LECTURE 07 – DECREASE & CONQUER

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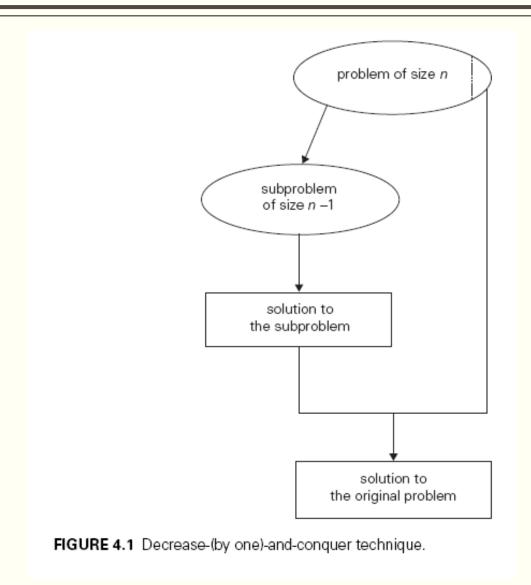
- Solve the instance of size k, using the same algorithm recursively.
- Use that solution to get the solution to the original problem.

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- Implement it recursively (top-down)
- Implement it iteratively (bottom-up)

Decrease & Conquer by one



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FIGURE 4.1 Decrease-(by one)-and-conquer technique.

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What about running time function complexity?

- How does the decrease-and-conquer algorithm differ from the Brute-force algorithm?
 - 1. the brute-force algorithm is more efficient
 - 2. the decrease-and conquer algorithm is more efficient
 - 3. the two algorithms are identical
 - 4. the two algorithms have the same asymptotic efficiency, but decrease and conquer has a better constant.

FIGURE 4.2 Decrease-(by half)-and-conquer technique.

$$a^{n} = \begin{cases} (a^{n/2})^{2} & \text{if } n \text{ is even and positive,} \\ (a^{(n-1)/2})^{2} \cdot a & \text{if } n \text{ is odd,} \\ 1 & \text{if } n = 0. \end{cases}$$

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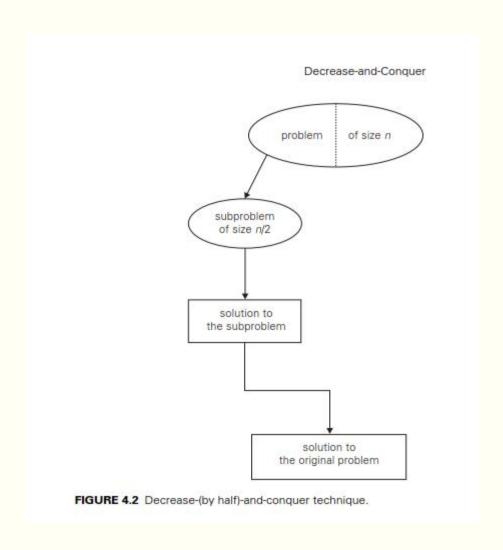


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$$\underbrace{A[0]\dots A[m-1]}_{\text{search here if}} A[m] \underbrace{A[m+1]\dots A[n-1]}_{\text{search here if}}.$$

- Search for an element K in an already sorted array A
- Compare K with array's middle element A[m]
- If they match stop. Otherwise do the same recursively for the first half if k < A[m], otherwise look at second half of the array

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```
ALGORITHM BinarySearch(A[0..n-1], K)

//Implements nonrecursive binary search

//Input: An array A[0..n-1] sorted in ascending order and

// a search key K

//Output: An index of the array's element that is equal to K

// or -1 if there is no such element

l \leftarrow 0; r \leftarrow n-1

while l \le r do

m \leftarrow \lfloor (l+r)/2 \rfloor

if K = A[m] return m

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- What is the basic operation?
- What about worst case complexity? C(n) = C(n/2) + 1, n > 1 C(1) = 1

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- All we have to do is put A[n-1] in the correct position.
 - Scan sorted subarray from right to left until you find a smaller element than A[n-1]
- Though insertion is based on a recursive idea its easier to implement bottom-up, iteratively.

89	45	68	90	29	34	17
45	89 1	68	90	29	34	17
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17	29	34	45	68	89	90

FIGURE 4.4 Example of sorting with insertion sort. A vertical bar separates the sorted part of the array from the remaining elements; the element being inserted is in bold.

```
ALGORITHM InsertionSort(A[0..n-1])
    //Sorts a given array by insertion sort
    //Input: An array A[0..n-1] of n orderable elements
    //Output: Array A[0..n-1] sorted in nondecreasing order
    for i \leftarrow 1 to n-1 do
         v \leftarrow A[i]
         j \leftarrow i - 1
         while j \ge 0 and A[j] > v do
             A[j+1] \leftarrow A[j]
             j \leftarrow j - 1
         A[j+1] \leftarrow v
```

Insertion Sort

Run animation: https://visualgo.net/bn/sorting

$$C(n) = \sum_{i=1}^{n-1} \sum_{j=0}^{i-1} 1 = \sum_{i=1}^{n-1} i = \frac{(n-1)n}{2} \in \Theta(n^2)$$

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almost-sorted files do arise in a variety of applications, and insertion sort preserves its excellent performance on such inputs.

- $C_{avg}(n) \approx \frac{n^2}{4} \in \Theta(n^2)$, average case
- In the worst case Insertion Sort makes same number of comparisons as Selection Sort

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- Is Insertion sort stable?
 - Yes, it is stable

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- How is gcd(96, 36) evaluated?
- \blacksquare gcd(96, 36) = gcd(36, 24) = gcd(24, 12) = gcd(12, 0) = 12