

# Merging, Merge Sort, and Divide-and-Conquer

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# Merging two Arrays

**Given:** Two sorted arrays  $B$  and  $C$ .

**Return:** A sorted array containing the elements of  $B$  and  $C$ .

Merge( $B, C$ )

```
 $p = B.length, q = C.length$   
create an empty array  $D$  of length  $p + q$   
 $i = 0, j = 0$   
while  $i < p$  and  $j < q$  do  
    if  $B[i] \leq C[j]$  then  
        append  $B[i]$  to  $D$   
         $i++$   
    else  
        append  $C[j]$  to  $D$   
         $j++$   
if  $i \geq p$  then  
    for  $k = j$  to  $q - 1$  do  
        append  $C[k]$  to  $D$   
else  
    for  $k = i$  to  $p - 1$  do  
        append  $B[k]$  to  $D$   
return  $D$ 
```

# Correctness of Merge( $B, C$ )

## Loop Invariant

At the beginning of each iteration of the **while** loop,  
 $\langle D[0], \dots, D[i + j - 1] \rangle$  consists of the elements of  $\langle B[0], \dots, B[i - 1] \rangle$   
and  $\langle C[0], \dots, C[j - 1] \rangle$  in sorted order.

## Correctness of Merge( $B, C$ )

Two possibilities at termination of the **for** loop:

- $i = p$ : Loop invariant implies that all elements of  $\langle B[0] \dots, B[p-1] \rangle = B$  and  $\langle C[0], \dots, C[j-1] \rangle$  are in sorted order  $D$ , but  $\langle C[j], \dots, C[q-1] \rangle$  remain to be inserted into  $D$ .
- $j = q$ : Loop invariant implies that all elements  $\langle C[0] \dots, C[q-1] \rangle = C$  and  $\langle B[0], \dots, B[j] \rangle$  are in sorted order in  $D$ , but  $\langle B[j], \dots, B[p-1] \rangle$  remain to be inserted into  $D$ .

The final **if-then-else** statement adds the un-inserted elements to  $D$ .

# Running Time of Merge

- Let  $n = p + q$ , where  $p = B.\text{length}$  and  $q = C.\text{length}$ .
- Since each iteration of the while takes constant time, the loop takes  $O(n)$  time.
- Appending the remainder of  $B$  or  $C$  to  $D$  also takes  $O(n)$  time.

$\Rightarrow$  Merge takes  $O(n)$  time.

## Note

Merge requires  $\Theta(n)$  additional space for temporary storage.

# Merge Sort

MergeSort( $A$ )

$n = A.length$

**if**  $n == 1$  **then**

**return**  $A$

$A_{left} = \langle A[0], \dots, A[\lfloor n/2 \rfloor - 1] \rangle$

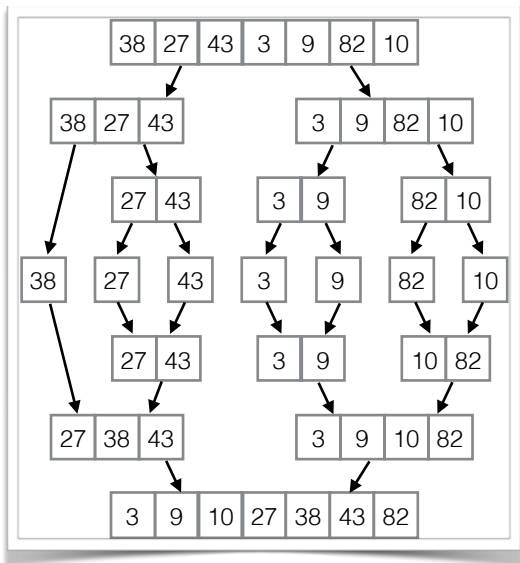
$A_{right} = \langle A[\lfloor n/2 \rfloor], \dots, A[n - 1] \rangle$

$A = \text{Merge}(\text{MergeSort}(A_{left}), \text{MergeSort}(A_{right}))$

**return**  $A$

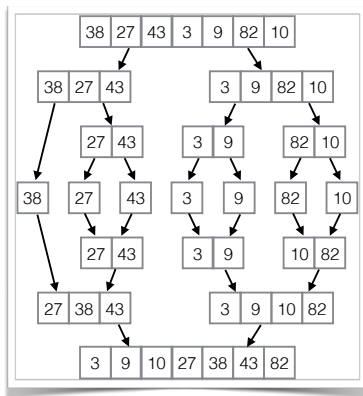


# Example



# Running Time of MergeSort

- Each level of the recursion tree involves  $O(n)$  operations.
- There are  $O(\log n)$  levels.
- Hence, MergeSort runs in  $O(n \log n)$  time.



# Divide-and-Conquer

# Divide-and-Conquer

## Components of a Divide-and-Conquer Algorithm

**Divide:** Break the problem into a number of subproblems that are smaller instances of the same problem.

**Conquer:** Solve subproblems recursively.

- If a subproblem is sufficiently small, solve it directly, without recursion, in  $\Theta(1)$  time.

**Combine:** Merge the solutions to the subproblems into the solution for the original problem.

# Divide-and-Conquer Recurrences

- A **recurrence equation** describes the running time of a recursive algorithm on problem of size  $n$  in terms of its running time on smaller inputs.
- Let  $T(n)$  be the running time on a problem of size  $n$ .
- If the problem is small enough, say  $n \leq c$  for some  $c$ , the algorithm takes  $\Theta(1)$  time.
- Suppose larger problems are divided into  $a$  subproblems, each of size  $n/b$ .
- Let  $D(n)$  be the time to divide the problem into subproblems and  $C(n)$  be the time to combine their solutions.
- Then,

$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq c, \\ aT(n/b) + D(n) + C(n) & \text{otherwise.} \end{cases}$$

# A Recurrence for MergeSort

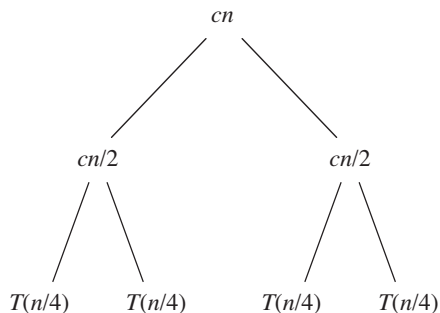
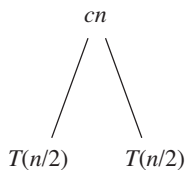
- MergeSort splits the input into  $a = 2$  subproblems.
- Each subproblem has size  $n/2$ , so  $b = 2$ .
- Splitting array in 2 and merging results after recursive calls takes  $cn$  time, for some constant  $c$ ; i.e.,  $C(n) + D(n) = cn$ .
- Thus,

$$T(n) = \begin{cases} c & \text{if } n \leq 1, \\ 2T(n/2) + cn & \text{if } n > 1. \end{cases}$$

- We can prove that  $T(n) = \Theta(n \log n)$  by expanding the recurrence.

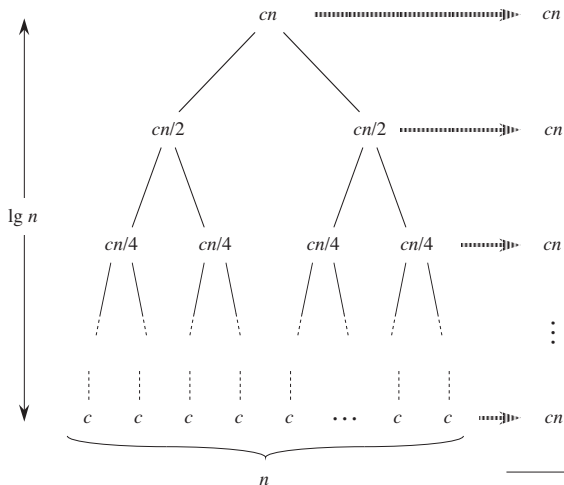
# Expanding the Recurrence

$T(n)$



Source: CLRS

# Expanding the Recurrence



(d)

Total:  $cn \lg n + cn$

Source: CLRS



## Floors and Ceilings

Our analysis of MergeSort assumed that  $n$  is always divisible by 2, so

$$T(n) = \begin{cases} c & \text{if } n \leq 1, \\ 2T(n/2) + cn & \text{if } n > 1. \end{cases} \quad (1)$$

But (1) is a simplification. A more accurate recurrence is

$$T(n) = \begin{cases} c & \text{if } n \leq 1 \\ T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + n & \text{if } n > 1. \end{cases} \quad (2)$$

It turns out that we lose nothing from the simplification.

### Fact

*If  $T(n)$  is defined by (2), then  $T(n) = \Theta(n \log n)$ .*

*See CLRS, Section 4.6.2.*

# Bibliography

# References

- [CLRS] Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest and Clifford Stein, *Introduction to Algorithms* (3rd edition), MIT Press, 2009.
- [KT] Jon Kleinberg and Éva Tardos, *Algorithm Design*, Addison-Wesley, 2006.