



Design and Analysis of Algorithms

Lecture - 7

Success is always inevitable with Hard Work and Perseverance

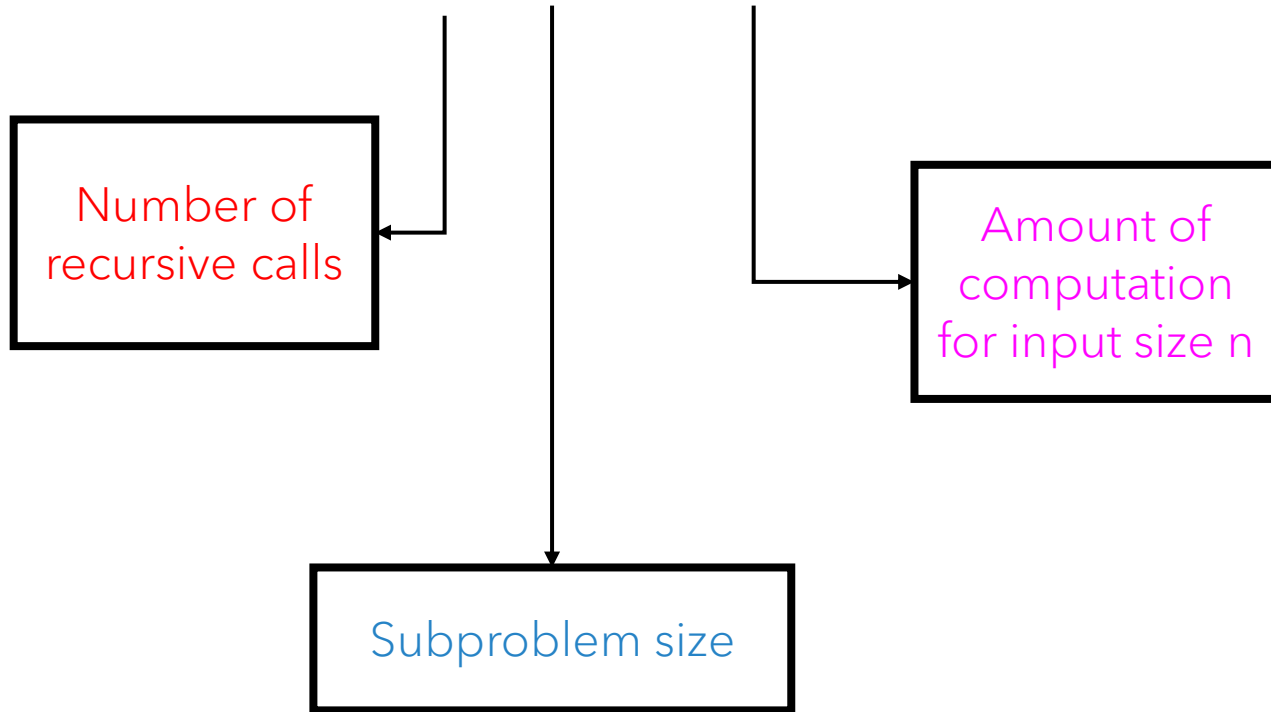
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Learning Objective

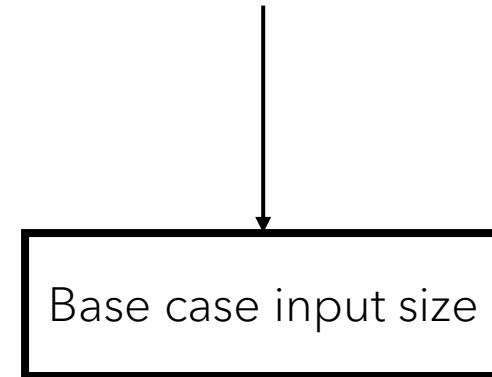
- Discuss the methods of recursion tree and Master theorem
- Understand when Master theorem is applicable

Recursion Tree Method

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



$$T(1) = 1$$



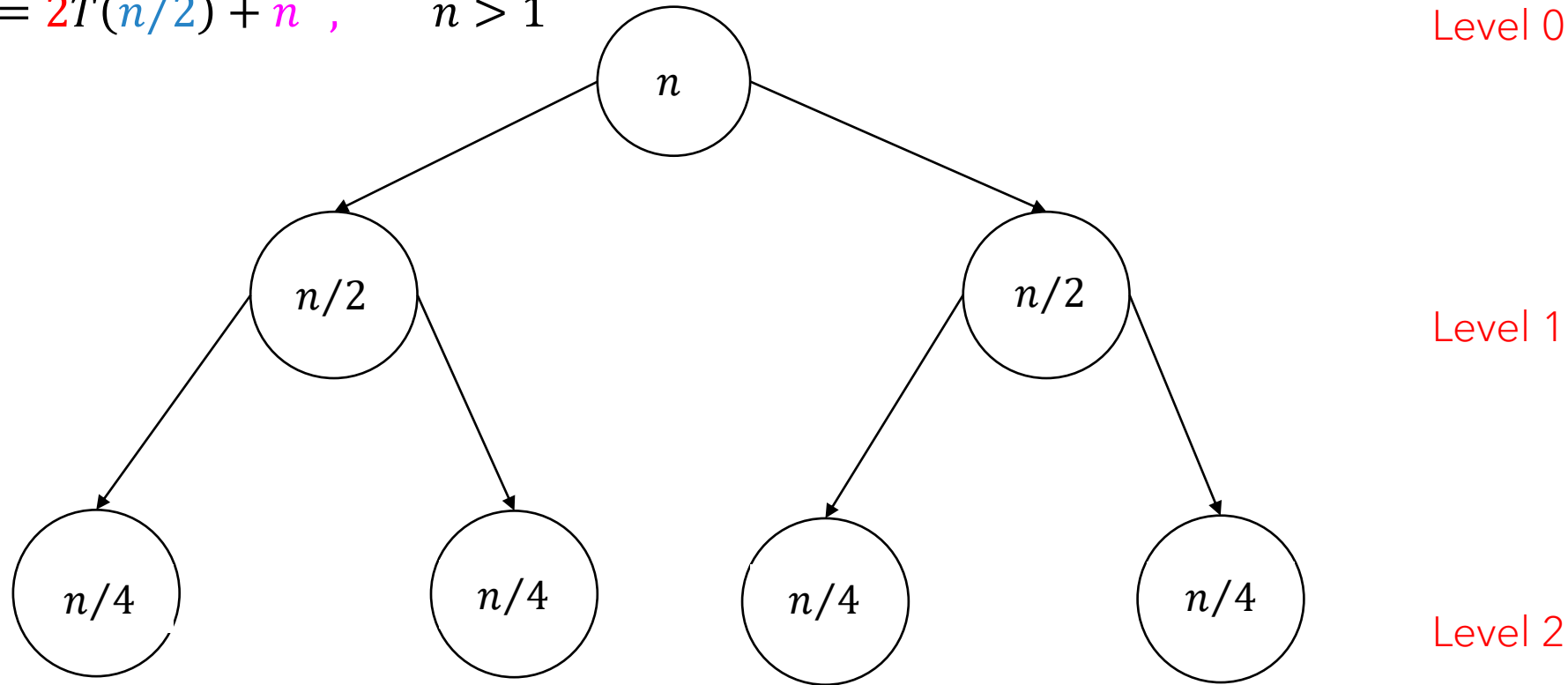
Recursion Tree Method

$$f(n) = n$$

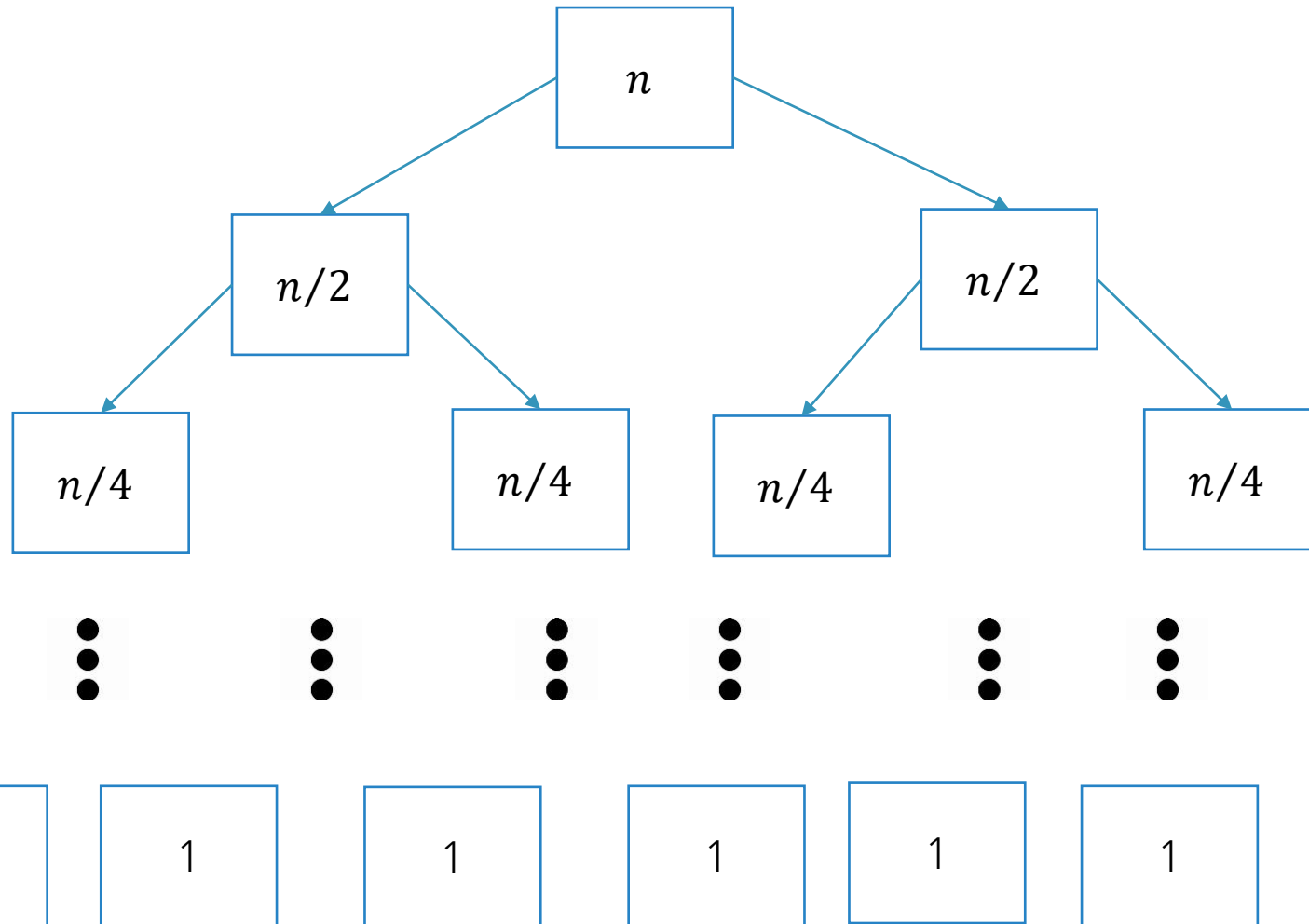
#recursive calls = 2

Size of sub problem = $n/2$

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



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



Problem size	# nodes	Amount of work done
n	1	n
$n/2$	2	$2(n/2) = n$
$n/4$	4	$4(n/4) = n$
1	n	$n * 1 = n$

Cost

$$\sum_{i=a}^b 1 = b - a + 1$$

Cost = Cost of internal nodes + Cost of leaf nodes

$$= [n + n + \dots] + n * 1$$

$$= \sum_{i=0}^{\log n - 1} n + n$$

$$= n \sum_{i=0}^{\log n - 1} 1 + n = n[\log n - 1 + 1] + n = O(n \log n)$$

Pause & Think

If problem of size (n) is reduced by a constant factor (1/3) after every recursive call, at what level does the problem get reduced to size of 1 ?

$$\frac{n}{3^i} = 1$$

$$i = \log_3 n$$

Recursion Tree Method

$$T(n) = 3T(n/4) + n^2, \quad n > 1$$

Number of
recursive calls

Amount of
computation
for input size n

Subproblem size

$$T(1) = 1$$

Base case input size

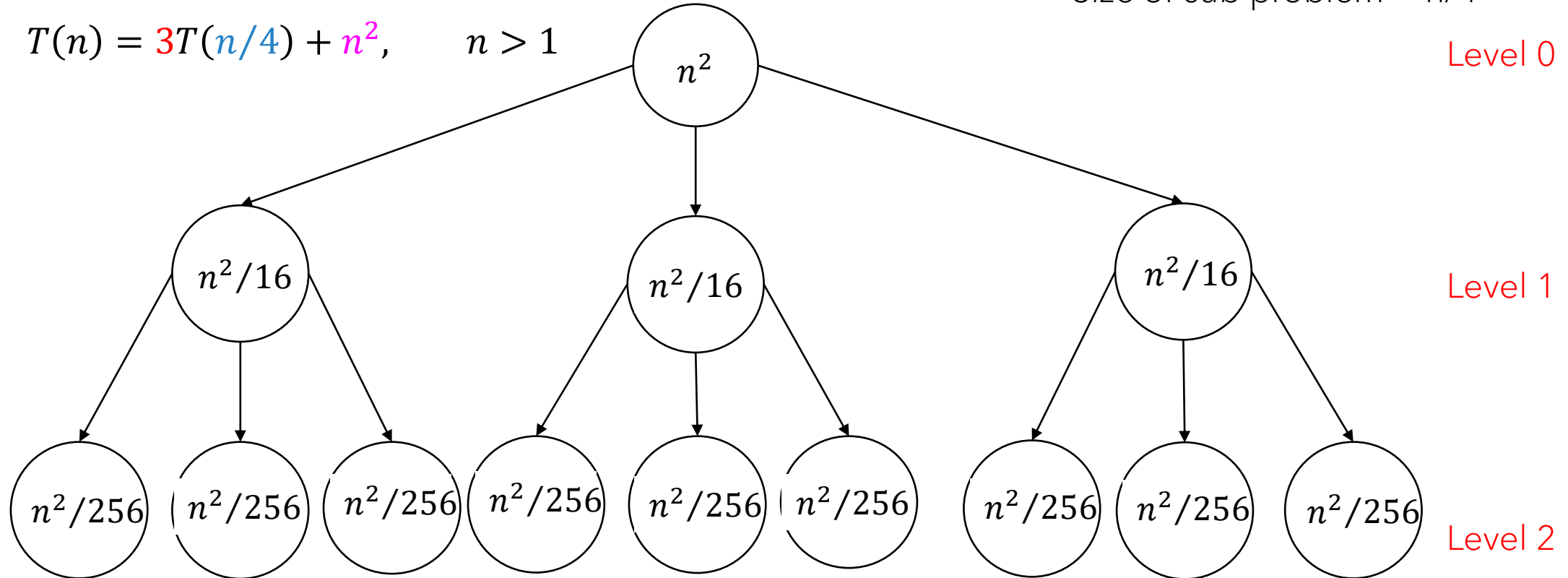
Recursion Tree Method

$$f(n) = n^2$$

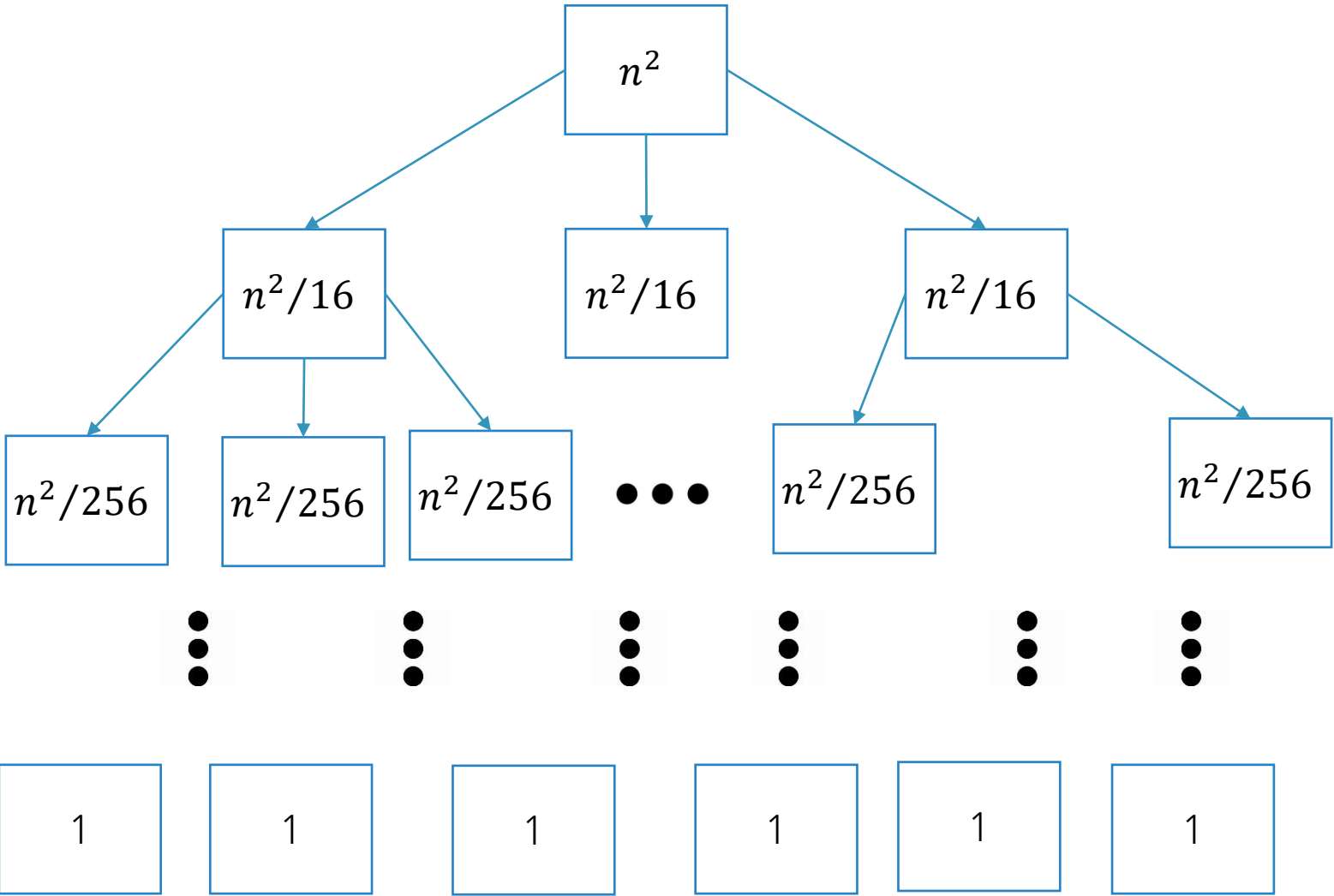
#recursive calls = 3

Size of sub problem = $n/4$

$$T(n) = 3T(n/4) + n^2, \quad n > 1$$



$$T(n) = 3T(n/4) + n^2$$



Problem size	# nodes	Amount of work done
n	1	n^2
$n/4$	3	$3(n^2/16)$
$n/16$	9	$9(n^2/256)$
1	$3^{\log_4 n}$	$n^{\log_4 3}$

Cost

$$\sum_{i=0}^{\infty} a^i = \frac{a}{1-a} \quad a < 1$$

Cost = Cost of internal nodes + Cost of leaf nodes

$$= [n^2 + 3(n^2/16) + 9(n^2/256) \dots] + n^{\log_4 3}$$

$$= \sum_{i=0}^{\log_4 n - 1} (3^i)(n^2/16^i) + n^{\log_4 3}$$

$$< n^2 \sum_{i=0}^{\infty} (3/16)^i + n^{\log_4 3} = (3/13)n^2 + n^{\log_4 3} = O(n^2)$$

Is there a shortcut ?

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



$$O(n \log n)$$

Master theorem

Theorem

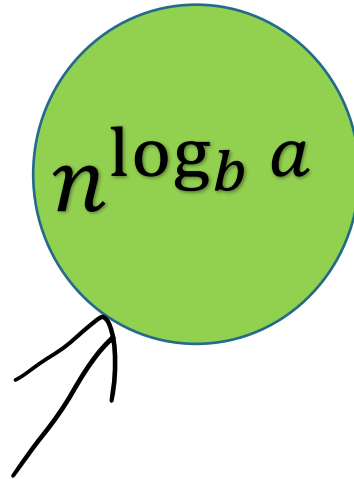
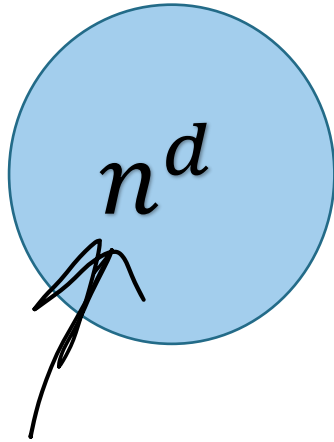
If $T(n) = aT(n/b) + (n^d)$ [for constants $a > 0$, $b > 1$, $d \geq 0$] then

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

$a.f(n/b) \leq c.f(n)$ for some $c \in [0, 1]$

Case

3



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

Master theorem

$$T(n) = 3T(n/4) + n^2, \quad n > 1$$

Number of
recursive calls
(a)

Amount of
computation
for input size n
(n^d)

Subproblem size
(n/b)

$$T(1) = 1$$

Base case input size

a	3
b	4
d	2
n^d	n^2
$n^{\log_b a}$	$n^{\log_4 3} = n^{0.79}$

It belongs to case 3

Regularity Condition

$$a \cdot f(n/b) \leq c \cdot f(n)$$

$$3 \left(\frac{n}{4} \right)^2 \leq c \cdot n^2$$

$$\frac{3}{16} n^2 = c \cdot n^2 \quad C = \frac{3}{16} [0.1875]$$

a	3
b	4
d	2
n^d	n^2
$n^{\log_b a}$	$n^{\log_4 3} = n^{0.79}$

It belongs to case 3 $T(n) = O(n^2)$

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

a	3
b	2
d	1
n^d	n
$n^{\log_b a}$	$n^{\log_2 3} = n^{1.58}$

It belongs to case 1 $T(n) = O(n^{\log_2 3})$

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

a	2
b	2
d	1
n^d	n
$n^{\log_b a}$	$n^{\log_2 2} = n$

It belongs to case 2 $T(n) = O(n \log n)$

When Master theorem is not applicable?

If $f(n)$ and $n^{\log_b a}$ does not differ by polynomial

$$T(n) = 3T(n/4) + n \log n$$

Pause & Think

How will you prove master theorem ?

Using recursion tree

Cost = cost of root node + cost of internal node + cost of leaf node

$$n^d + \sum_{i=0}^{\log n} n^d \left(\frac{a}{b^i}\right)^d + n^{\log_b a}$$

Case 1: $n^{\log_b a} > n^d$ [constant multiple of n^d] ($a > b^d$)

Case 2: $n^d = n^{\log_b a}$ [$n^d * \sum_{i=0}^{\log n} (1)^i$] = $n^d \log n$

Case 3: $n^d > n^{\log_b a}$ [$a.f(n/b) \leq c.f(n)$] ($a < b^d$)

Summary

- For problems that are reduced by constant factor on every recursive call with polynomial computation, learnt a shortcut for assessing time complexity.

Thank You
Happy Learning

Success is always inevitable with Hard Work and Perseverance