Introduction to Computer Systems

15-213, fall 2009

11th Lecture, Oct. 5th

Instructors:

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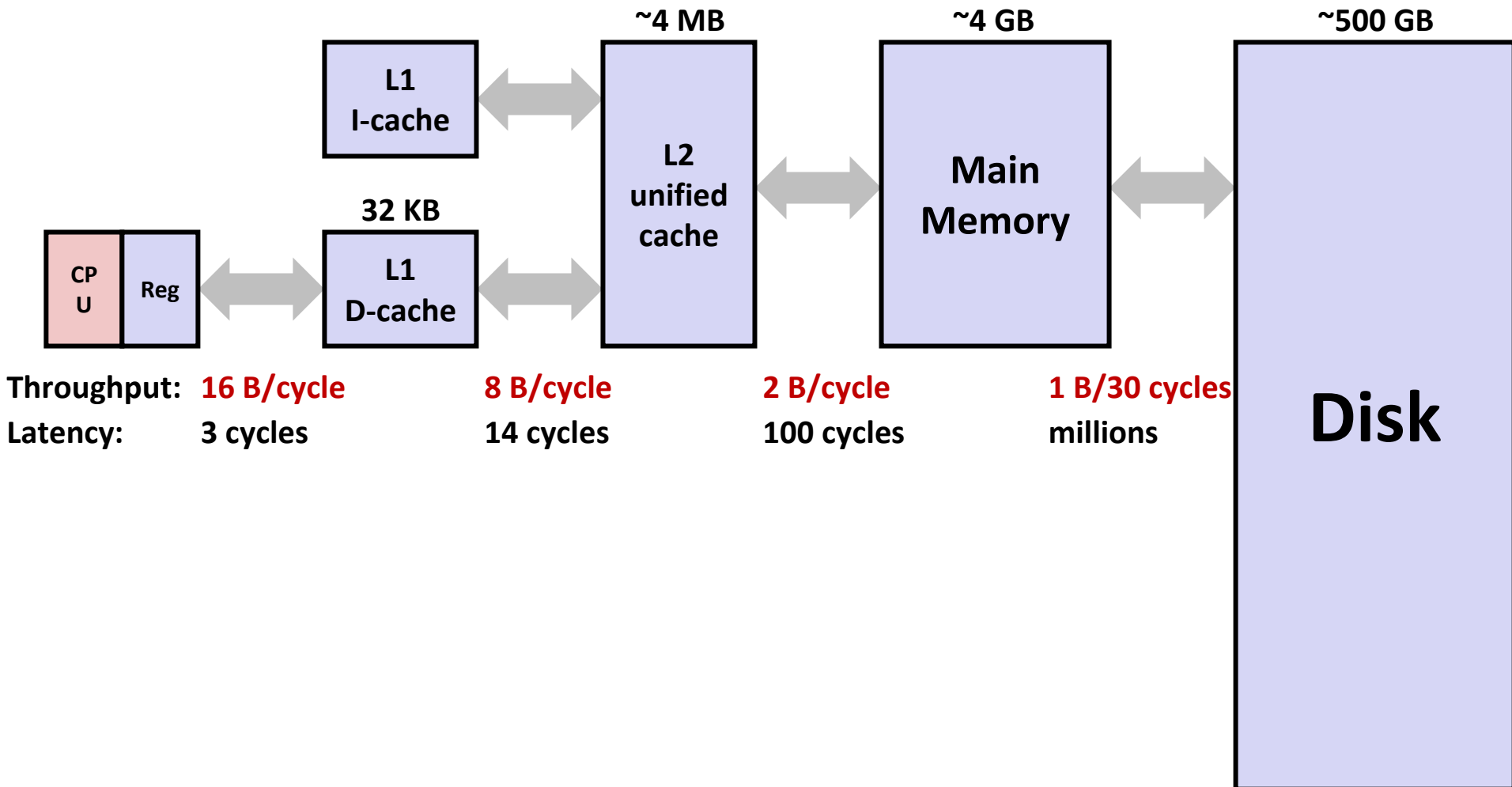
Last Time

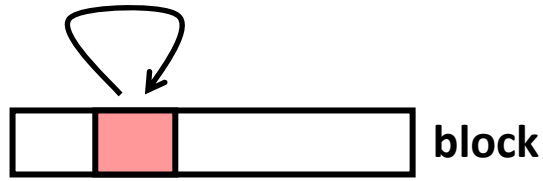
- Memory hierarchy (Here: Core 2 Duo)



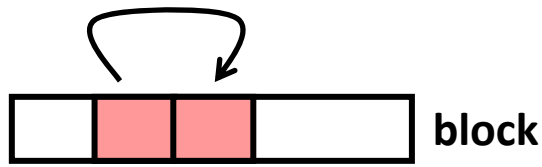
Last Time

- **Locality**

- **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

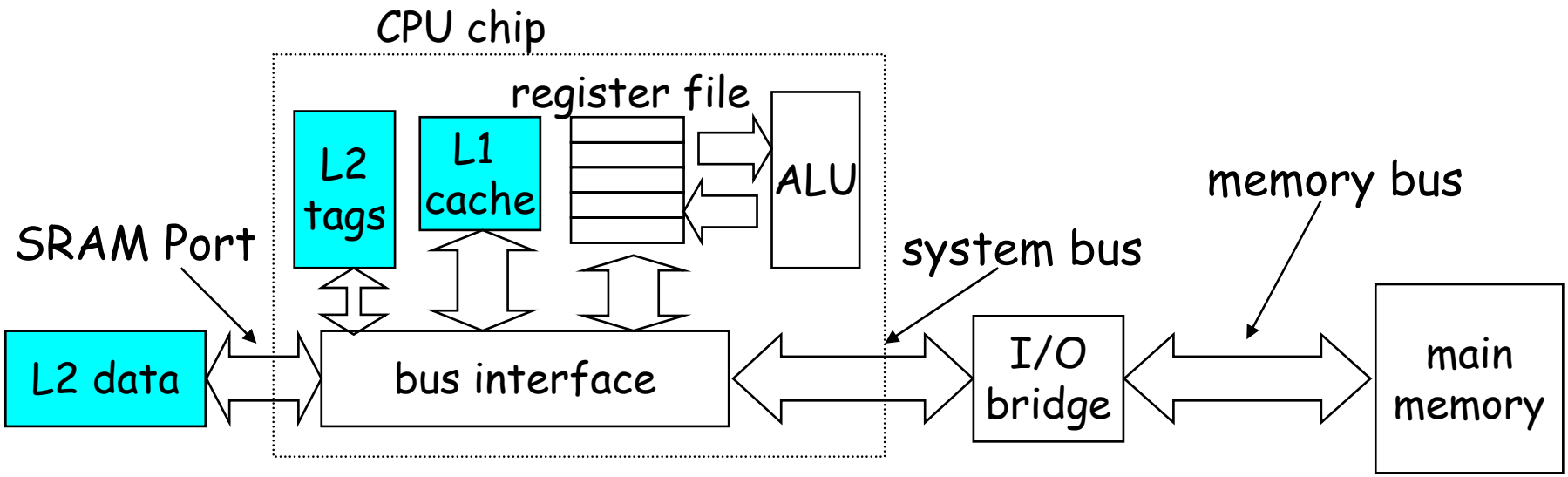


Today

- **Cache organization**
- **Memory Wall**
- **Program optimization (Matrix Multiplication):**
  - Cache optimizations
  - Cache Miss Analysis

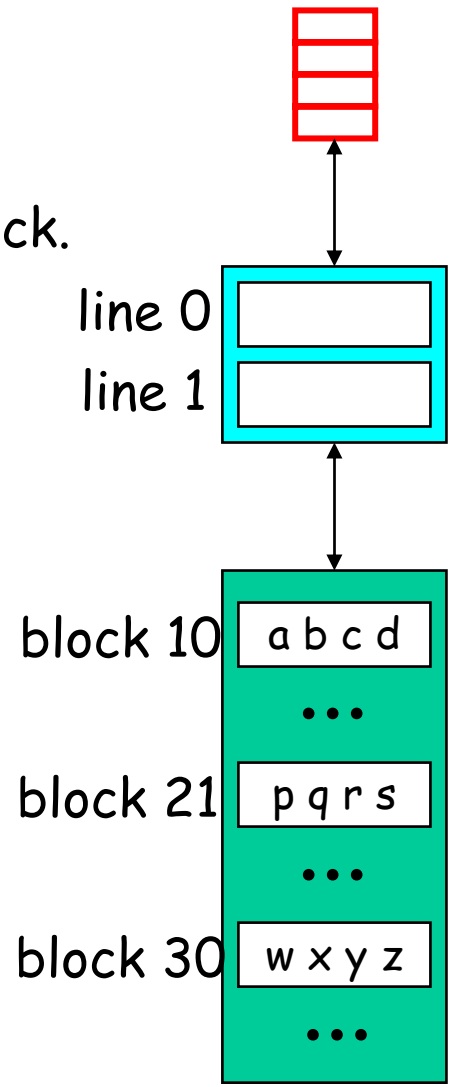
Cache Memories

- **Cache memories are small, fast SRAM-based memories managed automatically in hardware.**
  - Hold frequently accessed blocks of main memory
- **CPU looks first for data in L1, then in L2, then in main memory.**
- **Typical system structure:**



Inserting an L1 Cache Between the CPU and Main Memory

The transfer unit between the CPU register file and the cache is a 4-byte block.



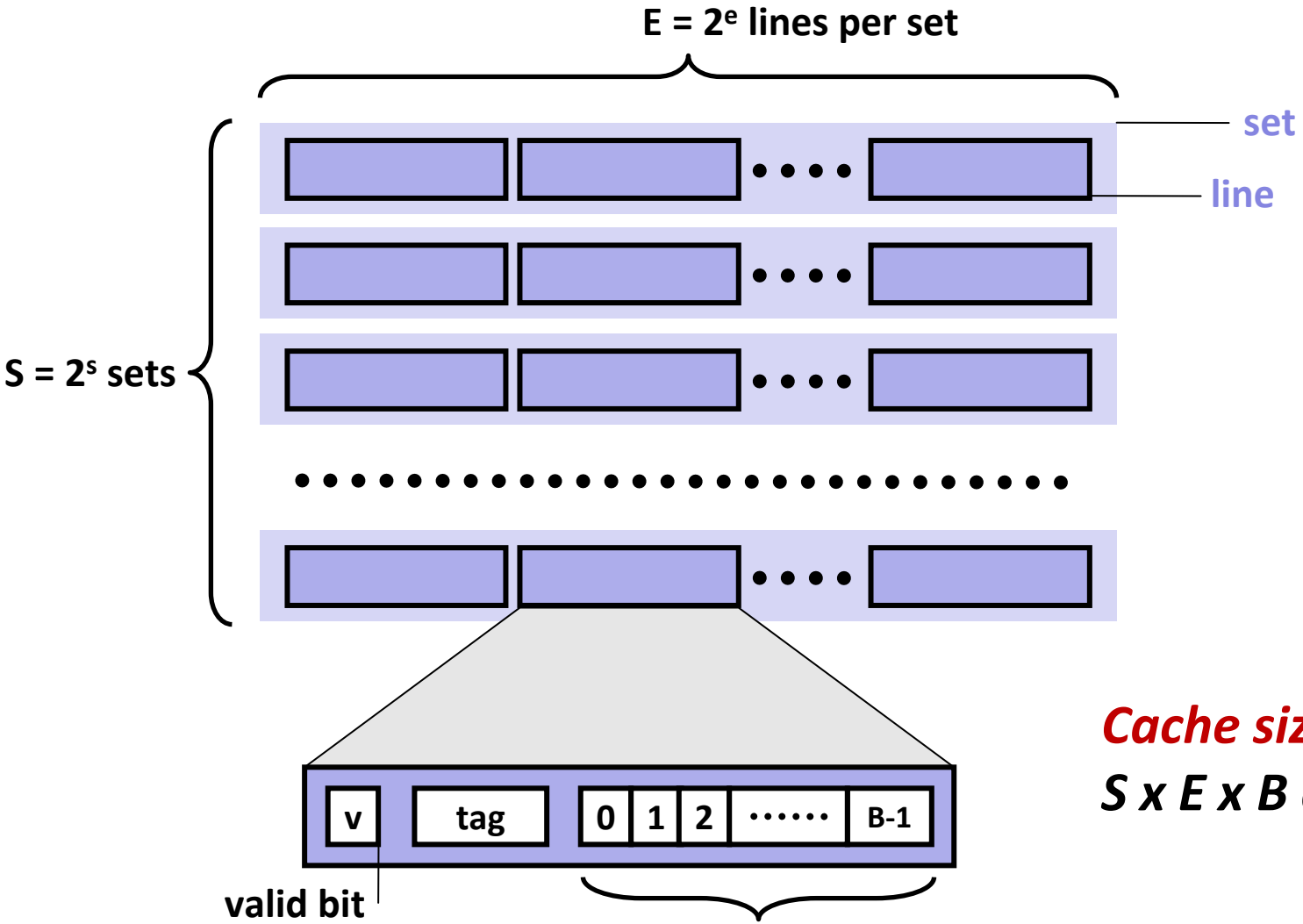
The tiny, very fast CPU register file has room for four 4-byte words.

The small fast L1 cache has room for two 4-word blocks.

The transfer unit between the cache and main memory is a 4-word block (16 bytes).

The big slow main memory has room for many 4-word blocks.

General Cache Organization (S, E, B)

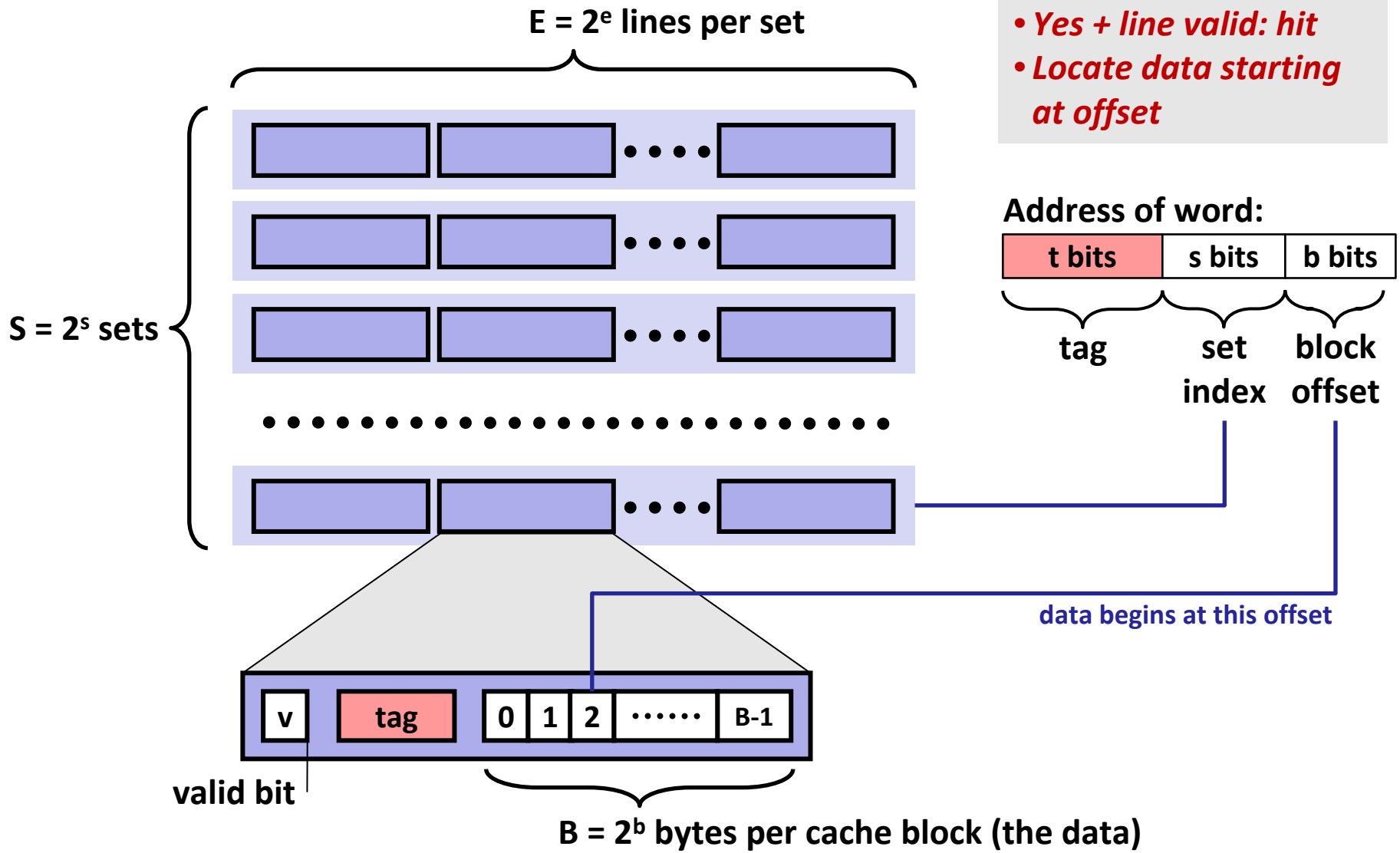


Cache size:

S x E x B data bytes

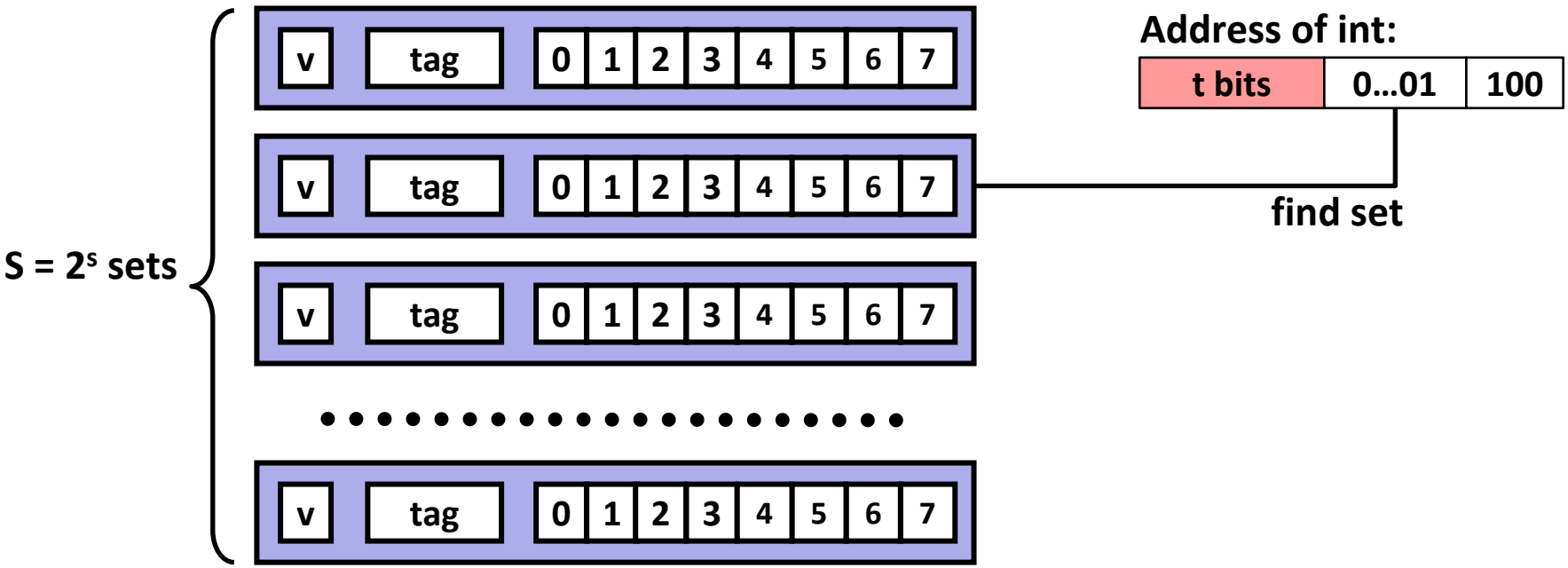
Cache Read

- *Locate set*
- *Check if any line in set has matching tag*
- *Yes + line valid: hit*
- *Locate data starting at offset*



Example: Direct Mapped Cache (E = 1)

Direct mapped: One line per set
Assume: cache block size 8 bytes



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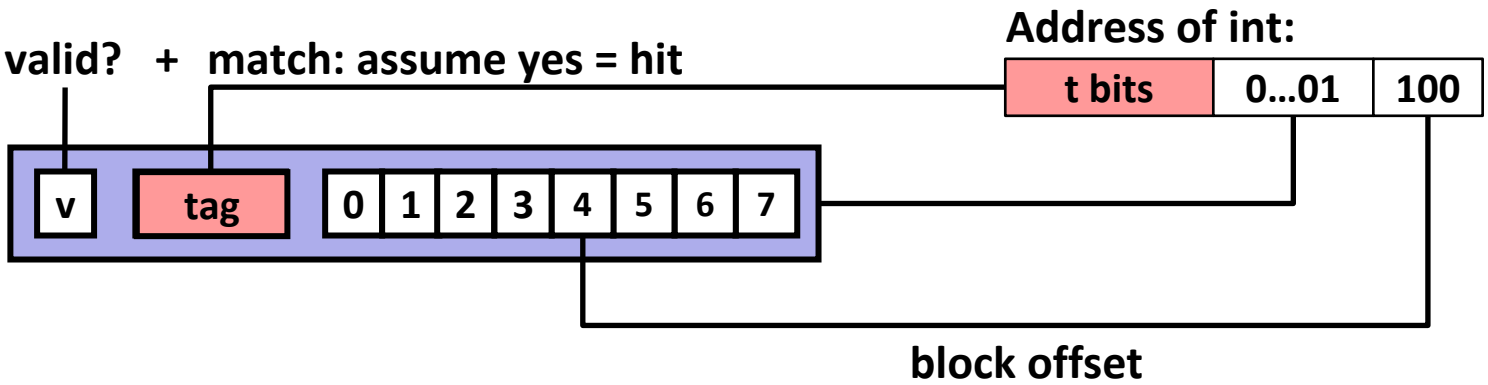

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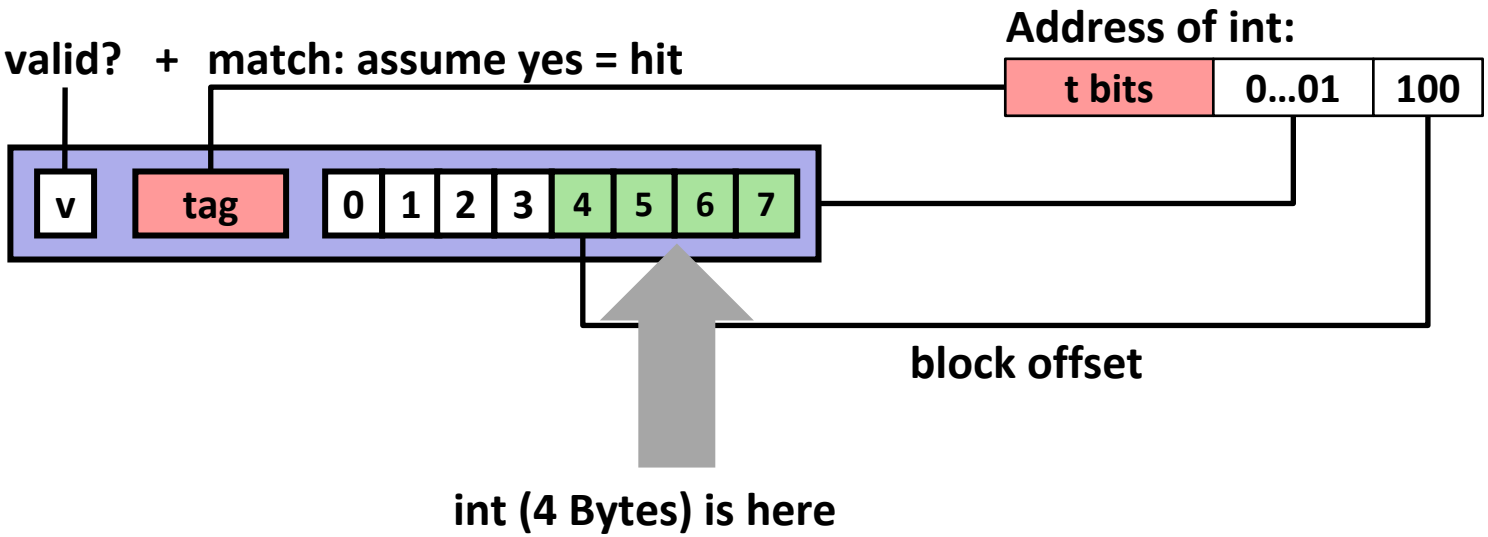


No match: old line is evicted and replaced

Direct-Mapped Cache Simulation

t=1	s=2	b=1
x	xx	x

M=16 byte addresses, B=2 bytes/block,
S=4 sets, E=1 entry/set

Address trace (reads):

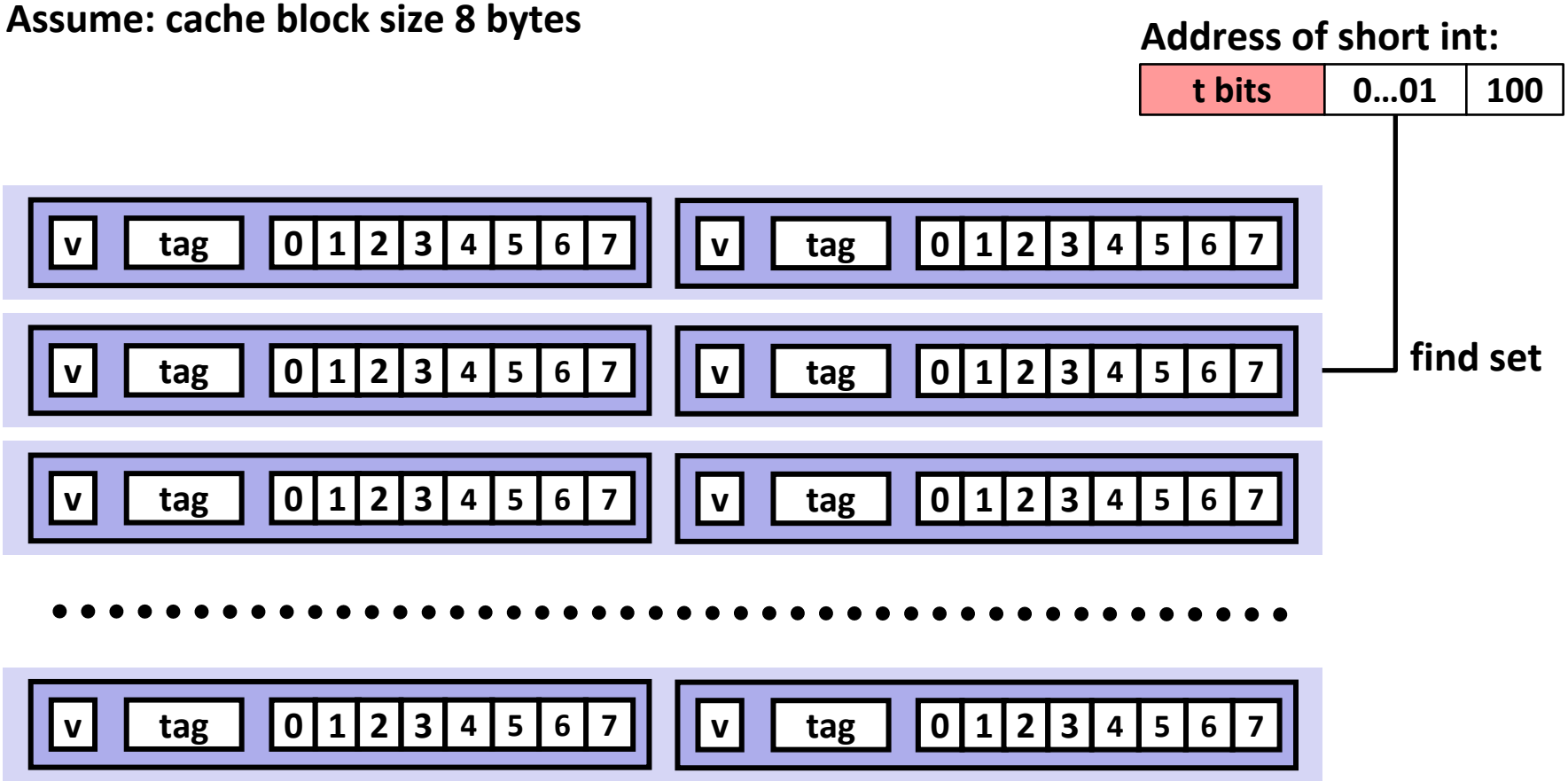
0	$[0000_2]$,	miss
1	$[0001_2]$,	hit
7	$[0111_2]$,	miss
8	$[1000_2]$,	miss
0	$[0000_2]$	miss

v	tag	data
1	0	M[0-1]
1	0	M[6-7]

E-way Set Associative Cache (Here: E = 2)

E = 2: Two lines per set

Assume: cache block size 8 bytes



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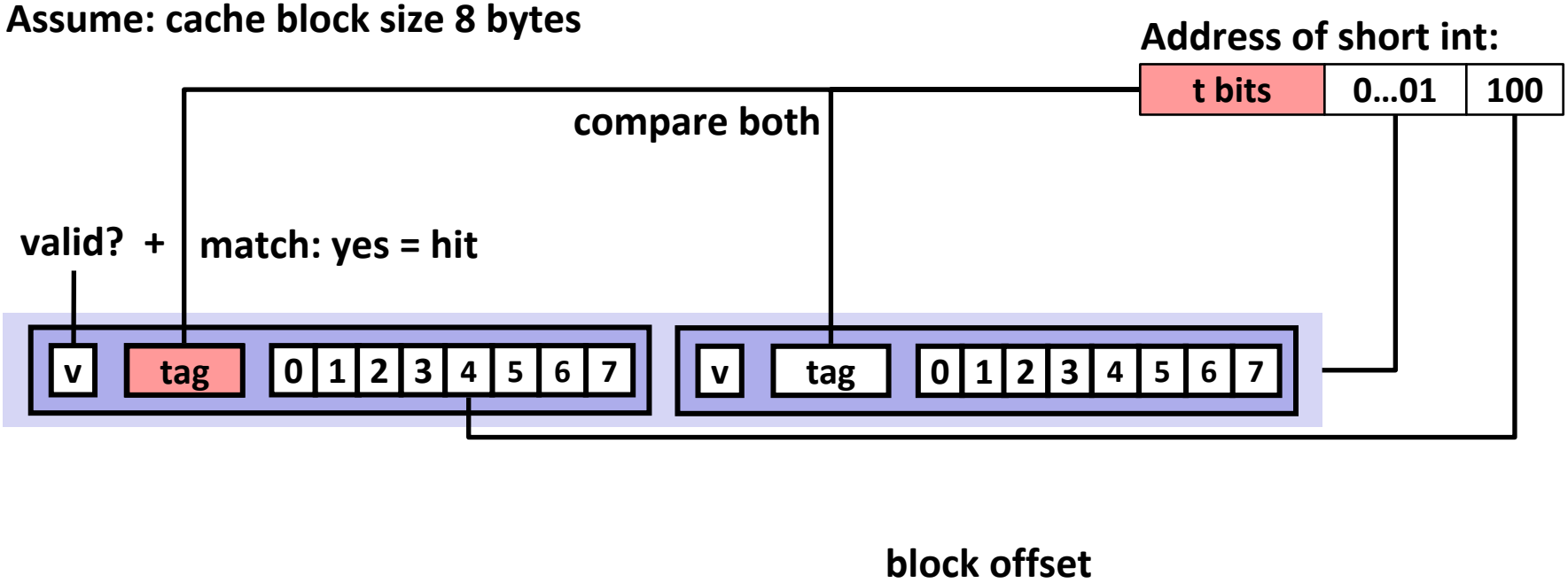
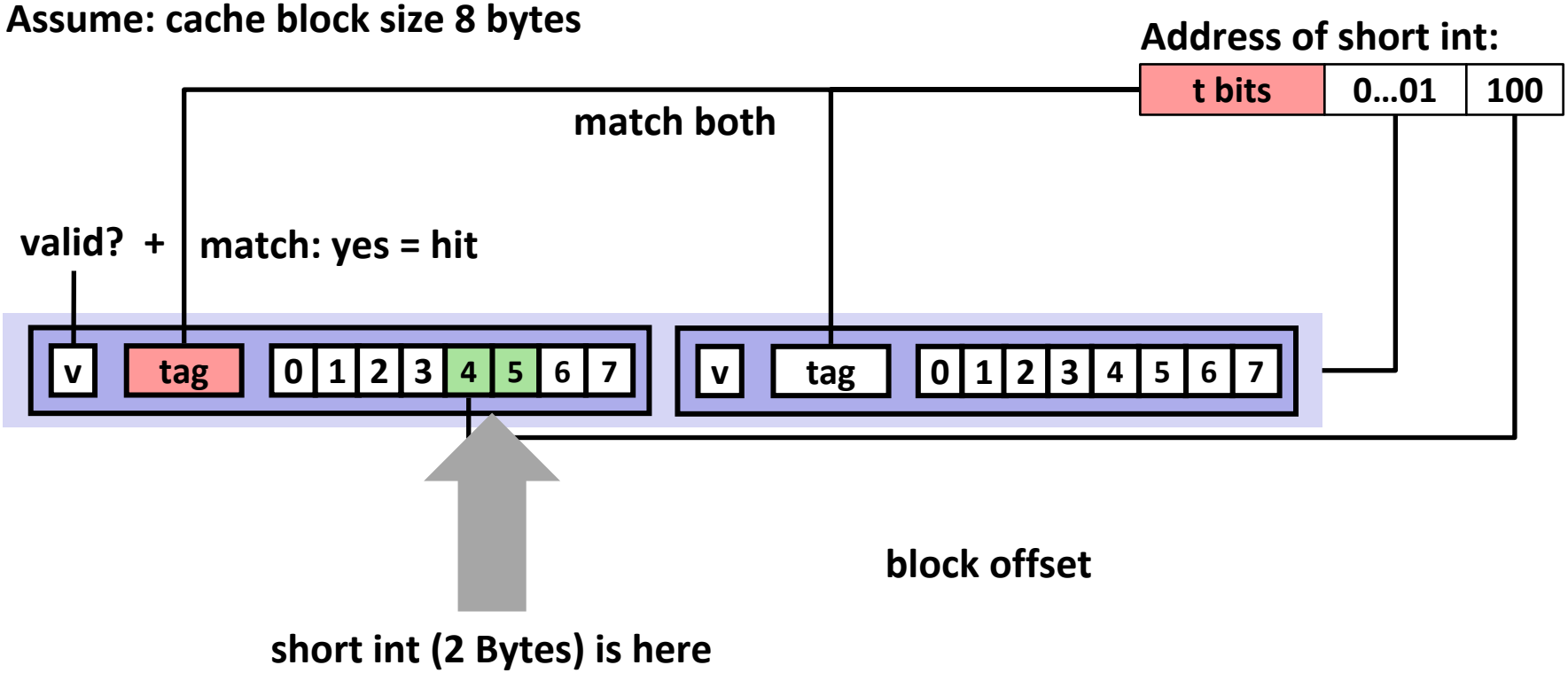


E-way Set Associative Cache (Here: E = 2)

E = 2: Two lines per set

Assume: cache block size 8 bytes



No match:

- One line in set is selected for eviction and replacement
- Replacement policies: random, least recently used (LRU), …

2-Way Associative Cache Simulation

t=2	s=1	b=1
xx	x	x

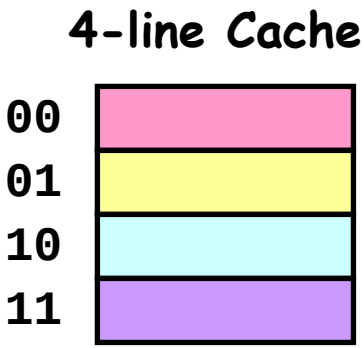
M=16 byte addresses, B=2 bytes/block,
S=2 sets, E=2 entry/set

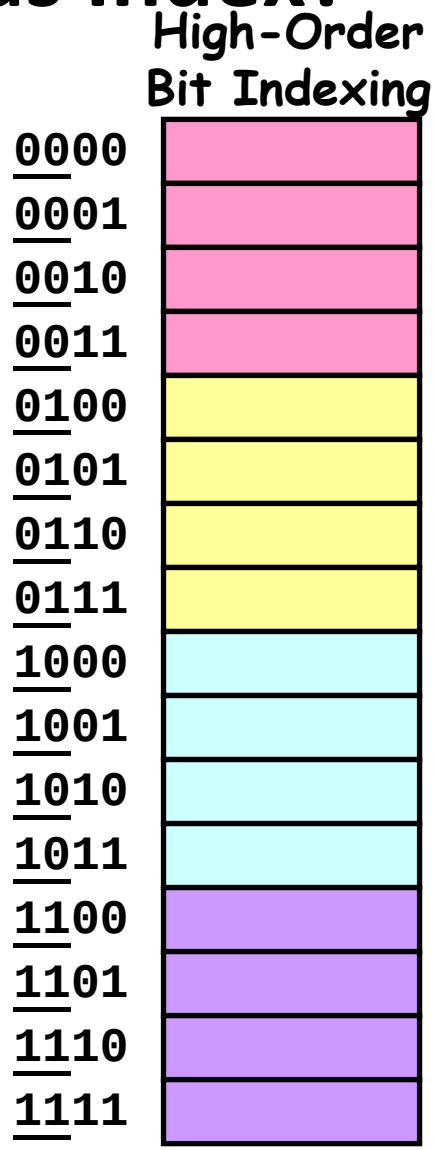
Address trace (reads):

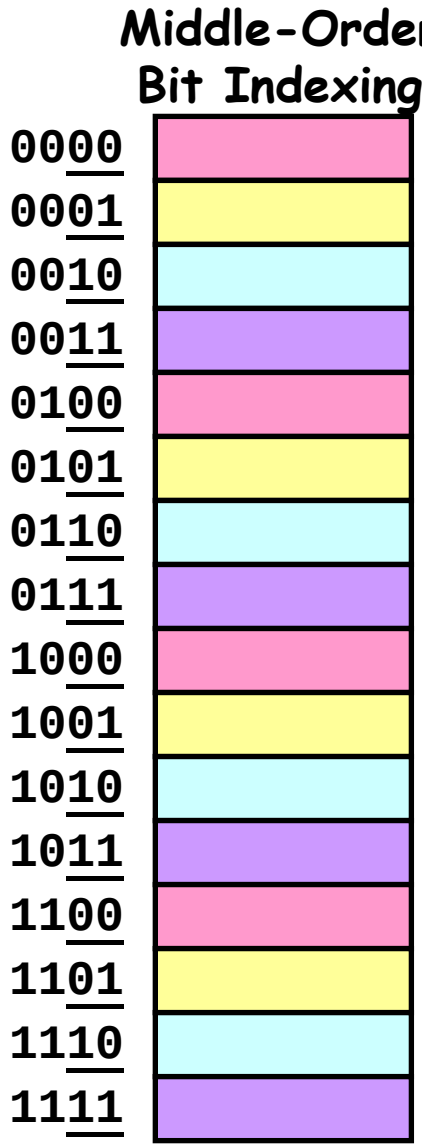
0	**[0000_2],**	miss
1	**[0001_2],**	hit
7	**[0111_2],**	miss
8	**[1000_2],**	miss
0	**[0000_2]**	hit

v	tag	data
1	00	M[0-1]
1	10	M[8-9]
1	01	M[6-7]
0		

Why Use Middle Bits as Index?







- **High-Order Bit Indexing**
  - Adjacent memory lines would map to same cache entry
  - Poor use of spatial locality
- **Middle-Order Bit Indexing**
  - Consecutive memory lines map to different cache lines
  - Can hold S*B*E-byte region of address space in cache at one time

What about writes?

- **Multiple copies of data exist:**
  - L1, L2, Main Memory, Disk
- **What to do on a write-hit?**
  - Write-through (write immediately to memory)
  - Write-back (defer write to memory until replacement of line)
    - Need a dirty bit (line different from memory or not)
- **What to do on a write-miss?**
  - Write-allocate (load into cache, update line in cache)
    - Good if more writes to the location follow
  - No-write-allocate (writes immediately to memory)
- **Typical**
  - Write-through + No-write-allocate
  - **Write-back + Write-allocate**

Cache Performance Metrics

- **Miss Rate**
  - Fraction of memory references not found in cache (misses / accesses) = 1 – hit rate
  - Typical numbers (in percentages):
    - 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
  - Typical numbers:
    - 1-2 clock cycle for L1
    - 5-20 clock cycles for L2
- **Miss Penalty**
  - Additional time required because of a miss
    - typically 50-200 cycles for main memory (Trend: increasing!)

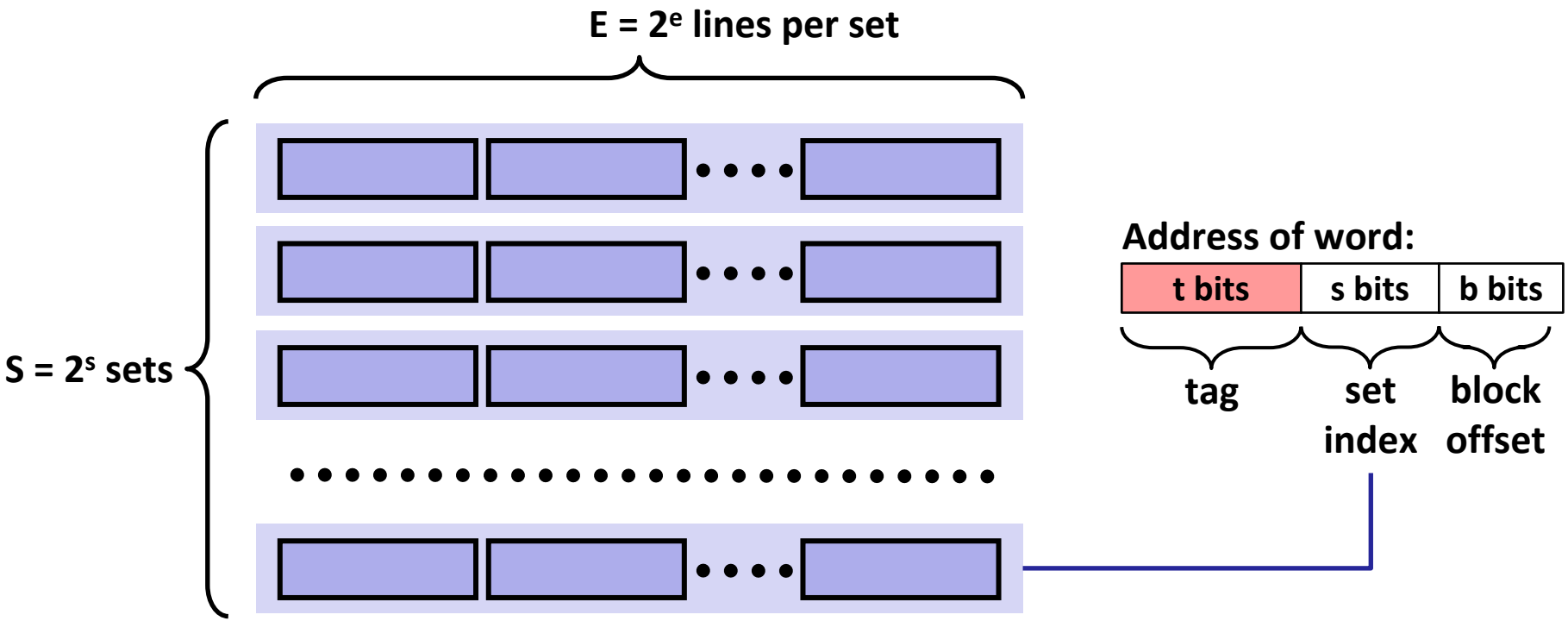
Aside for architects:
-Increasing cache size?
-Increasing block size?
-Increasing associativity?

Software Caches are More Flexible

- **Examples**
  - File system buffer caches, web browser caches, etc.

- **Some design differences**
  - Almost always fully associative
    - so, no placement restrictions
    - index structures like hash tables are common
  - Often use complex replacement policies
    - misses are very expensive when disk or network involved
    - worth thousands of cycles to avoid them
  - Not necessarily constrained to single “block” transfers
    - may fetch or write-back in larger units, opportunistically

Strided Access Question



- What happens if arrays are accessed in two-power strides?
- Example on the next slide

The Strided Access Problem (Blackboard?)

- **Example: L1 cache, Core 2 Duo**
  - 32 KB, 8-way associative, 64 byte cache block size
  - What is S, E, B?
    - *Answer:* B = 2^6, E = 2^3, S = 2^6.

- **Consider an array of ints accessed at stride 2^i, $i \geq 0$**
  - What is the smallest i such that only one set is used?
    - *Answer:* i = 10
  - What happens if the stride is 2^9?
    - *Answer:* two sets are used

- **Source of two-power strides?**
  - Example: Column access of 2-D arrays (images!)

Writing Cache Friendly Code

- Repeated references to variables are good (temporal locality)
- Stride-1 reference patterns are good (spatial locality)
- Examples:
  - cold cache, 4-byte words, 4-word cache blocks

```
int sum_array_rows(int a[M][N])
{
  int i, j, sum = 0;

  for (i = 0; i < M; i++)
     for (j = 0; j < N; j++)
        sum += a[i][j];
  return sum;
}
```

Miss rate = 1/4 = 25%

```
int sum_array_cols(int a[M][N])
{
  int i, j, sum = 0;

  for (j = 0; j < N; j++)
     for (i = 0; i < M; i++)
        sum += a[i][j];
  return sum;
}
```

Miss rate = 100%

Today

- Cache organization
- **Memory Wall**
- Program optimization (Matrix Multiplication):
  - Cache optimizations
  - Cache Miss Analysis

The Memory Mountain

- **Read throughput (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

```
/* The test function */
void test(int elems, int stride) {
    int i, result = 0;
    volatile int sink;

    for (i = 0; i < elems; i += stride)
        result += data[i];
    sink = result; /* So compiler doesn't optimize away the loop */
}

/* Run test(elems, stride) and return read throughput (MB/s) */
double run(int size, int stride, double Mhz)
{
    double cycles;
    int elems = size / sizeof(int);

    test(elems, stride);                    /* warm up the cache */
    cycles = fcyc2(test, elems, stride, 0);  /* call test(elems,stride) */
    return (size / stride) / (cycles / Mhz); /* convert cycles to MB/s */
}
```

Memory Mountain Main Routine

```
/* mountain.c - Generate the memory mountain. */
#define MINBYTES (1 << 10)  /* Working set size ranges from 1 KB */
#define MAXBYTES (1 << 23)  /* ... up to 8 MB */
#define MAXSTRIDE 16        /* Strides range from 1 to 16 */
#define MAXELEMS MAXBYTES/sizeof(int)

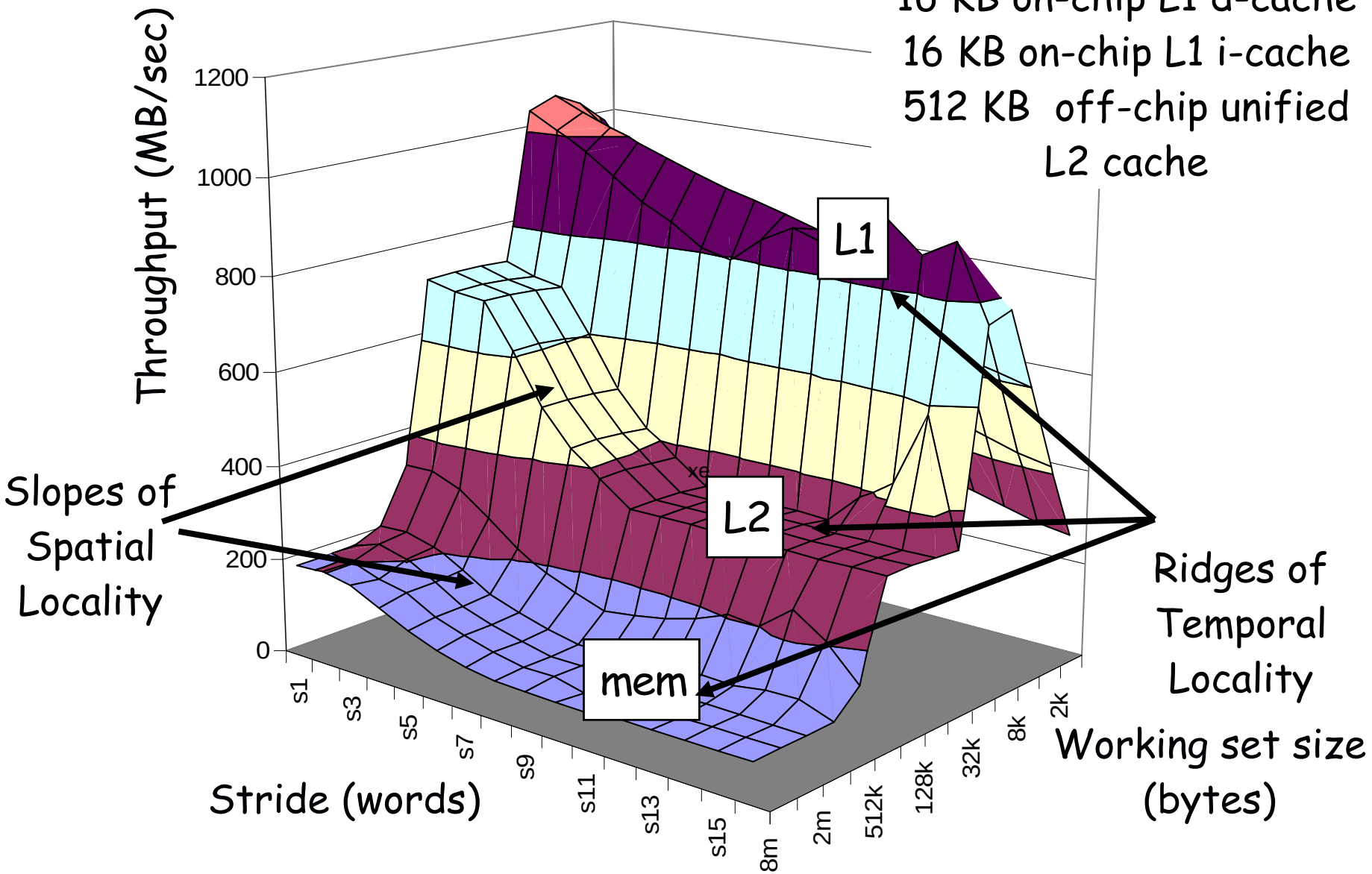
int data[MAXELEMS];         /* The array we'll be traversing */

int main()
{
    int size;        /* Working set size (in bytes) */
    int stride;      /* Stride (in array elements) */
    double Mhz;      /* Clock frequency */

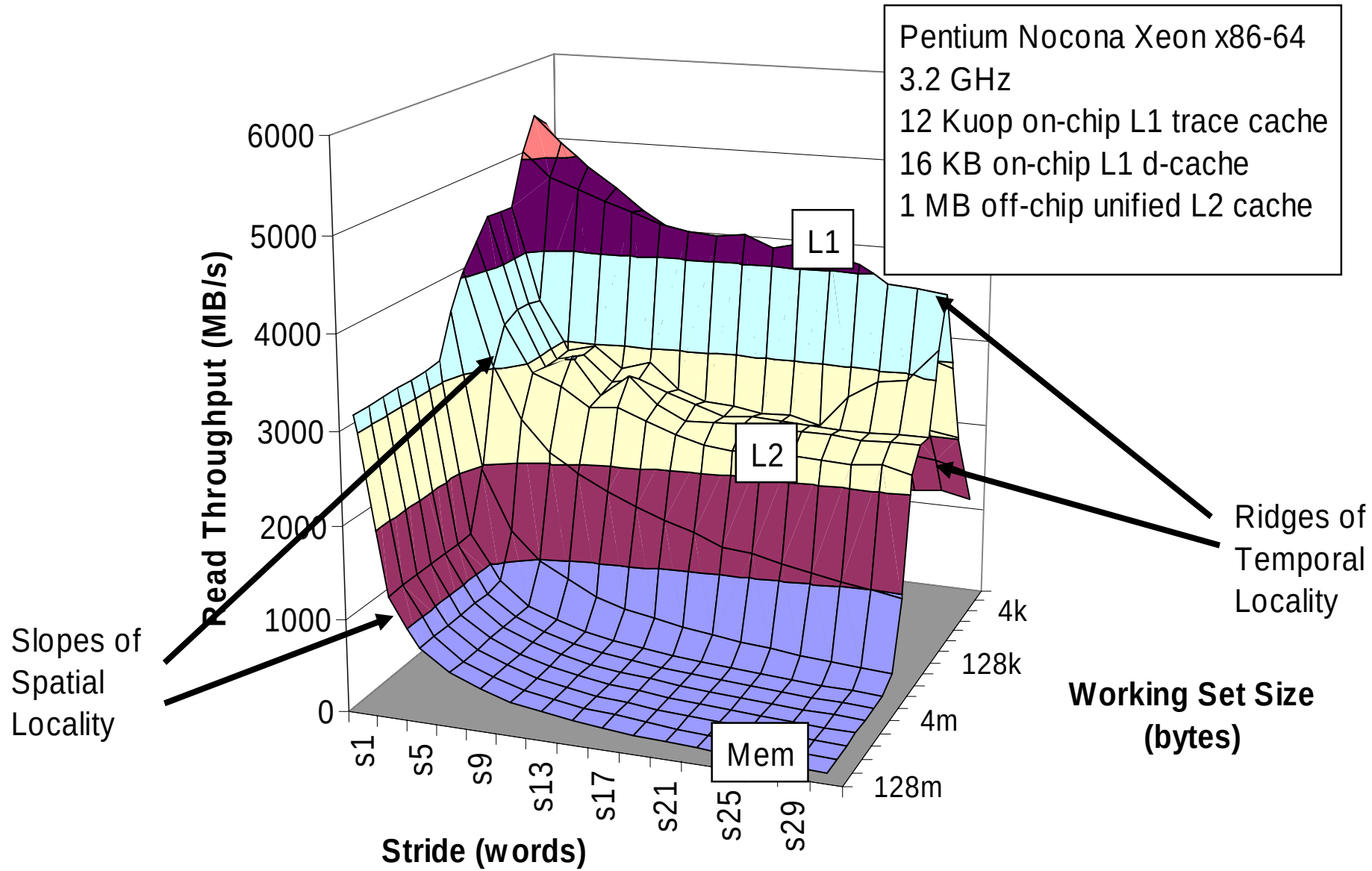
    init_data(data, MAXELEMS); /* Initialize each element in data to 1 */
    Mhz = mhz(0);              /* Estimate the clock frequency */
    for (size = MAXBYTES; size >= MINBYTES; size >>= 1) {
        for (stride = 1; stride <= MAXSTRIDE; stride++)
            printf("%.1f\t", run(size, stride, Mhz));
        printf("\n");
    }
    exit(0);
}
```

The Memory Mountain

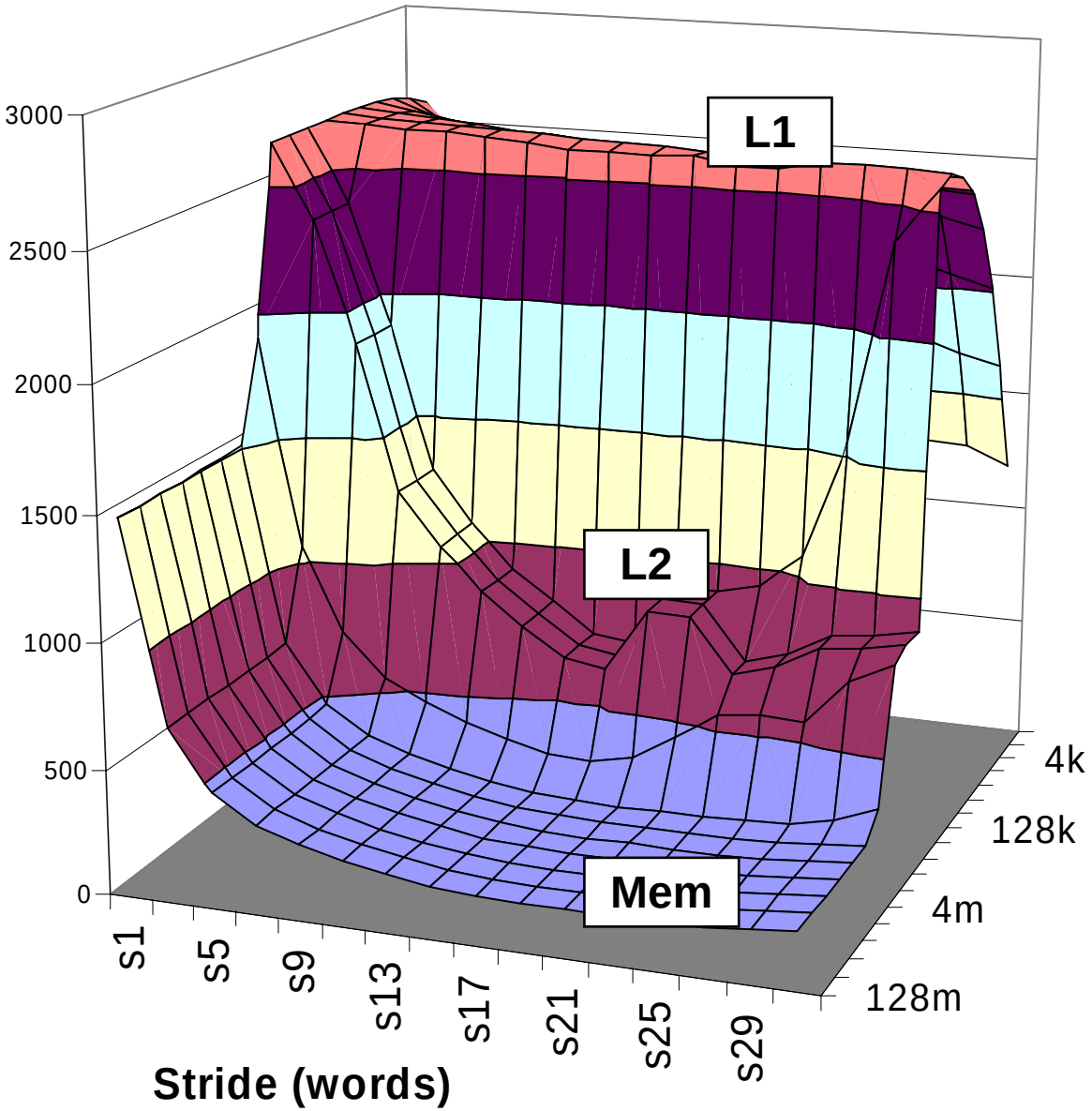
Pentium III
550 MHz
16 KB on-chip L1 d-cache
16 KB on-chip L1 i-cache
512 KB off-chip unified
L2 cache



X86-64 Memory Mountain



Opteron Memory Mountain

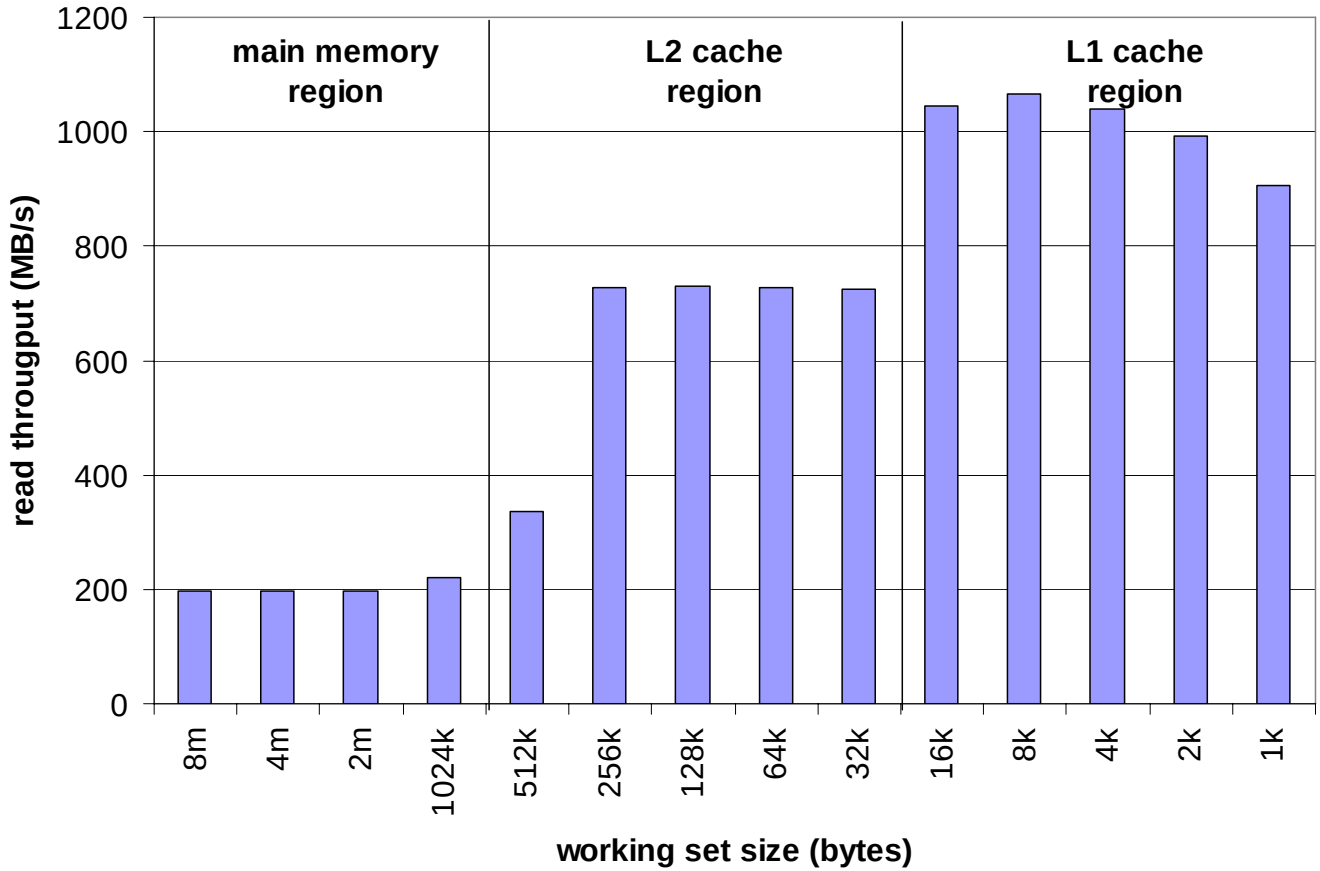


Read throughput (MB/s)

AMD Opteron
2 GHZ

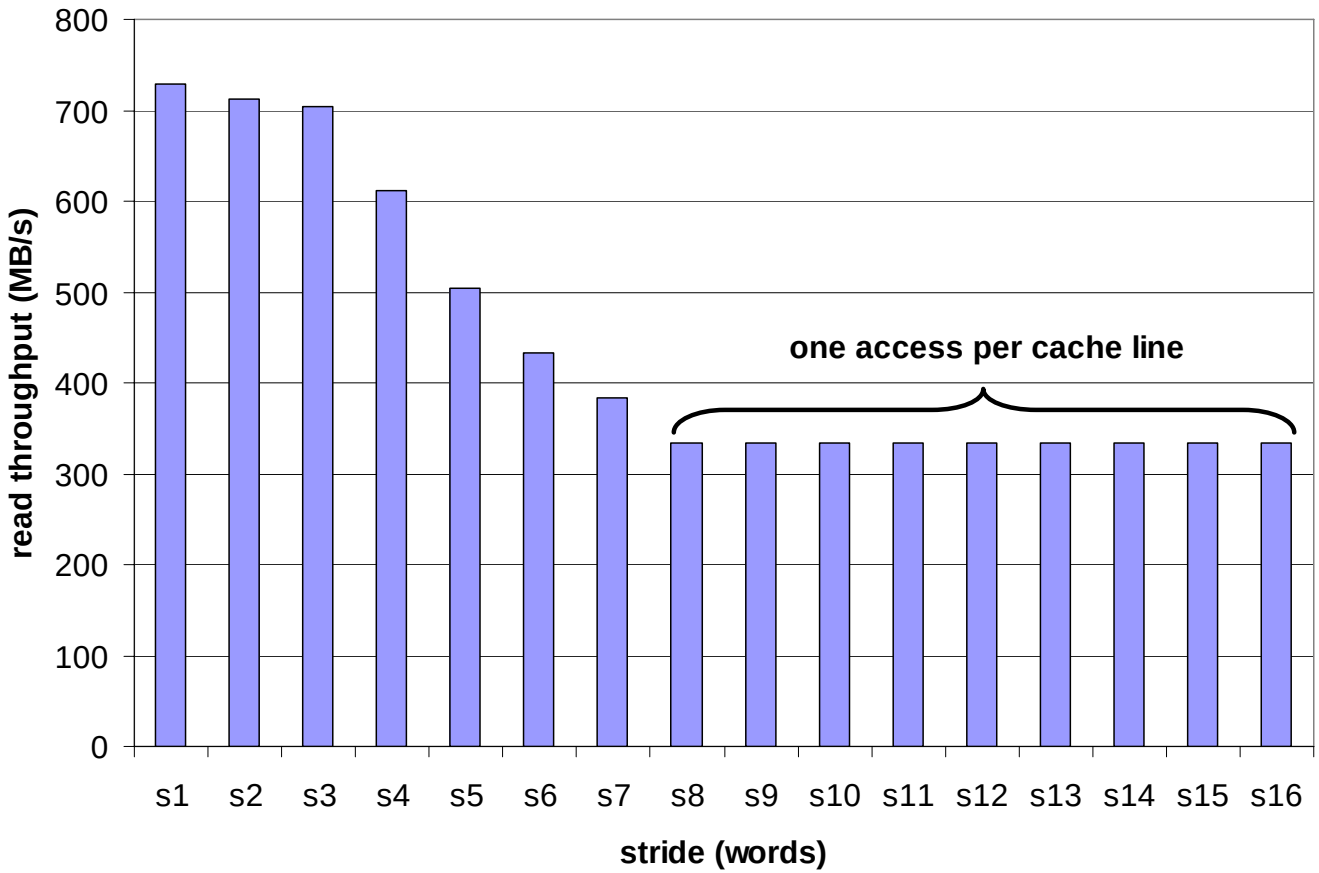
Ridges of Temporal Locality

- **Slice through the memory mountain with stride=1**
  - illuminates read throughputs of different caches and memory



A Slope of Spatial Locality

- **Slice through memory mountain with size=256KB**
  - shows cache block size.



Today

- Cache organization
- Memory Wall
- **Program optimization (Matrix Multiplication):**
  - Cache optimizations
  - Cache Miss Analysis

Matrix Multiplication Example

- **Major Cache Effects to Consider**
  - Total cache size
    - Exploit temporal locality and keep the working set small (e.g., use blocking)
  - Block size
    - Exploit spatial locality

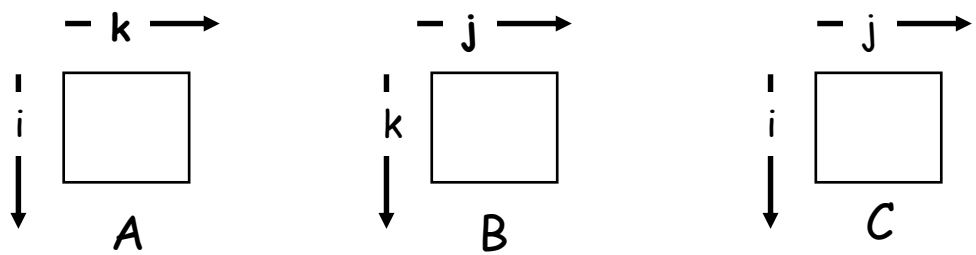
- **Description:**
  - Multiply N x N matrices
  - O(N3) total operations
  - Accesses
    - N reads per source element
    - N values summed per destination
      - but may be able to hold in register

```
/* 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;
  }
}
```

Variable **sum** *held in register*

Miss Rate Analysis for Matrix Multiply

- **Assume:**
  - Line size = 32B (big enough for four 64-bit words)
  - 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



Layout of C Arrays in Memory (review)

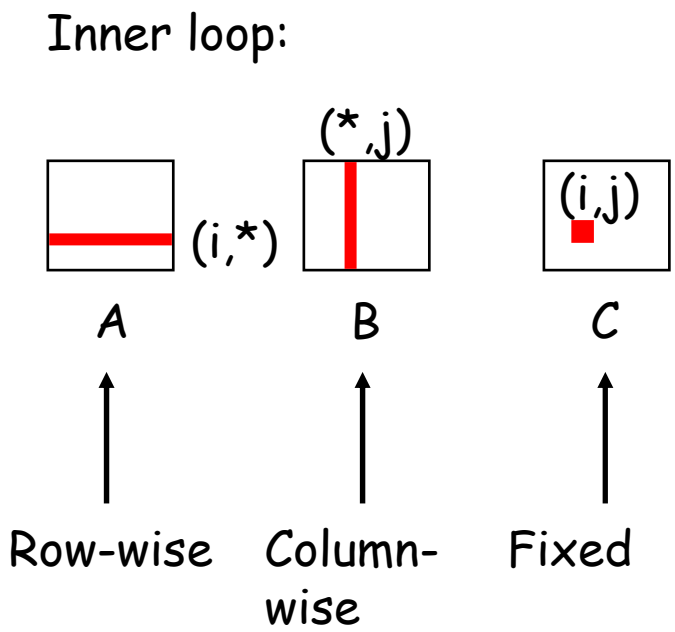
- **C arrays allocated in row-major order**
  - each row in contiguous memory locations
- **Stepping through columns in one row:**
  - `for (i = 0; i < N; i++)`

    `sum += a[0][i];`
  - accesses successive elements
  - if block size (B) > 4 bytes, exploit spatial locality
    - compulsory miss rate = 4 bytes / B
- **Stepping through rows in one column:**
  - `for (i = 0; i < n; i++)`

    `sum += a[i][0];`
  - accesses distant elements
  - no spatial locality!
    - compulsory miss rate = 1 (i.e. 100%)

Matrix Multiplication (ijk)

```
/* 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;
  }
}
```



Misses per Inner Loop Iteration:

A	B	C
0.25	1.0	0.0

Matrix Multiplication (jik)

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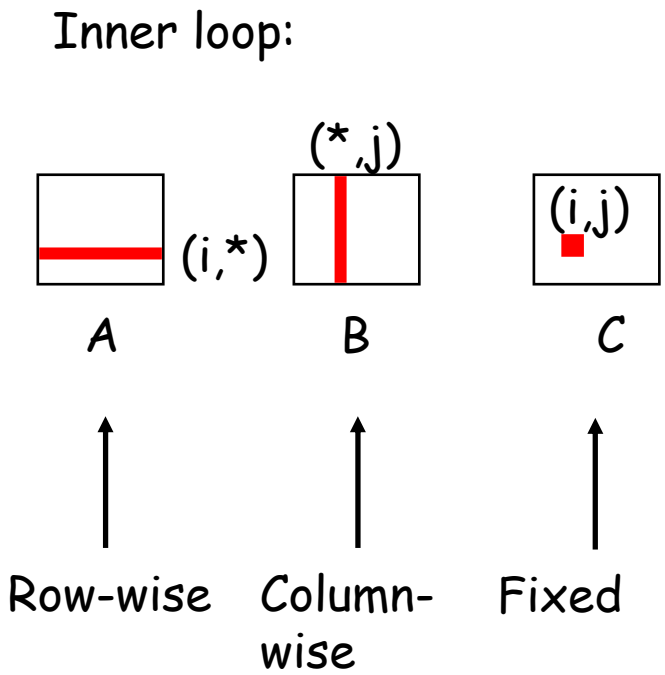
      sum += a[i][k] * b[k][j];
    c[i][j] = sum
  }
}
```



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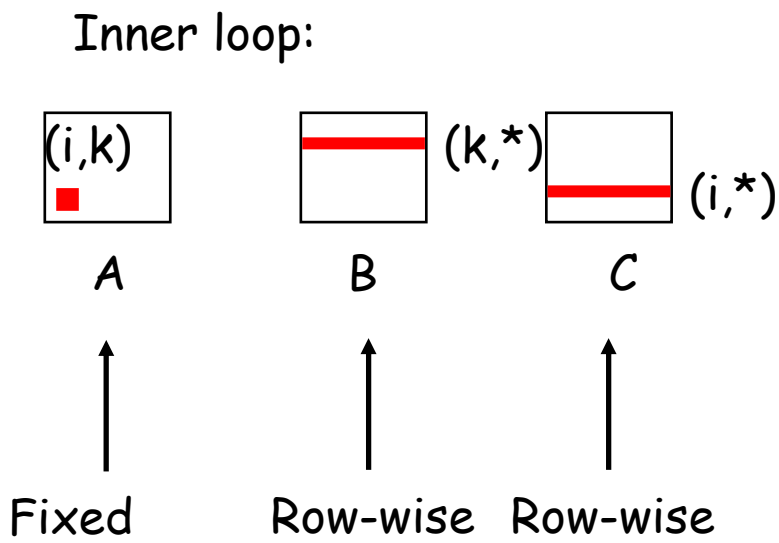
0.25	1.0	0.0

Matrix Multiplication (kij)

```
/* 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];
  }
}
```



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Matrix Multiplication (ikj)

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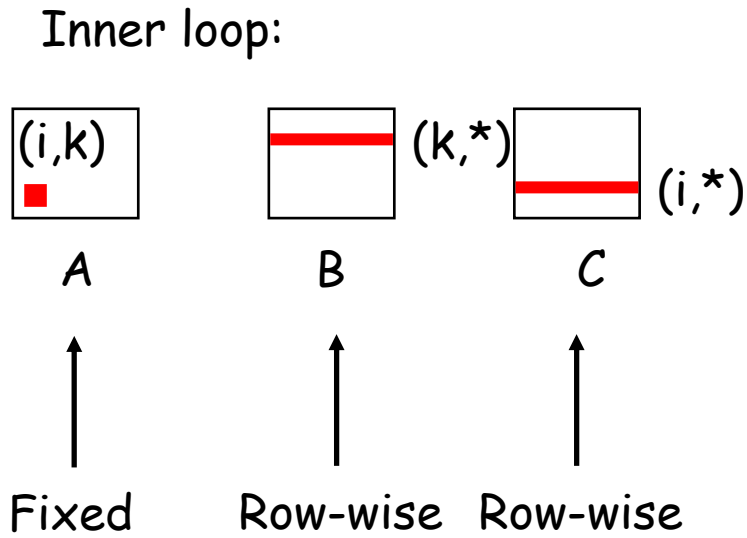
    for (j=0; j<n; j++)
      c[i][j] += r * b[k][j];
  }
}
```



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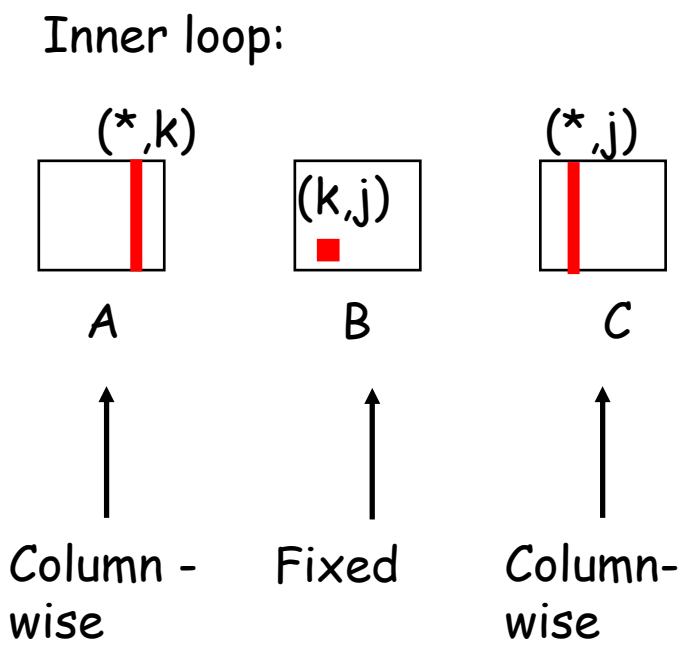

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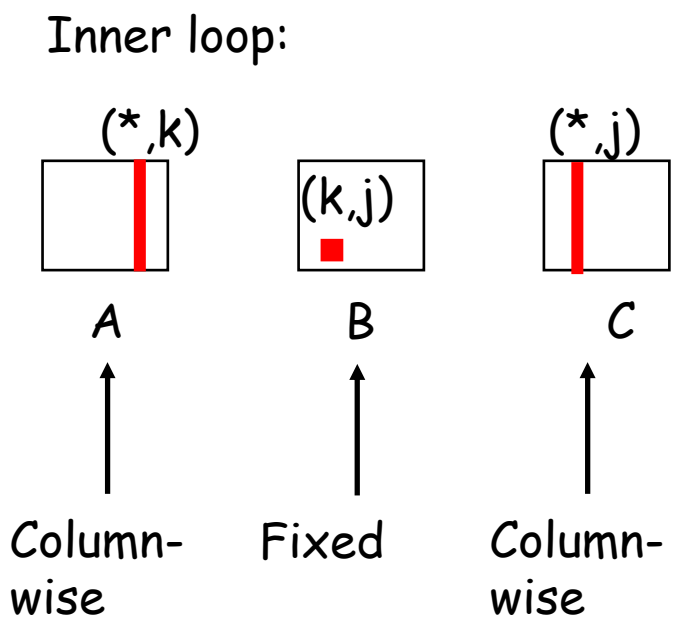


Misses per Inner Loop Iteration:

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;
 }
}
```

ijk (& jik):

- 2 loads, 0 stores
- misses/iter = **1.25**

```
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];
 }
}
```

kij (& ikj):

- 2 loads, 1 store
- misses/iter = **0.5**

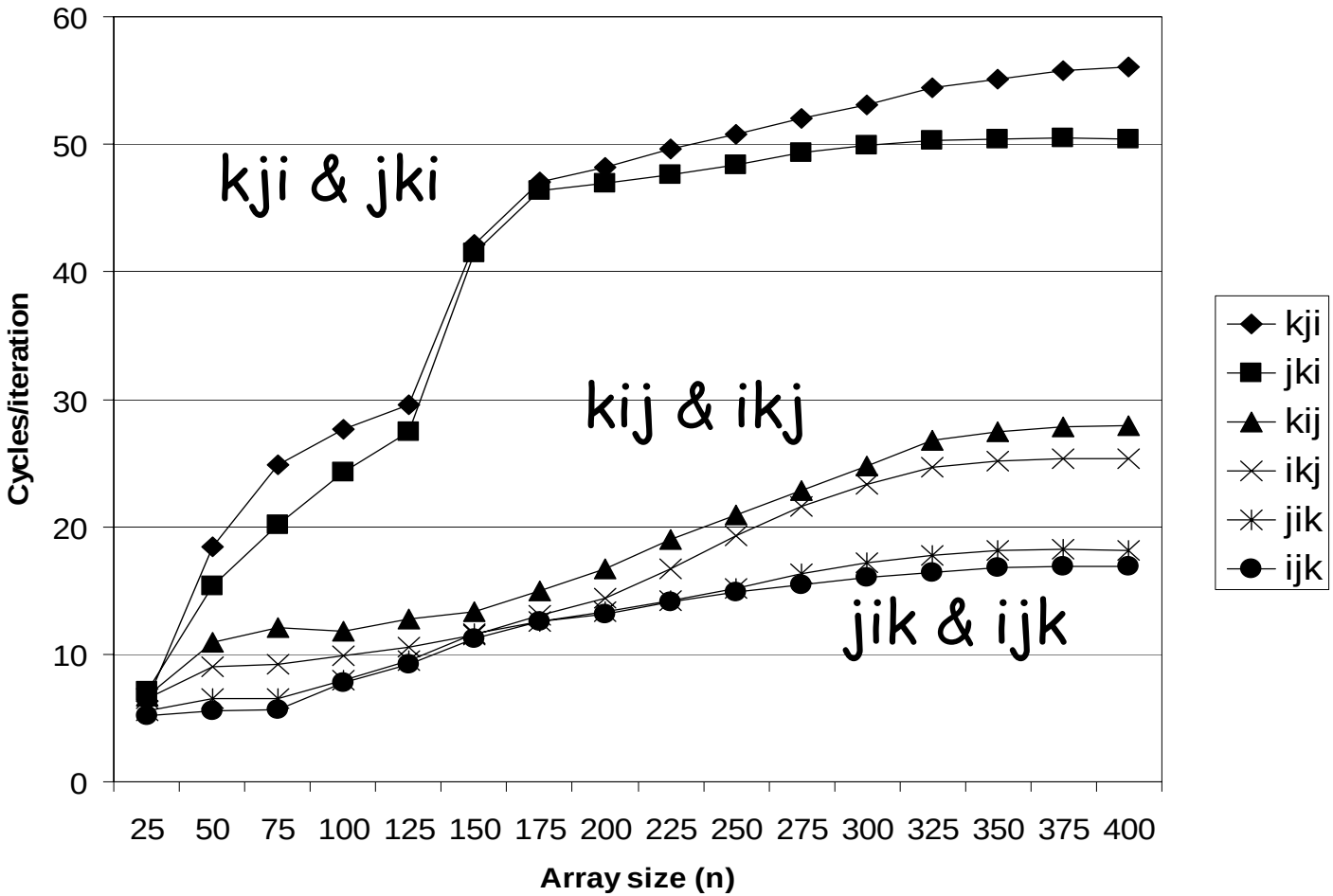
```
for (j=0; j<n; j++) {
 for (k=0; k<n; k++) {
   r = b[k][j];
   for (i=0; i<n; i++)
    c[i][j] += a[i][k] * r;
 }
}
```

jki (& kji):

- 2 loads, 1 store
- misses/iter = **2.0**

Pentium Matrix Multiply Performance

- **Miss rates are helpful but not perfect predictors.**
  - Code scheduling matters, too.



Improving Temporal Locality by Blocking

- **Example: Blocked matrix multiplication**
  - "block" (in this context) does not mean "cache block".
  - Instead, it mean a sub-block within the matrix.
  - Example: N = 8; sub-block size = 4

$$\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$$

Key idea: Sub-blocks (i.e., $\mathbf{A_{xy}}$) can be treated just like scalars.

$$C_{11} = A_{11}B_{11} + A_{12}B_{21} \qquad C_{12} = A_{11}B_{12} + A_{12}B_{22}$$

$$C_{21} = A_{21}B_{11} + A_{22}B_{21} \qquad C_{22} = A_{21}B_{12} + A_{22}B_{22}$$

Blocked Matrix Multiply (bijk)

```
for (jj=0; jj<n; jj+=bsize) {

  for (i=0; i<n; i++)
    for (j=jj; j < min(jj+bsize,n); j++)
      c[i][j] = 0.0;

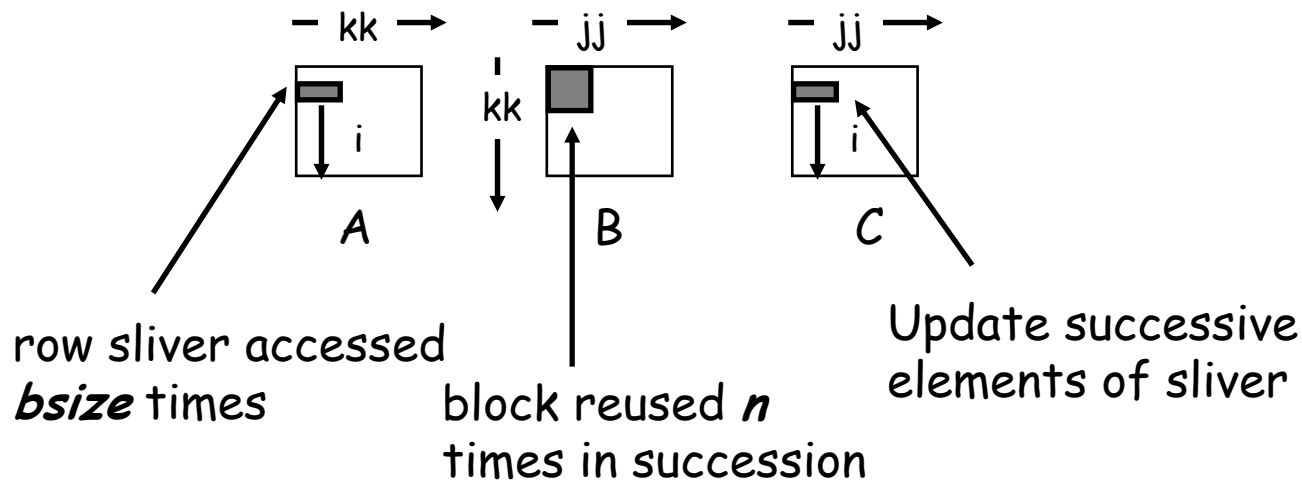
  for (kk=0; kk<n; kk+=bsize) {
    for (i=0; i<n; i++) {
      for (j=jj; j < min(jj+bsize,n); j++) {
        sum = 0.0
        for (k=kk; k < min(kk+bsize,n); k++) {
          sum += a[i][k] * b[k][j];
        }
        c[i][j] += sum;
      }
    }
  }
}
```

Blocked Matrix Multiply Analysis

- Innermost loop pair multiplies a *1 X bsize* sliver of *A* by a *bsize X bsize* block of *B* and accumulates into *1 X bsize* sliver of *C*
- Loop over *i* steps through *n* row slivers of *A* & *C*, using same *B*

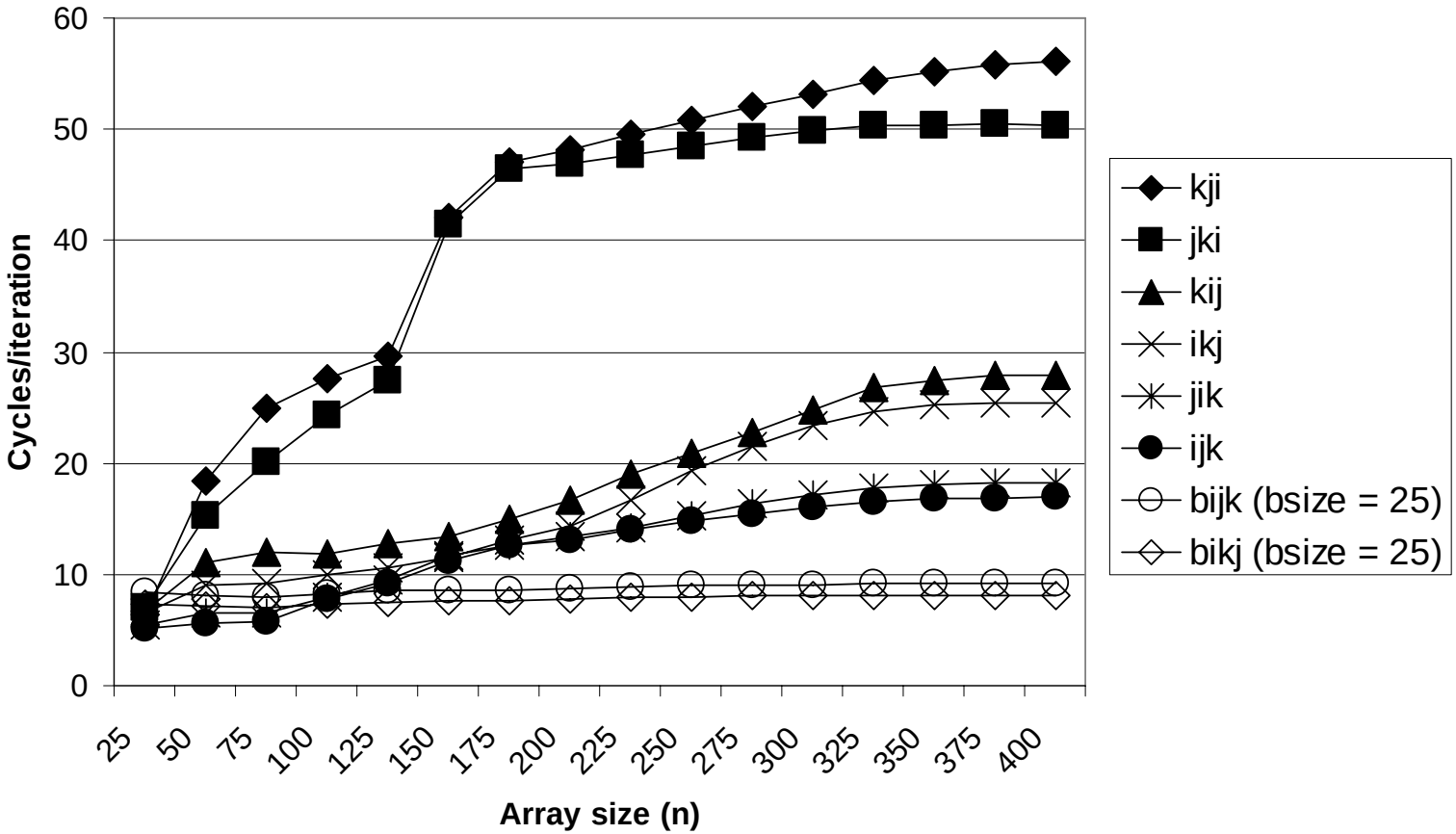
```
for (i=0; i<n; i++) {
    for (j=jj; j < min(jj+bsize,n); j++) {
      sum = 0.0
      for (k=kk; k < min(kk+bsize,n); k++) {
        sum += a[i][k] * b[k][j];
      }
      c[i][j] += sum;
    }
```

Innermost Loop Pair



Pentium Blocked Matrix Multiply Performance

- **Blocking (bijk and bikj) improves performance by a factor of two over unblocked versions (ijk and jik)**
  - relatively insensitive to array size.



Observations

- **Programmer can optimize for cache performance**
  - How data structures are organized
  - How data are accessed
    - Nested loop structure
    - Blocking is a general technique
- **All systems favor "cache friendly code"**
  - Getting absolute optimum performance is very platform specific
    - Cache sizes, line sizes, associativities, etc.
  - Can get most of the advantage with generic code
    - Keep working set reasonably small (temporal locality)
    - Use small strides (spatial locality)

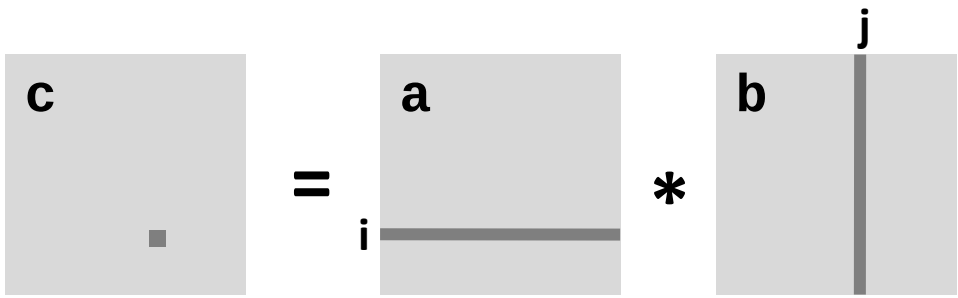
Today

- Cache organization
- Memory Wall
- **Program optimization (Matrix Multiplication):**
  - Cache optimizations
  - Cache Miss Analysis

Example: Matrix Multiplication

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

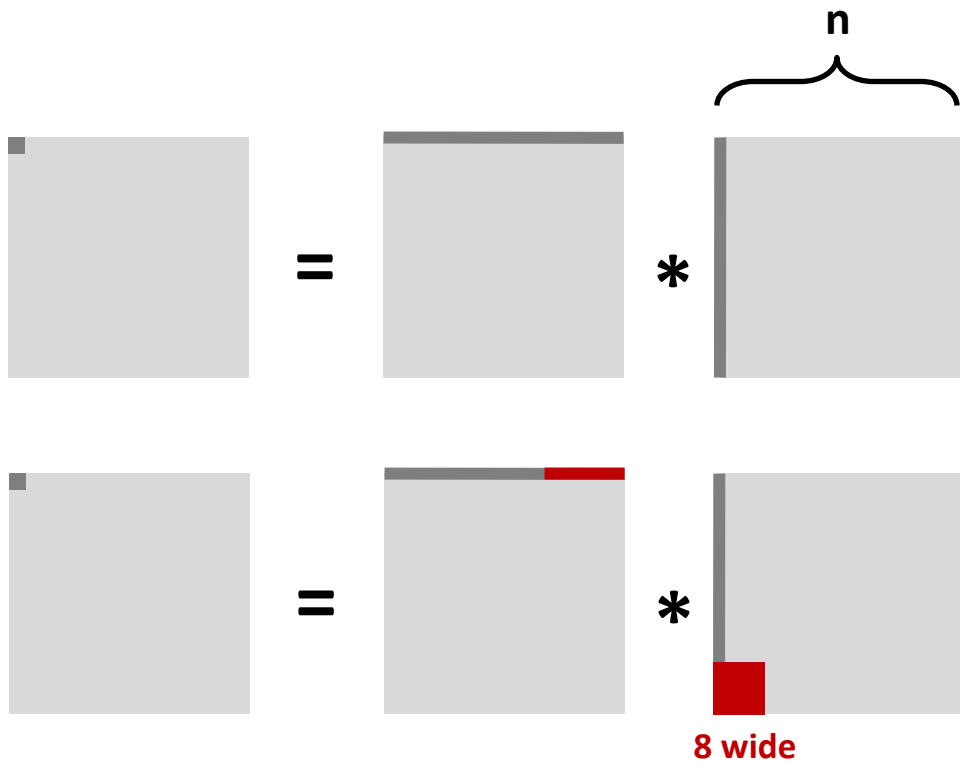
/* Multiply n x n matrices a and b  */
void mmm(double *a, double *b, double *c, int n) {
    int i, j, k;
    for (i = 0; i < n; i++)
        for (j = 0; j < n; j++)
            for (k = 0; k < n; k++)
                c[i*n+j] += a[i*n + k]*b[k*n + j];
}
```



Cache Miss Analysis

- **Assume:**
  - Matrix elements are doubles
  - Cache block = 8 doubles (64 B as in Core 2 Duo)
  - Cache size C << n (much smaller than n)

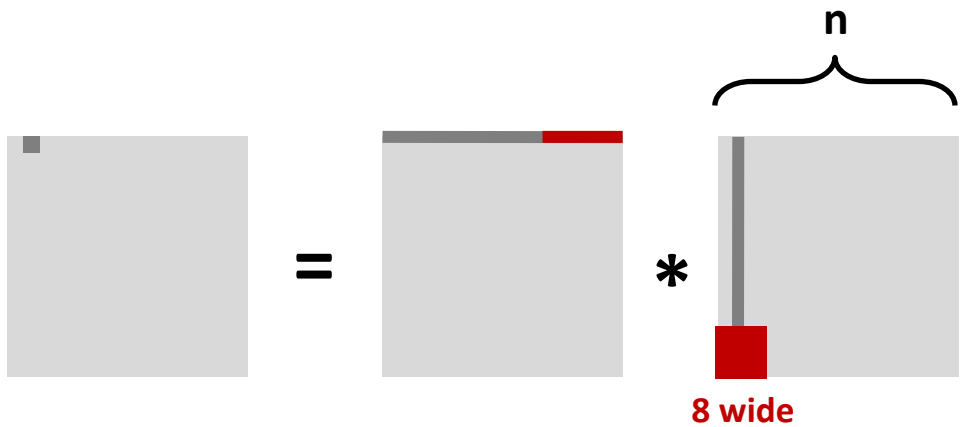
- **First iteration:**
  - n/8 + n = 9n/8 misses
  - Afterwards in cache: (schematic)



Cache Miss Analysis

- **Assume:**
  - Matrix elements are doubles
  - Cache block = 8 doubles
  - Cache size C << n (much smaller than n)

- **Second iteration:**
  - Again:
    n/8 + n = 9n/8 misses

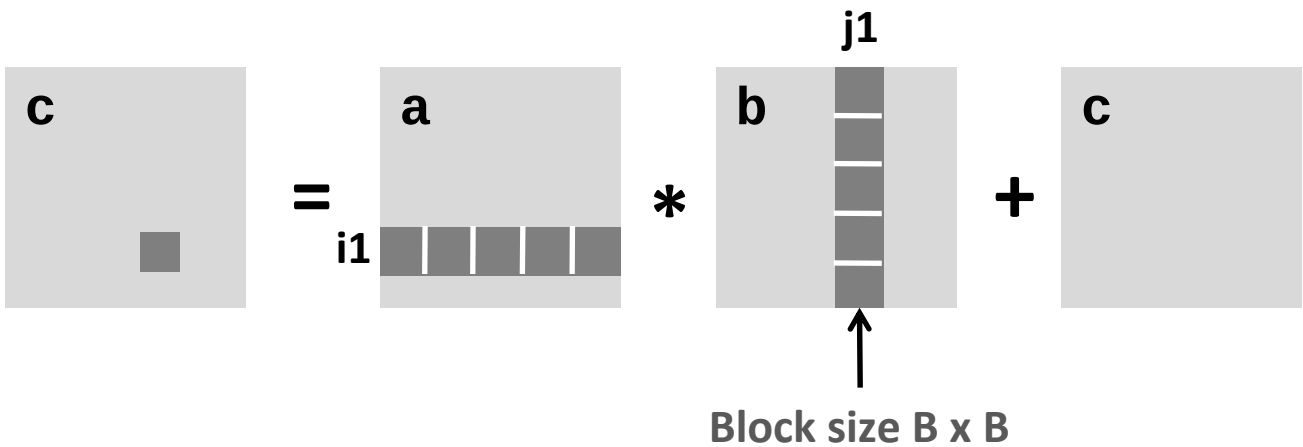


- **Total misses:**
  - $9n/8 * n^2 = (9/8) * n^3$

Blocked Matrix Multiplication

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

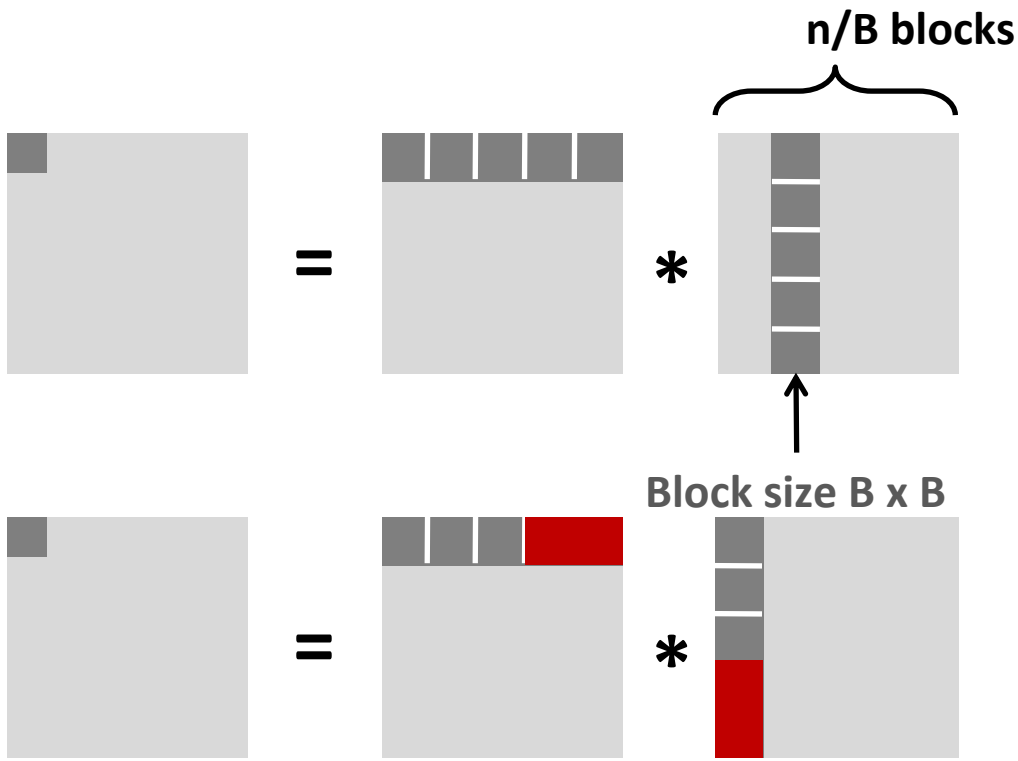
/* Multiply n x n matrices a and b  */
void mmm(double *a, double *b, double *c, int n) {
    int i, j, k;
    for (i = 0; i < n; i+=B)
        for (j = 0; j < n; j+=B)
            for (k = 0; k < n; k+=B)
                /* B x B mini matrix multiplications */
                for (i1 = i; i1 < i+B; 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 block = 8 doubles
  - Cache size C << n (much smaller than n)
  - Three blocks ■ fit into cache: $3B^2 < C$

- **First (block) iteration:**
  - $B^2/8$ misses for each block
  - $2n/B * B^2/8 = nB/4$ (omitting matrix c)

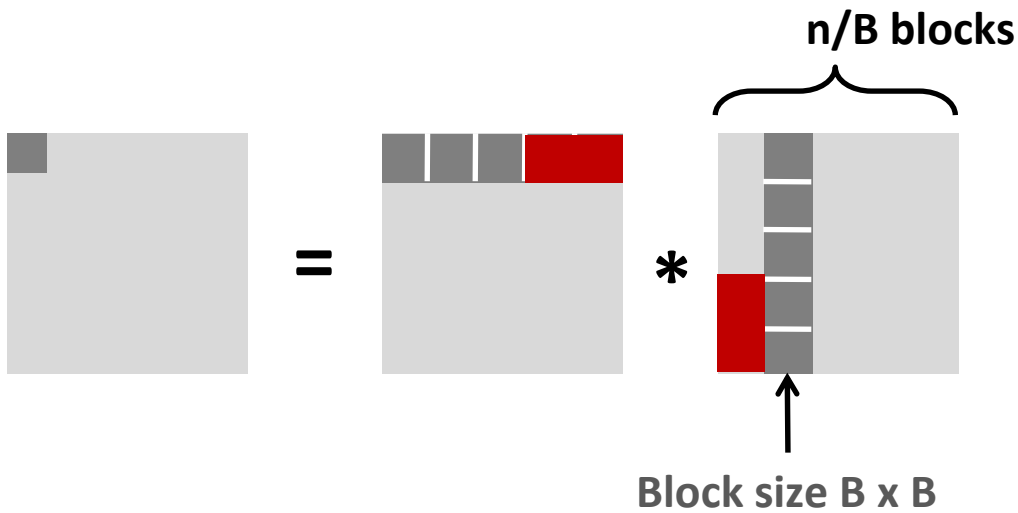


  - Afterwards in cache (schematic)

Cache Miss Analysis

- **Assume:**
  - Cache block = 8 doubles
  - Cache size C << n (much smaller than n)
  - Three blocks ■ fit into cache: $3B^2 < C$

- **Second (block) iteration:**
  - Same as first iteration
  - $2n/B * B^2/8 = nB/4$



- **Total misses:**
  - $nB/4 * (n/B)^2 = n^3/(4B)$

Summary

- **No blocking:** **$(9/8) * n^3$**
- **Blocking:** **$1/(4B) * n^3$**

- **Suggest largest possible block size B, but limit $3B^2 < C$!**
  (can possibly be relaxed a bit, but there is a limit for B)

- **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