

ADS - COURSEWORK 1 - 1/725018

1) a) Solve the recurrence $T(n) = 3T(\sqrt{n}) + \lg n$.

$$\text{let } m = \lg n \iff 2^m = n$$

$$\begin{aligned} \therefore T(n) &= 3T(\sqrt{n}) + \lg(n) \\ &= T(2^m) = 3T(2^{m/2}) + m \\ &= T(2^m) = 3T(2^{m/2}) + m \end{aligned}$$

now let $S(m) = T(2^m)$

$$\therefore T(2^m) = S(m) = 3S\left(\frac{m}{2}\right) + m$$

by using master theorem on $S(m) = 3S(m/2) + m$

we have $a=3$ $b=2$ $k=1$

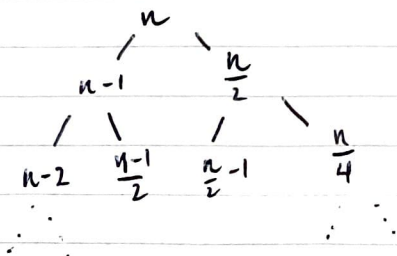
$c = \lg(3)$ since $c > 1$ then $S(m) = \Theta(m^{\lg 3})$

$$\begin{aligned} T(n) &= T(2^m) = S(m) = \Theta(m^{\lg 3}) \\ &= \Theta((\lg n)^{\lg 3}) \end{aligned}$$

□

b) Prove the asymptotic upper bound on:

$$T(n) = T(n-1) + T(n/2) + n$$



It is clear to see that the worst case scenario is when both sides decrease by 1. \therefore at each node we get 2^i elements. since it decreases by one our height will be n so as a result at the leaf node we will have 2^n elems. as a result this is the worst case run time $O(2^n)$

Similarly the best case run time is when we exclusively divide by $1/2$ and so our tree will have by letting $n = 2^k \rightarrow 2^k \therefore k = \lg n$.

$$\begin{aligned} T(2^k) &= T\left(\frac{2^k}{2}\right) + 1 \\ &= T(2^{k-1}) + 1 \\ &= (T(2^{k-2}) + 1) + 1 \\ &= T(2^{k-2}) + 2 \\ &= \vdots \\ &= T(2^{k-k}) + k \\ &= T(2^0) + k \\ &= T(1) + k \\ &= 1 + \lg n \end{aligned}$$

$$\therefore T(n) = O(1 + \lg n) = O(\lg n). \quad \square$$

e) we have that $T(n) = a T(n/4) + \Theta(n^2)$

in the worst case by the master theorem, $T(n) = O(n^{\log_4 a})$

$$\begin{aligned} \therefore \log_4 a &< \log_2 7 & (\text{so it is better than Strassen's}) \\ \log_2 \sqrt{a} &< \log_2 7 \\ \sqrt{a} &< 7 \\ a &< 49 \end{aligned}$$

$\therefore a$ can be at most 48.

\square

2) a) Compute DFT of vector $(0, 1, 2, 5)$

$$\text{DFT}_4 (0, 1, 2, 5)$$

$$A(x) = 0x^0 + 1x^1 + 2x^2 + 5x^3$$

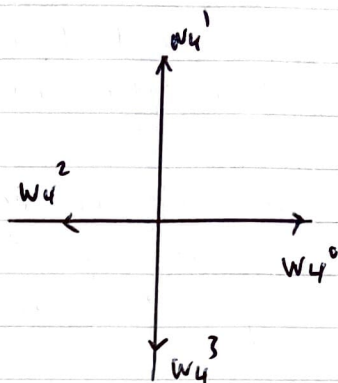
$$A(x) = x + 2x^2 + 5x^3$$

$$w_4^0 = (e^{2\pi i/4})^0 = e^0 = 1$$

$$w_4^1 = (e^{2\pi i/4})^1 = e^{\pi i/2} = i$$

$$w_4^2 = (e^{2\pi i/4})^2 = e^{\pi i} = -1$$

$$w_4^3 = (e^{2\pi i/4})^3 = e^{3\pi i/2} = -i$$



$$A(w_4^0) = 0 + 2(1)^2 + 5(1)^3 = 0 + 2 + 5 = 7$$

$$A(w_4^1) = i + 2(i)^2 + 5(i)^3 = i - 2 - 5i = -2 - 4i$$

$$A(w_4^2) = -1 + 2(-1)^2 + 5(-1)^3 = -1 + 2 - 5 = -4$$

$$A(w_4^3) = (-i) + 2(-i)^2 + 5(-i)^3 = -i - 2 + 5i = -2 + 4i$$

$$\text{DFT}_4 ((0, 1, 2, 5)) = (7, -2-4i, -4, -2+4i)$$

b) We know that $A(x) = q(x)(x - x_0) + r$ and as a result we can calculate r by horner's method in $O(n)$ time. We only need to calculate the coefficients of $q(x)$

$$A(x) = q(x)(x - x_0) + r$$

$$a_0 + a_1x^1 + \dots + a_nx^n = (b_0 + b_1x^1 + \dots + b_{n-1}x^{n-1})(x - x_0) + r$$

by construction: we have:

$$a_0 + a_1x^1 + \dots + a_nx^n = b_0x + b_1x^{1+1} + \dots + b_{n-1}x^n - x_0(b_0 + b_1x^1 + \dots + b_{n-1}x^{n-1}) + r$$

$$= a_0 + a_1 x + \dots + a_n x^n = b_0 x + b_1 x^2 + \dots + b_{n-1} x^n - x_0 (q(n)) + r$$

as a result we can see that:

$$a_0 = -x_0 b_0 + r$$

$$a_1 = b_0 - x_0 b_1$$

$$a_2 = b_1 - x_0 b_2$$

\vdots

$$a_{n-1} = b_{n-2} - x_0 b_{n-1}$$

$$a_n = b_{n-1}$$

$$\therefore b_0 = \frac{r - a_0}{x_0} \quad \text{and} \quad a_n = b_{n-1}$$

$$\text{and } b_i = \frac{b_{i-1} - a_i}{x_0}$$

which we can calculate directly for $i=2$ upto n and as a result this takes $O(n)$ time. We can do this since we have a_i , x_0 and r .

\therefore total time would be $O(n)$ for evaluating r and $O(n)$ for the coefficients $\therefore O(n) + O(n) = O(n)$ \square

c) Since $p(x) = 0 \quad \forall \quad z_0, z_1, \dots, z_{n-1}$

We can construct a polynomial of degree $n-1$ with $(x - z_0)(x - z_1) \dots (x - z_{n-1}) = Q(x)$

\therefore we know $p(x) = Q(x) + r(x)$ $r(x)$ of degree 2.

\therefore by polynomial division $p(x)/Q(x)$ using ppt we can get $r(x)$ in $O(n \log n)$ time.

now we need to solve $Q(x)$ to find the coefficients of $A(x)$

$$\text{since } Q(x) = (x - z_0)(x - z_1) \dots (x - z_{n-1})$$

we can use FFT to ~~find~~ multiply it out using polynomial multiplication where we use a divide and conquer strategy to recursively split the polynomial into 2 and multiply the resulting polynomial with FFT

\therefore Our recurrence looks like this:

$$T(n) = 2T(n/2) + O(n \lg n)$$

by master theorem $a=2$ $b=2$ $c = \log_2(2) = 1$

since $c = k$ of n \therefore our recurrence has

$$O(n \lg^{k+1} n) \text{ complexity} = O(n \lg^2 n)$$

which then we can use to read coefficients of ~~polynomial~~ the polynomial so total time = $O(n \lg^2 n) + O(n \lg n)$

$$= O(n \lg^2 n) \quad \square$$

d) let $A(x) = \sum_{i=0}^{10n} a_i x^i$ where i is the value between 0 and $10n$

~~a_i~~ $a_i = a_i = n^i$ of times value i occurred
and $i = \text{value}$

$$\text{like write for } B(x) = \sum_{i=0}^{10n} b_i x^i$$

by using FFT to multiply $A(x)$ and $B(x)$ we get it done in $O(n \log n)$ time.

where the coefficients of this new fraction are the number of times each element occurs and the power is the element

This is due to properties of matrix multiplication.

Since $A(x) \cdot B(x) = \sum_{i=0}^{2n-1} c_i x^i$ for example the first term would be: $b_0 \cdot a_0 \cdot x^{0+0}$ value in c .

↓
n^o of times it occurs in c

□