Numbersense: Clearing the Fog of Big Data

Kaiser Fung

INFORMS NYC Luncheon

9/18/2013

Big Data studies

- Observational data
- Co-opted
- Seemingly exhaustive N
- Fused data
- No controls

Flight Delay: the data

- U.S. domestic commercial flights
- 1987 to 2008
- 123 million records
- 29 variables

Wicklin (2009)

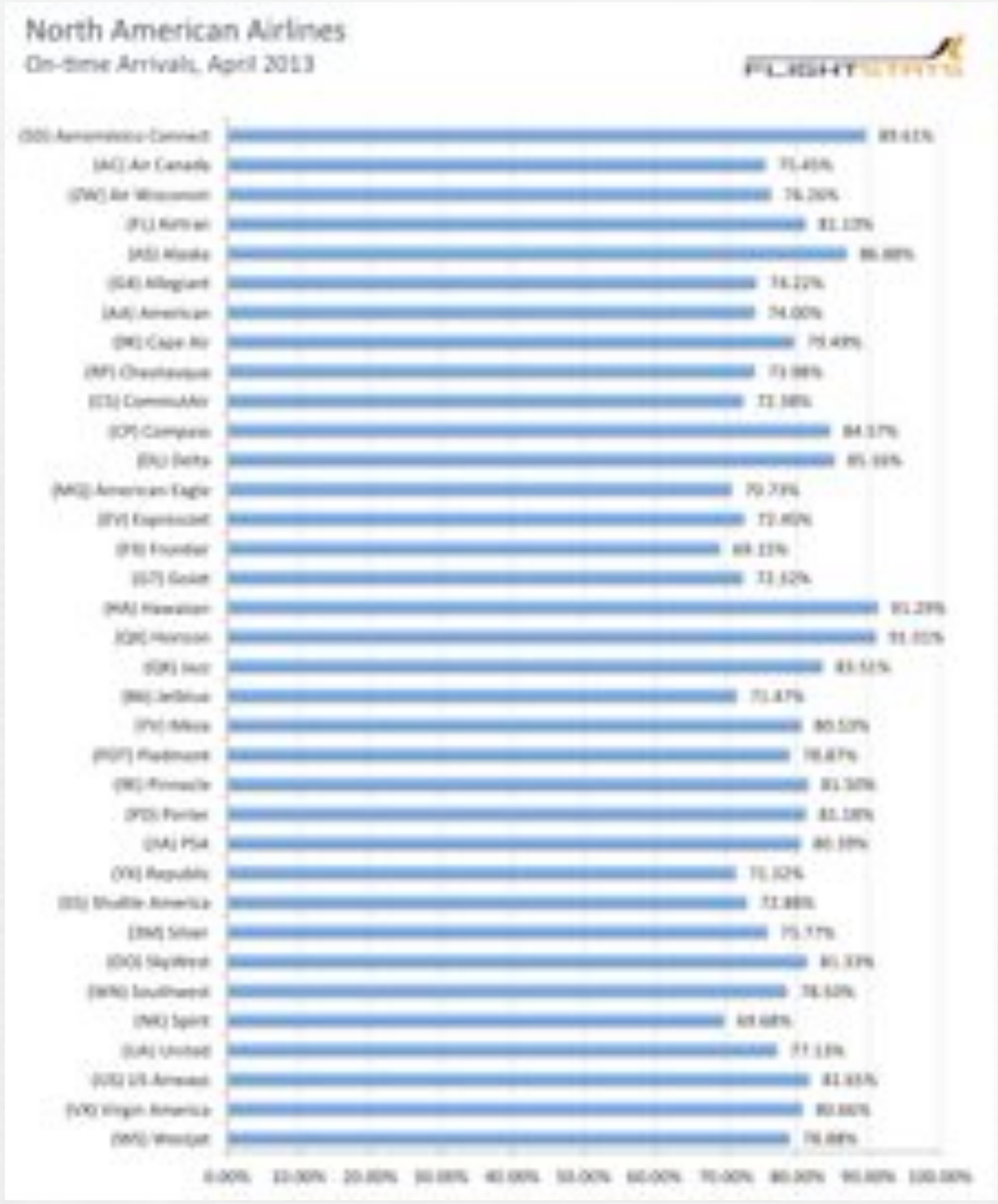
Which airline had a lower delay rate?

		Los Angeles	Phoenix	San Diego	San Francisco	Seattle
ALASKA	on time	497	221	212	503	1,841
	delayed	62	12	20	102	305
AM WEST	on time	694	4,840	383	320	201
	delayed	117	415	65	129	61

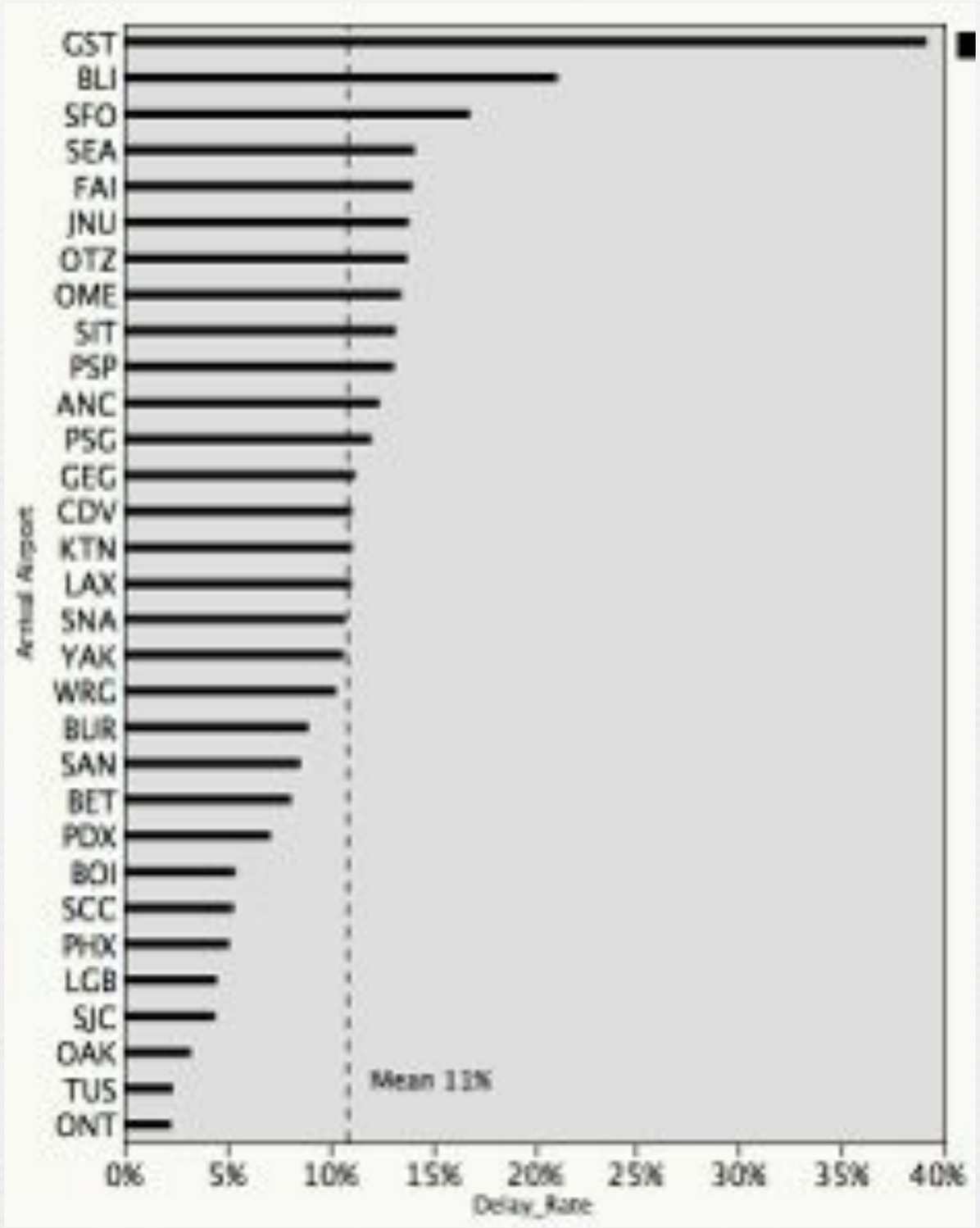
Which airline had a lower delay rate?

		Los Angeles	Phoenix	San Diego	San Francisco	Seattle	All 5 Airports	% delay
ALASKA	on time	497	221	212	503	1,841	3,274	13.3%
	delayed	62	12	20	102	305	501	
ALASKA	delay %	11.1%	5.4%	8.6%	16.9%	14.2%		
AM WEST	on time	694	4,840	383	320	201	6,438	10.9%
	delayed	117	415	65	129	61	787	
AM WEST	delay %	14.4%	7.9%	14.5%	28.7%	23.3%		

Ask the “right” question



Alaska's On-time Performance



10.9%

Alaska is the industry leader in on-time flights

Flight Delay: the data

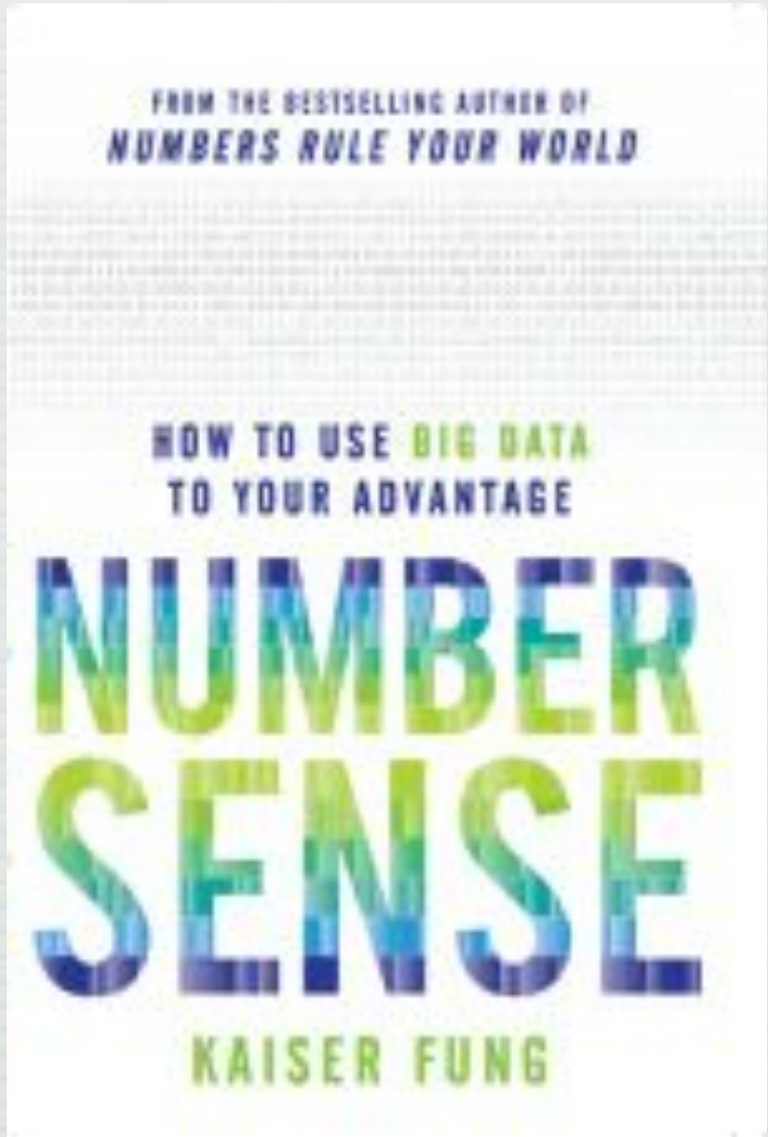
- U.S. domestic commercial flights
- 1987 to 2008
- 123 million records
- 29 variables

In which variable(s) does Simpson's paradox lurk?

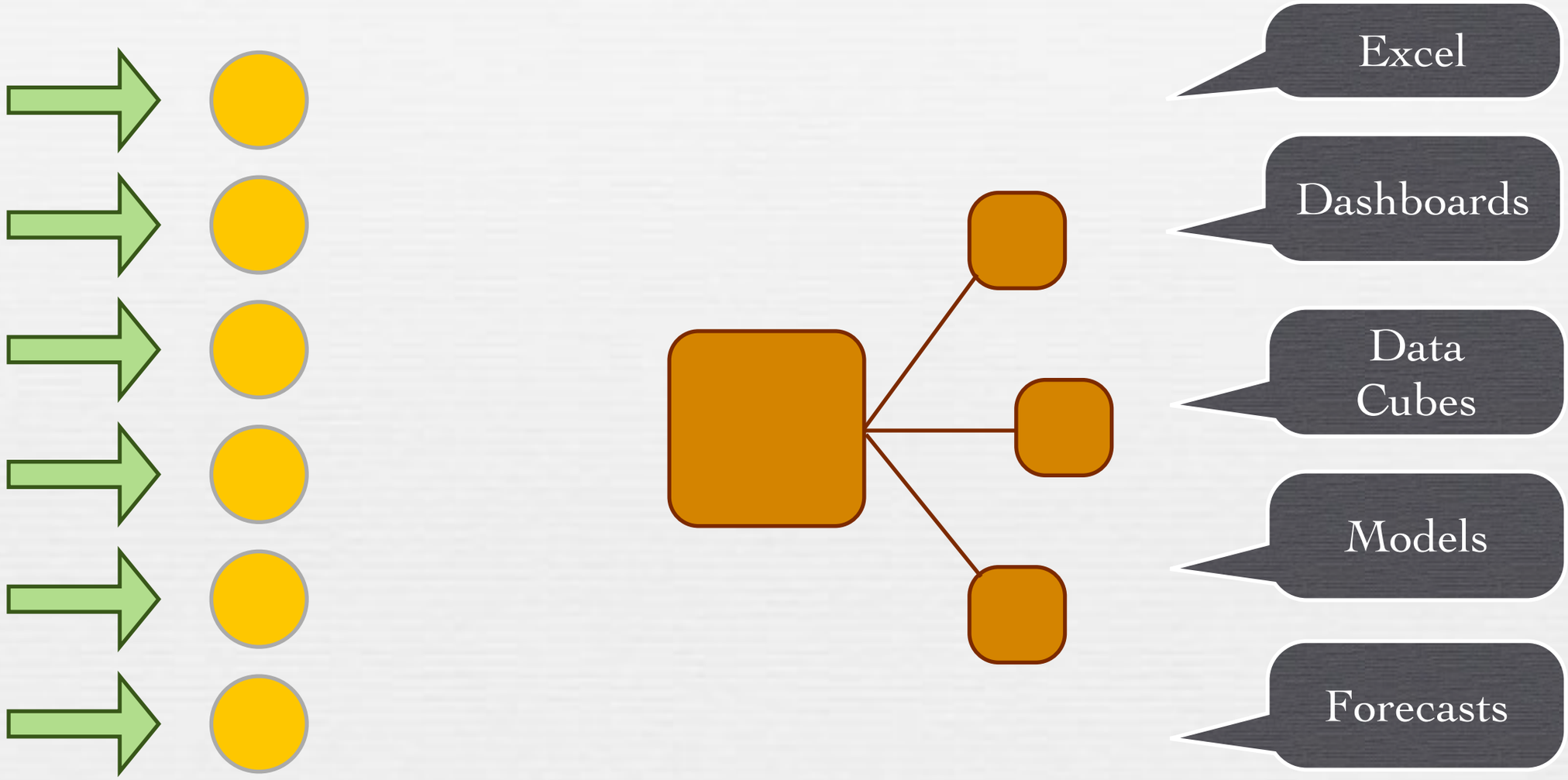
A priori?

Wicklin (2009)

When more people are performing more analyses more quickly, there are more theories, more points of view, more complexity, more conflicts and more confusion. There is less clarity, less consensus and less confidence.



Big Data: Producers



user interactions

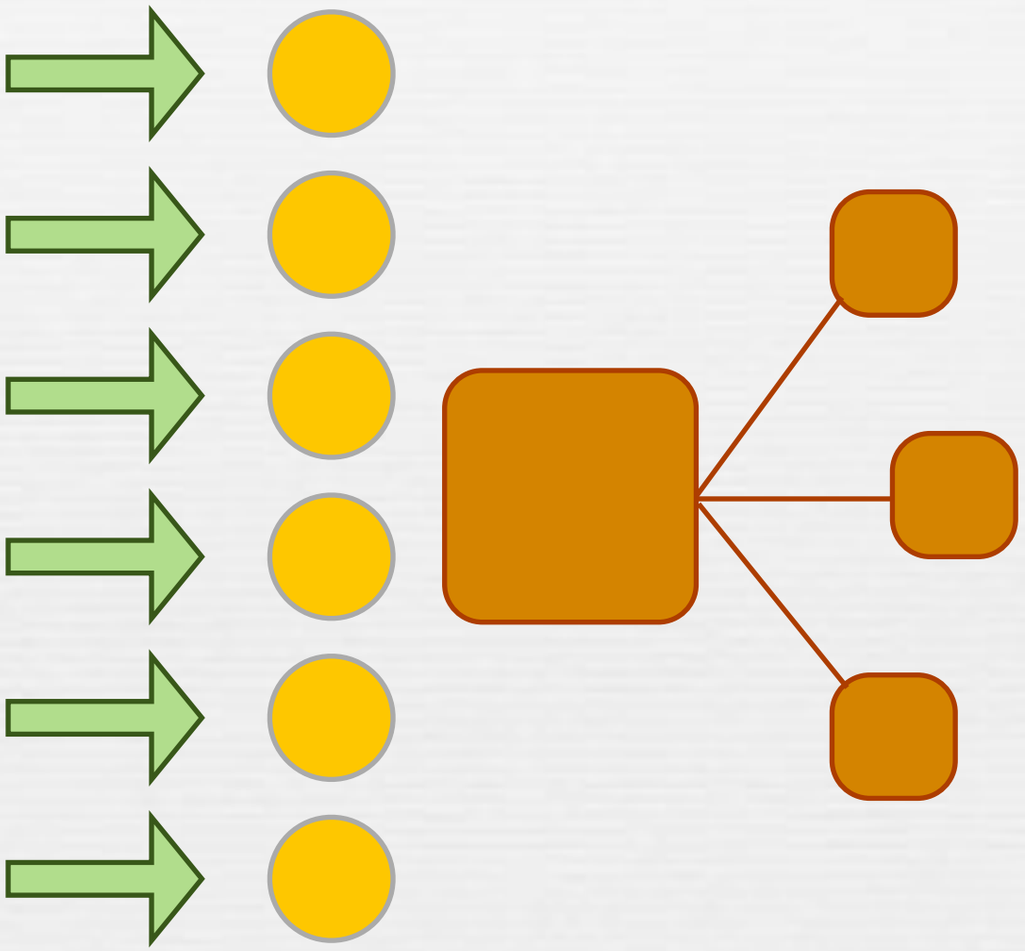
web logs: distributed servers

data ware house

data marts

displays

Big Data: Consumers



user interactions

web logs... datamarts

displays

Excel

Dashboards

Data Cubes

Models

Forecasts

Strategies

Tactics

Plans

Moneyball


MLB
Self-populating
INSTITUTION
OLD SCHOOL
SCOUTS
HUMANS
gut feelings
OUTCOME
ANTI-INTELLECTUALISM
ILLUSIONS
MAGIC
INSIDERS
CHANCE
LUCK
CAN HE HIT?
$
NEW YORK YANKEES, ETC.
DISPARITY
OAKLAND ATHLETICS, ETC.
COMPUTER
NEW SCHOOL
MATH
WHEN NUMBERS ACQUIRE THE SIGNIFICANCE OF LANGUAGE
STATS
BILL JAMES
REASON
SCIENCE
PROCESS
OUTSIDERS
SCIENTIFIC METHOD

FROM THE BESTSELLING AUTHOR OF
NUMBERS RULE YOUR WORLD

HOW TO USE BIG DATA
TO YOUR ADVANTAGE

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Statistics != Math

Brief Contents

David S. Moore, The Basic Practice of Statistics, ~2007

The Obesity Epidemic

The Obesity Epidemic

Quetelet's Index (1830)

BMI Critics (2000-)

obesity reviews

Viewpoint

Beyond body mass index

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Received 12 January 2001; *accepted* 16 January 2001

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Summary

Body mass index (BMI) is the cornerstone of the current classification system for obesity and its advantages are widely exploited across disciplines ranging from international surveillance to individual patient assessment. However, like all anthropometric measurements, it is only a surrogate measure of body fatness. Obesity is *defined* as an excess accumulation of body fat, and it is the amount of this excess fat that correlates with ill-health. We propose therefore that much greater attention should be paid to the development of databases and standards based on the *direct* measurement of body fat in populations, rather than on surrogate measures. In support of this argument we illustrate a wide range of conditions in which surrogate anthropometric measures (especially BMI) provide misleading information about body fat content. These include: infancy and childhood; ageing; racial differences; athletes; military and civil forces personnel;

BMI Critics (2000-)

obesity reviews

Viewp

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Why are doctors still measuring obesity with the body mass index?

By Jeremy Singer-Vine | Posted Monday, July 20, 2009, at 10:00 AM ET

| Posted Monday, July 20, 2009, at 10:00 AM ET

Slate.com

Beyond BMI

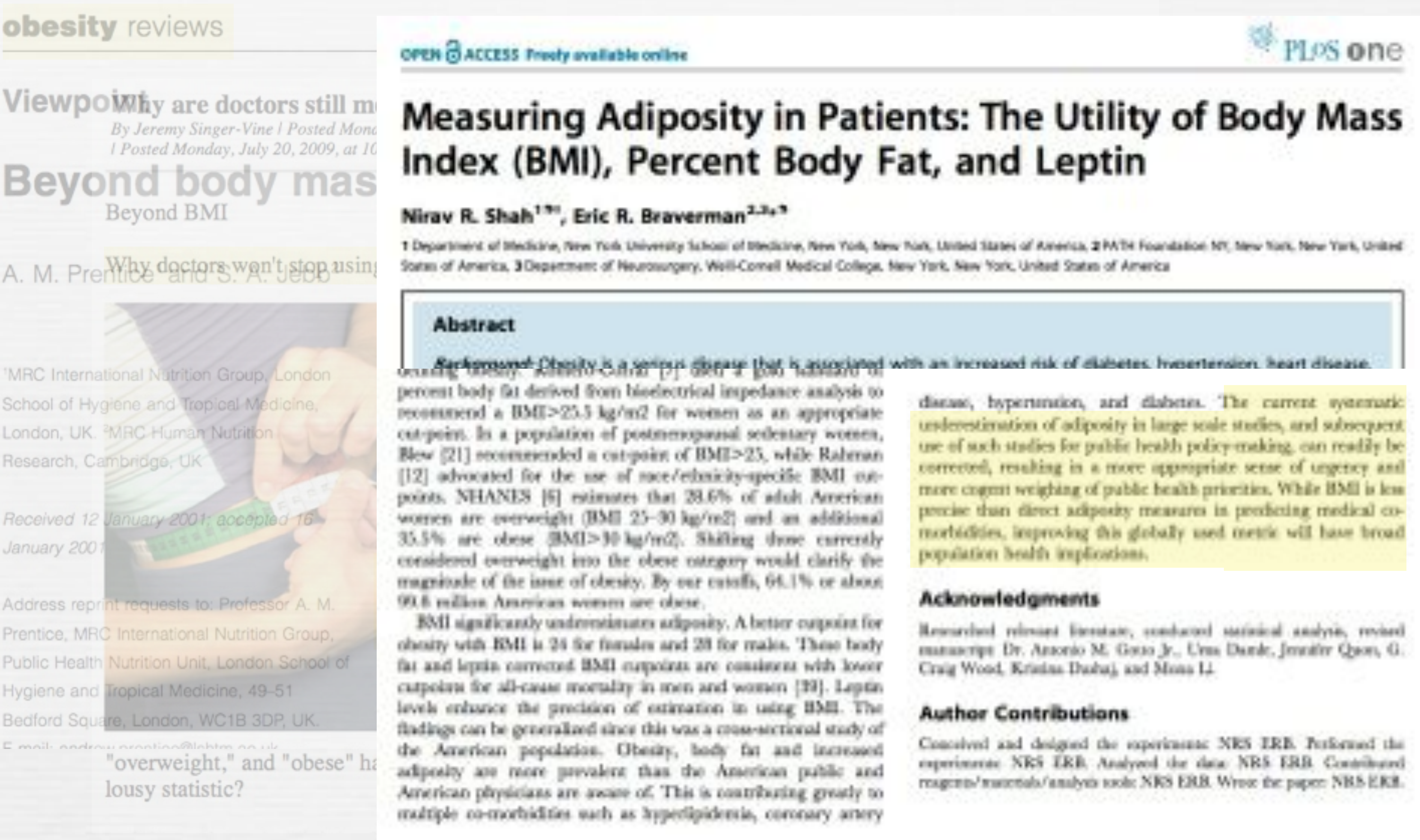
Why doctors won't stop using an outdated measure for obesity.

Why are doctors still measuring obesity with the body mass index?

A few extra pounds can extend your life. Or so chirped the press, reporting on a recent study from the journal Obesity. The new research, which supports earlier findings that being slightly overweight is associated with living longer, has added to an ongoing controversy over how we measure obesity. At the center of this debate is the body mass index, a simple equation (your weight in kilograms divided by the square of your height in meters) that has in the last decade claimed a near-monopoly on obesity statistics. Some researchers now argue that this flawed and overly reductive measure is skewing the results of research in public health.

For years, critics of the body mass index have griped that it fails to distinguish between lean and fatty mass. (Muscular people are often misclassified as overweight or obese.) The measure is mum, too, about the distribution of body fat, which makes a big difference when it comes to health risks. And the BMI cutoffs for "underweight," "normal," "overweight," and "obese" have an undeserved air of mathematical authority. So how did we end up with such a lousy statistic?

BMI Critics (2000-)



Taking eyes off the ball

OPEN ACCESS Freely available online

PLoS one

Measuring Adiposity in Patients: The Utility of Body Mass Index (BMI), Percent Body Fat, and Leptin

Nirav R. Shah[1,2*], **Eric R. Braverman**[2,3*]

1 Department of Medicine, New York University School of Medicine, New York, New York, United States of America, 2 PATH Foundation NY, New York, New York, United States of America, 3 Department of Neurosurgery, Weill-Cornell Medical College, New York, New York, United States of America

Abstract

Background: Obesity is a serious disease that is associated with an increased risk of diabetes, hypertension, heart disease, stroke, and cancer, among other diseases. The United States Centers for Disease Control and Prevention (CDC) estimates a 20% obesity rate in the 50 states, with 12 states having rates of over 30%. Currently, the body mass index (BMI) is most commonly used to determine adiposity. However, BMI presents as an inaccurate obesity classification method that underestimates the epidemic and contributes to failed treatment. In this study, we examine the effectiveness of precise biomarkers and duel-energy x-ray absorptiometry (DXA) to help diagnose and treat obesity.

Methodology/Principal Findings: A cross-sectional study of adults with BMI, DXA, fasting leptin and insulin results were measured from 1998–2009. Of the participants, 63% were females, 37% were males, 75% white, with a mean age = 51.4 (SD = 14.2). Mean BMI was 27.3 (SD = 5.9) and mean percent body fat was 31.3% (SD = 9.3). BMI characterized 26% of the subjects as obese, while DXA indicated that 64% of them were obese. 39% of the subjects were classified as non-obese by

Taking eyes off the ball

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"Although DXA is a direct measurement of fat and a better measure of adiposity than BMI, it is not a disease correlate."

The more things change

All Patients

BMI \ DXA	*Not Obese*	*Obese*	BMI Totals
Not Obese	35 %	39 %	74%
Obese	1 %	25 %	26%
DXA Totals	36 %	64 %	100%

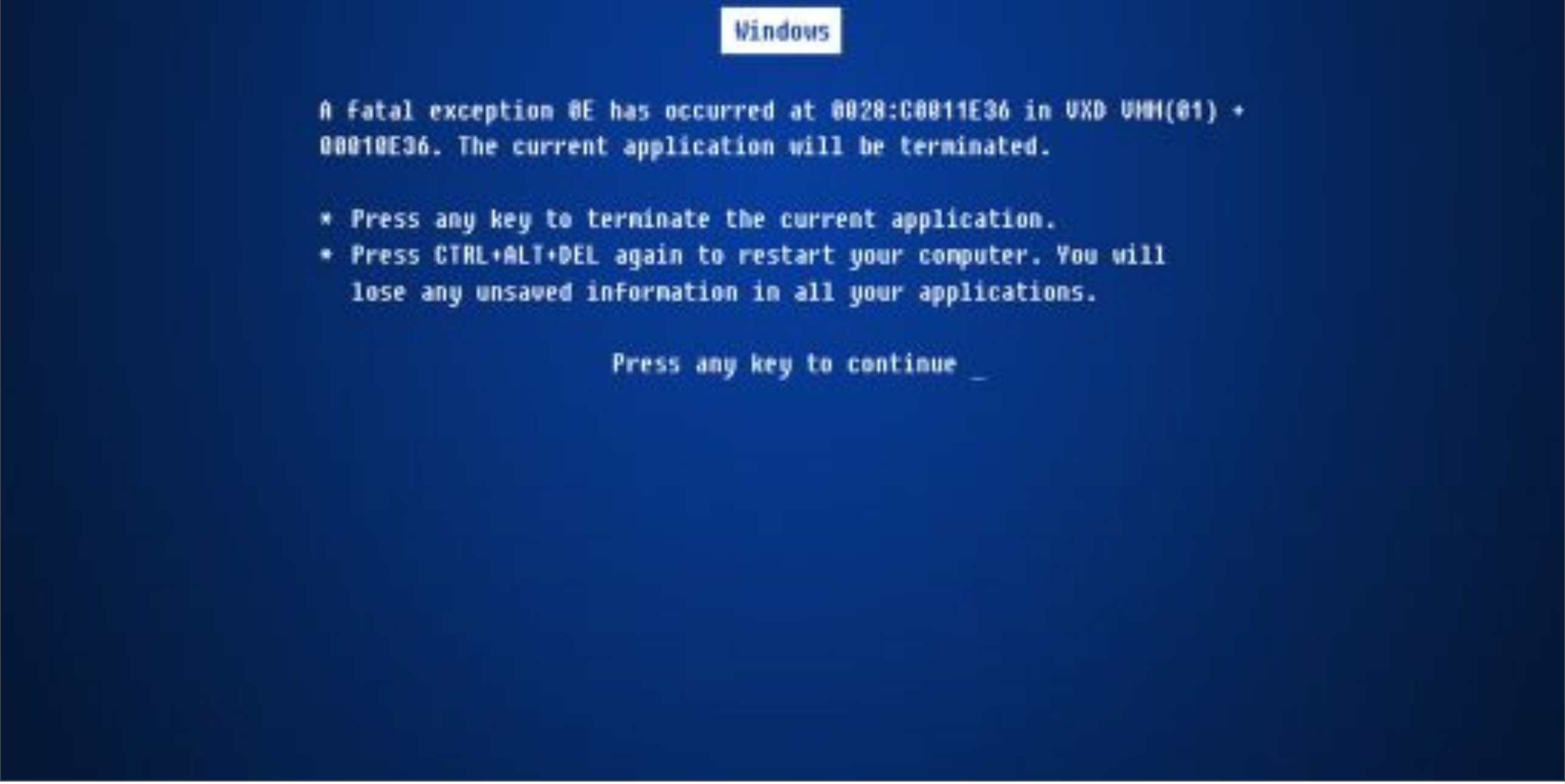
Female Patients

BMI \ DXA	*Not Obese*	*Obese*	BMI Totals
Not Obese	26 %	48 %	74%
Obese	0 %	26 %	26%
DXA Totals	26 %	74 %	100%

n-U-isance


Mortality Ratio
350
300
250
200
150
100
50
0
over weight
obese
extremely obese
15
20
25
30
35
40
Body Mass Index

Reinstall Windows



Trust, not Truth

Embarrassment of Riches

A team of psychologists performed personality tests on 100 professionals, of which 30 were engineers and 70 were lawyers.

Here is a brief description of one of the subjects:

> Jack is a 45-year-old man. He is married and has four children. He is generally conservative, careful, and ambitious. He shows no interest in political or social issues and spends most of his free time on his many hobbies, which include home carpentry, sailing, and mathematics.

Kahneman and Tversky (1974)

What is the probability that Jack is one of the 30 engineers?

A. 10 – 40 %

B. 41 – 60 %

C. 61 – 80 %

D. 81 – 100 %

Kahneman and Tversky (1974); Vanity Fair

The Law of Small Numbers is even more relevant in the era of Big Data

Target knows your daughter is pregnant

... before you do

Customer Acquisition

"Right around the birth of a child... parents are exhausted and overwhelmed and their shopping patterns and brand loyalties are up for grabs."

Customer Acquisition

"We knew that if we could identify them in their second trimester, there's a good chance we could capture them for years."

Brochure Design

"We started mixing in all these ads for things we knew pregnant women would never buy, so the baby ads looked random."

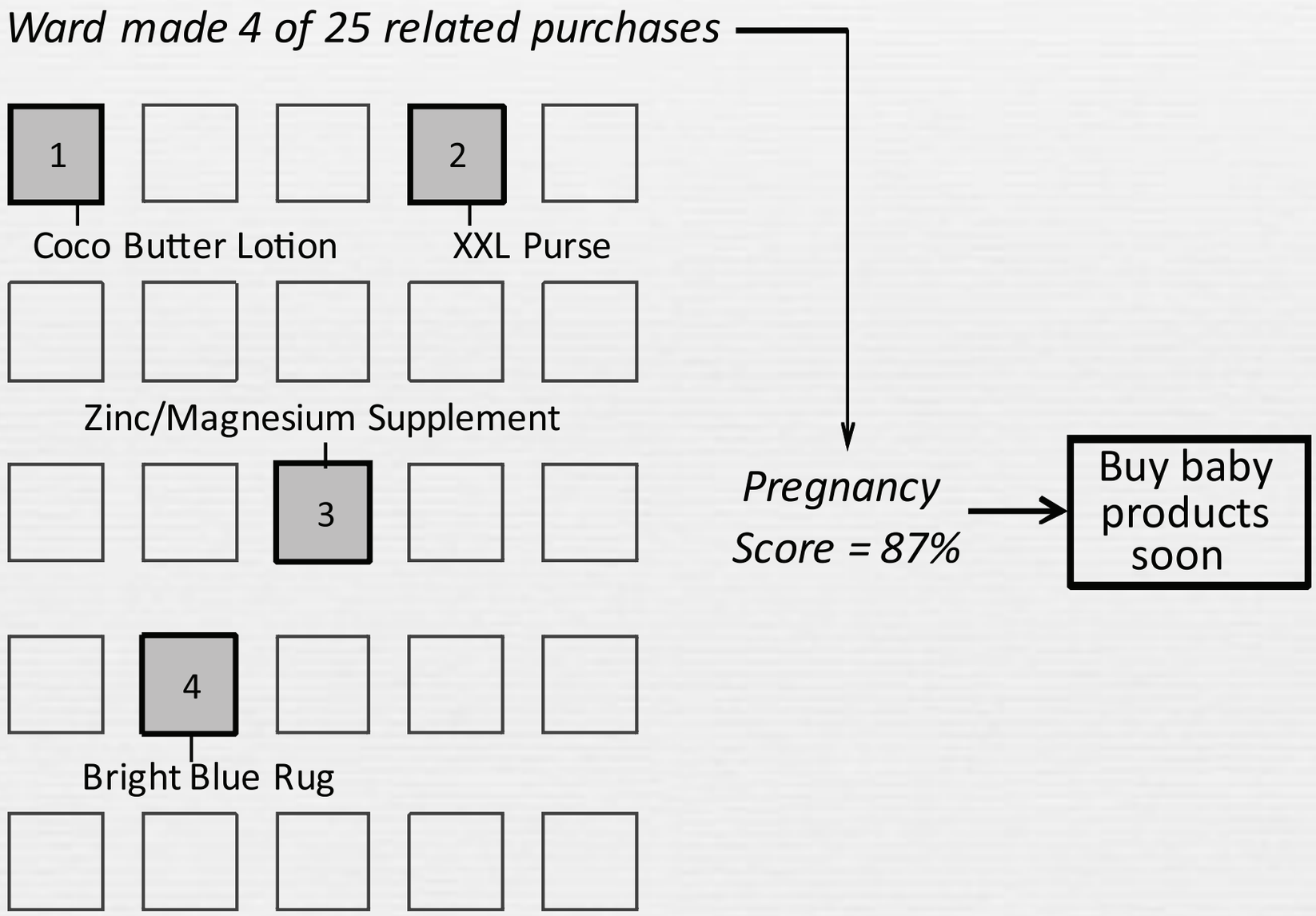
Brochure Design

"We'd put an ad for wineglasses next to infant clothes. That way, it looked like all the products were chosen by chance."

Brochure Design

"As long as a pregnant woman thinks she hasn't been spied on, she'll use the coupons."

The Model


Ward made 4 of 25 related purchases
1
2
Coco Butter Lotion
XXL Purse
Zinc/Magnesium Supplement
3
4
Bright Blue Rug
Pregnancy
Score = 87%
Buy baby products soon

Mad Dad

Model Says	Reality: Pregnant	Reality: Not	
Pregnant	6%	14%	20
Not	4%	76%	80
	10	90	100

Incidence: $\frac{10}{100} = 10\%$

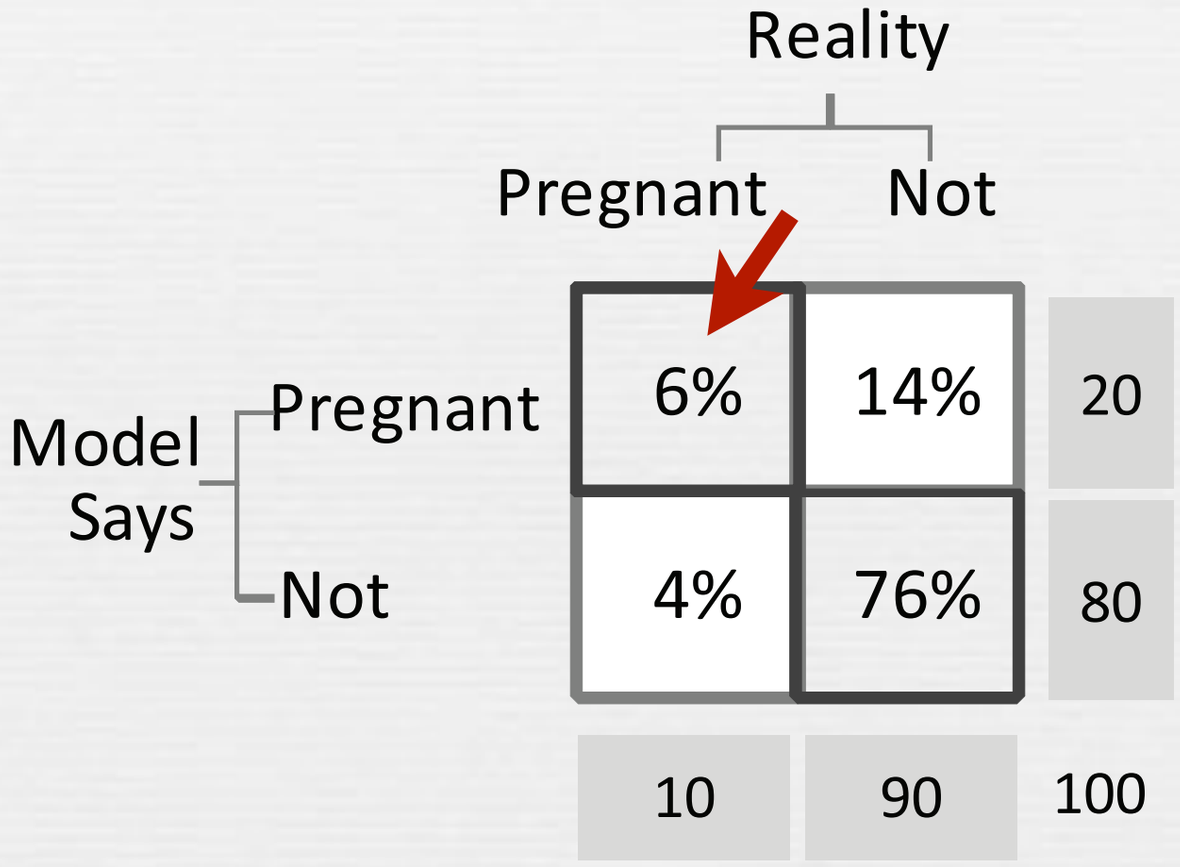
3x

Positive predictive value: $\frac{6}{20} = 30\%$

False positive rate: $\frac{14}{90} = 16\%$

False negative rate: $\frac{4}{10} = 40\%$

Sending Mixed Messages



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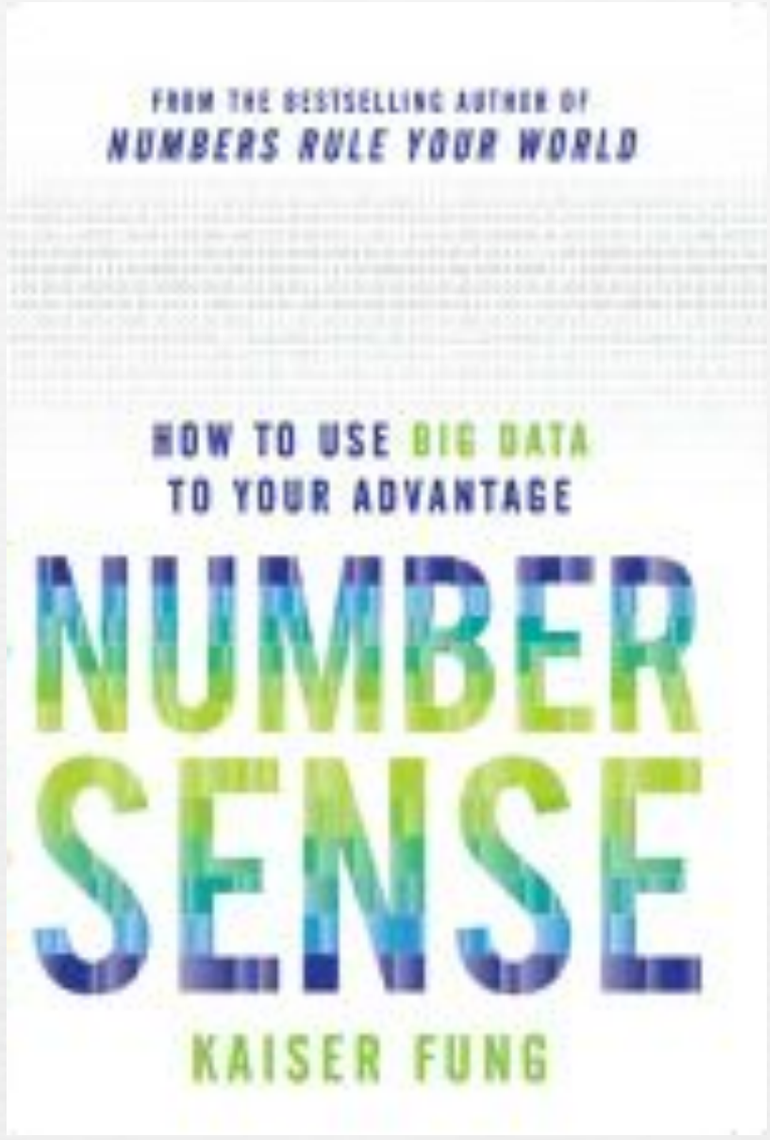
Intuition+Data

Data+Theory

Trust, ! Truth

Humans+Data

Law of small numbers



Thank you

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