



# **How Big is “Big Data” Across Disciplines: Workshop 1 Analysis and lessons learned**

**Gul Kremer, Program Officer  
Division of Undergraduate Education**

**Elizabeth Burrows, AAAS Big Data Fellow  
Division of Mathematical Sciences**

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# Observations

- ◆ Efficiency in scientific discovery through curation, analyses and interpretation of massive datasets
- ◆ Uptake level and concentration on “Big Data” opportunities are varied across disciplines
  - The nature of the data needed within disciplinary communities
  - Characterization using Velocity, Variety, Veracity, Volume typology

# Definition

“Big Data consists of extensive datasets primarily in the characteristics of volume, variety, velocity, and/or variability that require a scalable architecture for efficient storage, manipulation, and analysis.”

NIST

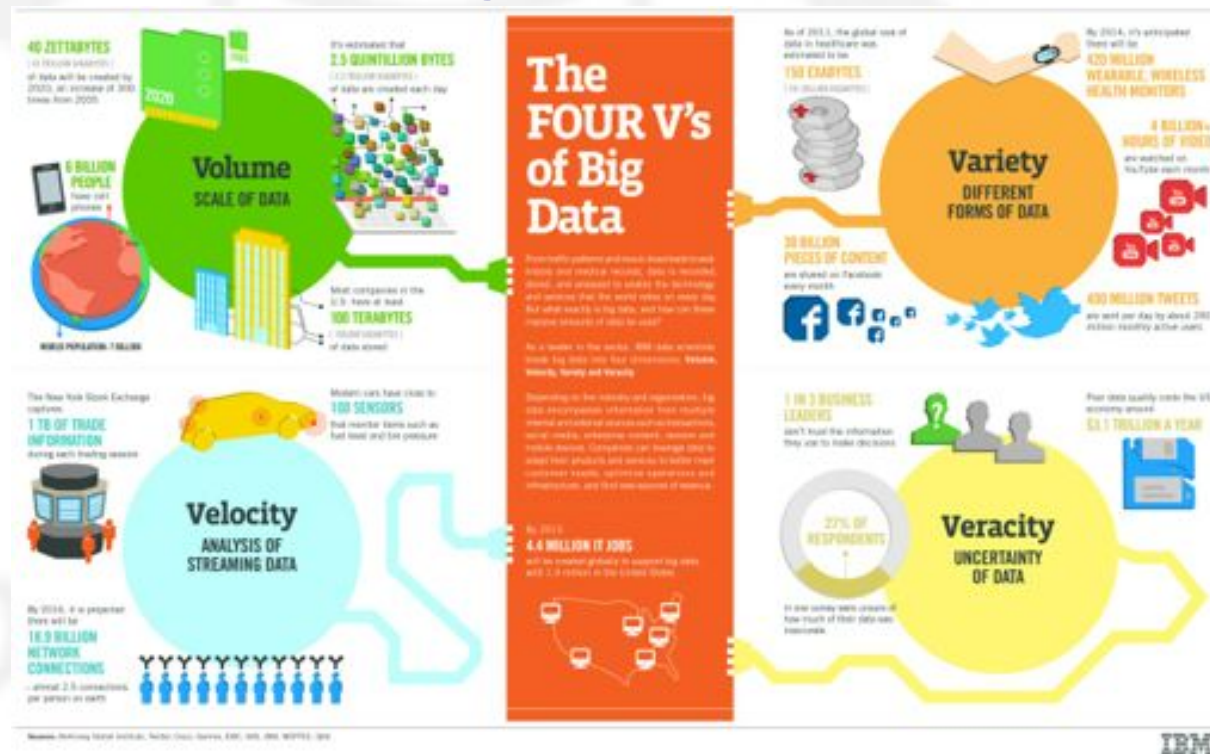
# Workshop 1 Overview

- **5 case studies** of effective partnerships  
**outside of education** btw producers & consumers
  - Earth Science
  - Biology
  - Astronomy
  - Health
  - Computer Science
- Breakouts
- Methods & Analytics

# Focus of Case Studies

For each case study we identified:

- What makes this “big data”?



From: <http://www-1.ibm.com/software/data/bigdata/>

# Projects that exemplify the V's

|  |   |
|--|---|
| <u><b>Volume</b></u> <ul style="list-style-type: none"><li>- LSST</li><li>- Climate</li></ul>                                | <u><b>Variety</b></u> <ul style="list-style-type: none"><li>- S &amp; C Health</li><li>- Plant Genomics</li></ul> |
| <u><b>Velocity</b></u> <ul style="list-style-type: none"><li>- Reality Deck</li><li>- SARS Outbreak</li><li>- LSST</li></ul> | <u><b>Veracity</b></u> <ul style="list-style-type: none"><li>- Climate</li></ul>                                  |

How do these projects relate to yours?



# Volume (from Borne):

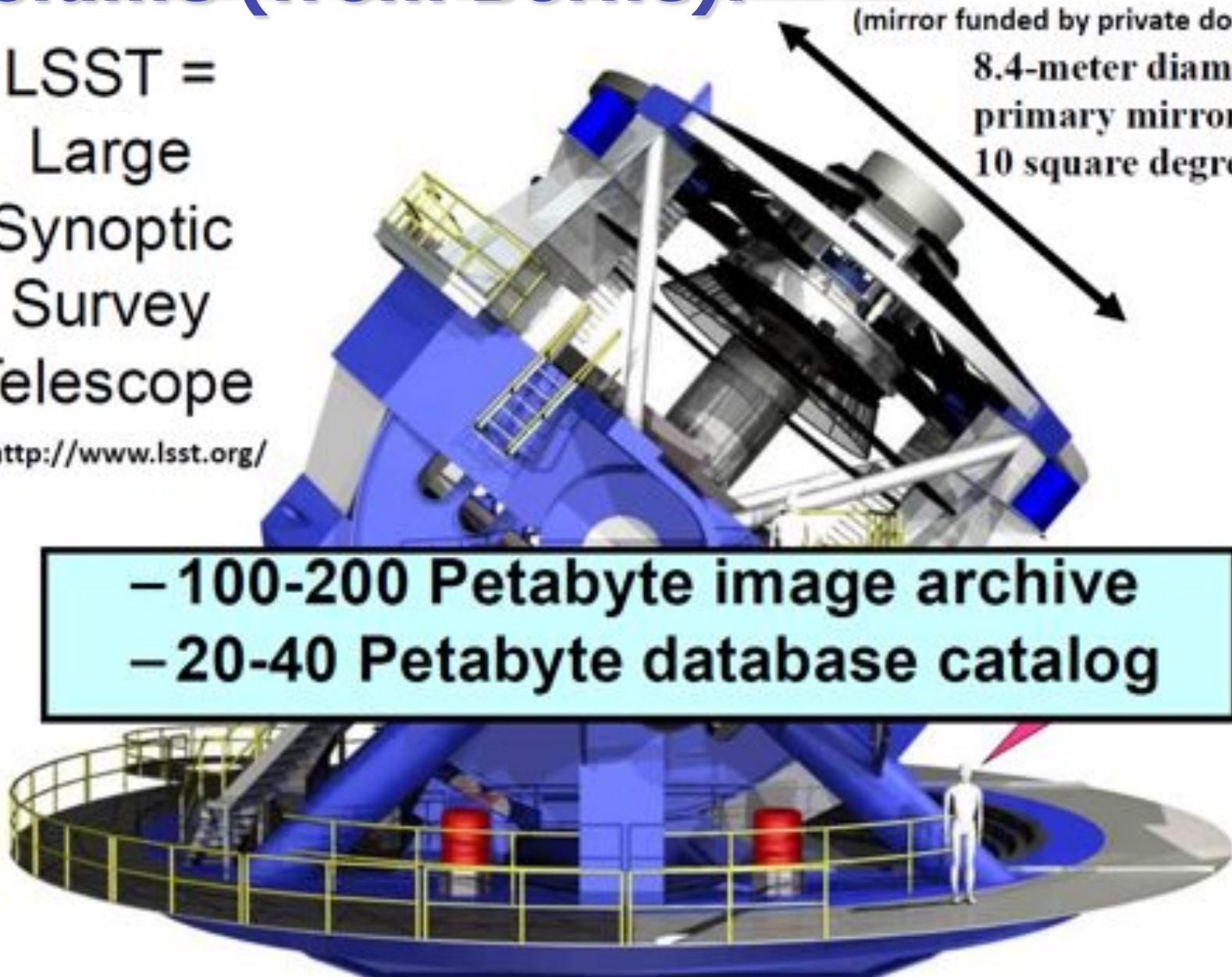
LSST =  
Large  
Synoptic  
Survey  
Telescope

<http://www.lsst.org/>

(mirror funded by private donors)

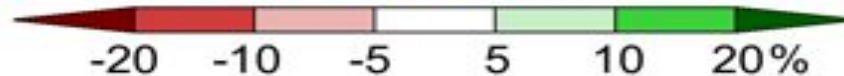
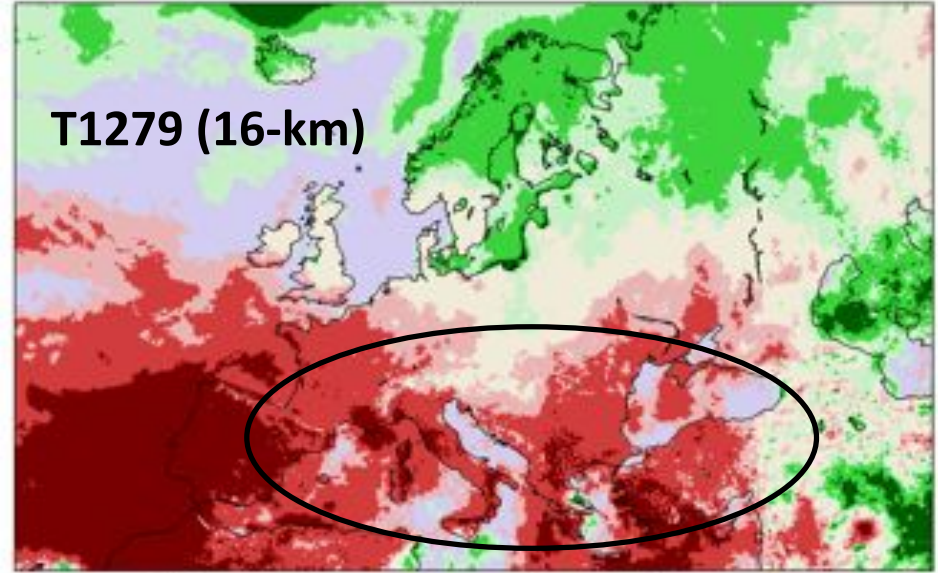
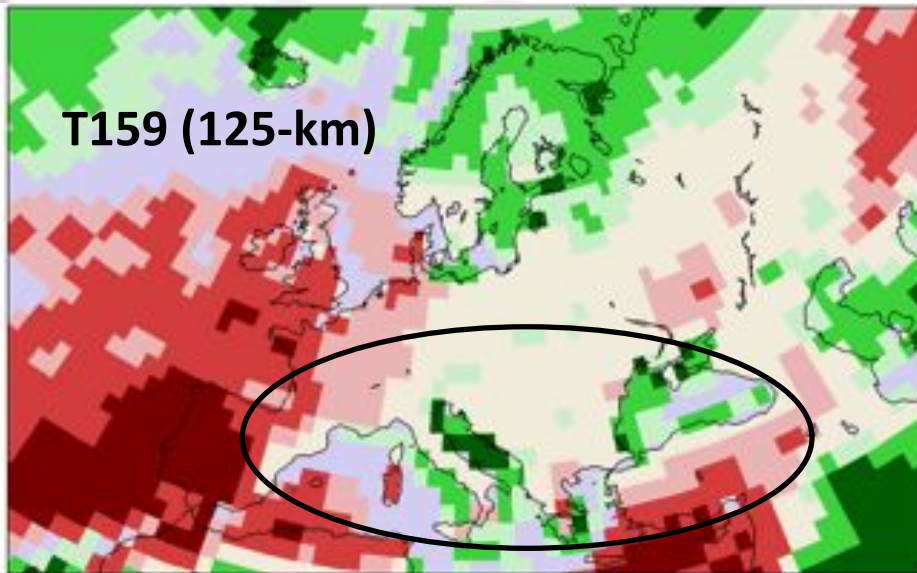
8.4-meter diameter  
primary mirror =  
10 square degrees!

- 100-200 Petabyte image archive
- 20-40 Petabyte database catalog



# Volume (from Cash, Project Minerva):

Growing Season Precip. Change: 20<sup>th</sup> C - 21<sup>st</sup> C



- IBM iDataplex, 72,280 cores, **1.5 petaFLOPS peak** performance
- #17 on June 2013 Top500 list of supercomputing sites
- **10.7 PB disk capability**
- **10x increase in FLOPS, 100x increase in storage over Athena**

This much data breaks everything: H/W, systems management policies, networks, apps S/W, tools, and shared archive space



# Velocity (from Borne):

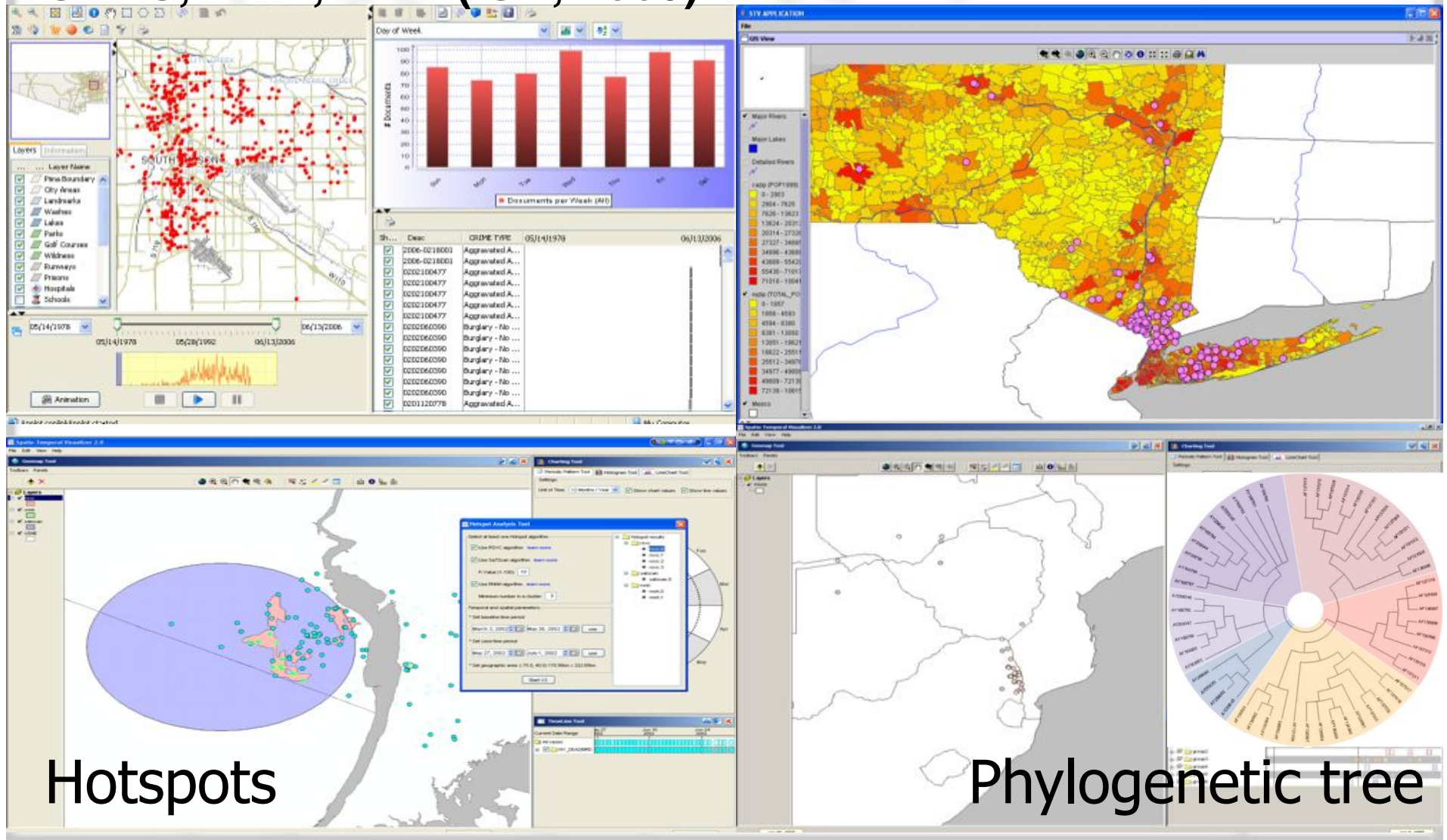
## *LSST Key Science Drivers: Mapping the Dynamic Universe*

- Complete inventory of the Solar System in **REAL-TIME**
  - **EVENT MINING:** ~10 million events per night, every night, for 10 years!
    - Follow-up observations required to classify these
    - Which ones should we follow up? ...
    - ... Decisions! Decisions! Data-to-Decisions!
  - Repeat images of the entire night sky every 3 nights
  - One 6-GB image every 20 seconds
    - Near-Earth Objects; killer asteroids???
    - Exploding supernovae



# Velocity (from Chen):

BioPortal: Infectious Disease Tracking and Visualization,  
SARS, WNV, FMD (ISR, 2009)



Hotspots

Phylogenetic tree

# Velocity (from Kaufman):

The Reality Deck at Stony Brook University

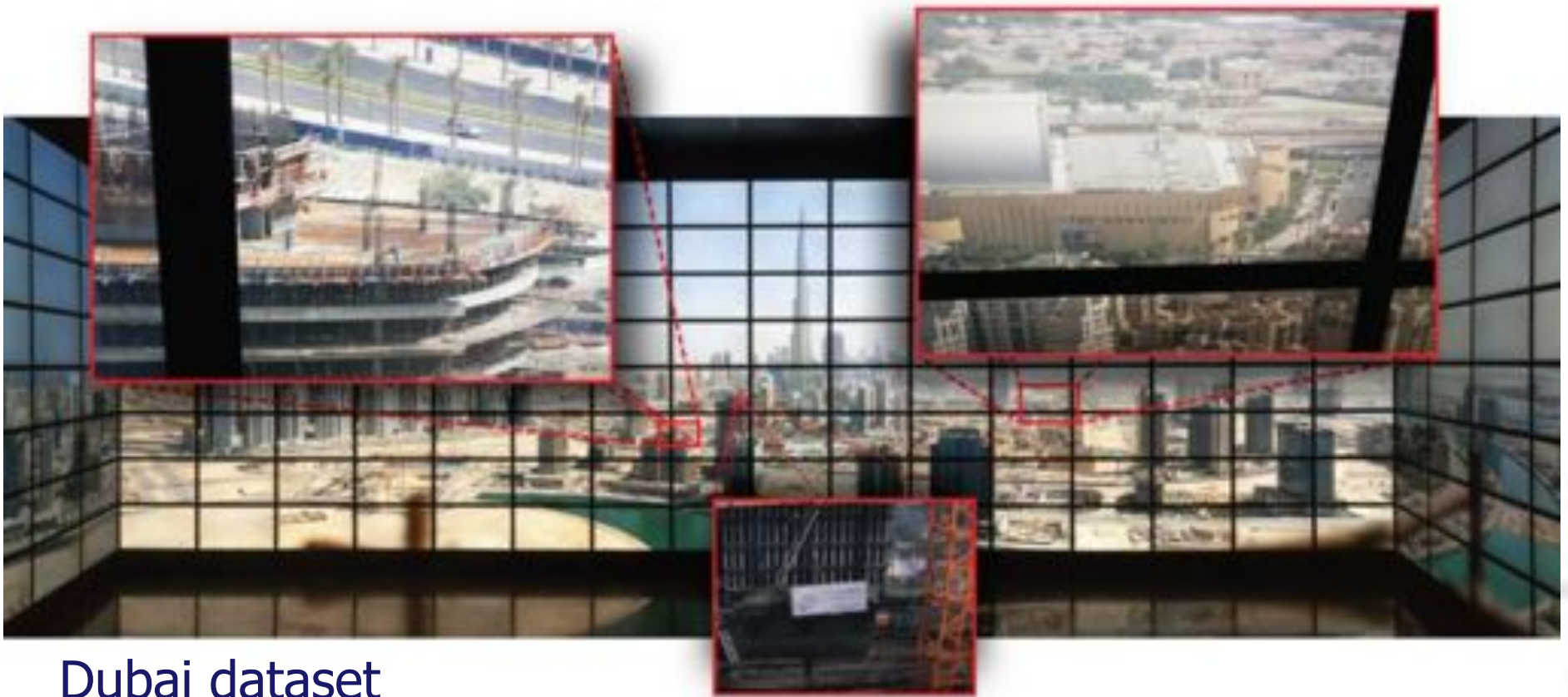




# Velocity (from Kaufman):

Reality Deck numbers:

- 1.5 Gigapixels
- 240 CPU cores: 2.3 TFLOPS, 1.2 TB distributed memory



Dubai dataset

# Variety (from Chen):

## Smart & Connected Health: From Medicine to Health

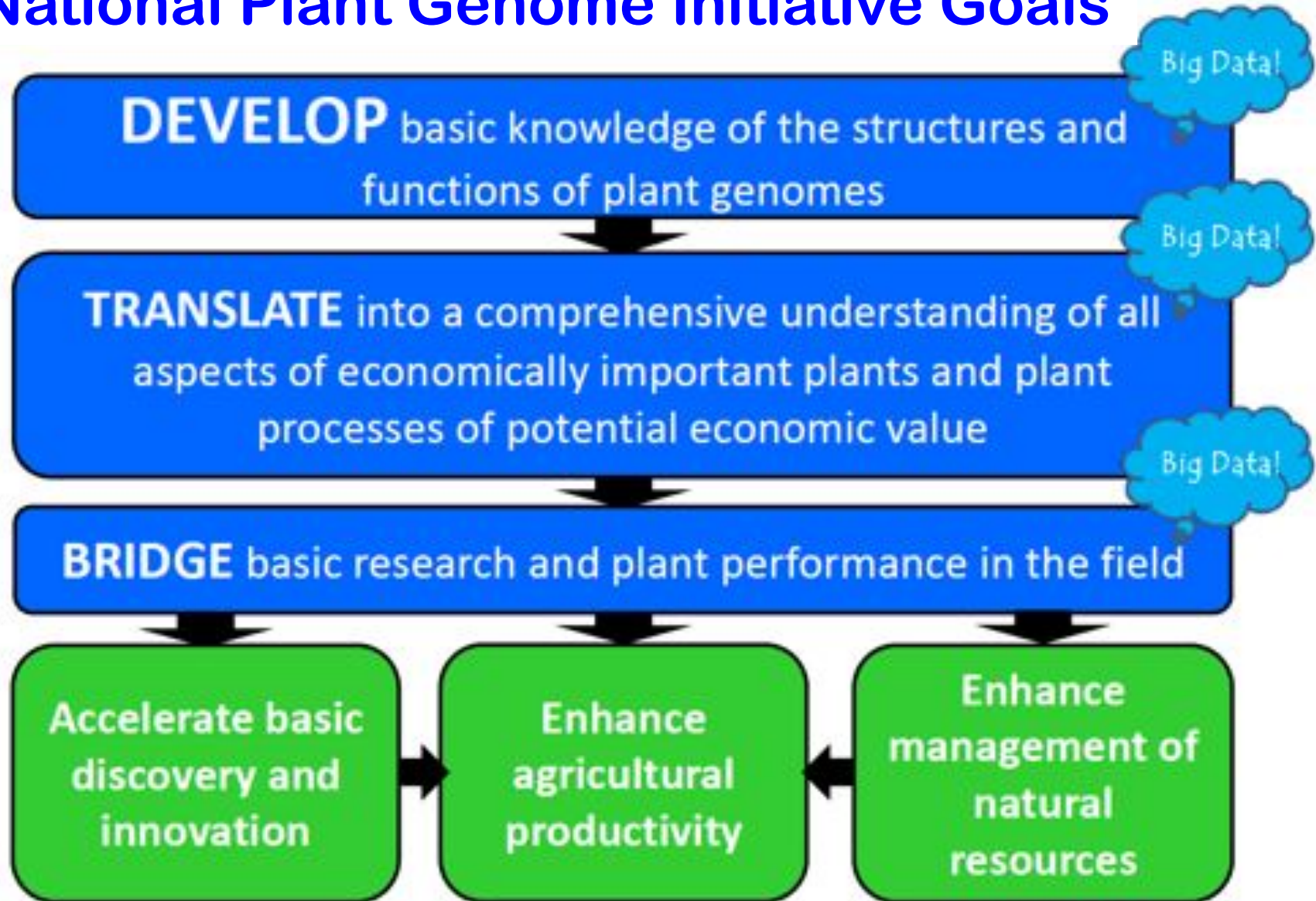


(Source: Dr. Howard Wactlar, IEEE IS, 2012; NSF)



# Variety (from Okamuro):

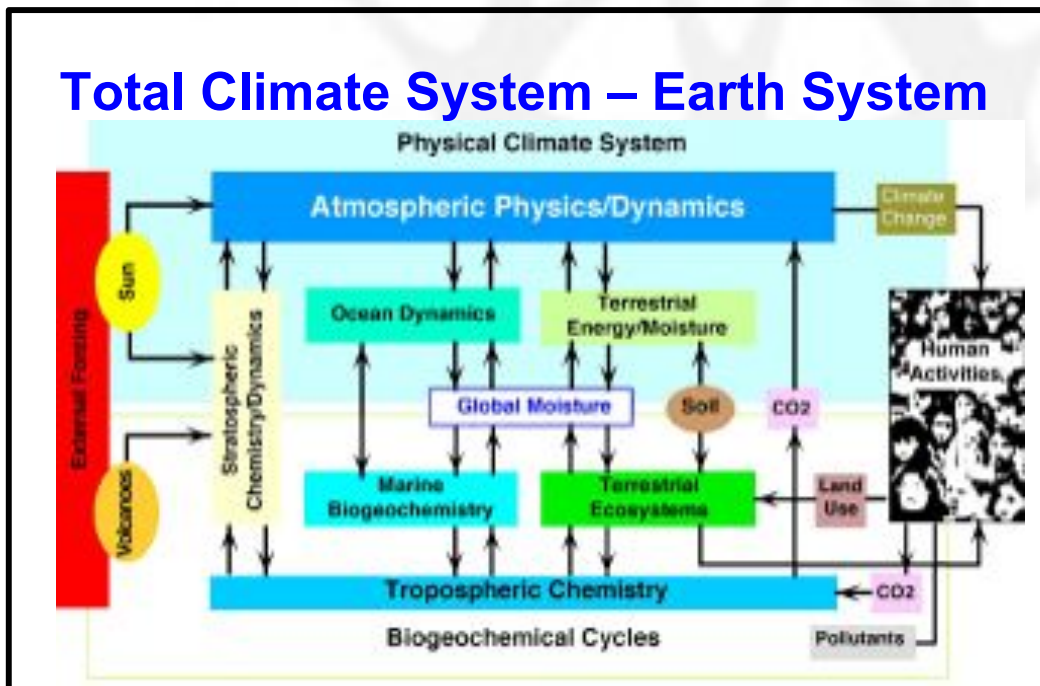
## National Plant Genome Initiative Goals



# Veracity (from Cash):



## Total Climate System – Earth System



(from Earth System Science: An Overview, NASA, 1988)

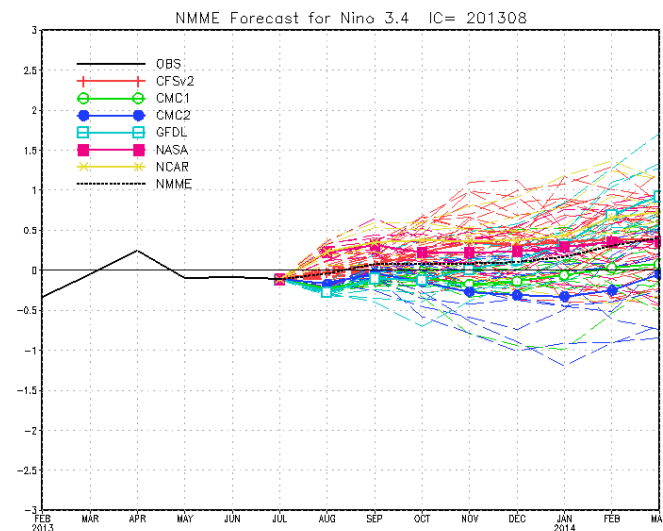


Fig. 12. Real-time Niño-3.4 predictions

Kirtman et al., 2014

Funded by NOAA, NSF, NASA, & DOE

# Some Lessons Learned

- Collaborate!
- Landscape rapidly Δing
- Get exaflood insurance ☺



Kirk Borne  
@KirkDBorne

Storage cost of 1GB:

1981 \$300K

1987 \$50K

1990 \$10K

2000 \$10

2004 \$1

2012 \$0.10

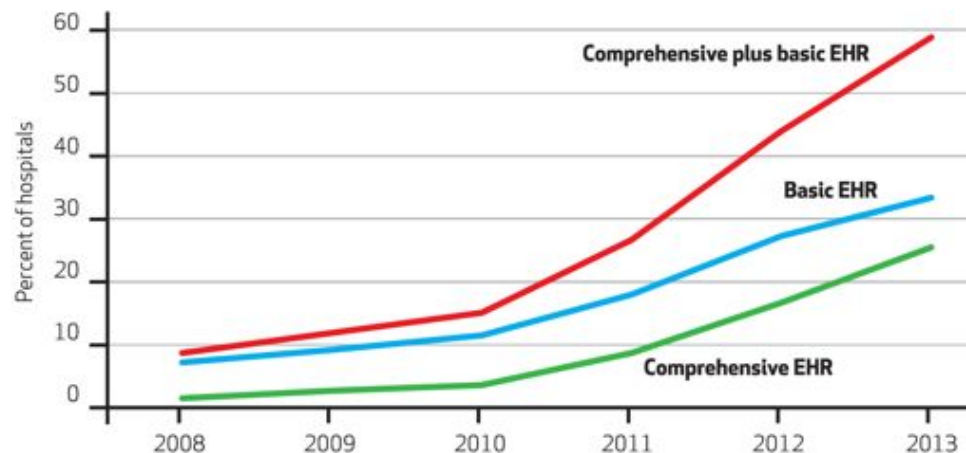
**2015 FREE**

50GB—BOX

15GB—GoogleDrive

5GB—iCloud

Hospitals' Adoption Of Electronic Health Record (EHR) Systems, 2008-13





Source: Adler-Milstein, J., DesRoches, C. M., Furukawa, M. F., Worzala, C., Charles, D., Kralovec, P., Stalley, S., and Jha, A. K. 2014. "More Than Half of US Hospitals Have At Least A Basic EHR, But Stage 2 Criteria Remain Challenging For Most," *Health Affairs* (33:9), pp. 1664–1671.



**Thank you!**

Questions?

Discussion may continue at the  
Producer/ Consumer Relationships  
breakout tomorrow





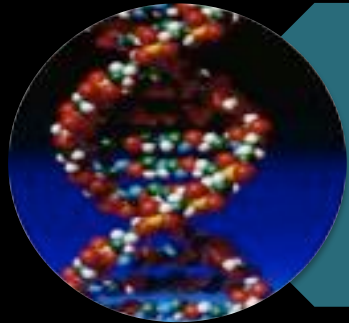
# Earth Sciences Lessons Learned

- Spatial resolution alone is not a panacea
- Validating high-resolution, high-complexity data pushes and in some cases exceeds observational capabilities
- Data from simulations can inform the veracity of the observational data.



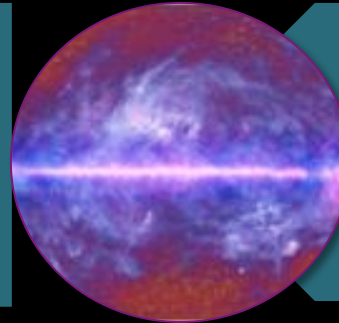
# **“Big Data” Challenges in Science**

## ***Volume, velocity, variety, and veracity***



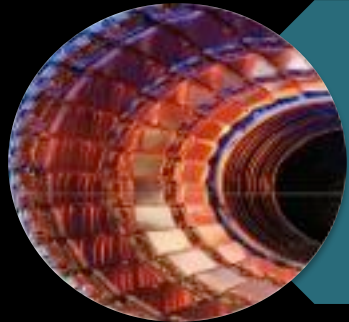
### **Biology**

- *Volume*: Petabytes now; computation-limited
- *Variety*: multi-modal analysis on bioimages



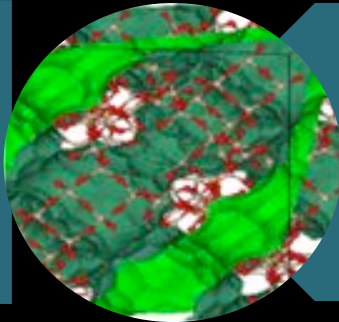
### **Cosmology & Astronomy:**

- *Volume*: 1000x increase every 15 years
- *Variety*: combine data sources for accuracy



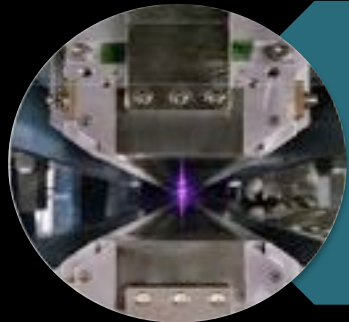
### **High Energy Physics**

- *Volume*: 3-5x in 5 years
- *Velocity*: real-time filtering adapts to intended observation



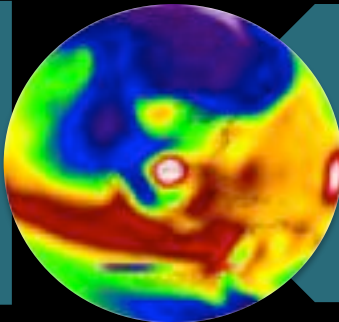
### **Materials:**

- *Variety*: multiple models and experimental data
- *Veracity*: quality and resolution of simulations



### **Light Sources**

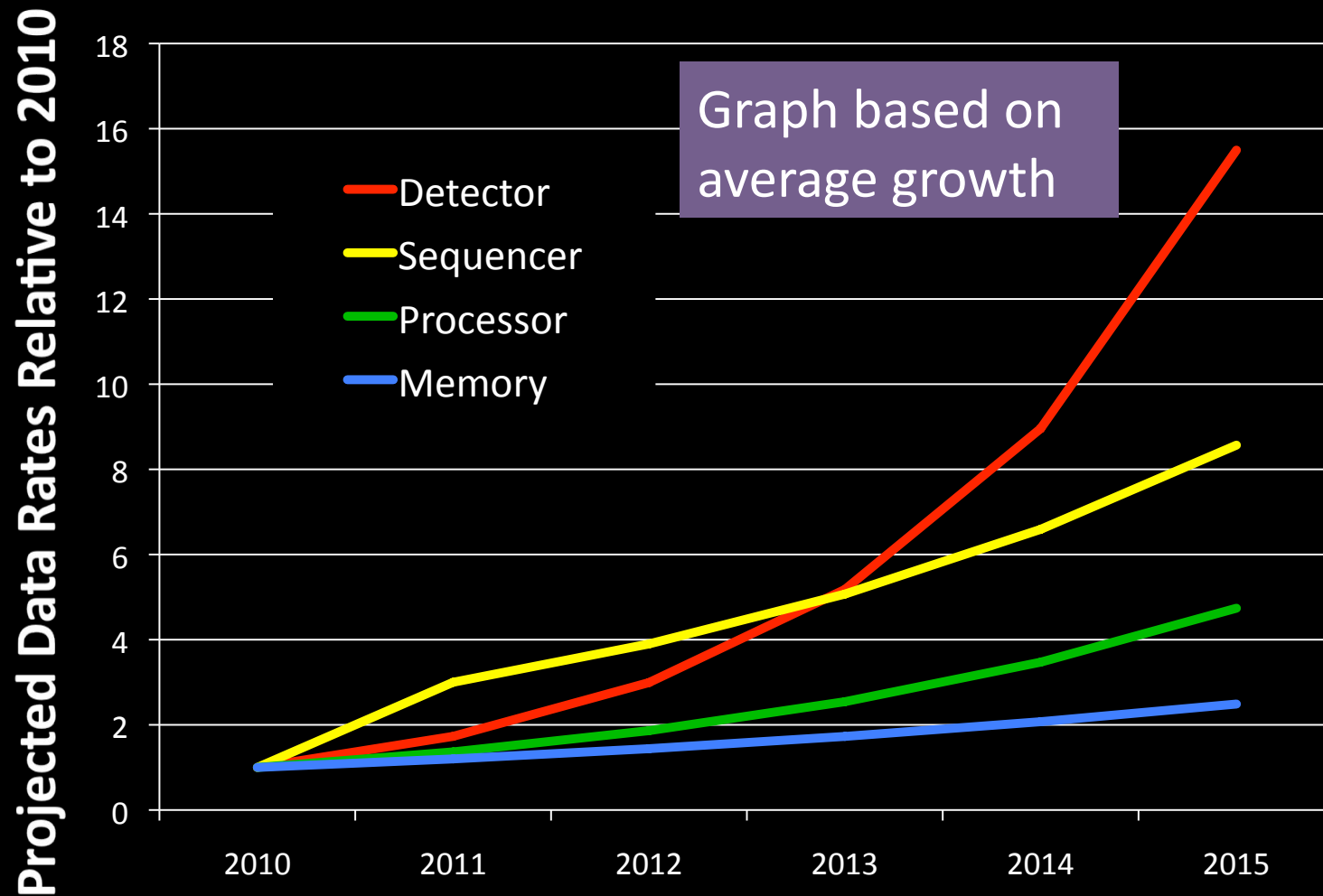
- *Velocity*: CCDs outpacing Moore's Law
- *Veracity*: noisy data for 3D reconstruction



### **Climate**

- *Volume*: Hundreds of exabytes by 2020
- *Veracity*: Reanalysis of 100-year-old sparse data

# Data Growth is Outpacing Computing Growth



Source: Kathy Yelick



Kirk Borne  
@KirkDBorne

### Storage cost of 1GB:

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2004 \$1

2012 \$0.10

**2015 FREE**

50GB—BOX

15GB—GoogleDrive

5GB—iCloud

|       |                  |               |
|-------|------------------|---------------|
| 2006  | CMIP3 (IPCC AR4) | 36 TB         |
| 2010  | Project Athena   | 1.2 <b>PB</b> |
| 2011  | CMIP5 (IPCC AR5) | 3 <b>PB</b>   |
| 2014  | Project Minerva  | 3+ <b>PB</b>  |
| 2011- | NMME             | 1 <b>PB</b>   |
| 2015- | COLA storage     | 1 <b>PB</b>   |

**Participants: ~50, 13 non-NSF**



# Workshop Goals

- Build capacity at NSF for using big data in education
- Articulate the conditions for success for effective usage of big data
- Study models of effective partnerships between sources of big data and its consumers
- Publish a volume that describes insights from the workshops



# Focus of Case Studies

- Types of data wanted
- What makes this “big data”?
- Infrastructure, funding, policies needed
- Issues of data standards & interoperability
- Issues of privacy, security, & ethics
- Methods used by producers & consumers
- Issues of limited capacity
- Partnerships involved
- How has big data changed your field?
- Advice for others

# Biology

- Doreen Ware, USDA & Cold Spring Harbor
  - Biology has become an information science
  - Genome sequencing – all big data issues apply
  - MAKER-P and Gramene project
- Diane Okamuro, Plant Genome Research Program
  - National Plant Genome Initiative – 17 yrs
    - 2014-2018 goals focused on open access and enhancing usability, big data
    - Companies are now ready to collaborate

# Astronomy

- Lucy Fortson, University of Minnesota
  - Zooniverse
    - Citizen science – galaxy classification
    - Not producer/consumer – seamless knowledge discovery system
- Kirk Borne, George Mason University
  - Data Literacy For All: Astrophysics and Beyond
  - Undergrad data science program since 2007
  - Creating and storing data as fast as capabilities
  - LSST: Large Synoptic Survey Telescope

# Methods & Analytics

- Barry Sloane, EHR/DRL
  - Analysis of numerical data
- Piotr Mitros, edX
  - Machine learning, analysis of non-numerical data

## **Breakout sessions**

- ◆ **Infrastructure, sharing, & standards**
  - Notetakers: Al, Jay, Brandi
- ◆ **Privacy, security, & ethics**
  - Notetaker: Renata
- ◆ **Capacity & producer/ consumer relationships**
  - Notetakers: Quincy & Lida



# Follow-up Activities

- Slides will be posted
  - Document produced
- Fellows will be invited to discussions
- Larger, education-based workshop May 13-15
- Volume published