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For Binary Search, the time complexity is given by the following recurrence: T(n) = O(1) + T(n/2) => T(n) = O(\log n)

For Merge Sort, the time complexity is given by the following recurrence: T(n) = 2.T(n/2) + O(n) => T(n) = O(n.\log(n))
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Solving Recurrence Relations --

- 1) Substitution method: Guess a solution, and then check whether it is correct.
 - Eg. Let us guess the solution for Binary Search as $T(n) = O(\log n)$, which means that we must have $T(n) <= c.(\log n)$ for large enough n (for all n >= n0).

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T(n) = O(1) + T(n/2)
<= O(1) + O(\log(n/2))
<= c1 + c2.(\log(n/2))
= c1 + c2.(\log(n) - \log(2))
= c1 + c2.\log(n) - c2
= c2.\log(n) - (c2 - c1)
<= c2.\log(n)
= O(\log n)
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- 2) Recurrence Tree method: Figure out the solution by studying the recurrence tree.
- 3) Master's Theorem method:

It is a direct method to get solutions for recurrences of the form $T(n) = a.T(n/b) + O(n^c)$, where a>=1 and b>1. Then, the following three cases are used to obtain the solution directly -

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(i) If c < log\_b(a), then T(n) = \Theta(n^(log\_b(a)))

Eg. - For the recurrence T(n) = 16.T(n/4) + O(n), we have: a=16, b=4, c=1

Therefore, 1 = c < log\_b(a) = log4(16) = 2, and so we have: T(n) = O(n^2)

(ii) If c = log\_b(a), then T(n) = O(n^c.log(n))

Eg. - For binary search recurrence T(n) = 1.T(n/2) + O(1), we have: a=1, b=2, c=0

Therefore, c = log\_b(a) = log2(1) = 0, and so we have: T(n) = O(n^0.log(n))

Eg. - For merge sort recurrence T(n) = 2.T(n/2) + O(n), we have: a=2, b=2, c=1

Therefore, c = log\_b(a) = log2(2) = 1, and so we have: T(n) = O(n^1.log(n))

(iii) If c > log\_b(a), then T(n) = O(n^c)

Eg. - For the recurrence T(n) = 2.T(n/4) + O(n^2), we have: a=1

Therefore, a=1
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T(n) = 2.T(n/2) + O(n)
= 2.(2.T(n/4) + O(n/2)) + O(n)
= 2.(2.(2.T(n/8) + O(n/4)) + O(n/2)) + O(n)
......
= 2^k.T(n/(2^k)) + (O(n) + 2.O(n/2) + 4.O(n/4) + ... + (2^k).O(n/(2^k))
= n.T(1) + (O(n) + O(n) + ..(k times)... + O(n)
= n + k.O(n)
= n + O(n).log(n) = O(n. log(n))
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