manipulating and analyzing data with dplyr

the tidyverse: an "umbrella" package



- **ggplot2**: a "grammar of graphics" by Hadley Wickham. Divide the data and the aesthetics. Create and modify the plots layer by layer
- dplyr: a way to lead with data frames, sql external data bases, written in C++
- readr: read data
- tidyr: format data frames
- **stringr**: deals with strings
- additional packages for other tasks: tibble, lubridate and many more

Most of R is still base-based and both philosophies communicate well with each other

reading data with readr

```
library(dplyr)
 library(readr)
 surveys <- readr::read_csv("data_raw/portal_data_joined.csv")</pre>
## Parsed with column specification:
## cols(
     record id = col double(),
##
     month = col double(),
##
    day = col_double(),
##
     year = col double(),
##
     plot_id = col_double(),
##
     species id = col_character(),
##
     sex = col_character(),
##
     hindfoot length = col double(),
##
```

inspect the data
str(surveys)

View(surveys)

surveys

some principal functions in dplyr

- **select** (columns)
- **filter** (rows)
- rename (columns)
- mutate (create new columns or modify existing columns)
- arrange to sort according to a column
- count cases of one or many columns

select columns

```
select(surveys, plot_id, species_id, weight)
```

- 1. there is no need to put quotes
- 2. there is no need to put variables between c()

base R still works in a tibble

```
surveys[, c("plot_id", "species_id", "weight")]
```

removing columns

```
select(surveys, -record_id, -species_id)
```

additional functions

```
select(surveys, -ends_with("id"))
```

filter rows

logical clauses!

```
surv_1995 <- filter(surveys, year == 1995)</pre>
```

No need to use \$ or brackets

```
surveys$year == 1995
surveys[surveys$year == 1995 , ]
```

mutate creates or modifies columns

```
surveys <- mutate(surveys, weight_kg = weight / 1000)

mutate(surveys,
    weight_kg = weight / 1000,
    weight_lb = weight_kg * 2.2)</pre>
```

group_by() and summarise()

• if you have a column factor (e.g. sex) and want to apply a function to the levels of this factor

another example:

mean_w

```
## # A tibble: 92 x 3
   # Groups:
                sex [3]
##
             species_id mean_weight
##
      sex
      <chr> <chr>
                                <dbl>
##
                                 9.16
##
    1 F
             BA
                                41.6
##
             DM
##
             DO
                                48.5
                               118.
##
             DS
##
             NL
                               154.
    5 F
                                31.1
             OL
##
                                24.8
             OT
##
##
             OX
                                21
##
             PB
                                30.2
    9 F
## 10 F
                                22.8
             PE
## # ... with 82 more rows
```

arrange sorts by a column

```
arrange(mean_w, mean_weight)
arrange(mean_w, desc(mean_weight))
```

the pipe operator



Classic syntax goes like this

```
object1
object2 <- function1(object1)
object3 <- function2(object2)</pre>
```

The pipe operator allows to apply functions sequentially:

```
object3 <- object1 %>% function1() %>% function2()
```

functions in the tidyverse work very well with pipes

select and filter

```
surveys2 <- filter(surveys, weight < 5)
surveys_sml <- select(surveys2, species_id, sex, weight)

surveys %>%
  filter(weight < 5) %>%
  select(species_id, sex, weight)
```

group_by() and summarize()

count

```
surveys %>%
    count(sex)

surveys %>%
    count(sex, species)

surveys %>%
    count(sex, species) %>%
    arrange(species, desc(n))
```

challenge

- How many animals were caught in each plot_type surveyed?
- Use **group_by()** and **summarize()** to find the mean, min, and max hindfoot length for each species (using **species_id**). Also add the number of observations (hint: see **?n**).

save data!

```
surveys <- readr::read_csv("data_raw/portal_data_joined.csv")</pre>
surveys_complete <- surveys %>%
  filter(!is.na(weight),
         !is.na(hindfoot_length),
         !is.na(sex))
species_counts <- surveys_complete %>%
    count(species id) %>%
    filter(n >= 50)
surveys_complete <- surveys_complete %>%
  filter(species_id %in% species_counts$species_id)
write_csv(surveys_complete, path = "data/surveys_complete.csv")
```

data visualization with ggplot2

ggplot2

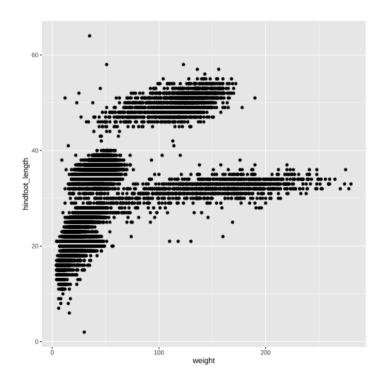
 ggplot2 separates the data from the aesthetics part and allows layers of information to be added sequentially with +

- data
- **mappings**: the specific variables (x, y, z, group...)
- **geom_xxx()**: functions for plotting options **geom_point()**, **geom_line()**

cheat sheet link

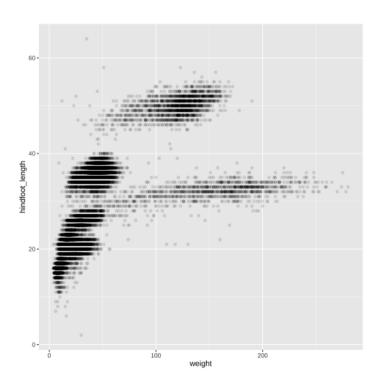
```
library(ggplot2)
library(readr)

surveys_complete <- read_csv("data/surveys_complete.csv")</pre>
```

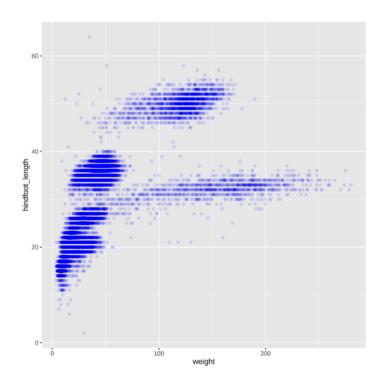


you can assign a plot to an object and build on it

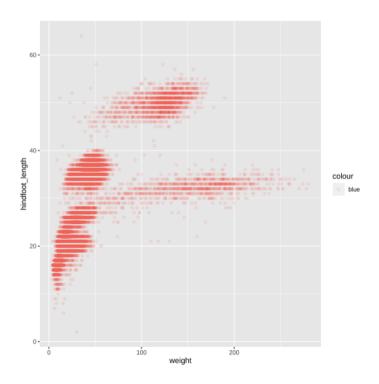
```
surveys_plot +
   geom_point(alpha = 0.1) #transparency
```



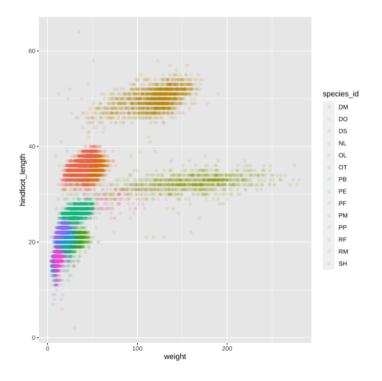
```
surveys_plot +
  geom_point(alpha = 0.1, color = "blue") #color
```



```
surveys_plot +
   geom_point(alpha = 0.1, aes(color = "blue")) #this is a mistake!
```

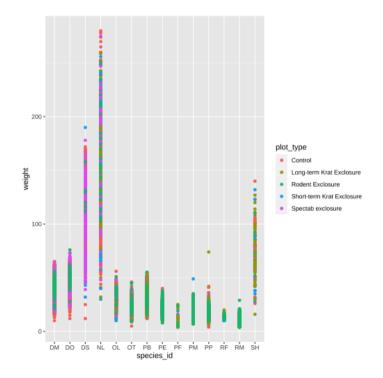


```
surveys_plot +
  geom_point(alpha = 0.1, aes(color = species_id))
```

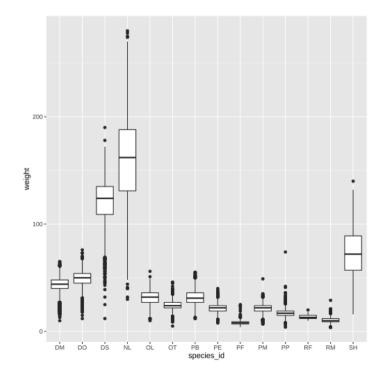


challenge: change x to categorial variable

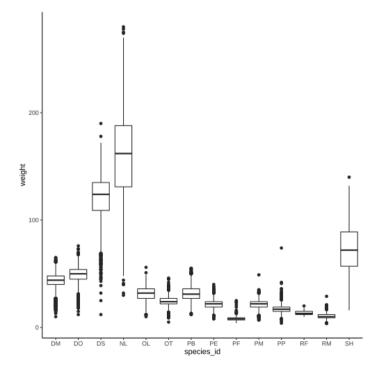
```
ggplot(data = surveys_complete,
    mapping = aes(x = species_id, y = weight)) +
    geom_point(aes(color = plot_type))
```



boxplots!



theme options theme_

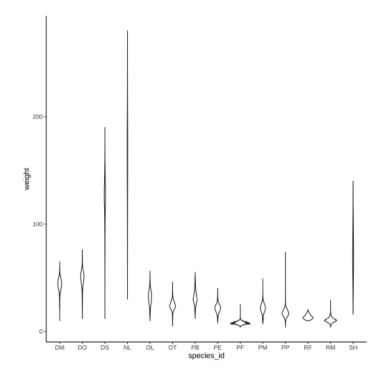


add jitter layer

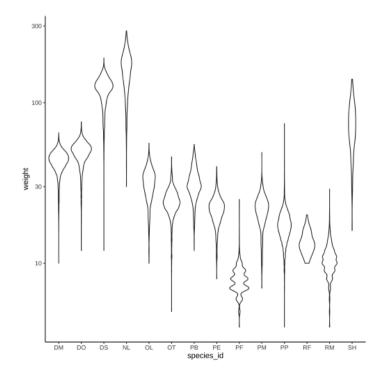
change plot order

violin plots

```
ggplot(data = surveys_complete,
    mapping = aes(x = species_id, y = weight)) +
    geom_violin() + theme_classic()
```



change scale (scale_xx options)



add title ggtitle()

```
p + #remember the plot can be an object
ggtitle("Nice violin plot")
```

