

Data Science Meets Open Science

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> Open Science Day EPFL 50th Anniversary 18 October 2019









privacy and ethical concerns throughout

What is Data Science?

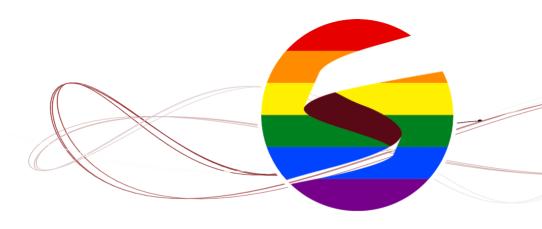
Definition: Data science is the study of extracting value from data.

Outline

- The Good, Bad, and Better News
- Technical and Non-Technical Challenges
- A Big Vision: Academic Cloud

The Good News

Open source software systems are indispensable to practicing data scientists, teaching data science, and researchers



Stan

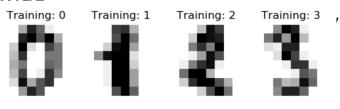
- Andrew Gelman, Columbia University
- Statistical modeling and analysis platform
- Probabilistic programming language
- Bayesian inference, MCMC built-in, Rbased
- Large and diverse user community worldwide academia, government, industry

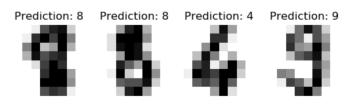
$$\log p(y) = -\frac{y^2}{2} - \log Z$$

leaun

```
images_and_labels = list(zip(digits.images, digits.target))
for index, (image, label) in enumerate(images_and_labels[:4]):
   plt.subplot(2, 4, index + 1)
   plt.axis('off')
   plt.imshow(image, cmap=plt.cm.gray_r, interpolation='nearest')
   plt.title('Training: %i' % label)
# To apply a classifier on this data, we need to flatten the image, to
# turn the data in a (samples, feature) matrix:
n_samples = len(digits.images)
data = digits.images.reshape((n_samples, -1))
# Create a classifier: a support vector classifier
classifier = svm.SVC(gamma=0.001)
# We learn the digits on the first half of the digits
classifier.fit(data[:n_samples // 2], digits.target[:n_samples // 2])
# Now predict the value of the digit on the second half:
expected = digits.target[n_samples // 2:]
predicted = classifier.predict(data[n samples // 2:])
```

- Andreas Mueller, Nicholas Hug, Thomas Fan, (core contributors), Columbia University
- Machine learning
- NumPy, SciPy, matplotlib
- Large and diverse user community worldwide

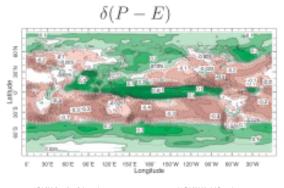


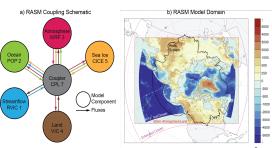


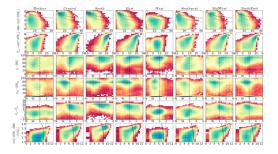


PANGEO

- Ryan Abernathy, Columbia University
- Platform for sharing data, code, and models for geoscience
- Jupyter-in-the-cloud, Python ecosystem
- Applications in meteorology, hydrology, oceanography, climate modeling









The Bad News

Much of the big datasets are locked up in companies.

Government funding agencies don't pay for our unsung heroes and their work.









Alibaba.com





Data is Locked Up

- Data is locked up for good reason
 - Privacy of customers
 - Data is company asset, and accrues to its bottom-line
- Consequences for science
 - Industry is ahead of academia, in some areas of science



- Academia can/should work on problems industry can't/won't
- Academics work inside company, leading to new models of academic-industry relations

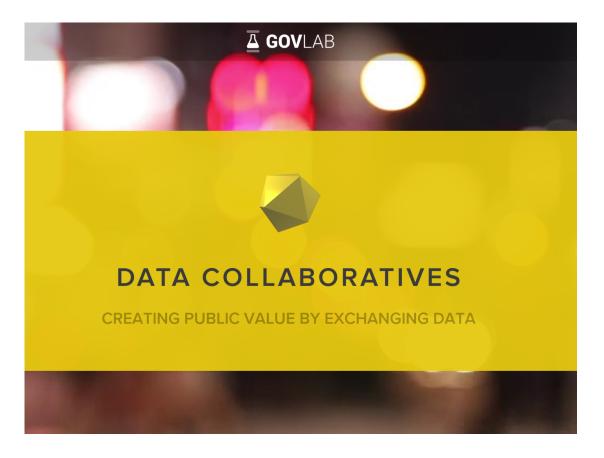
Open Source Developers

- Most government funding agencies in the US do not support software engineers
- Academia does not treat them as equal to tenure-track faculty

The Better News

New efforts support data sharing.

New funding sources and culture change for software developers.



Private companies (as well as government, non-profits, and academia) exchange data to create public value.

datacollaboratives.org



dataCommons.org

GNI Home

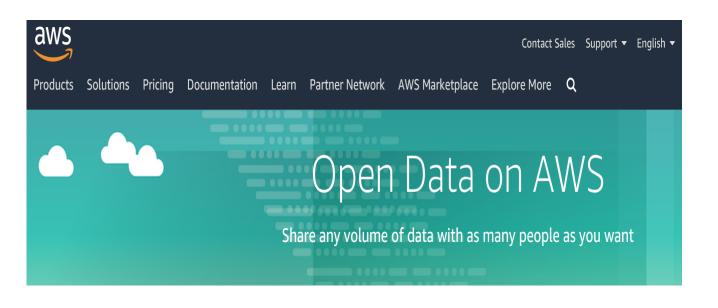
Browser

Documentation

Fact Checks

Welcome to dataCommons

Publicly available data from open sources (i.e. census.gov, NOAA, data.gov etc) are a vital resource for students and researchers in a variety of disciplines. Unfortunately, processing these datasets is often tedious and cumbersome.



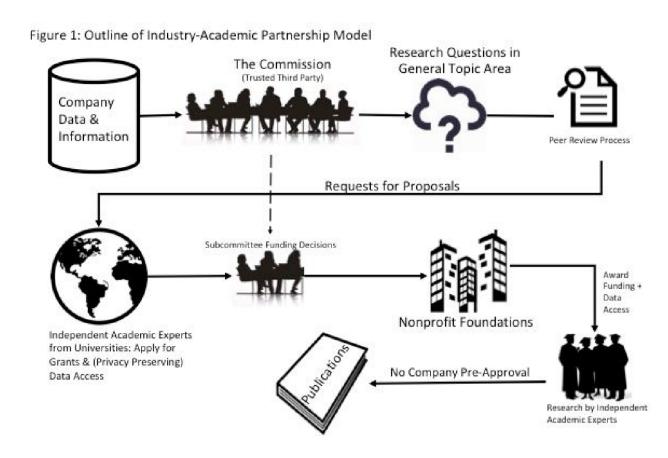












Gary King and Nathaniel Persily. Working Paper. "A New Model for Industry-Academic Partnerships". Copy at http://i.mp/2q1IQpH

Open Source Developers

- Foundations to the rescue!
 - Chan-Zuckerberg, Moore, Sloan, Schmidt, ...
- Forward-looking universities are developing new tracks (or reusing existing ones) to be equivalent to the tenuretrack
 - Carnegie Mellon: "systems science"
 - Columbia: "applied data scientist"

Challenges

What should the community focus on next?

What the Community Can Work on Together

- Non-technical
 - Define a "Data IRB": Institutional Review Board (IRB) process for data
 - Explore new models of academia-industry engagement
 - Create a culture change at universities to acknowledge contributions of open source developers, applied data scientists, etc.

Technical

- Make finding and publishing datasets as easy as using the Internet/ web/browser
- Explore "share back" model, to give back control of data to individual.
 See <u>Inverse Privacy paper</u>, by Gurevich, Hudis, and Wing

A Big Vision



Progress



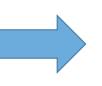


Academic Cloud for CISE workshop, January 8-9, 2018



"Enabling Computer and Information Science and Engineering Research and Education in the Cloud," Jennifer Rexford, Magdalena Balazinska, David Culler, and Jeannette M. Wing, ACM Digital Library, June 2018.





Coordinating "entity" (UW, UC San Diego, UC Berkeley) awarded August 2019

Academic Cloud = the next "Internet"

But, it needs to be for all disciplines, not just computer science, and ideally (eventually), for all academic Academic Cloud institutions worldwide, not just in the US.

