

Homework 4

Robbie McKinstry, Jack McQuown, Cyrus Ramavarapu

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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 .

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

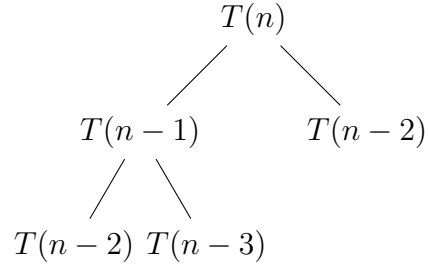
C:

Dynamic Programming

Problem 1:

A:

A direct implementation of this recurrence relation leads to an exponential runtime because every $T(i)$ requires 2^i recursive calls. This can be 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.

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

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

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 $O(n^2)$ operations are required.

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 $i : 1 \leq i \leq n-1$, 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.

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

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