### Module 1: Big Picture Ideas

Rebecca C. Steorts (slide and course adaptation from Maria Tackett)

#### **Announcements**

- Resources for extra R review(https://learnr.numbat.space/) (focus on Chapters 4 6)
- ▶ Reading: BMLR: Chapters 1 3.
- ► For Monday, make sure you have read/understood syllabus, course structure, and have access to all course materials!

Questions from last class?

### **Topics**

- ▶ Data analysis life cycle
- ► Reproducible data analysis
- Analyzing multivariable relationships

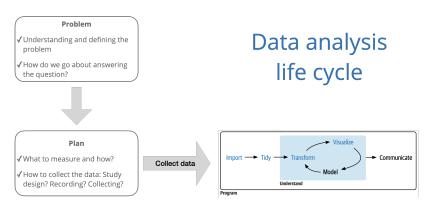


Figure 1: Source: *R for Data Science* with additions from *The Art of Statistics: How to Learn from Data*.

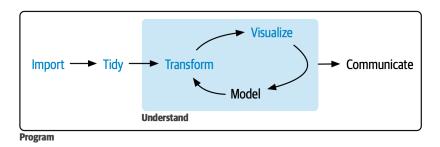


Figure 2: Source: R for Data Science

## Reproducibility

## Reproducibility checklist

What does it mean for an analysis to be reproducible?

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#### Near term goals:

- ► Can the tables and figures be exactly reproduced from the code and data?
- Does the code actually do what you think it does?
- ▶ In addition to what was done, is it clear why it was done?

### Reproducibility checklist

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#### Near term goals:

- ► Can the tables and figures be exactly reproduced from the code and data?
- Does the code actually do what you think it does?
- In addition to what was done, is it clear why it was done?

#### Long term goals:

- ► Can the code be used for other data?
- Can you extend the code to do other things?

## Why is reproducibility important?

- Results produced are more reliable and trustworthy [@ostblom2022]
- ► Facilitates more effective collaboration [@ostblom2022]
- Contributing to science, which builds and organizes knowledge in terms of testable hypotheses [@alexander2023]
- ▶ Possible to identify and correct errors or biases in the analysis process [@alexander2023]

# When things go wrong

Reproducibility		
error	Consequence	Source(s)
Limitations in	Loss of 16,000	(Kelion 2020)
Excel data	COVID case	
formats	records in the UK	
Automatic	Important	(Ziemann, Eren, and El-Osta
formatting in	genes	2016)
Excel	disregarded in scientific studies	
Deletion of a cell caused rows to shift	Mix-up of which patient group received	(Wallensteen et al. 2018)
	the treatment	
Using binary instead of	Mix-up of the intervention	(Aboumatar and Wise 2019)
explanatory	with the	

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#### **Toolkit**

- **► Scriptability** → R
- ▶ Literate programming (code, narrative, output in one place)
  → Rmarkdown
- **▶ Version control** → Git / GitHub

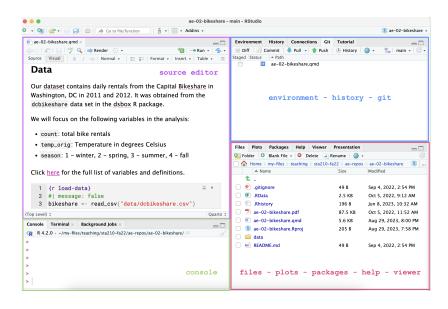
### R and RStudio

- R is a statistical programming language
- ► RStudio is a convenient interface for R (an integrated development environment, IDE)



Figure 3: Source: Statistical Inference via Data Science

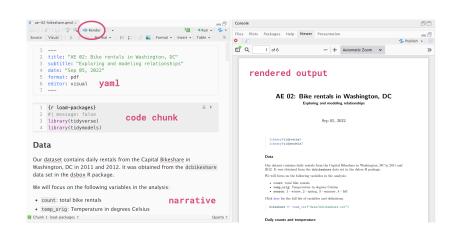
#### RStudio IDE



#### Rmarkdown

- ► Fully reproducible reports the analysis is run from the beginning each time you render
- Code goes in chunks and narrative goes outside of chunks
- Visual editor to make document editing experience similar to a word processor (Google docs, Word, Pages, etc.)

## Rmarkdown/Quarto

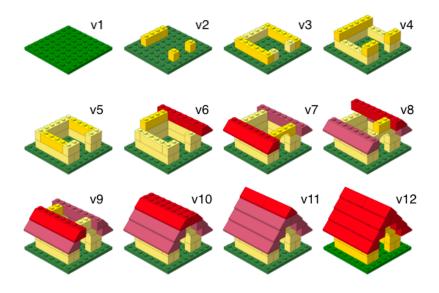


### How will we use Rmarkdown?

- Every assignment is written in Rmarkdown document
- ► You'll have a template in Rmarkdown to start with
- ▶ If you have used Quarto before, think about what the similarities/differences are between Rmarkdown/Quarto.

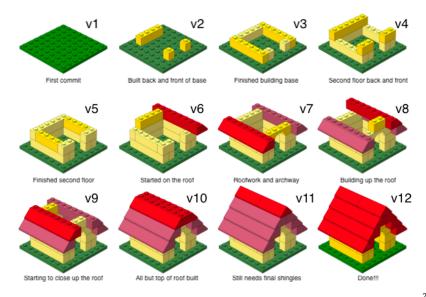
Version control with git and GitHub

## What is versioning?



### What is versioning?

#### with human readable messages



### Why do we need version control?

Provides a clear record of how the analysis methods evolved. This makes analysis auditable and thus more trustworthy and reliable. [@ostblom2022]

### git and GitHub



- ▶ **git** is a version control system like "Track Changes" features from Microsoft Word.
- ► **GitHub** is the home for your git-based projects on the internet (like DropBox but much better).
- ▶ There are a lot of git commands and very few people know them all. 99% of the time you will use git to add, commit, push, and pull.

### git and GitHub

► The entire materials for the semester can be found on github if you'd like to dig into them further at

https://github.com/resteorts/generalized-linear-models

▶ Video Tutorial of git/github: https://www.youtube.com/watch?v=RGOj5yH7evk

## Multivariable relationships

### Carbohydrates in Starbucks food

- ► Starbucks often displays the total calories in their food items but not the other nutritional information.
- Carbohydrates are a body's main fuel source. The Dietary Guidelines for America recommend that carbohydrates make up 45% to 65% of total daily calories.<sup>1</sup>
- Our goal is to understand the relationship between the amount of carbohydrates and calories in Starbucks food items. We'd also like to assess if the relationship differs based on the type of food item (bakery, salad, sandwich, etc.)

<sup>&</sup>lt;sup>1</sup>Source: Mayo Clinic

### Starbucks data

- ▶ **Observations**: 77 Starbucks food items
- ► Variables:
  - carb: Total carbohydrates (in grams)
  - calories: Total calories
  - bakery: 1: bakery food item, 0: other food type

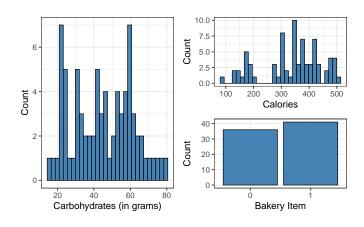
## **Terminology**

- carb is the response variable
  - variable whose variation we want to understand / variable we wish to predict
  - also known as outcome or dependent variable

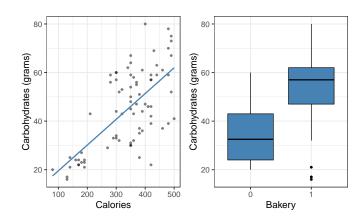
## **Terminology**

- carb is the response variable
  - variable whose variation we want to understand / variable we wish to predict
  - ▶ also known as *outcome* or *dependent* variable
- calories, bakery are the predictor variables
  - variables used to account for variation in the response
  - also known as explanatory, independent, or input variables

## Univariate exploratory data analysis



## Bivariate exploratory data analysis



## Function between response and predictors

$$carb = f(calories, bakery) + \epsilon$$

- ▶ **Goal**: Determine *f*
- ▶ How do we determine *f*?
  - ▶ Make an assumption about the functional form f (parametric model)
  - Use the data to fit a model based on that form

#### Determine *f*

- 1) Choose the functional form of f, i.e., choose the appropriate model given the response variable
- Suppose f takes the form of a linear model

$$y = f(\mathbf{X}) = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p + \epsilon$$

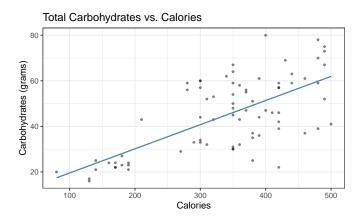
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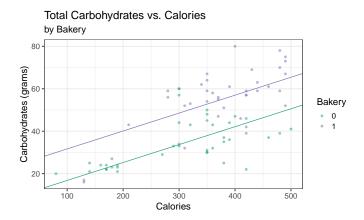
2. Use the data to fit (or train) the model, i.e, **estimate the** model parameters,  $\beta_0, \beta_1, \dots, \beta_p$ 

### Carb vs. Calories



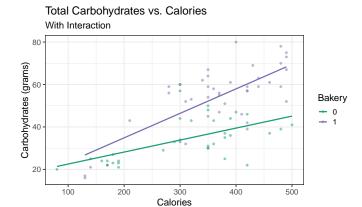
$$carb = \beta_0 + \beta_1 \ calories + \epsilon$$

### Carb vs. Calories + Bakery



$$\mathsf{carb} = \beta_0 + \beta_1 \ \mathsf{calories} + \beta_2 \ \mathsf{bakery} + \epsilon$$

## Carb vs. Calories + Bakery (with interaction)



$$carb = \beta_0 + \beta_1$$
 calories  $+ \beta_2$  bakery  $+ \beta_3$  calories  $\times$  bakery  $+ \epsilon$ 

### Statistical model vs. regression equation

### Statistical model (also known as data-generating model)

$$carb = \beta_0 + \beta_1$$
 calories  $+ \beta_2$  bakery  $+ \beta_3$  calories  $\times$  bakery  $+ \epsilon$ 

Models the process for generating values of the response in the population (function + error)

#### Regression equation

Estimate of the function using the sample data

$$\hat{\mathsf{carb}} = \hat{eta}_0 + \hat{eta}_1 \; \mathsf{calories} + \hat{eta}_2 \; \mathsf{bakery} + \hat{eta}_3 \; \mathsf{calories} imes \mathsf{bakery}$$

## Why fit a model?

- ▶ **Prediction:** Expected value of the response variable for given values of the predictor variables
- ► **Inference:** Conclusion about the relationship between the response and predictor variables
- What is an example of a prediction question that can be answered using the model of carb vs. calories and bakery?
- What is an example of an inference question that can be answered using the model of carb vs. calories and bakery?

### Recap

### Reproducibility

- lt is best practice conduct all data analysis in a reproducible way
- We will implement a reproducible workflow using R, Quarto, and git/GitHub

#### Multivariable relationships

- We can use exploratory data analysis to describe the relationship between two variables
- We make an assumption about the relationship between variables when doing linear regression
- ► The two main objectives for fitting a linear regression model are (1) prediction and (2) inference

## References