CSE 241 Class 10

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1 Computational Model of Sorting

The world is full of sorting algorithms.

- QUICKSORT: worst-case $\Theta(n^2)$, average $\Theta(n \log n)$
- Mergesort: worst-case $\Theta(n \log n)$
- Heapsort: worst-case $\Theta(n \log n)$
- Insertionsort: worst-case $\Theta(n^2)$

What do all these sorting algorithms have in common?

- They are all comparison sorts
- all based on comparing array elts to each other
- can be implemented given only this operation: Greater(A,i,j) {return A[i] > A[j]}
- (in particular, don't care about actual value in each array cell!)

2 How Fast Can Comparison Sort Run?

- worst-case time of every sort I've mentioned is $\Omega(n \log n)$
- want to show that no asymptotically faster comparison sort exists
- argument must cover all such sorts, including ones I've never imagined
- only info allowed:
 - 1. input array can be in arbitrary order
 - 2. output array must be sorted
 - 3. only permitted operations are moving elements and testing with Greater

3 Decision Trees

I don't know many lower bound techniques, but all the ones I do know are closely tied to the idea of a *decision tree*. (**not** a recursion tree)

- **decision tree**: graphically represents all possible computations by some algorithm A on inputs of size n
- example for some algorithm A sorting three elements:

- leaves of tree = possible outputs
- internal nodes of tree = constant-time compute operations
- edges of tree = outcomes for each operation

Every pair (A, n) has its own specific tree.

4 A Lower Bound Argument

Idea: lower bound related to size of decision tree

Intuition:

- A certain problem can produce many possible outputs.
- Our operations are of limited power can only do a little work to choose among possible outputs at each node.

• May need to do many operations to pick an output in worst case.

How does this map onto the tree?

- tree starts with single root node
- every internal node of tree can split w ways (e.g. two for >)
- suppose tree has at least t leaves (possible outputs)
- How tall must the tree be?
- \bullet Every level increases the number of nodes by at most a factor of w
- \bullet To get minimum tree height h, must solve

$$w^h > t$$
.

- Conclude that $h \ge \log_w t$
- Therefore, to produce output corresponding to deepest leaf, need $\Omega(\log_w t)$ computations (one per node on path from root down to deepest leaf).

5 Example Argument for Sorting

Step 1: how many leaves in any sorting algo's decision tree?

- output is one particular arrangement of elements
- input is any possible arrangement
- hence, must be able to compute every possible permutation of input (else we could not correctly sort some input)
- Given an array of size n, how many ways can we permute its elements?
- (From 240): n! (n factorial)
- \bullet Conclude: tree has at least n! leaves

Step 2: how much can tree grow at each level?

- only allowed operations are comparisons
- each comparison has two outcomes: greater / not greater
- hence, each node can split in two at each level
- conclude # nodes can at most double at each level

Step 3: Conclude that tree height is at least $\log(n!)$

• How big is $\log(n!)$ anyway?

- Want *lower* bound (not obvious)
- From text: use **Stirling's approximation**:

$$n! > \left(\frac{n}{e}\right)^n$$

• Therefore, we have

$$\log n! > \log \left(\frac{n}{e}\right)^n$$

$$= n(\log n - \log e)$$

$$= n \log n - cn$$

$$> c' n \log n.$$

- Conclude that tree height, and hence worst-case # of comparisons, is $\Omega(n \log n)$
- Time to sort is at least number of comparisons (could do other work, e.g. swaps)
- Hence, any sort that does only comparisons takes worst-case time $\Omega(n \log n)$ QED

6 Implications of Lower Bound

- $n \log n$ comparison sorts like MergeSort are asymptotically optimal in the worst case
- any asymptotically faster sort must use operations other than comparison (we'll see some in a little while)
- not covered: how much can we improve constant factor?
- not covered: algorithms that can fail to sort some input arrangements (e.g. Monte Carlo methods)

7 More Lower Bounds

What about searching a sorted array? (Note: we can't say "What about BSEARCH?" because lower bounds apply to *problems*, not particular algorithms!)

- General problem: given a sorted array of length n, find the location of a query value x in the array (or fail)
- suppose algorithm can only compare any two array elements, or any element with x (using = or >)
- how many possible outputs?
- x can be anywhere in array or nowhere: n+1 outputs!
- allow each operation to have two outcomes: \leq , > or =, \neq
- So, we have w=2, t=n+1. Tree must have height at least $\log_w t = \log_2(n+1)$.

• Conclude that worst-case time to find an element in an array is $\Omega(\log(n+1)) = \Omega(\log n)$

What about **closest pair**?

- \bullet General problem: given a collection of n points in the plane, find the closest pair
- suppose algorithm can only compute and compare any two interpoint distances (i.e. can test $d(p_i, p_j) < d(p_k, p_l)$?)
- how many possible outputs?
- any pair could be closest: n(n-1)/2 possible pairs
- each comparison can have two outcomes
- So, we have w = 2, t = n(n-1)/2. Tree must have height at least $\log_w t = \log_2 n(n-1)/2$.
- Conclude that worst-case time to find closest pair is $\Omega(\log n(n-1)/2) = \Omega(\log n)$

Whoa. Anyone know of any $\log n$ -time closest pair algorithms? How about linear time?

- we have a **gap** between fastest known c.p. algorithm $(O(n \log n))$ and lower bound $(\Omega(\log n))$
- true fastest algorithm might be anywhere in between
- (if we work harder, can actually show that $n \log n$ is lower bound too)