

Review

- · Last Class:
 - Basic cache optimizations
 - Address translation with caches
- Today's class:
 - Advanced cache optimizations
 - Advanced translation optimizations
- · Announcement and reminder
 - Project will be distributed tonight
 - In-class mid-term exam next Wednesday. Please mark you calendar

51

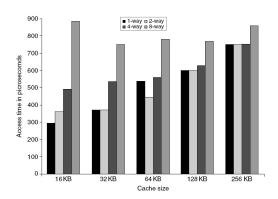


Advanced optimizations of cache performance (§ 2.2)



1. Small and Simple Caches to reduce hit time

- · Critical timing path:
 - · address tag memory, then compare tags, then select set
- Lower associativity
- Direct-mapped caches can overlap tag comparison and transmission of data



53



2. Way prediction to reduce hit time

- Combine fast hit time of Direct Mapped and the lower conflict misses of 2-way SA caches?
 - · check one way first (speed of direct mapped cache)
 - on a miss, check the other way, if it hits, call it a pseudo-hit (slow hit)
 - way prediction is a bit to indicate which half to check first (changes dynamically)



- · May extend prediction to more than 2-way SA caches
- · Saves power
- Drawback: CPU pipeline is hard if hit takes sometimes 1 and sometimes 2 cycles



3. Pipeline cache access to increase bandwidth

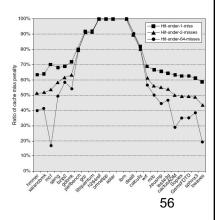
- · Examples:
 - » Pentium: 1 cycle
 - » Pentium Pro Pentium III: 2 cycles
 - » Pentium 4 Core i7: 4 cycles
- · Increases branch mis-prediction penalty
- Separate the tag compare and data access (in your HW)

55

4. Non-blocking caches to increase bandwidth



- <u>Hit under miss</u> allows data cache to continue to supply cache hits during a miss -- useful only with out-of-order execution.
- <u>Hit under multiple miss</u> or <u>miss under miss</u> may further lower the effective miss penalty by overlapping multiple misses
- Significantly increases the complexity of the cache controller (multiple outstanding memory accesses)
- Pentium Pro allows 4 outstanding memory misses





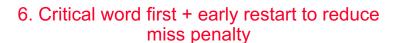
5. Multi-bank caches to increase bandwidth

- · Individual memory controller for each bank.
- · Each bank may have its own address and data lines.
- Banks used for independent accesses vs. faster sequential accesses.
 - ARM Cortex-A8 supports 1-4 banks for L2
 - Intel i7 supports 4 banks for L1 and 8 banks for L2
- · How blocks are interleaved affects performance.

Block	Block	Block	Block
address Bank 0	address Bank 1	address Bank 2	address Bank 3
0	1	2	3
4	5	6	7
8	9	10	11
12	13	14	15

Figure 2.6 Four-way interleaved cache banks using block addressing. Assuming 64 bytes per blocks, each of these addresses would be multiplied by 64 to get byte addressing.

57



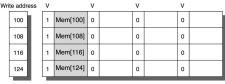


- Don't wait for full block to be loaded before restarting CPU
 - <u>Early restart</u> As soon as the requested word of the block arrives, send it to the CPU and let the CPU continue execution
 - <u>Critical Word First</u> Request the missed word first from memory and send it to CPU as soon as it arrives; Generally useful only in large blocks,
- Beneficial when we have long cache lines (blocks)
- · If want next sequential word, early restart may not be useful,

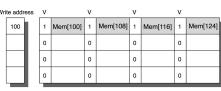


7. Merging write buffer to reduce miss penalty

No merging



Merging



- · Most useful in write through caches
- · Combine writing individual words into a block
- · Writing block is faster than writing individual words

59

8. Compiler optimizations to reduce miss rate



Instructions

- Reorder procedures in memory so as to reduce conflict misses
- Aligning basic blocks with cache blocks (lines)
- Profiling to look at conflicts.

Data

- Merging Arrays: improve spatial locality by single array of compound elements vs. 2 arrays
- Loop Fusion: Combine 2 independent loops that have same looping and some variables overlap
- Loop Interchange: change nesting of loops to access data in order stored in memory
- Blocking (Tiling): Improve temporal locality by accessing "blocks" of data repeatedly vs. going down whole columns or rows



Merging Arrays Example:

```
int val[SIZE];
int key[SIZE];

struct merge {
    int val;
    int key;
};
struct merge merged_array[SIZE];
```

Reducing conflicts between val & key improves spatial locality

61

Loop Fusion Example

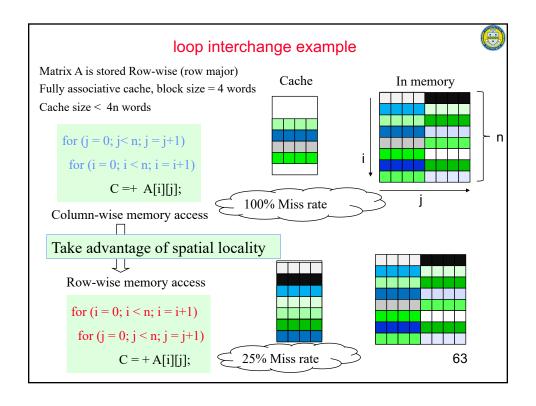


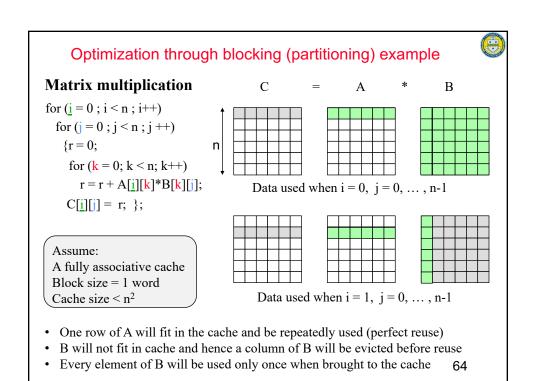
```
for (i = 0; i < N; i = i+1)
  for (j = 0; j < N; j = j+1)
        a[i][j] = 1/b[i][j] * c[i][j];

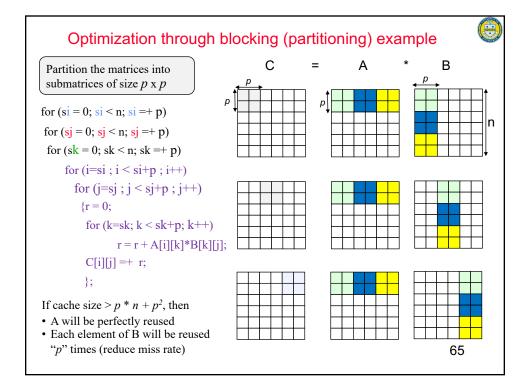
for (i = 0; i < N; i = i+1)
  for (j = 0; j < N; j = j+1)
        d[i][j] = a[i][j] + c[i][j];

for (i = 0; i < N; i = i+1)
  for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N; j = j+1)
        for (j = 0; j < N
```

One miss per access to a and c vs. two misses per access. Improves temporal locality



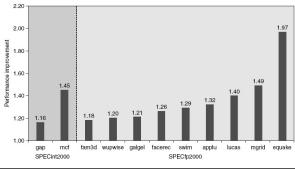




9. Hardware prefetching to reduce miss rate and penalty



- · Instruction Prefetching
 - Can fetch 2 (or more) blocks on a miss
 - Extra block placed in "stream buffer"
 - On miss, check stream buffer if found move to cache and prefetch next
- · Data Prefetching:
 - May have multiple stream buffers beyond the cache, each prefetching at a different address
- Relies on extra memory bandwidth that can be used without penalty.



10.Compiler prefetching to reduce miss rate and penalty

- · Data Prefetch
 - Load data into register
 - Cache Prefetch: load into cache
 - Special prefetching instructions should not cause premature page faults.
 - Issuing Prefetch Instructions takes time
 - Is cost of prefetch issues < savings in reduced misses?
- · Works only if can overlap prefetching with execution.
- Example: Assume that arrays a[] and b[] are aligned at block boundaries and that the cache block size is 4 words.

```
for (i=0; i < 100; i++)

if (i mod 4 = 0) prefetch (b[i+4], a[i+4])

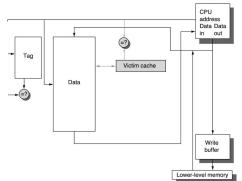
b[i] = c * a[i] + d * a[i+1];
```

if body of loop takes 20 cycles to execute and cache miss penalty is 80 cycles, then, after the first few iterations, data will be in cache when needed.

67

11. Using victim caches

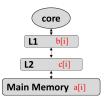
- A buffer to place data just evicted from cache
- A small number of fully associative entries
- Accessed in parallel with cache (no increase in hit time)
- On a hit in the VC, swap blocks in VC and cache

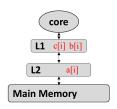


 When used with direct mapped caches, it has the effect of adding associativity to the most recently used cache blocks

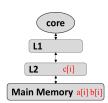


12. Computing with Near Data





- > a[i] in Sy can be re-computed by Sx.
- Should we perform the recomputation?
 - Based on the cost evaluation.
 - > Only re-computation if the cost is less.



69

Xulong Tang, Mahmut Taylan Kandemir, Hui Zhao , Myoungsoo Jung, Mustafa Karakoy. "Computing with Near Data." SIGMETRCIS-2019



- Replacement policy
 - Pseudo LRU
 - RL based
 - Prediction
 - · Hot-cold cache
- Multi-directional cache
 - Sumitha George, Minli Julie Liao, Huaipan Jiang, Jagadish B. Kotra, Mahmut T. Kandemir, Jack Sampson, Vijaykrishnan Narayanan: MDACache: Caching for Multi-Dimensional-Access Memories. MICRO 2018: 841-854
 - Minli Julie Liao, Jack Sampson: D-SOAP: Dynamic Spatial Orientation Affinity Prediction for Caching in Multi-Orientation Memory Systems. MICRO 2020: 581-595