# INTRO TO DATA SCIENCE LECTURE 1

NOVEMBER 19, 2014

DAT11-SF

#### INTRO TO DATA SCIENCE

# WELCOME

#### **LOGISTICS**

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Course Website: http://www.schoology.com/

**GitHub:** https://github.com/ga-students/DAT\_SF\_11

Course Times: 6:30pm-9:30pm, Mondays and Wednesdays

**Office Hours** 

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

LAB:
III. UNIX COMMAND LINE
IV. PYTHON TUTORIAL

# I. WHAT IS DATA SCIENCE?

#### WHAT IS DATA SCIENCE?

- A set of tools and techniques used to extract useful information from data.
- An interdisciplinary, problem-solving oriented subject.
- The application of scientific techniques to practical problems.
- A rapidly growing field.

#### WHAT IS DATA SCIENCE?

- Data mining
- Statistics
- Machine learning
- Information visualization
- Network analysis
- Natural language processing

- Algorithms
- Software engineering
- Databases
- Distributed systems
- Big data

## Harvard Business Review



SPOTLIGHT ON BIG DATA

## Data Scientist: The Sexiest Job Of the 21st Century

Meet the people who can coax treasure out of messy, unstructured data. by Thomas H. Davenport and D.J. Patil

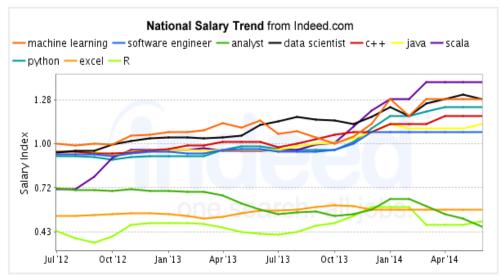
McKinsey estimates 140,000-190,000 shortage by 2018 I keep saying the sexy job in the next ten years will be statisticians. People think I'm joking, but who would've guessed that computer engineers would've been the sexy job of the 1990s?

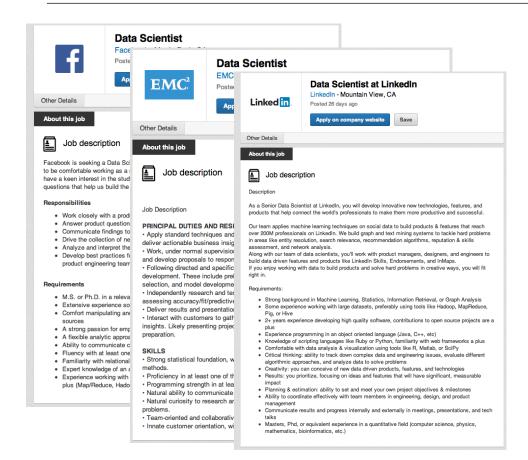
Hal Varian, Chief Economist at Google, The McKinsey Quarterly, January 2009

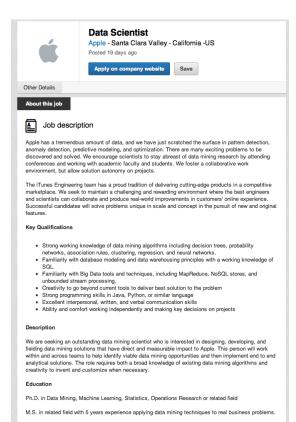
THE MOTIVATOR 10

#### Average Salary of Jobs with Titles Matching Your Search

































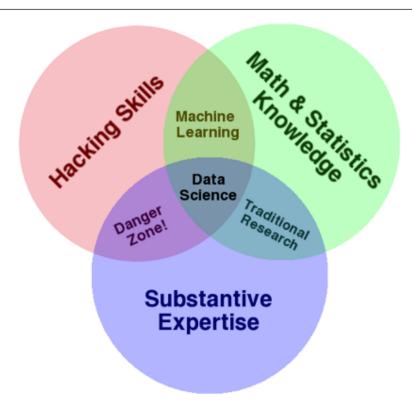




#### DATA ANALYSTS / DATA SCIENTIST

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#### THE QUALITIES OF A DATA SCIENTIST

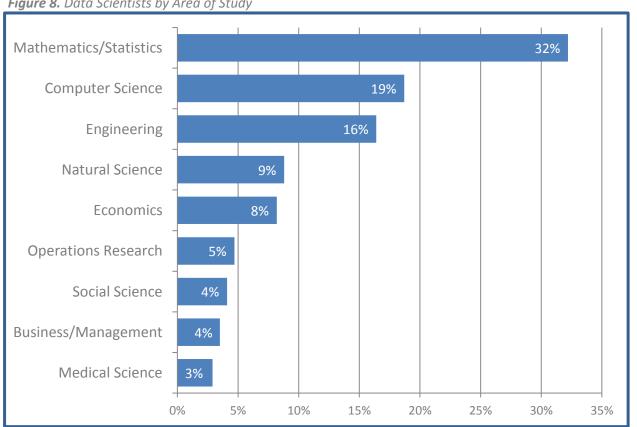


#### WHAT MAKES A GOOD DATA SCIENTIST?

- > Statistical and machine learning knowledge
- > Engineering experience
- > Academic curiosity
- Product sense
- Storytelling
- Cleverness

#### **WHO ARE DATA SCIENTISTS?**

**Figure 8.** Data Scientists by Area of Study



#### WHO ARE DATA SCIENTISTS?

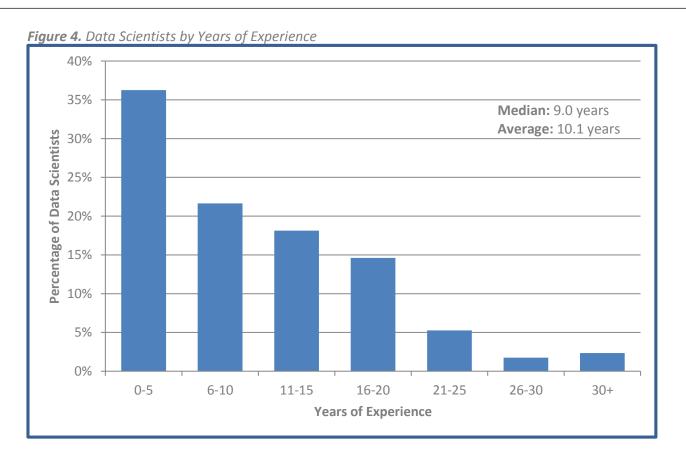
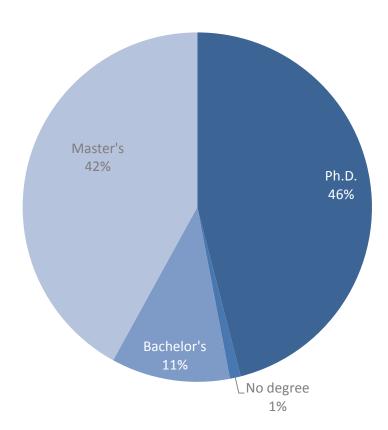
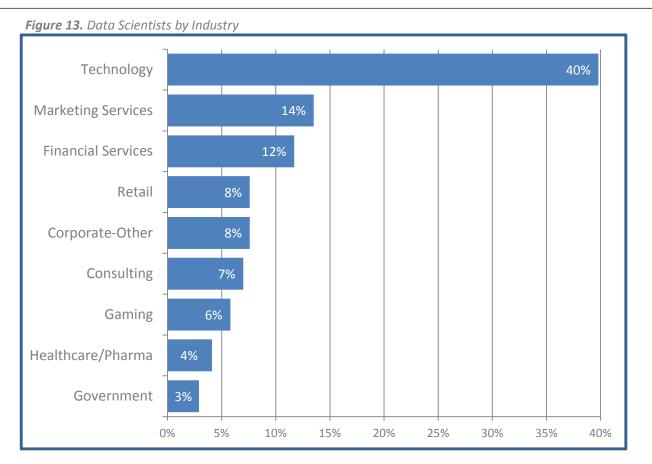
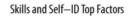
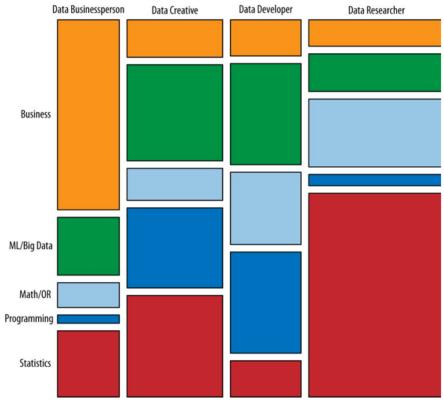


Figure 7. Data Scientists by Education









"Analyzing the Analyzers" by Harlan Harris, Sean Murphy, and Marck Vaisman , O'Reilly Strata 2012

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CURRICULUM 21

е	Date	Day	Topic	HW	Project
	Nov 19	W	Introduction to data science		
	Nov 24	M	Exploratory data analysis		
	Dec 1	M	Introduction to machine learning	1	
	Dec 3	W	Linear regression and regularization		
	Dec 8	M	Model selection and evaluation	2	
	Dec 10	W	Classification: kNN, decision trees		
	Dec 15	M	Classification: SVM	3	
	Dec 17	W	Ensemble methods: random forest		Title
			Christmas break		
	Jan 5	M	Intro to probability, naïve Bayes and logistic regression	4	Summary
	Jan 7	W	Feature engineering and selection		
	Jan 12	M	Clustering: k-means, hierarchical clustering	5	
	Jan 14	W	Dimensionality reduction: PCA and SVD		
	Jan 21	W	Text mining and information retrieval	6	
	Jan 26	M	Network analysis	7	
	Jan 28	W	Recommender systems		Proposal
	Feb 2	M	Relational databases, SQL	8	
	Feb 4	W	Big data storage and retrieval: noSQL, GraphDB		
	Feb 9	M	Big data distributed computing: map-reduce, spark rdd		
	Feb 11	W	Advanced: neural networks and deep learning		
	Feb 18	W	Guest lecture		
	Feb 23	M	Final projects presentations		Github
	Feb 25	W	Final projects presentations		

# II. THE DATA SCIENCE WORKFLOW

### Hilary Mason (bitly, HackNY, DatGotham conf, Accel)

- 1. **Obtain** pointing and clicking does not scale (APIs, Python, shell scripting)!
- 2. **Scrub** "Scrubbing data is the least sexy part of the analysis process, but often one that yields the greatest benefits" (Python, sed, awk, grep)!
- 3. Explore look at the data (visualizing, clustering, dimensionality reduction)!
- 4. **Model** "All models are wrong, but some are useful" / models are built to predict and interpret!
- 5. Interpret "The purpose of computing is insight, not numbers"

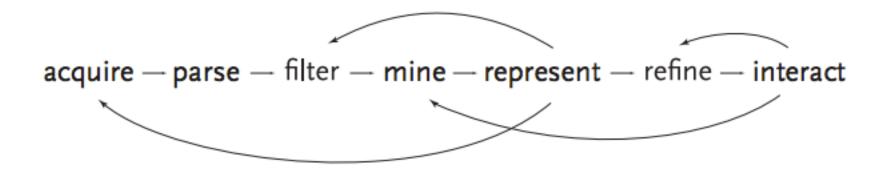
#### Jeff Hammerbacher (Facebook, Cloudera)

- ▶ 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

#### THE DATA SCIENCE WORKFLOW

### Ben Fry (author of "Visualizing Data")





## III. UNIX COMMAND LINE

#### **EXERCISE — WORKING AT THE UNIX COMMAND LINE**

#### **KEY OBJECTIVES**

- Navigate the filesystem
- Create, move, copy, and delete files & directories
- View & search files
- Edit & interact with files
- Combine steps
- Learn more

#### **TOOLS**

Files and directories:

- ls, cd, pwd
- cat, touch, mv, cp, mkdir, rm, rmdir

Manipulating data:

- head, tail, more, less, cut, paste, split
- tr, sort, uniq, wc, grep
- pipe (|), > , >>, <</pre>

Compressing data:

- compress, gzip, tar,

## IV. PYTHON TUTORIAL

EXERCISE — PYTHON 30

#### **KEY OBJECTIVES**

- Become familiar with iPython notebook
- Learn basic Python