

CS 5800: Algorithms SEC 05 Summer Full 2024 (Seattle)  
Homework 3

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**Question 1:** Section 2.2 describes a method for solving recurrence relations which is based on analyzing the recursion tree and deriving a formula for the work done at each level. Another method is to expand out the recurrence a few times, until a pattern emerges. For instance, let's start with the familiar  $T(n) = 2T(n/2) + \mathcal{O}(n)$ . Think of  $\mathcal{O}(n)$  as being  $\leq cn$  for some constant  $c$ , so:  $T(n) \leq 2T(n/2) + cn$ . By repeatedly applying this rule, we can bound  $T(n)$  in terms of  $T(n/2)$ , then  $T(n/4)$ , then  $T(n/8)$ , and so on, at each step getting closer to the value of  $T(\cdot)$  we do know namely,  $T(1) = \mathcal{O}(1)$ .

$$\begin{aligned} T(n) &\leq 2T(n/2) + cn \\ &\leq 2[2T(n/4) + cn/2] + cn = 4T(n/4) + 2cn \\ &\leq 4[2T(n/8) + cn/4] + 2cn = 8T(n/8) + 3cn \\ &\leq 8[2T(n/16) + cn/8] + 3cn = 16T(n/16) + 4cn \end{aligned}$$

$\vdots$

A pattern is emerging  $\dots$  the general term is

$$T(n) \leq 2^k T(n/2^k) + kcn.$$

Plugging in  $k = \log_2 n$ , we get  $T(n) \leq nT(1) + cn \log_2 n = \mathcal{O}(n \log n)$ .

**Part (a):** Do the same thing for the recurrence  $T(n) = 3T(n/2) + \mathcal{O}(n)$ . What is the general  $k$ th term in this case? And what value of  $k$  should be plugged in to get the answer?

**Solution:**

To find the recurrence of the given base case:  $T(n) = 3T(n/2) + \mathcal{O}(n)$

As we are given to assume that  $\mathcal{O}(n) \leq cn$ ,

Thus we get,

$$T(n) \leq 3T(n/2) + cn \quad (1)$$

We can write  $T(n/2)$  as,

$$T(n/2) \leq 3T(n/4) + (cn/2) \quad (2)$$

Similarly, we can write  $T(n/4)$  as,

$$T(n/4) \leq 3T(n/8) + (cn/4) \quad (3)$$

Similarly, we can write  $T(n/8)$  as,

$$T(n/8) \leq 3T(n/16) + (cn/8) \quad (4)$$

Now, substitute eqn(2) in eqn(1).

$$T(n) \leq 3T(n/2) + cn$$

$$\text{Hence, } T(n) \leq 3(3T(n/4) + (cn/2)) + cn$$

$$\implies \leq 9T(n/4) + 3(cn/2) + cn$$

$$\implies \leq 9T(n/4) + 5(cn/2)$$

We can rewrite this equation as,

$$\implies \leq 3^2 T(n/2^2) + (2cn((3/2)^2 - 1))$$

Hence, the general term for the  $k$ th expansion is:

$$T(n) \leq 3^k T(n/2^k) + kcn$$

Plugging in ( $k = \log_2 n$ ),

$$\implies 3^{\log_2 n} T(n/2^{\log_2 n}) + (\log_2 n)cn$$

$$\implies n^{0.6} T(n/n) + (\log_2 n)cn$$

$$\implies T(1) + (\log_2 n)cn$$

Therefore, the general term for the  $(k)$ th expansion is ( $T(n) \leq 3^k T(n/2^k) + kcn$ ), and plugging in ( $k = \log_2 n$ ), we obtain the time complexity as ( $T(n) = O(n \log n)$ ).

**Part (b): Now try the recurrence  $T(n) = T(n - 1) + \mathcal{O}(1)$ , a case which is not covered by the master theorem. Can you solve this too?**

**Solution:**

To solve this recurrence relation  $T(n) = T(n - 1) + \mathcal{O}(1)$ , we can unfold the above recurrence expression in the following pattern,

$$T(n) = T(n - 1) + \mathcal{O}(1)$$

$$T(n - 1) = T(n - 2) + \mathcal{O}(1)$$

$$T(n - 2) = T(n - 3) + \mathcal{O}(1)$$

$\vdots$

$$T(2) = T(1) + \mathcal{O}(1)$$

$$T(1) = \mathcal{O}(1)$$

From the above equations, we can express the general  $(k)$ th term as,

$$T(n) = T(n - k) + ck$$

Summing up the time complexity of each equation, we can find out the total time complexity,

$$T(n) = \mathcal{O}(1) + \mathcal{O}(1) + \mathcal{O}(1) + \dots + \mathcal{O}(1)(n \text{ times})$$

$$T(n) = \mathcal{O}(n)$$

Therefore, the general (k)th expansion for the given recurrence relation is  $T(n) = T(n - k) + ck$ , and the time complexity is  $T(n) = \mathcal{O}(n)$ , indicating a linear time complexity in terms of the input size n.

## Question 2: Solve the following recurrence relations to give a $\theta$ bound for each of them.

The Master Theorem is given as follows:

If  $T(n) = aT([n/b]) + \mathcal{O}(n^d)$  for some constants  $a > 0$ ,  $b > 1$ , and  $d \geq 0$ , then

$$\begin{cases} T(n) = \mathcal{O}(n^d) & \text{if } d > \log_b a \\ T(n) = \mathcal{O}(n^d \log n) & \text{if } d = \log_b a \\ T(n) = \mathcal{O}(n^{\log_b a}) & \text{if } d < \log_b a \end{cases}$$

Here,  $a \rightarrow$  no. of subproblems,

$b \rightarrow$  the factor by which the problem size is reduced,

$f(n) \rightarrow$  cost of dividing the problem and combining the results.

**Part (a):**  $T(n) = 2T(n/3) + 1$

**Solution:**

Comparing the given recurrence relation with the standard form, we have:

$$a = 2, b = 3, f(n) = 1, d = 0$$

$$\text{Calculating, } \log_b a = \log_3 2 = 0.63$$

$$\text{Since, } d < \log_b a \text{ i.e., } 0 < 0.63$$

This satisfies case 3 of the Master Theorem.

$$\text{Therefore, } T(n) = \theta(n^{\log_3 2}) = \theta(n^{0.63})$$

**Part (b):**  $T(n) = 5T(n/4) + n$

**Solution:**

Comparing the given recurrence relation with the standard form, we have:

$$a = 5, b = 4, f(n) = n, d = 1$$

$$\text{Calculating, } \log_b a = \log_4 5 = 1.16$$

$$\text{Since } d < \log_b a \text{ i.e., } 1 < 1.16$$

Thus, this falls into case 3 of the Master Theorem.

$$\text{Therefore, } T(n) = \theta(n^{\log_4 5}) = \theta(n^{1.16})$$

**Part (c):**  $T(n) = 7T(n/7) + n$

**Solution:**

Comparing the given recurrence relation with the standard form, we have:

$$a = 7, b = 7, f(n) = n, d = 1$$

$$\text{Calculating, } \log_b a = \log_7 7 = 1$$

$$\text{Since, } d = \log_b a \text{ i.e., } 1 = 1$$

Thus, this falls into case 2 of the Master Theorem.

$$\text{Therefore, } T(n) = \theta(n^{\log_b a} \log n) = \theta(n \log n)$$

$$\textbf{Part (d): } T(n) = 9T(n/3) + n^2$$

**Solution:**

Comparing the given recurrence relation with the standard form, we have:

$$a = 9, b = 3, f(n) = n^2, d = 2$$

$$\text{Calculating, } \log_b a = \log_3 9 = 2$$

$$\text{Since, } d = \log_b a \text{ i.e., } 2 = 2$$

Thus, this falls into case 2 of the Master Theorem.

$$\text{Therefore, } T(n) = \theta(n^{\log_b a} \log n) = \theta(n^2 \log n)$$

$$\textbf{Part (e): } T(n) = 8T(n/2) + n^3$$

**Solution:**

Comparing the given recurrence relation with the standard form, we have:

$$a = 8, b = 2, f(n) = n^3, d = 3$$

$$\text{Calculating, } \log_b a = \log_2 8 = 3$$

$$\text{Since, } d = \log_b a \text{ i.e., } 3 = 3$$

Thus, this falls into case 2 of the Master Theorem.

$$\text{Therefore, } T(n) = \theta(n^{\log_b a} \log n) = \theta(n^3 \log n)$$

$$\textbf{Part (f): } T(n) = 49T(n/25) + n^{3/2} \log n$$

**Solution:**

Comparing the given recurrence relation with the standard form, we have:

$$a = 49, b = 25, f(n) = n^{3/2} \log n, d = 3/2 = 1.5$$

$$\text{Calculating, } c = \log_b a = \log_{25} 49 = \log_5 7 = 1.2$$

$$d > \log_b a \text{ i.e., } 1.5 > 1.2$$

According to the Master Theorem:

If  $f(n) = \Omega(n^{c+\epsilon})$  for some  $\epsilon > 0$  and  $af(n/b) \leq kf(n)$ , for some  $k < 1$ , and sufficiently large  $n$ , then  $T(n) = \Theta(f(n))$ .

Let us verify  $f(n)$  to check if it satisfies this condition.

$$f(n) = n^{3/2} \log n$$

Here,  $c + \epsilon = 1.2 + \epsilon$

and we need  $3/2 > 1.2 + \epsilon$ . We can choose  $\epsilon = 0.25$  to satisfy  $1.5 > 1.2 + 0.25$ .

$$af(n/b) = 49(n/25)^{3/2} \log(n/25)$$

$$\implies (49/125)n^{3/2}(\log n - \log 25)$$

Ignoring constant coefficients,  
we can conclude that  $T(n) = \Theta(n^{3/2} \log n)$

**Part (g):**  $T(n) = T(n-1) + 2$

**Solution:**

We will expand the recurrence relation  $T(n) = T(n-1) + 2$ , to see the pattern,  
 $T(n) = T(n-1) + 2$

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

$$T(n-2) = T(n-3) + 2$$

$\vdots$

$$T(2) = T(1) + 2$$

$$T(1) = T(0) + 2$$

Summing up all these equations, we get,

$$T(n) = T(n-1) + T(n-2) + \dots + T(1) + T(0) + 2n$$

We can see that  $T(n)$  grows linearly with  $n$ .

Hence,  $T(n) = \theta(n)$

**Part (h):**  $T(n) = T(n-1) + n^c$ , where  $c \geq 1$  is a constant

**Solution:**

We will expand the recurrence relation to understand the pattern,

$$T(n) = T(n-1) + n^c$$

$$T(n-1) = T(n-2) + (n-1)^c$$

$$T(n-2) = T(n-3) + (n-2)^c$$

$\vdots$

$$T(2) = T(1) + 2^c$$

$$T(1) = T(0) + 1^c$$

Summing up these equations we get,

$$T(n) = T(0) + 2^c + 3^c + \dots + n^c$$

which gives us the solution to the recurrence relation as,  $T(n) = T(0) + \sum_{i=0}^n i^c$

$$\implies T(n) \approx T(0) + n^{c+1}/c + 1$$

Ignoring constant coefficients, we get,  $T(n) = \theta(n^{c+1})$

This means that  $T(n)$  grows asymptotically as  $n^{c+1}$ .

**Reflection:**

Through the exercise problems, I gained insights into solving recurrence relations by recognizing patterns. Additionally, I learned how to apply the Master Theorem to solve such relations.

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