

The Potential of Cloud Computing: Challenges, Opportunities, Impact

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Adapted from NAE Gilbreth Lecture, October 2011



Get Your Own Supercomputer

- 41.8 Teraflops on LINPACK benchmark
 - #451 on current Top500 list of supercomputers
- No batch queues—start your job in minutes
- No grantwriting—just a credit card
(from \$1300/hour, 1 hour minimum purchase)
- No need to predict usage—pay for what you use, add capacity on demand
- Lease several simultaneously, to run multiple experiments in parallel
- Matlab, Mathematica, other packages available



Warehouse Scale Computers

- Built to support consumer demand of Web services (email, social networking, etc.)
 - “Private clouds” of Google, Microsoft, Amazon,
- “Warehouse scale” buying power = 5-7x cheaper hardware, networking, administration cost



photos: Cnet News, Sun Microsystems (Oracle), datacenterknowledge.com



- The first demonstration of how to build really large Internet sites out of *clusters of commodity* computers was done by:
 - (a) Stanford
 - (b) Berkeley
 - (c) Yahoo!
 - (d) Google
 - (e) IBM



2008: Public Cloud Computing Arrives

Type & Price/Hour	1GHz core eqv.	RAM	Disk and/or I/O
Small - \$0.085	1	1.7 GB	160 GB
Large - \$0.34	4	7.5 GB	850 GB, 2 spindles
XLarge - \$0.68	8	15 GB	1690 GB, 3 spindles
Cluster ("HPC") - \$1.60	32*	23 GB	1690 GB + 10Gig Ethernet + optional 2x NVidia GPU

- Virtual machines from \$0.085/hr to \$2.10/hr
 - Pay as you go with credit card, 1 hr. minimum
 - Cheaper if willing to share or risk getting kicked off
 - Machines provisioned & booted in a few minutes

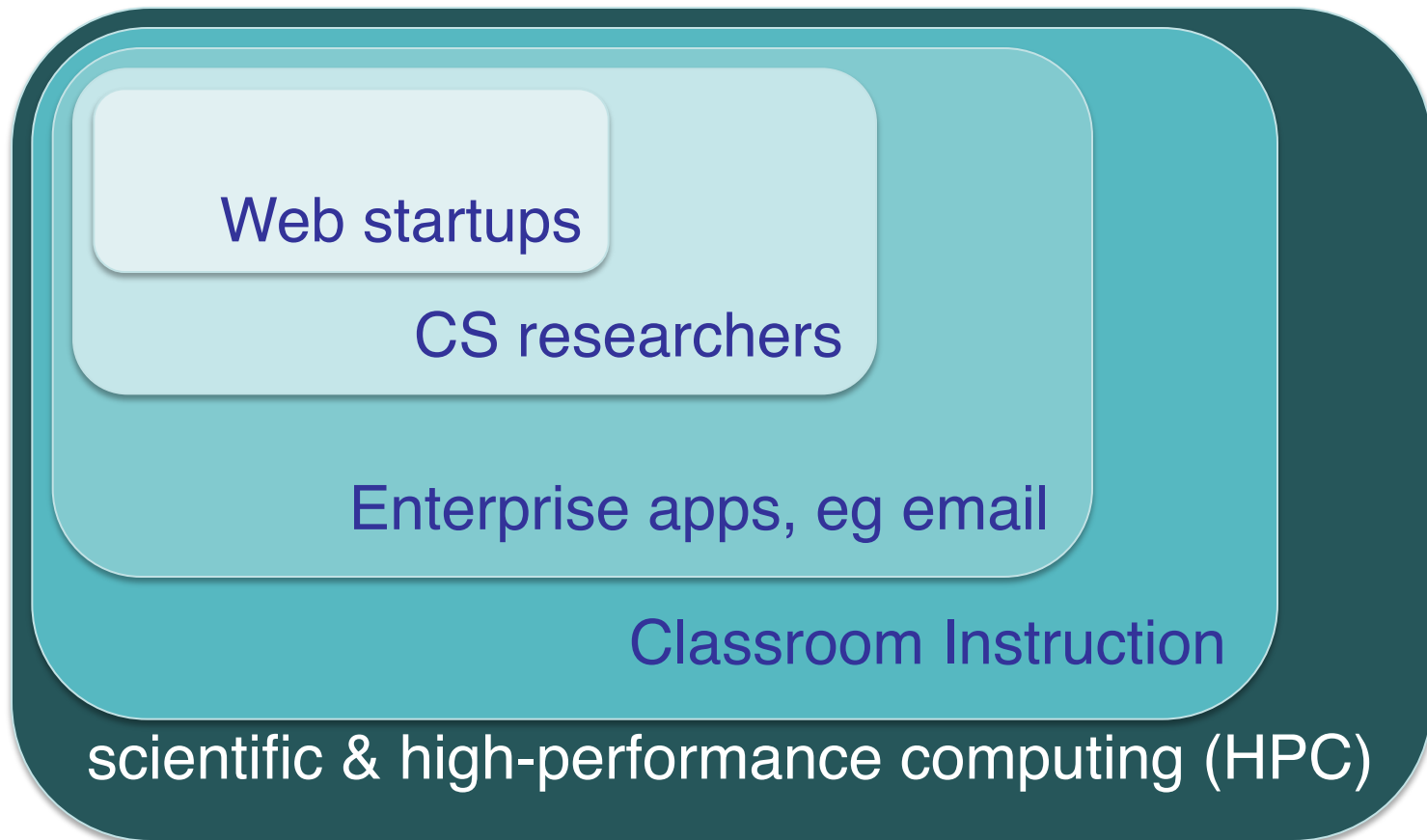
1,000 machines for 1 hr = 1 machine × 1,000 hrs

* 2x quad-core Intel Xeon (Nehalem)



What's In It For You?

Cloud Computing progress 2008-2011



What are the challenges to adoption?



Trivia Fact

- The first full Web browser running on a mobile device was developed by:
 - (a) Apple
 - (b) Stanford
 - (c) Berkeley
 - (d) Nokia
 - (e) Motorola



Hiding the Cloud's Complexity

- Sophisticated software can ***hide complexity & mask unreliability*** of cloud hardware
- Example: **MapReduce** (2004)
 - e.g.: “**Apply** $f(x)$ over all these x , **keep** K max values”
 - Google's MapReduce software automatically masks failures and stewards cloud resources
 - Hadoop (open source version) followed soon after

*“Warehouse scale” software engineering issues
hidden from application programmer*



MapReduce in Practice

- Example: classifying Twitter spam
 - training: 10^7 samples x 64KB data each = 640GB data
 - One heavy-duty server: ~270 hours
 - 100 servers in cloud: ~3 hours (~\$250)
- Rapid uptake in other scientific research
 - “top 5 pharma” molecular modeling: 3809 machines on Amazon cloud, **30K** cores, \$1279/hr* (Sep 2011)
 - Large-population genetic risk analysis & simulation (Harvard Medical School)
 - Genome sequencing (UNC Chapel Hill Cancer Ctr)
 - Compact Muon Solenoid Expt. (U. Nebraska Lincoln)
- What’s the downside?

* ArsTechnica, 9/20/11, <http://arst.ch/qy8>



Challenge: Cloud Programming

- Programmers must (re)express problems to ***expose parallelism***
- ⚠ parallel software hard to debug & operate, so build on existing successes
 - Pig (Yahoo! Research) & Hive (Apache Foundation) – transform database queries to MapReduce
 - Rhipe – include MapReduce operations in R programs
- ⚠ one abstraction doesn't fit all
 - e.g., clustering algorithms work poorly on MapReduce
 - **Opportunity**: common front-end to cloud & parallel programming tools (sejits.eecs.berkeley.edu)



Challenge: “non-cloudy” scientific codes

- ⚠ Existing scientific codes “supercomputer-centric”
 - reliability, static configuration, exclusive resource use...
- Time-to-answer may still be faster, since no wait!
- **Opportunity:** software that shares cloud among multiple frameworks (mesos.berkeley.edu)
- **Opportunity:** Vendors listening to HPC customers
 - Hardware: cloud-based “supercomputer” makes Top500 list (July 2010)
 - Software: MathWorks, Mathematica now support sending (parallel) computations to Amazon cloud



Challenge: Big Data

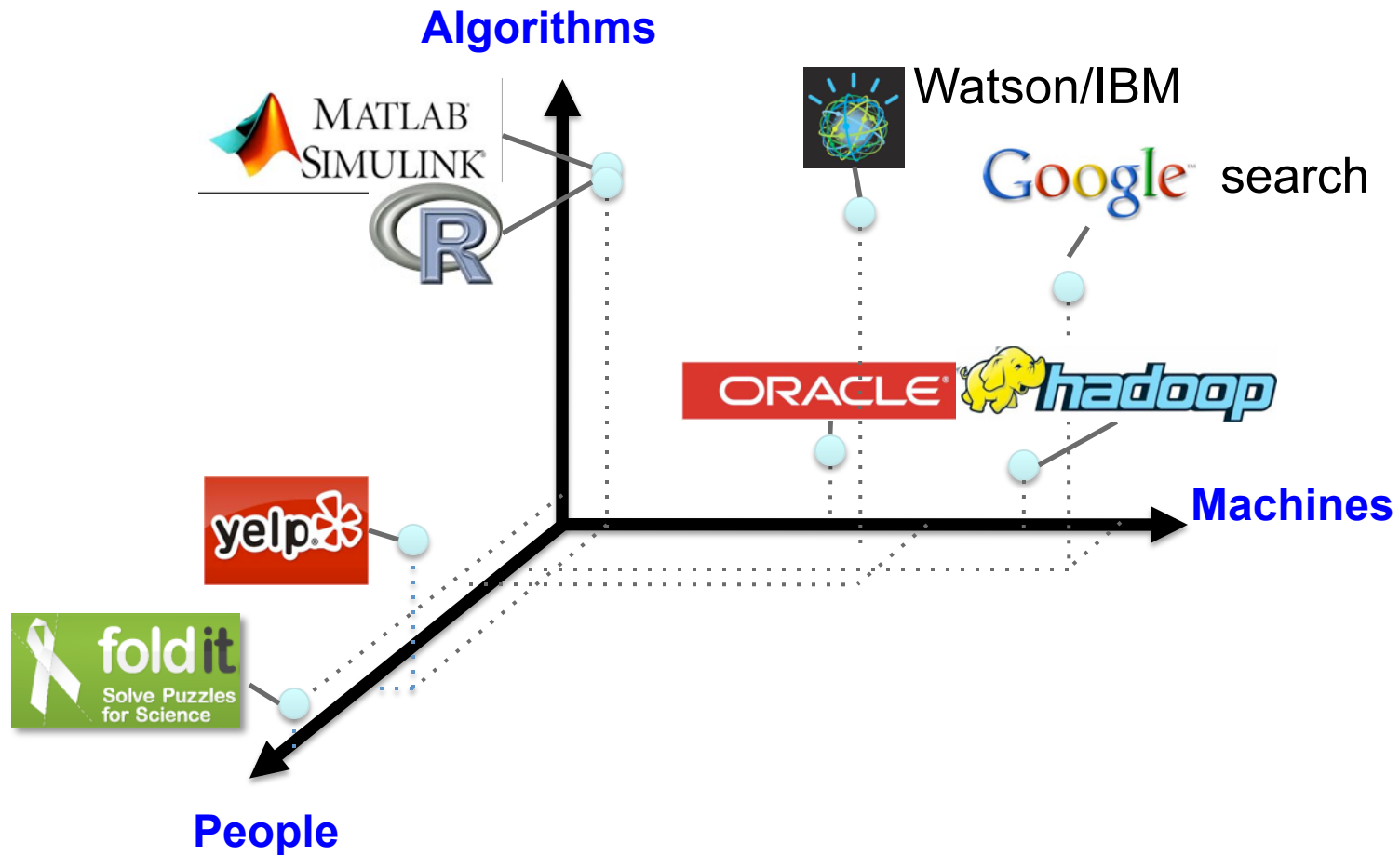
Application	Data generated per day* (1 TB = 10^{12} bytes)
Genome Sequencing	1 TB (Illumina HiSeq machine) Cancer Genome Atlas: 5PB online by end of 2011
Large Synoptic Survey Telescope	30 TB 400 Mbps sustained xfer between Chile & NCSA
Large Hadron Collider	60 TB

⚠ How to copy 8 TB (8×10^{12} bytes) from cloud?

- 2009: Download at ~ 20 Mbps \rightarrow 35 days, \$800
- 2010: Ship drive to Amazon \rightarrow 1 day, \$150
- 2011: Pay-as-you-go fast network \rightarrow <1 day, \$165
 - (~ 18 hours at 1 Gbps, 0.30/hr. + 0.02/GByte)

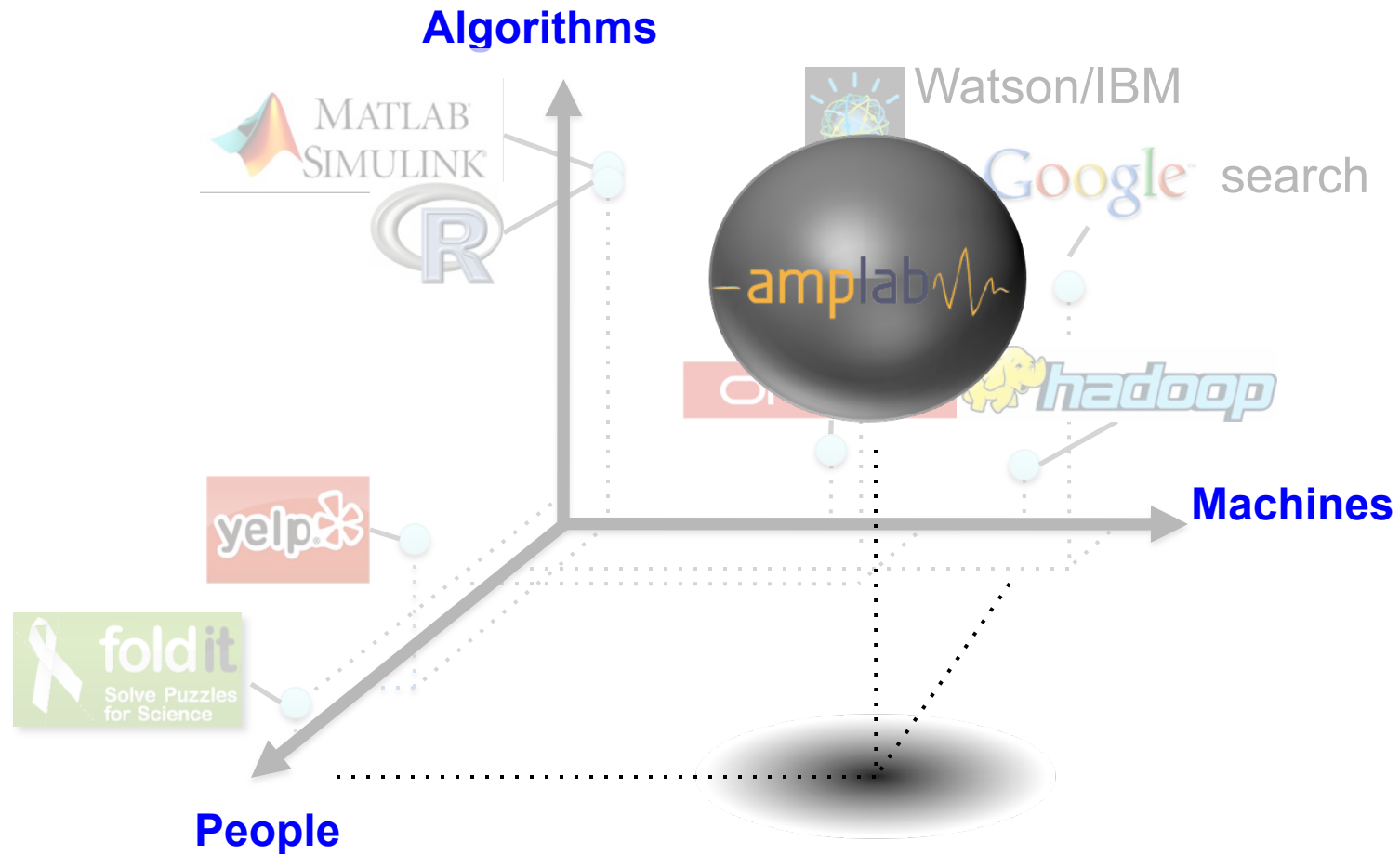


Algorithms, Machines, People





The Berkeley AMP Lab





Trivia fact

- The percentage of Americans who will face cancer in their lifetimes is approximately:
 - (a) 1%
 - (b) 5%
 - (c) 10%
 - (d) 20%
 - (e) 30%

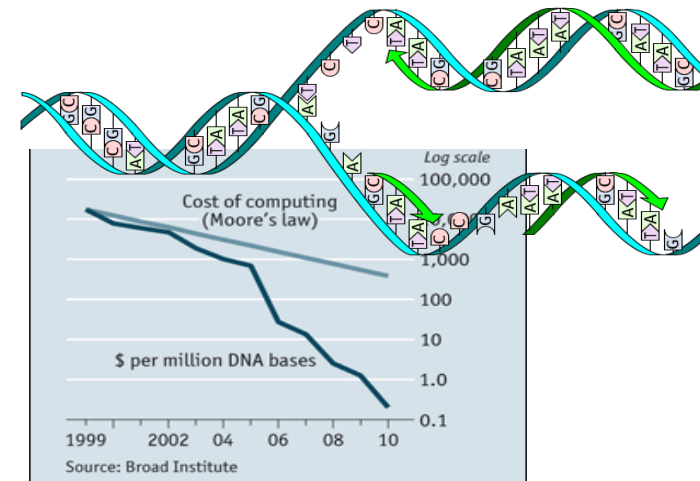


Trivia fact

- By next year, the “wet lab” cost of sequencing a human genome is expected to be about:
 - (a) \$1,000,000
 - (b) \$100,000
 - (c) \$10,000
 - (d) \$1,000
 - (e) None of the above



Opportunity: Application-Driven CS research



By working with domain experts, we will create better tools to support big-data applications on cloud computing



Cloud & UCB EECS

1. Berkeley research culture: integrate leading research into teaching at all levels
2. RAD Lab need for “killer apps” to show off infrastructure

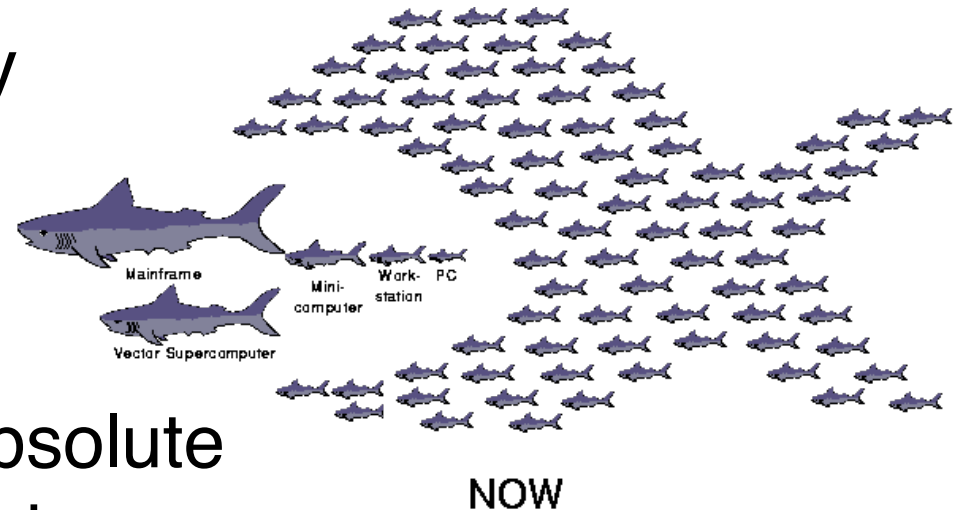
Current efforts (student counts approximate):

- Great Ideas in Computer Architecture (reinvented Fall 2010): 190 students
- Software Engineering for SaaS (Spr' 12 = 5th iteration): 50+50+50+70+100 students
- Operating Systems: 70 students
- Intro. Data Science: 30
- Adv. topics in HCI: 20 students
- Natural language processing: 20 students
- Cancer Genomics: 10 students



An Analogy: *Networks of Workstations (NOW)*, c.1995

- Clusters of Commodity PC's vs. symmetric multiprocessors
- *Potential* advantages: incremental scaling, absolute capacity, commodity price
- 1995: Web software architecture was SMP-centric
- 2010: Berkeley undergrads prototype Web apps in 6–8 weeks and deploy on cloud computing
- 2015: scientists routinely do 1000-machine e-science experiments using cloud computing





Conclusion

- Democratization of supercomputing capability
 - Time to answer may be faster even if hardware isn't
 - Writing a grant proposal around a supercomputer?
- Software & hardware for science-on-cloud improving at “commodity speed”
- Part of Computer Science “coming of age” with increasingly outward-looking research

***Potential democratizing impact
comparable to the microprocessor***



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RAD Lab
UC Berkeley

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