Lecture 4:
Parallel Programming Basics
This is an ISPC function.

It contains a loop nest.

Which iterations of the loop(s) are parallelized by ISPC? Which are not?

Answer: none
Parallel program instances were created when the \texttt{sinx()} ispc function was called.

```c
#include "sinx_ispc.h"

int N = 1024;
int terms = 5;
float* x = new float[N];
float* result = new float[N];

// initialize x here

// execute ISPC code
sinx(N, terms, x, result);
```

Each instance will run the code in the ispc function serially. (parallelism exists because there are multiple program instances, not the code that defines an ispc function)
Creating a parallel program

- Thought process:
  1. Identify work that can be performed in parallel
  2. Partition work (and also data associated with the work)
  3. Manage data access, communication, and synchronization

- Recall one of our main goals is speedup *
  For a fixed computation:

\[
\text{Speedup( P processors ) } = \frac{\text{Time (1 processor)}}{\text{Time (P processors)}}
\]

* Other goals include high efficiency (cost, area, power, etc.)
or working on bigger problems than can fit on one machine
Creating a parallel program

- **Problem to solve**
- **Decomposition**
- **Assignment**
- **Orchestration**
- **Mapping**

- **Subproblems** (a.k.a. “tasks”, “work to do”)
- **Parallel Threads** (“workers”)
- **Parallel program** (communicating threads)

- **Execution on parallel machine**

These responsibilities may be assumed by the programmer, by the system (compiler, runtime, hardware), or by both!

Adopted from: Culler, Singh, and Gupta
Problem decomposition

- Break up problem into tasks that can be carried out in parallel
- In general: create at least enough tasks to keep all execution units on a machine busy

Key challenge of decomposition: identifying dependencies (or... a lack of dependencies)
Amdahl’s Law: dependencies limit maximum speedup due to parallelism

- You run your favorite sequential program...

- Let $S$ = the fraction of sequential execution that is inherently sequential (dependencies prevent parallel execution)

- Then maximum speedup due to parallel execution $\leq \frac{1}{S}$
A simple example

- Consider a two-step computation on a N x N image
  - Step 1: double brightness of all pixels
    (independent computation on each pixel)
  - Step 2: compute average of all pixel values

- Sequential implementation of program
  - Both steps take $\sim N^2$ time, so total time is $\sim 2N^2$
First attempt at parallelism (P processors)

- **Strategy:**
  - Step 1: execute in parallel
    - time for phase 1: $N^2/P$
  - Step 2: execute serially
    - time for phase 2: $N^2$

- **Overall performance:**

  \[
  \text{Speedup} \leq \frac{2n^2}{n^2 + n^2/P}
  \]

  \[
  \text{Speedup} \leq 2
  \]
Parallelizing step 2

- **Strategy:**
  - Step 1: execute in parallel
    - time for phase 1: $\frac{N^2}{P}$
  - Step 2: compute partial sums in parallel, combine results serially
    - time for phase 2: $\frac{N^2}{P} + P$

- **Overall performance:**
  - Speedup $\leq \frac{2n^2}{\frac{2n^2}{P} + P}$

Note: speedup $\rightarrow P$ when $N \gg P$
Amdahl’s law

- Let $S =$ the fraction of total work that is inherently sequential
- Max speedup on $P$ processors given by:

$$\text{speedup} \leq \frac{1}{S + \frac{1 - S}{P}}$$

### Graph

- $S = 0.01$
- $S = 0.05$
- $S = 0.1$
Decomposition

- **Who is responsible for performing decomposition?**
  - In most cases: the programmer

- **Automatic decomposition of sequential programs continues to be a challenging research problem**
  - (very difficult in general case)
  - Compiler must analyze program, identify dependencies
    - What if dependencies are data dependent (not known at compile time)?
  - Researchers have had modest success with simple loop nests
  - The “magic parallelizing compiler” for complex, general-purpose code has not yet been achieved
Assignment

Problem to solve

Decomposition

Subproblems (a.k.a. "tasks", "work to do")

Assignment

Parallel Threads **
("workers")

Orchestration

Parallel program (communicating threads)

Mapping

Execution on parallel machine

** I had to pick a term
Assignment

- Assigning tasks to threads **
  - Think of “tasks” as things to do
  - Think of threads as “workers”

- Goals: balance workload, reduce communication costs

- Can be performed statically, or dynamically during execution

- Although programmer is often responsible for decomposition, many languages/runtime take responsibility for assignment.

** I had to pick a term (will explain in a second)
Assignment examples in ISPC

export void sinx(
    uniform int N,
    uniform int terms,
    uniform float* x,
    uniform float* result)
{
    // assumes N % programCount = 0
    for (uniform int i=0; i<N; i+=programCount)
    {
        int idx = i + programIndex;
        float value = x[idx];
        float numer = x[idx] * x[idx] * x[idx];
        uniform int denom = 6;  // 3!
        uniform int sign = -1;
        for (uniform int j=1; j<=terms; j++)
        {
            value += sign * numer / denom;
            numer *= x[idx] * x[idx];
            denom *= (2*j+2) * (2*j+3);
            sign *= -1;
        }
        result[i] = value;
    }
}

Decomposition of work by loop iteration

Programmer-managed assignment:
- Static assignment
  - Assign iterations to ISPC program instances in interleaved fashion

foreach construct exposes independent work to system
System manages assignment of iterations (work) to ISPC program instances (abstraction leaves room for dynamic assignment, but current ISPC implementation is static)
Example 2: static assignment using C++11 threads

```cpp
void my_thread_start(int N, int terms, float* x, float* results) {
    sinx(N, terms, x, result); // do work
}

void parallel_sinx(int N, int terms, float* x, float* result) {
    int half = N/2.

    // launch thread to do work on first half of array
    std::thread t1(my_thread_start, half, terms, x, result);

    // do work on second half of array in main thread
    sinx(N - half, terms, x + half, result + half);

    t1.join();
}
```

Decomposition of work by loop iteration

- **Programmer-managed static assignment**
  - This program assigns iterations to threads in a blocked fashion
    (first half of array assigned to the spawned thread, second half assigned to main thread)
Dynamic assignment using ISPC tasks

```c
void foo(uniform float* input,
         uniform float* output,
         uniform int N)
{
    // create a bunch of tasks
    launch[100] my_ispc_task(input, output, N);
}
```

ISPC runtime assign tasks to worker threads

List of tasks:

| task 0 | task 1 | task 2 | task 3 | task 4 | ... | task 99 |

Implementation of task assignment to threads: after completing current task, worker thread inspects list and assigns itself the next uncompleted task.
Orchestration

- Problem to solve
- Decomposition
- Assignment
- Orchestration
- Mapping

Subproblems (a.k.a. "tasks", "work to do")

Parallel Threads **
(“workers”)

Parallel program (communicating threads)

Execution on parallel machine

** I had to pick a term
Orchestration

- Involves:
  - Structuring communication
  - Adding synchronization to preserve dependencies if necessary
  - Organizing data structures in memory
  - Scheduling tasks

- Goals: reduce costs of communication/sync, preserve locality of data reference, reduce overhead, etc.

- Machine details impact many of these decisions
  - If synchronization is expensive, might use it more sparsely
Mapping to hardware

Problem to solve

Decomposition

Subproblems
(a.k.a. “tasks”, “work to do”)

Parallel Threads**
(“workers”)

Parallel program
(communicating threads)

Execution on
parallel machine

Assignment

Orchestration

Mapping

** I had to pick a term
Mapping to hardware

- Mapping “threads” (“workers”) to hardware execution units

- Example 1: mapping by the operating system
  - e.g., map pthread to HW execution context on a CPU core

- Example 2: mapping by the compiler
  - Map ISPC program instances to vector instruction lanes

- Example 3: mapping by the hardware
  - Map CUDA thread blocks to GPU cores (future lecture)

- Some interesting mapping decisions:
  - Place related threads (cooperating threads) on the same processor
    (maximize locality, data sharing, minimize costs of comm/sync)
  - Place unrelated threads on the same processor (one might be bandwidth limited and another might be compute limited) to use machine more efficiently
Example: last class I asked you a question about mapping

- Consider an application that creates two threads
- The application runs on the processor shown below
  - Two cores, two-execution contexts per core, up to instructions per clock, one instruction is an 8-wide SIMD instruction.

- Question: “who” is responsible for mapping the application’s pthreads to the processor’s thread execution contexts?
  
  **Answer:** the operating system

- Question: If you were implementing the OS, how would you map the two threads to the four execution contexts?

- Another question: How would you map threads to execution contexts if your C program spawned five threads?
A parallel programming example
A 2D-grid based solver

- Solve partial differential equation (PDE) on \((N+2) \times (N+2)\) grid
- Iterative solution
  - Perform Gauss-Seidel sweeps over grid until convergence

\[
\]

Grid solver example from: Culler, Singh, and Gupta
Grid solver algorithm
C-like pseudocode for sequential algorithm is provided below

const int n;
float* A; // assume allocated to grid of N+2 x N+2 elements

void solve(float* A) {
    float diff, prev;
    bool done = false;

    while (!done) { // outermost loop: iterations
        diff = 0.f;
        for (int i=1; i<n i++ ) { // iterate over non-border points of grid
            for (int j=1; j<n; j++) {
                prev = A[i,j];
                                A[i,j+1] + A[i+1,j]);
                diff += fabs(A[i,j] - prev); // compute amount of change
            }
        }

        if (diff/(n*n) < TOLERANCE) // quit if converged
            done = true;
    }
}
Step 1: identify dependencies
(problem decomposition phase)

Each row element depends on element to left.
Each row depends on previous row.

Note: the dependencies illustrated on this slide are element data dependencies in one iteration of the solver (in one iteration of the “while not done” loop)
Step 1: identify dependencies
(problem decomposition phase)

There is independent work along the diagonals!

Good: parallelism exists!

Possible implementation strategy:
1. Partition grid cells on a diagonal into tasks
2. Update values in parallel
3. When complete, move to next diagonal

Bad: independent work is hard to exploit
Not much parallelism at beginning and end of computation.
Frequent synchronization (after completing each diagonal)
Let’s make life easier on ourselves

- Idea: improve performance by changing the algorithm to one that is more amenable to parallelism

  - Change the order grid cell cells are updated

  - New algorithm iterates to same solution (approximately), but converges to solution differently
    - Note: floating-point values computed are different, but solution still converges to within error threshold

  - Yes, we needed domain knowledge of Gauss-Seidel method for solving a linear system to realize this change is permissible for the application
New approach: reorder grid cell update via red-black coloring

Update all red cells in parallel

When done updating red cells, update all black cells in parallel (respect dependency on red cells)

Repeat until convergence
Possible assignments of work to processors

Question: Which is better? Does it matter?

Answer: it depends on the system this program is running on
Consider dependencies (data flow)

1. Perform red update in parallel
2. Wait until all processors done with update
3. Communicate updated red cells to other processors
4. Perform black update in parallel
5. Wait until all processors done with update
6. Communicate updated black cells to other processors
7. Repeat
Communication resulting from assignment

Blocked Assignment

Interleaved Assignment

= data that must be sent to P2 each iteration

Blocked assignment requires less data to be communicated between processors
Three ways to think about writing this program

- Data parallel
- SPMD / shared address space
- Message passing (will wait until a future class)
Data-parallel expression of solver
const int n;

float* A = allocate(n+2, n+2));   // allocate grid

void solve(float* A) {
    bool done = false;
    float diff = 0.f;
    while (!done) {
        for_all (red cells (i,j)) {
            float prev = A[i,j];
            reduceAdd(diff, abs(A[i,j] - prev));
        }
        if (diff/(n*n) < TOLERANCE)
            done = true;
    }
}

Data-parallel expression of grid solver

Note: to simplify pseudocode: just showing red-cell update

Grid solver example from: Culler, Singh, and Gupta
Shared address space (with SPMD threads) expression of solver
Shared address space expression of solver
SPMD execution model

- Programmer is responsible for synchronization
- Common synchronization primitives:
  - Locks (provide mutual exclusion): only one thread in the critical region at a time
  - Barriers: wait for threads to reach this point
int n;                  // grid size
bool done = false;
float diff = 0.0;
LOCK myLock;
BARRIER myBarrier;

// allocate grid
float* A = allocate(n+2, n+2);

void solve(float* A) {
    int threadId = getThreadId();
    int myMin = 1 + (threadId * n / NUM_PROCESSORS);
    int myMax = myMin + (n / NUM_PROCESSORS);
    while (!done) {
        diff = 0.f;
        barrier(myBarrier, NUM_PROCESSORS);
        for (j=myMin to myMax) {
            for (i = red cells in this row) {
                float prev = A[i,j];
                                A[i+1,j], A[i,j+1]);
                lock(myLock)
                diff += abs(A[i,j] - prev));
                unlock(myLock);
            }
        }
        barrier(myBarrier, NUM_PROCESSORS);
        if (diff/(n*n) < TOLERANCE)            // check convergence, all threads get same answer
            done = true;
        barrier(myBarrier, NUM_PROCESSORS);
    }
}
int     n;                   // grid size
bool    done = false;
float   diff = 0.0;
LOCK    myLock;
BARRIER myBarrier;

// allocate grid
float* A = allocate(n+2, n+2);

void solve(float* A) {
    int threadId = getThreadId();
    int myMin = 1 + (threadId * n / NUM_PROCESSORS);
    int myMax = myMin + (n / NUM_PROCESSORS);

    while (!done) {
        diff = 0.f;
        barrier(myBarrier, NUM_PROCESSORS);
        for (j=myMin to myMax) {
            for (i = red cells in this row) {
                float prev = A[i,j];
                                A[i+1,j], A[i,j+1]);
                lock(myLock)
                diff += abs(A[i,j] - prev));
                unlock(myLock);
            }
        }
        barrier(myBarrier, NUM_PROCESSORS);
        if (diff/(n*n) < TOLERANCE)            // check convergence, all threads get same answer
            done = true;
        barrier(myBarrier, NUM_PROCESSORS);
    }
}
int n;                  // grid size
bool done = false;
float diff = 0.0;
LOCK myLock;
BARRIER myBarrier;

// allocate grid
float* A = allocate(n+2, n+2);

void solve(float* A) {
  float myDiff;
  int threadId = getThreadId();
  int myMin = 1 + (threadId * n / NUM_PROCESSORS);
  int myMax = myMin + (n / NUM_PROCESSORS);

  while (!done) {
    float myDiff = 0.f;
    diff = 0.f;
    barrier(myBarrier, NUM_PROCESSORS);
    for (j=myMin to myMax) {
      for (i = red cells in this row) {
        float prev = A[i,j];
                        A[i+1,j], A[i,j+1]);
        myDiff += abs(A[i,j] - prev));
      }
    }
    lock(myLock);
    diff += myDiff;
    unlock(myLock);
    barrier(myBarrier, NUM_PROCESSORS);
    if (diff/(n*n) < TOLERANCE)            // check convergence, all threads get same answer
      done = true;
  }
}
Barrier synchronization primitive

- `barrier(num_threads)`
- Barriers are a conservative way to express dependencies
- Barriers divide computation into phases
- All computations by all threads before the barrier complete before any computation in any thread after the barrier begins
  - In other words, all computations after the barrier are assumed to depend on all computations before the barrier
int n;               // grid size
bool done = false;
float diff = 0.0;
LOCK myLock;
BARRIER myBarrier;

// allocate grid
float* A = allocate(n+2, n+2);

void solve(float* A) {
    float myDiff;
    int threadId = getThreadId();
    int myMin = 1 + (threadId * n / NUM_PROCESSORS);
    int myMax = myMin + (n / NUM_PROCESSORS);

    while (!done) {
        float myDiff = 0.f;
        diff = 0.f;
        barrier(myBarrier, NUM_PROCESSORS);
        for (j=myMin to myMax) {
            for (i = red cells in this row) {
                float prev = A[i,j];
                                A[i+1,j], A[i,j+1]);
                myDiff += abs(A[i,j] - prev));
            }
            lock(myLock);
            diff += myDiff;
            unlock(myLock);
            barrier(myBarrier, NUM_PROCESSORS);
            if (diff/(n*n) < TOLERANCE)            // check convergence, all threads get same answer
                done = true;
        }
    }
}
int n;               // grid size
bool done = false;
LOCK myLock;
BARRIER myBarrier;
float diff[3];  // global diff, but now 3 copies

float *A = allocate(n+2, n+2);

void solve(float* A) {
    float myDiff;   // thread local variable
    int index = 0;  // thread local variable

    diff[0] = 0.0f;
    barrier(myBarrier, NUM_PROCESSORS);  // one-time only: just for init

    while (!done) {
        myDiff = 0.0f;
        // perform computation (accumulate locally into myDiff)
        //
        lock(myLock);
        diff[index] += myDiff;    // atomically update global diff
        unlock(myLock);
        diff[(index+1) % 3] = 0.0f;
        barrier(myBarrier, NUM_PROCESSORS);
        if (diff[index]/(n*n) < TOLERANCE)
            break;
        index = (index + 1) % 3;
    }
}

Idea:
Remove dependencies by using different diff variables in successive loop iterations
Trade off footprint for removing dependencies!
(a common parallel programming technique)

Grid solver example from: Culler, Singh, and Gupta
Solver implementation in two programming models

- **Data-parallel programming model**
  - **Synchronization:**
    - Single logical thread of control, but iterations of `forall` loop may be parallelized by the system (implicit barrier at end of `forall` loop body)
  - **Communication**
    - Implicit in loads and stores (like shared address space)
    - Special built-in primitives for more complex communication patterns: e.g., reduce

- **Shared address space**
  - **Synchronization:**
    - Mutual exclusion required for shared variables (e.g., via locks)
    - Barriers used to express dependencies (between phases of computation)
  - **Communication**
    - Implicit in loads/stores to shared variables
We will defer discussion of the message passing expression of solver to a later class.
Summary

- Amdahl’s Law
  - Overall maximum speedup from parallelism is limited by amount of serial execution in a program

- Aspects of creating a parallel program
  - Decomposition to create independent work, assignment of work to workers, orchestration (to coordinate processing of work by workers), mapping to hardware
  - We’ll talk a lot about making good decisions in each of these phases in the coming lectures (in practice, they are very inter-related)

- Focus today: identifying dependencies

- Focus soon: identifying locality, reducing synchronization