

CSE 101 Homework 4

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1. Maximum consecutive sum

Here is an algorithm that, given an array of integers (not necessarily positive) $A[1..n]$, finds the maximum sum of a consecutive subarray, $\max_{1 \leq I \leq J \leq n} \sum_{K=I}^J A[K]$.

- (a) Initialize $\text{MaxSumEndingAt}[1..n]$.
- (b) $\text{MaxSumEndingAt}[1] = A[1]$
- (c) For $J = 2$ to n do:
 - (d) $\text{MaxSumEndingAt}[J] = \max(A[J], A[J] + \text{MaxSumEndingAt}[J - 1])$
 - (e) $\text{MaxConsSum} = \text{MaxSumEndingAt}[1]$.
 - (f) For $J = 2$ to n do:
 - (g) $\text{MaxConsSum} = \max(\text{MaxConsSum}, \text{MaxSumEndingAt}[J])$
 - (h) Return MaxConsSum
- (a) For each $J \in \{1, \dots, n\}$,

$$\text{MaxSumEndingAt}[J] = \max_{1 \leq i \leq J} \sum_{k=i}^J A[k]$$

This means it stores the largest sum of any non-empty subarray that ends at position J . From the algorithm, this can also be written as:

$$\text{MaxSumEndingAt}[J] = \max(A[J], A[J] + \text{MaxSumEndingAt}[J - 1]).$$

- (b) **Loop invariant:** Before each loop iteration at index J ($J \geq 2$), $\text{MaxSumEndingAt}[t]$ holds the correct value for all $t \in \{1, \dots, J - 1\}$. After computing line 4 for J , the invariant also holds for index J .

Proof:

- *Base case ($J = 2$):* This is the state before the first loop iteration. Line 2 sets $\text{MaxSumEndingAt}[1] = A[1] = \max_{1 \leq i \leq 1} \sum_{k=i}^1 A[k]$.
- *Inductive step:* Assume the invariant holds up to $J - 1$. The best subarray ending at J either:
 - i. starts fresh at $A[J]$, or
 - ii. continues from the best subarray ending at $J - 1$.

Therefore,

$$\text{MaxSumEndingAt}[J] = \max(A[J], A[J] + \text{MaxSumEndingAt}[J - 1]),$$

which matches the formula, so the invariant holds for J .

- (c) The goal is to find:

$$\text{Ans} = \max_{1 \leq I \leq J \leq n} \sum_{k=I}^J A[k].$$

Since each $\text{MaxSumEndingAt}[J]$ gives the best sum ending at J ,

$$\text{Ans} = \max_{1 \leq J \leq n} \text{MaxSumEndingAt}[J].$$

The second loop keeps track of this maximum:

$$\text{MaxConsSum} = \max(\text{MaxConsSum}, \text{MaxSumEndingAt}[J]).$$

Initially (line 5), $\text{MaxConsSum} = \text{MaxSumEndingAt}[1]$. At the end (line 8), $J = n$ so $\text{MaxConsSum} = \max_{1 \leq J \leq n} \text{MaxSumEndingAt}[J] = \text{Ans}$.

Hence, the algorithm is correct.

- (d) The first loop (lines 3–4) runs $n - 1$ times, each taking constant time. The second loop (lines 6–7) also runs $n - 1$ times. Therefore, the total time is $\Theta(n)$.

2. Maximum within sliding window

You are given an array $A[1..n]$ of integers, and an integer $1 \leq k \leq n$. You wish to compute $\text{MaxInWindow}[I] = \max_{I \leq J \leq \min(I+k, n)} A[J]$ for all $1 \leq I \leq n$. By answering the questions below, you'll develop an efficient algorithm for this problem. A helpful abstraction will be to consider a window set at each I , $WS[I] = \{(J, A[J]) | I \leq J \leq \min(I+k, n)\}$.

- (a) In terms of the window set at I , what is $\text{MaxInWindow}[I]$?

$$\text{MaxInWindow}[I] = \max\{v \mid (J, v) \in WS[I]\}.$$

- (b) What is the difference between $WS[I]$ and $WS[I+1]$?

The window set $WS[I+1]$ differs from $WS[I]$ by one element leaving and one new element entering the window. Specifically, $(I, A[I])$ is removed from the set and $(I+k+1, A[I+k+1])$ is added, given that $(I+k+1 \leq n)$.

$$WS[I+1] = (WS[I] \setminus \{(I, A[I])\}) \cup \begin{cases} \{(I+k+1, A[I+k+1])\}, & \text{if } I+k+1 \leq n, \\ \emptyset, & \text{otherwise.} \end{cases}$$

- (c) Based on your answers to the first two questions, what data structure operations would we want a data structure for WS to support?

- `insert(x)` - In order to add the new element entering the window
- `delete(x)` - In order to delete the element exiting the window
- `get_max()` - To get the current maximum value within the window

- (d) What data structure could we use for WS ?

Data structures that support the needed operations include:

- Balanced Binary Search Trees (BST or RST)
- Double Ended Queue
- Priority Queue

- (e) What is the maximum size of this data structure?

At most $k+1$ elements (the window is $[I, \min(I+k, n)]$ inclusive).

- (f) How long do the different data structure operations take?

BST:

- `insert` = $O(\log k)$
- `delete` = $O(\log k)$
- `get_max` = $O(1)$ if we keep a pointer to the largest (or $O(\log k)$ to query the max).

Plain Dequeue (without monotone maintenance):

- push/pop at ends = $O(1)$
- get_max requires scanning = $O(k)$

Max-Heap:

- get_max = $O(1)$
- insert = $O(\log k)$ ($O(k)$ in worst case)
- delete = $O(\log k)$ ($O(k)$ in worst case)

(g) Give pseudo-code for an algorithm to solve this problem using data structure operations.

```
max_in_window():
    max_arr = [1...n]
    heap = initialize_max_heap( (i, A[i]) for i in range 0...k )

    for i in 1...n:
        max[i] = heap.peek()

        heap.remove(key=i)
        if (i + 1 + k <= n):
            heap.insert((i+1+k, A[i+1+k]))
```

(h) Give a time analysis for your algorithm, using answers to previous questions.

- The loop runs for each index of the array.
- The heap is initialized in $O(\log k)$
- heap.peek runs in $O(1)$
- heap.remove / heap.insert both run in $O(\log k)$ in the average case.

$$\begin{aligned} \text{Time Complexity} &\in O(\log k + n(2 \cdot \log k)) \\ &\in O(n \cdot \log k) \end{aligned}$$

In the worst case (insert/delete on the heap run in $O(k)$). Time complexity $\in O(n \cdot k)$

3. Subway stops

Underneath a city road, there are n subway stops s_1, \dots, s_n and k subway lines. Each subway line k stops at some of the stops, in the same order, with line k stopping at $s_{i_{k,1}}, \dots, s_{i_{k,t_k}}$, where $i_{k,1} < i_{k,2} < \dots < i_{k,t_k}$. You want to go from s_1 to s_n making as few transfers between lines as possible.

For example, there might be five stops, and three lines. The first line might stop at stops 1, 3, the second at 2, 4, 5, and the third at 2, 3, 4. Then we could board on line 1, take it to stop 3, switch to line 3, take it to stop 4, switch to line 2 and take it to the end.

Below, you will describe how to use a graph algorithm from class to solve the problem.

- (a) What graph will you use to solve the problem? Be sure to specify the set of vertices in your graph, the set of edges, whether the edges are directed or undirected, and what weights edges have, if any.
 - This will be a directed graph. Each node $S_{i,k}$ represents a subway station i reached by subway line k
 - There will be unweighted (weight = 0), directed edges representing getting from one subway station to another via line k
 - There will be weighted (weight = 1), undirected edges representing switching subway line at a single station. For example, these edges will go from vertex S_{i,k_1} to S_{i,k_2}

- (b) How will you create the graph from the information given? What format will you use for the graph? How long does it take to create the graph?

To create the graph from the given information:

- Create a grid of k rows by n columns of vertices. Label each $S_{row,col}$
- Draw unweighted, directed edges from $S_{i_1,k}$ to $S_{i_2,k}$ (collection of all stations reachable by line k) for every rail line k .
- Draw weighted (weight = 1), undirected edges between the rail stops at each specific station
- Create a single node at the start and end of the graph, with directed, unweighted edges to all (S_i, k) stops. This will serve as the start and end goals of the algorithm.
- Remove any unused rail stops. (a rail line that is not used at station i)

Given the answers in the following 2 questions, you must draw at most $n \cdot k + 2$ nodes, $(n - 1) \cdot k$ rail connections, $2 \cdot (k - 1) \cdot n$ rail line changes, and $2k$ edges from the first/last nodes to the first/last stations. Asymptotically, it will take $O(nk)$

- (c) How many vertices does your graph have, at most? Give this in terms of n and k .

$$|V| \leq n \cdot k + 2$$

- (d) How many edges does your graph have, at most? Give this in terms of n and k

- Directed Edges. If each rail line connects to every stations, there are at most $(n - 1) \cdot k$ edges

- Undirected Edges. To connect every rail line at every station requires 2 directed edges, so there are at most $2(k - 1) \cdot n$ edges
- Extra Edges. If there are k rail line connections at the first and last stations, then the first and last nodes will have k edges to and from them respectively.

$$|E| \leq (n - 1) \cdot k + 2(k - 1) \cdot n + 2 \cdot k$$

- (e) How do paths in your graph relate to ways of transferring? What is the relationship between the length of paths and the number of transfers?

Unweighted edges represent no transfers. Subway line transfers have a weight of 1. Thus the weight of the paths from S_1 to S_n is the number of transfers made during that trip.

- (f) What algorithm from class will you run on the graph? Be sure to specify all inputs to this algorithm, and say how you use the results.

The modified Dijkstras algorithm should be used to find the shortest weighted path between the first and last subway station. This will automatically minimize the number of transfers.

The algorithm would assume the goal is to get from station S_1 to station S_n . It would run the modified Dijkstras algorithm from class on the first node and try to reach the last node.

The fully weighted and directed graph would be input (either in adjacency list or matrix format). The result would either be the length of the shortest trip, or an array of previous nodes showing the shortest path.

- (g) What is the total time complexity of using this algorithm from class to solve the subway transfer problem? This should be given in terms of n and k .

The algorithm uses the same Dijkstras algorithm we used in class, and has the same time complexity: $O(|\text{RemoveMin}| |V| + |\text{ReduceKey}| |E|)$

For this problem specifically:

$$\begin{aligned} & O(|\text{RemoveMin}| \cdot (nk) + |\text{ReduceKey}| \cdot (nk)) \\ & \in O((nk) \cdot (|\text{RemoveMin}| + |\text{ReduceKey}|)) \end{aligned}$$

Depending on the underlying implementation of the min-heap, the exact time complexity would vary.

4. Vertex costs

Say we are given a graph G where both vertices and edges have positive integer weights and two vertices u, v . We want to find a path from u to v that minimizes the total weights of both edges and vertices along the path. Give an efficient algorithm to solve this problem. You can use any algorithm from class as given, but need to relate the correctness guarantee proved for that algorithm to correctness for this new problem. (10 points clear algorithm description, 10 points correctness argument, 5 points time analysis and efficiency)

Given an undirected graph G , we can convert each undirected edge to two directed edges. If the graph G is directed, we skip this step. Now with our directed graph we can create an adapted graph (G_2) where, for each vertex u in G , we can add the vertex weight u_ω to the edge weight $\omega(u, v)$ for all outgoing edges from vertex u . With this adapted graph, we can run Dijkstra's graph search algorithm to find the shortest weighted path from u to v .

Proof of correctness: For this problem, we only need to create the new edge weights for graph G_2 from G , and then can run the graph search algorithm Dijkstra's from class, and so the proof of correctness to find the shortest path is the same.

Time analysis and efficiency: To implement the graph, we will use the priority queue implementation. In our adapted graph (G_2), the number of vertices does not change from the original graph (G). If graph G was originally undirected, the number of edges in G_2 is doubled so the runtime of the algorithm is $O(\text{deletemin} * |V| + \text{decreasekey} * 2|E|)$. If the original graph was directed, the number of vertices and the number of edges in G_2 is the same as G , so the runtime is $O(\text{deletemin} * |V| + \text{decreasekey} * |E|)$ using a priority queue.