The memory hierarchy: Cache replacement, hierarchies, and optimization

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Locality: How to create illusion of fast access to capacious data

From the perspective of memory hierarchy, locality is using the data in at any particular level more frequently than accessing storage at next slower level.

First, let's experience the puzzling effect of locality in `sumArray.c`

- sumArrayRows()
- sumArrayCols()

Well-written programs maximize locality

- Spatial locality
- Temporal locality

Spatial locality

```
1  double dotProduct (
2    double a[N],
3    double b[N],
4  ) {
5    double sum = 0.0;
6    for(size_t i=0; i<N; i++){
7      sum += a[i] * b[i];
8    }
9    return sum;
10 }
```

Spatial locality

- Programs tend to access adjacent data.
- Example: stride 1 memory access in `a` and `b`.

Temporal locality

```
1  double dotProduct (
2    double a[N],
3    double b[N],
4  ) {
5    double sum = 0.0;
6    for(size_t i=0; i<N; i++){
7      sum += a[i] * b[i];
8    }
9    return sum;
10 }
```

Temporal locality

- Programs tend to access data over and over.
- Example: `sum` gets accessed N times in iteration.

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Cache placement policy (how to find data at address for read and write hit)

Several designs for caches

- Fully associative cache
- Direct-mapped cache
- N-way set-associative cache

Cache design options use m-bit memory addresses differently

- t-bit tag
- s-bit set index
- b-bit block offset

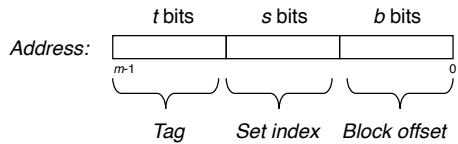


Figure: Memory addresses. Image credit CS:APP

Direct-mapped cache

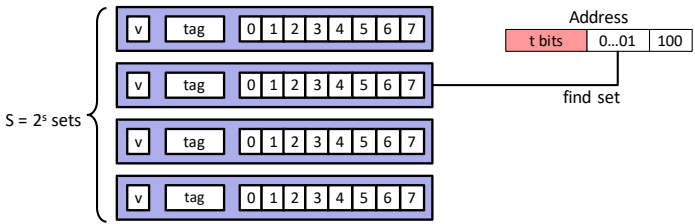


Figure: Direct-mapped cache. Image credit CS:APP

m-bit memory address split into:

- *t*-bit tag
- *s*-bit set index
- *b*-bit block offset

Direct-mapped cache

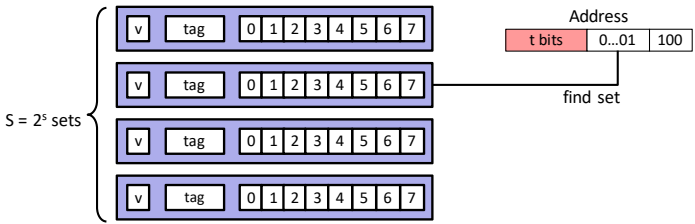


Figure: Direct-mapped cache. Image credit CS:APP

b-bit block offset

- here, $b = 3$
- The number of bytes in a block is $B = 2^b = 2^3 = 8$
- A block is the minimum number of bytes that can be cached
- Good for capturing spatial locality, short strides

Direct-mapped cache

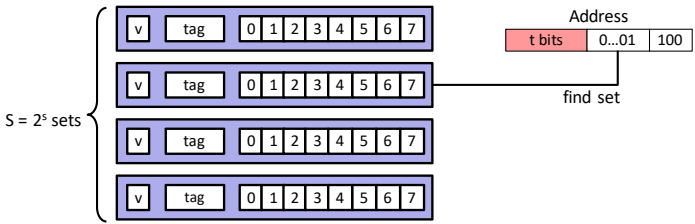


Figure: Direct-mapped cache. Image credit CS:APP

s-bit set index

- here, $s = 2$
- The number of sets in cache is $S = 2^s = 2^2 = 4$
- A hash function that limits exactly where a block can go
- Good for further increasing ability to exploit spatial locality, short strides

Direct-mapped cache

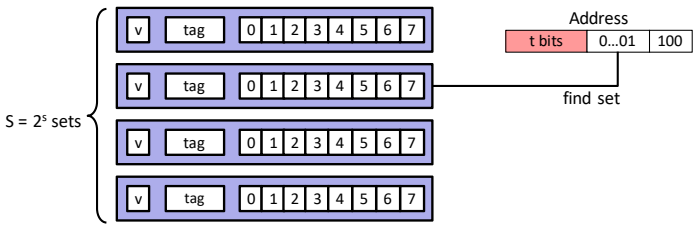


Figure: Direct-mapped cache. Image credit CS:APP

t-bit tag

- here, $t = m - s - b = m - 2 - 3$
- When CPU wants to read from or write to memory, all t-bits in tag need to match for read/write hit.

Direct-mapped cache

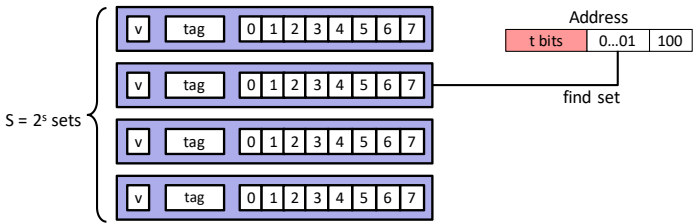


Figure: Direct-mapped cache. Image credit CS:APP

Direct mapping

- In direct-mapped cache, blocks can go into only one of $E = 1$ way

Direct-mapped cache

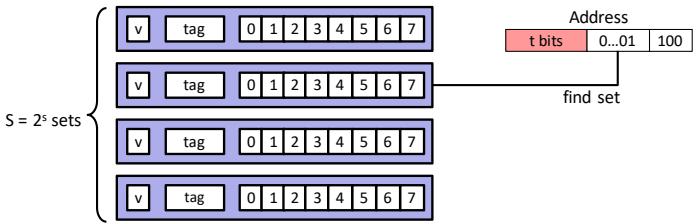


Figure: Direct-mapped cache. Image credit CS:APP

Capacity of cache

- Total capacity of fully associative cache in bytes:
  $C = SEB = 2^s * E * 2^b$
- Here, $C = 2^s * E * 2^b = 2^2 * 1 * 2^3 = 32$ bytes

Direct-mapped cache

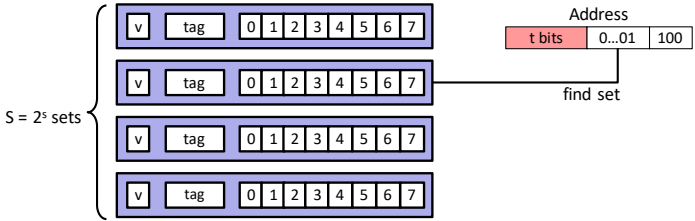


Figure: Direct-mapped cache. Image credit CS:APP

Strengths

- Simple to implement.
- No need to search across tags.

Weaknesses

- Can lead to surprising thrashing of cache with unfortunate access patterns.
- Unexpected conflict misses independent of cache capacity.

E-way set-associative cache

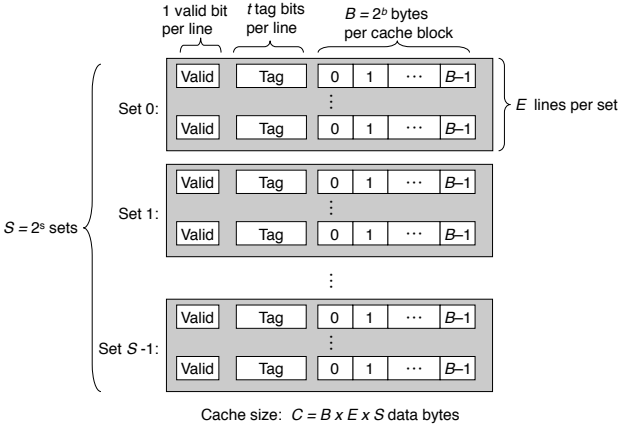


Figure: Direct-mapped cache. Image credit CS:APP

Strengths

- Blocks can go into any of E-ways, increases ability to support temporal locality, thereby increasing hit rate.
- Only need to search across E tags. Avoids costly searching across all valid tags.
- Avoids conflict misses due to unfortunate access patterns.

E-way set-associative cache

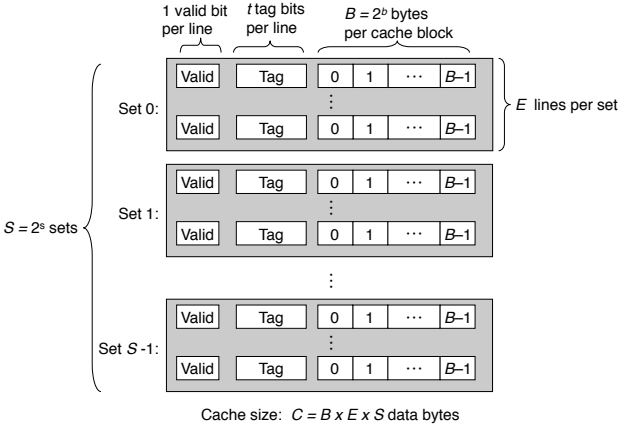
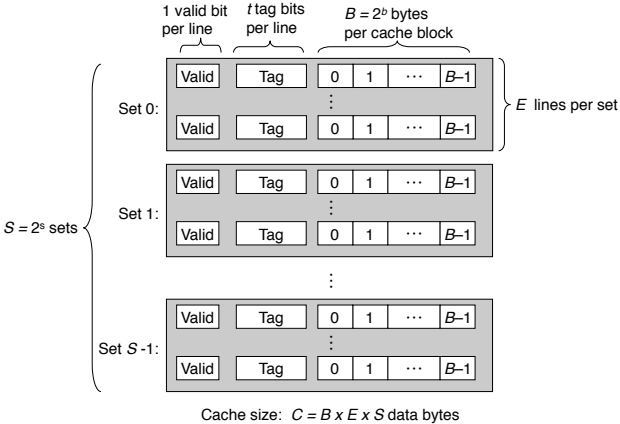


Figure: Direct-mapped cache. Image credit CS:APP

Used in practice in, e.g., a recent Intel Core i7:

- $C = 32$KB L1 data cache per core
- $S = 64 = 2^6$ sets/cache ($s = 6$ bits)
- $E = 8 = 2^3$ ways/set
- $B = 64 = 2^6$ bytes/block ($b = 6$ bits)
- $C = S * E * B$
- Assuming memory addresses are $m = 48$, then tag size $t = m - s - b = 48 - 6 - 6 = 36$ bits.

E-way set-associative cache



Let's see textbook slides for a simulation

Figure: Direct-mapped cache. Image credit CS:APP

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Cache hits

Memory access is serviced from cache

- Hit rate = $\frac{\text{Numberofhits}}{\text{Numberofmemoryaccesses}}$
- Hit time: latency to access cache (4 cycles for L1, 10 cycles for L2)

Cache misses: metrics

Memory access cannot be serviced from cache

- Miss rate = $\frac{\text{Numberofmisses}}{\text{Numberofmemoryaccesses}}$
- Miss penalty (miss time): latency to access next level cache or memory (up to 200 cycles for memory).
- Average memory access time = hit time + miss rate $\times$ miss penalty

Cache misses: Classification

Compulsory misses

- First access to a block of memory will miss because cache is cold.

Conflict misses

- Multiple blocks map (hash) to the same cache set.
- Fully associative caches have no such conflict misses.

Capacity misses

- Occurs when the set of active cache blocks (working set) is larger than the cache.
- Direct mapped caches have no such capacity misses.

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Direct-mapped cache

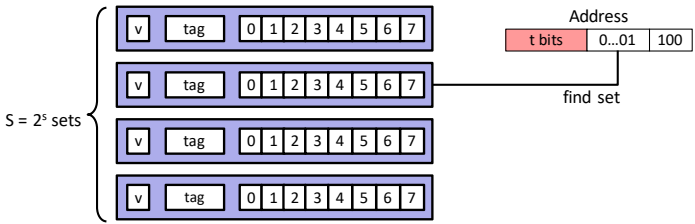

Figure: Direct-mapped cache. Image credit CS:APP

No need for replacement policy

- The number of sets in cache is $S = 2^s = 2^2 = 4$.
- A hash function that limits exactly where a block can go.
- In direct-mapped cache, blocks can go into only one of $E = 1$ way.
- No cache replacement policy is needed.

Associative caches

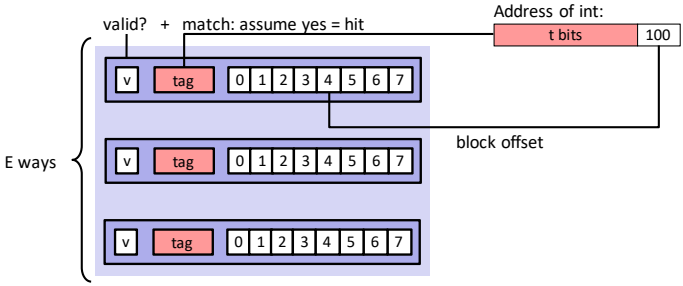

Figure: Fully associative cache. Image credit CS:APP

Needs replacement policy

- Blocks can go into any of E ways
- Here, $E = 3$
- Good for capturing temporal locality.
- If all ways/lines/blocks are occupied, and a cache miss happens, which way/line/block will be the victim and get evicted for replacement?

Cache replacement policies for associative caches

FIFO: First-in, first-out

- Evict the cache line that was placed the longest ago.
- Each cache set essentially becomes limited-capcity queue.

LRU: Least Recently Used

- Evict the cache line that was last accessed longest ago.
- Needs a counter on each cache line, and/or a global counter (e.g., program counter).

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Policies for writes from CPU to memory

How to deal with write-hit?

- **Write-through.** Simple. Writes update both cache and memory. Costly memory bus traffic.
- **Write-back.** Complex. Writes update only cache and set a dirty bit; memory updated only upon eviction. Reduces memory bus traffic. (For multi-core CPUs, motivates complex cache coherence protocols)

How to deal with write-miss?

- **No-write-allocate.** Simple. Write-misses do not load block into cache. But write-misses are not mitigated via cache support.
- **Write-allocate.** Complex. Write-misses will load block into cache.

Typical designs:

- **Simple:** write-through + no-write-allocate.
- **Complex:** write-back + write-allocate.

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Multilevel cache hierarchies

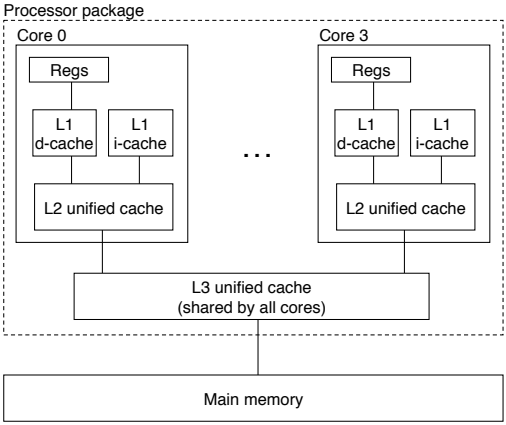

Figure: Intel Core i7 cache hierarchy. Image credit CS:APP

Small fast caches nested inside large slow caches

- L1 data and instruction cache: 32KB, 64 set, 8-way associative, 64B block, 4 cycle latency.
- L2 cache: 256KB, 512 set, 8-way associative, 64B block, 10 cycle latency.
- L3 cache: 8MB, 8192 set, 16-way associative, 64B block, 40-75 cycle latency.

Notice how latency cost increases as E-way associativity increases.

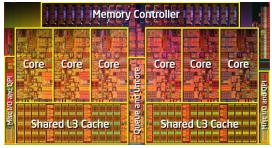

Figure: Intel 2020 Gulftown die shot. Image credit AnandTech

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Cache-friendly code

Algorithms can be written so that they work well with caches

- Maximize hit rate
- Minimize miss rate
- Minimize eviction counts

Do so by:

- Increasing spatial locality.
- Increasing temporal locality.

Advanced optimizing compilers can automatically make such optimizations

- GCC optimizations
- `https://gcc.gnu.org/onlinedocs/gcc/Optimize-Options.html`
- `-floop-interchange`
- `-floop-block`

Loop interchange

Refer to textbook slides on "Rearranging loops to improve spatial locality"

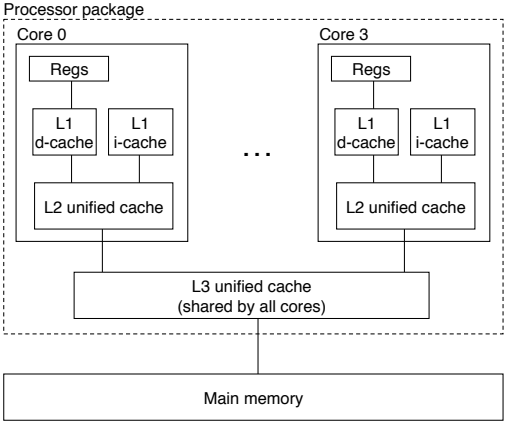
- Loop interchange increases spatial locality.
- In PA5, fourth part "cacheBlocking" you can explore the impact of this on matrix multiplication.
- In practice, programmers do not have to worry about this optimization.
- Optimized automatically in GCC by compiler flag `-floop-interchange` and `-O3`.

Cache blocking

Refer to textbook slides on "Using blocking to improve temporal locality"

- Cache blocking increases temporal locality.
- In PA5, fourth part "cacheBlocking" you can explore the impact of this on matrix multiplication.
- In practice, programmers do not have to worry about this optimization.
- Optimized automatically in GCC by compiler flag `-floop-block`. But it is not part of default optimizations such as `-O3` so you have to remember to set it.

Multilevel cache hierarchies



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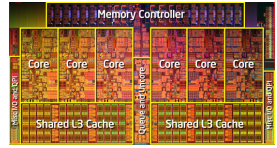

Figure: Intel 2020 Gulftown die shot. Image credit AnandTech

Cache oblivious algorithms

The challenge in writing code / compiling programs to take advantage of caches:

- Programmers do not easily have information about target machine.
- Compiling binaries for every envisioned target machine is costly.

Matrix transpose baseline algorithm: iteration

$$\mathbf{A} = \begin{bmatrix} a_{0,0} & a_{0,1} & a_{0,2} & a_{0,3} \\ a_{1,0} & a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,0} & a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,0} & a_{3,1} & a_{3,2} & a_{3,3} \end{bmatrix}$$

$$\mathbf{B} = \mathbf{A}^{\mathsf{T}} = \begin{bmatrix} a_{0,0} & a_{1,0} & a_{2,0} & a_{3,0} \\ a_{0,1} & a_{1,1} & a_{2,1} & a_{3,1} \\ a_{0,2} & a_{1,2} & a_{2,2} & a_{3,2} \\ a_{0,3} & a_{1,3} & a_{2,3} & a_{3,3} \end{bmatrix}$$

```
1 for ( size_t i=0; i<n; i++ ) {
2   for ( size_t j=0; j<n; j++ ) {
3     B[ j*n + i ] = A[ i*n + j ];
4   }
5 }
```

Matrix transpose via recursion

$$\mathbf{A} = \begin{bmatrix} A_{0,0} & A_{0,1} \\ A_{1,0} & A_{1,1} \end{bmatrix} = \left[\begin{array}{cc|cc} a_{0,0} & a_{0,1} & a_{0,2} & a_{0,3} \\ a_{1,0} & a_{1,1} & a_{1,2} & a_{1,3} \\ \hline a_{2,0} & a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,0} & a_{3,1} & a_{3,2} & a_{3,3} \end{array}\right]$$

$$\mathbf{B} = \mathbf{A}^{\mathsf{T}} = \begin{bmatrix} A_{0,0}^{\mathsf{T}} & A_{1,0}^{\mathsf{T}} \\ A_{0,1}^{\mathsf{T}} & A_{1,1}^{\mathsf{T}} \end{bmatrix} = \left[\begin{array}{cc|cc} a_{0,0} & a_{1,0} & a_{2,0} & a_{3,0} \\ a_{0,1} & a_{1,1} & a_{2,1} & a_{3,1} \\ \hline a_{0,2} & a_{1,2} & a_{2,2} & a_{3,2} \\ a_{0,3} & a_{1,3} & a_{2,3} & a_{3,3} \end{array}\right]$$

Strategy:

- Divide and conquer large matrix to transpose into smaller transpositions.
- After some recursion, problem will fit well inside cache capacity.
- Once enough locality exists withing subroutine, switch to plain iterative approach.

Advantages:

- No need to know about cache capacity and parameters beforehand.
- Works well with deep multilevel cache hierarchies: different amounts of locality for each cache level.

Memory hierarchy implications for software-hardware abstraction

It is not entirely true the architecture can hide details of microarchitecture

Even less true going forward. What to do?

Application level recommendations

- Use industrial strength, optimized libraries compiled for target machine.
- Lapack, Linpack, Matlab, Python SciPy, NumPy...
- `https://people.inf.ethz.ch/markusp/teaching/263-2300-ETH-spring11/slides/class08.pdf`

Algorithm level recommendations

Deploy cache-oblivious algorithm implementations.

Compiler level recommendations

- Enable compiler optimizations—*e.g.*, `-O3`, `-floop-interchange`, `-floop-block`.
- `https://gcc.gnu.org/onlinedocs/gcc/Optimize-Options.html`