

# DIVIDE AND CONQUER PARADIGM

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# THE PARADIGM I

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Solves a problem **recursively** applying at each level of the recursion

- ▶ **DIVIDE** the problem into smaller sub-problems.
- ▶ **CONQUER** via recursive calls.
- ▶ **COMBINE** solutions of sub-problems into one for the original problem.

Also consider a **base case** when the sub-problems become small enough.

# THE PARADIGM II

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## MERGE SORT: RETAKEN

MERGE( $A, p, q, r$ )

MERGE-SORT( $A, p, r$ )

```
1: if  $p < r$  then  
2:    $q = \lfloor (p + r) / 2 \rfloor$   
3:   MERGE-SORT( $A, p, q$ )  
4:   MERGE-SORT( $A, q, r$ )  
5:   MERGE( $A, p, q, r$ )  
6: end if
```

```
1:  $B = 1^{st}$  part of array.  
2:  $C = 2^{nd}$  part of array.  
3:  $i = 1, j = 1$   
4: for  $k = 1$  to  $n$  do  
5:   if  $B[i] < C[j]$  then  
6:      $A[k] = B[i++]$   
7:   else  
8:      $A[k] = C[j++]$   
9:   end if  
10: end for
```

# THE PARADIGM III

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## INSIGHTS

- ▶ Sub-problems sizes can be any! e.g.  $1/2$ ,  $1/3$ , etc.
- ▶ Base case is often too naive.
- ▶ Generally, in the third step relies the good performance.
- ▶ Trying to make third step simple is the best way to tackle the problem.

# RECURRENCES I

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## A RECURRENCE

- ▶ provide a natural way to describe the running time of divide-and-conquer algorithms.
- ▶ describes  $T(n)$  in terms of the running time of recursive calls.
- ▶ consider a base case for which

$$T(n) \leq a$$

- ▶ consider the larger inputs as

$$T(n) \leq f(n)$$

# RECURRENCES II

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## COMPUTING RECURRENCES: MERGE SORT

- Base case: sort two numbers. Swap positions in the worst case.

$$T(1) = 1$$

- Larger inputs: two sub-problems of size  $n/2$  plus one call to merge with size  $n$ , i.e.

$$T(n) = 2T(n/2) + \Theta(n)$$

# RECURRENCES III

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leading to

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ 2T(n/2) + \Theta(n) & \text{if } n > 1 \end{cases}$$

help us to get the running time

$$T(n) = \Theta(n \log_2 n)$$

and for the properties of asymptotic notation  $T(n) = O(n \log n)$

# EXAMPLE: COUNTING INVERSIONS I

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## MOTIVATION

- ▶ List of ranked objects  $A = [a_1, a_2, a_3, \dots, a_n]$  and a reference list  $B$ .
- ▶ We want to figure out how similar these lists are.
- ▶ Application [Collaborative Filtering](#).

Let  $B$  be the sorted list of  $A$ , then what we seek is the number of inversions in list  $A$ , i.e. the number of pairs  $(i, j)$  such that

$$i < j \text{ and } A[i] > A[j]$$

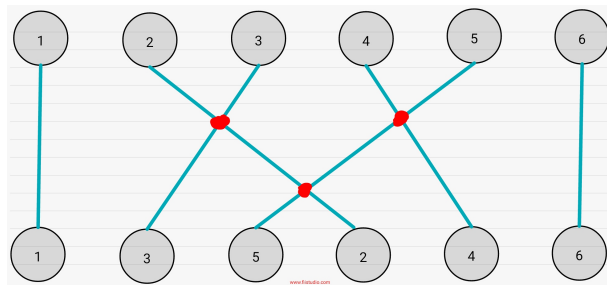


## EXAMPLE: COUNTING INVERSIONS II

EXAMPLE:  $A = [1, 3, 5, 2, 4, 6]$

Inversions:

$(3, 2), (5, 2), (5, 4)$



In general the upper bound of number of inversions is  $n(n - 1)/2$ .

## EXAMPLE: COUNTING INVERSIONS III

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A brute force algorithm

BRUTE-FORCE( $A, n$ )

```
1: count = 0
2: for  $i = 1, \dots, n$  do
3:   for  $j = i + 1, \dots, n$  do
4:     if  $A[i] > A[j]$  then
5:       count = count + 1
6:     end if
7:   end for
8: end for
9: return count
```

It is correct but

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

We can do better!

## EXAMPLE: COUNTING INVERSIONS IV

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The key idea is to use divide and conquer. Let be  $(i, j)$  an inversion with  $i < j$

1. Let's call it a left inversion if  $i, j \leq n/2$ .
2. Let's call it a right inversion if  $i, j > n/2$ .
3. Let's call it a split inversion if  $i \leq n/2 < j$

compute 1 and 2 recursively and implement other routine for 3.

## EXAMPLE: COUNTING INVERSIONS V

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COUNT-INVERSIONS( $A, n$ )

```
1: if  $n = 1$  then  
2:   return 0  
3: end if  
4:  $x = \text{COUNT-INVERSIONS}(\text{first half of } A, n/2)$   
5:  $y = \text{COUNT-INVERSIONS}(\text{second half of } A, n/2)$   
6:  $z = \text{COUNT-SPLIT-INVERSIONS}(A, n/2)$   
7: return  $x + y + z$ 
```

The goal is to implement efficiently COUNT-SPLIT-INVERSIONS for good performance.

## EXAMPLE: COUNTING INVERSIONS VI

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The key idea to implement COUNT-SPLIT-INVERSIONS is to consider sorting, more precisely the merge sub-routine of merge-sort

MERGE( $A, p, q, r$ )

- ▶  $B$  = left part of array.
- ▶  $C$  = second part of array.
- ▶  $A$  = output array.

```
1:  $i = 1, j = 1$ 
2: for  $k = 1$  to  $n$  do
3:   if  $B[i] < C[j]$  then
4:      $A[k] = B[i++]$ 
5:   else
6:      $A[k] = C[j++]$ 
7:   end if
8: end for
```

## EXAMPLE: COUNTING INVERSIONS VII

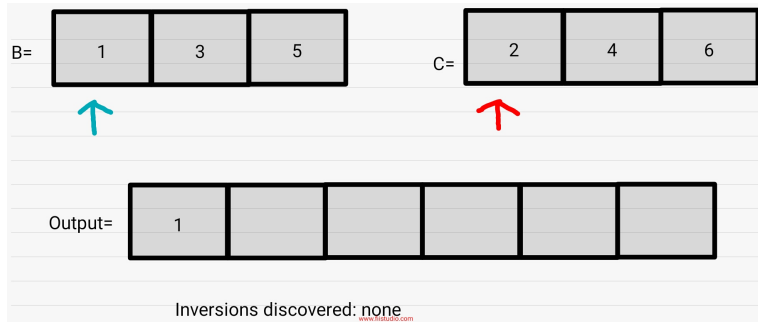
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Now suppose  $A$  has no split inversions, then following properties holds

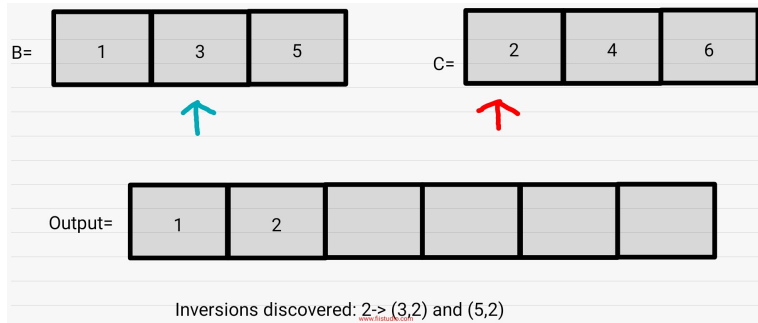
- ▶ All elements in  $B$  are less than the elements in  $C$ .
- ▶ All elements in  $B$  are copied back to  $A$  before the elements in  $C$ .

The second half has the clue to discover the inversions!

# EXAMPLE: COUNTING INVERSIONS VIII

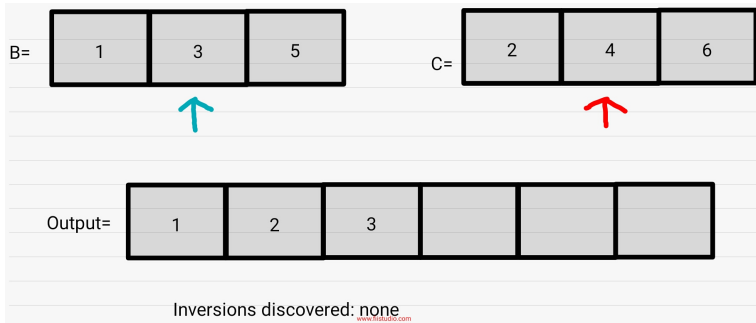


# EXAMPLE: COUNTING INVERSIONS IX

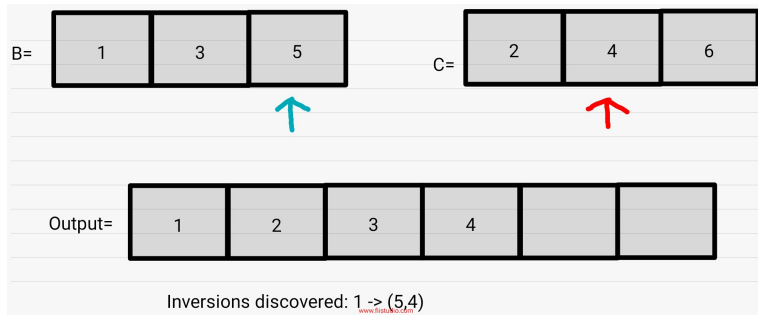




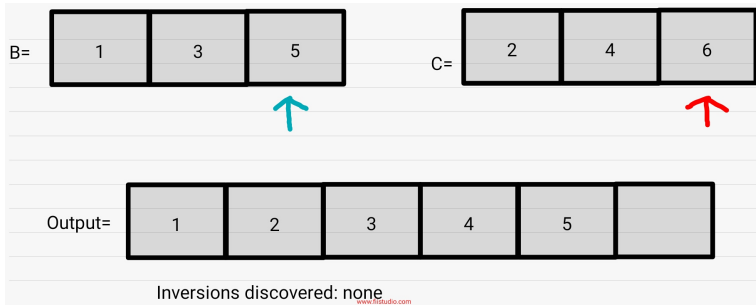
# EXAMPLE: COUNTING INVERSIONS X



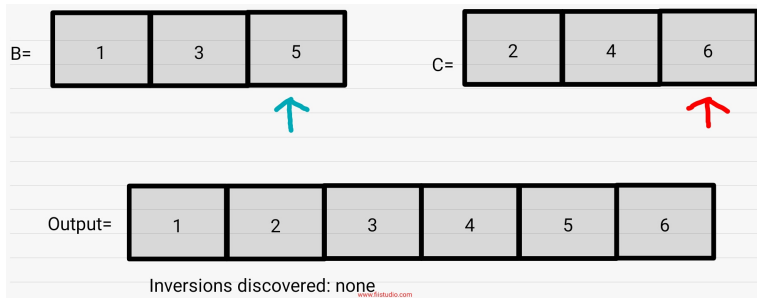
# EXAMPLE: COUNTING INVERSIONS XI



# EXAMPLE: COUNTING INVERSIONS XII



# EXAMPLE: COUNTING INVERSIONS XIII



## EXAMPLE: COUNTING INVERSIONS XIV

The split inversions involving an element  $y_i \in C$  is the number of remaining positions from  $i$  in  $B$  when it is copied to the output array  $D$ .



# EXAMPLE: COUNTING INVERSIONS XV

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## THE ALGORITHM

COUNT-INVERSIONS( $A, n$ )

- 1: **if**  $n = 1$  **then**
- 2:     **return** 0
- 3: **end if**
- 4:  $B, x = \text{COUNT-INVERSIONS}(\text{first half of } A, n/2)$
- 5:  $C, y = \text{COUNT-INVERSIONS}(\text{second half of } A, n/2)$
- 6:  $D, z = \text{COUNT-SPLIT-INVERSIONS}(B, C, n/2)$
- 7: **return**  $x + y + z, D$

# EXAMPLE: COUNTING INVERSIONS XVI

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COUNT-SPLIT-INVERSIONS( $B, C, n$ )

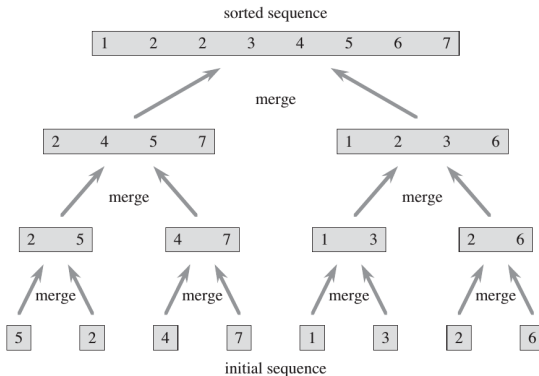
```
1:  $i = 1, j = 1, inv = 0$ 
2: for  $k = 1$  to  $n$  do
3:   if  $B[i] < C[j]$  then
4:      $D[k] = B[i++]$ 
5:   else
6:      $inv++ = n/2 - i$ 
7:      $D[k] = C[j++]$ 
8:   end if
9: end for
10: return  $inv, D$ 
```

The recurrence is

$$T(n) = \begin{cases} O(1) & \text{if } n = 1 \\ 2T(n/2) + O(n) & \text{if } n > 1 \end{cases}$$

What is the running time?

# SOLVING RECURRENCES : RECURSION TREE



At each level  $j = 0, 1, \dots, \log_2(n)$ , there are  $2^j$  subproblems of size  $n/2^j$ .



# SOLVING RECURRENCES : THE MASTER METHOD I

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- ▶ Also known as the Master Theorem.
- ▶ Black box for solving recurrences.
- ▶ Take as input few parameters to get the solution.
- ▶ **Assumption** all sub-problems have the same size.
- ▶ We will only consider the method that yields upper bounds, i.e.  $O()$ .

# SOLVING RECURRENCES : THE MASTER METHOD II

## RECURRENCE FORMAT

- ▶ Base case:  $T(n) \leq \alpha$  for sufficiently small  $n$ .
- ▶ For larger  $n$

$$T(n) \leq aT(n/b) + O(n^d)$$

where

$a$  is the number of recursive calls ( $\geq 1$ ).

$b$  is the factor by which the input size shrinks ( $> 1$ ).

$d$  exponent in summing time of combining step ( $\geq 0$ ).

and  $a, b, c \perp n$

# SOLVING RECURRENCES : THE MASTER METHOD III

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## THE METHOD

$$T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \text{ (1)} \\ O(n^d) & \text{if } a < b^d \text{ (2)} \\ O(n^{\log_b a}) & \text{if } a > b^d \text{ (3)} \end{cases}$$

# SOLVING RECURRENCES : THE MASTER METHOD IV

## COUNTING INVERSIONS SOLUTION

The parameters

- ▶ Two recursive calls to COUNT-INVERSIONS:  $a = 2$ .
- ▶ Divide the array into two sub-problems:  $b = 2$ .
- ▶ COUNT-SPLIT-INVERSIONS runs in linear time  $d = 1$ .
- ▶  $a = b^d?$  :  $2 = 2^1$ .

Falls into case # 1

$$T(n) = O(n \log n)$$

# SOLVING RECURRENCES : THE MASTER METHOD V

## MORE: BINARY SEARCH

BINARY-SEARCH( $A, a, i, p, r$ )

```
1: if  $p < r$  then  
2:    $q = \lfloor (p + r)/2 \rfloor$   
3:   if  $A[q] == a$  then  
4:      $i = q$   
5:   end if  
6:   if  $A[q] > a$  then  
7:     BINARY-  
       SEARCH( $A, a, i, p, q - 1$ )  
8:   else  
9:     BINARY-  
       SEARCH( $A, a, i, q + 1, r$ )  
10:  end if  
11: end if
```

The parameters

- ▶  $a = 1$ .
- ▶  $b = 2$ .
- ▶  $d = 0$ , no combining step.

Falls into case # 1

$$T(n) = O(\log n)$$

# SOLVING RECURRENCES : THE MASTER METHOD VI

## MORE: STRASSEN ALGORITHM FOR MATRICES MULTIPLICATION

The recurrence is

$$T(n) = \begin{cases} O(1) & \text{if } n = 1 \\ 7T(n/2) + O(n^2) & \text{if } n > 1 \end{cases}$$

The parameters:  $a = 7, b = 2, d = 2$ .  $a > b^d$ , i.e.  $7 > 4$ , falls into case # 3

$$T(n) = O(n^{\log_2 7}) = O(n^{2.81})$$

and beats the naive  $O(n^3)$ .