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:

Time	1	2	3	4	5	6	7
Job	J_1	J_1	J_1	J_1	J_2	J_2	J_1

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.

Time	1	2	3	4	5	6	7
Job	J_1	J_1	J_1	J_1	J_1	J_2	J_2

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:

Time	1	2	3	4
Job	J_2	J_1	J_1	J_2

C:

The correctness of this algorithm can be shown by the following exchange argument.

Let Alg be the process by which this algorithm operates. Assume that there exists some input I such that $Alg(I)$ is incorect. Let $Opt(I)$ be the optimal solution for input I that agrees with the most number of steps with $Alg(I)$.

As a result, there must be some first point of disagreement between $Alg(I)$ and $Opt(I)$. Label this step u and let $Alg_u(I)$ be J_i and $Opt_u(I)$ be J_j.

The process by which Alg operates is to schedule the job with the least laxity, defined as

$$l = d_i - t - x_i(t)$$

This quantity represents the urgency with which a job must be scheduled, with Alg scheduling the most urgent job at every step. Therefore, since at step u $Alg(I)$ chose J_i whereas $Opt(I)$ has J_j at step u, J_i must have a lower laxity than J_j.

Alternatively, laxity can be viewed as the number of available slots a job has at a time t and its deadline. Since J_j has a higher laxity than J_i, it must have free slots in which another job can be scheduled.

Therefore, define $Opt'(I)$ as $Opt(I)$ except at step u schedule J_i and schedule J_j down by one unit, thereby using one of its lax spaces.

To show that $Opt'(I)$ is at least as optimal as $Opt(I)$ it needs to be recognized that J_i must have fewer lax slots than J_j, otherwise Alg would have picked J_j at step u. This includes the case in which J_j can't be shifted because it would violate its deadline d_j. Additionally, if J_i and J_j must overlap due, both $Opt'(I)$ and $Opt(I)$ would produce 0 for no viable schedule since only a single unit is changed in $Opt(I)$ and deadlines remain constant.

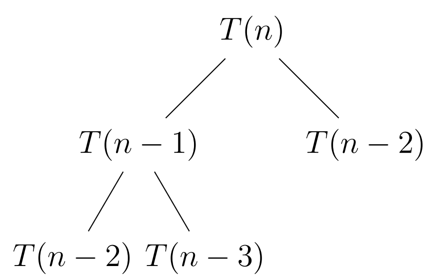
As a result, $Opt'(I)$ agrees with $Alg(I)$ for more steps than $Opt(I)$, contradicing the initial assuptions. Therefore there is not an input I such that $Alg(I)$ is incorrect.

Dynamic Programming

Problem 1:

A:

A direct implementation of this recurrence relation leads to an exponential runtime because every $T(i)$ requires 2^i recurvsive 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 $\mathrm{O}\big(2^i\big)$ calculations.

B:

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

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

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

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

T (i)	T (2)	T (3)	T (4)	T (5)	T (6)
Ops	1	3	5	7	9

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 $\mathrm{O}\big(n^2\big)$ operations are required.

C:

A $\mathrm{O}\big(n\big)$ 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 $i : 1 \leq i \leq n-1$, therefore subtracting the two sums leaves only the 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 $O(n)$ loop.

$Array:\ T$
$T[0] = 2$
$T[1] = 2$
for $i \leftarrow 2$ *to* n **do**
$|\quad T[i] = T[i-1] * (1 + T[i-2])$
end
$Output:\ T[n]$