The Data Lifecycle #harvarddatafest

Mercè Crosas

Chief Data Science and Technology Officer Institute for Quantitative Social Science Harvard University

@mercecrosas

Data

Facts that can be analyzed or used in an effort to gain knowledge or make decisions; information.

Latin <u>Datum</u> "Something Given"

(The American Heritage Dictionary)





FRIDAY, JANUARY 13, 2017, 7PM

Giza Excavation 1902-1947



THE HARVARD UNIVERSITY BOSTON MUSEUM OF FINE ARTS EXPEDITION

Archeological Finds as Data



Click here to order a publication quality version of this image.

Click here to download a study copy of this image.

Photo ID number: B5789_NS

Photo subjects: Object(s) photograph: Site: Giza; view: G 7632, G 7510, street G 7000, street G 7300, street G 7400, street G 7500, avenue G 2

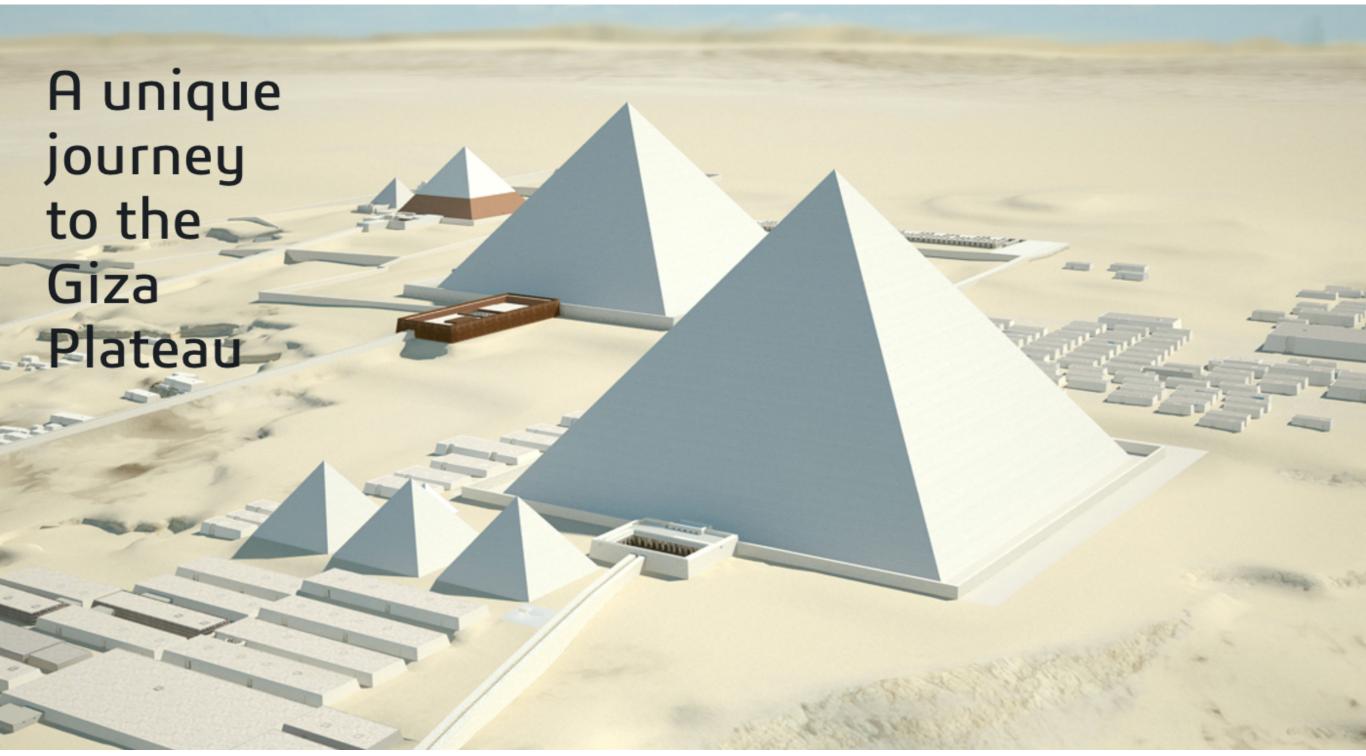
Description: Bronze objects: top row: [G 7632 A II] 25-2-822 (= MFA 25.2436, amulet of Horus), [G 7510 G] 25-2-916a (= MFA 25.2465, amulet of ichneumon), [street G 7500] 25-2-296 (= MFA 25.2211, Apis bull amulet), [street G 7300] 25-1-143 (= MFA 25.1591, Apis bull amulet), [avenue G 2] 25-1-1363 (= MFA 25.2066, Apis bull amulet), [street G 7500] 25-2-297 (= MFA 25.2212, amulet of Anubis); second row, left: [street G 7500] 25-1-1385 (= MFA 25.2082, ladle); second row, right: [G 7130-7140: G 7130 A] 24-12-513 (= MFA 24.2866, unidentified deity, fragment), [avenue G 2] 25-1-1364 (= MFA 25.2067, fragment of ring); third row, right: [street G 7000] 24-11-39 (= MFA 24.2271, fish on standard); bottom row: [street G 7000] 24-11-838 (= MFA 24.2638, cat on standard), [G 7510 D] 25-3-252 (= MFA 25.2672, flail), [street G 7400] 25-1-620 (= MFA 25.1766, beard), [street G 7000] 24-11-223 (= MFA 24.2371, amulet of Mehit and Inher), [avenue G 2] 25-1-1424 (= MFA 25.2103, uraeus), [street G 7400] 25-1-842 (= MFA 25.1853, uraeus)

Photo date: 10/21/1925

The Giza Archive and catalogue:

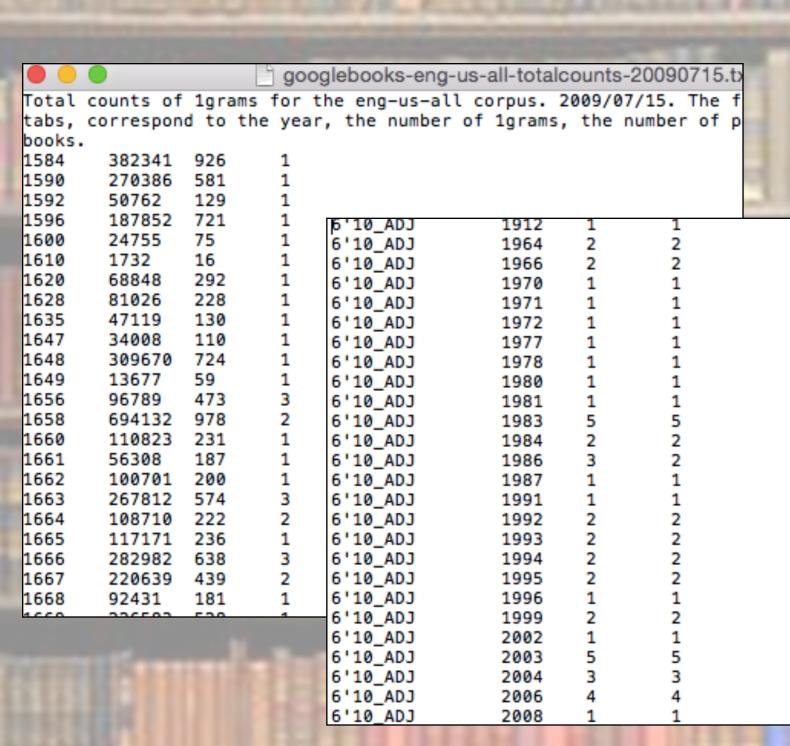
- 21,000 finds
- 3,000 photos
- 3000 tomb and monument records
- 3,000 diary pages
- 7,000 maps
- books, manuscripts
- metadata

3D Visualization to Re-Explore Giza



Peter Der Manuelian, Harvard's Visualization Center (Geological Museum), DARTH, and Dassault Systèmes in Paris

Books as Data

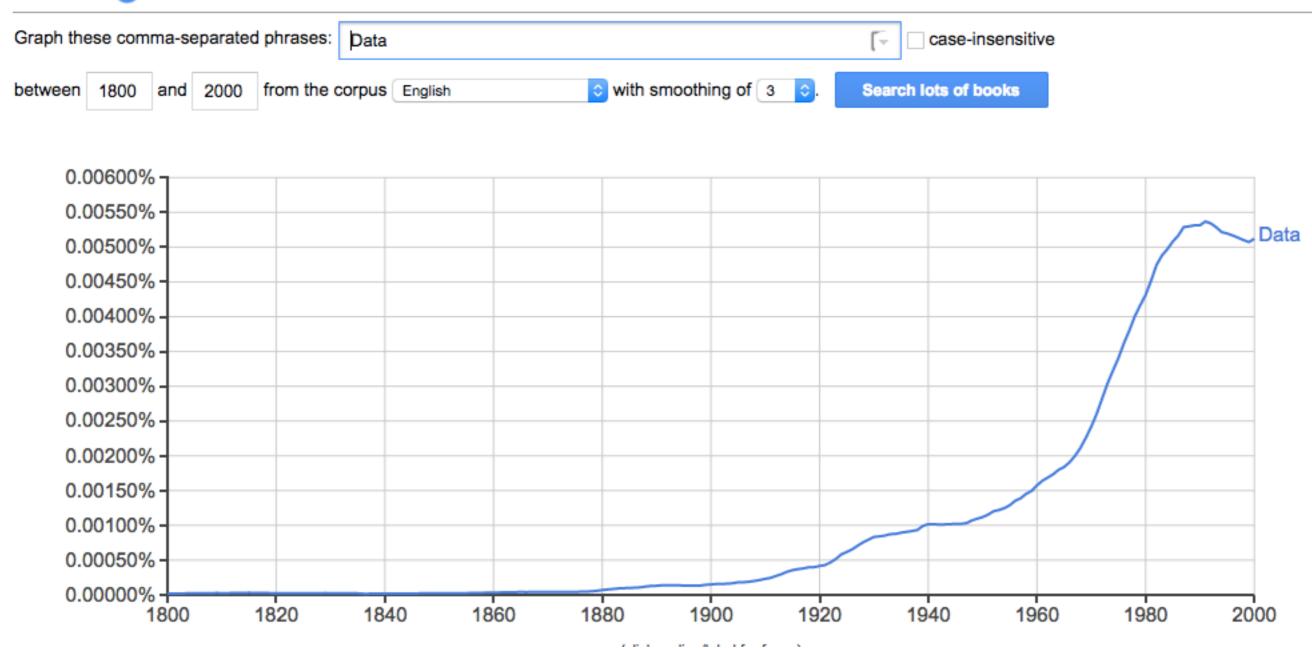


Harvard Cultural
Observatory (Aiden,
Michel, SEAS) + Google:

- 5 million digitized books (from I 30 million total)
- From year 1500 to 2008
- Optical Character
 Recognition (OCR)
- Quantify unstructured text to n-gram count matrix

Frequency of Words with Time

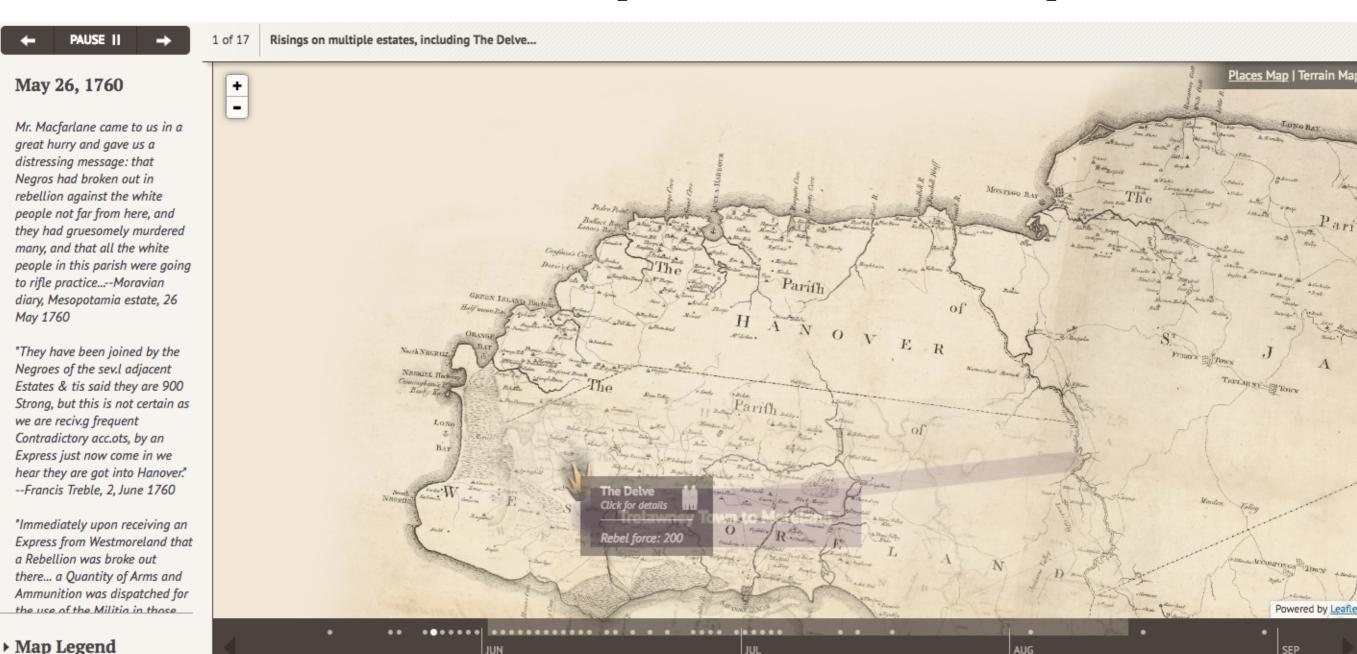
Google Books Ngram Viewer



Historical events as Data

18.3648, -76.8852 to NW Trinity mispl 18.3542, -76.8990 NW	aced to	1 Conspiracy 1 Rebels	Frontier Trinity	4/7/1760 4/7/1760	2	5 Rebel conspira 100 Rebel force	ators	18.3648 18.3542	-76.8852 -76.899		1	
Still glitchy	start			4/8/1760	0							
						1						
POSITION	Places Notes	Terrain Notes Needs static event map			ID		TYPE		NAME	DA	DATE 4/6/1760	
	Needs static event map											
												4/7/1760
18.3648, -76.8852	Frontier misplaced to NW						1	Consp	oiracy	Frontier		4/7/1760
18.3542, -76.8990	Trinity misplaced to NW						1	Rebels	s	Trinity		4/7/1760
										1		4/8/1760
18.3542, -76.8990 to Ft. Haldane	Still glitchy; start should be from Trinity						1	Rebel	e	Ft. Haldane		4/8/1760
to i d rididano	database fixed?							Clash		Ft. Haldane		4/8/1760
	Trinity misplaced to							Oldon		T E Flandario		1,0,1100
18.3542, -76.8990	NW						1	Rebels	s	Trinity		4/8/1760
												4
	_									-76.8882], [18.2709, -76.8851],		
Rebels sho move from 18.3223, -76.8875 Ballard's V: to Esher Esher		1 Rebels	Esher	4/8/1760	6			18.3223	-76.8875	-7-0.885], [18.2704, -7-0.8822], [18.2703, -7-0.8773], [18.2722, -7-0.8736], [18.2783, -7-0.8703], [18.2809, -7-0.8675, -7-0.8675]	1	
Eliminate movement; appear at s	To extent possible, governent should follow roads (dotted line) Militia ngle	1 Clash	Esher	4/8/1760	7		5 whites killed	18.2809	-76.8675		1	
point, Balla 18.3223, -76.8875 Valley	rd's To extent possible, movement should follow roads (dotted line)	2 Militia	Ballard's Valley	4/8/1760	8			18.3223	-76.8875	[[18.2809, -76.8675], [18.2792, -76.8694], [18.2766,	•	
	Geospat	tial.		orc	lina	tos	ı			-76.8720], [18.2740, -76.8736], [18.2708,	Tir	ne
	Jeospai	LIAI			III Ia	ICE2)			[16.2706, -76.8796], [18.2706, -76.8839],		IIE
	•									[18.2809, -76.8906], [18.2885,		
										-76.8921], [18.2908, -76.8964],		
										[18.2919, -76.8995], [18.2952, -76.8995],		

Jamaica Slave Revolt in Space and Time (1760 - 1761)

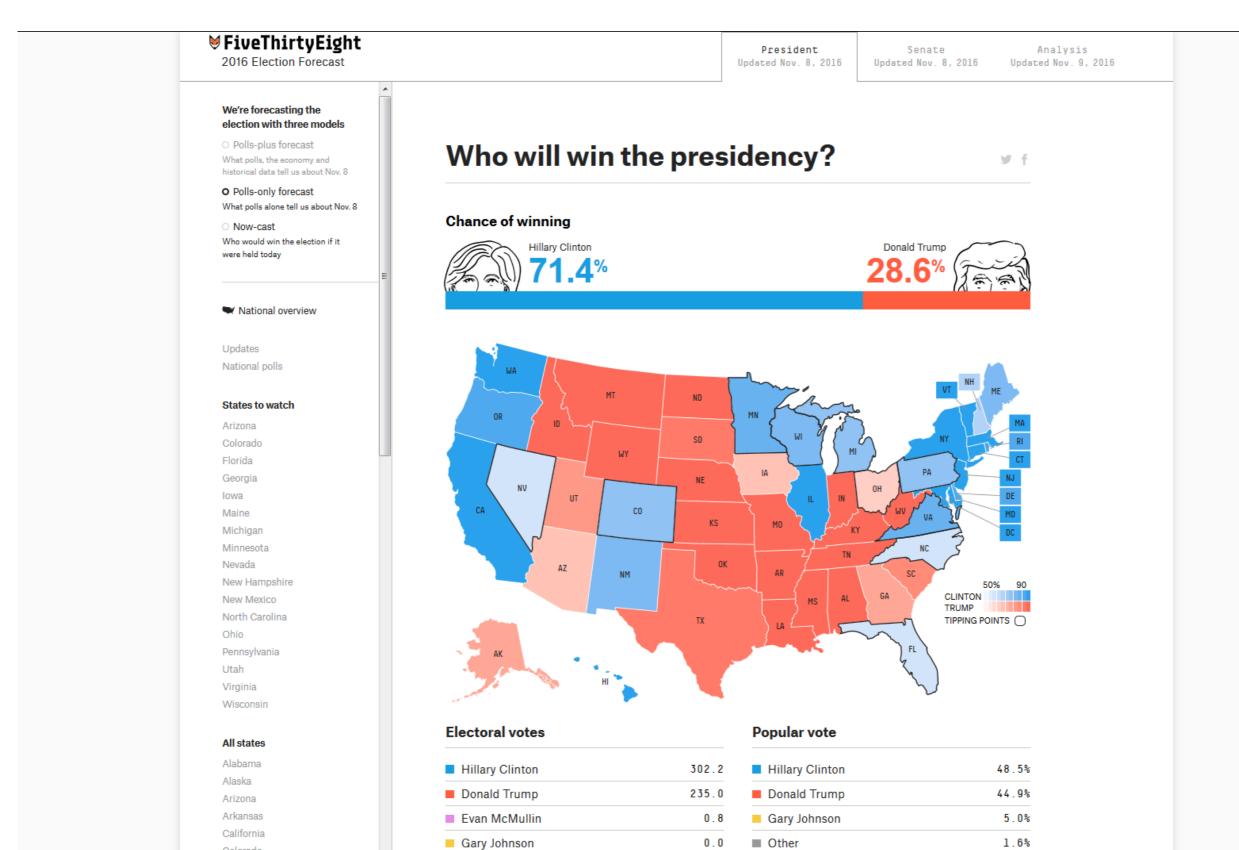


Animated maps to narrate and explore history data by Vincent Brown (Harvard's History Department)

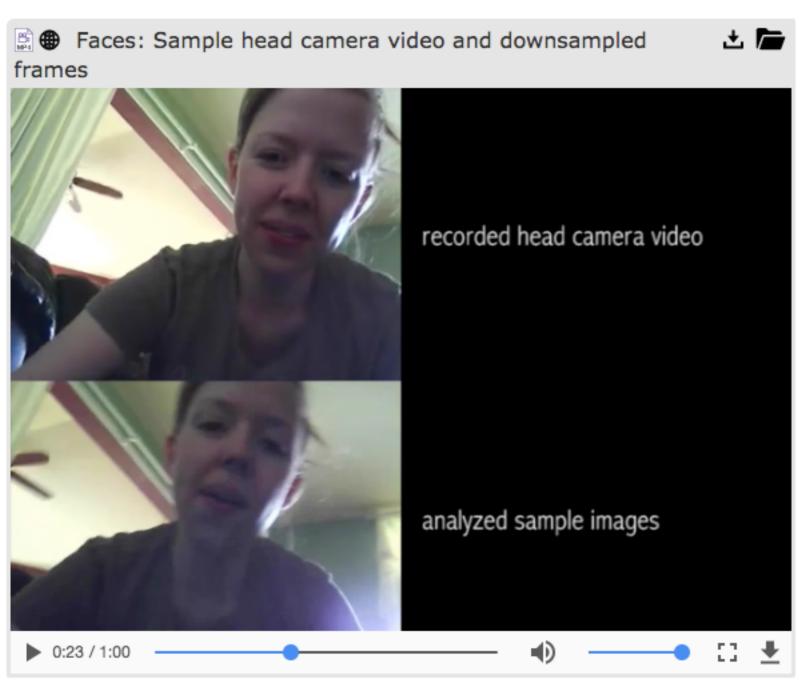
Polls as Data

	A	D	-	U	-	P	G	п		J	N.	L	[V]	IN	U	
1	Trump	Clinton	Johnson	Other		poll_slug	survey_hous			question_tex sam	-	observations n			partisanship partisan_af	fili
2	41	45	5	4		3 insights-wes				As you may k Likel		940	3.2	Internet	Nonpartisan None	
3	6	89	9	1		insights-wes				As you may k Likel				Internet	Nonpartisan None	
4	82		7	3		insights-wes				As you may k Likel				Internet	Nonpartisan None	
5	38		3	8		7 insights-wes	_			As you may k Likel				Internet	Nonpartisan None	
6	43		L	7		ibd-tipp-268		11/4/16			ly Voters	1107		Live Phone	Nonpartisan None	
7	42					ibd-tipp-268		11/4/16		Likel	ly Voters	1107	3.1	Live Phone	Nonpartisan None	
8	41			5		4 yougov-econ										
9	6	90)	1		yougov-econ				201			Lia 1		40.	
10	84		-	4		4 yougov-econ				_ Z U I (O C	:ieci	uon	I レa	lla:	
11	44			9	9 8	yougov-econ										
12	45				7		YouGov/Ecor				$I \cap$	Λ	•	\.:C	II-	
13	6	91			2		YouGov/Ecor				ΙU	US S	cier	ITITIC	polls	
4	88				5		YouGov/Ecor				. •		.	. •	- P	
15	49			1	13		YouGov/Ecor				N 4 ·				Ī	ī
16	40			5		7 lucid-the-tim	-					Xed	mo	dela	s and sa	mole
7	43	46	5			bloomberg-s			11/6/16			\mathcal{N}	1110	CIC	aria sa	\cdots
8	6		5			bloomberg-s			11/6/16		_					-
19	85	5	5			4 bloomberg-s	_			lf ·	fra	mes				
20	44	38	3	1		B bloomberg-s			11/6/16	lf:	II a	11163				
21	41	44	4	4		B bloomberg-s				lf ·		_	Ā			
22	43	47	7	4	4 1	abc-post-267	ABC/Post	11/3/16	11/6/16			interv	iewer	~admi	inistered	
23	46	49	9		3 2	2 abc-post-267	ABC/Post	11/3/16	11/6/16							
24	44	50)	4	3 1	1 monmouth-	Monmouth l	11/3/16	11/6/16	lf:	4	teleph	one r	alla		
25	44	50)		5 1	1 monmouth-u	Monmouth l	11/3/16	11/6/16			celepii	one h	JO113		
26	44	48	3	3	4 1	1 fox-26784	FOX	11/3/16	11/6/16			intoro	ctivo	voico	-response p	Solls
27	4	91	L	2	3 (fox-26784	FOX	11/3/16	11/6/16			iiicei a	Cuve-	VOICE	-i esholise t	JOIIS
28	86	9	9	2	3 1	1 fox-26784	FOX	11/3/16	11/6/16				L:I:4.,	b	مالم حام مربر	
29	42	36	5	9	9 5	fox-26784	FOX	11/3/16	11/6/16			probai	Dility-	-basec	l web polls	
30	44	48	3		2 7	7 fox-26784	FOX	11/3/16	11/6/16		_ '		1.		1 1 11.	
31	4	90)		1 4	4 fox-26784	FOX	11/3/16	11/6/16			DOOD (qualit	y non	-probability	/ based
32	85	8	3		1 6	6 fox-26784	FOX	11/3/16	11/6/16				•	,	1 /	
33	43	37	7		3 18	8 fox-26784	FOX	11/3/16	11/6/16		•	web p	olls			
34	43	41	L	6	4 5	ibd-tipp-267	IBD/TIPP	11/3/16	11/6/16			.				
35	42	43	3		5 9	ibd-tipp-267	IBD/TIPP	11/3/16	11/6/16		11				f	4:
36	39	44	1		9 8	ipsos-reuters	Ipsos/Reuter	11/2/16	11/6/16	lf:	INT	erer	ıce	and	torecas	Sting
37	7	84	1		5 5	ipsos-reuters	lpsos/Reuter	11/2/16	11/6/16	lf:		.				00
38	81	6	5		7	7 ipsos-reuters	lpsos/Reuter	11/2/16	11/6/16	If the 2016 p Likel	ly Voters -	Republican		Internet	Nonpartisan None	
39	24	29	9	2	28 19	ipsos-reuters	lpsos/Reuter	11/2/16	11/6/16	If the 2016 p Likel	ly Voters -	independent		Internet	Nonpartisan None	
10	39	42	2	6	6 7	7 ipsos-reuters	lpsos/Reuter	11/2/16	11/6/16	If the 2016 p Likel	ly Voters	2195	2.4	Internet	Nonpartisan None	
41	7	81	L	4	3 4	ipsos-reuters	lpsos/Reuter	11/2/16	11/6/16	If the 2016 p Likel	ly Voters -	Democrat		Internet	Nonpartisan None	
42	81	6	5	5	3 5	ipsos-reuters	lpsos/Reuter	11/2/16	11/6/16	If the 2016 p Likel	ly Voters -	Republican		Internet	Nonpartisan None	
3	24	28	3 1	5 1	16 17	7 ipsos-reuters	lpsos/Reuter	11/2/16	11/6/16	If the 2016 p Likel	ly Voters -	independent		Internet	Nonpartisan None	
14	43	45	5	4	5 4	4 rasmussen-2	Rasmussen	11/2/16	11/6/16	If the 2016 p Likel	ly Voters	1500	2.5	IVR/Online	Nonpartisan None	
15	43	47	7		5 5	cbs-times-26	CBS/Times	11/2/16	11/6/16	Likel	ly Voters	1426	3	Live Phone	Nonpartisan None	
16	7	88	3		1 4	4 cbs-times-26	CBS/Times	11/2/16	11/6/16	Likel	ly Voters -	Democrat		Live Phone	Nonpartisan None	
47	0.0				-	1 -b - 4: 20	CDC/Times	11/2/16	11/0/10	1.311	h. M-4	Danieliane.		Live Diverse	Managettana Mana	

(no comment)



Video as Data

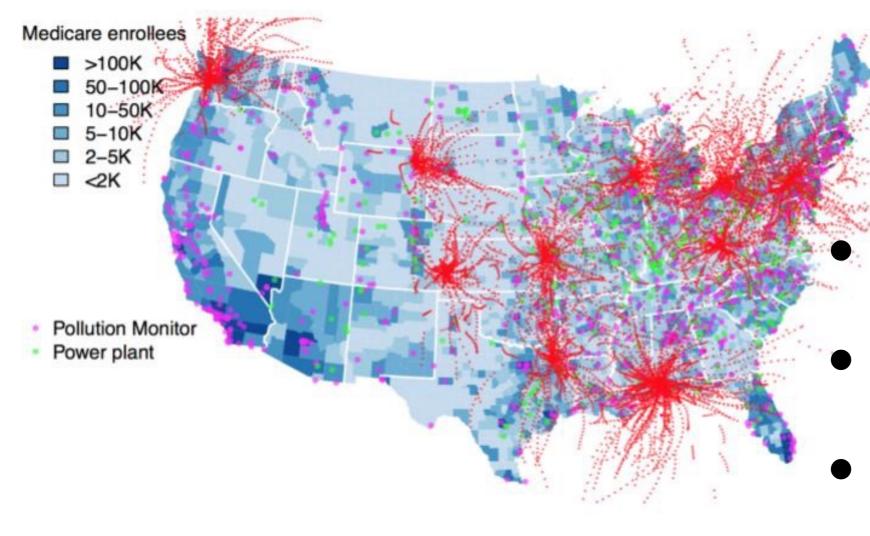


Video in developmental psychology:

- I43 hours of infantperspective scenes, collected from 34 infants aged I month to 2 years
- Coding and annotations

Fausey, C.M., Smith, L.B. & Jayaraman, S. (2015). From faces to hands: Changing visual input in the first two years. Databrary

Pollution measurements as Data



Combine data from diverse sources:

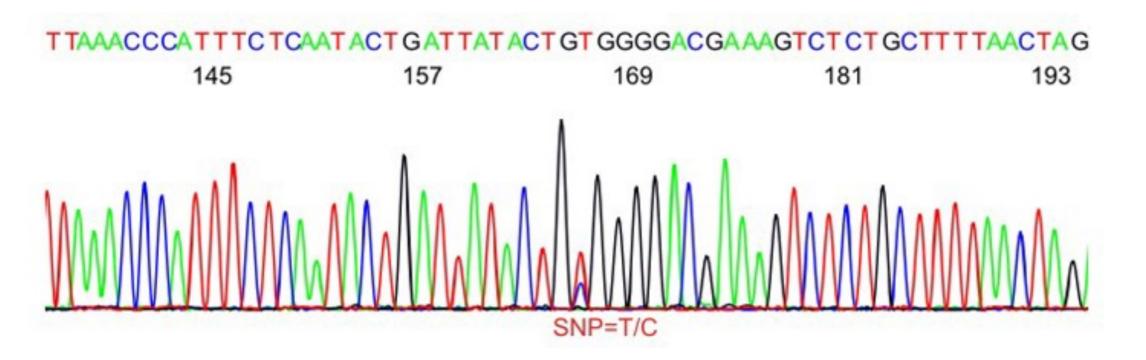
- 3000 air pollution monitors
- 1800 power plants SO₂ emissions
- claims data from 60 M
 Medicare beneficiaries
- air pollution trajectories based on weather models

Zigler, Dominici, Wang, Kim, Choirat (Harvard Chan School of Public Health)

Your Genome as Data

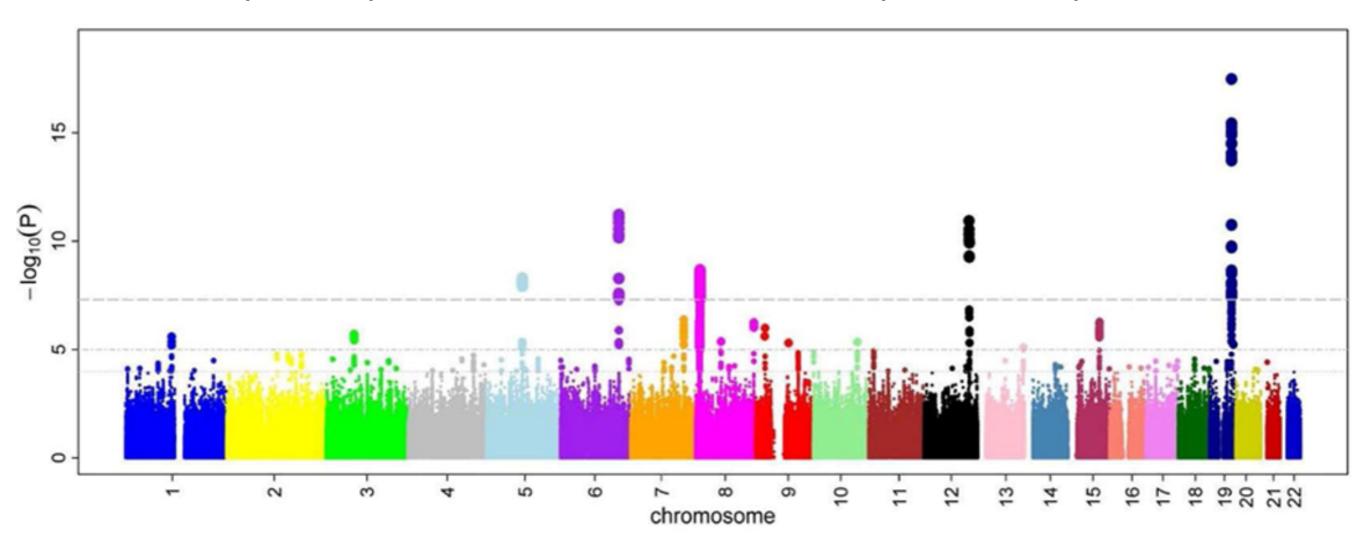


Measurement of Single-Nucleotide Polymorphisms (SNPs) with Sequencer instrument



Genome Wide Association Study

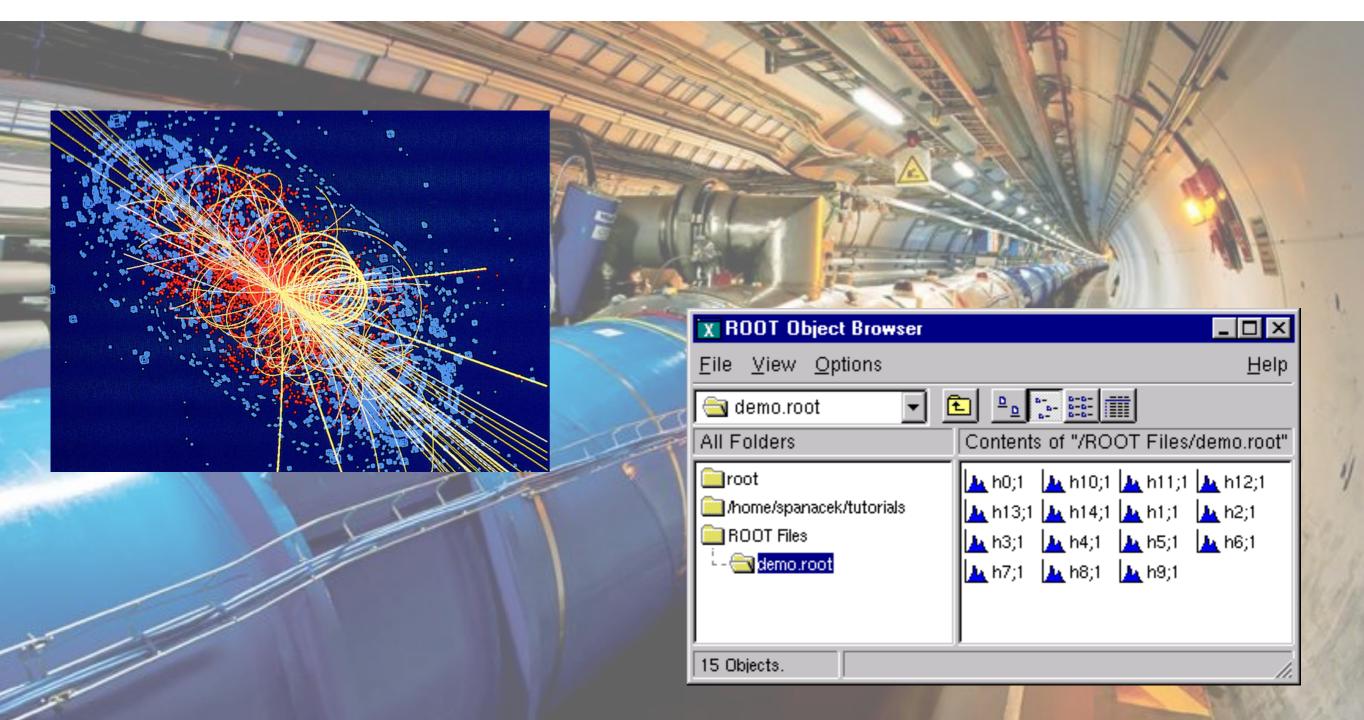
Comparing genomes from individuals with (cases) and without disease (controls)



Genetic variants more often found in individuals with constrictions in small blood vessels (GWAS, Wikipedia)

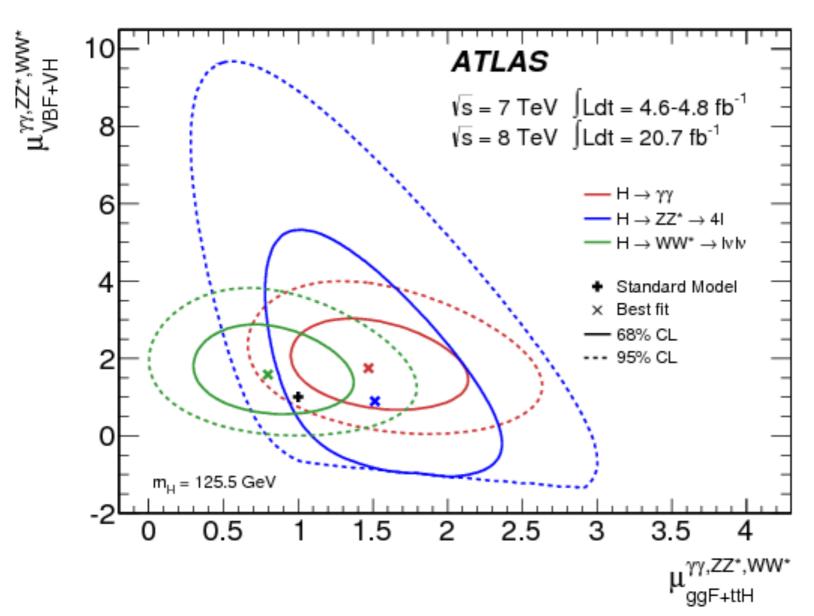
Particle Detections as Data

The Large Hadron Collider (HLC), CERN



Yes, there is a Higgs Boson!

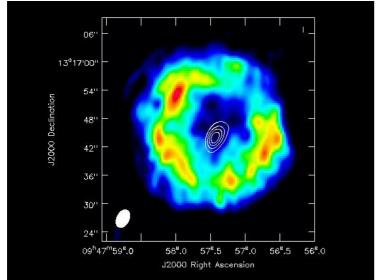
(but is it the one we expected?)

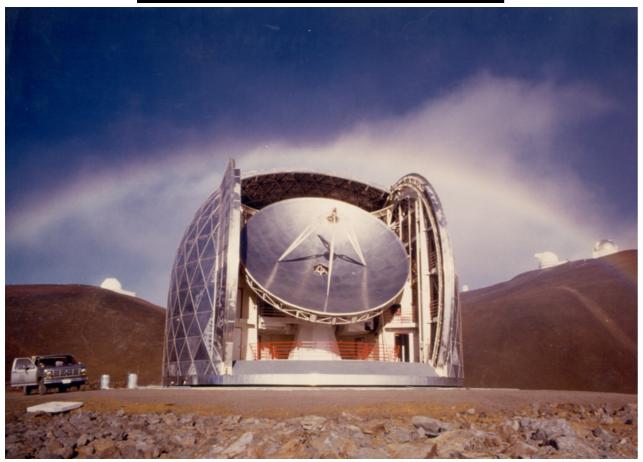


Likelihoods as result of comparing LHC observations with predictions from particle physics Standard Model

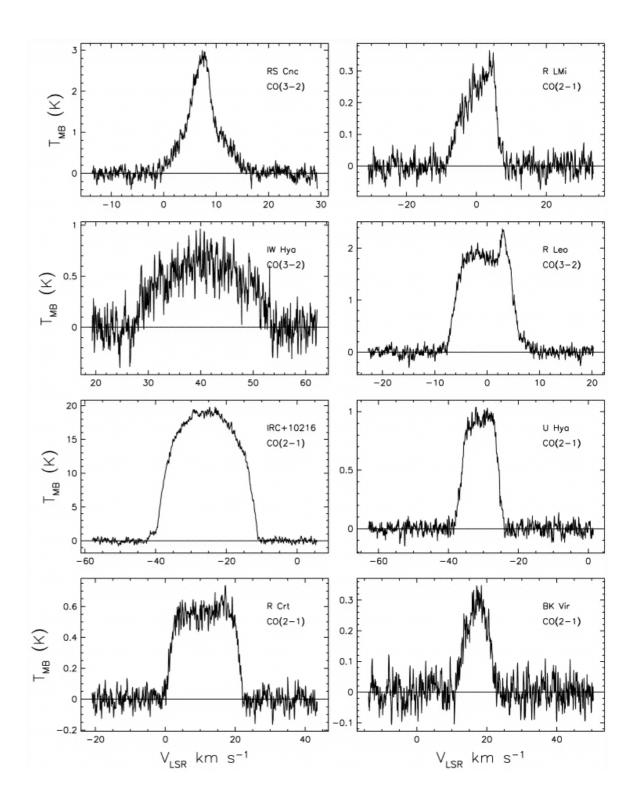
Measurements of Higgs boson production and couplings in diboson final states with the ATLAS detector at the LHC (Atlas Collaboration, 2013, with 2923 authors). Acknowledgment Kyle Cranmer (NYU)

Radio Emissions as Data





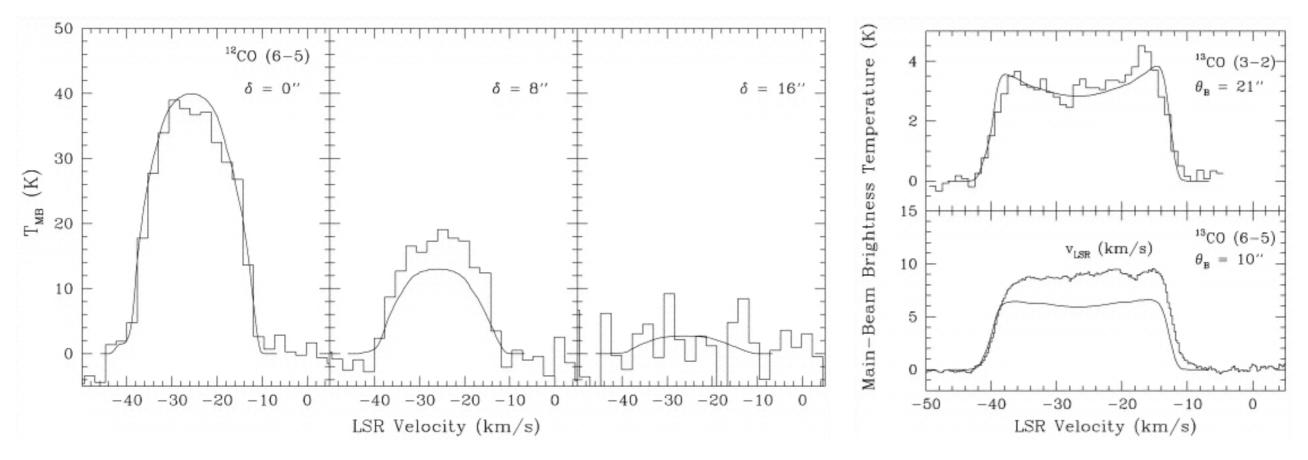
Caltech Submillimeter Observatory, California



Temperature, Mass, and Distance derived from Stellar Molecular Lines

$$\frac{dT}{dr} = (2 - 2\gamma) \left(1 + 0.5 \frac{d \ln v}{d \ln r} \right) \frac{T}{r} + \frac{\gamma - 1}{n_{\text{H}_2} k v} (H - C) . \quad (11)$$

Monte Carlo simulations of radiative transfer compare observations with predictions from physical laws



Crosas, Menten, Physical Parameters of the IRC+10216 Circumstellar Envelope (ApJ, 1997)

raw or primary data



order, transform, and analyze them

Catalog Classify Visualize Quantify Summarize Geo reference Inference Missing data **Forecast** Causal Inference Coding **Annotations Associations** Likelihoods Compare with theory

gain knowledge, make decisions

Learn about the whole from a part.

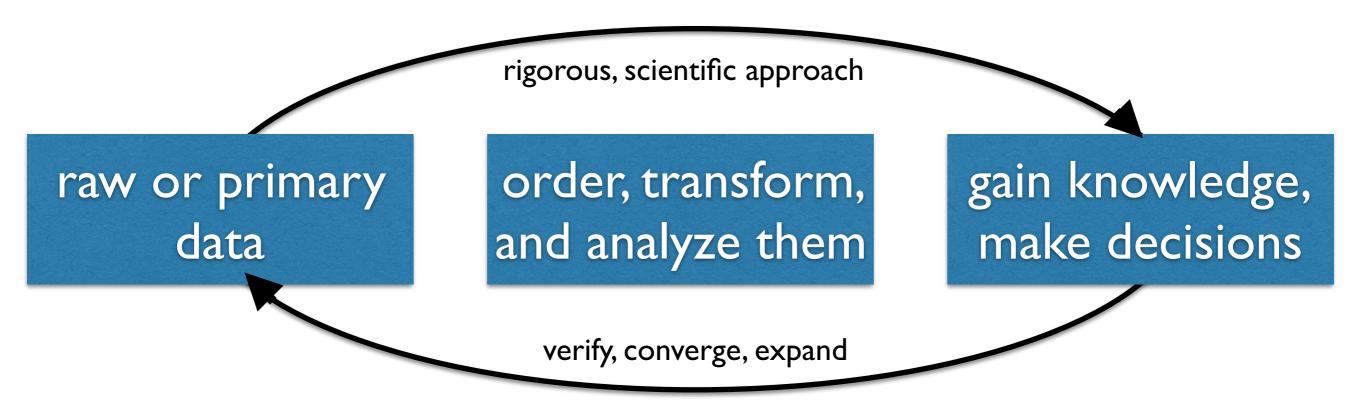
Tell a story.

Make a prediction.

Ultimately explain.

"Nullius in Verba"

(1665, the Royal Society motto and origin of Philosophical Transactions)



- Replication*: Independent scientific experiments to validate findings
- Reproducibility*: Calculation of quantitative results by others using original datasets and methods

(Stodden, Leisch, Peng, Implementing Reproducible Research, 2014)

* Replication and reproducibility definitions vary across disciplines

"Answering even a simple scientific question requires lots of choices that can shape the results"



Problems with scientific research

How science goes wrong

Scientific research has changed the world. Now it need

Oct 19th 2013 | From the print edition



A SIMPLE idea underpins science: "trust, but verify". Resu challenge from experiment. That simple but powerful idea h knowledge. Since its birth in the 17th century, modern scie beyond recognition, and overwhelmingly for the better.

INSIGHTS

Design principles for synthetic ecology p. 1425 >

Whacking hydrogen into metal p. 1429



SCIENTIFIC INTEGRITY

Self-correction in science at work

Improve incentives to support research integrity

By Bruce Alberts, 'Ralph J. Cicerone,'
Stephen E. Fienberg, 'Alexander Kamb,
Marcia McNutt, ''s Robert M. Nerem,
Randy Schekman, 'Richard Shiffrin,'
Victoria Stodden, 'Subra Suresh, ''
Maria T. Zuber, '' Barbara Kline Pope, ''
Kathleen Hall Jamieson.''

eek after week, news outlets carry word of new scientific discoveries, but the media sometimes give is, but the media sometimes give suspect science equal play with substantive discoveries. Careful qualifications about what is known are lost in categorical headlines. Rare instances of misconduct or instances of irreproducibility are translated into concerns that science is broken. The Octoonerns that science is broken. The Octoonerns

ber 2013 Economist headline proclaimed "Trouble at the lab: Scientists like to think of science as self-correcting. To an alarming degree, it is not" (I). Yet, that article is also rich with instances of science both policing itself, which is how the problems came to The Economist's attention in the first place, and addressing discovered lapses and irreproducibility concerns. In light of such issues and efforts, the U.S. National Academy of Sciences (NAS) and the Annenberg Retreat at Sunnylands convened our group to examine ways to remove some of the current disincentives to high standards of integrity in science.

Like all human endeavors, science is imperfect. However, as Robert Merton noted more than half a century ago "the ous policing, to leled in any oth a result, as Popp of the very few

POLICY

of success, not is onstrate the und nisms of science Still, as in an writ large does n als. Although att Wakefield study ∀ FiveThirtyEight Science





Science Isn't Broken

It's just a hell of a lot harder than we give it credit for.

By CHRISTIE ASCHWANDEN
Graphics by RITCHIE KING

I fyou follow the headlines, your confidence in science may have taken a hit lately. Peer review? More like self-review. An investigation in November uncovered a scam in which researchers were rubber-stamping their own work, circumventing peer review at five high-profile

"When possible, make data, methods, and code open to verify"

"Science/research might be imperfect, but is self-correcting"

"It's not unreliable, but more challenging that we give it credit for"





Altmetric: 442

Views: 29,187

Citations: 32

More detail >>

Comment | OPEN

The FAIR Guiding Principles for scientific data management and stewardship

Mark D. Wilkinson, Michel Dumontier, IJsbrand Jan Aalbersberg, Gabrielle Appleton, Myles Axton, Arie Baak, Niklas Blomberg, Jan-Willem Boiten, Luiz Bonino da Silva Santos, Philip E. Bourne, Jildau Bouwman, Anthony J. Brookes, Tim Clark, Mercè Crosas, Ingrid Dillo, Olivier Dumon, Scott Edmunds, Chris T. Evelo, Richard Finkers, Alejandra Gonzalez-Beltran, Alasdair J.G. Gray, Paul Groth, Carole Goble, Jeffrey S. Grethe, Jaap Heringa, Peter A.C 't Hoen, Rob Hooft, Tobias Kuhn, Ruben Kok, Joost Kok, Scott J. Lusher, Maryann E. Martone, Albert Mons, Abel L. Packer, Bengt Persson, Philippe Rocca-Serra, Marco Roos, Rene van Schaik, Susanna-Assunta Sansone, Erik Schultes, Thierry Sengstag, Ted Slater, George Strawn, Morris A. Swertz, Mark Thompson, Johan van der Lei, Erik van Mulligen, Jan Velterop, Andra Waagmeester, Peter Wittenburg, Katherine Wolstencroft, Jun Zhao & Barend Mons

Findable
Accessible
Interoperable
Reusable

(data, code, methods)

Harvard Data Resources

Harvard Library Services and Archives

Harvard Dataverse (IQSS, Library, HUIT)

Research Computing Council

Harvard Data Group (OVPR)

VPAL Research Group

HMS Data group

Harvard Chan Core

DARTH Arts Management Bioinformatics and Humanities Computing

Geographic **Analysis**

Center for IQSS Data Science HBS Research Services & Survey Research

Computing **Services**

Services and Resources

Orchestra (HMS)

Odyssey (FAS)

RCE (IQSS)

HBS RC (HBS)

Computing & Storage Infrastructure

DataFest

Follow your sessions using SCHED

Data Concepts

Project Planning a Data Science Workflows

Processing and Analyzing Data*

Dissemination

Plan how to address your research question

Address it!

Share your results, data, code with others

*Does not include handling sensitive data

Don't miss!

Continental Breakfast Computing Help Desk 8:30am Coffee Break Computing Help Desk 10:00am Archive-a-thon signup Computing Help Desk 12:00pm Lunch Archive-a-thon signup

5:30pm

Reception

If you leave DataFest thinking "I need to learn more about X", we have already succeeded. (however, we welcome your feedback, using SCHED)

Harvard DataFest 2017

Schedule * Speakers Attendees About Tickets Search



Troy Adair Harvard Business School Senior Director, Research Computing Services



Jeff Blossom Harvard University



Christine Choirat



Simo Goshev

IQSS, Harvard University

Daina Bouquin

Harvard-Smithsonian Center for

Astrophysics

Head Librarian

IQSS, Harvard University



Christopher Gandrud IQSS



Nicole Alexander



Sonia Barbosa

Alex Caracuzzo

Baker Library, Harvard Business

School

Collections and Data Management

Librarian

Gustavo Durand

Wendy Guan



Lex Berman Center for Geographic Analy Web Services Mgr, GIS Specia

□ □ □ □ ▼ f



Kareem Carr





Bob Freeman Harvard Business School Dir, Research TechOps



Chase Harrison



Judson Harward Harvard University Director, Arts&Humanities Research Computing



Ben Lewis



Jessica Pierce



Hugh Truslow Harvard University Head, Social Sciences and



Amir Karger Harvard Medical School Associate Director of Research Computing



Derek Miller Harvard University Assistant Professor of English



Diane Sredl

Maps, Media, Data, and

Government Information, Lamont

Library

Scott Yockel

FAS Research Computing

Sr. Team Lead for HPC



Caroline Shamu Harvard Medical School Assistant Professor



Steven Worthington



Radhika Khetani



Scott Lapinski Harvard Medical School Scholarly Communication &



David Osterbur Harvard Medical School, County Library of Medicine Director of Public and Access Services



Dustin Tingley Harvard University Professor of Government



Ista Zahn The Institute for Quantitative Social Science at Harvard