



# Statistics 442

## Spring 2021

### **Course Description: Introduction to Bayesian Data Analysis**

The course will introduce data analysis from the Bayesian perspective to undergraduate students. We will cover important concepts in Bayesian probability modeling as well as estimation using both optimization and simulation-based strategies. Key topics covered in the course include hierarchical models, mixture models, hidden Markov models and Markov Chain Monte Carlo.

### **Prerequisites:**

1. A course in probability (Statistics 430 or equivalent)
2. A course in statistical inference (Stat 102, Stat 112, Stat 431 or equivalent)
3. Basic knowledge of linear algebra (e.g. matrix multiplication and inversion, etc.)
3. Experience with the statistical software R (at the level of Stat 405 or Stat 470)

### **Professor:**

Dr. Shane Jensen  
[stjensen@wharton.upenn.edu](mailto:stjensen@wharton.upenn.edu)

JMHH 463  
215-573-2211

**Lectures:** Section 1: TTh 1:30-3pm (Virtual: links will be posted on Canvas)  
Section 2: TTh 3-4:30pm (Virtual: links will be posted on Canvas)

**Office Hours:** Wed 2:30-4:30pm (Virtual: links will be posted on Canvas)

### **Recommended Textbook:**

Bayesian Data Analysis (3<sup>rd</sup> Edition) by Gelman, et.al.

### **Required Software:**

The R statistical package is needed and can be downloaded at [www.r-project.org](http://www.r-project.org)

### **Course Website:**

All course materials will be available on Canvas

## Evaluation

Your course grade will be calculated as:

- 50% homeworks
- 25% midterm exam
- 25% final exam

**Midterm Exam** will be a **24 hr take home exam** starting at 1:30pm on **Thu March 4**

**Final Exam** will be a **24 hr take home exam**, **date and time TBA**

### Notes about Grading:

- *No late homeworks will be accepted, for any reason whatsoever.*
- **No make-up midterm will be given**

## Course Topics

1. Introduction to Bayesian Inference (Ch.1)
2. Simple Parametric Models (Ch. 2, 3)
3. Regression Models from the Bayesian Perspective (Ch. 14,15)
4. Hierarchical and Mixture Models (Ch. 5)
5. Optimization Algorithms for Model Estimation (Ch. 13)
6. Monte Carlo Simulation Algorithms for Model Estimation (Ch. 10,12,13)
7. Model Checking (Ch. 6,7)
8. Logistic Regression Models
9. Nonparametric and Semiparametric Bayesian models (Ch. 23)
10. Hidden Markov Models
11. Bayesian Tree Models