## **Data Science 100**

Principles & Techniques of Data Science

Slides by:

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### Enrolling & the Waitlist

We apologize for issues with the waitlist and CalCentral ®

- > I have notified 96 students that they have a spot
- Several are having CalCentral issues
   Several are no longer interested in taking the class
- > At 11:00 AM I notified 25 additional students that they are at the top of the wait list
- > If you are not in these groups please try to visit office hours or section and let us know you are still interested.

### Questions for Today

- > Why am I excited about Data Science?
- > What is Data Science?
- > Who are we?
- > What does it mean to be a data scientists today?
- > What will I learn and how?

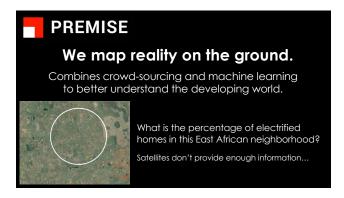
Slides from lecture available online at http://ds100.org















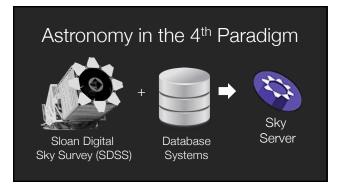


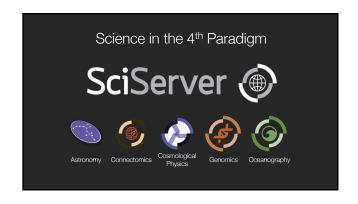






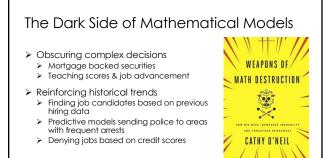












Dangers of Targeting and Personalization

| Facebook stailure did false press and polarized pola

## But ... I am optimistic

- > Knowledge is empowering
- > Data science offers **immense potential** to tackle challenging problems facing society
- > The future is in **your hands** and I believe

You will use your knowledge for good.

... I am thrilled to teach DS100!

Why are you excited about Data Science & DS100?

What are your concerns?



### Is Data Science ...

- ➤ Statistics?
  - > Yes! Use data to infer properties of the world
- ➤ Machine Learning?
  - > Yes! Use data to build algorithms that make predictions
- > Computer Science?
  - Yes! Use computational thinking and abstraction to manage complexity
- > Science, Art, and Engineering?
  - Yes! Combines the scientific method, creative thinking, and the ability to solve challenging problems ...

How can we possibly teach all of this!??



### My Definition for Data Science

The application of **data centric**, **computational**, and **inferential thinking** to

understand the world

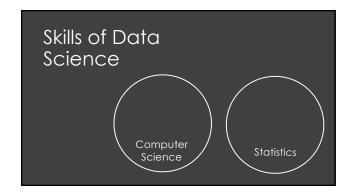
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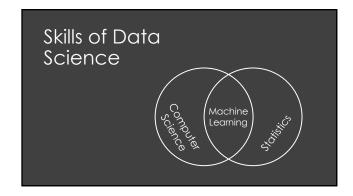
solve problems

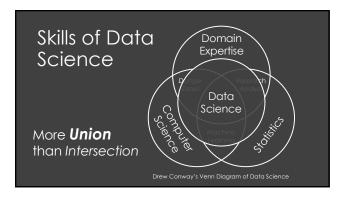
Science

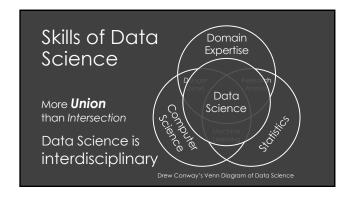
Engineering

> Data science is fundamentally interdisciplinary









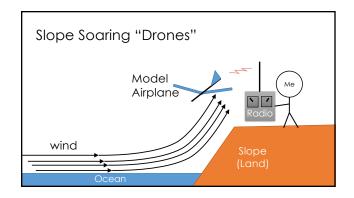


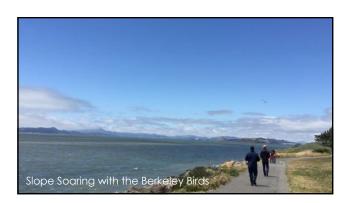
## Joey Gonzalez

Joined EECS at UC Berkeley in 2016

Research Area: Machine Learning & Data Systems

- > Study design of scalable systems for machine learning
- > Algorithms: Developed parallel algorithms for statistical inference
- Abstractions: vertex centric programming abstraction
- Systems: created the GraphLab and parts of Apache Spark
- > Co-Founder of Turi Inc.
  - > Python tools for scalable data science
  - > sold to Apple Inc. in 2016
- > Something interesting about me ...





### Joe Hellerstein

Joined UC Berkeley in 1996

Research Area: Database systems + ...

- Data-centric computer systems design
   Programming at scale: Cloud (Bloom), ML (Apache MADlib)
   Visualization: especially interactive visualizations

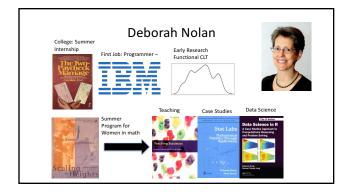
  - Foundations: systems internals & theory
- ➤ Co-Founder of Trifacta
  - > Visual data wrangling for end users
- ightharpoonup Something interesting about me ...

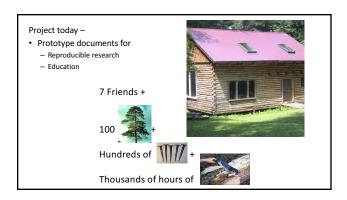


### Two Truths and a Lie

- 1. Taught Bill Gates how to scroll on a MacBook
- 2. Married Prof. Gonzalez' Ph.D. advisor
- 3. Advised President Obama on bank bailouts





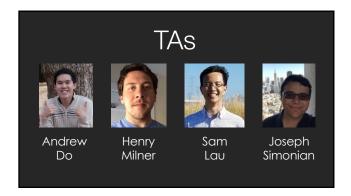


## Bin Yu, Professor of Stats and EECS

Started at Berkeley in 1993

- Research Areas: Statistics, machine learning, causal inference, collaborative research in neuroscience, genomics, precision medicine, remote sensing, ...
- Approach: Solving data problems via statistical and machine learning methods/algorithms, domain knowledge, and theory, while embedding students/postdocs in labs.
- > Other things that I love to do...







## Intermission

5 Minute Break.

Ask a neighbor:

What is your name?

Emacs or Vim or ...?

What are the ethics of data science?

Can data do harm?

What do you want to get out of DS100?

### Homework 1 will be Released Today

- > Setting up your computer for data science
  - > We want you to know how to start from scratch
- > Warmup Homework:
  - Reviewing python & numpyFun prediction exercise
- > Covered in **section** and **lab** this week
  - > try to bring your computer ...
- > Due next Tuesday at midnight (1 Week)

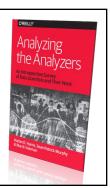
What does it mean to be a data scientist today?

How can we answer this question?



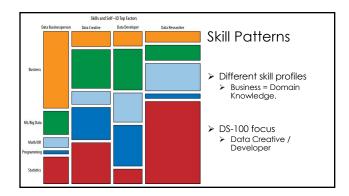
### Analyzing the Analyzers

- Surveyed 250 people in 2012
   Self reported as data scientists
  - Self reported as data scientists
     Invited at data science meetups
- > Asked to rank skills and activities
- Asked for job group:
  - Data Businessperson
  - Data Creative
  - > Data Developer
  - Data Researcher



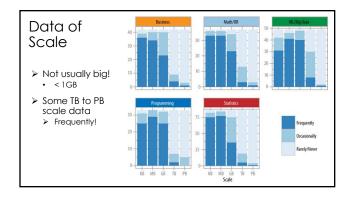
## What are the skills that best define each group?

Data (Businessperson | Creative | Developer | Researcher)



There is a lot of excitement around Big Data

... how big is the data?

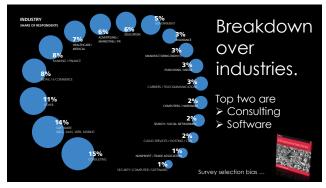


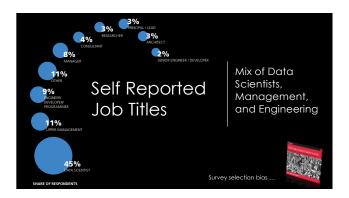
### O'REILLY Data Science Salary Survey

- > Conducted annually
- Anyone can take the surveyPromoted at O'Reilly Events
- ➤ The 2016 Survey Sample:
  - 64 Questions
  - 983 Respondents
  - 45 Countries (61% US)
- ➤ Results ...

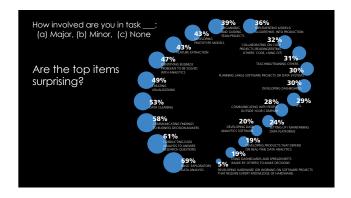


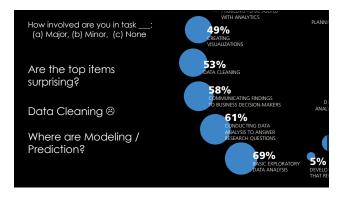


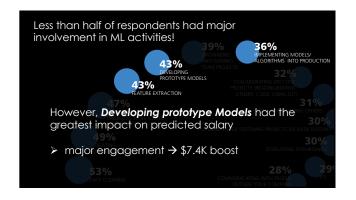




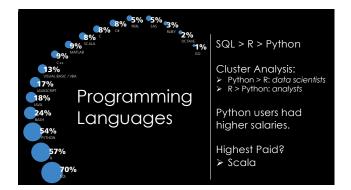


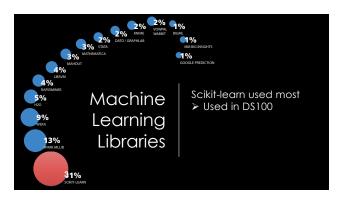












How much are they paid?



# What are your goals for D\$100?

>What do you want to learn?

➤ How does this class fit into your future plans?

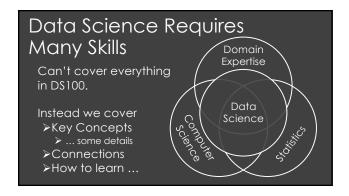
#### Our Goals

Prepare students for advanced Berkeley courses in datamanagement (CS186), machine learning (CS189), and statistics (Stat-154), by providing the necessary foundation and context

**Enable** students to start careers as data scientists by providing experience in working with <u>real</u> data, tools, and techniques.

**Empower** students to apply **computational** and **inferential thinking** to tackle real-world problems

What will I learn?



## Big Concepts in Data Science

- > Data preparation and representation
- > Efficient data processing
- > Question formulation and experimental design
- > Exploratory data analysis
- Modeling, parameter estimation, and statistical inference
- Various prediction methods: generalized linear models, decision trees, neural networks, clustering, PCA, ...
  - Overfitting, regularization, and cross validation

### Principles Computer Science in Data Science

- > Software Design & Debugging
  - > How do we develop and maintain reliable & repeatable analysis?
- > Abstraction and Algorithm Design
- How do we break big problems into small problems?
- Computational Complexity
- > How do we tradeoff time and space to compute efficiently?
- > Parallelism & Locality
- ➤ How do we divide computation across resources?
- Others?

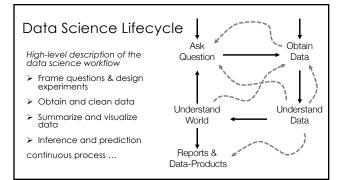
### Principles Statistics in Data Science

- ➤ Experimental Design & Sampling
  - How do we collect data to accurately answer questions?
- > Probability & Uncertainty
  - How do we quantify what we don't know?
- ➤ Modeling
  - > How do we distill the essential structure of complex phenomena?
- > Inference & Prediction
  - > How do we use the known to reason about the unknown?

## Domain Knowledge

- > What are the key questions/problems in the domain?
- ➤ What is the context of the data?
  - > What data is already available?
  - How and why was it collected?
  - What is the schema and limitations of the data?
  - How can more data be collected/obtained?
- ➤ What is the underlying process that generates the data? > causal structure, dependencies, ...

Data scientists must be inquisitive and learn new domains quickly ...



## Working with Real Data

Homework, labs, and in class examples will build on real data:

Twitter, Speeches, Scientific Data, Maps, Surveys, Images, ...

The data will be:

- > messy and you will have to clean it
- big(ish) and you will have to be a little clever to process it
- > complicated and you will have to learn about the domain

### **Using Real Tools**

- ightharpoonup Focus on Python programming language
- > We will use various different technologies Jupyter notebooks, pandas, numpy, matplotlib, SQL Server, github, Wrangler, plotly, tableau, Spark?, ...
- > We won't teach you everything ...
  - You will learn to read documentation
  - > You will learn to teach yourself
- > BETA WARNING: Things will break ...
  - You will learn how to debug
  - You will learn how to get help (Piazza)

## Reading and Reference Materials

No single great book on data science

- Lectures slides and materials will be available online
   Use online reference materials

Several eBooks we will occasionally reference (optional)

- Joel Grus. "Data Science from Scratch" [eBook Link]
- Cathy O'Neil and Rachel Schutt. "Doing Data Science [eBook Link]
- G. James, D. Witten, T. Hastie and R. Tibshirani. "An Introduction to Statistical Learning." [pdf Link]
- ➤ Wes McKinney. "Python for Data Analysis" [pdf link]





### Grades

[40%] 7 Homework: be a data-scientist

1 to 2 week long programming assignments

[13%] 13 Vitamins: don't fall behind

> Mini quizzes (1 per week of instruction)

[7%] 13 Labs: improve computing skills

➤ Completion graded

[15%] 1 Midterm: checkpoint on progress
➤ In class, healthy checkpoint

[25%] 1 Final

### On Time Policy (don't be late)

- 5 days of "slip-time" to be used on homework for unforeseen circumstances (e.g., get sick or deadline conflicts)
- After you have used your slip-time budget
   20% per day for each late day
- > If you are having trouble finishing assignments on time let us know!

## Collaboration Policy: Don't Cheat!

- > Data Science is a collaborative activity
- > You may discuss problems with friends
  - List their names at the top of your assignments
  - We may periodically analyze the collaboration networks
- > You must write your solutions individually

### **Don't Cheat**

- > Content in the homework and vitamins will be on the midterm and final
- > If you are struggling let us know so we can help!

### Staying Up to Date

- Communication will be largely through Piazza
   http://piazza.com/berkeley/spring2017/ds100
- We will also be updating the website with links to homework, lectures, and vitamins
  - http://www.ds100.org/sp17/



