Divide and Conquer

Lecture 13

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https://www.khanacademy.org/computing/computerscience/algorithms/sorting-algorithms/a/sorting

Warmup: Recursive array max

- Design a recursive algorithm called *findArrayMax* that returns the maximum value in an array
- Formally:

Input: array A of length n >= 1
Output: max value of A

Examples: Input: A = {4, 13, 21, 5, 2}) Output: 21
Input: A = {-1, -3, -8, -5, -12} Output: -1
Input: A = {5} Output: 5

Recursive array max: stop condition

Algorithm findArrayMax(A): input: a NONEMPTY array, A output: A's maximum element Stop condition?



return 0

В

С

Α

if A.length == 1

if A.length == 1 return 1

if A.length == 1 return A[0]

• A

- **B**
- C
- None of the above
- More than one is correct

Recursive array max: recursive step

```
if
       A.length == 1:
       return A[0]
    *
   if A[0] < A[1]:
       A = A - A[0]
       return findArrayMax(A)
Α
   else:
*
       A = A - A[1]
       return findArrayMax(A)
   if A[0] < A[1]:
       A = A - A[1]
B
       return findArrayMax(A)
   else:
*
       A = A - A[0]
       return findArrayMax(A)
```

Algorithm findArrayMax(A):



• B

• A

None of the above

Recursive array max: solution

```
Algorithm findArrayMax(A):
   input: a NONEMPTY array, A
       output: A's maximum element
   if A.length == 1:
       return A[0]
   if A[0]<A[1]:
       A = A - A[0]
       return findArrayMax(A)
   else:
       \mathbf{A} = \mathbf{A} - \mathbf{A}[\mathbf{1}]
       return findArrayMax(A)
```

Sorting Problem

Input:Sequence A of n elementsOutput:Permutation A' of elements in Asuch that all elements of A'are in non-decreasing order.

Sorting Problem



https://www.cs.usfca.edu/~galles/visualization/ComparisonSort.html

https://www.toptal.com/developers/sorting-algorithms

Why Sorting?

- Sorting data is an important step of many efficient algorithms
- Sorted data allows for more efficient queries (binary search)

We will use Divide-and-conquer technique

- 1. Break into *non-overlapping* subproblems *of the same type*
- 2. Solve subproblems
- 3. Combine results

Divide: break apart



Conquer: solve subproblems



Combine





Idea: merge sort

725371316split the array into two halves

Idea: merge sort



Idea: merge sort



merge the sorted halves into one array

1 2 3 5 6 7 7 13

Algorithm MergeSort (array A[1...n])

- if *n* = 1: return *A* # already sorted
- $m \leftarrow \lfloor n/2 \rfloor$
- $B \leftarrow MergeSort(A[1 ... m])$
- $C \leftarrow \text{MergeSort}(A[m + 1 \dots n])$
- $A' \leftarrow merge(B, C)$

return A'

Merging Two Sorted Arrays

Algorithm Merge(*B*[1...*p*], *C* [1...*q*])

B and C are sorted $D \leftarrow empty array of size p + q$ while *B* and *C* are both non-empty: $b \leftarrow$ the first element of B $c \leftarrow$ the first element of C if b < c: move b from B to the end of D else: move c from C to the end of D move what remains of B or C to the end of D

return D

















k









k



k







Merge sort: running time

Subproblem size at each level





The height of the tree is...



The height of the tree is *log n*

Work at each level: all the work during merge



Work at each level: **O(n)**



Total: $O(n)^* \log n = O(n \log n)$

Merge Sort: running time

The running time of MergeSort(A[1 . . . n]) is $O(n \log n)$.

We can prove that this running time is *optimal* if we consider **sorting based on comparing pairs of numbers**

We can **not** do (asymptotically) faster. **Can we do better in practice?**

Idea: Quicksort

Divide array A into 2 subarrays

Recursively fully sort each subarray



divide

□ Combine the sorted subarrays by a simple concatenation

combine

Quicksort



Select an element called *pivot*

- Divide elements into 2 groups L (less or equal), and G (greater than pivot)
- 2. Conquer: recursively sort *L* and *G*
- 3. Combine: concatenate $L \rightarrow E \rightarrow R$

Example: quick sort

Example: quick sort

Rearrange elements with respect to x = A[0]



Example: quick sort

6 is in its final position 1 4 2 3 4 6 6 9 7 8 9

sort the two parts recursively

QuickSort(A, l, r) if $\ell \geq r$: m is the final element A[ℓ] return $m \leftarrow \text{Partition}(A, \ell, r)$ # A[m] is in the final position QuickSort(A, l, m - 1) QuickSort(A, m + 1, r)

 \Box the **pivot** is $x = A[\ell]$ \Box loop *i* from ℓ +1 to *r* maintaining the following invariant: $\Box A[k] \leq x$ for all $\ell + 1 \leq k \leq j$ $\Box A[k] > x$ for all $j + 1 \le k \le i$ 9 2 3 9 8 4 7 6 1













□ the pivot is x = A[ℓ]
□ move *i* from ℓ+1 to *r* maintaining the following invariant:
□ A[k] ≤ x for all ℓ+1 ≤ k ≤ j
□ A[k] > x for all j+1 ≤ k ≤ i
□ in the end, move A[ℓ] to its final place j



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□ A[k] > x for all j+1 ≤ k ≤ i
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Algorithm Partition(A, l, r)

```
x \leftarrow A[\ell]  # pivot
j ← ł
for i from \ell + 1 to r:
   if A[i] \leq x:
      i \leftarrow i + 1
      swap A[i] and A[i]
swap A[l] and A[j]
return j
# A[\ell + 1 \dots j] \le x, A[j + 1 \dots i] > x
```

Running time of Quick Sort

If we happen to choose the pivot *x* in such a way that after the partitioning the array *A* is split into even halves:

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

This is the same as in Merge sort, only here *n* comes from *partitioning*, and in merge sort *n* comes from combine (*merge*)

The running time of Quicksort is $O(n \log n)$

Quick Sort: summary

Simple
 Comparison-based
 Very fast in practice

Which choice of pivot would yield an **optimal** partitioning of *A*?

- A. 7
- B. 6
- C. 5
- D. 1
- E. None of the above



Which choice of pivot would yield the **worst** partitioning of *A*?

- A. 7
- B. 6
- C. 5
- D. 1
- E. None of the above



Unlucky choice of pivot

If we choose a pivot in such a way that **all values are greater than it**, then in each recursive step we decrement a size of the problem only by 1:



T(n) = O(n) + T(n-1)

Quick Sort: worst case complexity

Input size at each level Work at each level T(n) = n + T(n-1)



Total: $n^*n = O(n^2)$

Quick Sort: worst case complexity

Input size at each level Work at each level T(n) = n + T(n-1)



Total: $n^*n = O(n^2)$

Pathological case

$$T(n) = \mathcal{O}(n^2)$$

It requires $O(n^2)$ time to process the already sorted array which seems very inefficient since the array is already sorted!

Choosing random pivot

We can show that if we choose x randomly there is at least 50% chance that a good pivot will be chosen!

We can prove this using the expectation and the probabilities of random events

☐ If we choose all pivots at random, then half the times we do decrease the input sizes by a factor

This implies that the height of the recursive tree will be (2 log n) and the running time becomes O(n log n)

RandomizedQuickSort(A, l, r)

- if $\ell \geq r$:
 - return
- $k \leftarrow \text{random number between } \ell \text{ and } r$ swap $A[\ell]$ and A[k] $m \leftarrow \text{Partition}(A, \ell, r)$ # A[m] is in the final position RandomizedQuickSort($A, \ell, m - 1$) RandomizedQuickSort(A, m + 1, r)

Randomized Quick sort: Summary

- Randomized Quick sort is a comparison-based algorithm based on random partitioning
- \Box Expected running time: $O(n \log n)$
- □ Still $O(n^2)$ in the worst case
- □ Very fast in practice