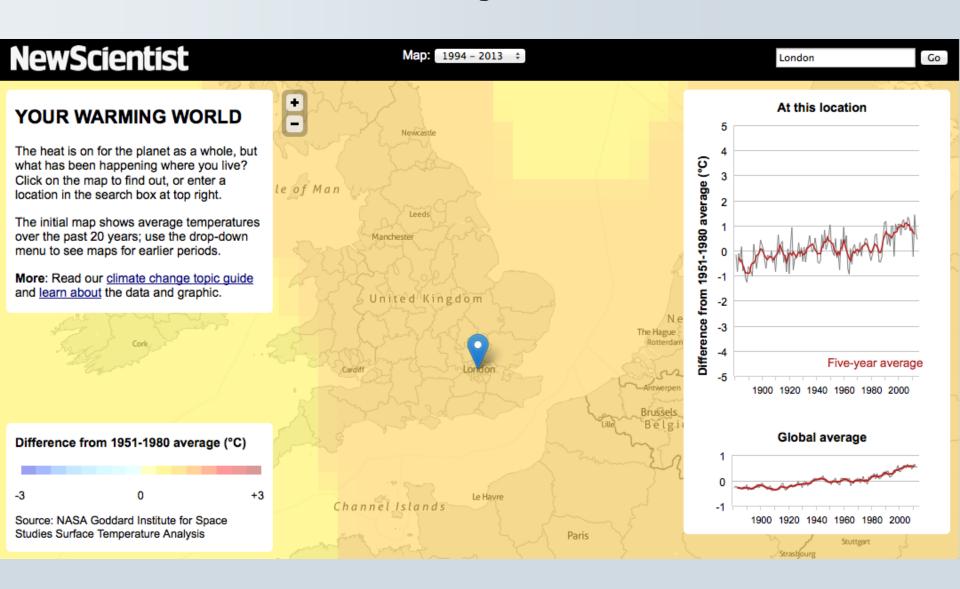
Investigating science

(with help from data)

Peter Aldhous, Science reporter, BuzzFeed News

<u>peter@peteraldhous.com</u> <u>@paldhous</u>

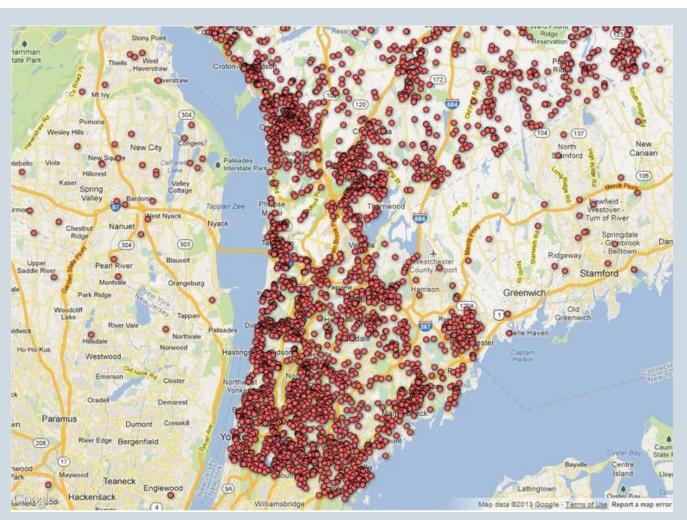
Is this data journalism?



Is this data journalism?

Map: Where are the gun permits in your neighborhood?

FILED UNDER - News / Local News I 3:51 PM, Dec. 22, 2012



Is this data journalism?

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POLITICS EDUCATION TEXAS

BREAKDOWN Death and disarray at America's racetracks

Mangled Horses, Maimed Jockeys

The new economics of horse racing are making an always-dangerous game even more so, as lax oversight puts animal and rider at risk.

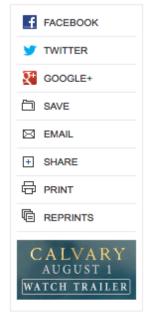
By WALT BOGDANICH, JOE DRAPE, DARA L. MILES and GRIFFIN PALMER Published: March 24, 2012

☐ 481 Comments

RUIDOSO, N.M. — At 2:11 p.m., as two ambulances waited with motors running, 10 horses burst from the starting gate at Ruidoso Downs Race Track 6,900 feet up in New Mexico's Sacramento Mountains.

Nineteen seconds later, under a brilliant blue sky, a national champion jockey named Jacky Martin lay sprawled in the furrowed dirt just past the finish line, paralyzed, his neck broken in three places. On the ground next to him, his frightened horse, leg broken and chest heaving, was minutes away from being euthanized on the track.

For finishing fourth on this early September day last year, Jacky Martin got about \$60 and possibly a lifetime tethered to a respirator.





A Jockey's Story

Jockeys are on the front lines of the risky business of horse racing. Twenty-four horses a week die at racetracks around the country.

Video by Matthew Orr/The New York Times

Data + journalism = story



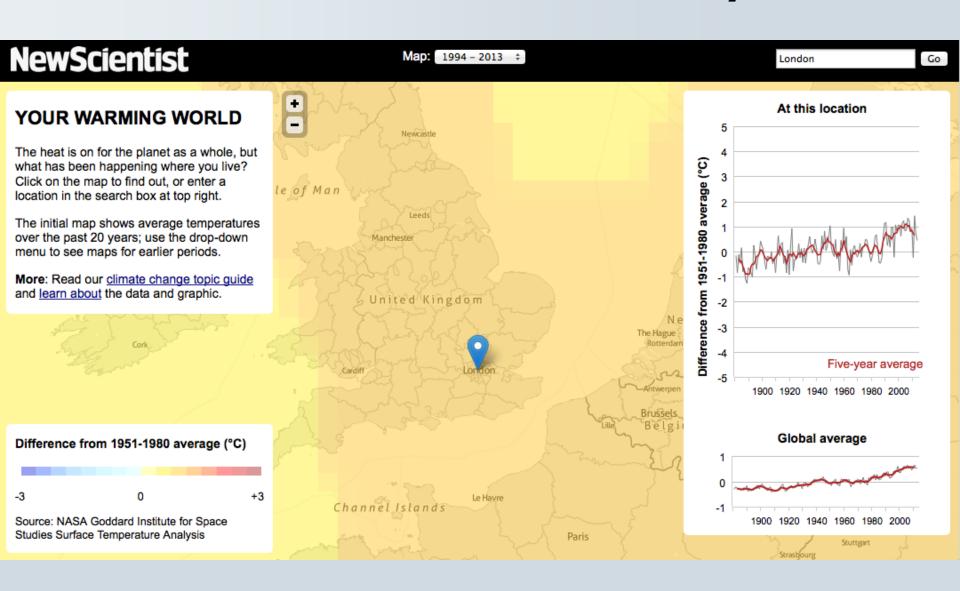
BREAKDOWN Death and disarray at America's racetracks

Mangled Horses, Maimed Jockeys

The new economics of horse racing are making an always-dangerous game even more so, as lax oversight puts animal and rider at risk.

But an investigation by The New York Times has found that industry practices continue to put animal and rider at risk. A computer analysis of data from more than 150,000 races, along with injury reports, drug test results and interviews, shows an industry still mired in a culture of drugs and lax regulation and a fatal breakdown rate that remains far worse than in most of the world.

... which can be told visually



But it's not only maps and graphs!

Data can be used in both reporting and storytelling.

And visualization can be a powerful for both finding and telling stories.

But think carefully about what you need to show to your audience. Some of the best data-driven stories contain little in the way of numbers or graphs.

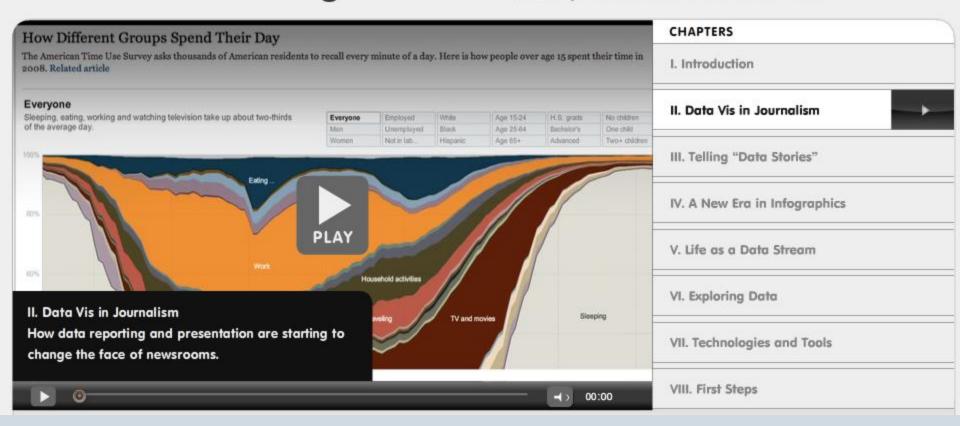
Have you ...

- Filed a request for public records?
- Worked on a story that required a spreadsheet to analyze data?
- Worked on a story that relied on querying an online database?
- Obtained a database and analyzed with database management software, or built your own database?
- Used mapping software for a story?
- Used statistical analysis for a story?
- Studied connections between people/organizations using network analysis?

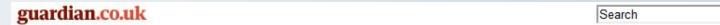
A new thing?

Journalism in the Age of Data

A video report on data visualization as a storytelling medium Produced during a 2009-2010 Knight Journalism Fellowship Total Running Time: 54 Minutes; with related information and links



A thing of the future?



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Analysing data is the future for journalists, says Tim Berners-Lee

Inventor of the world wide web says reporters should be hunting for stories in datasets





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Charles Arthur The Guardian, Monday 22 November 2010 Article history



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Digital media -Journalism education

Technology

Tim Berners-Lee

More features

More on this story



Berners-Lee: Facebook could fragment web Founder of world wide



Banks Forced To Forgive Credit Card Debt

New Credit Laws Allow San Francisco Consumers to Reduce Debt up to 60%...



Mom's \$5 Wrinkle Secret!

San Francisco: Dermatologists DON'T Want You Knowing This Skin Care Trick!



What's Your Credit Score?

The Average Credit Score is 678. Find Out Your Score For Free.

ADS BY YABUKA.COM >

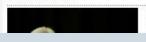


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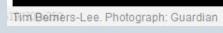
Zeitgeist

Latest

Last 24 hours



1. Julian Assange: Whoever leaked US



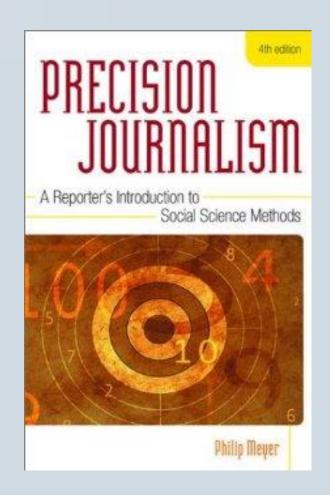
The pioneer: Philip Meyer



Now emeritus professor of journalism, University of North Carolina at Chapel Hill.

Pioneered use of quantitative methods in journalism with Knight Newspapers in 1960s.

Author of *Precision Journalism*, first published 1973.



A Pulitzer for data journalism: 1967 Detroit riot



Data: Survey conducted in the immediate aftermath of the riot.

Findings: One theory held that the rioters were reacting to being stuck at the foot of the economic ladder. Another blamed southern blacks who had moved to Detroit. But Philip Meyer showed that college graduates were as likely to have rioted as high-school dropouts, and that those born in the South were less likely to have participated.

Attention turned instead to pervasive racial discrimination in policing and housing in Detroit.

What's in it for me?

- Place your other reporting in context. Less "he said; she said."
- Visualize complex stories: Fresh understanding; new points of entry
- Find original stories, new angles

Where do I start?

Usually, with a question you want to answer, or a point you want to demonstrate.

Good data journalism rarely starts by aimlessly poking at a dataset.

Approach data like you would an interview: What do you and your audience want to know?

The data frame of mind

- When you start working on a story, think "What sources of data are available?" as well as "Who can I speak to about this?"
- Assume the data you need exists and is open to the public until proven otherwise.
- Make it a regular practice to learn about sources of data related to your beat.
- If necessary, have a plan for acquiring data at regular intervals. Some data may require public records requests.

The data frame of mind

This very different to:

"I've written my story. Now I'd better find some numbers for a graph."

A note of caution: data is often 'dirty'

Data can be seductive, but never simply assume that it is correct and consistent. Examine any data you obtain to see how it is organized, and scan for potential errors.

You will almost always need to reformat and edit data to suit your purposes; frequently you will have to do extensive data "cleaning".

Please clean me!

d	А	В	С	D	E	F	G	Н	1	J	K	L	M	N
	REVIEWER		118111111	MIDDLE							111			
1	ID	LAST NAME	FIRST NAME	INITIAL	RANK	DEGREE	SITE	STREET ADDRESS	CITY	STATE	ZIP CODE	COUNTRY	RECEIPT DATE	TYPE
2	459203	%BENN%	TERRY	L	NG	MD	RANDOLPH FAMILY PRACTICE	1918 RANDOLPH RD STE 275	CHARLOTTE	NC	28207	US	12/5/2001	DEM
3	533704	%EL-GHOROURY%	MOHAMMAD		NG	MD	NG	22201 MOROSS STE 150	DETROIT	MI	48236	US	2/11/2011	DEM
4	512096	%GUENTHER	RAINER		NG	MD	UNIVERSITATSKLINIKUM SCHLE	SCHITTENHELMSTR 12	KIEL	NG	24105	GM	11/19/2007	DEM
5	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	3/7/2000	DEM
6	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	5/5/2000	DEM
7	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	8/21/1981	DEM
8	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	9/11/2003	DEM
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13	534551	%TAUTH	JEFFREY		NG	MD	NG	180 MEDICAL PARK DRIVE	HOT SPRINGS	AR	71901	US	4/11/2011	DEM
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17	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	5/15/2008	IRB
18	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	5/20/2008	IRB
19	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	1/9/2009	IRB
20	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	3/23/2009	IRB
21	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	4/27/2010	IRB
22	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	11/5/2009	IRB
23	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	3/10/2011	IRB
24	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	2/18/2011	IRB
25	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	10/16/2009	IRB
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28	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	7/2/2009	IRB
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But science journalists are lucky: Lots of clean, well curated data ...

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	*****		*****	******	*****	/****	* *											
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June	28	12 UT	rc 20.3N	93.1W	325	deg	4	mph	7	kph	35	mph	55	kph	1006	mb		
June	28	18 UT	rc 20.7N	93.5W	315	deg	5	mph	9	kph	40	mph	65	kph	1006	mb	Tropical	Storm
June	29	0 UT	rc 21.0N	93.9W	310	deg	4	mph	7	kph	40	mph	65	kph	1005	mb	Tropical	Storm
June	29	6 UT	rc 21.2N	94.5W	290	deg	5	mph	9	kph	40	mph	65	kph	1003	mb	Tropical	Storm
June	29	12 UT	rc 21.3N	95.3W	280	deg	8	mph	12	kph	50	mph	85	kph	1000	mb	Tropical	Storm
June	29	18 UT	rc 21.4N	95.6W	290	deg	2	mph	3	kph	60	mph	95	kph	998	mb	Tropical	Storm
June	30	0 UT	rc 21.6N	96.1W	295	deg	5	mph	9	kph	60	mph	95	kph	996	mb	Tropical	Storm
June	30	6 UT	rc 21.6N	97.0W	270	deg	9	mph	14	kph	65	mph	100	kph	994	mb	Tropical	Storm
June	30	12 UT	rc 21.6N	97.3W	270	deg	2	mph	3	kph	65	mph	100	kph	993	mb	Tropical	Storm
June	30	18 UT	rc 21.5N	98.1W	260	deg	8	mph	12	kph	50	mph	85	kph	998	mb	Tropical	Storm
July	1	0 UT	rc 21.1N	98.7W	235	deg	6	mph	11	kph	35	mph	55	kph	1002	mb	Tropical	Depression
Storm			is number		_													
Month		Hour		Long.		ir.		Spe	ed			Wir	nd		Press	sure		Туре

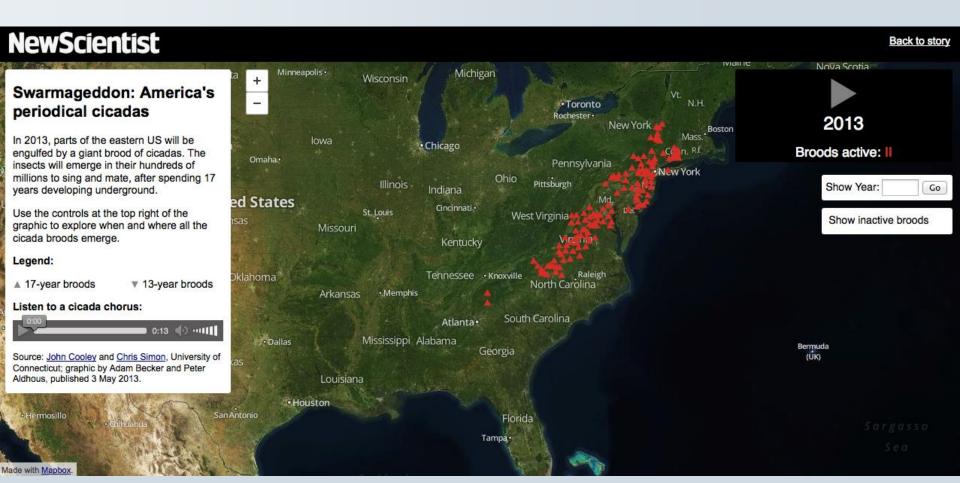
16 6 UTC 30.7N 79.7W -- deg -- mph -- kph 25 mph 35 kph 1014 mb

16 12 UTC 30.3N 79.4W 145 deg 4 mph 7 kph 25 mph 35 kph 1014 mb

July

July

...which scientists may be willing to share



The basics (OK, I have some data. What now?)

Sort

Largest to smallest; Alphabetical etc

Aggregate

Count, Sum, Mean, Median, Maximum, Minimum etc

Filter

Select a defined subset of the data

Join

Merge entries from two or more datasets based on common field(s), e.g. unique ID number, last name and first name

Think of those operations as 'interviewing' the data

Tools and stories: online databases



that researchers in Australia had created a strain of mousepox that killed even animals that had been vaccinated (19 January 2001, p. 4). The scientists, hoping to control plagues of mice, engineered a mousepox virus

support of the biosecurity specialists we asked to consider their risks and

benefits. Only one of those mentioned

above – the anthrax project – generated serious objections from

some of our experts.

20 | NewScientist | 14 October 2006 | NewScientist | 21 www.newscientist.com | 14 October 2006 | NewScientist | 21

Tools and stories: work with databases on your own computer

Revealed: Pfizer's payments to censured doctors

> 16:18 22 April 2010 by Peter Aldhous and Jim Giles

They are billed as "healthcare professionals who spend years building expertise in their fields". Using materials firmly grounded in science, they educate their peers in the risks and benefits of drugs.

This is how Pfizer, the pharmaceuticals giant, describes the experts it hires to lead educational forums in which doctors are lectured on the use of its products.

Yet New Scientist has found that some of Pfizer's experts have been disciplined for deficiencies in patient care, while others have been reprimanded for how they conducted drug research trials.

The findings add to a growing controversy surrounding the pharmaceutical industry's efforts to market drugs by influencing patterns of prescribing.

Unknown influence

Doctors paid to educate peers are a particular worry, argues Sidney Wolfe of consumer advocacy group Public Citizen in Washington DC. "They are doing things that may be influencing your doctor and you have no way of knowing about it," he says. "It's made worse by the fact that some of them have been disciplined."

Data: Pfizer's records of payments to doctors, scraped from the web. Data on disciplinary actions from state medical boards in four largest states; FDA warning letters to clinical investigators

Findings: Some of Pfizer's "experts" had questionable records for patient safety



NewScientist

Tools and stories: databases















Why Are Dope-Addicted, **Disgraced Doctors Running Our Drug Trials?**

By Peter Aldhous

Photographs by Grant Cornett

Data: US Food and Drug Administration's database of clinical investigators, joined to data on disciplinary actions from state medical boards. (Also FDA's list of disqualified investigators, and its database of inspections of clinical sites.)

Findings: Dozens of doctors selected to work on clinical trials over the past five years had previously been censured by state medical boards for problems with patient care; some had their own problems with substance abuse.

Tools and stories: databases

Home | Opinion | Health | Science in Society | News

My 'non-human' DNA: a cautionary tale

-) 15:02 26 August 2009 by Peter Aldhous
-) For similar stories, visit the Genetics Topic Guide

"This is a strange question, but are you sure this is Homo sapiens?"

It's not every day that an expert queries whether your DNA is human, so when I received this comment by email earlier this month I was somewhat bemused.

I am not in fact the result of a coupling between human and alien, nor the product of some twisted genetic experiment. Instead, Blaine Bettinger, who blogs as The Genetic Genealogist, had been baffled by a DNA profile generated in error by deCODEme, a leading commercial "personal genomics" service provided by Decode Genetics in Reykjavik, Iceland. The false profile seems to be the fault of a software bug.

No harm was done, but the incident serves as a cautionary tale for personalised medicine. As we move towards a future in which readouts from our genomes will routinely be queried by computer systems to help doctors make important clinical decisions, similar glitches could cause prescribing errors — with patients being given drugs at the wrong dose, drugs that won't work, or ones that could even trigger serious side effects in people with a

Data: Downloads of my own genetic scans, performed by 23andMe and DeCode Genetics. Corresponding data for my DNA markers read from the same companies' online "genome browsers".

Findings: DeCode had a glitch in its database software that could cause the presentation of an erroneous mitochondrial DNA profile in its genome browser.



NewScientist

DeCode's genome browser

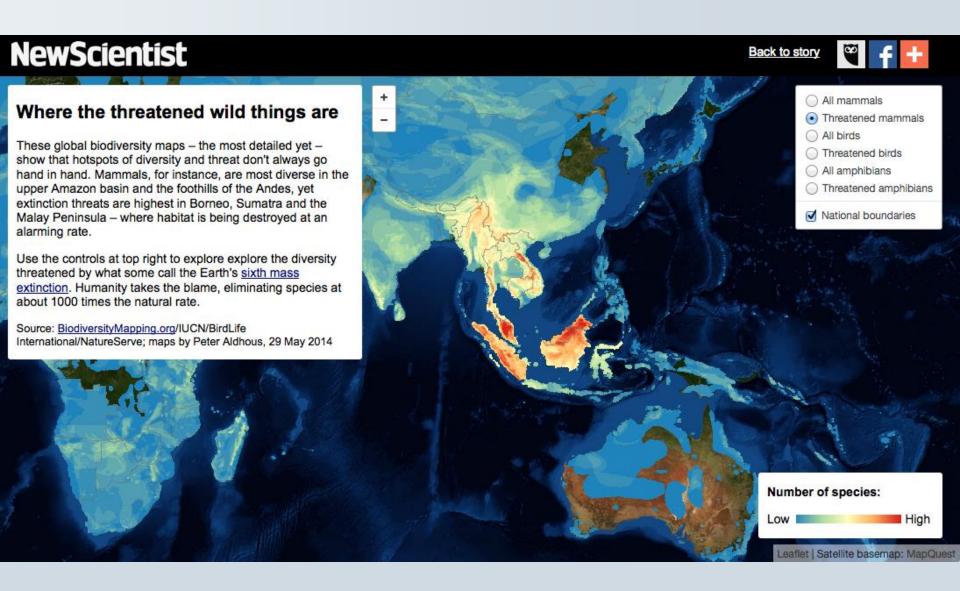




NewScientist

4	A B	С	D	E	F	G	Н	I	J	K	L	M	N
1	Position (Human NCBI Build Record 36)	Position (CRS)	23andMe ID	23andMe Variation	23andMe Download Genotype		23andMe Consistency	DeCodeMe ID		DeCodeMe Download Genotype	DeCodeMe Browser Genotype	DecodeMe Consistency	Consistency between 23andMe and DeCodeMe Downloads
20	19 2887	2885	rs2854130	C/T	T	T	Consistent	MitoT2887C	C/T	T	No data	Data in dowr	Consistent
21	20 3012	3010	rs3928306	A/G	Α	Α	Consistent	MitoG3012A	A/G	Α	A	Consistent	Consistent
22	21 3198	3197	rs2854131	C/T	T	T	Consistent	MitoT3198C	C/T	T	T	Consistent	Consistent
23	22 3349	3348	rs41423746	A/G	Α	Α	Consistent	MitoA3349G	A/G	Α	A	Consistent	Consistent
24	23 3395	3394	rs41460449	C/T	T	T	Consistent	MitoT3395C	C/T	T	С	Mismatch	Consistent
25	24 3481	3480	rs28358584	A/G	Α	Α	Consistent	MitoA3481G	A/G	Α	G	Mismatch	Consistent
26	25 3595	3594	rs2854134	C/T	С	C	Consistent	MitoC3595T	C/T	C	T	Mismatch	Consistent
27	26 3667	3666	rs28357968	A/G	G	G	Consistent	MitoG3667A	A/G	G	G	Consistent	Consistent
28	27 3721	3720	rs41355750	A/G	Α	A	Consistent	MitoA3721G	A/G	Α	G	Mismatch	Consistent
29	28 3916	3915	rs41524046	A/G	G	G	Consistent	MitoG3916A	A/G	G	G	Consistent	Consistent
30	29 3919	3918	rs28357972	A/G	G	G	Consistent	MitoG3919A	A/G	G	G	Consistent	Consistent
31	30 3971	3970	rs28357973	C/G/T	C	C	Consistent	MitoC3971T	C/T	C	T	Mismatch	Consistent
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33	32 4025	4024	i1000001	A/G	Α	Α	Consistent	MitoA4025G	A/G	Α	A	Consistent	Consistent
34	33 4337	4336	i3001462	C/T	T	T	Consistent	MitoT4337C	C/T	Т	С	Mismatch	Consistent
35	34 4562	4561	i1000011	C/T	T	T	Consistent	MitoT4562C	C/T	T	С	Mismatch	Consistent
36	35 4770	4769	rs3021086	A/G	G	G	Consistent	MitoG4770A	A/G	G	Α	Mismatch	Consistent
37	36 4821	4820	rs28357977	A/G	G	G	Consistent	MitoG4821A	A/G	G	G	Consistent	Consistent
38	37 4825	4824	rs28357978	A/G	Α	Α	Consistent	MitoA4825G	A/G	Α	No data	Data in dowr	Consistent
39	38 4884	4883	rs28357979	C/T	С	С	Consistent	MitoC4884T	C/T	С	T	Mismatch	Consistent
40	39 4918	4917	rs28357980	A/G	Α	Α	Consistent	MitoA4918G	A/G	Α	Α	Consistent	Consistent
41	40 4978	4977	rs28357981	C/T	T	T	Consistent	MitoT4978C	C/T	Т	С	Mismatch	Consistent

Tools and stories: putting data onto maps



Tools and stories: putting data onto maps

The Scattle Times

Logging and landslides: What went wrong?

When Weyerhaeuser began clear-cutting the Douglas firs on the slopes surrounding Little Mill Creek, local water officials were on edge. Some of these lands had slid decades ago, after an earlier round of logging. They worried new slides could dump sediments into the mountain stream and overwhelm a treatment plant. Those fears came true last December.

By Hal Bernton and Justin Mayo

Seattle Times staff reporters

BOISTFORT VALLEY, Lewis County — When Weyerhaeuser began clear-cutting the Douglas firs on the slopes surrounding Little Mill Creek, local water officials were on edge.

Some of these lands had slid decades ago, after an earlier round of logging. They worried new slides could dump sediments into the mountain stream and overwhelm a treatment plant.

Those fears came true last December when a monster storm barreled in from the Pacific, drenching the mountains around the Chehalis River basin and touching off hundreds of landslides. Little Mill Creek, filled with mud and debris, turned dark like chocolate syrup.

More than three months passed before nearly 3,000 valley residents could drink from their taps again.

"I have never seen anything like this before, and I hope I never do again," said Fred Hamilton, who works for the Boistfort Valley Water Corp.

State forestry rules empower the Department of Natural Resources (DNR) to restrict logging on



STEVE RINGMAN / THE SEATTLE TIMES

Data: GIS data on clear-cuts, landslides and prior studies of the hazards from the state Department of Natural Resources; logging company Weyerhaeuser's logging permits.

Findings: With little scrutiny from state geologists, Weyerhaeuser was allowed to clear-cut unstable slopes.

Using mapping software, the reporters showed that clear-cut sites that had at least half of their acreage in a moderate- to high-hazard zone accounted for a disproportionate number of landslides in December 2007 storms.

Statistical analysis, network analysis, etc:

Not just for our scientist sources



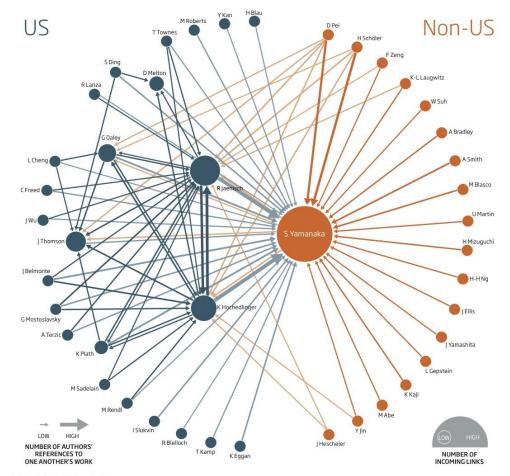
NewScientist

SPECIAL REPORT

THE STEM CELL WARS

When a Nobel prize is up for grabs, do scientists across the globe compete on a level playing field? **Peter Aldhous** investigates

The most influential players in cellular reprogramming are revealed by recording how many times the scientists have referred to each other's work. Each link shows where one researcher cited another four or more times in papers in leading journals (for analysis, see "The strongest link"; below right)

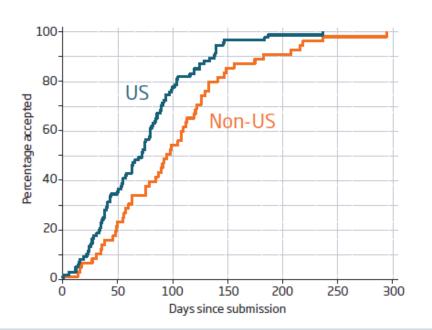


12 | NewScientist | 12 June 2010

100612_N_Special_Report.indd 12 8/6/10 14:10:48

What's the hold-up?

In a sample of 148 papers from high-profile journals, those from scientists outside the US took longer to be accepted for publication



Data: Time-to-acceptance for papers involving "induced pluripotent stem cells" – an exciting alternative to embryonic stem cells which later won their discoverer a Nobel prize.

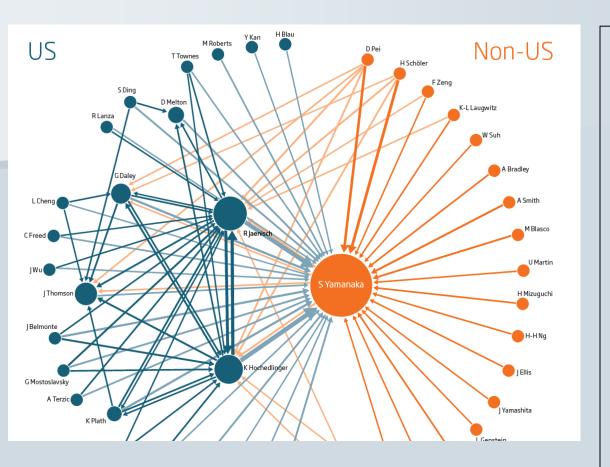
Data preparation and analysis:

Searches and downloads from Web of Science database. Annotation of spreadsheet to document time of receipt, acceptance and publication, and location of primary author. Statistical and graphical analysis of time-to-acceptance data. See methods.

Findings: Papers from corresponding authors outside the US took significantly longer to be accepted for publication



NewScientist





Data: Citations between primary authors of papers on iPS cells.

Data preparation and analysis: Citation information extracted from Web of Science using academic bibliometric software; database queries to link the citations by primary author. Manual checking for errors. Then analysis of network graph. See methods.

Findings: The citation network graph maps influence and connections in the field. Does this help explain why non-US scientists seem to be losing the race to publish?



Data: Metadata for 34,000+ papers published in PNAS from 2004-2013, plus citation counts, scraped from the journal's website.

Findings: Few academy members "contribute" papers at close to the maximum rate, but this group includes several members of the journal's editorial board. Contributed papers are cited less often than those reviewed in the normal way although the gap has narrowed in recent years.



The inside track

Members of the US National Academy of Sciences have long enjoyed a privileged path to publication in the body's prominent house journal. Meet the scientists who use it most heavily.

BY PETER ALDHOUS

n April, the US National Academy of Sciences elected 105 new members to its ranks. Academy membership is one the most prestigious honours for a scientist, and it comes with a tangible perk: members can submit up to four papers per year to the body's high-profile journal, the venerable Proceedings of the National Academy of Sciences (PNAS), through the 'contributed' publication track. This unusual process allows authors to choose who will review their paper and how to respond to those reviewers' com-

For many academy members, this privileged path is central to the appeal of PNAS. But to some scientists, it gives the journal the appearance of an old boys' club. "Sound anachronistic? It is," wrote biochemist Steve Caplan of the University of Nebraska, Omaha, in a 2011 blog post that suggested the contributed track could be used as a "dumping ground" for certain

papers. Editors at PNAS have strived to dispel that perception.

With PNAS currently celebrating its centenary, the news team at Nature decided to examine the contributed track, both to assess its scientific impact and to see which members use it most heavily and why. After analysing a decade's worth of PNAS papers, we found that only a small number of scientists have used the track at close to the maximum allowable rate. The group includes some of the biggest names in science, and six of them sit on the editorial board at the journal. These scientists say the main motivator for using the contributed track is an intense frustration with the peer-review process at other high-profile journals, which they argue has become excessive and labori-

Our analysis also suggests that the efforts by PNAS to prevent abuse of the contributed track and to boost the quality of papers published by this route are bearing fruit. Although contributed PNAS papers attract fewer citations than those handled through the journal's standard review process, the gap has narrowed in recent years. "We have worked really hard at this," says Alan Fersht, a biophysicist at the University of Cambridge, UK, one of PNAS's associate editors and a heavy user of the contributed track.

A PRIVILEGE TO PUBLISH

An inside track to publication for academy members rests deep in PNAS's DNA. The journal was established in 1914 with the explicit goal of publishing members' "more important contributions to research" in addition to "work that appears to a member to be of particular importance". That remit led to the creation of two publishing tracks: contributed and 'communicated' papers -- manuscripts sent by non-members to colleagues in the academy, who shepherded them through review.

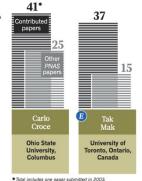
Who are the power users?

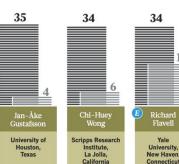
Just 13 members of the US National Academy of Sciences consistently published three or more papers per year in the 'contributed track' at PNAS during the past decade. 'Other' papers include direct submissions, reviewed in the normal way, and papers contributed or communicated by other



Nobel prizewinner Member of PNAS

Former member of editorial board





Johns Hopkins University, University, Baltimore,

Maryland



Data preparation and analysis: After the web scraping, extensive data cleaning to remove variants of authors' names, giving one name format for each academy member. Database queries to count papers of different types from each academy member. Statistical analysis to analyse citation rates of different classes of paper. See methods.

TABLE 1

		Contributed papers				
Period	Effect size	95% confidence interval	P	Effect size	95% confidence interval	P
2004–11	-4.43%	-5.39% to -3.47%	<0.001	-0.53%	-1.70% to 0.65%	0.37
2004	-6.60%	-9.39% to -3.72%	<0.001	-3.45%	-6.37% to -0.44%	<0.05
2005	-7.39%	-10.01% to -4.70%	<0.001	-1.02%	-3.94% to 1.99%	0.5
2006	-7.01%	-9.63% to -4.32%	<0.001	-1.42%	-4.23% to 1.47%	0.33
2007	-5.81%	-8.52% to -3.01%	<0.001	-3.67%	-6.48% to -0.78%	<0.05
2008	-3.41%	-6.14% to -0.60%	<0.05	-0.12%	-3.25% to 3.11%	0.94
2009	-1.89%	-4.65% to 0.94%	0.19	1.53%	-1.85% to 5.03%	0.38
2010	-3.48%	-6.06% to -0.83%	<0.05	-2.50%	-6.58% to 1.76%	0.25
2011	-1.14%	-3.70% to 1.48%	0.39	N/A	N/A	N/A

Beware running with scissors: Seek expert help if you need rigorous statistical analysis!

DIY statistical analysis: experience the thrill of touching real data

The story of one man's efforts to re-analyse the stats behind a BBC report on bowel cancer is a heartwarmingly nerdy one





Ben Goldacre guardian.co.uk, Friday 28 October 2011 17.31 EDT Comments (60)

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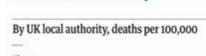
Media

BBC

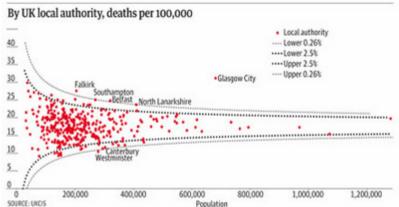
Series Bad science

Related

18 Apr 2007 Aspirin linked to lower risk of bowel and prostate cancer



Bowel cancer mortality



A funnel plot of bowel cancer mortality rates in different areas of the UK

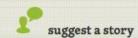
The BBC has found a story: "Threefold variation' in UK bowel cancer rates". The average death rate across the UK from bowel cancer is 17.9 per 100,000 people, but in some places it's as low as 9, and in some places it's as high as 30. What can be causing this?

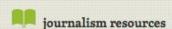
Journalists tend to find imaginary patterns in statistical noise, which we've covered many times before. But this case is particularly silly, as you will see, and it has a heartwarming, nerdy twist.

Even Nate Silver can get his fingers burned ...

Tracker

Peer review within science journalism









Nate Silver's new FiveThirtyEight dishes out statistical nonsense on health coverage.



Author: Paul Raeburn



Share



Nate Silver's fivethirtyeight.com relaunched yesterday at its new home--ESPN--vowing to focus its coverage on five areas: politics, economics, life, sports--and science.

The inclusion of science was a surprise to me. And possibly a mistake, unless FiveThirtyEight can quickly improve the quality of the "science" it's publishing. The lead story on the relaunched site's first day--"Finally, a Formula for Decoding Health News"--was abysmal.

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... and be forced to backtrack

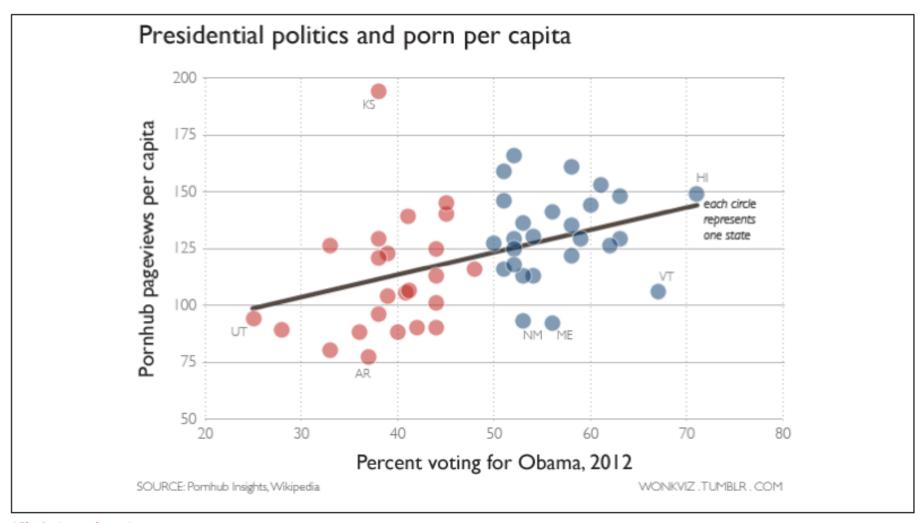


FiveThirtyEight relaunched less than two weeks ago. It's been a heck of a learning experience. When you're trying something new, it's going to take some time to get everything right, and you're going to get criticism from all quarters.



DISTRUST YOUR DATA

Jacob Harris on Six Ways to Make Mistakes with Data



(Chris Ingraham)

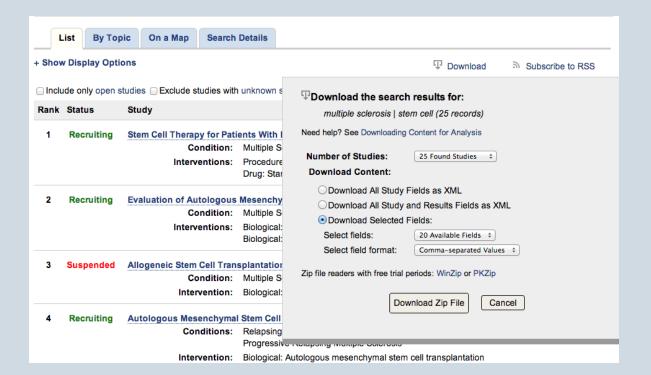
Read this before committing an act of data journalism!

Why aren't we seeing more data-driven science journalism?

Some basic data resources for science reporters

Using web search forms

- Look for the advanced search page, which will offer options to customize your search.
- Read the Help or FAQs to learn how the search works. Does it use Boolean logic (AND, OR, NOT)? Do quote marks allow you to search for a specific phrase? Is there a wildcard character, such as * or %, that allows you to look for variations on a search term?
- Look for download options once you've found the data you need:



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Information on more than 187,000 clinical trials in US and beyond

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http://www.isrctn.com/

Information on almost 13,000 registered clinical trials

WHO International Clinical Trials Registry Platform

http://apps.who.int/trialsearch/default.aspx

Investigating science

(with help from data)

Peter Aldhous, Science reporter, BuzzFeed News

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