#### COMP 790-033 - Parallel Computing

Lecture 2 August 24, 2022

#### The PRAM model and its complexity measures

• Reading for next class (Wed Aug 31): PRAM handout secns 3.6, 4.1

#### **First class summary**

- In this course we study how to speed up large computational problems using parallel computing
  - in theory and in practice
- We study various parallel programming models
  - Initially we consider a theoretical model, the Parallel Random Access Machine (PRAM)
    - study algorithms and their asymptotic complexity
  - Subsequently we focus on practical models and their implementation on current hardware
    - shared memory multiprocessors, accelerators, and distributed memory clusters
      - examine execution model, hardware operation, programming constructs, performance analysis
      - illustrate principles using various case studies

# **Topics today**

#### PRAM model

- execution model
- programming model
- Work-Time model
  - programming model
  - complexity metrics
  - Brent's theorem: translation to PRAM programs
- Parallel prefix algorithm
  - derivation
  - applications

# **PRAM model of parallel computation**

- PRAM = Parallel Random Access Machine
  - *p* processors
  - shared memory
  - each processor has a unique identity  $1 \le i \le p$
  - synchronous PRAM model
    - Single Instruction, Multiple Data (SIMD)
    - each processor may be active (✓) or inactive (×)
    - each instruction is executed by active processors only
    - each instruction completes in unit time



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### **PRAM program**

- PRAM program
  - sequential program
  - expressions involving processor id *i* have a unique value in each processor
    - *i* can be used as an array index

```
X[i] := 10 * i
```

• conditionals specify active processors



### **Concurrent memory access - Read**

#### • Concurrent reads (CR)

#### - all readers of a given location see the same value

```
X[i] := y
X[i] := B[ [i/2] ]
```

value of  $y\,$  read concurrently by all p processors the first p/2 elements of B are read concurrently by two processors

- Eliminating bounded-degree concurrent reads
  - replace x[i] := B[[i/2]] with



concurrent read is eliminated but number of steps is doubled

#### **Concurrent memory access - Write**

#### • Concurrent writes (CW)

- final value depends on the arbitration policy among writes to the same destination:
  - Arbitrary CW
    - nondeterministic choice among values written
  - Common CW
    - processors that write a value to the same destination must write the same value, else error
  - Priority CW
    - value written by processor with lowest processor id
  - Combining Write
    - all values combined using a specified associative operation (e.g. "+")
- **Example** (p = 6)
  - y := X[i]
  - B[ $\lceil i/2 \rceil$ ] := X[i]



#### **Concurrent writes:**

- Let B[1:p] be an array of boolean values and define  $c = B_1 \lor B_2 \lor \ldots \lor B_p$ 
  - use p processors and concurrent writes to compute c in a constant number of steps
    - a) with combining CW

b) with a CW policy other than combining CW (which?)

#### **Concurrent memory access**

- PRAM variants
  - EREW, CREW, ERCW, CRCW
  - differ in performance, not expressive power
    - EREW < CREW < CRCW
  - loosely reflect difficulty of model implementation
- The following are considered EREW
  - references to
    - processor id i
    - number of processors p
    - problem size n
  - references to local variables

**local** h; h := 2\*i + 1; X[h] := X[i]

- expression evaluation is synchronous, e.g.

```
X[i] := X[i] + X[i+1]
is EREW
```

# A PRAM program

- Simple problem: vector addition
  - given V,W vectors of length n
  - compute Z = V + W
- PRAM program
  - constructed to operate with arbitrary
    - problem size n
    - number of processors p
  - work to be performed must explicitly be "scheduled" across processors
  - time complexity with p procs
    - T<sub>c</sub>(n,p) =
  - PRAM model?



*Input*: V[1:n], W[1:n] in shared memory *Output*: Z[1:n] in shared memory

# **Work-Time paradigm**

- W-T parallel programming model
  - high-level PRAM programming model
    - specifies available parallelism
    - no explicit scheduling of parallelism over processors
  - simplifies algorithm presentation and analysis
  - W-T programs can be mechanically translated to PRAM programs
- W-T program
  - sequential program
  - forall construct
    - specification of available parallelism
    - number of processors is not a parameter of the model !

#### WT program for vector addition

```
Input: V[1:n], W[1:n]
Output: Z[1:n]
```

```
forall i in 1:n do
```

```
Z[i] := V[i] + W[i]
```

```
enddo
```

# **Programming notation for the W-T framework**

- standard sequential programming notation
  - statements
    - assignment
    - statement composition
    - alternative construct (if ... then ... else ...)
    - repetitive construct (for, while)
  - expressions
    - arithmetic and logical functions
    - variable reference
    - (recursive) function and procedure invocation
- forall statement
  - specifies statement T may be executed simultaneously for each value of i in D
  - no restriction on T
    - can be a sequence of statements
    - can invoke (recursive) functions
    - can be another (nested) forall statement



# **W-T complexity metrics**

- Work complexity W(n)
  - total number of operations performed (as a function of input size n)
- Step complexity S(n)
  - number of steps required (as a function of input size n)
  - assuming unbounded parallelism
- Inductively defined over constructs of W-T programming notation

#### W-T complexity measures: simple example



### Work and Step Complexity of the forall construct

- How to define work and time complexity of the forall construct?
  - P: forall i in D do body T depending on i enddo
  - assume we can determine  $W(T_i)$  and  $S(T_i)$  for each i in D
    - W(P) =
    - S(P) =

#### W-T complexity measures: vector summation

• let  $n = 2^k$ 

```
forall i in 1:n/2 do
        S[i] := S[2i - 1] + S[2i]
enddo
```

```
for h := 1 to k do
    forall i in 1:n/2<sup>h</sup> do
        S[i] := S[2i - 1] + S[2i]
        enddo
enddo
```



### W-T complexity measures: vector summation

- Vector summation (sum - reduction)
  - given V[1..n],  $n = 2^k$
  - compute s = sum(V[1:n])
  - optimal sequential time  $T(n) = \Theta(n)$

Complexity
 W(n) =

S(n) =

```
Input: V[1:n] vector of integers, n = 2<sup>k</sup>
Output: s = sum(V[1:n])
```

```
P1: forall i in 1:n do
B[i] := V[i]
```

```
enddo
```

```
P2: for h := 1 to k do
        forall i in 1:n/2<sup>h</sup> do
        B[i] := B[2i-1]+B[2i]
        enddo
        enddo
```

P3:s := B[1]

#### PRAM model needed?

# Brent's theorem and $T_c(n,p)$

• Brent's theorem schedules a W-T program for a *p*-processor PRAM

– idea

- simulate each parallel step in W-T program using p processors
- the work  $W_i(n)$  to be performed in step i can be completed using p processors in time  $\left[\underline{W_i(n)}\right]$
- bound concurrent runtime  $T_C(n,p)$  of resultant PRAM program
  - by summing over all S(n) steps

$$T_{c}(n,p) = \sum_{i=1}^{S(n)} \left\lceil \frac{W_{i}(n)}{p} \right\rceil \leq \sum_{i=1}^{S(n)} \left( \left\lfloor \frac{W_{i}(n)}{p} \right\rfloor + 1 \right) \leq \left\lfloor \sum_{i=1}^{S(n)} \frac{W_{i}(n)}{p} \right\rfloor + S(n) = \left\lfloor \frac{W(n)}{p} \right\rfloor + S(n)$$
$$\left\lceil \frac{W(n)}{p} \right\rceil = \left\lceil \sum_{i=1}^{S(n)} \frac{W_{i}(n)}{p} \right\rceil \leq \sum_{i=1}^{S(n)} \left\lceil \frac{W_{i}(n)}{p} \right\rceil = T_{c}(n,p)$$

# **Scheduling W-T vector summation algorithm**

W-T vector summation algorithm

```
Input: V[1:n] vector of integers, n = 2<sup>k</sup>
Output: s = sum(V[1:n])
```

```
P1: forall i in 1:n do
B[i] := V[i]
```

enddo

```
P2: for h := 1 to k do
    forall i in 1:n/2<sup>h</sup> do
        B[i] := B[2i-1]+B[2i]
        enddo
    enddo
P3: s := B[1]
```

PRAM vector summation algorithm

```
Input: V[1:n] vector of integers, n = 2<sup>k</sup>
Output: s = sum(V[1:n])
p > 0 processor PRAM; processor index i
local integer j, r;
P1: for j := 1 to [n/p] do
```

```
r := (j-1)•p + i
if r ≤ n then B[r] := V[r] endif
enddo
```

```
P2: for h := 1 to k do
for j := 1 to \lceil (n/2^h)/p \rceil do
r := (j-1) \cdot p + i
if r \le n/2^h then
B[r] := B[2r-1]+B[2r]
endif
enddo
enddo
```

```
P3: if i \leq 1 then s := B[1] endif
```

# Performance of translated W-T program

- Count steps needed to perform the additions
  - Brent's theorem predicts

$$T_c(n,p) = O\left(\left\lfloor \frac{n-1}{p} \right\rfloor + \lg n\right)$$

counts for various p

 $\frac{p}{p=1} \frac{T_c(n,p)}{(n-1)/p}$   $p > n \qquad \qquad \log n$   $p = 3, n = 2^k, k \text{ even } \approx \lfloor (n-1)/p \rfloor + \frac{1}{2} \lg n$ 

- Upper bound is tight (for this program)
- translation retains EREW model

#### PRAM vector summation algorithm

```
Input: V[1:n] vector of integers, n = 2^k
Output: s = sum(V[1:n])
p > 0 processor PRAM; processor index i
```

```
local integer j, r;
P1: for j := 1 to [n/p] do
    r := (j-1)•p + i
    if r ≤ n then B[r] := V[r] endif
enddo
```

```
P2: for h := 1 to k do
for j := 1 to \lceil (n/2^h)/p \rceil do
r := (j-1) \cdot p + i
if r \le n/2^h then
B[r] := B[2r-1] + B[2r]
endif
enddo
P3: if i \le 1 then s := B[1] endif
```

# Parallel prefix-sum

- Prefix sum
  - Input
    - Sequence X of  $n = 2^k$  elements, binary associative operator +
  - Output
    - Sequence S of  $n = 2^k$  elements, with  $S_i = x_1 + ... + x_i$
  - Example:
    - X = [1, 4, 3, 5, 6, 7, 0, 1]
    - S = [1, 5, 8, 13, 19, 26, 26, 27]
  - $T_{S}(n) = \Theta(n)$
- Uses of prefix sum
  - efficient parallel implementation of sequential "scan" through consecutive actions
    - ex: Given series of bank transactions T[1:n], with T[i] positive or negative, and T[1] the opening deposit > 0
      - Was the account ever overdrawn?
  - explicit or implicit component of many parallel algorithms

### **Prefix sum algorithm**

#### • Recursive solution

- Xi stands for X[i] and Xij stands for X[i]+X[i+1]+... +X[j]



• W-T complexity

$$- W(n) = W\left(\frac{n}{2}\right) + O(n), W(1) = O(1) \implies ?$$
$$- S(n) = S\left(\frac{n}{2}\right) + O(1), S(1) = O(1) \implies ?$$

#### Parallel prefix sum algorithm – WT model

```
Input: X[1..n] vector of integers Output: S[1..n]
```



```
par_prefix_sum( X[1..n] ) =
   var Y[1..n/2], Z[1..n/2], S[1..n];
   S[1] := X[1];
   if n > 1 then
      forall 1 \le i \le n/2 do
         Y[i] := X[2i-1] + X[2i]
      enddo
      Z[1..n/2] := par_prefix_sum(Y[1..n/2]);
      forall 2 \le i \le n do
         if even(i) then
            S[i] := Z[i/2]
         else
            S[i] := Z[(i-1)/2] + X[i]
         endif
      enddo
   endif
   return S[1..n]
```

#### **Balanced trees in arrays**

- Balanced Tree Ascend / Descend
  - Key idea
    - view input data as balanced binary tree
    - sweep tree up and/or down
  - "Tree" not a data structure but a control structure (e.g., recursion)
- Example
  - vector summation





#### In-place prefix sum



- → + ascend phase
- + descend phase
- retained value
  - S(n)
  - W(n)
  - Space
  - PRAM model

#### In-place prefix-sum algorithm – WT model



```
Input: X[1..n] vector of values, n = 2^k
Output: S[1..n] vector of prefix sums
parallel_prefix_sum( X[1..n] ) =
   forall i in 1:n do
       S[i] := X[i]
   enddo
   for h = 1 to k do
       forall i in 1:n/2<sup>h</sup> do
            S[2^{h_{i}}] := S[2^{h_{i}} - 2^{h-1}] + S[2^{h_{i}}]
       enddo
   enddo
   for h = k downto 1
       forall i in 2:n/2<sup>h-1</sup> do
           if odd(i) then
               S[2^{h-1}i] := S[2^{h-1}i - 2^{h-1}] + S[2^{h-1}i]
           endif
       enddo
   enddo
```

#### **Scan-based primitives**

- Scan operations (parallel prefix operations) can be used to implement many useful primitives
  - Suppose we are given SCAN to compute prefix sum of integer sequences

```
seq<int> SCAN(seq<int>)
```

- step complexity is  $\Theta(\lg n)$
- work complexity is  $\Theta(n)$
- PRAM model is EREW
- The next three examples have the same complexity as SCAN

# **COPY (or DISTRIBUTE)**

```
seq<int> COPY(int v, int n) ){
    seq<int> V[1:n];
    V[1] = v;
    forall i in 2 : n do
        V[i] := 0;
    enddo
    return SCAN(V);
}
```

$$v = 5$$
  
n = 7  
V = 5 0 0 0 0 0 0  
Res = 5 5 5 5 5 5 5 5

```
seq<int> ENUMERATE(seq<bool> Flag){
    seq<int> V[1:#Flag];
    forall i in 1 : #Flag do
        V[i] := Flag[i] ? 1 : 0;
    enddo
    return SCAN(V);
}
```

```
seq<T> PACK(seq<T> A, seq<bool> Flag){
    seq<T> R[1:#A];
    P := ENUMERATE(Flag);
    forall i in 1 : #Flag do
        if Flag[i] then R[P[i]] := A[i] endif;
    enddo
    return R[1:P[#Flag]];
}
```

#### **Radix Sort**

```
Input: A[1:n] with b-bit integer elements
Output: A[1:n] sorted
Auxiliary: FL[1:n], FH[1:n], BL[1:n], BH[1:n]
for h := 0 to b-1 do
     forall i in 1:n do
         FL[i] := (A[i] bit h) == 0
         FH[i] := (A[i] bit h) != 0
     enddo
     BL := PACK(A, FL)
     BH := PACK(A, FH)
     m := #BL
     forall i in 1:n do
           A[i] := if (i \le m) then BL[i] else BH[i-m] endif
     enddo
enddo
```

S(n) = W(n) =

# **Complexity measures for W-T algorithms**

- Asymptotic time complexity measures
  - (optimal) sequential time complexity  $T_s(n)$
  - parallel time complexity  $T_c(n,p)$
- Speedup
  - definition

$$SP(n,p) = \frac{T_s(n)}{T_c(n,p)}$$

- limitation

$$SP(n,p) = \frac{T_s(n)}{T_c(n,p)} \le \frac{T_s(n)}{W(n)/p} = \frac{pT_s(n)}{W(n)} = O(p)$$

- Average available parallelism
  - definition

$$AAP(n) = \frac{W(n)}{S(n)}$$

### **Objectives in the design of W-T algorithms**

- Goal 1: construct work efficient algorithms
  - a W-T algorithm is work efficient if  $W(n) = \Theta(T_s(n))$
  - work-inefficient parallel algorithms have limited appeal on a PRAM with a fixed number of processors p

$$\lim_{n \to \infty} SP(n, p) \leq \lim_{n \to \infty} \frac{pT_s(n)}{W(n)} = p \lim_{n \to \infty} \frac{T_s(n)}{W(n)} = 0$$

## **Objectives in the design of W-T algorithms**

- Goal 2: minimize step complexity
  - get optimal speedup using  $AAP(n) = T_s(n) / S(n)$  processors

$$SP(n, AAP(n)) = \Theta\left(\frac{T_s(n)}{T_c(n, AAP(n))}\right) = \Omega\left(\frac{T_s(n)}{\frac{T_s(n)}{AAP(n)} + S(n)}\right)$$
$$= \Omega\left(\frac{T_s(n)}{S(n) + S(n)}\right) = \Omega(AAP(n))$$

- when S(n) is decreased, AAP(n) is increased
  - with fixed problem size
    - can use more processors to get greater speedup
  - with fixed number of processors
    - reach optimal speedup at smaller problem size

### W-T model advantages

- Widely developed body of techniques
- Ignores scheduling, communication and synchronization
  - "easiest" parallel programming
- Source-level complexity metrics
  - Work and step complexity
  - related to running time via Brent's theorem
- Good place to start
  - many "real-world" algorithms can be derived starting from W-T algorithms