## Example Solution: Divide And Conquer and Dynamic Programming

Feel free to work in groups of at most 4 for these - if you have a group of more than 4, please run it by me first. If you do work in a group, please include the names of those that you worked with. However: each student should submit a separate copy where the solutions have been written by yourself.

1. In this problem, we are given an array A of length n and want to count the number of inversions within the list. Recall that an inversion is a pair (i, j) such that i < j and A[i] > A[j]. Construct an algorithm that determines the number of inversions in  $O(n \log n)$  time.

Solution: We will use a DIVIDE AND CONQUER strategy to solve this problem. Given A of length n, we will create two arrays  $B = A \left[1 : \frac{n}{2}\right]$  and  $C = A \left[\frac{n}{2} : n\right]$ . We will recurse on B and C to determine the number of inversions in each, and then count the number of inversions that cross B and C and return the total sum of inversions. As part of our recursion, we will sort B and C to make it easier to count the inversions that cross them. To count the inversions, we loop through both B and C, comparing corresponding elements. Since both B and C are already sorted by means of our recursion, when we have an inversion, we know that all remaining elements in B must also form inversions with the current element in B. On the other hand, if the current element in B doesn't form an inversion with the current element in B, then no remaining elements in B would form an inversion with the current element of B either. The following pseudocode implements this idea:

```
def inversionCount(A):
      if len(a) == 1:
          return (A, 0)
      else:
4
          (B, I1) = inversionCount(A[1:n/2])
          (C, I2) = inversionCount(A[n/2:n])
          Initialize i = 0 and j = 0 as the start of B and C
          Initialize sortedA as an empty list
          Initialize I3 as 0
          while the ends of both B and C are not reached:
10
              if B[i] < C[j]:
                   Add B[i] to sortedA and increment i
12
              else if B[i] > C[j]:
13
                   Add C[j] to sortedA, increment j,
14
                       and increase I3 by len(B)-i
15
              else: # If they are equal
16
                   Add C[j] to sortedA and increment j
17
          Add whatever is left in B and C to sortedA
18
          return (sortedA, I1 + I2 + I3)
```

Let T(n) be the runtime of the above algorithm for an input of length n. Then  $T(n) = 2T\left(\frac{n}{2}\right) + \text{however much time it will take to merge and count the inversions in$ 

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the middle. Since we are only iterating through B and C once each at most, this gives us  $T(n) = 2T(\frac{n}{2}) + O(n)$ , which we know from class gives us  $T(n) = O(n \log n)$ .

2. In this problem, we are given a list of intervals  $I_1 = (s_1, e_1, v_1)$ , ...,  $I_n = (s_n, e_n, v_n)$  where  $s_i$ ,  $e_i$  and  $v_i$  are the start time, end time, and value of  $I_i$ . Construct an algorithm that determines the maximum value non-conflicting collection of intervals from the input.

**Solution:** Let OPT(j) denote the sum of the optimal collection of intervals from the sublist  $I_1, ..., I_j$ . Note that in the optimal solution, either  $I_j$  was included in the sum or it wasn't. If it was not included, then OPT(j) = OPT(j-1). Otherwise, OPT(j) = OPT(p(j)) where p(j) is the index of the latest job that ends before  $I_j$  starts, or 0 if no such job exists. Hence

$$Opt(j) = \max(Opt(j-1), v_j + Opt(p(j))). \tag{1}$$

To calculate Opt(n), we create an array A such that A[j] = Opt(j). We initialize A[0] = 0 (since 0 value can be obtained from 0 intervals), and then set  $A[j] = max(A[j-1], v_j + A[p(j)])$  for j = 1 to n (in that order). We can precalculate p in linear time (left as an exercise, obviously a student cannot leave this open but it's a good problem for you all, so I recommend thinking about it). Hence, since it takes only a constant amount of time to calculate A[j] if A[0], ..., A[j-1] have all already been calculated, we have that A can be computed in linear time. Since A[j] follows the recursive formula in Eq.1, we have that A[n] = Opt(n), giving us a linear time solution.