Q&A Material

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What's the Deal with Fall Coding Lab?

Two tracks:

Accelerated: 2 lessons

- ▶ 2 lessons covering loops and functions.
- No final project (you already did it).

Not accelerated: 5 lessons.

- 3 lessons review summer camp material.
- 2 lessons covering loops and functions.
- Final project:
 - Find a data set that speaks to you.
 - Try to uncover something interesting. Graph it and tab it.
 - We'll give you feedback.

Logistics for Both Tracks

- Instructors Ari and Terence + wonderful TAs.
- ▶ 80 minutes per week: brief review and Q&A, then work in groups.
- Not graded.
- Access to TAs for coding specific problems throughout the quarter.
 - ► TA office hours 30 minutes before and after lecture time.
- Github website with all material.
 - We'll post solutions (eventually).
- Use Piazza for questions.
 - Rules of engagement: coding questions only, no Stats homework!
 - How to ask a good question?

Poll: How much coding experience do you have?

- First timer.
- ► Beginner.
- ► Intermediate.
- Proficient.

Please include your email.

Class 1: Why R? & Vectors

Key Points: R Basics

- Rstudio has a console to access R and a text editor to write code for reproducible projects.
 - Analogy: R is to RStudio as Tony Stark is to Iron Man's suit.
- R extensible through packages.
 - use install.packages("") once and then library() each session.
- ▶ Use <- to assign any object to a name.
- Functions take inputs and return outputs.
 - Input "understood" based on position or name.
 - Find out more about functions with ? (e.g. ?filter).

Questions

▶ Any questions on this? Feel free to ask on chat.

Key points: Vectors

- Vectors are the fundamental way to store data in R.
- ▶ We can operate on vectors element-by-element without loops.
 - dplyr verbs rely on this!
- We introduced built-in functions to build vectors and do operations on vectors.
- NAs are sticky!

Key points: Data Types and Coercion

- ► (Atomic) Vectors have a single data type.
 - ▶ Most often: logical, integer, double, or character.
- Certain operations expect a certain data type and R will try to coerce the data if it can.
- Usually, simpler types can be coerced to more complex types.
 - ▶ logical < integer < double < character.
 - Example on slides: paste0(1L, "ing").
- Caution! Coercion can lead to unexpected behavior such as making NAs.

One More Thing

Logicals are coercible to numeric or character. This is very useful!

Determine the rule for how R treats TRUE and FALSE in math.

```
TRUE + 4
```

[1] 5

FALSE + 4

[1] 4

Questions?

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Warm up

➤ Solve the questions at the beginning of the lab in small (random) groups.

Lab 1

- Two "types" of breakout room:
 - ▶ Work along: larger group with more guidance from a TA.
 - Small groups: 4 people, TAs will come in and out to answer questions. Use the help button!
- ► Add "(work along)" or "(small group)" to your Zoom name so we can sort you.
- Get as far as you can, then finish it up after class.
- ▶ Before you leave, fill out the exit poll.

Lab 1: Exit poll

- What does hist() return?
 - A histogram plot of the data you give it.
 - ► A history of the commands you've run.

Class 2: Reading files and 'dplyr'

Course logistics:

- When should we start working on the final project?
 - Start looking for a dataset now.
 - Write code to read it into R and start investigating with dplyr verbs.
 - Ask simple questions that can be addressed with your current tools.

Key points: Reading files

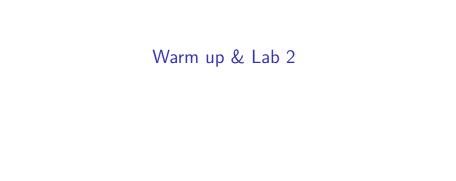
- Tabular data is stored in a lot of different formats.
 - e.g. .csv, .xlsx, .dta
- Read tabular data of a given type with the proper function.
 - e.g. for csvs we have read_csv()
 - ► If you get a new type, Google "How to read xxx files into R tidyverse".
- We need to be aware of the file path and can setwd().
- We know there are useful tools built into the read_xxx() functions.
 - Though we just scratched the surface.

Key points: Manipulating data with dplyr()

- ► Choose columns with select().
- Choose rows based on a match criteria with filter().
 - ▶ We were introduced to comparison operators like == and %in%.
- ► Make new columns with mutate().
- Sort data with arrange() and arrange(desc()) or arrange(-x).
- Create summary statistics with summarize().

Key points: Grouped analysis with group_by()

- Groups are a set of rows that belong together.
 - group_by() adds information about groups without changing the "data".
- Use group_by() with summarize() to create summary data at group-level.
 - Use with functions that reduce data from a vector to a single value per group.
 - Expected output: a table with one row per group and one column per summary statistic and one column per grouping column.
- ▶ We can also use group_by to do grouped analysis with:
 - mutate with window functions or to add a summary stat as column for further analysis.
 - It also can impact arrange and filter.



Warm up & Lab 2

- Solve the questions at the beginning of the lab in small (random) groups.
- ► After: add "(small group)" in front of your name if you want to work in one, if you want to work along just stay in the main room.
- Exit poll:
- 1. What dplyr command allows you to create or modify variables?
- 2. What dplyr command allows you to sort your data?

Class 3: Control Flow

Key points: control flow with if () and ifelse ()

- ▶ Review how logical operators (i.e. !, |, &) and comparison operators work.
 - ▶ If this all still seems confusing a refresher in propositional logic will seriously pay off!
- Use if (), else, and else if () to control when and how an action is completed.
 - ► Can use shorthand for conditions: if (thing_thats_true) is equivalent to if (thing_thats_true == TRUE).
- ifelse() as a vectorized, NA friendly alternative.
- Use ifelse() with mutate() to create new columns contingently.

Two more things (1/2): case_when()

case_when() allows you to chain many conditions and outputs:

Somewhat equivalent to:

```
ifelse(condition_1, value_1,
    ifelse(condition_2, value_2,
        ifelse(condition_3, value_3, value_4)))
```

Two more things (2/2): NA

▶ How to think about logicals and NA? If the answer is ambiguous \rightarrow NA. If the answer isn't \rightarrow not NA.

```
TRUE | NA
```

```
## [1] TRUE
```

```
FALSE & NA
```

```
## [1] FALSE
```

Warm up & Lab 3

Warm up & Lab 3

- ► This warm up is a little longer. Aim to get up to the "Vectorized booleans" part!
- ► After: add "(small group)" in front of your name if you want to work in one.
- Exit poll:
- 1. What is FALSE | NA?
- 2. What is TRUE & NA?

Acc Class 1 / Class 4: Functions

What's the Deal with Fall Coding Lab?

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Accelerated: 2 lessons

- 2 lessons covering loops and functions.
- No final project (you already did it).
- Occurs every other week.
 - Ideally the sections are balanced.

Not accelerated: 5 lessons.

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- Final project:
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Today

- Review what functions are
- ▶ Do a warm-up
- ▶ Provide background for the lab
- ▶ Do the lab

Functions

What are some of the key take aways you learned about functions from the video?

Key points: Functions

- Write functions when you want o use a set of operations repeatedly
 - Don't Repeat Yourself (DRY)
- Functions consist of arguments and a body and are usually assigned to a name.
 - The arguments are things that you want to change in the code
 - The body is what you would be repeating if you wrote sloppy code
- Functions are for humans
 - pick names for the function and arguments that are clear and consistent
- Debug your code as much as you can as you write it.
 - if you want to use your code with mutate() test the code with vectors

Warm up

➤ Solve the questions at the beginning of the lab in small (random) groups.

Background for the lab

Concepts: r/q/p/d functions

R has built-in functions for working with distributions.

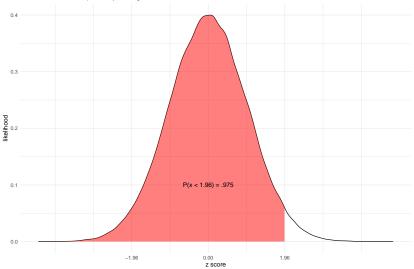
	example	what it does?
r	rnorm(n)	generates a random sample of size n
p	pnorm(q)	returns CDF value at q
q	qnorm(p)	returns inverse CDF (the quantile) for a given probability
d	dnorm(x)	returns pdf value at x

We should already be familiar with r functions like rnorm() and runif().

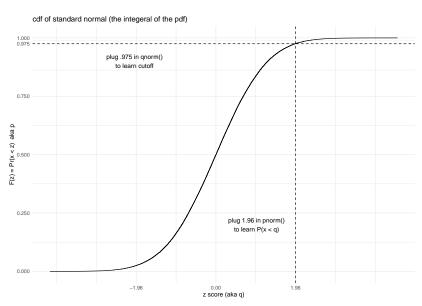
► These concepts will be taught in stats. We don't expect you to learn them here. We can help you reason through the material enough to do the coding.

What are p and q?





What are p and q?



What are p and q?

 ${\tt pnorm}$ returns the probability we observe a value less than or equal to some value q.

```
pnorm(1.96)
## [1] 0.9750021
pnorm(0)
## [1] 0.5
```

 ${\tt qnorm}$ returns the inverse of ${\tt pnorm}.$ Plug in the probability and get the cutoff.

```
qnorm(.975)
## [1] 1.959964
```

```
qnorm(.5)
```

```
# [1] C
```

Monte Carlo experiments

Monte Carlo is a world gambling hub.

- ► Gamblers know that roulette wheels are not made perfectly.
- ► If you watch the wheel long enough and take notes you can figure out the emprical probability



Monte Carlo experiments

Statisticians use the same idea.

- ► If we're not sure how to calculate something exactly, we can simulate it and get the result.
- Often used for difficult to compute integrals.

An example

In real life experiements we usually have one sample.

```
# Setting a seed ensures replicability
set.seed(4)
# we set our parameters
true mean <- 0.5
N < -30
# We simulate and observe outcomes
simulated data <- rnorm(N, mean = true mean)
obs mean <- mean(simulated_data)</pre>
obs mean
```

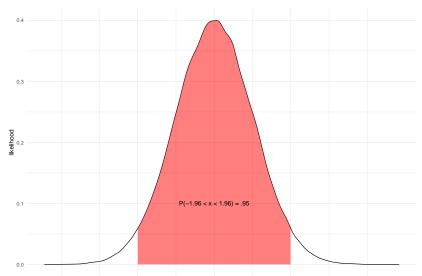
```
## [1] 0.9871873
```

Put that number in perspective with a z score

```
obs_sd <- sd(simulated_data)
zscore <- (obs_mean - true_mean) / (obs_sd / sqrt(N))
zscore
## [1] 3.303849
1 - pnorm(zscore)
## [1] 0.000476836</pre>
```

Our monte carlo experiment

Show that the sample mean from a sample drawn from a normal distribution falls outside the 95 percent confidence region 5 percent of the time.



How to do a Monte Carlo Simulation

- 1. Generate random samples of data using a known process (e.g. rnorm()).
- 2. Make calculations based on the random sample.
- 3. Aggregate the results.

```
do_monte_carlo <- function(N, true_mean, B, alpha){
   sample_statistics <- make_mc_samples(N, true_mean, B)
   z_scores <- get_zscores(sample_statistics$mean, true_mean
   test_significance(z_scores, alpha) %>% mean()
}
```

Lab 4

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