

How to Make Loop-level Data Dependence Profiling Fast and Memory Efficient?

Hongtao Yu, Zhiyuan Li

Department of Computer Science
Purdue University

The bigger picture

- *How to retrofit existing application software for multicore processors?*
 - *Difficulties with programmer writing parallel programs*
 - *Race conditions*
 - *Memory consistency model*
 - *Execution model (process, thread, stream, vector, ...)*
 - *Granularity*
 - *Difficulties with using compilers to help*
 - *Alias problems*
 - *Symbolic analysis*
 - *Input-dependent behaviors*
 - *etc*

- *In both cases, a key issue is to understand the inherent data dependences thoroughly*

Loop-level data dependence

```
while (...) {
```

```
...
```

```
for (i = 0; i <= 65536; i++)
```

```
    ftab[i] = 0;
```

```
c1 = block[-1];
```

```
for (i = 0; i <= last; i++) {
```

```
    c2 = block[i];
```

```
    ftab[(c1 << 8) + c2]++;
```

```
    c1 = c2;}
```

```
for (i = 1; i <= 65536; i++)
```

```
    ftab[i] += ftab[i-1];
```

```
c1 = block[0];
```

```
for (i = 0; i < last;
```

```
    i++) {
```

```
    c2 = block[i+1];
```

```
    j = (c1 << 8) +
```

```
    c2;
```

```
    c1 = c2;
```

```
    ftab[j]--;
```

```
    zptr[ftab[j]] = i;}
```

```
...
```

```
for () {
```

```
    ... = zptr[..]
```

```
    } /* end while */
```

- `Zptr[]` and `ftab[]` are both privatizable;
- No flow dependence actually exists across the while-loop iterations;
- While loop is a *doacross* kind of loop w/ significant parallelism
- Current compilers cannot determine yet

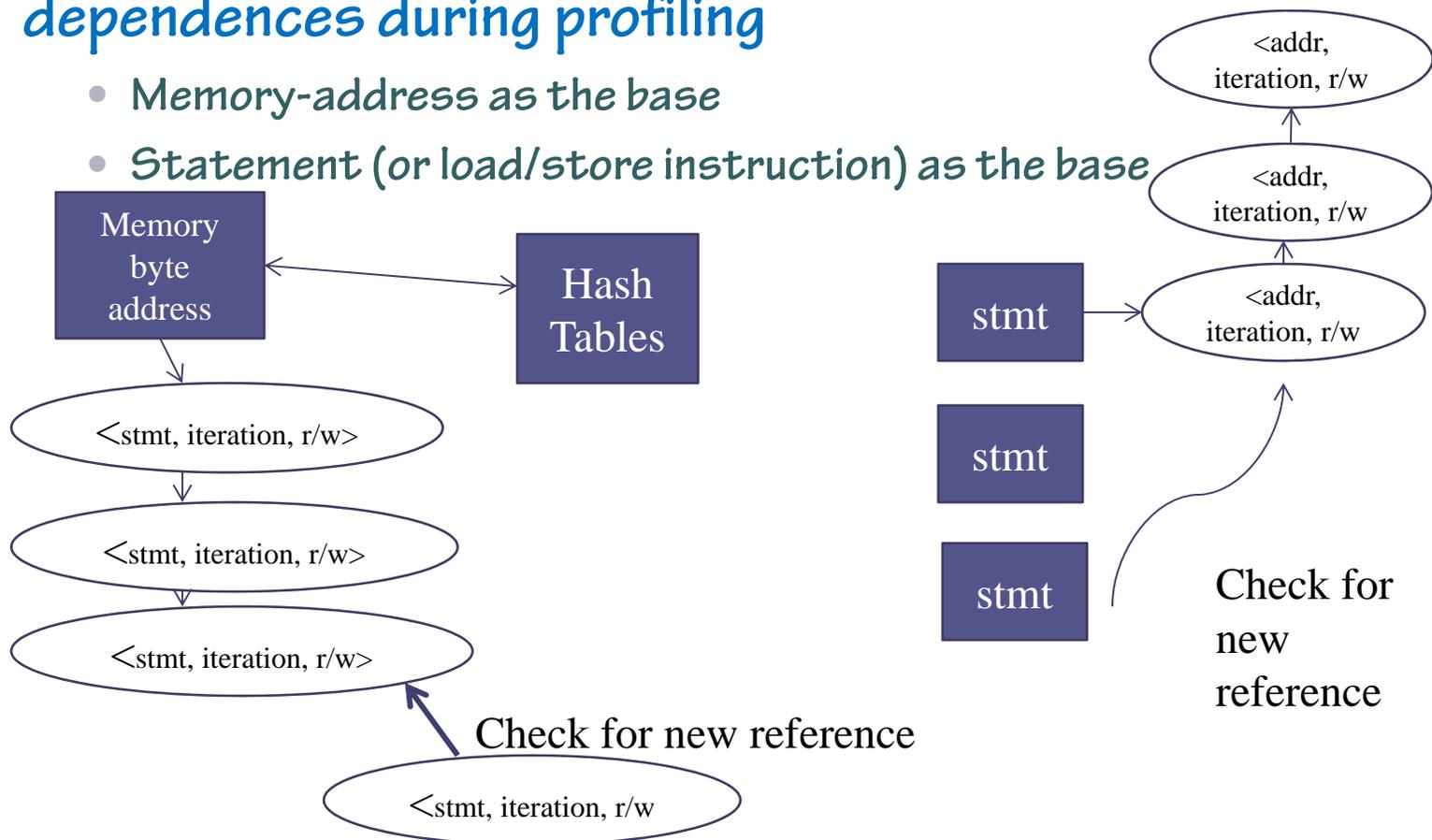
- **Data dependence profiling can be an important tool for such understanding**
 - Compiler writers can discover areas for improvement
 - Programmers can find code segments to rewrite
 - With speculative parallel execution support, we'll know which loop to speculate
- **Unfortunately, when done in a straightforward way, data dependence profiling is extremely time consuming**
 - Which is perhaps why it has not been widely used in practice
- **We look for ways to make it practical**
 - Reduce memory use (otherwise profiling may not even be feasible)
 - Reduce time (by one or two orders of magnitude)

Techniques to discuss today

- **Basic mechanisms**
 - Hash tables for memory addresses
 - Regular strides (and limitation)
- **Obvious compiler-based techniques to reduce cost**
 - Some works but some don't
- **“Multi-slicing” (or parallel profiling based on partitioning)**
 - Partitioning by alias classes
 - Partitioning by potential dependence edges
 - Input-dependent specialization (“thinned analysis”)
- **To make profiling much more practical**
 - Partial profiling that is sufficient for loop parallelization
- **A more efficient mechanism (memory-tags)**
 - Need compiler support

Basic Mechanisms

- Fundamentally there are two ways to check dependencies during profiling
 - Memory-address as the base
 - Statement (or load/store instruction) as the base



Regular strides

- **Method called SD3**

- M. Kim, H. Kim, and C.-K. Luk. *Sd3: A scalable approach to dynamic data-dependence profiling*. In MICRO, 2010
- When it works, it can speed up profiling by 100x

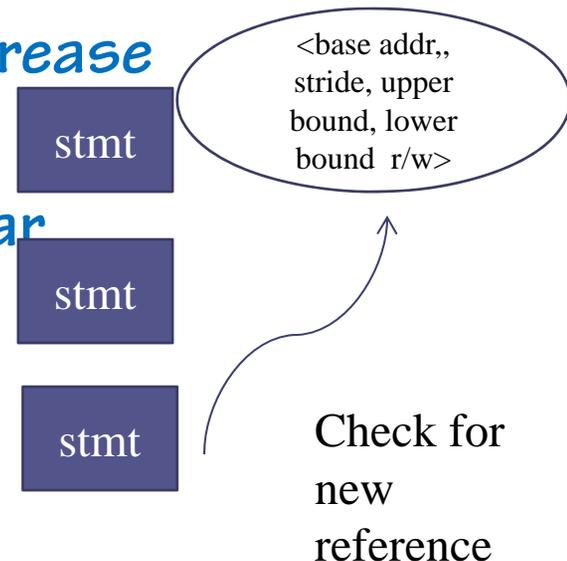
- **Unfortunately, when loop levels increase**

- Strides are often not regular anymore

- **When aliasing relationship is unclear**

- It is unsafe to use strides

- **Fall back to hash tables**



Compiler-based improvement

- Obviously, if the dependence question for a pair of load/store operations can be answered by the compiler,
 - then it does not need to be checked by profiling
- However, under the (memory-address) hash-table mechanism
 - If a load/store operation has a possible dependence with any other operation,
 - then it must be registered at (and checked against) the hash table
- Therefore in C programs, we often end up with registering and checking every memory reference
- Nonetheless, there exists a simple opportunity for improvement

Equivalence class and its representative for profiling

- Need symbolic analysis to make sure the address is the same in every iteration
- Need control flow information to identify upward-exposed use

A code segment from 456.hmm

```
133:   for (k = 1; k <= M; k++) {
134:     mc[k] = mpp[k-1] + tpmm[k-1];
135:     if ({sc = ip[k-1] + tpim[k-1]} > mc[k]) mc[k] = sc;
136:     if ({sc = dpp[k-1] + tpdm[k-1]} > mc[k]) mc[k] = sc;
137:     if ({sc = xmb + bp[k]} > mc[k]) mc[k] = sc;
138:     mc[k] = mc[k] + ms[k];
139:     if {mc[k] < -INFTY} mc[k] = -INFTY;
140:
141:     dc[k] = dc[k-1] + tpdd[k-1];
142:     if ({sc = mc[k-1] + tpmd[k-1]} > dc[k]) dc[k] = sc;
143:     if {dc[k] < -INFTY} dc[k] = -INFTY;
144:
145:     if (k < M) {
146:       ic[k] = mpp[k] + tpmi[k];
147:       if ({sc = ip[k] + tpim[k]} > ic[k]) ic[k] = sc;
148:       ic[k] += is[k];
149:       if {ic[k] < -INFTY} ic[k] = -INFTY;
150:     }
151:   }
```

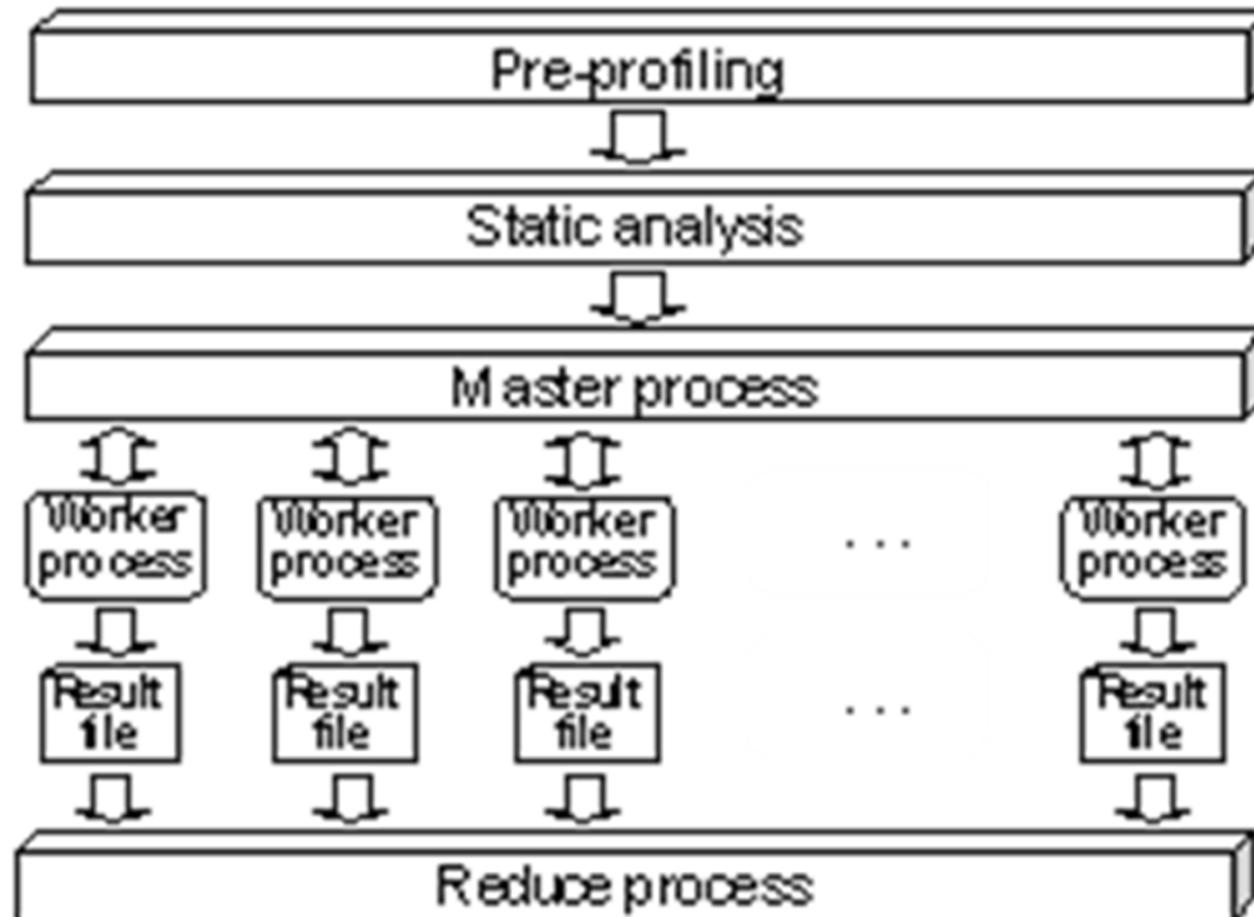
Static count of dependence edges

Benchmark	Original	Dependence checking	Equivalence Classification
401.bzip2	498202	294665	43344
429.mcf	12346	9478	844
433.milc	126336	93711	7358
456.hmmmer	632792	487546	71202
458.sjeng	5288671	4057566	482911
462.libquantum	62163	30113	2826
464.h264ref	5097323	4588307	64744
482.sphinx3	101688	61969	9278

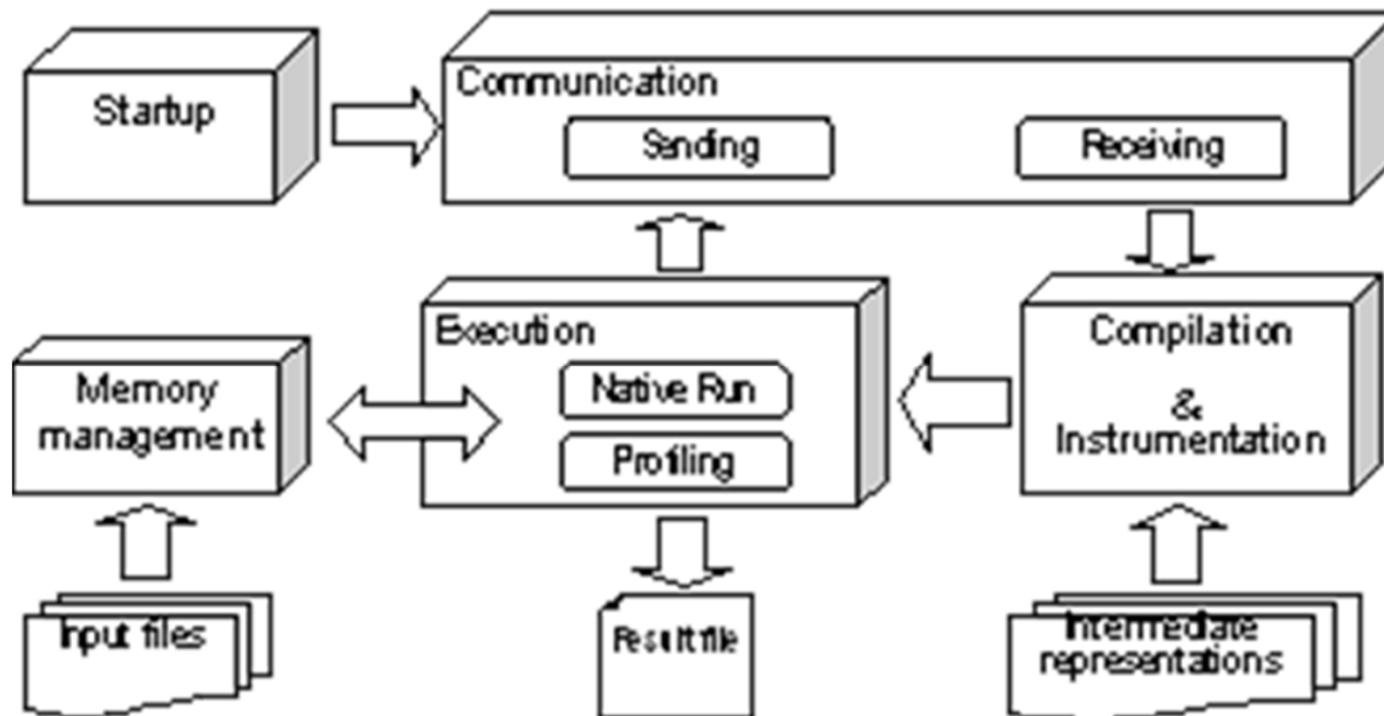
Parallel profiling

- **Basic idea**
 - **Instrumentation for a subset of potentially dependent memory references**
 - Run instrumentation and profiling for different subsets independently
 - Merge the results to produce the final dependence graph
- **Overhead due to extra instrumentation and program execution**
 - Controlling granularity and scheduling carefully for load balance
- **Partitioning by alias classes**
- **Partitioning by potential dependence edges**
 - Grouping dependence edges that must be profiled together (to recognize loop-independent dependences)
 - Granularity controlled by available computing resource
 - Dynamic scheduling by work stealing

A high-level overview of the multi-slicing approach



Details of worker processes



Implementation

- In the *GCC* compiler, version 4.6.
- Experiments with *SPECint 2006* benchmark suite
 - 401.bzip2, 429.mcf, 433.milc, 445.gobmk, 456.hmmer, 458.sjeng, 462.libquantum, 464.h264ref, and 482.sphinx3.
- “Test” size input

Visualizing DD graph by DDGrapher



Effect of thinned analysis

Benchmark	Ordinary Analysis		Thinned Analysis	
	ACP	#Edge class	ACP	#Edge class
401.bzip2	17.7	25661	13.9	21752
429.mcf	7.3	439	7.2	284
433.milc	7.4	3967	5.7	3415
456.hmmmer	4.4	49129	1.4	1487
458.sjeng	10.1	209034	10.0	172156
462.libquantum	4.2	515	2.9	229
464.h264ref	6.6	17975	5.9	17975
482.sphinx3	4.8	5336	1.3	3045

ACP (Alias Class Population): the average size of an alias class, counted by the number of load/stores.

Memory usage during profiling (MB)

Benchmark	Native	SD ³	Max Alias	Max Edge
401.bzip2	24	3.8G	532	70
429.mcf	154	3.6G	491	170
433.milc	9.3	>3G	>337	14
456.hmmer	1.1	3G	170	53
458.sjeng	174	>28G	>5.5G	191
462.libquantum	1.4	>721	223	14
464.h264ref	26	>1.1G	>352	179
482.sphinx3	30	>1.7G	>301	94

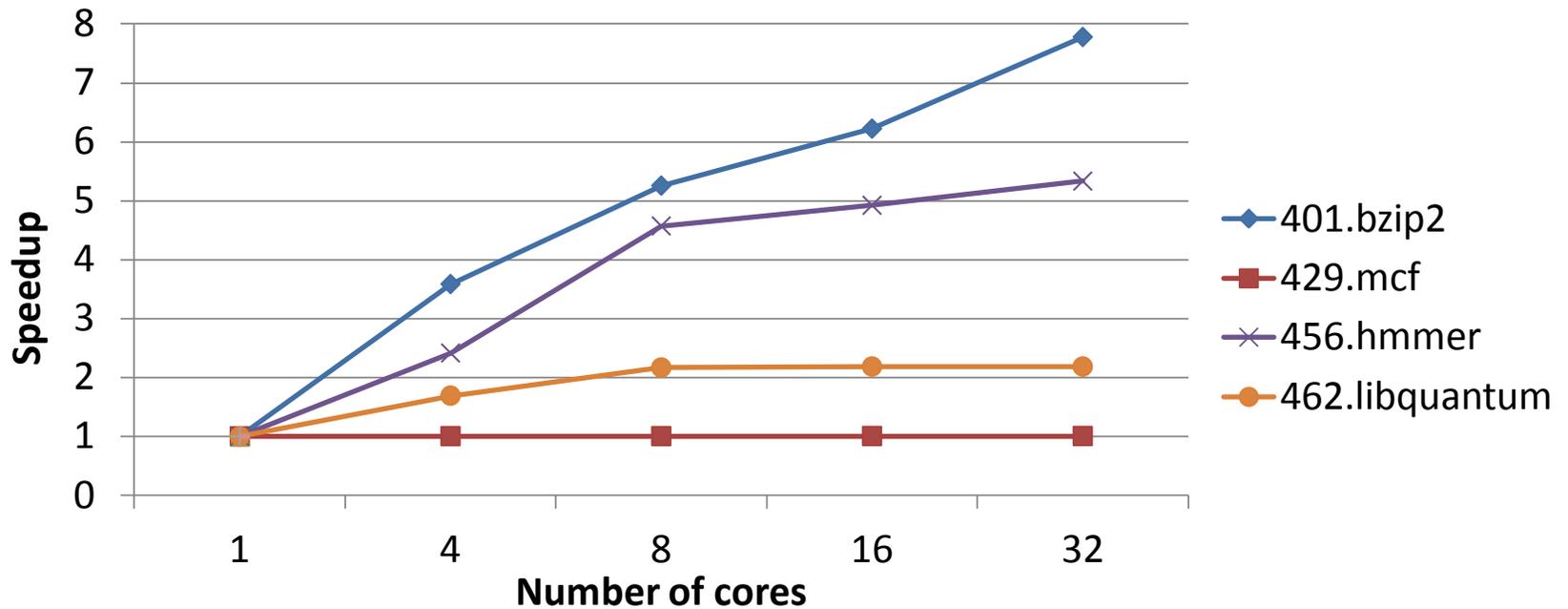
Profiling time with the multi-slicing approach

Benchmark	Native	SD ³	Alias Class					Edge Class						
			1	4	8	16	32	1	4	8	16	32	64	128
401.bzip2	3.7s	180h	77.8h	21.7h	14.8h	12.5h	10h	103.2h	27.4h	15.5h	7.7h	3.8h	3.2h	2.8h
429.mcf	6.7s	957h	489.9h	488.5h	488.2h	487.6h	487.3h	327.7h	111h	72.1h	43.7h	35.4h	30.7h	29.5h
433.milc	19s	×	×	×	×	×	×	42.5h	14.7h	7.7h	3.5h	1.8h	1h	42m
456.hmmer	16.5s	142h	12.8h	5.3h	2.8h	2.6h	2.4h	19.5h	9.6h	5h	2.5h	1.3h	55m	47m
458.sjeng	11s	×	×	×	×	×	×	914.3h	311h	153.8h	73.7h	37h	18.4h	9.8h
462.libquantum	7.6s	×	28.4h	16.8h	13.1h	13h	13h	24.5m	9.4m	6.5m	2.7m	1.6m	1.4m	1.1m
464.h264ref	22.6s	×	×	×	×	×	×	629.3h	214h	108.6h	52.5h	26.9h	13.5h	6.8h
482.sphinx3	6.5s	×	×	×	×	×	×	17.1h	6.1h	3.1h	1.5h	44m	23m	11.8m

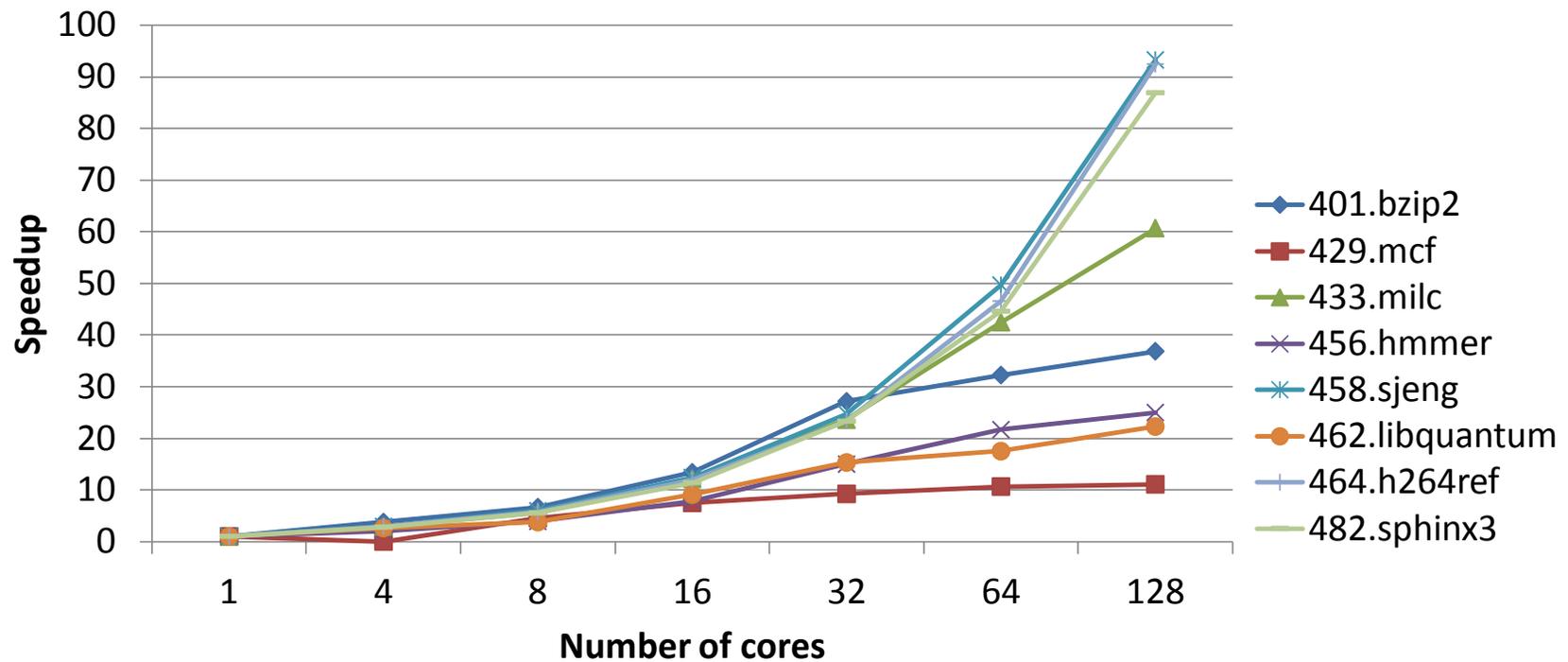
All versions compiled with “O0” option.

For profiling, “O0” allows precise drawing of dependence graph at source level

Multi-slicing parallel efficiency (alias-class partitioning)



Multi-slicing parallel efficiency (edge partitioning)



Characteristics of the Benchmarks ("reference" input size)

Benchmark	#KLOC	#File	#Loop	Loop profiled	#Run	#Iter	%Time
400.perlbench	170	126	1361	perl.c, 4751:4833	279	1.96	1%
401.bzip2	8	12	234	spec.c, 329:377	1	3	99.71%
403.gcc	521	278	4621	c-parse.c, 1874:4510	1	609709	99.80%
429.mcf	3	25	53	mcf.c, 48:104	1	6	99.33%
433.milc	15	89	445	update.c, 40:94	2	2	90.82%
445.gobmk	197	96	1359	interface/gtp.c, 82:135	1	484	99.98%
456.hmmcr	36	72	897	hmmcalibrate.c, 499:518	1	500000	99.99%
458.sjeng	14	23	306	epd.c, 313:334	1	9	99.99%
462.libquantum	4	31	102	expn.c, 45:54	1	21	97.81%
464.h264ref	51	81	1963	lencod.c, 305:424	1	62	99.95%
470.lbm	1.3	6	23	main.c, 39:51	1	300	99.16%
482.sphinx3	25	94	599	spec_main_live_pretend.c:167,189	1	24	99.81%

All versions compiled with "O0" option.

For profiling, "O0" allows precise drawing of dependence graph at source level

Timing and memory usage result of different profiling techniques (“reference” input size)

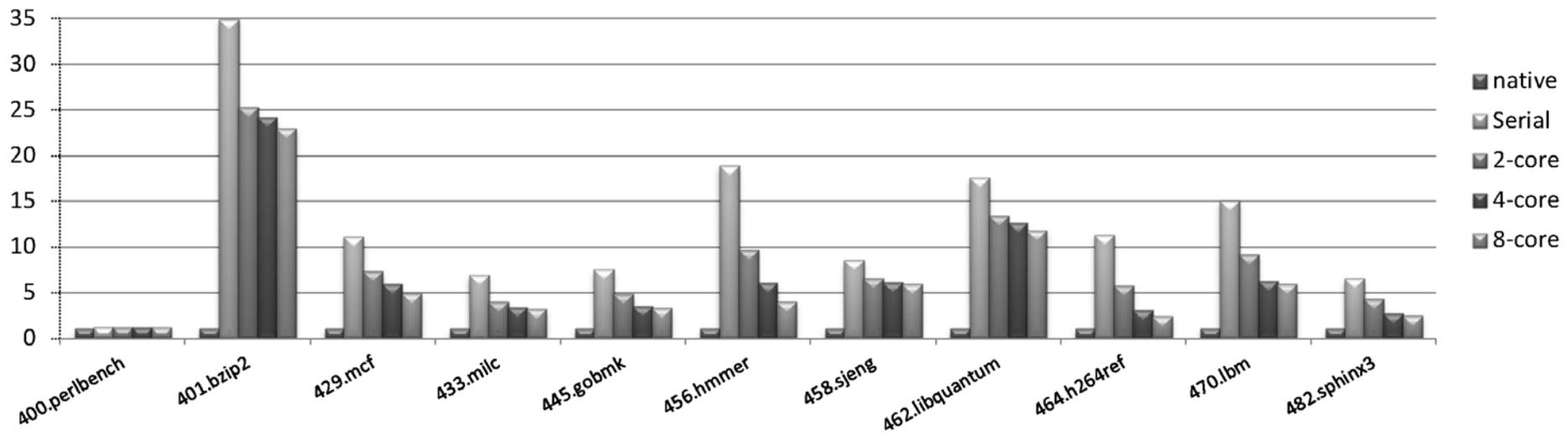
Benchmark	Time(Minutes)				Memory(MB)			
	Native	SD ³	Hash	Tag	Native	SD ³	Hash	Tag
400.perlbench	13.26	34.15	16.26	15.48	227	234	275	249
401.bzip2	9.16	>15000	364.46	319.65	849	×	1536	1487
403.gcc	4.97	>15000	×	×	846	×	>28G	>28G
429.mcf	18.31	>15000	×	203.43	1600	×	>28G	4812
433.milc	48.22	>15000	×	331.72	679	×	>28G	2048
445.gobmk	4.46	>15000	306.06	33.49	31	×	5222	99
456.hmmer	43.83	>15000	7858.14	829.61	3.8	×	1536	18
458.sjeng	39.72	>15000	2025.18	337.64	175	×	1945	629
462.libquantum	45.71	>15000	>15000	804.93	64	×	×	241
464.h264ref	7.84	>15000	634.07	88.54	18	×	730	63
470.lbm	2.88	>15000	×	43.38	409	×	>28G	1200
482.sphinx3	86.72	>15000	4687.52	564.38	41	×	18432	119

“Thinned analysis” and “partial profile” are both implemented in the “Hash” and “Tag” versions

Intel Review Seminar (Santa Clara, CA)

Parallel profiling time (minutes) with the memory-tag approach

Benchmark	Serial	2-core	4-core	8-core
400.perlbench	15.48	15.38	15.22	15.26
401.bzip2	319.65	231.6	221.07	210.22
429.mcf	203.43	133.40	108.26	89.9
433.milc	331.72	191.31	163.61	151.68
445.gobmk	33.49	21.55	15.64	14.46
456.hmmer	829.61	424.67	264.88	175.41
458.sjeng	337.64	260.18	240.89	233.61
462.libquantum	804.93	612.90	576.39	550.26
464.h264ref	88.54	45.07	24.08	18.64
470.lbm	43.38	26.41	17.99	16.91
482.sphinx3	564.38	368.87	231.89	212.40



Idea of partial but sufficient profiling

- *A partial dependence graph*
 - *contains only the latest data dependence concerned with each statement*
 - *The access record for each monitored memory address keeps only the most recent load and store.*
- *Under such simplification, we are able to capture the most recent instance of loop-independent and loop-carried dependence.*
- *We also keep the shortest dependence distance*

Algorithm 1 Partially Check Address.

```
1: procedure check_address(addr, opid, kind)
2: begin
3: if iterno > 0 then
4:    $\langle id_1, iterno_1, kind_1 \rangle = hash[addr].last\_read$ 
5:    $\langle id_2, iterno_2, kind_2 \rangle = hash[addr].last\_write$ 
6:   check whether  $\langle addr, opid, kind \rangle$  depends on ei-
   ther/both of the above triplets
7:   if kind == "read" then
8:      $hash[addr].last\_read = \langle opid, iterno, kind \rangle$ 
9:   else
10:     $hash[addr].last\_write = \langle opid, iterno, kind \rangle$ 
11:   end if
12: end if
13: end
```

Idea of partial but sufficient profiling

- Theorem 1 (Fundamental Theorem of Dependence) :

“Any reordering transformation that preserves every dependence in a program preserves the meaning of that program.”

- Theorem 2 :

A transformation or a loop execution schedule violates a dependence in the partial dependence graph if and only if it violates a dependence in the complete dependence graph

Memory tagging

- **Avoid hash table operations**
 - Especially to avoid byte-level hashing
- **Embed memory tags in the original data structure**
 - Direct comparison between memory tags
 - Fixed offset from the base address
- **Requires careful type analysis**
 - Especially in the presence of type cast, union, aliasing
- **Look for “promotable” objects**
 - Set up promotion rules

Recap

- **Basic mechanisms**
 - Hash tables for memory addresses
 - Regular strides (and limitation)
- **Obvious compiler-based techniques to reduce cost**
 - Some works but some don't
- **“Multi-slicing” (or parallel profiling based on partitioning)**
 - Partitioning by alias classes
 - Partitioning by potential dependence edges
 - Input-dependent specialization (“thinned analysis”)
- **To make profiling much more practical**
 - Partial profiling that is sufficient for loop parallelization
- **A more efficient mechanism (memory-tags)**
 - Need compiler support
- **Future work:**
 - We are yet to combine partial profiling and memory tagging with edge partitioning
 - Use profiling result to produce parallel code automatically

- *To be continued ...*