

Thinking About Data

MMED

African Institute for the Mathematical Sciences Muizenberg, South Africa May, 2019

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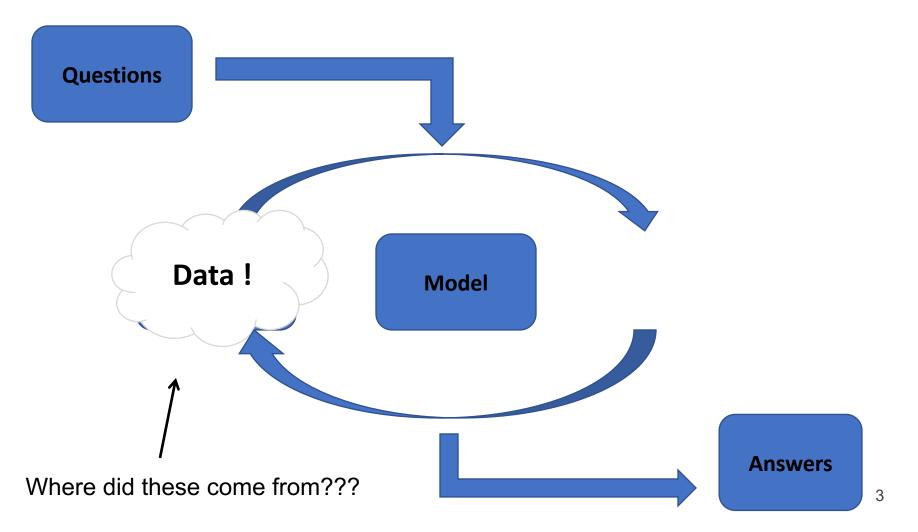
Slide set citation: https://doi.org/10.6084/m9.figshare.5044615.v4

Goals:

- Recognize that data are better than anecdotes
- Understand that how the data are collected matters
- Be aware of variability don't be fooled by it
- Consider confounding and other forms of bias
- Healthy skepticism



Why do we care about data?







"Data, data, data!", he cried impatiently. "I can't make bricks without clay"

- Sherlock Holmes

Source: Statistics 3rd, ed. Pisani Purves, Freedman

The New york Times



February 3, 2010

Journal Retracts 1998 Paper Linking Autism to Vaccines

By GARDINER HARRIS

A prominent British medical journal on Tuesday retracted a 1998 research paper that set off a sharp decline in <u>vaccinations</u> in <u>Britain</u> after the paper's lead author suggested that vaccines could cause <u>autism</u>.

The retraction by <u>The Lancet</u> is part of a reassessment that has lasted for years of the scientific methods and financial conflicts of Dr. Andrew Wakefield, who contended that his research showed that the combined <u>measles</u>, <u>mumps</u> and <u>rubella</u> vaccine may be unsafe.



"The story became credible because it was published in The Lancet," Alison Singer, president of the Autism Science Foundation, said Tuesday. "It was in The Lancet, and we really rely on these medical journals."

THE LANCET • Vol 351 • February 28, 19

Consequences

National MMR vaccination catch-up programme announced in response to increase in measles cases



Commons.wikimedia.org

A national catch-up programme to increase MMR vaccination uptake in children and teenagers is announced today by Public Health England, NHS England and the Department of Health.

Experts believe the rise in measles cases can be mostly attributed to the proportion of unprotected 10-16 year-olds who missed out on vaccination in the late 1990s and early 2000s when concern around the discredited link between autism and the vaccine was widespread. At this time measles had been eliminated in the UK, but coverage fell nationally to less than 80% in 2005, with even lower uptake in some parts of the country. After many years of low vaccination uptake, measles became re-established in 2007.

https://www.gov.uk/government/news/national-mmr-vaccination-catch-up-programme-announced-in-response-to-increase-in-measles-case9



How the Data are Collected Matters

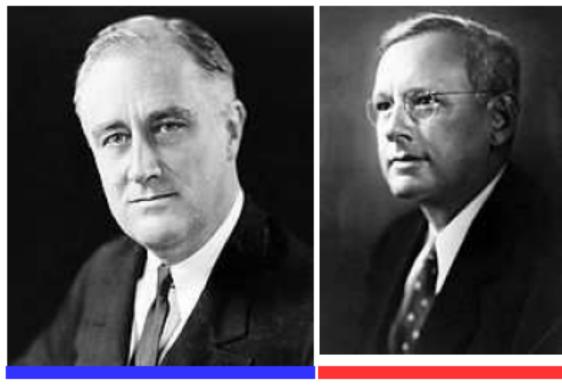
 "Always do right. This will gratify some people, and astonish the rest"

- Mark Twain

Beware: All data are not created equal



1936 Presidential Election



Franklin D. Roosevelt

Alf Landon

1936 Presidential Election

- 1936 Literary Digest Poll
- Literary Digest had predicted the winner of every US presidential election since 1916.
- In 1936, Literary Digest mailed questionnaires to 10 million people (25% of voters).
- 2.4 million people responded
- Returned questionnaires:

Landon: 1,293,668 57%

• FDR: 972,897 43%

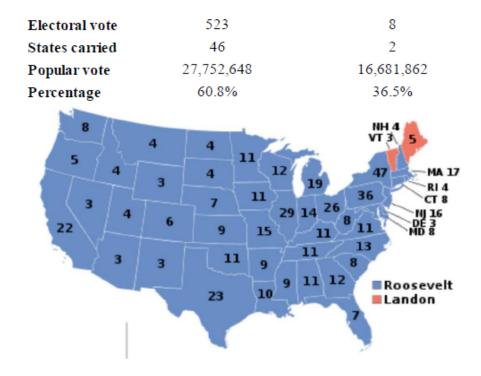
Source: http://historymatters.gmu.edu/d/5168/

(ICI3D)

Results

Actual Result: Roosevelt 61%, Landon 37%.

One of the biggest landslides in U.S. history



What went wrong?

- How were the data collected?
 - Those who received the questionnaire were systematically different than those who didn't
 - 10 million sent out (~25% of voters)
 - 2.3 million returned sample of convenience
 - Not representative
 - Sampling frame:
 - Telephone books
 - Automobile registries

Bias and Variability

 Sample size doesn't matter if the data collection scheme is flawed

Observed value = The TRUTH + Chance Error

Another example?

- IBM created Watson
 - Uses data to make predictions finds patterns in knowledge database
- Trained Watson to play Jeopardy
- Crushed all humans
- Triumph for A.I.
 -mostly.....



U.S. Cities



Its largest airport is named for a World War II hero; its second largest, for a World War II battle

Watson: What is..... Toronto(?!)



Oops

Big Data Limitations



IN CS, IT CAN BE HARD TO EXPLAIN THE DIFFERENCE BETWEEN THE EASY AND THE VIRTUALLY IMPOSSIBLE.

Variability

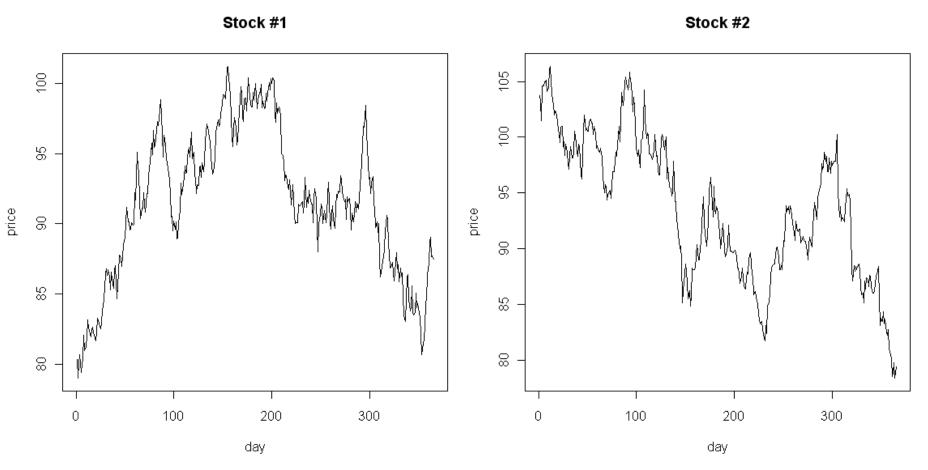
"When the facts change, I change my mind.
 What do you do sir?"

- John Maynard Keynes

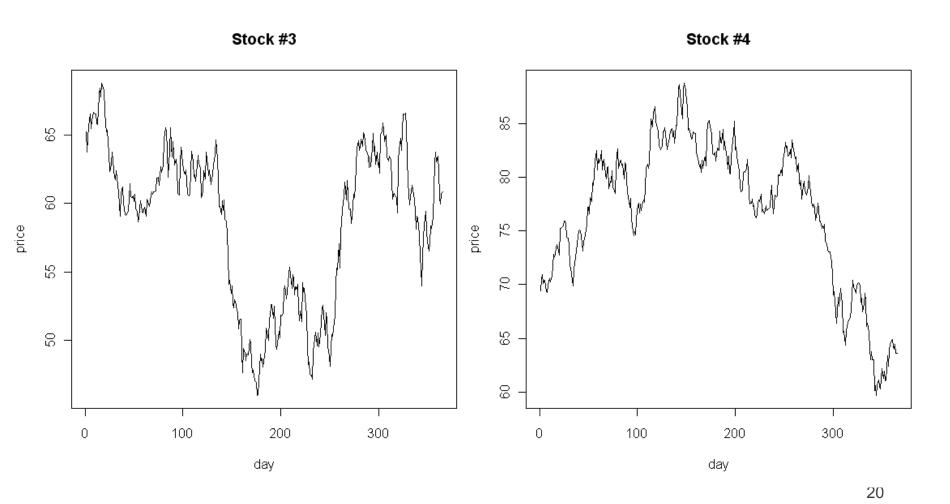
Variation is everywhere

Observed value = Truth + Bias + Random Error

Time Series of four "stock prices"

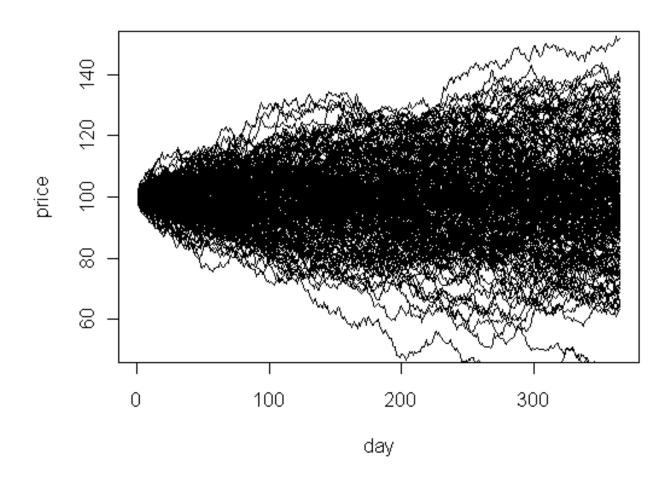


Time series of four "stock prices"

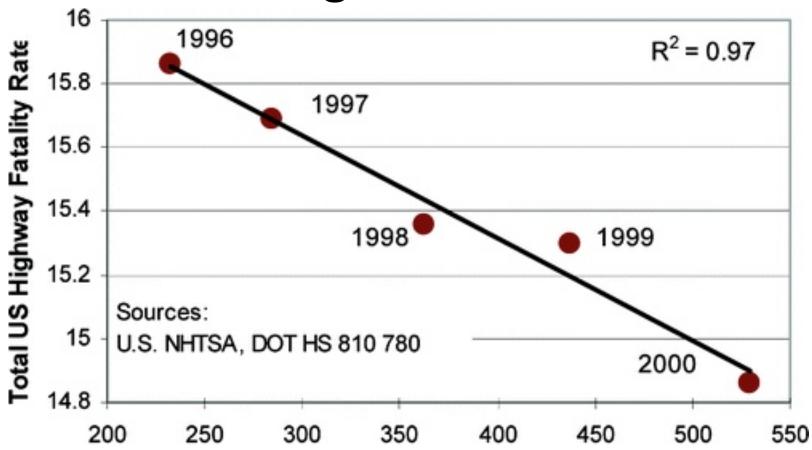


Don't be fooled by randomness

1000 Random Stocks



Confounding



Per Capita Expenditures on Road Maintenance(?)

What is the conclusion?

January 2009
Percent of Planes Delayed from City of Origin

Continental				United		
<u>Airport</u>	<u>Late</u>	<u>Total</u>	<u>%</u>	<u>Late</u>	<u>Total</u>	<u>%</u>
Newark	957	3998	23.9	100	399	25.1
LaGuardia	62	356	<i>17.4</i>	113	573	19.7
Pittsburg	8	60	<i>13.3</i>	17	119	14.3
Detroit	16	145	11.0	16	139	11.5
Totals	1043	4559	22.9	246	1230	<u>20.0</u>

slide credit: Jeff Witmer, data source: www.bts.gov



How about now?

. logistic delay continental

Logistic regre	ssion			Number LR chi2		; = =	5789 4.72
				Prob >	chi2	=	0.0298
Log likelihood	= -3067.3063	3		Pseudo	R2	=	0.0008
delay		Std. Err.	z	P> z	[95%	Conf.	Interval]
continental		.0943644	2.15	0.031	1.015	318	1.38672

Unadjusted OR = 1.19 95% CI = (1.02, 1.39)

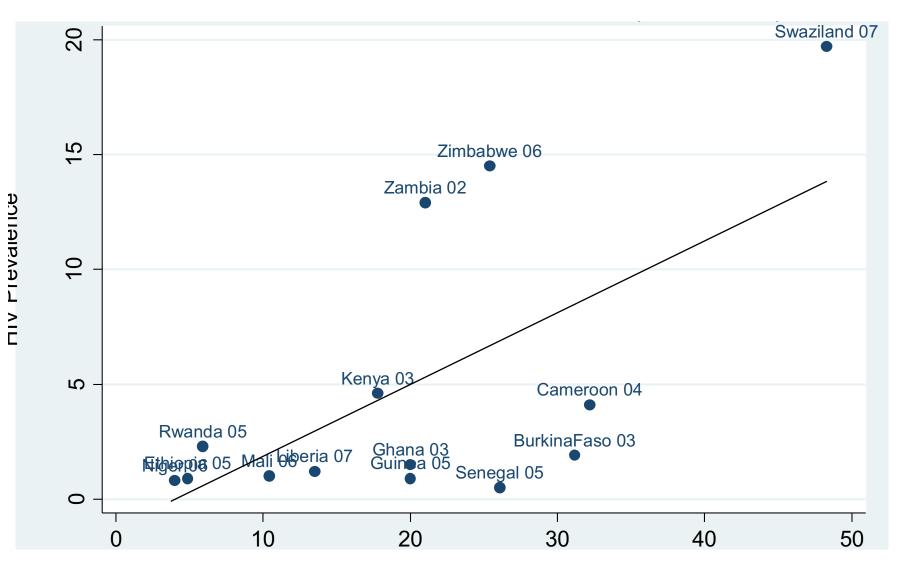
. logistic delay continental laguardia newark pittsburg

Logistic regression	Number of obs	=	5789
	LR chi2(4)	=	46.30
	Prob > chi2	=	0.0000
Log likelihood = -3046.5183	Pseudo R2	=	0.0075

delay	 Odds Ratio +	Std. Err.	z	P> z	[95% Conf.	Interval]
continental		.0859693	-0.93	0.351	.7622593	1.101156
laguardia	.9161694	.3722533	2.88	0.004	1.207464	2.706607
newark	1.807797	.5031265	4.87	0.000	1.762527	3.783036
pittsburg	2.58219	.360524	0.80	0.421	.7183302	2.20692

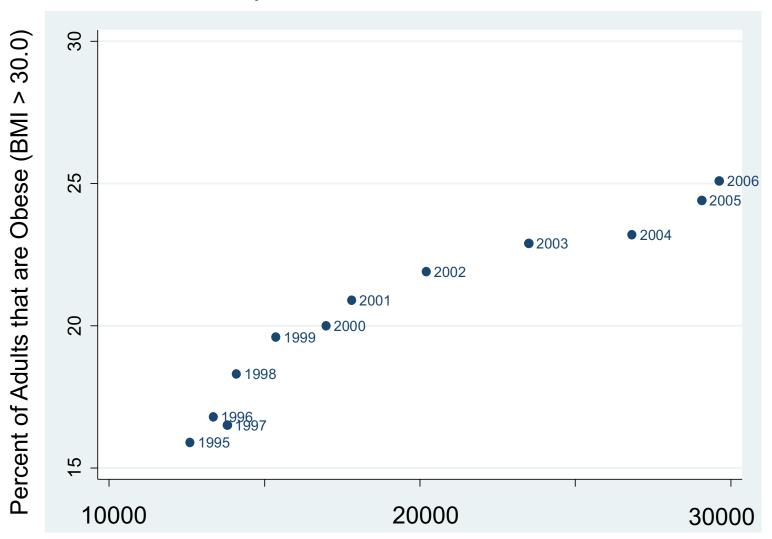
Adjusted OR = 0.92 95% CI = (0.76, 1.10)







Obesity in the United States: 1995 - 2006



Number of Gyms in the U.S.

What do the data say?

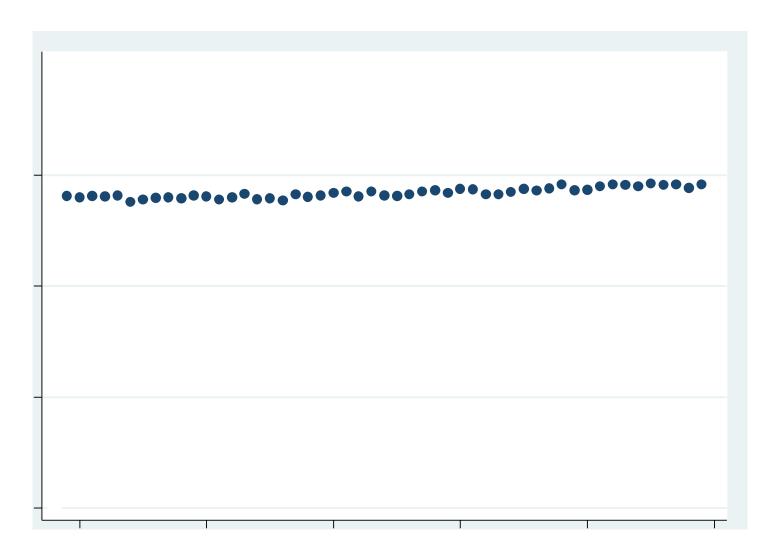
• Caution:

This may require thoughtful consideration



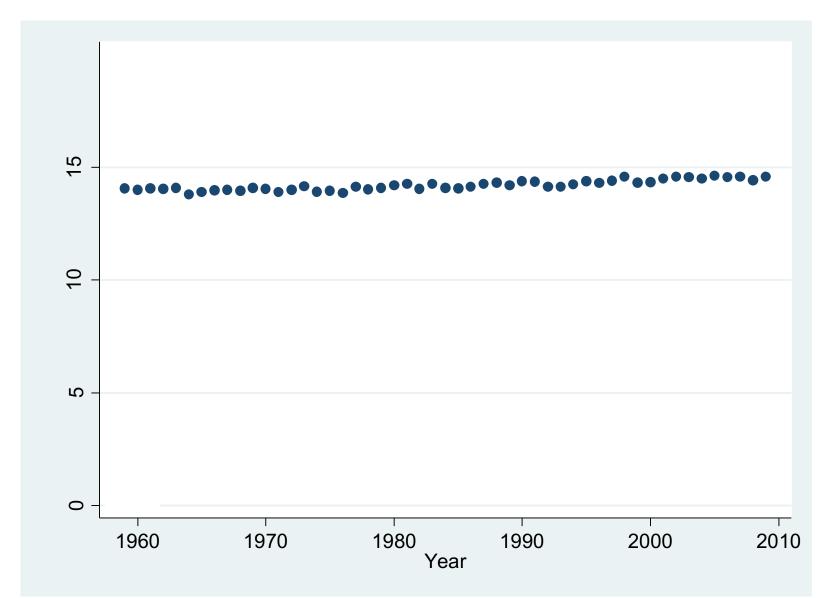


What is the relationship?



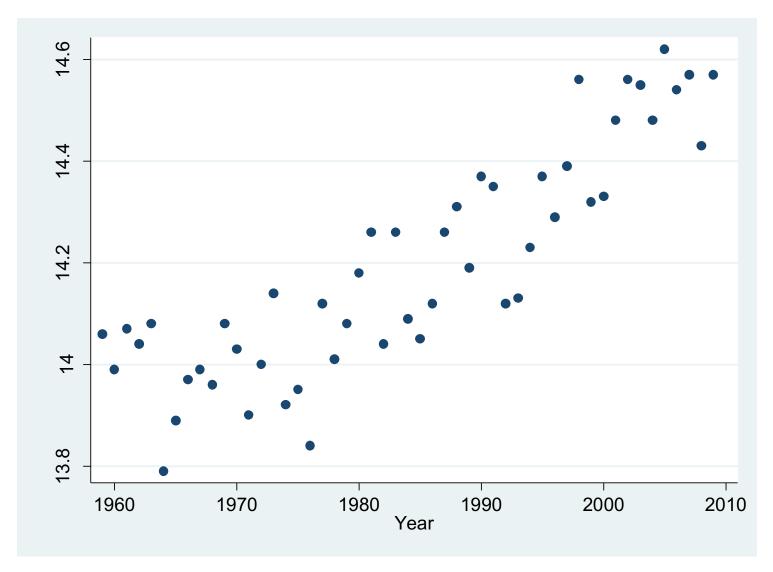


What is the relationship?

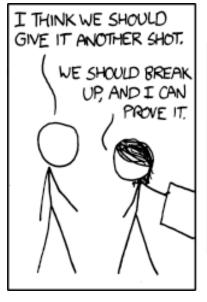


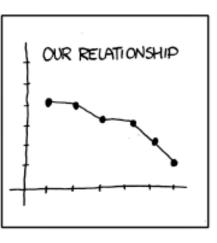


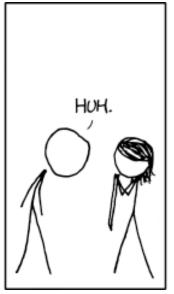
Median Global Temperature During the Past 50 Years

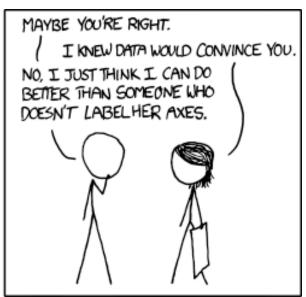


Love statistics(?)





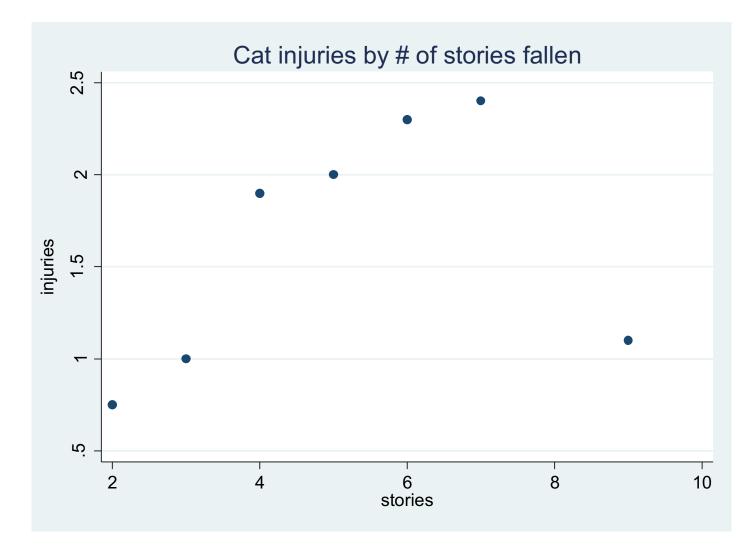








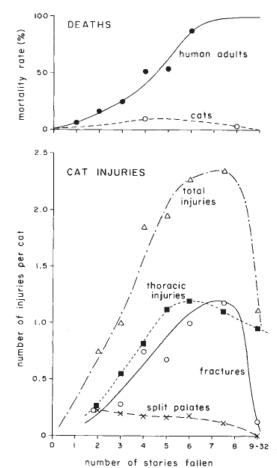
Are you a believer?



Cat conundrum

As long as it experiences acceleration, the cat probably extends its limbs reflexly, but on reaching terminal velocity it may relax and extend the limbs more horizontally in flying-squirrel fashion, thus not only reducing the velocity of fall but also absorbing the impact over a greater area of its body. This may explain the paradoxical decrease of mortality and injury in cats that fall more than 100 feet.

	Stories Fallen	# of cats
_	1	0
	2	8
	3	14
	4	27
	5	34
	6	21
	7-8	9
	9-32	13



Mortality rates for falling adult humans and cats (above), and number of total injuries and various types of injury per falling cat (below), as a function of number of stories fallen. (Based on the work by Waring and Demling and by Whitney and Mehlhoff.)

What do you think?

ORIGINAL INVESTIGATION

A Randomized, Controlled Trial of the Effects of Remote, Intercessory Prayer on Outcomes in Patients Admitted to the Coronary Care Unit

Arch Intern Med. 1999:159:2273-2278

Table 4. Effects of Intercessory Prayer on Mid America Heart Institute–Cardiac Care Unit (MAHI-CCU) Scores and Length of Stay in the CCU and in the Hospital*

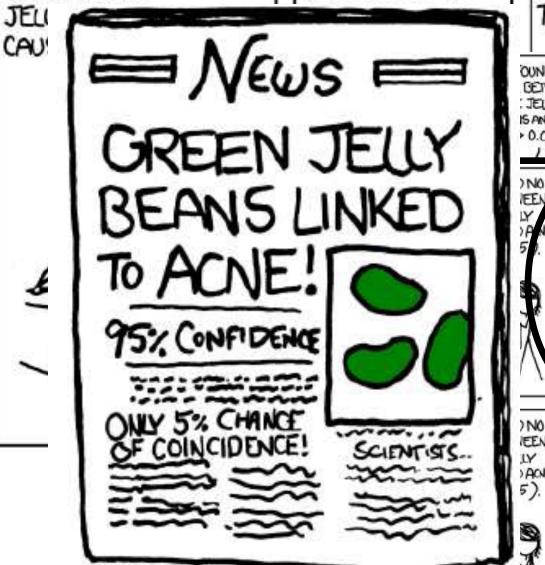
	Mean	± SEM		
	Usual Care Group (n = 52)	Prayer Group (n = 466)	Percentage Change	P
MAHI-CCU score	7.13 ± 0.27	6.35 ± 0.26	-11	.04
Unweighted MAHI-CCU score†	3.00 ± 0.10	2.70 ± 0.10	-10	.04
Length of CCU stay, d‡		1.12 ± 0.08	-9	.28
Length of hospital stay, d‡	5.97 ± 0.29	6.48 ± 0.54	+9	.41

Table 3. Effects of Intercessory Prayer on Individual Components of the Mid America Heart Institute—Cardiac Care Unit (MAHI-CCU) Score*

	No. (%) o		
MAHI-CCU Score Component	Usual Care Group (n = 524)	Prayer Group (n = 466)	P
Antianginal agents	59 (11.3)	47 (10.1)	.62
Antibiotics	82 (15.6)	77 (16.5)	.77
Unstable angina	4 (0.8)	1 (0.2)	.3
Arterial monitor	42 (8.0)	32 (6.9)	.5
Catheterization	180 (34.4)	162 (34.8)	.9
Antiarrhythmics	56 (10.7)	50 (10.7)	.9
Inotropes	76 (14.5)	69 (14.8)	.9
Vasodilation	78 (14.9)	59 (12.7)	.3
Diuretics	112 (21.4)	97 (20.8)	.8
Pneumonia	10 (1.9)	12 (2.6)	.6
Atrial fibrillation	17 (3.2)	12 (2.6)	.6
Supraventricular tachycardia	6 (1.1)	2 (0.4)	.2
Hypotension	7 (1.3)	8 (1.7)	.8
Anemia/transfusion	66 (12.6)	50 (10.7)	.4
Temporary pacer	16 (3.0)	13 (2.8)	.9
Third-degree heart block	1 (0.2)	2 (0.4)	.6
Readmit to cardiac care unit	22 (4.2)	25 (5.4)	.4
Swan-Ganz catheter	172 (32.8)	123 (26.4)	.0
Implanted cardiac defibrillator	6 (1.1)	10 (2.1)	.3
Electrophysiology study	15 (2.9)	10 (2.1)	.6
Radiofrequency ablation	8 (1.5)	2 (0.4)	.1
Extension of infarct	2 (0.4)	0 (0.0)	.5
Gastrointestinal bleed	12 (2.3)	5 (1.1)	.2
Interventional coronary procedure	155 (29.6)	121 (26.0)	.2
PTCA alone	69 (13.2)	62 (13.3)	.9
PTCA with stent and/or rotablator	86 (16.4)	59 (12.7)	.1
Permanent pacer	21 (4.0)	12 (2.6)	.2
Congestive heart failure	17 (3.2)	19 (4.1)	.6
Ventricular fibrillation/tachycardia	12 (2.3)	10 (2.1)	.9
Intra-aortic balloon pump	20 (3.8)	12 (2.6)	.3
Major surgery	76 (14.5)	51 (10.9)	.1
Sepsis	7 (1.3)	7 (1.5)	.9
Intubation/ventilation	27 (5.2)	26 (5.6)	.8
Cardiac arrest	6 (1.1)	5 (1.1)	.8
Death	46 (8.8)	42 (9.0)	.9



ICI3D



THAT SETTLES THAT.

OUND NO BETWEEN JELLY IS AND ACKE 0.05)

WE FOUND NO LINK BETWEEN TEAL JELLY BEANS AND ACNE (P>0.05)

R

WE FOUND A LINK BETVEEN GREEN JELLY BEANS AND ACNE P < 0.05)



WE FOUND NO

LINK BETWEEN

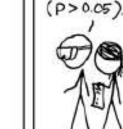
BEANS AND ACKE

PEACH JELLY

WE FOUND NO INK BETWEEN AUVE JELLY TANS AND ACNE > 0.05)



)NO EEN LY ACKE



WE FOUND NO LINK BETWEEN ORANGE JELLY BEANS AND ACNE (P>0.05)



xkcd

Skepticism vs Openness

- "It seems to me what is called for is an exquisite balance between two conflicting needs: the most skeptical scrutiny of all hypotheses that are served up to us and at the same time a great openness to new ideas ...
- If you are only skeptical, then no new ideas make it through to you ...
- On the other hand, if you are open to the point of gullibility and have not an ounce of skeptical sense in you, then you cannot distinguish the useful ideas from the worthless ones."

Summary

- We use data to help form and revise models
- Data collection should not be haphazard
- Data are not easily obtained especially good data
- Think about how the data were collected and be sure to consider:
 - Bias and confounding
 - Variability
 - How the data are presented
- Be skeptical but open





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Title: Thinking About Data. DOI: 10.6084/m9.figshare.5044615.v4

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Clinic on the Meaningful Modeling of Epidemiological Data

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