

X-Informatics Introduction: What is Big Data, Data Analytics and X-Informatics? Part II

July 6 2013

Geoffrey Fox

gcf@indiana.edu

<http://www.infomall.org/X-InformaticsSpring2013/index.html>

Associate Dean for Research, School of Informatics and
Computing

Indiana University Bloomington

2013

Big Data Ecosystem in One Sentence

Use **Clouds** running **Data Analytics Collaboratively**
processing **Big Data** to solve problems in
X-Informatics (or e-X)

X = Astronomy, Biology, Biomedicine, Business, Chemistry, Climate,
Crisis, Earth Science, Energy, Environment, Finance, Health,
Intelligence, Lifestyle, Marketing, Medicine, Pathology, Policy, Radar,
Security, Sensor, Social, Sustainability, Wealth and Wellness with
more fields (physics) defined implicitly
Spans Industry and Science (research)

Education: **Data Science** see recent New York Times articles
<http://datascience101.wordpress.com/2013/04/13/new-york-times-data-science-articles/>



Climate Informatics
network

How Wealth Informatics can help
with your financial freedom?



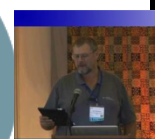
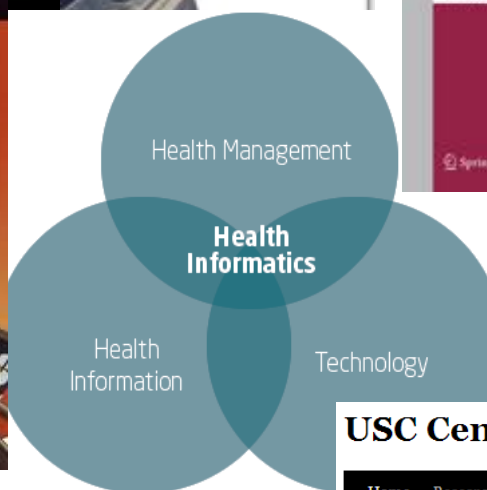
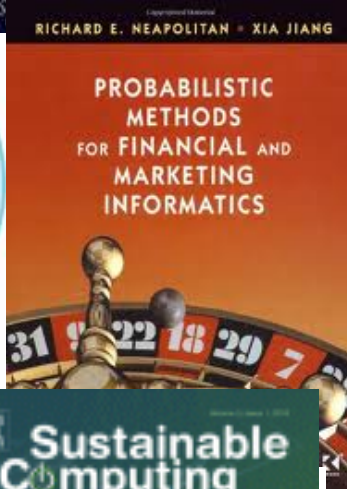
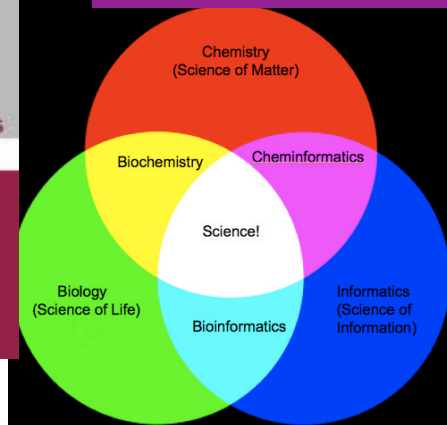
Biomedical Informatics

Computer Applications in Health Care
and Biomedicine

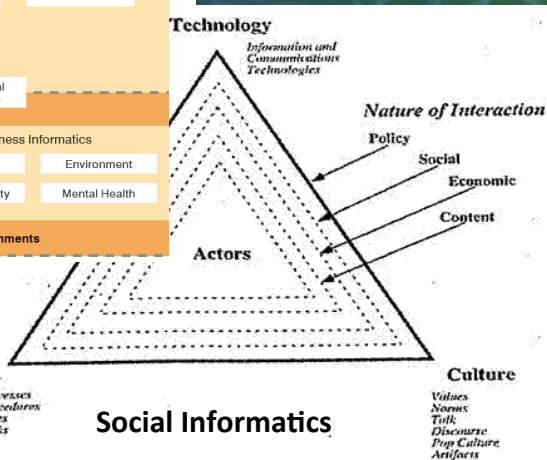
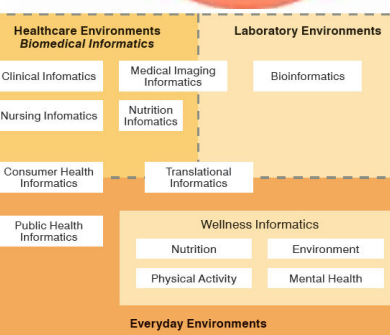
AstroInformatics2012

Redmond, WA, September 10 - 14, 2012

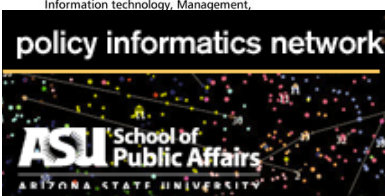
Journal of
Pathology
Informatics



Opportunities and Challenges
in Crisis Informatics



Noella Penelope Greer (Ed.)
Business Informatics
Information technology, Management,



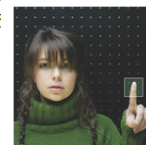
USC Center For Energy Informatics

Home Research Publications Sm

About the Center

Welcome to the Center For Energy Informatics (CEI) at USC, an Organized Research Unit (ORU) housed in the [Viterbi School of Engineering](#). Energy Informatics is the application of inf

Lifestyle Informatics



Applications of LI
How is the training classified?
Occupation Pr
Further study
Student at the
Watch the mov
Studying Abro

Admission and registration
VU Honours Programme

ENVIRONMENTAL
INFORMATICS

Lifestyle Informatics: Let people l

The study Lifestyle Informatics is about s
this bachelor including applied psycholog
knowledge about language and informatic
short better. Lifestyle Informatics: let peo
[Lifestyle Informatics](#)

combine
body,
healthier,
[aining](#)

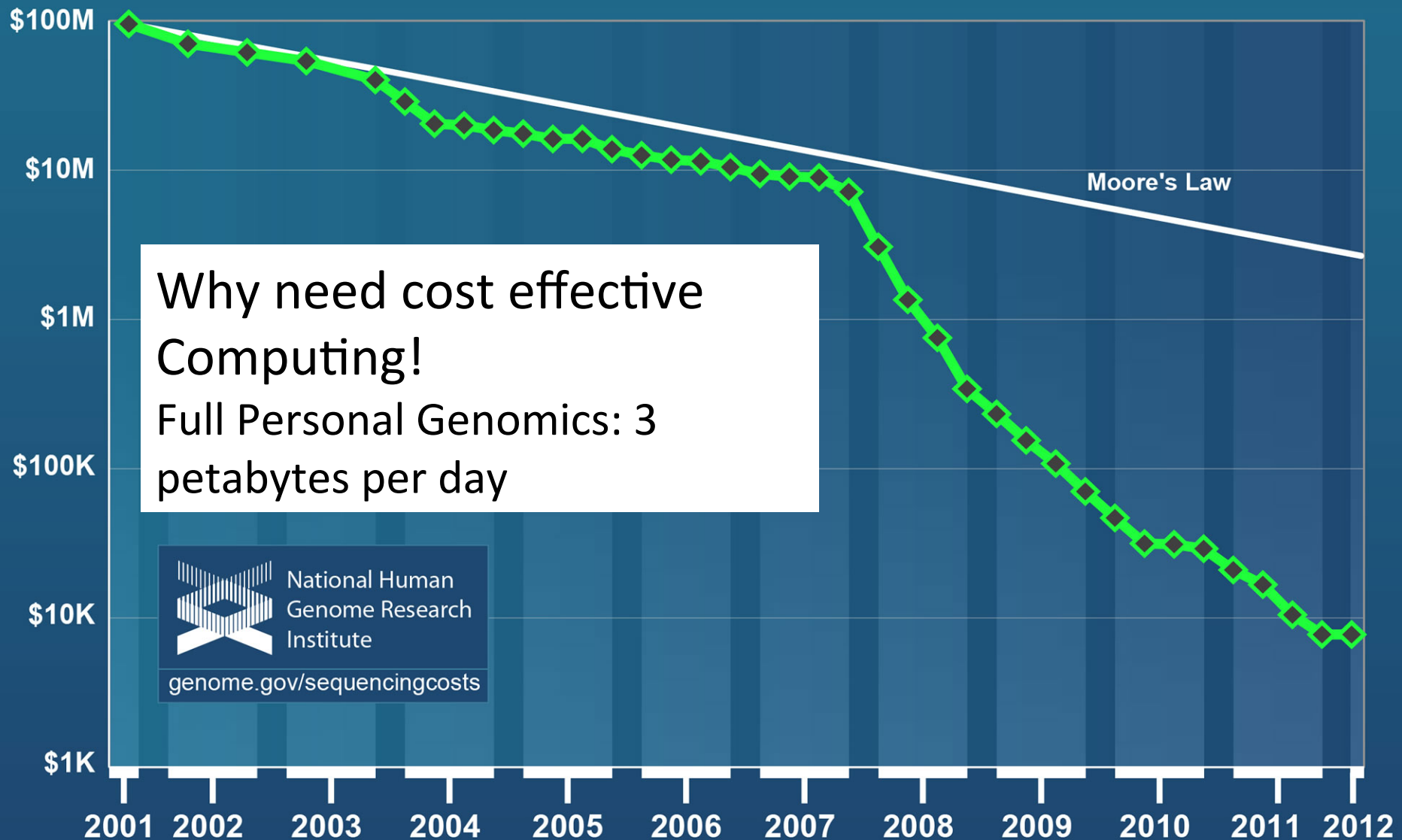


Data Deluge

Science & Research

- WHEN the Sloan Digital Sky Survey started work in 2000, its telescope in New Mexico collected more data in its first few weeks than had been amassed in the entire history of astronomy. Now, a decade later, its archive contains a whopping 140 terabytes of information. A successor, the Large Synoptic Survey Telescope, due to come on stream in Chile in 2016, will acquire that quantity of data every five days.
- Such astronomical amounts of information can be found closer to Earth too. Wal-Mart, a retail giant, handles more than 1m customer transactions every hour, feeding databases estimated at more than 2.5 petabytes—the equivalent of 167 times the books in America's Library of Congress (see article for an explanation of how data are quantified).
- Facebook, a social-networking website, is home to 40 billion photos. And decoding the human genome involves analysing 3 billion base pairs—which took ten years the first time it was done, in 2003, but can now be achieved in one week.

Cost per Genome



Why need cost effective
Computing!
Full Personal Genomics: 3
petabytes per day



<http://www.genome.gov/sequencingcosts/>

Ninety-six percent of radiology practices in the USA are filmless and Table below illustrates the annual volume of data across the types of diagnostic imaging; this does not include cardiology which would take the total to over 10⁹ GB (an Exabyte).

<http://grids.ucsf.edu/ptliupages/publications/Where%20does%20all%20the%20data%20come%20from%20v7.pdf>

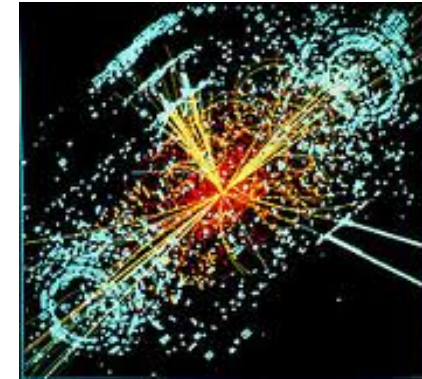
| Modality | Part B non HMO | All Medicare | All Population | Per 1000 persons | Ave study size (GB) | Total annual data generated in GB |
|-------------------------------|----------------|--------------|----------------|------------------|---------------------|-------------------------------------|
| CT | 22 million | 29 million | 87 million | 287 | 0.25 | 21,750,000 |
| MR | 7 million | 9 million | 26 million | 86 | 0.2 | 5,200,000 |
| Ultrasound | 40 million | 53 million | 159 million | 522 | 0.1 | 15,900,000 |
| Interventional | 10 million | 13 million | 40 million | 131 | 0.2 | 8,000,000 |
| Nuclear Medicine | 10 million | 14 million | 41 million | 135 | 0.1 | 4,100,000 |
| PET | 1 million | 1 million | 2 million | 8 | 0.1 | 200,000 |
| Xray, total incl. mammography | 84 million | 111 million | 332 million | 1,091 | 0.04 | 13,280,000 |
| All Diagnostic Radiology | 174 million | 229 million | 687 million | 2,259 | 0.1 | 68,700,000 68.7 PETAbytes |



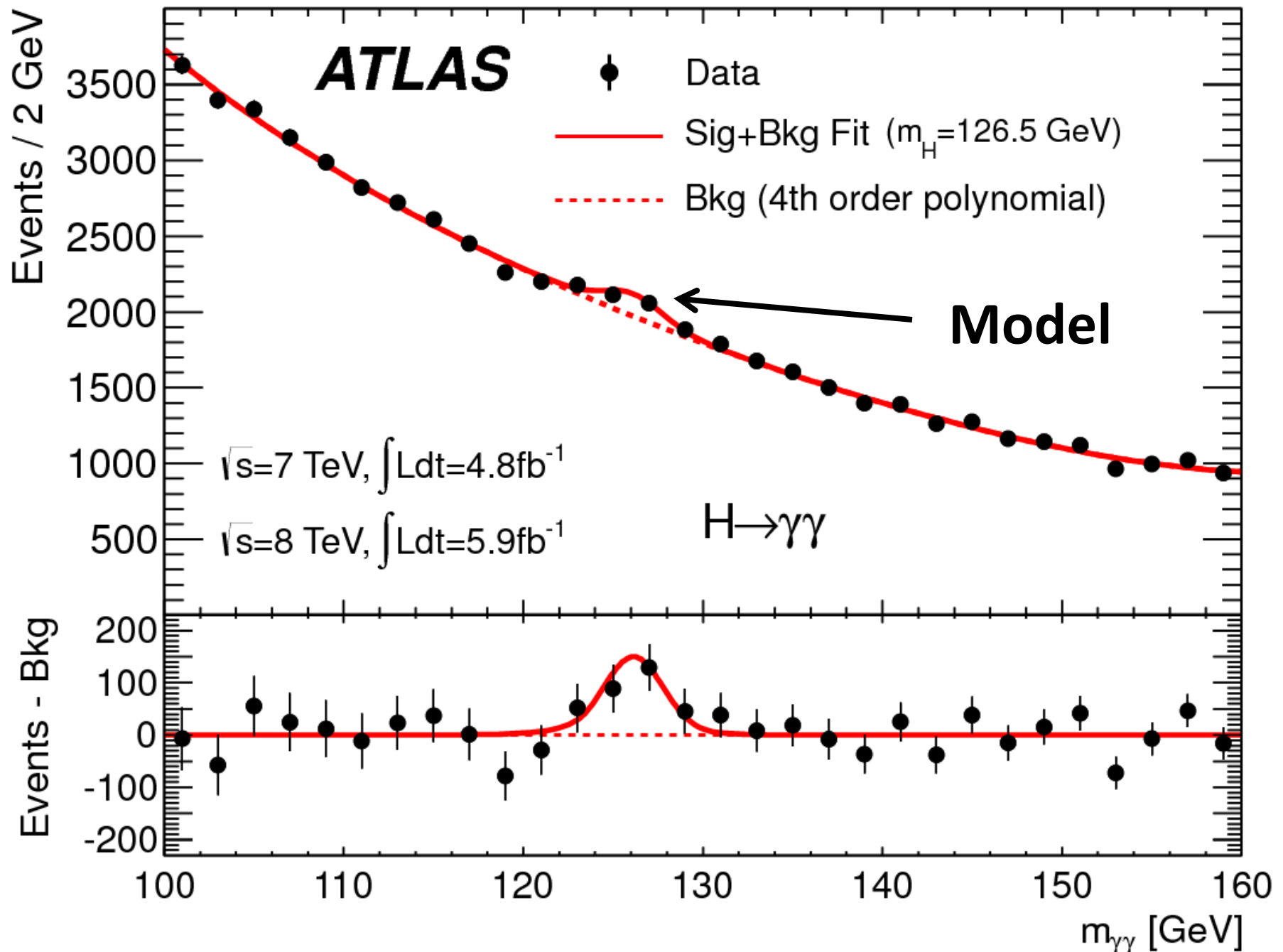
This analysis raw data → reconstructed data → AOD and TAGS → Physics is performed on the multi-tier LHC Computing Grid. Note that every event can be analyzed independently so that many events can be processed in parallel with some concentration operations such as those to gather entries in a histogram. This implies that both Grid and Cloud solutions work with this type of data with currently Grids being the only implementation today.



ATLAS Expt
Note LHC lies in a tunnel 27 kilometres (17 mi) in circumference
Higgs Event



The LHC produces some 15 petabytes of data per year of all varieties and with the exact value depending on duty factor of accelerator (which is reduced simply to cut electricity cost but also due to malfunction of one or more of the many complex systems) and experiments. The raw data produced by experiments is processed on the LHC Computing Grid, which has some 200,000 Cores arranged in a three level structure. Tier-0 is CERN itself, Tier 1 are national facilities and Tier 2 are regional systems. For example one LHC experiment (CMS) has 7 Tier-1 and 50 Tier-2 facilities.



```

1      SUBROUTINE ZERHPS(ICOM,IA)
2      ZERHPS 2
3      ZERHPS 3
4      ZERHPS 4
5      ZERHPS 5
6      C      ZERO OUT HISTOGRAM SUM OR SCATTERPLOT STORED IN IA(1,INO)
7      C      ICOM IS COMMAND NUMBER
8      C      KIO1
9      C      KIO1
10     C      STORAGE FOR INFORMATION USED BY DPLOT == POINTERS
11     C      COMMON/KIO1/NUMCOM(200),TYPCOM(200),IPTCOM(200),HPSCD(200),NOCOM,
12     C      1 NOTE5T,NEED,MASK5,MASK6
13     C      KIO1 2
14     C      KIO1 3
15     C      KIO1 4
16     C      KIO1 5
17     C      KIO1A 2
18     C      KIO1A 3
19     C      KIO1A 3
20     C      ZERHPS 8
21     C      KIO2 2
22     C      KIO2 3
23     C      KIO2 4
24     C      KIO2A 2
25     C      KIO2A 3
26     C      ZERHPS 11
27     C      KIOPRE 2
28     C      KIOPRE 3
29     C      KIOPRE 4
30     C      KIOPRE 5
31     C      KIOPRE 6
32     C      KIOPRE 7
33     C      KIOPRE 8
34     C      KIOPRE 9
35     C      KIOPRE 10
36     C      ZERHPS 13
37     C      ZERHPS 14
38     C      ZERHPS 15
39     C      ZERHPS 16
40     C      ZERHPS 17
41     C      ZERHPS 18
42     C      ZERHPS 19
43     C      ZERHPS 20
44     C      ZERHPS 21
45     C      ZERHPS 22
46     C      ZERHPS 23
47     C      ZERHPS 24
48     C      ZERHPS 25
49     C      ZERHPS 26
50     C      ZERHPS 27
51     C      ZERHPS 28
52     C      ZERHPS 29
53     C      ZERHPS 30
54     C      ZERHPS 31
55     C      ZERHPS 32
56     C      ZERHPS 33
57     C      ZERHPS 34
58     C      ZERHPS 35
59     C      ZERHPS 36
60     C      ZERHPS 37
61     C      ZERHPS 38
62     C      ZERHPS 39
63     C      ZERHPS 40
64     C      ZERHPS 41
65     C      ZERHPS 42
66     C      ZERHPS 43

```

SUBROUTINE ZERHPS(ICOM,IA)
 ZERO OUT HISTOGRAM SUM OR SCATTERPLOT STORED IN IA(1,INO)
 ICOM IS COMMAND NUMBER
 STORAGE FOR INFORMATION USED BY DPLOT == POINTERS
 COMMON/KIO1/NUMCOM(200),TYPCOM(200),IPTCOM(200),HPSCD(200),NOCOM,
 1 NOTE5T,NEED,MASK5,MASK6
 INTEGER TYPCOM,HPSCD
 STORAGE FOR INFORMATION USED BY DPLOT ==A) COMMUNAL ARRAYS
 COMMON/KIO1A/ISTCOM(1600),AQTCOM(400)
 VARIABLES USED TO SPECIFY HDS BUT NOT USED BY DPLOT == POINTERS
 COMMON/KIO2/NDHPS,ONEHPS(150),A88COM(150)
 INTEGER ONEHPS,A88COM
 VARIABLES USED TO SPECIFY HDS BUT NOT USED BY DPLOT == COMMUNAL ARRAYS
 COMMON/KIO2A/IMP8(2000),AMP8(400)
 PRESET CONSTANTS
 COMMON/KIOPRE/NDLEN,NIDLEN,NJOLEN,MAXNM1,MAXNM2,MAXNM3,NOTE5T,
 1 MAXHPS,MAXNAM,MOCOM,MSTOR1,MSTOR2,MXHP51,MXHP52,MAXTIT,
 2 NOCBIT,MACHIN,IOBFL,NDBLOC,IFILP,IRRFIL,PRETAR,
 3 NOMODE,ICORE(4),JCORE1,JCORE2,JCORE3,JSUMRY,IPERM,ISUMRY,
 4 LCHONE,LCHTOT,PAPHOD,MAXCOI,MAXTYP,MAXCRD,ICRFIL,
 5 NOUNSP,IPL0T1,IPL0T2,IPL0T3,IPL0T4,IPL0T5
 INTEGER WRDMOD(1),WRDUNS(5),PAPHOD,PRETAR
 EQUIVALENCE (WRDMOD,ICORE), (WRDUNS,IPL0T1)
 COMMON/ZERCOM/AVAL
 DIMENSION IIVAL(1)
 EQUIVALENCE (IIVAL,AVAL)
 DIMENSION IA(1)
 ITYPE= TYPCOM(ICOM)
 IF(ITYPE.LE.3) GO TO 1
 RETURN
 1 J=IPYCOM(ICOM)
 INO=ISTCOM(J+8)
 I1=NUMCOM(ICOM)
 I1=ONEHPS(I1)
 KIN=0
 GO TO (11,12,13),ITYPE
 12 HISTOGRAM
 IF(IMP8(I1+6).EQ.2) GO TO 17
 KIN=INO-14*IOBFL
 GO TO 11
 13 SCATTERPLOT
 IF(IMP8(I1+6).EQ.2) GO TO 17
 KIN=INO-30*IOBFL
 KIN INTEGER AND KFL FLOATING WORDS TO INITIALIZE
 KFL=(INO-KIN)/IOBFL
 AVAL=0.
 L=0



Geoffrey Fox

Professor of Informatics and Computing and Physics

[Cyberinfrastructure](#) - [Distributed Systems](#) - [Clouds](#) - [Parallel Computing](#) - [Particle Physics](#)

Verified email at [indiana.edu](#)

[Homepage](#)

Go

Search

My City

[« Back to list](#) [Edit](#) [Export](#) [Delete](#)

| | |
|------------------|---|
| Title | Quantum-chromodynamic approach for the large-transverse-momentum production of particles and jets |
| Authors | RP Feynman, RD Field, GC Fox |
| Publication date | 1978/11/1 |
| Journal name | Physical Review D |
| Volume | 18 |
| Issue | 9 |
| Pages | 3320 |
| Publisher | American Physical Society |
| Description | I. INIODOUCTION% e investigate whether the present experimental behavior of mesons with large transverse mo-mentum in hadron-hadron collisions is consistent with the theory of quantum-chromodynamics(QCD) with asymptotic freedom, atleast as the theory is now partially understood. It is shown that if things behave more or less according to current theoretical ideas, the experimental data at high P~ would be explicable with reasonable choices for currently unknown quantities (such as the dis-tribution of gluons in the proton and the ... |

Total citations [Cited by 373](#)

Citations per year



Scholar articles

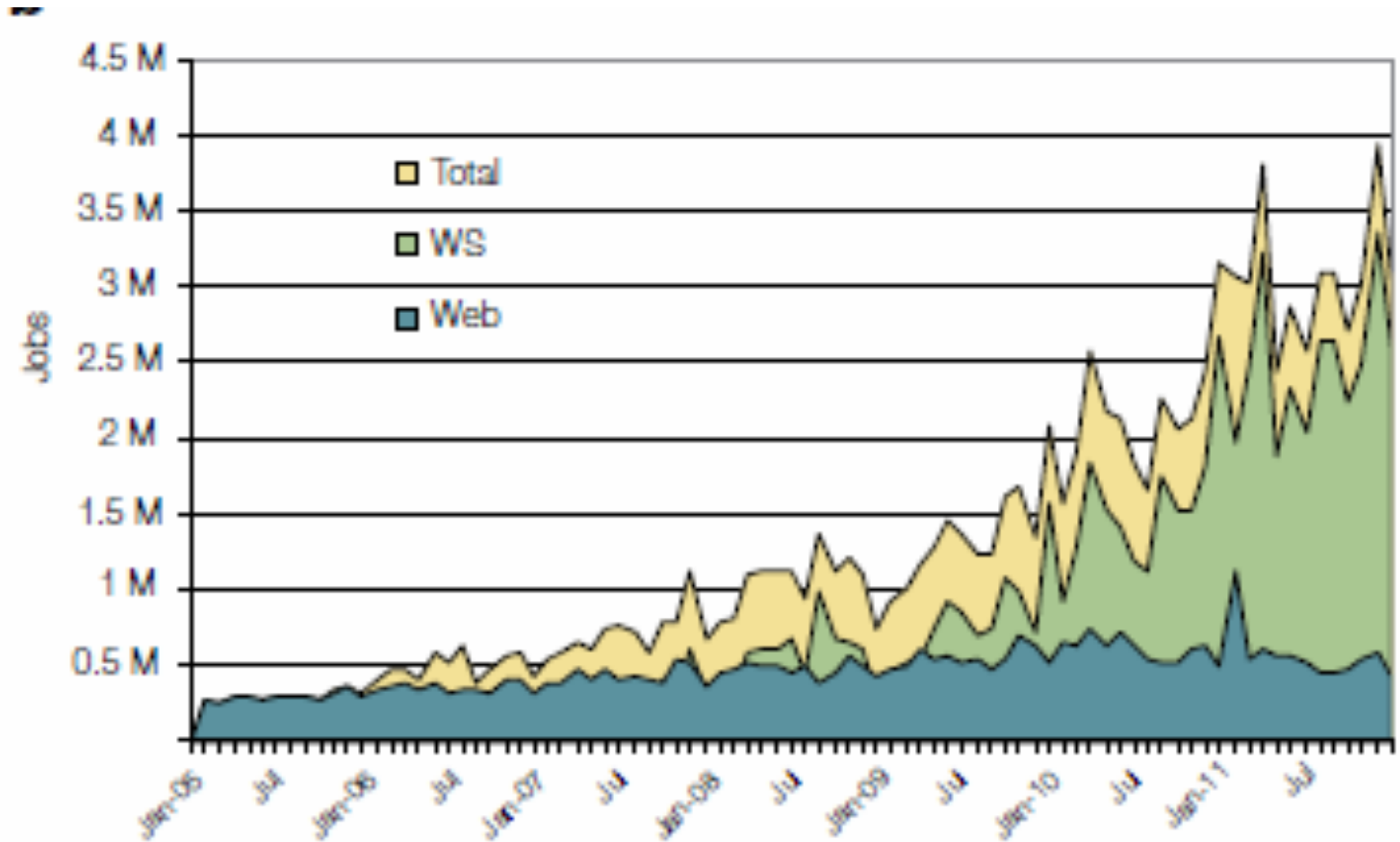
[Quantum-chromodynamic approach for the large-transverse-momentum production of particles and jets](#)
RP Feynman, RD Field, GC Fox - Physical Review D, 1978
[Cited by 373](#) - [Related articles](#) - [All 12 versions](#)

- Newton's laws and Einstein's special theory
- Physicists just discovered whose existence
- Its search was how a model is needed
- A model is a hypothesis approach that are fit to experiment

<http://en.wikipedia.org>

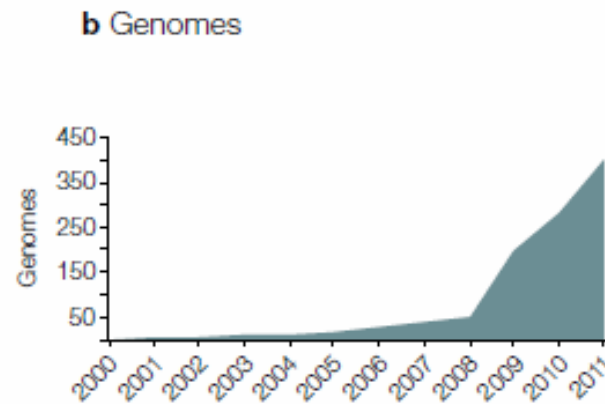
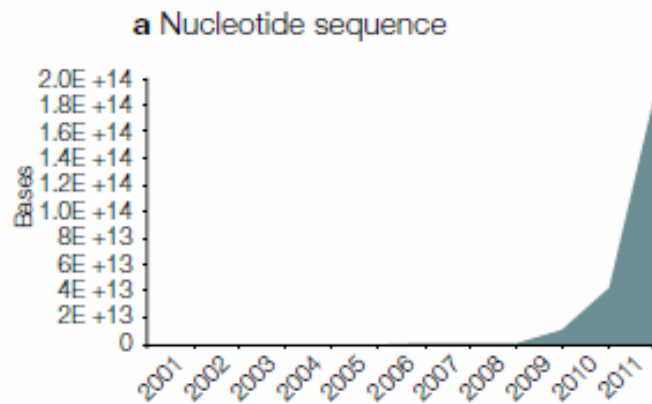
[Simple linear regression](#)

macroeconomics is a simple linear regression dependent variable (presumed to be in a the changes in the un

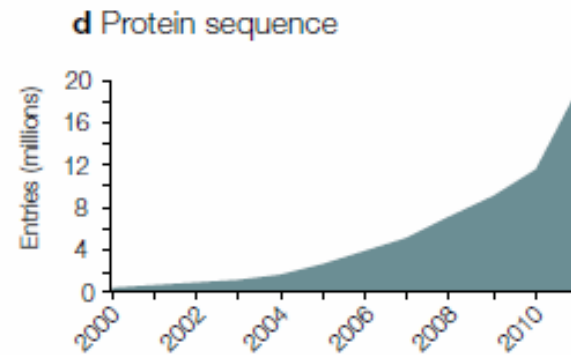
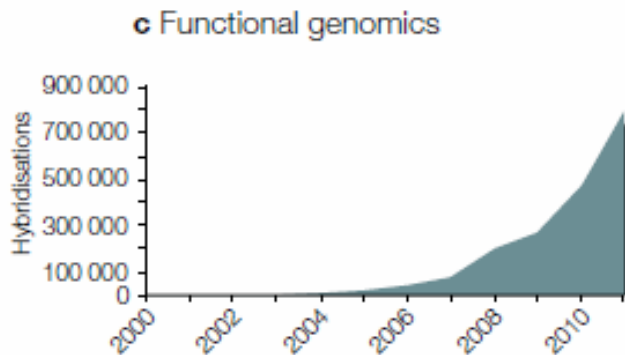


2005-20011 Job request at European Bioinformatics Institute EBI for Web hits and automated services WS

<http://www.ebi.ac.uk/Information/Brochures/>



2005-20011 Data stored
at European
Bioinformatics Institute
EBI



[http://www.ebi.ac.uk/
Information/Brochures/](http://www.ebi.ac.uk/Information/Brochures/)

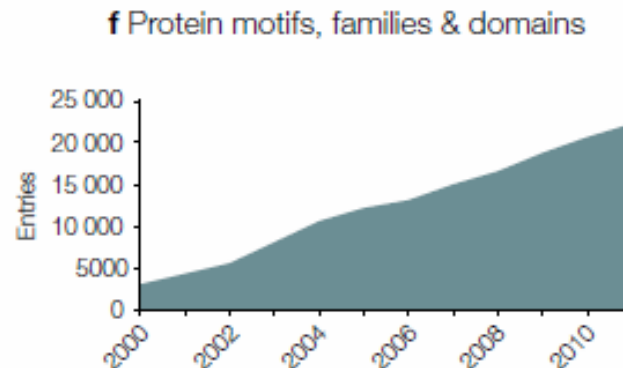
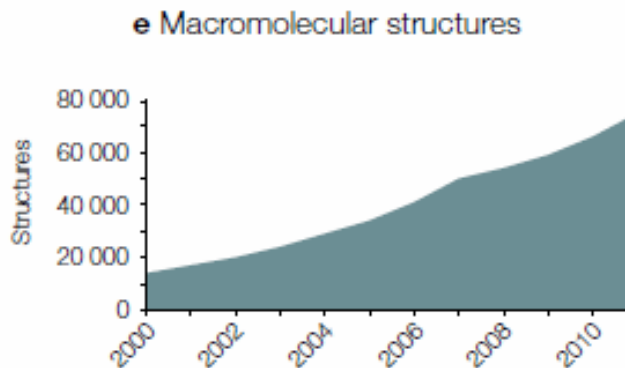


Figure 2. Growth of EMBL-EBI's core data resources from 2000 to 2011. (a) Nucleotide sequence (bases in the European Nucleotide Archive); (b) genomes (entire genomes in Ensembl plus Ensembl Genomes combined); (c) functional genomics (assays in the ArrayExpress Archive); (d) protein sequence (protein sequences in UniParc); (e) macromolecular structures (structures in PDB); (f) protein families, motifs and domains (entries in InterPro).

Data Deluge

Implications for Scientific Method

The End of Science

The quest for
knowledge used
to begin with
grand theories.
Now it begins
with massive
amounts of data.
Welcome to the
Petabyte Age.



The 4 paradigms of Scientific Research

1. Theory
2. Experiment or Observation
 - E.g. Newton observed apples falling to design his theory of mechanics
3. Simulation of theory or model
4. Data-driven (Big Data) or The Fourth Paradigm: Data-Intensive Scientific Discovery (aka Data Science)
 - <http://research.microsoft.com/en-us/collaboration/fourthparadigm/>
 - A free book
 - More data; less models

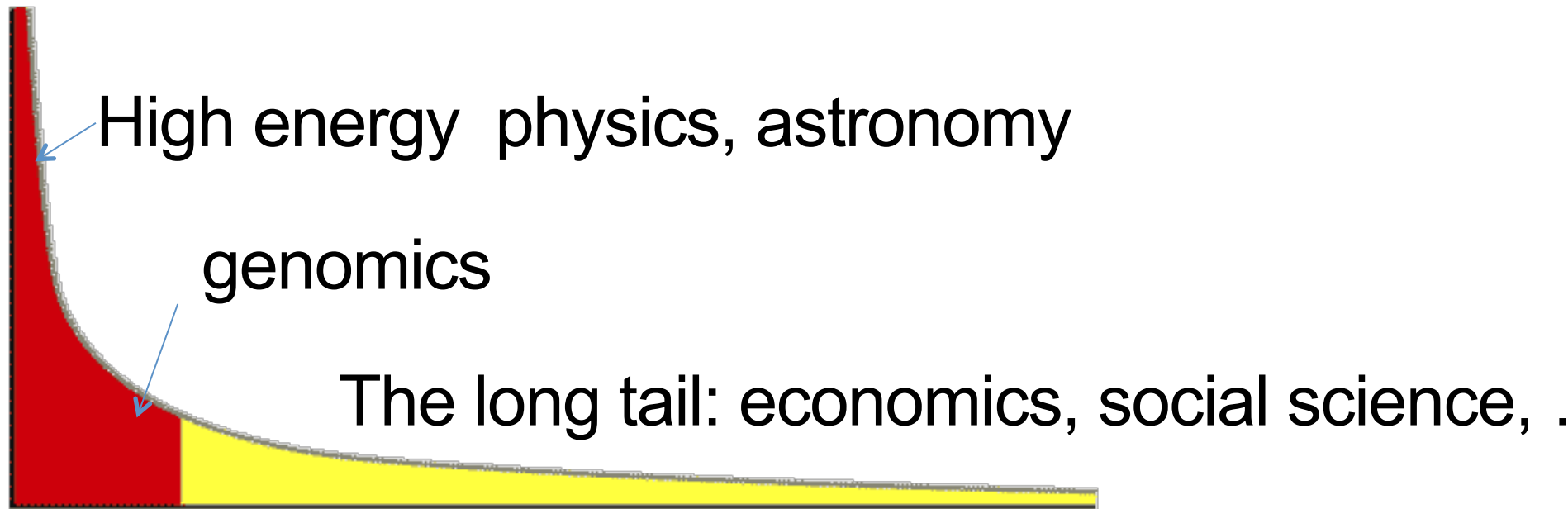
Another Personal Note

- In 1990, only methods 1 and 2 were recognized but due to increasing power of computers, method 3 (computation science) was being recognized
- I tried to persuade Caltech to adopt a “computational science curriculum” but failed
 - I left Caltech partly for this reason
- I now realize that perhaps not such a good idea as not huge numbers of jobs in area.
- However starting in 2005-2010, method 4 and data science emerges
 - There are lots of jobs in data science so curricula perhaps more interesting

Data Deluge

Long Tail of Science

The Long Tail of Science



Collectively “long tail” science is generating a lot of data
Estimated at over 1PB per year and it is growing fast.

80-20 rule: 20% users generate 80% data but not necessarily 80% knowledge

Data Deluge

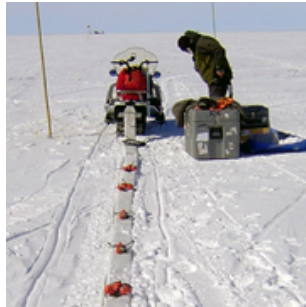
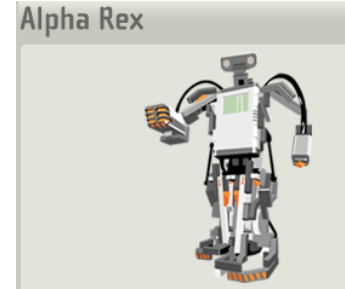
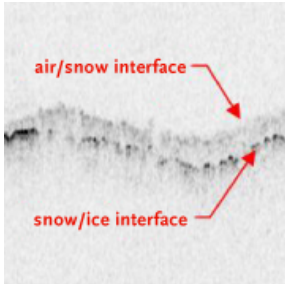
Internet of Things

Internet of Things and the Cloud

- It is projected that there will be **24 billion devices** on the Internet by 2020. Most will be small sensors that send streams of information into the cloud where it will be processed and integrated with other streams and turned into knowledge that will help our lives in a multitude of small and big ways.
- The **cloud** will become increasingly important as a controller of and **resource provider for the Internet of Things**.
- As well as today's use for smart phone and gaming console support, "Intelligent River" "smart homes and grid" and "ubiquitous cities" build on this vision and we could expect a growth in cloud supported/controlled **robotics**.
- Some of these "things" will be supporting science
- Natural parallelism over "things"
- "Things" are distributed and so form a Grid

Sensors (Things) as a Service

Output Sensor



A larger sensor

