CS360 Assignment 4

15.1-2 Show by means of a counterexample that the following "greedy" strategy does not always determine an optimal way to cut rods. Define the density for a rod of length i to be p_i/i and cut off pieces of length i with the maximal density.

Consider the following pricing table for rods of length 4 with computed densities

length	1	2	3	4
p_i	1	20	33	36
p_i/i	1	10	11	9

The greedy strategy would say to initially select a length 3 piece (since it has the highest density) and then the remaining length 1 piece giving a total revenue for the greedy strategy of

$$r_a = p_3 + p_1 = 33 + 1 = 34$$

However, clearly an optimal solution would be to take two pieces of length 2 giving the optimal total revenue of

$$r' = p_2 + p_2 = 20 + 20 = 40$$

Note that even simply selecting the entire rod with no cuts would give a revenue of 36 which is better than the greedy strategy. This problem is very similar to the knapsack problem where simply selecting the most valuable item intially may preclude selecting several slightly less valuable items which would have produced a greater total amount, and demonstrates why simply employing a greedy strategy must be used with caution.

15-4 Printing neatly.

Step 1 Characterize optimal solution

Define a variable $e(i, j) = M - j + i - \sum_{k=i}^{j} l_k$ (where $l_i \leq M$ is the length of each word) as the number of spaces at the end of a line starting with word i and ending with word j (which may be negative if the total length of the line exceeds the allowable number of spaces M). Then define a cost for the line as

$$lc(i,j) = \begin{cases} \infty & \text{if } e(i,j) < 0\\ 0 & \text{if } j = n, e(i,j) \ge 0\\ e(i,j)^3 & \text{otherwise} \end{cases}$$

The first condition handles the case where the words don't fit by preventing them from being the minimum by setting the cost to infinity, the second condition is for the last line where we don't want the extra spaces at the end to count, and the third condition is the penalty defined for words that fit on lines other than the last.

Step 2 Define top-down recursive solution

We now note that the optimal solution for words 1..n where the last line starts at word i must contain an optimal solution for words 1..(i-1).

Hence we can define the recursion for the solution by defining a minimum cost for n words as c(n) given by the following formula

$$c(n) = \left\{ \begin{array}{ll} 0 & n=0 \\ \min_{1 \leq i \leq n} (c(i-1) + lc(i,n)) & n>0 \end{array} \right.$$

i.e. the minimum is 0 if we have no words, otherwise it is the minimum of the first i-1 words and the cost of the last line containing words i to n.

To show optimal substructure, assume optimality occurs when i = k such that c(n) = c(k-1) + lc(k,n) where c(n) is minimum. Now assume that c(k-1) is not optimal, therefore we can find another spacing for the first k-1 words such that c'(k-1) < c(k-1).

Then $c'(n) = c'(k-1) + lc(k,n) < c(k-1) + lc(k,n) = c(n) \Rightarrow c'(n) < c(n)$. However, this **contradicts** our assumption that c(n) was the minimum, and thus c(k-1) must also be an optimal minimum.

Step 3 Computing bottom-up solution

Replacing n by j and starting with the base case $n = 0 \Rightarrow c(0) = 0$ gives

$$c(j) = \begin{cases} 0 & j = 0\\ \min_{1 \le i \le j} (c(i-1) + lc(i,j)) & j > 0 \end{cases}$$

Which is then computed by looping from j = 1 to j = n storing the values of c(j) in an array.

Pseudocode for this implementation is

```
NEAT_PRINTING(1,n,M)
    Create e[n,n], lc[n,n], c[n], and p[n]
    // compute e[i,j]
    for i = 1 to n
3.
        e[i,i] = M - 1[i]
4.
5.
        lsum = l[i]
        for j = i+1 to n
6.
7.
             lsum = lsum + l[j]
             e[i,j] = M - j + i - lsum
8.
9.
     // compute lc[i,j]
10.
     for i = 1 to n
         for j = i to n
11.
              if e[i,j] < 0
12.
                  lc[i,j] = INF
13.
              else if j = n
14.
15.
                  lc[i,j] = 0
16.
              else
                  lc[i,j] = e[i,j]^3
17.
     // compute c(j)
18.
     c[0] = 0
19.
20.
     for j = 1 to n
21.
         c[j] = INF
         for i = 1 to j
22.
              if c[j] > c[i-1] + lc[i,j]
23.
                  c[j] = c[i-1] + lc[i,j]
24.
                  p[j] = i
25.
```

Note in the above algorithm we include an extra array p[] which keeps track of which word each line breaks at which can be used to print the formatted paragraph. Clearly the run time of this algorithm is dominated by the nested loops which run in $O(n^2)$, e.g. lines 10-17 have two nested O(n) loops. Again there are many simplifications that can be made to improve on this basic algorithm in terms of space usage.

16.1-4 Given a set of activities to schedule among a large number of lecture halls, we wish to schedule all the activities using as few lecture halls as possible. Give a greedy algorithm to solve the problem.

One solution is to simply use the activity selection algorithm to assign maximal sets of compatible activities to successive halls. In other words, starting with the set of all activities $S^{(1)}$, let $A^{(1)}$ be the maximum set of compatible activities in $S^{(1)}$ and schedule those in lecture hall 1. Then let $S^{(2)} = S^{(1)} - A^{(1)}$ (i.e the remaining activities) and find $A^{(2)}$ to schedule in lecture hall 2. Continue

by finding $A^{(i)}$ for $S^{(1)} = S^{(i-1)} - A^{(i-1)}$ until $S^{(i)} = 0$ (i.e. all activities are scheduled).

This gives an optimal solution since at any step if there would be an activity that could be scheduled in a previous hall, it would have been in the solution to that hall's subproblem. Alternatively, at each step the only remaining activities are ones that occur concurrently and hence need to be scheduled in separate halls.

Since each subproblem runs in O(n) (assuming presorting of the inputs) and there are at most n subproblems (all activities overlap), the algorithm runs in $O(n^2)$.

HOWEVER THERE IS A BETTER WAY! (A greedy, greedy algorithm)

Initially sort <u>all</u> times (both start and finish) in increasing order (with finish times first in the case of ties).

Create an increasing order priority queue of occupied halls and a stack of available halls. Initially push all halls onto the stack.

Going through the list of sorted times, whenever a start time occurs, pop a hall from the stack and put it in the queue using its finish time as the priority key. The activity can be marked with this hall if necessary to produce the final schedule.

Whenever a finish time occurs, dequeue the first element of the queue and push it back on the stack (i.e. the hall has become available). This ensures that halls that have already been used are reused first.

Keeping track of the queue size gives the maximum number of halls needed (which corresponds to the maximum number of simultaneous activities occurring at any one time).

Initial sorting of 2n times takes $O(n \lg n)$. The stack push/pop operations take O(1). The queue enqueue/dequeue operations take $O(\lg n)$. So for the 2n times we get a total running time of $O(n \lg n) + 2nO(1) + nO(\lg n) + nO(\lg n) = O(n \lg n)!$

B.4-1 Handshaking lemma.

If we consider each pair of professors that shake hands as a pair of vertices in a graph, then we can represent the handshake as an edge between the vertices. If we represent the graph as an adjacency list, then each handshake will appear as two edges. Hence the total number of edges in the adjacency lists will be 2|E|. But by definition, the degree of a vertex is the number of entries in its adjacency list (since these are all the vertices it is adjacent to) so the sum of the lengths of the adjacency list lengths, i.e. degrees, of all the vertices $\sum_{v \in V} \text{degree}(v) = 2|E|$. Therefore when the department chair adds up the total number of times all the professors shook hands, the result will be even.