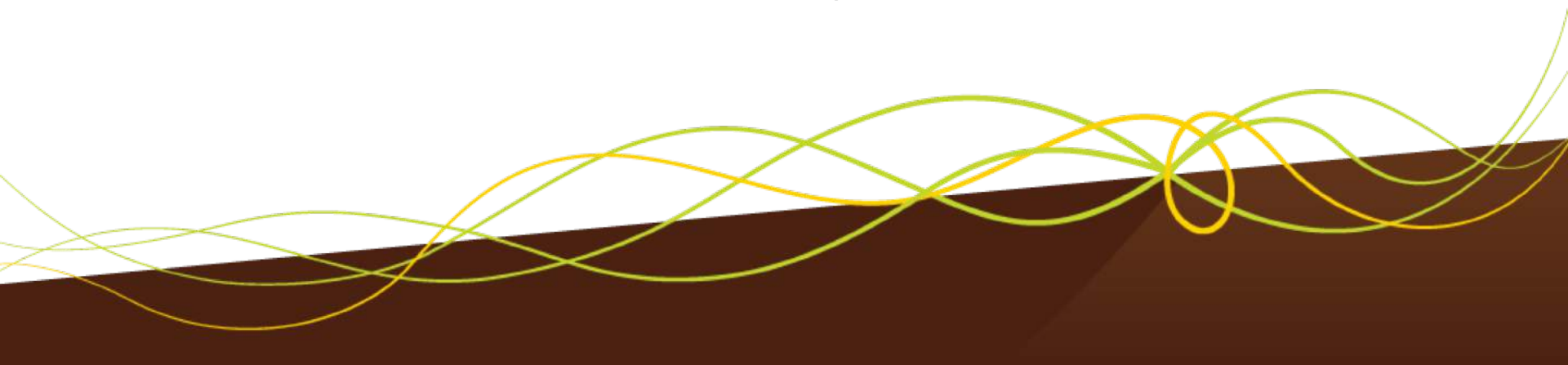


A Project-Driven Introduction to Data Science

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Outline

- Course Overview & Context
- Why projects? Why early?
- How to have ‘successful’ projects
- Things I haven’t figured out yet
- Where to find resources
- Questions

Institutional Context

- Valparaiso University:
 - About 3500 undergrads
 - About 800 engineering students (28 faculty)
 - Lutheran (faith-based) institution
 - 15 full-time Math/Stat faculty (14 Tenure-Track)
 - 9 full-time CIS faculty (4* Tenure-Track)
- Data Science Program
 - Housed in Mathematics & Statistics (MST)
 - *Director holds affiliate appointment in Computing and Information Sciences (CIS)
 - Started in Fall 2016
 - Graduate Program in Analytics and Modeling since 2011

Course Overview/Context

- First* course in “Data Science” (Major)
 - Minimal prerequisites
- Offered in SPRING semesters
- Typically taken by:
 - Freshmen/Sophomores (majors or Stat/Math)
 - Junior/Seniors (CS or Engineers)
- 15 Weeks, Offered as 2+3:
 - ‘Lecture’ 2x a week (50 min)
 - Lab/Project Time 2x a week (75 min)
 - All sessions meet in a computer lab

Course Overview/Context: Learning Goals

- Course Goals:
 - Understand the fundamental concepts of data science and knowledge discovery
 - Apply and perform the basic algorithmic and computational tasks for data science
 - Develop and improve analytical thinking for problem formation and solution validation

Course Overview/Context: Learning Goals

- Topical Objectives:
 - Gain an overview of the field of knowledge discovery
 - Learn introductory data mining algorithms
 - Be able to distinguish and translate between data, information, and knowledge
 - Apply algorithms for inductive and deductive reasoning
 - Apply information filtering and validation on real world datasets
 - Understand the social, ethical, and legal issues of informatics and data science
 - Apply data mining, statistical inference, and machine learning algorithms to a variety of datasets

Why Projects?

- Evidenced-based best practice for engaging minorities (and students in general)¹
- Real data is messy!²
- Prepares them for realistic workforce experiences
- Portfolio/Resume building

[1]Corbett, C. and Hill, C. 2015. Solving the equation: the variables for women's success in engineering and computing. DC: AAUW. (2015).

[2]Committee on Envisioning the Data Science Discipline: The Undergraduate Perspective et al. 2018. *Data Science for Undergraduates: Opportunities and Options*. National Academies Press.

Why early?

- Provide context for topics/learning in future classes
- Builds important communication and self-management skills
- Evidenced-based best practice for improving retention! (real data)
- Establishes resume/portfolio for early internship applications
- Better prepares them for doing undergraduate research
- (Potentially) provides a unified topic/dataset for many future classes
- **It's fun..**

And why they to be Data Scientists!

HAVING 'SUCCESSFUL' PROJECTS

The background features a large, abstract shape that curves from the bottom left towards the top right. This shape is divided into two main color sections: a bright yellow section on the left and a dark brown section on the right. The overall composition is clean and modern, with the text positioned in the upper left quadrant.

(1) MANAGE EVERYONE'S EXPECTATIONS!

This is hard. But by far, the most important.

(a) Carefully define “success”

- For Students
- For Clients
- *From an Instructor perspective

(b) Established expected project outputs/deliverables early

(2) Start getting projects EARLY
(even earlier than you think you need to)

This year I started in mid-November...
(‘war story’ of issues with this)

(3) Have several mile-stones during semester

Each mile-stone has 'sub'-deliverables.

I use 5 phases:

1. Proposal & Design Sketch
2. Data Processing & Design Specifications
3. Algorithm Plans (and Problem Revisit)
4. Basic Data Pipeline/Product
5. Final Paper (with 3 sub-stages here)

(4) (Carefully) Assign Teams

Students rank projects.

I assign teams, striving for:

- Balance of grade level/majors
- Mix of (previous) skills needed for projects
- Non-isolation of minorities/gender
- Personalities

Things I changed in Year 2...

- Heavier front-loading of programming assignments
- Required use of GitHub (sorta...)
- Included significant in-class time for working on projects
- Had clients present the outcomes/use of projects at the end of the semester

Things I haven't figured out:

- Getting students to have more professional engagement with clients
 - Limit number of interactions
 - Better scheduling and student commitment
- Course is still a LOT more work than most introductory courses
 - FOR EVERYONE – Students and faculty
- Where/how (website/wiki) to have students post materials for general sharing
- How much programming to front-load
- How much project-management concepts to front-load

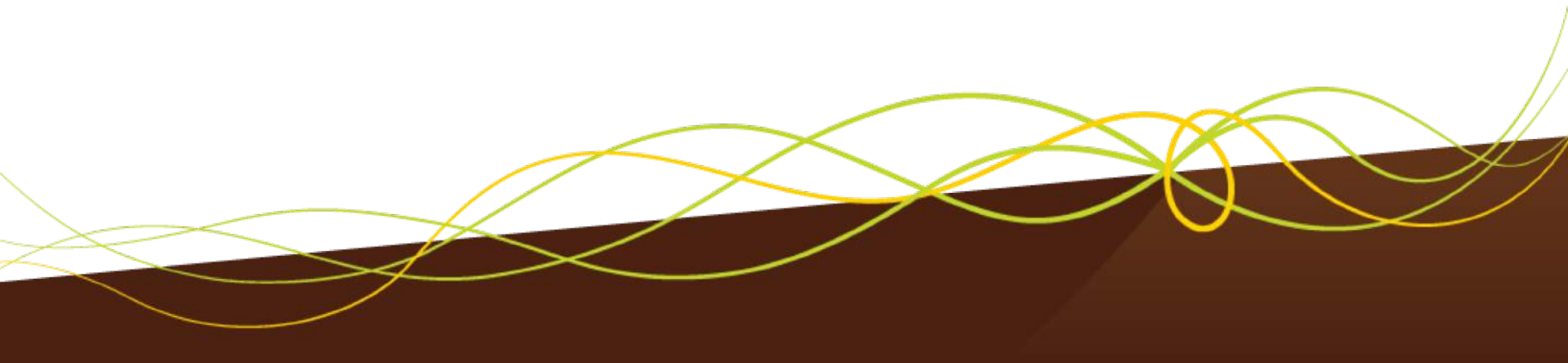
Things I'm changing next year:

- Senior student project managers
 - Former students (as TAs), majors, etc.
- Explicit and limited (5-6) meetings between students and clients
- Getting projects even earlier!
 - Especially the data!
- Extended in-class discussions of each phase/mile-stone expectations
 - Provide exemplars (and failures)
- (Maybe) Start working on projects a little later (last time, Week 2 of course)

Resources

- SIGCSE and SIG-STATED List-Serves
- Projects:
 - Kaggle.com
 - Challenge.gov
 - Riipen.io and Telanto.com
- <http://www.teachingdatascience.org/>
 - Sort of defunct
- ‘From the Director’s Desk’
 - blogs.valpo.edu/datadesk
 - A blog about data science education and curriculum

Thank you for listening!



Questions?

