L18: Parallelism and Concurrency

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Before We Begin

- Please go to this link and try if Hangout Meet works ok
 - meet.google.com/bgu-zhup-zby

 I want to make sure everything works fine in case we need to move the class to an online format

Parallelism



This means you do multiple things at once

- It is a simultaneous execution of computations
 - Does not have to be related
 - Can be related

- Warning: Some examples here are in c++
 - Rayon and parallel constructs in Rust naturally enable these

Concurrency

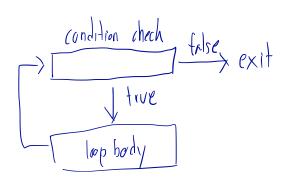
- This is handling multiple things at once-Things done on CPU1 cannot delay CPU2
- The thing computer do concurrently should be independent from each other

- Think of it this way (and, of course, I will tie HW in this)
 - Concurrency is a process of truly running things independently on the hardware level
 - Think Amdahl's law with critical path that we discussed

https://blog.golang.org/concurrency-is-not-parallelism

Parallel For-loop

- In C, you can use Cilk execute in parallel
 - cilk_for(int i=0; i<n; i++)B[i] = A[i] + 1



- You can use OpenMP annotation Make many threads on each streeting
 - #pragma omp for for(int i=0; i<n; i++)
 B[i] = A[i] + 1

$$B[i] : B[i-1]$$

(each iteration is

indepen dent)

(annot Parallel - Have to wait for B[1-1]

before start new one

Notice how there are no dependency?

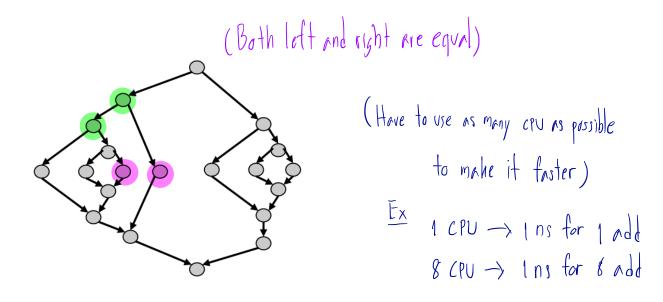
Fork-join

- (A | | B) means A and B can be run in parallel
- Use rayon for this
 int fib(int n) {
 if n < 2 { return n; }
 let (x, y) = rayon::join(|| fib(n-1), || fib(n-2))
 return x + y;
 }</pre>

How does the execution looks like in dataflow?

Nested Parallelism

- Basically combines parallel loop and fork join
- What is the maximum concurrent computation we get out of this parallel task?



 With this dependency graph, two tasks are parallel if there is not path between each other

Cost Analysis - How to compute the cost

- W = work
 - The total number of operations (How many instructions)
- D = depth (or span)
 - The longest chain of dependency (Data Flow Graph)
- Parallelism = work/depth
 - This determine the number of processors that can be effectively used
- Remember Amdahl's Law?

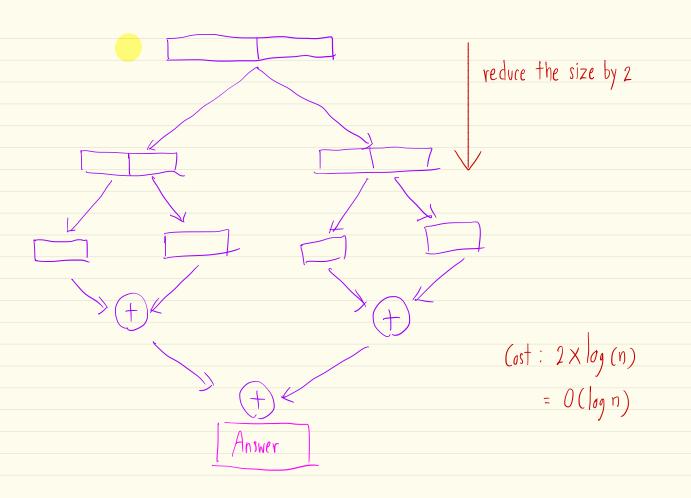
- Lingering question from a HW/system person (me):
 - Well, what do we do with the CPUs that are doing nothing?

Example

```
• fn seq sum(v: &[i32]) -> i32 {
  let mut total = 0i32;
  for num in v {
   total += num
  total
• What is W? - 0(n)
                        Printleism = O(n) = 1
• What is D? - 0(n)
              L> No parallel
```

Example #2

```
• fn par sum(v: &[i32]) -> i32 {
  if v.len() <= 1 {
  return seq sum(v);
                             (Split left sum and right sum)
  let (left, right) =
 v.split at(v.len()/2);
  use rayon::join;
  let (left sum, right sum) =
    join(|| par_sum(left), || par_sum(right));
  left sum + right sum
                              Paralleism = O(n) = better than 1
• What is W? - O(n) - Add n items
• What is D? - log(n)
```



Why This Cost Model?

- Simple (just draw the dependency graph)
- Brent's Theorem:

- number of processors (How many (PU needed)
- Can schedule in O(W/P + D) time on P processors
- See how this follow exactly as Amdahl's law?
- Lower bound: How much time do we need if we have P processors?
- When you design a parallel algorithm
 - Work efficient: Work should be ~ the same as sequential running time
 - A polynomial parallelism of O(n^{1/2}) is generally good

Example: Not Very Good Quicksort

```
def qs(xs: Seq[Int]): Seq[Int] = \{
  if (xs.length <= 1) xs
  else {
     p = RNG. choice (xs) - Random Pick number in the list
     s0 = [efor ein xs if e < p]
     s1 = [e for e in xs if e == p]
     s2 = [efor ein xs if e > p]
      (r0, r2) = par(qs(s0) | | qs(s2))
                               M = O(n)
     r0 ++ s1 ++ r2
           Sorted Right
                               D = 0( |09 h)
```

Normal Quick Sort Pivot Left Part will smaller than Right Part Pivot Pivet

Better Quick Sort Pivot Pivot Pivet

A Bad Parallel Program

- Consider this example: what if
 - Partitioning and concatenation is done in O(n) work and O(logn) depth
 - But you perform serial recursive calls Finish one before another
 - What is my complexity?



Hint 1: Try drawing the dependency graph

Hint 2: This is not very good

A Better Parallel Program

- Consider this example: what if
 - Partitioning and concatenation is done in O(n) work and O(logn) depth
 - The recursive calls are not made in parallel
- What is the work?
- What is the depth/span?

Which one gets lower? Why?

What is the parallelism?

Designing a Good Parallel Program

- This in terms of operations on collections
 - map, reduce, filter ...
 - Why are these good?

```
> Do it with every elements on the same time
```

- Map: applies a function f to every elt of the collection
- Reduce: pairwise combines elements in a tree until you have 1 using a (n assoctive) binary operator
- Filter: retains an element if pred(e) returns true

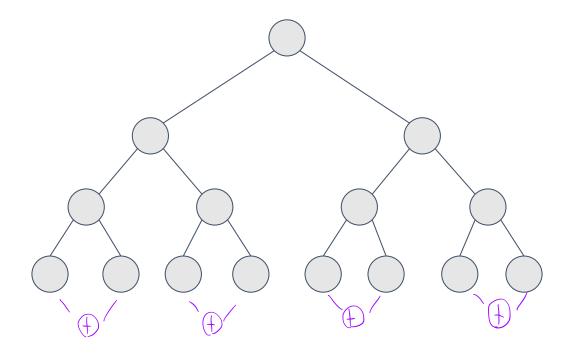
Map's Parallelism

Apply a function on all elements

• What is W and what is D?

Reduce's Parallelism

Combine until get i olement



• What is W and what is D? Why?

Filter's Parallelism - Apply function, if true heep'it, false throw it away

- Major challenges
 - How many are selected?
 - How to retain and put into the output?
- 3 Pan't Linow

- Filter can be done in parallel if designed properly
 - Complexity depends on the elements being filtered
 - Convert this into the dataflow graph is a good way to tell its parallelism



In-class Exercise 17

 What is the dependency graph looks like and what is the complexity of our parallel quicksort (both W and D)?