# Homework 4

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# **Greedy Problems**

### Problem 12:

This greedy algorithm can be shown to be correct by an exchange argument.

Let Alg be the process by which the greedy algorithm operates. Assume that there is some input I such that Alg(I) is incorrect. Let Opt(I) be an optimal solution that agrees with the most number of steps with Alg(I).

Opt(I) and Alg(I) must have a first point of disagreement which must occur at some row, since Alg(I) works row by row. Label the row in which the disagreement occurs  $r_i$ .

A general case of this disagreement can be demonstrated by the following two matrices.

$$Alg(I)_{n,n} = \begin{pmatrix} a_{1,1} & \cdots & a_{1,u} & \cdots & a_{1,v} & \cdots & a_{1,n} \\ a_{2,1} & \cdots & a_{2,u} & \cdots & a_{1,v} & \cdots & a_{1,n} \\ \vdots & \ddots & a_{i,u} & \ddots & a_{i,v} & \ddots & \vdots \\ a_{n,1} & \cdots & a_{n,u} & \cdots & a_{n,v} & \cdots & a_{n,n} \end{pmatrix}$$

$$Opt(I)_{n,n} = \begin{pmatrix} o_{1,1} & \cdots & o_{1,u} & \cdots & o_{1,v} & \cdots & o_{1,n} \\ o_{2,1} & \cdots & o_{2,u} & \cdots & o_{1,v} & \cdots & o_{1,n} \\ \vdots & \ddots & o_{i,u} & \ddots & o_{i,v} & \ddots & \vdots \\ o_{n,1} & \cdots & o_{n,u} & \cdots & o_{n,v} & \cdots & o_{n,n} \end{pmatrix}$$

Without loss of generality, allow the disagreement to be between columns u and v. In the matrices above, this could mean  $a_{i,u} = 1$  and  $a_{i,v} = 0$  whereas  $o_{i,v} = 0$  and  $o_{i,u} = 1$ .

Naievely defining Opt'I as OptI except  $o_{i,v} = 1$  and  $o_{i,u} = 0$  does make Opt'I agree with AlgI for an additional step and maintains the necessary sum of row  $r_i$ , labeled  $|r_i|$ .

However, this swap invalidates the column sums for  $c_u$  and  $c_v$ , respectively  $|c_u|$  and  $|c_v|$ , since they previously aggreed in Opt, but the swap altered the values to  $|c_u|' = |c_u| + 1$  and  $|c_v|' = |c_v| - 1$ , where  $|c_u|'$  and  $|c_v|'$  are the analogous column values in Opt'(I).

The column sums in Opt'(I) can be rectified, while still maintaining the row sums and the additional step agreement with Opt(I) if there exists a row,  $r_j$ , such that i < j, where  $o_{j,v} = 1$  and  $o_{j,u} = 1$ .

Since Alg operates row-by-row placing 1s in those columns which have the greatest need,  $c_i$  required more 1s than  $c_j$  at  $r_i$ . Since Opt(I) deprived  $c_i$  a 1 at  $r_i$ , it must give a 1 to  $c_i$  at some  $r_k$  such that i < k. Additionally, by giving  $c_j$  a 1 at  $r_i$ , Opt(I) is guaranteed to satisfy value  $|c_j|$  with fewer 1s than needed to satisfy  $|c_i|$ . This ensures the existence of a row with the above mentioned property.

Therefore redefine Opt'(I) such that Opt(I) except  $o_{i,v} = 1$ ,  $o_{i,u} = 0$  and  $o_{j,v} = 0, o_{j,u} = 1$  where j is the row where in Opt(I),  $o_{j,v} = 1$  and  $o_{j,u} = 1$ . This gives an optimal solution that agrees with at least 1 more step than Opt(I) does with Alg(I), which contradicts the original assumption that there exists an I such that Alg is incorrect. Therefore Alg is correct.

## Problem 18:

#### **A**:

The following counter example shows that this algorithm does not work.

Let the jobs be

$$J_1 = (1, 5, 5)$$

$$J_2 = (4, 2, 100)$$

This algorithm will do the following scheduling:

Since  $J_1$  was scheduled needed to finish after its deadline, the algorithm outputs 0. However, there is a feasible schedule for these two jobs.

#### B:

The following counter example show that this algorithm does not work.

Let the jobs be

$$J_1 = (1, 2, 2)$$

$$J_2 = (1, 1, 100)$$

This algorithm will do the following scheduling:

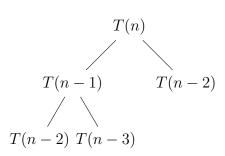
 $\mathbf{C}$ :

# **Dynamic Programming**

## Problem 1:

#### **A**:

A direct implementation of this recurrence relation leads to an exponential runtime because every T(i) requires  $2^i$  recurvive calls. This can seen using a tree diagram.



This tree will have a depth of i and at every point branches by into two at every node. Hence the  $O(2^i)$  calculations.

#### **B**:

To show that only  $O(n^2)$  operations are needed if every duplicate T(i) is calculated only once, begin by expanding the sum in the recurrence.

$$T(n) = \sum_{i=1}^{n-1} T(i)T(i-1)$$

$$T(n) = T(1)T(0) + T(2)T(1) + T(3)T(2) \cdot \cdot \cdot + T(n-2)T(n-3) + T(n-1)T(n-2) + T(n-2)T(n-3) + T(n-2)T(n-2) + T(n$$

Since every T(i) will only be calculated once, following sequence can be observed by counting the number of operations needed to determine each T(i).

It can be shown that the T(i+1) element of the sum requires two additional operations to calculate: a multiplication and an addition. Hence, this sequence will continue. It can be proven inductively that a closed form expression for the sum of operations required is  $n^2$ . Therefore, in this case  $O(n^2)$  operations are required.

 $\mathbf{C}$ :

A O(n) algorithm can be derived from the original recurrence relationship by first eliminating the summation by calculating T(n+1) in the following manner.

$$T(n+1) = \sum_{i=1}^{n} T(i)T(i-1)$$
 
$$T(n) = \sum_{i=1}^{n-1} T(i)T(i-1)$$
 
$$T(n+1) - T(n) = \sum_{i=1}^{n} T(i)T(i-1) - \sum_{i=1}^{n-1} T(i)T(i-1)$$

T(n+1) and T(n) overlap for all values  $i: 1 \le i \le n-1$ , therefore subtracting the two sums leaves only the final in the sum for T(n+1).

$$T(n+1) - T(n) = T(n)T(n-1)$$

The values for n can be shifted by setting n = m - 1.

$$T(m) - T(m-1) = T(m-1)T(m-2)$$

However, the label m is without meaning, so label m=n.

$$T(n) - T(n-1) = T(n-1)T(n-2)$$

Equivalently,

$$T(n) = T(n-1)[1 + T(n-2)]$$

This expression is easily expressed as a single  $\mathrm{O} \left( n \right)$  loop.

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\begin{array}{l} Array: \ T \\ T[0] = 2 \\ T[1] = 2 \\ \textbf{for } i \leftarrow 2 \ to \ n \ \textbf{do} \\ \mid \ T[i] = T[i-1] * (1+T[i-2]) \\ \textbf{end} \\ Output: \ T[n] \end{array}
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