POS 4931: ELECTION DATA SCIENCE

University of Florida — Fall 2020 Online, T 8:30AM-10:25AM; TR 9:35AM-10:25AM

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and by appointment
Course Github Repository

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Course Description

It is increasingly said that data is power. Politics is, at its heart, the study of power—and elections are venues where some of that power is contested. Perhaps it should be no surprise then that there is such fruitful ground to be explored at the intersection of the two...

This course is fundamentally about data: How to find it, wrangle it, model it, and extract practical insight from it. It will be unlike many standard political science courses. Most political science courses expose students to important themes, findings, and ideas in the multifaceted study of power. Most of the effort *here*, though, is less about ideas in (and of) politics and more about developing a particular skill-set. We will be using the programming language R—one of the most popular programming languages in data science and statistical computing—and learning how gain a myriad of different insights from machine-readable data. You will get a hands-on understanding of basic-to-intermediate R programming. You will gain an practical appreciation of, and introductory insight to, data collection, spatial data, and prominent machine learning algorithms. You will learn

the logic of exploratory, inferential, and causal data analysis. You will learn these things the only way they *can* be learned: Through doing and guided practice. You will be able to walk away at the end of this class having learned a set of valuable and in-demand skills.

But this course is also fundamentally about elections. (And, therefore, poltics.) Because the dirty little secret about data is that it doesn't just appear from thin air. Data isn't important due to some intrinsic, magical property, but because it relates to things we already deem important. As denizens of a nation committed to authorizing governing power through democratic means, we hold elections as important. And machine-readable data can be gathered at all stages and venues of the democratic process. State voter files can help us give candidates a cutting edge and/or identify instances of unfair election practices. Census geographic and demographic data can help us shape district boundaries to be competitive and fair. Data from well-crafted experiments can tell us how we can get voters to participate more. And data from high-quality survey projects can give us insight into the thoughts and motivations of voters before, during, and after the election transpires.

So, in this course, we will not just learn about data. We will learn to be critical of data, ethical with data, and competent enough in the practice of analyzing it to gain and apply practical, actionable insight. And we will not just learn about elections. We will learn how to apply these practices and insights to aid the functioning of the electoral process overall. We will, in short, explore the intersection of two popular and prominent wellsprings of power.

I am excited to see where it takes us.

Structure of the Course

This course will guide students through the use of reading assignments, weekly lecture (Tuesdays), a weekly hands-on lab (Thursday), and practical projects to help students acquire the basic skills needed to do election data science. Due to the Coronavirus pandemic, all class lectures, labs, and meetings will be held online. These meetings will be held synchronously through Zoom and the Tuesday lectures will be published to YouTube for those who were unable to attend the lecture during class time. These videos will be private and viewable only to those enrolled in the class and will be available until January 1st, 2021.

Data science is all about transparency. So, in that spirit, I'd like to explain the logic for these structural components.

- Weekly Readings: The weekly readings are meant to be an initial introduction to the concepts and coding techniques required to excel as a election data science practitioner. The examples in the texts are usually geared towards general audiences and not towards political science. But they are excellent introductions to the fundamentals..
- Lectures: The lectures are meant to reinforce and expand that initial introduction. They will be broken down into three parts. The first part will be dedicated to going over standing questions pertaining to the prior week's lab. The second will largely be more "theoretical." It will break-down the underlying logic of the week's topic with the aim of giving your a better grasp of the core concepts at work. It will also provide examples of these concepts in practice in the US elections context. The second will be more of a tutorial. I will demonstrate how

to apply some of these techniques using real election data. Students are expected to follow along and are strongly encouraged to frequently ask questions.

- Labs: Labs are meant to provide you with the opportunity to strengthen (and ideally solidify) the skills that we have learned about during that week and over the ones preceding. These are tasks and challenges that will be made available before class on Thursday. Some of these may be tough, some of them easy, all of them are doable. Here, you will learn through doing—but doing in a supportive environment where I and your fellow learners are available to help should you want/need it.
- Practical projects: If you are taking this class, there is a good chance that you are interested in elections, data, and/or electoral data. You might be so interested that you can see yourself having an enjoyable career in it. (Or, there's a chance you just needed a credit. I get it! No judgments here.) As your instructor, I want all of you to be happy and successful in your endeavors. For data science/data analyst positions, most people do not get hired unless they have some kind of tangible example of their skill that is readily findable by prospective employers. You need to have a portfolio. You don't need to reinvent the wheel or solve the optimal stopping problem, but you need to have something that proves you know what you're doing. These practical experiences are the seeds of that portfolio. You will be writing blog posts that demonstrate your ability to question, communicate, visualize, and write code. (It will also allow you to claim some rudimentary experience with GitHub.) Your final project will show that you understand how to think of a research question, how to pursue it using the techniques of data analysis, and how to communicate it coherently using images and text. Ideally, you might also be able to get a publication credit under your belt. (More on that in class.) All in all, these are to help you take your first steps into doing this professionally.

Required Readings

There are required readings for the class but there are none that you are required to purchase. All of the reading materials are available freely online or will be provided through Canvas.

- Hands on Programming with R by Garrett Grolemund.
- R Markdown: The Definitive Guide by Yihui Xie, J. J. Allaire, and Garrett Grolemund
- blogdown: Creating Websites with R Markdown by yihui Xie, Amber Thomas, and Alison Presmanes Hill
- Statistical Inference via Data Science: A ModernDive into R and the Tidyverse by Chester Ismay and Albert Y. Kim
- R for Data Science by Hadley Wickham and Garrett Grolemund
- Data Wrangling by Sara Altman and Bill Behrman
- Web Scraping with R by Steve Pittard
- Geocomputation with R by Robin Lovelace, Jakub Nowosad, and Jannes Muenchow
- Exploratory Data Analysis with R by Roger Peng

A few notes about the readings:

- Most of these readings include snidbits of code so that the reader can type them into their own copy of R (or copy/paste them) and then run them on their own. While I highly encourage you to do this as you read, you are not required to. Your grades come from the assessments and tasks listed within the syllabus.
- It is perfectly natural to not understand every bit of every detail presented in these books. By assigning these readings, I am not asking you to commit all of their details and code to memory. I am asking your to extract the *logic* of their individual examples with an aim of being able to apply it to election contexts.
- It is a normal part of a programmer's progression to simply copy-paste code and tweak it slightly to fit their needs before. Do not feel like you must become as skilled as these writers are in wrangling R over the course of this semester. Imitation comes before understanding comes before creative synthesis comes before mastery. It's a process.

I strongly encourage you to read sites such as R-Bloggers and Towards Data Science in your free time—especially if you're interested in translating this into a career. These resources are like immersion for language learning. You will learn R and data science far faster, and be far more adroit, if you read materials on these topics regularly aimed at different levels of understanding and skill.

I also encourage students to subscribe to *The New York Times*' "Morning Briefing" email newsletter. It's free and provides news and information on the United States, Canada, and the Americas, each weekday morning. I also recommend following a combination of major news sources (e.g., *Wall Street Journal, NPR, Washington Post*), political news sources (e.g., *The Hill* and *Politico*), and academic-oriented blogs (e.g., *Mischiefs of Faction* and *Monkey Cage*). I also challenge you to make a concerted effort to read from a source that cuts against your prior beliefs. If you see yourself as a political liberal (or leaning that way), I recommend that you try to read from places such as *The Economist* and/or *The National Review*. If you see yourself as a political conservative (or leaning that way), I recommend that you add outlets such as *Vice*, and/or *The Atlantic* to your reading diet. It's important to understand the roots of others' positions—especially when studying how/why they behave the way they do.

Learning Outcomes

By the end of the semester, students enrolling in the course should be able to:

- Understand the logic of various data science techniques and their applicability to elections;
- Apply the tools gained in this class to various kinds of machine-readable data;
- Describe, to both technical and lay audiences, the insights constructed from the analysis of data—as well as intermediate-to-advanced facets of statistical programming in R;
- Think critically about data, its sources, its analysis, its audience, and its impact–especially in the domain of US elections.

Assignments & Course Requirements

There are a number of different kinds of assignments that students are responsible for throughout the semester. The course will be graded out of a total of 1000 possible points. Letter grades will be assigned per the following numerical scales:

```
>= 930
             Α
900 - 920
             Α-
870 - 890
            B+
830 - 860
             В
800 - 820
             B-
770 - 790
            C+
730 - 760
             \mathbf{C}
             C-
700 - 720
670 - 690
            D+
630 - 660
             D
600 - 620
             D-
<=600
             Е
```

Points will be rounded to the nearest whole-number (e.g., $869.5 \rightarrow 870; 869.4 \rightarrow 869$). Additional extra credit will not be provided, aside from that stipulated below.

- Attendance: Attendance will be worth 5 percent of the total grade, or 50 points. The grade will comprise of two main elements.
 - 1. The first is attendance to be "taken" every lecture class, via Google Forms. Each lecture will have a unique "key-phrase" that I will provide to the students. Students will then input the key-phrase and UFID into the Google Form. (If you are unable to attend the virtual lectures live due to Internet or personal issues, you can get the key-phrase by watching the lectures posted on YouTube.) Students are allowed up to 2 unexcused absences throughout the semester. Excused absences include those taken for medical purposes, school-sanctioned events (however many exist during this time of COVID), and family emergencies. They also include notable family and life events (weddings, birthdays, etc). If you know in advance that you will miss a class, please let me know ahead of time. In the event of an unforeseen absence that you want excused, please contact me within one week of the date missed and we will discuss it. Exceptions to this rule are at my discretion. Each unexcused day after those two will result in a loss of 10 points apiece to the participation grade.
 - 2. The second is office hours. Students are required to attend at least one office hour session throughout the semester. Not attending at least one office hours session will result in a 20 point loss to your attendance grade. The intent of this policy is so I can get to know you as students better so that I can help you should you need a letter of recommendation, a reference for a job or internship, etc. If you aren't able to make any of the posted office hours, please e-mail me and we'll set up an appointment at a time that works for both of us.
- Check-In Quizzes: Each week before class, students are expected to respond to a check-in quiz on Canvas, comprised of three questions: One will ask students to reflect on the material that they were assigned to read before that week; another will ask them if there were any

questions from the lab that stumped them; and a third asks if there are any concepts that they are still having a hard time understanding. These are all intended to help you deepen your understanding and skills over the course of the semester. These are due **the Monday before class at 11:59 PM**. (Note: There will not be a check-in quiz covering the first week of the semester. Instead, there will have a 10 question syllabus quiz due **Thursday**, **September 10th**.) Each quiz is weighted the same and all of the quizzes, cumulatively, amount to **5 percent of the total grade**, or **50 points**.

- Weekly Labs: Every Thursday, students will be provided with a set of tasks to be accomplished. These will be an extension of the concepts and skills learned from the lecture and reading. Each assignment is due the following Tuesday at 11:59 PM through Canvas. Students are given the whole of Thursday's class to complete the assignment. If you run into a problem, you are encouraged to ask me and your classmates for assistance. You are welcome to consult the almighty oracles (Google and Stack Exchange) for assistance. Each question will come with an additional "challenge." Challenges are worth extra credit applied to that week's lab; students can earn a maximum of 10 percent above the original maximum score for the assignment. Each challenge accomplished will provide extra credit for that week's assignment. Labs are equally weighted and worth 30 percent of the overall grade, for a total of 300 points.
- Blog Posts: Throughout the semester, students will be responsible for making their own blog through the blogdown package in R and hosting it on a personal GitHub page. They will also be responsible for populating it with at least three posts. Each post must contain the following attributes:
 - It must be a critique, reflection, or extension of the concepts covered in the class/lab/readings.
 - It must include highlighted code that is functional (i.e., it would run if someone followed the steps you describe).
 - It must contain a visualization that clarifies or illuminates the post's point. This can be
 a table, graph, interactive plot, dashboard, etc. So long as it is more than simply prose
 on the page.

Blog posts will be worth approximately 7 percentage points apiece for a total of 20 percent of the final grade, or 200 points. (If students catch the blogging bug and wish to write more, I will allow them to identify the posts they want me to grade.) Links to blog posts will be submitted through Canvas. The links can be submitted throughout the semester but are due to me at latest on December 10th

- Final Project: Students will be responsible for completing an original Election Data Science project. Students can work alone or in groups of up to 4. (If students opt to work in groups, we will need to schedule a time to meet to make sure the work is divided fairly.) The project requires the following attributes:
 - Topic: The topic of the project must be original and must have at least two of the following four focuses: Description; Inference; Prediction; and/or Programming.
 - Data: The project must center around machine-readable data with clear pertinence to elections and/or electoral behavior. This can include voter registration data, survey data, experimental data, or social media data.

- Replicability: In theory, it should be possible for someone to take your data and your code and near-perfectly replicate your analyses.

Students are expected to make two deliverables. Both are worth 15 percent of the total grade, or 150 points apiece

- A Technical Report: This is to be made as an R Markdown document. The report should describe the motivation behind the project, sources of data, the theory driving your approach, challenges faced in the analysis process, how you surmounted them, and the results of your hard work. This should provide all the code chunks and visualizations that allow a technically-inclined reader to understand what your project is, how you did it, and why it's important. The report should be between 1,000-2,000 words (not including code chunks and text embedded in the visualization), written as an RMarkdown document, and is to be submitted to me via Canvas no later than the course's final exam day (December 15th).
- A Public-Facing Output: This is so you can describe the project, the general logic, its importance, and your findings to a non- technical audience. This is to be shared/presented to the class. The default option is a PowerPoint (or Slidy/Beamer/etc.) presentation, but you have full creative freedom in how you approach this. It can be a Medium blog post, a video, a dashboard, a scrolly-telling story, a podcast—sky's the limit; go for it! This should take between 5-12 minutes to present. Presentations will begin the week after Thanksgiving and will (ideally) be completed prior to the start of reading days. However, we will use the Final Exam block to wrap-up presentations if need be. The grade for this component will be ascertained by the accuracy and clarity that the student(s) conveyed the technical material.
- Students/groups that use GitHub for version control and/or code sharing will receive 20 points extra credit on the assignment.

Students will submit a short proposal (roughly 1–2 double-spaced pages, or 100-500 words) on what they would like to do through Canvas by **Tuesday, October 13th.** (If students are working as a group, only one proposal for the entire group is needed. However, all members of the group must be identified in the proposal). The proposal will be worth **50 points.** Students will also submit a short (roughly 1–2 double-spaced pages, or 100-500 words) status update through Canvas by **Tuesday, November 10th.** This will also be worth **50 points.** (If students are working as a group, only one status-update for the entire group is needed.)

- Extra Credit Opportunities: Students will have the following opportunities to earn extra credit throughout the semester.
 - Create/update a LinkdIn profile: I want you to get gainful employment. LinkdIn is becoming increasingly necessary for that. (15 points.)
 - Data Science Resume: Same rationale as above. (15 points.)
 - R Bloggers: Submit your blog to R-bloggers. This will help give your work some exposure and it can be seen favorably to future employers (10 points.)
 - Original Election Data Science Memes: 2020 is a dumpster fire. Let's make each other laugh. Submitted memes will be featured prior to lecture. 1 point for every submitted meme with an extra 1.5 points given if the meme is original. (Soft-cap of 10 points with

a hard cap of 15 for originality.) Note: Memes using racist, sexist, and/or prejudicial humor will not only *not* be given extra-credit, they will be referred for disciplinary action.

Course Policies

- Office Hours: During the semester, office hours are an opportunity to get help with assignments and discuss course material, including grading. Students will be required to attend one office hour session (see assignments above). Aside from this, students are strongly encouraged to take advantage of this time to ask questions or just swing by so I can get to know you better. The better I know about you, the more I can write for a letter of recommendation or talk about as a reference to a potential employer. Appointments are also available.
- Lecture: You are expected to attend the synchronous lectures as best as can reasonably be expected. Normally, these class meetings be required. But, also, normally we'd be meeting in an actual class without a pandemic going on. Some of you might not have reliable internet to get to classes live. Some of you might be in a timezone where an 8:30AM class is unreasonable. (e.g., if you're living in California, you don't need to wake up at 5:00AM just for my class. Get some sleep; watch it on YouTube.) "Attendance" will be taken in a way that accommodates both those who could attend live and those who are unable to.
- Late Assignments: Without any notice of mitigating circumstances, I will deduct 10 points for every day that an assignment is late beyond the original due date. Most assignments will be due at midnight. However, I do not start work until 8:00 AM—so consider the period between 12:00–8:00 AM as an "emergency buffer." This means that I won't hold it against you if an assignment is occasionally submitted during that time frame. Do not make a habit of it though. I reserve the right to count submissions as late if they are submitted during this time-frame if it appears that the student is abusing this policy. If you have a mitigating circumstance that prohibited you from reasonably completing the assignment on time, please reach out to let me know. I can't help if I don't know.
- Children in Class: For students with children (or living with younger siblings), it is understandable that unforeseen disruptions in childcare can occur. While not a long-term solution, bringing a child to "class" with you when such disruptions occur is acceptable. In those cases, please be courteous to your fellow students and mute your microphone if they are becoming too disruptive. Likewise, all other students should work together to create a welcoming environment for both the parent/caregiver and child—which means being kind and encouraging as well as patient of minor disruptions. If you are unable to come to a long-term solution, please contact me via email. We will work towards a solution.
- Disruptions in Class: Please silence your phone, emails, and chat applications during class meetings. If you are in a noisy environment, please mute your microphone when you are not talking or asking/answering a question. We are meeting early so I will not be offended if you eat/drink breakfast during the scheduled time—but please be sure to mute your microphone should you decide to do that. Most importantly though: Remember that we are all trying to do our best in a very weird time. Please be patient with your fellow students if a disruption occurs. Also: Never apologize if your pet, child, or younger sibling spontaneously pops-in despite your best efforts. Animals and small children are objectively adorable and quick adorable distractions are always welcome.

- Class Demeanor: Students are expected to arrive to class on time and behave in a manner that is respectful to the instructor and to fellow students. Derogatory comments, personal attacks on others, or interrupting the class will not be accepted. Please avoid the use of cell phones and electronics for extraneous purposes. Opinions held by other students should be respected in discussion, and conversations that do not contribute to the discussion should be held at minimum, if present at all.
- Recording: Our Tuesday lessons (after the first class) will be visually recorded for students in the class to refer back and for enrolled students who are unable to attend live. Students who participate with their camera engaged or utilize a profile image are agreeing to have their video or image recorded. If you are unwilling to consent to have your profile or video image recorded, be sure to keep your camera off and do not use a profile image. Likewise, students who un-mute during class and participate orally are agreeing to have their voices recorded. If you are not willing to consent to have your voice recorded during class, you will need to keep your mute button activated and communicate exclusively using the "chat" feature, which allows students to type questions and comments live. The chat will not be recorded or shared. As in all courses, unauthorized recording and unauthorized sharing of recorded materials is prohibited.
- **Technology:** You are expected to have a laptop and/or computer that is capable of running R and RStudio. Please limit all non-class related activity during class time as much as possible.
- Course Evaluation: Students are expected to provide feedback on the quality of instruction in this course by completing online evaluations via GatorEvals at ufl.bluera.com/ufl. Students will be notified when the evaluation period opens at the end of the semester. Summary results are available to students at gatorevals.aa.ufl.edu/public-results.
- Subject to Change: This syllabus is subject to change at the discretion of the instructor to accommodate instructional and/or student needs. Proper notification will be provided to students of relevant changes.

University Policies

- Accommodation: Students requesting accommodations should first register with the Disability Resource Center (352-392-8565) by providing appropriate documentation. Once registered, students will receive an accommodation letter which must be presented to the instructor when requesting accommodation. Students with disabilities should follow this procedure as early as possible in the semester.
- Non-Medical Accommodation: Many students are involved involved with school sanctioned extracurricular activities and/or have other professional commitments (e.g., work, military obligations, etc.). If your schedule for these activities/obligations will conflict with the meetings for this class, please let me know ahead of time.
- Plagiarism: Plagiarism that is, lifting without giving credit from something someone else has written such as a published book, article, or even a student paper is forbidden. As a writer, I can tell you that the words people pen are more than just ink on a page or pixels behind a screen. They are the result of deliberate effort and represent part of their contribution to the cumulative human experience. Taking those words and passing it off

as your own is deeply unethical and will not be tolerated. Modern technology means that plagiarism is fairly easily detected. Do not do it. You will get caught. There will be consequences.

- Academic Honesty: Academic honesty and integrity, more broadly, are fundamental values of the University community. Cheating in any form undermines the integrity and mutual trust essential to a community of learning and places at a comparative disadvantage those students who respect and work by the rules of that community. It is understood that any work a student submits is indeed his/her own. There are other, more obvious forms of academic dishonesty, such as turning in work completed by someone else, and offering or receiving whispered, signaled, or other forms of assistance during an exam. Working with fellow students in exam/study groups is not only acceptable but also encouraged so long as one is refining ideas that are essentially their own. Please review the University's policies regarding student conduct and conflict resolution, available through the Dean of Students Office. Any violations of the Student Honor Code will result in a failing grade for the course and referral to Student Judicial Affairs.
- Communication Courtesy: Per university policy, all members of the class are expected to follow rules of common courtesy in all messages and other electronic communications. Under Florida law (FS 119.07), GatorLink emails are public records. If you do not want your email to be released in response to a public records request, contact the instructor in person. Per university and federal policies, grades may not be discussed via e-mail or over the phone. Please allow 24-48 hours for a response.
- Counseling and Wellness Center: Contact information for the Counseling and Wellness Center: counseling.ufl.edu, 352-392-1575; and the University Police Department: 352-392-1111 or 9-1-1 for emergencies.

Important Dates

Aug. 31	Classes Begin
Sep. 4	Last Day to Add or Drop a Course
Sep. 7	No Classes (labor day)
Sep. 10	Syllabus Quiz Due
Sep. 18	S\U grade option
Oct. 2 & 3	Homecoming
Oct. 13	Proposal Due
Oct. 24	Early Voting Begins in Florida
Nov. 3	Election Day
Nov. 11	No Classes (veterans day)
Nov. 13	Progress Report Due
Nov. 23	Last Day to Drop with a "W"
Nov. 24	Course Evaluations Open
Nov. 25–28	Thanksgiving Break
Dec. 1	Presentations Begin
Dec. 10	Blog Posts Due

Dec. 10 & 11 Reading Days

Dec. 15 Final Assignment Due

Dec. 15 Final Exam Day

Dec. 23 Final Grades Available

Course Schedule

The following outline reflects my expectations before the class begins, but weekly coverage may change depending on the progress of the class.

Week 1, August 31– September 6: Course Structure and What Is Data Science In God we trust: All others must bring data — W. Edwards Demming

- T: No assignment due; first day of class
- TH: Week 1 lab
- TH: The Quartz Bad Data Guide

Week 2, September 7–13: Programming & Reporting Basics

Life is what happens when your code is compiling.

- T: Grolemund, Chapters 1–3
- **T:** Xie et al., Chapter 2.1–2.6
- **T:** Xie et al., Chapters 1–2
- TH: Week 2 lab
 - Syllabus Quiz Due

Week 4, September 14–20: Data Manipulation & Munging

Facts are stubborn, but statistics are more pliable—Mark Twain

- T: Ismay and Kim, Chapter 3
- T: Wickham and Grolemund, Chapter 5
- TH: Week 4 lab

Week 5, September 21–27: Data Acquisition

"Data! Data!" He cried impatiently. "I Can't Make bricks without clay!"—Sherlock Holmes

- T: Ismay and Kim, Chapter 4
- T: Altman and Behrman, Chapters 1–2
- T: Altman and Behrman, Chapter 12
- TH: Week 5 lab

Week 6, September 28-October 4: Data Visualization

Above all else, show the data.—Edward Tufte

• T: Wickham and Grolemund, Chapter 3

- T: Rules for Making Great Charts by Martin Telefont
- TH: Week 6 lab

Week 7, October 5–11: Mapping

"We actually made a map of the country, on the scale of a mile to the mile!"

"Have you used it much?" I inquired.

"It has never been spread out yet." —Alice Through the Looking Glass

- T: Wickham and Grolemund, Chapter 3
- T: Lovelace et al. Chapters 2 & 8
- TH: Week 7 lab

Week 8, October 12–18: Exploration I

"Would you tell me, please, which way I ought to go from here?"

"That depends a good deal on where you want to get to."

"I don't much care where."

"Then it doesn't much matter which way you go."-Alice in Wonderland

- T: Wickham and Grolemund, Chapter 7
- T: Peng Chapters 4 & 6
- TH: Week 8 lab

Week 9, October 19–25: Inference I

A pebble in the water makes a ripple effect; every action in this world will bear a consequence.—**The**Red Jumpsuit Apparatus

- T: StatQuest:Fitting a line to data
- T: Canvas Readings (Woolridge)
- TH: Week 9 lab

Week 10, October 26- November 1: Prediction I

It's tough to make predictions, especially about the future—Yogi Bera

- **T:** Boehmke & Greenwell, Chapters 1, 2, and 4.
 - Short proposal due on Canvas.
- **TH:** Week 10 lab

Week 11, November 2-8: Exploration II

Judging a book by all the surrounding covers.

• **T:** *Peng*, Chapters 12–13.

- T: Boehmke & Greenwell, Chapter 8.
- **TH:** Week 11 lab

Week 12, November 9–15: Prediction II

Those who have knowledge don't predict. Those who predict, don't have knowledge.-Lao Tzu

- **T**: *Peng*, Chapters 5, 6, 14
- T: Use of election forecasts in campaign coverage can confuse voters and may lower turnout.
- **TH:** Week 12 lab
 - Check-in report due.

Week 13, November 16-22: Inference II

I love fools' experiments. I am always making them-Charles Darwin

- T: Canvas Reading (Gerber and Green)
- **TH:** Week 13 lab

Week 14, November 23-29: Thanksgiving

Consider the turkey that is fed every day. Every single feeding will firm up the bird's belief that it is the general rule of life to be fed every day by friendly members of the human race...On the afternoon of the Wednesday before thanksgiving, something unexpected will happen to the turkey. It will incur a revision of belief.—Nassim N. Taleb

• No class. Stay safe.

Week 16, November 30– December 6: Presentations I

The numbers never speak for themselves.

- **T:** Presentations
- **TH:** Presentations

Week 17, December 7- December 13: Presentations II

Tell me, and I forget; teach me, and I may remember; involve me, and I'll understand-**Xunzi**

- **T:** Presentations
- TH: No class; reading day

Week 18, December 14- December 20: Final Exam Day

The further one goes, the less one knows.-Lao Tzu

- **T:** Final exam day. (Possible presentation day).
 - Final projects due.

My deepest appreciation goes to Stephen Philips, who graciously allowed me to recycle materials from his prior syllabi. (See how easy it is to not plagiarize!)

After having reviewed the entire syllabus, do not forget the syllabus quiz due **September 10**. To confirm that you've read the syllabus all the way through, please e-mail me your favorite picture involving a dinosaur. If you don't have a favorite picture of a dinosaur, there's never a wrong time to find one! Doing so will give you 5 points towards your final grade.