

Energy Efficiency via the N-way Model

Romain Cledat and Santosh Pande



Current trends

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- Moore's law is still in force:
 - More **cores** instead speed

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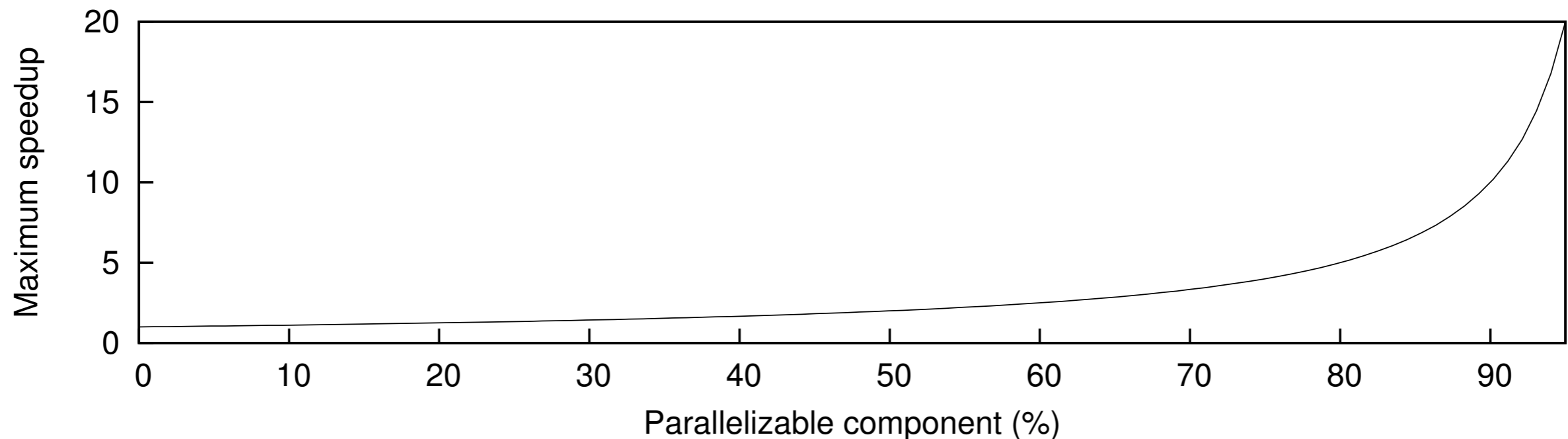
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Research questions

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- **Parallel codes efficiently utilize parallel resources**

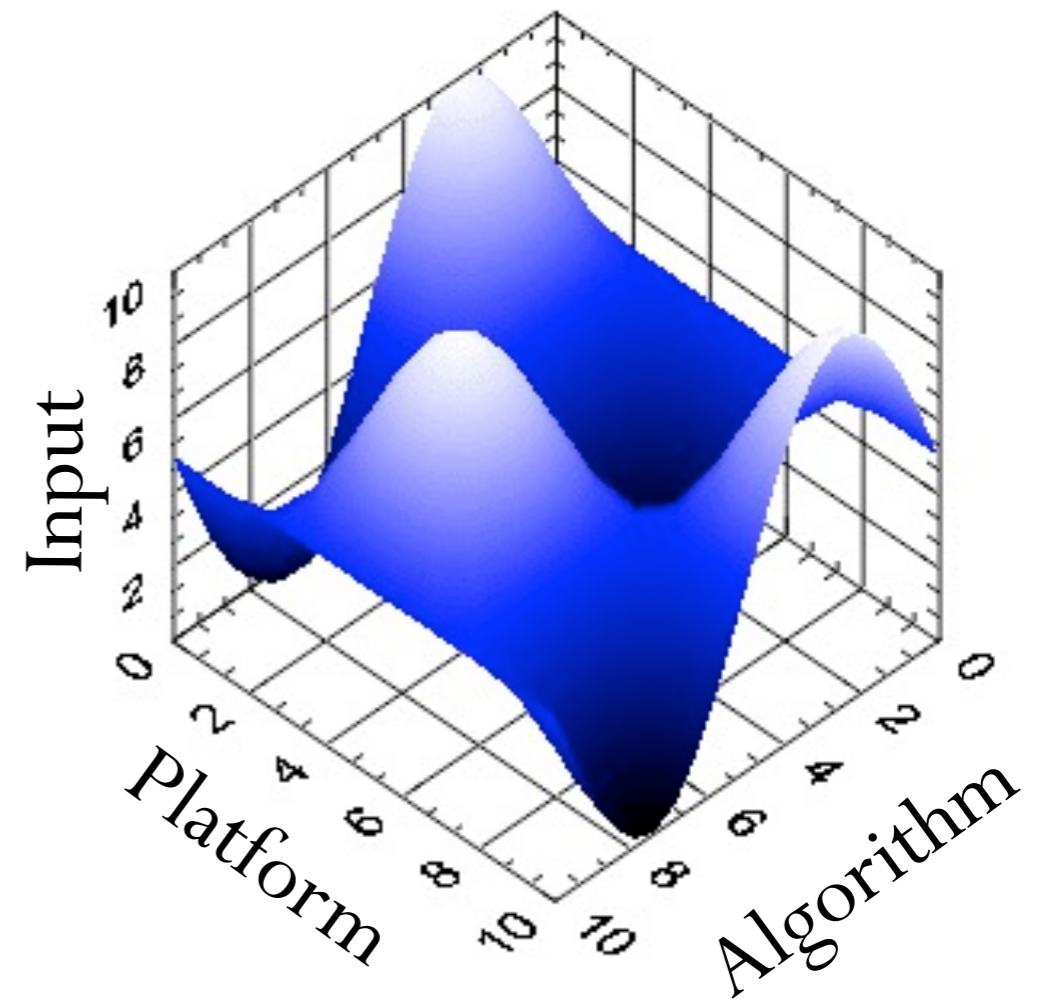
Research questions

- **Parallel codes efficiently utilize parallel resources**
- **For sequential codes:**
 - **Can parallel resources be utilized?**
 - ILP uses in-core parallelism
 - Other cores can be used to speculate, prefetch, etc.
 - **Can they be used efficiently in terms of energy?**

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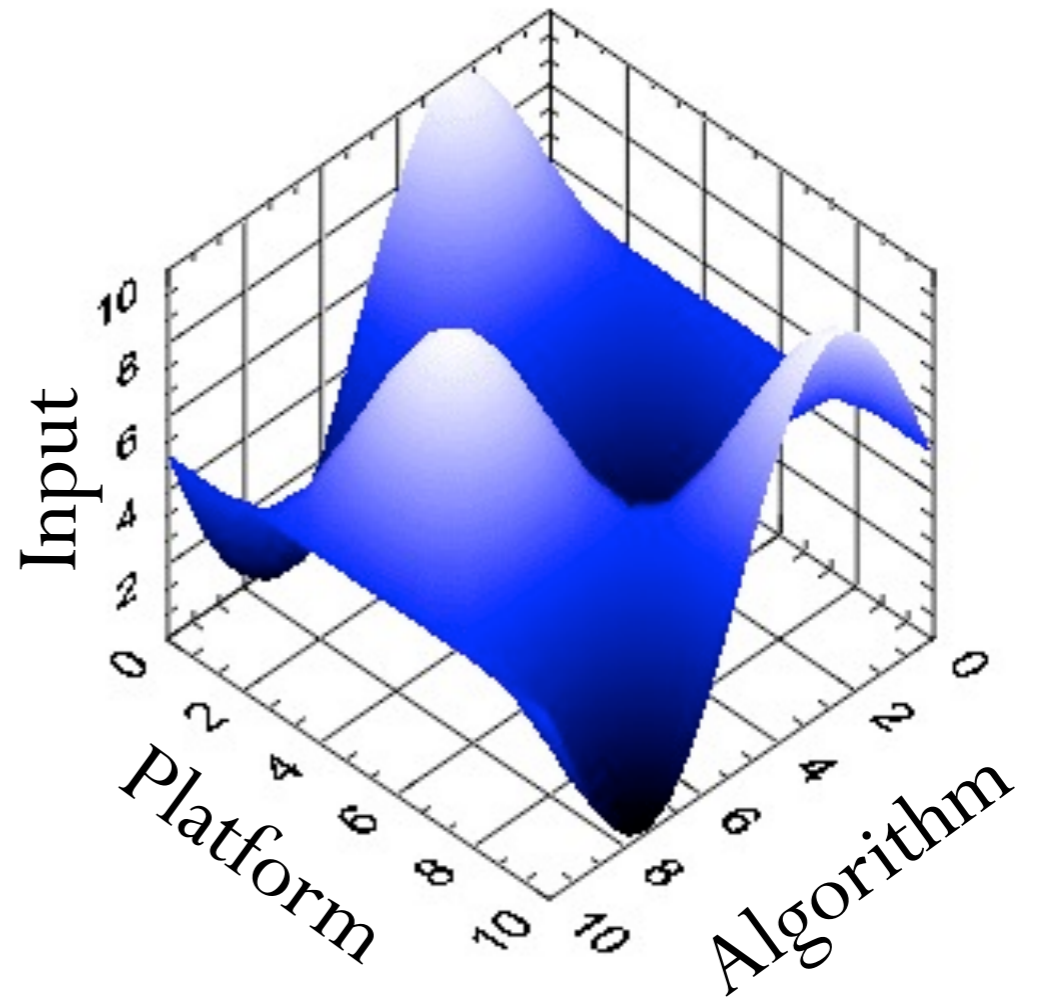
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 - Can parallel resources be utilized?
 - ILP uses in-core parallelism
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 - Can they be used efficiently in terms of energy?
- In this work, we present a use of parallel cores to improve the *algorithmic energy efficiency* of sequential algorithms

Algorithms, platforms and inputs



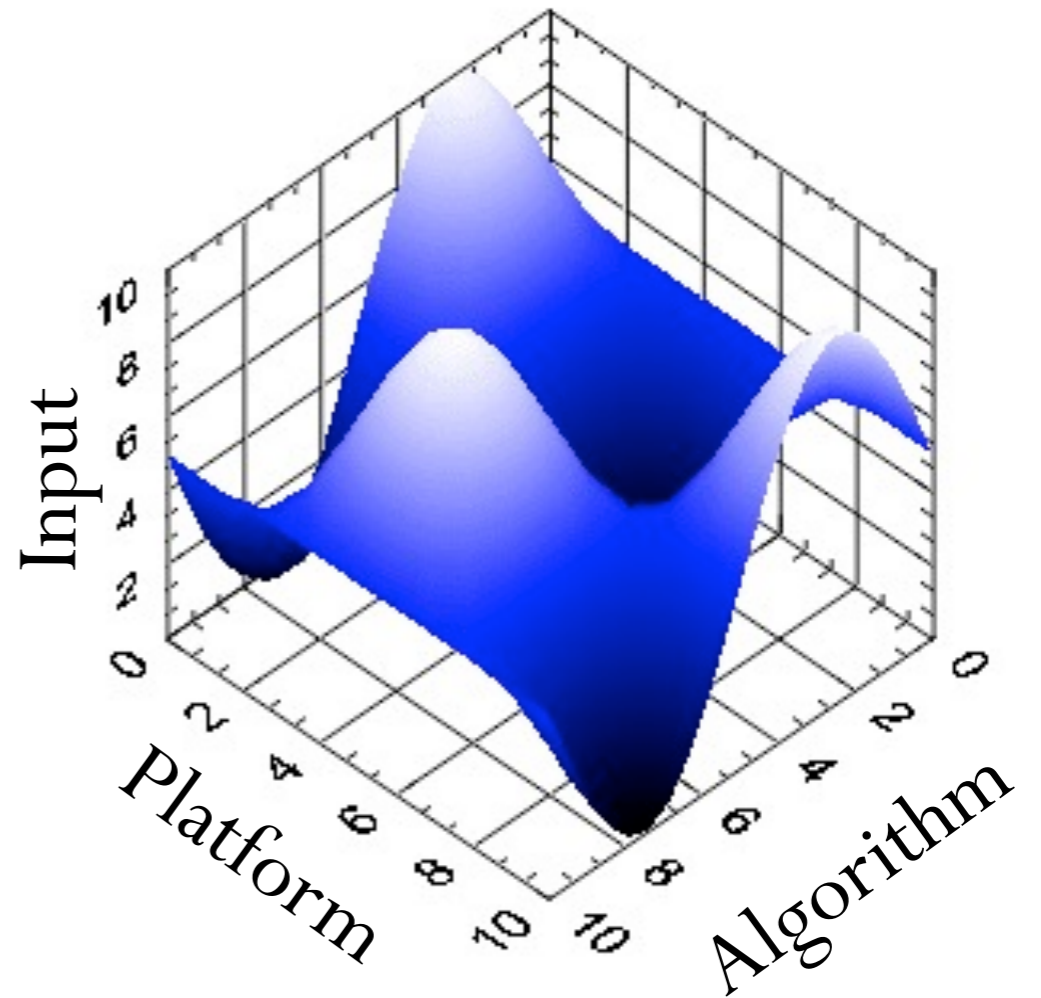
Algorithms, platforms and inputs

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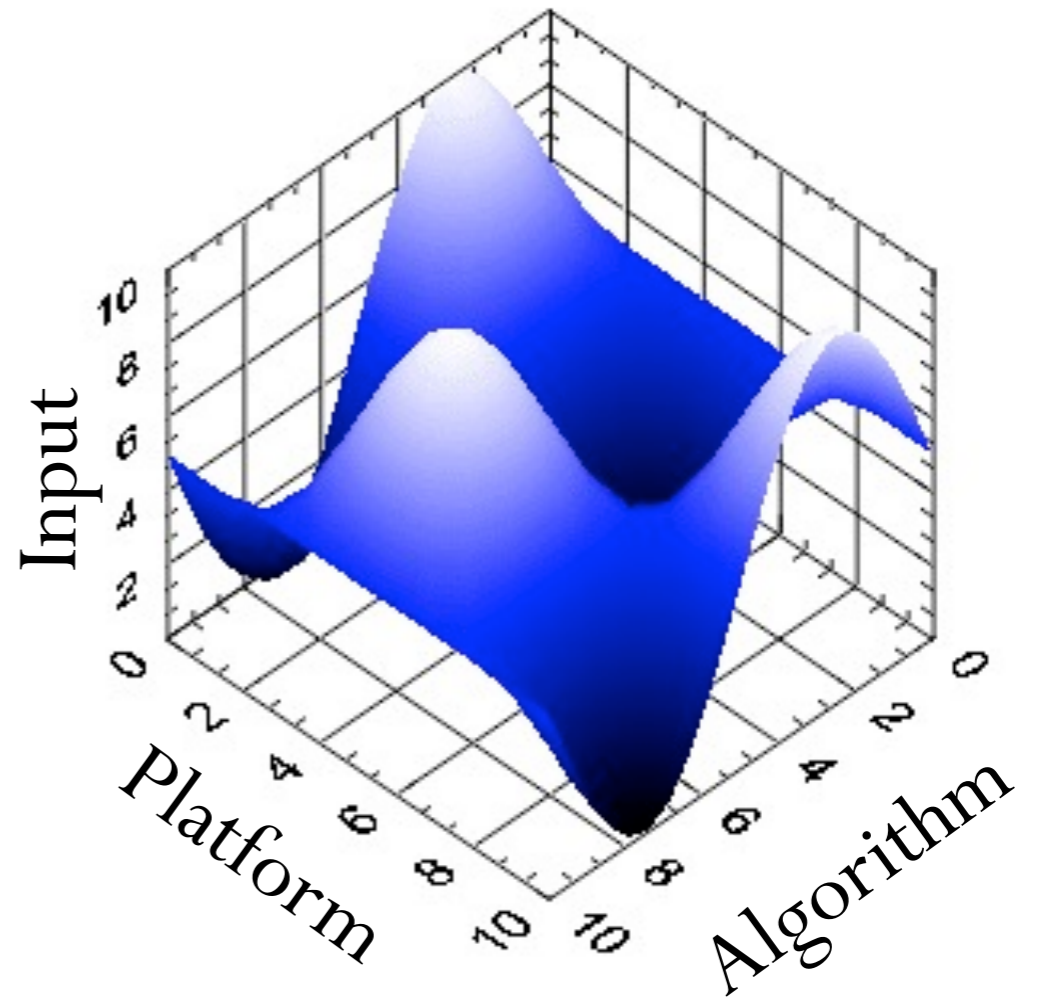
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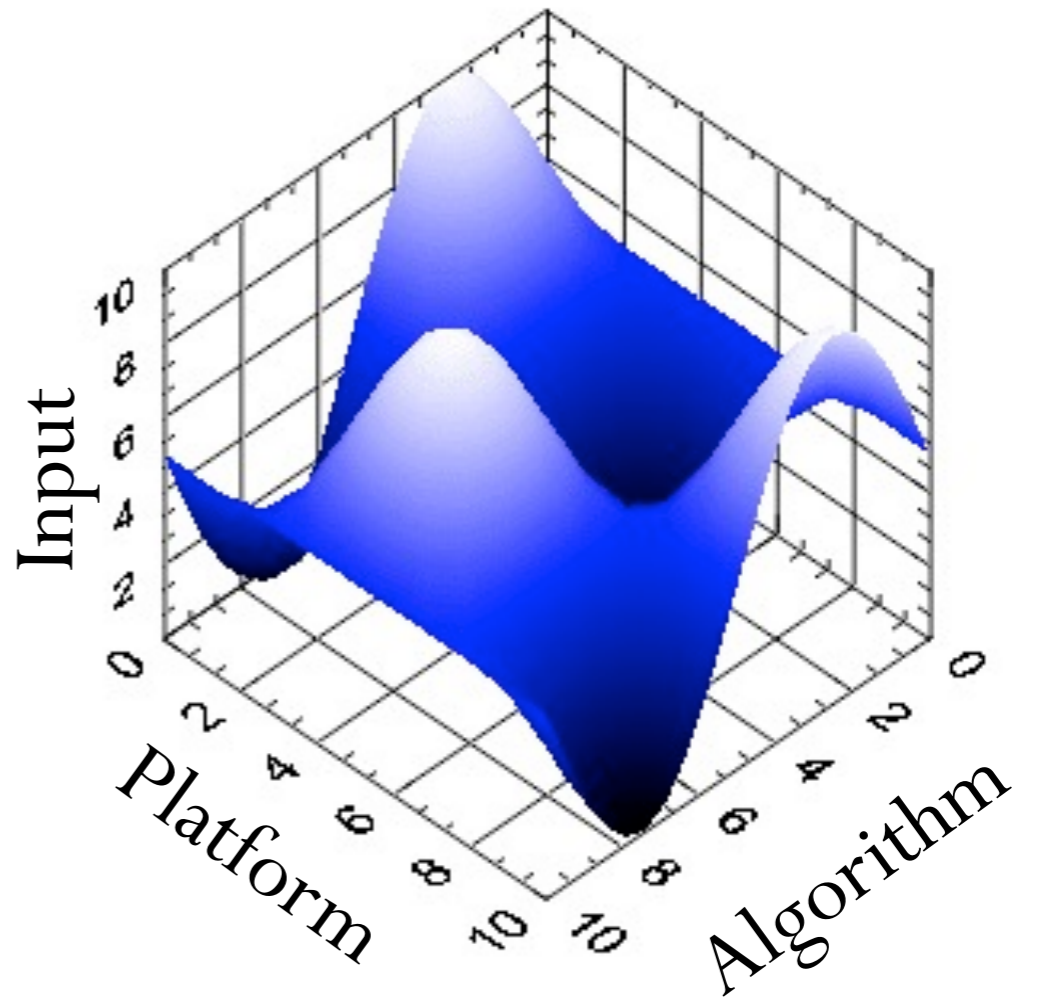
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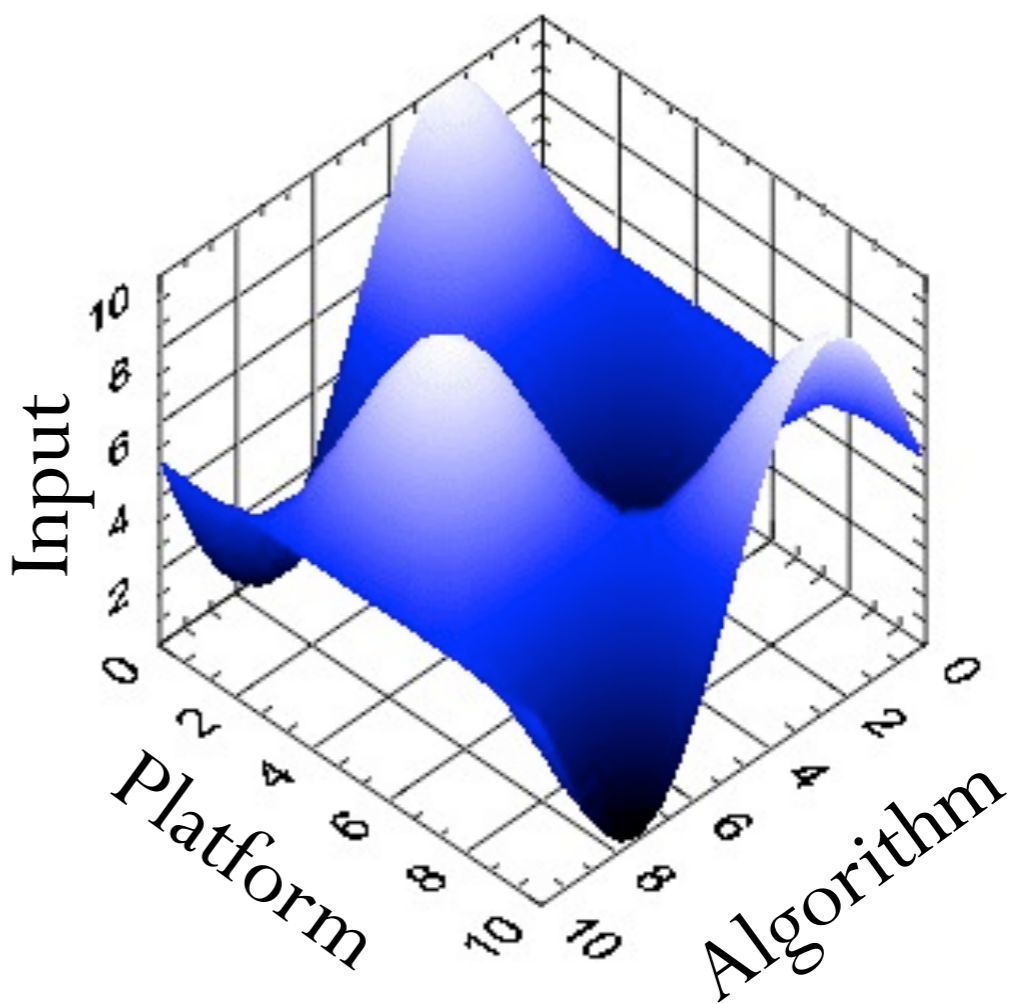


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- For a given problem and input, what is the best match?
 - For a given input, a particular algorithm may be better.

Outline

- The challenges of sequential code on multi-cores
- **Using diversity to expose parallelism**
- Measures of energy efficiency
 - Progress measure
 - Power measure
- The n-way programming model
- Preliminary results and future work

Diversity in platforms

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 - ASICs and FPGAs were optimized for certain algorithms
- Today, chips are becoming heterogeneous
 - Asymmetric multi-cores to save space and power
 - Specialized accelerators (GPUs, network processors...)
- The efficiency of each chip is different and depends on the algorithm
 - GPUs adapted for massive data parallelism
 - Cell SPUs excel at SIMD code

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- Cores can differ:
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- Tools to match code to cores are rare (GLIMPSES for SPUs:
<http://sourceforge.net/projects/glimpses/>)

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- Diversity present:
 - Across algorithms
 - In tuning parameters for algorithms (CPLEX for example)
 - Within an algorithm (randomization for example)
- The choice for the “best” algorithm is not always clear statically or even when the input is known

Diversity across algorithms

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- Solutions across algorithms may be different but all are correct
- Examples: SAT solvers, path finding algorithms, ...

Diversity within an algorithm

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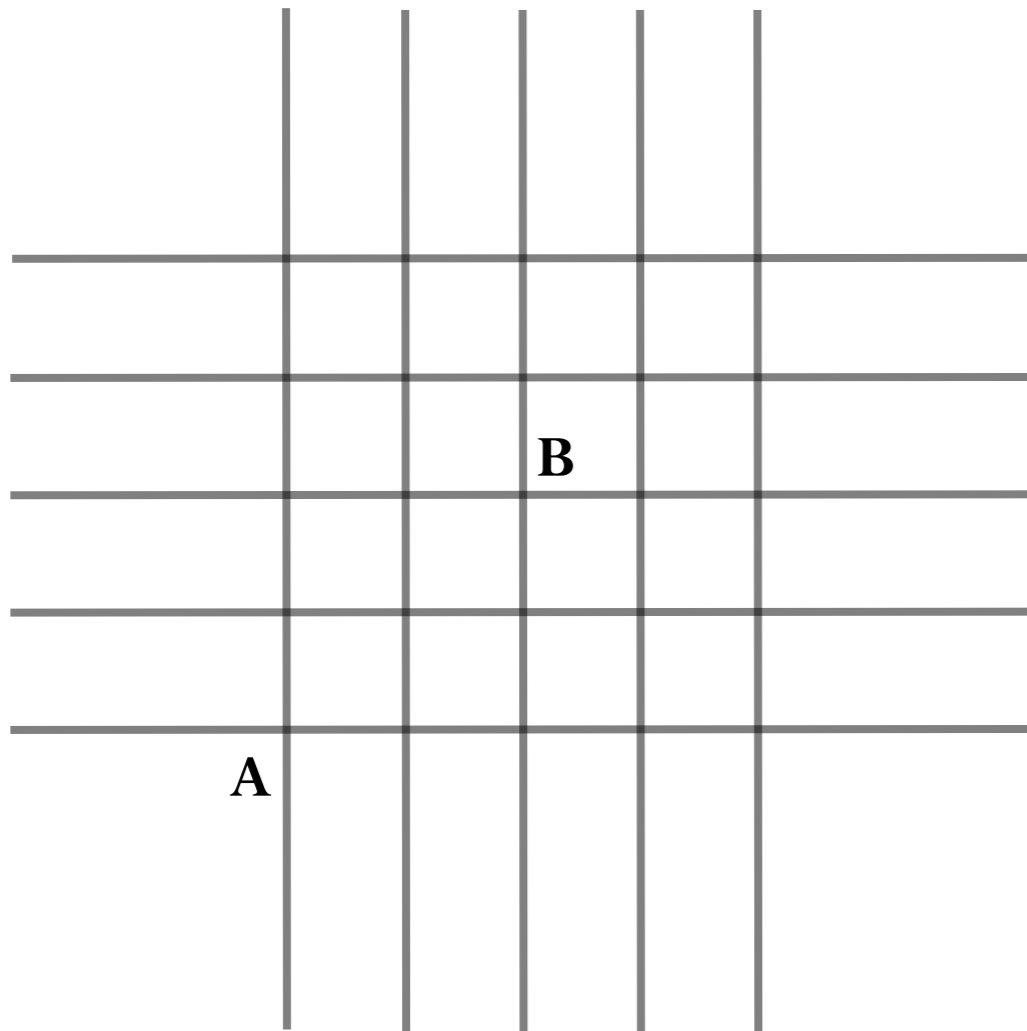
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 - **Distinct executions will lead to distinct execution paths and possibly distinct solutions**

Diversity within an algorithm

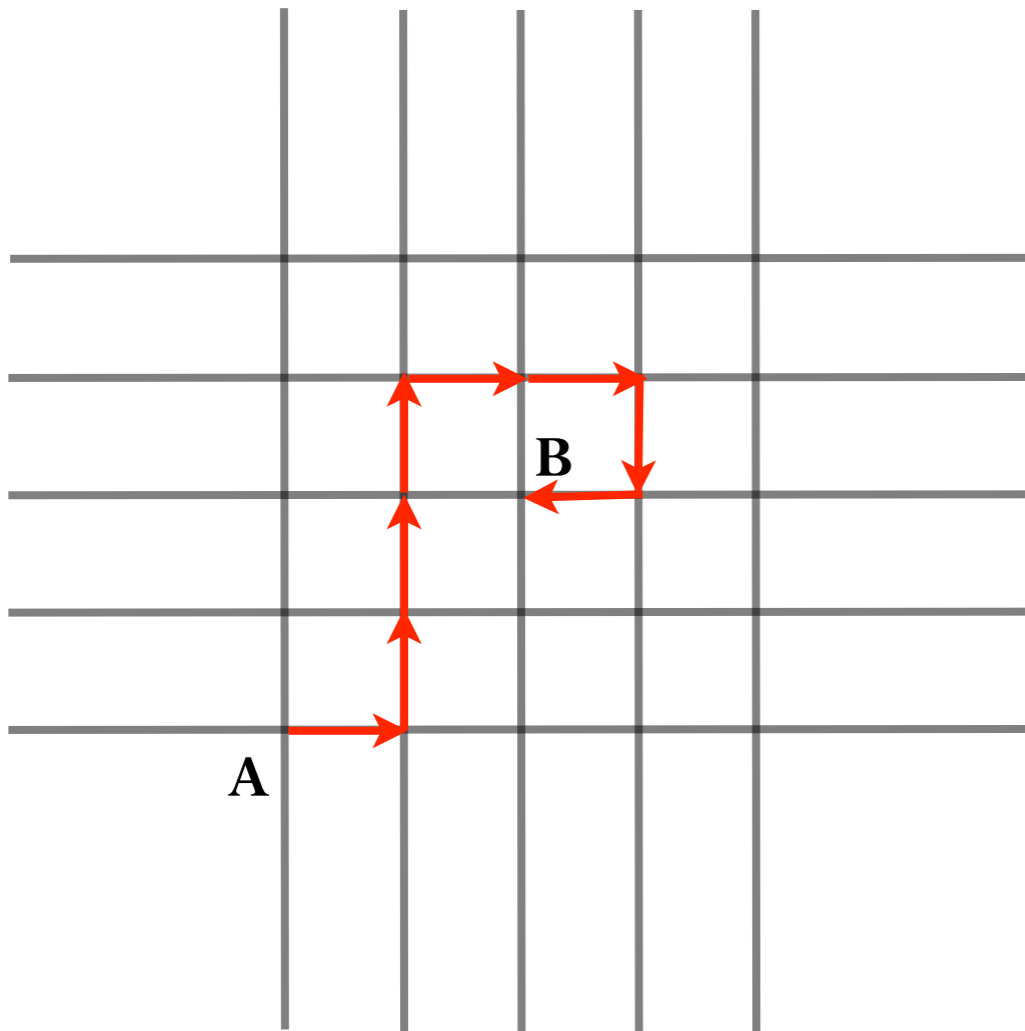
- **Certain algorithms are also intrinsically diverse:**
 - **Distinct executions will lead to distinct execution paths and possibly distinct solutions**
- **Intuition: randomized algorithms**
 - **Utilized for their simplicity and quicker execution**
 - **Random choices lead to different executions each time**

Example: random walk

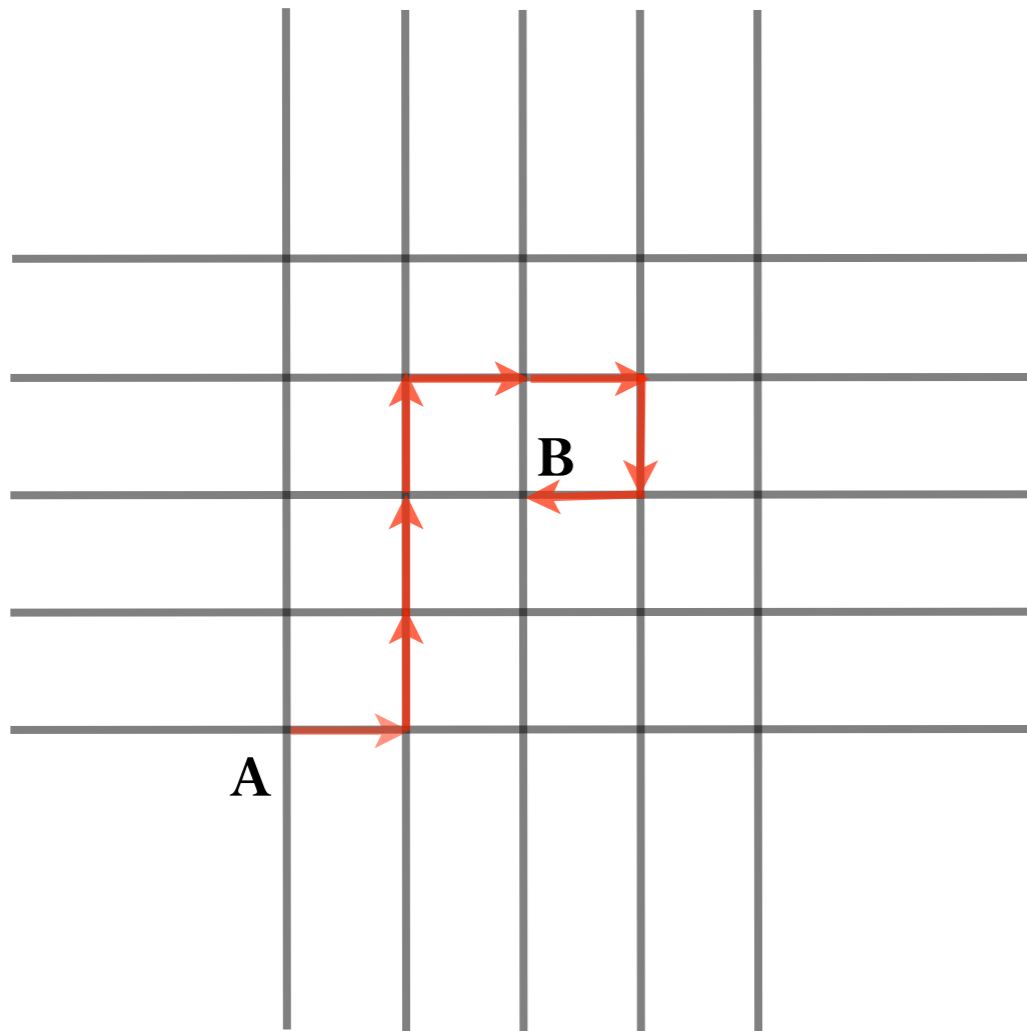
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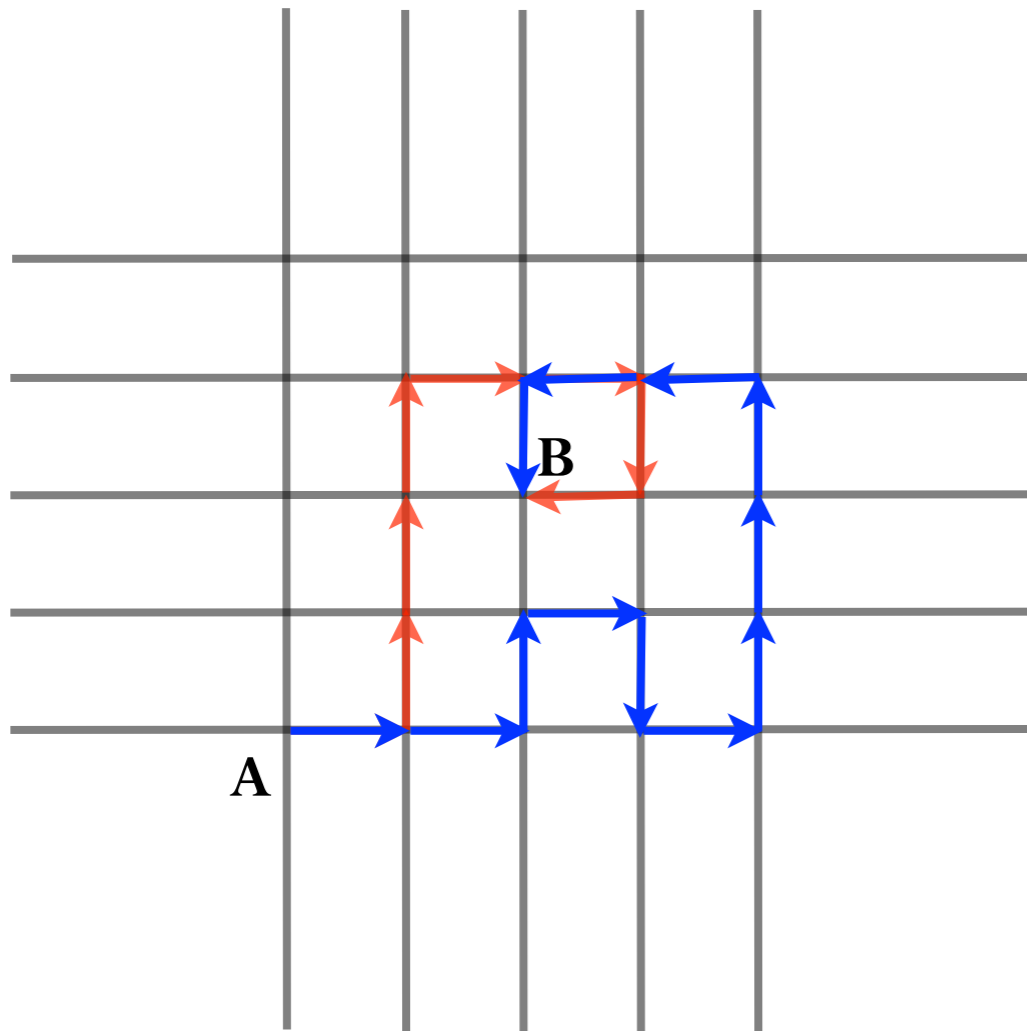
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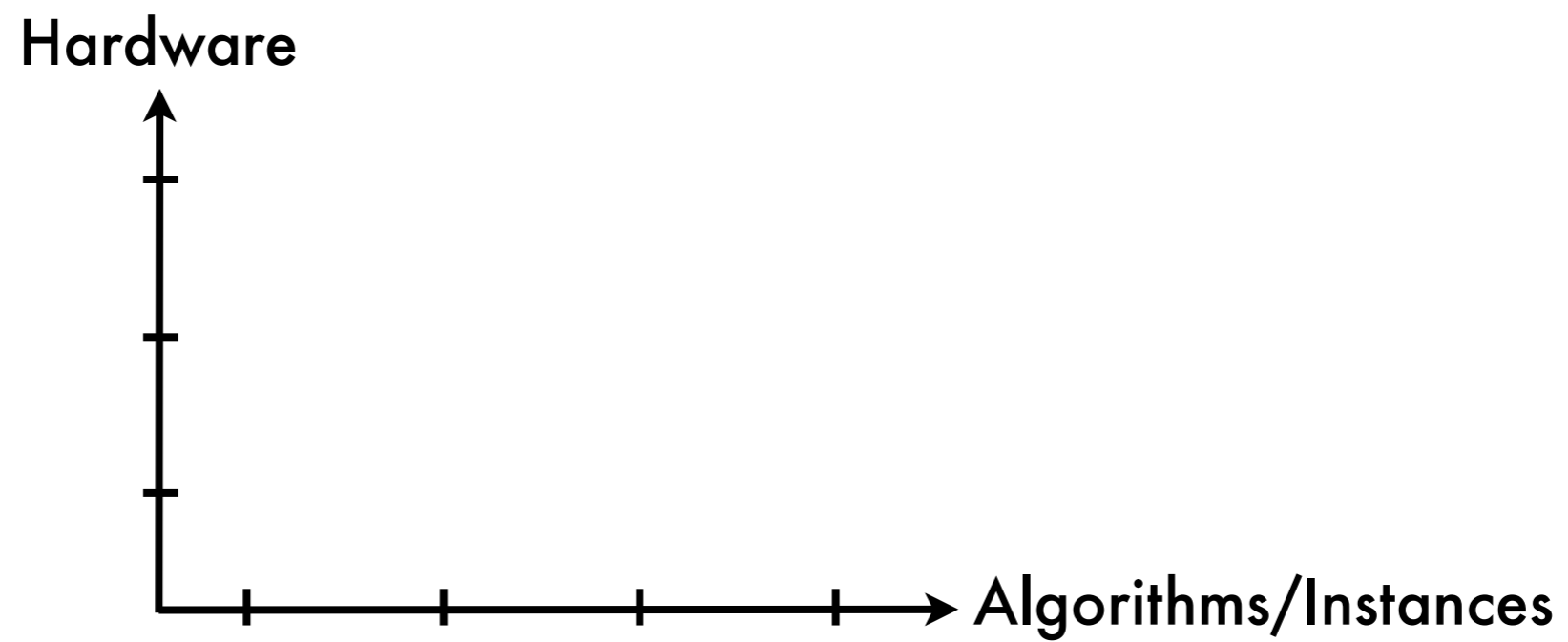
Example: random walk



- Differences in:
 - Memory accesses
 - Total compute steps
- Energy needs will be different
- Highly unpredictable

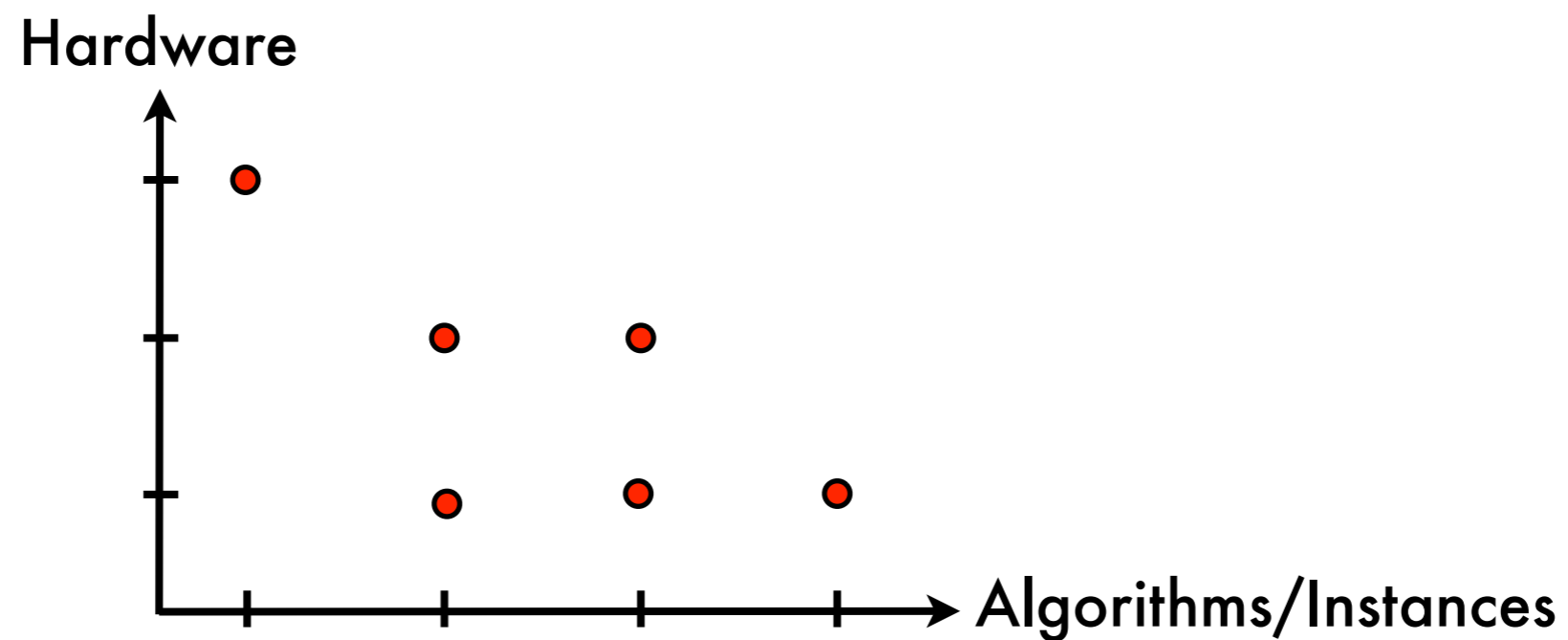
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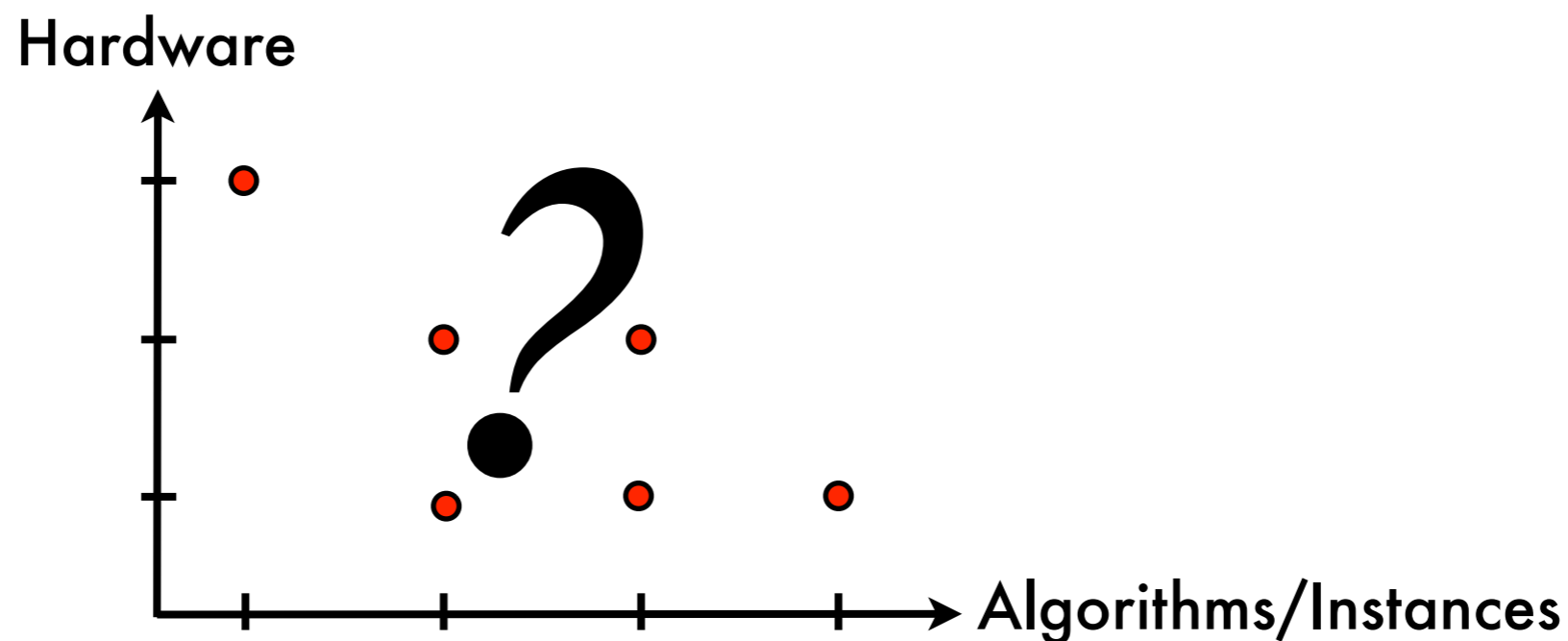
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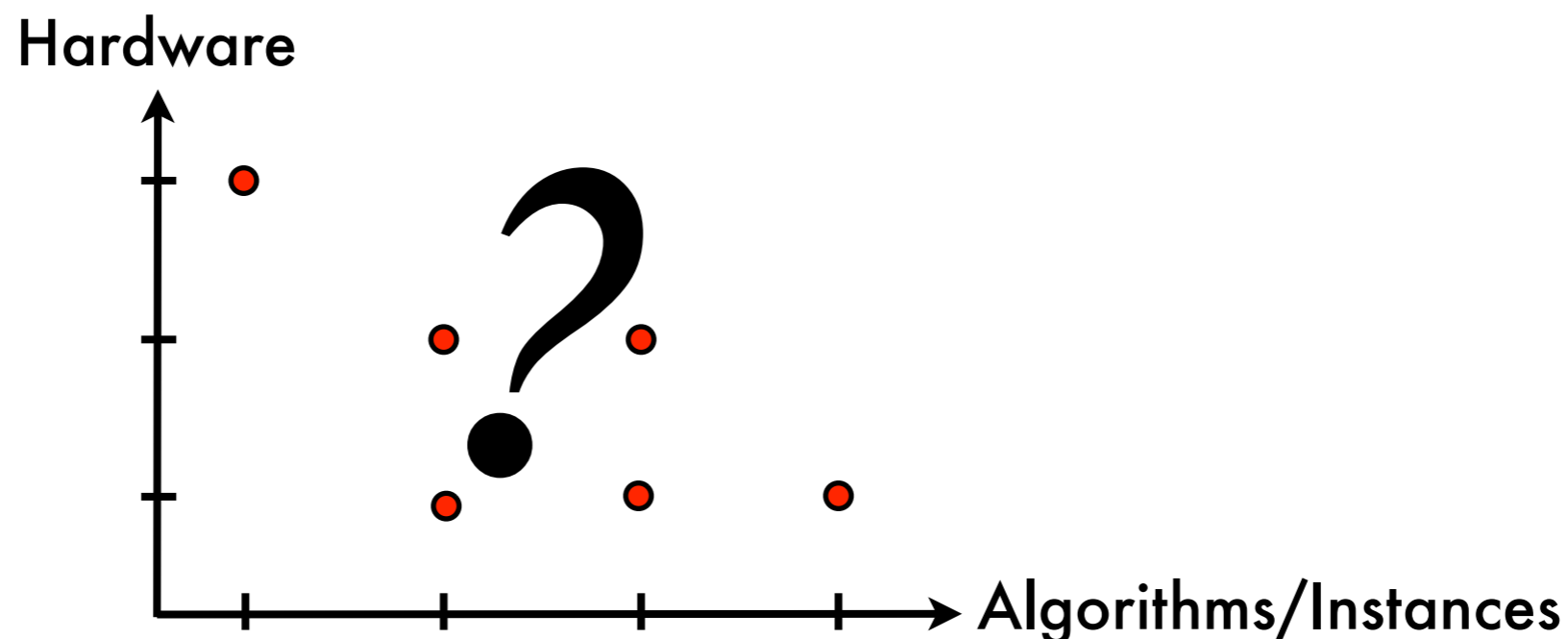
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- We want to utilize multi-cores to dynamically determine the best platform/algorithm match for a given input
- Multi-cores allow this exploration to be done without any assumptions
 - Just-in-time decision



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Energy efficiency

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- ***Efficient***: achieving maximum productivity with minimum wasted effort
- For an algorithm:
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- An algorithm is therefore energy efficient if it makes the most progress per unit of energy used

Progress measure

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- ***Progress* is the amount of work done towards finding the solution to the problem:**
 - **Defined in terms of the problem, not the algorithm**
 - **Very problem dependent**

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 - Defined in terms of the problem, not the algorithm
 - Very problem dependent
- Progress needs to be:
 - Monotonic: cannot undo progress
 - Comparable

Progress examples

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 - Number of elements correctly placed

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Progress examples

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 - Number of elements correctly placed
- Greedy constructive algorithms:
 - Natural notion of progress
- Search-based problems (SAT problems):
 - The amount of space explored
- Not always evident but widely applicable

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 - Run micro-benchmarks exercising PMCs
 - Measure power consumption with a Watt meter
 - Correlate PMCs and consumption and build a model

Energy measure

- Energy is hard to measure in real-time
- Approximate models are required
- Performance monitoring counters (PMCs) can be used:
 - Run micro-benchmarks exercising PMCs
 - Measure power consumption with a Watt meter
 - Correlate PMCs and consumption and build a model
- Estimating power in real-time is thus possible

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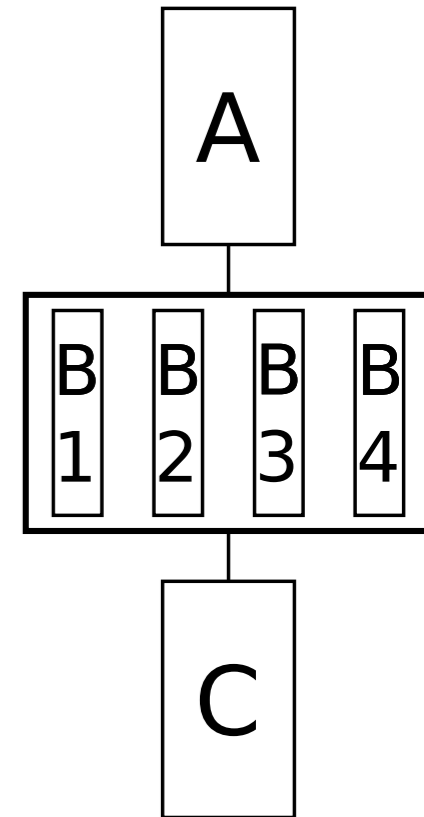
Diversity and the n-way model

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 - **To speedup or improve the QoR [HotPar 2009]**
 - **To find the best match just-in-time [PESPMA 2010]**

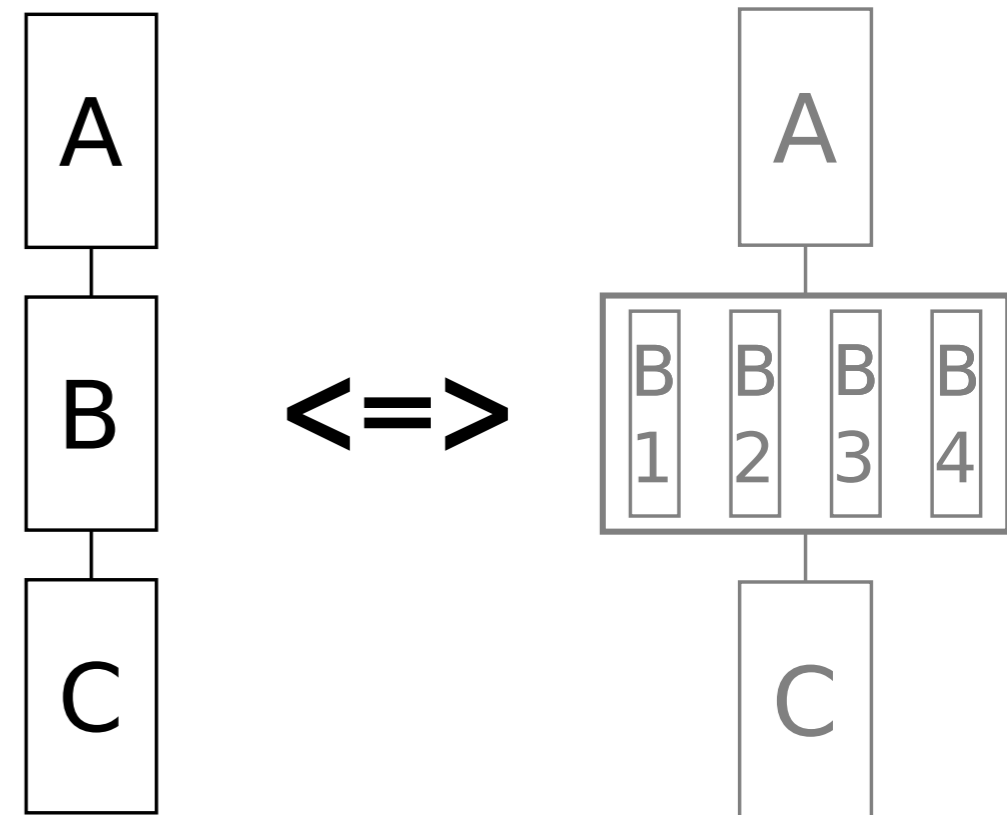
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- **Dynamic choice and culling**

- The most energy efficient way is automatically selected
- It must be late enough to have gathered enough information
- But also early enough to save energy

Available API

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- Identify a “problem” and attach “ways”
 - All ways solve the same problem

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```
/* Each way has its own “runWay” function */  
void runWay(NVWayMetrics& progressMetric, ...) {  
    <code for the way>  
    <updates progressMetric along the way>  
}
```

```
/* The main code */  
NVGoal myGoal;  
myGoal.attach(NVWay(runWay));  
<attach more ways>  
  
myGoal.run();
```

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 - Chosen for illustration purposes
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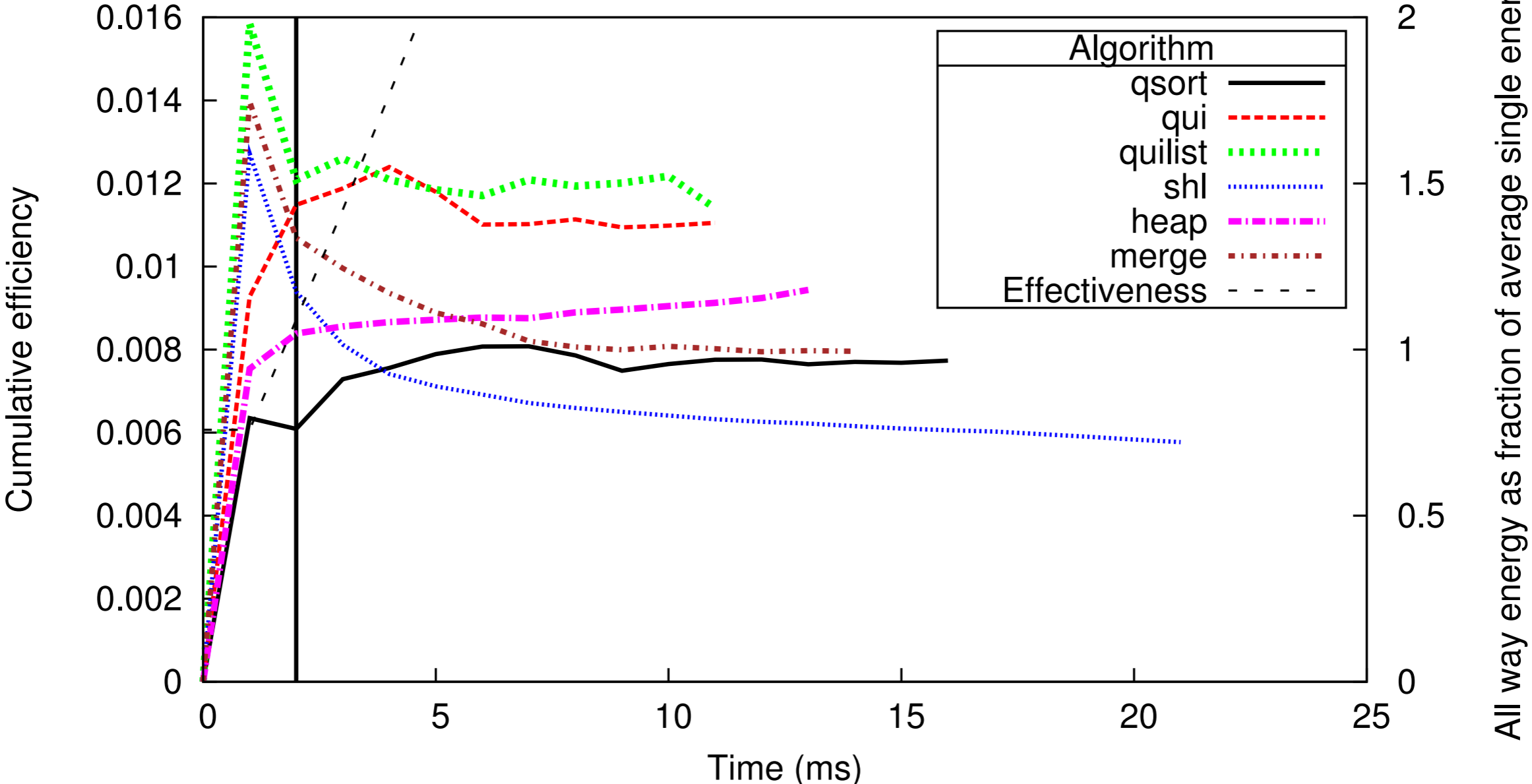
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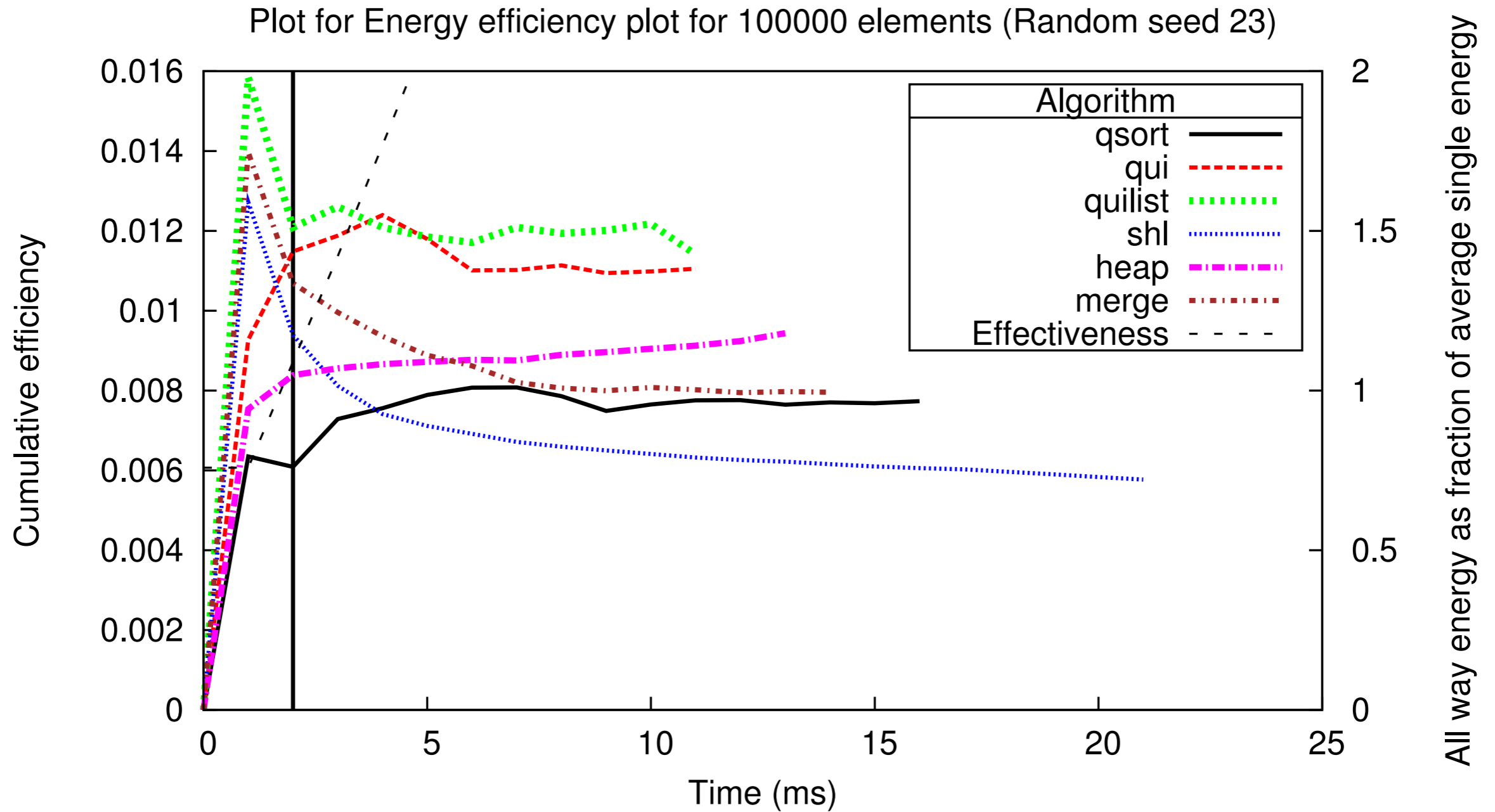
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- Used an Intel Core 2 Duo Q6700 with 2 GB of RAM
 - Patched 2.6.26 Linux kernel for PMC monitoring with PAPI

Preliminary results

Plot for Energy efficiency plot for 100000 elements (Random seed 23)



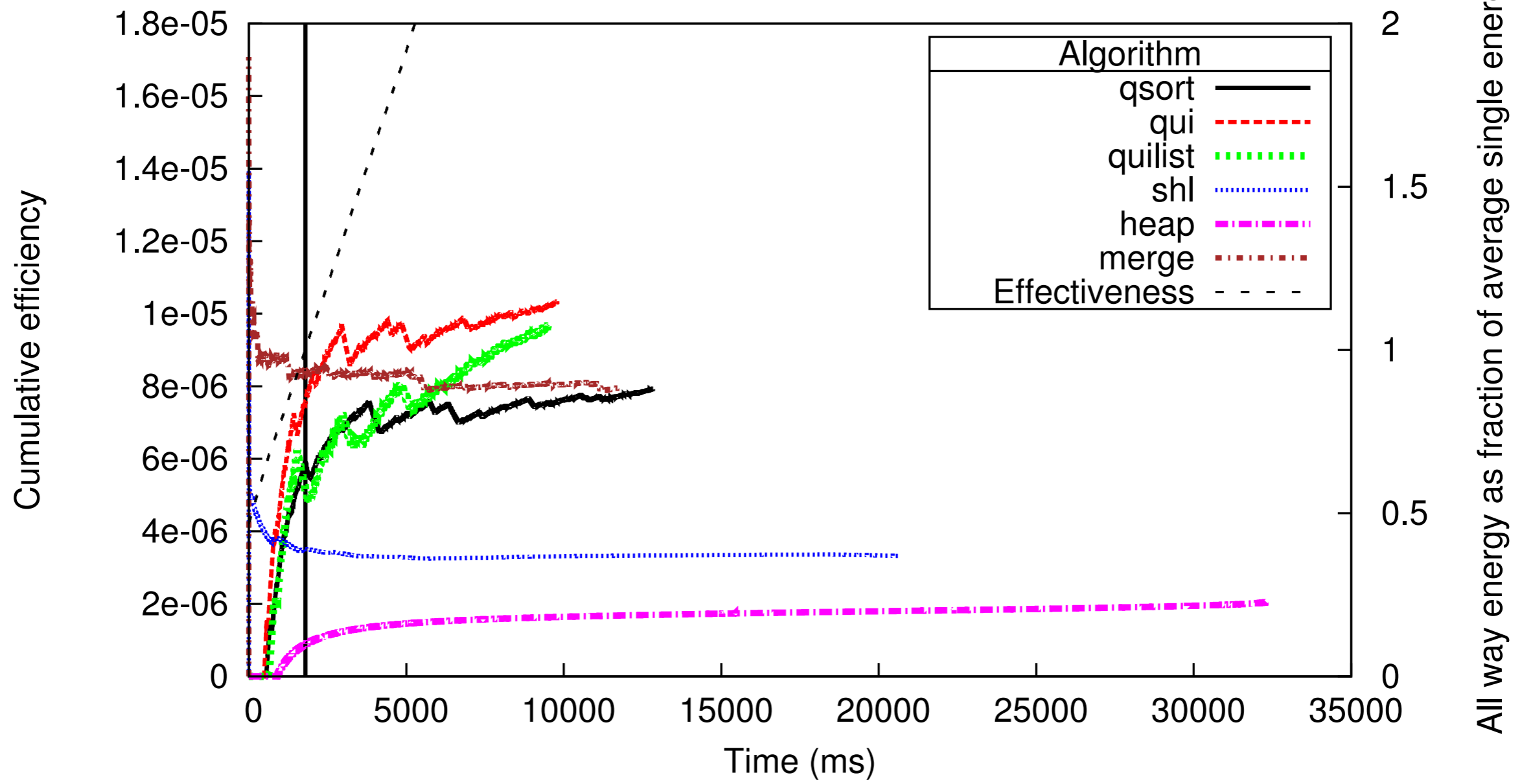
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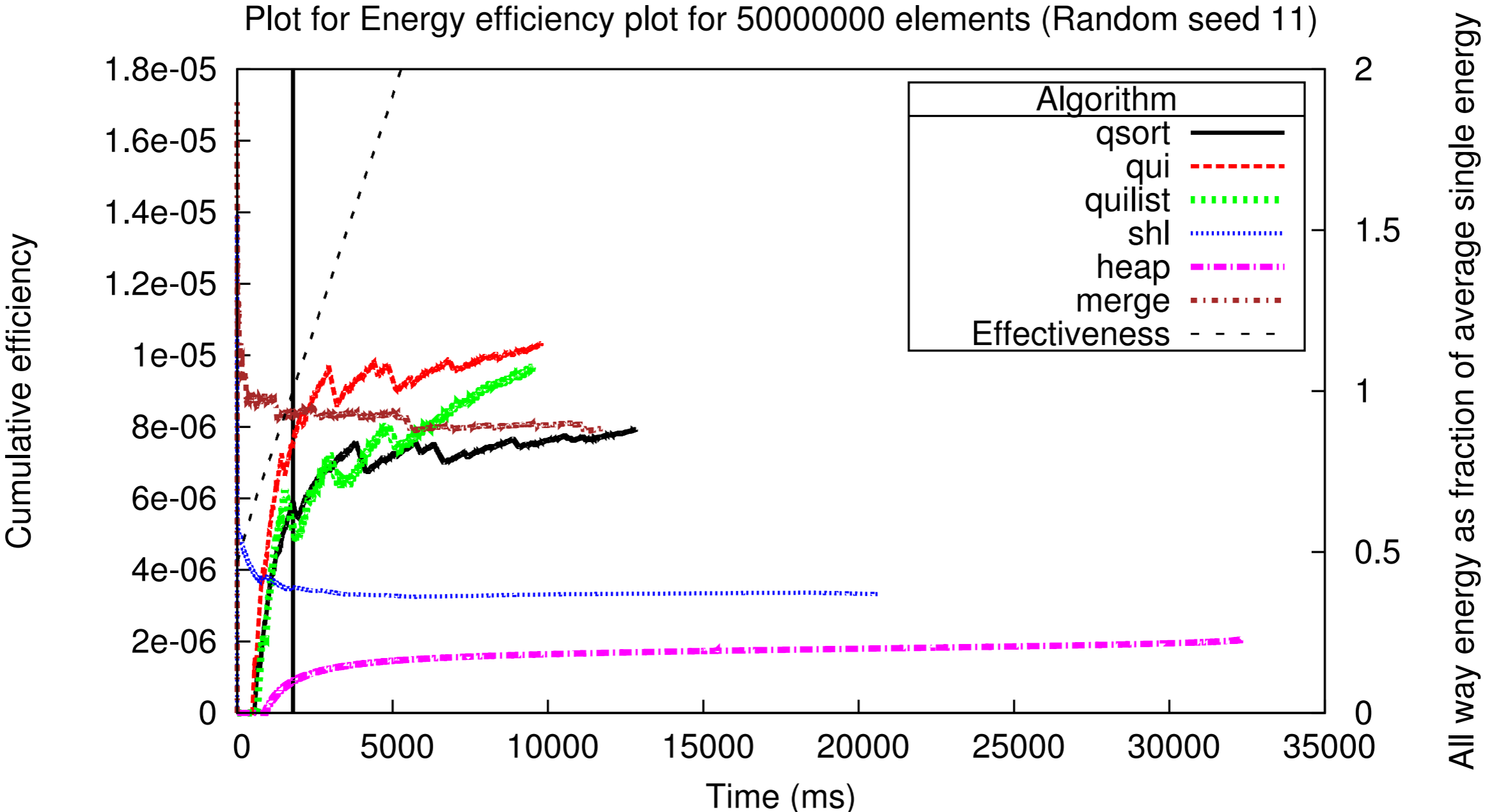
- Pick with good effectiveness possible

Preliminary results

Plot for Energy efficiency plot for 50000000 elements (Random seed 11)



Preliminary results



- "Profiling" becomes required because useful work is not equally distributed

Challenges

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- **Non-uniform energy efficiency**
- **Different algorithms have different *usage profile***
- **Possible to pre-profile applications and apply corrective factor**

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- **Non-uniform energy efficiency**
 - Different algorithms have different *usage profile*
 - Possible to pre-profile applications and apply corrective factor
- **Accurate monitoring**
 - Energy needs to be accurately modeled
 - Challenges with shared resources

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- Progress metric is important to define (monotonic, etc)
- Burst of energy use allows longer period when unused resources can be turned off
- Currently works with algorithms with similar “phases”
 - Looking for input on this and “power-models” for machine and algorithms

Thank you!

Reference: *Opportunistic Computing: a new paradigm for scalable realism on many cores*
R. Cledat, T. Kumar, J. Sreeram, S. Pande [HotPar 2009]