

# Divide and Conquer

CSC 411

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# Review: Decrease-and-Conquer

1. Reduce problem instance to **smaller** instance of the same problem
  2. **Solve** smaller instance
  3. **Extend solution** of smaller instance to obtain solution to original instance
- Can be implemented either top-down or bottom-up
  - (Richard doesn't love this terminology.)

# 3 Types of Decrease and Conquer

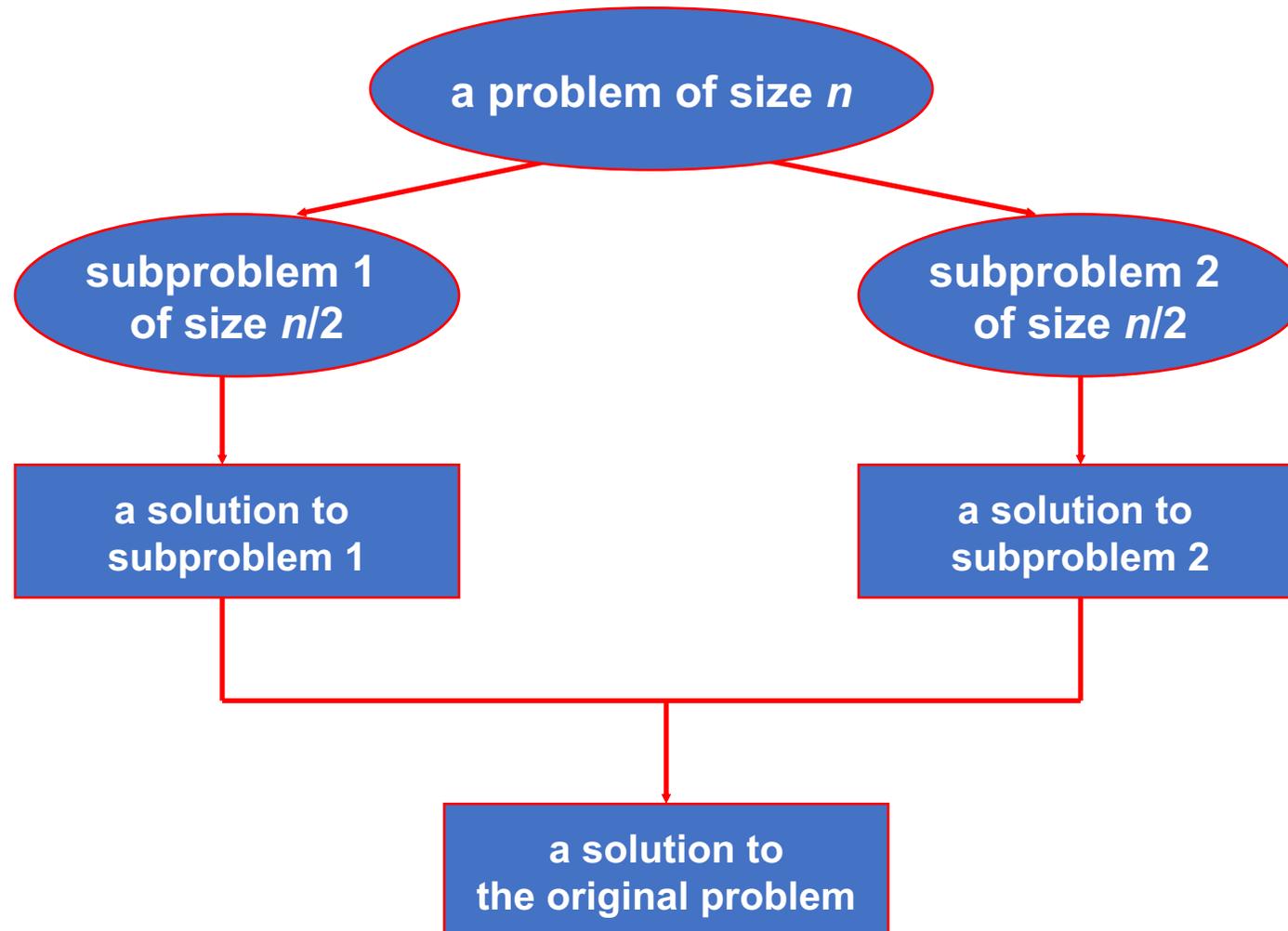
- Decrease by a constant (usually by 1):
  - insertion sort
  - graph traversal algorithms (DFS and BFS)
  - topological sorting
  - algorithms for generating permutations
- Decrease by a constant factor (usually by half)
  - binary search and bisection method
  - exponentiation by squaring
- Variable-size decrease
  - Euclid's algorithm
  - selection by partition

# Divide-and-Conquer

most-well known algorithm design strategy:

1. **Divide** instance of problem into two or more smaller instances
2. Solve smaller instances **recursively**
3. Obtain solution to original (larger) instance by **combining** these solutions

# Divide-and-Conquer Technique (cont.)



# Divide-and-Conquer Examples

- Sorting: mergesort and quicksort
- Binary tree traversals
- Multiplication of large integers and Matrix multiplication: Strassen's algorithm
- Closest-pair and convex-hull algorithms.

# Review: Mergesort

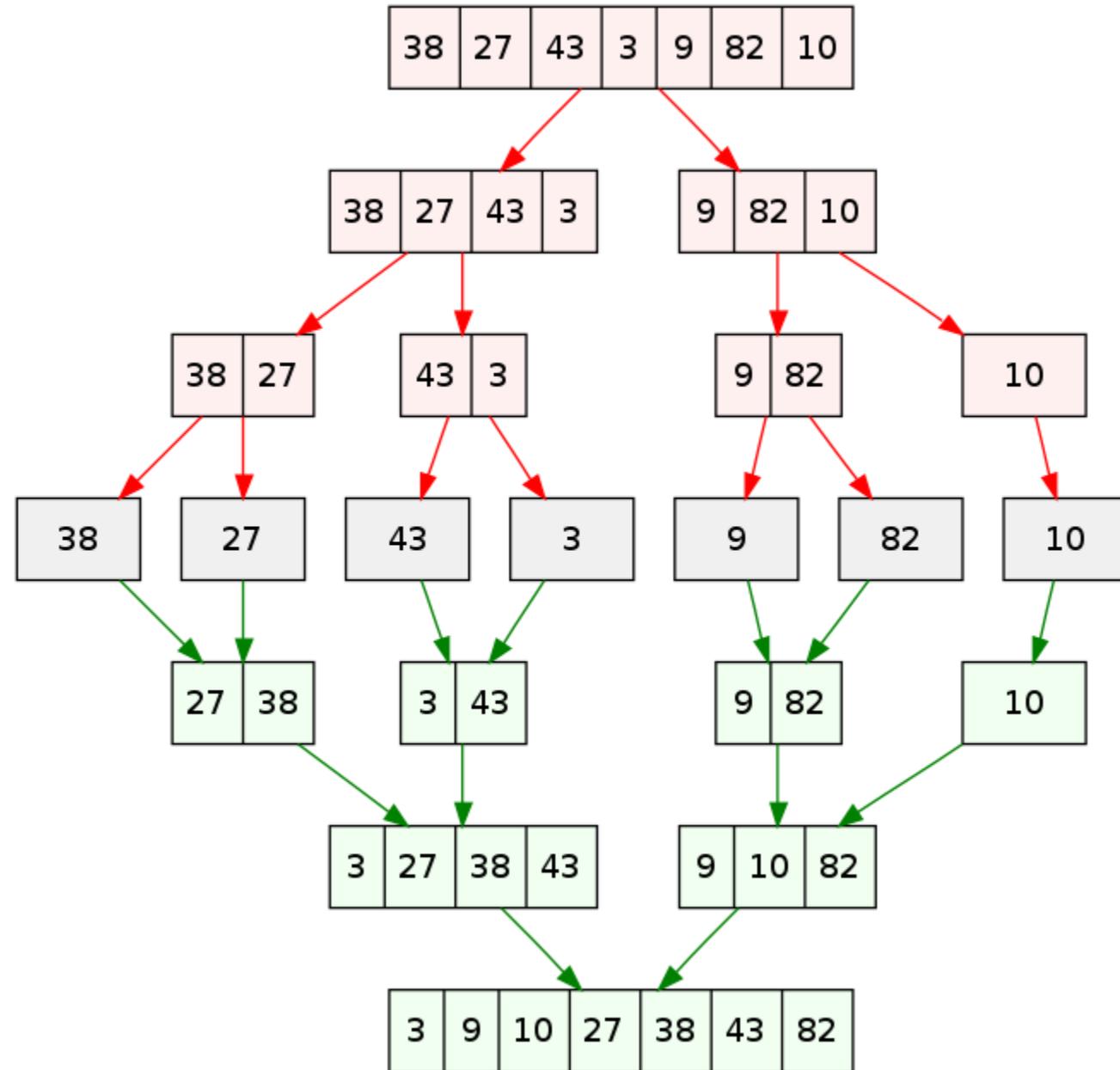
- Split array  $A[0..n-1]$  in **two about equal halves** and make copies of each half in arrays B and C
- Sort arrays B and C recursively
- **Merge** sorted arrays B and C into array A as follows:
  - Repeat the following until no elements remain in one of the arrays:
    - compare the first elements in the remaining unprocessed portions of the arrays
    - copy the smaller of the two into A, while incrementing the index indicating the unprocessed portion of that array
  - Once all elements in one of the arrays are processed, copy the remaining unprocessed elements from the other array into A.

# Pseudocode of Mergesort

**ALGORITHM** *Mergesort*( $A[0..n - 1]$ )  
//Sorts array  $A[0..n - 1]$  by recursive mergesort  
//Input: An array  $A[0..n - 1]$  of orderable elements  
//Output: Array  $A[0..n - 1]$  sorted in nondecreasing order  
**if**  $n > 1$   
    copy  $A[0..\lfloor n/2 \rfloor - 1]$  to  $B[0..\lfloor n/2 \rfloor - 1]$   
    copy  $A[\lfloor n/2 \rfloor..n - 1]$  to  $C[0..\lceil n/2 \rceil - 1]$   
    *Mergesort*( $B[0..\lfloor n/2 \rfloor - 1]$ )  
    *Mergesort*( $C[0..\lceil n/2 \rceil - 1]$ )  
    *Merge*( $B, C, A$ )

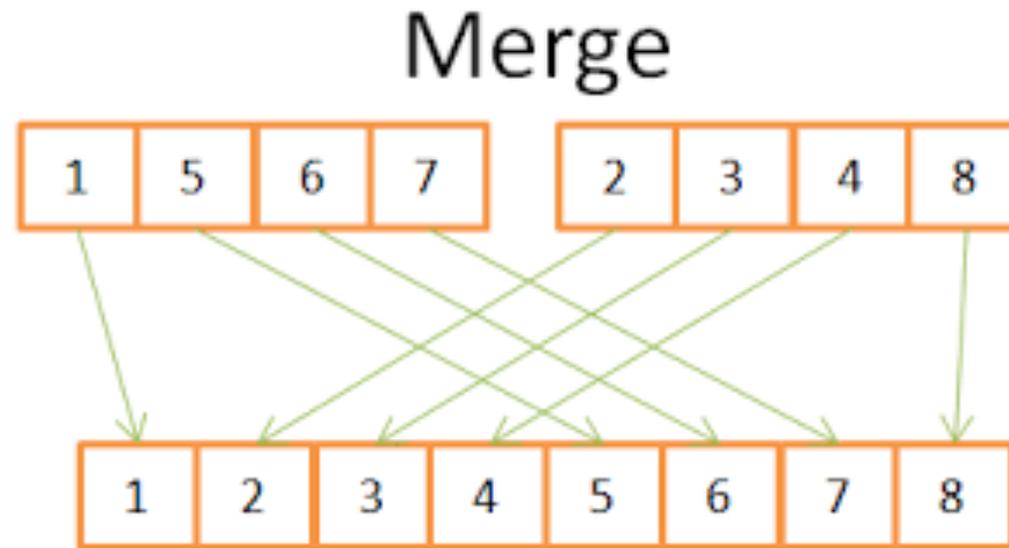
Does anything about this seem wasteful?

# Example



# Merging Two Sorted Arrays

- To merge two sorted array into a third (sorted) array, repeatedly **compare the two least elements** and copy the smaller of the two onto the third array.



# of comparison:  
7 in this example

(1, 2)

(5, 2)

(5, 3)

(5, 4)

(5, 8)

(6, 8)

(7, 8)

# Pseudocode of Merge

**ALGORITHM**  $Merge(B[0..p-1], C[0..q-1], A[0..p+q-1])$

//Merges two sorted arrays into one sorted array

//Input: Arrays  $B[0..p-1]$  and  $C[0..q-1]$  both sorted

//Output: Sorted array  $A[0..p+q-1]$  of the elements of  $B$  and  $C$

$i \leftarrow 0; j \leftarrow 0; k \leftarrow 0$

**while**  $i < p$  **and**  $j < q$  **do**

**if**  $B[i] \leq C[j]$

$A[k] \leftarrow B[i]; i \leftarrow i + 1$

**else**  $A[k] \leftarrow C[j]; j \leftarrow j + 1$

$k \leftarrow k + 1$

**if**  $i = p$

    copy  $C[j..q-1]$  to  $A[k..p+q-1]$

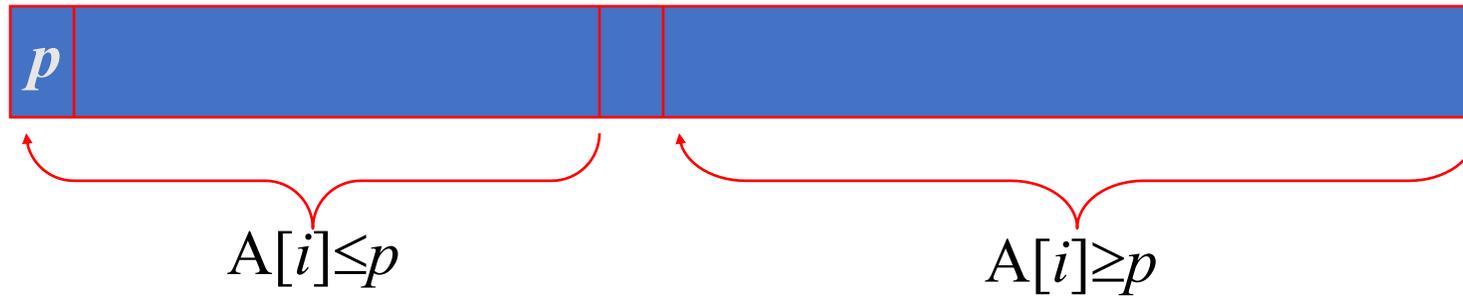
**else** copy  $B[i..p-1]$  to  $A[k..p+q-1]$

# Analysis of Mergesort

- All cases have same efficiency:  $\Theta(n \log n)$
- Why? See below...

# Quicksort

- Select a *pivot* (partitioning element) – here, the first element

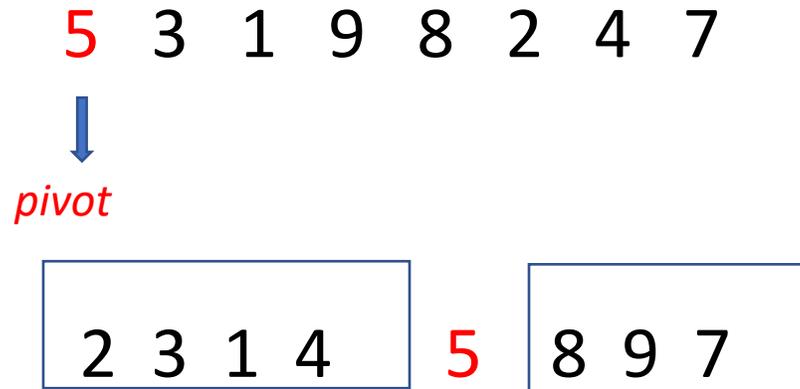


- Rearrange the list so that all the elements in the first  $s$  positions are smaller than or equal to the pivot and all the elements in the remaining  $n-s$  positions are larger than or equal to the pivot (see next slide for an algorithm)
- Exchange the pivot with the last element in the first (i.e.,  $\leq$ ) subarray — the pivot is now in its final position
- Sort the two subarrays recursively

# Quicksort Pseudocode

**ALGORITHM** *Quicksort*( $A[l..r]$ )  
//Sorts a subarray by quicksort  
//Input: A subarray  $A[l..r]$  of  $A[0..n - 1]$ , defined by its left and right indices  
//  $l$  and  $r$   
//Output: Subarray  $A[l..r]$  sorted in nondecreasing order  
**if**  $l < r$   
     $s \leftarrow \text{Partition}(A[l..r])$  //  $s$  is a split position  
    *Quicksort*( $A[l..s - 1]$ )  
    *Quicksort*( $A[s + 1..r]$ )

# Quicksort Example



the best case



the worse case

# The slickest Quicksort Implementation: Haskell

```
quicksort [] = []
quicksort (x:xs) =
    quicksort [y | y <- xs, y <= x]
  ++ [x] ++
    quicksort [y | y <- xs, y > x]
```

But probably not one you should ever use...

# Partitioning Algorithm

```
Algorithm Partition( $A[l..r]$ )  
//Partitions a subarray by using its first element as a pivot  
//Input: A subarray  $A[l..r]$  of  $A[0..n - 1]$ , defined by its left and right  
//      indices  $l$  and  $r$  ( $l < r$ )  
//Output: A partition of  $A[l..r]$ , with the split position returned as  
//      this function's value  
 $p \leftarrow A[l]$   
 $i \leftarrow l; j \leftarrow r + 1$   
repeat  
    repeat  $i \leftarrow i + 1$  until  $A[i] \geq p$   
    repeat  $j \leftarrow j - 1$  until  $A[j] < p$   
    swap( $A[i], A[j]$ )  
until  $i \geq j$   
swap( $A[i], A[j]$ ) //undo last swap when  $i \geq j$   
swap( $A[l], A[j]$ )  
return  $j$ 
```

# Analysis of Quicksort

- Best case: split in the middle  $\Theta(n \log n)$
- Why?

# Analysis of Quicksort

- Best case: split in the middle —  $\Theta(n \log n)$
  - Worst case: ??? —  $\Theta(n^2)$
  - Average case: random arrays —  $\Theta(n \log n)$
- 
- What is the worst-case for quicksort?
  - How do we analyze mergesort and quicksort?

# Master Theorem

Let  $T$  be an increasing function that satisfies the recurrence relation:

$$T(n) = a T(n/b) + cn^d$$

whenever  $n = b^k$ , where  $k$  is a positive integer,  $a \geq 1$ ,  $b$  is an integer greater than 1,  $c$  is a positive real number, and  $d$  is a non-negative real number.

Then:

$T(n)$	=	$O(n^d)$	if $a < b^d$	case 1
		$O(n^d \log_b n)$	if $a = b^d$	case 2
		$O(n^{\log_b a})$	if $a > b^d$	case 3

$a \geq 1$ : The number of subproblems

$b > 1$ : Amount by which problems shrink

$cn^d$ : Amount of nonrecursive work at each level of recursion.

The cases are (effectively) comparing  $cn^d$  with  $n^{\log_b a}$

# Analysis of Mergesort

- All cases have same efficiency:  $\Theta(n \log n)$
  - Why? What is its growth function?
    - $T(n) = 2 T(n/2) + f(n) = ?$  The time needed for merging
    - $T_{\text{worst}}(n) = 2 T_{\text{worst}}(n/2) + n - 1$  for  $n > 1, T(1) = 0$
    - Using Master Theorem
      1. If  $a < b^d$ ,  $T(n) \in \Theta(n^d)$
      2. If  $a = b^d$ ,  $T(n) \in \Theta(n^d \log n)$
      3. If  $a > b^d$ ,  $T(n) \in \Theta(n^{\log_b a})$
- $a = 2, b = 2, d = 1$   
 $\Rightarrow$  Case 2,  $T(n) \in \Theta(n \log n)$

# Analysis of Quicksort

- Best case: split in the middle  $\Theta(n \log n)$

- Why?

- $T_{\text{best}}(n) = 2 T_{\text{best}}(n/2) + f(n)$  =? The time needed for partition

- $T_{\text{best}}(n) = 2 T_{\text{best}}(n/2) + n - 1$  for  $n > 1$ ,  $T(1) = 0$

- Using Master Theorem

1. If  $a < b^d$ ,  $T(n) \in \Theta(n^d)$

2. If  $a = b^d$ ,  $T(n) \in \Theta(n^d \log n)$

3. If  $a > b^d$ ,  $T(n) \in \Theta(n^{\log_b a})$

$$a = 2, b = 2, d = 1$$

$$\Rightarrow \text{Case 2, } T(n) \in \Theta(n \log n)$$

# Analysis of Quicksort

- Best case: split in the middle —  $\Theta(n \log n)$
- Worst case: sorted array! —  $\Theta(n^2)$
- Average case: random arrays —  $\Theta(n \log n)$