

Principles for Experimental Algorithmics

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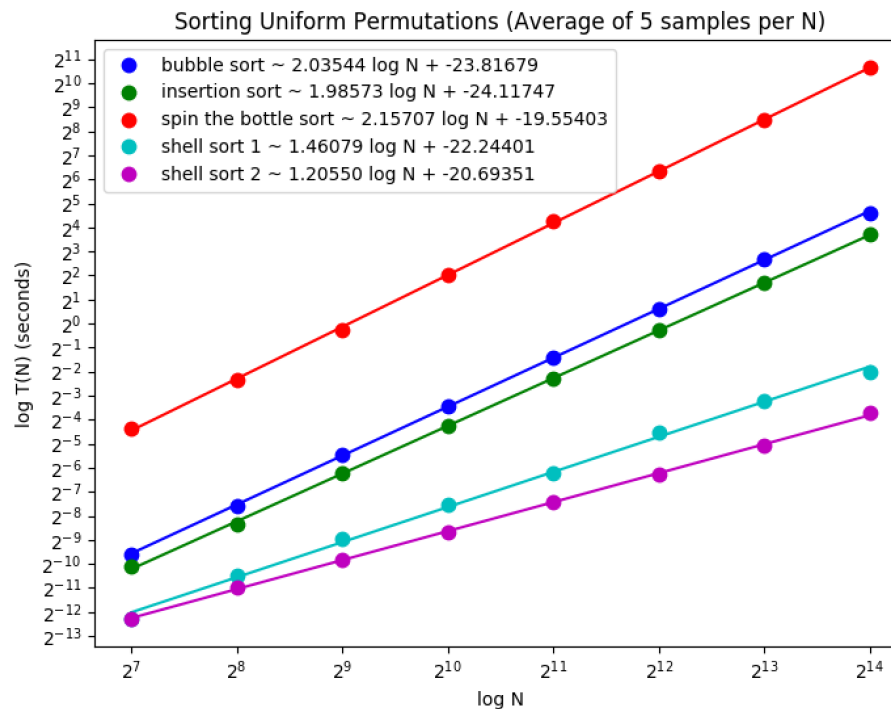
based, in part, on the following papers:

Towards a Discipline of Experimental Algorithmics, by Bernard M.E. Moret

A Theoretician's Guide to the Experimental Analysis of Algorithms, by David S. Johnson

Experimental Algorithmics

- Experimental Algorithmics studies algorithms and data structures by joining experimental studies with the traditional theoretical analyses.
 - Scientists do experiments because they have no choice
 - In experimental algorithmics we combine theoretical analysis with experimentation.

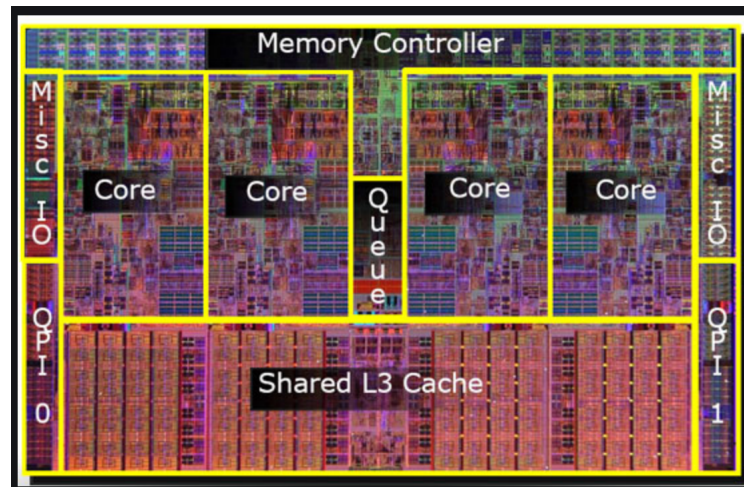


Experimental Algorithmics

- Experimentation with algorithms and data structures can prove to be indispensable for the following tasks:
 - The assessment of heuristics for hard problems
 - The characterization of asymptotic behavior of complex algorithms
 - The comparison of competing designs for tractable problems
 - The formulation of new combinatorial conjectures
 - The evaluation of optimization criteria
 - The transfer of research results from paper to production code

Perform Worthwhile Experiments

- Ask questions worth asking
 - New problems
 - New algorithms
 - New types of input distributions
 - New types of computer hardware

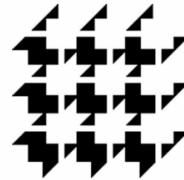


Measuring Actual Performance

- Random instances should be motivated from real-world data
- Also use real-world data when possible

DIMACS

*Center for Discrete Mathematics and Theoretical Computer Science
Founded as a National Science Foundation Science and Technology Center*



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- <http://dimacs.rutgers.edu/programs/challenge/>

By Jure Leskovec

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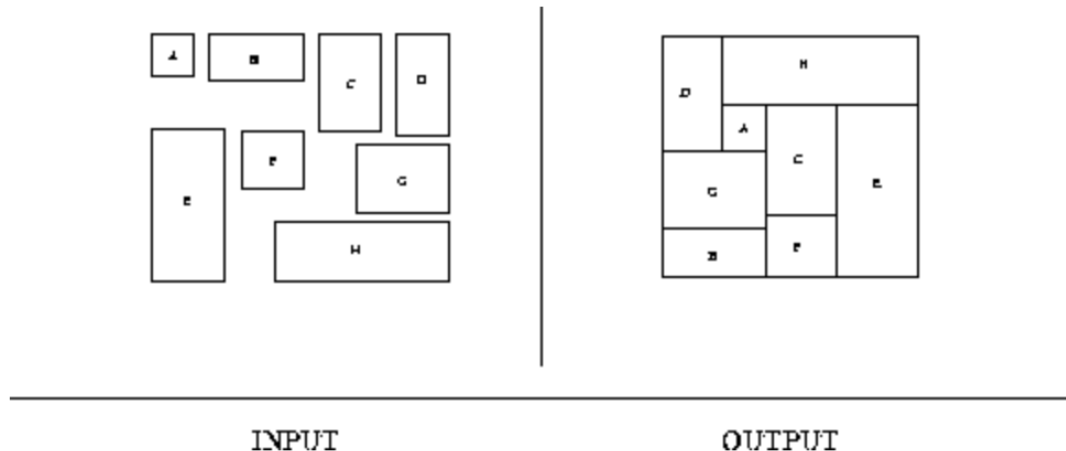


Stanford Network Analysis Project

- <http://snap.stanford.edu/>

Testing the Quality of Solutions

- Find a parameter that can be effectively tested experimentally
 - Waste in bin packing
 - Closeness to a known lower bound

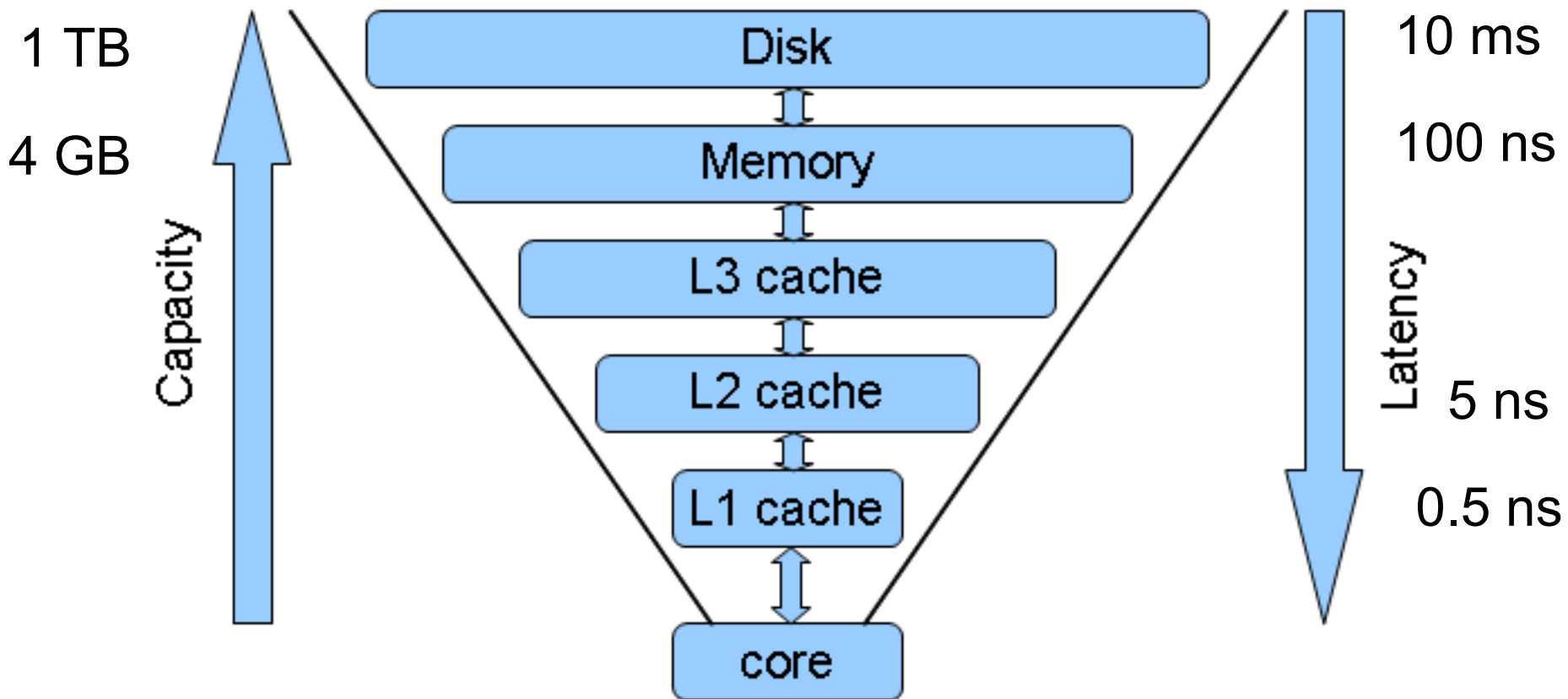


Experimental Setup

- Have clear objectives
- Gather data to answer the questions posed
- Choose hardware appropriately
- Code solutions consistently to allow for good conclusions
- Generate useful problem instances
- Analyze your data

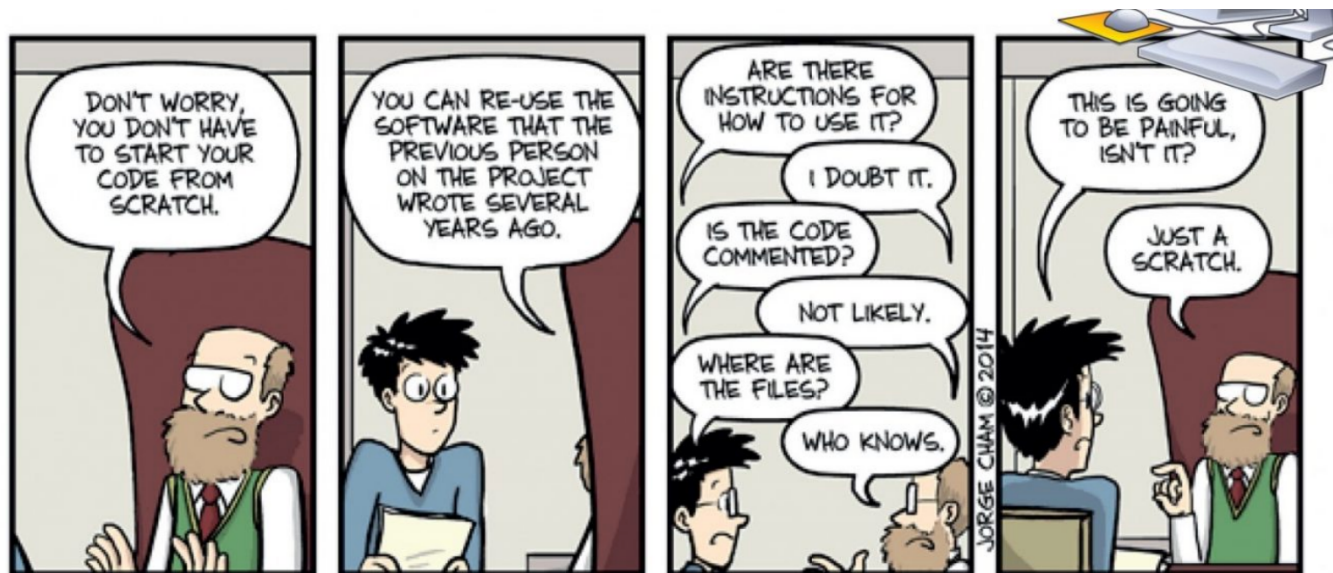
Understand Your Hardware: The Memory Hierarchy

- The trade-off of size and speed



Ensure Reproducibility

- Unless it is truly confidential, post your code for others to use.
- For random data, post how you generated it
- For real-world data, post how to get it



Ensure Comparability

- Perform all experiments on the same hardware
- Report the type of hardware used
- Code all algorithms with the same level of code optimizations and tuning

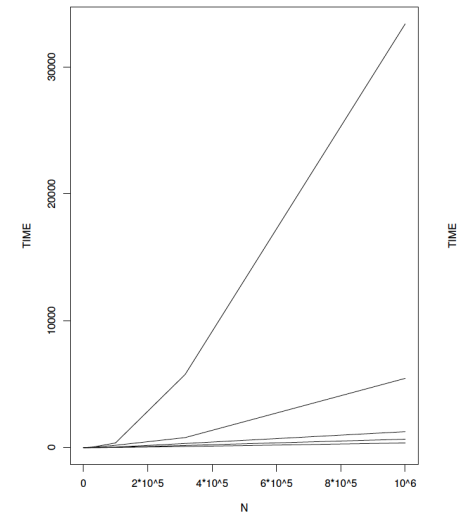


Present your Data in Meaningful Ways

- Use tables only for small data sets

	100	316	1,000	3,162	10,000	31,623	100,000	316,227	1,000,000
Algorithm A	0.00	0.02	0.08	0.29	1.05	5.46	23.0	89.6	377
Algorithm B	0.00	0.03	0.11	0.35	1.38	6.50	30.6	173.3	669
Algorithm C	0.01	0.06	0.21	0.71	2.79	10.98	42.7	329.5	1253
Algorithm D	0.02	0.09	0.43	1.64	6.98	37.51	192.4	789.7	5465
Algorithm E	0.03	0.14	0.57	2.14	10.42	55.36	369.4	5775.0	33414

Running Time versus Instance Size



Log Log Scale

