

Types of recursions -

① Linear

② Divide & conquer

→ DIVIDE & CONQUER RECURRENCES

dividing in sub-problems

whole problem

form^o, form of program $T(n) = a_1 T(b_1 n + \epsilon_1(n)) + a_2 T(b_2 n + \epsilon_2(n)) + \dots + a_k T(b_k n + \epsilon_k(n)) + g(n)$

conquer

combining

time complexity required to do something with previous tasks/calls. $n \geq n_0$ constant

Relating this form to binary search formula, relation

$$T(n) = T\left(\frac{n}{2}\right) + \frac{c}{\text{constant}}$$

Here, $a_1 = 1, b_1 = \frac{1}{2}, g(n) = c$

$$\epsilon_1(n) = 0$$

→ D&C is a recursive strategy.

→ This technique divided into three parts -

- ① Divide - dividing problems into smaller sub-problems.
- ② Conquer - solve sub-problems by calling recursively until solved.
- ③ Combining - combine sub problems to get final solⁿ of whole problem.

* Sub-problems should be same as main problem, for example if the main problem is of sorting, so the sub-problems should also be sorting only which can be ~~work~~ recursively.

It should not be like you're doing another thing instead of sorting in sub-problems. That shouldn't be count in Divide & conquer technique.

GENERAL METHOD FOR DIVIDE & C.

```
D&C(P){  
  if (small(P)) { // P is small  
    solve(P);  
  }  
  else {  
    divide P into  $P_1, P_2, P_3, \dots, P_k$   
    Apply D&C( $P_1$ ), D&C( $P_2$ ), ....  
    combine(D&C( $P_1$ ), D&C( $P_2$ ), ...)  
  }  
}
```


Recurrence Relation for Dividing $\#1$

- * $T(n) = 2T(n/2) + 1 \rightarrow O(n \log n) \rightarrow O(n)$
- * $T(n) = 4T(n/2) + n \rightarrow O(n^2)$
- * $T(n) = 2T(n/2) + n \rightarrow O(n \log n)$
- * $T(n) = 4T(n/2) + n^2 \rightarrow O(n^2 \log n)$
- * $T(n) = 4T(n/2) + n^2 \log^2 n \rightarrow O(n^2 \log^3 n)$

Example of computing

$$\begin{aligned} T(n) &= 2T(n/2) + 1 \\ a &= 2 \quad f(n) = O(1) \\ b &= 2 \quad = O(n^0 \log n) \\ k &= 0, \quad p = 0 \end{aligned}$$

from this,
 $\log_2 2 = 1, k = 0$

case 1) $O(n^{\log_2 a}) \rightarrow O(n^{\log_2 2}) \rightarrow O(n^1)$

MASTER THEOREM for dividing $\#2$

General $\Rightarrow T(n) = aT(n/b) + f(n)$
 $a \geq 1, f(n) = O(n^k \log^p n)$
 $b > 1$

Case 1 - if $\log_b a > k$ ^{power}
then $O(n^{\log_b a})$ equ

Case 2 - if $\log_b a = k$

if $p > -1 \rightarrow O(n^k \log^{p+1} n)$

if $p = -1 \rightarrow O(n^k \log \log n)$

if $p < -1 \rightarrow O(n^k)$

Case 3 - if $\log_b a < k$,

if $p \geq 0 \rightarrow O(n^k \log^p n)$

if $p < 0 \rightarrow O(n^k)$