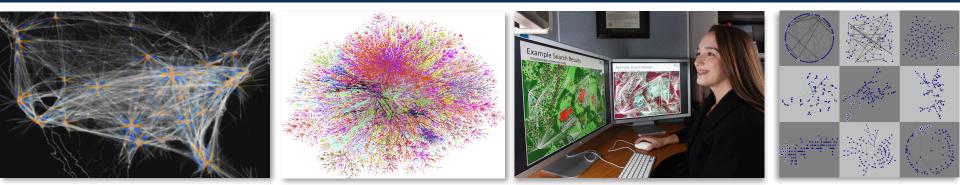
Exceptional service in the national interest





Increasing Coherence Between Simulation and Data Analytics

Chesapeake Large Scale Data Analytics Conference Annapolis, MD **Rob** Leland October 25, 2016

Vice President, Science & Technology Chief Technology Officer Sandia National Laboratories



Sandia National Laboratories is a multi-mission laboratory managed and operated by Sandia Corporation, a wholly owned subsidiary of Lockheed Martin Corporation, for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-AC04-94AL85000.



- A tale of two visions
- Some background
- A charge from the National Strategic Computing Initiative

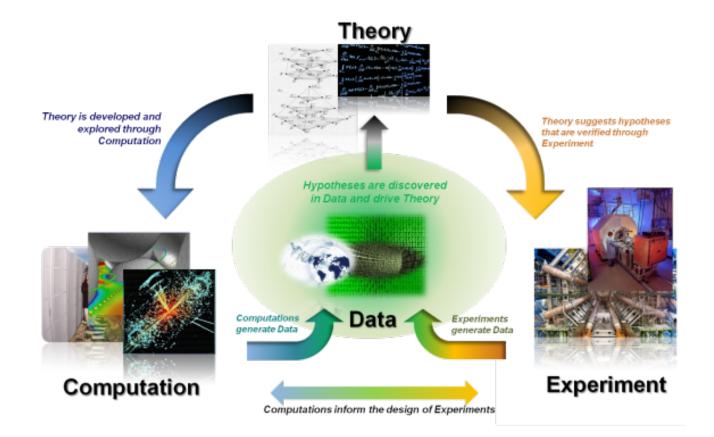
Answers to three key questions

- Why is an increasing coherence between simulation and analytics important?
- What is really meant by "increasing coherence" between the two?
- How might coherence be furthered in practice?

A unifying vision

Vision 1: From a scientific perspective



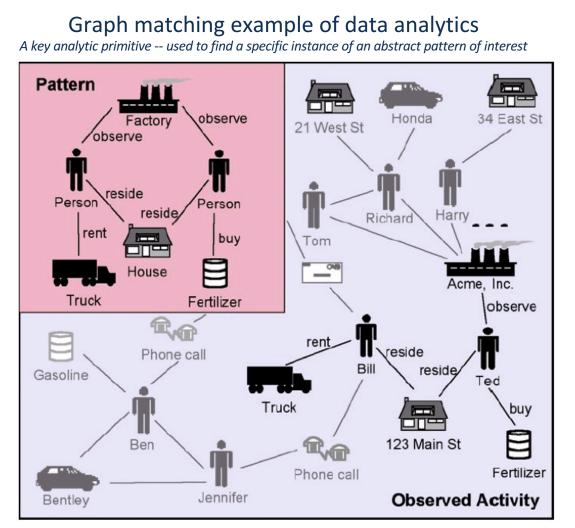


Data analysis complements theory, experiment, and computation

From The Fourth Paradigm: Data-Intensive Scientific Discovery by Jim Gray

Vision 2: From a national security perspective





From Coffman, Greenblatt, and Marcus, *Graph-Based Technologies for Intelligence Analysis*, Communications of the ACM, 47, March 2004.

Some background



Simulation

Computations to understand physical phenomena or conduct engineering

Large Scale Data Analytics (LSDA)

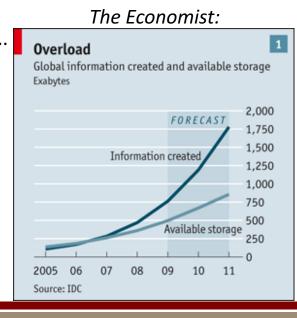
- Data Analytics = Discovering meaningful patterns in data
- Large Scale = Requiring leading-edge processing and storage capabilities

LSDA is increasing in importance

- Pervasive
 - Commerce, finance, health care, science, engineering, national security, ...
- Lasting societal significance
 - Internet search, genomics, climate modeling, Higgs particle, ...

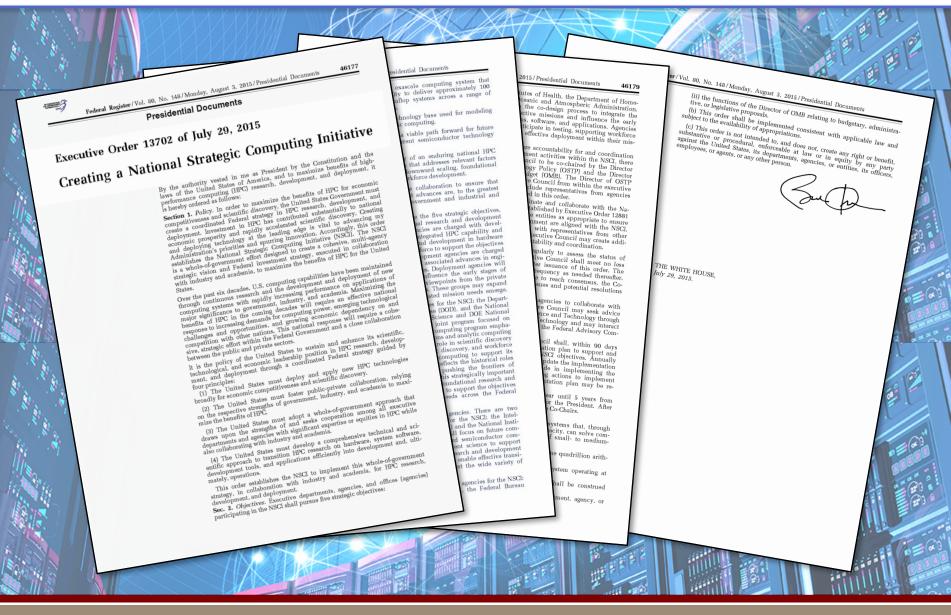
LSDA is getting "harder"

- Captured data growing exponentially with time
- Individual analysis becoming more sophisticated
- More people examining more data more frequently
- Aggregate work growing much faster than Moore's Law



National Strategic Computing Initiative (NSCI)





NSCI Strategic Objectives



- (1) Accelerating delivery of a capable exascale computing system that integrates hardware and software capability to deliver approximately 100 times the performance of current 10 petaflop systems across a range of applications representing government needs.
- (2) Increasing coherence between the technology base used for modeling and simulation and that used for data analytic computing.
- (3) Establishing, over the next 15 years, a viable path forward for future HPC systems even after the limits of current semiconductor technology are reached (the "post-Moore's Law era").
- (4) Increasing the capacity and capability of an enduring national HPC ecosystem by employing a holistic approach that addresses relevant factors such as networking technology, workflow, downward scaling, foundational algorithms and software, accessibility, and workforce development.
- (5) Developing an enduring public-private collaboration to ensure that the benefits of the research and development advances are, to the greatest extent, shared between the United States Government and industrial and academic sectors.

Q1: Why is increasing coherence between simulation and analytics important?



For simulation

- HPC simulation must ride on some commodity curve
- Larger market forces behind analytics
- Can exploit commodity component technology from analytics

For analytics

- Large Scale Data Analytics problems becoming ever more sophisticated
- Requiring more coupled methods
- Can exploit architectural lessons from HPC simulation

For both: Integration of simulation and analytics in the same workflow

- Automation of analysis of data from simulation
- Creation of synthetic data via simulation to augment analysis
- Automated generation and testing of hypothesis
- Exploration of new scientific and technical scenarios

• ...

Mutual inspiration, technical synergy, and economies of scale in the creation, deployment, and use of HPC resources

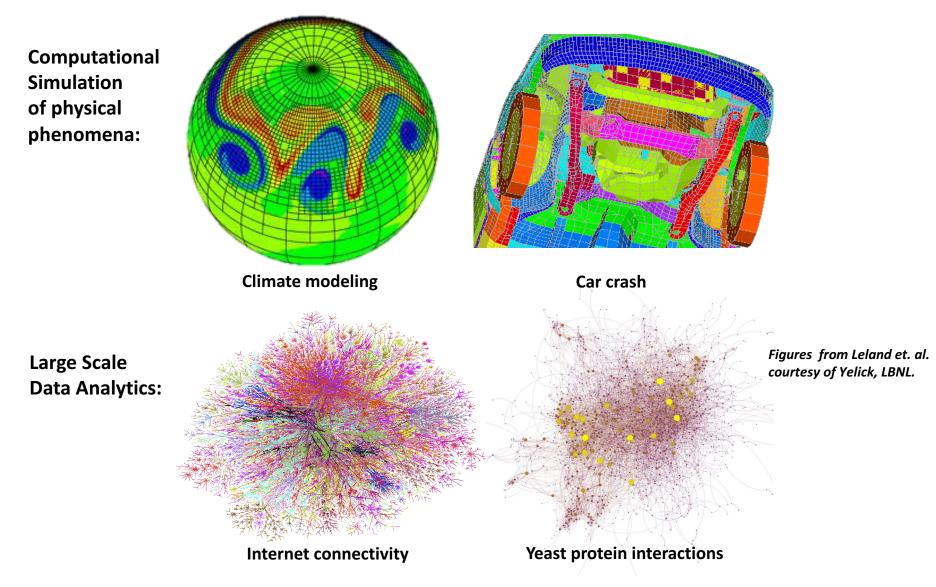


A challenge because simulation and analytics differ in many respects ...

Data structures describing simulation and analytics differ

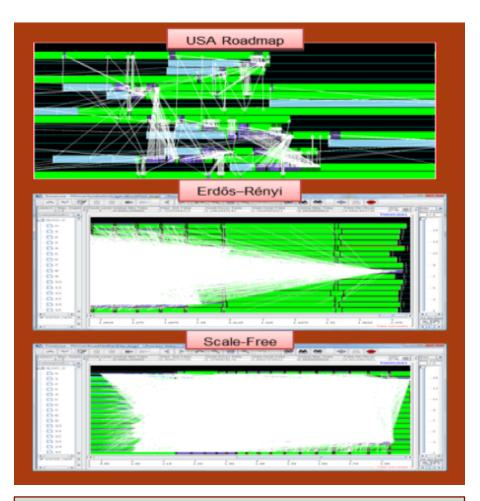
Graphs from simulations may be irregular, but have more locality than those derived from analytics





Computation and communication patterns differ





Black = time spent computing
Green = time spent communicating
White = time spent waiting for data to be communicated

The U.S. roadmap, which has spatial locality and is thus most similar of the three in structure to computational patterns that would arise in typical physical simulations.

The *Erdős-Rényi* graph, a well-studied example in graph theory work.

A scale-free graph, an example more reflective of real-world networks.

Figure from Leland et. al. courtesy of Johnson, PNNL.

Memory performance demands differ

A key differentiator in the performance of simulation and analytics



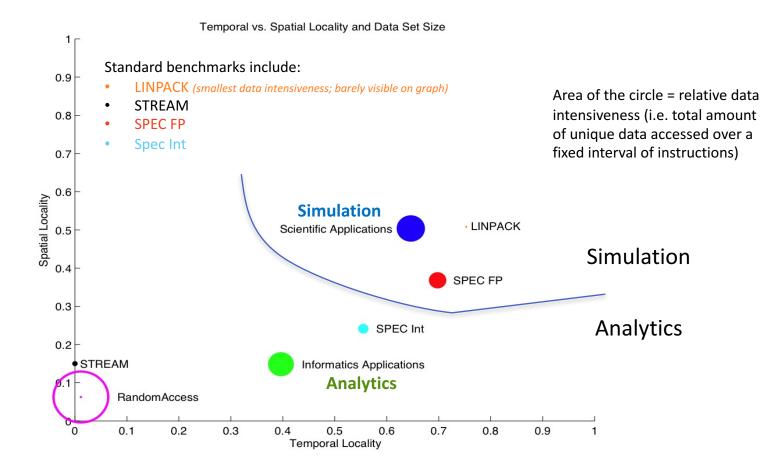


Figure from Murphy & Kogge with adjustment to double radius of Linpack data point to make it visible.

Application code characteristics differ



Contrasting properties:

Application code property	Simulation	Analytics
Spatial locality	High	Low
Temporal locality	Moderate	Low
Memory footprint	Moderate	High
Computation type	May be floating-point dominated*	Integer intensive
Input-output orientation	Output dominated	Input dominated

* Increasingly, simulation work has become less floating-point dominated

Q2: So what do we really mean by "increasing coherence" between simulation and analytics?



- NOT one system ostensibly optimized for both simulation and analytics
- Greater commonality in underlying componentry and design principles
- Greater interoperability, allowing interleaving of both types of computations

... A more common hardware and software roadmap between simulation and analytics



And yet, there is hope ...

Simulation and analytics are evolving to become more similar in their architectural needs

Current challenges for the LSDA community

- Data movement
- Power consumption
- Memory/interconnect bandwidth
- Scaling efficiency

Instruction mix for Sandia's HPC engineering codes

- Memory operations 40%
- Integer operations 40%
- Floating point 10%
- Other 10%

Common design impacts of energy cost trends

- Increased concurrency (processing threads, cores, memory depth)
- Increased complexity and burden on
 - system software, languages, tools, runtime support, codes

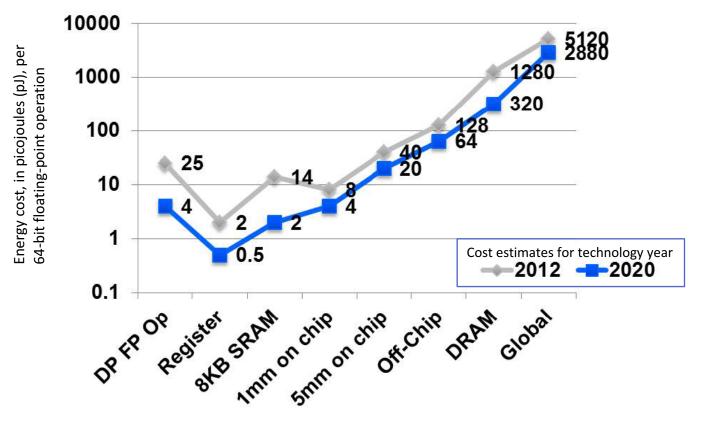
... similar to HPC simulation



... similar to LSDA



Energy cost for various common operations



From Dan McMorrow, Technical Challenges of Exascale Computing, JSR-12-310, JASON, MITRE Corporation, April 2013.

Emerging architectural and system software synergies



Similar needs:

Architectural Characteristic	Simulation	Analytics
Computation	Memory address generation dominated	Same
Primary memory	Low power, high bandwidth, semi-random access	Same
Secondary memory	Emerging technologies may offset cost, allowing much more memory	require extremely large memory spaces
Storage	Integration of another layer of memory hierarchy to support checkpoint/restart	to support out-of-core data set access
Interconnect technology	High bisection bandwidth, (for relatively coarse-grained access)	(for fine-grained access)
System software (node-level)	Low dependence on system services, increasingly adaptive, resource management for <u>structured</u> parallelism	highly adaptive, resource management for <u>unstructured</u> parallelism
System software (system-level)	Increasingly irregular workflows	Irregular workflows

Q3: How might coherence be furthered in practice? The Sandia National Laboratories

Making it an element of national strategy

- Check via the NSCI
- Building this in to exascale computing efforts
 - Also a component of the NSCI
- Communicating with and enlisting the technical communities concerned
 - This forum and similar events
- Further developing the vision
 - Today's dialogue session!

Acknowledgements



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Large-Scale Data Analytics and Its Relationship to Simulation Robert Leland, Richard Murphy, Bruce Hendrickson, Katherine Yelick,

John Johnson, and Jonathan Berry

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