Welcome to Data Bootcamp

Joseph Adler, Drew Conway, Jake Hofman, Hilary Mason

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@jadler, @hmason, @drewconway, @jakehofman

Joseph Adler

LinkedIn

Joseph Adler has many years of experience in data mining and data analysis at companies including Double Click, American Express, and Verßign. He graduated from MIT with an B.Sc. and M.Eng in Computer Science and Electrical Engineering. He is the inventor of several patents for computer security and cryptography, and the author of "Baseball Hacks" and "R in a Mutshell". Currently, he is a senior data scientist at Linkedin.



Hilary Mason

bit.lv

Hilary is the lead scientist at bit.ly, where she is finding sense in vast data sets. She is a former computer science professor with a background in machine learning and data mining, has published numerous academic papers, and regularly releases code on her personal site, www.hilarymason.com. She has discovered two new species, loves to bake cookies, and asks way too many questions.



Web site

Drew Conway

New York University

Drew Conway is a PhD student in political science at New York University. Drew studies terrorism and armed conflict; using tools from mathematics and computer science to gain a deeper understanding of these phenomena.



▶ Web site

Jake Hofman

Yahoo!

Jake Hofman is a member of the Human Social Dynamics group at Yahoo! Research. His work involves data-driven modeling of social data, focusing on applications of machine learning and statistical inference to large-cacel data. He holds a B.S. in Electrical Engineering from Boston University and a Ph.D. in Physics from Columbia University.



Please do try this at home

All of the materials from today's tutorial are available on Github:

```
Clone the repository for data/code/slides
```

git clone https://github.com/drewconway/strata_bootcamp.git

Disclaimer

You may be bored if you already know how to ...

- Acquire data from APIs
- ► Clean/explore/visualize data
- Classify and cluster image and text data
- Work with large data sets
- Build simple mashups
- Work with Python, R, SciPy/NumPy, etc.
- Work with unix tools on the command line, e.g.
 - \$ sed -e 's/<[^>]*>//g' < page.html > page.txt

Data-dependent products



○ Not Interested

Our best guess for Jake: 5 stars

Average of 4 Z/15 920 ratings: 3.8 stars

The Big Lebowski

1998 R 117 minutes

Slacker Jeff "The Dude" Lebowski (Jeff Bridges) gets involved in a gargantuan mess of events when he's mistaken for another man named Lebowski, whose wife has been kidnapped and is being held for ransom. All the while, Dude's friend, Walter (John Goodman), sitis the pot. Brothers Joel Coen and Ethan Coen write and direct this cult comedy classic that also stars Steve Buscemi, Philip Seymour Hoffman, Julianne Moore and John Turturro.

> Cast: Jeff Bridges, John Goodman, Philip Seymour Hoffman, Steve Buscerni, Julianne Moore, Tara Reid, Peter Stormare, David Huddleston, Philip Moor, Mark Pellegrino, Flea, Torsten Voges, Jimmie Dale Gillmore, Jack Kehler, John Turturno, James G. Hoosler, Richard Gant, Christian Clemenson, David Thewlis, Peter Siragusa, Sam Elliott, Ben Gazzara, Jon Polito, Asia Carrera, Paris Themmen

Director: Joel Coen

Genres: Comedy, Cult Comedies, Universal Studios Home

Entertainment, Blu-ray

This movie is: Quirky, Witty

Format: DVD and streaming (Blu-ray availability date unknown) (HD available)

Play

Add to Instant Queue

Add to DVD Queue

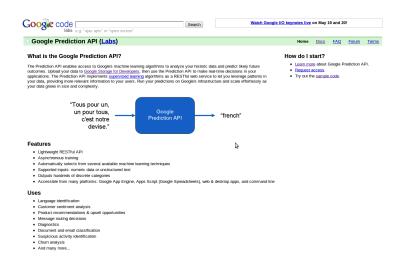
Play Trailer

Recommended based on your interest in: Fargo, O Brother, Where Art Thou? and No Country for Old Men

Data-dependent products

- Effective/practical systems that learn from experience impact our daily lives, e.g.:
 - Recommendation systems
 - Spam detection
 - Optical character recognition
 - Face recognition
 - Fraud detection
 - ► Machine translation
 - ▶ ...

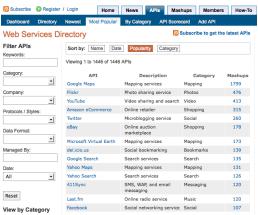
Black¹-boxified?



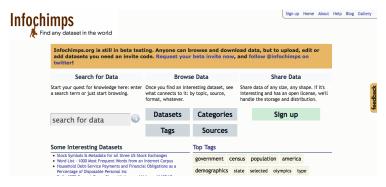
 $^{^{1}}$ s/black/blue/g

Web service APIs expose lots of data





Many free, public data sets available online



Roadmap?

Step 1: Have data

Step 2: ???

Step 3: Profit

Learning by example

Fwd: Yahoo! supercomputing cluster RFP - i have no idea. i have no idea. O non urgent - whoops! yes that's what i meant, thanks for decoding my questi-SourceForge.net: variational bayes for network modularity - can i get admin Byline - iPhone Apps, iPhone 3G apps and iPod touch Applications Gallery a Laurence J. Peter: Facts are stubborn things, but statistics are more pliable. Re: JAFOS 2008, Applied Math Session - yes. the listening post dude. On N Access to over 5,000 Health Plan Choices! - Affordable health insurance. Ins More effective - If you are having trouble viewing this email click here. Thurs Special Offer! Cialis, Viagra, VicodinES! - Order all your Favorite Rx~Medica Financial Aid Available: Find Funding for Your Education - Get the financial a Find The Perfect School and Financal Aid for your College Degree - HI! It has **PHARMA viagra PHARMA cialis** - Wanted: web store with remedies. N

Learning by example

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- ► How did you solve this problem?
- Can you make this process explicit (e.g. write code to do so)?

Learning by example



We learn quickly from few, relatively unstructured examples ... but we don't understand how we accomplish this

Everything old is new again²

- Many fields ...
 - Statistics
 - ▶ Pattern recognition
 - ► Data mining
 - Machine learning
- ... similar goals
 - Extract and recognize patterns in data
 - Interpret or explain observations
 - Test validity of hypotheses
 - Efficiently search the space of hypotheses
 - Design efficient algorithms enabling machines to learn from data

²http://cbcl.mit.edu/publications/theses/thesis-rifkin.pdf

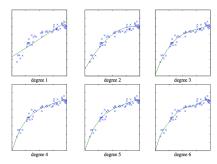
Statistics vs. machine learning³

Glossary				
Machine learning	Statistics			
network, graphs	model			
weights	parameters			
learning	fitting			
generalization	test set performance			
supervised learning	regression/classification			
unsupervised learning	density estimation, clustering			
$large\ grant = \$1,000,000$	large grant= \$50,000			
nice place to have a meeting:	nice place to have a meeting:			
Snowbird, Utah, French Alps	Las Vegas in August			

³http://anyall.org/blog/2008/12/statistics-vs-machine-learning-fight/

Philosophy

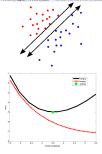
- We would like models that:
 - Provide predictive and explanatory power
 - Are complex enough to describe observed phenomena
 - ▶ Are simple enough to generalize to future observations



Roadmap, take 24

- 1. Get data
- 2. Visualize/perform sanity checks
- 3. Clean/filter observations
- 4. Choose features to represent data
- 5. Specify model
- 6. Specify loss function
- 7. Develop algorithm to minimize loss
- 8. Choose performance measure
- 9. "Train" to minimize loss
- 10. "Test" to evaluate generalization

Feet Month approximating dates (PRI - 1 have in life.) I have no life. I have no life. I have no life. I have no life. I have have a life. I have have a former have been decoding my case Source-From and very life. I have a first first first first first first have a little form thing, but statistics we may life. I have a first first first have little first first, but statistics we may place. I have a first first first first we little first first, but statistics we may place. Moreover, a first f



⁴http://www.dataists.com/2010/09/a-taxonomy-of-data-science/



How does Cinematch do it?

Straightforward statistical linear models with a lot of data conditioning. But a real-world system is much more than an algorithm, and Cinematch does a lot more than just optimize for RMSE. After all, we have a website to support. In production we have to worry about system scaling and performance, and we have additional sources to data we can use to guide our recommendations. But, as mentioned in the Rules and just to be perfectly clear, for the purposes of the Prize the RMSE values we report here do not use any of this extra data.

Shipping = Feature

Add an asymmetric frequency feature $\mathbf{y}_{j,f_{ut}}^{(3)}$: SBRAMF-UTB-UTF-MTF-AFF

$$\widehat{\mathbf{r}_{uit}} = \mu_i + \mu_u + \mu_{u,t} + \mu_{i,bin(t)} + \left(\mathbf{p}_i^{(1)} + \mathbf{p}_{i,bin(t)}^{(2)} + \mathbf{p}_{i,bin(t)}^{(3)} + \mathbf{p}_{i,f_{ui}}^{(3)}\right)^T \left(\mathbf{q}_u^{(1)} + \mathbf{q}_{u,t}^{(2)} + \frac{1}{\sqrt{|N(u)|}} \sum_{j \in N(u)} \left(\mathbf{y}_j^{(1)} + \mathbf{y}_{j,bin(t)}^{(3)} + \mathbf{y}_{j,f_{ui}}^{(3)}\right)\right)$$
(34)

Model extension (+)	epoch time	#epochs	probeRMSE, $k = 50$ features
SBRMF - SVD with biases	17[s]	69	0.9054
SBRAMF - asymmetric part	50[s]	30	0.8974
+UTB - user time bias	61[s]	50	0.8919
+UTF - user time feature	62[s]	38	0.8911
+MTF - movie time feature	74[s]	37	0.8908
+ATF - asymmetric time feature	74[s]	44	0.8905
+MFF - movie frequency feature	149[s]	46	0.8900
+AFF - asymmetric frequency feature	206[s]	45	0.8886 (0.8846 with k = 1000)



Data jeopardy

Regardless of scale, it's difficult to find the right questions to ask of the data

Data hacking

Cleaning and normalizing data is a substantial amount of the work (and likely impacts results)

Data hacking

The ability to iterate quickly, asking and answering many questions, is crucial

Data hacking

Hacks happen: sed/awk/grep are useful, and scale

Data "science"

Simple methods (e.g., linear models) work surprisingly well, especially with lots of data

Data "science"

It's easy to cover your tracks—things are often much more complicated than they appear

References

