15-451/651 Algorithm Design & Analysis Fall 2022, Recitation #1

Objectives

- Practice writing recurrence relations to analyze the runtime of algorithms
- Understand the concept of a lower bound for the running time of an algorithm
- Understand the different proof techniques that can be used to establish lower bounds
- Practice identifying flawed lower bounds proofs and writing correct ones

Recitation Problems

- 1. Quartile of quartiles Wurzelbrunft is taking 451, but he hates medians, and so he comes up with a new algorithm for deterministic selection that pivots on the first quartile of the medians (the element 25% of the way through in order), rather than the median of the medians:
 - 1. Group the array into n/5 groups of size 5 and find the median of each group. (For simplicity, we will ignore integrality issues.)
 - 2. Recursively, find the true quartile of the medians. Call this p.
 - 3. Use p as a pivot to split the array into subarrays LESS and GREATER.
 - 4. Recurse on the appropriate piece.

He tasks you, his group member, with analyzing his new algorithm before the oral.

(a) Write the recurrence relation for Wurzelbrunft's algorithm and compute its time complexity.

	(b) Can you improve the time complexity of Wurzelbrunft's algorithm by changing only the elements used in the recursive call?
2.	(A recipe book for lower bounds) In lecture, we established a set of tools for proving deterministic lower bounds in the comparison model. Briefly describe some of them and how they work.

- 3. (Sorting few distinct numbers) Suppose you have a list of n numbers a_1, a_2, \ldots, a_n , but there are at most D distinct numbers in this list. You would like to design a comparison-based algorithm for sorting these n numbers, (i.e., the only operation allowed is to choose two input numbers a_i and a_j and check whether or not $a_i \leq a_j$, at a cost of one operation).
 - (a) Describe an algorithm that can sort a list containing at most D distinct numbers in $O(n \log D)$ comparisons.

- (b) Now we would like to prove a lower bound for the problem. Consider the following attempted proofs. Some of them are correct and some are incorrect. Identify which are which, and for the incorrect ones, explain why they are incorrect¹.
 - In a list of size n consisting of at most D distinct numbers, each number has at most D possibilities, so the number of possible input lists is D^n . Using the information-theoretic argument, since $\log(D^n) = n \log D$, we must have an $\Omega(n \log D)$ lower bound in the comparison model.

- Consider a set of possible inputs of size n with D distinct elements constructed as follows. Create n/D contiguous chunks of size D. Each chunk contains the numbers 1...D in a random order. Since each chunk is size D and is in a random order, there are D! possible orders for each chunk. Since there are n/D chunks, the total number of inputs in the set is $(D!)^{\frac{n}{D}}$. Now observe that for each of these input sequences, no two of them are ever sorted by the same permutation. This is because each permutation selects exactly one item from each chunk for each element, and the elements of the chunks are unique. Therefore, we require at least $(D!)^{\frac{n}{D}}$ different possible permutations to sort them all.

¹You should be looking for mistakes in the *overall logic* of the proof. There are no intentional mistakes hidden in the math, so don't overthink the math parts. You can assume they are correct

Taking the log of this quantity, using an information-theoretic argument, we have a lower bound of

$$\log\left((D!)^{\frac{n}{D}}\right) = \frac{n}{D}\log(D!) = \Omega\left(\frac{n}{D}(D\log(D))\right) = \Omega(n\log D),$$

where we have used the fact from lecture that $\log D! = \Theta(D \log D)$.

- Consider the following set of possible sorted outputs to this problem:

$$\underbrace{1,1,\ldots,1}_{x_1},\underbrace{2,2,\ldots,2}_{x_2},\ldots,\underbrace{D,D,\ldots,D}_{x_D}$$

where $x_1, x_2, \ldots, x_D \ge 0$ and $x_1 + x_2 + \cdots + x_D = n$. By the "stars and bars" combinatorial argument, we can say that the number of such outputs is

$$\binom{n+D-1}{D-1}$$

and since no possible input can legally produce multiple of these outputs, any algorithm must correctly distinguish the 1 exact valid output among them. Therefore, by an information-theoretic argument we get a lower bound of

$$\log \binom{n+D-1}{D-1} \ge \log \left(\frac{n+D-1}{D-1}\right)^{D-1} \ge \Omega \left(D \log \left(\frac{n}{D}\right)\right)$$

- Consider the class of input sequences in which each of the D elements occurs the same number of times, i.e., there are n/D copies of each of the D elements. Now note that each of our n elements can go in any of at most n locations in the array. This will create $\left(\frac{n}{D}!\right)^D$ copies of each array due to permuting the n/D identical elements of all D values. Since each non-duplicate value

among these needs a distinct output (there is only one valid sorted order), by an information-theoretic argument we get a lower bound of

$$\log \frac{n^n}{\left(\frac{n}{D}!\right)^D} \ge \frac{n^n}{\left(\left(\frac{n}{D}\right)^{\frac{n}{D}}\right)^D} = \frac{n^n}{\left(\frac{n}{D}\right)^n} = \Omega(n\log D)$$

- Consider the class of input sequences in which each of the D elements occurs the same number of times, i.e., there are n/D copies of each of the D elements. Recall that for a sequence of n distinct elements, there are n! distinct permutations. Note that for a sequence with n/D copies of each of the D elements, permuting any subsequence of equal elements results in an also correctly sorted sequence. Therefore for each distinct element, there are (n/D)! ways they could be permuted and still be sorted. Combining all of these possibilities over the D elements, for any input sequence, there are exactly $((n/D)!)^D$ permutations that all correctly sort it. These sets of permutations are all disjoint, and hence there are

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

different permutations required to sort each of the possible input sequences. Taking the log of this quantity and making an information-theoretic argument, we get a lower bound of

$$\log\left(\frac{n!}{\left(\frac{n}{D}!\right)^D}\right) \ge \log\left(\frac{D}{e}\right)^n = \Omega(n\log D).$$

Further review

- 1. (Short answer / multiple choice)
 - (i) What is the solution of $A(n) = 3A(n-1) + 3^n$ with A(1) = 1
 - (a) $\Theta(3^n)$
 - (b) $\Theta(n3^n)$
 - (c) $\Theta(3^{n^2})$
 - (d) $\Theta(9^n)$
 - (e) $\Theta(3^n \log n)$
 - (ii) For c > 1, consider the summation $S = c + c^2 + c^3 + ... + c^n$. Which of the following are true? (Multiple may be true)
 - (a) $S = \Omega(nc^n)$
 - (b) $S = \Omega(nc)$
 - (c) $S = O(c^n)$
 - (d) S = O(nc)
 - (e) $S = \Theta(c^n)$
 - (f) $S = \Theta(nc^n)$
 - (g) $S = \Theta(nc)$
 - (iii) Given a number a > 1 and n, the sum of the series $1 + a + a^2 + ... + a^n$ is
 - (a) a^{n+1}
 - (b) $(a^{n+1}+1)/(a-1)$
 - (c) $(a^{n+1}-1)/(a+1)$
 - (d) $(a^{n+1}-1)/(a-1)$
 - (iv) What is a correct bound on A(n), where

$$A(n) = \begin{cases} 8A(n/2) + n^2 & \text{for } n \ge 2, \\ 1 & \text{for } 0 \le n \le 1. \end{cases}$$

- (a) $A(n) = \Theta(n^2 \log n)$
- (b) $A(n) = \Theta(n^3)$
- (c) $A(n) = \Theta(n^3 \log n)$
- (d) $A(n) = \Theta(n^4)$
- (e) $A(n) = \Theta(n^4 \log n)$
- (v) What is a correct asymptotic bound on T(n), where

$$T(n) = \begin{cases} \left\lfloor \sqrt{T(n-1)} \right\rfloor + 1, \text{ for } n \ge 2, \\ 1 & \text{for } 0 \le n \le 1 \end{cases}$$

(a) $\Theta(1)$

- (b) $\Theta(\log n)$
- (c) $\Theta(\sqrt{n})$
- (d) $\Theta(n/\log n)$
- (e) $\Theta(n)$
- (vi) What is a correct bound for T(n), where

$$T(n) = \begin{cases} 4T(n/2) + \sqrt{n} & \text{for } n \ge 2, \\ 1 & \text{for } 0 \le n \le 1. \end{cases}$$

- (a) $T(n) = \Theta(\sqrt{n})$
- (b) $T(n) = \Theta(\sqrt{n} \log n)$
- (c) $T(n) = \Theta(n)$
- (d) $T(n) = \Theta(n \log n)$
- (e) $T(n) = \Theta(n^2)$
- (f) $T(n) = \Theta(n^2 \log n)$
- (vii) In the DeterministicSelect algorithm, the median of medians procedure
 - (a) recursively calls itself
 - (b) calls DeterministicSelect
 - (c) Both of the above
- (viii) What would be the worst-case running time of the Quicksort algorithm if we pick an arbitrary (fixed) element as the pivot?
 - (a) $\Theta(n)$
 - (b) $\Theta(n \log n)$
 - (c) $\Theta(n^2)$
 - (d) $\Theta(n^2 \log n)$
 - (ix) What would be the **expected** worst-case running time of the **Quicksort** algorithm if we pick a random element as the pivot?
 - (a) $\Theta(n)$
 - (b) $\Theta(n \log n)$
 - (c) $\Theta(n^2)$
 - (d) $\Theta(n^2 \log n)$
 - (x) What would be the worst-case running time of the Quicksort algorithm if we used DeterministicSelect to select the pivot element?
 - (a) $\Theta(n)$
 - (b) $\Theta(n \log n)$
 - (c) $\Theta(n^2)$
 - (d) $\Theta(n^2 \log n)$
 - (xi) Regarding lower bounds for problems in the comparison model, which of the following statements are true and which are false?

- (a) $\exists M$ distinct inputs \implies lower bound of $\lg M$
- (b) $\exists M$ distinct inputs that require distinct outputs \implies lower bound of $\lg M$
- (c) $\exists M$ distinct outputs, each required by some input \implies lower bound of $\lg M$
- (d) $\exists M$ distinct outputs, each possible for some input \implies lower bound of $\lg M$
- (xii) Out of the following choices, what is the largest lower bound on the number of comparisons required to find the second largest element in an array?
 - (a) $n + \lg n 2$
 - (b) n
 - (c) n-1
 - (d) 2n-3
 - (e) $n \lg n$
- 2. (Recurrences) Solve the following recurrences in Θ notation by the tree/brick method (assume all base cases have T(1) = 0):
 - (a) T(n) = 1 + 2T(n-1).
 - (b) T(n) = n + T(n/2).
 - (c) T(n) = n + 2T(n/2).
- 3. (More recurrences) Solve the following recurrences, assuming a base case of T(0) = 1, T(1) = 1
 - (a) $T(n) = 3T(n/2) + n^2$.
 - (b) $T(n) = 4T(n/2) + n^2$.
 - (c) $T(n) = 5T(n/2) + n^2$.
- 4. (Probability Facts) Let's follow the convention of using uppercase letters X, Y, Z to denote random variables (which we often abbreviate to r.v.s).

Independence: Random variables X, Y are *independent* if for any values a, b, we have

$$\Pr[X = a, Y = b] = \Pr[X = a] \cdot \Pr[Y = b].$$

(a) Consider flipping two fair coins. Let $X \in \{0,1\}$ be the outcome of the first coin (where we think of heads as 1 and tails as 0) and let $Y \in \{0,1\}$ be the outcome of the second coin. Let $Z = X + Y \mod 2$. Are X and Y independent? What about X and Z?

Linearity of expectation: Given any two r.v.s X, Y,

$$\mathbf{E}[X+Y] = \mathbf{E}[X] + \mathbf{E}[Y].$$

This is true even if X and Y are not independent!

(b) Suppose we take n books numbered 1 to n, and place them on a shelf in a (uniformly) random order. What is the expected number of books that end up in their correct location? ("correct" = location it would be in if they were sorted).

Markov's inequality: given a non-negative random variable X with mean/expectation $\mu = \mathbf{E}[X]$,

$$\Pr[X > c \cdot \mu] \le \frac{1}{c}.$$

- (c) Show that if X is allowed to take negative values then the above inequality is no longer true.
- 5. (Summations) Prove the following theorem: For x < 1

$$\sum_{i=0}^{\infty} x^i = \frac{1}{1-x}$$

- 6. (**Deciding the maximum**) In lecture we used an adversary argument to obtain a lower bound of n-1 comparisons for finding the maximum element of a list.
 - (a) Suppose we instead used an information-theoretic argument. What lower bound would we obtain for the problem? Is it useful?
 - (b) Suppose we instead use decision trees to analyze the problem
 - i. How many outputs does the root node of any decision tree for this problem contain?
 - ii. How many outputs does each internal node of any decision tree for this problem contain?
 - iii. Are the leaves of these decision trees unique?
 - iv. How many nodes do each of the possible decision tree have?
 - v. What lower bound is implied by the structure of these decision trees?
- 7. (Lower bounds via reductions) In 15-251, you learned how to use reductions to show that one problem is just "as hard" as another problem. 251 taught you how to show that a problem is loosely in the same complexity class as another (e.g., NP-hard). This technique is not only applicable to loose complexity classes, but can also be used to establish tight lower bounds!

Lets do an example. Recall the partitioning algorithm used as a subroutine of Quicksort, which, given an array A and some "pivot" element A[i] of the array, rearranges the elements of A so that all elements that are $\leq A[i]$ come before all elements that are > A[i]. What if we generalize this problem to pick k distinct pivot elements, $A[i_1], A[i_2], ..., A[i_k]$? (The pivots do not have to be given in sorted order). The problem then becomes to rearrange the array so that all elements that are at most as large as the smallest pivot come first, followed by those that are larger but at most as large as the second-smallest pivot and so on...

- (a) Give an algorithm that solves k-partitioning in $\Theta(n \log k)$ comparisons.
- (b) Show that there can not possibly exist an algorithm that performs k-partitioning in fewer than $\Theta(n \log k)$ comparisons. To do so, show that if you assume such an algorithm exists, then it must be possible to sort in fewer than $\Theta(n \log n)$ comparisons.

- 8. (Breaking Eggs) Say I choose a number between 1 and N. You want to guess it in as few questions as possible (each time you make an incorrect guess, I'll tell you if it is too high or too low). As we know, the strategy that minimizes the worst-case number of guesses is to do binary search: it takes $\lceil \log_2 N \rceil$ guesses.
 - But, what if you are only allowed **one** guess that is too high? Can you still solve the problem in o(N) guesses? [If you were not allowed **any** guesses that are too high, the only option you would have would be to guess 1,2,3,... in order].
 - (a) Show how to figure out how many centimeters high you can drop an egg without it breaking, using $2\sqrt{N}$ guesses, when you only have two eggs.
 - (b) Can you show an $\Omega(\sqrt{N})$ lower bound for any deterministic algorithm?