Lecture 27: Multiprocessors

- Today's topics:
  - Shared memory vs message-passing
  - Simultaneous multi-threading (SMT)
  - GPUs

Shared-Memory Vs. Message-Passing

Shared-memory:

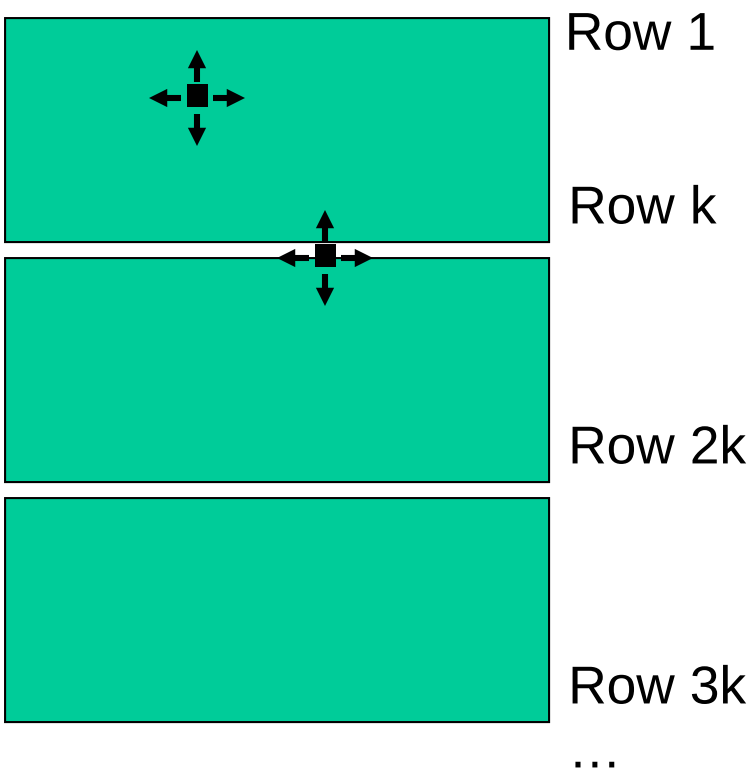
- Well-understood programming model
- Communication is implicit and hardware handles protection
- Hardware-controlled caching

Message-passing:

- No cache coherence → simpler hardware
- Explicit communication → easier for the programmer to restructure code
- Software-controlled caching
- Sender can initiate data transfer

Ocean Kernel

```
Procedure Solve(A)
begin
  diff = done = 0;
  while (!done) do
      diff = 0;
      for i ← 1 to n do
         for j ← 1 to n do
            temp = A[i,j];
            A[i,j] ← 0.2 * (A[i,j] + neighbors);
            diff += abs(A[i,j] – temp);
         end for
      end for
      if (diff < TOL) then done = 1;
  end while
end procedure
```



Shared Address Space Model

```
int  n, nprocs;
float  **A, diff;
LOCKDEC(diff_lock);
BARDEC(bar1);


main()
begin
   read(n); read(nprocs);
   A ← G_MALLOC();
   initialize (A);
   CREATE (nprocs,Solve,A);
   WAIT_FOR_END (nprocs);
end main
```

```
procedure Solve(A)
    int i, j, pid, done=0;
    float temp, mydiff=0;
    int mymin = 1 + (pid * n/procs);
    int mymax = mymin + n/nprocs -1;
    while (!done) do
       mydiff = diff = 0;
       BARRIER(bar1,nprocs);
       for i ← mymin to mymax
          for j ← 1 to n do
             …
          endfor
       endfor
       LOCK(diff_lock);
       diff += mydiff;
       UNLOCK(diff_lock);
       BARRIER (bar1, nprocs);
       if (diff < TOL) then done = 1;
       BARRIER (bar1, nprocs);
    endwhile
```

Message Passing Model

```
main()
  read(n); read(nprocs);
  CREATE (nprocs-1, Solve);
  Solve();
  WAIT_FOR_END (nprocs-1);

procedure Solve()
  int i, j, pid, nn = n/nprocs, done=0;
  float temp, tempdiff, mydiff = 0;
  myA ← malloc(…)
  initialize(myA);
  while (!done) do
    mydiff = 0;
    if (pid != 0)
      SEND(&myA[1,0], n, pid-1, ROW);
    if (pid != nprocs-1)
      SEND(&myA[nn,0], n, pid+1, ROW);
    if (pid != 0)
      RECEIVE(&myA[0,0], n, pid-1, ROW);
    if (pid != nprocs-1)
      RECEIVE(&myA[nn+1,0], n, pid+1, ROW);
    for i ← 1 to nn do
      for j ← 1 to n do
        …
      endfor
    endfor
    if (pid != 0)
      SEND(mydiff, 1, 0, DIFF);
      RECEIVE(done, 1, 0, DONE);
    else
      for i ← 1 to nprocs-1 do
        RECEIVE(tempdiff, 1, *, DIFF);
        mydiff += tempdiff;
      endfor
      if  (mydiff < TOL)  done = 1;
      for i ← 1 to nprocs-1  do
        SEND(done, 1, I, DONE);
      endfor
    endif
  endwhile
```

Multithreading Within a Processor

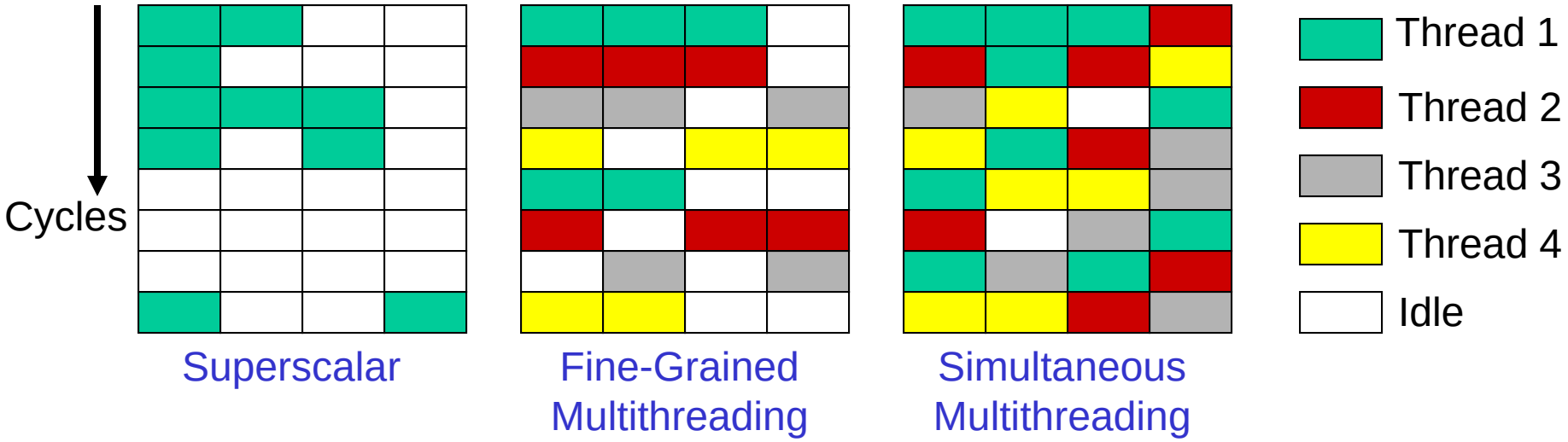
- Until now, we have executed multiple threads of an application on different processors – can multiple threads execute concurrently on the same processor?

- Why is this desireable?
  - inexpensive – one CPU, no external interconnects
  - no remote or coherence misses (more capacity misses)

- Why does this make sense?
  - most processors can't find enough work – peak IPC is 6, average IPC is 1.5!
  - threads can share resources → we can increase threads without a corresponding linear increase in area

How are Resources Shared?

Each box represents an issue slot for a functional unit. Peak thruput is 4 IPC.



- Superscalar processor has high under-utilization – not enough work every cycle, especially when there is a cache miss
- Fine-grained multithreading can only issue instructions from a single thread in a cycle – can not find max work every cycle, but cache misses can be tolerated
- Simultaneous multithreading can issue instructions from any thread every cycle – has the highest probability of finding work for every issue slot

Performance Implications of SMT

- Single thread performance is likely to go down (caches, branch predictors, registers, etc. are shared) – this effect can be mitigated by trying to prioritize one thread

- With eight threads in a processor with many resources, SMT yields throughput improvements of roughly 2-4

SIMD Processors

- Single instruction, multiple data
- Such processors offer energy efficiency because a single instruction fetch can trigger many data operations
- Such data parallelism may be useful for many image/sound and numerical applications

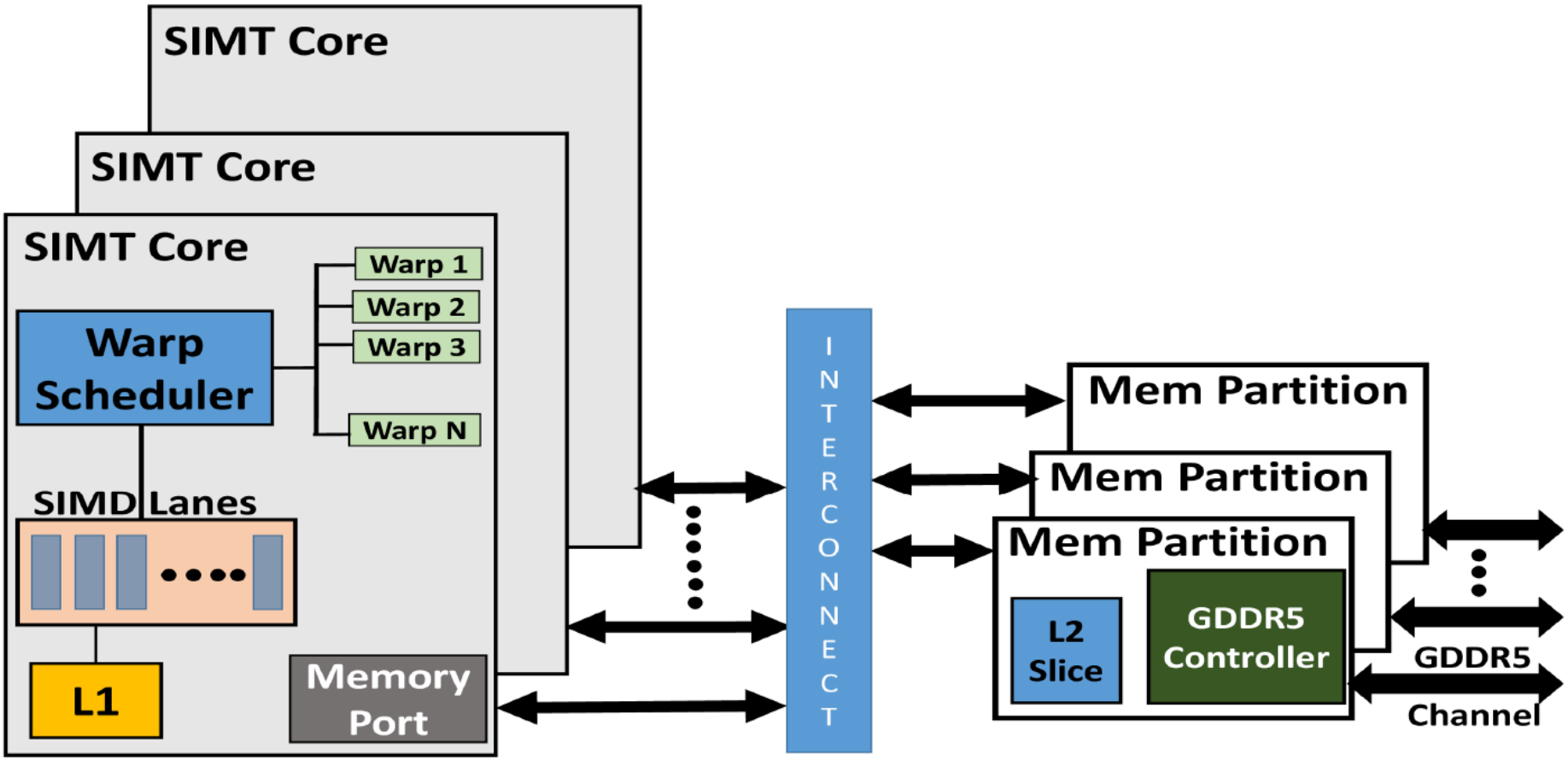
GPUs

- Initially developed as graphics accelerators; now viewed as one of the densest compute engines available
- Many on-going efforts to run non-graphics workloads on GPUs, i.e., use them as general-purpose GPUs or GPGPUs
- C/C++ based programming platforms enable wider use of GPGPUs – CUDA from NVidia and OpenCL from an industry consortium
- A heterogeneous system has a regular host CPU and a GPU that handles (say) CUDA code (they can both be on the same chip)

The GPU Architecture

- SIMT – single instruction, multiple thread; a GPU has many SIMT cores
- A large data-parallel operation is partitioned into many thread blocks (one per SIMT core); a thread block is partitioned into many warps (one warp running at a time in the SIMT core); a warp is partitioned across many in-order pipelines (each is called a SIMD lane)
- A SIMT core can have multiple active warps at a time, i.e., the SIMT core stores the registers for each warp; warps can be context-switched at low cost; a warp scheduler keeps track of runnable warps and schedules a new warp if the currently running warp stalls

The GPU Architecture



Architecture Features

- Simple in-order pipelines that rely on thread-level parallelism to hide long latencies
- Many registers (~1K) per in-order pipeline (lane) to support many active warps
- When a branch is encountered, some of the lanes proceed along the “then” case depending on their data values; later, the other lanes evaluate the “else” case; a branch cuts the data-level parallelism by half (branch divergence)
- When a load/store is encountered, the requests from all lanes are coalesced into a few 128B cache line requests; each request may return at a different time (mem divergence)

GPU Memory Hierarchy

- Each SIMT core has a private L1 cache (shared by the warps on that core)
- A large L2 is shared by all SIMT cores; each L2 bank services a subset of all addresses
- Each L2 partition is connected to its own memory controller and memory channel
- The GDDR5 memory system runs at higher frequencies, and uses chips with more banks, wide IO, and better power delivery networks
- A portion of GDDR5 memory is private to the GPU and the rest is accessible to the host CPU (the GPU performs copies)

Title

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