14.0  2023 wet
2 Raquette     Mid stream Zr      0.111     7.61  2019 wet
3 Raquette     Down stream Zr      NA      7.36  2020 wet
4 St. Regis    Up stream  Zr      0.889     7.94  2021 sunny
5 St. Regis    Mid stream Zr      0.778     9.28  2022 snowy
6 St. Regis    Down stream Zr      0.667    10.1  2023 cloudy
```

	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

# 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.

# 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
```

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

Oops?

# col\_names

- Geolocator data

## Declare no headings

```
1 my_data <- read_csv("data/geolocators.csv",
2                       col_names = FALSE)
3 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

```
1 my_data <- read_csv("data/geolocators.csv",
2                       col_names = c("date", "light"))
3 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

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

```
# A tibble: 36 × 7
  `DATA DOWNLOAD: 2015-09-23` ...2 ...3 ...4 ...5 ...6 ...7
  <chr> <chr> <chr> <chr> <chr> <chr>
1 SYSTEM 001 <NA> <NA> <NA> <NA> <NA>
2 LOGGER X <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

```
1 my_data <- read_tsv("data/grain_size.txt", skip = 3)
2 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 18.4 6.28
7    3183 CSP08 CSP08-P-1-1        10 11.6 17.1 8.18
8    3184 CSP08 CSP08-P-1-2        27 15.4 16.2 6.76
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

# Your turn: Load this data set

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

```
1 telemetry <- read_csv("data/Sta A Data 2006-11-07.dmp", skip = 2)
2 telemetry
```

# 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

**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() the data

## skim() is from skimr

- 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 ( ) categories



## count ( ) is from dplyr\*

- Check for sample sizes and potential typos in categorical columns
- Assess missing values

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

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

```
1 count(penguins, sex)
```

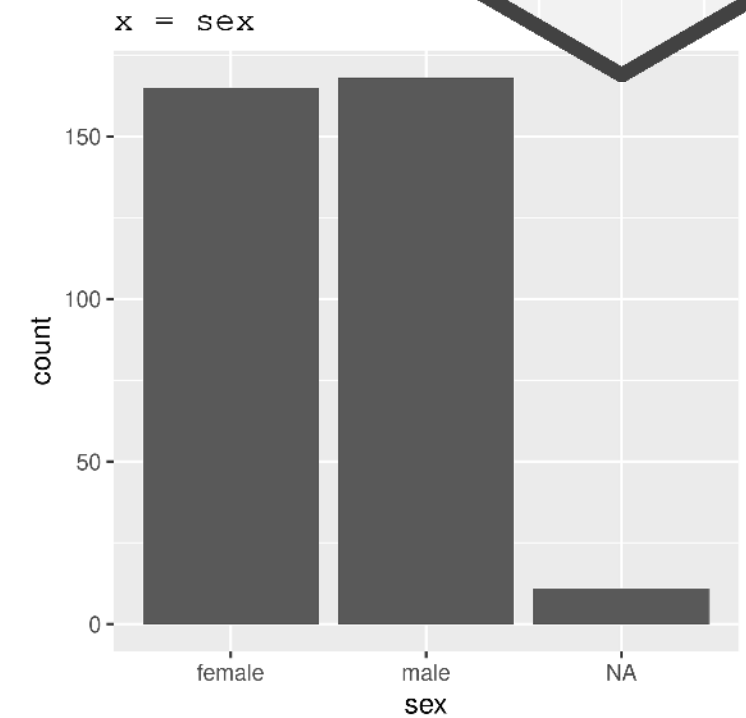
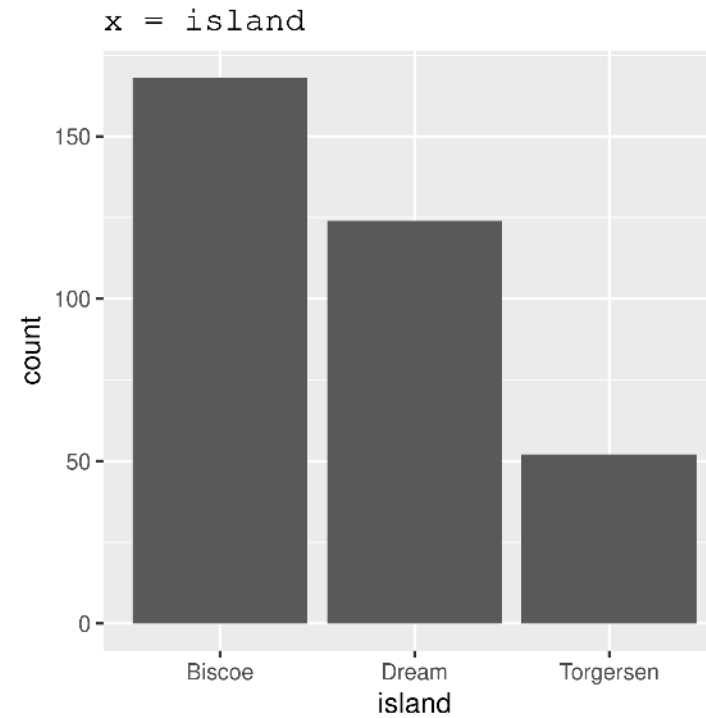
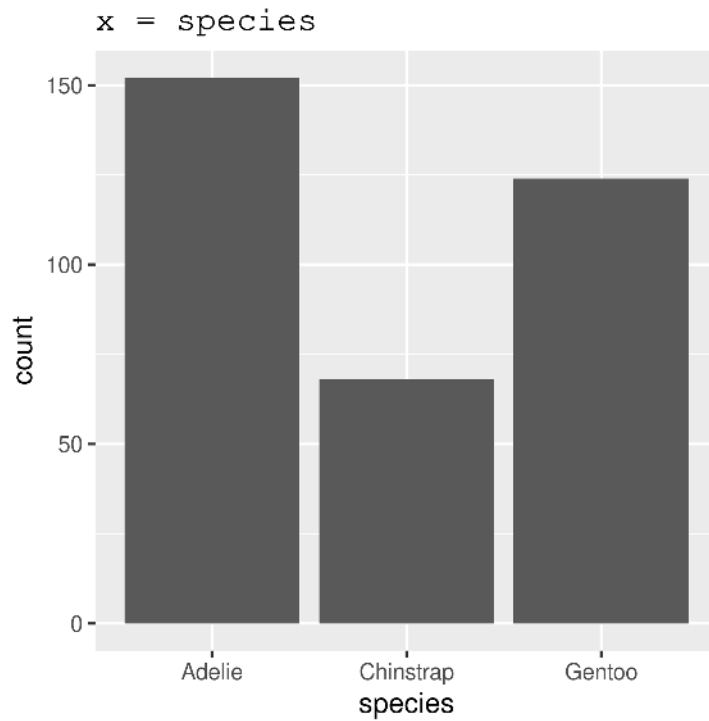
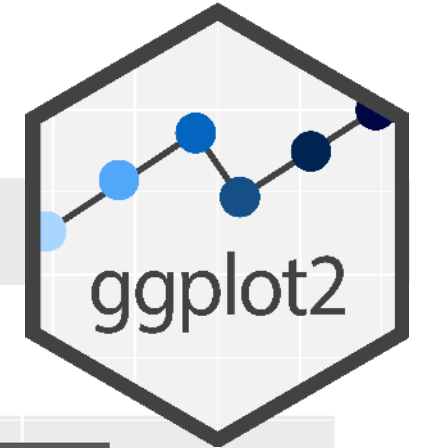
```
# A tibble: 3 × 2  
  sex      n  
  <fct> <int>  
1 female 165  
2 male   168  
3 <NA>    11
```

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

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

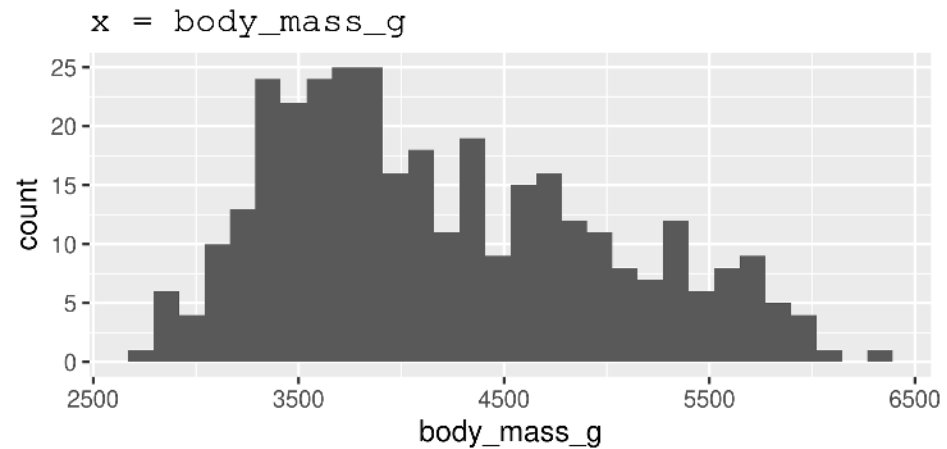
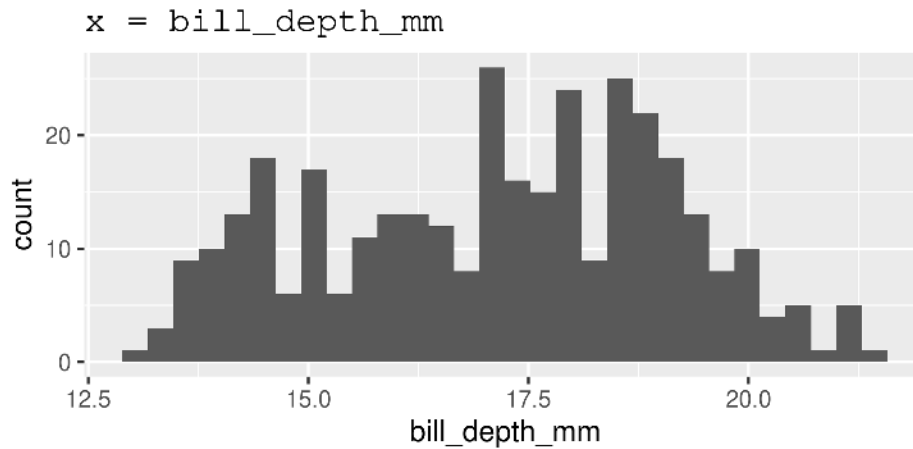
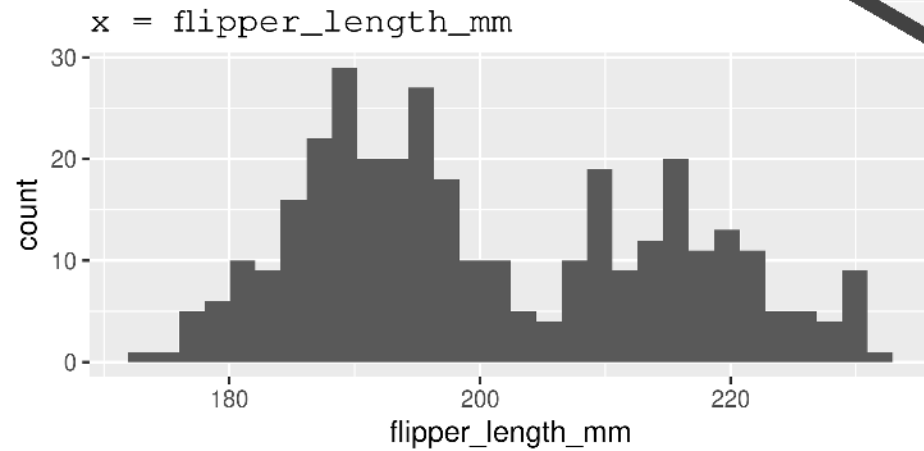
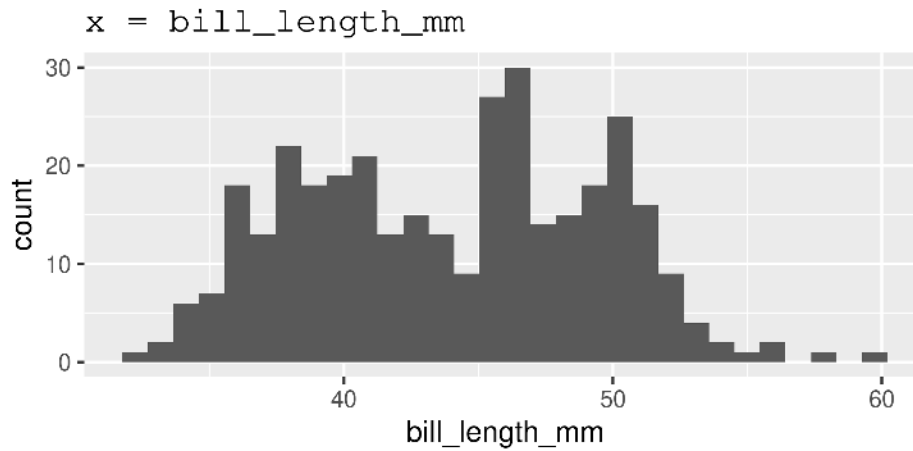
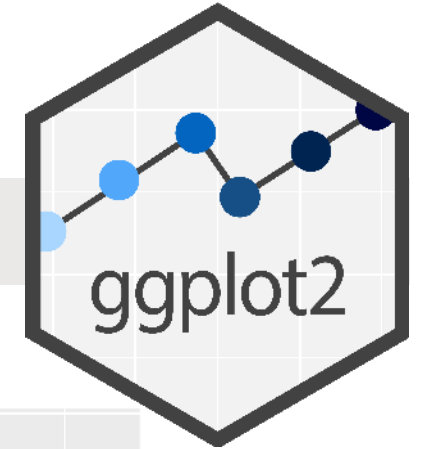
# Plot categories

```
1 ggplot(data = penguins, aes(x = COLUMN)) +  
2   geom_bar()
```



# Plot numbers

```
1 ggplot(data = penguins, aes(x = COLUMN)) +  
2   geom_histogram() ()
```



# Example of problematic data

`rivers_correct.csv`

# Look at the data

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

# A tibble: 300 × 7

	`River Name` <chr>	Site <chr>	Ele <chr>	Amo <dbl>	`Temperature C°` <dbl>	Year <dbl>	Wea <chr>
1	Grasse	Up stream	Al	0.606	10.9	2019	cloudy
2	Grasse	Mid stream	Al	0.425	8.68	2020	cloudy
3	Grase	Down stream	Al	0.194	8.75	2021	snowy
4	Oswegatchie	Up stream	Al	1	0.791	2022	sunny
5	Oswegatchie	Mid stream	Al	0.161	9.32	2023	cloudy
6	Oswegatchie	Down stream	Al	0.0333	10.6	2019	cloudy
7	Raquette	Up stream	Al	0.292	4.01	2020	sunny
8	Raquette	Mid stream	Al	0.0389	5.96	2021	cloudy
9	Raquette	Down stream	Al	NA	6.21	2022	cloudy
10	St. Regis	Up stream	Al	0.681	8.02	2023	cloudy

# i 290 more rows

- Column names are not great (**River Name** and **Temperature C°** are not R-friendly) or obvious (what is **Ele**?)
- At least one typo in River (**Grase** should be **Grasse**)

# Looking for problems

## Your Turn!

- `skim()` the data
- `count()` some columns
- Perhaps make some `ggplot()`s

Find any problems?

# skim() the data

```
1 skim(rivers)
```

— Data Summary —

Name

Number of rows

Number of columns

Values

rivers

300

7

Column type frequency:

character

numeric

4

3

Group variables

None

— Variable type: character —

skim\_variable

n\_missing

complete\_rate

min

max

empty

n\_unique

whitespace

1 River Name

2 Site

3 Ele

4 Wea

0

0

0

0

1

1

1

1

5

9

1

3

11

11

2

6

0

0

0

0

7

3

25

4

0

0

0

0

— Variable type: numeric —

skim\_variable

n\_missing

complete\_rate

mean

sd

p0

p25

p50

p75

p100

hist

1 Amo

2 Temperature C°

3 Year

39

0

0

0.87

1

1

0.429

9.17

2021

0.299

11.5

1.42

0.00656

-99

2019

0.169

7.54

2020

0.379

10.3

2021

0.643


12.7


2022


1

20.9

2023









# count() categories

```
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, Wea)
```

```
# A tibble: 4 × 2
  Wea      n
  <chr> <int>
1 cloudy  75
2 snowy   82
3 sunny   65
4 wet     78
```

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

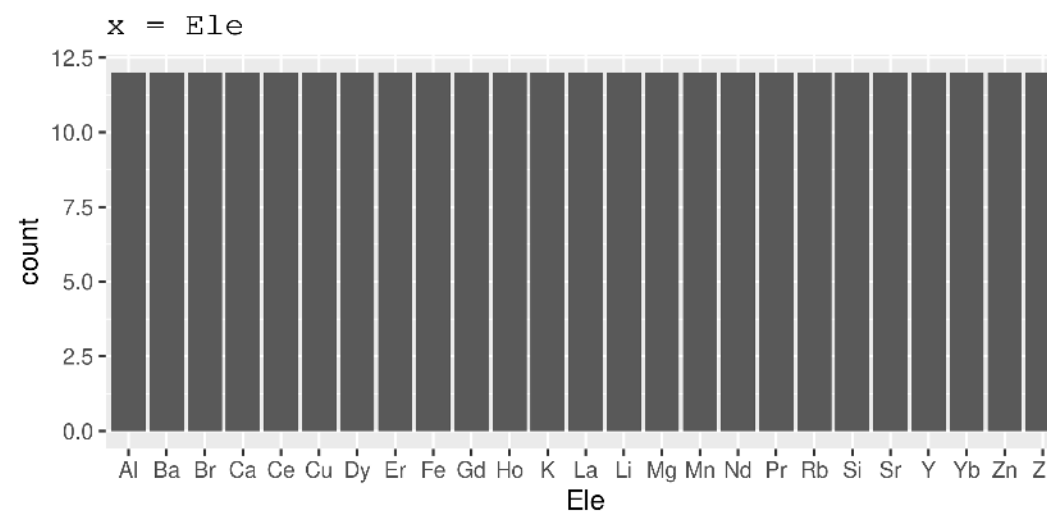
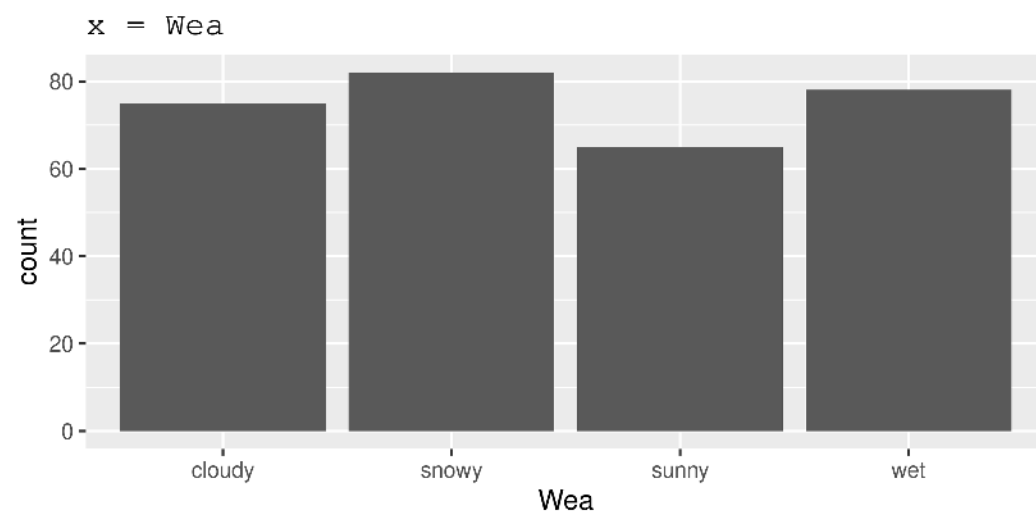
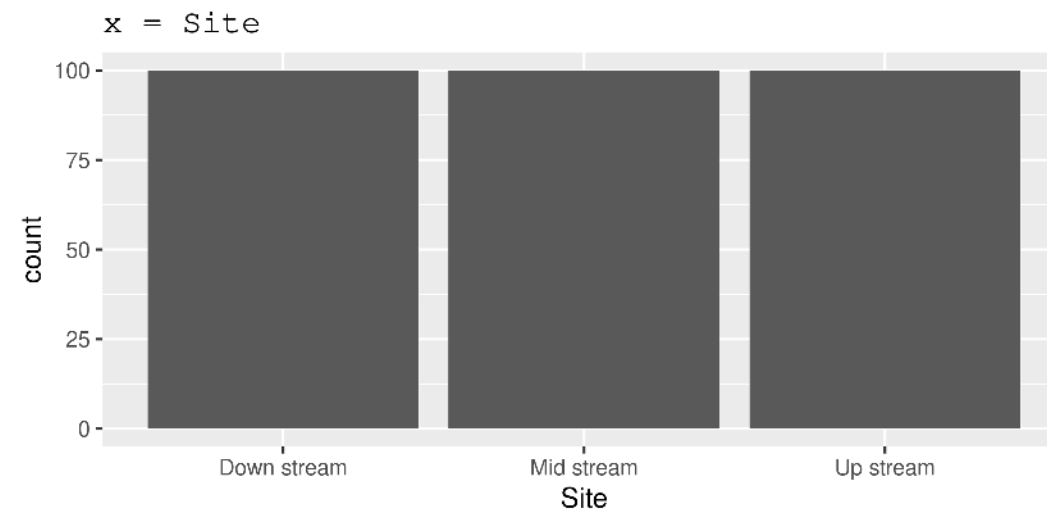
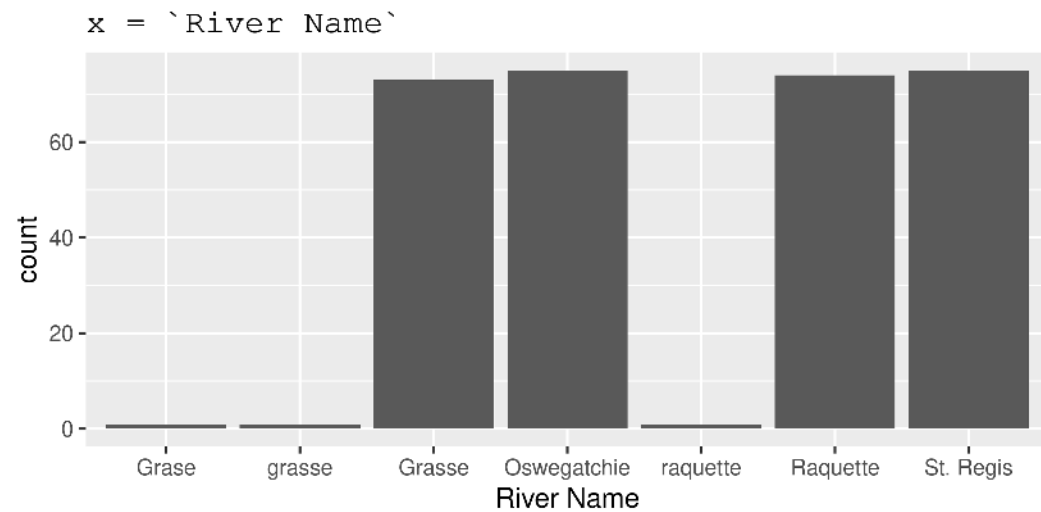
```
# A tibble: 3 × 2
  Site      n
  <chr> <int>
1 Down stream  100
2 Mid stream  100
3 Up stream   100
```

```
1 count(rivers, Ele)
```

```
# A tibble: 25 × 2
  Ele      n
  <chr> <int>
1 Al     12
2 Ba     12
3 Br     12
4 Ca     12
5 Ce     12
6 Cu     12
7 Dy     12
8 Er     12
9 Fe     12
10 Gd     12
# i 15 more rows
```

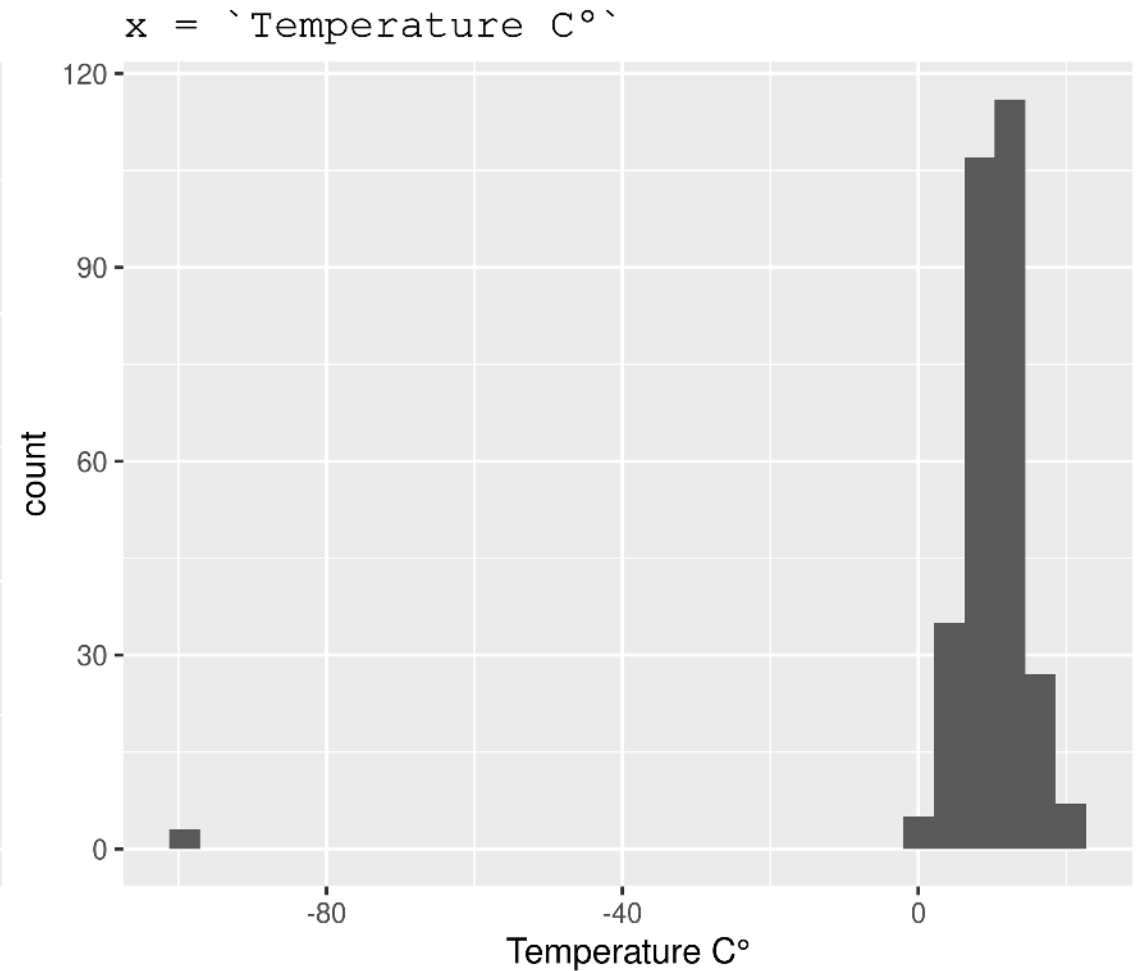
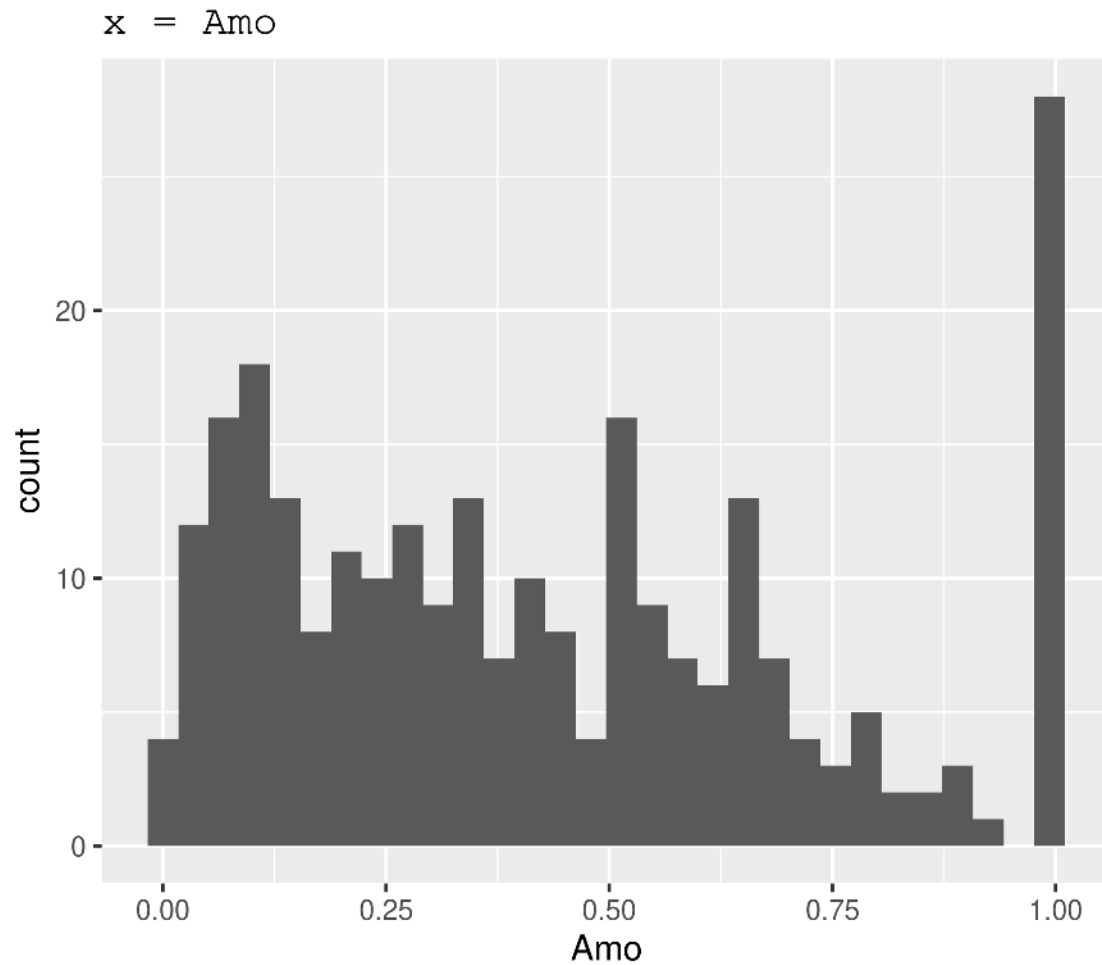
# Plot categories

```
1 ggplot(data = rivers, aes(x = COLUMN)) +  
2   geom_bar()
```



# Plot numbers

```
1 ggplot(data = rivers, aes(x = COLUMN)) +  
2   geom_histogram()()
```





# Fixing problems

`clean_names()` is from **janitor**\*

```

1 # A tibble: 300 × 7
2   river_name site      ele      amo temperature_c year wea
3   <chr>      <chr>      <chr>    <dbl>      <dbl> <dbl> <chr>
4 1 Grasse    Up stream    Al      0.606      10.9    2019 cloudy
5 2 Grasse    Mid stream   Al      0.425       8.68    2020 cloudy
6 3 Grasse    Down stream  Al      0.194       8.75    2021 snowy
7 4 Oswegatchie Up stream    Al      1          0.791    2022 sunny
8 5 Oswegatchie Mid stream   Al      0.161       9.32    2023 cloudy
9 6 Oswegatchie Down stream  Al      0.0333     10.6    2019 cloudy
10 7 Raquette  Up stream    Al      0.292       4.01    2020 sunny
11 8 Raquette  Mid stream   Al      0.0389       5.96    2021 cloudy
12 9 Raquette  Down stream  Al      NA          6.21    2022 cloudy
13 10 St. Regis Up stream    Al      0.681       8.02    2023 cloudy
14 # i 290 more rows

```



# Cleaning column names

`rename()` is from `dplyr`\*

`rename()` columns



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

# A tibble: 300 × 7
   river_name site      element amount temperature year wea
   <chr>      <chr>      <chr>    <dbl>      <dbl> <dbl> <chr>
1 Grasse     Up stream    Al      0.606      10.9    2019 cloudy
2 Grasse     Mid stream   Al      0.425       8.68    2020 cloudy
3 Grasse     Down stream  Al      0.194       8.75    2021 snowy
4 Oswegatchie Up stream    Al      1          0.791    2022 sunny
5 Oswegatchie Mid stream   Al      0.161       9.32    2023 cloudy
6 Oswegatchie Down stream  Al      0.0333     10.6    2019 cloudy
7 Raquette   Up stream    Al      0.292       4.01    2020 sunny
8 Raquette   Mid stream   Al      0.0389       5.96    2021 cloudy
9 Raquette   Down stream  Al      NA          6.21    2022 cloudy
10 St. Regis  Up stream    Al      0.681       8.02    2023 cloudy
# i 290 more rows
```



# Subsetting columns



`select()` is from **dplyr**\*

`select()` columns you want

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

OR, `unselect()` columns you don't want

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

```
# A tibble: 300 × 6
  river_name site      element amount temperature year
  <chr>      <chr>      <chr>      <dbl>      <dbl> <dbl>
1 Grasse    Up stream    Al         0.606      10.9  2019
2 Grasse    Mid stream   Al         0.425       8.68  2020
3 Grasse    Down stream  Al         0.194       8.75  2021
4 Oswegatchie Up stream    Al          1         0.791  2022
5 Oswegatchie Mid stream   Al         0.161       9.32  2023
6 Oswegatchie Down stream  Al         0.0333     10.6  2019
7 Raquette  Up stream    Al         0.292       4.01  2020
8 Raquette  Mid stream   Al         0.0389       5.96  2021
9 Raquette  Down stream  Al          NA        6.21  2022
10 St. Regis Up stream    Al         0.681       8.02  2023
# i 290 more rows
```

# Cleaning columns

## Put it all together

```
1 rivers <- read_csv("data/rivers_correct.csv")
2 rivers <- clean_names(rivers)
3 rivers <- rename(rivers, element = ele, amount = amo, temperature = temperature_c)
4 rivers <- select(rivers, -wea)
5 rivers
```

```
# A tibble: 300 × 6
  river_name site      element amount temperature year
  <chr>      <chr>      <chr>      <dbl>      <dbl> <dbl>
1 Grasse    Up stream    Al         0.606      10.9  2019
2 Grasse    Mid stream  Al         0.425       8.68  2020
3 Grasse    Down stream Al         0.194       8.75  2021
4 Oswegatchie Up stream  Al          1         0.791  2022
5 Oswegatchie Mid stream  Al         0.161       9.32  2023
6 Oswegatchie Down stream Al         0.0333     10.6  2019
7 Raquette  Up stream    Al         0.292       4.01  2020
8 Raquette  Mid stream  Al         0.0389       5.96  2021
9 Raquette  Down stream Al          NA        6.21  2022
10 St. Regis Up stream    Al         0.681       8.02  2023
# i 290 more rows
```

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

# Fixing typos

```
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

# Fixing typos



**`if_else()`**, **`case_when()`**, **`mutate()`** from **dplyr** package\*

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

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

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

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

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

```
1 case_when(condition1 ~ value_if_true1,  
2           condition2 ~ value_if_true2,  
3           condition3 ~ value_if_true3,  
4           TRUE ~ default_value)
```

# Fixing typos

## Replace typos

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

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

## Check that it's gone:

```
1 filter(rivers, river_name == "Grase")  
  
# A tibble: 0 × 6  
# i 6 variables: river_name <chr>, site <chr>, element <chr>, amount <dbl>, temperature <dbl>, year <dbl>
```

# 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

```
1 rivers <- read_csv("data/rivers_correct.csv")
2 rivers <- clean_names(rivers)
3 rivers <- rename(rivers, element = ele, amount = amo, temperature = temperature_c)
4 rivers <- select(rivers, -wea)
5 rivers <- mutate(rivers, river_name = if_else(river_name == "Grase", "Grasse", river_name))
6
7 rivers <- mutate(???, ??? = ???)
```

**Too Easy?**

Examine and fix problems in your own data

# Your Turn: Fix another “Grasse” typo

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

```
1 rivers <- read_csv("data/rivers_correct.csv")
2 rivers <- clean_names(rivers)
3 rivers <- rename(rivers, element = ele, amount = amo, temperature = temperature_c)
4 rivers <- select(rivers, -wea)
5 rivers <- mutate(rivers, river_name = if_else(river_name == "Grase", "Grasse", river_name))
6
7 rivers <- mutate(rivers, river_name = if_else(river_name == "grasse", "Grasse", river_name))
```



# Fixing typos

To be more efficient, fix all typos at once

```
1 rivers <- read_csv("data/rivers_correct.csv")
2 rivers <- clean_names(rivers)
3 rivers <- rename(rivers, element = ele, amount = amo, temperature = temperature_c)
4 rivers <- select(rivers, -wea)
5 rivers <- mutate(rivers,
6                   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

# Fixing typos

## One last typo to fix

```
1 rivers <- read_csv("data/rivers_correct.csv")
2 rivers <- clean_names(rivers)
3 rivers <- rename(rivers, element = ele, amount = amo, temperature = temperature_c)
4 rivers <- select(rivers, -wea)
5 rivers <- mutate(rivers,
6                   river_name = if_else(river_name %in% c("Grase", "grasse"), "Grasse", river_name),
7                   river_name = if_else(river_name == "raquette", "Raquette", river_name))
```

## Combine with **case\_when()**

```
1 rivers <- read_csv("data/rivers_correct.csv")
2 rivers <- clean_names(rivers)
3 rivers <- rename(rivers, element = ele, amount = amo, temperature = temperature_c)
4 rivers <- select(rivers, -wea)
5 rivers <- mutate(rivers,
6                   river_name = case_when(river_name %in% c("Grase", "grasse") ~ "Grasse",
7                                         river_name == "raquette" ~ "Raquette",
8                                         TRUE ~ river_name))
```

**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'

```
1 rivers <- read_csv("data/rivers_correct.csv")
2 rivers <- clean_names(rivers)
3 rivers <- rename(rivers, element = ele, amount = amo, temperature = temperature_c)
4 rivers <- select(rivers, -wea)
5 rivers <- mutate(rivers,
6                   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:

```
1 rivers <- read_csv("data/rivers_correct.csv")
2 rivers <- clean_names(rivers)
3 rivers <- rename(rivers, element = ele, amount = amo, temperature = temperature_c)
4 rivers <- select(rivers, -wea)
5 rivers <- mutate(rivers,
6                   river_name = case_when(river_name %in% c("Grase", "grasse") ~ "Grasse",
7                                           river_name == "raquette" ~ "Raquette",
8                                           TRUE ~ river_name))
```

We have:

```
1 rivers <- read_csv("data/rivers_correct.csv") |>
2   clean_names() |>
3   rename(element = ele, amount = amo, temperature = temperature_c) |>
4   select(-wea) |>
5   mutate(river_name = case_when(river_name %in% c("Grase", "grasse") ~ "Grasse",
6                                 river_name == "raquette" ~ "Raquette",
7                                 TRUE ~ river_name))
```

# Play around

Take a moment to play with this code in your console

## Convert this:

```
1 rivers <- read_csv("data/rivers_correct.csv")
2 rivers <- clean_names(rivers)
3 rivers <- rename(rivers, element = ele, amount = amo, temperature = temperature_c)
4 rivers <- select(rivers, -wea)
5 rivers <- mutate(rivers,
6                   river_name = case_when(river_name %in% c("Grase", "grasse") ~ "Grasse",
7                                           river_name == "raquette" ~ "Raquette",
8                                           TRUE ~ river_name))
```



## To this:

```
1 rivers <- read_csv("data/rivers_correct.csv") |>
2   clean_names() |>
3   rename(element = ele, amount = amo, temperature = temperature_c) |>
4   select(-wea) |>
5   mutate(river_name = case_when(river_name %in% c("Grase", "grasse") ~ "Grasse",
6                                 river_name == "raquette" ~ "Raquette",
7                                 TRUE ~ river_name))
```

# Dealing with NAs

Data that *is* missing

Data that *should* be missing

# Exploring NAs

- We saw missing values in `amount`
- Use `filter()` to take a closer look

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

```
# A tibble: 39 × 6
```

	river_name	site	element	amount	temperature	year
	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>
1	Raquette	Down stream	Al	NA	6.21	2022
2	Raquette	Up stream	Ba	NA	5.23	2022
3	Raquette	Up stream	Br	NA	-99	2019
4	Oswegatchie	Up stream	Ca	NA	4.76	2023
5	Raquette	Down stream	Ce	NA	13.9	2020
6	Grasse	Up stream	Cu	NA	9.13	2019
7	Raquette	Down stream	Dy	NA	4.98	2019
8	Raquette	Down stream	Er	NA	3.07	2021
9	Raquette	Down stream	Fe	NA	7.20	2023
10	Raquette	Down stream	Gd	NA	4.73	2020

```
# i 29 more rows
```



# Omitting NAs

If we want to omit NAs

`drop_na()` is from **tidyr**\*

```
1 rivers_no_na <- drop_na(rivers, amount)
```

(Or if you want to omit **all** NAs from all columns)

```
1 rivers_no_na <- drop_na(rivers)
```



Check...

```
1 filter(rivers_no_na, is.na(amount))
```

```
# A tibble: 0 × 6
```

```
# i 6 variables: river_name <chr>, site <chr>, element <chr>, amount <dbl>, temperature <dbl>, year <dbl>
```

```
1 nrow(rivers_no_na)
```

```
[1] 261
```

No more NAs!

Fewer rows

# Replacing NAs

If we want to replace NAs

**replace\_na()** is from **tidyr**\*

```
1 rivers_no_na <- mutate(rivers, amount = replace_na(amount, 0))
```



Check...

```
1 filter(rivers_no_na, is.na(amount))  
  
# A tibble: 0 × 6  
# i 6 variables: river_name <chr>, site <chr>, element <chr>, amount <dbl>, temperature <dbl>, year <dbl>  
  
1 nrow(rivers_no_na)  
  
[1] 300
```

No more NAs!  
Same number of rows

# Converting to NA

Remember the problem with `temperature`?



```
1 filter(rivers, temperature < -10)
```

# A tibble: 3 × 6

	river_name	site	element	amount	temperature	year
	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>
1	Raquette	Up stream	Br	NA	-99	2019
2	Oswegatchie	Mid stream	K	0.426	-99	2020
3	St. Regis	Mid stream	La	0.367	-99	2023

`na_if()` is from **dplyr**\*

```
1 rivers <- mutate(rivers, temperature = na_if(temperature, -99))
```

Check...

```
1 filter(rivers, is.na(temperature))
```

# A tibble: 3 × 6

	river_name	site	element	amount	temperature	year
	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>
1	Raquette	Up stream	Br	NA	NA	2019
2	Oswegatchie	Mid stream	K	0.426	NA	2020
3	St. Regis	Mid stream	La	0.367	NA	2023

# Fixing formats

# Changing classes

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

**Your turn, try the following.** We'll deal with dates and times later...

```
1 a <- c("hi", "hello", "bonjour")
```

```
1 as.character(a)
2 as.numeric(a)
3 as.logical(a)
4 as.factor(a)
```

```
1 b <- c(1, 0, 20)
```

```
1 as.character(b)
2 as.numeric(b)
3 as.logical(b)
4 as.factor(b)
```

# Changing classes

Your turn, try the following...

```
1 a <- c("hi", "hello", "bonjour")
```

```
1 as.character(a)
```

```
[1] "hi"      "hello"    "bonjour"
```

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

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

```
1 as.logical(a)
```

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

```
1 as.factor(a)
```

```
[1] hi      hello    bonjour  
Levels: bonjour hello hi
```

```
1 b <- c(1, 0, 20)
```

```
1 as.character(b)
```

```
[1] "1"      "0"      "20"
```

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

```
[1] 1 0 20
```

```
1 as.logical(b)
```

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

```
1 as.factor(b)
```

```
[1] 1 0 20  
Levels: 0 1 20
```

# Look for problems

```
1 rivers

# A tibble: 300 × 6
  river_name site      element amount temperature year
  <chr>      <chr>      <chr>    <dbl>         <dbl> <dbl>
1 Grasse    Up stream    Al      0.606         10.9  2019
2 Grasse    Mid stream   Al      0.425          8.68  2020
3 Grasse    Down stream  Al      0.194          8.75  2021
4 Oswegatchie Up stream    Al      1            0.791  2022
5 Oswegatchie Mid stream   Al      0.161          9.32  2023
6 Oswegatchie Down stream  Al      0.0333        10.6  2019
7 Raquette  Up stream    Al      0.292          4.01  2020
8 Raquette  Mid stream   Al      0.0389          5.96  2021
9 Raquette  Down stream  Al      NA            6.21  2022
10 St. Regis Up stream    Al      0.681          8.02  2023
# i 290 more rows
```

Year could be categorical (factor)  
Better for plotting!

# Convert to categorical

```
1 rivers <- mutate(rivers, year = factor(year))
2 rivers
```

```
# A tibble: 300 × 6
  river_name site      element amount temperature year
  <chr>      <chr>      <chr>    <dbl>      <dbl> <fct>
1 Grasse    Up stream    Al      0.606      10.9  2019
2 Grasse    Mid stream   Al      0.425       8.68  2020
3 Grasse    Down stream  Al      0.194       8.75  2021
4 Oswegatchie Up stream    Al      1          0.791  2022
5 Oswegatchie Mid stream   Al      0.161       9.32  2023
6 Oswegatchie Down stream  Al      0.0333     10.6   2019
7 Raquette  Up stream    Al      0.292       4.01  2020
8 Raquette  Mid stream   Al      0.0389       5.96  2021
9 Raquette  Down stream  Al      NA          6.21  2022
10 St. Regis Up stream    Al      0.681       8.02  2023
# i 290 more rows
```



# Put it all together...

```
1 rivers <- read_csv("data/rivers_correct.csv") |>
2   clean_names() |>
3   rename(element = ele, amount = amo, temperature = temperature_c) |>
4   select(-wea) |>
5   mutate(river_name = case_when(river_name %in% c("Grase", "grasse") ~ "Grasse",
6                                 river_name == "raquette" ~ "Raquette",
7                                 TRUE ~ river_name),
8         amount = replace_na(amount, 0),
9         temperature = na_if(temperature, -99),
10        year = factor(year))
```

## 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

# Put it all together...

## Feel free to annotate within a pipe

```
1 rivers <- read_csv("data/rivers_correct.csv") |>
2   # Fix column names
3   clean_names() |>
4   rename(element = ele, amount = amo, temperature = temperature_c) |>
5   select(-wea) |>
6   mutate(
7     # Correct typos
8     river_name = case_when(river_name %in% c("Grase", "grasse") ~ "Grasse",
9                           river_name == "raquette" ~ "Raquette",
10                          TRUE ~ river_name),
11     # Missing amounts should be 0
12     amount = replace_na(amount, 0),
13     # Problems with temperature logger, -99 is a mistake
14     temperature = na_if(temperature, -99),
15     # Convert for plotting
16     year = factor(year))
```

# 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 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> <dtm>
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 `dtm` (Date/Time)

# lubridate package \*

Generally, only the order of the year, month, day, hour, minute, or second matters.



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

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

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

```
1 my_data <- my_data |>           # Pass my_data
2   mutate(my_column1 = if_else(...), # No passing (no pipes!)
3         my_column2 = if_else(...),  # Instead, give 3 arguments to mutate:
4         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](#)

