

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

$$\alpha^2 = 3\alpha - 2$$

$$x(n) = c_1 \cdot 2^n + c_2 \cdot 1^n$$

$$\alpha^2 - 3\alpha + 2 = 0 \quad (\alpha - 2)(\alpha - 1)$$

$$\begin{array}{l} n=1 \quad c_1 \cdot 2 + c_2 = 1 \\ n=2 \quad c_1 \cdot 4 + c_2 = 2 \end{array} \quad \left. \begin{array}{l} c_1 = 1/2 \\ c_2 = 0 \end{array} \right\}$$

$$x(n) = 2^{n-1} \in \underline{\underline{O(2^n)}}$$

b) $T(n) = T(n/2) + 1$, $T(1) = 1$, $T(2) = 2$

$$n=2 \rightarrow T(2) = T(1) + 1$$

$$n=4 \rightarrow T(4) = T(2) + 1$$

$$n=8 \rightarrow T(8) = T(4) + 1$$

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

$$2^x = n$$

$$x = \log_2 n \text{ times}$$

$$\rightarrow T(n) = T(1) + \log_2 n$$

$$T(n) = 1 + \log_2 n \in O(\log n)$$

c) $T(n) = 4T(n-1) - 4T(n-2) + 3n$
 $= 4(4T(n-2) - 4T(n-3) + 3(n-1)) - 4T(n-2) + 3n$
 $= 4^2 T(n-2) - 4^2 T(n-3) + 15n - 12$
 $= 4^3 T(n-3) - 4^3 T(n-4) + 15 \cdot 4(n-2) - 12$
 $= 4^3 T(n-3) - 4^3 T(n-4) + 60n - 108$

$$T(n) = 4^k \cdot T(n-k) - 4^k \cdot T(n-k-1) + 3k \cdot 4^{k-1} - 12(1 + \dots + k-1)$$

$$n = 4^k \rightarrow k = \log_4 n$$

$$T(n) = 4^{\log_4 n} \cdot T(1) - 4^{\log_4 n} T(0) + 3 \log_4 n \cdot 4^{\log_4 n - 1} - 12(1 + \dots + \log_4 n - 1)$$

$$T(n) = n \cdot T(1) - n \cdot T(0) + 3n \cdot \log_4 n - 12 \in O(n \log n)$$

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

$$a=4, b=2, d=2 \rightarrow \text{Master theorem}$$

$$a=b^d \Rightarrow x(n) = O(n^2 \cdot \log n)$$

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

$$a=2, b=2, f(n) = O(n) \rightarrow \text{Master theorem}$$

$$T(n) = O(n^{\log_2 2}) = O(n^1) = O(n)$$

$$\begin{aligned}
 f) \quad T(n) &= T(n/2) + T(n/4) + n \\
 &= (T(n/4) + T(n/8) + n/2) + (T(n/8) + T(n/16) + n/4) + n \\
 &= T(n/4) + 2 \cdot T(n/8) + T(n/16) + n/2 + n/4 + n \\
 &\vdots \\
 &= T(n/2^k) + k \left(\frac{n}{2^k} + \frac{n}{2^{k-1}} + \dots + 1 \right)
 \end{aligned}$$

$$\frac{n}{2^k} = 1 \rightarrow k = \log_2 n$$

$$\begin{aligned}
 T(n) &= T(1) + \log_2 n \left(\frac{n}{2^{\log_2 n}} + \dots + 1 \right) \\
 &\Rightarrow T(1) + \log_2 n \cdot O(\log n) \Rightarrow O(\log^2 n)
 \end{aligned}$$

$$g) \quad T(n) = T(n/2) + n, \quad T(1) = 1, \quad T(2) = 3$$

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

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

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

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

$$T(n) = T(1) + (2^1 + 2^2 + \dots + n) \quad 2^k = n$$

$$= T(1) + (2^1 + 2^2 + \dots + 2^k)$$

$$= T(1) + \sum_{i=1}^k 2^i \Rightarrow T(1) + 2^{k+1} - 1$$

$$= 1 + \frac{2^k}{2} \cdot 2 - 1$$

$$T(n) = 2n \rightarrow \in O(n)$$

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

$$2n = \frac{2n}{2} + n \quad \checkmark$$

$$h) \quad T(n) = 2 T(\sqrt{n}) + 1, \quad T(1) = 1, \quad T(4) = 3$$

$n = 2^m \rightarrow$ perfect squares.

$$T(2^m) = 2 \cdot T(2^{m/2}) + 1$$

$$S(m) = T(2^m) \rightarrow \text{Let's assume}$$

$$S(m) = 2 S\left(\frac{m}{2}\right) + 1 \rightarrow \text{Master theorem}$$

$$a=2, b=2, f(n)=1$$

$$T(2^m) = O((\log 2^m)^1) = O(m)$$

\Downarrow

$$T(n) = O(\log 2) \rightarrow \in O(\log n)$$

2)

a) is-balanced: Recurrence method. We divide into 2 parts to tree in every node. So:

$$T(n) = T(n/2) + O(1) \rightarrow \in O(\log n)$$

b) height-of-tree: This method also recurrence. Algorithm is very similar to a.

$$T(n) = T(n/2) + O(1) \rightarrow \in O(\log n)$$

3)

a) $T(n) = 5 T(n/2) + O(n^3)$

\rightarrow master theorem. $a=5, b=2 \Rightarrow O(n^{\log_2 5})$

b) $T(n) = 2 T(n-2) + O(n)$

\rightarrow master theorem, $a=2, k=1 \Rightarrow O(n^3)$

c) $T(n) = 3 T(n/2) + O(n^2)$

\rightarrow master theorem, $a=3, k=2 \Rightarrow O(n^{\log_2 3})$

\Rightarrow C has the min complexity, But B is more stable. n's power is a number. So B needs to be preferred choice.

4) Question's says definition of Hopcroft-Karp algorithm. It's a polynomial-time algorithm for finding a maximum cardinality matching in bipartite graphs.

algorithm:

1. Initialize maximal matching M as empty.

2. While there exist an Augmenting Path P

- Remove matching edges of P from M and add not-matching edges of P to M.
- Increase size of M by 1 as P starts and ends with a free vertex

3. Return M.

\rightarrow n : # of vertices } in bipartite graph
 m : # of edges }

\rightarrow in worst case: It need to check every node in graph. It become $O(n \cdot m)$.
 If we say $m=n$, $\in O(n^2)$

\rightarrow in best case, matching pairs is very large. BFS explore small area so $\Omega(n+m)$
 If we say $m=n$, $\in \Omega(2n) \rightarrow \in \Omega(n)$.

\rightarrow in average case, It depends on finding augmented path times. As we visit, fewer nodes will remain unvisited.



\rightarrow # of augmented path so $O(n \cdot m) \rightarrow$ if $n=m$, $\in O(n^2)$

(3)

5) } foo(n):
 if $n \leq 1$:
 return 1
 else:
 for i in range(n):
 print("a");
 return foo(n/2) + foo(n/2);

→ parameter is n → # of print "a" in every calling.
 → 2 times call function itself with half of n.

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

in master theorem

$$a=2, b=2, d=1 \Rightarrow O(n^d \cdot \log n)$$

$$\Rightarrow O(n \cdot \log n)$$

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

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

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

$$\Rightarrow \underbrace{2^x}_{n} \cdot T(1) + \underbrace{\left[n + 2 \cdot \frac{n}{2} + 4 \cdot \frac{n}{4} \dots \right]}_{\log_2^n \text{ step}}$$

$$\Rightarrow n \cdot \underbrace{T(1)}_1 + \log_2^n$$

$$T(n) \Rightarrow n + \log_2^n$$

We have \log_2^n step for calling this function, for every tour, running for loop n times. So we write "a" $n \cdot \log_2^n$ times.