

332

Advanced Computer Architecture

Chapter 3: Caches and Memory Systems Part 2: miss rate reduction using software

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These lecture notes are partly based on the course text, Hennessy and Patterson's *Computer Architecture, a quantitative approach* (3rd, 4th, 5th and 6th eds), and on the lecture slides of David Patterson and John Kubiawicz's Berkeley course

Average memory access time:

$$\text{AMAT} = \text{HitTime} + \text{MissRate} \times \text{MissPenalty}$$

There are three ways to improve AMAT:

1. Reduce the miss rate,
2. Reduce the miss penalty, or
3. Reduce the time to hit in the cache

In hardware
In software

A diagram consisting of two arrows pointing from the text 'Reduce the miss rate,' to the words 'In hardware' and 'In software'. The words 'In software' are circled in black.

We now look at each of these in turn...

Reducing misses by software prefetching

- Some processors have instructions to trigger prefetching explicitly, in software
- *Almost never worth using on sophisticated processors with good hardware prefetching*
- May be useful on simpler processors →
- Some care is needed to ensure prefetch accesses don't have unwanted side effects
 - Eg memory-mapped i/o registers
 - (this is the function of the **R10KCBARRIER** macro)
- Prefetch instructions may target addresses that would cause a page fault or protection violation
 - Prefetches of addresses that would result in a page fault or exception are silently squashed

```

1:
R10KCBARRIER(0(ra))
LOAD(t0, UNIT(0)(src), .Ll_exc\@)
LOAD(t1, UNIT(1)(src), .Ll_exc_copy\@)
LOAD(t2, UNIT(2)(src), .Ll_exc_copy\@)
LOAD(t3, UNIT(3)(src), .Ll_exc_copy\@)
SUB    len, len, 8*NBYTES
LOAD(t4, UNIT(4)(src), .Ll_exc_copy\@)
LOAD(t7, UNIT(5)(src), .Ll_exc_copy\@)
STORE(t0, UNIT(0)(dst), .Ls_exc_p8u\@)
STORE(t1, UNIT(1)(dst), .Ls_exc_p7u\@)
LOAD(t0, UNIT(6)(src), .Ll_exc_copy\@)
LOAD(t1, UNIT(7)(src), .Ll_exc_copy\@)
ADD    src, src, 8*NBYTES
ADD    dst, dst, 8*NBYTES
STORE(t2, UNIT(-6)(dst), .Ls_exc_p6u\@)
STORE(t3, UNIT(-5)(dst), .Ls_exc_p5u\@)
STORE(t4, UNIT(-4)(dst), .Ls_exc_p4u\@)
STORE(t7, UNIT(-3)(dst), .Ls_exc_p3u\@)
STORE(t0, UNIT(-2)(dst), .Ls_exc_p2u\@)
STORE(t1, UNIT(-1)(dst), .Ls_exc_p1u\@)
PREFS( 0, 8*32(src) )
PREFD( 1, 8*32(dst) )
bne    len, rem, 1b
nop
  
```

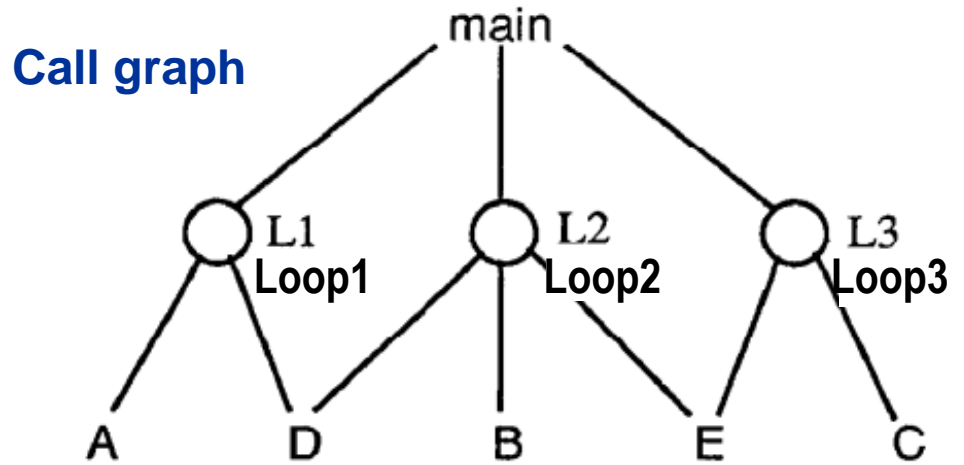
- Example: MIPS “memcpy” library code – handwritten assembler – unrolled 12 times, manually scheduled, with prefetching to initiate loading the source and destination cache lines into cache (heavy use of macros)
- From <https://elixir.bootlin.com/linux/v5.9.2/source/arch/mips/lib/memcpy.S>

Reducing instruction-cache misses

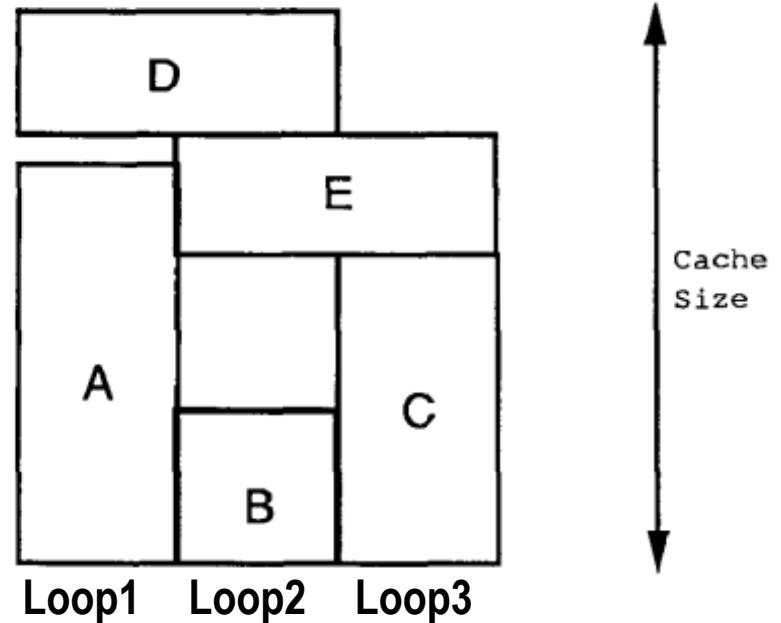
- McFarling [1989]* reduced instruction cache misses by 75% on 8KB direct mapped cache, 4 byte blocks in software

● Instructions

- By choosing instruction memory layout based on callgraph, branch structure and profile data
- Reorder procedures in memory so as to reduce conflict misses
- (actually this really needs the *whole* program – a link-time optimisation)



Packing code for each function into the I-cache



time →

Function E is placed to avoid conflicts with B and C, but can be placed in addresses that conflict with A

* "Program optimization for instruction caches", ASPLOS89, <http://doi.acm.org/10.1145/70082.68200>

● Storage layout transformations

- **Merging Arrays**: improve spatial locality by single array of compound elements vs. 2 arrays
- **Permuting a multidimensional array**: improve spatial locality by matching array layout to traversal order

- Improve *spatial* locality

● Iteration space transformations

- **Loop Interchange**: change nesting of loops to access data in order stored in memory
- **Loop Fusion**: Combine 2 independent loops that have same looping and some variables overlap
- **Blocking**: Improve temporal locality by accessing “blocks” of data repeatedly vs. going down whole columns or rows (wait for Chapter 4)

- Can also improve *temporal* locality

Array Merging - example

```
/* Before: 2 sequential arrays */  
int val[SIZE];  
int key[SIZE];
```

```
/* After: 1 array of structures */  
struct merge {  
    int val;  
    int key;  
};  
struct merge merged_array[SIZE];
```

“Array of Structs” vs
“Struct of Arrays”

(AoS vs SoA)

Reducing conflicts between val & key (example?)

Improve spatial locality (counter-example?)

- **whether this is a good idea depends on access pattern**

(actually this is a transpose: $2 \times \text{SIZE} \rightarrow \text{SIZE} \times 2$)

Consider matrix-matrix multiply (tutorial ex)

● MM1:

```

for (i=0;i<N;i++)
  for (j=0;j<N;j++)
    for (k=0;k<N;k++)
      C[i][j] += A[i][k] * B[k][j];


```

● MM2:

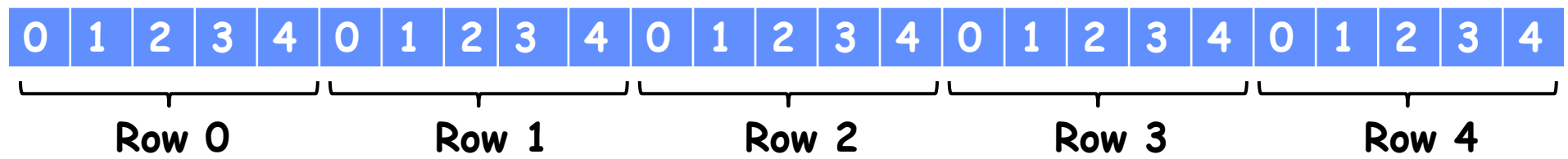
```

for (i=0;i<N;i++)
  for (k=0;k<N;k++)
    for (j=0;j<N;j++)
      C[i][j] += A[i][k] * B[k][j];

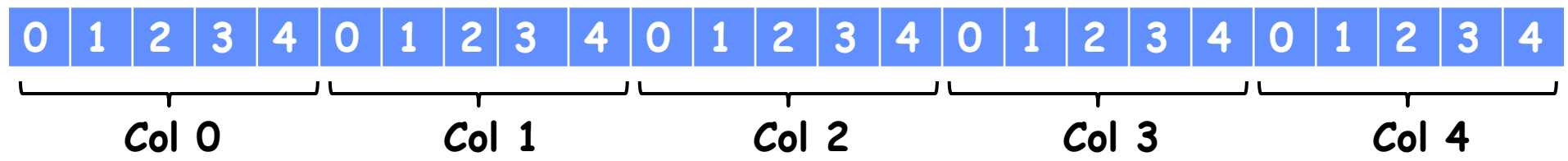
```

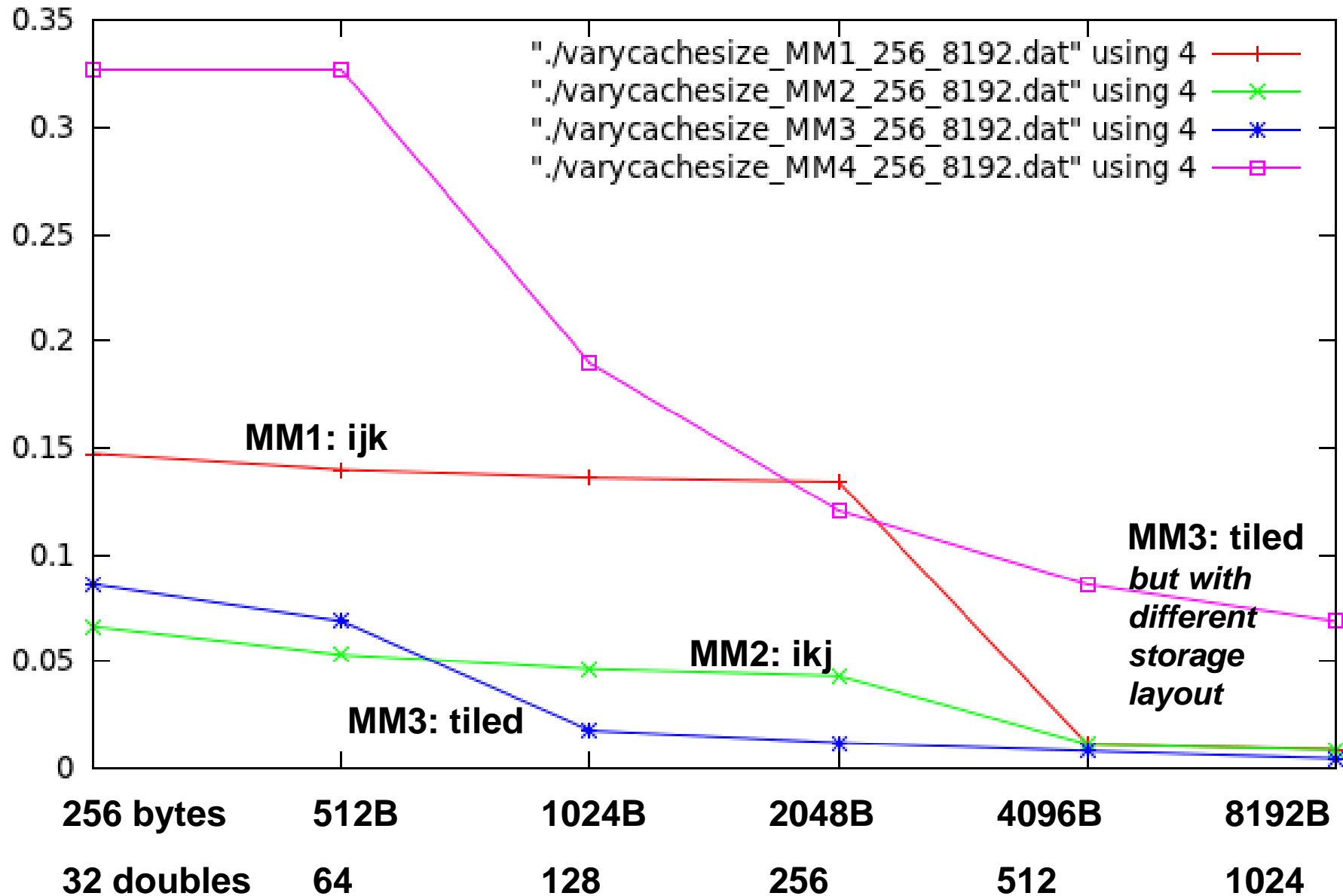


● Row-major storage layout (default for C):



● Column-major storage layout (default for Fortran):

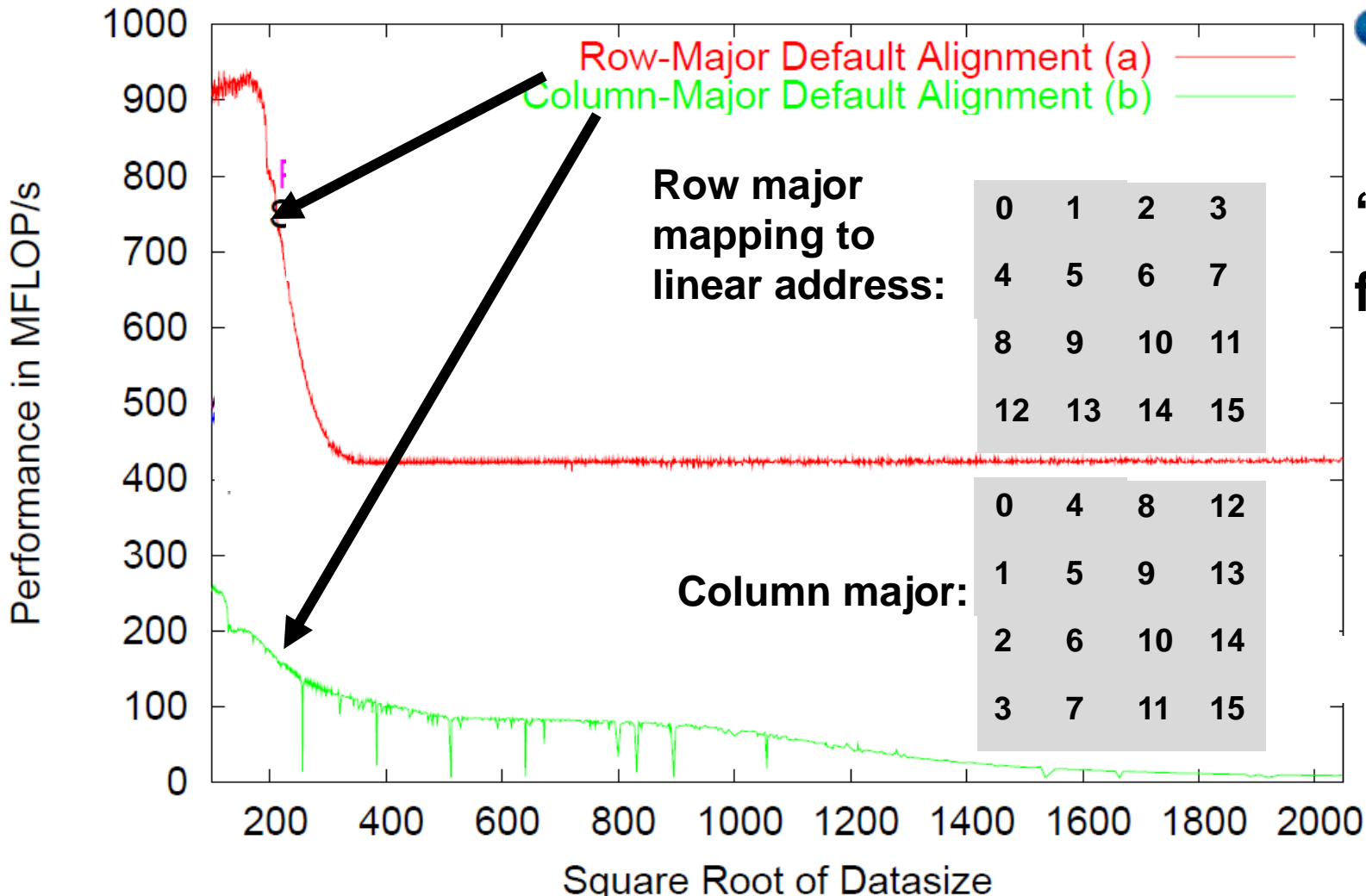




Problem size: 192 doubles, 1536 bytes per row

Permuting multidimensional arrays to improve spatial locality¹

MMikj on P4: Performance in MFLOP/s



● Matrix-matrix multiply on Pentium 4

“ikj” variant:

for i

for k

for j

$C[ij] += A[ik]$

$*B[kj]$

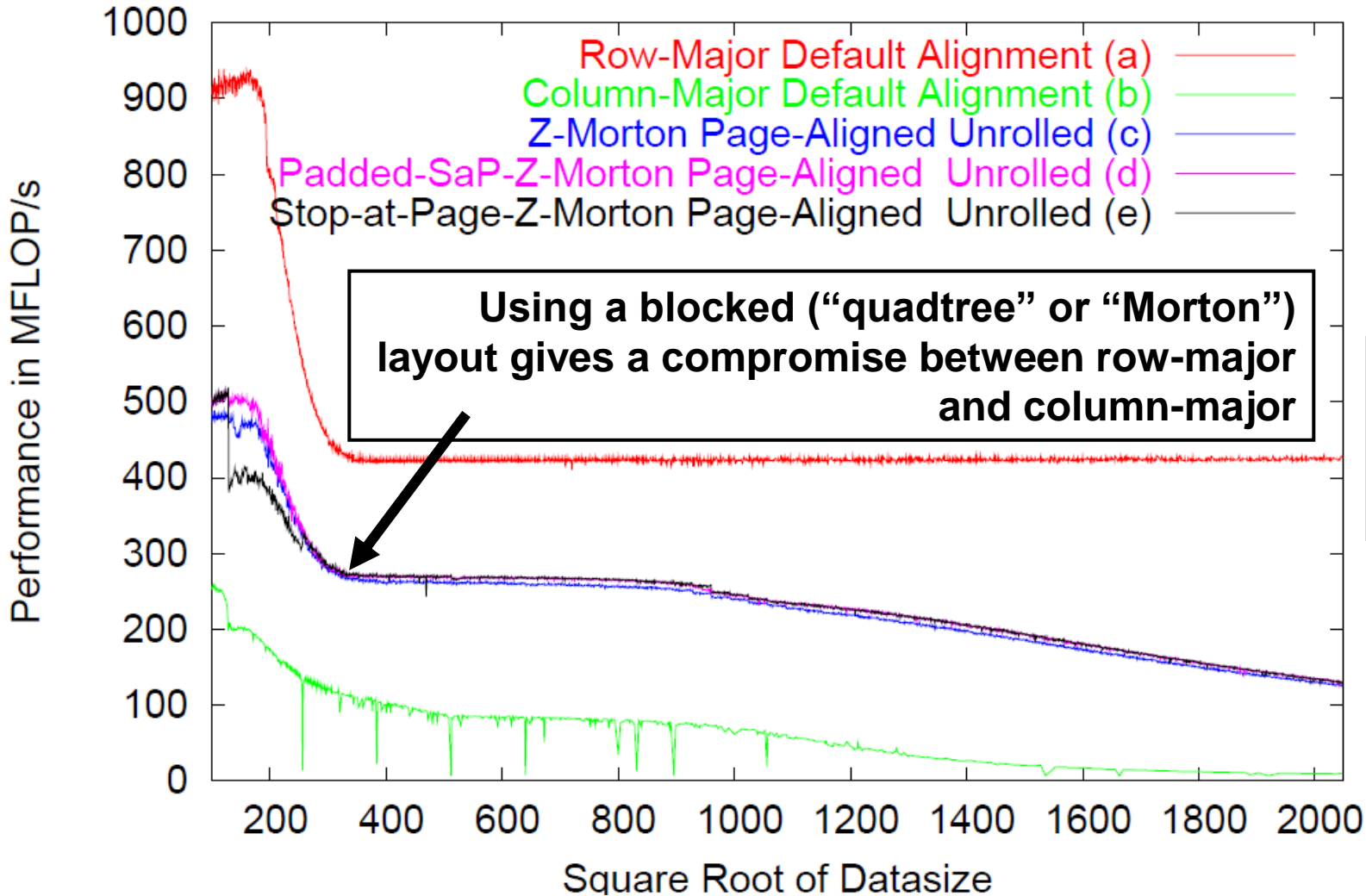
● Traverses B and C in row-major order

● Which is great if the data is stored in row-major order

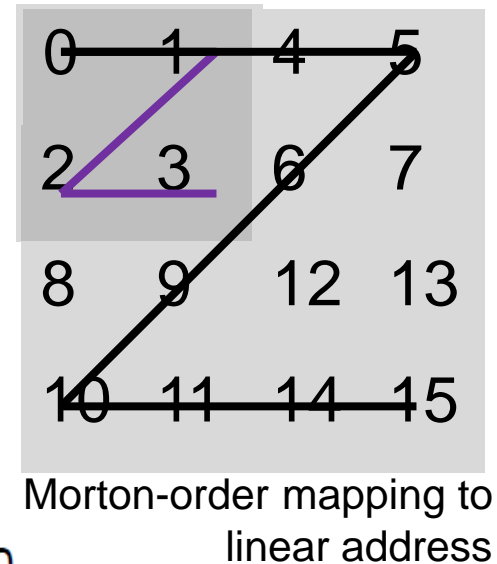
● If data is actually in column-major order...

Permuting multidimensional arrays to improve spatial locality

MMikj on P4: Performance in MFLOP/s



A variant of Morton-order layout is used for texture caching in some GPUs



- Blocked layout offers compromise between row-major and column-major
- Some care is needed in optimising address calculation to make this work (Jeyan Thiyaalingam's Imperial PhD thesis)


Loop Interchange: example

/* Before */

```
for (k = 0; k < 100; k = k+1)
  for (j = 0; j < 100; j = j+1)
    for (i = 0; i < 5000; i = i+1)
      x[i][j] = 2 * x[i][j];
```

/* After */

```
for (k = 0; k < 100; k = k+1)
  for (i = 0; i < 5000; i = i+1)
    for (j = 0; j < 100; j = j+1)
      x[i][j] = 2 * x[i][j];
```



Sequential accesses: instead of striding through memory every 100 words; improved spatial locality

Loop Fusion: example

```

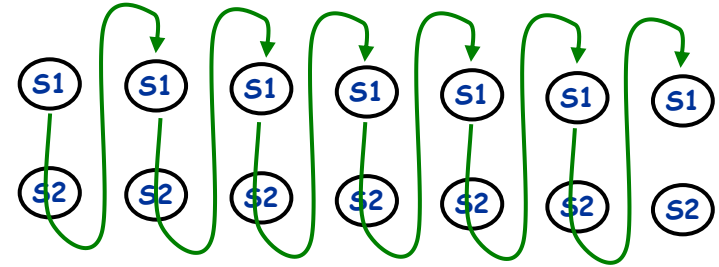
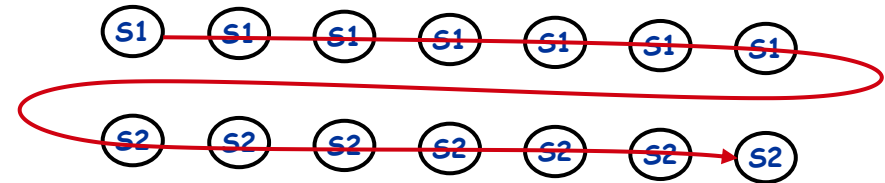
/* Before */
for (i = 0; i < N; i = i+1)
  for (j = 0; j < N; j = j+1)
    S1: a[i][j] = 1/b[i][j] * c[i][j];
for (i = 0; i < N; i = i+1)
  for (j = 0; j < N; j = j+1)
    S2: d[i][j] = a[i][j] + c[i][j];

```

```

/* After fusion */
for (i = 0; i < N; i = i+1)
  for (j = 0; j < N; j = j+1)
    { S1: a[i][j] = 1/b[i][j] * c[i][j];
      S2: d[i][j] = a[i][j] + c[i][j]; }

```



```

/* After array contraction */
for (i = 0; i < N; i = i+1)
  for (j = 0; j < N; j = j+1)
    { cv = c[i][j];
      S1: a = 1/b[i][j] * cv;
      S2: d[i][j] = a + cv; }

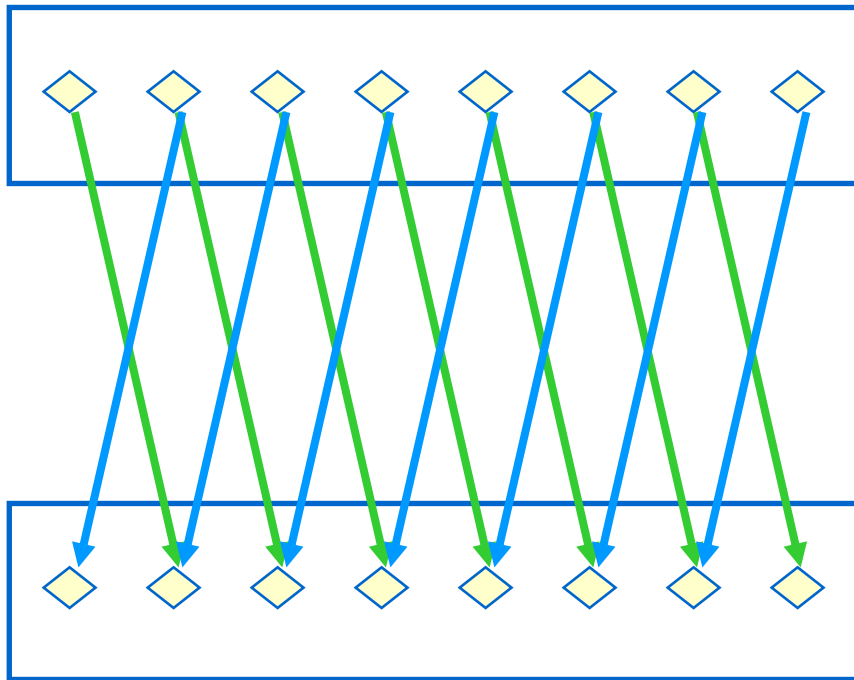
```

2 misses per access to a & c vs.
one miss per access; improve
spatial locality

The real payoff comes if
fusion enables **Array
Contraction**: values
transferred in scalar
instead of via array

Fusion is not always so simple

- Dependences might not align nicely
- Example: one-dimensional convolution filters



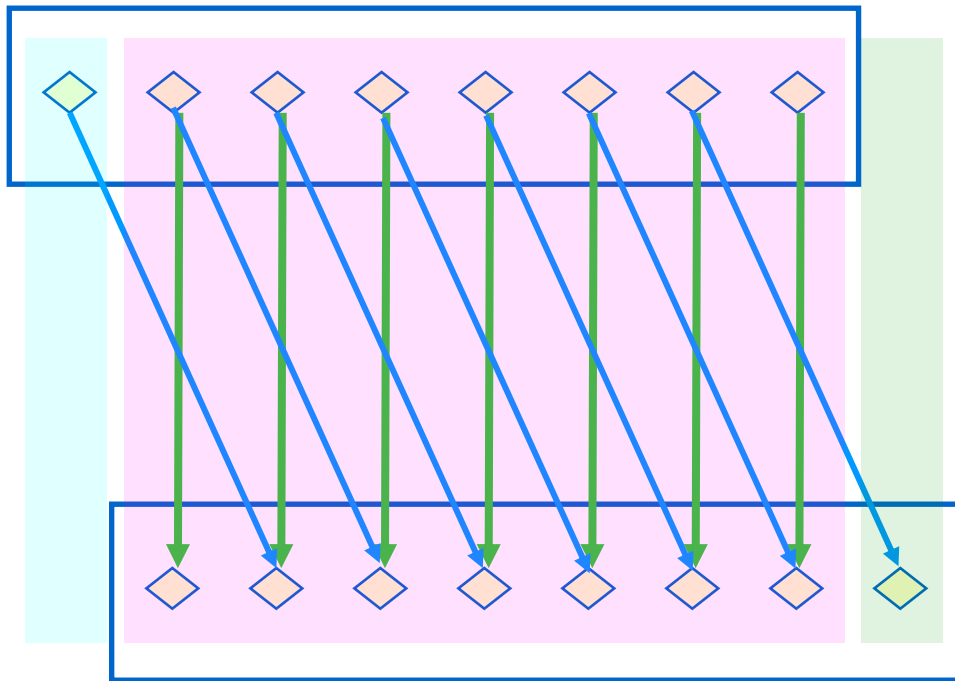
```
for (i=1; i<N; i++)  
    V[i] = (U[i-1] + U[i+1])/2
```

```
for (i=1; i<N; i++)  
    W[i] = (V[i-1] + V[i+1])/2
```

- “Stencil” loops are not directly fusable

Loop fusion – code expansion

- We make them fusable by shifting:



$$V[1] = (U[0] + U[2])/2$$

```
for (i=2; i<N; i++) {
```

$$V[i] = (U[i-1] + U[i+1])/2$$

$$W[i-1] = (V[i-2] + V[i])/2$$

```
}
```

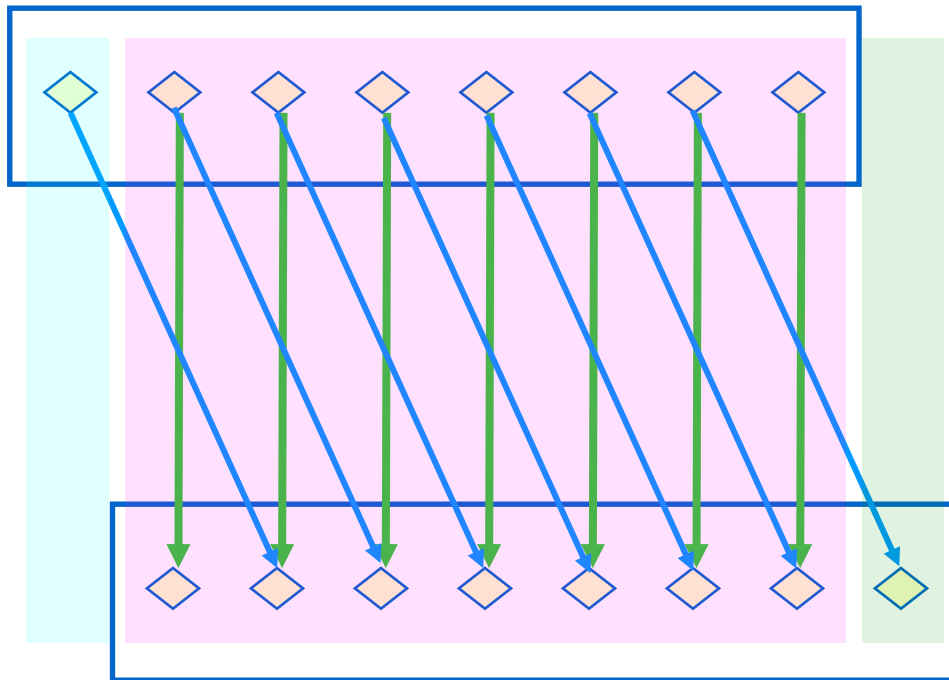
$$W[N-1] = (V[N-2] + V[N])/2$$

- The middle loop is fusable
- We get lots of little edge bits

Loop fusion – code expansion

This transformation is important in image-processing filters, finite difference solvers, and convolutional neural networks

- We make them fusable by shifting:



$$V[1] = (U[0] + U[2])/2$$

```
for (i=2; i<N; i++) {  
    V[i%4] = (U[i-1] + U[i+1])/2  
    W[i-1] = (V[(i-2)%4] + V[i%4])/2  
}
```

$$W[N-1] = (V[(N-2)%4] + V[N%4])/2$$

- The middle loop is fusable
- We get lots of little edge bits

- Contraction is trickier
- We need the last *two* Vs
- We need 3 V locations
- Quicker to round up to four

Summary

We can reduce the miss rate at the software level

- By using prefetch instructions
 - If they work better than predictive prefetch hardware
- By transforming storage layout
 - Might help with *spatial* locality
 - Might help with associativity conflicts
 - Can't help with *temporal* locality
- Storage layout optimisations are disruptive – they affect all the code that might use that data
- Loop interchange, fusion, tiling
 - Can get *really* messy to implement by hand
 - Can lead to a large space of possible schedules – it can be hard to know what will work best
 - Loop fusion can be very powerful but often breaks abstraction boundaries

Further reading

Algorithms and locality: cache-oblivious algorithms:

- https://en.wikipedia.org/wiki/Cache-oblivious_algorithm

Compilers that optimise for locality:

- Michael E. Wolf and Monica S. Lam. 1991. **A data locality optimizing algorithm.** PLDI91.
- Kathryn S. McKinley, Steve Carr, and Chau-Wen Tseng. 1996. **Improving data locality with loop transformations.** ACM Trans. Program. Lang. Syst. 18, 4 (July 1996)
- Uday Bondhugula, Albert Hartono, J. Ramanujam, and P. Sadayappan. 2008. **A practical automatic polyhedral parallelizer and locality optimizer.** PLDI08

Programming Abstractions for Data Locality

- <https://sites.google.com/a/lbl.gov/padal-workshop/>

Optimisations for convolutional neural networks

- Yizhi Liu, Yao Wang, Ruofei Yu, Mu Li, Vin Sharma, Yida Wang. **Optimizing CNN model inference on CPUs.** USENIX ATC'19.