Homework 5

Robbie McKinstry, Jack McQuown, Cyrus Ramavarapu

12 September 2016

Greedy Problems

Problem 7:

A greedy algorithm that will minimize evictions is to select the page that will be accessed furthest in the future from the current time.

For example, assuming k=4 let the current state of the system be the following:

In this situation, the algorithm will evict page B, since it is not used until time 5. Any other choice will result in an earlier eviction.

Proof by Exchange:

Let Alg be the process by which the above algorithm operates. Assume there exists an input I such that Alg(I) is incorrect. Let Opt(I) be the optimal result for I that agrees with the greatest number of steps with Alg(I). Alg(I) and Opt(I) must have a first point of disagreement. Label this time t_i .

At time i, Alg(I) must have evicted some page u and Opt(I) has some page

v evicted. Since Alg(I) always picks the page that will be used furthest in the future, the next time page u will be used must be further into the future than the next time page v is used. Let these times be respectively t_u and t_v .

Since Alg(I) and Opt(I) both agreed upto time t_i , pages u and v must be in the fast memory of Opt(I) at t_i . Therefore, define Opt'(I) as Opt(I) except at time t_i evict page u instead of v.

To show that Opt'(I) is at least as optimal as Opt(I) it needs to be recognized that since $t_u > t_v$, page v will have to be brought back into memory at least 1 more time before page u is brought back into memory.

Since Opt'(I) is at least as optimal as Opt(I) and agrees with Alg(I) for 1 more step a contradiction has be reached. Therefore, there is not input for which Alg is incorrect.

Problem 17:

A greedy algorithm that will determine if there is enough information available to fairly partition the goods is as follows:

Given a set of goods $\mathcal{G} = G_1, G_2, \ldots, G_n$, and two orderings on this set, $\mathcal{H} = G_a > G_b > \cdots > G_n$ and $\mathcal{W} = G_i > G_k > \cdots > G_m$, partition the goods by initially giving \mathcal{W} and \mathcal{H} their maximal elements. Remove the element given to the opposing list from each list. If this is not possible because the maximal elements for both \mathcal{W} and \mathcal{H} are the same, conclude that there is not enough information guarantee a fair partitioning.

While there are elements in \mathcal{G} , continue giving \mathcal{W} and \mathcal{H} their second most desired element. Ties are broken arbitrarily. For example, if at a given step both \mathcal{W} and \mathcal{H} want G_c , arbitrarily give G_c to \mathcal{W} and cross it off in the ordering for \mathcal{H} and give \mathcal{H} the next element in the ordering.

After all elements have been assigned, first evenly parition \mathcal{H} and \mathcal{W} into two equal portions. If this is not possible because $|\mathcal{G}|$ is odd, conclude that there is not enough information available to fairly partition the goods. If the upper half of \mathcal{H} and \mathcal{W} is full, conclude that there is enough information to

fairly partition.

If the upper half of \mathcal{H} or \mathcal{W} are not full, begin iterating through both orderings and check to see that items given alternate with items not given. Repeat this process for alternation in groups of 2, 3,..., $|\mathcal{G}|/2$. Every assigned element must belong to at least 1 of the alternating patterns. If an assignment was reach such that elements from \mathcal{G} can be given to \mathcal{H} and \mathcal{W} such that both pass at least one test, conclude that there is enough information to fairly partition the goods. Otherwise, conclude that there is not enough information available.

Proof by Exchange:

This algorithm can be shown to be correct by an exchange argument. Let Alg be the process by which the above algorithm operates. Assume there exists some input I such that Alg(I) is incorrect. Let Opt(I) be the optimal output for input I that aggrees with Alg(I) for the most number of steps. Therefore, Alg(I) and Opt(I) must have a first point of disagreement. Label this step s_i .

At s_i , Alg(I) must have given a different pair of goods to \mathcal{H} and \mathcal{W} than Opt(I). This could mean that Alg(I) respectively gave \mathcal{H} and \mathcal{W} G_a , G_b whereas Opt(I) gave G_c , G_d . However, since Alg attempts to assign the maximal element at each step to \mathcal{H} and \mathcal{W} and since Alg(I) and Opt(I) agree upto step s_i , the items G_a , G_b must be assigned in Opt(I) at a step, s_j , after s_i . As a result, define Opt'(I) as Opt(I) except at s_i assign G_a and G_b to the same lists as Alg(I) and move the items originally assigned in Opt(I) at s_i to s_j .

To check that Opt'(I) is at least as optimal as Opt(I), the case in which Opt(I) determines that there is enough information to partition the goods and the case in which it finds the information insufficient need to be considered separately. In the first case, the condition for enough information requires the sequence of assigned goods and unassigned goods follow an alternating pattern. Since Alg(I) and Opt(I) agreed until step s_i alternation must have already started and to continue, G_a , G_b and G_c , G_d must have been in alternating positions. Exchanging these values therefore does not change the pattern, giving the same result.

In the case that Opt(I) finds the ordering information insufficient to fairly partition the goods, there will be an element that does not belong to any alternating pattern of the elements assigned to \mathcal{H} and \mathcal{W} . Since the elements assigned by Opt(I) at s_i are not the maximal elements available, swaping G_a , G_b instead of G_c , G_b , will not change the patterning since the elements are ordered. As a result, Opt'(I) will also conclude that there is not enough information available.

Since in both cases Opt'(I) is at least as optimal as Opt(I) and agrees with Alg(I) for one more step than Opt(I), a contradiction has been reached. Therefore there does not exist an input for which Alg(I) is incorrect.

Dynamic Programming

Problem 2:

The longest common subsequence between the three can be initially defined recursively by considering the last letter in common between the three strings and then seeing if this letter is in the longest common subsequence of one letter shorter.

$$S_A = A_1, A_2, A_3, \dots, A_{i-1}, A_i$$

 $S_B = B_1, B_2, B_3, \dots, B_{j-1}, B_j$
 $S_C = C_1, C_2, C_3, \dots, C_{k-1}, C_k$

Reading right to left, the first letter common to each string will be the last letter in the longest common subsequence. As a result, once this letter is found, problem can be redefined in terms of the shorter substrings formed by ignoring the first common point and all succeeding letters.

This analysis leads to the following recursive algorithm:

```
Function: LCS
Input: int \ i, \ int \ j, \ int \ k
if i \equiv j \equiv k \equiv 0 then
| \ return \ 0
end
if A_i \equiv B_j \equiv C_k then
| \ LCS(i-1, j-1, k-1) + A_i
else if (A_i \equiv B_j) \neq C_k then
| \ \max(LCS(i-1, j-1, k), \ LCS(i, j, k-1))
else if (A_i \equiv C_k) \neq B_j then
| \ \max(LCS(i-1, j, k-1), \ LCS(i, j-1, k))
else if A_i \neq (B_j \equiv C_k) then
| \ \max(LCS(i, j-1, k-1), \ LCS(i-1, j, k))
else
| \ \max(LCS(i, j-1, k-1), \ LCS(i-1, j, k-1), \ LCS(i-1, j-1, k))
end
```

Although this algorithm produces the longest common subsequence for three given strings, it runs in exponential time due to the numerous recursive calls that operate on a problem of only 1 letter smaller.

By moving to an array based solution and changing the recursive calls to array look-ups, a polynomial runtime algorithm can be developed.

```
Function: Array LCS
Input: string A, String B, String C
Initialization:
Array[ ][ ][ ][ LCS]
for i \leftarrow 0 to len(A) do
 LCS[i][0][0] = 0
end
for j \leftarrow 0 to len(B) do
 |LCS[0][j][0] = 0
end
for k \leftarrow 0 to len(C) do
 |LCS[0][0][k] = 0
end
Array Calculations:
for i \leftarrow 0 to len(A) do
   for j \leftarrow 0 to len(B) do
       for k \leftarrow 0 to len(C) do
           if A_i \equiv B_i \equiv C_k then
               LCS[i\check{j}][j][k] = LCS[i-1][j-1][k-1] + 1
           else if (A_i \equiv B_i) \neq C_k then
               LCS[i][j][k] = \max(LCS[i-1][j-1][k], LCS[i][j][k-1])
           else if (A_i \equiv C_k) \neq B_i then
               \max(LCS[i][j][k] = LCS[i-1][j][k-1], LCS[i][j-1][k])
           else if A_i \neq (B_i \equiv C_k) then
               LCS[i][j][k] = \max(LCS[i][j-1][k-1], LCS[i-1][j], [k])
           else
               LCS[i][j][k] = \max(LCS[i][j-1][k-1],
               LCS[i-1][j][k-1], \ LCS[i-1][j-1][k]
           end
       \quad \text{end} \quad
    end
end
```

By tracing backwards from LCS[len(A) - 1][len(B) - 1][len(C) - 1] following the path of the largest lengths, the *String* value of the longest common subsequence can be recovered.