Advanced Algorithms Problems and Solutions

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Contents

1	Hog	gwarts	2
	1.1	Solution 1: Preprocessing-then-Dijkstra	2
		1.1.1 Pseudo-code	2
	1.2	Solution 2: HogwartsDijkstra	3
		Solution 3: BFS-like traversal	
2	Paletta		
	2.1	Solution 1: Split and count-inversions	5
3	Rar	nge updates	7
		Solution 1: Fenwick lazy a-b sums	7
3	Depth of a node in a random search tree		
	_	Recursive balanced proof	1

1 Hogwarts

The Hogwarts School¹ is modeled as a graph G = (V, E) where V is the set of castle's rooms and $E \subseteq V \times V$ is the set of the stairs. Each stair is labelled with the time of appearance and disappearance, and can be walked in both directions, therefore the graph is undirected. The goal is to find, if possible, the minimum amount of time required to go from the first to the last room.

1.1 Solution 1: Preprocessing-then-Dijkstra

Dijkstra is able to find the shortest path in a graph with non-negative weights on its edges. Our main idea is to create a Dijkstra compatible graph through a *normalize* function, then apply Dijkstra to it in order to find the shortest path. The core of the preprocessing is the normalize function which computes traversal times between nodes at a given time time:

```
1: function NORMALIZE(from, to, time):
 2:
       t = \infty
 3:
       if start[v'] \le t < end[v'] then
                                                                 ▷ No waiting time
 4:
          t = t + 1
 5:
 6:
       else if t < start[v'] then
                                                                    ▶ Waiting time
 7:
           t = start[v'] + 1
 8:
 9:
       else
10:
          t=\infty

    Available time already expired

11:
       end if
12:
13: return t
14: end function
```

Computational cost: $\Theta(n^2)$ if the vertex set is implemented as an array. $O(|E| + |V| \log |V|)$ with Fibonacci heap.

The normalize function is then applied to a node traversal:

1.1.1 Pseudo-code

 $^{^{1}} http://didawiki.cli.di.unipi.it/lib/exe/fetch.php/magistraleinformatica/alg2/algo2_16/hogwarts.pdf$

```
6:
       if edge \in visited\_edges then return
 7:
       end if
8:
       traversal\_time = \infty
9:
10:
       for each neighbor \in neighbors\_of\_node do
           traversal_time = traversal_time(node, neighbor, time)
11:
           E'[0][node] = traversal\_time
                                              ▷ E'[i][j] holds the weight/traversal
12:
                                              ▷ time for the stair between i and j
13:
           for each new\_neighbor \in neighbors\_of\_neighbor do
14:
              normalize(neighbor, new_neighbor, traversal_time)
15:
           end for
16:
       end for
17:
       dijkstra\_graph = \{V, E'\}
18:
       t = dijkstra(dijkstra\_graph)
19:
20:
       if t == \infty then return -1;
21:
       elsereturn t;
22:
       end if
23:
24: end function
```

Computational cost. See the previous section.

1.2 Solution 2: HogwartsDijkstra

```
1: function HogwartsDijkstra(G):
        create vertex set Q of unvisited nodes
 2:
        for each vertex v \in V do
                                                                            ▶ initialization
 3:
            time[v] \leftarrow \infty
                                                     ▷ unknown time from source to v
 4:
 5:
            add v to Q
                                                                ▶ all nodes initially in Q
        end for
 6:
        time[0] \leftarrow 0
                                                           ▶ time from source to source
 7:
        while Q \neq \emptyset do
 8:
            u \leftarrow x \in Q \text{ with } \min\{time[x]\}
 9:
            remove u from Q
10:
11:
            for each neighbor v of u do:
                if time[u] \leq appear[v] then
12:
                    alt \leftarrow appear[v] + 1
                                                     ▶ wait the appearance of the stair
13:
                else if time[u] < disappear[v] then
14:
                    alt \leftarrow time[u] + 1
                                                                            ▶ use the stair
15:
16:
                else
17:
                    alt \leftarrow \infty

    b the stair has already disappeared

                end if
18:
                if alt < time[v] then
19:
                    time[v] \leftarrow alt
                                                 \triangleright a quicker path to v has been found
20:
```

```
      21:
      end if

      22:
      end for

      23:
      end while

      24:
      return time[|N|-1]

      25:
      end function
```

Computational cost. See the previous section.

1.3 Solution 3: BFS-like traversal

```
1: function REACH(N, M, A[], B[], appear[], disappear[])
        for i = 0 to M - 1 do
2:
           edges\_[A[i]].push\_back(make\_pair(i, B[i]))
3:
           edges\_[B[i]].push\_back(make\_pair(i,A[i]))
4:
       end for
5:
       for i = 0 to N - 1 do
6:
           done_{-}[i] \leftarrow false
7:
           distance_{-}[i] \leftarrow \infty
8:
        end for
9:
        reached_{-}[0].push_{-}back(0)
10:
        distance\_[0] \leftarrow 0
11:
        for t = 0 to MAX\_TIME do
12:
           for v \in reached_{-}[t] do
13:
               if not done_{-}[v] then
14:
                   for edge \in edges_{-}[v] do
15:
                       staircase \leftarrow edge.first
16:
                       neighbor \leftarrow edge.second
17:
                       time \leftarrow \max(distance[v], appear[staircase]) + 1
18:
                       if not done_[neighbor]
19:
                                            and distance_{-}[v] < disappear[staircase]
20:
                                              and time < distance\_[neighbor] then
21:
                           distance\_[neighbor] \leftarrow time
22:
                           reached\_[time].push\_back(neighbor)
23:
                       end if
24:
25:
                   end for
                   done\_[v] \leftarrow true
26:
               end if
27:
           end for
28:
        end for
29:
        return (distance[N-1] == \infty)? -1 : distance[N-1]
30:
31: end function
```

```
Computational cost: O(m + MAX\_TIME).
```

2 Paletta

Paletta ordering² is a peculiar ordering technique: given a 3-tuple of elements, paletta takes the central element as pivot and swaps the two elements right before and next to it. To make an example:

$$\{3, 2, 1\} \xrightarrow{paletta} \{1, 2, 3\} \tag{1}$$

We now want to develop an algorithm to order any array through paletta ordering with the minimal number of swaps. You should see as not every array can be ordered like this:

$$\{3, , 1\} \xrightarrow{paletta} \{1, 2, 3\} \tag{2}$$

2.1 Solution 1: Split and count-inversions

We should note that the following properties hold:

- 1. Every element can be a pivot, but the first and the last one, as they have respectedly no elements before and after them.
- 2. Every element can be swapped as many times as necessary, but only with elements of the same 2-remainder (numbers in even positions can only be swapped with numbers in even positions, the same holds for odd indexes). More formally $|a| = N, \forall i \in [1, N-2], \neg \exists swap(i, j) : i \mod 2 \neq i \mod j$
- 3. The least number of swaps does not backtrack any element. Formally, let k be the minimal number of swaps applied to an array, backtracks included. By hypothesis, k is minimal, but at least m, m > 0 backtrack swaps have been operated, therefore we found a k' = k m : k' < k, a new minimal number of swaps: contradiction.

Given item 2, we can split our array in two, even and odd numbers, and order them counting the swaps. In our example we'll use mergesort, as it runs in $\log_2 n$, does backtrack elements, and is very well-known. Clearly, given an array, a swap happens when an element is pushed back, pulling the one between its new position and the old one ahead: we can map this behaviour in the merge routine of mergesort: the array merged is able to push back elements from its right pointer to the new array, moving them back of (m-i)+(j-m) positions, where m is the dimension of the current two sub-arrays to merge. Provided that our edited version of mergesort ran successfully on both the even-index and odd-index, we now need to verify if by merging them we obtain an ordered array. Intuitively, the merged array will start with the first element of the even-index arrays, followed by the first of the odd-index array, followed by the second of the even-index array, and so on. To check for these elements is pretty trivial and can be done in linear time. Follows the pseudo-code for the edited version and $snake_check$ function:

 $^{^2} http://didawiki.cli.di.unipi.it/lib/exe/fetch.php/magistraleinformatica/alg2/algo2_16/paletta.pdf$

```
1: function MERGE_WITH_PALETTA(left, right, k):
2:

ightharpoonup merge instructions
3:
      if right > left then
4:
5:
          paletta\_count = paletta\_count + 1
6:
      end if
7:
8: end function
1: function SNAKE_CHECK:
      even, odd = 0
2:
      for ; even, odd < N even = even + 1, odd = odd + 1 do
3:
4:
         if a[even] > a[odd] then
5:
             \mathbf{return} \ \textbf{-} 1
6:
         end if
7:
      end for
       return paletta_count
9: end function
```

 $\textbf{Computational cost: } \Omega(n\log_2(n)), \Theta(n\log_2(n)), O(n\log_2(n)).$

3 Range updates

Consider an array ${\cal C}$ of n integers, initially all equal to zero. We want to support the following operations:

- update(i, j, c): where $0 \le i \le j \le n-1$ and c is an integer: it changes C such that C[k] = C[k] + c for every $i \le k \le j$.
- query(i) where $0 \le i \le n-1$: it returns the value of C[i].
- sum(i,j) where $0 \le i \le j \le n-1$: it returns $\sum_{k=1}^{j} (C[k])$.

Design a data structure that uses O(n) space and implements each operation above in $O(\log(n))$ time. Note that query(i) = sum(i,i) but it helps to reason. [Hint to further save space: use an implicit tree such as the Fenwick tree (see wikipedia).]

3.1 Solution 1: Fenwick lazy a-b sums

Let T be a segmented binary tree over a continuous interval I:[0,N-1] s.t. its leafs are the points in I, and the parent of two nodes comprises of their interval:

$$n' \cup n'' = n, n' \cap n'' = \emptyset$$
 s.t. n is the father of n', n"

T will keep track of the prefix sums for every interval. We define a function

$$s': [0, n-1] \to \mathbb{N} \tag{3}$$

that given a node in T returns the value associated with I, namely the cumulative sum of that interval.

In order to reduce the computational cost, we introduce a lazy algorithm that doesn't propagate sums over T as they are streamed in the input, which means s'(i) might not be accurate at a given time t for any of the requested operation.

We'll instead either compute over T or update T as necessary. Let us define a function to do so:

$$l: \mathbb{N} \to (\mathbb{N} \cup \{\epsilon\}, \mathbb{N}) \tag{4}$$

to keep track of our lazy sums:

$$s(n) = \begin{cases} \epsilon, \text{ _} & \text{if no lazy prefix sum is in that interval} \\ k, m & \text{if a lazy sum of k is to be propagated to m} \end{cases}$$

The *query* function is then trivial:

```
1: function QUERY(I, i, sum):
2:
3: if I.size == 1 then
4: return I.sum
5: end if
```

```
6:
       if lazy(I), i \in I.left, i \neg \in I.right then
                                                                ▷ Lazy on left child
7:
8:
           lazy(I) = False
           query(I.left, i, sum + I.sum)
9:
10:
       end if
11:
       if lazy(I), i \in I.right, i \neg \in I.left then
                                                              ▷ Lazy on right child
12:
           lazy(I) = False
13:
           query(I.right, i, sum + I.sum)
14:
       end if
15:
16:
       if lazy(I), i \in I.right, i \in I.left then
                                                                     ▶ Lazy on both
17:
           lazy(I) = False
18:
           query(I.right, i, j, sum + I.sum) + query(I.left, i, j, sum + I.sum)
19:
       end if
20:
21:
       if !lazy(I), i \in I.left then
                                                            ⊳ Not lazy on left child
22:
           query(I.left, i, sum)
23:
       end if
24:
25:
       if !lazy(I), i \in I.right then
                                                          ⊳ Not lazy on right child
26:
           query(I.right, i, sum)
27:
       end if
28:
29:
       if !lazy(I), i \in I.right, i \in I.left then
                                                                 ▷ Not lazy on both
30:
31:
           sum(I.right, i, sum)
       end if
32:
33: end function
 1: function SUM(I, i, j, sum):
2:
       if I.size == 1 then
                                                              ▷ Return found value
3:
4:
           return I.sum + sum
       end if
5:
6:
       if lazy(I), i \in I.left, i \neg \in I.right then
                                                                ▶ Lazy on left child
7:
           lazy(I) = False
8:
           sum(I.left, i, j, sum + I.sum)
9:
10:
       end if
11:
                                                              \triangleright Lazy on right child
       if lazy(I), i \in I.right, i \neg \in I.left then
12:
           lazy(I) = False
13:
           sum(I.right, i, j, sum + I.sum)
14:
       end if
15:
16:
17:
       if lazy(I), i \in I.right, i \in I.left then
                                                                     ▷ Lazy on both
```

```
lazy(I) = False
18:
           sum(I.right, i, j, sum + I.sum) + sum(I.left, i, j, sum + I.sum)
19:
       end if
20:
21:
22:
       if !lazy(I), i \in I.left then
                                                             ⊳ Not lazy on left child
           sum(I.left, i, sum)
23:
       end if
24:
25:
       if !lazy(I), i \in I.right then
                                                           ▷ Not lazy on right child
26:
           sum(I.right, i, sum)
27:
       end if
28:
29:
       if !lazy(I), i \in I.right, i \in I.left then
                                                                 \triangleright Not lazy on both
30:
           sum(I.right, i, sum)
31:
       end if
32:
33: end function
1: function UPDATE(I, i, j, k):
2:
       if I.size == 1 then
                                                               ▷ Return found value
3:
           return I.val + = update
4:
       end if
5:
6:
       if lazy(I), i \in I.left, i \neg \in I.right then
                                                                 ▷ Lazy on left child
7:
           lazy(I.left) = True
8:
           I.left.val = k
9:
       end if
10:
11:
       if lazy(I), i \in I.right, i \neg \in I.left then
                                                               ▷ Lazy on right child
12:
           lazy(I.right) = True
13:
           I.right.val = k
14:
       end if
15:
16:
       if lazy(I), i \in I.right, i \in I.left then
                                                                     \triangleright Lazy on both
17:
           lazy(I) = True
18:
           I.val = k
19:
       end if
20:
21:
22:
       if !lazy(I), i \in I.left then
                                                             ▷ Not lazy on left child
           update(I.left, i, update)
23:
       end if
24:
25:
       if !lazy(I), i \in I.right then
                                                           ▷ Not lazy on right child
26:
           update(I.right, i, update)
27:
28:
       end if
29:
```

```
30: if !lazy(I), i \in I.right, i \in I.left then \triangleright Not lazy on both 31: update(I.right, i, update) 32: end if 33: end function
```

4 Depth of a node in a random search tree

A random search tree for a set S can be defined as follows: if S is empty, then the null tree is a random search tree; otherwise, choose uniformly at random a key $k \in S$: the random search tree is obtained by picking k as root, and the random search trees on $L = x \in S : x < kandR = x \in S : x > k$ become, respectively, the left and right subtree of the root k. Consider the randomized QuickSort discussed in class and analyzed with indicator variables CLRS 7.3, and observe that the random selection of the pivots follows the above process, thus producing a random search tree of n nodes. Using a variation of the analysis with indicator variables, prove that the expected depth of a node (i.e. the random variable representing the distance of the node from the root) is nearly $2\log_2(n)$. Prove that the probability that the expected depth of a node exceeds $2\log_2(n)$ is small for any given constant c > 1. [Note: the latter point can be solved after we see Chernoff's bounds.]

4.1 Recursive balanced proof

Let n be the number of nodes in the input list l, $h = \log_2(n)$ the height of a balanced tree over l, T(p) the tree built over the permutation p of pivots, d(m) be the positional distance of a value m of a partition from the median value of the said partition. Then the following holds:

- $height(T) = h \iff |T.left| = |T.right| \pm 1$ Trivially, let r be the root of a 3-nodes partition: then, if the partition is unbalanced, the lesser one will comprise of 0 nodes, while the greater one of 2, which implies that height(T.right) == 2.
- P = pivot, $d(m) = \pm k \implies height(T.left) = height(T.right) \pm k$. Recursively from the previous statement, a partition unbalanced of one element generates subtrees whose levels differ on a factor of 1. By iterating recursively, their subtrees, if unbalanced by 1, will yield one more level difference. Over k unbalanced pivots on a single subtree, at most k levels will be added to k.
- By the previous statement, it follows that $\nexists T, T' : height(T) >= height(T')$, T balanced, T' unbalanced. As stated, let T', T be the unbalanced/balanced tree respectively; let us cheat with T and switch the root pivot with the first element in its subtree. Now, let us prove by contradiction that T can't stay balanced and that its height will increase. By shifting the tree to the left we have deprived T.right of either 0 levels (in case T.right is able to switch every pivot in its tree with its right subtree root, ending with the rightmost leaf in its subtree) or 1, in case no rightmost leaf is present. Therefore height(T) <= height(T').
- The completely unbalanced tree is the tree with the most levels. By taking partitions of size 0 we costantly force, at each level, one subtree to

disappear. Therefore, its level(s) has to be necessarly transferred to its brother. We then have exactly one node per level, therefore n levels.

Behaviour on random permutations Now let us analyse how the tree depth varies according to random pivot selection. We start by applying the 4.1k-distance to a tree T with n=3 nodes. Trivially, height(T) with balanced tree is equal to two. Now, let us pick either the lowest or the greatest pivot possible: the tree is unbalanced towards either the left or the right, but height(T)=2 in both cases. As the reader can see from 4.1, the distance works in absolute value; it is then clear how, at every permutation for a pivot p, out of the n, there are 2 that generate a tree of the same height: p=d(P)+k, p=d(P)-k. Given that at every iteration a node x in a completely unbalanced tree T has a probability of $\frac{1}{n-i}$, we can define the probability of x being a pivot at level x as:

$$P(x_k) = \frac{1}{n-l} \tag{5}$$

Now, in order for x not to be chosen as pivot in the previous l-1 levels we have:

$$P(x_k) = \sum_{k=1}^{l-1} \left(\frac{1}{n-l+1}\right) \tag{6}$$

Given the height of T, the (harmonic) partial series converges to $\ln(n)+1$. Now let i be the node in T whose height we just found, and j the node Now, since every node has equal probability of being picked as next pivot, we have that given a node k at height h', the height of $T \leq h' + \max\{h(k.left), h(k.right)\}$. Moreover, being the nodes picked randomly, the probability of each of them to be picked is equal, and the weighted average trivially translates to an arithmetic one.