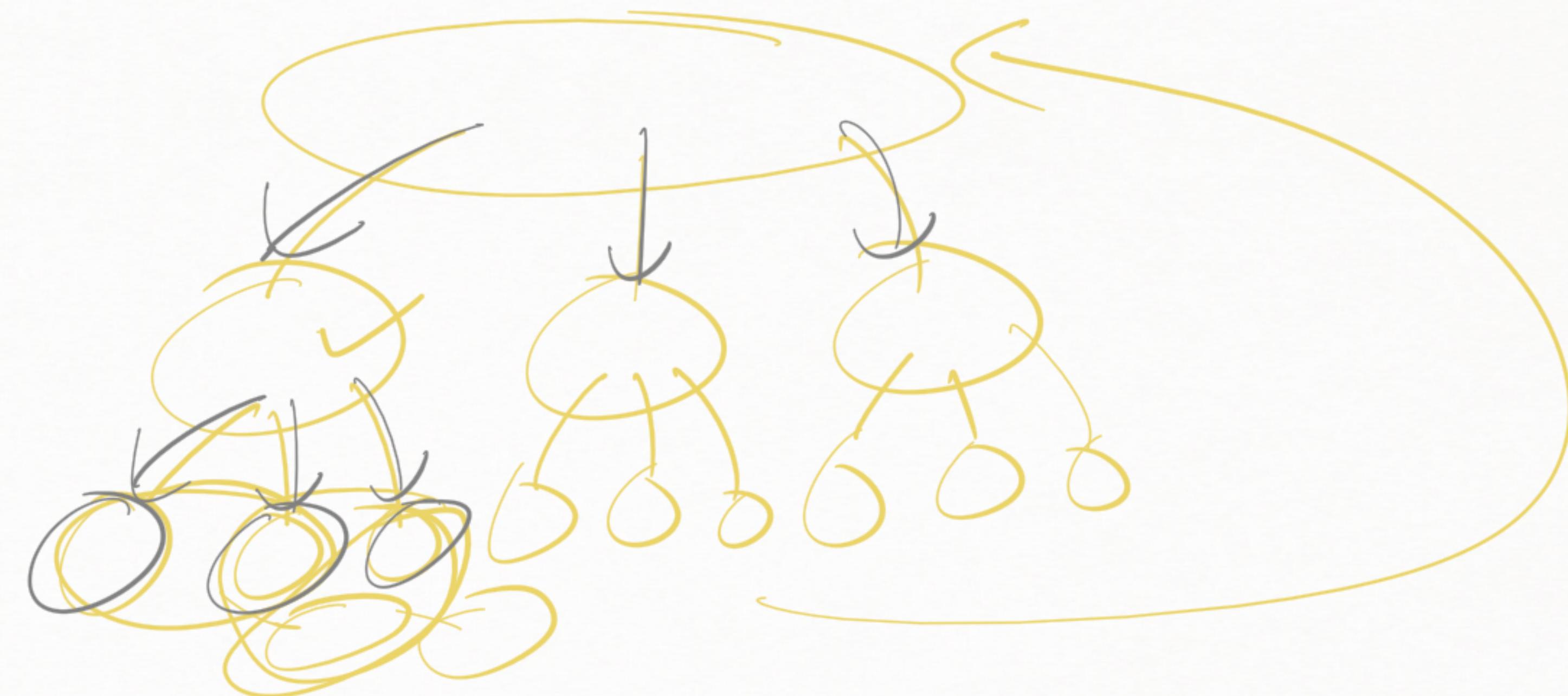


# Greedy  $\rightarrow$  optimisation  $\leq_{\text{B1B}}$   
Laugst No Problem  
295 Greedy  
 $\downarrow$  choice  
952 Next  
Subproblem  
 $\geq$

## # Divide & Conquer

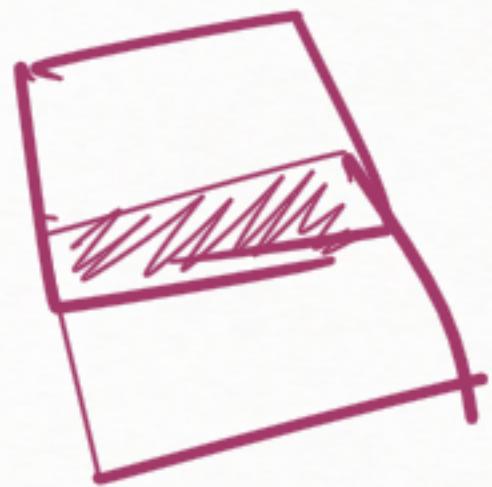
Divide a problem into overlapping subproblems to solve them recursively and combine them in the end.

Bu<sup>fish</sup>  
India  
BIT



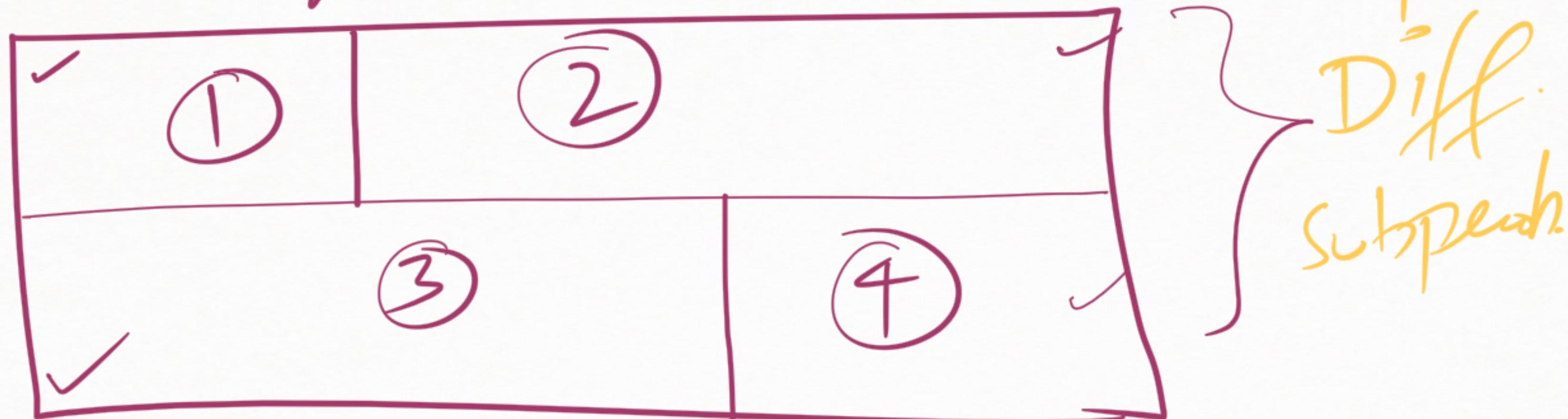
Divide

A problem to be solved

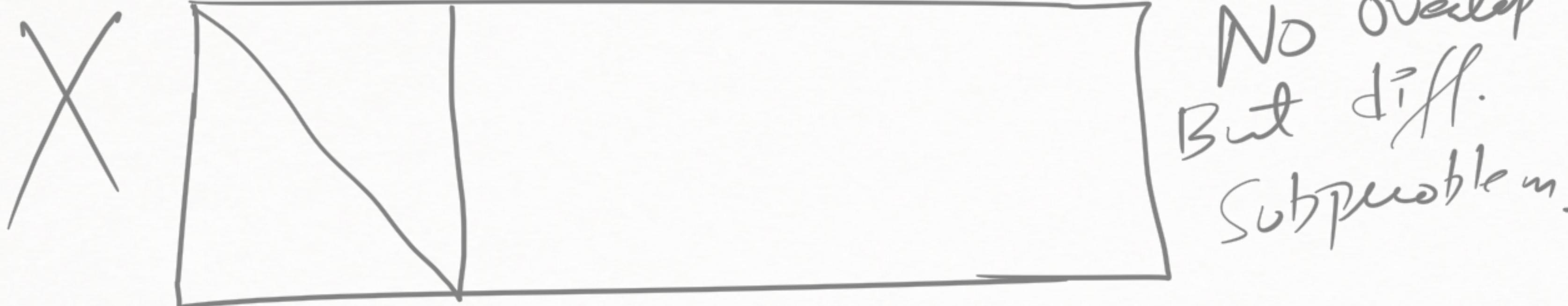
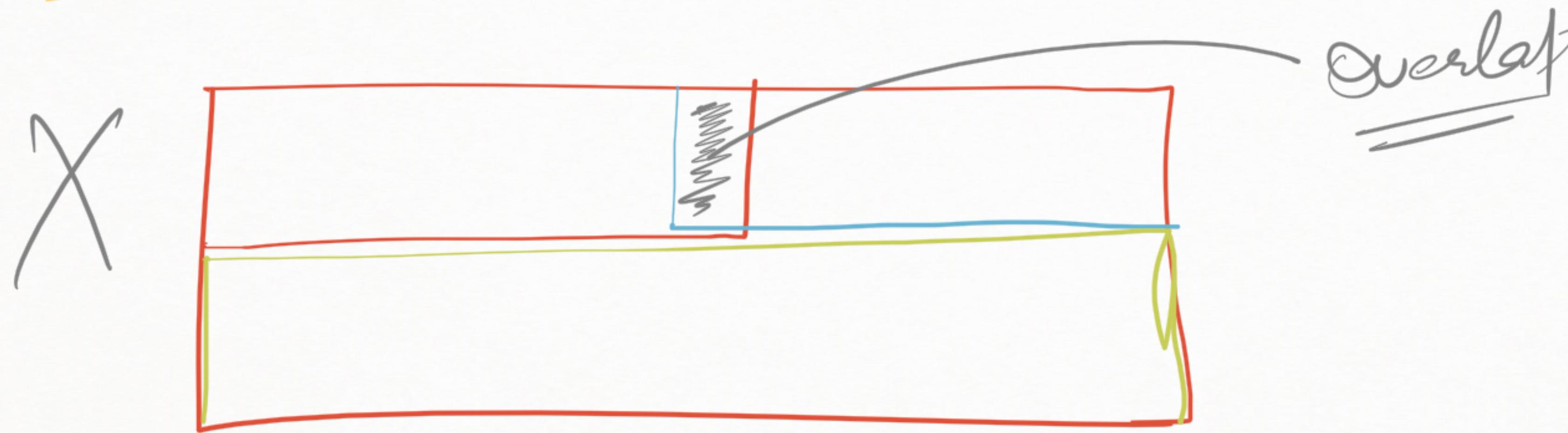


Break Problem into non-overlapping subproblems  
the same type

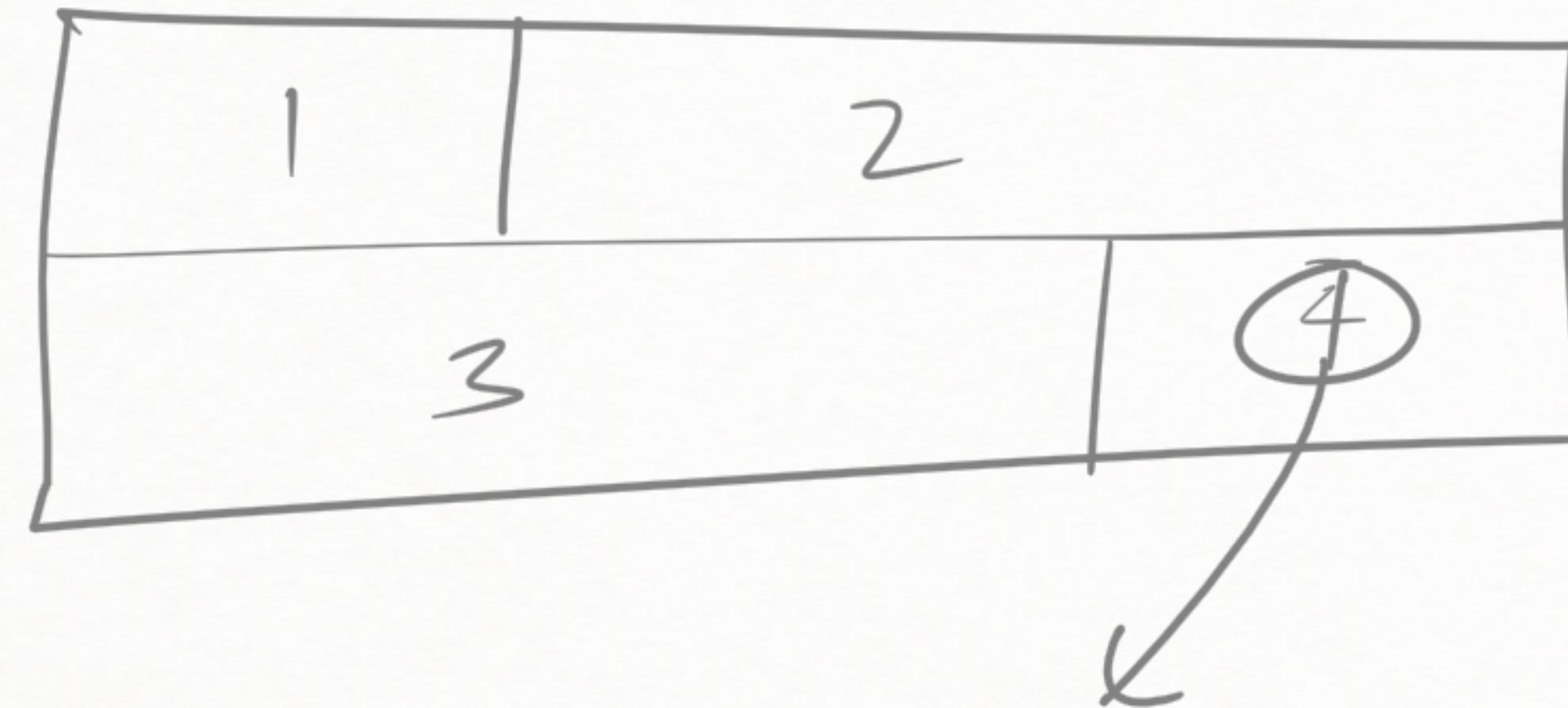
and Divide



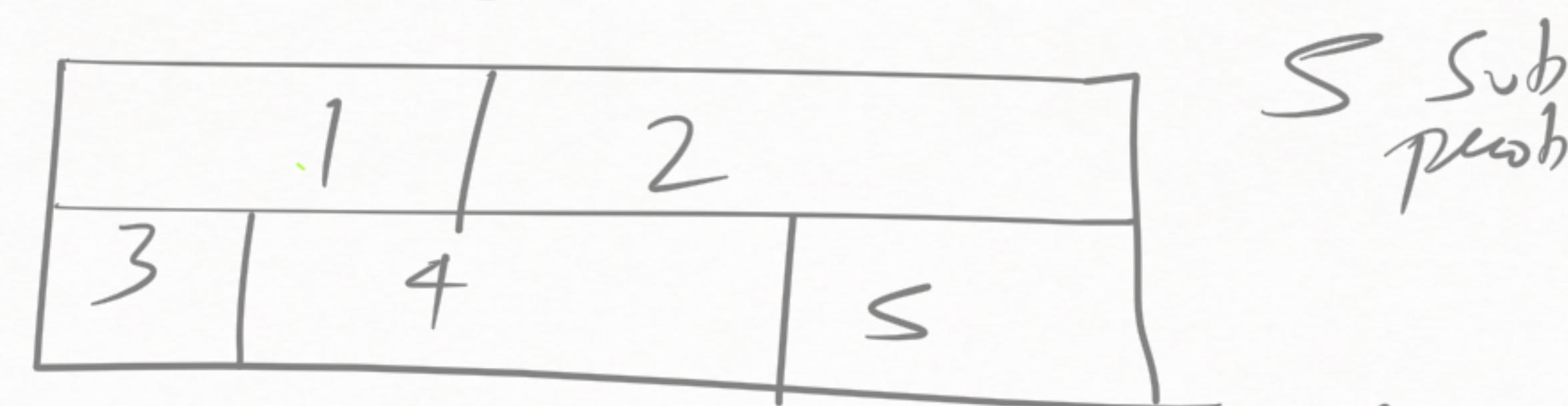
~~Invalid Divide~~



# How each subproblem is solved?



4 subprob.

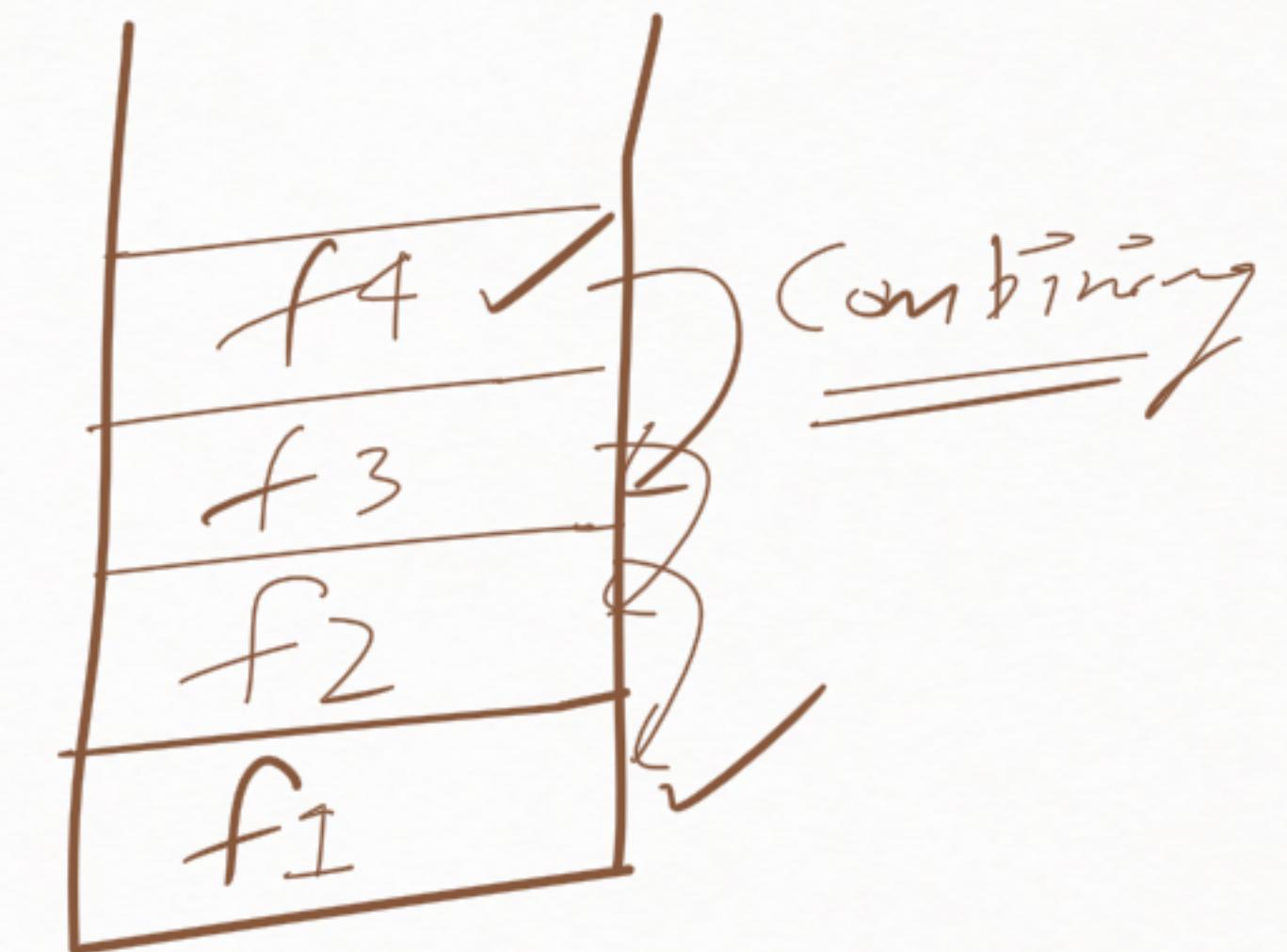
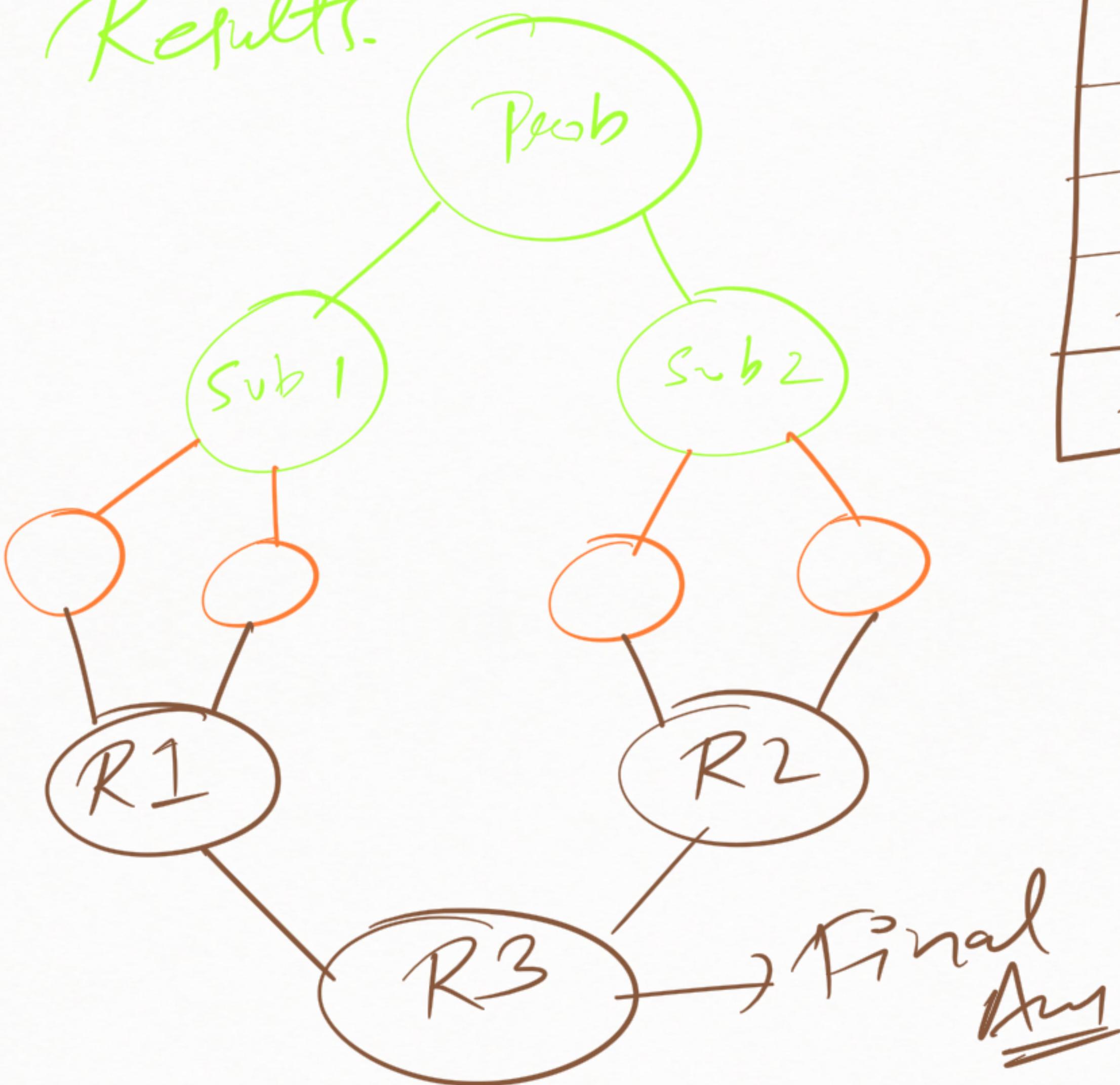


5 subprob

# Do it till Base case is arrived  
Terminating Condition for recursion.

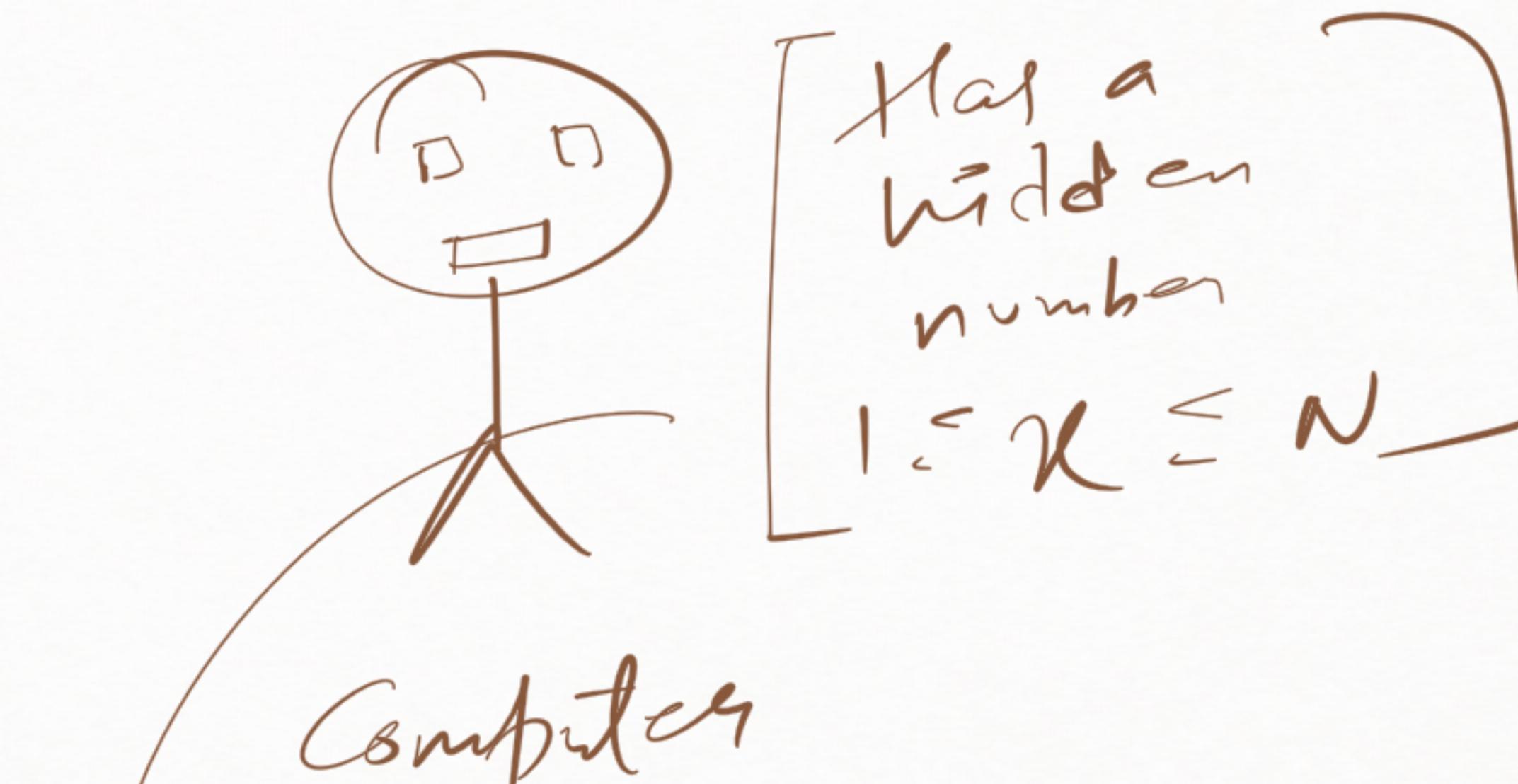
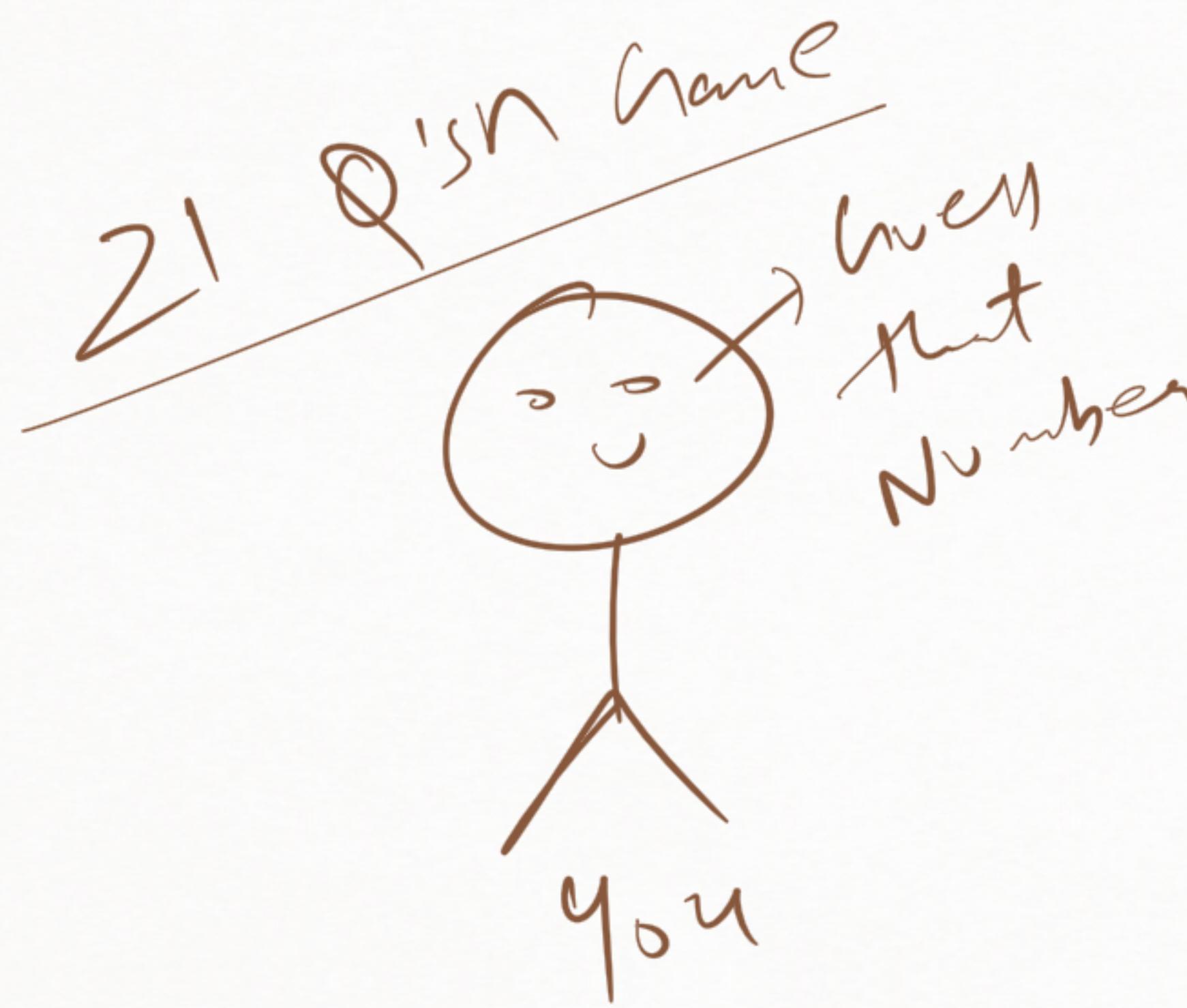
## # Combine Results.

Bare cases  
starting from  
of combine stage.



## # Divide & Conquer Strategy

- ① Break into non-overlapping subproblems of the same type.
- ② Solve subproblems.
- ③ Combine Results.

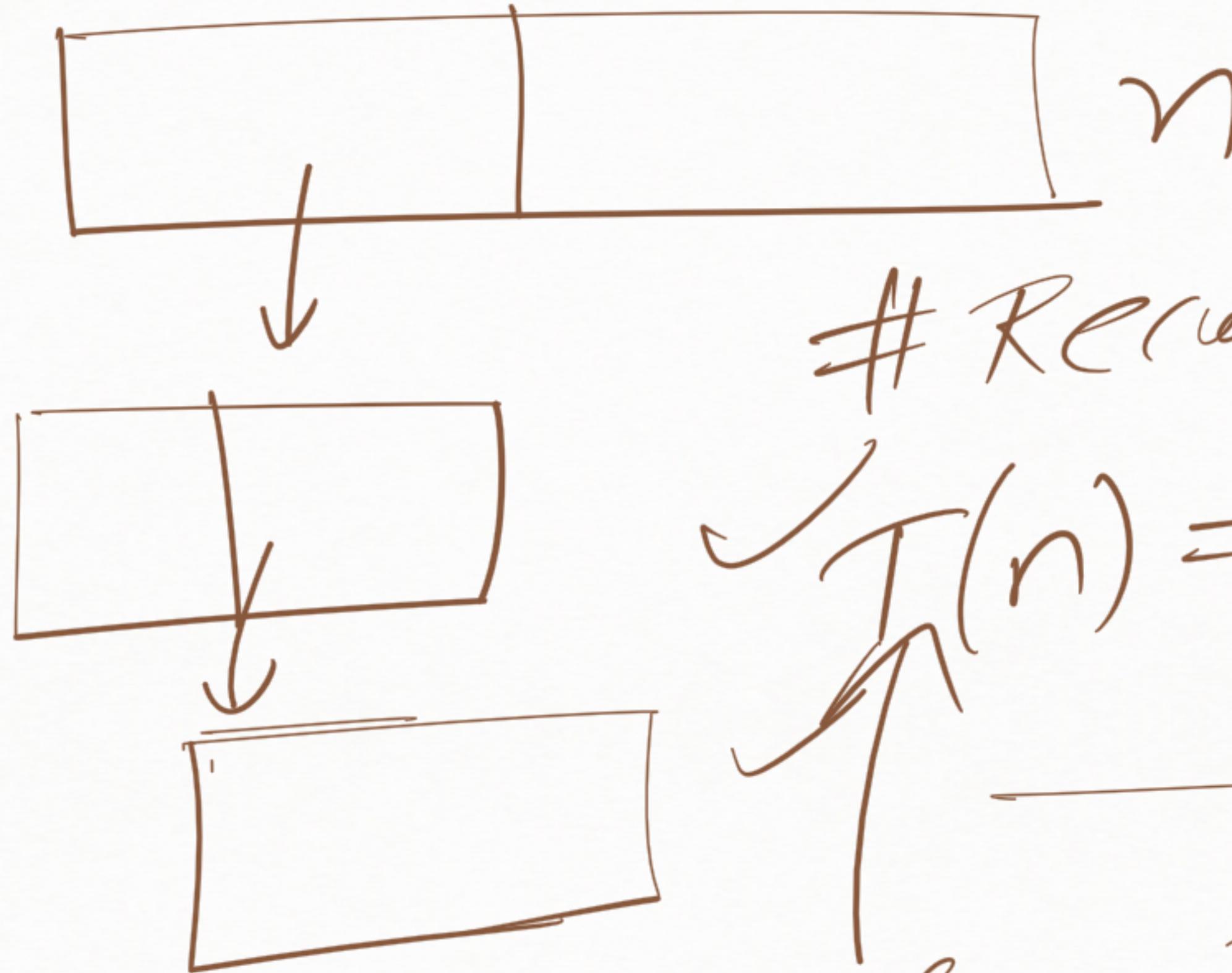


Computer will ans  
if your guessed no  
is equal, less, or  
greater than the  
hidden no.

# Binary Search

Exact Some  
Subproblem

# Divide

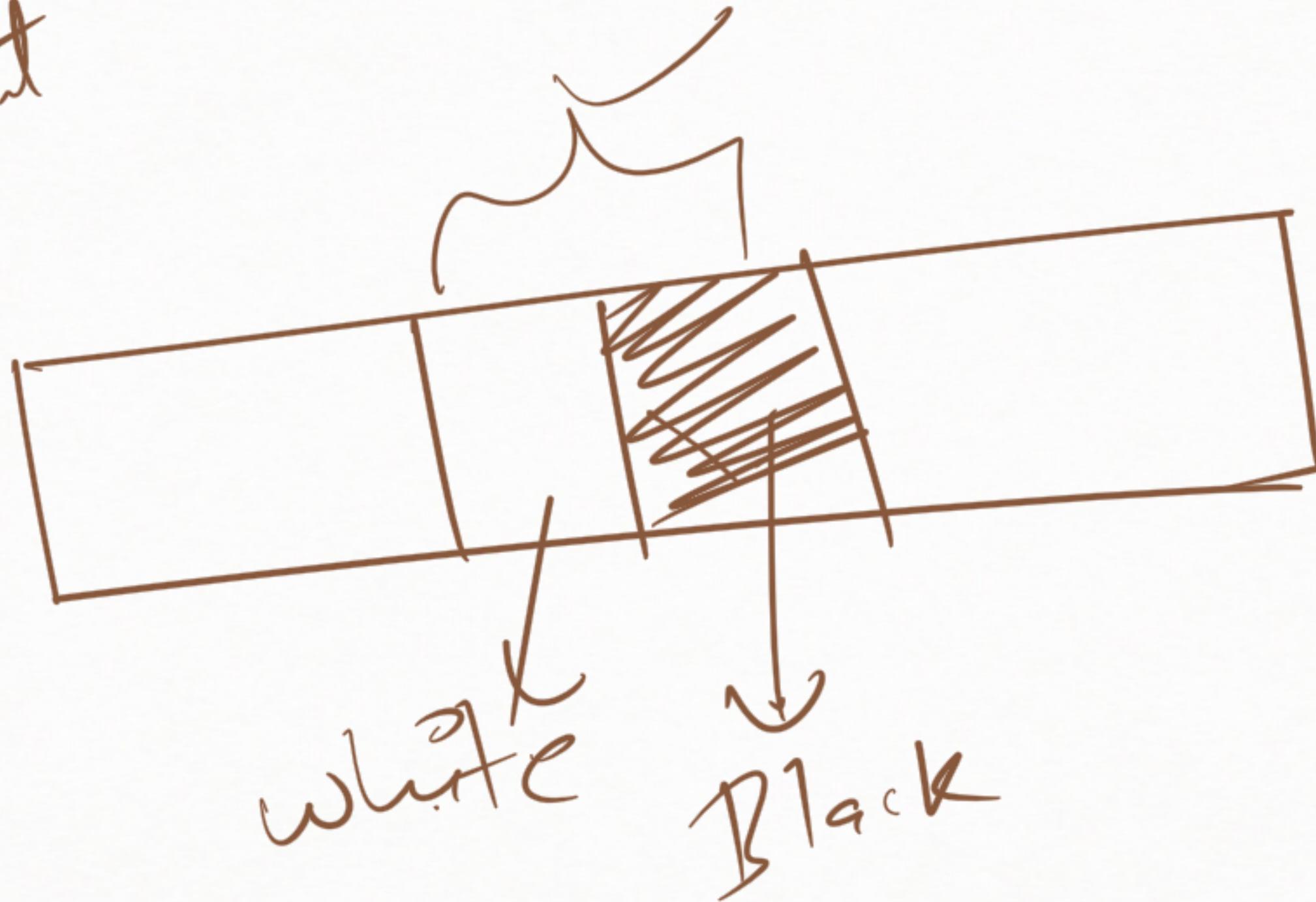


# Recurrence Relation

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

Time complexity

# Two adjacent  
cell game



Binary Search

# # Polynomial Multiplication (Divide & Conquer)

$$A(x) = \textcircled{3}x^2 + \textcircled{2}x + \textcircled{5}$$

$$B(x) = \textcircled{5}x^2 + \textcircled{1}x + \textcircled{2}$$

$$A(x) \cdot B(x) = 15x^4 + \textcircled{3}\textcircled{1}x^3 + 6x^2 +$$

$$\textcircled{1}\textcircled{0}x^3 + 2x^2 + 4x +$$

$$25x^2 + 5x + 10$$

~~$x^3 + 10$~~

~~$B$~~

$$= 15x^4 + 13x^3 + 33x^2 + 9x + 10$$

✓ Page rank dep  
 → IB Search  
 ← engine  
 → Google

## Multiplying polynomials

Input: Two  $n - 1$  degree polynomials:

$$a_{n-1}x^{n-1} + a_{n-2}x^{n-2} + \dots + a_1x + a_0$$

$$b_{n-1}x^{n-1} + b_{n-2}x^{n-2} + \dots + b_1x + b_0$$

Output: The product polynomial:

$$c_{2n-2}x^{2n-2} + c_{2n-3}x^{2n-3} + \dots + c_1x + c_0$$

where:

$$c_{2n-2} = a_{n-1}b_{n-1}$$

$$c_{2n-3} = a_{n-1}b_{n-2} + a_{n-2}b_{n-1}$$

...

$$c_2 = a_2b_0 + a_1b_1 + a_0b_2$$

$$c_1 = a_1b_0 + a_0b_1$$

$$c_0 = a_0b_0$$

Variablen

Coefficients:

$$\mathcal{B} = 1+1+1, \\ 0+2+1,$$

$$x^2 \rightarrow P_0 - s$$

$$x^2$$

$$A = 3x^2 + 2x + 5$$

$$B = 5x^2 + 2x$$

$$C = 6x^2 + 2x$$

$$Q = 1+1, 0+2 \\ 1+0 \\ 2+0$$

## Multiplying Polynomials

### Example

Input:  $n = 3, A = (3, 2, 5), B = (5, 1, 2)$

$$A(x) = 3x^2 + 2x + 5$$

$$B(x) = 5x^2 + x + 2$$

$$A(x)B(x) = 15x^4 + 13x^3 + 33x^2 + 9x + 10$$

$$\text{Output: } C = (15, 13, 33, 9, 10)$$

3	2	5
---	---	---

$$3x^2 + 2x + 5$$

$x^4$	$x^3$	$x^2$
5	1	2

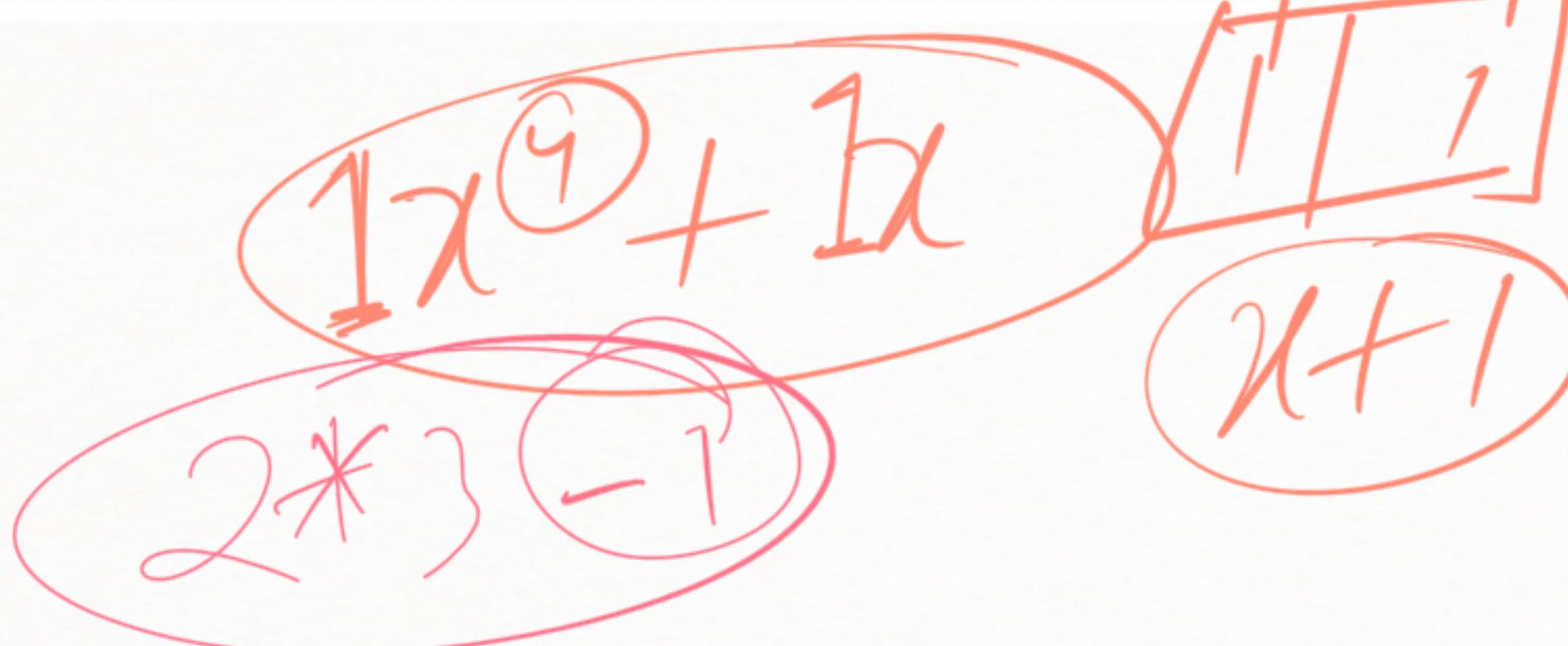
4	3	2	1	0
1	0	0	1	0

$\geq \text{size}$

$$5x^2 + 1x + 2$$

$x^4$	$x^3$	$x^2$	$x^1$	$x^0$
1	3	7	2	8

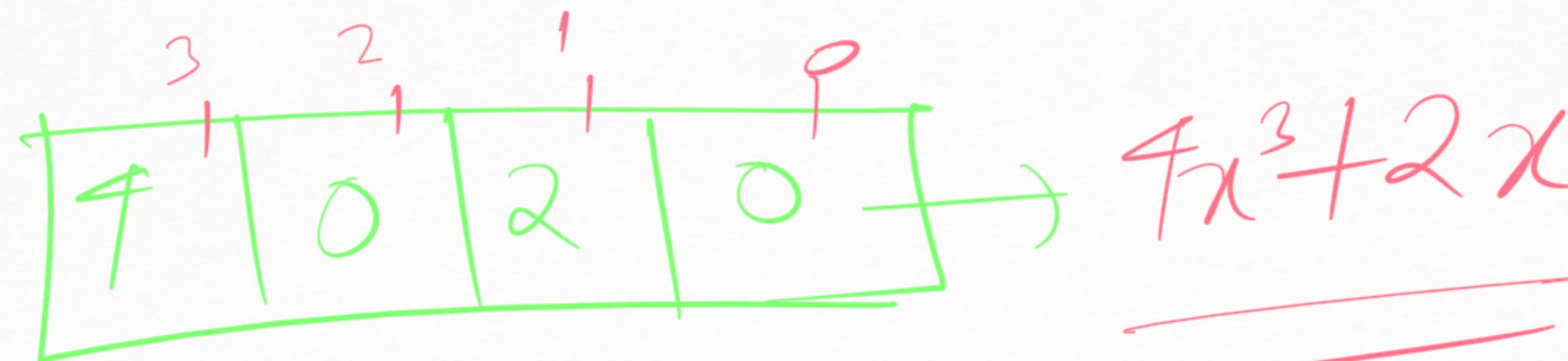
$$x^4 + 3x^3 + 7x^2 + 2x + 8$$



$$3x^2 + 7x^0 + 2$$

Every polynomial can be written in an array.

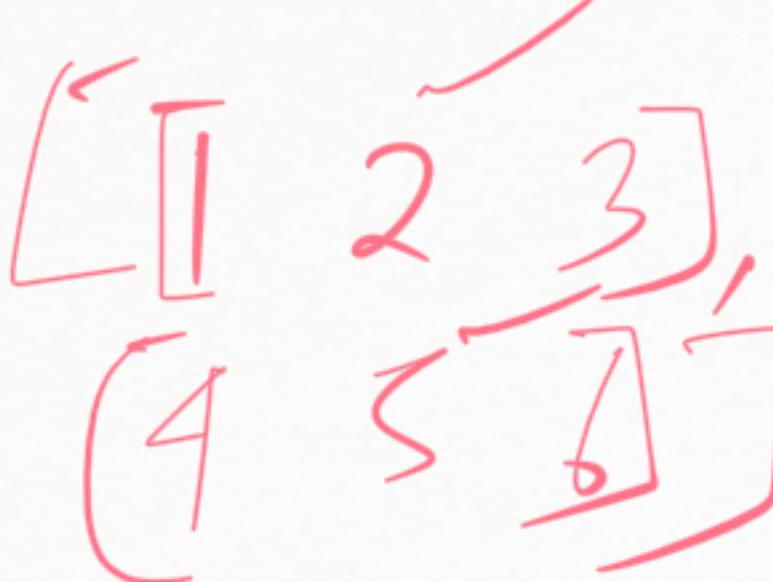
(S)	4	3	2	1	0
7	0	0	3	0	2



- # Naïve → Algo, without any efficient approach, just as the way we can solve a problem.
  - w/o applying any algo-technique like DP, Greedy, Divide & Conq.
- # Efficient → Opposite of the above VDP,  
Clean DS

MultPoly( $A, B, n$ )

```
pair  $\leftarrow$  Array[n][n]
for i from 0 to  $n - 1$ :
    for j from 0 to  $n - 1$ :
        pair[i][j]  $\leftarrow$  A[i] * B[j]
product  $\leftarrow$  Array[ $2n - 1$ ]
for i from 0 to  $2n - 1$ :
    product[i]  $\leftarrow$  0
for i from 0 to  $n - 1$ :
    for j from 0 to  $n - 1$ :
        product[i + j]  $\leftarrow$  product[i + j] + pair[i][j]
return product
```



#N<sup>o</sup>ve

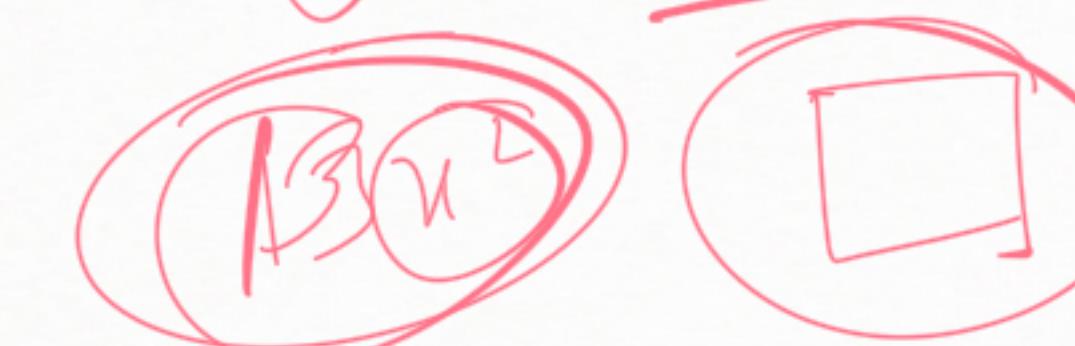
Input:  $A, B \rightarrow$  Polynomial  
~~n~~  $\rightarrow$  Size of Array.

Time for the earlier problem

Pair[i][j]  $\rightarrow$   $2^D$  Accuracy

[ ] [ ]

Product Array ( $2n - 1$ )



# Time Complexity of Naïve Approach.

$O(n^2)$  → multiplications

$O(n^2)$  → Addition

Divide &  
Conquer APP:  
 $O(n^{1.58})$

$O(n^2)$  ✓  
 $n, \log n$  ✗

# Why  $O(n^{1.58})$  makes it better sense?

→ We talk about  $n$  values which are very very huge.

→  $n = 10 \text{ billion}$

$$(10 \text{ billion})^{1.58} \ll (10^9)^2$$

# # Master Theorem. (Time Complexity topic)

Theorem

If  $T(n) = aT\left(\frac{n}{b}\right) + O(n^d)$  (for constants  $a > 0, b > 1, d \geq 0$ ), then:

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

Reurrence Relation  
(Recursion Algo)

if  $d > \log_b a$       ①

if  $d = \log_b a$       ②

if  $d < \log_b a$       ③

$\log_b a$

$$\checkmark T(n) = 4T\left(\frac{n}{2}\right) + O(n) = O(n^2)$$

### Master Theorem Example 1

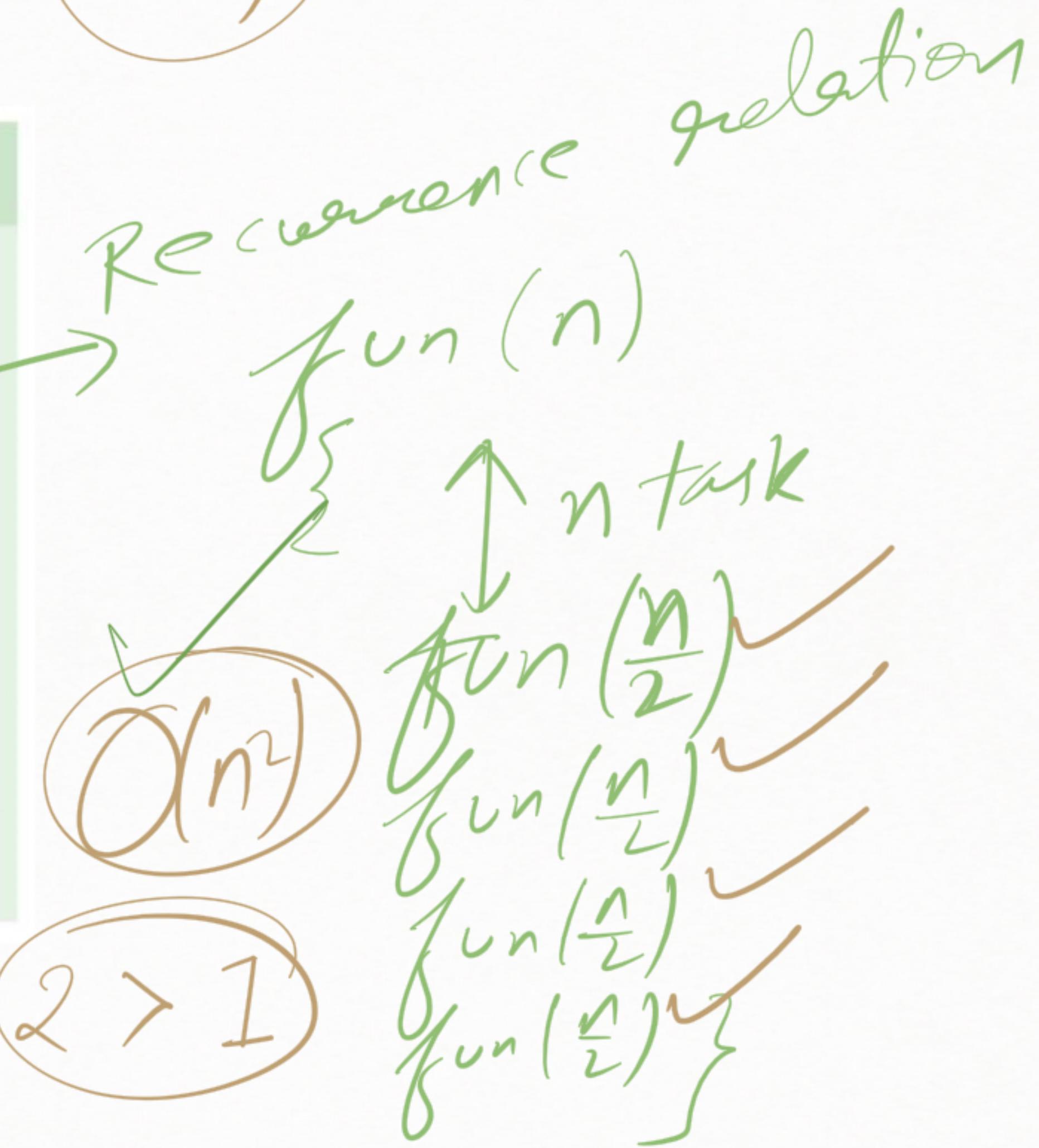
$\checkmark T(n) = 4T\left(\frac{n}{2}\right) + O(n)$

$\log_b a \quad \{ \begin{array}{l} a = 4 \\ b = 2 \end{array}$

$d = 1$

Since  $d < \log_b a$ ,  $T(n) = O(n^{\log_b a}) = O(n^2)$

$$\log_2^4 = \log_2^{2^2} = 2 > 1$$



## # Recurrence Relation [Recursive Algo]

function( $n$ )

{



}

Suppose  $T(n)$  is the time taken  
 $T(n) = 1 + T(\frac{n}{2})$

#  
Linear  
Search

①	②	③	④	⑤
7	3	16	15	9

①	②
7	3

2 units  
of time

≤ units  
of time  
 $O(n^2) \leq$   
~~Algo~~

function(n)

{  
    for (i = 0 to n)  
        { point }

function( $\frac{n}{2}$ )

function( $\frac{n}{2}$ )

function( $\frac{n}{4}$ )

}

}  $O(n)$

Suppose it takes

$T(n)$  time

$T(\frac{n}{2})$

$T(\frac{n}{2})$

$T(\frac{n}{4})$

$T(\frac{n}{4})$

$T(n)$

$T(\frac{n}{2})$

$T(\frac{n}{4})$

$T(\frac{n}{4})$

$T(n)$

$T(\frac{n}{4})$

$T(n)$

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$T(\frac{n}{2})$

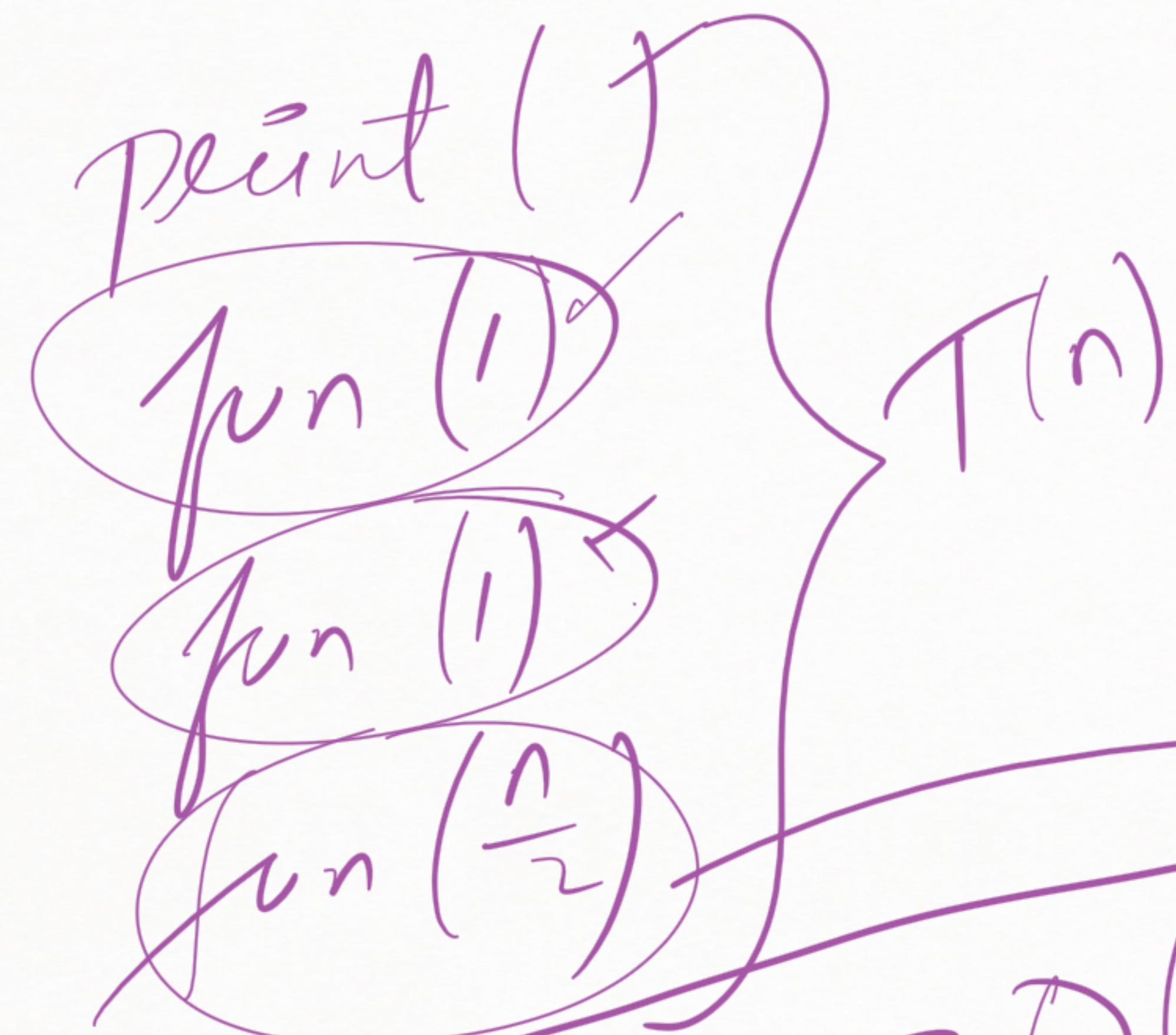
$T(\frac{n}{4})$

$T(\frac{n}{4})$

$T(n)$

$T(\frac{n}{4})$

$\text{fun}(n)$



$O(n)\sqrt{n}$  units of time  
But  $T(n)$  can take an amount of time, this upon  $n$ .

just a math

$$T(n) = 3O(1) + T\left(\frac{n}{2}\right)$$

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

function(n){

for i = 0 to n

{for j = 0 to n  
print(i,j)}

}

function ( $\frac{n}{4}$ )

$O(n^2)$   
 $T(n)$

$T\left(\frac{n}{4}\right)$

## Master Theorem Example 2

case where  $d < b$

$$\rightarrow T(n) = 3T\left(\frac{n}{2}\right) + O(n)$$

$$a = 3$$

$$b = 2$$

$$d = 1$$

$$d < \log_b a, \\ = O(n^{\log_b a}) = O(n^{\log_2 3})$$

$$d = 1$$

$$> 1 \\ < 2$$

$$a = 3 \quad = \\ \log_b b = 2$$

$$0$$

### Master Theorem Example 3

$\sim$

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

$$a = 2$$

$$b = 2$$

$$d = 1$$

$$I = \log_b a,$$

$$= O(n^d \log n) = O(n \log n)$$

$$d = 1$$

$$\log b = 1$$
$$a = 2 = 1$$
$$O(n^1 \log^1)$$

### Master Theorem Example 4

$$T(n) = T\left(\frac{n}{2}\right) + O(1)$$

$$a = 1$$

$$b = 2$$

$$d = 0$$

$$d = \log_b a, T(n) = O(n^d \log n) = \\ \log n = O(\log n)$$

$$a = 1$$

$$b = 2$$

$$d = 0$$

$$n^0 = 1$$

$$O(n^0) = O(1)$$

$$O(n^0 \log^1) = O(\log n)$$

### Chebyshev Theorem Example 5

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

$$a = 2$$

$$b = 2$$

$$d = 2$$

$$d > \log_b a, T(n) = O(n^d) = O(n^2)$$

$$a = 2$$

$$b = 2$$

$$d = 2$$

$$2 > 1$$

$$= 1$$

$$O(n^d)$$

$$O(n^2)$$

orem

$T(n) = aT\left(\left\lceil \frac{n}{b} \right\rceil\right) + O(n^d)$  (for constants  
 $b > 1, d \geq 0$ ), then:

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

The bigger one  
is kept

$O(n^{\text{bigger}})$

~~Dev~~ Webhooks

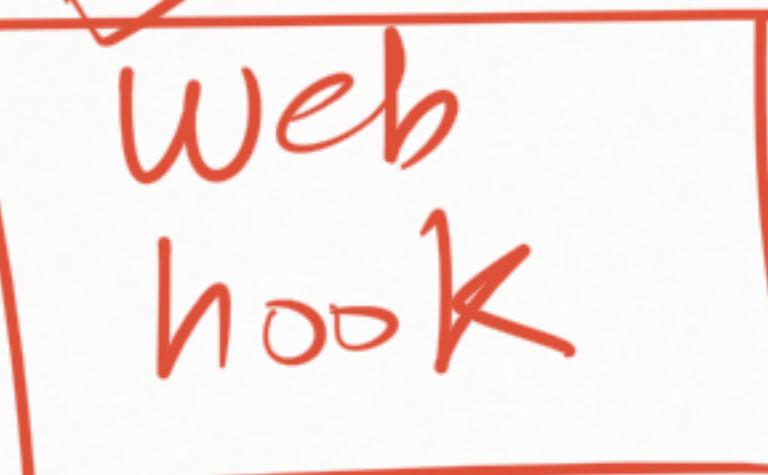
Payload



Output / Response

Reverse API

Event



Payload

on  
on

