INTRO TO DATA SCIENCE LESSON 1: DATA EXPLORATION

INTRO TO DATA SCIENCE

WELCOME

LOGISTICS 3

CONTACT

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Class M/W 6:30-9:30

LOGISTICS 4

OFFICE HOURS

9-9:30 and Google Hangouts (schedule accordingly)

COURSE NOTES, WIKI, AND LAB/PROJECT SUBMISSIONS teachingdatascience.github.io/data-science-course/

AGENDA 5

I. WHAT IS DATA SCIENCE? II. THE DATA MINING WORKFLOW

LAB:
III. COMPUTER SETUP
IV. DATA PRACTICE

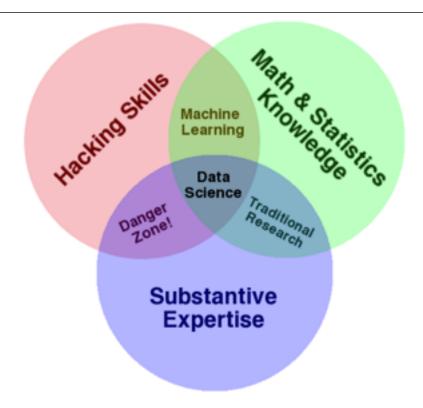
I. WHAT IS DATA SCIENCE?

A set of tools and techniques used to extract useful information from data.

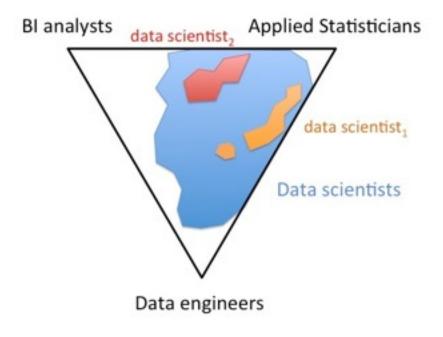
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An interdisciplinary, problem-oriented subject.

THE QUALITIES OF A DATA SCIENTIST



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An interdisciplinary, problem-solving oriented subject.

The application of scientific techniques to practical problems.

A rapidly growing field.



















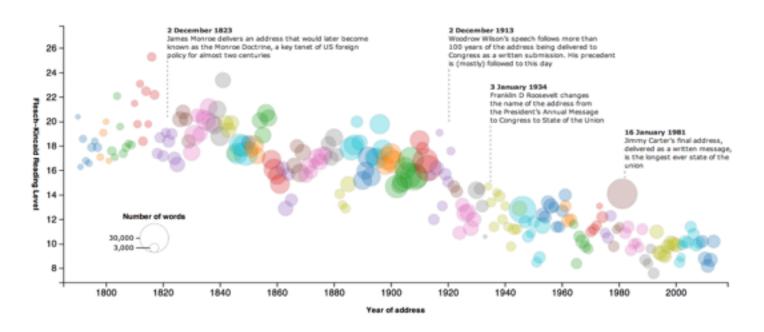




The state of our union is ... dumber:

How the linguistic standard of the presidential address has declined

Using the Flesch-Kincaid readability test the Guardian has tracked the reading level of every state of the union



- "Data Scientist' is a Data Analyst who lives in California"
- "A data scientist is someone who is better at statistics than any software engineer and better at software engineering than any statistician."
 - "A data scientist is a business analyst who lives in New York."
 - "A data scientist is a statistician who lives in San Francisco."
 - "Data Science is statistics on a Mac."



Michael E. Driscoll @medriscoll



Following

Data scientists: better statisticians than most programmers & better programmers than most statisticians bit.ly/NHmRqu @peteskomoroch











- Statistical and machine learning knowledge
- Computer Science and Engineering experience
- Academic curiosity
- Product sense
- Storytelling and communication skills

Dataists blog

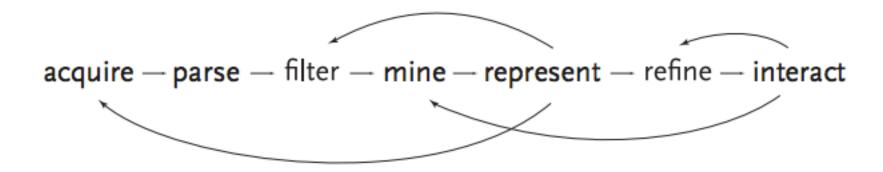
- 1. Obtain
- 2. Scrub
- 3. Explore
- 4. Model
- 5. Interpret

Jeff Hammerbacher: Chief Scientist, Cloudera

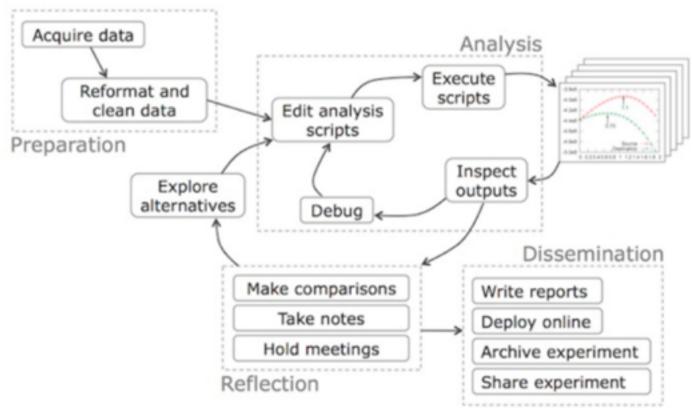
- 1. Identify problem
- 2. Instrument data sources
- 3. Collect data
- 4. Prepare data (integrate, transform, clean, impute, filter, aggregate)
- 5. Build model
- 6. Evaluate model
- 7. Communicate results

Ben Fry: Principal, Fathom





Zip Decode http://benfry.com/zipdecode/



BUILDING AN ANALYTICS TEAM

- 1. Define the top priorities of the organization
- 2. Determine the data you'd like to collect

What will your greatest challenges be?
What products could you build?
What studies could you run?
How would these influence the organization?

PROBLEM: WHAT ARE THE LEADING INDICATORS THAT A USER WILL MAKE A NEW PURCHASE?

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- 4. Extract new meaning to predict if user would purchase again
- 5. Share results (and probably also go back to the drawing board)

PROBLEM: HOW TO DEFINE "MORE ITEMS TO CONSIDER" IN AMAZON?

10 Minutes: In a small group, define the flow an Amazon Data Scientist would work through to curate the "More items to consider" list for a particular user.

III. COMPUTER SETUP

LAB. DATA WORKFLOW

INTRO TO DATA SCIENCE

DISCUSSION