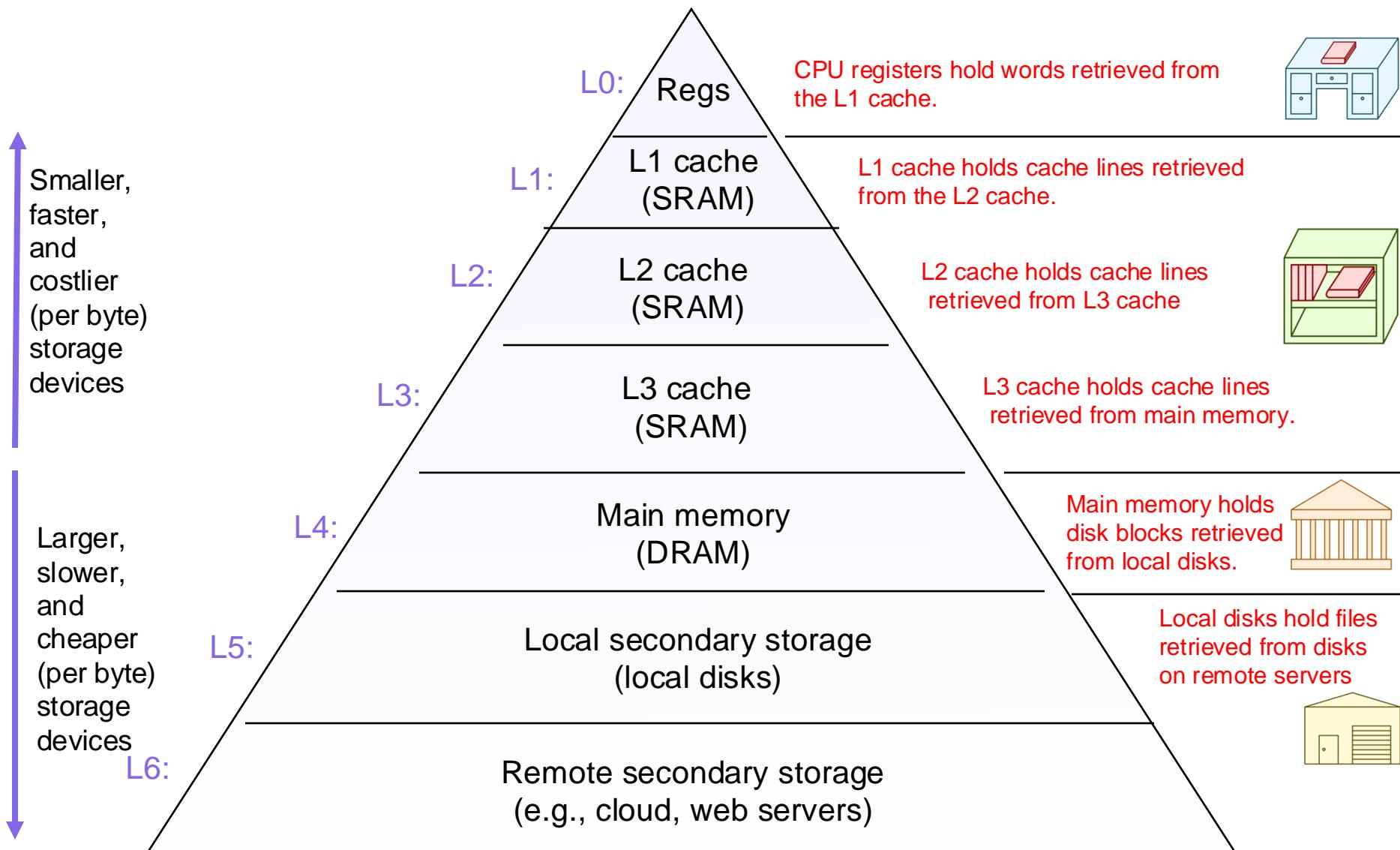


Lecture 13: Optimization with Caches

CS 105

Fall 2024

Review: Memory Hierarchy

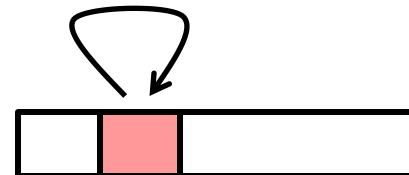


Review: Principle of Locality

Programs tend to use data and instructions with addresses near or equal to those they have used recently

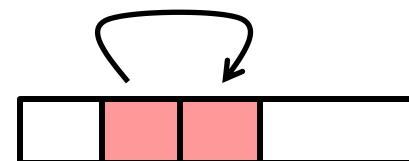
- **Temporal locality:**

- Recently referenced items are likely to be referenced again in the near future



- **Spatial locality:**

- Items with nearby addresses tend to be referenced close together in time

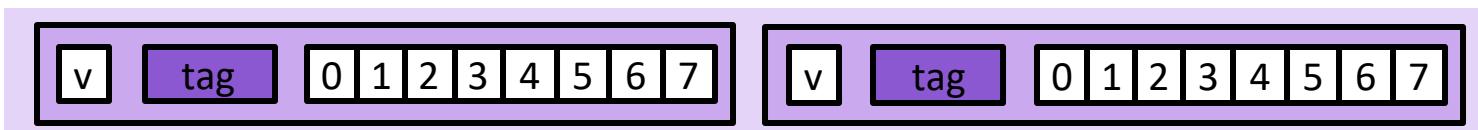
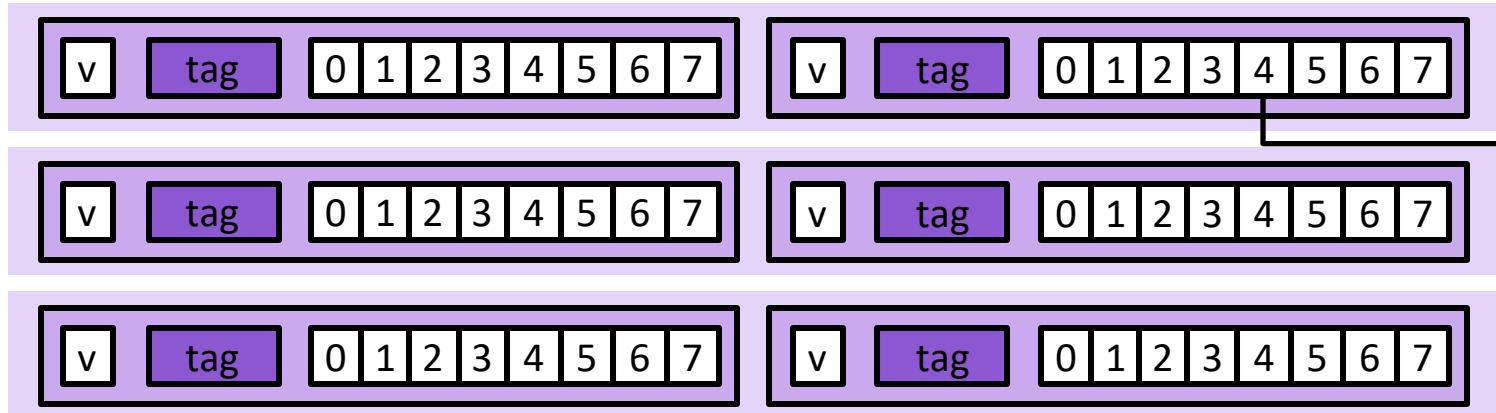


Review: An Example Cache

$E = 2$: Two lines per set

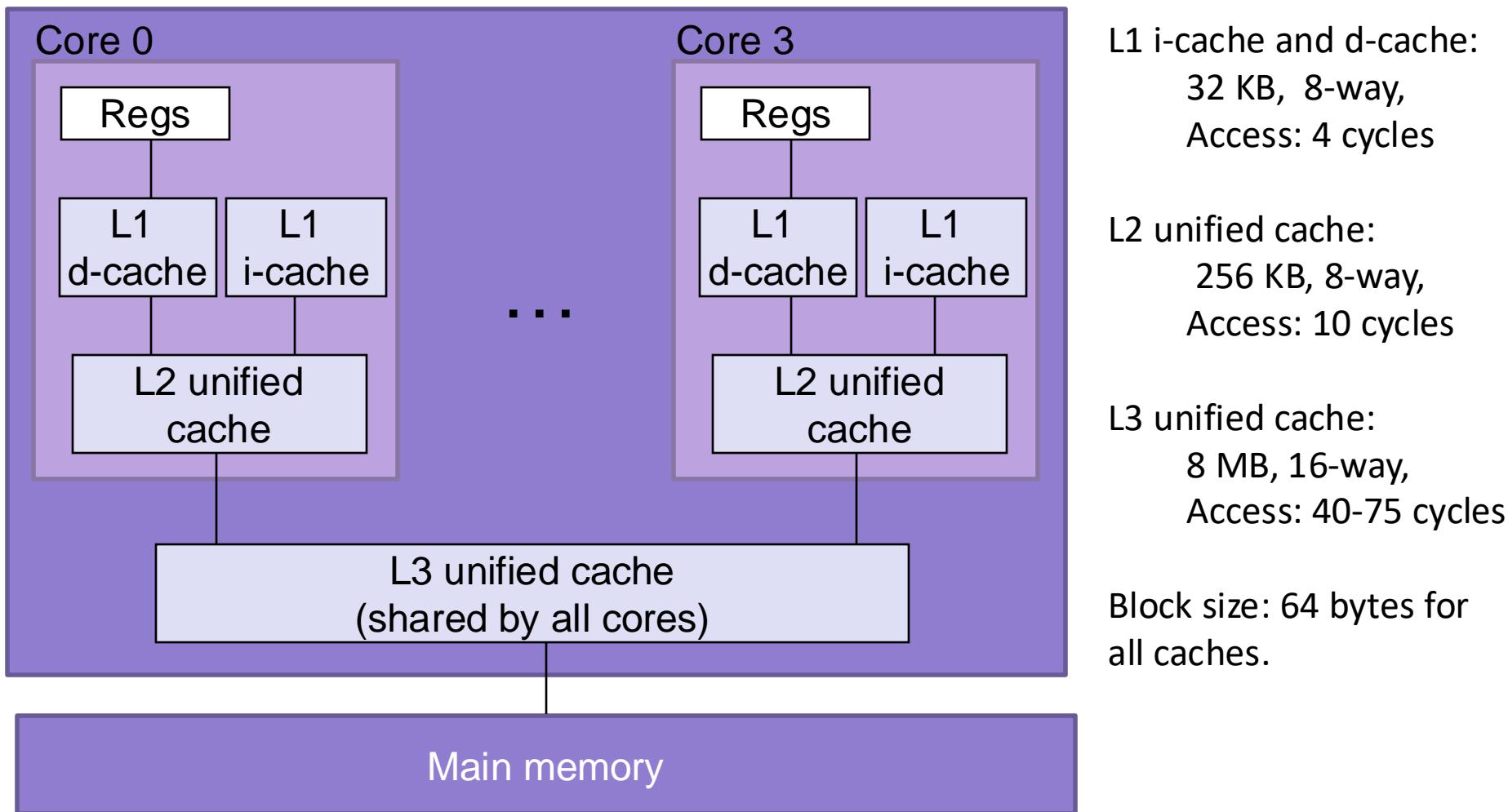
Assume: cache block size 8 bytes

Address of data:



Typical Intel Core i7 Hierarchy

Processor package



Cache Performance Metrics

- Miss Rate
 - Fraction of memory references not found in cache (misses / accesses)
 - Typically 3-10% for L1
 - can be quite small (e.g., < 1%) for L2, depending on size, etc.
- Hit Time
 - Time to deliver a line in the cache to the processor
 - includes time to determine whether the line is in the cache
 - Typically 4 clock cycles for L1, 10 clock cycles for L2
- Miss Penalty
 - Additional time required because of a miss
 - typically 50-200 cycles for main memory (Trend: increasing!)

Memory Performance with Caching

- **Read throughput (aka read bandwidth):** Number of bytes read from memory per second (MB/s)
- **Memory mountain:** Measured read throughput as a function of spatial and temporal locality.
 - Compact way to characterize memory system performance.

Memory Mountain Test Function

Call test() with many combinations of elems and stride.

For each elems and stride:

1. Call test() once to warm up the caches.

2. Call test() again and measure the read throughput (MB/s)

```
long data[MAXELEMS]; /* Global array to traverse */

/* test - Iterate over first "elems" elements of
 *      array "data" with stride of "stride", using
 *      using 4x4 loop unrolling.
 */
int test(int elems, int stride) {
    long i, sx2=stride*2, sx3=stride*3, sx4=stride*4;
    long acc0 = 0, acc1 = 0, acc2 = 0, acc3 = 0;
    long length = elems, limit = length - sx4;

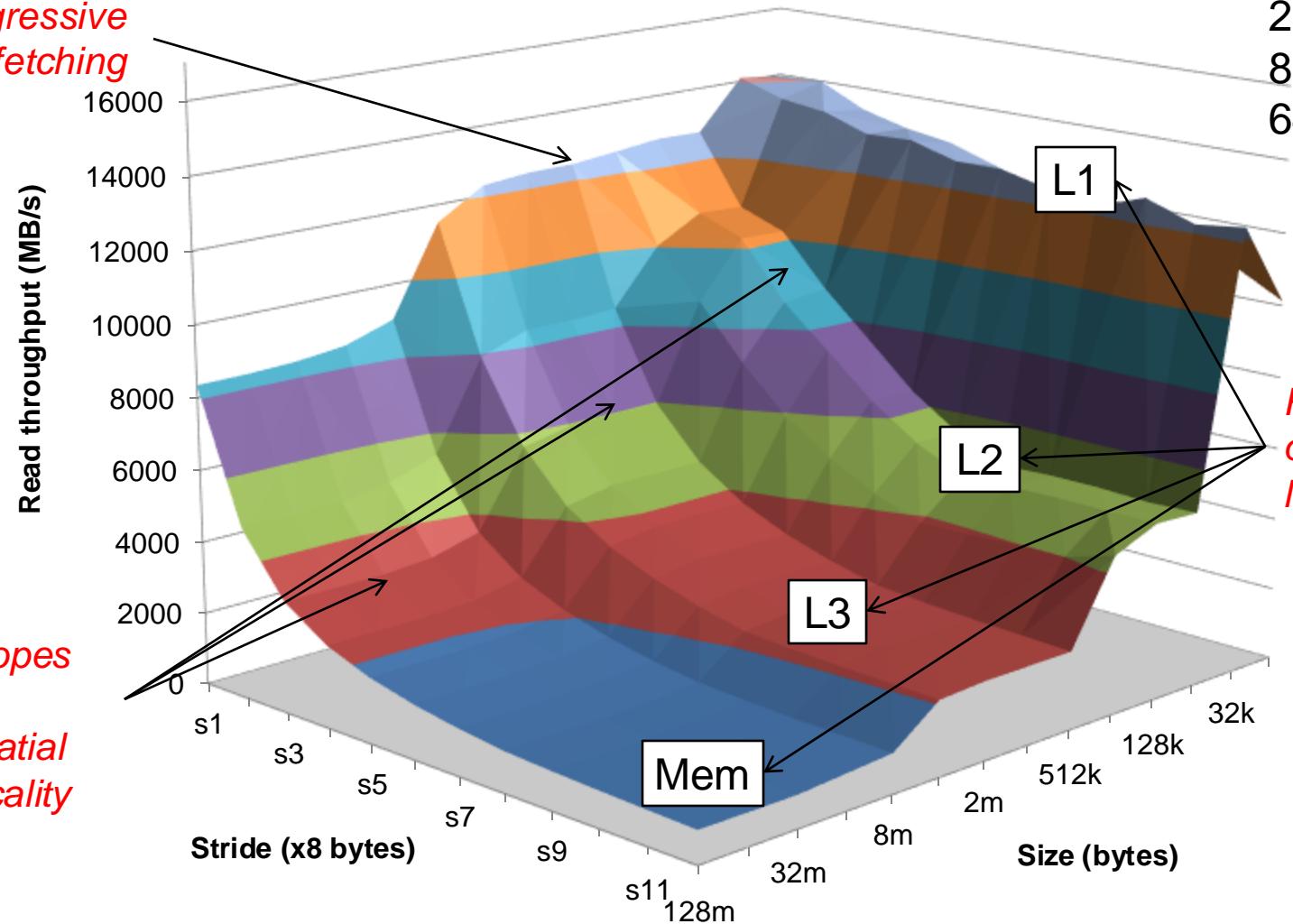
    /* Combine 4 elements at a time */
    for (i = 0; i < limit; i += sx4) {
        acc0 = acc0 + data[i];
        acc1 = acc1 + data[i+stride];
        acc2 = acc2 + data[i+sx2];
        acc3 = acc3 + data[i+sx3];
    }

    /* Finish any remaining elements */
    for (; i < length; i++) {
        acc0 = acc0 + data[i];
    }
    return ((acc0 + acc1) + (acc2 + acc3));
}
```

The Memory Mountain

Core i7 Haswell
2.1 GHz
32 KB L1 d-cache
256 KB L2 cache
8 MB L3 cache
64 B block size

*Aggressive
prefetching*



*Slopes
of
spatial
locality*

Exercise 1: Locality

- Which of the following functions is better in terms of locality with respect to array src?

```
void copyij(int src[2048][2048],  
           int dst[2048][2048])  
{  
    int i,j;  
    for (i = 0; i < 2048; i++)  
        for (j = 0; j < 2048; j++)  
            dst[i][j] = src[i][j];  
}
```

```
void copyji(int src[2048][2048],  
           int dst[2048][2048])  
{  
    int i,j;  
    for (j = 0; j < 2048; j++)  
        for (i = 0; i < 2048; i++)  
            dst[i][j] = src[i][j];  
}
```

Writing Cache-Friendly Code

- Make the common case go fast
 - Focus on the inner loops of the core functions
- Minimize the misses in the inner loops
 - Repeated references to variables are good (**temporal locality**)
 - Stride-1 reference patterns are good (**spatial locality**)

Exercise: Miss Rate Analysis

```
int sum_array(int* array, int n){  
    int sum = 0;  
  
    for(int i = 0; i < n; i++){  
  
        sum += array[i];  
  
    }  
  
    return sum;  
}
```

assume n, sum and i are stored in registers and only array is stored in memory, assume n=16

assume array = 0x600090

assume 256 byte direct-mapped cache
w/ 16-byte cache lines

Exercise: what is the sequence of memory accesses made by this program?

Exercise: what is the hit rate of this program?

Review: Matrix Multiplication

- $a = \begin{bmatrix} 2.0 & 1.5 \\ 1.0 & 0.5 \end{bmatrix}$
- $b = \begin{bmatrix} 1.0 & 2.0 \\ 3.0 & 4.0 \end{bmatrix}$
- What is $a \times b$?

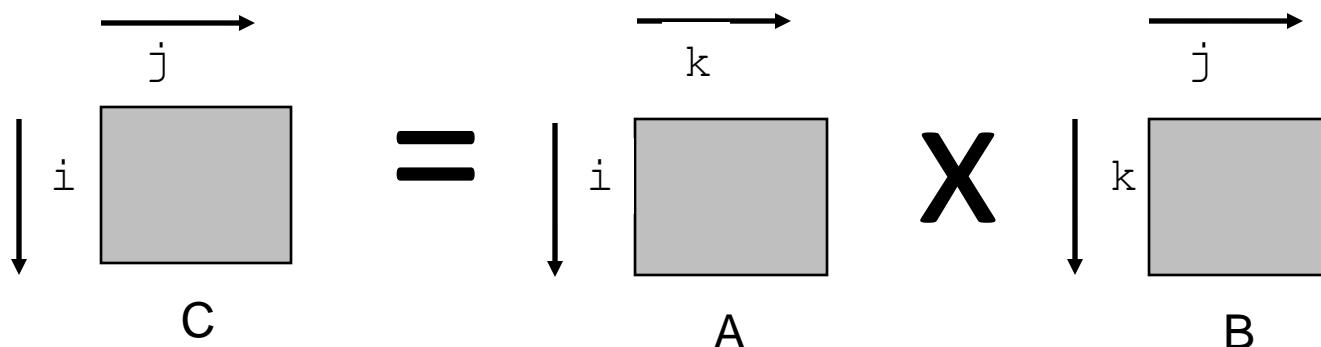
Example: Matrix Multiplication

- Multiply $N \times N$ matrices
- Matrix elements are doubles (8 bytes)
- $O(N^3)$ total operations
- N reads per source element
- N values summed per destination

```
/* ijk */
for(int i=0; i<n; i++) {
    for(int j=0; j<n; j++) {
        sum = 0.0;
        for(int k=0; k<n; k++) {
            sum += a[i][k] * b[k][j];
        }
        c[i][j] = sum;
    }
}
```

Miss Rate Analysis for Matrix Multiply

- Assume:
 - datablock size = 32 bytes (big enough for four doubles)
 - Matrix dimension (N) is very large
 - Approximate $1/N$ as 0.0
 - Cache is not even big enough to hold multiple rows
- Analysis Method:
 - Look at access pattern of inner loop



Review: Layout of C Arrays in Memory

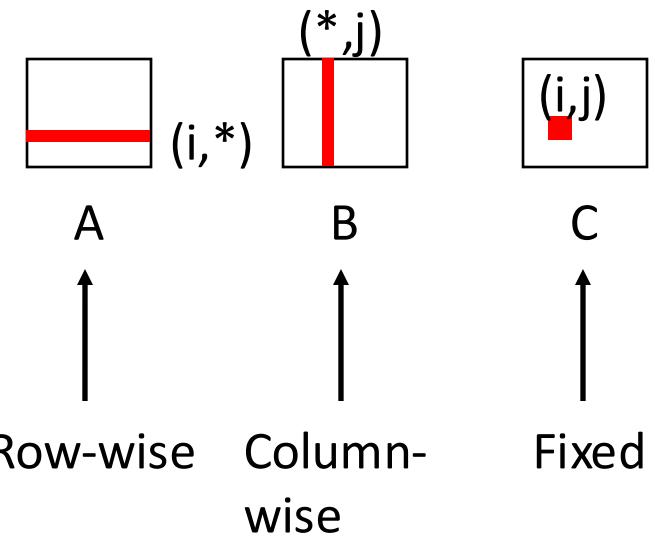
- C arrays allocated in row-major order
 - each row in contiguous memory locations
- Stepping through columns in one row:
 - accesses successive elements
 - if data block size (B) > $\text{sizeof}(a_{ij})$ bytes, exploit spatial locality
 - miss rate = $\text{sizeof}(a_{ij}) / B$
- Stepping through rows in one column:
 - accesses distant elements
 - no spatial locality!
 - miss rate = 1 (i.e. 100%)

Matrix Multiplication (ijk)

(jik is similar)

```
/* ijk */  
for (i=0; i<n; i++) {  
    for (j=0; j<n; j++) {  
        sum = 0.0;  
        for (k=0; k<n; k++)  
            sum += a[i][k] * b[k][j];  
        c[i][j] = sum;  
    }  
}
```

Inner loop:



Average Misses per inner loop iteration:

<u>A</u>	<u>B</u>	<u>C</u>	<u>Total</u>
.25	1.0	0.0	1.25

2 reads, 0 writes
per inner loop iteration

Exercise: Alternative Matrix Multiplication Algs

- $a = \begin{bmatrix} 2.0 & 1.5 \\ 1.0 & 0.5 \end{bmatrix}$
- $b = \begin{bmatrix} 1.0 & 2.0 \\ 3.0 & 4.0 \end{bmatrix}$
- Use this algorithm to compute $a \times b$

```
/* kij */  
for (k=0; k<n; k++) {  
    for (i=0; i<n; i++) {  
        r = a[i][k];  
        for (j=0; j<n; j++)  
            c[i][j] += r * b[k][j];  
    }  
}
```

```
/* jki */  
for (j=0; j<n; j++) {  
    for (k=0; k<n; k++) {  
        r = b[k][j];  
        for (i=0; i<n; i++)  
            c[i][j] += a[i][k] * r;  
    }  
}
```

```

/* kij */
for (k=0; k<n; k++) {
    for (i=0; i<n; i++) {
        r = a[i][k];
        for (j=0; j<n; j++)
            c[i][j] += r * b[k][j];
    }
}

```

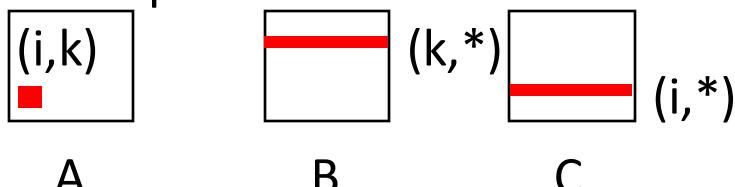
```

/* jki */
for (j=0; j<n; j++) {
    for (k=0; k<n; k++) {
        r = b[k][j];
        for (i=0; i<n; i++)
            c[i][j] += a[i][k] * r;
    }
}

```

2 reads, 1 write per inner loop iteration

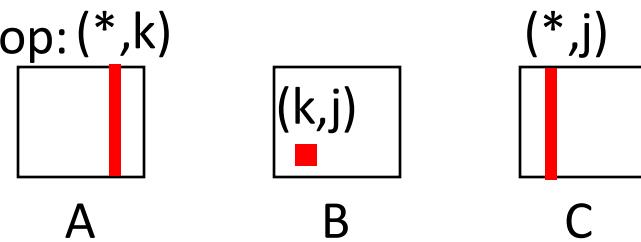
Inner loop:



Misses per inner loop iteration:

A	B	C
0.0	0.25	0.25

Inner loop: (*,k)



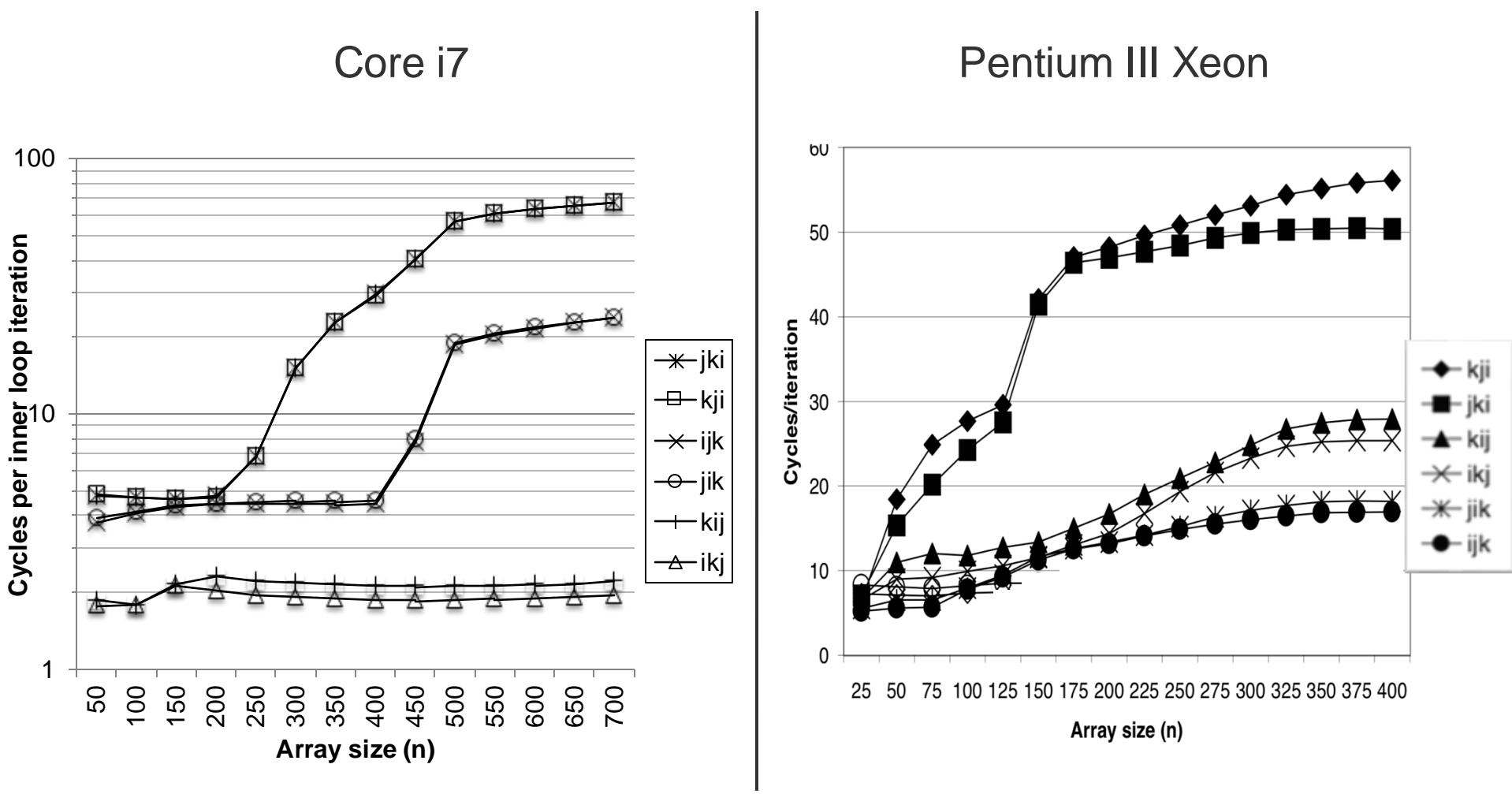
Misses per inner loop iteration:

A	B	C
1.0	0.0	1.0

Summary of Matrix Multiplication

```
for (i=0; i<n; i++) {  
    for (j=0; j<n; j++) {  
        sum = 0.0;  
        for (k=0; k<n; k++)  
            sum += a[i][k] * b[k][j];  
        c[i][j] = sum;  
    }  
}  
  
for (k=0; k<n; k++) {  
    for (i=0; i<n; i++) {  
        r = a[i][k];  
        for (j=0; j<n; j++)  
            c[i][j] += r * b[k][j];  
    }  
}  
  
for (j=0; j<n; j++) {  
    for (k=0; k<n; k++) {  
        r = b[k][j];  
        for (i=0; i<n; i++)  
            c[i][j] += a[i][k] * r;  
    }  
}
```

Matrix Multiply Performance

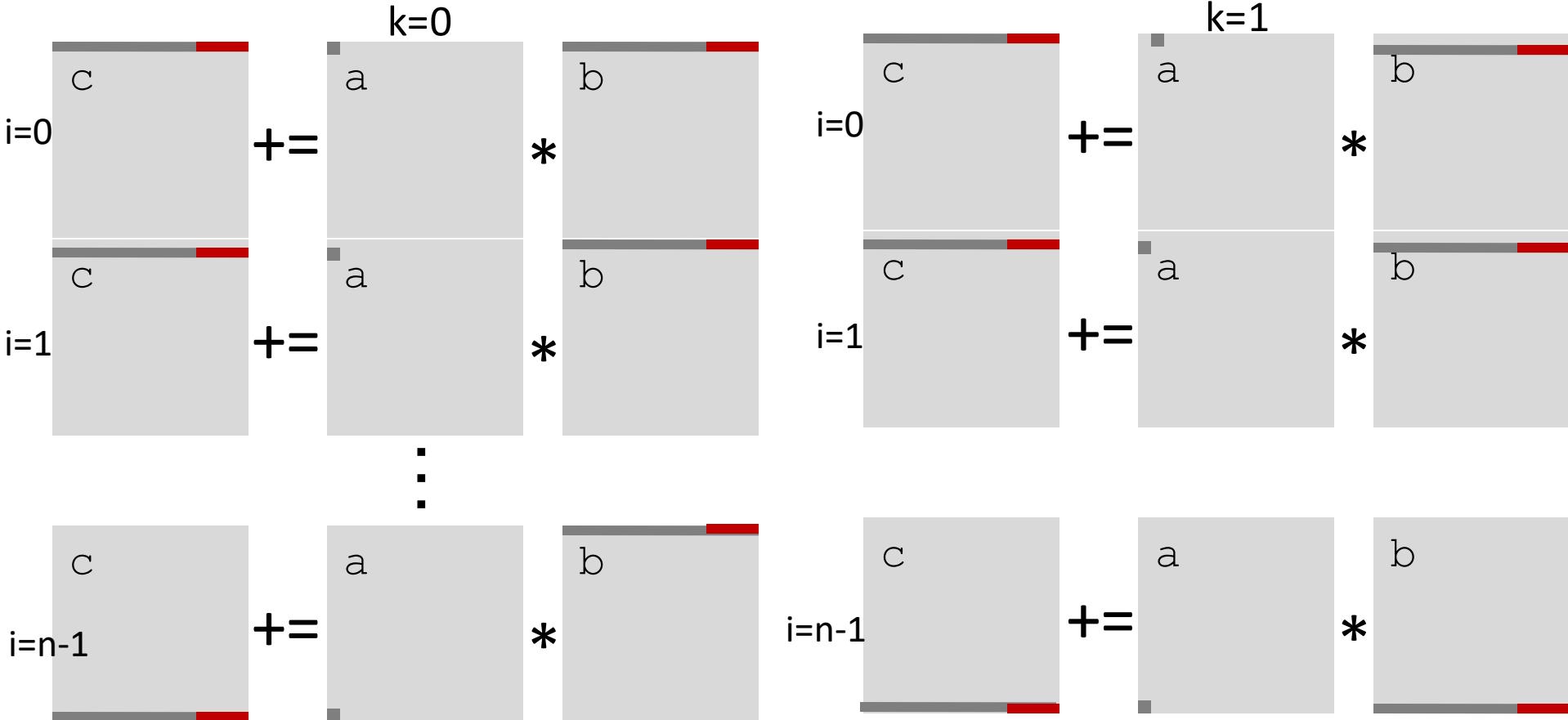


Can we do better?

Assume:

- Matrix elements are doubles
- Cache data block = 4 doubles
- Cache size C << n

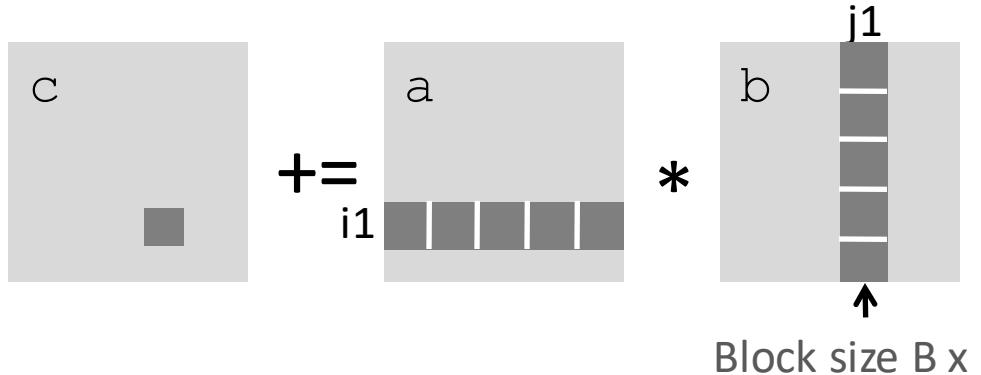
```
/* kij */  
for (k=0; k<n; k++) {  
    for (i=0; i<n; i++) {  
        r = a[i][k];  
        for (j=0; j<n; j++)  
            c[i][j] += r * b[k][j];  
    }  
}
```



Blocked Matrix Multiplication

```
c = (double *) calloc(sizeof(double), n*n);

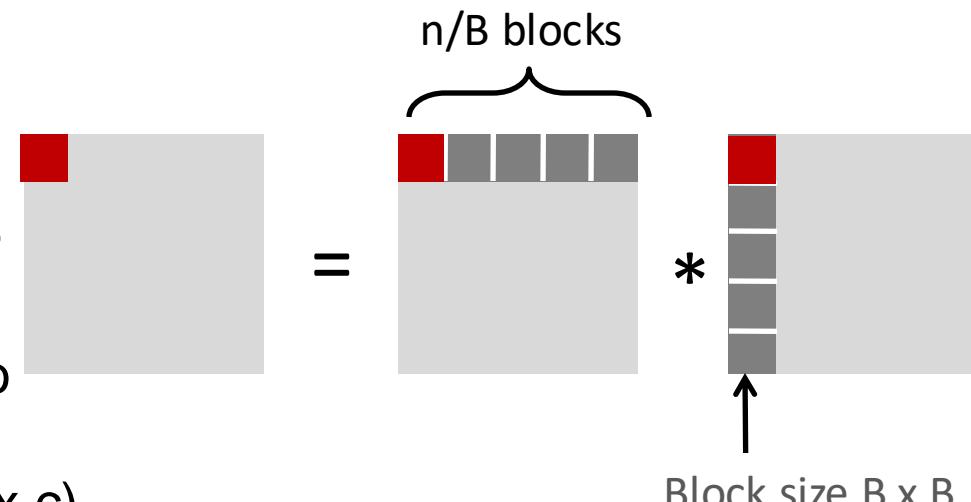
/* Multiply n x n matrices a and b */
void mmm(double* a, double* b, double* c, int n) {
    int i, j, k;
    for (i = 0; i < n; i+=B)
        for (j = 0; j < n; j+=B)
            for (k = 0; k < n; k+=B)
                /* B x B mini matrix multiplications */
                for (i1 = i; i1 < i+B; i++)
                    for (j1 = j; j1 < j+B; j++)
                        for (k1 = k; k1 < k+B; k++)
                            c[i1*n+j1] += a[i1*n + k1]*b[k1*n + j1];
}
```



Cache Miss Analysis

- Assume:
 - Cache data block = 4 doubles
 - Cache size $C \ll n$ (much smaller than n)
 - Three blocks ■ fit into cache: $3B^2 < C$

- First (block) iteration:
 - B^2 elements in each block, so $B^2/4$ misses for each block
 - n/B blocks in each row/col, so $2 * n/B * B^2/4 = nB/2$ misses in first iteration(omitting matrix c)



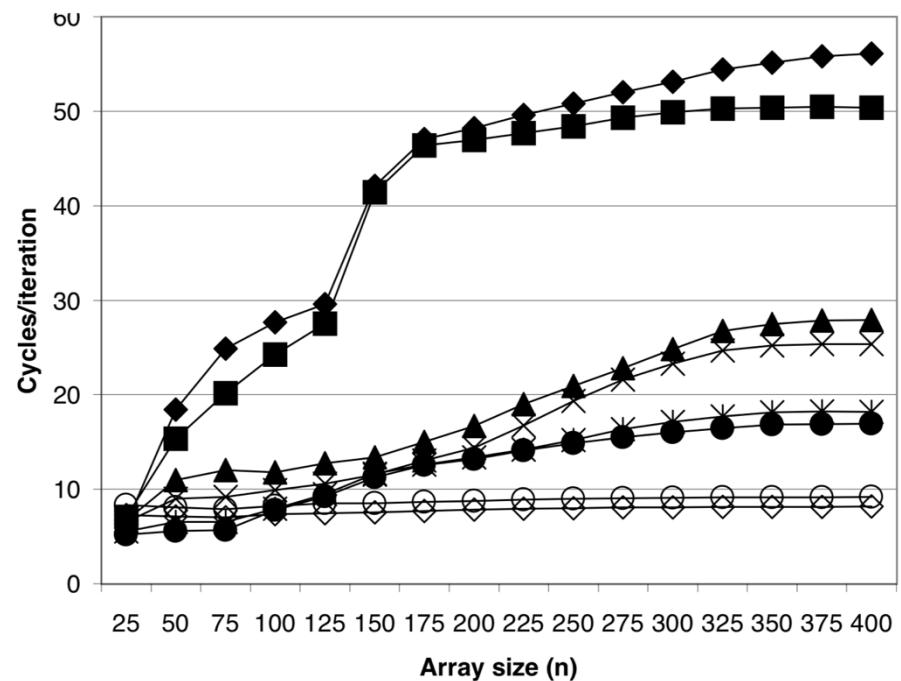
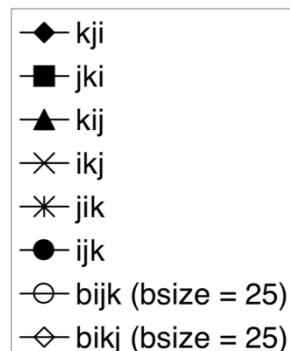
- Total misses:
 - $nB/2 * (n/B)^2 = n^3/(2B)$

Blocking Summary

- No blocking: $n^3 / 2$
- Blocking: $n^3 / (2B)$
- Suggest largest possible block size B, but limit $3B^2 < C!$
- Reason for dramatic difference:
 - Matrix multiplication has inherent temporal locality:
 - Input data: $3n^2$, computation $2n^3$
 - Every array elements used $O(n)$ times!
 - But program has to be written properly

A reality check

- This analysis only holds on some machines!
- Intel Core i7 does aggressive pre-fetching for one-stride programs, so blocking doesn't actually improve performance
- But on a Pentium III Xeon:



And that's the end of Part 1

