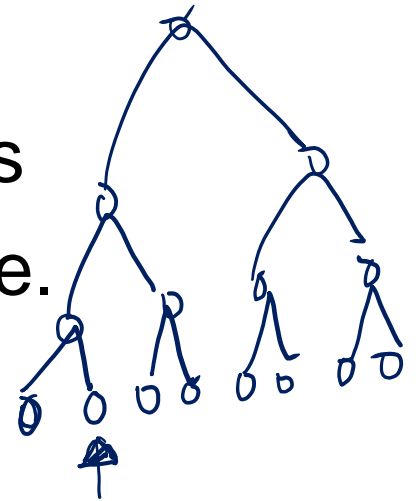


The Power of CRCW – Minima

- Two points of interest
 - Illustrate the power of CRCW models
 - Illustrate another optimality technique.
- Find the minima of n elements.
 - Input: An array A of n elements
 - Output: The minimum element in A .
- From what we already know:
 - Standard sequential algorithm not good enough
 - Can use an upward traversal, with min as the operator at each internal node. Time = $O(\log n)$, work = $O(n)$.



$O(n)$ time

The Power of CRCW – Minima

- Our solution steps:
 - Design a $O(n^2)$ work, $O(1)$ time algorithm.
 - Gain optimality by sacrificing runtime to $O(\log \log n)$.

Friday

An $O(1)$ Time Algorithm

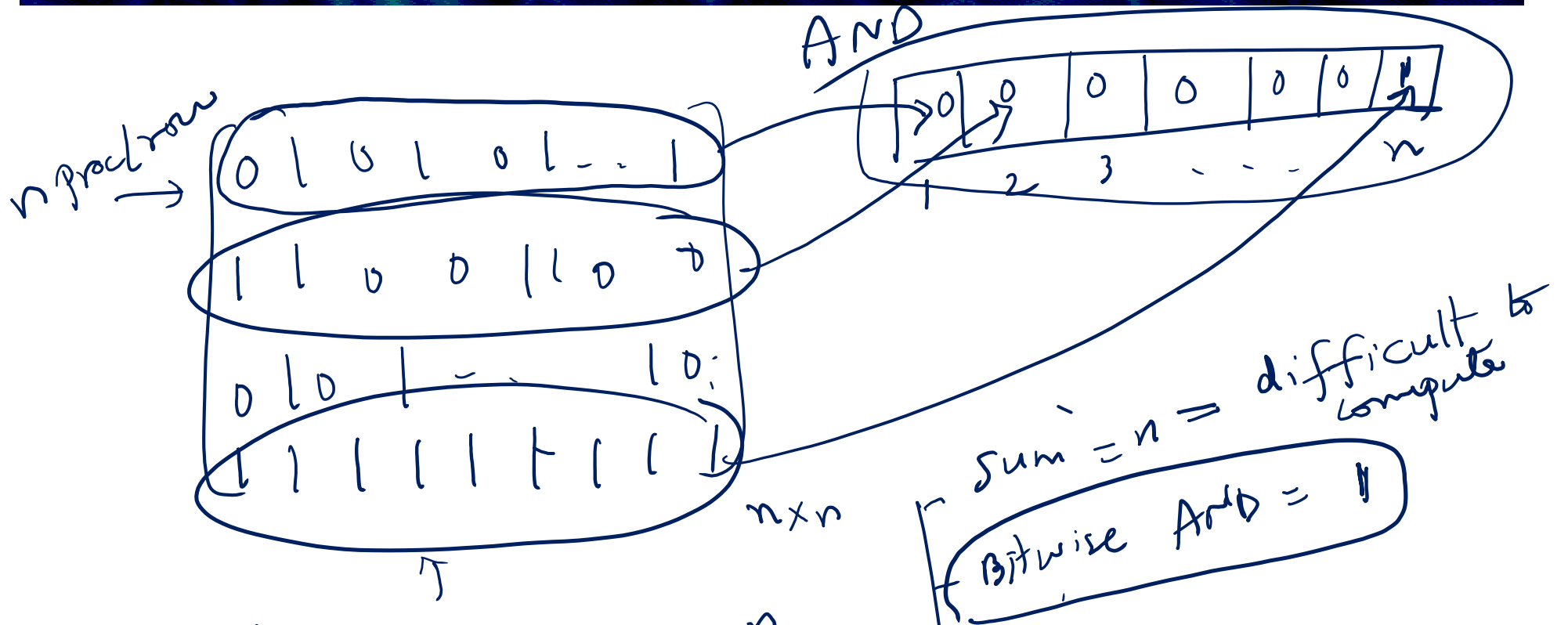
A	12	17	8	18	26
12	--	1	0	1	1
17	0	--	0	1	1
8	1	1	--	1	1
18	0	0	0	--	1
26	0	0	0	0	--

$n \times n$

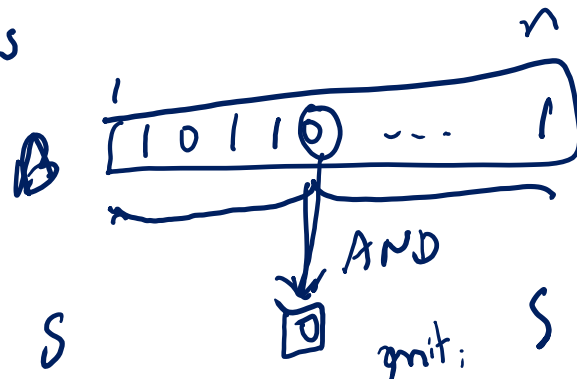
- Use n^2 processors.
- Compare $A[i]$ with $A[j]$ for each i and j .
- Now can identify the minimum.

$$\underline{\underline{A(i,j) = 0 \text{ if } A(i) > A(j)}}$$

n^2 Processes



CRCW helps

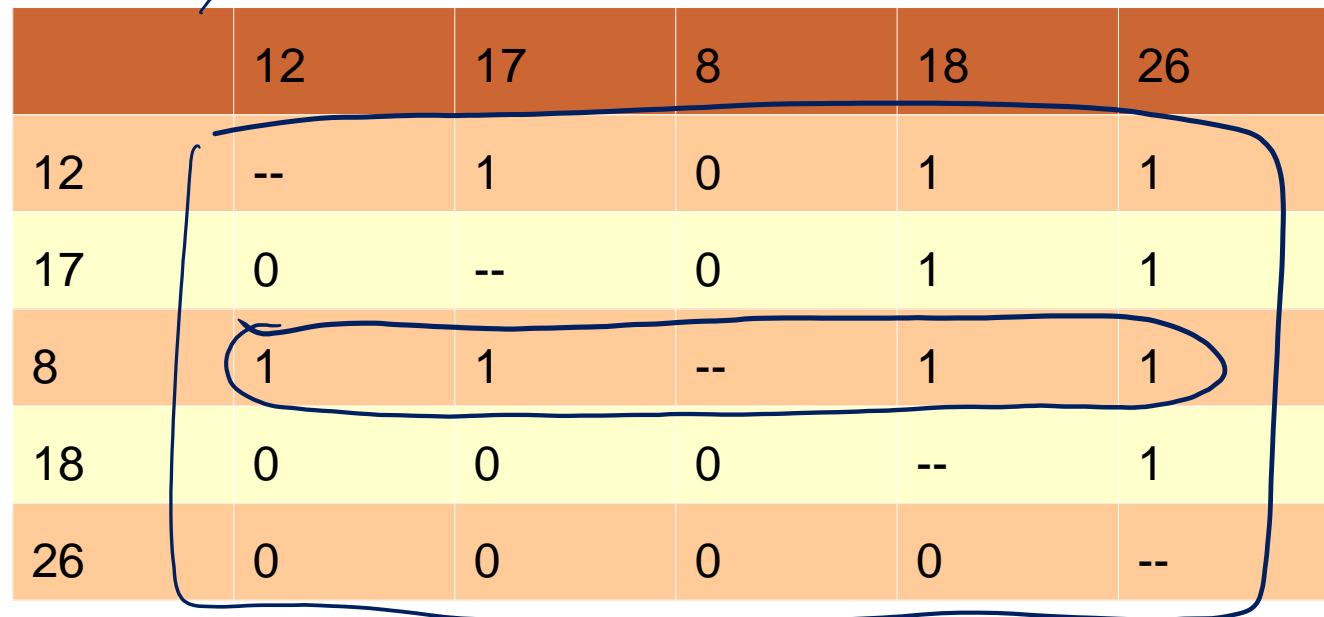


if $B(i) = 0$ write 0 to S

$S = \begin{cases} 1 & \text{if all } B(i) = 1 \\ 0 & \text{if any } B(i) = 0 \end{cases}$

// for all i in parallel

An $O(1)$ Time Algorithm



	12	17	8	18	26
12	--	1	0	1	1
17	0	--	0	1	1
8	1	1	--	1	1
18	0	0	0	--	1
26	0	0	0	0	--

- Use n^2 processors.
- Compare $A[i]$ with $A[j]$ for each i and j .
- Now can identify the minimum.
 - How?

An $O(1)$ Time Algorithm

	12	17	8	18	26
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- Use n^2 processors.
- Compare $A[i]$ with $A[j]$ for each i and j .
- Now can identify the minimum.
 - How?
- Where did we need the CRCW model?

work = n^2 ↓↓
time = $O(1)$

Towards Optimality

- The earlier algorithm is heavy on work.
- To reduce the work, we proceed as follows.
- We derive an $O(n \log \log n)$ work algorithm running in $O(\log \log n)$ time.
- For this, use a doubly logarithmic tree.
 - Defined in the following.

Work : $n^2 \rightarrow \underline{\underline{n \log \log n}}$ not optimal

Time : $O(1) \rightarrow \underline{\underline{O(\log \log n)}}$

↓

optimal

start with a smaller problem size

$$x = O(n/\log n)$$

$$x \log x = n$$

$$x = n$$

$$\log x + \log \log x = \log n$$

LHS $\uparrow\uparrow$

$$x < n$$

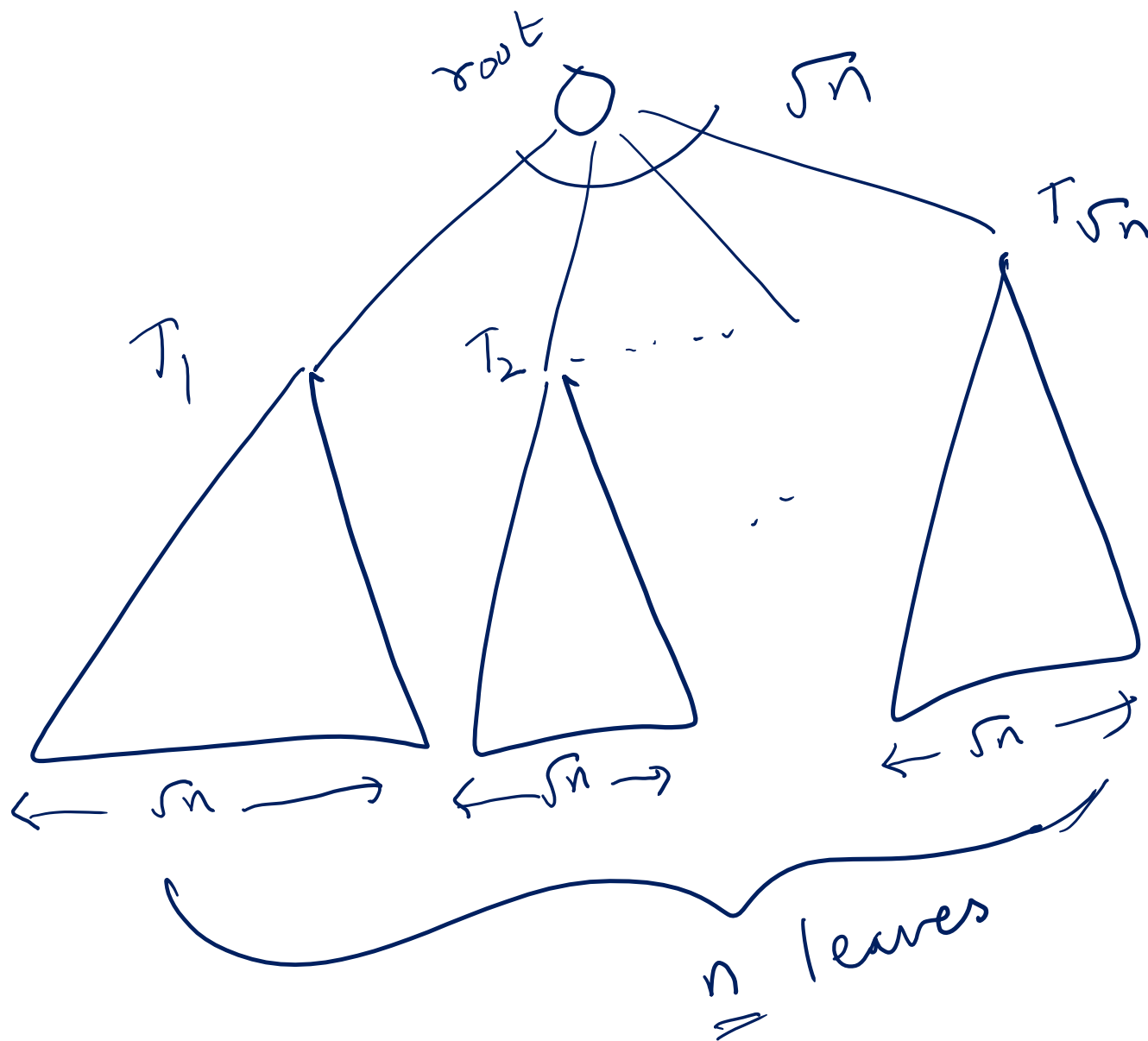
RHS $\uparrow\uparrow$

$$x = \frac{n}{\log n}$$

LHS:

$$\frac{n}{\log n} \times \log\left(\frac{n}{\log n}\right) = \frac{n}{\log n} (\log n - \log \log n)$$

$$= \cancel{n} - \frac{n \log \log n}{\log n}$$



Root : \sqrt{n} children

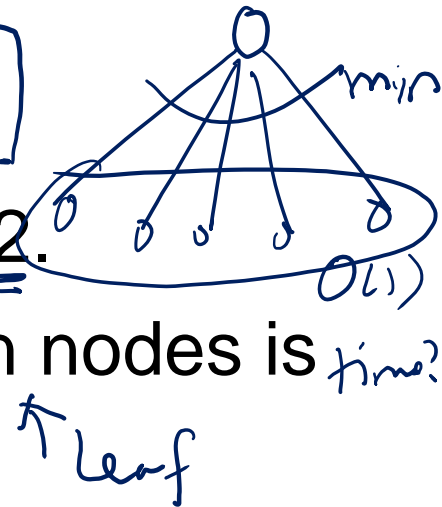
$\rightarrow \sqrt{n}$ subtrees

T_1 : a D.L. tree for \sqrt{n} nodes as leaves

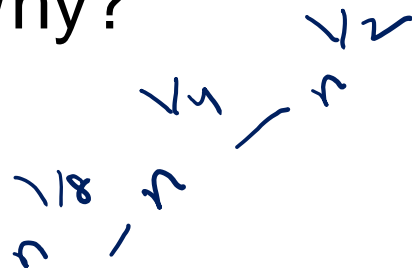
Doubly Logarithmic Tree

$$n = 2^{2^k}$$

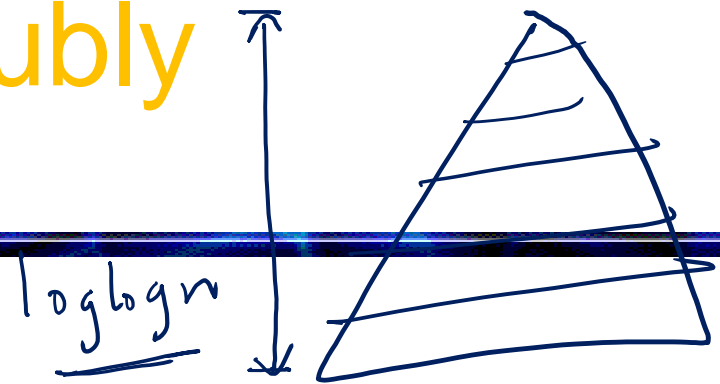
- Some claims: *induction on i*
 - Number of nodes at level i is $2^{2^k - 2^{k-i}}$.
 - Number of nodes at the k th level is $n/2$.
 - Depth of a doubly logarithmic tree of n nodes is $k+1 = \log \log n + 1$.



- To compute the minimum using a doubly logarithmic tree:
 - Each internal node performs the min operation does not suffice.
 - Why?




Minima Using the Doubly Logarithmic Tree



- Intuition:

- Should spend only $O(1)$ time at each internal node.
- Use the $O(1)$ time algorithm at each internal node.

•  At each internal node of level i , if there are c_i children, use c_i^2 processors.

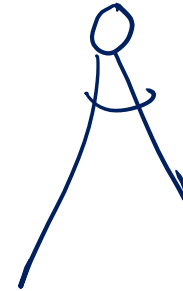
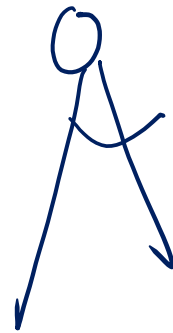
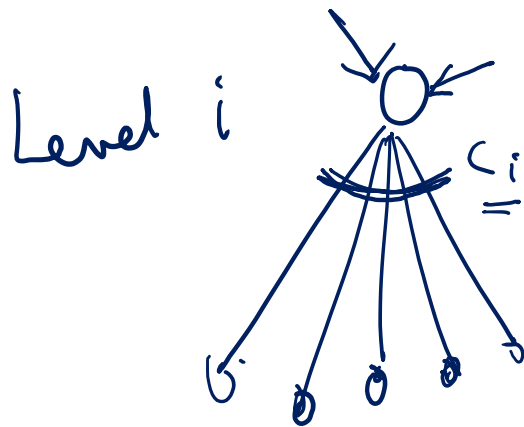
- Minima takes $O(1)$ time at each level.
- Also, No. of nodes at level i x No. of processors used = $2^{2^k - 2^{k-i}} \cdot (2^{2^{k-i-1}})^2 = 2^{2^k} = n$.

$$\text{degree} = C_i$$

$$\# \text{Proc's} = C_i^2$$

$$\# \text{Proc's at level } i = n$$

$$\text{Time for minima} = O(1)$$



} # nodes at level i
 $= n_i$

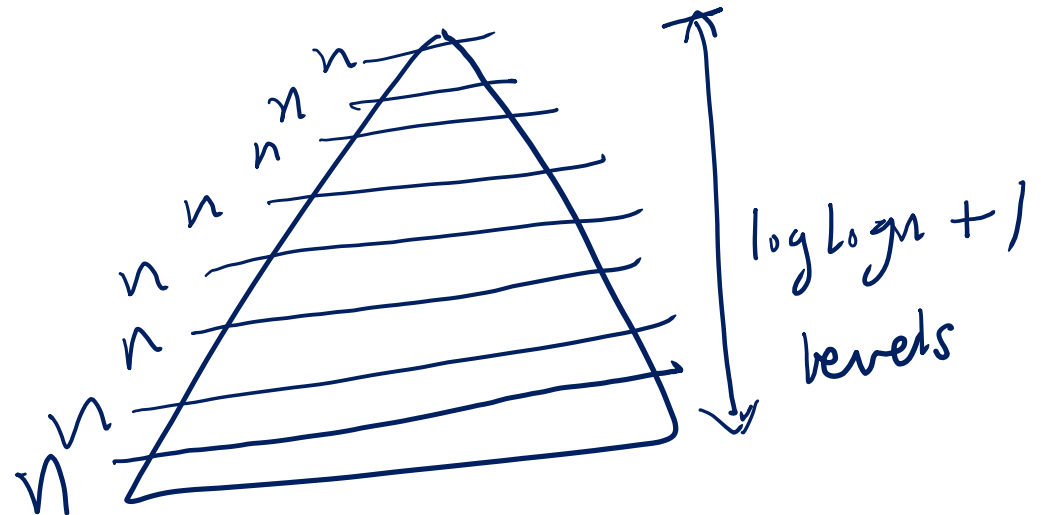
* n_i 's
 2^2 comparisons

$$n_i \times C_i^2 \text{ Processors} = 2^{2^k - 2^{k-i}} \times \left(2^{2^{k-i-1}} \right)^2$$

$$= 2^{2^k - 2^{k-i}} \times 2^{2^{k-i}} = 2^{2^k} = n$$

Minima Using a Doubly Logarithmic Tree

- Second, slightly improved result:
 - With n processors, can find the minima of n numbers in $O(\log \log n)$ time.
 - Total work = $O(n \log \log n)$.
- Still suboptimal by a factor of $O(\log \log n)$.
- We now introduce a technique to achieve optimality.



Accelerated Cascading



- Our two algorithms:
 - Algorithm 1: A slow but optimal algorithm. *optimal*
 - Binary tree based: $O(\log n)$ time, $O(n)$ work.
 - Algorithm 2: A fast but non-optimal algorithm
 - Doubly Logarithmic tree based: $O(\log \log n)$ time, $O(n \log \log n)$ work.
- The accelerated cascading technique suggests combining two such algorithms to arrive at an optimal algorithm *for the problem* \uparrow *for the same problem*
 - Start with the slow but optimal algorithm till the problem is small enough
 - Switch over to the fast but non-optimal algorithm.

Accelerated Cascading

non optimal by
a factor of $O(\log \log n)$

- The binary tree based algorithm starts with an input of size n .
- Each level up the tree reduces the size of the input by a factor of 2.
- In $\log \log \log n$ levels, the size of the input reduces to $n/2^{\log \log \log n} = \underline{n/\log \log n}$.
- Now switch over to the fast algorithm with $n/\log \log n$ processors, needing $O(\log \log (n/\log \log n))$ time.

$$\begin{aligned} \text{Time} &= O(\log \log n') \\ &= O(\log \log n) \end{aligned}$$

$$\begin{aligned} n' \text{ Procs} &= \frac{n}{\log \log n} \\ \text{work} &= O(n' \log \log n') = O\left(\frac{n \times \log \log n}{\log \log n}\right) \\ &= O(n) \end{aligned}$$

Final Result

$$n' = \frac{n}{2^{\log \log \log n}} = \frac{n}{\log \log n}$$

- Total time = $O(\log \log \log n)$ + $O(\log \log n)$.
 - Total work = $O(n)$.
 - Need CRCW model.
 - Where did we need the CRCW model?
- input size after $\log \log \log n$ levels of bin tree = n'
- DL tree alg on an input of size = n'
- work = $O(n' \log \log n') = O\left(\frac{n}{\log \log n} \log \log \left(\frac{n}{\log \log n}\right)\right)$
- $\log \log \left(\frac{n}{\log \log n}\right) \approx \log \log n$
- $\log \log n = O(1)$
- $\log \log \left(\frac{n}{\log \log n}\right) = O(1)$