Space Profiling for Parallel Functional Programs

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Improving Performance – Profiling Helps!

Profiling improves functional program performance.

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Good performance in parallel programs is also hard.

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This work: space profiling for parallel programs

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Naïve NESL code for matrix multiplication 

function dot(a,b) = sum (\{ a * b : a; b \})

function prod(m,n) = \{ \{ dot(m,n) : n \} : m \}
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Given a parallel functional program, can we determine,

"How much space will it use?"

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Given a parallel functional program, can we determine,

"How much space will it use?"

Short answer: It depends on the implementation.

Scheduling Matters

Parallel programs admit many different executions

▶ not all impl. of matrix multiply are $O(n^3)$

Determined (in part) by scheduling policy

lots of parallelism; policy says what runs next

Semantic Space Profiling

Our approach: factor problem into two parts.

- 1. Define parallel structure (as graphs)
 - circumscribes all possible executions
 - deterministic (independent of policy, &c.)
 - include approximate space use
- 2. Define scheduling policies (as traversals of graphs)
 - used in profiling, visualization
 - gives specification for implementation

Contributions

Contributions of this work:

- cost semantics accounting for...
 - scheduling policies
 - space use
- semantic space profiling tools
- extensible implementation in MLton

Talk Summary

Cost Semantics, Part I: Parallel Structure

Cost Semantics, Part II: Space Use

Semantic Profiling

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Program Execution as a Dag

Model execution as directed acyclic graph (dag)

One graph for all parallel executions

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- ▶ limit number of nodes visited in each step

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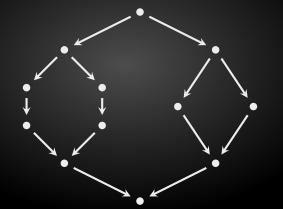
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A policy determines schedule for every program

Program Execution as a Dag (con't)



Program Execution as a Dag (con't)

Graphs are NOT...

- control flow graphs
- explicitly built at runtime

Graphs are...

- derived from cost semantics
 - unique per closed program
 - independent of scheduling

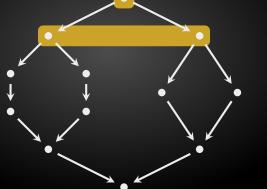
Breadth-First Scheduling Policy

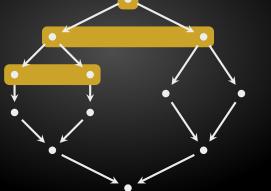
Scheduling policy defined by:

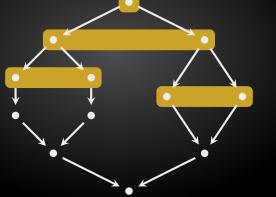
- breadth-first traversal of the dag(i.e. visit nodes at shallow depth first)
- break ties by taking leftmost node
- visit at most p nodes per step(p = number of processor cores)

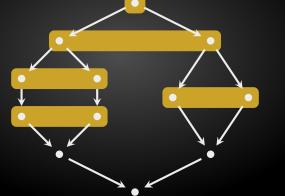


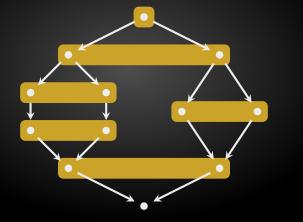


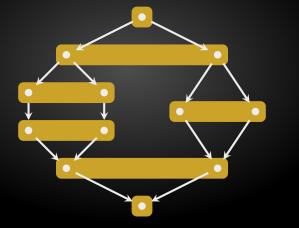












Breadth-First Scheduling Policy

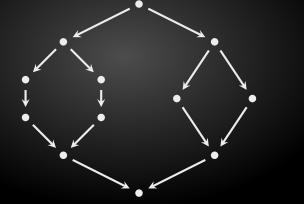
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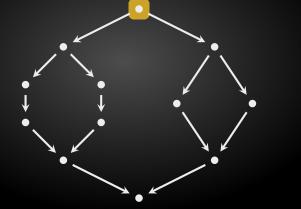
- breadth-first traversal of the dag
 (i.e. visit nodes at shallow depth first)
- break ties by taking leftmost node
- visit at most p nodes per step(p = number of processor cores)
- Variation implicit in impls. of NESL
 - & Data Parallel Haskell
 - vectorization bakes in schedule

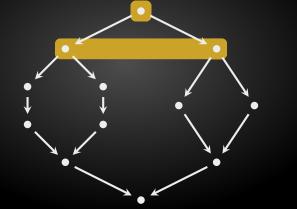
Depth-First Scheduling Policy

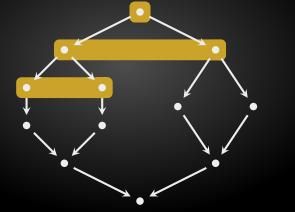
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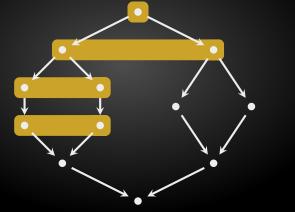
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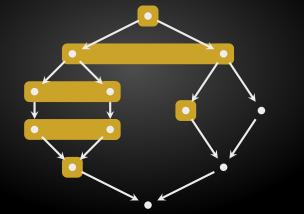


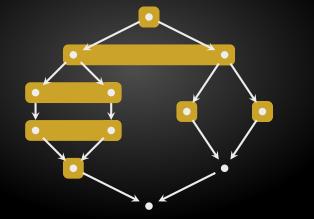




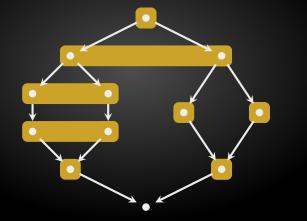




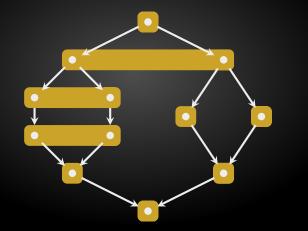




Depth-First Illustrated (p = 2)



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Depth-First Scheduling Policy

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 (i.e. favor children of recently visited nodes)
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Sequential execution

= one processor depth-first schedule

Work-Stealing Scheduling Policy

- "Work-stealing" means many things:
 - idle procs. shoulder burden of communication
 - specific implementations, e.g. Cilk
 - implied ordering of parallel tasks
- For the purposes of space profiling, ordering is important
 - briefly: globally breadth-first, locally depth-first

Computation Graphs: Summary

- Cost semantics defines graph for each closed program
 - ▶ i.e.. defines parallel structure
 - call this graph computation graph
- Scheduling polices defined on graphs
 - describe behavior without data structures, synchronization, &c.

Talk Summary

Cost Semantics, Part I: Parallel Structure

Cost Semantics, Part II: Space Use

Semantic Profiling

Heap Graphs

- Goal: describe space use independently of schedule
 - our innovation: add heap graphs
- Heap graphs also act as a specification
 - constrain use of space by compiler & GC
 - just as computation graph constrains schedule

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- Goal: describe space use independently of schedule
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- Heap graphs also act as a specification
 - constrain use of space by compiler & GC
 - just as computation graph constrains schedule
- Computation & heap graphs share nodes.
- think: one graph w/ two sets of edges

Generate costs for parallel pair,

```
\{e_1, e_2\}
```

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$$\{e_1,e_2\}$$



Generate costs for parallel pair,

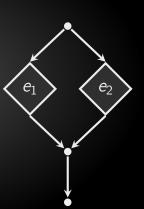
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$$e_1$$

 e_2

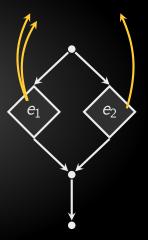
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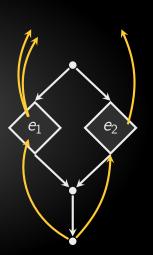
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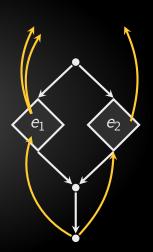
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Generate costs for parallel pair,

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(see paper for inference rules)



From Cost Graphs to Space Use

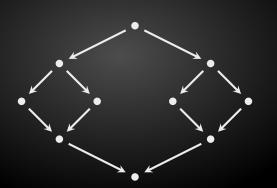
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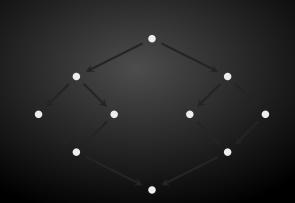
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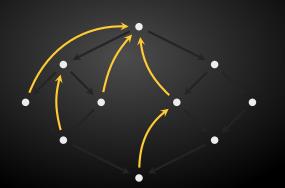
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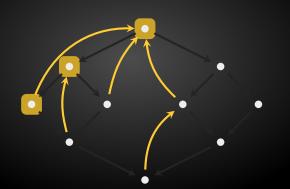
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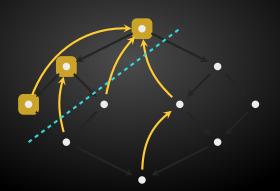
- visiting p nodes per step to simulate p processors
- Each step of traversal divides set of nodes into:
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 - 2. notes to be executed in future
- Heap edges crossing from future to past are "roots"
- ▶ i.e. future uses of existing values

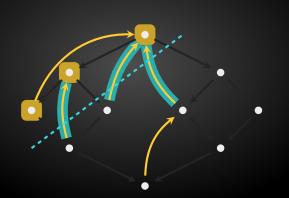












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if e_1 then e_2 else e_3

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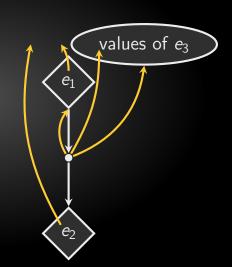
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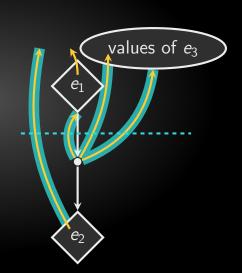
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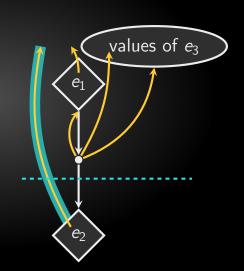
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Heap Graphs: Summary

Heap edge from B to A indicates a dependency on A . . . given knowledge up to time corresponding to B

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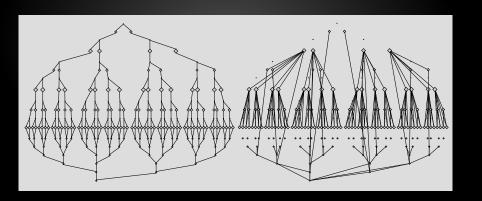
Some push back on semantics from implementation

- semantics must be implementable
 - ► *e.g.*, "true" vs. "provable" garbage

Example Graphs

Matrix multiplication

computation graph on left; heap on right



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not a static analysis

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 $Semantic \Rightarrow independent \ of \ implementation$

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Analysis of costs

- not a static analysis
- Semantics yields one set of costs per input
 - run program over many inputs to generalize
- ${\sf Semantic} \Rightarrow {\sf independent} \ {\sf of} \ {\sf implementation}$
 - ***** loses some precision
 - acts as specification

Visualizing Schedules

Distill graphs, focusing on parallel structure

- coalesce sequential computation
- use size, color, relative position
- omit less interesting edges

Visualizing Schedules

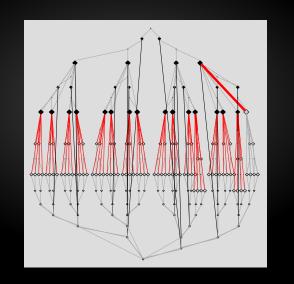
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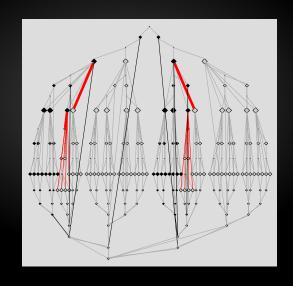
Graphs derived from semantics, ... compressed mechanically,

...then laid out with GraphViz

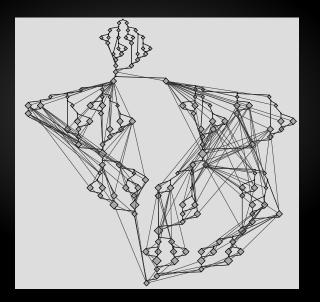
Matrix Multiply (Breadth-First, p = 2)



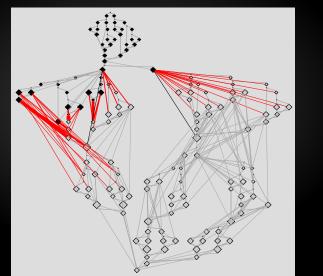
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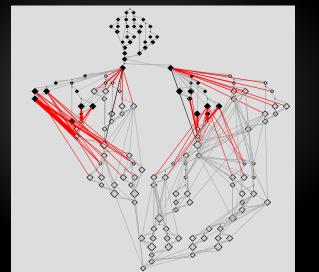
Quick Hull



Quick Hull (Depth First, p = 2)

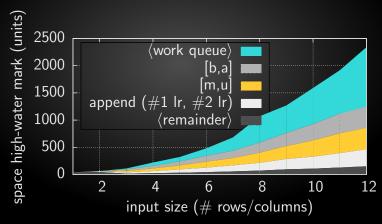


Quick Hull (Work Stealing, p = 2)



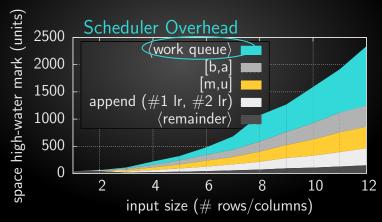
Space Use By Input Size

Matrix multiply w/ breadth-first scheduling policy:



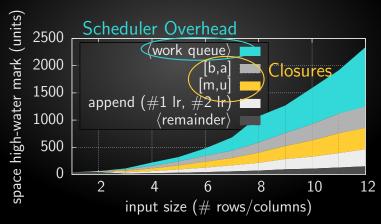
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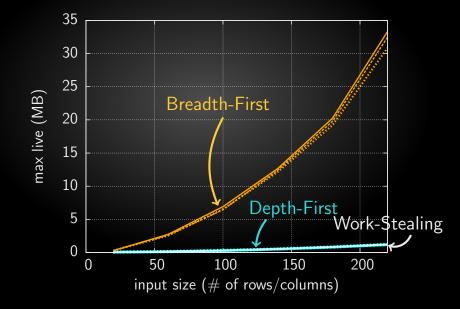
Verifying Profiling Results

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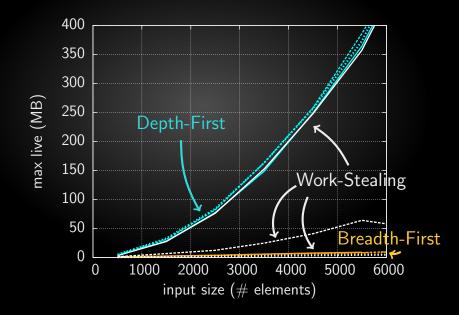
Implemented a parallel extension to MLton

- including three different schedulers
- compared predicted and actual space use

Matrix Multiply – MLton Space Use

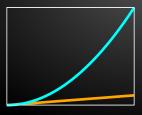


Quicksort - MLton Space Use



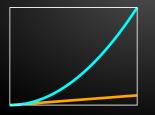
Initial Quicksort Results

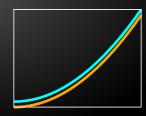
predicted: breadth-first outperforms depth-first



Initial Quicksort Results

- predicted: breadth-first outperforms depth-first
- initial observation: same results!





Space Leak Revealed

Cause: reference flattening optimization (representing reference cells directly in records)

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Now fixed in MLton source repository

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Without a cost semantics, there is no bug!

Also in the Paper

More details, including. . .

- rules for cost semantics
- discussion of MLton implementation
 - efficient method for space measurements
- more plots (profiling, speedup, &c.)
- application to vectorization (in TR)

Selected Related Work

- Cost semantics
 - ► Sansom & Peyton Jones. POPL '95
 - ▶ Blelloch & Greiner. *ICFP '96*
- Scheduling
 - ▶ Blelloch, Gibbons, & Matias. JACM '99
 - ▶ Blumofe & Leiserson. *JACM '99*
- Profiling
 - ▶ Runciman & Wakeling. *JFP '93*
 - ▶ ibid. Glasgow FP '93

Conclusion

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Semantic profiling for parallel programs...

- accounts for scheduling, space use
- constrains implementation (and finds bugs!)
- supports visualization & predicts actual performance

Thanks!

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Thanks to MLton developers, and Thank you for listening!
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Questions?
spoons@cmu.edu
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Download binaries, source code, papers, slides: http://www.cs.cmu.edu/~spoons/parallel/svn co svn://mlton.org/mlton/... branches/shared-heap-multicore mlton
```