

CS 250B: Modern Computer Systems

Cache And Memory System



Sang-Woo Jun

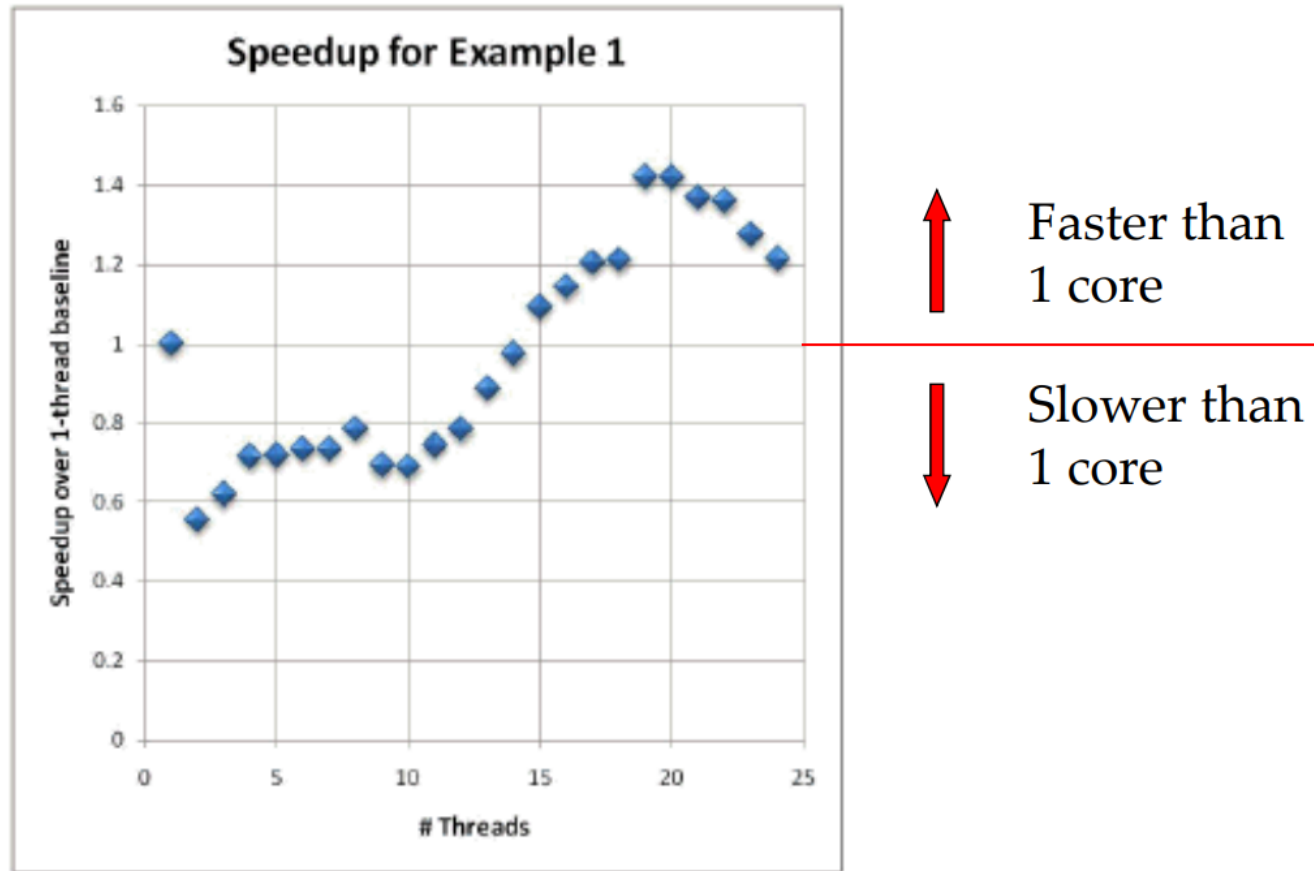
Motivation Example: An Embarrassingly Parallel Workload

- A very simple example of counting odd numbers in a large array

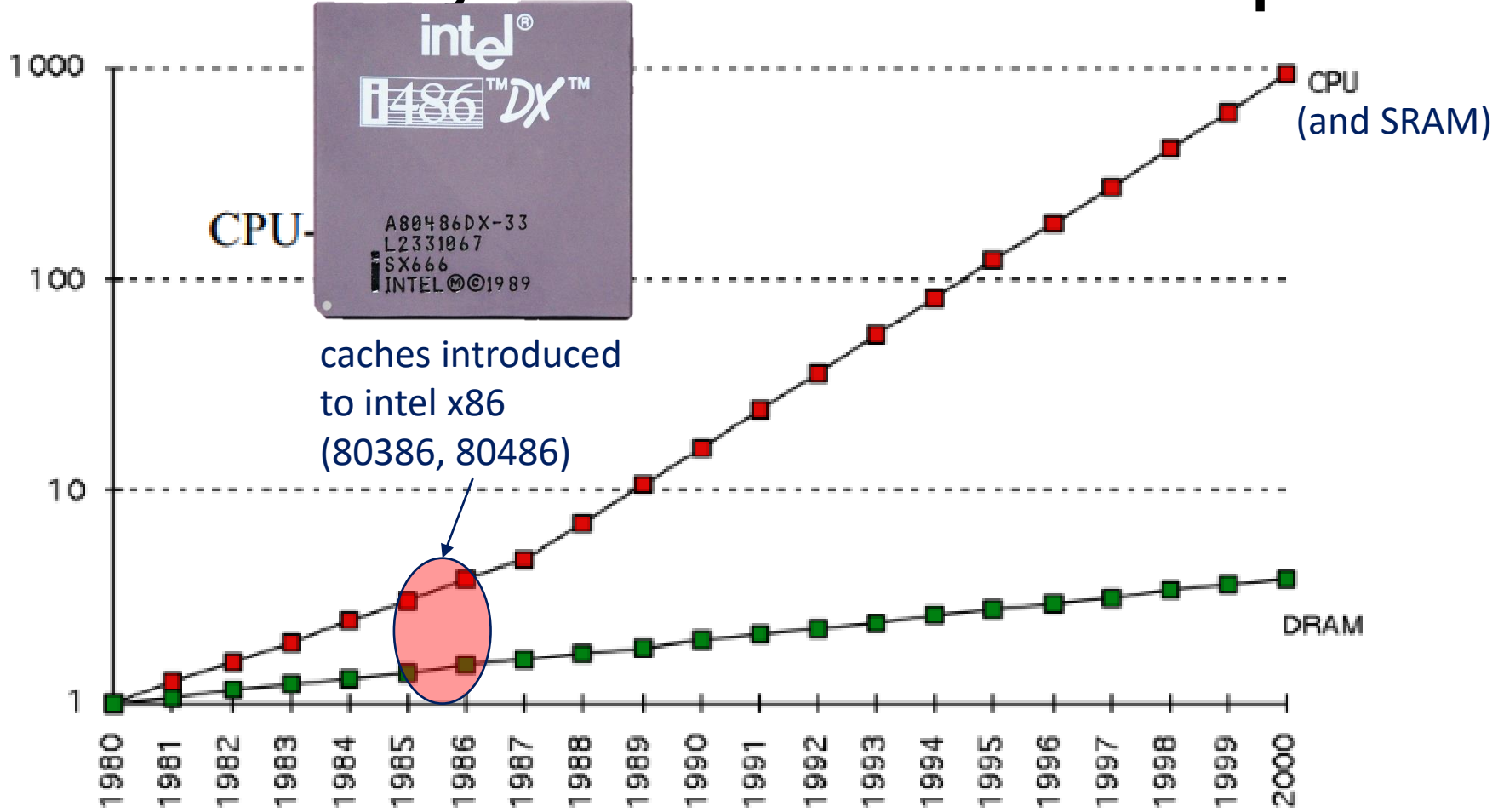
```
int results[THREAD_COUNT];  
void worker_thread(...) {  
    int tid = ...;  
    for (e in myChunk) {  
        if ( e % 2 != 0) results[tid]++;  
    }  
}
```

Do you see any performance red flags?

Scalability Unimpressive



History of The Processor/Memory Performance Gap



What is the Y-axis? Most likely normalized latency reciprocal

Source: Extreme tech, "How L1 and L2 CPU Caches Work, and Why They're an Essential Part of Modern Chips," 2018

Purpose of Caches

- ❑ The CPU is (largely) unaware of the underlying memory hierarchy
 - The memory abstraction is a single address space
 - The memory hierarchy automatically stores data in fast or slow memory, depending on usage patterns
- ❑ Multiple levels of “caches” act as interim memory between CPU and main memory (typically DRAM)
 - Processor accesses main memory through the cache hierarchy
 - If requested address is already in the cache (address is “cached”, resulting in “cache hit”), data operations can be fast
 - If not, a “cache miss” occurs, and must be handled to return correct data to CPU

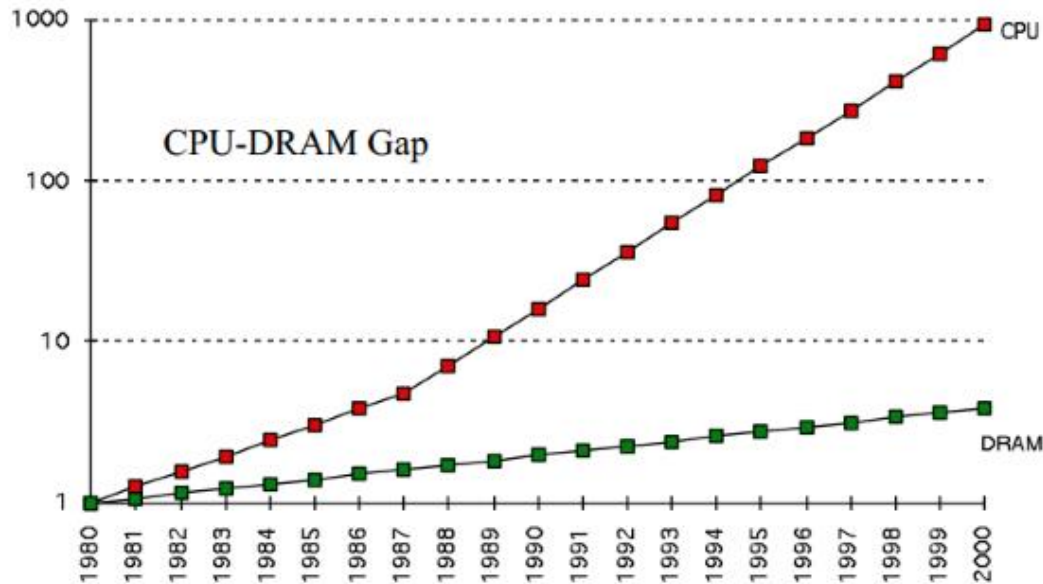
Caches Try to Be Transparent

- ❑ Software is (ideally) written to be oblivious to caches
 - Programmer should not have to worry about cache properties
 - Correctness isn't harmed regardless of cache properties

- ❑ However, the performance impact of cache affinity is quite high!
 - Performant software cannot be written in a completely cache-oblivious way

History of The Processor/Memory Performance Gap

Processor vs Memory Performance



1980: no cache in microprocessor;
1995 2-level cache

What is the Y-axis? Most likely normalized latency reciprocal

- 80386 (1985) :
Last Intel desktop CPU with no on-chip cache
(Optional on-board cache chip though!)
- 80486 (1989) : 4 KB on-chip cache
- Coffee Lake (2017) :
64 KiB L1 Per core
256 KiB L2 Per core
Up to 2 MiB L3 Per core (Shared)

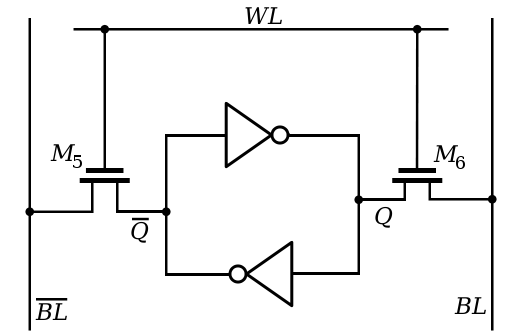
Why The Gap? SRAM vs. DRAM

❑ SRAM (Static RAM) – Register File, Cache

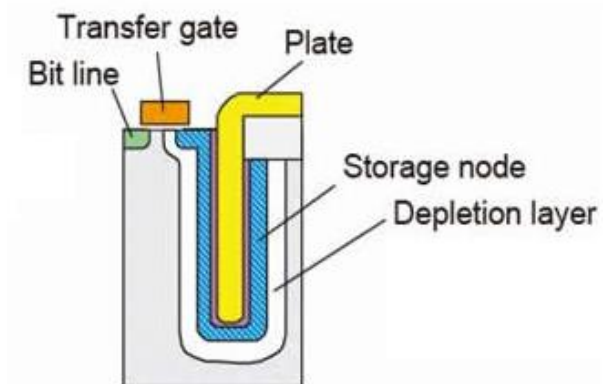
- Built using transistors, which processor logic is made of
- As fast as the rest of the processor

❑ DRAM (Dynamic RAM)

- Built using capacitors, which can hold charge for a short time
- Controller must periodically read all data and write it back (“Refresh”)
 - Hence, “Dynamic” RAM
- Requires fabrication process separate from processor
- Reading data from a capacitor is high-latency
 - EE topics involving sense amplifiers, which we won’t get into



Source: Inductiveload, from commons.wikimedia.org

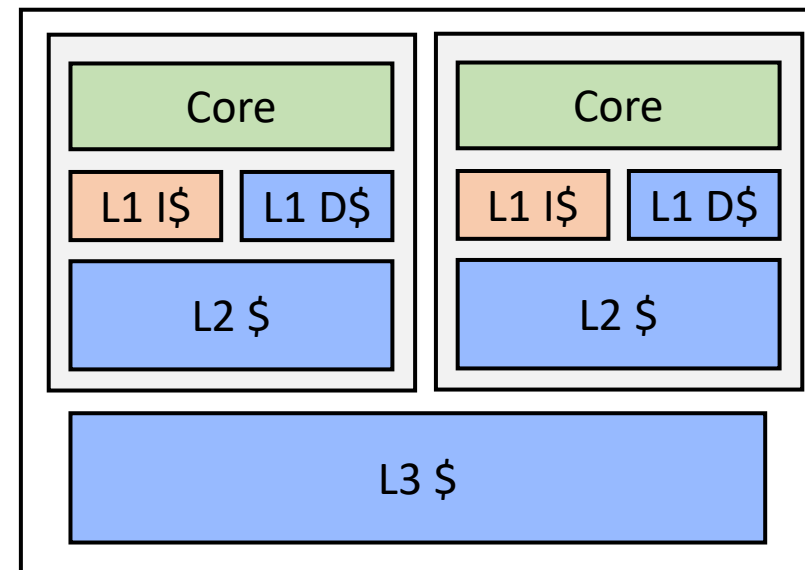


Note: Old, “trench capacitor” design

Multi-Layer Cache Architecture

Numbers from modern Xeon processors (Broadwell – Kaby lake)

Cache Level	Size	Latency (Cycles)
L1	64 KiB	< 5
L2	256 KiB	< 20
L3	~ 2 MiB per core	< 50

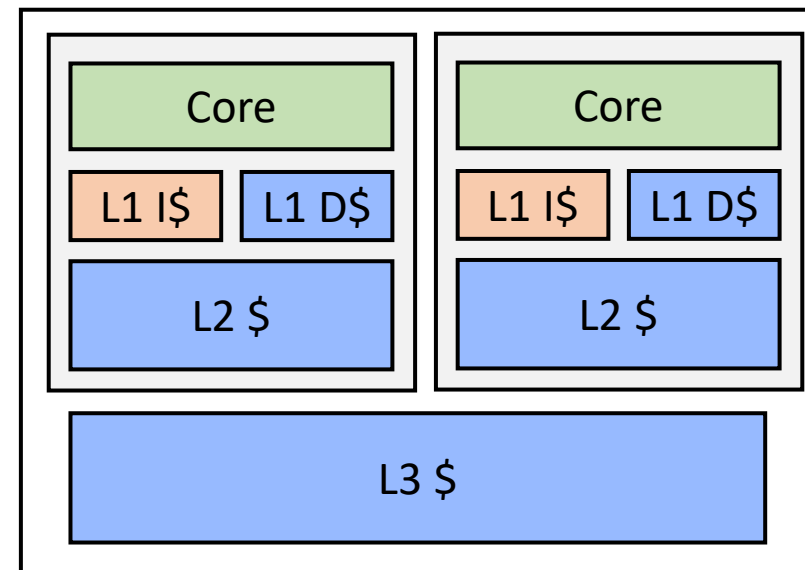


- ❑ Even with SRAM there is a size-performance trade-off
 - Not because the transistors are any different!
 - Cache management logic becomes more complicated with larger sizes
- ❑ L1 cache accesses can be hidden in the pipeline
 - Modern processors have pipeline depth of 14+
 - All others take a performance hit

Multi-Layer Cache Architecture

Numbers from modern Xeon processors (Broadwell – Kaby lake)

Cache Level	Size	Latency (Cycles)
L1	64 KiB	< 5
L2	256 KiB	< 20
L3	~ 2 MiB per core	< 50
DRAM	100s of GB	> 100*



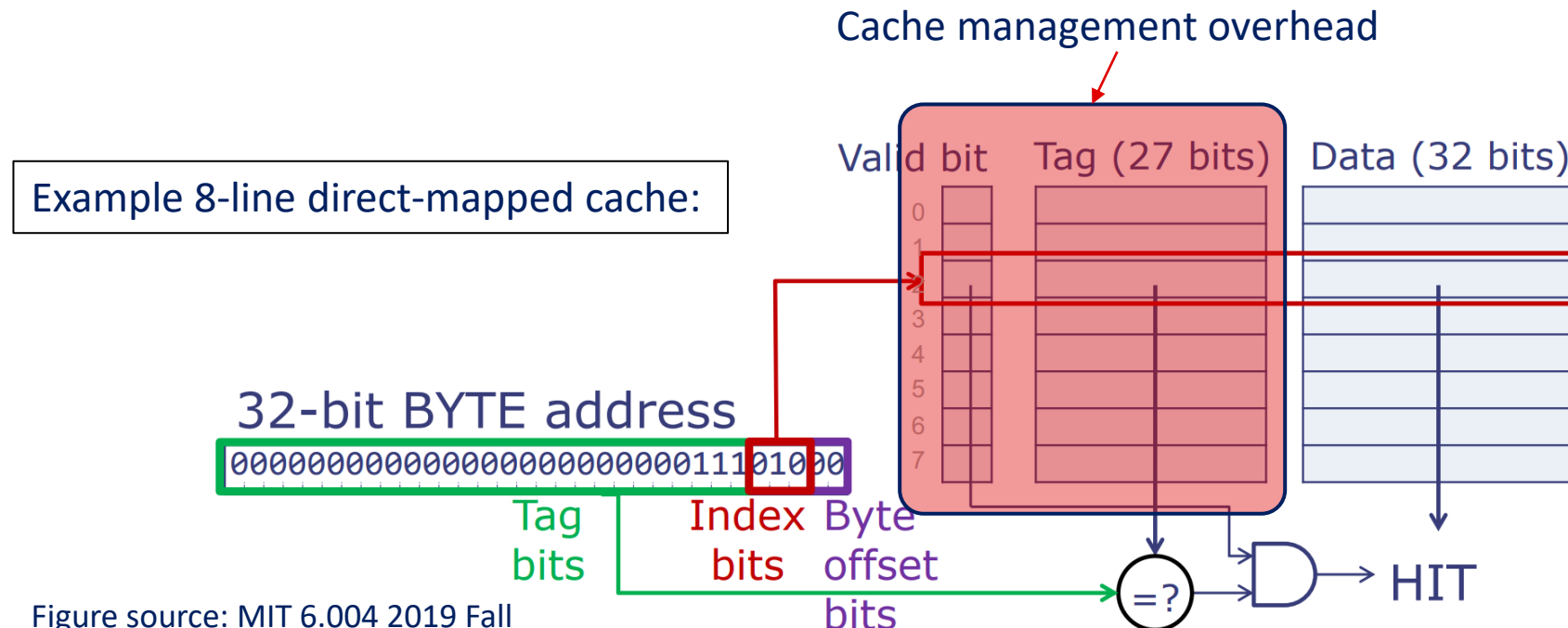
- ❑ *This is in an ideal scenario
 - Actual measurements could be multiple hundreds or thousands of cycles!
- ❑ DRAM systems are complicated entities themselves
 - Latency/Bandwidth of the same module varies immensely by situation...

Cache Line Unit of Management

- ❑ CPU Caches are managed in units of large “Cache Lines”
 - Typically 64 bytes in modern x86 processors
- ❑ Why not smaller units?
 - Word-size management is natural to reason about. Why not this?

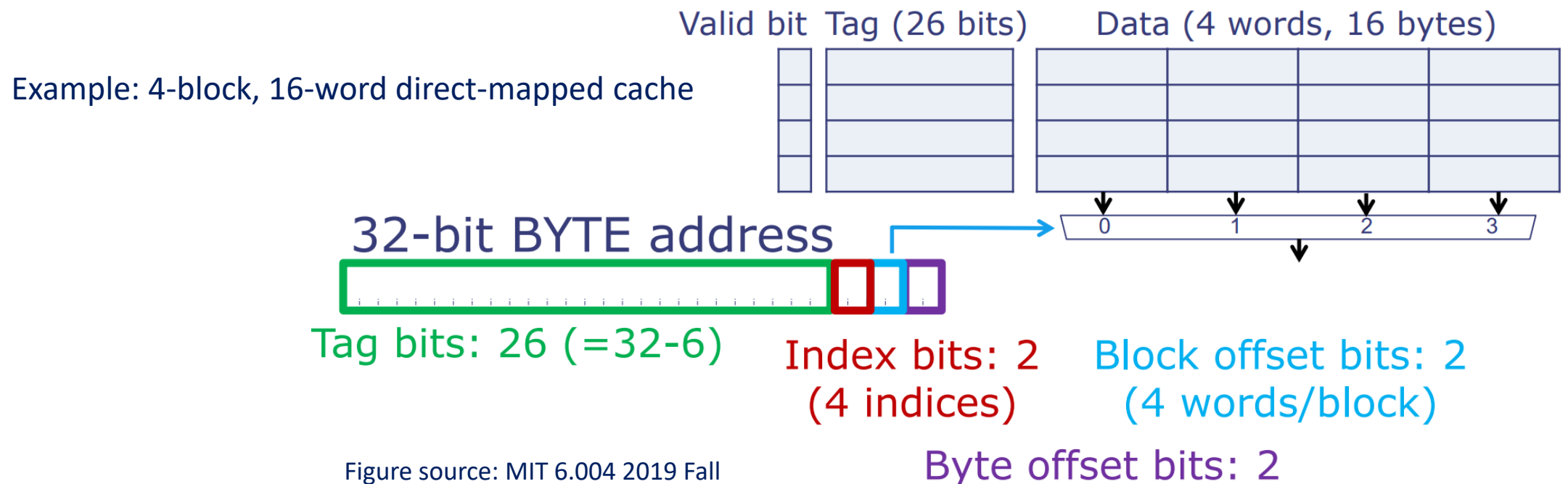
Reminder: Direct Mapped Cache Access

- For cache with 2^W cache lines
 - Index into cache with W address bits (the index bits)
 - Read out valid bit, tag, and data
 - If valid bit == 1 and tag matches upper address bits, cache hit!



Larger Cache Lines

- ❑ Reduce cache management overhead: Store multiple words per data line
 - Always fetch entire block (multiple words) from memory
 - **+ Advantage:** Reduces size of tag memory
 - **- Disadvantage:** Fewer indices in the cache -> Higher miss rate



Larger Cache Lines

- ❑ Caches are managed in *Cache Line* granularity
 - Typically 64 Bytes for modern CPUs
 - 64 Bytes == 16 4-byte integers
 - Balance of performance and on-chip SRAM usage
- ❑ Reading/Writing happens in cache line granularity
 - Read one byte not in cache -> Read all 64 bytes from memory
 - Write one byte -> Eventually write all 64 bytes to memory
 - Inefficient cache access patterns really hurt performance!

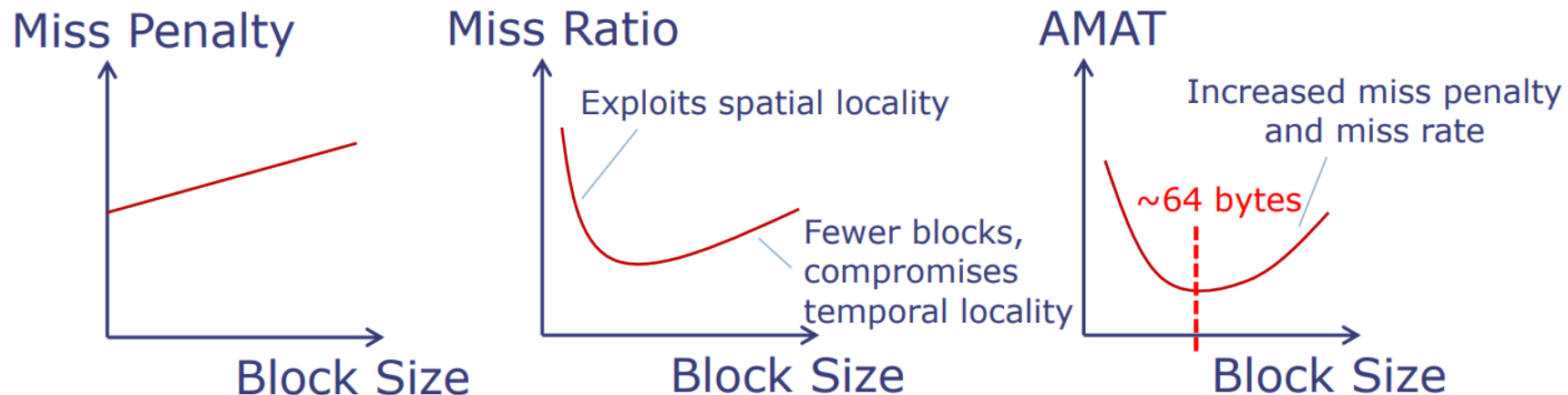
Block Size Trade-Offs

❑ Larger block sizes...

- Take advantage of spatial locality (also, DRAM is faster with larger blocks)
- Incur larger miss penalty since it takes longer to transfer the block from memory
- Can increase the average hit time (more logic) and miss ratio (less lines)

❑ AMAT (Average Memory Access Time)

- = $\text{HitTime} + \text{MissPenalty} * \text{MissRatio}$



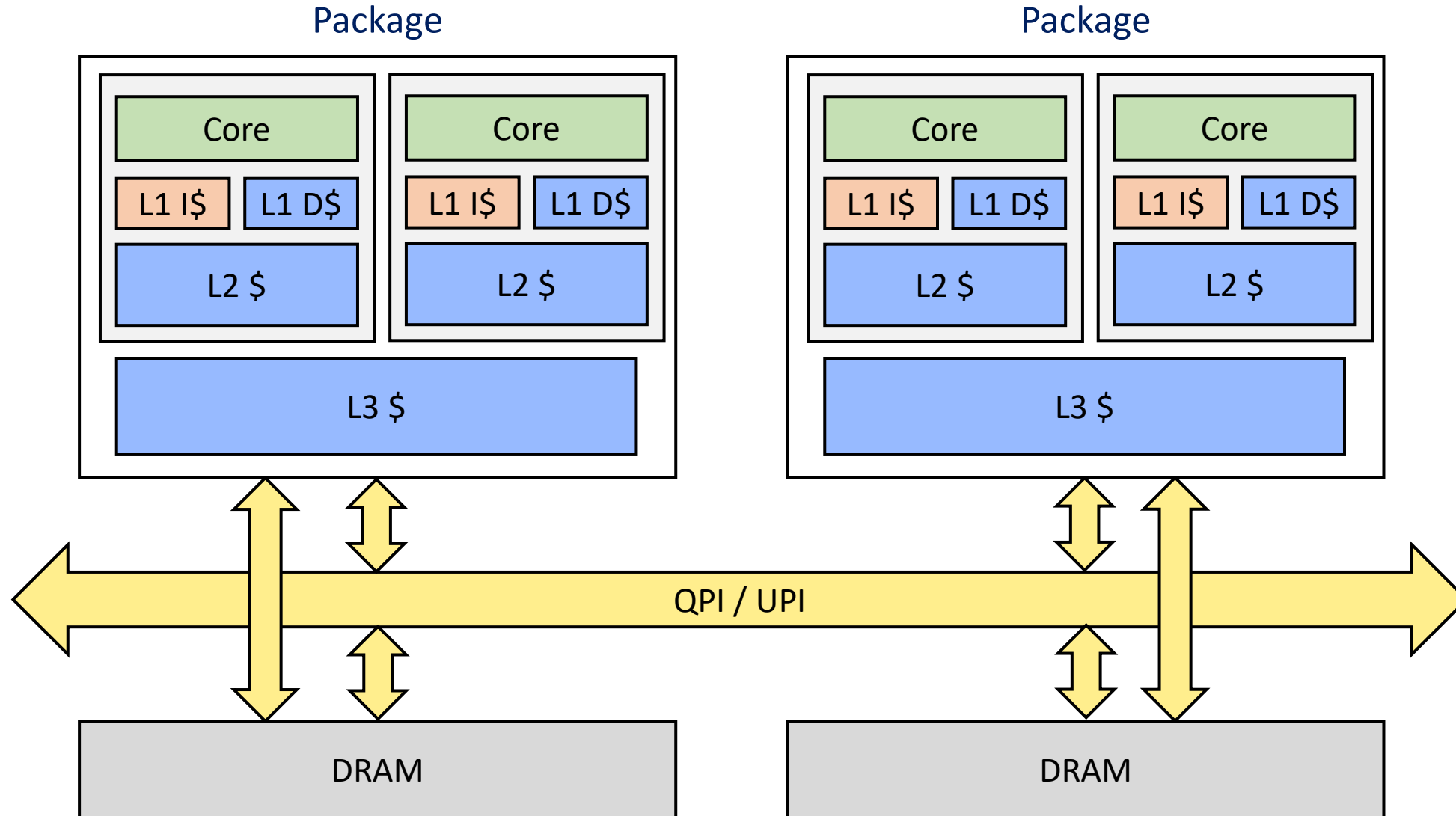
An Analytical Example: Two 4 KiB Caches

- ❑ 4-way set-associative, cache line size of 16 bytes
 - Each set == 64 bytes -> 64 sets
 - Assuming 32 bit addresses: 22 bit tag + valid + dirty = 24 bits per line
 - 768 bytes of overhead per 4 KiB cache
 - Total SRAM requirement: 4 KiB + 768 bytes = **4864** bytes

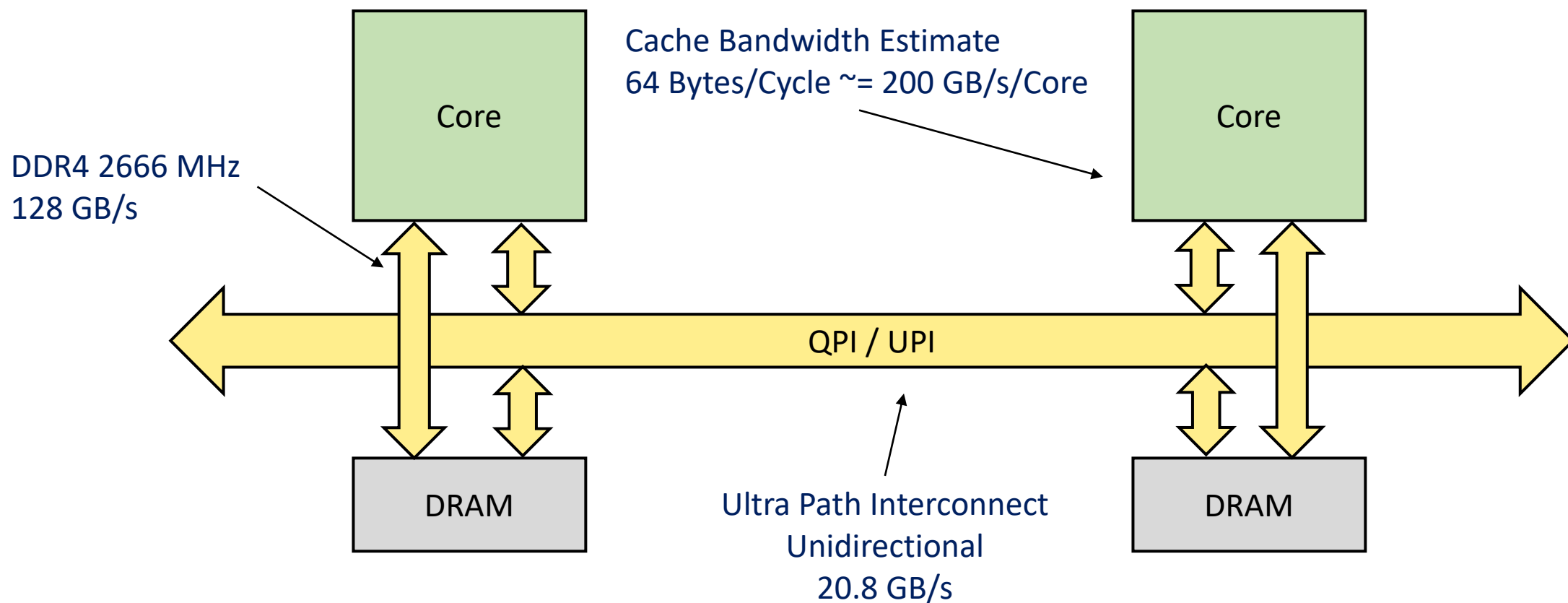
- ❑ Direct-mapped, cache line size of 4 bytes
 - Each line == 4 bytes -> 1024 lines
 - Assuming 32 bit addresses: 20 bit tag + valid + dirty = 22 bits per line
 - 2816 bytes of overhead per 4 KiB cache
 - Total SRAM requirement: 4 KiB + 2816 bytes = **6912** bytes

Memory System Architecture

Two packages make up a
NUMA (Non-Uniform Memory Access)
Configuration



Memory System Bandwidth Snapshot



Memory/PCIe controller used to be on a separate "North bridge" chip, now integrated on-die
All sorts of things are now on-die! Even network controllers! (Specialization!)

Reminder: Cache Coherency

- ❑ Cache coherency
 - Informally: Read to each address must return the most recent value
 - Typically: All writes must be visible at some point, and in proper order
- ❑ Coherency protocol implemented between each core's private caches
 - MSI, MESI, MESIF, ...
 - Won't go into details here
- ❑ Simply put:
 - When a core writes a cache line
 - All other instances of that cache line needs to be invalidated
- ❑ Emphasis on *cache line*

Cache Prefetching

- ❑ CPU speculatively prefetches cache lines
 - While CPU is working on the loaded 64 bytes, 64 more bytes are being loaded
- ❑ Hardware prefetcher is usually not very complex/smart
 - Sequential prefetching (N lines forward or backwards)
 - Strided prefetching
- ❑ Programmer-provided prefetch hints
 - `__builtin_prefetch(address, r/w, temporal locality?);` for GCC
 - Will generate prefetch instructions if available on architecture

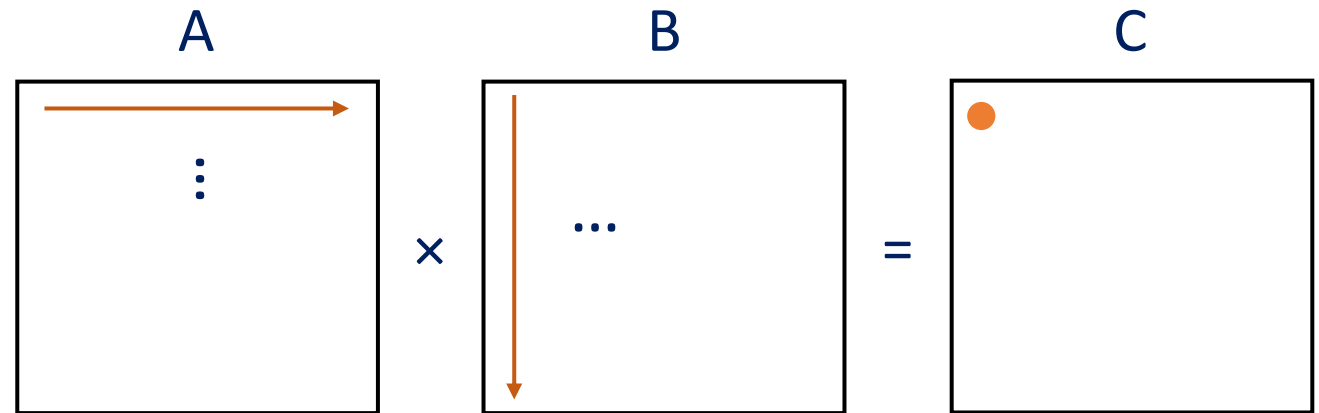
Now That's Out of The Way...

Cache Efficiency Issue #1: Cache Line Size

Matrix Multiplication and Caches

- ❑ Multiplying two NxN matrices ($C = A \times B$)

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



2048*2048 on a i5-7400 @ 3 GHz using GCC -O3 = 63.19 seconds

is this fast?

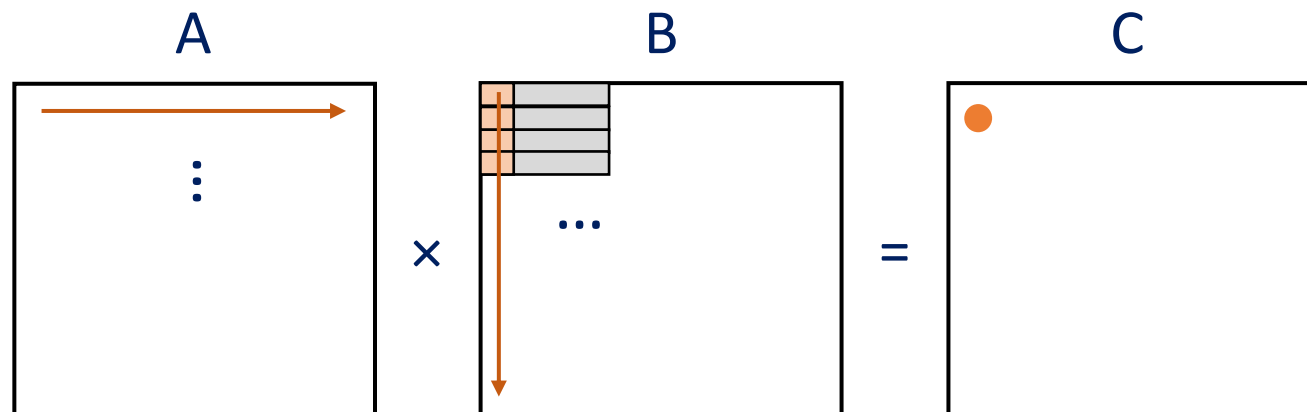
Whole calculation requires $2K * 2K * 2K = 8$ Billion floating-point mult + add
At 3 GHz, ~5 seconds just for the math. Over 1000% overhead!

Cache Efficiency Issue #1: Cache Line Size

Matrix Multiplication and Caches

- ❑ Column-major access makes inefficient use of cache lines
 - A 64 Byte block is read for each element loaded from B
 - 64 bytes read from memory for each 4 useful bytes
- ❑ Shouldn't caching fix this? Unused bits should be useful soon!
 - 64 bytes x 2048 = 128 KB ... Already overflows L1 cache (~32 KB)

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

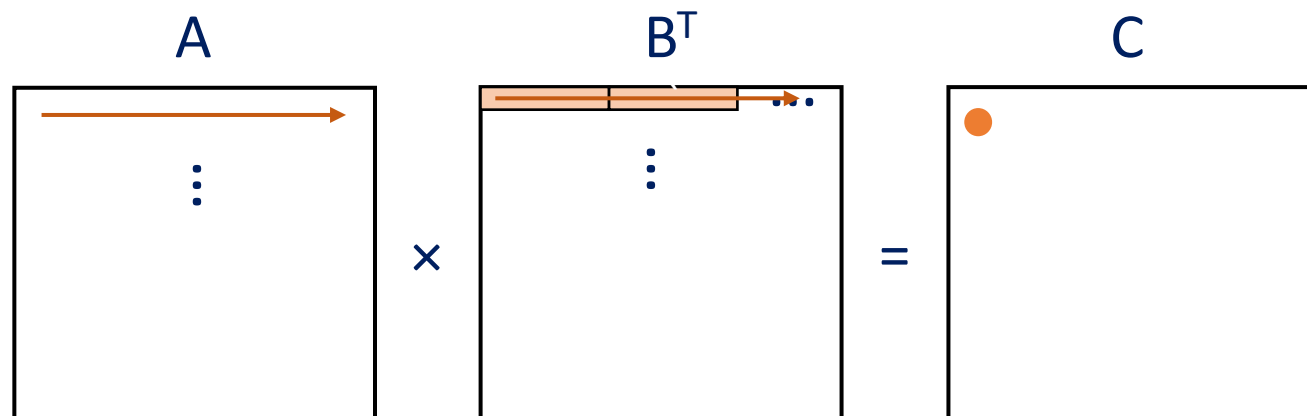


Cache Efficiency Issue #1: Cache Line Size

Matrix Multiplication and Caches

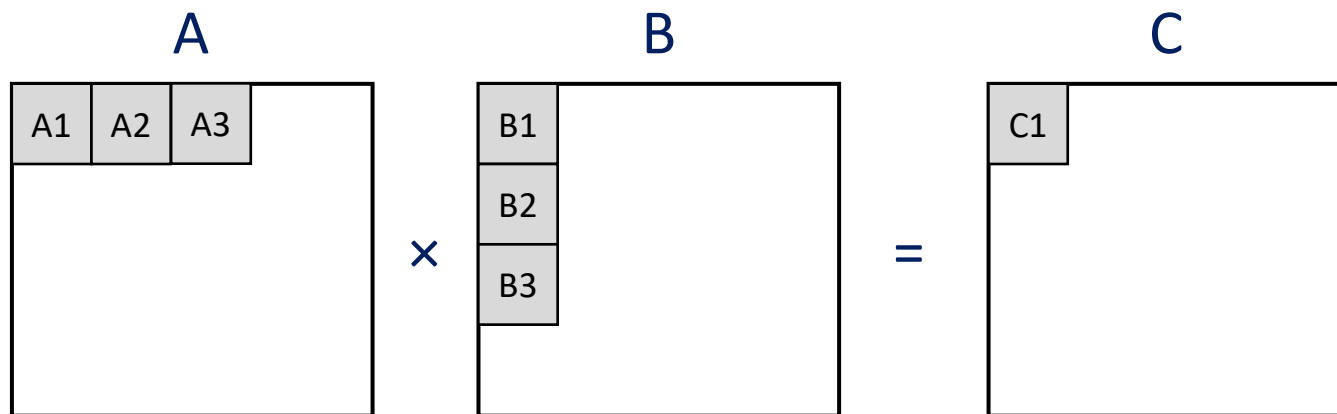
- ❑ One solution: Transpose B to match cache line orientation
 - Does transpose add overhead? Not very much as it only scans B once
- ❑ Drastic improvements!
 - Before: 63.19s
 - After: 10.39s ... 6x improvement!
 - But still not quite ~5s

```
for (i = 0 to N)
  for (j = 0 to N)
    for (k = 0 to N)
      C[i][j] += A[i][k] * Bt[j][k]
```



Cache Efficiency Issue #2: Capacity Considerations

- ❑ Performance is best when working set fits into cache
 - But as shown, even 2048×2048 doesn't fit in cache
 - $\rightarrow 2048 * 2048 * 2048$ elements read from memory for matrix B
- ❑ Solution: Divide and conquer! – Blocked matrix multiply
 - For block size $32 \times 32 \rightarrow 2048 * 2048 * (2048/32)$ reads



$$C1 \text{ sub-matrix} = A1 \times B1 + A2 \times B2 + A3 \times B3 \dots$$

Blocked Matrix Multiply Evaluations

Benchmark	Elapsed (s)	Normalized Performance
Naïve	63.19	1
Transposed	10.39	6.08
Blocked Transposed	7.35	8.60

- ❑ Blocked Transposed bottlenecked by computation
 - Peak theoretical FLOPS for my processor running at 3 GHz \approx 3 GFLOPS
 - 7.35s for matrix multiplication \approx 2.18 GFLOPS
 - Not bad, considering need for branches and other instructions!
 - L1 cache access now optimized, but not considers larger caches
- ❑ This chart will be further extended in the next lectures
 - Normalized performance will reach 57 (\approx 1 second elapsed)

Writing Cache Line Friendly Software

- ❑ (Whenever possible) use data in coarser-granularities
 - Each access may load 64 bytes into cache, make use of them!
 - e.g., Transposed matrix B in matrix multiply, blocked matrix multiply
- ❑ Many profilers will consider the CPU “busy” when waiting for cache
 - Can't always trust “CPU utilization: 100%”

Aside:

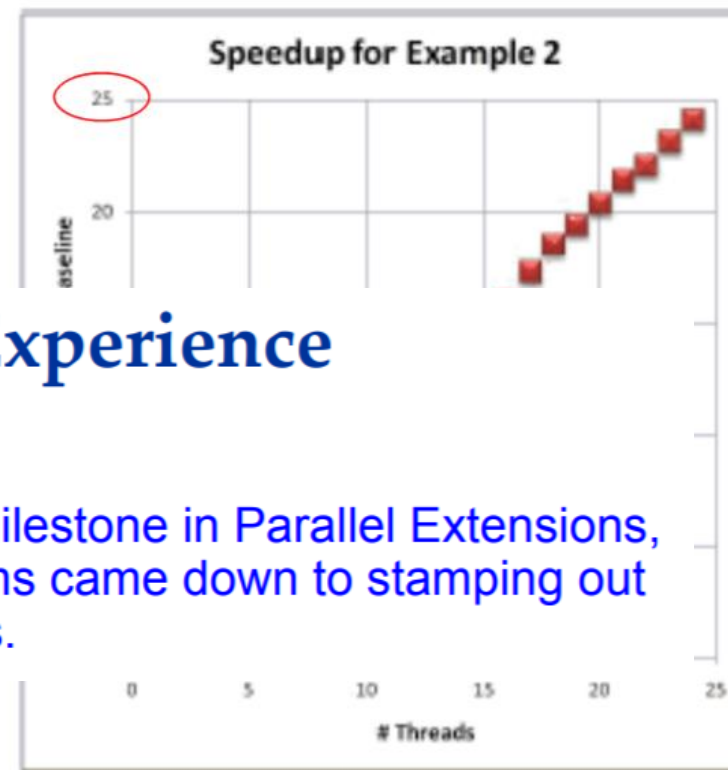
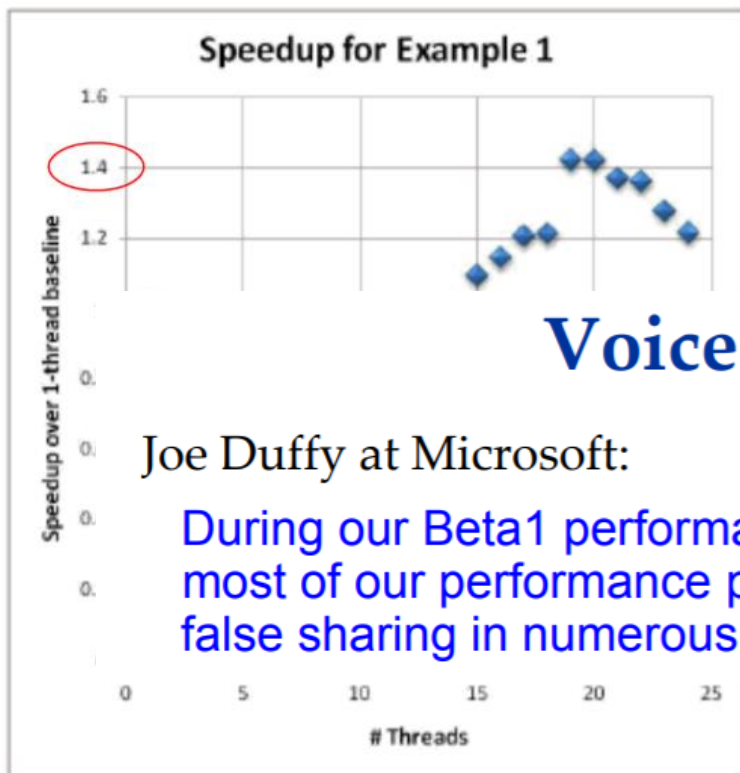
Object-Oriented Programming And Caches

- ❑ OOP wants to collocate all data for an entity in a class/struct
 - All instance variables are located together in memory
- ❑ Cache friendly OOP
 - All instance variables are accessed whenever an instance is accessed
- ❑ Cache unfriendly OOP
 - Only a small subset of instance variables are accessed per instance access
 - e.g., a “for” loop checking the “valid” field of all entities
 - 1 byte accessed per cache line read!
- ❑ Non-OOP solution: Have a separate array for “valid”s
 - Is this a desirable solution? Maybe...

Cache Efficiency Issue #3: False Sharing

- ❑ Different memory locations, written to by different cores, mapped to same cache line
 - Core 1 performing “results[0]++;”
 - Core 2 performing “results[1]++;”
- ❑ Remember cache coherence
 - Every time a cache is written to, all other instances need to be invalidated!
 - “results” variable is ping-ponged across cache coherence every time
 - Bad when it happens on-chip, terrible over processor interconnect (QPI/UPI)
- ❑ Simple solution: Store often-written data in local variables

Removing False Sharing



Voice of Experience

Joe Duffy at Microsoft:

During our Beta1 performance milestone in Parallel Extensions, most of our performance problems came down to stamping out false sharing in numerous places.

With False Sharing

Without False Sharing

Aside: Non Cache-Related Optimizations: Loop Unrolling

- ❑ Increase the amount of work per loop iteration
 - Improves the ratio between computation instructions and branch instructions
 - Compiler can be instructed to automatically unroll loops
 - Increases binary size, because unrolled iterations are now duplicated code

Normal loop	After loop unrolling
<pre>int x; for (x = 0; x < 100; x++) { delete(x); }</pre>	<pre>int x; for (x = 0; x < 100; x += 5) { delete(x); delete(x + 1); delete(x + 2); delete(x + 3); delete(x + 4); }</pre>

Source: Wikipedia "Loop unrolling"

Aside: Non Cache-Related Optimizations: Function Inlining

- ❑ A small function called very often may be bottlenecked by call overhead
- ❑ Compiler copies the instructions of a function into the caller
 - Removes expensive function call overhead (stack management, etc)
 - Function can be defined with “inline” flag to hint the compiler
 - “inline int foo()”, instead of “int foo()”
- ❑ Personal anecdote
 - Inlining a key (very small) kernel function resulted in a 4x performance boost

Issue #4

Instruction Cache Effects

- ❑ Instructions are also stored in cache
 - L1 cache typically has separate instances for instruction and data caches
 - In most x86 architectures, 32 KiB each
 - L2 onwards are shared
 - Lots of spatial locality, so miss rate is usually very low
 - On SPEC, ~2% at L1
 - But adversarial examples can still thrash the cache
- ❑ Instruction cache often has dedicated prefetcher
 - Understands concepts of branches and function calls
 - Prefetches blocks of instructions without branches

Optimizing Instruction Cache

❑ Instruction cache misses can effect performance

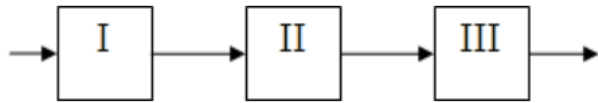
- “Linux was routing packets at ~30Mbps [wired], and wireless at ~20. Windows CE was crawling at barely 12Mbps wired and 6Mbps wireless.
- [...] After we changed the routing algorithm to be more cache-local, we started doing 35Mbps [wired], and 25Mbps wireless – 20% better than Linux.
– Sergey Solyanik, Microsoft
- [By organizing function calls in a cache-friendly way, we] achieved a 34% reduction in instruction cache misses and a 5% improvement in overall performance.
-- Mircea Livadariu and Amir Kleen, Freescale

Improving Instruction Cache Locality #1

- ❑ Careful with loop unrolling
 - They reduce branching overhead, but reduces effective I\$ size
 - When gcc's -O3 performs slower than -O2, this is usually what's happening
- ❑ Careful with function inlining
 - Inlining is typically good for very small* functions
 - A rarely executed path will just consume cache space if inlined
- ❑ Move conditionals to front as much as possible
 - Long paths of no branches good fit with instruction cache/prefetcher

Improving Instruction Cache Locality #2

- ❑ Organize function calls to create temporal locality



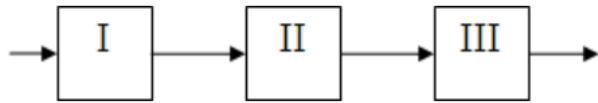
```
for (i=0;i<N;i++)
{
    temp=stage_I(input[i]);
    temp=stage_II(temp);
    output[i]= stage_III(temp);
}
```

If the functions `stage_I`, `stage_II`, and `stage_III` are sufficiently large, their instructions will thrash the instruction cache!

Baseline: Sequential algorithm

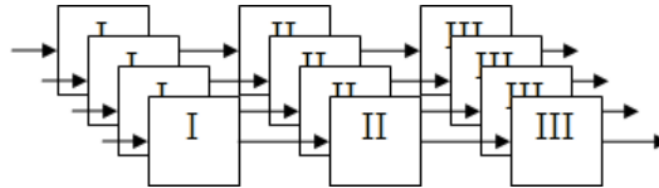
Improving Instruction Cache Locality #2

- ❑ Organize function calls to create temporal locality



```
for (i=0;i<N;i++)  
{  
    temp=stage_I(input[i]);  
    temp=stage_II(temp);  
    output[i]= stage_III(temp);  
}
```

Baseline: Sequential algorithm



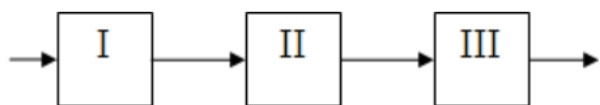
```
for (i=0;i<N;i++)  
    temp[i]=stage_I(input[i]);  
for (i=0;i<N;i++)  
    temp[i]=stage_II(temp[i]);  
for (i=0;i<N;i++)  
    output[i]= stage_III(temp[i]);
```

Ordering changed for
cache locality

New array “temp” takes up
space. N could be large!

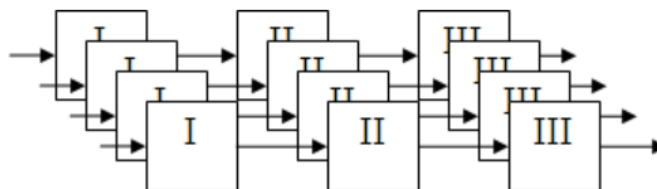
Improving Instruction Cache Locality #2

- Organize function calls to create temporal locality



```
for (i=0;i<N;i++)  
{  
    temp=stage_I(input[i]);  
    temp=stage_II(temp);  
    output[i]= stage_III(temp);  
}
```

Baseline: Sequential algorithm



```
for (i=0;i<N;i++)  
    temp[i]=stage_I(input[i]);  
for (i=0;i<N;i++)  
    temp[i]=stage_II(temp[i]);  
for (i=0;i<N;i++)  
    output[i]= stage_III(temp[i]);
```

Ordering changed for
cache locality

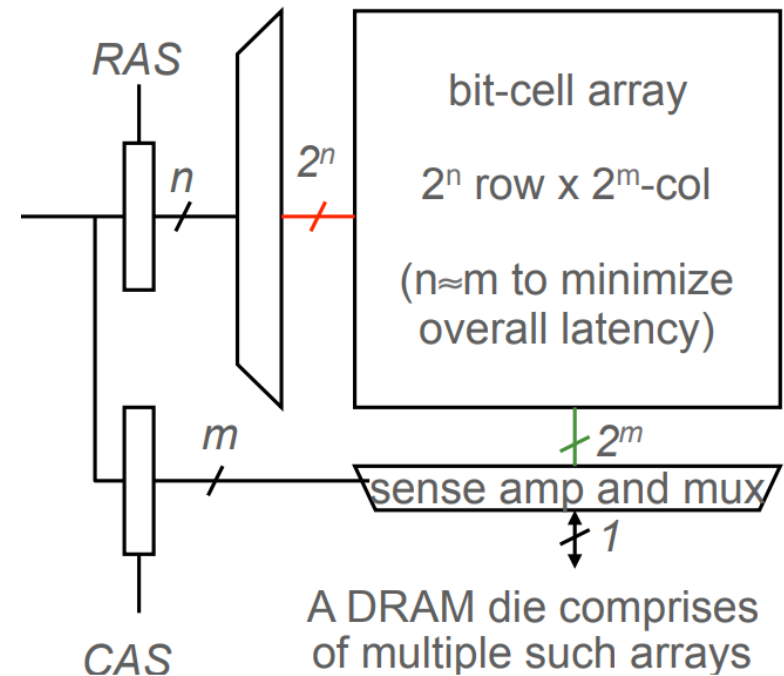
```
for (j=0;j<N;j+=M)  
{  
    for (i=0;i<M;i++)  
        temp[i]=stage_I(input[j+i]);  
    for (i=0;i<M;i++)  
        temp[i]=stage_II(temp[j+i]);  
    for (i=0;i<M;i++)  
        output[i]= stage_III(temp[j+i]);  
}
```

Balance to reduce
memory footprint

Questions?

Some Details On DRAM

- ❑ DRAM cell access latency is very high
 - Electrical characteristics of the capacitors and the circuitry to read their state
 - To mitigate this, accesses are done at a very coarse granularity
 - Might as well spend 10 ns to read 8 KiB, instead of only 4 bytes
- ❑ DRAM is typically organized into a rectangle (rows, columns), called a “bank”
 - Reduces addressing logic, which is a high overhead in such dense memory
 - Whole row must be read whenever data in new row is accessed
 - As of 2022, typical row size ~8 KB



Some Details On DRAM

- ❑ Accessed row temporarily stored in DRAM “row buffer”
 - Fast when accessing data in same row
 - Much slower when accessing small data across rows
- ❑ The off-chip memory system is also hierarchical
 - A DRAM chip consists of multiple banks
 - A DRAM card consists of multiple chips
 - A memory system (typically) consists of multiple DRAM cards
- ❑ Row buffer exists for each bank
 - Total size of all row buffers in a system is quite large
 - Inter-bank parallelism

Exploiting Inter-Bank Parallelism

- ❑ Ideally, accesses should be grouped within a row
- ❑ When this is not possible, access to the same bank must be avoided
 - Access cannot be serviced until the previous (high-latency) access is done
- ❑ The processor hardware tries to automatically handle this via address mapping
 - LSB of the address used for column index
 - MSB of the address used for row index
 - Everything in the middle spread across card/chip/bank/...

Questions?