

Parallel Algorithms

Lecture 7: Divide And Conquer

Ronald Veldema

Divide and Conquer (1)

- Take a problem 'X'
 - Subdivide in subproblems of equal type
 - Input, output domain, etc partitioning
 - Subproblems are independent of each other
 - Solve the subproblems
 - Recursively
 - sequentially
 - Combine subproblems' solutions
 - Wait for all subproblems to complete ?

Divide and Conquer (2)

- Also known as "recursive decomposition"
- Top down reasoning:
 - split large jobs into smaller jobs until so small that it can be best solved sequentially.
 - merge results to solve the bigger task which in turn can be used to solve the next larger task.
- Take a recursive program
 - I.S.O. recursive calls, create parallel tasks
 - when result needed, wait for subtask to finish

Divide and Conquer (3)

- Communication only occurs when
 - Spawning a new task
 - Gathering/reducing results
- pro: no dependence in program on
 - the number of processors used
 - network topology
 - scheduling
- con: hard to tune algorithms by hand as hardly anything to tune.

Complexity....

- Say we have an D&C alg with a branchout of '2'
 - 2 subproblems are solved recursively
 - seq. time $\text{alg}(N) = \text{time alg}(N/2) + \text{time alg}(N/2)$
 - par. time $\text{alg}(N) = \text{time alg}(\text{smallest granularity})$
 - Neglect time needed to traverse the binary tree

Example: fibonacci numbers (1)

$$\text{fib}(n) = \begin{cases} N & , \text{ if } N < 2 \\ \text{fib}(n-1) + \text{fib}(n-2), & \text{ otherwise} \end{cases}$$

$$\text{fib}(n) = (n < 2) ? n : \text{fib}(n - 1) + \text{fib}(n - 2)$$

```
long sequential_fib(long n) {  
  
    if(n < 2) return n;  
  
    long x = fib(n-1);  
    long y = fib(n-2);  
    return x + y;  
}
```

Example: fibonacci numbers (2)

$fib(n) = (n < 2) ? n : fib(n - 1) + fib(n - 2)$

```
long parallel_fib(long n) {
    if(n < 2) return n;

    long x = spawn task fib(n-1);
    long y = spawn task fib(n-2);
    <wait for sub tasks>
    return x + y;
}
```

Example: Fibonacci Numbers (3)

- optimization: more sequential execution to make tasks more coarse grained.
 - How to choose CUT_OFF ?

```
long parallel_fib(long n) {
    if(n < CUT_OFF) return sequential_fib(n);
    long x = spawn fib(n-1);
    long y = spawn fib(n-2);
    <wait for sub tasks>
    return x + y;
}
```

When to stop dividing ?

- granularity
 - how much parallel ?
 - how much sequential ?
 - smaller granularity →
 - more communication but better load balancing...
 - More overhead
 - Sub task administration
 - Method call overheads

Example: Binary Search (1)

// A sorted table. The table is sorted on 'keys'
// using the KeyType.smaller_than() method

```
KeyType keys[];
ValueType values[];

ValueType binary_search(KeyType key, int start, int end)
{
    int middle := (start+end)/2;

    if (keys[middle].equals(key))
        return value[middle];
    else (keys[middle].smaller_than(key))
        return binary_search(X, start, middle-1);
    else
        return binary_search(key, middle+1, end);
}
```

Example: Binary Search (2)

- Any ideas on how to parallelize ?
 - Spawning recursive iterations won't help

Example: Binary Search (3)

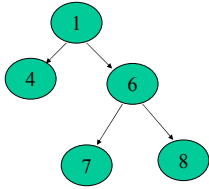
- When spawning job [0..n/2]
 - Spawn parallel jobs [0..n/4] and [n/4..n/2]

```
ValueType binary_search(KeyType key, int start, int end)
{
    int middle := (start+end)/2;

    if (keys[middle].equals(key))
        return value[middle];
    else (keys[middle].smaller_than(key))
        return parallel
            binary_search(X, start, (start+middle-1)/2)
            binary_search(X, (start+middle-1)/2, middle-1)
    else return parallel
        binary_search(key, middle+1, (middle+1+end)/2);
        binary_search(key, (middle+1+end)/2, end);
}
```

Example: Heap Sort (1)

- heap = balanced binary tree with value at any node smaller than that of its children



Example: Heap Sort (2)

Sequential_heap_sort(**unordered** set of numbers L)

```
{  
    Heap H = new Heap( L );  
    List X = new List();  
    while (H not empty)  
        P = remove lowest number from H;  
        // note: lowest number = top of the tree  
        X += P;  
    return X  
}
```

Example: Heap Sort (3)

- Sequential Heap construction:

```
BinaryTree t = new BinaryTree();
```

```
For each number in list:  
    find a good spot to insert element
```

Example: Heap Sort (4)

- Parallel_heap_sort(List L)
 - BalancedUnsortedBinaryTree T(L);
 - Heapify(T)

Example: Heap Sort (4)

```
Heapify(Tree T)  
    if T = empty  
        return;  
  
    heapify(T.left);  
    heapify(T.right);  
  
    if (left.value < T.value)  
        swap_data(left, T);  
    if (right.value < T.value)  
        swap_data(right, T);
```

Example: Raytracer (1)

- For each pixel on the screen
 - Color[x,y] = render_pixel()
- Some pixels take longer to compute than others
- Pixels are independent

Example: Raytracer (2)

```
void render_screen(int width, int height) {
    for (int x=0;x<width;x++)
        for (int y=0;y<height;y++) {
            Color color = render(x,y);
            screen.plot_pixel(x, y, color);
        }
}
```

Example: Raytracer (3)

```
void render_screen_rectangle(rectangle r) {
    if (r.is_large()) {
        render_screen_rectangle(r.upper_left());
        render_screen_rectangle(r.upper_right());
        render_screen_rectangle(r.lower_left());
        render_screen_rectangle(r.lower_right());
        return;
    }
    for (int x=r.x; x<r.width;x++)
        for (int y=r.z; y<r.height;y++) {
            Color color = render(x,y);
            screen.plot_pixel(x, y, color);
        }
}
```

Dynamic Programming (DP) (0)

- Solve a problem recursively
(Just like divide and conquer)
- Before recomputing a subproblem
 - See if you've already computed it
 - Return 'cached' answer
 - Recompute otherwise

Dynamic Programming (DP) (1)

- DP = recursion with memory
- where divide and conquer has no dependencies between tasks, dynamic programming assumes that there is
- bottom up reasoning I.S.O. top down reasoning.
 - solve smallest tasks first and record results.
 - try and solve next larger task and record result
 - goto 2 while not found solution.

Dynamic Programming (2)

- tasks that have already been computed are assumed to return the same value on another invocation
- solving the next bigger task can be viewed as combining sub-results using a function:
- result = f(s(i1), s(i2) ... s(in))
- monadic combining functions: contain 1 's' in f(..)
- polydic combining functions: contain >1 's' in f(..)
 - F(a(i1),b(i2), c(i2), etc)
- work can be represented in a directed graph.
- serial DP formulation: depend only on tasks that are one level deeper in graph
- DP requires communication to store intermediate results where simple div&conquer requires none.

Dynamic Programming (3)

```
// parallel fib with dynamic programming
long cached_value[] = {-1, -1, -1, etc};

public long fib(long n) {
    if(n < 2) return n;
    long x = spawn dp_fib(n-1);
    long y = spawn dp_fib(n-2);
    sync();
    return x + y;
}

long dp_fib(long n) {
    if (cached_value[n] != -1)
        return cached_value[n];
    else
        cached_value[n] = fib(n);
    return valued_value[n];
}
```

Dynamic Programming (4)

- Knapsack problem:
 - given items 'i' of weight w(i) and profit p(i), try to fit as many objects in knapsack of capacity c while maximizing profit.
 - solution vector = (0/1)*, where 0 if not in sack, 1 if in sack.
 - $F(i, x) = \max$ profit solution for i objects in sack with x capacity
 - DP solution = 0 if $x \geq 0, i=0$
 - -inf if $x < 0, i=0$
 - $\max(F(i-1,x), F(i-1,x-w(i)) + p(i))$ if $1 < i \leq n$
 - parallel:
 - $F(i-1,x)$ can be computed on another machine in parallel to $F(i-1,x-w(i))$
 - results of computations on other machines can be broadcasted.

Knapsack problem (1)

Item	0	1	2	3	4	5	6	7	8
Value	5	0	3	3	5	6	3	2	4
Weight	2	4	5	4	4	3	5	2	2

Knapsack can hold 15 kilos

- try and put as many objects in the knapsack as possible

Knapsack problem (2)

```

Solution knapsack(int bitvector, int item) {
  if (weight(bitvector) > capacity) return INF;
  if (item == max_item) return new Solution(bitvector);
  // put item in knapsack
  int bit_vector1 = bitvector | (1<<item);
  // don't put item in knapsack
  int bit_vector2 = bitvector;

  Solution s1 = knapsack(bitvector1, item+1);
  Solution s2 = knapsack(bitvector2, item+1);

  if better(s1,s2) return s1
  else return s2;
}
  
```

Knapsack problem (3)

```

Solution knapsack(int bitvector, int item) {
  if (weight(bitvector) > capacity) return INF;
  if (item == max_item) return new Solution(bitvector);
  // put item in knapsack
  int bit_vector1 = bitvector | (1<<item);
  // don't put item in knapsack
  int bit_vector2 = bitvector;

  Solution s1 = spawn knapsack(bitvector1, item+1);
  Solution s2 = spawn knapsack(bitvector2, item+1);
  sync;
  if better(s1,s2) return s1
  else return s2;
}
  
```

Knapsack problem (4)

- Problems
 - If a better solution has been found, the algorithm will still look at knapsacks that are worse
 - There is no idea of a 'global optimum' until the very end
- Solution: maintain a global 'best capacity' value

Knapsack problem (5)

```

Solution knapsack(int bitvector, int item) {
  if (weight(bitvector) > best_found) return INF;
  best_found = min(best_found, weight(bitvector));
  if (item == max_item) return new Solution(bitvector);
  // put item in knapsack
  int bit_vector1 = bitvector | (1<<item);
  // don't put item in knapsack
  int bit_vector2 = bitvector;

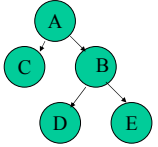
  Solution s1 = spawn knapsack(bitvector1, item+1);
  Solution s2 = spawn knapsack(bitvector2, item+1);
  sync;
  if better(s1,s2) return s1
  else return s2;
}
  
```

Where is the bug ?

Dynamic programming (5)

- Problems:
 - Every processor needs concurrent read/write to the 'cache' of already found answers
 - Possible solutions:
 - Each processor maintains his own private cache
 - Partition cache & maintain a separate 'lock' for each partition
 - 'cache' can become very large if ALL answers are stored
 - Pushing out old return-values may help

Task Graphs (1)

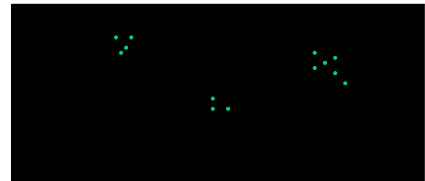
- task A spawns B & C.
 - task B spawns D & E.
- 
- no incoming edges = start nodes
 - no outgoing edges = end nodes
 - critical path = max start → end node.
 - critical path length = length(critical path)
 - interaction graph = if data exchanges between tasks I & J, then edge between I & J.

Task Graphs (2)

- Divide & conquer: from parent to child (parameter & return value)
- Dynamic programming: from each internal node that may use or produce a cachable value

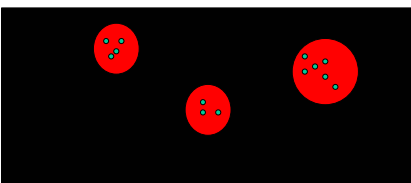
Example: cluster search (1)

- Given a number points in a 2 dimensional space, try and group spatially close points.
- Clustering is best if
 - Size of each cluster is minimal



Example: cluster search (2)

- Given a number points in a 2 dimensional space, try and group spatially close points.
- Clustering is best if
 - Size of each cluster is minimal



Example: cluster search (5)

```
void clusterize(point2d P[], int start, int end)
{
    // make a cluster entry for pair of points
    for each point X in P
        for each point Y in P
            if X != Y
                clusters += new cluster(X,Y)
    // see if the total surface area decreases when merging
    // with another area
    for each cluster A in clusters
        for each cluster B clusters
            if A != B and surface(A)+surface(B) > surface(A+B)
                clusters -= A
                clusters += cluster(A+B)
}
```

Example: cluster search (6)

- Problems
 - Not recursive = not divide and conquer

Example: cluster search (7)

```
int len = length(a);
foo(a, len);
-----
for each i in a do
  statement(i)
void foo(Data []a, int len) {
  if (len == 1) statement(i)
  else {
    foo(a[0..len/2], len/2);
    foo(a[len/2 .. len-1], len/2);
  }
}
```

Questions

- Can each loop be rewritten to use recursion ?
 - ? YES: each jump to start of the loop = recursive invocation
- Can each loop be rewritten as above ?
 - ? NO: depends on data-dependencies

CILK (1)

- ANSI C extension
 - spawn call(parameters)
 - sync;
- For shared memory multiprocessors
- Translates CILK to plain C

```
cilk long parallel_fib(long n) {
  if(n < 2) return n;
  long x = spawn fib(n-1);
  long y = spawn fib(n-2);
  sync;
  return x + y;
}
```

CILK (2)

- Each CPU has a private 'job' stack
- Each spawn puts a descriptor on his job stack
 - The parameters of the spawned invocation
 - Where to store return value of call afterwards
- Each sync pops descriptors from his stack until stack exhausted
 - Because of divide & conquer
 - Big tasks on bottom of stack
 - Smallest tasks on top
 - Current processor picks work from top-of-the-stack
 - Others try and "steal" jobs from bottom of stack
 - "work stealing" algorithm

CILK (3)

- Who do you steal jobs from ?
 - Random work stealing ?
 - Hierarchical work stealing ?
 - Match stealing pattern to that of network ?
- Do you sometimes push jobs to other machines when you have too many jobs ?
 - Push jobs to whom ?

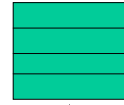
CILK (4)

- Matrix multiplication using divide and conquer

$$c_{ik} = \sum_{j=1}^n a_{ij} \times b_{jk}$$

CILK (4aa)

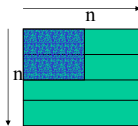
Improving Matrix Memory Layout



Access element $A[i][j]$ of $N \times N$ matrix using:
 $\text{ptr} + (N * I) + J$

CILK (4ab)

Improving Matrix Memory Layout



Access element $A[i][j]$ of $N \times N$ matrix using:
 $\text{ptr} + (n * I) + J$

CILK (4b)

$$c_{ik} = \sum_{j=1}^n a_{ij} \times b_{jk}$$

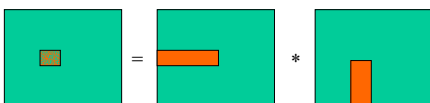
```
void iter_matmul(double *A, double *B, double *C, int n)
{
    int i, j, k;

    for (i = 0; i < n; i++)
        for (k = 0; k < n; k++) {
            double c = 0.0;
            for (j = 0; j < n; j++)
                c += A[i * n + j] * B[j * n + k];
            C[i * n + k] = c;
        }
}
```

CILK (4c)



Is this the same



As this ?



CILK (4d)

```
cilk void rec_matmul(double *A, double *B, double *C,
                    int m, int n, int p, int ld, boolean add)
{
    if ((m + n + p) <= 64) { /* base case */
        if (add) {
            for (int i = 0; i < m; i++)
                for (int k = 0; k < p; k++) {
                    double c = 0.0;
                    for (int j = 0; j < n; j++)
                        c += A[i * ld + j] * B[j * ld + k];
                    C[i * ld + k] += c;
                }
        } else {
            for (int i = 0; i < m; i++)
                for (int k = 0; k < p; k++) {
                    double c = 0.0;
                    for (int j = 0; j < n; j++)
                        c += A[i * ld + j] * B[j * ld + k];
                    C[i * ld + k] = c;
                }
        }
    } else spawn_sub_matrix_computations(A, B, C, m, n, p, ld, add);
}
```


CILK (4e)

```
void spawn_sub_matrix_computations(double *A, double *B, double *C,
    int m, int n, int p, int ld, boolean add) {
    if (m >= n && n >= p) {
        int m1 = m / 2;
        spawn rec_matmul(A, B, C, m1, n, p, ld, add);
        spawn rec_matmul(A + m1 * ld,
            B,
            C + m1 * ld, m - m1,
            n, p, ld, add);
    } else if (n >= m && n >= p) {
        int n1 = n / 2;
        spawn rec_matmul(A, B, C, m, n1, p, ld, add);
        sync;
        spawn rec_matmul(A + n1,
            B + n1 * ld,
            C, m, n - n1, p, ld, 1);
    } else {
        int p1 = p / 2;
        spawn rec_matmul(A, B, C, m, n, p1, ld, add);
        spawn rec_matmul(A,
            B + p1,
            C + p1, m, n, p - p1, ld, add);
    }
}
```

CILK (6)

- Abort mechanism
 - During search space exploration you may want to abort all branches that have become superfluous
 - Dynamic programming does not stop already running superfluous tasks, only prevents new unneeded tasks
 - One task kills
 - A sibling task
 - ...and all its children
 - Where are those tasks running (other machines...)

JSR-166 (1)

- Future Java extension: *java.util.concurrent.**
 - Instead of extending *java.lang.Thread* extend *java.util.concurrent.FJTask*
 - FJTask instances are managed by *java.util.concurrent.FJTaskRunner* instances (themselves *java.lang.Thread* instances)

JSR-166 (2)

```
Step1: in class heading:
class Fib extends FJTask {
    int number; // result & FIB number this task is to compute

Step2: in main:
FJTaskRunnerGroup g = new FJTaskRunnerGroup(cpus);
Fib f = new Fib(num);
g.invoke(f);

Step 3: in run:
int n = number;
if (n <= 1) {
    // Do nothing: fib(0) = 0; fib(1) = 1
} else if (n <= sequentialThreshold) {
    number = seqFib(n);
} else {
    Fib f1 = new Fib(n - 1);
    Fib f2 = new Fib(n - 2);
    fork(f1); fork(f2);
    join();

    // Combine results:
    number = f1.number + f2.number;
}
```

JSR-166 (3)

//Conceptually:

```
FJTask.fork(FJTask t)
    stack.push(t);
```

```
FJTask.join();
    while (! stack.empty())
        FJTask t = stack.pop();
        t.run();
```

Satin (1)

- An alternative Java extension for divide and conquer programming
- Designed for
 - Clusters of workstations
 - Clusters of clusters (aka the “computational grid”)
- Objects implementing a ‘Spawnable’ interface have their methods invoked in parallel on other machines
- All parameters to methods of ‘Spawnable’ are call by copy
 - Distributed memory machines !

Satin (2)

```
interface FibInter extends satin.Spawnable {
    int fib(int n);
}

class Fib extends satin.SatinObject
    implements FibInter {
    public int fib(int n) {
        if (n<=2) return n;
        long x = fib(n-1); // spawns new task
        long y = fib(n-2); // spawns new task
        sync();
        return x + y;
    }

    public static void main(String args[]) {
        Fib f = new Fib();
        int result = f.fib(17);
        f.sync();
        System.out.println("result = " + result);
    }
}
```

Satin (3)

- Implements
 - Random stealing
 - Once a machine is idle
 - try and perform “work stealing” from some other machine
 - Cluster-aware random Stealing
 - Once a machine is idle
 - Send a work request asynchronously to a remote cluster
 - Try and perform “work stealing from some other machine within own cluster in parallel to work request