

# ADA Assignment 2

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## 1 City of Computopia

### 1.1 Problem

The police department in the city of Computopia has made all the streets one-way. But the mayor of the city still claims that it is possible to legally drive from one intersection to any other intersection.

- (a) Formulate this problem as a graph theoretic problem and explain why it can be solved in linear time.
- (b) Suppose it was found that the mayor's claim was wrong. She has now made a weaker claim: "if you start driving from town-hall, navigating one-way streets, then no matter where you reach, there is always a way to drive legally back to the town-hall. Formulate this weaker property as a graph theoretic problem and explain how it can be solved in linear time.

### 1.2 Solution Part (a)

#### 1 Formulation as a graph problem

Consider the initial city as a graph  $G(V, E)$  with intersections as nodes and streets as edges between them as shown in Fig 1.

More formally, any intersection  $\mathbf{int}_i \in V$ , and  $(\mathbf{int}_i, \mathbf{int}_j) \in E$  there is a road connecting intersections  $\mathbf{int}_i$  and  $\mathbf{int}_j$

Now, the police turning all the streets to one-way means converting all the undirected edges into directed ones. as shown in Fig 2

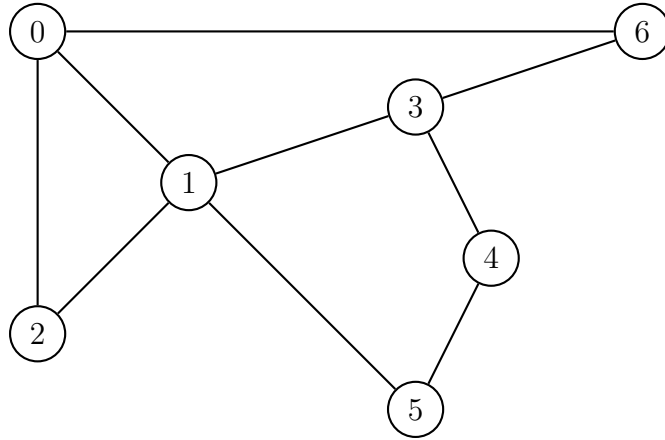


Figure 1: two-way streets

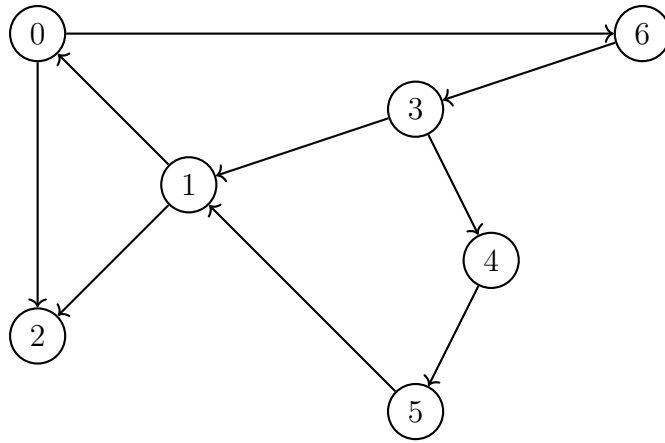


Figure 2: One way streets instead of 2 way

## 2 Linear time solution

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**Algorithm 1:** All intersections reachable

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```

1 procedure connectedGraph( $G$ )  $\rightarrow$  boolean
  /* Returns true if mayor's claim is true, i.e. graph is
   strongly connected. Otherwise, returns false */
2  visited  $\leftarrow$  boolean array of size  $|V(G)|$ , and initialize all to false
3  pick any arbitrary node  $\text{int}_i \in V(G)$ 
4  perform DFS with  $\text{int}_i$  as the starting node
5  set  $\text{vis}_j = \text{true}$  if  $\text{int}_j$  is visited in the DFS  $\forall \text{int}_j \in V(G)$ 
6  for  $\text{vis}_i$  in visited do
7    if  $\text{vis}_i = \text{false}$  then
8      return false // In case any node is not visited
9    end
10 end
11 return true
12 end

```

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The main idea behind the algorithm is to check if the graph is strongly connected or not (implying that all intersections are reachable). It takes  $\mathcal{O}(|V| + |E|)$  operations to perform DFS on a graph, and  $\mathcal{O}(|V|)$  further operations to check if all nodes have been visited. Therefore, overall time complexity of the algorithm is  $\mathcal{O}(|V| + |E|)$  i.e. linear.

### 1.3 Solution Part (b)

#### 1 Formulation as a graph problem

The formulation is the same as part (a) when converting all the streets into one-way roads.

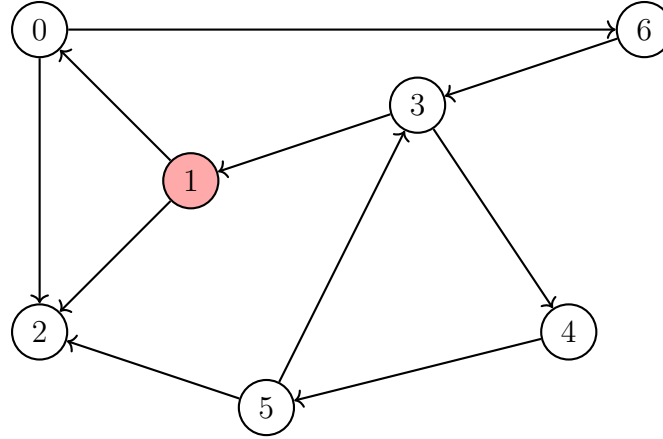


Figure 3: Cycle in a directed graph, Node 1 is the starting node or town hall

#### 2 Linear time solution

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**Algorithm 2:** All intersections reachable

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```

1 procedure connectedGraph( $G$ )  $\rightarrow$  boolean
2   /* Returns true if there is a path from town hall to itself, false otherwise */
3   Convert  $G$  into a condensation graph using Kosaraju's algorithm.
4   Let the new graph be  $C(V, E)$ 
5   Let  $U \in V(C)$  be of the form  $U = \{u_1, u_2, \dots, u_k\}$  where  $u_i \in V(G)$ 
6   Let  $T \in V(C)$  be the condensation node which contains the town-hall
7   for  $U \in V(C)$  do
8     for  $u \in U$  do
9       | set label( $u$ ) =  $U$ 
10    end
11  end
12  for  $(u_1, u_2) \in E(G)$  do
13    | if label( $u_1$ ) =  $T \wedge$  label( $u_2$ )  $\neq T$  then
14      |   return false
15    | end
16  end
17  return true

```

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## 2 Smallest weight in a cycle

### 2.1 Problem

Given an edge-weighted connected undirected graph  $G = (V, E)$  with  $n + 20$  edges. Design an algorithm that runs in  $\mathcal{O}(n)$ -time and outputs an edge with the smallest weight contained in a cycle of  $G$ . You must give a justification for why your algorithm works correctly

### 2.2 Solution

#### 1 Linear Time Algorithm

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**Algorithm 3:** Smallest weight contained in a cycle of  $G$

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```
// global variables
1  $min_w \leftarrow \infty$ 
2  $visited \leftarrow$  boolean array of size  $|V(G)|$ , and initialize all to false
3  $curr_{min} \leftarrow \infty$ 
4
5 procedure parentDFS( $v$ )
6   set  $vis_v = \text{true}$ 
7
8   for  $w$  in  $Adj(v)$  do
9     if  $vis_w = \text{false}$  then
10        $parent(w) = v$ 
11        $curr_{min} = \min(curr_{min}, (w, v))$  // (u,v) is the weight of the
12       Run parentDFS( $w$ )
13     end
14
15     if  $vis_v = \text{true} \wedge parent(v) \neq u$  then
16        $min_w = \min(min_w, curr_{min})$ 
17        $curr_{min} \leftarrow \infty$ 
18     end
19   end
20 end
21
22 procedure smallestCycleEdge( $G$ )  $\rightarrow$  int
23   /* Returns smallest weight contained in a cycle of  $G$ .
24   Returns  $\infty$  if there is no cycle in  $G$  */
25   consider any arbitrary vertex  $u$  as the starting node
26   Run parentDFS( $u$ )
27   return  $min_w$ 
28 end
```

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The main idea behind Algorithm 3 is to keep track of current minimum edge, and then update the global minimum edge when it detects a loop.

## 2 Correctness Proof

**Assumption:** A cycle contains 3 or more vertices, i.e. no self-loops and edge being considered as loop.

Any undirected connected graph  $G = (V, E)$  can be cyclic or acyclic.

**Case 1: Acyclic** - In this case, it will never be possible for a vertex  $v \in V(G)$  to be both visited and have its parent not be the current node, i.e.  $u \in V(G)$ .

1. If  $v$  is not visited, then its parent will be set to  $u$  and marked visited.
2. If  $v$  is visited, its parent must be  $u$ . Assume by contradiction that  $v$  is visited, and its parent is  $w \in V(G)$ . This implies a cycle since we know a path from  $u$  to  $w$  (the graph is known to be connected). We have just encountered the edge  $(u, v) \because v \in \text{Adj}(u)$ , and  $\text{parent}(v) = w$ , so there is also an edge  $(v, w)$ , thus completing the cycle.

Thus, the value of  $\min_w$  remains  $\infty$  implying there is no cycle in the graph and therefore no minimum edge

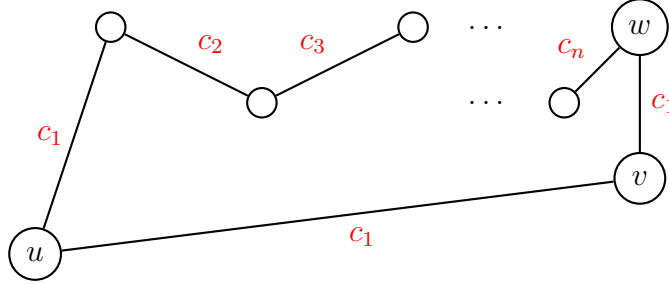


Figure 4: Cycle condition

**Case 2: Cyclic** If there is a cycle, there will always be a case where  $v$  is visited and  $\text{parent}(v) \neq u$ . There will always be a node  $u$  with a path from it to itself, as shown in Fig 3. We always maintain the minimum edge along the cycle and update it with  $\min_w$  when we reach the cycle's starting point, giving us the smallest edge weight in all cycles of  $G$ .

## 3 Time Complexity

## 3 Probability of reaching sink

### 3.1 Problem

Suppose that  $G$  be a directed acyclic graph with the following features

- $G$  has a single source  $s$  and several sinks  $t_1, \dots, t_k$
- Each edge  $(v \rightarrow w)$  (i.e. an edge directed from  $v$  to  $w$ ) has an associated weight  $Pr(v \rightarrow w)$  between 0 and 1.
- For each non-sink vertex  $v$ , the total weight of all the edges leaving  $v$  is

$$\sum_{(v \rightarrow w) \in E} Pr(v \rightarrow w) = 1$$

The weights  $Pr(v \rightarrow w)$  define a random walk in  $G$  from the source  $s$  to some sink  $t_i$ ; after reaching any non-sink vertex  $v$ , the walk follows the edge  $v \rightarrow w$  with probability  $Pr(v \rightarrow w)$ .

All the probabilities are mutually independent. Describe and analyze an algorithm to compute the probability that this random walk reaches sink  $t_i$  for every  $i \in 1, \dots, k$ . You can assume that an arithmetic operation takes  $\mathcal{O}(1)$  time.

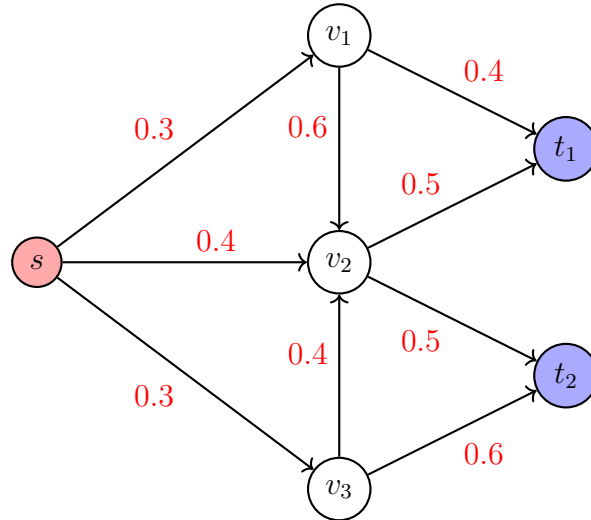


Figure 5: An Example Network Flow Graph. Probabilities are mentioned on every edge. The source vertex is coloured red, and sinks are coloured blue

### 3.2 Solution

#### 1 Sub-problem

Probability of reaching the sink  $t_i$  from vertex  $v$ , denoted as  $P(v, t_i)$ .

#### 2 Recurrence Relation

**Base Case:**  $P(t_i, t_i) = 1$ , and  $P(v, t_i) = 0 \forall v \neq t_i$

## Recurrence Relation :-

$$P(v, t_i) = \sum_{(v \rightarrow w) \in E} Pr(v \rightarrow w) \times P(w, t_i)$$

In other words, the probability of reaching from  $v$  to  $t_i$  is the sum of the probability of going from  $v$  to  $w$  times the probability of reaching from  $w$  to  $t_i$

### 3 Subproblem which gives solution

$P(s, t_i)$ , which gives the probability of reaching sink  $t_i$  from  $s$

### 4 Algorithm

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**Algorithm 4:** Probability of reaching sinks

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```
1 procedure computeProb( $G, u, v, P$ )
2   if  $u = v = t_i$  then
3     return
4   end
5   curr_val  $\leftarrow 0$  // for storing current value of  $P(u, v)$ 
6   for  $w \in \text{Adj}(u)$  do
7     if  $P(w, v) = -1$  then
8       Run computeProb( $G, w, v, P$ )
9     end
10    curr_val +=  $Pr(u \rightarrow w) \times P(w, v)$ 
11  end
12   $P(u, v) = \text{curr\_val}$ 
13 end
14
15 procedure pSinks( $G$ )  $\rightarrow$  array
16   /* Returns an array pathProb where pathProb[i] is probability of reaching sink  $t_i$ 
17    from source  $s$ . Assuming the graph has  $n$  sinks  $t_1, t_2, \dots, t_n$  */
18    $P \leftarrow$  2-D array to store value of  $P(u, v) \forall u, v \in V(G)$ 
19   pathProb  $\leftarrow$  array to store value of  $P(s, t_i)$ 
20   Initialize all values in pathProb to 0 and in  $P$  to -1
21
22   for  $i = 1$  to  $n$  do
23      $P(t_i, t_i) = 1$ 
24   end
25   for  $i = 1$  to  $n$  do
26     Run computeProb( $G, s, t_i, P$ )
27     pathProbs[i] =  $P(s, t_i)$ 
28   end
29   return pathProbs
30 end
```

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## 5 Time Complexity

For every sink  $t_i$ , We will traverse through the entire graph in a DFS-type manner and then backtrack to fill the values of the edges already visited, as can be seen in the recursive nature of the algorithm. Therefore, the time complexity for finding  $P(s, t_i)$  is  $\mathcal{O}(|V| + |E|)$

If there are  $n$  such sinks, then the algorithm will be repeated  $n$  times for each sink. Therefore, total time complexity of the algorithm is  $\mathcal{O}(n(|V| + |E|))$