

3: Divide and Conquer (Analyzing Recursive ..): Binary Search, Recurrences, Repeated Substitution Method, Merge sort,

Recursive algorithms and recurrences, Merge sort.

Divide big problems into small problems of the same nature. Then solve the sub-problems recursively until it is so small that you can solve the problem trivially.

factorial n!

- Recursive algorithm

INPUT: n – a non-negative integer.
OUTPUT: fac – a non-negative integer that equals n!

```
FACTORIAL (n)
int fac=1;
if n==1 then fac=1
else fac=n*FACTORIAL(n-1)
return fac
```

Trivial case

Divide & Conquer

- Non-recursive algorithm

INPUT: n – a non-negative integer.
OUTPUT: fac – a non-negative integer that equals n!

```
FACTORIAL (n)
int fac=1;
for i from 1 to n
    fac=fac * i
return fac
```

Divide-and-conquer

Divide-and-conquer method for algorithm design:

- If the problem size is small enough to solve it in a straightforward manner, solve it. Otherwise, do the following:
 - **Divide:** Divide the problem into a number of *disjoint* subproblems.
 - **Conquer:** Use divide-and-conquer recursively to solve the subproblems.
 - **Combine:** Take the solutions to the sub-problems and combine these solutions into a solution for the original problem.

Binary Search

Non-recursive version:

INPUT: $A[1..n]$ – a sorted (non-descending) array of integers, q – an integer.
OUTPUT: an index j such that $A[j] = q$; 0, if $\forall j (1 \leq j \leq n): A[j] \neq q$

```

Left=1
Right=n
do
  j = ⌊(left+right)/2⌋
  if A[j]==q then return j
  else if A[j]>q then right=j-1
  else left=j+1
while left<=right
return 0

```

Recursive version:

INPUT: $A[1..n]$ – a sorted (non-descending) array of integers, q – an integer, l – an integer, left bound, r – an integer, right bound.
OUTPUT: an index j such that $A[j] = q$; 0, if $\forall j (1 \leq j \leq n): A[j] \neq q$

Binary-search(A, l, r, q):

```

if l == r then
  if A[l] == q then return l
  else return 0

```

Trivial case

```

m = ⌊(l+r)/2⌋

```

Divide & Conquer

```

if A[m] ≥ q then return Binary-search(A, l, m, q)
else return Binary-search(A, m+1, r, q)

```

Recurrences

- Running times of algorithms with **recursive calls** can be described using recurrences.
- A **recurrence** is an equation or inequality that describes a function in terms of its value on smaller inputs.
- Assume that
 - If the problem size is small enough, the problem can be solved in constant time, i.e., $\Theta(1)$.
 - The division of problem yields a sub-problems and each subproblem is $1/b$ the size of the original.
- We have:

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

Conquer
Divide
Combine

Recurrence on Binary Search

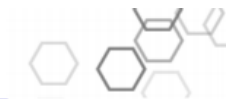
- $a = 1, b=2$, having one sub-problem with half elements in the array.
- $D(n) = \Theta(1)$, computing the middle index, constant time.
- $C(n) = 0$, no need to combine.

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

The Repeated Substitution Method

- Solving recurrences with the repeated substitution method
 - Substitute
 - Expand
 - Substitute
 - Expand
 - ...
- Observe a *pattern* and write how your expression looks after the i -th substitution.
- Find out what the value of i should be to get the base case of the recurrence $T(1)$.
- Insert the value of $T(1)$ and the expression of i into your expression.

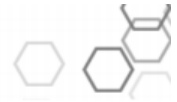
Example



- $T(n) = \begin{cases} e & \text{if } n=1 \\ T(n/2) + f & \text{otherwise, i.e., if } n>1 \end{cases}$

*e, f are constants.
 e : cost for solving a trivial case.
 f : cost for dividing*
- $T(n) = T(n/2) + f$
 - $= (T(n/4) + f) + f = T(n/4) + 2*f$
 - $= (T(n/8) + f) + 2*f = T(n/8) + 3*f$
 - $= (T(n/16) + f) + 3*f = T(n/16) + 4*f$
 - ...
 - $= T(n/2^i) + i * f$
- When $i = \lg n$, $T(n/2^i) = T(1) = e$, and thus,
 - $T(n) = e + f * \lg n$
- Drop low order terms and ignore leading constants
 - $T(n) = \Theta(\lg n)$

Mini-quiz (on Moodle)



- Try the repeated substitution method on
- $T(n) =$ Eq. A.5, CLRS p. 1147
 - a if $n=1$
 - $2T(n/2) + b$ otherwise, i.e., if $n>1$
- which is only slightly different from what we have done for
- $T(n) =$
 - e if $n=1$
 - $T(n/2) + f$ otherwise, i.e., if $n>1$
- Let's check whether it is still $\Theta(\lg n)$? If not, what is the complexity?
- $T(n) = 2T(\frac{n}{2}) + b$
- $= 2(2T(\frac{n}{2^2}) + b) + b = 2^2T(\frac{n}{2^2}) + (2+1)b$
- $= 2^2(2T(\frac{n}{2^3}) + b) + (2+1)b = 2^3T(\frac{n}{2^3}) + (2^2+2+1)b$
- $= 2^3(2T(\frac{n}{2^4}) + b) + (2^2+2+1)b = 2^4T(\frac{n}{2^4}) + (2^3+2^2+2+1)b$
- $= 2^iT(\frac{n}{2^i}) + (2^{i-1} + \dots + 2^3 + 2^2 + 2 + 1)b$
- $= 2^iT(\frac{n}{2^i}) + b * \sum_{k=0}^{i-1} 2^k$ Geometric series
CLRS p. 1147
- $= 2^iT(\frac{n}{2^i}) + b \frac{2^{(i-1)+1}-1}{2-1}$ For real $x \neq 1$, the summation
 $\sum_{k=0}^n x^k = 1 + x + x^2 + \dots + x^n$
is a *geometric* or *exponential series* and has the value
 $\sum_{k=0}^n x^k = \frac{x^{n+1} - 1}{x - 1}$.
- $= 2^iT(\frac{n}{2^i}) + b(2^i - 1)$

- $T(n) = 2^i T\left(\frac{n}{2^i}\right) + b(2^i - 1)$
- To use $T(1)=a$, we set $n/2^i=1$ and get $i=\lg n$.
- $T(n)=2^{\lg n} a + b(2^{\lg n}-1)$
- $=a*n + b*n - b$
- $=(a+b)n-b$
- $=\Theta(n)$

a = cost of solving trivial case, b = cost of dividing.

Merge Sort

- An algorithm that is able to solve the sorting problem and uses the divide-and-conquer technique.
- Assume that we are going to sort a sequence of numbers in array A .
- **Divide**
 - If A has at least two elements (nothing needs to be done if A has zero or one elements), remove all the elements from A and put them into two sequences, A_1 and A_2 , each containing about half of the elements of A . (i.e. A_1 contains the first $\lfloor n/2 \rfloor$ elements and A_2 contains the remaining $\lceil n/2 \rceil$ elements).
- **Conquer**
 - Sort sequences A_1 and A_2 using Merge Sort.
- **Combine**
 - Put back the elements into A by merging the sorted sequences A_1 and A_2 into one sorted sequence

Merge Sort: Algorithm

```

Merge-Sort(A, p, r)
  if p < r then
    q = ⌊(p+r)/2⌋
    Merge-Sort(A, p, q)
    Merge-Sort(A, q+1, r)
    Merge(A, p, q, r)

```

Merge(A, p, q, r)

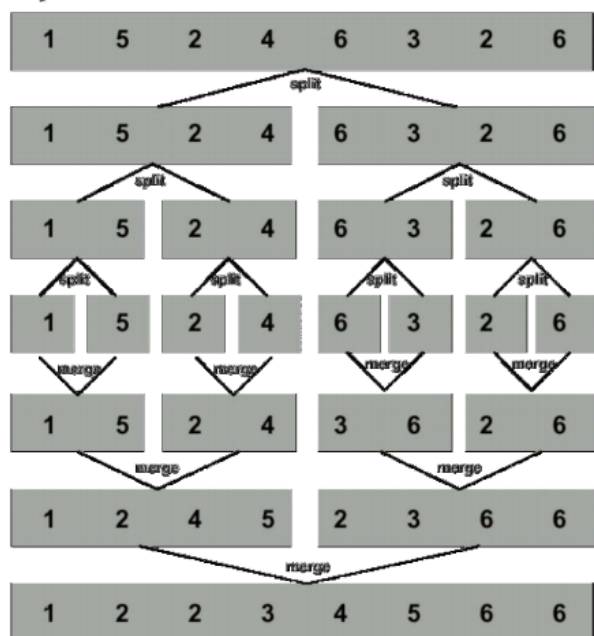
Take the smallest of the two topmost elements of sequences $A[p..q]$ and $A[q+1..r]$ and put into the resulting sequence. Repeat this, until both sequences are empty. Copy the resulting sequence into $A[p..r]$.

Merge Sort Summarized



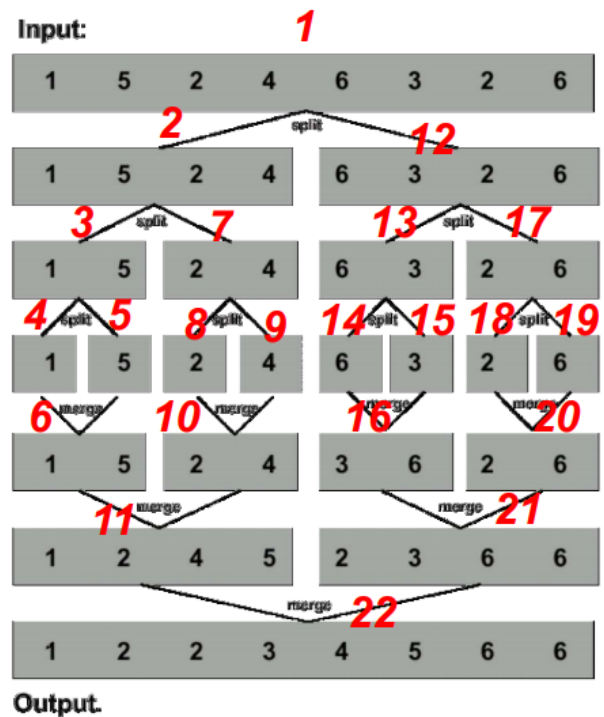
- To sort n numbers
 - if $n=1$, done!
 - recursively sort 2 lists of numbers $\lceil n/2 \rceil$ and $\lfloor n/2 \rfloor$ elements
 - merge 2 sorted lists in $\Theta(n)$ time
- Strategy
 - break problem into similar (smaller) subproblems
 - recursively solve subproblems
 - combine solutions to answer

Input:



Output:

1. Merge-Sort(A, 1, 8)
2. Merge-Sort(A, 1, 4)
3. Merge-Sort(A, 1, 2)
4. Merge-Sort(A, 1, 1)
5. Merge-Sort(A, 2, 2)
6. Merge(A, 1, 1, 2)
7. Merge-Sort(A, 3, 4)
8. Merge-Sort(A, 3, 3)
9. Merge-Sort(A, 4, 4)
10. Merge(A, 3, 3, 4)
11. Merge(A, 1, 2, 4)
12. Merge-Sort(A, 5, 8)
13. Merge-Sort(A, 5, 6)
14. Merge-Sort(A, 5, 5)
15. Merge-Sort(A, 6, 6)
16. Merge(A, 5, 5, 6)
17. Merge-Sort(A, 7, 8)
18. Merge-Sort(A, 7, 7)
19. Merge-Sort(A, 8, 8)
20. Merge(A, 7, 7, 8)
21. Merge(A, 5, 6, 8)
22. Merge(A, 1, 4, 8)



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Running Time of Merge Sort

- Write the recurrences
 - Solving the trivial problem: constant time, $\Theta(1)$
 - Dividing: constant time, $\Theta(1)$
 - Combining: linear time, $\Theta(n)$
 - Each division, we get two sub-problems with half size.
- Thus, we have $T(n)=$
 - $\Theta(1)$ if $n=1$
 - $2T(n/2) + \Theta(n)$ if $n>1$

Mini Quiz

- What is the running time of merge sort?
- By solving the recurrence
- $T(n)=$
 - $\Theta(1)$ if $n=1$
 - $2T(n/2) + \Theta(n)$ if $n>1$

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

$$\begin{aligned} T(n) &= 2T(n/2) + n && \text{substitute} \\ &= 2(2T(n/4) + n/2) + n && \text{expand} \\ &= 2^2T(n/4) + 2n && \text{substitute} \\ &= 2^2(2T(n/8) + n/4) + 2n && \text{expand} \\ &= 2^3T(n/8) + 3n && \text{observe the pattern} \\ T(n) &= 2^iT(n/2^i) + in \\ &= 2^{\lg n}T(n/n) + n \lg n = n + n \lg n \end{aligned}$$

$$\Theta(n \lg n)$$

Better than insertion sort $\Theta(n^2)$

- Can we say merge sort is better than insertion sort?
- Yes, run time $\Theta(n \lg n)$ vs. $\Theta(n^2)$
- No, additional space $\Theta(n)$ vs. $\Theta(1)$
 - The merge step requires an additional array with size n
 -
- We will see another sorting algorithm, Heap Sort, that has run time $\Theta(n \lg n)$ and also $\Theta(1)$ additional space.