

**SIMULATION INFORMATICS**  
SAND2011-3203C  
*A data-driven methodology for  
synthesis from simulation results*

*David F. Gleich*

*John von Neumann Postdoctoral Fellow*

*Sandia National Laboratories*

*Livermore, CA*

# *A bit more background on me*

## **Before**

*Undergrad* Harvey Mudd College (Joint CS/Math)

*Internships* [Yahoo!](#) and [Microsoft](#) Internships

*Graduate* Stanford (Computation and Mathematical Engineering)

Thesis: Models and Algorithm's for PageRank Sensitivity

*Internships* [Intel](#), Joint [Library of Congress](#)/Stanford, [Microsoft](#)

*Post-doc* / Univ. of British Columbia

## **Now**

*Post-doc* 2 2010 John von Neumann fellow @ Sandia

## **In August**

Tenure-Track position at Purdue's CS department

*It's time to ask:  
What can science  
learn from Google?*

– *Wired Magazine (2008)*

[http://www.wired.com/science/discoveries/magazine/16-07/pb\\_theory](http://www.wired.com/science/discoveries/magazine/16-07/pb_theory)

# *A tale of two computers ...*



***Cielo (#10 in top 500)***  
Separate storage system

*This platform is awesome  
for simulating physical systems*

*Cost* ~\$50 million



***SNL/CA Nebula Cluster***  
2TB/core storage

*This platform is awesome  
for working with data*

*Cost* \$150k

# *Simulation: The Third Pillar of Science*

*21<sup>st</sup> Century Science in a nutshell*

Experiments are not practical / feasible.  
Simulate things instead.



But do we trust the simulations?

**We're trying!**

Model Fidelity

Verification & Validation (V&V)

Uncertainty Quantification (UQ)



***Insight and confidence requires multiple runs.***

# The problem Simulation ain't cheap!

*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 (again)

Input  
Parameters  
→  
 $s$



Time history  
of simulation  
→  
 $f$

## 21. 1<sup>st</sup> Century Science in a nutshell?

Simulations are too expensive.  
Let data provide a surrogate.

The Database



# Our CSRF project *Simulation Informatics*

## **The People**

*Myself*

Paul G. Constantine

(2009 von Neumann)

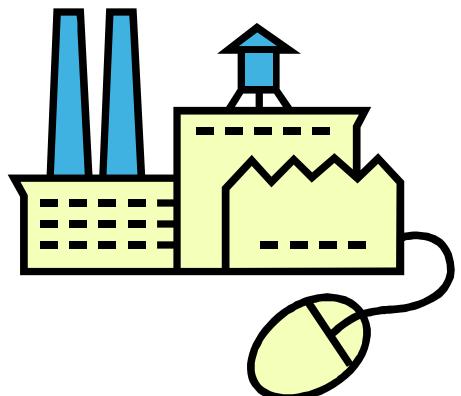
To enable analysts and engineers to hypothesize from **inexpensive data computations** instead of expensive HPC computations.



Stefan Domino  
Jeremy Templeton  
Joe Ruthruff

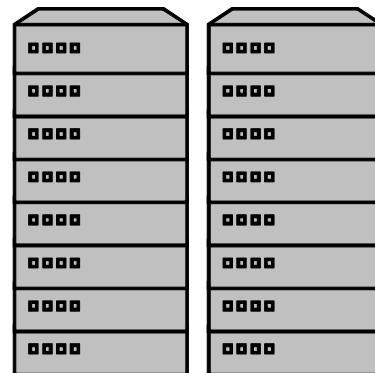
# *The idea and vision: Store the runs!*

Supercomputer



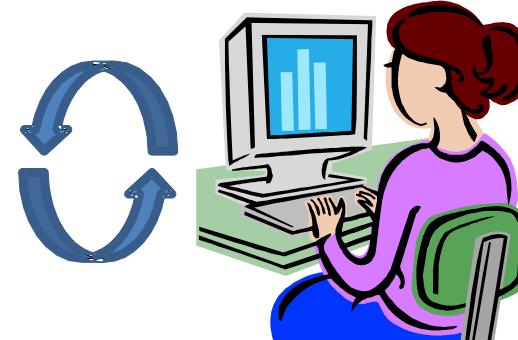
*Each multi-day HPC simulation generates gigabytes of data.*

MapReduce cluster



*A MapReduce cluster can hold hundreds or thousands of old simulations ...*

Engineer



*... enabling engineers to query and analyze months of simulation data for statistical studies and uncertainty quantification.*

# MapReduce

Originated at Google for indexing web pages and computing PageRank

## *The idea*

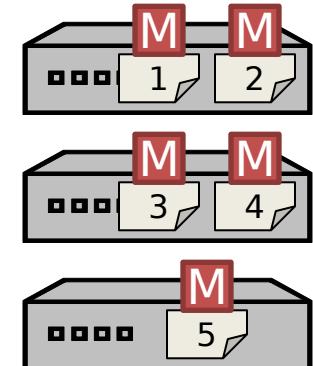
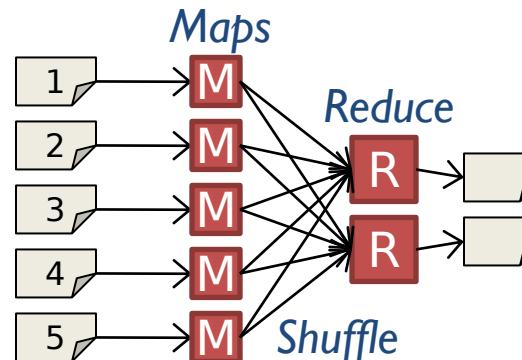
*Bring the computations to the data.*

Express algorithms in data-local operations.

Implement one type of communication: **shuffle**.

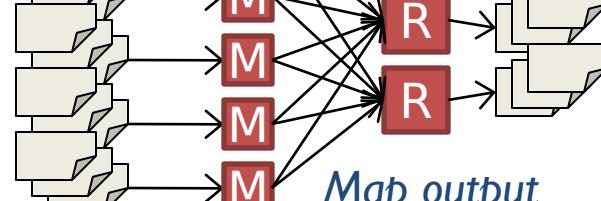
**Shuffle** moves all data with the same key to the same reducer

## **Data scalable**



## **Fault tolerance by design**

*Input stored in triplicate*

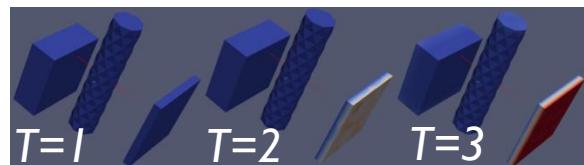


*Map output persisted to disk before shuffle*

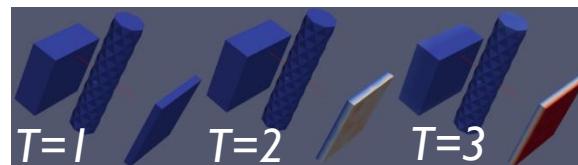
# Mesh point variance in MapReduce

Three simulation runs for three time steps each.

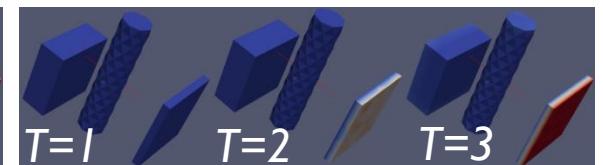
Run 1



Run 2

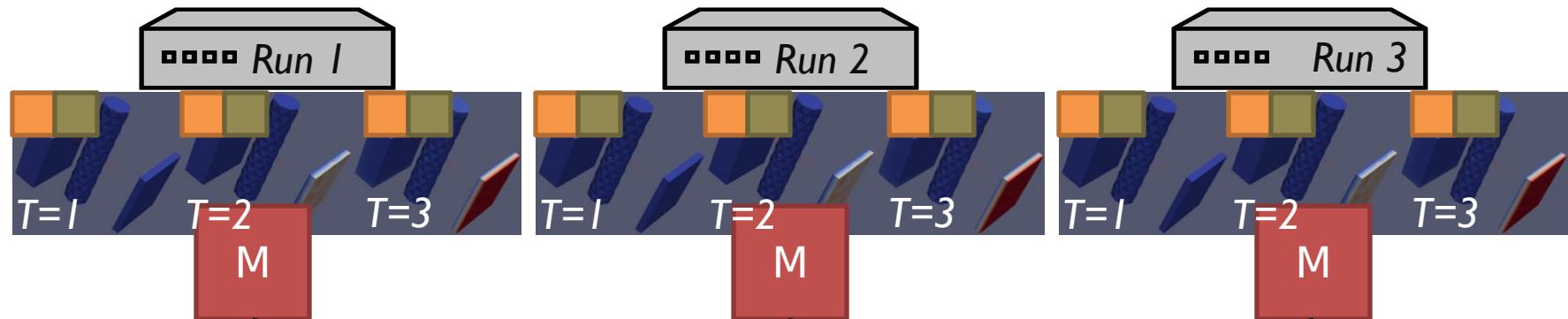


Run 3



# Mesh point variance in MapReduce

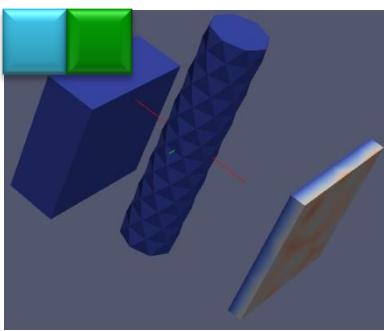
Three simulation runs for three time steps each.



1. Each mapper outputs the mesh points with the same key.

2. Shuffle moves all values from the same mesh point to the same reducer.

3. Reducers just compute a numerical variance.

Bring the computations to the data! 

# Hadoop???

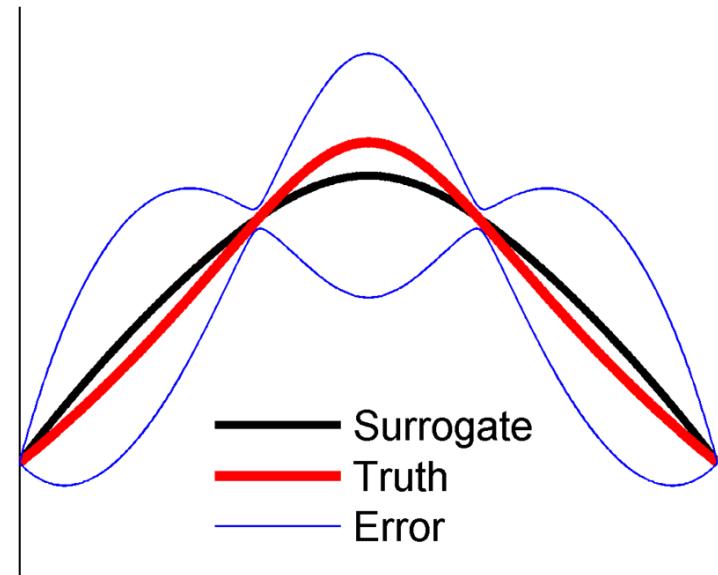
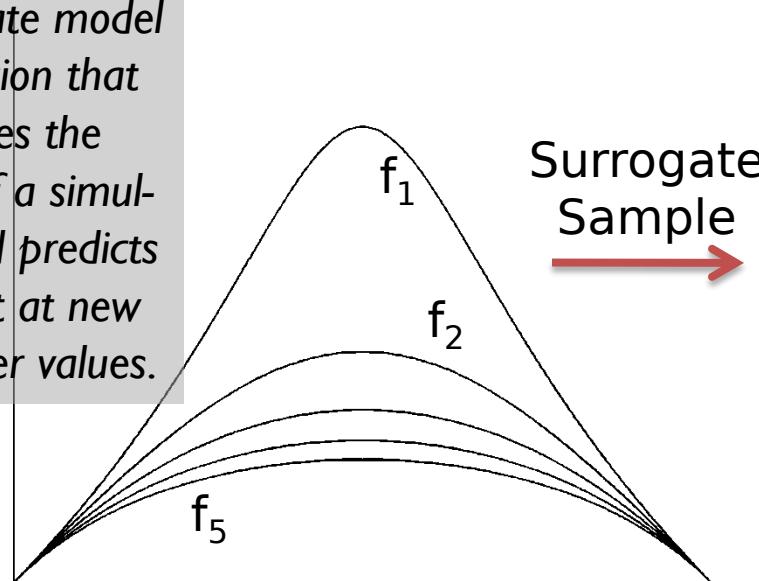
Hadoop: an open-source  
MapReduce system  
supported by Yahoo!



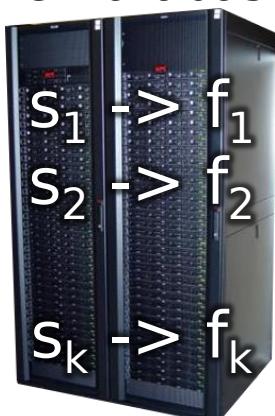
*Photo credit, New York Times (2009)*  
<http://www.nytimes.com/2009/03/17/technology/business-computing/17cloud.html>

# MapReduce and Surrogate Models

A surrogate model is a function that reproduces the output of a simulation and predicts its output at new parameter values.

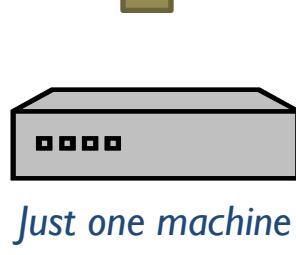


The Database



Extraction

The Surrogate



Interpolation

New Samples



On the MapReduce cluster

On the MapReduce cluster

# MapReduce + Simulations

## **Thermal Race problem**

Joe Ruthruff and Jeremy Templeton

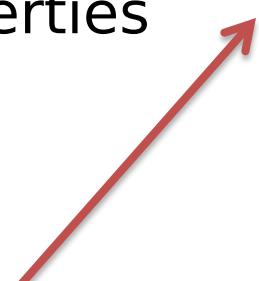
Heating an unclassified NW model to simulate a fire

**Goal** make sure the weak link fails  
before the strong link

**Objective** determine where  
parameters are interesting

**Parameters** material properties

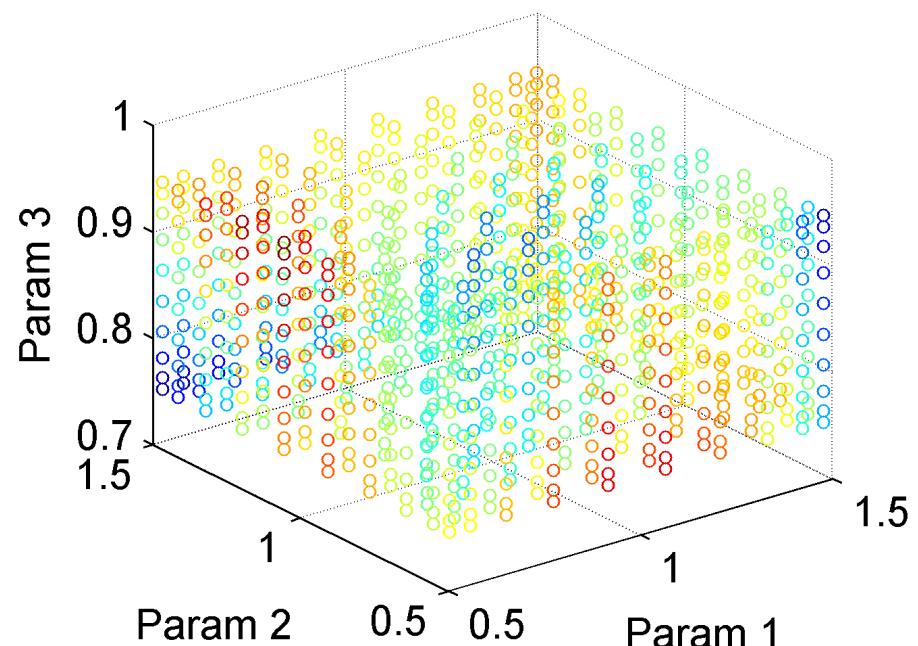
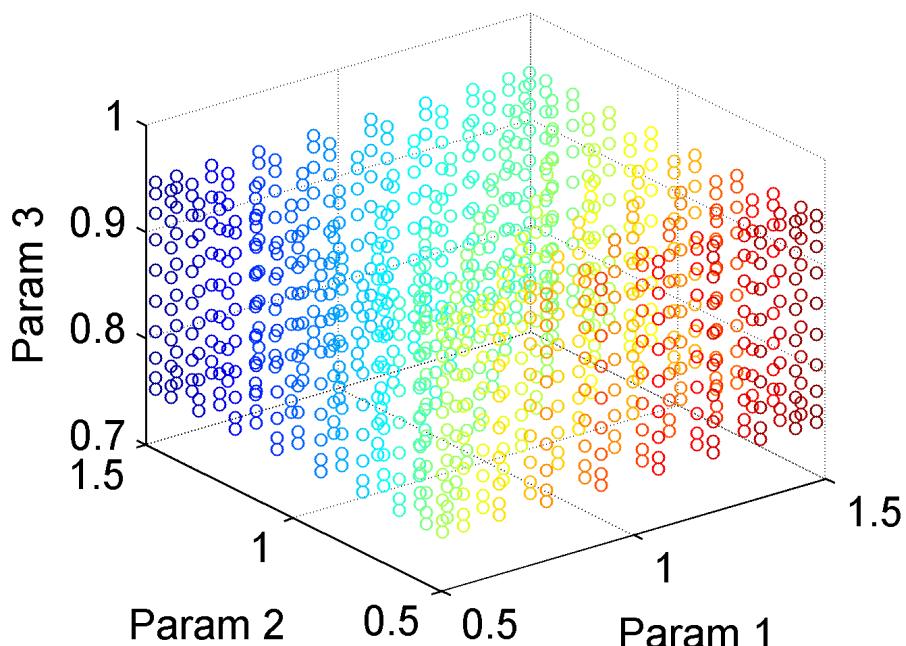
1000 sample  
parameter  
study using



30 min per run  
700GB of Raw Data  
~64GB of data for  
MapReduce

**Our results**  
Extraction **30 min**  
Interpolation **5 min**

# *Data driven Green's functions*



The simulation varies most in the corners of the parameter space.

*These analyses will help understand which parameters matter, and where to continue running the simulations.*

# *What can science learn from Google?*

## **The importance of data.**

For more about this, see a talk  
and paper by Peter Norvig

“Internet-Scale Data Analysis”  
[http://www.stanford.edu/group/  
mmds/slides2010/Norvig.pdf](http://www.stanford.edu/group/mmds/slides2010/Norvig.pdf)

Halevy et al. “The unreasonable  
effectiveness of data.” IEEE  
Intelligent systems, 2009.

# The future?

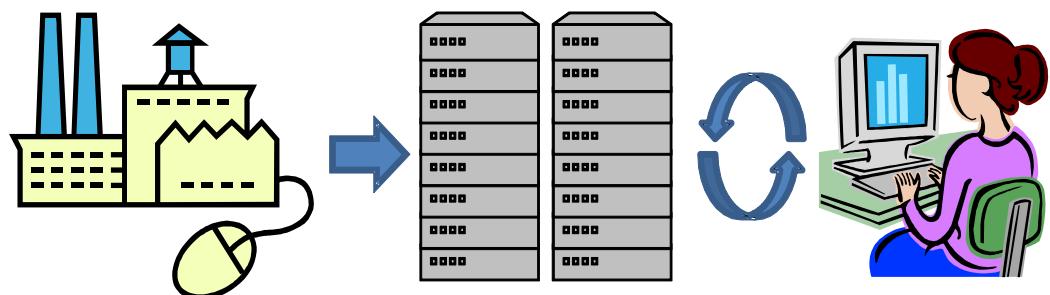


*MapReduce computers embedded within super computers as the parallel file system and data analysis platform.*

***The best of both worlds.***

## Our work

*Laying the foundation for useful surrogate models that will make data based simulation a compelling addition to the simulation revolution.*



*Any questions?*