

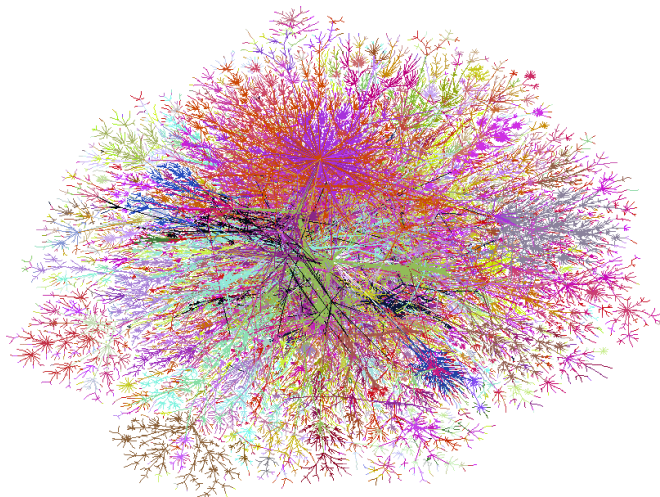
Internet Measurement and Data Analysis (1)

Kenjiro Cho

2013-09-25

introduction

how does the entire Internet look like?



lumeta internet mapping <http://www.lumeta.com>

<http://www.cheswick.com/ches/map/>

introduction (cont'd)

how does the entire Internet look like?

- ▶ no one knows
- ▶ but, everyone is interested

the theme of the class

- ▶ looking at the Internet from different views
 - ▶ how to measure what is difficult to measure
 - ▶ how to extract useful information from huge data sets

this kind of approach will be increasingly important in the future
information society

Internet measurement and data analysis

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<imda2013f@sfc.wide.ad.jp>
- ▶ textbooks, references: the lecture slide materials will be provided online.
- ▶ programming: data processing exercises by Ruby
- ▶ evaluation: 2 assignments and a final report

what you will learn in the class

- ▶ how to understand statistical aspects of data, and how to process and visualize data
 - ▶ which should be useful for writing thesis and other reports
- ▶ programming skills to process a large amount of data
 - ▶ beyond what the existing package software provide
- ▶ ability to suspect statistical results
 - ▶ the world is full of dubious statistical results and information manipulations
 - ▶ (improving literacy on online privacy)

Big Data everywhere



Big Data is a Big Deal

Posted by Tom Kalit on March 29, 2012 at 09:23 AM EDT

E-Mail | Tweet
[Editor's Note: Watch <http://live.science360>

Today, the Obama Admini
our ability to extract kno
promises to help accelera
transform teaching and l

To launch the initiative, 1
commitments that, toget
glean discoveries from t
address the challenges

We also want to challen
most of the opportunitie
President calls an "all ha

Some companies are ait
research. Universities a
generation of "data scie
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forum to highlight new p

Tom Kalit is Deputy Dir



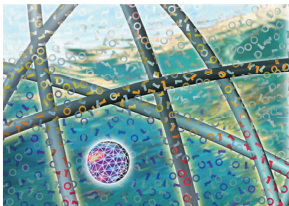
Special report: Managing information

Data, data everywhere

Information has gone from scarce to superabundant. That i
benefits, says Kenneth Cukier (interviewed here)—but also

Feb 25th 2010 | from the print edition

Like 30



The New York Times Sunday Review | The Opinion Pages

WORLD U.S. N.Y. / REGION BUSINESS TECHNOLOGY SCIENCE HEALTH

NEWS ANALYSIS The Age of Big Data

By STEVE LCH
Published: February 11, 2012

GOOD with numbers? Fascinated by data? The sound you hear is opportunity knocking.



Mo Zhou was snapped up by I.B.M. last summer, as a freshly minted Yale M.B.A., to join the technology company's fast-growing ranks of data consultants. They help businesses make sense of an explosion of data — Web traffic and social network comments, as well as software and sensors that monitor shipments, suppliers and customers — to guide decisions, trim costs and lift sales. "I've always had a love of numbers."

McKinsey Global Institute

Research | People | In the news | Contact us

Report

Big data: The next frontier for innovation, competition, and productivity

May, 2011 | by James Manyika, Michael Chui, Brad Brown, Jacques Bughin, Richard Dobbs, Charles Roxburgh

Download | Executive Summary PDF-822KB | Full Report PDF-6MB | Kindle MOBI-5MB | eBook EPUB-3MB

The amount of data in our world has been exploding, and analyzing large data sets—so-called big data—will become a key basis of competition, underpinning new waves of productivity growth, innovation, and consumer surplus, according to research by MGI and McKinsey's Business Technology Office. Leaders in every sector will have to grapple with the implications of big data, not just a few data-oriented managers. The increasing volume and detail of information captured by enterprises, the rise of multimedia, social media, and the Internet of Things will fuel exponential growth in data for the foreseeable future.

big data by cloud computing

- ▶ “big data” becomes a trendy word, especially for marketing
- ▶ most technologies are not new
 - ▶ have been used in search ranking, online recommender systems, etc.
- ▶ big data processing used to be limited to big organizations that could collect, manage, and analyze data in-house
- ▶ now, anyone can easily use big data with cloud services
- ▶ package tools are available for collecting and analyzing online customer behaviors
- ▶ customer information can be easily used for marketing with minimal initial investment

the age of data

- ▶ big data is not just for marketing
- ▶ technological innovations known as the data revolution are occurring in every field
- ▶ previously difficult applications become possible
 - ▶ access to huge amount of data, analysis of data constantly being updated, and applications to non-linear models
- ▶ big data analysis becomes an indispensable research method in all areas of science and technology

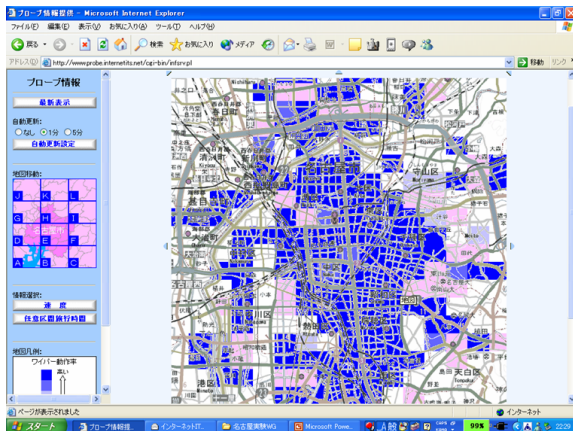
example: impact to science

e-science: paradigma shift?

- ▶ theory
- ▶ experiment
- ▶ simulations (enabled by computer)
- ▶ data-driven discovery (enabled by big data)

example: Internet vehicle experiments

- ▶ by WIDE Project In Nagoya in 2001
 - ▶ location, speed and wiper usage data from 1,570 taxis
 - ▶ blue areas indicate high ratio of wiper usage, showing rainfall in detail



Japan Earthquake

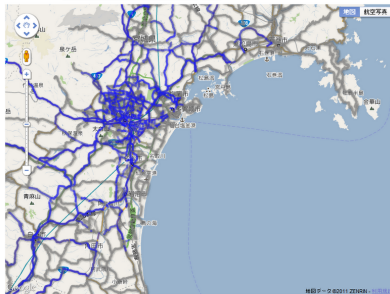
- ▶ the system is now part of ITS
- ▶ usable roads info released 3 days after the quake
 - ▶ data provided by HONDA (TOYOTA, NISSAN)

Google Crisis Response 自動車・通行実績情報マップ

a google.org project

下記マップ中に青色で表示されている道路は、前日の0時～24時の間に通行実績のあった道路を、灰色は同期間に通行実績のなかった道路を示しています。
(データ提供: 本田技研工業株式会社)

住所を入力して検索:



この自動車・通行実績情報マップは、被災地内での移動の参考となる情報を提供することを目指しています。ただし、個人が現地に向かうことは、余震の危険・交通渋滞の発生などの可能性がありますので、ご注意ください。

このマップは、Googleが、本田技研工業株式会社(Honda)から提供を受けた、Hondaが運営する「インターネット・ナビゲーション」サービスが提供する「ナビゲーション」が作成した「通行実績情報」を参照して作成されています。Hondaは、24時間以内に通行実績情報を更新する予定であり、Googleは更新後の情報更新の頻度、可否の通やのに情報提供を予定しています。

なお、通行実績がある道路でも、現在通行できることを保証するものではありません。実際の道路状況は、このマップと異なる場合があります。緊急交通路に指定される等、通行が規制されている可能性もあります。事前に、国土交通省、警察、東日本高速道路株式会社等の情報をご確認ください。

Google's Chief Economist Hal Varian on Statistics

The McKinsey Quarterly, January 2009

"I keep saying the sexy job in the next ten years will be statisticians. People think I'm joking, but who would've guessed that computer engineers would've been the sexy job of the 1990s? The ability to take data — to be able to understand it, to process it, to extract value from it, to visualize it, to communicate it — that's going to be a hugely important skill in the next decades, not only at the professional level but even at the educational level for elementary school kids, for high school kids, for college kids. Because now we really do have essentially free and ubiquitous data. So the complimentary scarce factor is the ability to understand that data and extract value from it."



self-introduction

Kenjiro Cho

- ▶ positions

- ▶ Research Director, IIJ Research Lab
- ▶ Guest Professor, Keio SFC
- ▶ Adjunct Professor, JAIST
- ▶ Board member, WIDE Project

- ▶ bio

- ▶ BE in electronics from Kobe University in 1984.
 - ▶ started as a hardware engineer at Canon, Inc, then became interested in operating systems
- ▶ M.Eng in computer science from Cornell University in 1993
 - ▶ studied computer science and distributed systems
- ▶ Researcher at Sony Computer Science Labs from 1996
 - ▶ research on the Internet
- ▶ Ph.D. (Media and Governance) from Keio University in 2001
- ▶ Researcher at IIJ from 2004

- ▶ research topics

- ▶ Internet measurement and management
- ▶ networking support in operating systems
- ▶ datacenter management and cloud services

class overview

It becomes possible to access a huge amount of diverse data through the Internet. It allows us to obtain new knowledge and create new services, leading to an innovation called "Big Data" or "Collective Intelligence". In order to understand such data and use it as a tool, one needs to have a good understanding of the technical background in statistics, machine learning, and computer network systems.

In this class, you will learn about the overview of large-scale data analysis on the Internet, and basic skills to obtain new knowledge from massive information for the forthcoming information society.

class overview (cont'd)

Theme, Goals, Methods

In this class, you will learn about data collection and data analysis methods on the Internet, to obtain knowledge and understanding of networking technologies and large-scale data analysis.

Each class will provide specific topics where you will learn the technologies and the theories behind the technologies. In addition to the lectures, each class includes programming exercises to obtain data analysis skills through the exercises.

Prerequisites

The prerequisites for the class are basic programming skills and basic knowledge about statistics.

In the exercises and assignments, you will need to write programs to process large data sets, using the Ruby scripting language and the Gnuplot plotting tool. To understand the theoretical aspects, you will need basic knowledge about algebra and statistics. However, the focus of the class is to understand how mathematics is used for engineering applications.

class schedule (1/4)

- ▶ Class 1 Introduction (9/25)
 - ▶ Big Data and Collective Intelligence
 - ▶ Internet measurement
 - ▶ Large-scale data analysis
 - ▶ exercise: introduction of Ruby scripting language
- ▶ Class 2 Data and variability (10/2)
 - ▶ Summary statistics
 - ▶ Sampling
 - ▶ How to make good graphs
 - ▶ exercise: graph plotting by Gnuplot
- ▶ Class 3 Data recording and log analysis (10/9)
 - ▶ Network management tools
 - ▶ Data format
 - ▶ Log analysis methods
 - ▶ exercise: log data and regular expression

class schedule (2/4)

- ▶ Class 4 Distribution and confidence intervals (10/16)
 - ▶ Normal distribution
 - ▶ Confidence intervals and statistical tests
 - ▶ Distribution generation
 - ▶ exercise: confidence intervals
 - ▶ **assignment 1**
- ▶ Class 5 Diversity and complexity (10/23)
 - ▶ Long tail
 - ▶ Web access and content distribution
 - ▶ Power-law and complex systems
 - ▶ exercise: power-law analysis
- ▶ Class 6 Correlation (10/30)
 - ▶ Online recommendation systems
 - ▶ Distance
 - ▶ Correlation coefficient
 - ▶ exercise: correlation analysis

class schedule (3/4)

- ▶ Class 7 Multivariate analysis (11/6)
 - ▶ Data sensing and GeoLocation
 - ▶ Linear regression
 - ▶ Principal Component Analysis
 - ▶ exercise: linear regression
- ▶ Class 8 Time-series analysis (11/13)
 - ▶ Internet and time
 - ▶ Network Time Protocol
 - ▶ Time series analysis
 - ▶ exercise: time-series analysis
 - ▶ **assignment 2**
- ▶ Class 9 Topology and graph (11/27)
 - ▶ Routing protocols
 - ▶ Graph theory
 - ▶ exercise: shortest-path algorithm
- ▶ Class 10 Anomaly detection and machine learning (12/4)
 - ▶ Anomaly detection
 - ▶ Machine Learning
 - ▶ SPAM filtering and Bayes theorem
 - ▶ exercise: naive Bayesian filter

class schedule (4/4)

- ▶ Class 11 Data Mining (12/11)
 - ▶ Pattern extraction
 - ▶ Classification
 - ▶ Clustering
 - ▶ exercise: clustering
- ▶ Class 12 Search and Ranking (12/18)
 - ▶ Search systems
 - ▶ PageRank
 - ▶ exercise: PageRank algorithm
- ▶ Class 13 Scalable measurement and analysis (12/25)
 - ▶ Distributed parallel processing
 - ▶ Cloud computing technology
 - ▶ MapReduce
 - ▶ exercise: MapReduce algorithm
- ▶ Class 14 Privacy Issues (1/8)
 - ▶ Internet data analysis and privacy issues
 - ▶ Summary of the class
- ▶ Class 15 Special Topics (TBD)

network measurement and Internet measurement

- ▶ network measurement (engineering)
 - ▶ measurement in limited environment
 - ▶ snapshot at a time
- ▶ Internet measurement
 - ▶ measurement of the Internet as a large-scale open system
 - ▶ large-scale distributed system
 - ▶ open system (continuously changing)
 - ▶ behavior of users on the Internet (social science)

goals of measurement and data analysis

- ▶ operational goals
 - ▶ trouble-shooting
 - ▶ tuning for performance and reliability
 - ▶ monitoring the usage, usage reports
 - ▶ long-term planning, cost evaluation of network capacity and equipment
- ▶ engineering goals (software, hardware, protocol design and implementations)
 - ▶ design trade-offs (e.g., buffer size and its cost)
 - ▶ testing and evaluation
 - ▶ observing unexpected behaviors (in complex systems)
- ▶ research goals (theory, modeling, new findings)
 - ▶ characteristics of network behaviors
 - ▶ modeling (e.g., behavior of web services)
 - ▶ behaviors of complex systems
 - ▶ abundant data and tools
- ▶ inputs for policy or investment plans

importance of measurement

measurement is a basis of all technologies

- ▶ for networking, it is an attempt to observe invisible networks
- ▶ needed for operation, design, implementation, and research
- ▶ however, it has become difficult by commercialization of the Internet and widespread use
 - ▶ traffic data is confidential for providers and will not be disclosed
 - ▶ risks of leaking private information

Internet measurement – measuring unmeasurable Internet

- ▶ need for generic measurement data for the Internet
 - ▶ example: typical packet size distribution
- ▶ the Internet is an open system continuously changing, evolving, and expanding
 - ▶ no central point, representative locations, different behaviors are observed depending on observing location and time
 - ▶ seeking for generality of the Internet: measuring unmeasurables
- ▶ for operation of the Internet, for development of protocols, equipment and services
 - ▶ seeking for the best estimates, predicting the future, and revisiting the existing knowledge
- ▶ user behavior: need to consider not only from technical aspects but also from social, political and economical aspects

characteristics of network data and behavior

- ▶ skewed distributions with large variance
 - ▶ inherent mechanism to make burst transfer
 - ▶ skewed utilization: e.g., a handful users generate most traffic
- ▶ anomalies everywhere
 - ▶ bugs, mis-configurations, spec mismatches, accidents, maintenance's
- ▶ interferences among various mechanisms
 - ▶ e.g., congestion control: Ethernet's collision avoidance, packet queueing, TCP's congestion control, capacity provisioning
- ▶ traffic aggregation
 - ▶ complex behavior as a whole (more than the sum of the individual components)
- ▶ limitations of network measurement
 - ▶ many practical issues and limitations exist
 - ▶ measurement affects the observed behavior

measurement needs combined skills

- ▶ goals could be operational, engineering, scientific
 - ▶ all inseparable, all skills required
 - ▶ knowledge of operational environment
 - ▶ engineering of measurement tools
- ▶ output can be facts, findings, new ideas
 - ▶ new ideas are not always necessary
 - ▶ facts, especially long-term measurement, are valuable
- ▶ but you should have clear goals
 - ▶ better to start with real problems to solve
 - ▶ there are many issues and problems but some are more important than others

why measurement of Internet is so hard?

- ▶ massive, diverse and changing measurements
- ▶ incomplete noisy data
 - ▶ need to use statistical inference to understand the nature of data
- ▶ mechanisms at different layers in different time scale
 - ▶ interact with each other
- ▶ dynamics
 - ▶ Internet mechanisms are adaptive and resilient
 - ▶ traditional measurement techniques are often not applicable
- ▶ pathological traffic is not unusual
 - ▶ by bugs, misconfigurations, errors, mismatches, accidents
- ▶ we still don't have good understanding

massive volume

- ▶ unprecedented scale with unprecedented growth
 - ▶ e.g., traffic volume: 1Gbps traffic
 - ▶ 120MB/sec 7GB/minute 420GB/hour 9.8TB/day
- ▶ far more data than we can analyze
 - ▶ techniques needed to reduce data size
 - ▶ filtering: e.g., record only TCP SYN packets
 - ▶ aggregation: e.g., flow-based accounting
 - ▶ sampling: e.g., record 1 in n packets
 - ▶ also, techniques needed to reduce dimensionality
- ▶ still, details matter
 - ▶ a big impact often comes
 - ▶ from small fraction
 - ▶ from minor differences

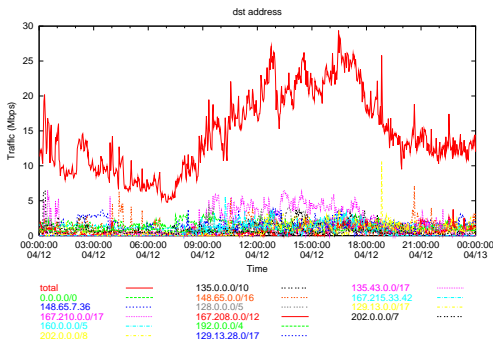
diverse data

- ▶ large variation in traffic mix between sites
- ▶ backbone vs. access links
 - ▶ access line types: fiber, ADSL, modem, wireless, satellite
 - ▶ differences in bandwidth, delay, loss
- ▶ various applications and users

typical data doesn't exist!

constant change

- ▶ daily, weekly traffic pattern
- ▶ trend changes over time
 - ▶ web in 90s and p2p/video in 2000s completely changed traffic pattern
- ▶ hard to predict future!



limitations of Internet measurement

- ▶ problems often occur at boundaries of different networks
 - ▶ cooperation needed but not easy
- ▶ measurement affects the behavior of the observed network
- ▶ need understanding and help from operators
 - ▶ need to understand operational requirements and find suitable methods for measurement
- ▶ cost: measurement doesn't come free
 - ▶ limitations to measure high-end routers with a PC
- ▶ privacy and confidential information in data
 - ▶ barriers for researchers to access commercial data

possible topics to be studied in the class

- ▶ search ranking (PageRank), online recommender systems(collaborative filtering)
- ▶ connections among SNS users, popular keyword extraction, shortest path search, online privacy
- ▶ SPAM filtering, MapReduce, Geolocation services, Web server log analysis
- ▶ Internet traffic, Internet topology, Packet capturing

summary

Internet measurement and data analysis

- ▶ measurement is basis for all technologies
- ▶ for networking, it is an attempt to observe invisible networks
- ▶ incomplete noisy data, statistical inference
- ▶ need to consider not only from technical aspects but also from social, political and economical aspects

theme of the class

- ▶ Internet measurement and data analysis as case studies
- ▶ learn how to measure what is difficult to measure
- ▶ learn how to extract useful information from huge data sets

Introduction to Ruby

Ruby

- ▶ a scripting language for object-oriented programming
- ▶ supports wide range of functions for text processing and system management
- ▶ free software started in 1993
- ▶ original author: Yukihiro Matsumoto
- ▶ became popular for Ruby on Rails (a web application framework)

Ruby information

Ruby official site: <http://www.ruby-lang.org/>

Ruby reference manual: <http://www.ruby-lang.org/en/documentation/>

Ruby の歩き方: <http://jp.rubyist.net/magazine/?FirstStepRuby>

Ruby characteristics

- ▶ interpreter language: no need to compile for execution
- ▶ highly portable: runs on most platforms
- ▶ simple syntax
 - ▶ no predefined data type for variables, variables can store any data and are dynamically typed
 - ▶ no need to declare variables, variable types (local variables, global variables, instance variables) can be inferred from variable names
- ▶ garbage collection: users do not need to manage memory
- ▶ object-oriented
 - ▶ everything is an object
 - ▶ class, inheritance, methods
 - ▶ iterator and closure
 - ▶ control structures and procedures can be written in object-oriented manner
- ▶ powerful string operations/regular expressions
- ▶ built-in support for large integers
- ▶ Ruby's shortcomings: a bit slower than its competitors

Ruby commands

- ▶ irb: Ruby's interactive interface

```
$ irb --simple-prompt
>> puts "Hello"
Hello
```

- ▶ ruby: Ruby main program

```
$ ruby test.rb
```

or,

```
$ ruby -e 'puts "Hello".reverse'
olleH
```

exercise: a program to count text lines

count the number of text lines in a file given by the argument

```
filename = ARGV[0]
count = 0
file = open(filename)
while text = file.gets
  count += 1
end
file.close
puts count
```

write to “count.rb” and then run it

```
$ ruby count.rb foo.txt
```

rewrite it in a more rubyish way

```
#!/usr/bin/env ruby
count = 0
ARGV.each_line do |line|
  count += 1
end
puts count
```

next class

Class 2 Data and variability (10/2)

- ▶ Summary statistics
- ▶ Sampling
- ▶ How to make good graphs
- ▶ exercise: graph plotting by Gnuplot

references

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