Teaching R in DASC 1104

R Part II

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Learning R

- Introduction to tidy statistics
- RStudio primers
- Software carpentries
    - R for Reproducible Analyses
    - Programming with R
- R for data science

Materials

- Available on gitHub at https://github.com/jtipton25/dasc-1104-teaching
- Labs and HWs available by request: email: jrtipton@uark.edu

Course make file

- In the loaded `dasc1104-teaching` project, the following command will generate all code

```
# note: this will take a while, maybe 15-20 minutes
source(here::here("make.R"))
```

Lectures

- unit_4_09_intro_R_Rmarkdown.Rmd
- unit_4_R_10_intro_plotting.Rmd
- unit_4_R_11_File_IO_and_basic_data_wrangling.Rmd
- unit_4_R_12_Data_types_and_objects.Rmd
- unit_4_R_13_Functional_programming.Rmd
- unit_4_R_14_Intro_to_modeling.Rmd

Brief recap of each unit

Intro to R and Rmarkdown

- Focus on reproducibility and workflow
- Heavily inspired by R for Reproducible Analyses
- Much of this was covered in part I

Data visualization and plotting

- Goal: to get students interested and excited when working with data
    - Give enough background and language for students to seek out help online for data visualization
    - Emphasis on flexibility and exploration of data through visualization
    - Worked through a couple datasets to demonstrate how the visualization process works

Data Visualization

- Using `ggplot` and the `mtcars` data, generate a scatterplot of fuel milage (`mpg`) as a function of engine power (`hp`)
- Explore how other variables in the `mtcars` dataset influence fuel milage

File IO

- Read the `game_goals.csv` file from the `data` folder with `read_csv()` (from the `tidyverse` library) and give this `data.frame` the name `dat`
- Make sure you properly define the filepath
- Many more specifics about reading files
    - skipping lines
    - defining column variable types
    - different file delimiters
    - missing values and `NA`s
    - data comments

Data manipulation

- Most important `dplyr` (a part of the `tidyverse`) commands
  - Using pipes `%>%` (or `|>`)
  - Choose observations (rows) based on conditional values with `filter()`
  - Reorder the variables with `arrange()`
  - Select variables by name with `select()`
  - Create new variables with `mutate()`
  - Create summary variables with `summarize()`
    - `summarise()` if you are British
  - Group variables with `group_by()`

Pipes

> The itsy bitsy spider went up the water spout
>
> Down came the rain and washed the spider out
>
> Out came the sun and dried up all the rain
>
> And the itsy bitsy spider went up the spout again

- multiple variable names, likely to make a typo

```
itsy_bitsy_spider_1 <- went(itsy_bitsy_spider, up = "waterspout")
itsy_bitsy_spider_2 <- rain_down(itsy_bitsy_spider_1, washed = "out")
itsy_bitsy_spider_3 <- sun(itsy_bitsy_spider_2, dried = "rain")
itsy_bitsy_spider_4 <- went(itsy_bitsy_spider_3, up = "spout again")
```

- overwriting the data, likely to make errors

```
itsy_bitsy_spider <- went(itsy_bitsy_spider, up = "waterspout")
itsy_bitsy_spider <- rain_down(itsy_bitsy_spider, washed = "out")
itsy_bitsy_spider <- sun(itsy_bitsy_spider, dried = "rain")
itsy_bitsy_spider <- went(itsy_bitsy_spider, up = "spout again")
```

Pipes

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- hard to read

```
went(
    sun(
        rain_down(
            went(itsy_bitsy_spider, up = "waterspout"),
            washed = "out"),
        dried = "rain"),
    up = "spout again"
)
```

- Much better

```
itsy_bitsy_spider %>%
    went(up = "waterspout") %>%
    rain_down(washed = "out") %>%
    sun(dried = "rain") %>%
    went(up = "spout again")
```

Pipes

- How does the pipe work?

```
library(tidyverse)
x <- 1:10
x %>%
  mean()
```

```
## [1] 5.5
```

is equivalent to

```
x %>%
  mean(.)
```

```
## [1] 5.5
```

where the `.` is a placeholder for the variable `x` which results in

```
mean(x)
```

```
## [1] 5.5
```

Pipes

- Pipes are not always the optimal choice
- Try to avoid super long chains of pipes
    - Instead: break up the pipe into intermediate objects
- Pipes can't handle multiple inputs/outputs
- How to build a chain of pipes
    - Start by outlining a plan of attack
    - Write one small subset of the pipe chain at a time and verify the results
- Challenge: Ask me questions about the `diamonds` data

Data manipulation

- Using the NHL goals dataset from before (named `dat`), `filter()` out all records where the `player` is `"Alex Ovechkin"` or `"Sidney Crosby"`. How many observations are there?
- Using the NHL goals dataset, `arrange()` the observations in decreasing order (the `desc()` function) using the `penalty_min` variable. Who on the list has two games in the top 10 penalty minutes of the dataset?
- Using the NHL dataset, `select()` the variables `player`, `shots`, and `goals`. Then, `mutate()` a new variable `shots_per_goal` by dividing the number of `shots` by the number of `goals`. Did you get any warnings/errors?
- Using the NHL dataset, `summarize()` the total `goals` for each `player`. Using `arrange()`, who are the top 5 players in all time goals (in the data)?

Relational data

- Joining multiple datasets together
- `left_join()`
- `right_join()`
- `inner_join()`
- `outer_join()`

Processing strings

- Use the `stringr` package (part of the `tidyverse`)
- Can use regular expressions

```
(string1 <- "This is a string")
```

```
## [1] "This is a string"
```

```
(string2 <- 'This is also a string')
```

```
## [1] "This is also a string"
```

```
(string3 <- 'This is a "quoted" string')
```

```
## [1] "This is a \"quoted\" string"
```

String processing

- `str_length()`
- `str_c()`
- `str_sub()`
- others

Factors

- Factors are an atomic type for categorical data
  - By default, these are unordered
  - Ordered factors can be used (good - better - best)
- `factor()`
- `levels()`
- `fct_recode()`
- `fct_order()`

Dates (not covered)

Functional programming (currently not covered, but should be)

- Code is for **humans** not computers
- Functions make code readable to humans

Keep it DRY

- Don't Repeat Yourself
- If you copy and paste code 3 or more times, write a function
- Generate some data

```
df <- data.frame(
  x = rnorm(10),
  y = rnorm(10),
  z = rnorm(10)
)
```

- The above isn't DRY, but there is no chance of errors and the code is easy to understand

DRY

- Transform each variable to mean 0, standard deviation 1

```
x_scaled <- (df$x - mean(df$x, na.rm = TRUE)) / sd(df$x, na.rm = TRUE)
y_scaled <- (df$y - mean(df$y, na.rm = TRUE)) / sd(df$y, na.rm = TRUE)
z_scaled <- (df$z - mean(df$z, na.rm = TRUE)) / sd(df$y, na.rm = TRUE)
```

- Can you spot the error?

```
z_scaled <- (df$z - mean(df$z, na.rm = TRUE)) / sd(`df$y`, na.rm = TRUE)
```

- write a function to do this

```
rescale_variable <- function(x) {
  mean_x <- mean(x, na.rm = TRUE)
  sd_x   <- sd(x, na.rm = TRUE)
  ## save the variable for the output
  out <- (x - mean_x) / sd_x
  return(out)
}
```

- Much less likely to make an error

```
x_scaled <- rescale_variable(df$x)
y_scaled <- rescale_variable(df$x)
z_scaled <- rescale_variable(df$z)
```

Naming Functions

- Name functions for human readers!

```
# bad name -- what does this do?
f()

# what does this do?
func()

# longer names -- use autocomplete
fill_missing_values()
pre_process_data()
```

- Don't name your functions over functions or variables which already exist!

```
mean <- function(x) {}
TRUE <- function(x, y) {}
```

Naming Functions

- Pick a naming convention and stick with it (I prefer snake case)

```
# snake case
my_function_name <- function(x, y) {}
# camel case
myFunctionName <- function(x, y) {}
```

- Try to reuse leading names -- allows for autocomplete

```
# better
check_input()
check_parameters()
check_inits()

# not as good
input_check()
parameters_check()
inits_check()
```

For loops

- for loops in R

```
value <- 0
for (i in 1:100) {
  value <- value + i
}
value
```

```
## [1] 5050
```

Writing functions

- given a `data.frame`, calculate the mean of each column
- write a function that returns the mean of each column