

Lecture 25

Tuesday, 12 October 2021 10:01 AM

$$a, d > 0 \quad b \geq 1$$

Recurrence:

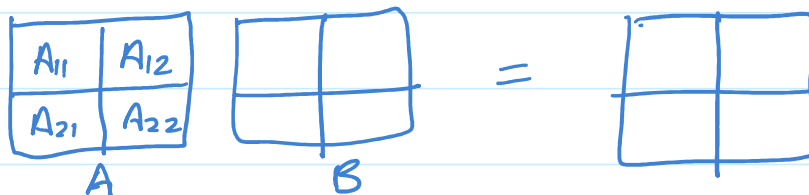
$$T(n) = \underbrace{a}_{\substack{\text{No of} \\ \text{subproblems}}} T\left(\underbrace{\frac{n}{b}}_{\substack{\text{size of the} \\ \text{subproblem}}}\right) + \underbrace{O(n^d)}_{\text{Divide + Combine step}}$$

Eg.

Merge/Quick Sort: $T(n) = 2T(n/2) + O(n)$

Min Pairwise Distance: $T(n) = 2T(n/2) + O(n)$

Matrix Product: $T(n) = 7T\left(\frac{n}{2}\right) + O(n^2)$



Lemma 1 : For $x < 1$, $1 + x + x^2 + x^3 + \dots \leq \frac{1}{1-x} = O(1)$ if x is constant

Lemma 2 : For $x > 1$, $1 + x + x^2 + \dots + x^{n-1} = \frac{x^n - 1}{x - 1} \leq \frac{x^n}{x - 1} = O(x^n)$ if x is constant

Lemma 3 : $a^{\log_b(n)} = n^{\log_b(a)}$

(Proof: If you take $\log_b(\)$ on both sides, then)
LHS = $\log_b(n) \log_b(a) = \text{RHS}$)

$$T(n) = a T(n/b) + c n^d$$

$$= a \left(a T\left(\frac{n}{b^2}\right) + c \left(\frac{n}{b}\right)^d \right) + c n^d$$

$$= a^2 T\left(\frac{n}{b^2}\right) + c n^d \left(1 + \frac{a}{b^d} \right)$$

$$= a^2 T\left(\frac{n}{b^2}\right) + cn^d \left(1 + \frac{a}{b^d}\right)$$

$$= a^2 \left(a T\left(\frac{n}{b^3}\right) + c \left(\frac{n}{b^2}\right)^d \right) + cn^d \left(1 + \frac{a}{b^d}\right)$$

$$= a^3 T\left(\frac{n}{b^3}\right) + cn^d \left(1 + \frac{a}{b^d} + \frac{a^2}{b^{2d}}\right)$$

⋮

$$i = \log_b n \quad = a^i T\left(\frac{n}{b^i}\right) + cn^d \left(1 + \frac{a}{b^d} + \frac{a^2}{b^{2d}} + \dots + \frac{a^{i-1}}{b^{(i-1)d}}\right)$$

$$= \underbrace{a^{\log_b n} \cdot T(1)}_{n^{\log_b a}} + \boxed{\dots}$$

$$T(n) = \begin{cases} n^{\log_b a} + n^d \log_b n & \text{if } \frac{a}{b^d} = 1 \\ n^{\log_b a} + n^d \cdot 1 & \text{if } \frac{a}{b^d} < 1 \\ n^{\log_b a} \cdot \dots & \text{if } \frac{a}{b^d} > 1 \end{cases}$$

$$\left[n^{\log_b a} + n^d \cdot \left(\frac{a}{b^d} \right)^{\log_b n} \right] \quad \text{if } \frac{a}{b^d} > 1$$

Case 1 + Case 2: $a \leq b^d$

$$\log_b a \leq \log_b b^d = d$$

$$\underbrace{n^{\log_b a}}_{\text{LHS}} \leq n^d$$

Case 3:

$$\text{RHS} = \frac{n^d \cdot a^{\log_b n}}{b^{\log_b n^d}} = \frac{n^d \cdot n^{\log_b a}}{(n^d)^{\log_b b}} \quad \text{|| LHS}$$

<u>Master's Theorem.</u> $T(n) =$	{	$O(n^d \log_b n)$	if $\frac{a}{b^d} = 1$
		$O(n^d)$	if $\frac{a}{b^d} < 1$
		$O(n^{\log_b a})$	if $\frac{a}{b^d} > 1$

Quick Sort: $T(n) = 2T\left(\frac{n}{2}\right) + O(n)$

$$a = 2 \quad b = 2 \quad d = 1 \quad \frac{a}{b^d} = 1$$

$$T(n) = O(n \log_2 n)$$

Matrix Product: $T(n) = 7 T(n/2) + O(n^2)$

$$a = 7 \quad b = 2 \quad d = 2 \quad \frac{a}{b^d} = \frac{7}{4} > 1$$

$$O(n^{\log_2 7})$$

Integer Product

Method I

$$a = a_1 2^{n/2} + a_0$$

$$b = b_1 2^{n/2} + b_0$$

$$0 - 2^1 \times \dots \times 2^0$$

$$ab = \underbrace{(a_1, b_1)}_{\textcircled{ii}} 2^n + (a_1, b_0 + a_0, b_1) * 2^{\frac{n}{2}} + \underbrace{a_0, b_0}_{\checkmark \textcircled{ii}}$$

$$\underbrace{(a_1 + a_0)(b_1 + b_0)}_{\textcircled{i}} - \underbrace{a_1, b_1}_{\textcircled{ii}} - \underbrace{a_0, b_0}_{\textcircled{iii}}$$

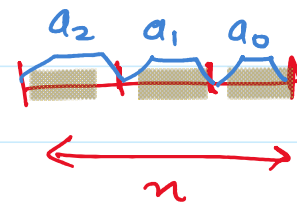
$$T(n) = 3 T\left(\frac{n}{2}\right) + O(n)$$

$$= O(n^{\log_2 3}) \quad (\text{By Master's Thm})$$

Method II

$$a = a_2 2^{\frac{2n}{3}} + a_1 2^{\frac{n}{3}} + a_0$$

$$b = b_2 2^{\frac{2n}{3}} + b_1 2^{\frac{n}{3}} + b_0$$



Assume \$n = \text{pow}(3)\$

$$ab = (a_2, b_2) 2^{\frac{4n}{3}} + (a_2, b_1 + a_1, b_2) 2^{\frac{3n}{3}} + () 2^{\frac{2n}{3}} + () 2^{\frac{n}{3}} + a_0, b_0$$

$$ab = (a_2 b_2) 2^{\frac{4n}{3}} + (a_2 b_1 + a_1 b_2) 2^{\frac{3n}{3}} + () 2^{\frac{2n}{3}} + () 2^{\frac{n}{3}} + a_0 b_0$$

H. W. - Represent all five terms as add/sub of five products.

$$\begin{aligned} T(n) &= 5 T\left(\frac{n}{3}\right) + O(n) \\ &= O(n^{\log_3 5}) \end{aligned}$$

Method III
FFT

$$\begin{aligned} a &= a_n 2^n + a_{n-1} 2^{n-1} + \dots + a_1 2 + a_0 \quad \left] \leq \begin{matrix} n+1 \\ \text{bits} \end{matrix} \right. \\ a(x) &= a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0 \end{aligned}$$

$$\begin{aligned} b &= b_n 2^n + b_{n-1} 2^{n-1} + \dots + b_1 2 + b_0 \quad \left] \leq \begin{matrix} n+1 \\ \text{bits} \end{matrix} \right. \\ b(x) &= b_n x^n + b_{n-1} x^{n-1} + \dots + b_1 x + b_0 \end{aligned}$$

Compute $c(x) = a(x) \cdot b(x)$ in $O(n \log n)$ time

Compute $C(x) = a(x) \cdot b(x)$ in $O(n \log n)$ time
 $= C_{2n} x^{2n} + \dots + C_1 x + C_0$

$$C_i = \underbrace{a_i b_0 + a_{i-1} b_1 + \dots + a_1 b_{i-1} + a_0 b_i}_{2 \text{ binary bits}} \leq i+1$$

$$\text{Output } C(2) = C_{2n} 2^{2n} + C_{2n-1} 2^{2n-1} + \dots + C_2 2^2 + C_1 2 + C_0$$

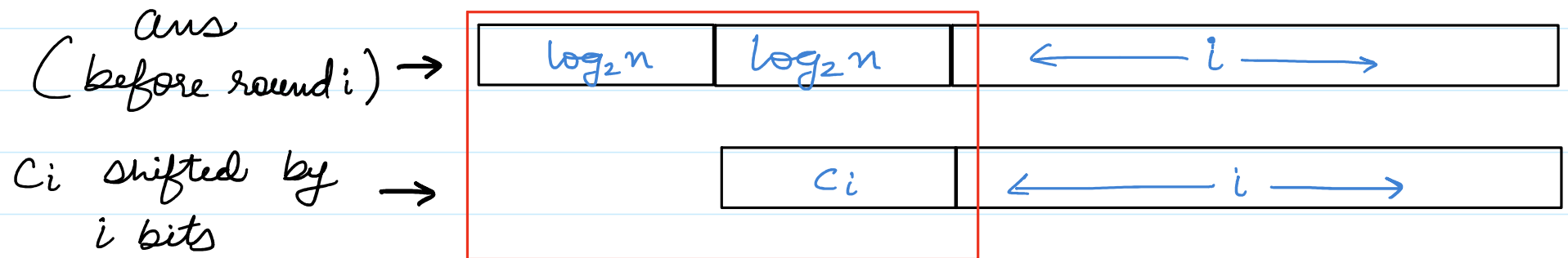
$$\text{Trivial} = O(n) * O(n) = O(n^2)$$

Algorithm

- ① $ans = C_0$
- ② For $i=1$ to $2n$: $ans = ans + C_i \cdot \underbrace{2^i}_{\text{bit-shifting}}$
- ③ Return ans .

Before Round i : $\text{Ans} = C_{i-1} 2^{i-1} + \dots + C_1 2 + C_0 \leq i^2 \cdot 2^i$

Before Round i , number of bits taken by ans $\leq i + 2 \log_2 n$



Computing sum takes
 $\triangleleft O(\log_2 n)$ time

Time taken by step 2 - $O(\log n)$

\Rightarrow Time to compute $C(2) = O(n \log_2 n)$

REMARK: Product of 2 n -bit numbers take $O(n \log n)$ time as long as n is small.

In practice, FFT involves *rounding errors* of complex numbers, so to tackle it there is extra overhead in time complexity.

See this for further reference:

https://en.wikipedia.org/wiki/F%26C3%BCr%27s_algorithm