

# Data Structures and Algorithms

(CS210A)

Semester I – 2014-15

## Lecture 13:

- Algorithm paradigms
- Algorithm paradigm of Divide and Conquer
- Majority element : Proof of correctness of the algorithm from last class

# Algorithm Paradigms

# Algorithm Paradigm

## Motivation:

- Many problems whose algorithms are based on a common approach.
- A need of a systematic study of the characteristics of such widely used approaches.

## Algorithm Paradigms:

- **Divide and Conquer**
- **Greedy Strategy**
- **Dynamic Programming**
- **Local Search**

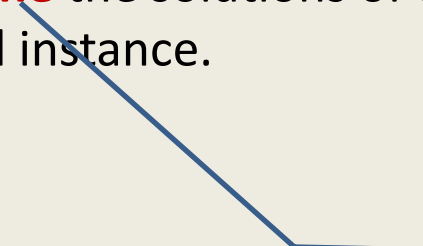
# **Divide and Conquer** paradigm for **Algorithm Design**

# Divide and Conquer paradigm

## An Overview

A problem in this paradigm is solved in the following way.

1. **Divide** the problem instance into two or more instances of the same problem.
2. Solve each smaller instances recursively (base case suitably defined).
3. **Combine** the solutions of the smaller instances to get the solution of the original instance.



This is usually the main **nontrivial** step in the design of an algorithm using divide and conquer strategy

# Example 1

## Sorting

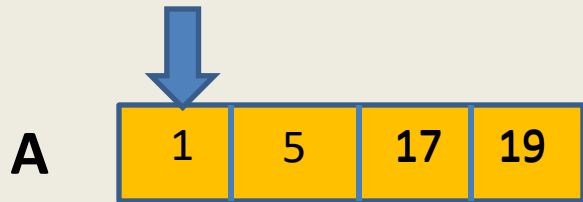
# A problem in Practice sheet 1 (8 August)

## Merging two sorted arrays:

Given two sorted arrays **A** and **B** storing  $n$  elements each, Design an  $O(n)$  time algorithm to output a sorted array **C** containing all elements of **A** and **B**.

**Example:** If **A**={1,5,17,19} **B**={4,7,9,13}, then output is **C**={1,4,5,7,9,13,17,19}.

# Merging two sorted arrays A and B





# Pesudo-code for Merging two sorted arrays

**Merge**(A[0.. $n-1$ ], B[0.. $m-1$ ], C) // Merging two sorted arrays **A** and **B** into array **C**.

{  $i \leftarrow 0$ ;  $j \leftarrow 0$ ;

$k \leftarrow 0$ ;

**While**( $i < n$  and  $j < m$ )

{     **If**(A[ $i$ ] < B[ $j$ ]) {     C[ $k$ ]  $\leftarrow$  A[ $i$ ];  $k++$ ;  $i++$      }

**Else**             {     C[ $k$ ]  $\leftarrow$  B[ $j$ ];  $k++$ ;  $j++$      }

}

**While**( $i < n$ ) { C[ $k$ ]  $\leftarrow$  A[ $i$ ];  $k++$ ;  $i++$  }

**While**( $j < m$ ) { C[ $k$ ]  $\leftarrow$  B[ $j$ ];  $k++$ ;  $j++$  }

return C;

}

**Correctness** : An exercise

Time Complexity =  
 **$O(n+m)$**

# Divide and Conquer based sorting algorithm

**MSort**(A,  $i, j$ ) // Sorting the subarray A[ $i..j$ ].

```
{ If (  $i < j$  )  
  {  $mid \leftarrow (i+j)/2$ ;  
    MSort(A,  $i, mid$ );  
    MSort(A,  $mid+1, j$ );  
    Create temporarily C[ $0..j - i$ ]  
    Merge(A[ $i..mid$ ], A[ $mid+1..j$ ], C);  
    Copy C[ $0..j - i$ ] to A[ $i..j$ ]  
  }  
}
```

Divide step

Combine/conquer step



This is **Merge Sort**  
algorithm

# Divide and Conquer based sorting algorithm

```

MSort(A, i, j) // Sorting the subarray A[i..j].
{
  If ( i < j )
  {
    mid ← (i+j)/2;
    MSort(A, i, mid); ← T(n/2)
    MSort(A, mid+1, j); ← T(n/2)
    Create temporarily C[0..j - i]
    Merge(A[i..mid], A[mid+1..j], C); } c n
    Copy C[0..j - i] to A[i..j]
  }
}
    
```

Time complexity:

If  $n = 1$ ,

$T(n) = c$  for some constant  $c$

If  $n > 1$ ,

$$\begin{aligned}
 T(n) &= c n + 2 T(n/2) \\
 &= c n + c n + 2^2 T(n/2^2) \\
 &= c n + c n + c n + 2^3 T(n/2^3) \\
 &= c n + \dots (\log n \text{ terms}) \dots + c n \\
 &= O(n \log n)
 \end{aligned}$$

# Proof of correctness of Merge-Sort

**MSort**(A,  $i, j$ ) // Sorting the subarray A[ $i..j$ ].

```
{ If (  $i < j$  )  
  {  $mid \leftarrow (i+j)/2$ ;  
    MSort(A,  $i, mid$ );  
    MSort(A,  $mid+1, j$ );  
    Create temporarily C[ $0..j-i$ ]  
    Merge(A[ $i..mid$ ], A[ $mid+1..j$ ], C);  
    Copy C[ $0..j-i$ ] to A[ $i..j$ ]  
  }
```

**Question:** What is to be proved ?

**Answer:** **MSort**(A,  $i, j$ ) sorts the subarray A[ $i..j$ ]

**Question:** How to prove ?

**Answer:**

- By **induction** on the length ( $j - i + 1$ ) of the subarray.
- Use correctness of the algorithm **Merge**.

## **Example 2**

**Faster algorithm for multiplying two  
integers**

# Addition is **faster** than multiplication

**Given:** any two  $n$ -bit numbers  $X$  and  $Y$

**Question:** how many **bit-operations** are required to compute  $X+Y$  ?

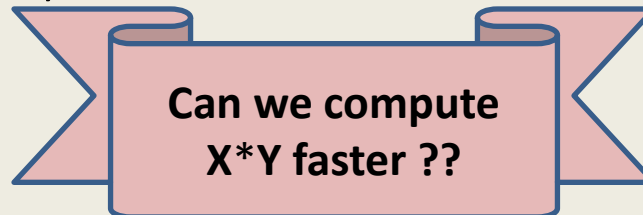
**Answer:**  $O(n)$

**Question:** how many **bit-operations** are required to compute  $X * 2^n$  ?

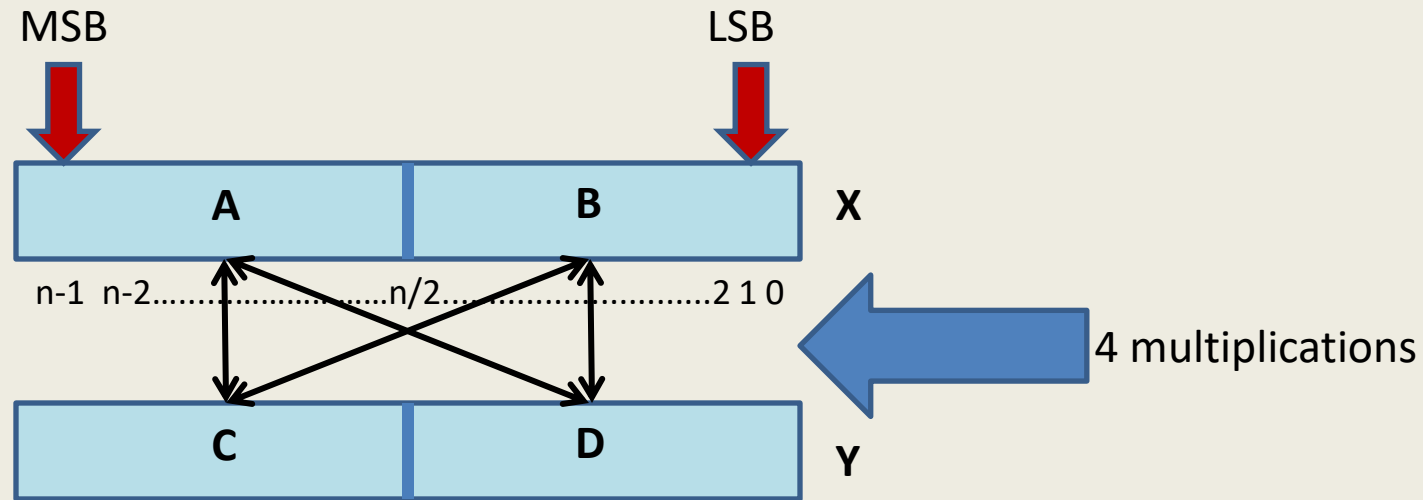
**Answer:**  $O(n)$  [left shift the number  $X$  by  $n$  places, (do it carefully)]

**Question:** how many **bit-operations** are required to compute  $X * Y$  ?

**Answer:**  $O(n^2)$  (why ??)



# Pursuing Divide and Conquer approach



**Question:** how to express  $X*Y$  in terms of multiplication/addition of  $\{A,B,C,D\}$  ?

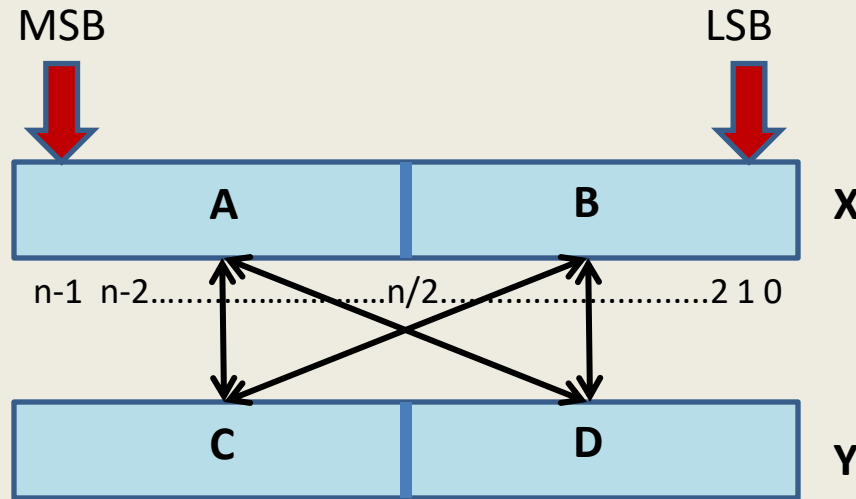
**Hint:** First Express  $X$  and  $Y$  in terms of  $\{A,B,C,D\}$ .

$$X = A * 2^{n/2} + B \quad \text{and} \quad Y = C * 2^{n/2} + D$$

Hence ...

$$X*Y = (A*C)*2^n + (A*D + B*C)*2^{n/2} + B*D$$

# Pursuing Divide and Conquer approach



$$X * Y = (A * C) * 2^n + (A * D + B * C) * 2^{n/2} + B * D$$

Let  $T(n)$  : time complexity of multiplying  $X$  and  $Y$  using the above equation.

$$T(n) = c n + 4 T(n/2) \text{ for some constant } c$$

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

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

$$= c n + 2c n + 4c n + 8c n + \dots + 4^{\log_2 n} T(1)$$

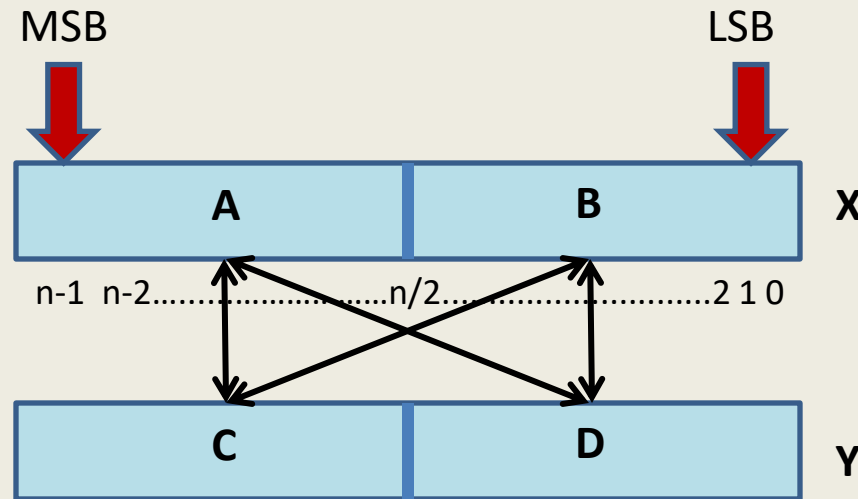
$$= c n + 2c n + 4c n + 8c n + \dots + c n^2$$



$O(n^2)$  time algo



# Pursuing Divide and Conquer approach



$$X * Y = (A * C) * 2^n + (A * D + B * C) * 2^{n/2} + B * D$$

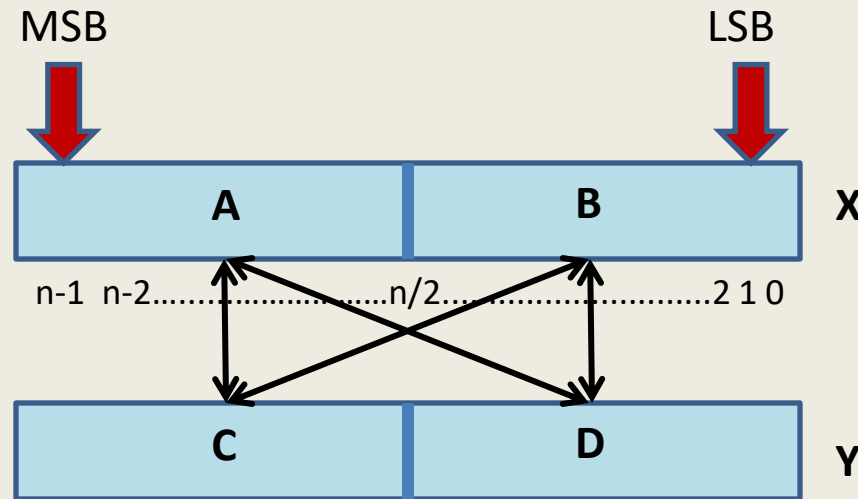
**Observation:**  $A * D + B * C = (A - B) * (D - C) + A * C + B * D$

**Question:** How many multiplications do we need now to compute  $X * Y$  ?

**Answer:** 3 multiplications :

- $A * C$
- $B * D$
- $(A - B) * (D - C)$  .

# Pursuing Divide and Conquer approach



$$X * Y = (A * C) * 2^n + ((A - B) * (D - C) + A * C + B * D) 2^{n/2} + B * D$$

Let  $T(n)$  : time complexity of the **new algo** for multiplying two  $n$ -bit numbers

$$T(n) = c n + 3 T(n/2) \text{ for some constant } c$$

$$= c n + 3 c \frac{n}{2} + 3^2 T(n/2^2)$$

$$= c n + 3 c \frac{n}{2} + 9 c \frac{n}{4} + \dots + 3^{\log_2 n} T(1)$$

$$= O(n^{\log_2 3}) = O(n^{1.58})$$

# Conclusion

**Theorem:** There is a **divide and conquer** based algorithm for multiplying any two  $n$ -bit numbers in  $O(n^{1.58})$  time (**bit operations**).

## Note:

The fastest algorithm for this problem runs in almost  $O(n \log n)$  time.  
One such algorithm was designed in **2008** at CSE, IIT Kanpur.

By (Dey, Kurur, Saha, and Saptharishi).

## Example 3

Counting the number of “*inversions*”  
in an array

# Counting Inversions in an array

## Problem description

**Definition (Inversion):** Given an array **A** of size **n**, a pair  $(i,j)$ ,  $0 \leq i < j < n$  is called an inversion if  $A[i] > A[j]$ .

**Example:**

	0	1	2	3	4	5	6	7
<b>A</b>	3	15	8	19	9	67	11	27

**Inversions are :**  $(1,2)$ ,  $(1,4)$ ,  $(3,4)$ ,  $(1,6)$ ,  $(3,6)$ ,  $(5,6)$ ,  $(5,7)$

**AIM:** An efficient algorithm to count the number of inversions in an array **A**.

# Counting Inversions in an array

## Problem familiarization

Trivial-algo( $A[0..n-1]$ )

```
{ count  $\leftarrow$  0;  
  For(j=1 to n-1) do  
  {   For( i=0 to j-1 )  
      {   If ( $A[i]>A[j]$ ) count  $\leftarrow$  count + 1;  
      }  
  }  
}
```

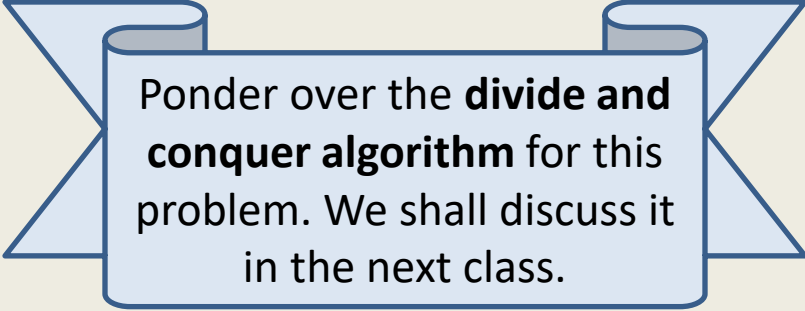
Time complexity:  $O(n^2)$

**Question:** What can be the max. no. of inversions in an array  $A$  ?

**Answer:**  $\binom{n}{2}$ , which is  $O(n^2)$ .

**Question:** Is the algorithm given above optimal ?

**Answer:** No, our aim is not to report all inversions but to report the count.




Ponder over the **divide and conquer algorithm** for this problem. We shall discuss it in the next class.

# **Proof of correctness**

**Algorithm for majority element**

# Algorithm for majority element

```
Algo-majority(A){  
    count ← 0;  
    for(i=0 to n-1)  
    {    if (count=0)  
        {    x ← A[i];  
            count ← 1;  
        }  
        else if(x ≠ A[i])  
            count ← count - 1;  
        else  
            count ← count + 1;  
    }  
    if x appears more than n/2, then  
        print(x is majority element)  
    else print(no majority element)
```



**Question:** What is to be proved ?

**Answer:** For every possible instance of **A**, the output of algorithm is correct.

**Observation:** If **A** does not have any majority element, the output of the algorithm is correct.



**Inference:** To prove correctness, it suffices to prove the following:

If **A** has a majority element, say  **$\alpha$** , then at the end of the **For** loop,  **$x = \alpha$**

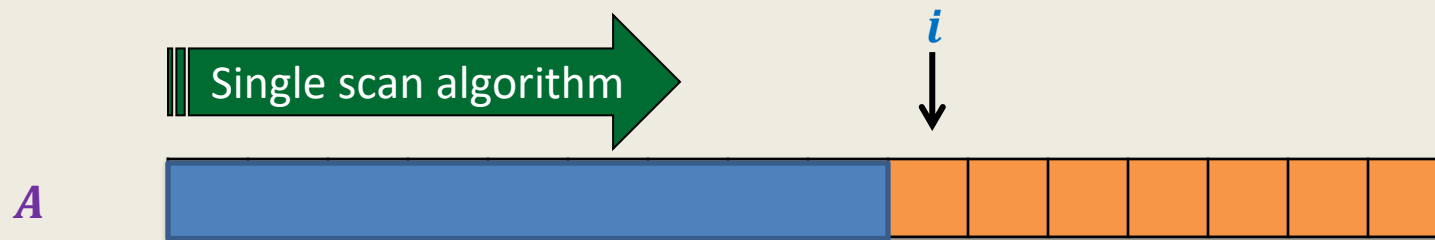


# Algorithm for majority element

```
Algo-majority(A){  
    count ← 0;  
    for(i=0 to n-1)  
    {    if (count=0)  
        {    x ← A[i];  
            count ← 1;  
        }  
        else if(x <> A[i])  
            count ← count - 1;  
        else  
            count ← count + 1;  
    }  
    if x appears more than n/2, then  
        print(x is majority element)  
    else print(no majority element)
```

**Inference:** To prove correctness, it suffices to prove the following:

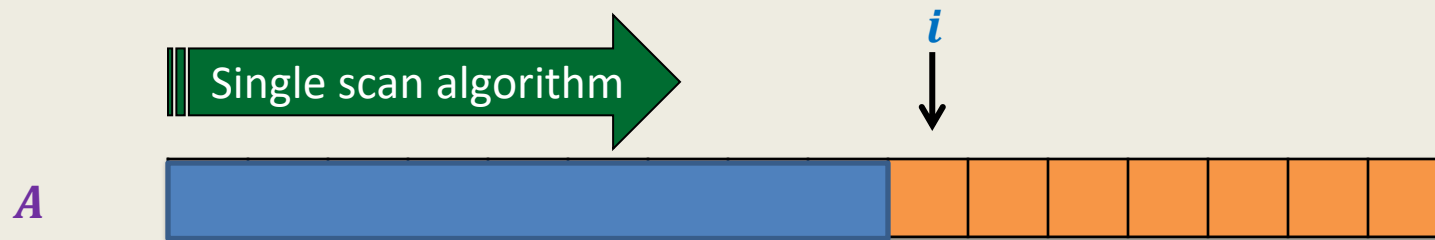
If **A** has a majority element, say  **$\alpha$** , then at the end of the **For** loop, **x =  $\alpha$**



**Question:** What assertion holds at the end of  $i$ th iteration ?

**A wrong answer:**  $x$  is a majority element of  $\{A[0], A[1], \dots, A[i-1]\}$  ?





**Question:** What assertion holds at the end of  $i$ th iteration ?

**Answer:**

$P(i) : \alpha$  is a majority element of  $\{x, \dots, \text{count times } x, A[i], \dots, A[n-1]\}$  ?

**Question:** What is  $P(n)$  ?

**Answer:**  $\alpha$  is a majority element of  $\{x, \dots, \text{count times } x\}$

$\rightarrow x = \alpha$

