

$$1000! = 1000 * 999 * 998 * \dots * 1$$

**Multiplication of Two n-digit numbers**

$Z = x * y$ ;  $O(1)$  is it?

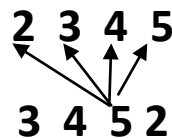
`int x[ ], y[ ], z[ ];`

`x[] = 2345678912345678`

`y[] = 2145678912245676`

`z[] = mul(x * y)`

**School level algorithm for multiplication**



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..... n muls + n-1 addition

.....x "

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$O(n^2)$

**Can we do it better?**

**How?**

**Yes, using divide and conquer approach**

$O(n^{1.59})$

## Multiplication of Two n-digit numbers

**x: n-bit number**  
**y: n-bit number**

$$x = \boxed{x_L} \boxed{x_R} = 2^{n/2} x_L + x_R$$

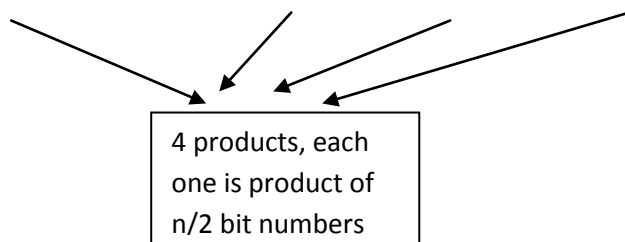
$$y = \boxed{y_L} \boxed{y_R} = 2^{n/2} y_L + y_R$$

**x=10110110,  $x_L=1011$ ,  $x_R=0110$**

**$x=(1011) * 2^4 + 0110$**

$$xy = (2^{n/2} x_L + x_R) (2^{n/2} y_L + y_R)$$

$$= 2^n x_L y_L + 2^{n/2} (x_L y_R + x_R y_L) + x_R y_R$$



The original problem is subdivided into four sub-problems and the some problems are of similar type and size of each sub-problem is half of the original problem

$T(n)$ : running time of multiplying two n-bit numbers

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

running time for combining the solutions of sub probs

where  $T(n/2)$  = running time for computing product of two  $n/2$  bit numbers

## Recurrence relations

How to solve this recurrence equation?

### (1) Substitution method

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

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

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

$$T(n) = 4[4T(n/4) + O(n/2)] + O(n)$$

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

$$= 4^2 T(n/2^2) + (2^2 - 1) O(n) = 4^2 [4T(n/8) + O(n/4)] + 3O(n)$$

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

$$= 4^3 T(n/2^3) + (2^3 - 1) O(n)$$

.....

.....

$$= 4^k T(n/2^k) + (2^k - 1) O(n)$$

$$= 4^{\log_2 n} T(1) + (2^{\log_2 n} - 1) O(n)$$

$$= 4^{\log_2 n} + (2^{\log_2 n} - 1) O(n) \quad [T(1)=1]$$

$$= n^{\log_2 4} + (n-1)O(n)$$

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

$$= O(n^2)$$

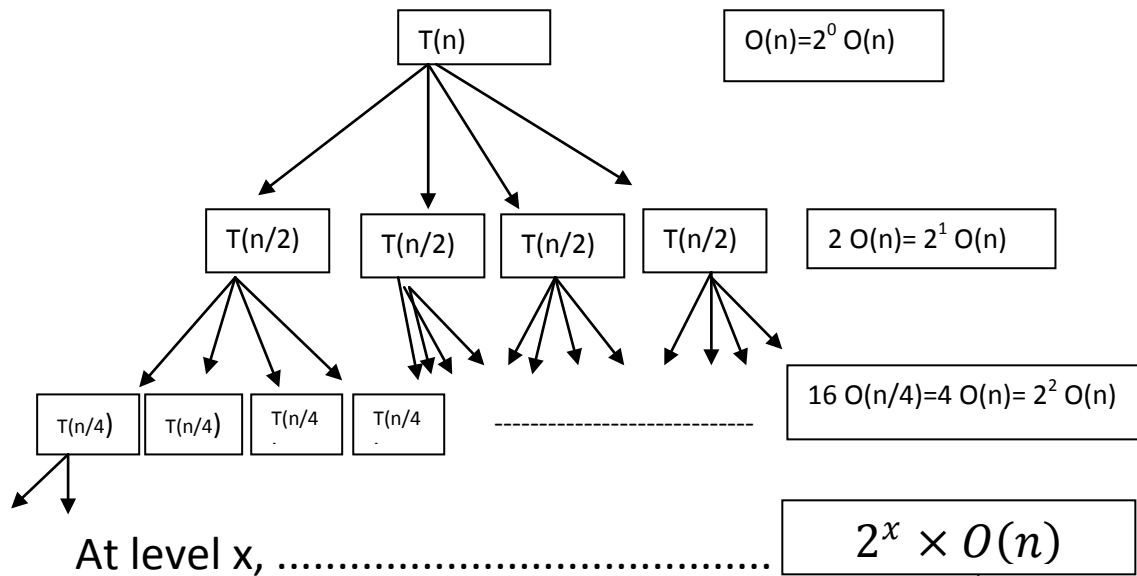
Say at step k, n  
reduces to 1  
 $T(n/2^k) = T(1)$   
 $n/2^k = 1$   
 $2^k = n$   
 $k = \log_2 n$

No improvement over school level algorithm

How can

## (2) Recursion Tree

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



Say, at level  $x$ , prob size reduces to 1, that is,  $n/2^x = 1$

$x = \log_2 n$  (depth of the tree)

At level 1, number of sub problems = 4

At level 2, number of sub-problems =  $4^2$

At level  $x$ , number of sub-problems of size 1 =  $4^x$

Time spent at level  $x = 4^x \cdot T(1)$

$= 4^x \cdot 1$  (assume  $T[1] = 1$ )

$= 4^x \cdot O\left(\frac{n}{2^x}\right)$  [since  $1 = \frac{n}{2^x}$ ]

$= \left(\frac{4}{2}\right)^x \times O(n) = 2^x \times O(n)$

Say, at level  $x$ , size of the sub problem reduces to 1

$$T(n) = \sum_{i=0}^x 2^i O(n)$$

$$T(n) = \sum_{i=0}^{\log_2 n} 2^i O(n)$$

$$= O(2^{\log_2 n} O(n)) \quad [\text{since the last term is the largest}]$$

$$= O(n \cdot O(n)) = O(n^2)$$

Can we have an algorithm which is better than this?

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

Here dividing the problem into 4 sub-problems

If we want to do better, we need to concentrate on how to reduce number of sub-problems?

1777-1855

C. F gauss

$(a + bi)(c + di) = ac - bd + (bc + ad)i$  (4 different products leading to 4 sub-problems)

$bc + ad = (a+b)(c+d) - ac - bd$

$(a + bi)(c + di) = ac - bd + \{(a+b)(c+d) - ac - bd\}i$  (3 products leading to 3 sub-problems)

1 product gain.

How this idea can be used in designing an efficient algorithm for multiplication!!

**x: n-bit number**

**y: n-bit number**

$$x = \boxed{x_L} \boxed{x_R}$$

$$= 2^{n/2} x_L + x_R$$

$$y = \boxed{y_L} \boxed{y_R}$$

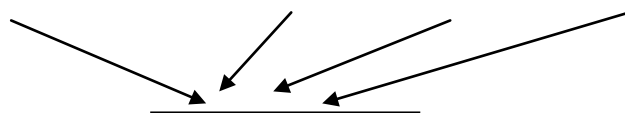
$$= 2^{n/2} y_L + y_R$$

$$x = \text{10110110}, x_L = 1011, x_R = 0110$$

$$x = (1011) * 2^4 + 0110$$

$$xy = (2^{n/2} x_L + x_R) (2^{n/2} y_L + y_R)$$

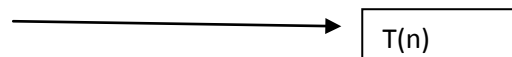
$$= 2^n x_L y_L + 2^{n/2} (x_L y_R + x_R y_L) + x_R y_R$$



$$x_L y_R + x_R y_L = (x_L + x_R)(y_L + y_R) - x_L y_L - x_R y_R =$$

**P3-P1-P2**

Function multiply(x,y)



Input: n-bit positive integers x and y

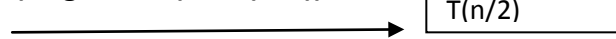
output: product

if(n==1): return xy

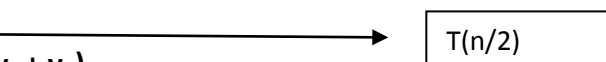
$x_L, x_R = \text{leftmost } \lceil n/2 \rceil, \text{rightmost } \lfloor n/2 \rfloor$

$y_L, y_R = \text{leftmost } \lceil n/2 \rceil, \text{rightmost } \lfloor n/2 \rfloor$

P1 = multiply(x<sub>L</sub>, y<sub>L</sub>)



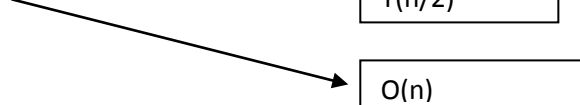
P2 = multiply(x<sub>R</sub>, y<sub>R</sub>)



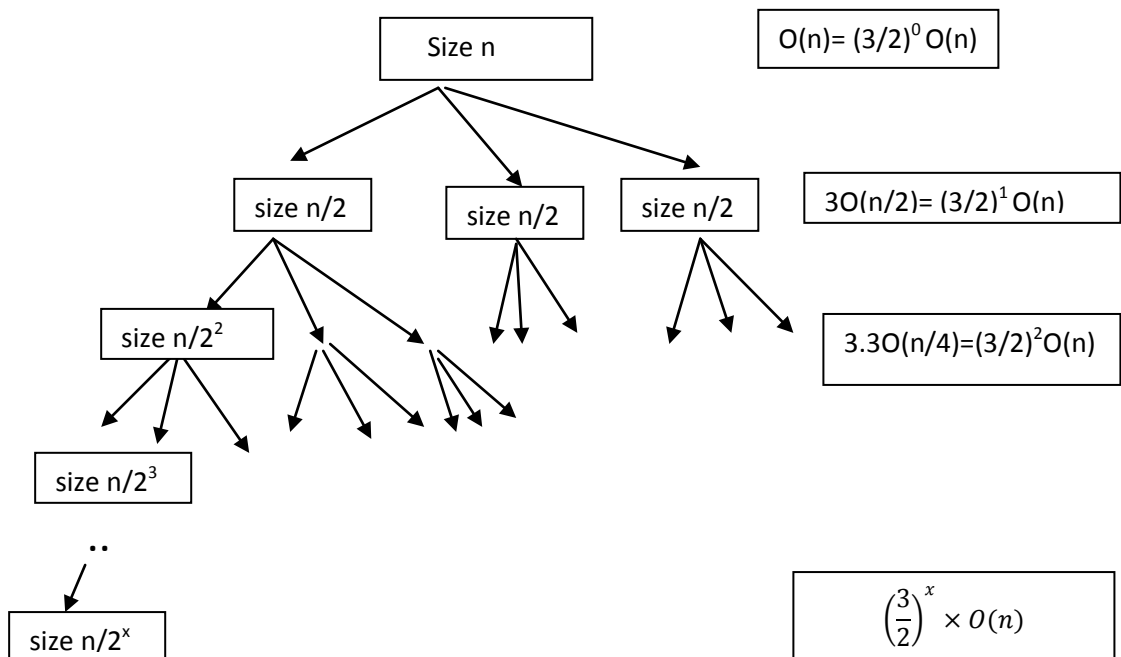
P3 = multiply(x<sub>L</sub> + x<sub>R</sub>, y<sub>L</sub> + y<sub>R</sub>)



return P1 \* 2<sup>n</sup> + (P3 - P1 - P2) \* 2<sup>n/2</sup> + P2



$$T(n) = 3T(n/2) + O(n)$$



Say, at level  $x$ , prob size reduces to 1, that is,  $n/2^x = 1$

$$x = \log_2 n$$

At level 1, number of sub problems = 3

At level 2, number of sub-problems =  $3^2$

At level  $x$ , number of sub-problems of size 1 =  $3^x$

Time spent at level  $x = 3^x \cdot T(1)$

$$= 3^x \cdot 1 \quad (\text{assume } T[1] = 1)$$

$$= 3^x \cdot O\left(\frac{n}{2^x}\right) \quad [\text{since } 1 = \frac{n}{2^x}]$$

$$= \left(\frac{3}{2}\right)^x \times O(n)$$

$$T(n) = \sum_{i=0}^{\log_2 n} \left[ \left(\frac{3}{2}\right)^i \times O(n) \right]$$

$$= O\left(\left(\frac{3}{2}\right)^{\log_2 n} \times O(n)\right) \quad [\text{since last term of } T(n) \text{ is the largest}]$$

$$= O\left(\frac{3^{\log_2 n}}{2^{\log_2 n}} \times O(n)\right)$$

$$= O(3^{\log_2 n}) = O(n^{\log_2 3}) = O(n^{1.59})$$

Mergesort Algorithm