CS 180: Homework 4

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Discussion 1B

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When their respective sport is not in season, UCLA's student-athletes are very involved in their community, helping people and spreading goodwill for the school. Unfortunately, NCAA regulations limit each student-athlete to at most one community service project per quarter, so the athletic department is not always able to help every deserving charity. For the upcoming quarter, we have S student-athletes who want to volunteer their time, and B buses to help get them between campus and the location of their volunteering. There are F projects under consideration; project i requires s_i student-athletes and b_i buses to accomplish, and will generate $g_i > 0$ units of goodwill for the university.

Use dynamic programming to produce an algorithm to determine which projects the athletic department should undertake to maximize goodwill generated. For full-credit, your algorithm should run in time O(SBF) but you don't have to prove its correctness or analyze the time complexity.

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S student-athletes = \{s_1, s_2, \dots, S\}

B buses = \{b_1, b_2, \dots, B\}

F projects = \{1, 2, \dots, F\}

G goodwill = \{g_1, g_2, \dots, G\}

G oal: maximize G
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Let O be an optimal solution calculated using OPT(project, student, bus):

If project n is not in O, we must solve the problem for projects $\{1, \ldots, n-1\}$ using OPT(n-1, s, b)If project n is in O, we must solve the problem for projects $\{1, \ldots, n-1\}$, but we know that $OPT(n, s_n, b_n) = g_n + OPT(n-1, S-s_n, B-b_n)$

MAX-GOODWILL-COMPUTE-OPT:

Else:

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    Set OPT(0, s, b) = 0 for all students s = {s<sub>1</sub>, s<sub>2</sub>,..., S} and all buses b = {b<sub>1</sub>, b<sub>2</sub>,..., B}.
    Set OPT(i, 0, 0) = 0 for all i = {1,...,n}.
    For i = {1,..., F}: O(F)
    For s = {s<sub>1</sub>,..., S}: O(S)
    For b = {b<sub>1</sub>,..., B}: O(B)
    If s<sub>i</sub> > S or b<sub>i</sub> > B:
    OPT(i, s, b) = OPT(i - 1, s, b)
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$$OPT(i, s, b) = \max \begin{cases} OPT(i - 1, s, b) \\ g_i + OPT(i - 1, s - s_i, b - b_i) \end{cases}$$

$$OPT(F, S, B).$$

3. Output OPT(F, S, B).

In order to find the subset that maximizes goodwill, we use the following algorithm: FIND-SUBSET(i, s, b):

- 1. If $s_i > s$ or $b_i > b$:
- RETURN FIND-SUBSET(i-1, s, b).
- 2. Else:

If
$$g_i + OPT(i-1, s-s_i, b-b_i) > OPT(i-1, s, b)$$
:
RETURN $\{i\} \cup$ FIND-SUBSET $(i-1, s-s_i, b-b_i)$.

Else:

RETURN FIND-SUBSET(i-1, s, b).

You have a knapsack of total weight capacity W and there are n items with weights w_1, \ldots, w_n respectively. Give an algorithm to compute the number of different subsets that you can pack safely into the knapsack. In other words, given integers w_1, \ldots, w_n, W as input, give an algorithm to compute the number of different subsets $S \subseteq [n]$ such that $\sum_{i \in S} w_i \leq W$. For full-credit, your algorithm should run in time O(nW) but you don't have to prove its correctness or analyze the time complexity.

Let M be the set of items to be put in the knapsack with total weight $\leq W$.

Let FEASIBLE(i, w) return the set of items in the knapsack with total weight $\leq W$.

If $w_n > W$:

$$FEASIBLE(n, w) = FEASIBLE(n - 1, w)$$

Else:

If n not in M:

$$FEASIBLE(n, w) = FEASIBLE(n - 1, w)$$

Else n in M:

$$FEASIBLE(n, w) = n \cup FEASIBLE(n - 1, w - w_n)$$

The algorithm to compute all possible subsets is therefore:

- 1. Set FEASIBLE(0, w) = 0 for all weights (w_1, \ldots, w_n)
- 2. Initialize stack S to be empty.
- 3. For i = 1, ..., n:

For
$$w = (w_1, ..., w_n)$$
:

Compute FEASIBLE(i, w) using the recurrence.

If FEASIBLE(i, w) returns a set that is not currently in S:

Push
$$FEASIBLE(i, w)$$
 onto S .

Else:

Do nothing.

4. RETURN the total number of elements in S.

Just like the knapsack algorithm, this algorithm runs in O(nW).

Given two strings $X = x_1x_2 \cdots x_m$ and $Y = y_1y_2 \cdots y_n$, let deleteScore(X,Y) be the least number of characters you have to delete from X,Y so that you get two same strings. For example, if X = goodman and Y = goldmann, then deleteScore(X,Y) = 3 (you can delete 'o' from X, 'l' and one 'n' from Y to get the same string - "godman").

Give an algorithm that given two strings X, Y computes deleteScore(X, Y). For full-credit, your algorithm should run in polynomial time.

We define the recurrence relation as:

$$deleteScore(X,Y) = \min \begin{cases} case \ 1:2 + deleteScore(X-1,Y-1) \ if \ (X_m \neq Y_n) \\ case \ 2:0 + deleteScore(X-1,Y-1) \ if \ (X_m = Y_n) \\ case \ 3:1 + deleteScore(m-1,n) \\ case \ 4:1 + deleteScore(m,n-1) \end{cases}$$

This works much like the edit distance algorithm.

Case 1: Instead of incrementing the cost by 1 when $X_m \neq Y_n$, we increment it by 2. This is because the cost to create two matching strings requires the deletion of X_m and Y_n .

Case 2: When X_m and Y_n match, we don't need to delete anything.

Case 3: When we insert a blank for Y_n in the edit distance algorithm, this is equivalent to deleting X_m .

Case 4: When we insert a blank for X_m in the edit distance algorithm, this is equivalent to deleting Y_n .

There are four types of brackets: (,), <, and >. We define what it means for a string made up of these four characters to be *well-nested* in the following way:

- (a) The empty string is well-nested.
- (b) If A is well-nested, then so are $\langle A \rangle$ and $\langle A \rangle$.
- (c) If S, T are both well-nested, then so is their concatenation ST.

Devise an algorithm that takes as input a string $s = (s_1, s_2, ..., s_n)$ of length n made up of these four types of characters. The output should be the length of the shortest well-nested string that contains s as a subsequence.

We use the same algorithm as RNA sequencing shown in class, but do not have the condition that a character can only be paired with another character more than 4 spaces away. Instead, we allow a closing bracket to be paired with a corresponding opening bracket at any prefix position.

Let OPT(i, j) be the length of the shortest well-nested string that contains s.

$$OPT(i, j) = \min \begin{cases} case \ 1 : 2 + OPT(i, j - 1) \\ case \ 2 : 1 + OPT(i, t - 1) + OPT(t + 1, j - 1) \end{cases}$$

Case 1: s_i is either a closing bracket that is not paired to an opening bracket or an opening bracket that was never paired with a closing bracket, so we +1 for the unpaired bracket and +1 for inserting the bracket that completes the pair.

Case 2: s_i is a closing bracket that is paried to an opening bracket s_t , so we +1 for the closing bracket and calculate the optimal solution for the two substrings (s_i, \ldots, s_{t-1}) and $(s_{t+1}, \ldots, s_{j-1})$.

We evaluate the subproblems in the order of small difference (j-i) to bigger difference, just like the RNA sequencing algorithm:

ALGORITHM:

- 1. Initialize OPT(i, k) = 0 for i = 0 and k = 0.
- 2. For k = 1, ..., n 1:

For
$$i = 1, ..., n - k$$
:

$$j = i + k$$

Compute OPT(i, j) using the recurrence.

3. RETURN OPT(1, n).

Just like the RNA sequencing algorithm, this algorithm runs in $O(n^3)$.