CS344: MidTerm Solution

November 5, 2017

1. (Rapid Answers)

(a) If f = O(g(n)) and $f = \Omega(g(n))$, then f(n) = g(n). True or False? Provide reasons.

False. f(n) = 10g(n) would satisfy both properties and yet $f(n) \neq g(n)$.

(b) Solve the recurrence $T(n) = 3T(n/4) + n \log n$ and T(1) = 1, using any method you prefer.

We have

$$T(n) = 3(3T(n/16) + (n/4)\log(n/4)) + n\log n < 9T(n/16) + (3n/4)\log n + n\log n$$

Continuing this way with each iteration, we will have

$$n \log n + (3/4)n \log n + (3/4)^2 n \log n + \cdots$$

so in total $T(n) = O(n \log n)$.

(c) We implemented dictionary using uniform hashing, collision resolution by chaining, and hash functions that can be calculated in O(1) time each. If we are happy with $O(\log n)$ time on average for search, what will be m, the number of hash buckets you will choose for storing n elements?

Average search time is O(1+n/m), so if we want this to $O(\log n)$, we can choose $m = O(n/\log n)$.

(d) Given a polynomial of 10 variables, degree 20 and suppose we choose randomly values for the variables from a set S of size 1000, what is an upper bound on the probability that we get a false outcome (that the polynomial is NOT identically 0, but these values make the polynomial turn out to be 0).

We showed that the probability is at most d/|S| where d is the degree. So, the probability is at most 20/1000, independent of number of variables.

(e) Muthu has a binary string s of length n and you have a binary string t of length n. What should Muthu communicate to you to check if s = t, and how many bits are needed? Assume we are good with a small probability of error.

We should use Karp-Rabin fingerprint f(x) for string x. In this case, Muthu should communicate q (the random prime), and f(s). Both of them are $O(\log n)$ in size.

- 2. (Dynamic Programming) A tree T consists of a root node and zero or more children, each of which is a tree. Its vertex cover is a subset S of nodes such that for each (node u,child v) pair in T, at least one of u or v is in S. The problem is to find the minimum sized vertex cover. Solve it using dynamic programming.
 - \bullet For each node v in T, what function will you compute?

For every v in T, we compute two functions:

- $-f_1(v)$, the smallest vertex cover for tree rooted at v when v IS in the vertex cover, and
- $-f_2(v)$, the smallest vertex cover for tree rooted at v when v is NOT in the vertex cover.
- What is the recursive definition for that function. Consider the two cases when v is part of the minimum vertex cover and when v is NOT a part of it.

We have

$$f_1(v) = \left(\sum_{w \text{ children of } v} \min\{f_1(w), f_2(w)\}\right) + 1$$

and

$$f_2(v) = \sum_{w \text{ children of } v} f_1(w)$$

• Complete the dynamic programming solution by stating in what order you evaluate this function and how much time, space does the whole algorithm take?

We calculate this bottom-up in T so $f_1(w)$ and $f_2(w)$ will be available for all w when we evaluate for v. Like we discussed in the maximum independent case, time and space is O(n), the number of nodes in T.

- 3. (Divide and Conquer) 30 points Suppose you are given an array of positive integers A[1, n]. Each A[i] is an element.
 - For $1 \le i \le j \le n$, a sub-array A[i, j] of A is $A[i, i+1, \ldots, j]$
 - A prefix of A is a sub-array A[1, 2, ..., i]
 - A suffix of A is a sub-array A[i, i+1, ..., n].
 - The sum of a sub-array A[i,j] is the sum of all its elements $\sum_{k=i}^{k=j} A[k]$.

Design a divide and conquer algorithm that counts the number of sub-arrays with even sum.

• A naive approach is to calculate the sum of all possible sub-arrays (A[i,j] for all i,j) and then returning the number of these of even sum. What is the running time of this algorithm?

There are $O(n^2)$ such sub-arrays, and for each, it takes O(n) time to calculate the sum, so the running time is $O(n^3)$.

• Show how to compute the sums of all prefixes of A in O(n) time. Given these values, show that the sum of any sub-array A[i,j] can be computed in O(1) time. How long does it take now to go through all possible sub-arrays and return the ones of even sum like in the case above?

Let sum of the *i*th prefix $A[1,2,\ldots,i]$ be $P[i]=\sum_{k=1}^{k=i}A[k]$. We have

$$P[i+1] = P[i] + A[i+1]$$

so all P's can be computed in O(n) time.

The sum of any sub-array A[i, j] is

$$P[j] - P[i-1]$$

so it can be computed in O(1) time from P's.

The running time of the algorithm above is $O(n^2)$ because the sum of each of the $O(n^2)$ sub-arrays can be computed in O(1) time as shown.

• We will use divide-and-conquer approach. Suppose that you divide A into two disjoint sub-arrays L and R of length n/2, and determine the number of sub-arrays of even sum for both of these two sub-arrays recursively. Why is this information not sufficient to compute the number of even sum sub-arrays of A? Be concise.

There are sub-arrays of A like $A[i, i+1, \ldots, n/2, n/2+1, \ldots, j]$ where for example, the sum of the sum-array

$$A[i, i + 1, \dots, n/2]$$

and the sum of the sub-array

$$A[n/2+1,\ldots,j]$$

can both be odd, and the subarray

$$A[i, i+1, \ldots, n/2, n/2+1, \ldots, j]$$

has even sum, which means it is not sufficient to only determine the sub-arrays of even sum in L or R.

• Suppose now that your recursive algorithm computes and returns extra information beyond the number of even sum sub-arrays: what constant size extra information can be returned so that from the recursive solutions for L and R one can recover in O(1) time (1) the number of even sum sub-arrays of A and (2) also the extra information to return when you pop out of the recursion for A? (Hint: Think about the number of prefixes and suffixes with even and odd sum in R and L.)

For the subproblem on L, return

- $-L_1$, the number of sub-arrays with even sum,
- $-L_2$, the number of suffixes with odd sum,
- $-L_3$, the number of suffixes with even sum.
- $-L_4$, the number of prefixes with odd sum,
- $-L_5$, the number of prefixes with even sum

Do the same for R. Then the number of even sum sub-arrays is

$$L_1 + L_2 R_4 + L_3 R_5 + R_1$$

(be careful about corner cases if any). Now we can devise similar formula for all other sums we need recursively.

• Write a recurrence for your divide-and-conquer algorithm above. Solve the recurrence using your favorite method. What is the running time of the algorithm?

The discussion above gives the recurrence for a subarray in terms of the 5 terms we calculate and return for each of L and R. These recurrences can be calculated in O(1) time once the L and R problems have been solved. So, the recurrence is

$$T(n) = 2T(n/2) + O(1)$$

which solves to O(n).