



100 Years of Expertise,
Insight & Solutions

Big Data: What is it?

And what does it mean for the insurance industry?

CAS Centennial Meeting

New York City

November 10, 2014

In the past few years we
have produced more data than
in all of human history

Data is the new oil

Data Science: the Sexiest job of the 21st century

The 2012 presidential election
was the world's first big data
election.

Big data is one of the greatest sources of power in the 21st century

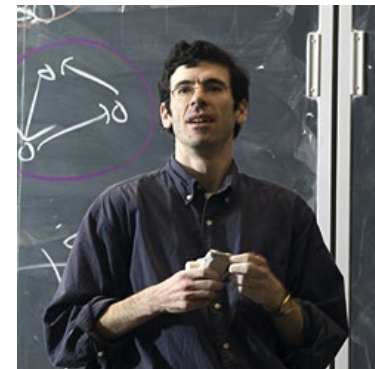
Big Data: a revolution that will
transform how we live, work,
and think

It will make you rich

The potential to transform everything

“The term itself is vague, but it is getting at something that is real... Big Data is a tagline for a process that has the potential to transform everything.”

— Jon Kleinberg, Cornell University



nature International weekly journal of science

Computational social science: Making the links

From e-mails to social networks, the digital traces left by life in the modern world are transforming social science.

Themes

What is big data?

A few examples of big data in action

Technology and analytic tools for big data

Big data and behavioral data

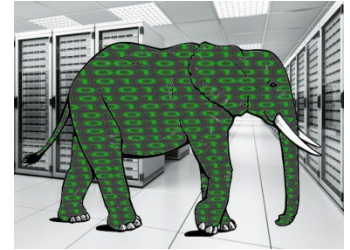
The dark side: ethics, business and liability

Data as a positive force: a new mindset for big data

So, what is it?

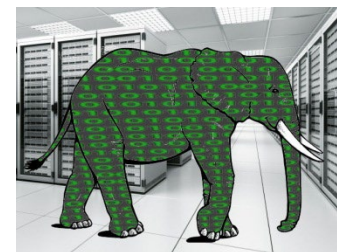
Three definitions of big data

1. Data sets with sizes beyond the capability of standard IT tools to capture, process, and analyze in reasonable time frames.



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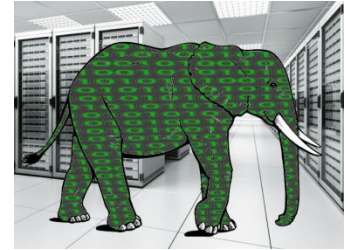
2. Data with high Volume, Velocity, Variety

- Huge datasets
- ... emanating continuously from smart phones, sensors, cameras, GPS devices, computers, TVs, ...
- ... involving all manner of numeric, text, photographic data



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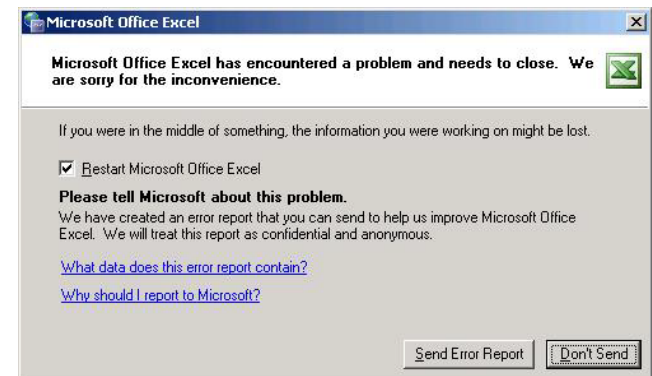


2. Data with high Volume, Velocity, Variety

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3. “Anything that doesn’t fit in Excel”

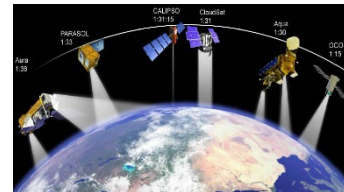


Traditional insurance data vs. new “big” data

Traditional Data



New “Big” Data

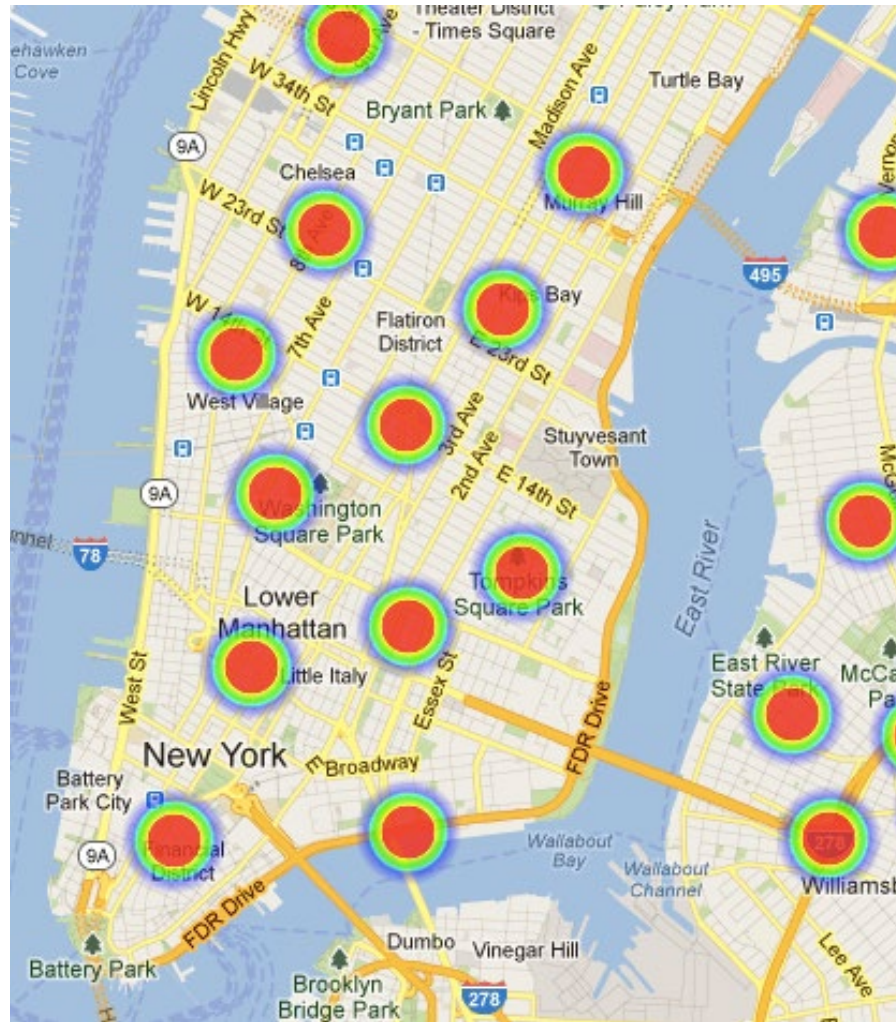


Expensive to collect, high value
Static, mixed within type

“Free” user-content, low value
Dynamic, fixed within type

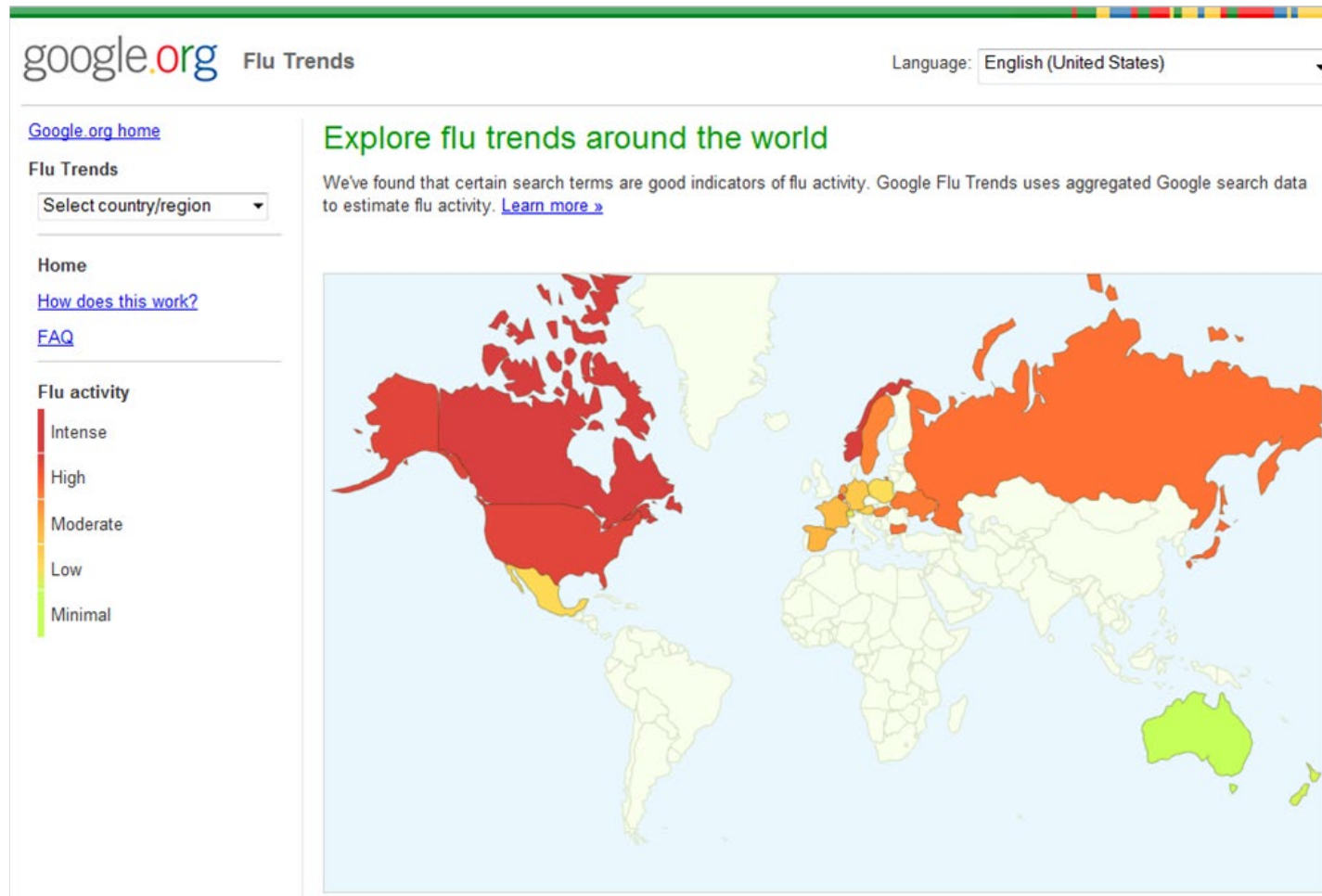
Some celebrated
examples of big data
in action

The big apple does actuarial science



*Data-driven building inspections in New York City:
Prospectively identify safety risks/violations*

Taking the temperature of the population



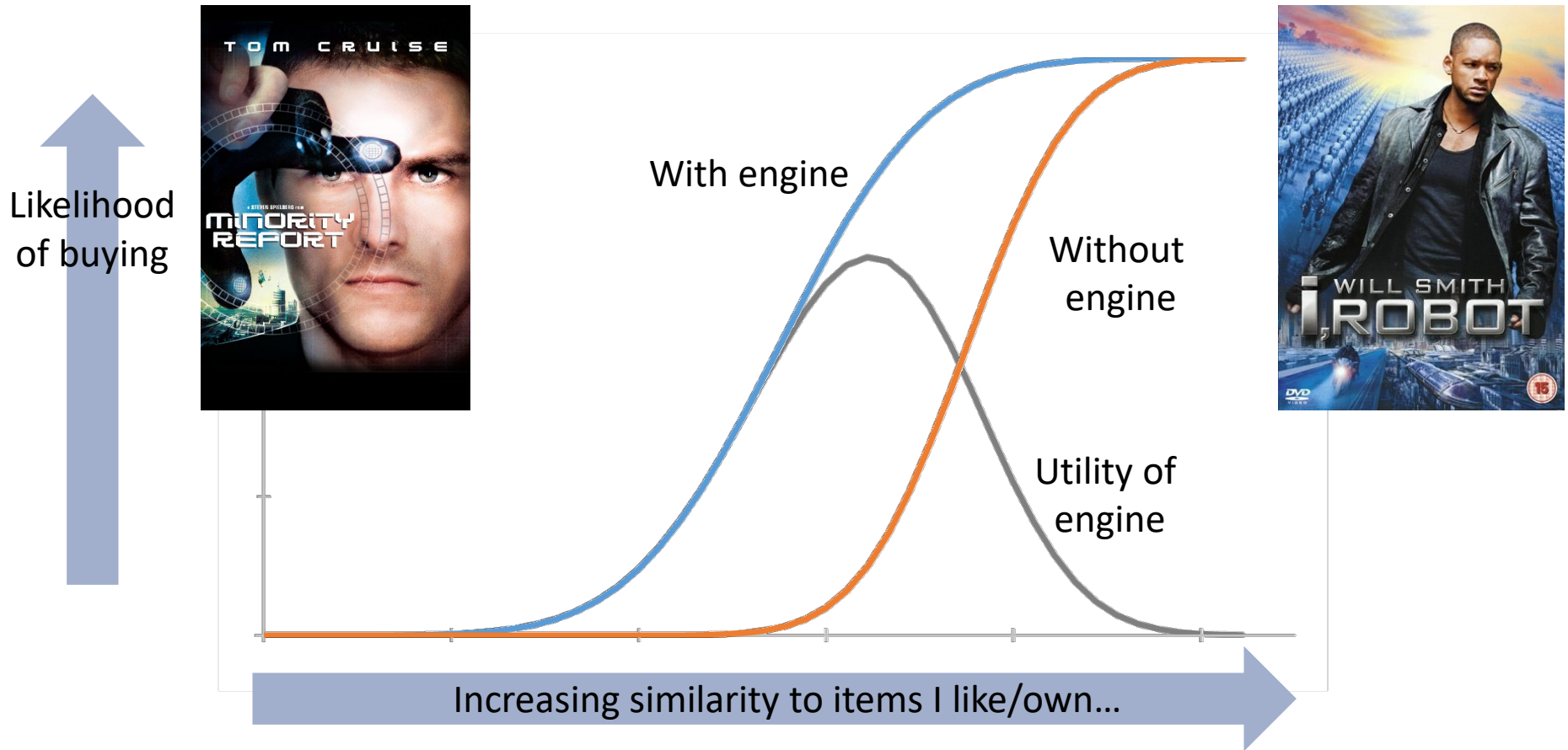
Similar ideas for economics, weather, risk hot-spots, or the “mood” of a customer base or populace

Better viewing through “datafication”



House of cards was actually built on a solid foundation of data

You may also like...



Recommendation engines must look beyond the obvious to delight and surprise customers

Technology and
analytics transform
data into insight

We need more than data

In general, people are not interested in data. What people want are answers.

-- David Hand, Data not Dogma



Data

Information

Insight

Work and some magic transforms data into information



**Transformation
= IT Problem**

Tools
~~Excel~~
SQL, Python
Hadoop

Old & new tools and methods compared

Old World	New World
Expensive , customized hardware	Cheap , commodity hardware
Model data	Dump data
Pre-optimize (index) to expected queries	Post-optimize requested ad hoc queries
Early binding types	Late binding types
Compute everything, dimensional DW	Compute what you want, on demand
Move data to the code	Move the code to the data
Inflexible, changes slow	Flexible, built to change



Magic = fundamentally different
operating model, 4x speed-up

How new methods can radically improve on old

Are two database records equal?

Record	Age	Sex	Ht	Wt	SSN	Zip	Etc.	Etc.
Memory	00101010101010100101010100100010111...							

Old School

- Age1 = Age2
- Sex1 = Sex2
- Wt1 = Wt2
- SSN1 = SSN2
- Etc.

- Implement for each record type
- Slow to execute

New School

- Compare binary numbers

- Works for ALL record types
- Quick to execute: primitive CPU function

Analysis transforms information into insight



Methodology

Exploratory analysis
Data adjustments
Variable selection
Model validation
Hold-out samples
...

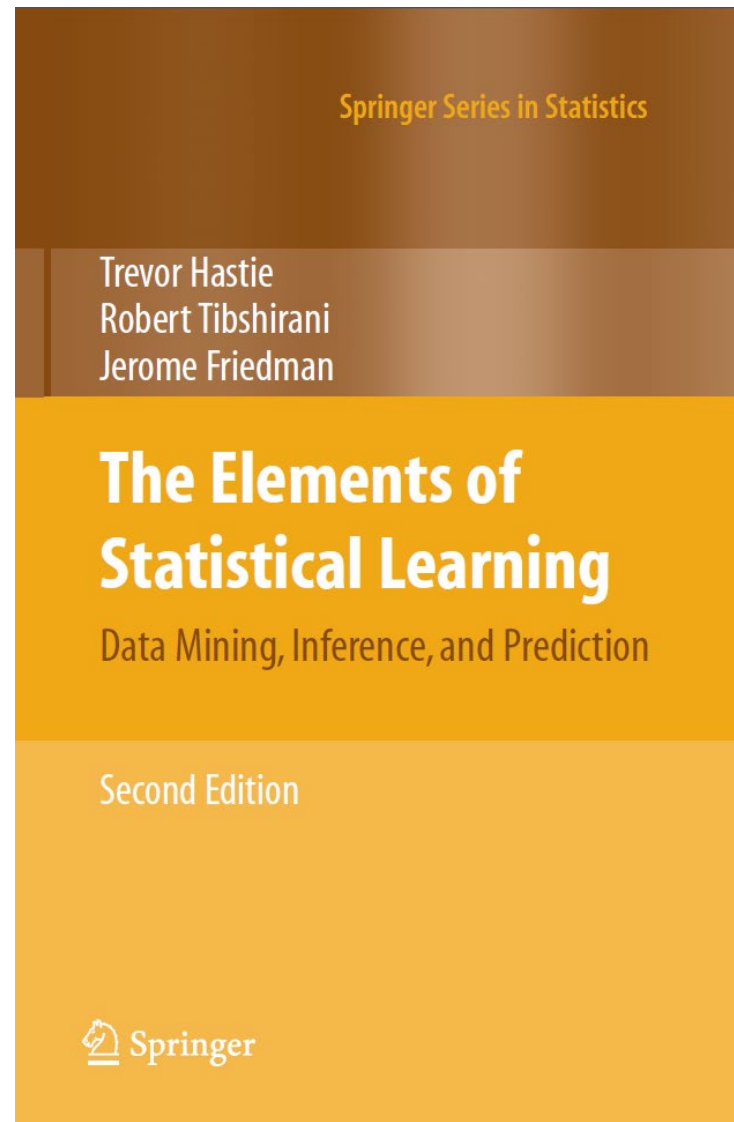
Algorithms

Machine Learning
SVM · NLP
Signal processing
Topological models
GAM · Splines
...

Tools

R
SAS
Matlab
Python
Prolog
...

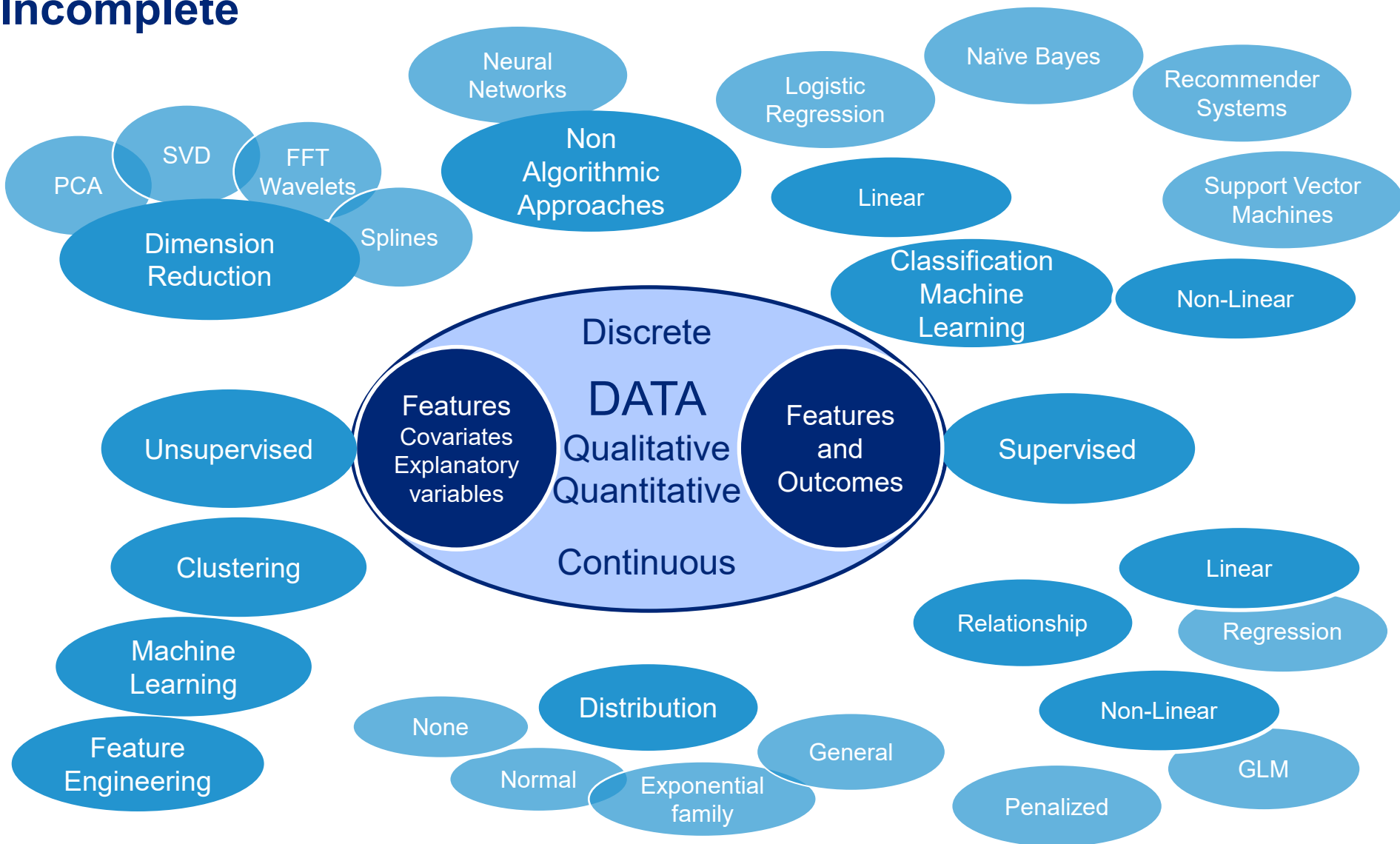
Techniques and algorithms



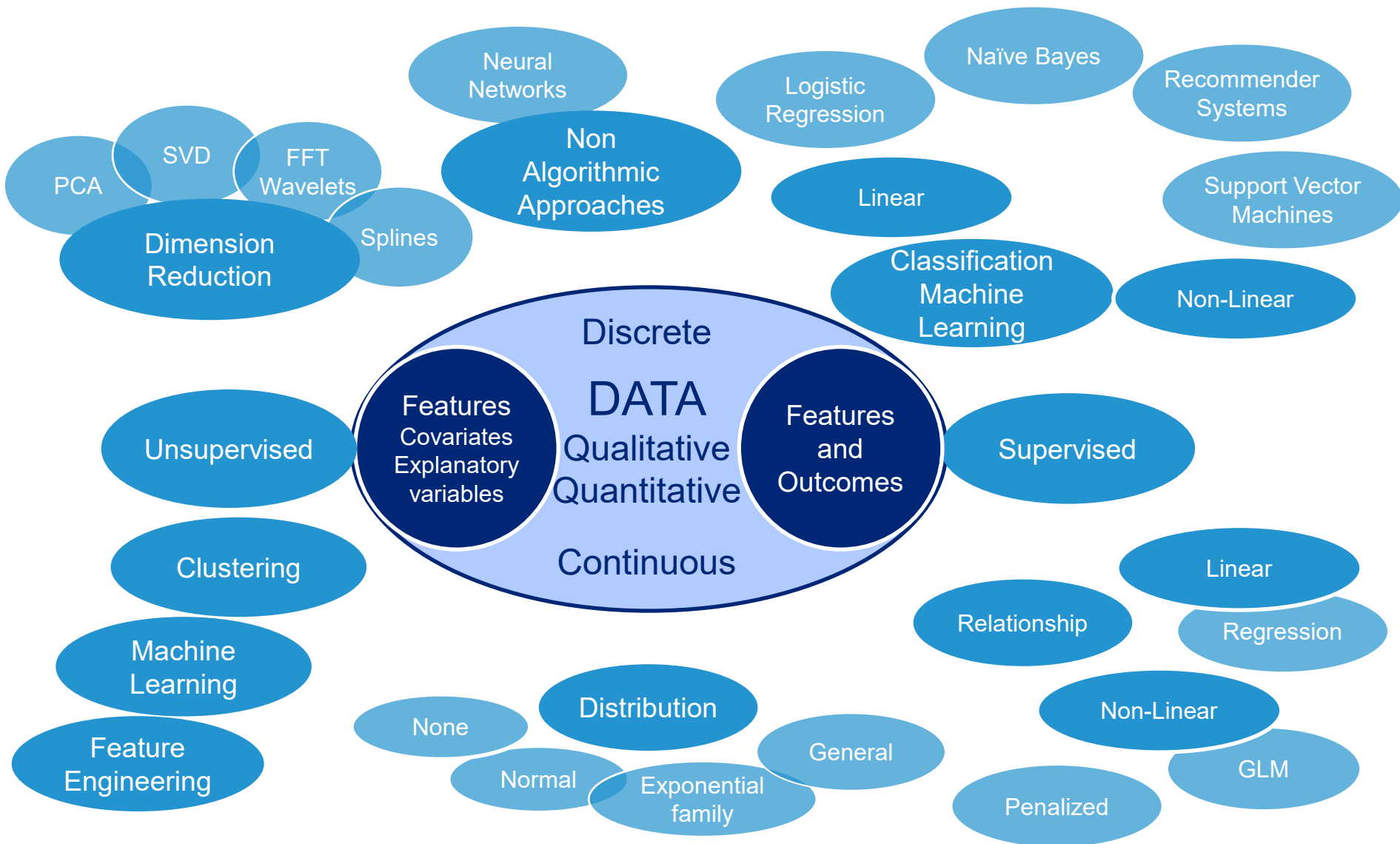
http://web.stanford.edu/~hastie/local.ftp/Springer/OLD/ESLII_print4.pdf

Schematic of analytic techniques and algorithms

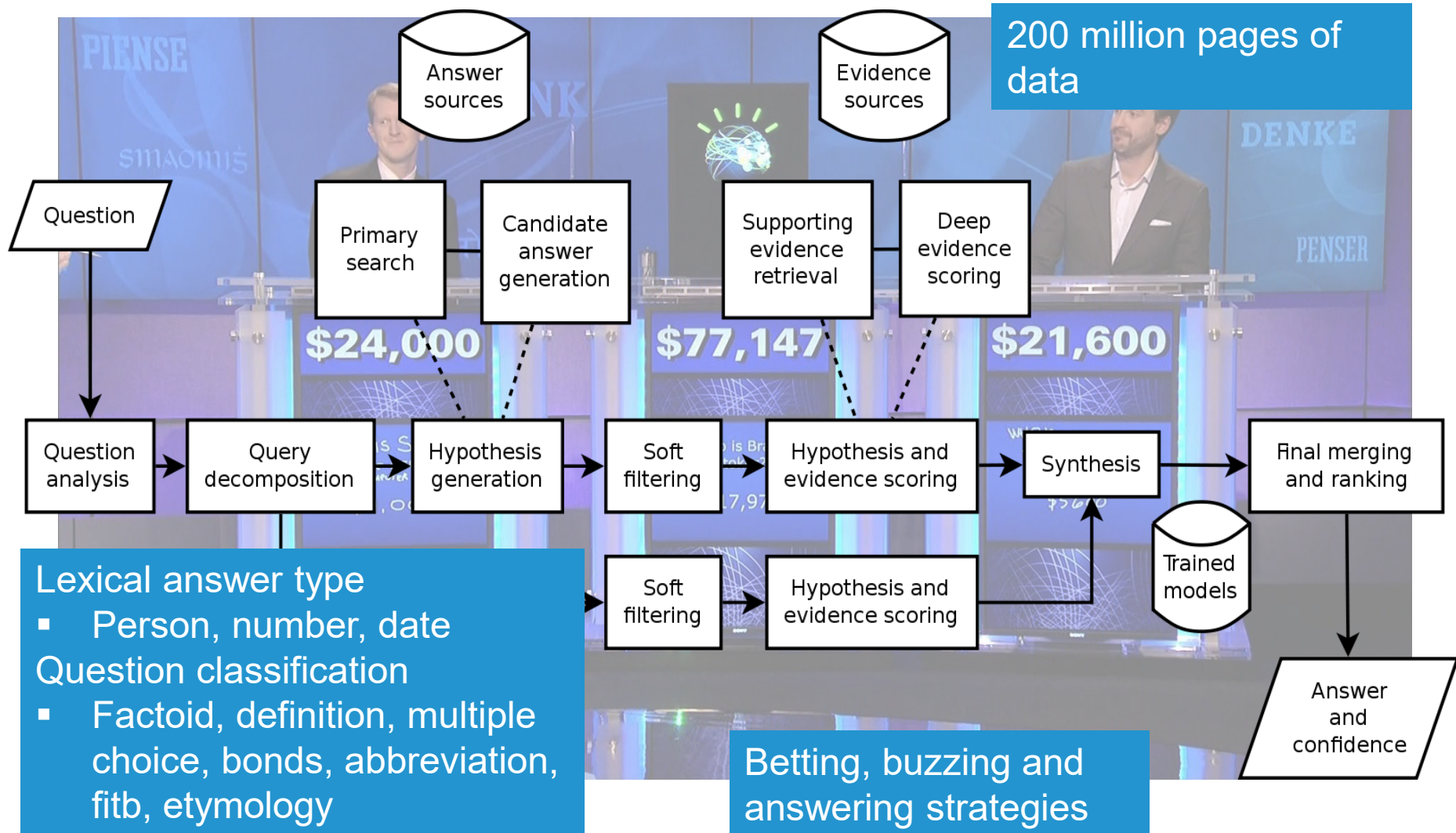
Incomplete



Schematic of analytic techniques and algorithms



Watson Wins Jeopardy! Training model a “significant effort”



Google Translate & the “unreasonable effectiveness of data”

A major big data success story!

“Learning from text at web scale”

- Non-parametric – no probability assumptions
- Uses unstructured text corpuses “in the wild”
- N-word matching
- “follow the data”



Translate



The Unreasonable Effectiveness of Data

Alon Halevy, Peter Norvig, and Fernando Pereira, Google

“Invariably, simple models and a lot of data trump more elaborate models based on less data...”

Currently, statistical translation models consist mostly of large memorized phrase tables that give candidate mappings between specific source- and target-language phrases.”

From machine translation to meme

“There is now a better way. Petabytes allow us to say: “Correlation is enough.” We can stop looking for models. We can analyze the data without hypotheses about what it might show. We can throw the numbers into the biggest computing clusters the world has ever seen and let statistical algorithms find patterns where science cannot.”

WIRED MAGAZINE: 16.07

SCIENCE : DISCOVERIES

The End of Theory: The Data Deluge Makes the Scientific Method Obsolete

By Chris Anderson 06.23.08



Illustration: Marian Bantjes

THE PETABYTE AGE:

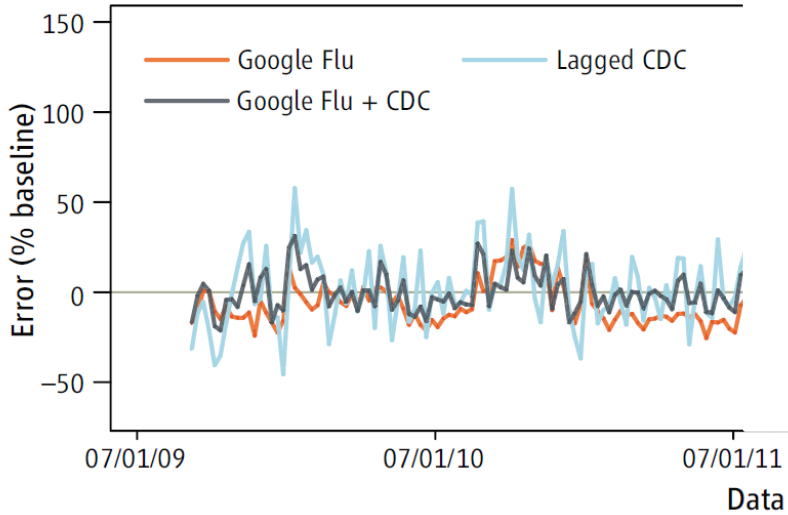
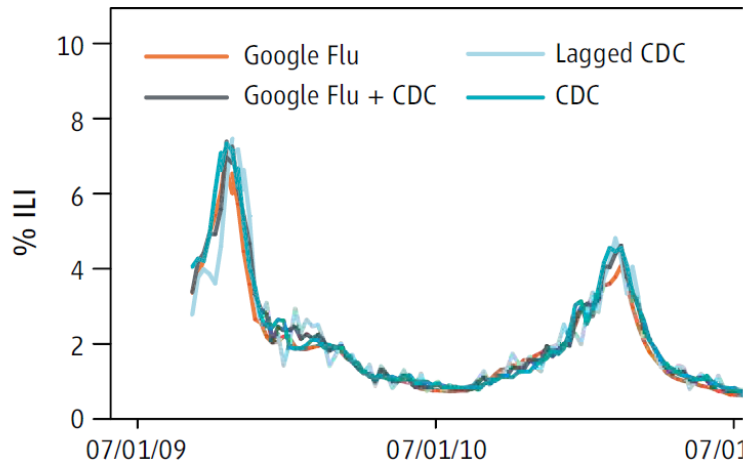
Sensors everywhere. Infinite storage. Clouds of processors. Our ability to capture, warehouse, and understand massive amounts of data is

"All models are wrong, but some are useful."

So proclaimed statistician George Box 30 years ago, and



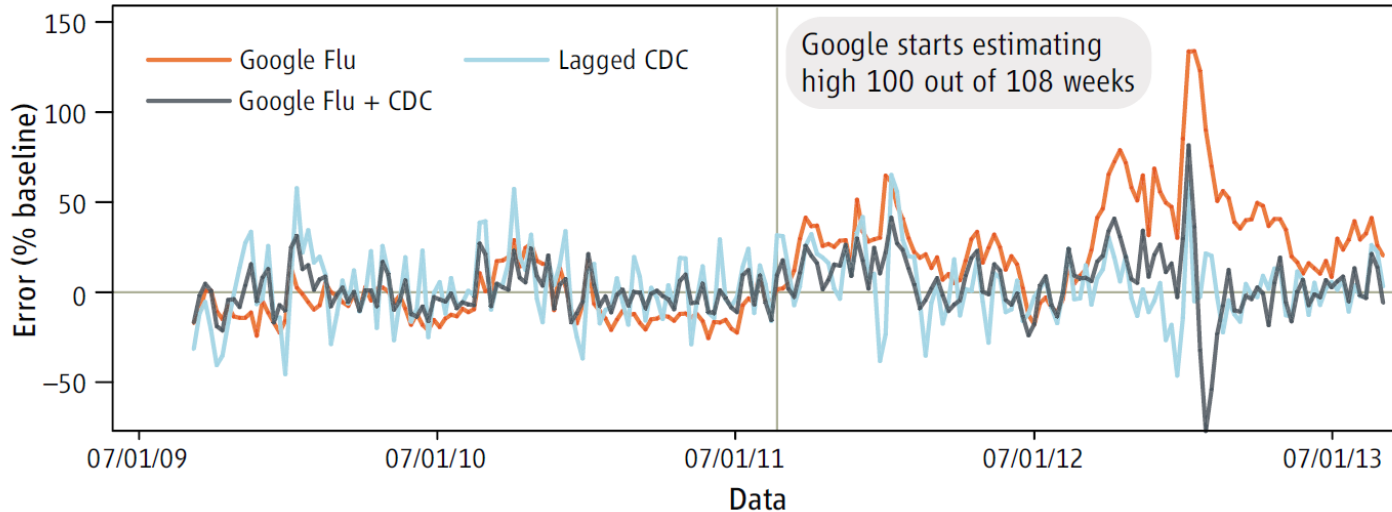
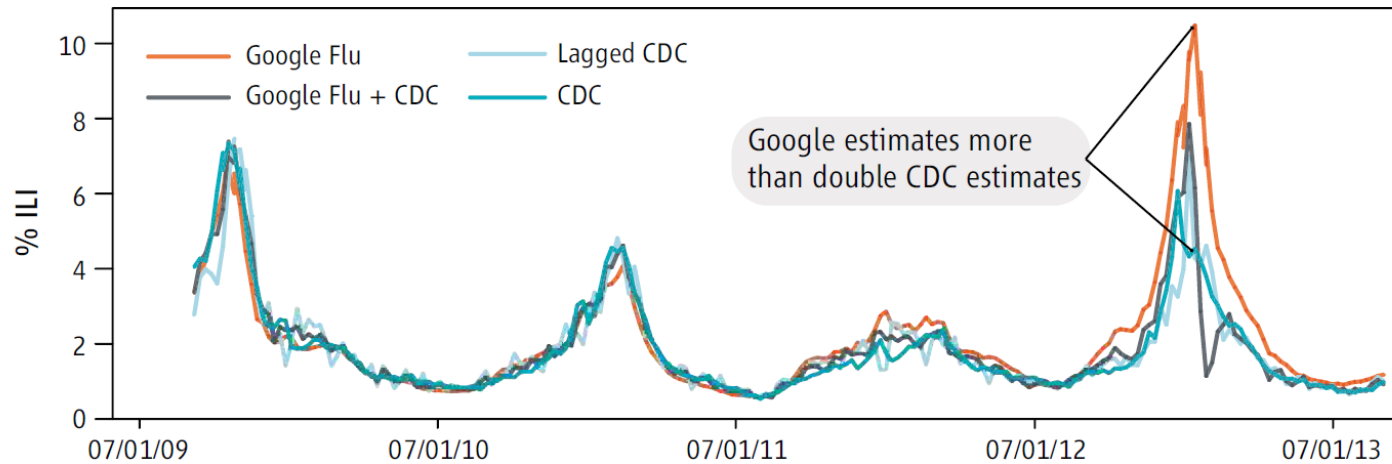
From poster child...



From poster child...to parable

The Parable of Google Flu: Traps in Big Data Analysis

David Lazer,^{1,2*} Ryan Kennedy,^{1,3,4} Gary King,³ Alessandro Vespignani^{3,5,6}



Google Flu Trends and “big data hubris”

The Parable of Google Flu: Traps in Big Data Analysis

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Big Data Hubris

“Big data hubris” is the often implicit assumption that big data are a substitute for, rather than a supplement to, traditional data collection and analysis.

Google Flu Trends and “big data hubris”

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Big Data Hubris

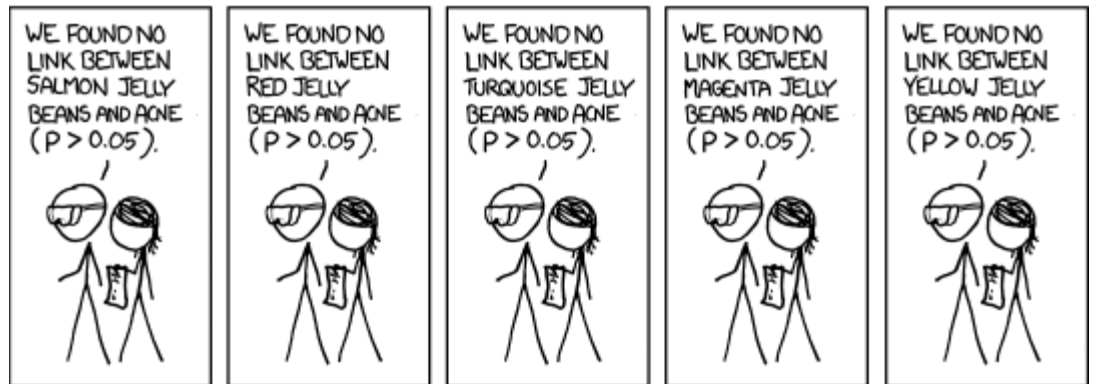
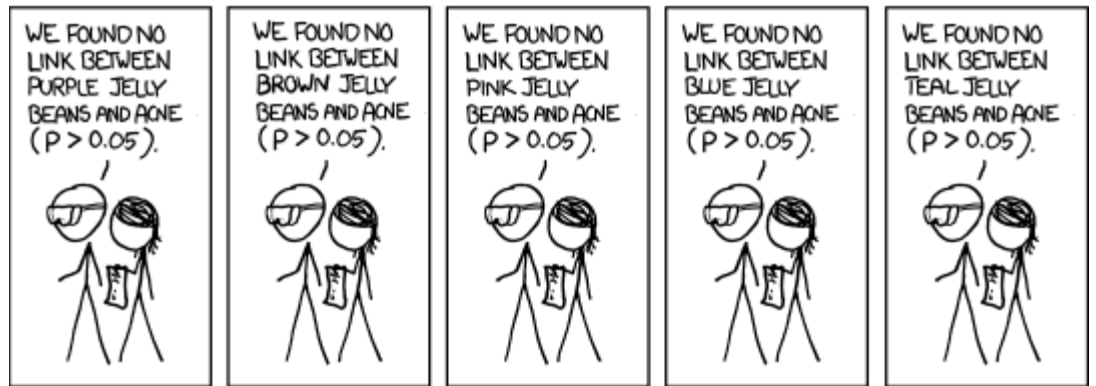
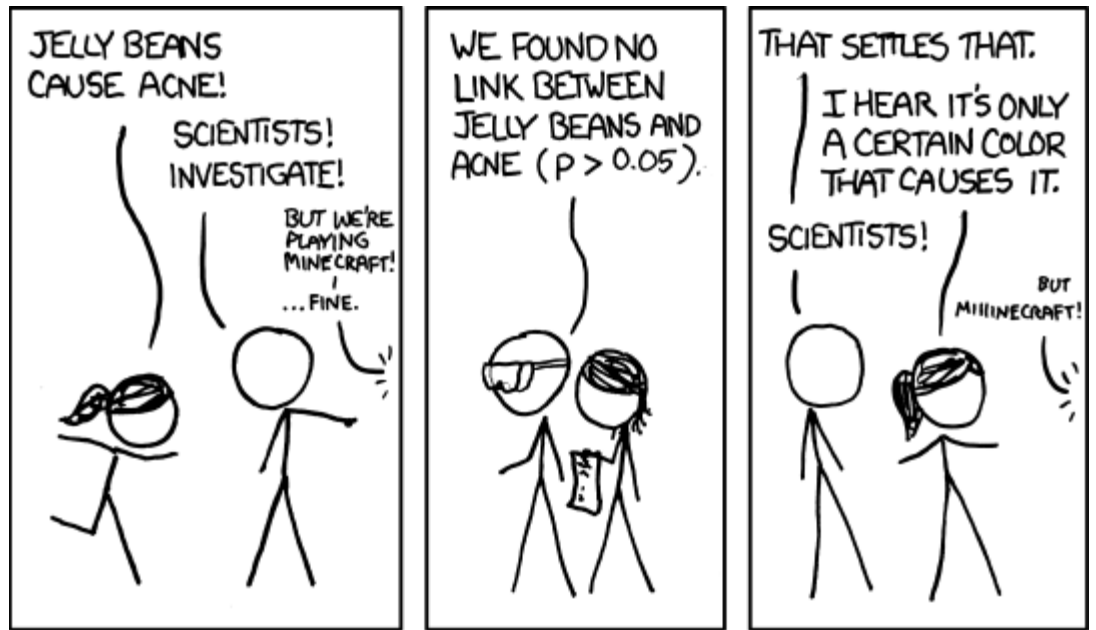
“Big data hubris” is the often implicit assumption that big data are a substitute for, rather than a supplement to, traditional data collection and analysis.

Big data offer enormous possibilities for understanding human interactions at a societal scale, with rich spatial and temporal dynamics, and for detecting complex interactions and nonlinearities among variables. We contend that these are the most exciting frontiers in studying human behavior. However, traditional “small data” often offer information that is not contained (or containable) in big data, and the very factors that have enabled big data are enabling more traditional data collection.

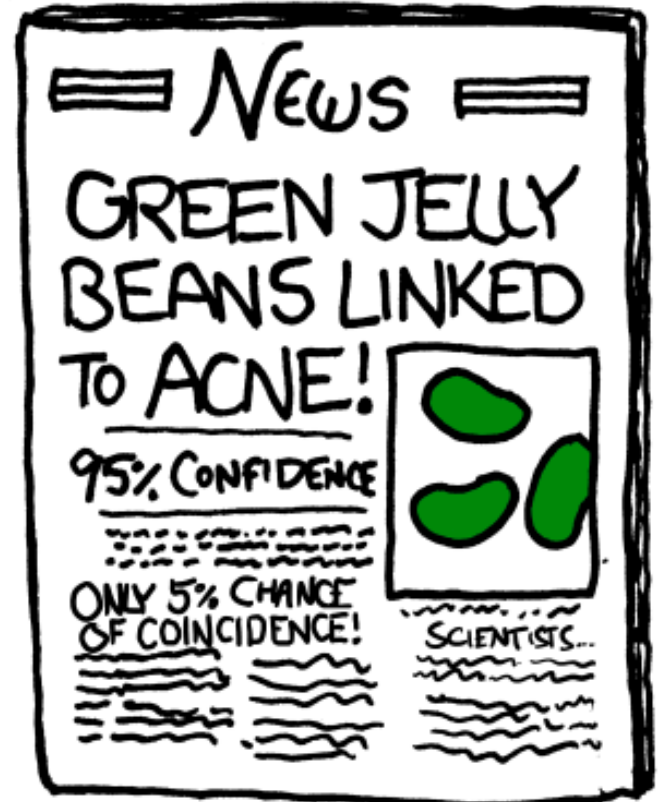
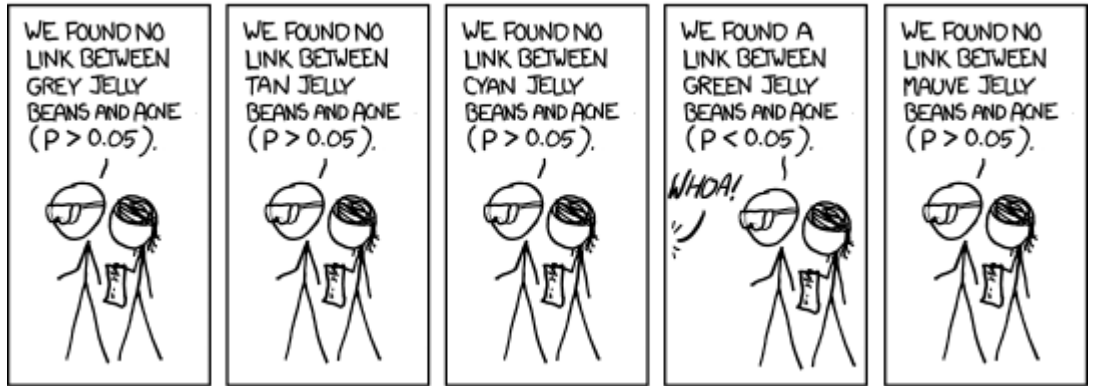
History doesn't repeat itself but it does rhyme



We've all bean there



We've all been there



Big data and behavioral data in insurance today

An early example of business analytics

CREDIT SCORE FACTORS



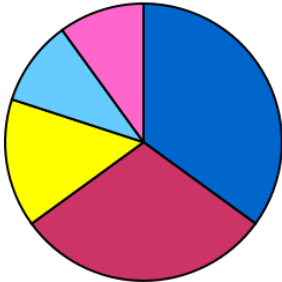
- On-time payments
- Capacity used
- Length of credit history
- Types of credit used
- Past credit applications

(This we know)



A more striking correlation

CREDIT SCORE FACTORS



- On-time payments
- Capacity used
- Length of credit history
- Types of credit used
- Past credit applications



More food for thought



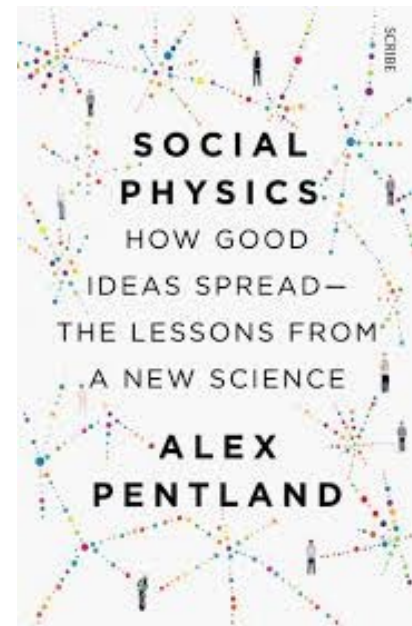
The real reason why big data is a big deal

“I believe that the power of Big Data is that it is information about people's behavior instead of information about their beliefs... This sort of Big Data comes from things like location data off of your cell phone or credit card, it's the little **data breadcrumbs** that you leave behind you as you move around in the world.



...those breadcrumbs tell... the story of your life... Big data is increasingly about real behavior, and by analyzing this sort of data, scientists can tell an enormous amount about you. They can tell whether you are the sort of person who will pay back loans. They can tell you if you're likely to get diabetes”

—Sandy Pentland, MIT Media Lab
“Reinventing Society in the Wake of Big Data”
edge.org conversation

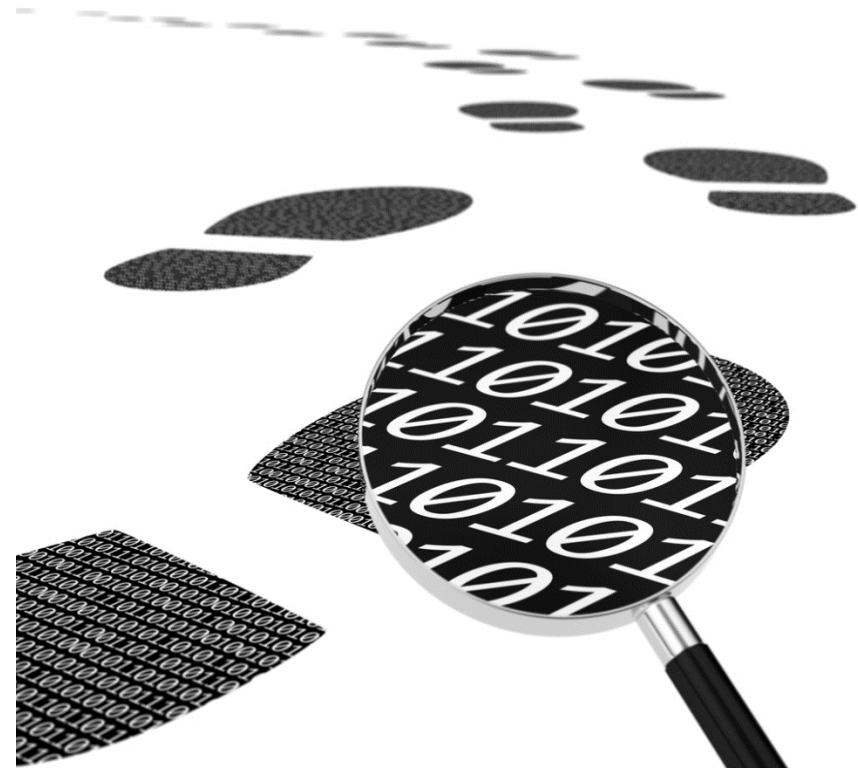


Looking ahead to ...?
New Mindset for Data
Science

Digital breadcrumbs, today's cleaner digital exhaust

*Our daily activities are increasingly digitally mediated...
We leave behind traces of*

- How we drive
- What we buy
- What we eat
- What we watch, read
- What and how we opine
- Where we travel
- Who we know
- Who we call
- How we socialize
- How we surf the web
- Where we are going next
- What is really on our minds

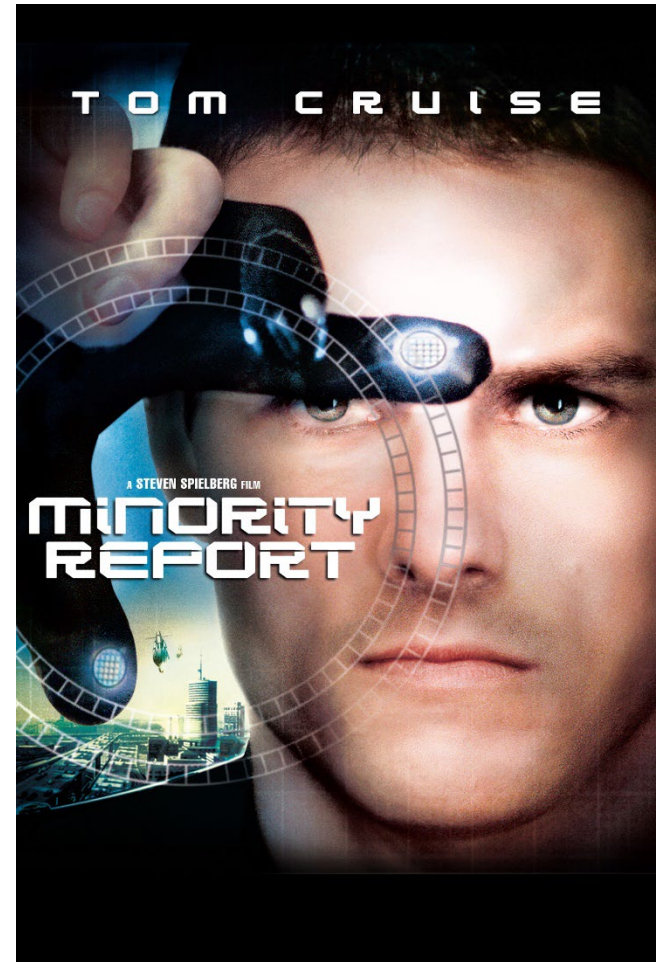


Evolving data privacy challenges

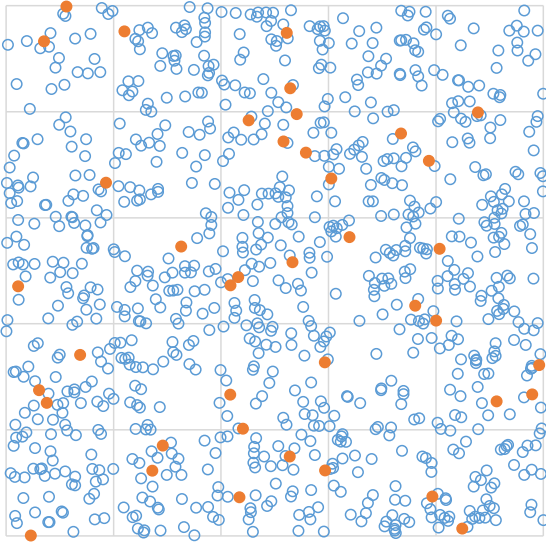
“We envision a very different privacy framework for the big-data age, one focused less on individual consent at the time of collection and more on holding data users accountable for what they do. In such a world, firms will formally assess a particular reuse of data based on the impact it has on individuals...

... sloppy assessments or poor implementation of safeguards will expose data users to legal liability, and regulatory actions such as mandates, fines, and perhaps even criminal prosecution.”

-- Viktor Mayer-Schönberger and Kenn Cukier



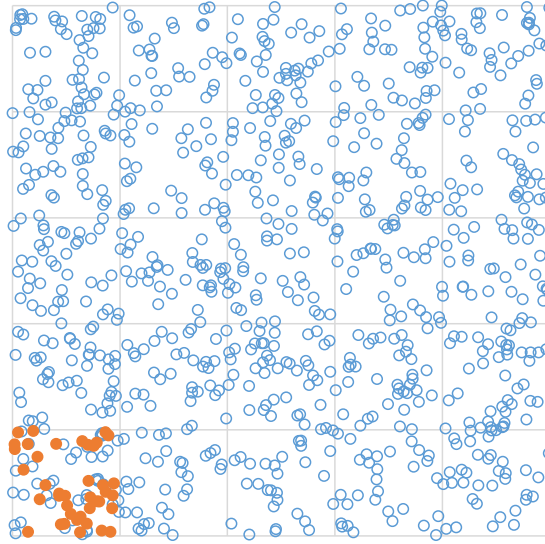
Big data and insurance: be careful what you wish for



Insurable

Old School

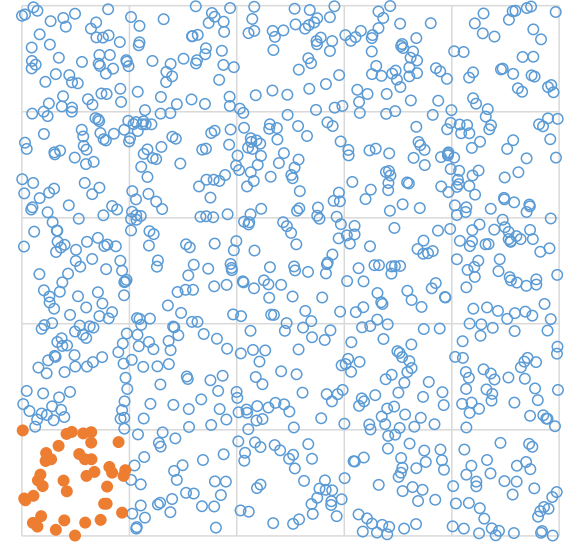
4% of 100%



Insurable, but expensive

Flood

50% of 8%



Not Insurable

Genetics

100% of 4%

The potential to
transform everything

Driving behavioral change

Actuaries now use telematics data to better segment and price insurance policyholders in terms of their utilization and riskiness

But could the data be used to create new products and services...

... periodic or real-time reports that serve as behavioral nudges...

Ideas

- *Detailed feedback reports to help student drivers learn and older drivers stay behind the wheel longer and safer*
- *Feedback prompting carbon footprint improvements through peer effects*



A healthy regard for one's policyholders

U1 Group follows



Medibank @medibank

1d

Track your steps towards better health with a free Fitbit Flex Activity Tracker.

 Promoted by Medibank



Medibank Health Insurance

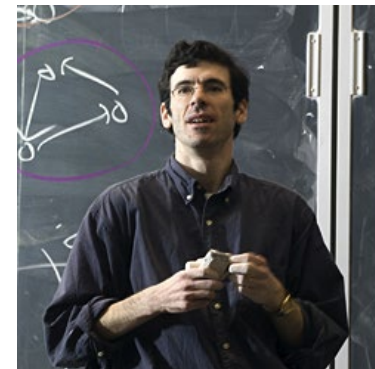
[Read more](#)

  5  14 

The potential to transform everything

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nature International weekly journal of science

Computational social science: Making the links

From e-mails to social networks, the digital traces left by life in the modern world are transforming social science.

Contact Information



James Guszcza

US Chief Data Scientist
Deloitte Consulting, LLP
Los Angeles

jguszcza@deloitte.com



Stephen Mildenhall

Global CEO of Analytics Aon
Center for Innovation and
Analytic, Singapore

stephen.mildenhall@aon.com

Sing cell: +65 9233 0670

US cell: +1 312 961 8781

Incredibly useful...

Translate



English Spanish Bengali Detect language ▾



English Polish Bengali ▾

Translate

dumbed-down escapist fare



tyle przystępne escapist taryfy



Incredibly useful... though not perfect

Translate



English Spanish Bengali Detect language



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dumbed-down escapist fare



tyle przystępne escapist taryfy



Translate



English Polish Bengali Detect language



English Polish Bengali

Translate

tyle przystępne escapist taryfy



affordable enough escapist fare



Credit as a psychobehavioral proxy

“... we delineate that basic chemical and psychobehavioral characteristics (e.g. a sensation-seeking personality type) are common to individuals exhibiting both higher insured automobile loss costs and poorer credit scores...”

© *The Journal of Risk and Insurance*, 2007, Vol. 74, No. 1, 23-63

BIOLOGICAL AND PSYCHOBEHAVIORAL CORRELATES OF CREDIT SCORES AND AUTOMOBILE INSURANCE LOSSES: TOWARD AN EXPLICATION OF WHY CREDIT SCORING WORKS

Patrick L. Brockett
Linda L. Golden

ABSTRACT

The most important new development in the past two decades in the personal lines of insurance may well be the use of an individual's credit history as a classification and rating variable to predict losses. However, in spite of its obvious success as an underwriting tool, and the clear actuarial substantiation of a strong association between credit score and insured losses over multiple methods and multiple studies, the use of credit scoring is under attack because there is not an understanding of *why* there is an association.