### Introduction to EPIB 704

Mabel Carabali

EBOH, McGill University

**EPIB 704** 

(updated: 2025-08-27)

#### **Outline**

- 1. Presentations
- 1. Course description
- 1. Reproducible research
- 1. Descriptive epidemiology

#### Who are we?

#### Mabel

- Social and Infectious Diseases epidemiologist
- "Recovered" clinician; ES(T)L
- Twitter/FB/IG-less
- Website

#### Gabrielle Jacob

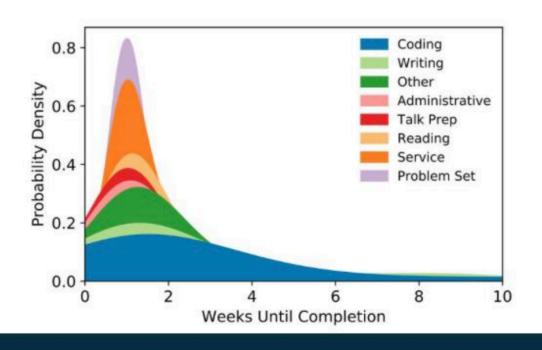
#### Julia Brillinger

# Who are you?

What are the things you dislike the most about EPI?

#### How good are we at getting things done?





### Course description - more specifically

- Emphasis on epidemiologic theory and the estimation of epidemiologic effect measures and uncertainty in different study designs
- Emphasis on causal inference concepts and its distinction from, and yet simultaneously dependency, on statistical models
- Emphasizes the limitations of mainstream null hypothesis statistical significance (NHST) paradigm
- Encourages the recognition of the value of an estimation (Bayesian) paradigm for optimal inferences

### Course description (3)

- Attention will also be given to exploratory data analysis (tabular and graphical), data interpretation, critical examination of assumptions and reproducible research
- Focus mainly on cumulative incidence measures for categorical outcomes, with attention to model checking and screening for confounders and effect measure modifiers
- Regression models covered will include linear, logistic, Poisson, survival, & meta-analytical (hierarchical) models
- Experimental and quasi-experimental designs will be discussed
- Miscellaneous topics will include attributable fractions, selection bias, sensitivity analyses, bootstrapping, matched data, missing data, and misclassified data

### **Objectives:**

- 1) Identify common threats of the validity or biases,
- 2) Identify and interpret their effects in the measurement of associations, and
- 3) Identify opportunities to investigate, correct, and address these biases.

### **Learning Outcomes**

By the end of this course you should be able to:

- 1. Contrast advantages and disadvantages of Frequentist and Bayesian statistical inference
- 2. Recommend analytical approaches to each (every) different study designs
- 3. Identify the presence of potential threats to the validity of a statistical analysis
- 4. Design a basic statistical analysis plan to answer a given research question accounting for threats to the validity

#### **Inbstructional Methods**

- "Hands-on" approach whereby the various key statistical concepts covered will be illustrated by computer coding, the new calculus for modern epidemiologic methods
- ullet Counting has historically been the essential background to epidemiologic research and remains so in the  $21^{st}$  century but is now often best accomplished by simulation and sampling of posterior probability distributions
- Examples of the required computing code will be extensively provided.

#### Reference textbooks

- 1) What if? by Miguel Hernán and James Robins, available here (Chapters 1-11)
- 2) *Modern Epidemiology* by Timothy L. Lash, Tyler J. VanderWeele, Sebastien Haneuse, Kenneth J. Rothman, available at the McGill Library here
- 3) *Regression and Other Stories* by Andrew Gelman, Jennifer Hill, and Aki Vehtari, available here Optional deeper Bayesian dive consider the excellent
- 4) **Statistical rethinking: a Bayesian course with examples in R and Stan** by Richard McElreath available at the McGill Library here (Chapters 1-10)
  - Associated website: https://xcelab.net/rm/statistical-rethinking/ and YouTube lectures here

Other readings will be assigned each week from published journal articles

### Computing language

The computing language of choice for the course is R.

There are several reasons for this choice including its open source and rich online community which means help is often only a Google away.

We believe R has become the *lingua franca* for much of the epidemiology/biostatistical universe. Of course, other languages including Julia or Python will give the same results but unfortunately, we can't supply the necessary support for those languages.

Stata is a popular proprietary software, but we consider the scripting and reproducibility offered by R-Markdown provides additional advantages for its choice.

#### R - Related resources

We understand that not everyone is familiar with R as a computing language and therefore **strongly recommend** starting, reviewing and or strengthening your knowledge on the software.

**In-person:** McGill's Computational and Data Systems Initiative R Summer Camp: August 19 to 23, 2024.

Other resources from the CDSI here

#### R - Related resources

#### **Online**

- R-Studio Education: Independent learning (https://education.rstudio.com/learn/)
- R-Studio/posit: Practice-based learning (https://posit.co/products/enterprise/academy/)
- R-related Books: https://www.rstudio.com/resources/books/
- Princeton site to explore R: https://exploringr.princeton.edu/self-learning-resources-for-r/
- R-resources list: https://thatdatatho.com/r-resources-beginner-advanced/

#### **Course Structure**

- Lectures (mostly discussions!)
- Assignments (submitted through *mycourses*)
- Rapid Reviews

### **Assessment/Evaluation**

	%	Rubrics
Attendance & Participation	5%	Presence, attention, active participation 📲 😌 🕳
Homework Assignments (5 x 15% each)	75 %	Accuracy, demonstrated skills, timely delivery
Rapid Review (RR)	15%	Critical appraisal, time management 🦁 🍃 💆 🔎
RR Participation	5%	Meaningful contribution on RRs from peers A 👓
Total	100 %	

### **Attendance & Participation**

We expect to see everyone in class in-person.

• Since life happens: *Zoom* attendance can be an option **under special circumstances ONLY**.

Attendance will be graded of 5 points to make up for the 5% of the total grade:

• Attend class, pay attention to the lectures and your classmates' presentations, and to participate actively of the in-class discussion.

Doing the suggested lecture **readings BEFORE the lectures** will assure an optimal learning experience.

### **Attendance & Participation**

To safeguard the well being of everyone, **if you present respiratory-like symptoms or any signs or symptoms of communicable conditions** that may endanger your health, the health of classmates or instructors, **please consult a medical professional and inform the instructors as early as possible** to make the required accommodations.

Please see McGill's guidelines for attendance: Guidelines and Policies

### Rapid Review

Every enrolled student must sign up for one *rapid review* which is a 10-minute presentation of a **methodological critique** about any new paper (**publication date since July 2025**) published in the biomedical literature (i.e., indexed in PUBMED).

• Presenters should comment on study design, validity, precision, causal interpretation, modeling strategy and other concepts from the course.

The comments can be praises or critiques but **must** demonstrate thoughtful application of the principles discussed in the lectures as applied to new work in the field.

The rapid review will be graded out of 15 points to make up for the 15% of the total grade.

### Guidelines for the rapid review

 Selection of the document: Identify the published manuscript considering the date of publication, your personal/thesis interest and the methodological approach. [5/15 total points]

Consider the **focus on methods**, consider exploring new methodological approaches or novel use of existent epidemiological methods.

- E-pubs or online publications ahead and *pre-prints* are allowed.
- Please provide a pdf of your target paper at least two weeks before the presentation to be shared with the class.

### Guidelines for the rapid review

- **All students** must participate in the review of the paper through the Perusall platform.
  - You must have received an email to access the EPIB 704 Perusall platform.
  - You could also access it through this link
- This activity will have an individual grade of 5 points to make **5% of the total grade**.

### Guidelines for the rapid review (2)

• **Critical appraisal:** Please read and critically appraise the selected manuscript considering key aspects of the design and analytic plan.

To illustrate your understanding of the selected document, **you must re-write the abstract**, as follows:

- Write a 250-words (maximum) structured abstract: Background, Objectives, Methods, Results, and Conclusion.
- Provide a **50-words** (maximum) section of **strengths and limitations**.

--

• Finally, provide a brief statement (50-words) indicating if the methodological approach used in the reviewed paper is something that you will do, or something that you would like to do, or something that you would definitely not do, and why.

### Guidelines for the rapid review (3)

- In-Class presentation format: In a maximum total time of 10 minutes present to the class the results of your review [5/15 total points], considering the following:
  - Power point / slides or any visual aids are welcome, but there should not be more than 5 slides/pages.
  - After each presentation there would be a brief period for Q&As (approximately 5 minutes).
  - Interactive discussions are highly encouraged but a **maximum of 15 minutes** are allocated for each presentation and discussion in every class.

#### **Dates assignment**

select an article and a date (last date Nov 28) <sup>1</sup> and **inform the instructor by September 9, 2025** 1st come 1st served.

Date	Name	Article/Manuscript
09 Sep	TAs	TBD
11 Sep		
16 Sep		
18 Sep		
23 Sep		
25 Sep		
30 Sep		
02 Oct		
07 Oct		
09 Oct		
21 Oct		

### Homework Assignments

The homework will make 75% of the total grade. There will be 5 homework assignments (each worth 15%) throughout the 13 weeks of the semester, which will be graded by the TAs. The schedule of dates for when these will be assigned and handed in and handed back is as follows:

HW #	Assigned	Due	Returned
1	Sept 11	Sept 18	Sept 25
2	Sept 25	Oct 02	Oct 09
3	Oct 09	Oct 23	Oct 30
4	Oct 30	Nov 06	Nov 13
5	Nov 06 *	Nov 20	Nov 27

<sup>\*</sup> Last assignment is given 15 working days in advance. See Policy on Assessment of Student Learning (PASL) subsection 6.7. in Point 6. COMMUNICATION OF COURSE ASSESSMENT TASKS.

#### **Howework Grading**

Submitted assignments should be your **individual effort**, even if you consult with other students about your strategy for obtaining these solutions (see plagiarism note below).

To reinforce the concept of scientific **reproducibility** all assignments should be submitted as R-Markdown files. Data files for the assignments are available on GitHub and may be installed directly into R.

Late assignments carry a very severe penalty of **20% off per day late**. This is primarily to protect the time of the TAs.

Pleas of mercy for extenuating circumstances will be accepted only with written documentation. Taking extra time to do a better job is probably not a worthwhile strategy, because the late penalty is so costly.

#### Some Admin & Logistics

- Important dates (McGill): Add/Drop class: Sep 9th, 2025
  - Key dates: Assignments and Rapid Reviews
- Communication channels: *mycourses* & emails : Be Mindful of time schedules, office hours
  - We all have other activities outside the classroom
- Course Evaluations: Feedback throughout the course is MORE THAN WELCOME!

Read the syllabus!:)

#### **Some Admin & Logistics**

- Academic Integrity: Plagiarism, misconducts, & any form of discrimination, aggressive behavior or harassment towards ANY member of the community is **UNACCEPTABLE**
- Policy for the academic accommodation of religious observances
- Guidelines for the Academic Accommodation of Pregnant Students and Students Caring for Dependants
- Policy concerning the rights to academic accommodations for students with dissabilities
- Charter of Students' Rights
- Other resources at Student Wellness Hub: FEEL FREE TO ASK!

# Where are we on our Epi knowledge?

#### **Baseline Assessment Quizz**



#### Some medical research bingo we will cover

unclear analysis aims causality, prediction or description: it makes a difference	evidence of absence fallacy not significant = no effect? if only things were that easy	data dredging "look what I found that turned out significant"	noisy data fallacy that what doesn't kill it makes it stronger? i wish!
dichotomania categorize everything, chapter 1 of data torture handbook	table 2 fallacy interpret each regression coefficient as adjusted for confounding? yeah, that is probably wrong	regression to the mean yes, measurements at follow-up are different than at baseline. doesn't mean much	ignoring dependent observations the data are probably nested, and that is important
poor reporting yes, there are reporting guidelines for that	small sample size gets these confidence intervals nice and wide	ignoring missing data can't miss what you never had? think again	data driven variable selection post selection inference, yikes!
only apparent predictive performance that is optimistic. could have tried a cross- validation or a bootstrap, perhaps?	point estimate is the effect why did you even bother calculating the confidence interval?	collider is it a cause? is it a confounder? no, it is a collider here to ruin your inferences!	multivariate model what you mean to say is: multivariable model

### Descriptive Epidemiology

- What is Descriptive Epidemiology?
- How much bias is *acceptable*? ( •• )
- Descriptive vs. Causal (<sup>((2))</sup>)

Matthew P Fox, Eleanor J Murray, Catherine R Lesko, Shawnita Sealy-Jefferson, On the Need to Revitalize Descriptive Epidemiology, American Journal of Epidemiology, Volume 191, Issue 7, July 2022, Pages 1174–1179, https://doi.org/10.1093/aje/kwac056

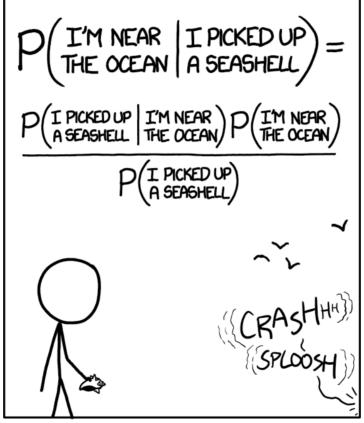
Not every research needs to be causal,

but everything should be robust (!)

## Framework for descriptive Epidemiology

- 1. Research Question and Background
- 2. Population and Data Source
- 3. Outcome Ascertainment/Covariates
- 4. Analysis Plan: Targeted Measure of Occurrence
- 5. Results (Presentation of)
- 6. Bias, Limitations and Interpretations

#### A little bit of Bayes



STATISTICALLY SPEAKING, IF YOU PICK UP A SEASHELL AND DON'T HOLD IT TO YOUR EAR, YOU CAN PROBABLY HEAR THE OCEAN.

**QUESTIONS?** 

**COMMENTS?** 

**RECOMMENDATIONS?** 

### Other resources

### Reproducible research

Can you reproduce a project that you completed 3 weeks ago? How about 3 months ago? 3 years ago?

What if your data changed, even minimally, could you easily redo your analyses? How long would it take you to update your manuscript?

# Getting setup and starting a project in R Studio

- 1. Put all the files you received into a folder and give it a name (e.g., 704\_course)
- 2. Choose File > New Project
- 3. Associate the project with the course folder (e.g., 704\_course)
- 4. From now on, you can open your project by clicking directly on the .Rproj file in that folder (or using the drop-down menu on the top right)
- 5. Now create an R Notebook file (File > New File > R Notebook)

## A note on computer code and syntax

- There are two main "dialects" of R: base R and tidyverse
- I generally use tidyverse syntax
- There are three main differences:
  - The pipe %>%
  - dplyr verbs (e.g., select(), filter(), mutate())
  - graphing with ggplot2 (but this antedated the Tidyverse)
- However a case can be made that base R is kinder, gentler and more efficient
- This course is not about learning computer syntax so plenty of sample code will be provided
- It is about understanding what code to use and understanding its output

# Tidyverse examples

- Instead of tidy(lm(y ~ x)), type lm(y ~ x) %>% tidy()
- For data management, you can easily filter and select columns:

```
data_subset <- data %>%
  filter(female == 0) %>%  # keep female respondents
  select(income, education) %>%  # keep two variables
  mutate(educ_2 = education^2) %>%  # create edu^2
  drop_na()  # listwise delete missings
```

# A few useful keyboard shortcuts

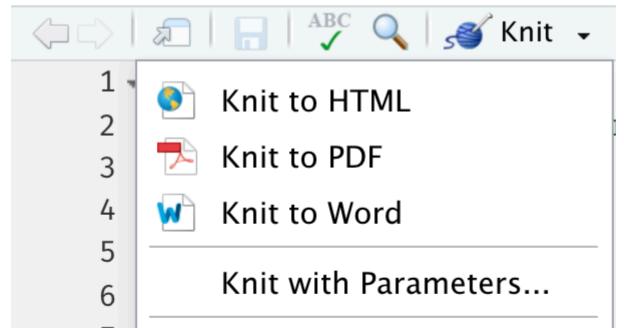
Object/Action	Windows	MacOS
<-	Alt + -	Option + -
%>%	Ctrl + Shift + M	Cmd + Shift + M
New code block	Ctrl + Alt + I	Cmd + Option + I
Run highlight	Ctrl + Enter	Cmd + Enter
Clear console	Ctrl + L	Ctrl + L

# R can help reproducibility

Can integrate code and text.

Uses RMarkdown which is simple intuitive language which can be out put to different formats

- HTML
- pdf
- LaTex
- Word



## What is R Markdown?

- 1. "An authoring framework for data science." (✔)
- 2. A document format (.Rmd). (✓)
- 3. An R package named rmarkdown. (✓)
- 4. "A file format for making dynamic documents with R." (✓)
- 5. "A tool for integrating text, code, and results." (✓)
- 6. "A computational document." (✓)
- 7. Wizardry. (🏖)

### R Markdown

R Markdown depends on knitr and Pandoc

knitr executes the computer code embedded in Markdown, and converts R Markdown to Markdown

Pandoc renders Markdown to the output format you want (such as PDF, HTML, Word, etc)

- R Markdown: The Definitive Guide
- RStudio has a Markdown quick reference drop down help menu

# Basic R-Markdown anatomy

- 1. The **metadata (YAML)**
- 2. The **text**
- 3. The code
- 4. The **output**

## **Gentle tutorial**

R Markdown for writing reproducible scientific papers

## R Chunk

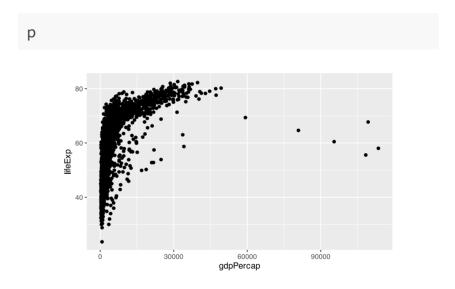
```
```{r echo=TRUE, results='hide'}
glimpse(mockdata)
```

#### default options

```
## List of 54
## $ eval
                 : logi TRUE
              : logi FALSE
## $ echo
## $ results : chr "markup"
  $ tidy
                : logi FALSE
  $ tidy.opts
                : NULL
   $ collapse : logi FALSE
  $ prompt : logi FALSE
  $ comment : chr "##"
  $ highlight : logi TRUE
##
   $ size
          : chr "normalsize"
   $ background : chr "#F7F7F7"
  $ strip.white : 'AsIs' logi TRUE
   $ cache
           : logi FALSE
   $ cache.path : chr "L1_EPIB704_25_cache/html,
   $ cache.vars
                : NULL
  $ cache.lazy : logi TRUE
  $ dependson : NULL
##
  $ autodep : logi FALSE
##
   $ cache.rebuild: logi FALSE
   $ fig.keep : chr "high"
##
  $ fig.show : chr "asis"
##
                                     46 / 49
                : chr "center"
## $ fig.align
```

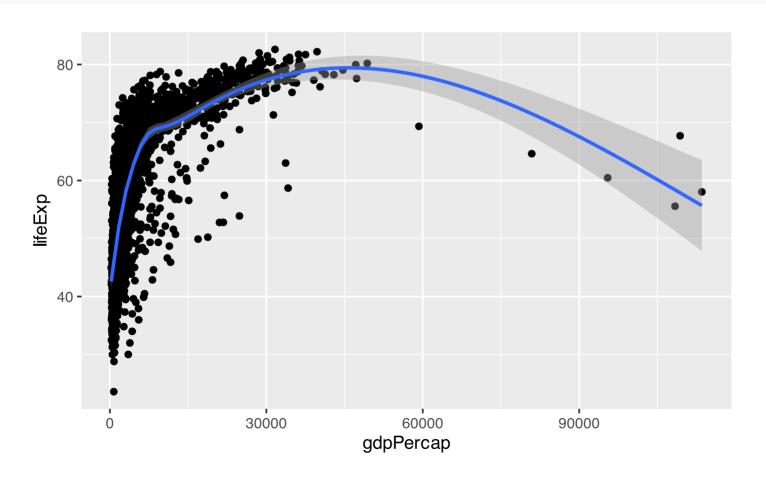
## **Visualizations**

Great way to understand and share insights into your data R is great for visualizations - either base R or tidyverse with **ggplot2** package



# **Build plots with layers**

p + geom\_smooth()



# Increasing complexity

