

# CO 353 COURSE NOTES

COMPUTATIONAL DISCRETE OPTIMIZATION

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# 1 Shortest Paths

## 1.1 Preliminaries on Graphs

An **(undirected) graph**  $G$  is a pair  $(V, E)$ , where  $E$  is a set of unordered pairs of elements in  $V$ . The elements of  $V$  are called **vertices** or **nodes**; the elements of  $E$  are called **edges**.

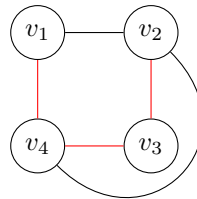
Let  $u, v \in V$  and let  $e = uv \in E$  be an edge.

- We say that  $e$  is **incident** to  $u$  and  $v$ .
- The vertices  $u$  and  $v$  are said to be **adjacent**.
- We call  $u$  and  $v$  the **endpoints** of  $e$ .

By default, we assume that there are no parallel edges (i.e. two edges  $e = uv$  and  $e' = u'v'$  in  $E$  with  $\{u, v\} = \{u', v'\}$ ) and no loops (i.e. an edge  $e = uv \in E$  with  $u = v$ ).

For distinct  $u, v \in V$ , a  $u, v$ -**path** is a sequence of vertices  $w_1, \dots, w_k$  such that  $w_1 = u$ ,  $w_k = v$ , and  $w_i w_{i+1} \in E$  for all  $i = 1, \dots, k-1$ .

For example, consider the following graph  $G = (V, E)$  with vertices  $V = \{v_1, v_2, v_3, v_4\}$  and edges  $E = \{v_1 v_2, v_1 v_4, v_2 v_3, v_2 v_4, v_3 v_4\}$ .



The lines in red form a  $v_1, v_2$ -path, namely  $v_1, v_4, v_3, v_2$ . Another  $v_1, v_2$ -path can be obtained by simply traversing the edge  $v_1 v_2$ .

A **cycle** in  $G$  is a sequence of vertices  $w_1, \dots, w_{k+1}$  such that  $w_i w_{i+1} \in E$  for all  $i = 1, \dots, k$ , the vertices  $w_1, \dots, w_k$  are all distinct, and  $w_1 = w_{k+1}$ .

Finally, a graph  $G$  is **connected** if for any pair of distinct vertices  $u, v \in V$ , there exists a  $u, v$ -path in  $G$ .

## 1.2 Shortest Paths Problem

Given a *directed* graph  $G = (V, E)$  with edge lengths  $\ell_e \geq 0$  for each  $e \in E$  and a distinguished start vertex  $s \in V$ , we wish to find shortest paths from  $s$  to every other vertex in  $V$ . Note that when we work with directed graphs, we will denote the directed edges with  $(v_1, v_2)$  as opposed to  $v_1 v_2$  in the case of undirected graphs, where the order of the vertices did not matter.

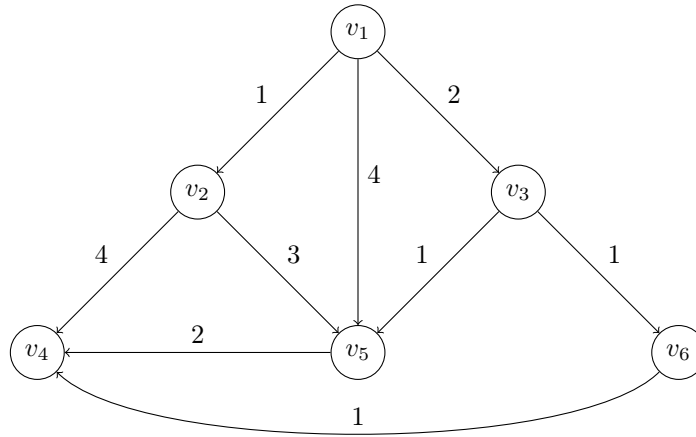
The **length** of a path  $P$  given by the sequence  $w_1, \dots, w_k$  is given by

$$\ell(P) := \sum_{i=1}^{k-1} \ell_{(w_i, w_{i+1})} = \sum_{e \in P} \ell_e,$$

where the second sum makes sense because there are no parallel edges. Then the **shortest-path distance** from  $s$  to a vertex  $u \in V$  is defined to be

$$d(u) := \min_{s, u\text{-paths } P} \ell(P).$$

For example, we can consider the following instance of an undirected graph with given edge lengths and starting vertex  $s = v_1$ .



In this case, we have  $d(v_2) = 1$ , since the only possible path from  $v_1$  to  $v_2$  is by taking the edge  $(v_1, v_2)$ . There are multiple paths from  $v_1$  to  $v_5$ ; the shortest one is  $v_1, v_3, v_5$  giving  $d(v_5) = 3$ .

Note that we always set  $d(s) = 0$ . We now make some observations:

- (i) If  $(u, v) \in E$ , then  $d(v) \leq d(u) + \ell_{(u,v)}$ , since such an  $s, v$ -path is always an option.
- (ii) For every  $v \in V$  distinct from  $s$ , there exists  $w \in V$  such that  $d(v) = d(w) + \ell_{(w,v)}$  and  $(w, v) \in E$ . This can be seen by chopping off the last edge from a shortest path from  $s$  to  $v$ .

### 1.3 Dijkstra's Algorithm

In 1959, Dijkstra came up with the following algorithm to solve the shortest paths problem. The main idea is to maintain a set  $A \subseteq V$  of "explored" nodes; that is, a set of nodes for which we already know the shortest-path distances. We'll also maintain labels  $d'(v)$  for  $v \in V \setminus A$  with upper bounds on the shortest-path distances from  $s$ .

**Input.** A directed graph  $G = (V, E)$ , edge lengths  $\ell_e \geq 0$  for all  $e \in E$ , and a start vertex  $v \in V$ .

**Output.** For all  $v \in V$ , the length  $d(v)$  for the shortest-path from  $s$  to  $v$ .

Step 1. **Initialization.**  $A \leftarrow \{s\}$ ,  $d(s) \leftarrow 0$ , and  $d'(v) \leftarrow \infty$  for all  $v \in V \setminus A$ .

Step 2. While  $A \neq V$ :

Step 2.1. **Push down the upper bounds.** For each  $v \in V \setminus A$ , compute

$$d'(v) \leftarrow \min \left\{ d'(v), \min_{\substack{u \in A \\ (u,v) \in E}} \{d(u) + \ell_{(u,v)}\} \right\}.$$

Step 2.2. **Add a new vertex.** Set  $w \leftarrow \arg \min_{v \in V \setminus A} d'(v)$ ,  $A \leftarrow A \cup \{w\}$ , and  $d(w) \leftarrow d'(w)$ .

Suppose that for each vertex  $w \in V$ , we keep track of the node  $u$  determining its upper bound  $d'(w)$ . That is, the node  $u$  is such that  $(u, w) \in E$  and  $d'(w) = d(u) + \ell_{(u,w)}$ . Then at the end of the algorithm, a shortest path from  $s$  to  $w$  can be obtained as a shortest path from  $s$  to  $u$  adjoined with the edge  $(u, w) \in E$ . Moreover, these edges selected by Dijkstra's algorithm form an arborescence, which is a nice graph structure that we'll discuss more later.

Next, let's prove the correctness of Dijkstra's algorithm. In particular, we need to show that for every  $v \in V$ , the distance from  $s$  to  $v$  is computed correctly. We'll assume that the graph is connected; that is, for every  $v \in V$ , there is an  $s, v$ -path in  $G$ . (Note that the algorithm won't terminate otherwise, but it can be adjusted to deal with this.)

### Proof of correctness of Dijkstra's algorithm.

We proceed by induction on  $|A|$ , and show that at each point in time,  $d(v)$  is computed correctly for all  $v \in A$ . The case where  $|A| = 1$  is clear because at the start of the algorithm, we initialize  $A = \{s\}$  with  $d(s) = 0$ , which is correct.

Assume that  $d(v)$  is computed correctly for every  $v \in A$  when that  $|A| = k$ . Suppose that we are adding a new vertex  $w$  to  $A$  in Step 2.2 of the algorithm. Consider the vertex  $u \in A$  such that  $(u, w) \in E$  and

$$d'(w) = d(u) + \ell_{(u,w)}.$$

Specifically, this is the vertex  $u$  determining the upper bound  $d'(w)$  which we discussed in the paragraph following the description of the algorithm.

For the sake of contradiction, assume that the distance from  $s$  to  $w$  is not  $d'(w)$ . Let  $P_u$  be a shortest path from  $s$  to  $u$ , and let  $P'$  be a shortest path from  $s$  to  $w$ . Then by our assumption, we know that

$$\ell(P') < \ell(P_u) + \ell_{(u,w)} = d'(w).$$

Now, let  $x, y \in V$  be such that  $(x, y) \in E$  lies on the shortest path  $P'$  from  $s$  to  $w$ , with  $x \in A$  and  $y \in V \setminus A$ . (This exists because at some point, the path must exit  $A$  to get from  $s$  to  $w$ .) Then we obtain

$$d'(y) \leq d(x) + \ell_{(x,y)} \leq \ell(P') < \ell(P_u) + \ell_{(u,w)} = d'(w),$$

where the first inequality is because of how  $d'(y)$  is computed in Step 2.1, and the second inequality is because the shortest path from  $x$  to  $y$  adjoined with the edge  $(x, y)$  is part of the path  $P'$ , noting that  $\ell_e \geq 0$  for all  $e \in E$ . But this contradicts our choice of  $w = \arg \min_{v \in V \setminus A} \{d'(v)\}$  in Step 2.2 since  $y \in V \setminus A$  but  $d'(y) < d'(w)$ .  $\square$