Resources, Capabilities, and Strategies for Data Science Learners



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This presentation is available at (.odp, .pptx, .pdf): https://github.com/wsphd/datajam2018

Overview

- Thinking as a Data Scientist
 - Positioning Yourself in the Contemporary Paradigm
- Learning as a Data Scientist
 - What the Best Autodidacts (self-learners) Know
- Acting as a Data Scientist
 - Earning Career Success

- Q&A
- Give away some goodies...

Motivating Question

```
Are you a data science learner?
  ...quantitative reasoning...
          ...big data...
   ...statistical computing...
        ...applied math...
   ...business intelligence...
    ...predictive analytics...
...decision support modeling...
     ...machine learning...
   ...artificial intelligence...
```

Information Dynamics

- Wisdom
 - Extraordinary Insight (Explanation) for Foresight (Prediction)
 - Restaurant: How should our menu change in the future to best optimize nightly sales?
- Knowledge
 - Combination of Explicit Information and Tacit Information
 - Restaurant: What action led to the change in last night's sales?
- Information
 - Meaningful Data
 - Restaurant: How does last night's sales compare to that night the previous year? How does last night's sales compare to our goals?
- Data
 - Raw, atomic, basic
 - Restaurant: What were the total sales for last night?

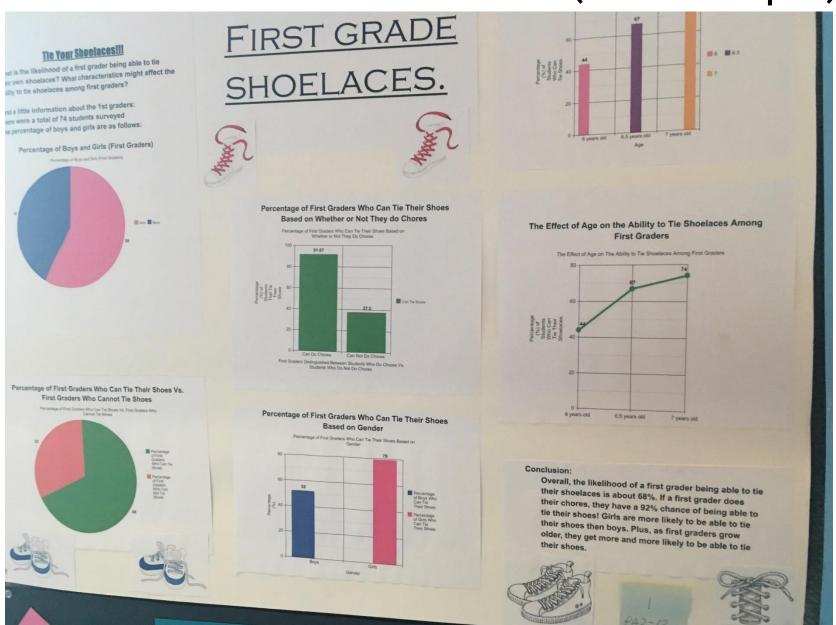
Analytics for Decision-making

- Prescriptive Analytics
 - What should we do?
 - HR Department: What should we (the HR Department) do to meet or exceed the organization's hiring and retention goals for next year? What data/information/knowledge/wisdom should we provide to our hiring and technical managers to help? What are we missing?
- Predictive Analytics
 - What is likely to happen?
 - HR Department: How many new employees will our organization need next year?
 How will the mix change? What is our competition likely to do?
- Diagnostic Analytics
 - Why did it happen?
 - HR Department: Did our emphasis on recruiting from campus A (over campus B, etc.) matter? What do the managers of these entry-level employees think?
- Descriptive Analytics
 - What happened?
 - HR Department: How many entry-level professionals did we hire last year? How many of them are still with us now?

Common Core (e.g., a graph in Grade 3)

ASA ID	award Category	Title	LocalID	City
4-3550	1 GR10-12	Charting Chart Toppers	4-3550	Andover
MI4-194	2 GR10-12	Influenza 2018	MI4-194	
PA 4-75	3 GR10-12	The Pen is Mightier than the Mouse	PA 4- 75	Glen Mills
CT10-12D13	HM GR10-12	The Cost of Milk Production	CT10-12D	13 Higganum
PA 2-12	1 GR4-6	First Grade Shoelaces	PA 2- 12	Abington
2-3654	2 GR4-6	Screen Time	2-3654	Appleton
DC 2-3718	3 GR4-6	Sugary Soda Showdown	DC 2-3718	- Grannbia
PA 2-61	HM GR4-6	DuckDuckRabbit? Pressure for Good Grades: Does It Exist?	PA 2- 61 OH 7-9 10	Abington
OH 7-9 10	1 GR7-9	Do High School Students Perform Acts of Kindness?	3-3712	Lyndhurst West Nyack
3-3712	2 GR7-9	Starbucks Consumption = mmm!	DC 3-3729	Williamsburg
DC 3-3729	3 GR7-9 HM GR7-9	What's Up with Binge Watching?	MI3-121	
MI3-121	1 GRK-3	Which Hand Rules	1-3635	Bar Harbor
1-3635	2 GRK-3	Are You a Square or a Rectangle Do You Know that Orangutans Are Endangered?		Chagrin Falls (
OH K-3 9	3 GRK-3	Speedcubing Statistics	DC 1-3939	Plano 7 Fairfax V
1-3552 DC 1-393	9 HM GRK-3	Speedcubing States		Tantax V
DC 1-00				
OTH	×			
ОН	×			
PA	×			
OTH	×			

"First-Grade Shoelaces" (Grades 4-6)



"First-Grade Shoelaces" (Grades 4-6) (results)

Conclusion:

Overall, the likelihood of a first grader being able to tie their shoelaces is about 68%. If a first grader does their chores, they have a 92% chance of being able to tie their shoes! Girls are more likely to be able to tie their shoes then boys. Plus, as first graders grow older, they get more and more likely to be able to tie their shoes.

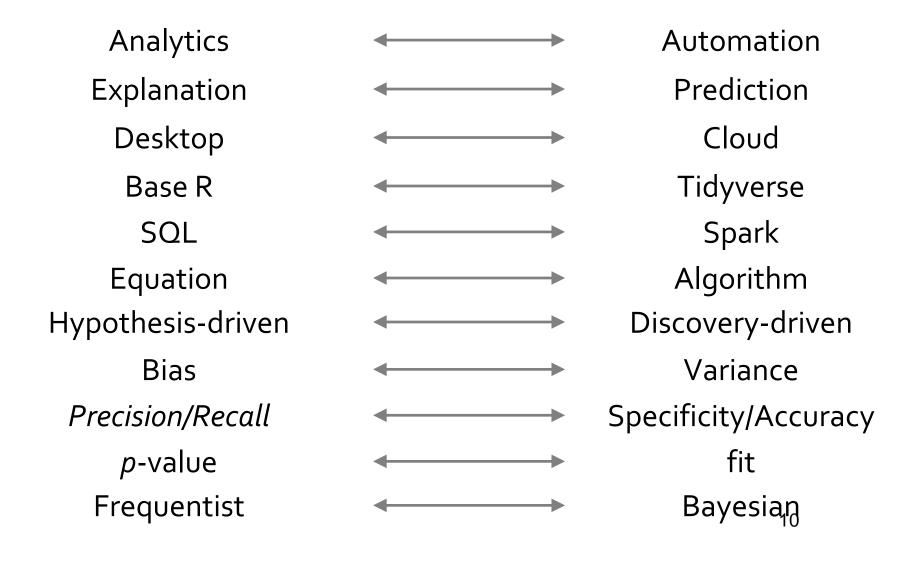


Next-Generation Science Standards

There are *seven* "cross-cutting" skills to support *four* core subject domains: Physical Science, Life Science, Earth and Space Science, and Engineering.

	Cross-cutting skill	Data Science analogue	
1	Patterns	Observations, Variables, and Pattern Matching	
2	Cause and Effect	Causality	
3	Scale, Proportion, and Quantity	Measure Theory	
4	Systems and System Models	Model Building	
5	Energy and Matter	Systems Dynamics, Moments, and Entropy	
6	Structure and Function	Hierarchical, Structural, and Latent Variable Analysis	
7	Stability and Change	Variation, Complexity, and Interactions	

Persistent Tensions



What is Data Science?

- Data Science is "the science of planning for, acquisition, management, analysis of, and inference from data."
- Data science is case-based, hands-on, and interdisciplinary.
- Building upon <u>experimentation</u>, <u>modeling</u>, and <u>computation</u>, there are some that believe that <u>data science</u> is, in fact, a new, type of scientific discovery.

- Data scientists demonstrate mastery of skills and concepts, including many traditionally associated with the fields of <u>Mathematics</u>, <u>Statistics</u>, and <u>Computer Science</u> and. Data Science blends much of the pedagogical content from all three disciplines, but it is neither the simple intersection, nor the superset of the three.
- There is a fourth area of demonstrated mastery too: <u>Subject-Domain</u> expertise. ¹¹

Math/Stat/CompSci/Subject Domain

What can we leverage well for data science success?

Fvidence-...Operational ...Truth in ...Approach ...Language Based... Scale Action **Parametric** Algorithms, Various— Equations, Literate **Professional** tests, **Practice Theorems** Regressions Standards Programming Computa-Various— General Probability, tional Theory Research-**Matrices** Linear Model **Thinking** based Subject Computer Discipline Math **Statistics** Matter Science (Domain)

Practical Applications III: Deep (Machine) Learning

Mt. Sinai Hospital (NY) 2015 Research Program: "Deep Patient"

- 1. Tested on 700,000 patient records

 Able to predict disease far better than traditional methods
- 2. Better than humans at predicting onset of schizophrenia

 Not even physicians can accurately predict that psychiatric disorder
- 3. Algorithm was able to detect a pattern never before discovered Not only is pattern latent, so is its detection method ("black-box")

Balance Breadth & Depth (Software Tools)

OLS Regression

lm, glm

Statsmodels (sm.OLS)

Juliastats, GLM, linreg

t-test for ind.

t.test, wilcox.test, MASS

numpy, scipy(stats)

Juliastats (pvalue, confint)

Data Visualization

Base R, lattice, ggplot2

pandas, matplotlib Plots, Gadfly, JuliaGraphs

Software

R

Python

Julia

Software Tools (cont.)

- Development Environments
 - R (RStudio)
 - Python (Spyder, PyCharm)
 - Julia (JUNO, Atom)
 - Jupyter Notebooks (for all of the above)
- Text editors (e.g., Notebook++, VIM, or just plain Notepad)
- Weave documentation, code, and output together to make dynamic documents (e.g., knitr, sweave)
- Generate publication-quality, reproducible results (e.g., markdown, LaTex)
- Many, many other software (and many packages for each)
 - FOSS: STAN, QGIS, Gephi, Gretl, Scala, Apache Spark, KNIME...
 - COTS: SPSS, SAS, Stata, Matlab, Nvivo, Gurobi, Tropes, ArcGIS...

Balance Breadth & Depth (Analytical Techniques)

- More general?
 - GIS
 - effect sizes
 - logistic regression, k-means clustering, classification
 - supervised learning/resampling
 - LP/IP/MIP optimization
- More nichey?
 - spatial analytics
 - penalty-based regression
 - neural networks
 - unsupervised learning/topic modeling
 - convex optimization
- Text analysis, network analysis, genomic analysis, Bayesian analysis
- CranViews
 - https://cran.r-project.org/web/views/

Reproducibility ("academic perspective")

Results reproducible?

assumptions, interpretations, platforms, equivocality

Code available?

versioning, documentation, dialects, code "rot"

Data available?

curation, archival, funding fiats, data format "rot"

Paper (public)?

access scope, licensing/rights, journal rules, tenure

Paper (traditional)?

keep our own drafts, updates, data, and code

less

more

Workflow ("professional perspective") How does Disney do it? Personas: Business

- Information Worker
 - Excel, Powerpoint
 - Prepared BI reports
 - Light Statistics
- Business Analyst
 - Excel, Powerpoint
 - COTS Reporting tool
 - Light Statistics
- Data Analyst
 - Excel, Powerpoint
 - COTS Reporting tool
 - o SQL

Workflow ("How does Disney do it?") Personas: Data Scientist

Data Scientist

- o Required: Python & SQL; Nice to have: Java, Scala
- Machine Learning, Statistics, Deep Learning
- Data wrangling skills
- Distributed systems & algorithms
- Data Sampling, approximate aggregations, extrapolation
- Scientific Method Notebooks
- Data communication, visualization
- Cloud services, Linux CLI
- Bonus: NLP, image recognition

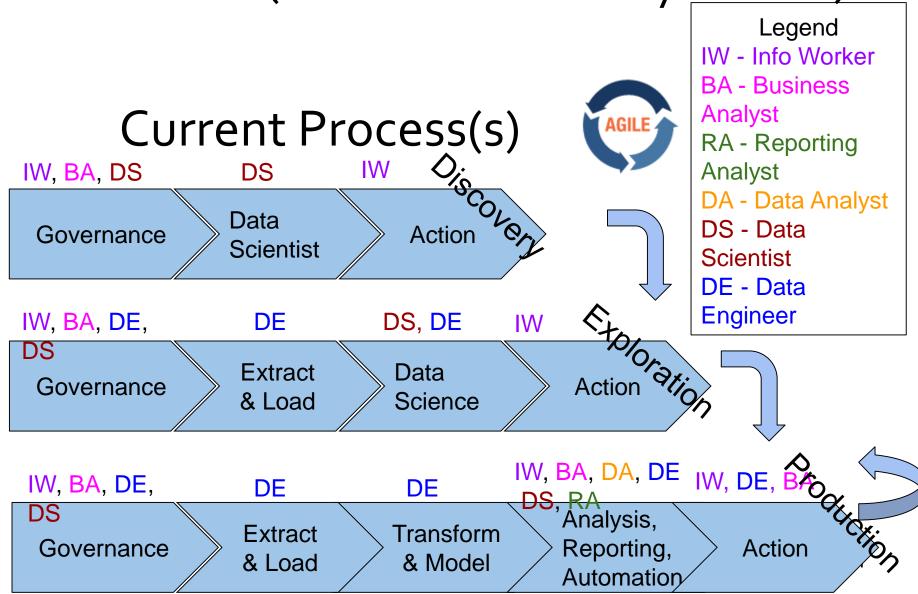
Workflow ("How does Disney do it?") Personas: Technical

- Data Engineer
 - Distributed Systems, Stream Processing
 - Tools, Infrastructure,
 Frameworks, Services
 - Java, Scala, SQL, Python, R, Bash/Zsh
 - o Linux, Git, DevOps, Cloud
 - Medium Stats
 - Medium ML/DL
 - Hadoop, Yarn, HDFS,
 ElasticSearch
- Reporting Programmer

Analyst

COTS Reporting tool SQL

Workflow ("How does Disney do it?")



References, more...

- Twitter
 - #rstats
- Podcasts
 - https://www.analyticsvidhya.com/blog/2018/01/10-data-science-machinelearning-ai-podcasts-must-listen/
- Tutorials
 - http://tutorials.iq.harvard.edu/R/Rintro/Rintro.html
 - <u>lynda.csun.edu</u>
- Niche-specific (e.g., "Psycho")
 - Blog (<u>https://neuropsychology.github.io/psycho.R/</u>)
 - Papers (http://joss.theoj.org/papers/10.21105/joss.00470)
 - Packages (<u>https://github.com/neuropsychology/psycho.R</u>)
- More references:
 - https://smithw.org/datajam



Thursday, September 6, 2018, 7:00 PM

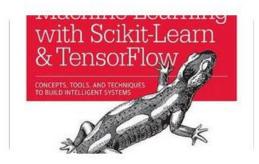
Book Club: Hands-On Machine Learning with Scikit-Learn and TensorFlow



Hosted by Cassie Borish

Join our biweekly meetup as we work through Hands-On Machine Learning with Scikit-Learn & TensorFlow (https://amzn.to/2jpUoiN). Jupyter notebooks accompanying the text can be found here: https://github.com/ageron/handson-ml ***** Tonight we are discussing and working through exercises from Chapter 8: Dimensionality Reduction. ***** Haven't done the previous readings? That's ok! Summaries of previous chapters are posted here:





Attend

- Zappbuddy Technologies
 25000 Avenue Stanford · Santa Clarita, CA
- 7 comments

What we're about

This meetup is designed to create a meeting point in Santa Clarita for those

Upcoming Meetups

See all





Past Meetup

Join Persian Women In Tech LA for our January 2018

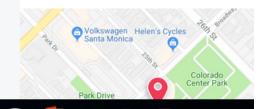


Hosted by Kevin Mehrabi and Elno From Deep Learning | Los Angeles Public group 2

This Meetup is past 12 people went



- Thursday, January 25, 2018 6:30 PM to 9:00 PM
 - Hulu Santa Monica HQ 2500 Broadway Street Suite 200 · Santa Monica, ca



Career Homework

- TechFest (CSUN Career Center)
 - Marketed mostly to Eng/CompSci, but a data science learner goes...
- Career Fairs (various) (CSUN Career Center and off-campus)
 - Also, if a display sign doesn't list "data science" (or whatever) on the sign, you ask about it...
- Indeed.com
 - "data science intern", "data science entry level"
 - E.g., 25 miles from CSUN
 - Do your homework from at least 20 job descriptions
 - What skills do I have? What skills don't I have but I can get? What skills don't I have that I don't even know what the skill is?

25

Cloud Services

- Free-tier
 - IBM Watson Cloud (you want the "no time restrictions" option)
 - https://console.bluemix.net/registration/free/
 - Amazon Web Services (you want the "non-expiring" offer)
 - https://aws.amazon.com/free/
 - Google Cloud (you want the "always free" option)
 - https://cloud.google.com/free/
 - Microsoft Azure (you want the "start free" option)
 - https://azure.microsoft.com/en-us/free/
- CSUN too (<u>https://www.csun.edu/it/ibm-cloud-services-csun</u>)
- Amazon certificate (https://laedc.org/2018/08/09/amazon-los-angeles-colleges-cloud/)
 - SMC (http://www.smc.edu/NewsRoom/Pages/Cloud-Computing-Certificate.aspx)
- But keep in mind threats to both reproducibility and workflow

Canvas API Example

```
# do once
install.packages( "devtools" )
library( devtools )
install_github( "daranzolin/rcanvas" )
library( rcanvas )
set_canvas_token( "your-40-character-token-from-Account-Settings-here" )
set canvas domain( "https://canvas.csun.edu" )
# get course items
get user items( course id = 12345, item = "assignments" )
get user items( course id = 12345, item = "missing submissions" )
# get course analytics
get user items( course id = 12345, item = "activity" )
# upload a file
                                                                        27
upload_course_file( course_id = 12345, file_name = "testfile.pdf" )
```

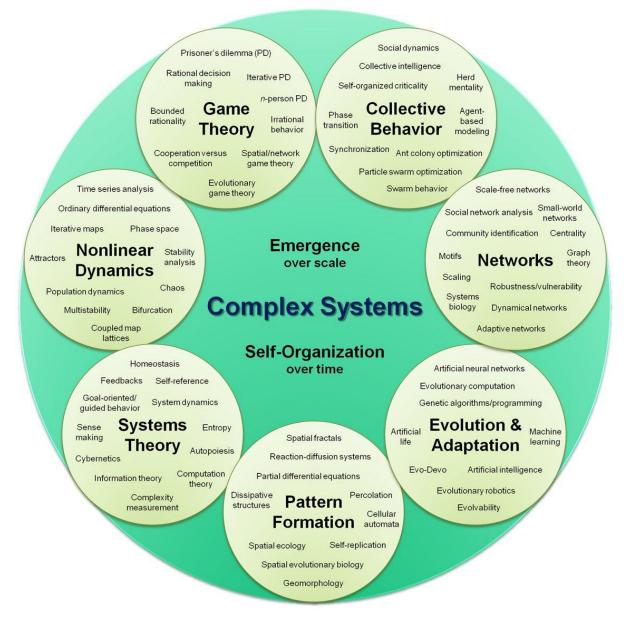
Some Examples

- LA Crime
 - https://datascienceplus.com/analysis-of-los-angeles-crimewith-r/
- LA Homelessness
 - https://source.opennews.org/articles/how-we-found-newpatterns-la-homeless-arrest/
- LA Reduce Vehicle v. Pedestrian Deaths
 - http://visionzero.lacity.org/wpcontent/uploads/2015/08/VisionZero LosAngeles.pdf
- You can do this yourself:
 - LA City (https://data.lacity.org/)
 - LA County (<u>https://data.lacounty.gov/</u>)
 - CA (https://data.ca.gov/)
 - US (https://www.data.gov/)
 - Data Science for the Social Good (https://dssg.uchicago.edu/projects/)

Counterpoint

- Hernandez, D., and Greenwald T. (August 11, 2018), "IBM Has a Dilemma", Wall Street Journal.
- Muller, J. (2018), *The Tyranny of Metrics*, Princeton University Press.
- O'Neill, C. (2017), Weapons of Math Destruction: How Big Data Increases Inequity and Threatens Democracy, Broadway Books.
- Pearl, J. (2018), The Book of Why: The New Science of Cause and Effect, Basic Books.
- Tenner, E. (2018), *The Efficiency Paradox: What Big Data Can't Do*, Alfred A. Knopf.

Macro-level (Complexity)



Micro-level (A New Language for conversations)

- Student peers
- Professional contacts

- Most important in the short-run—Professors
 - What kinds of research questions have you worked on?
 - What kinds of data have you used?
 - What kinds of analytical methods have you used?
 - What kinds of software tools have you used?
 - What would you like to learn in the near future?
 - How do you learn new things related to using data?

fin

• Again: Are you a data science learner?

• Questions?

Goodies

References

- Automation vs. Analytics
 - Davenport, T. (2009), "Make Better Decisions", *Harvard Business Review*, Nov. 87(11), p. 117-123.
- Curriculum Guidelines for Undergraduate Programs in Data Science
 - https://www.stat.berkeley.edu/~nolan/Papers/Data.Sci ence.Guidelines.16.9.25.pdf
- Complex Systems chart
 - By Hiroki Sayama, D.Sc. Created by Hiroki Sayama, D.Sc., Collective Dynamics of Complex Systems (CoCo) Research Group at Binghamton University, State University of New York, CC BY-SA 3.0, https://commons.wikimedia.org/w/index.php?curid=12191267