Computing in the Statistics Curriculum: Lessons Learned from the Educational Sciences

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If you have found these slides online, here are talk videos

- UCSOTS talk (5 minutes): https://youtu.be/8kqlGHcnVNY
- JSM talk (15 minutes): https://youtu.be/ZsYJ81TwGW8

Computing is fundamental to contemporary statistical practice and scientific inquiry and should be explicitly taught



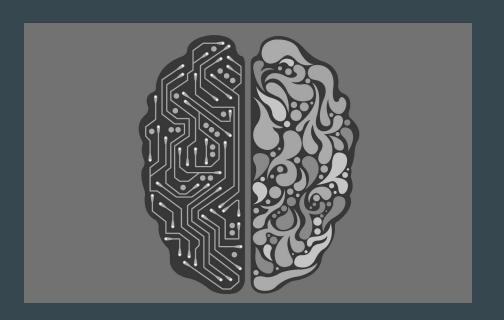
(ASA, 2017; Horton, 2015; NASEM, 2018; Nolan & Temple Lang, 2010)

Cognitive Load: Impediment to Learning

There is a finite amount of information that can be processed or stored in working memory at a time (cognitive load).

Different types of cognitive load (Hermans, 2021):

- *Intrinsic:* characteristics of the information being learned.
- *Extraneous:* the way information is presented.



Statistical Computing and Cognitive Load

- Statistical computing adds cognitive load to the learning process in statistics (e.g., Woodard & Lee, 2021)
 - Computational considerations
 - Syntax
 - Syntactical structure
 - Debugging
 - Computational thinking
 - Coding seems to be difficult for many students
- We can work to manage and lessen cognitive load by thoughtfully considering the specific coding content we teach and how we teach it.

Cognitive *Un-*loading: Make Purposeful Pedagogical Decisions

Pedagogical decisions need to be made about coding content.

- What logistical considerations do you need to account for?
- What will be taught (scope)?
- How will it be sequenced?



Example Pedagogical Decisions

- How will students compute in the course?
 - Desktop / Cloud / Both
- Where and when do students need practice with code?
 - In-class / out-of-class
 - Individual / group
 - Templates / Blank documents (RMD, R script)
- How will coding be introduced in class?
 - Live coding/ Worked examples / Group activities
- What code content do you start the course with?
 - Data structures (e.g., vectors, data frames) / "Cake" (data visualization, EDA)

Cognitive *Un-*loading: Use Consistent Syntactic Structure

Using code with the same syntactic structure (common grammar) lessens cognitive load

 Can focus on learning new functions (verbs) and their purpose

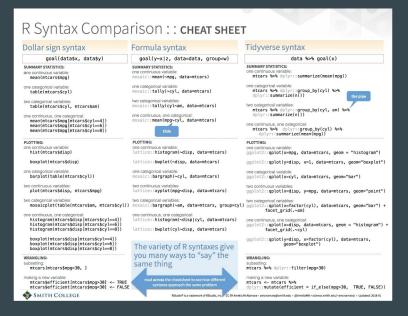
Some syntactic structures can emphasize the relationship between syntax and concepts

• E.g., Roles of variables



Example Pedagogical Decision

- What syntax or packages will be used? In R:
 - base
 - formula (mosaic, ggformula, lattice)
 - tidyverse (ggplot2, dplyr)



Make a histogram of the bill_length_mm variable from the penguins dataset:

```
ggplot(penguins) + geom_histogram(aes(x = bill_length_mm))
histogram( ~ bill_length_mm, data = penguins)
gf_histogram( ~ bill_length_mm, data = penguins)
hist(penguins$bill_length_mm)
```

```
library(psych)
library(BHH2)
nhanes2017= read.csv("nhanes2017.csv", as.is = F)
table(nhanes2017$exerciseGT60)
par(mfrow = c(1, 2))
hist(nhanes2017$pulse[nhanes2017$exerciseGT60 == "YES"], xlim = c(min(nhanes2017$pulse), max(nhanes2017$pulse)))
hist(nhanes2017$pulse[nhanes2017$exerciseGT60 == "NO"], xlim = c(min(nhanes2017$pulse), max(nhanes2017$pulse)))
par(mfrow = c(1, 1))
tapply(nhanes2017$pulse, nhanes2017$exerciseGT60, summary)
tapply(nhanes2017$pulse, nhanes2017$exerciseGT60, describe)
t.test(pulse ~ exerciseGT60, data = nhanes2017)
```

Authentic example: Analyze the difference in pulse by exerciseGT60 from NHANES

```
# Load libraries
library(psych)
# Import data
nhanes2017= read.csv("nhanes2017.csv")
# Get levels and sample sizes
table(nhanes2017$exerciseGT60)
# Plot histograms
par(mfrow = c(1, 2))
hist(nhanes2017$pulse[nhanes2017$exerciseGT60 == "YES"], xlim = c(min(nhanes2017$pulse), max(nhanes2017$pulse)))
hist(nhanes2017$pulse[nhanes2017$exerciseGT60 == "NO"], xlim = c(min(nhanes2017$pulse), max(nhanes2017$pulse)))
par(mfrow = c(1, 1))
# Compute summary statistics
tapply(nhanes2017$pulse, nhanes2017$exerciseGT60, FUN = describe)
# Carry out two-sample t-test
t.test(pulse ~ exerciseGT60, data = nhanes2017)
```



Use comments

```
# Load libraries
library(mosaic)
# Import data
nhanes2017 = read_csv("nhanes2017.csv")
# Get levels and sample sizes
tally(~ exerciseGT60, data = nhanes2017)
# Plot histograms
gf_histogram(~ pulse | exerciseGT60, data = nhanes2017)
# Compute summary statistics
favstats(~ pulse | exerciseGT60, data = nhanes2017)
# Carry out two-sample t-test
t_test(~ pulse | exerciseGT60, data = nhanes2017)
```



The Most Important Template

The following template is important because we can do so much with it.

It is useful to name the components of the template:

We're hiding a bit of complexity in the template, and there will be times that we will want to gussy things up a bit. We'll indicate that by adding ... to the end of the template. Just don't let ... become a distractor early on.

Cognitive *Un-*loading: An Example with Histograms in R

What you teach:

- ggplot(penguins) + geom_histogram(aes(x = bill_length_mm))
- histogram(~ bill_length_mm, data = penguins)
- gf_histogram(~ bill_length_mm, data = penguins)
- hist(penguins\$bill_length_mm)

The order you teach it in:

- Variables -> Histograms -> R intro -> syntax
- R intro -> Variables -> Histograms -> syntax

How it's introduced:

- RMarkdown file with a worked example
- Live coding session
- Combination-- RMarkdown with some worked, some blanks

Cognitive *Un*-loading: Model Computational Thinking Norms

Modeling computational thinking and norms

- Debugging (e.g., McCauley et al., 2008)
- Normalizing emotional reactions to coding (e.g., Not panicking (e.g., DiNapoli, 2018))



Questions to help you revise how you teach coding:

- What syntactic structure makes the most sense for my students/course/goals?
- Is the code being presented to students consistent in its structure?
- How does new code connect with previous content?
- Will students see/use this code more than once?
 - "Stitch in time saves 9"
- How will students encounter code?
 - Live coding, scaffolded documents, cheatsheet?

Resources and Places to Start

- <u>Introductory statistics labs in R</u>, Amelia McNamara (formula or tidyverse)
- Speaking R, Amelia McNamara (guidance for live coding and reading code)
- <u>Statistical Modeling and Computation for Educational Scientists</u>, Andrew Zieffler (tidyverse)
- <u>Simulation Based Inference</u>, Randy Pruim (formula)
- <u>Data Science in a Box</u>, Mine Çetinkaya-Rundel (tidyverse)



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Thank you!

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