

# CS301 - HW3

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May 2, 2021

- (a)  $\delta(s, v)$  is the formula to go from  $s$  to  $v$ .

$$\delta(s, s) = 0$$

$$\delta(s, v) = \min(\delta(u, v) + w(s, u)) \text{ where there is an edge between } s \text{ and } u$$

**Subproblems:** Starting from source, program will return the minimum road from neighbours of the source to target. Hence, subproblems are adjacents of the caller.

**Optimal Substructure:** For each subproblem, program will ensure that from source to any intermediary node is the optimal solution. Because, it is not possible to go longer path from  $u$  (an intermediary node between source and target) to target, since it will break the optimality of general problem, as well.

**Overlapping Computations:** When there is edge between at least two of the parents ( $u_1$  and  $u_2$ , let's say) (closer to source from target), one of them will call the other one ( $u_1$  will call  $u_2$ ) and subproblem  $\delta(s, u_2)$  will be solved more than once.

**Topological order:** source,  $u_1, u_2, \dots, u_n, v_{1.1}, \dots, v_{1.n}, v_{2.n}, \dots, v_{2.n}, \dots, target$  ( $u_{1\dots n}$  are children of *source* and  $v_{2.1\dots n}$  are children of  $u_2$ )

Hence, the problem is to find the shortest path from source to target, dividing the problem into subproblems which are children of source.

- (b) **procedure** shortestRoute( $x, visited, target$ ):

**if**  $x$  equals  $target$ :

**return** 0

**if** no neighbour left:

**return** inf

    distlist = [ ]

**for** each *adjacent* and not in *visited*:

**add**  $dist(x-adjacent) + shortestRoute(adjacent, visited, target)$  to *distlist*

**add** *adjacent* to *visited* //to avoid cycles (going back to the parent calls)

**return**  $min(distlist)$

**for** every *city* except the *Istanbul*:

        shortestRoute(*Istanbul*, [ ], *city*)

**Time complexity:**

Target will be every other city than Istanbul =  $\Theta(V)$

From source to target at most  $|V| - 1$  vertices may exist (height of the tree) =  $O(|V|)$

Let  $n$  be the number of calls in each step: In total there will be  $n^{|V|}$  calls.

Thus, time complexity will be  $O(|V| \cdot n^{|V|})$

**Space complexity:**

Every call will have its own distlist which may at most consist of  $|V| - 1$  elements.

Since there is  $n^{|V|}$  calls, complexity will be  $O(|V| \cdot n^{|V|})$

(c) **procedure** shortestRouteDP(*vertices, edges, source*):

distlist = *inf* **for** size of  $|V|$

parent = *null* **for** size of  $|V|$

distlist[source] = 0

**loop**  $|V| - 1$  times:

**for** each *edge*( $u, v$ ) in *edges*:

**if** distlist[u] + weight of edge < distlist[v]:

      distlist[v] = distlist[u] + weight of edge

      parent[v] = u

**return** distlist, parent

**Time complexity:**

Greater loop iterates  $|V| - 1$  times =  $\Theta(|V|)$

Inner loop iterates over each edge =  $\Theta(|E|)$

Thus in total  $\Theta(|V| \cdot |E|)$

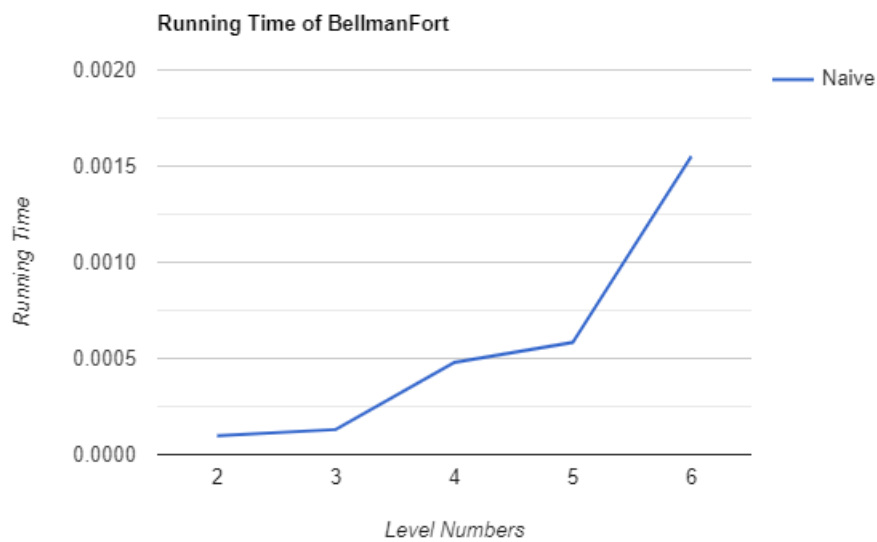
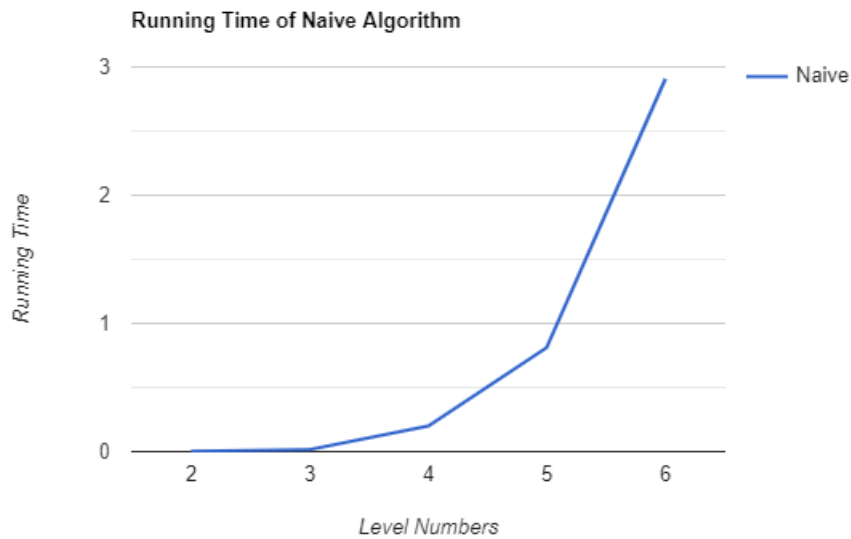
**Space complexity:**

Both distlist and parent has  $|V|$  elements =  $\Theta(|V|)$

(d)

Computer properties: Windows10 OS, 16GB RAM, intel i7-770HQ CPU (2.8 GHz)

| Algorithm   | level = 2 | level = 3 | level = 4 | level = 5 | level = 6 |
|-------------|-----------|-----------|-----------|-----------|-----------|
| Naive       | 0.0031006 | 0.0157071 | 0.199261  | 0.812308  | 2.9061138 |
| BellmanFort | 0.0000974 | 0.0001302 | 0.0004793 | 0.0005832 | 0.0015508 |



As expected, naive algorithm grows exponentially, even though the input size is small, the exponential trend can be observed.

On the other hand, BellmanFort is expected to show a trend between linearly and quadratic, however its data plots are quite messy. It can be steemed from the small input sizes. Even though, it is scattered (huge jump) from 5 to 6, if a trend is drawn, it will be near to a quadratic line.