

Homework 10

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Assume the vertices are ordered via a breadth first search or a depth first search. This ensures a list of vertices from v_1 to v_n such that any vertex v_i is connected to one and only one vertex in $v_1 \dots v_{i-1}$. v_i cannot be connected to more than one vertex in $v_1 \dots v_{i-1}$, otherwise there would be a cycle and T would not be a tree. v_1 must be connect to a vertex in $v_1 \dots v_{i-1}$ by definition of a tree search. Let $neighbor(v_i)$ represent the sole neighbor of v_i of $v_1 \dots v_i$. Let w_i be the weight

Base case: When T has one vertex, the most profitable (and only) thing to do is stay put. Return 0 for the best profit.

MostProfitableEndpoints[1] = (1, 1)

MaximumProfit[1] = 0

MostProfitablePathRootedAt[1] = v_1 # The path is (1, 1): the start and end point

RootedProfit[1] = 0

for $i := 2$ to n do:

 if RootedProfit[neighbor(v_i)] + $w_i > 0$ then:

 RootedProfit[i] = RootedProfit[neighbor(v_i)]

 MostProfitablePathRootedAt[i] = MostProfitablePathRootedAt[neighbor(v_i)]

 else: # Stay put

 RootedProfit[i] = 0

 MostProfitablePathRootedAt[i] = v_i

Let $u, v = \text{MostProfitableEndpoints}[i - 1]$

 if neighbor(v_i) $\in (u, v)$ and $w_i > 0$ then: # v_i adds value to the optimal solution

 if neighbor(v_i) == u then: MostProfitableEndpoints[i] = (v_i, v)

 else then neighbor(v_i) == v : MostProfitableEndpoints[i] = (u, v_i)

 MaximumProfit[i] = MaximumProfit[i - 1] + w_i

 else if RootedProfit[i] \nless MaximumProfit[i - 1]: # v_i could possibly still be in the best path.

 MaximumProfit[i] = RootedProfit[i]

 MostProfitablePathRootedAt[i] = MostProfitablePathRootedAt[i - 1]

 else: # v_i doesn't contribute at all to the maximum profit path

 MaximumProfit[i] = MaximumProfit[i - 1]

 MostProfitableEndpoints[i] = MostProfitableEndpoints[i - 1]

Output MostProfitableEndpoints[n]

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An algorithm for this problem is a modified form of the Optimal Binary Search Tree dynamic algorithm. Like the OBST dynamic programming algorithm, this algorithm caches the minimum weight of a tree with keys K_l and K_r in a 2D array, Weight[l][r]. It also stores the key K_i where $l \leq i \leq r$ that should be used as the root of the OBST for keys from K_l to K_r in a 2D array, Root[l][r]. Our algorithm differs from the OBST algorithm in that there is a

third 2D array, $\text{Height}[l][r]$ which stores the height of the OBST with keys from K_l to K_r .

The base case is the same as with the OBST algorithm. When $l = r$, that is, on the diagonal of the 2D array, the only key that can be considered as the (childless) root is K_l . So $\text{Root}[l][r] = K_l$ and $\text{Weight}[l][r] = w_l$. Moreover, since the sole key does not have any children, $\text{Height}[l][r] = 1$.

The recursive case considers all keys from K_l to K_r . Like the OBST algorithm, consider the key K_i that has the minimum weight of the sum of its subtrees $\text{Weight}[l][i-1] + \text{Weight}[i+1][r]$ plus the sum of the weight of all the keys K_l through K_r , and make $\text{Root}[l][r] = K_i$. But do not consider K_i if $\text{Height}[l][i-1] - \text{Height}[i+1][r] \geq 1$ or ≤ -1 . Discard K_i as a candidate root if this is the case.

As with OBST, output the tree from K_0 to K_n . This algorithm is very similar, only it requires caching a third property of the subtrees, and has greater scrutiny over candidate roots.