

Data Structure and Algorithms

Lab4

Submitted to: Ms. Sehar Waqar

Submitted by: SABA

Roll no: 2019-CE-04

Department of Computer Engineering

UET,Lahore

Homework Question:

Matrix multiplication: Suppose that we have $n \times n$ matrices X and Y and we'd like to multiply them.

- a) What is the running time of the standard algorithm (that computes the inner product of rows of X and columns of Y)? You can assume that simple arithmetic operations, like multiplication, take constant time ($O(1)$).

SQUARE-MATRIX-MULTIPLY(A,B)

- | | |
|--|--|
| 1. $n = A.rows$ | → cost = c_0 , 1 time |
| 2. let C be a new $n \times n$ matrix | |
| 3. for $i = 1$ to n | → cost = c_1 , n times |
| 4. for $j = 1$ to n | → cost = c_2 , $(n \cdot n)$ times |
| 5. $c_{ij} = 0$ | → cost = c_3 , $(n \cdot n)$ times |
| 6. for $k = 1$ to n | → cost = c_4 , $(n \cdot n \cdot n)$ times |
| 7. $c_{ij} = c_{ij} + a_{ik} \cdot b_{kj}$ | → cost = c_5 , $(n \cdot n \cdot n)$ times |
| 8. return C | → cost = c_6 , 1 time |

$$T(n) = c_0 + c_6 + c_1n + (c_2 + c_3)(n \cdot n) + (c_4 + c_5)(n \cdot n \cdot n)$$

$$T(n) = O(n^3)$$

Because each of the triply-nested for loops runs exactly n iterations, and each execution of line 7 takes constant time, the SQUARE-MATRIX-MULTIPLY procedure takes $O(n^3)$ time.

- b) Now let's divide up the problem into smaller chunks like this, where the eight $n/2 \times n/2$ sub-matrices (A, B, C, D, E, F, G, H) are each quarters of the original matrices, X and Y : We now have a divide and conquer strategy! Find the recurrence relation of this strategy and the runtime of this algorithm.

SQUARE-MATRIX-MULTIPLY-RECURSIVE (A,B)

1. $n = A.rows$
2. let C be a new $n \times n$ matrix
3. if $n == 1$
4. $c_{11} = a_{11} \cdot b_{11}$
5. else partition A, B and C
6. $C_{11} = \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{11}, B_{11}) + \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{12}, B_{21})$
7. $C_{12} = \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{11}, B_{12}) + \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{12}, B_{22})$
8. $C_{21} = \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{21}, B_{11}) + \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{22}, B_{21})$
9. $C_{22} = \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{21}, B_{12}) + \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{22}, B_{22})$

Ans: Recurrence relation:**Base Case :** When $n = 1$, we perform just one scalar multiplication.So, $T(1) = \Theta(1)$ **Recursive Case :** When $n > 1$, we partitioning the matrices in $\Theta(1)$ time, and recursively call SQUARE-MATRIX-MULTIPLY-RECURSIVE (A,B) a total of 8 times.Time taken by one recursive call = $T(n/2)$ Time taken by eight recursive calls = $8T(n/2)$ Time taken by addition of matrices = $\Theta(n^2)$

$$T(n) = \begin{cases} \Theta(1), & \text{if } n = 1 \\ 8T\left(\frac{n}{2}\right) + \Theta(n^2), & \text{if } n > 1 \end{cases}$$

Running Time by using Master theorem: $a = 8$, $b=2$, $d = 2$ $b^d = 2^2 = 4$ As, $a > b^d \rightarrow T(n) = O(n^{\log_b a}) = O(n^{\log_2 8})$ **$T(n) = O(n^3)$** **c)** Can we do better? It turns out we can by calculating only 7 of the subproblems:

$P1 = A(F - H)$

$P2 = (A + B)H$

$P3 = (C + D)E$

$P4 = D(G - E)$

$P5 = (A + D)(E + H)$

$P6 = (B - D)(G + H)$

$P7 = (A - C)(E + F)$

And we can solve XY by

$$XY = \begin{bmatrix} P4 + P5 - P2 + P6 & P1 + P2 \\ P3 + P4 & P1 + P5 - P3 - P7 \end{bmatrix}$$

We now have a more efficient divide and conquer strategy! What is the recurrence relation of this strategy and what is the runtime of this algorithm?

Ans: We use Strassen's method. The key to Strassen's method is to make the recursion tree slightly less bushy. That is, instead of performing eight recursive multiplications of $n/2 \times n/2$ matrices, it performs only seven.

So, the recurrence relation is :

$$T(n) = \begin{cases} \Theta(1), & \text{if } n = 1 \\ 7T\left(\frac{n}{2}\right) + \Theta(n^2), & \text{if } n > 1 \end{cases}$$

Running Time by using Master theorem: $a = 7$, $b=2$, $d = 2$ $b^d = 2^2 = 4$ As, $a > b^d \rightarrow T(n) = O(n^{\log_b a}) = O(n^{\log_2 7})$ **$T(n) = O(n^{2.81})$**

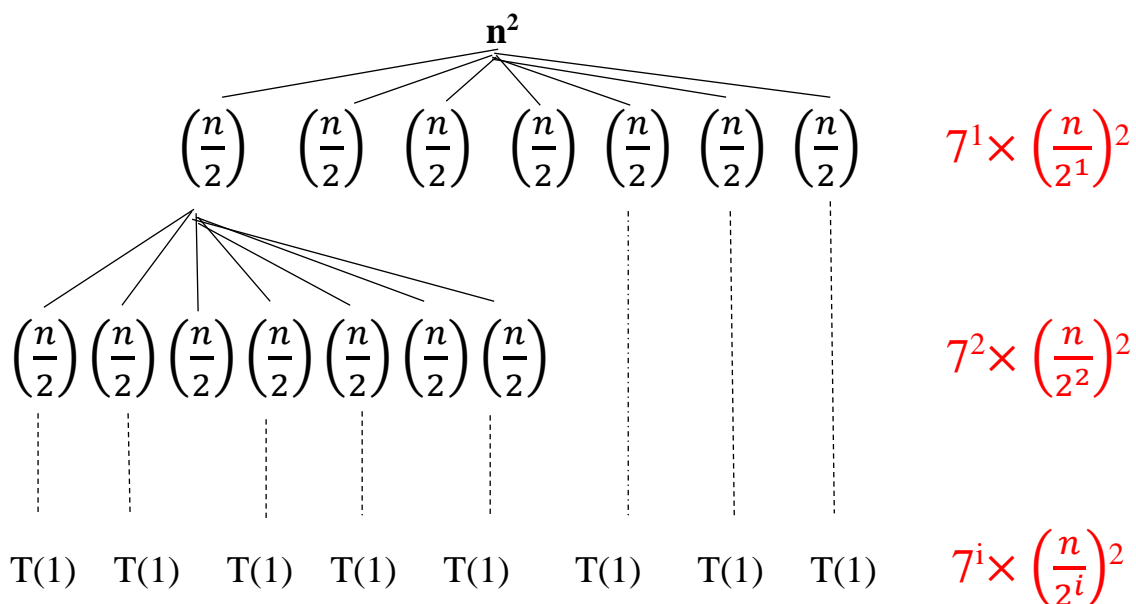
d) Claim: The algorithm runs in time $T(n) = O(n^2 \log(n))$.

Proof: At the top level, we have 7 operations adding/subtracting $n/2 \times n/2$ matrices, which takes $O(n^2)$ time. At each subsequent level of the recursion, we increase the number of subproblems by a constant factor (7), and the subproblems only become smaller. Therefore, the running time per level can only increase by a constant factor. By induction, since for the top level it is $O(n^2)$, it will be $O(n^2)$ for all levels. There are $O(\log(n))$ levels, so in total the running time is $T(n) = O(n^2 \log(n))$. What is the error in this reasoning?

Ans: Error: The above proof says that it is $O(n^2)$ for all levels which is wrong.

Proof:

$$T(n) = \begin{cases} \Theta(1), & \text{if } n = 1 \\ 7T\left(\frac{n}{2}\right) + \Theta(n^2), & \text{if } n > 1 \end{cases}$$



Branching factor: 7

Cost at each level: $7^i \times \left(\frac{n}{2^i}\right)^2$

Depth of tree: $\frac{n}{2^i} = 1 \rightarrow i = \log n$

Number of leaves: $7^{\log n} = n^{\log 7} = n^{2.81}$

Total cost = $\sum_{i=0}^{\log(n)-1} 7^i \times \left(\frac{n}{2^i}\right)^2 + O(n^{2.81})$

$T(n) = O(n^{2.81})$

Hence, we prove that cost at each level is not $O(n^2)$.

Recurrence Questions:

Question1:

There exist different ways to solve the recurrence relation $T(n) = 2 \cdot T(n/2) + n$ with $T(1) = 1$. From lectures, we have seen that $T(n)$ is exactly $n(1 + \log(n))$ when n is a power of 2.

a) What is the exact solution to $T(n) = 3T(n/3) + n$ with $T(1) = 3$, when n is a power of 3?

$$\begin{aligned}\text{Ans: } T(n) &= 3T\left(\frac{n}{3}\right) + n \rightarrow 3^1 T\left(\frac{n}{3^1}\right) + 1n \\ &= 3 \left[3T\left(\frac{n}{9}\right) + \left(\frac{n}{3}\right) \right] + n \\ &= 9T\left(\frac{n}{9}\right) + 2n \rightarrow 3^2 T\left(\frac{n}{3^2}\right) + 2n \\ &= 9 \left[3T\left(\frac{n}{27}\right) + \left(\frac{n}{9}\right) \right] + 2n \\ &= 27 T\left(\frac{n}{27}\right) + 3n \rightarrow 3^3 T\left(\frac{n}{3^3}\right) + 3n\end{aligned}$$

General equation:

$$T(n) = 3^i T\left(\frac{n}{3^i}\right) + in$$

$$\text{Let, } \frac{n}{3^i} = 1 \rightarrow n = 3^i \rightarrow i = \log_3 n$$

So, now general equation becomes

$$\begin{aligned}T(n) &= 3^{\log_3 n} T(1) + n \log_3 n \\ &= 3n + n \log_3 n \quad \therefore T(1) = 3\end{aligned}$$

$$T(n) = n(3 + \log_3 n)$$

b) What is the exact solution to $T(n) = 3T(n/3) + 3n$ with $T(1) = 1$, when n is a power of 3?

$$\begin{aligned}\text{Ans: } T(n) &= 3T\left(\frac{n}{3}\right) + 3n \rightarrow 3^1 T\left(\frac{n}{3^1}\right) + 3 \times 1n \\ &= 3 \left[3T\left(\frac{n}{9}\right) + n \right] + 3n \\ &= 9T\left(\frac{n}{9}\right) + 6n \rightarrow 3^2 T\left(\frac{n}{3^2}\right) + 3 \times 2n \\ &= 9 \left[3T\left(\frac{n}{27}\right) + \left(\frac{n}{3}\right) \right] + 6n \\ &= 27 T\left(\frac{n}{27}\right) + 9n \rightarrow 3^3 T\left(\frac{n}{3^3}\right) + 3 \times 3n\end{aligned}$$

General equation:

$$T(n) = 3^i T\left(\frac{n}{3^i}\right) + 3in$$

$$\text{Let, } \frac{n}{3^i} = 1 \rightarrow n = 3^i \rightarrow i = \log_3 n$$

So, now general equation becomes

$$\begin{aligned}T(n) &= 3^{\log_3 n} T(1) + 3n \log_3 n \\ &= n + 3n \log_3 n \quad \therefore T(1) = 1\end{aligned}$$

$$T(n) = n(1 + 3 \log_3 n)$$

Question2:

Use any of the methods we've seen in class so far to give big-Oh solutions to the following recurrence relations. You may treat fractions like $n/2$ as either $\text{floor}(n/2)$ or $\text{ceiling}(n/2)$, whichever you prefer.

a) $T(n) = 3T(n/9) + \sqrt{n}$ for $n \geq 9$, and $T(n)=1$ for $n < 9$.

Ans: $a=3$, $b=9$, $d=1/2$

$$b^d = 9^{1/2} = 3$$

$$\text{As, } a = b^d \rightarrow T(n) = O(n^d \log n)$$

So, $T(n) = O(\sqrt{n} \log_3 n)$

b) $T(n) = T(n-4) + n$ for $n \geq 4$, and $T(n) = 1$ for $n < 4$.

Ans: $T(n) = T(n-4) + n$

$$= [T(n-8) + (n-4)] + n$$

$$= [T(n-12) + (n-8)] + (n-4) + n$$

$$= T(n-16) + (n-12) + (n-8) + (n-4) + n$$

General equation:

$$T(n) = T(n-k) + (n-(k-4)) + (n-(k-8)) + \dots + (n-4) + n$$

Let, $n-k=0 \rightarrow n=k$

$$T(n) = T(n-n) + (n-n+4) + (n-n+8) + \dots + (n-4) + n$$

$$= T(0) + 4 + 8 + 12 + \dots + n$$

$$= 1 + 4[1+2+3+\dots + n]$$

$$= 1 + 4 \left(\frac{n(n+1)}{2} \right)$$

$$= 1 + 2n(n+1)$$

$$= 1 + 2n^2 + 2n$$

$$= 2n^2 + 2n + 1$$

Hence, $T(n) = O(n^2)$

c) $T(n) = 6T(n/4) + n^2$ for $n \geq 4$, and $T(n) = 1$ for $n < 4$.

Ans : $a=6$, $b=4$, $d=2$

$$b^d = 4^2 = 16$$

$$\text{As, } a < b^d \rightarrow T(n) = O(n^d)$$

So, $T(n) = O(n^2)$

d) $T(n) = 5T(n/2) + n^2$ for $n \geq 2$, and $T(n) = 1$ for $n < 2$.

Ans: $a = 5$, $b=2$, $d=2$

$$b^d = 2^2 = 4$$

$$\text{As, } a > b^d \rightarrow T(n) = O(n^{\log_b a})$$

So, $T(n) = O(n^{\log_2 5}) = O(n^{2.32})$

Question3:

Consider the following algorithm, which takes as input an array A:

def printStuff(A):

 n = len(A)

 if n <= 4:

 return

 for i in range(n):

 print(A[i])

 printStuff(A[:n/3]) # recurse on first n/3 elements of A

 printStuff(A[2*n/3:]) # recurse on last n/3 elements of A

 return

What is the asymptotic running time of printStuff?

Ans: Recurrence relation:

$$T(n) = \begin{cases} \Theta(1), & \text{if } n \leq 4 \\ 2T\left(\frac{n}{3}\right) + \Theta(n), & \text{if } n > 5 \end{cases}$$

a=2, b=3, d=1

$b^d = 3^1 = 3$

As, $a < b^d \rightarrow T(n) = O(n^d)$

So, $T(n) = O(n)$
