

The Master Method:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n), \quad a \geq 1 \text{ and } b > 1$$

$$1) c < \log_b a \Rightarrow T(n) = \Theta(n^{\log_b a})$$

$$2) c = \log_b a \Rightarrow T(n) = \Theta(n^c \log n)$$

$$3) c > \log_b a \Rightarrow T(n) = \Theta(f(n))$$

If $f(n) = \Theta(n^c)$

Q-1) a) $T(n) = \underbrace{27}_a T\left(\underbrace{n/3}_b\right) + n \underbrace{(2)^c}_c$

$\Rightarrow \log_3 27 = \log_3 3^3 = 3$ and $3 > c=2$ so case 1 is valid for the situation. $T(n) = \Theta(n^3)$

b) $T(n) = 9T(n/4) + n$

$\log_4 9 \approx 1.58$ and $c=1$, so; $1.58 > 1$, case 1

$$T(n) = \Theta(n^{1.58})$$

c) $T(n) = 2T(n/4) + \sqrt{n}$

$\log_4 2 \approx 0.5$, $c = \frac{1}{2}$

so; $0.5 = 0.5$, case 2

$$\Rightarrow T(n) = \Theta(n^{0.5} \log n)$$

d) $T(n) = 2T(n/2) + 17$

$\log_2 2 = 1$, $c=0$

so; $1 > 0$, case 1

$$\Rightarrow T(n) = \Theta(n)$$

$$e) T(n) = 2T(\sqrt{n}) + 1$$

Let's assume that $\boxed{n=2^k}$ then the expression transforms to

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

2) let's assume that $T(2^k) = S(k)$ so;

$$\boxed{T(2^k) = S(k) = 2S(\frac{k}{2}) + 1} \rightarrow \text{Use master theorem;}$$

$$\log_2 > 0 \Rightarrow \boxed{\Theta(k) = S(k) = T(2^k) = T(n)}$$

→ We need to convert back to the expression from k to n .

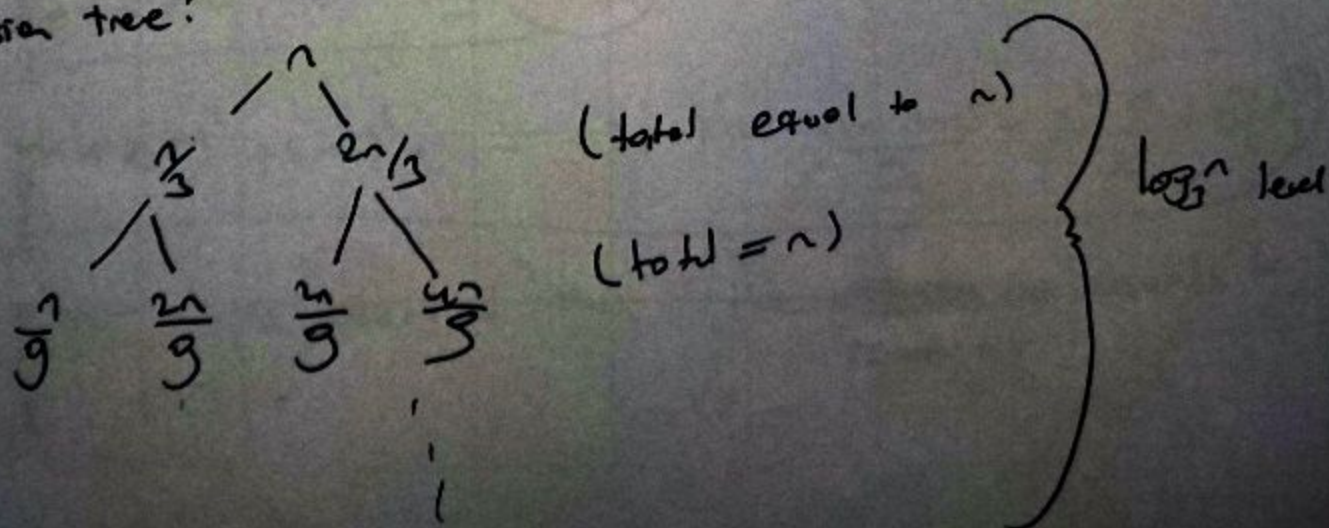
if $n = 2^k$ then $\underline{k = \log n}$ so the $\boxed{\Theta(k) = \Theta(\log n)}$

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

$$\log_{\frac{4}{2}} > 1 \Rightarrow T(n) = \Theta(n^2)$$

g) $T(n) = T(n/3) + T(2n/3) + \Theta(n)$

Recursion tree:



→ left most node represents best case and rightmost represents worst case.

J- cont.) \Rightarrow Total value on each level is n and there are $\log_3 n$ level to reach end of the tree. ~~Thus~~ Thus;

$$T(n) \geq n \cdot \log_3 n = \Omega(n \log n)$$

\Rightarrow there are $\log_{3/2} n$ level on the worst case and n subproblem on the each level at the worst case.

$$T(n) \leq n \cdot \log_{3/2} n = O(n \log n)$$

So; if Ω and O is the same $T(n) = \Theta(n \log n)$

h-

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

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

See the Faulhabers formula. \downarrow This equation is equal to n^{c+1} on the worst case.

for example; let say $c=1$ the equation equals to;

$$n^1 + (n-1)^1 + (n-2)^1 + \dots + 2^1 + 1^1 = \frac{n \cdot (n-1)}{2} = \frac{n^2 - n}{2} = \Theta(n^2)$$

If the c is equal 1 then the ~~recurrence~~ equals to

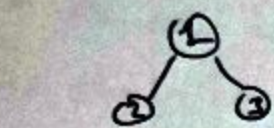
$$\Theta(n^{c+1})$$

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

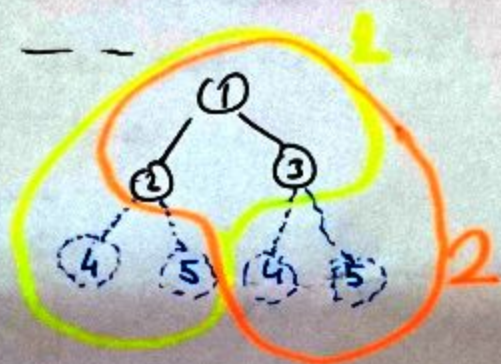
$$T(n) = c^n + c^{n-1} + c^{n-2} + \dots + c^2 + c^1 + c^0 = \frac{c^{n+1} - 1}{c - 1}$$

this relation equals to c^{n+1} on both worst and best case so $\boxed{\Theta(c^{n+1}) = T(n)}$

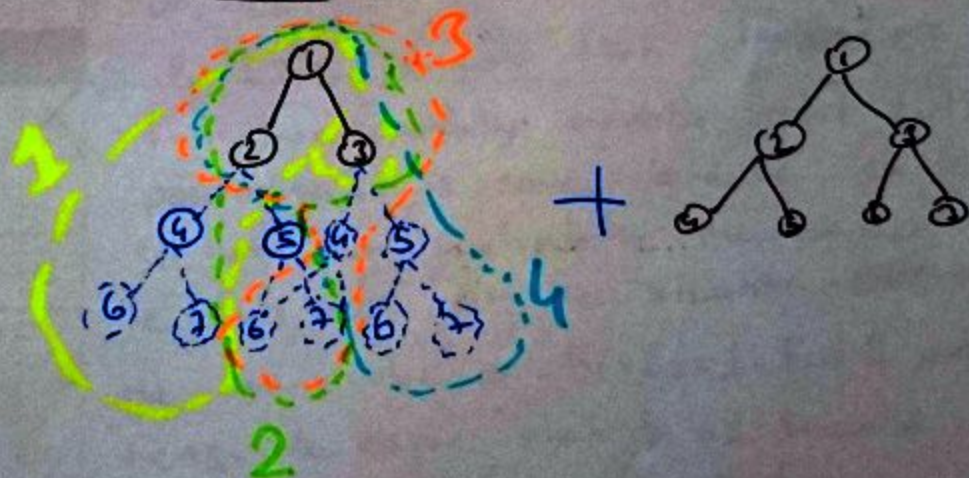
2-Q) $\Rightarrow B_3 = 1$ because we can construct only one full tree with 3 vertex



$B_3 = 2$

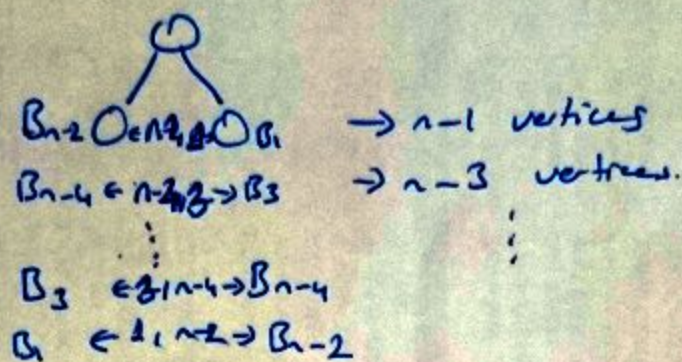


$B_7 = 5$



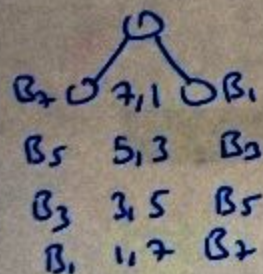
* There couldn't be any full tree with even numbers of vertices because full tree states for "each node must have 0 or 2 children" and that means there must be $k+1$ vertices in the tree where $\boxed{k \div 2 = 0}$. If a number can be divided 2 without any remainder; that means the number is even and any odd number is equal to adding 1 to any even number. So $k+1$ is an odd number that's why we cannot use even numbers to calculate B_n .

2-b) We need to consider combinator logic to calculate number of binary trees with n vertices.
 → let's assume that we have n vertices;



* We can divide the ~~the~~ vertices into two subclass (branch) as shown on the illustration below. So, if we sum all the possibilities then we get the total number of possible subtrees (full binary trees).

→ E.g. let's say $n=8 \rightarrow B_8$;



8 vertices

→ for $[7, 1]$; it means that there are B_7 left full tree possibility and B_1 right full tree possibilities. It is some kind of cartesian product. So we can calculate the number of possible full trees for the situation ~~with~~ by multiplying the two values $B_1 \times B_7 = 1 \times 5 = 5$ possible full trees. We assume that $B_1 = 1$ because 1 vertices with zero children is a full tree.

→ do the process for each situation;

$$B_7 \times B_1 = 5$$

$$B_5 \times B_3 = 2$$

$$B_3 \times B_5 = 2$$

$$B_1 \times B_7 = 5$$

$$= 14$$

2-b-cont.)

→ let's calculate a familiar B_n like B_7 :



$$\begin{aligned} B_5 &\in 5, 1 \rightarrow B_1 \\ B_3 &\in 3, 3 \rightarrow B_3 \\ B_1 &\in 1, 5 \rightarrow B_5 \end{aligned} \Rightarrow \begin{aligned} &= 2 \times 1 = 2 \\ &= 1 \times 1 = 1 \\ &= 2 \times 2 = 2 \\ &= \underline{\underline{5}} \end{aligned}$$

⇒ So we can formulate the relation as:

$$B_n = (B_{n-2} \times B_1) + (B_{n-4} \times B_3) + \dots + (B_1 \times B_{n-2})$$

$$= \sum_{k=1}^{n-2} B_k \times B_{n-k-1}, \text{ for } k, k \text{ is odd.}$$

2-c) Unfortunately this question is wrong.

3-)

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

a = Count of the sub-problems.

$\frac{n}{b}$ = Size of the sub-problems.

a) "dividing into 7 subproblems with one-third of the size" — "when combining them in quadratic time"

$$T(n) = 7T\left(\frac{n}{3}\right) + n^2$$

by master theorem;

$$\log_3 7 \approx 1.77 \text{ and } c=2$$

$$\text{so; } \log_3 7 < 2 \Rightarrow$$

$$T(n) = \Theta(n^2) = O(n^2)$$

$$3-b) T(n) = T(n-1) + n$$

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

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

⋮

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

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

$$T(n) = T(0) + \left(\frac{n \cdot (n+1)}{2}\right) = \frac{n^2 + n}{2} + T(0) \rightarrow T(0) \text{ is a constant.}$$

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

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

Masters theorem's

$$\log_2 4 = 2 = c = 2$$

$$\text{So, } T(n) = \Theta(n^2 \log n) = \text{ ~~} \Theta(n^2) \text{ }~~$$

$$\boxed{= O(n^2 \log n)}$$

\Rightarrow So we need to compare the algorithms

$$O(n^2) = O(n^2) < O(n^2 \log n)$$

(1)

(2) ✓ (3)

$$\lim_{n \rightarrow \infty} \frac{n^2 \log n}{n^2} = \infty$$

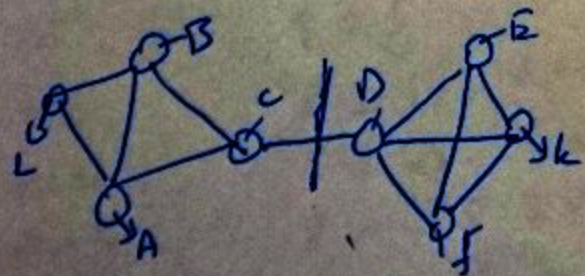
\therefore So we can pick both 1 and 2 algorithms.

Q4) There is no polynomial-time algorithm to solve the min-cut problem exactly. The question asks to give a polynomial-time algorithm, it doesn't ask the optimal solution for the problem because the problem is NP-complete and therefore there is no polynomial-time algorithm for the solution of the problem.

* Any offered algorithm which has polynomial running time will be considered as a correct answer.

An example solution for the problem?

① We can offer a greedy algorithm to solve the problem in polynomial-time. Our greedy condition ~~considers~~ consider the vertex degrees to solve the problem. Our solution ~~considers~~ ~~wants~~ wants to find smallest degree vertex and accept the vertex as set one and $V -$ the vertex is the second set. This solution doesn't guarantee the optimal solution but still it is a reasonable solution for our problem. E.g. the algorithm can not find the right solution on the following example.



* The algorithm finds vertex "L" as the vertex with smallest degree and provides a solution like:

$$S_1 = \{L\}, S_2 = \{A, B, C, D, E, F, K\}$$

→ But the optimal solution divides the graph into two sets by cutting the edge between C and D.

⇒ Optimal solution only cuts 1 edge but our solution cuts 2 edge at least.

3) The code is very similar to an implementation of Catalan numbers but they are not the same. When we consider the relation between n and number of prints, we get some relation like the following:

$n=0$	print count = 1
$n=1$	$u = 1$
$n=2$	$u = 5 =$
$n=3$	$u = 15 = 3 \times P(2)$
$n=4$	$u = 45 = 3 \times P(3) = 3 \times 3 \times P(2)$
$n=5$	$u = 135 = 3 \times P(4) = 3 \times 3 \times P(3) = 3 \times 3 \times 3 \times P(2)$
$n=6$	$u = 405 = 3 \times 3 \times 3 \times 3 \times P(2)$
	\vdots
$n=n$	$u = 3^{n-2} \times P(2)$

↓

$$T(n) = P(2) \times 3^{n-2} = \left(\frac{5}{9}\right) \cdot 3^n$$

① for all $k < \frac{5}{9}$; $k \cdot 3^n < \frac{5}{9} 3^n$

$$\boxed{\text{So } T(n) = \Omega(3^n)}$$

② for all $k \geq \frac{5}{9}$; $k \cdot 3^n > \frac{5}{9} 3^n$

$$\boxed{\text{So } T(n) = O(3^n)}$$

Since $T(n) = \Omega(3^n)$ and $T(n) = O(3^n)$

$$\boxed{T(n) = \Theta(3^n)}$$