CSCI 570 - Spring 2018 - HW 2 Solution

1 Graded Problems

1. Solve Kleinberg and Tardos, Chapter 2, Exercise 3.

Answer: We know from the text that polynomials grow slower than exponentials. Thus, we will consider f_1, f_2, f_3, f_6 as a group, and then put f_4 and f_5 after them. For polynomials f_i and f_j , we know that f_i and f_j can be ordered by comparing the highest exponent on any term in f_i to the highest exponent on any term in f_j . Thus, we can put f_2 before f_3 before f_1 . Now, since f_6 grows faster than n^2 , and we know that logarithms grow slower than polynomials, so f_6 grows slower than n^c for any c > 2. Thus we can insert f_6 between f_3 and f_1 . Finally come f_4 and f_5 , we know that exponential can be ordered by their bases, so put f_4 before f_5 . Therefore, in ascending order of growth, the list is f_2 , f_3 , f_6 , f_1 , f_4 , f_5 .

2. Solve Kleinberg and Tardos, Chapter 2, Exercise 4.

Answer: In ascending order of growth, the list is $g_1(n)$, $g_3(n)$, $g_4(n)$, $g_5(n)$, $g_2(n)$, $g_7(n)$, $g_6(n)$.

3. Solve Kleinberg and Tardos, Chapter 2, Exercise 5.

Answer: Assume that functions f(n) and g(n) take nonnegative values.

(a) False. Consider for example $f(n) = 2, \forall n \text{ and } g(n) = 1, \forall n$.

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Clearly, f(n) = \mathcal{O}(g(n)). Observe that \log_2(f(n)) = 1, \forall n and \log_2(g(n)) = 0, \forall n. Hence \log_2(f(n)) \neq \mathcal{O}(\log_2(g(n))).
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Note: If we further add the constraint that $\exists N$ such that $g(n) \geq 2, \forall n > N$, then the statement becomes true.

- (b) False. Consider for example f(n) = 2n and g(n) = n. Clearly 4^n is not $\mathcal{O}(2^n)$.
- (c) True. Since $f(n) = \mathcal{O}(g(n))$, there exists positive constants c and n_0 such that $f(n) \leq cg(n), \forall n \geq n_0$. This implies $f(n)^2 \leq c^2g(n)^2, \forall n \geq n_0$, which in turn implies that $f(n)^2 = \mathcal{O}(g(n)^2)$.

4. Which of the following statements are **true**?

- (a) If f, g, and h are positive increasing functions with f in O(h) and g in $\Omega(h)$, then the function f+g must be in $\Theta(h)$.
- (b) Given a problem with input of size n, a solution with O(n) time complexity always costs less in computing time than a solution with $O(n^2)$ time complexity.
- (c) $F(n) = 4n + \sqrt{3n}$ is both O(n) and $\Theta(n)$.
- (d) For a search starting at node s in graph G, the DFS Tree is never as the same as the BFS tree.
- (e) BFS can be used to find the shortest path between any two nodes in a non-weighted graph.

Answer:

- (a) False. Consider for instance f(n) = n and $g(n) = n^2$. Clearly $n + n^2$ is not in $\Theta(n)$.
- (b) False. The O(n) notation merely gives an upper bound on the running time. To claim one is faster than other, you need an upper bound for the former and a lower bound (Ω) for the latter.
- (c) True. The dominant term is 4n, which is obviously both O(n) and $\Omega(n)$
- (d) False. For instance, G being a tree is a counterexample.
- (e) True.

5. Solve Kleinberg and Tardos, Chapter 3, Exercise 2.

Answer: Without loss of generality assume that G is connected. Otherwise, we can compute the connected components in $\mathcal{O}(m+n)$ time and deploy the below algorithm on each component.

Starting from an arbitrary vertex s, run BFS and obtain a BFS tree (call it T). If G = T, then G is a tree and has no cycles. Otherwise, G has a cycle and hence there exists an edge e = (u, v) such that e is in G but not in T. Find the least common ancestor of u and v in the tree. Call the least common ancestor w. There exist a unique path (call P_1) in T from u to w (and likewise a unique path P_2 in T from v to w). These paths can be constructed in $\mathcal{O}(m)$ time by starting from u (respectively from v) and going up the tree until w is reached. Output the cycle e concatenated with P_2 concatenated with P_1 . Here P_1 denotes P_1 in the reverse order.

2 Practice Problems

- 1. Reading Assignment: Kleinberg and Tardos, Chapter 2 and 3.
- 2. Solve Kleinberg and Tardos, Chapter 2, Exercise 6.

Answer:

- (a) The outer loop of the given algorithm runs for exactly n iterations, and the inner loop of the algorithm runs for at most n iterations. Therefore, the line of code that adds up array entries A[i] through A[j] (for various is and js) is executed at most n^2 times. Adding up any array entries A[i] through A[j] takes O(j-i+1) operations, which is O(n). Store the results in B[i,j] requires only constant time. Therefore, the running time of the entire algorithm is at most $n^2 \cdot O(n)$, and so the algorithm runs in $O(n^3)$.
- (b) Consider the times during the execution of the algorithm when $i \leq \frac{n}{4}$ and $j \geq \frac{3n}{4}$. In this case, $j-i+1 \geq \frac{3n}{4}-\frac{n}{4}+1 > \frac{n}{2}$. Therefore, adding up the array entries A[i] through A[j] takes at least $\frac{n}{2}$ operations. How many times during the execution of the algorithm do we encounter such cases $(i < \frac{n}{4} \text{ and } j > \frac{3n}{4})$? There are $\left(\frac{n}{4}\right)^2$ pairs of (i,j) with $i < \frac{n}{4}$ and $j > \frac{3n}{4}$. The given algorithm enumerates over all of them, and as shown above, it must perform at least $\frac{n}{2}$ operations for each such pair. Therefore, the algorithm perform at least $\frac{n}{2} \cdot \left(\frac{n}{4}\right)^2 = \frac{n^3}{32}$ operations. This is $\Omega(n^3)$.
- (c) Consider the following algorithm:

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\begin{array}{l} \textbf{for } i = 1, 2, \cdots, n \ \textbf{do} \\ \text{Set } B[i, i+1] \ \textbf{to } A[i] + A[i+1] \\ \textbf{end for} \\ \textbf{for } k = 2, 3, \cdots, n-1 \ \textbf{do} \\ \textbf{for } i = 1, 2, \cdots, n-k \ \textbf{do} \\ j = i+k \\ B[i,j] \ \textbf{to be } B[i,j-1] + A[j] \\ \textbf{end for} \\ \textbf{end for} \end{array}
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This algorithm works since the values B[i, j-1] were already computed in the previous iteration of the outer for loop, when k was j-1+i, since j-1-i < j-i. It first computes B[i, i+1] for all i by summing A[i] with A[i+1]. This requires O(n) operations. For each k, it then computes all B[i, j] for j-i=k by setting B[i, j] = B[i, j-1] + A[j]. For each k, this algorithm performs O(n) operations since there are at most n B[i, j]'s such that j - i = k. There are less than n values of k to iterate over, so this algorithm has running time $O(n^2)$.

3. Solve Kleinberg and Tardos, Chapter 3, Exercise 6.

Answer: Assume that G contains an edge e=(x,y) that does not belong to T. Since T is a DFS tree and (x,y) is an edge of G that is not an edge of T, one of x or y is ancestor of the other. On the other hand, since T is a BFS tree if x and y belong to layer L_i and L_j respectively, then i and j differ by at most 1. Notice that since one of x or y is an ancestor of the other, we have that $i \neq j$ and hence i and j differ by exactly 1. However, combining that one of x or y is ancestor of the other and that i and j differ by 1 implies that the edge (x,y) is in the tree T. It contradicts the assumption that e=(x,y) that does not belong to T. Thus G cannot contain any edges that do not belong to T.