

## Module 0: Welcome to STA 310

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

Welcome!

# Teaching Team

**Instructor:**

Professor Rebecca Steorts  
Old Chem 208  
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**Teaching assistants**

Deva Bag, PhD Student

## Announcements

- ▶ We will meet on Friday, January 9th for lecture, so please come prepared for lecture and not lab (as the PhD student is traveling and not available for lab). I will be sure to give you this lecture back at the end of the semester.
- ▶ The course webpage (<https://resteorts.github.io/teach/generalized.html>) will be updated on roughly a weekly basis, so please check this frequently for any updates.

# Course logistics

## Lectures

Tuesday and Thursday, 11:45 - 1:00 pm, Perkins 127

## Labs (Office Hour or Alternate Lecture Time)

Lab 01: Friday, 11:45 - 1:00 pm, Old Chemistry 001

# Generalized Linear Models

*In statistics, a generalized linear model (GLM) is a flexible generalization of ordinary linear regression. The GLM generalizes linear regression by allowing the linear model to be related to the response variable via a link function and by allowing the magnitude of the variance of each measurement to be a function of its predicted value.<sup>1</sup>*

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## Example: Logistic regression

$$\begin{aligned}\pi = P(y = 1|x) &\Rightarrow \text{Link function: } \log\left(\frac{\pi}{1 - \pi}\right) \\ &\Rightarrow \log\left(\frac{\pi}{1 - \pi}\right) = \beta_0 + \beta_1 \times\end{aligned}$$

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By the end of the semester, you will be able to ...

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- ▶ analyze real-world data by fitting and interpreting GLMs.
- ▶ use R for analysis and write reports
- ▶ effectively communicate results from statistical analyses to a general audience in writing.

# Course topics

## **Generalized Linear Models**

- ▶ Review of distributions, likelihoods, and regression
- ▶ Introduce models for non-normal response variables
- ▶ Estimation, interpretation, and inference
- ▶ Mathematical details of GLMs as a unified framework

# GLMs in practice



JSLHR ■

## Research Article

### Intersectionality of Race and Question-Asking in Women After Right Hemisphere Brain Damage

Danai Kasambira Fannin,<sup>a</sup> Jada Elleby,<sup>a</sup> Maria Tackett,<sup>b</sup> and Jamila Minga<sup>c</sup>

<sup>a</sup>Department of Communication Sciences and Disorders, North Carolina Central University, Durham <sup>b</sup>Department of Statistical Science, Duke University, Durham, NC <sup>c</sup>Department of Head and Neck Surgery & Communication Sciences and Department of Neurology, Vascular and Stroke Division, Duke University School of Medicine, Durham, NC

“... we used **negative binomial regression** to model the association between the number of questions produced, race, and group after adjusting for the additional covariates age and years of education. **Poisson and zero-inflated Poisson regression models** were also considered... the negative binomial model was a good fit for the data given the **overdispersion** in the distribution of number of questions asked.”<sup>2</sup>

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<sup>2</sup>Fannin, D. K., Elleby, J., Tackett, M., & Minga, J. (2023). Intersectionality of Race and Question-Asking in Women After Right Hemisphere Brain Damage. *Journal of Speech, Language, and Hearing Research, 66*(1), 314-324.

# GLMs in practice

Papers

## Officiating bias: The effect of foul differential on foul calls in NCAA basketball

Kyle J. Anderson & David A. Pierce

Pages 687-694 | Accepted 07 Jan 2009, Published online: 20 May 2009

Cite this article <https://doi.org/10.1080/02640410902729733>

Full Article

Figures & data

References

Citations

Metrics

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Read this article

" . . . a **logistic regression model** is used to test how the likelihood of a foul is affected by which team is the home team, the foul differential, and the score differential. . . The logistic regression was run under several specifications . . . using **clustered observation standard errors**, with each game as a cluster. This is done as an attempt to adjust for the fact that **observations may not be independent** as required under the logistic specification.<sup>3</sup>

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<sup>3</sup> Anderson, K. J., & Pierce, D. A. (2009). Officiating bias: The effect of foul differential on foul calls in NCAA basketball. *Journal of sports sciences*, 27(7), 687-694.

## Meet your classmates!

- ▶ Get in groups of 2 - 3
- ▶ Each person in the group...
  - ▶ Introduce yourself
  - ▶ Share a boring fact about yourself
- ▶ Everyone will introduce one person from your group to the class

## Course details

# Pre-reqs

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STA 210 and STA 230 / STA 240

## Background knowledge

### Statistical methods

- ▶ Linear and logistic regression
- ▶ Statistical inference
- ▶ Basic understanding of random variables

### Computing

- ▶ Using R for data analysis
- ▶ Writing reports using Rmd
- ▶ Understanding of github
- ▶ Understanding reproducibility

# Course toolkit

## **Course webpage:**

<https://resteorts.github.io/teach/generalized.html>

- ▶ Course information and course schedule

## **Canvas**

- ▶ Changes to Schedule
- ▶ Ed Discussion
- ▶ Homework uploads

## **Gradescope** (link on course webpage)

- ▶ Homework uploads (make sure to upload to Canvas as well).

## **Ed Discussion** (link on course webpage)

- ▶ Course discussion

# Class Meetings

## Lectures

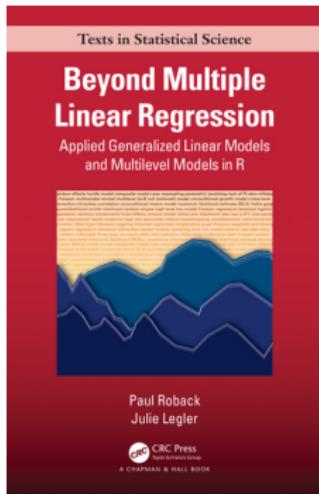
- ▶ Some traditional lecture
- ▶ Short individual and group activities
- ▶ Bring fully-charged laptop / tablet to use R

## Labs (start January 10)

- ▶ Work on class assignments with TA support
- ▶ Time for clarifying questions regarding course material
- ▶ Alternative lecture time when needed

**Attendance is strongly expected (if you are healthy!)**

# Readings



- ▶ Primary textbook: *Beyond Multiple Linear Regression* by Roback and Legler
- ▶ Other texts:
  - ▶ *R for Data Science (2nd edition)* by Wickham, Çetinkaya-Rundel, and Grolemund
  - ▶ Tidy Modeling with R by Kuhn and Silge
  - ▶ Articles and videos periodically assigned

## Computing toolkit

## R and RStudio

- ▶ Install R and RStudio on your laptop
- ▶ Click [here](#) for instructions to install RStudio and configure git

or

Access RStudio through Docker container provided by Duke OIT

- ▶ Reserve a generic **RStudio** container (there is no course specific container)

## Canvas and Gradescope

- ▶ All homework assignments will be uploaded to Gradescope and Canvas.
- ▶ Gradescope allows more fair and balanced grading.
- ▶ Canvas allows us to check the reproducibility of your work.
- ▶ Unfortunately, there is no platform that does both (to my knowledge).
- ▶ Feedback will be given in Gradescope and is individual and private.

## Ed Discussion

- ▶ Online discussion forum (like Piazza, etc.)
- ▶ Platform to ask questions about course content, logistics, assignments, etc.
- ▶ Content organized by channels. Before posting, please browse previous posts to see if your question has already been answered. If not, please post your question in the relevant channel.
- ▶ Questions about grades, absences, and other private matters should be emailed to me with “STA 310” in the subject line.

## Activities & Assessment

## Homework (40%)

- ▶ Individual assignments
- ▶ Combination of conceptual questions, guided analyses, and open-ended analyses
- ▶ Lowest homework grade is dropped

## Quizzes (60%)

- ▶ Individual online quizzes
- ▶ Covers content since the previous quiz, including readings, lecture notes, in-class activities, and homework
- ▶ Lowest quiz grade is dropped

## Grading

Final grades will be calculated as follows

Category	Percentage
Homework	40%
Quizzes	60%

See the course syllabus for letter grade thresholds.

## Course community

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- ▶ Uphold the Duke Community Standard:
  - ▶ *I will not lie, cheat, or steal in my academic endeavors;*
  - ▶ *I will conduct myself honorably in all my endeavors;*
  - ▶ *I will act if the Standard is compromised.*
- ▶ Commit to respect, honor, and celebrate our diverse community
- ▶ Commit to being part of a learning environment that is welcoming and accessible to everyone

## Accessibility

- ▶ The Student Disability Access Office (SDAO) is available to ensure that students are able to engage with their courses and related assignments.
- ▶ If you have documented accommodations from SDAO, please send the documentation within the first week to make sure all accommodations can be put in place as quickly as possible!
- ▶ I am committed to making all course activities and materials accessible. If any course component is not accessible to you in any way, please don't hesitate to let me know.

# Support

- ▶ **Office hours** to meet with a member of the teaching team.
  - ▶ Find the course schedule on the course webpage
  - ▶ Office hours begin January 16
  - ▶ Please see me after class if you have questions before then.
- ▶ **Ed Discussion** for questions about course logistics, content, and assignments
- ▶ **Email** for questions not appropriate for Ed Discussion, e.g., regarding personal matters or grades
  - ▶ Please put **STA 310** in the subject line

See the syllabus regarding additional academic and mental health and wellness resources.

## Latex Resources

1. <https://wch.github.io/latexsheet/latexsheet.pdf>
2. <https://www.bu.edu/math/files/2013/08/LongTeX1.pdf>
3. <https://www.docx2latex.com/tutorials/mathematical-equations-latex/>

# Questions

Questions?