

# Analysis in Epidemiology



## Contact Information & Course Website

- Office: Allen Center 311
- Office Hours: By Appointment
- Email: [Eric.Lofgren@wsu.edu](mailto:Eric.Lofgren@wsu.edu)

Course material will be posted on GitHub at:  
<https://github.com/epimodels/AnalysisInEpi>

Generally I will try to have material up the day before class.



## Purpose of this Course

- Build a functional vocabulary in regression models
- Let you engage with the literature, work on your own problems, etc.
- Know when and how to get help



## Some Resources for the Curious

- *Modern Epidemiology 3<sup>rd</sup> Edition*
- [stats.stackexchange.com](https://stats.stackexchange.com)
  - Generally speaking posting homework questions there is discouraged, but this can be a valuable resource for the future
- CISER
- The usual advice: Get help EARLY!



Center for Interdisciplinary Statistical  
Education and Research



**Questions?**



## Types of Observational Data

- Most of these got covered in Module 1
- Binary/Continuous
- People/Time/Events
- The decision on *how* to model is largely dictated by *what* you are modeling



## Why Regression?

- Module 1 taught you how to calculate measures of association with 2x2 tables, long division, etc.
- Adjusting for confounding using stratification
- All of this seemed to work well enough, can be done in Excel, on a whiteboard, etc. – why bother with regression?





# Stratification

100	50
40	220

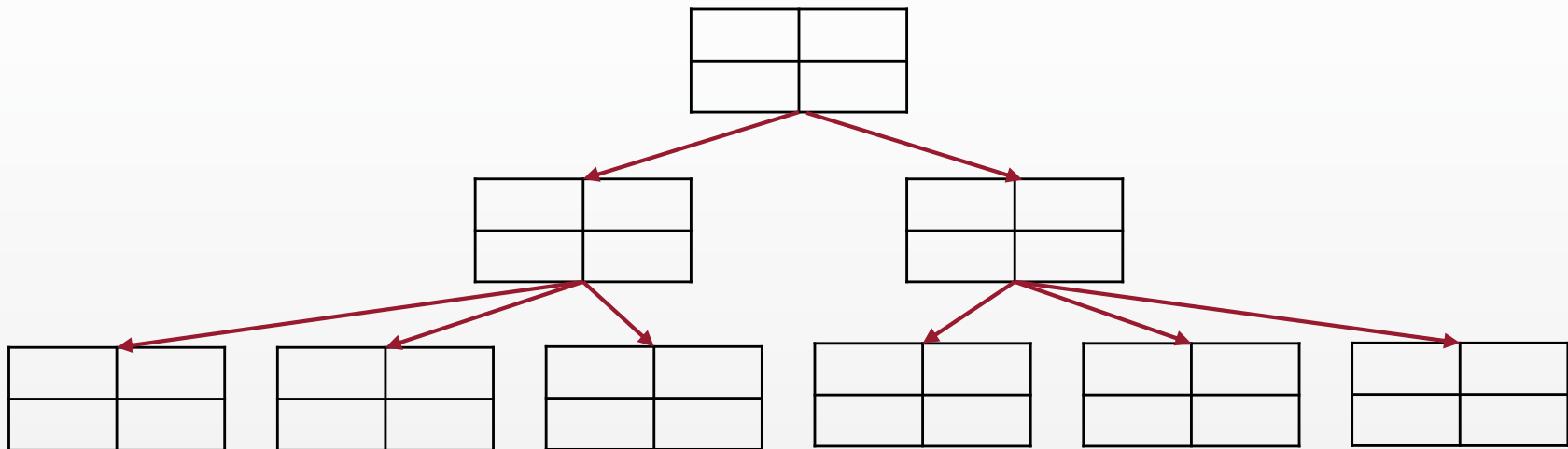
60	20
20	120

40	30
20	100





# Stratification



And this is just for two variables...what about 16?



## Strengths of Regression

- Can handle adjustment by *many* variables
- Can handle non-categorical data
- Can smooth/spackle over empty cells
  - If you know what happens to 26 year olds and 28 year olds, you can guess what happens to 27 year olds
- You can predict
  - Given  $m$ ,  $X$  and  $b$ , solve for  $Y$
  - Regression is the foundational toolset of machine learning/data science



## What Regression Isn't





## Regression Can't...

- Automatically fix bad data collection
- Control for bias that it (or you) don't know about
- Solve your sample size problems for you
- ...solve *any* of your problems **for** you – regression is a tool, and a dumb one at that



# Assumptions and Problems of Regression

- Positivity: An individual has a non-zero probability of having any combination of parameter values
  - Regression assumes cells with 0's happened by chance – what if those cells are impossible?
- Model misspecification: Missing confounders, the wrong distribution, etc. will give you the wrong answer
  - This is, I would argue, the biggest problem in observational epi
- Nonidentifiability: Two (or more) combinations of parameters are equally supported by the data, and there is no “best fit”
- Others we will discuss as the class goes on
- There are *more* assumptions necessary for causal inference, which is beyond the scope of this module



## Reading a Regression Equation

- Regression is essentially progressively more complex versions of  $y = mx + b$

$$Y = \beta_0 + \beta_1 A + \varepsilon$$

Linear Predictor

$$Y = \alpha + \beta_1 A + \gamma \mathbf{Z}$$





## What's a Link Function?

- A link function is a function that describes the relationship between  $Y$  and the rest of the equation
- Linear Predictor:  $\mathbf{X}\beta$
- Link function:  $g(Y) = \mathbf{X}\beta$
- Identity:  $Y = \mathbf{X}\beta$
- Log:  $\ln(Y) = \mathbf{X}\beta$
- Logit:  $\frac{Y}{1-Y} = \mathbf{X}\beta$





## What is a Distribution?

- Linear regression assumes things came from a normal distribution
- This is *often* not true
- Other distributions are common
- Binomial: Binary data
- Poisson/Negative Binomial: Counts and rates
- Exponential/Weibull/Gamma: Time
- When unspecified, it is often assumed to be normal



## Least Squares and Maximum Likelihood

- Two ways to estimate the best fitting parameter
- Linear regression often uses least squares
- Most of the other models we will discuss use some form of maximum likelihood