Homework 8, Solution set

1. Val[i, j] is an array that stores the value for a knapsack of size j for the first i items, with  $0 \le i \le n$  and  $0 \le j \le t$ . It is initialized to be all zero; similarly, the array Done[i, j] is initialized to False for the same range of i, j values. Following this, OptKnap(n, t) is called.

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\begin{aligned} &\operatorname{OptKnap}(n,t) \\ &\mathbf{if} \ n=0 \ \operatorname{or} \ t=0 \ \mathbf{then} \ \operatorname{Val}[n,t] \leftarrow 0 \\ &\mathbf{else} \\ &\mathbf{if} \ \ \mathbf{not}(\operatorname{Done}[n-1,t]) \ \mathbf{then} \ \operatorname{OptKnap}(n-1,t) \\ &\mathbf{end} \ \mathbf{if} \\ &\mathbf{if} \ \ \mathbf{not}(\operatorname{Done}[n-1,t-s_n]) \ \operatorname{and} \ t \geq s_n \ \mathbf{then} \ \operatorname{OptKnap}(n-1,t-s_n) \\ &\mathbf{end} \ \mathbf{if} \\ &\operatorname{Val}[n,t] \leftarrow \max\{\operatorname{Val}[n-1,t],\operatorname{Val}[n-1,t-s_n] + v_n\}; \\ &\mathbf{end} \ \mathbf{if} \\ &\operatorname{Done}[n,t] \leftarrow \operatorname{True} \end{aligned}
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There are up to  $(n+1) \cdot (t+1)$  recursive calls that might be made. Each recursive call has an O(1) cost for its non-recursive computation. Thus the overall cost is O(nt).

2. We introduce a new array Root[i, k] with  $1 \le i \le n$  and  $1 \le k \le n$ ; it will store the index of the best choice for the item to go at the root of the BST for items  $e_i, e_{i+1}, \ldots, e_k$ .

The following is the enhanced recursive procedure.

```
OptBST(i, k)
if i = k + 1 then Cost[i, k] \leftarrow 0
else
   Cost[i, k] \leftarrow MaxInt;
   RootCost \leftarrow Tot[k] - Tot[i-1];
   for j = i to k do
       if not(Done[i, j-1]) then OptBST(i, j-1)
       end if
       if not(Done[j+1,k]) then OptBST(j+1,k)
       end if
       if Cost[i, k] > Cost[i, j - 1] + Cost[j + 1, k] + RootCost then
           \text{Root}[i,k] \leftarrow j; (* this is the code enhancement *)
           Cost[i, k] \leftarrow Cost[i, j - 1] + Cost[j + 1, k] + RootCost
       end if
   end for
end if
Done[i, k] \leftarrow \text{True}
```

After the run of OptBST(1, n) we can build an optimal BST with a call to BuildOptBST(1, n, T), which returns T, or more precisely, a pointer to its root.

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\begin{array}{l} \operatorname{BuildOptBST}(i,k,v) \text{ (* returns $v$, a pointer to the root of the tree *)} \\ \textbf{if $i>k$ then $v\leftarrow \text{nil}$} \\ \textbf{else} \\ & \operatorname{GetNode}(v); \\ j \leftarrow \operatorname{Root}[i:k]; \\ v. \operatorname{item} \leftarrow e_j; \\ \operatorname{BuildOptBST}(i,j-1,w); \\ \operatorname{BuildOptBST}(j+1,k,x); \\ v. \operatorname{left} \leftarrow w; \\ v. \operatorname{right} \leftarrow x \\ \textbf{end if} \end{array}
```

The procedure call OptBST(1, n) makes up to n(n + 1) recursive calls (including the initial call). The call OptBST(i, k) has non-recursive cost O(k - i + 1) = O(n). Thus the overall cost is  $O(n^3)$  (and actually is  $\Theta(n^3)$ , though we have not shown that).

The procedure BuildOptBST(1, n, T) runs in O(n) time. The reason is that each recursive call takes O(1) time, and there is one recursive call per node in the tree. But a distinct item is placed at each node, so there are n nodes in total, resulting in an overall O(n) runtime.

3.a. We define  $\operatorname{Val}(m, j, u)$  to be a function that returns True if one can use j or fewer coins with values any of  $v_1, v_2, \ldots, v_m$  to make a total value of u, and returns False otherwise.  $\operatorname{Value}[m, j, u]$  is an array that stores this Boolean value, for  $1 \leq m \leq n$ ,  $1 \leq j \leq k$ ,  $1 \leq u \leq v$ .

$$\operatorname{Val}(m, j, u) = \begin{cases} \operatorname{Val}(m, j - 1, u - v_m) & m = 1, u \ge v_m \\ \operatorname{Val}(m, j - 1, u - v_m) \vee \operatorname{Val}(m - 1, j, u) & m > 1, u \ge v_m \\ \operatorname{Val}(m - 1, j, u) & m > 1, u < v_m \\ \operatorname{True} & u = 0 \\ \operatorname{False} & m = 1, u < v_m \end{cases}$$

This follows because either one uses an instance of the mth coin or not, with base cases occurring when m = 1 or  $v_m > u$ .

- b. There are (n+1)kv possible recursive calls. In an efficient implementation, each of them performs O(1) non-recursive work, yielding a total O(nkv) runtime.
- c. For each recursive call we need to record which choice, if any, yields an outcome of True. The choices are: none, use another coin of value  $v_m$ , don't use another coin of value  $v_m$ .
- 4. Let  $L(i,k) = \sum_{j=i}^{k} |L_j|$ , the total length of the *i*th through *k*-th lists. Note that L(i,k) = Lnth[k] Lnth[0], where  $\text{Lnth}[j] = \sum_{h=1}^{j} |L_h|$ . The values Lnth[0:n] are readily computed in O(n) time.

The recursive formulation is given by:

$$M(i,k) = \left\{ \begin{array}{ll} \min_{i \leq j \leq k} \left\{ M(i,j) + M(j+1,k) \right\} + L(i,k) & k > i \\ 0 & k = i \end{array} \right.$$

The first line follows because the cost of the final merge is  $\sum_{j=1}^{k} |L_j|$  regardless of which subsets of lists are being merged in this step, so it suffices to minimize over all choices of j, i.e. of recursively merging  $L_i, \ldots, L_j$  and  $L_{j+1}, \ldots, L_k$ .

- b. As  $1 \le i \le k \le n$ , there are at most  $\frac{1}{2}n(n-1)$  recursive calls, or more loosely,  $O(n^2)$  recursive calls. In an efficient implementation, the recursive call M(i,k) performs O(k-i+1) = O(n) non-recursive work, Therefore the total runtime is  $O(n^3)$ .
- c. For each recursive call M(i,k), we need to record the value j that yields an optimal outcome for that subproblem.