

# Homework #2

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## Question #7

a)  $T(n) = 3T(n-1) - 2T(n-2)$

$$x^2 = 3x - 2$$

$$x^2 - 3x + 2 = 0$$

$$(x-2)(x-1)$$

$$2 \quad 1$$

root

$$n=1 \quad T(1) = 1$$

$$1 = 3T(1/2) + C_2$$

$$n=2 \quad T(2) = C_1 \cdot 4 + C_2$$

$$T(n) = C_1 x_1^n + C_2 x_2^n$$

$$T(n) = C_1 (2)^n + C_2 (1)^n$$

$$T(n) = C_1 (2)^n + C_2$$

$$T(n) = \frac{1}{2} 2^n = 2^{n-1}$$

0(n)

$$17 = 2C_1 + C_2$$

$$2 = 4C_1 + C_2$$

$$7 = 2C_1$$

$$\frac{7}{2} = C_1$$

$$C_2 = 0$$

b)  $T(n) = T(n/2) + 7$

$$T(n) = T(n/2) + 7$$

$$T(n) = T(n/4) + 7$$

$$T(n) = T(n/8) + 7$$

$$T(2) = T(1) + 7$$

$$T(2) = T(1) + 7 = 7 + 7 = 14$$

$$T(4) = T(2) + 7 = 14 + 7 = 21$$

$$T(8) = T(4) + 7 = 21 + 7 = 28$$

$$T(16) = T(8) + 7 = 28 + 7 = 35$$

$$T(2^k) = k + 7$$

$$2^k = n$$

$$\log_2 n = k$$

$$T(n) = \log_2 n + 7$$

$$O(\log_2 n)$$

$$c) T(n) = 4T(n-1) - 4T(n-2) + 3n$$

$$T(n) - 4T(n-1) + 4T(n-2) = 0$$

$$r^2 - 4r + 4 = 0$$

$$(r-2)^2 = 0 \quad r = 2$$

$$T(n) = (6n + \frac{1}{2})2^n$$

$$T(n) = (6n + \frac{1}{2})2^n + 2n + 1$$

$$T(n) = 6n + 1$$

$$d) T(n) = 4T(n/4) + n^2$$

$$a=4, \quad b=2$$

$$T(n) = n^2 \quad n^2 = \Theta(n^{\log_4 4})^2$$

$$a=4, \quad b=2$$

$$c = \log_4 2 = \frac{1}{2} \Rightarrow \Theta(n^{\log_4 2} \log n) = \Theta(n^{\frac{1}{2}} \log n)$$

Master Theorem

$$\Theta(n^{\log_4 4}) \quad T(n) = \Theta(n^{\log_4 4 - 1})$$

$$\Theta(n^{\log_4 4} \log n) \quad T(n) = \Theta(n^{\log_4 4})$$

$$\Theta(T(n)) \quad T(n) = \Omega(n^{\log_4 4 + \epsilon})$$

$$e) T(n) = 2T(\frac{n}{2}) + T(n)$$

$$T(n) = 2T(\frac{n}{2}) + O(n) \quad a=2, \quad b=2$$

$$T(n) = O(n) \quad c=1$$

$$n^{\log_2 2}$$

$$f(n) = O(n) \quad n$$

2. case for Master theorem

$$\Theta(n^{\log_2 2} \log n) = \Theta(n \log n)$$

$$7) T(n) = T(n/2) + T(n/4) + n$$

$$r^2 - r - 1 = 0$$

$$x = \frac{-1 \pm \sqrt{1+4}}{2}$$

$$r = \frac{1 \pm \sqrt{5}}{2} = \frac{1 \pm \sqrt{5}}{2}$$

$$T(n) = \left(\frac{1+\sqrt{5}}{2}\right)^n + \left(\frac{1-\sqrt{5}}{2}\right)^n + \ln n + c$$

$$9) T(n) = T(n/2) + n$$

$$= [T(n/4) + n/2] + n$$

$$= T(n/4) + n/2 + n$$

$$= T(n/8) + n/4 + n/2 + n$$

$$= T(n/2^k) + n/2^{k-1} + n/2^{k-2} + \dots + n/2$$

$$= T(n/2^k) + n(1/2 + 1/4 + \dots + 1/2^{k-1})$$

$$= T(n/2^k) + n(2 - (1/2)^{k-1})$$

$$= T(n/2^k) + n(2 - 1/2^{k-1})$$

$$= T(n/2^k) + n(2 - 1/2^{k-1})$$

$$= T(n/2^k) + 2n - \frac{n}{2^{k-1}} = 2n - \frac{n}{2^{k-1}}$$

$$T(n) = T(n/2) + n \Rightarrow T(n) = 2n - \frac{n}{2^{k-1}}$$

$$h) T(n) = 2T(n/2) + n$$

$$T(2^k) = 2T(2^{k/2}) + 2^k \quad \text{Master's recursion relation}$$

$$S(k) = 2S(k/2) + 2^k$$

$$T(2^k) = 2S(2S(k/4) + 2^k) + 2^k$$

$$T(2^k) = (S(2S(k/4)) + 2^k) + 2^k$$

$$T(2^k) = (S(2S(k/8) + 2^k) + 2^k) + 2^k$$

$$T(2^k) = (S(2S(k/16) + 2^k) + 2^k) + 2^k$$

$$T(2^k) = (S(2S(k/32) + 2^k) + 2^k) + 2^k$$

$$T(2^k) = (S(2S(k/64) + 2^k) + 2^k) + 2^k$$

$$T(2^k) = (S(2S(k/128) + 2^k) + 2^k) + 2^k$$

$$h) S(\frac{n}{2}) = 2^{\log_2(\frac{n}{2})} S(1) + \frac{n}{2} - 1 = \frac{n}{2} - 1$$

$$S(n) = T(2^{\frac{n}{2}}) \quad \frac{n}{2} = \log_2 n \quad \text{Transform}$$

$$T(n) = 2 \log_2(n) - 1$$

$$T(n) = 2 \log_2 n - 1 = \Theta(\log_2 n)$$

0) is-balanced(node)  
 if node is null:  
 return true, 0

left-balanced, left-height = is-balanced(node.left)  $T(2/n)$

right-balanced, right-height = is-balanced(node.right)  $T(2/n)$

balanced = abs(left-height - right-height)  $\leq 1$   $O(1)$

height = max(left-height, right-height) + 1  $O(1)$

return balanced and left-balanced and right-balanced, height

$T(n) = 2T(n/2) + O(1)$  we use master theorem

$O(1) \leq n^{\log_2 2} \Rightarrow$  we use case 1. of master theorem

$O(n^{\log_2 2}) = O(n)$

5) height\_of\_tree (node)  
if node is null:  
return 0

left\_height = height\_of\_tree (node.left)  $T(n/2)$

right\_height = height\_of\_tree (node.right)  $T(n/2)$

height = max (left\_height, right\_height) + 1  $O(1)$

return height

$T(n) = 2T(n/2) + O(1)$  we use master theorem

$$T(n) \leq O(2^{log_2 n}) \Rightarrow O(n^{log_2 2}) \Rightarrow O(n^{1+0}) = O(n)$$

3.2)  $T(n) = 5T(n/2) + O(n^3)$  Algorithm Recursion Relation ex 6/5

we use master theorem

$$T(n) = O(n^{log_2 5})$$

$$n^3 > O(n^{log_2 5}) \quad \text{we select 3 case so} \\ O(T(n)) = O(n^3)$$

$$4) T(n) = 2T(n-2) + O(n) \quad T(0) = 1 \quad n=2k$$

$$T(2k) = 2T(2k-2) + O(2k)$$

$$T(2k-2) = 2T(2k-4) + O(2k-2)$$

$$T(2k-4) = 2T(2k-6) + O(2k-4)$$

$$T(2k-6) = 2T(2k-8) + O(2k-6)$$

$$T(0) = 1$$

$$T(2k) = 2(2T(2k-2) + O(2k)) + O(2k)$$

$$= 2^2 T(2k-2) + 3O(2k)$$

$$T(2k) = 2^3 T(2k-4) + 2^2 O(2k-2) + 3O(2k)$$

$$T(2k) = 2^4 T(2k-4) + 2^3 O(2k-4) + 2^2 O(2k-2) + 3O(2k)$$

$$T(2k) = 2^k T(0) + 2^{k-1} O(2) + 2^{k-2} O(4) + \dots + 2^2 O(2k-2) + 3O(2k)$$

$$\begin{aligned}
&= 2^k T(0) + 2^{k-1} T(1) + 2^{k-2} T(2) + \dots + 2^1 T(n-2) + 2^0 T(n-1) \\
&= 2^k + \sum_{i=1}^k 2^{k-i} T(i) \\
&= 2^k + 0(1) + (2^k - 2^1) T(1) + 0(2) \\
&= 2^k + 0(1) - 0(2) + 3 \cdot 0(2) \\
&= 2^k (1 + 0(2)) + 2 \cdot 0(2) \\
&= 0(2^k) + 0(2) \\
&= 0(2^k) \Rightarrow T(n) = O(2^k) \Rightarrow O(2^{\frac{n}{2}}) =
\end{aligned}$$

$n = 2k$

$$c) T(n) = 3T\left(\frac{n}{2}\right) + O(n^4)$$

we can use master theorem.

$$T(n) = O(n^4) \quad \text{work} < 3.$$

$$T(n) > O(\log n) \Rightarrow O(T(n)) = O(n^2)$$

Summary

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

$$b(n) = O(2^{\frac{n}{2}})$$

$$c(n) = O(n^2)$$

I select c(n) algorithm because more efficient than other algorithm.

4) There are several algorithms for the maximum cardinality matching problem. Some of these algorithms have worst-case, best-case and average-case time complexity.

These algorithms is below

**Simulation:** Attempts to find a matching using augmenting paths. It has a worst-case time complexity of  $O(V^2 E)$ , where  $V$  is the number of nodes and  $E$  is the number of edges. This algorithm can be inefficient, especially for large graphs.

**Edmond-Karp Algorithm:** (BFS-finds augmenting path). This algorithm finds augmenting paths using Breadth First Search (BFS). It has a worst-case time complexity of  $O(V^2 E)$ . In the best case the complexity can be better. For example, when there are few nodes or few augmenting paths in the graph.

**Hopcroft-Karp Algorithm:** This algorithm has a better time complexity of  $O(E \sqrt{V})$ . This algorithm is often more effective for larger graphs.

**Dinic Algorithm:** This algorithm can have a worst-case time complexity of  $O(V^2 E)$  but it often performs better in real applications. In the best case it can have a complexity of  $O(E \sqrt{V} \log V)$ .

\* I select to Hopcroft-Karp algorithm, because better than other. Pseudo code is below

```
Function HopcroftKarp(Graph G)
    For each vertex u in G.LeftNodes do
        Pair[u] = Null
    MaxMatching = 0
    while (DFS()) do
        For each vertex u in G.LeftNodes do
            if (Pair[u] == Null and DFS(u)) then
                MaxMatching = MaxMatching + 1
    return MaxMatching
```

$O(E \sqrt{V})$



5) Foo(n)

if  $n \leq 7$

return 7

else

for  $i$  in range(n) :  
    print("0")

return Foo(n/2) + Foo(n/2)

$T(n/2) + T(n/2)$

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

we use master theorem

$$f(n) = n$$

$$n^{\log_2 2}$$

$$0 < 2$$

$$0 < 2$$

$$n \in \Theta(f(n))$$

case 2

$$\Theta(n^{\log_2 2} \cdot \log n) = \Theta(n \cdot \log n)$$