

COMP 543: Tools & Models for Data Science

Course Overview

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This Class is about *Data Science*

- What is THAT?
- Extraction of actionable knowledge from large volumes of data
 - Encompasses methods from:
 - Computer science
 - Statistics
 - Optimization/Applied Math
 - Also includes
 - Domain knowledge
 - Communication skills
 - Data management

Examples of Data Science Tasks

- Given a huge set of per-customer sales data, build a model to predict customer “churn”
- Given a large graph of Medicare payout data, find suspicious (potentially fraudulent) referral patterns
- Given a set of EMR data, find previously unknown side effects (ex: Vioxx and heart disease)
- Given data from an online learning tool find markers that are an early sign of later academic achievement problems
- Many, many more!

Both Tools and Models are Important

- Back in the day...
 - You had statisticians who dealt primarily with small data sets
 - You had computer scientists who were interested in advanced modeling
- But in the “Big Data” era, the two can’t live in isolation
 - You need advanced models to solve challenging prediction/analysis tasks
 - You need computer systems that can scale those models to the largest data sets
 - You need computer tools that make it easy to implement complicated models

Important Disclaimer

- 543 is a hard-core computer science class!
- This is not “tools and models” from a naive user’s perspective
 - No learning to be an end-user of classical analytics packages
 - This is not a “Get to know R” class
 - Nor is it a “Get to know SAS” class
 - No plugging data into a standard software package and writing a report on the results
 - A class covering such topics WOULD be useful
 - But that’s simply not this class
- Lots of focus on algorithms and engineering

- Focus on data / problems in the biomedical domain
- Familiarity is helpful, but not required
- The models and tools are general, but we will explore applications in medicine and bioinformatics
 - Gene-gene interactions
 - Prescribing practices
 - Genetic sequence similarities
 - Medical publication abstract analysis

When We Say “Models”

- Strong focus on the math foundations of data science
- Lots of optimization theory, probability, statistics
- Even some continuous mathematics
- Here's a slide from one of the later lectures:

- Nasty!! Or is it? Consider just one variable, z_1 ; try to separate out. Write as:

$$\begin{aligned} &= c_{1,1} (x_i \log(p_1) + (10 - x_i) \log(1 - p_1)) \sum_{\langle z_2, z_3, \dots \rangle} a(\langle z_2, z_3, \dots \rangle) \\ &+ \sum_{\langle z_2, z_3, \dots \rangle} a(\langle z_2, z_3, \dots \rangle) \sum_{i=2}^n b(\langle z_2, z_3, \dots \rangle) \\ &+ c_{1,2} (x_i \log(p_2) + (10 - x_i) \log(1 - p_2)) \sum_{\langle z_2, z_3, \dots \rangle} a(\langle z_2, z_3, \dots \rangle) \\ &+ \sum_{\langle z_2, z_3, \dots \rangle} a(\langle z_2, z_3, \dots \rangle) \sum_{i=2}^n b(\langle z_2, z_3, \dots \rangle) \\ &= c_{1,1} (x_i \log(p_1) + (10 - x_i) \log(1 - p_1)) + \text{other terms w/o } z_1 \\ &+ c_{1,2} (x_i \log(p_2) + (10 - x_i) \log(1 - p_2)) + \text{other terms w/o } z_1 \end{aligned}$$

When We Say “Tools”

- We mean tools for manipulating large data sets
- Tools for scalable, distributed computation
- Focus is on “Big Data”!
- Specifically, we’ll learn about:
 - SQL databases
 - Python programming (NumPy, SciPy)
 - Hadoop (MapReduce software, Big Data file system)
 - Spark (distributed Big Data manipulation software)
 - TensorFlow (tool for building learning algorithms)

Example Use Case for Your 543 Skill Set

- Imagine...
 - You are working at a hospital
 - You collect 5TB of patient monitoring data each day...
 - And want a software to predict what will happen to a patient in the next hour
 - Such a software does not exist...
 - How to build it?
- Key questions to answer:
 - How will you process the raw data?
 - What model will you use to do prediction?
 - How will you train the model?
 - How will you scale to 5TB per day?
- After 543, you'll have the answers!

As Such, this Class...

- Will give an introduction to modern data management software...
 - First half of the class
 - Relational database systems and SQL
 - No-SQL systems such as Hadoop and Spark
- Will give an introduction to models for modern data analysis...
 - Second half of the class
 - Basic optimization theory
 - Supervised learning (linear models, support vector machines)
 - Unsupervised learning (clustering, matrix factorization)
 - Text mining
- Projects will focus on implementing the models using the tools

Skills You Need to Take this Class

- Should be a proficient programmer
 - Really good in a modern, general-purpose language
 - Python preferred
 - Two assignments use SQL (no knowledge assumed)
 - Four assignments use Python

More Skills You Need to Take this Class

- Should not be afraid of a bit of math
 - Some background in probability/statistics
 - Common distributions (e.g. Gaussian)
 - Expected value
 - Variance, covariance
 - Norms (e.g. L_1, L_2)
 - Some calculus (partial derivatives & the chain rule should not freak you out!)
 - Linear algebra
 - Vectors and scalars
 - Matrix inversion
 - Matrix transposition
 - Dot products
- Fluency in English to be able to read research papers, evaluate them critically, and find related papers

What About Overlap with Other Classes?

- COMP 533—biggest overlap
 - First three weeks of class are going to strictly be review
 - As will be the first two assignments (a lot like COMP 533 assignments)
- COMP 440/502/540/602
 - Many/all of the methods we'll cover will also be covered in those classes
- So, what's the point of taking this class?
 - The only place where you can get an overview of all of this in one place
 - Focus on big data and tools that operate on big data

Teaching Assistants

- Gau Pan
- Sean Wang
- Gabe Vacaliuc
- Office hours will be posted on Piazza

- Communication...
- Grading and Evaluation...
- Exams...
- Academic misconduct...
- Assignments... (more on the next slide)

Assignments

- Exercises

- 4 short programming exercises designed to reinforce in-class concepts

- Labs

- 7 one-hour activities to get initial hands-on experience with a practical concept

- Programming Assignments

- 6 in-depth programming assignments

- Research

- 7 research / writing assignments to increase understanding of a key topic and how it has been used in a domain of interest

- **Assignment & Exercise Due Dates**
 - Typically due at 11:55 PM
 - 1 second – 24 hours late = 10% penalty
 - 24 hours + 1 second – 48 hours late = 20% penalty
 - > 48 hours late: NOT ACCEPTED
 - Last assignment may **NOT** be submitted late
- Canvas is the time keeper – if Canvas says it's late, it's late
- Exceptions will only be made for EXTENDED Canvas outages
- Submit early!

Class Policies – Extensions

- Must be requested ≥ 1 week in advance
- Exceptions possible for very extenuating circumstances, with proper documentation

Class Policies – Regrades

- Must be requested within 1 week of assignment being returned
- Intended for errors in grading or MINOR errors
- Not a week-long extension to the assignment
- Process
 - Talk to Risa, after class or during office hours
 - **Type** up request
 - Submit in person or under the door to DH 2062

Questions?

- If there's time: on to databases!!!
 - What's a database system?