

## 1 Outline

In this lecture, we study

- the shortest path problem,
- maximum  $st$ -flow,
- minimum  $st$ -cut,
- the max-flow min-cut theorem,
- bipartite matching.

## 2 Shortest path problem

Given a directed graph  $D = (N, A)$  and two distinct nodes  $s, t \in N$ , a **(directed)  $st$ -path** is a sequence of nodes  $v_0, v_1, \dots, v_\ell$  such that

- $v_0 = s$  and  $v_\ell = t$ ,
- $v_0, \dots, v_\ell$  are distinct nodes,
- $(v_{i-1}, v_i) \in A$  for  $i = 1, \dots, \ell$ .

Here we often call  $s$  the **origin node** and  $t$  the **destination node**. We can define an  $st$ -path with arcs. A directed  $st$ -path can be defined as a sequence of arcs  $a_1, \dots, a_\ell$  such that

- $a_i = (v_{i-1}, v_i)$  for  $i = 1, \dots, \ell$  for some nodes  $v_0, \dots, v_\ell$ ,
- $v_0 = s$  and  $v_\ell = t$ ,
- $v_0, \dots, v_\ell$  are distinct nodes.

In general, a (directed) path is an  $st$ -path where  $s$  and  $t$  are the first and the last nodes in the path. Let  $c_{uv}$  be the length of arc  $(u, v) \in A$ . Then the length of a path  $P$  is

$$\sum_{(u,v) \in P} c_{uv}$$

where  $(u, v) \in P$  means that arc  $(u, v)$  is on the path  $P$ . Now the problem is to find a shortest  $st$ -path, that is, a directed  $st$ -path of the minimum length.

We will show that the problem of finding a shortest  $st$ -path can be posed as an instance of the minimum cost flow problem. Let  $x_{uv} \in \{0, 1\}$  denote the variable for arc  $(u, v)$  to indicate whether

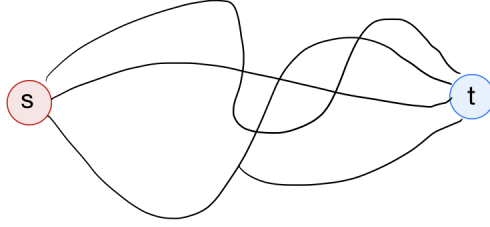


Figure 13.1:  $st$ -path as a unit flow

arc  $(u, v) \in A$  is chosen to be part of my path. Then we may look at  $x \in \{0, 1\}^A$  whose components correspond to the arc set  $A$ . If  $x$  corresponds to the arc set of an  $st$ -path, then

$$\sum_{(u,v) \in A} c_{uv} x_{uv}$$

would be the length of the path.

When does a 0,1 vector  $x \in \{0, 1\}^A$  correspond to an  $st$ -path? Observe the following.

- The origin node  $s$  has an outgoing arc on the path. No other arc of the path is incident to  $s$ . We may model this as

$$\sum_{j \in N: (s,j) \in A} x_{sj} - \sum_{k \in N: (k,s) \in A} x_{ks} = 1.$$

- The destination node  $t$  has an incoming arc on the path. No other arc of the path is incident to  $t$ . We may model this as

$$\sum_{j \in N: (t,j) \in A} x_{tj} - \sum_{k \in N: (k,t) \in A} x_{kt} = -1.$$

- Let  $i \in N \setminus \{s, t\}$ . If  $i$  is on the path, then  $i$  has an incoming arc and an outgoing arc on the path. No other arc is incident to  $i$ . If  $i$  is not on the path, then no arc of the path is incident to  $i$ . This implies that the number of arcs going into  $i$  and the number of arcs going out of  $i$  are the same. This can be modeled as

$$\sum_{j \in N: (i,j) \in A} x_{ij} - \sum_{k \in N: (k,i) \in A} x_{ki} = 0.$$

Therefore, an  $st$ -path can be viewed as the source node  $s$  sending one unit of flow to the sink node  $t$ . More precisely, the origin node  $s$  has supply 1, and the destination node  $t$  has demand 1. The other nodes have 0 net supply, meaning that they are transshipment nodes. Then the problem can

be formulated as

$$\begin{aligned}
 \min \quad & \sum_{(u,v) \in A} c_{uv} x_{uv} \\
 \text{s.t.} \quad & \sum_{j \in N: (s,j) \in A} x_{sj} - \sum_{k \in N: (k,s) \in A} x_{ks} = 1 \\
 & \sum_{j \in N: (t,j) \in A} x_{tj} - \sum_{k \in N: (k,t) \in A} x_{kt} = -1 \\
 & \sum_{j \in N: (i,j) \in A} x_{ij} - \sum_{k \in N: (k,i) \in A} x_{ki} = 0, \quad \forall i \in N \setminus \{s, t\} \\
 & x_{ij} \in \mathbb{Z}_+, \quad \forall (i, j) \in A
 \end{aligned}$$

The formulation is an instance of the minimum cost flow formulation. Therefore, solving this linear program will return a solution  $x^*$  that has integer entries only, which corresponds to a shortest  $st$ -path.

A **directed cycle** is a sequence of nodes  $v_0, v_1, \dots, v_\ell$  such that

- $v_0, \dots, v_{\ell-1}$  are distinct nodes,
- $v_\ell = v_0$ .

**Remark 13.1.** If  $D$  contains a directed cycle of negative length, then the linear program is unbounded. If  $D$  contains no directed cycle of negative length, then the linear program would have an optimal solution.

### 3 Maximum $st$ -flow

The minimum cost flow model we learned does not have a designated source or a sink. In this section, we discuss a network flow model with a **sink node** and a **source node**. Let  $s$  and  $t$  be the source node and the sink node, respectively. The source node  $s$  sends flows, and the sink node receives the flows sent by the source. The other nodes are transshipment node, meaning that the other nodes have 0 net supply. Each arc in the given network has an upper bound on the amount of flows that it can take, i.e.

$$0 \leq x_{ij} \leq c_{ij}, \quad (i, j) \in A.$$

Then the problem is to compute the maximum amount of flows that the source node  $s$  can send to the sink node  $t$  while obeying the flow capacities of arcs.

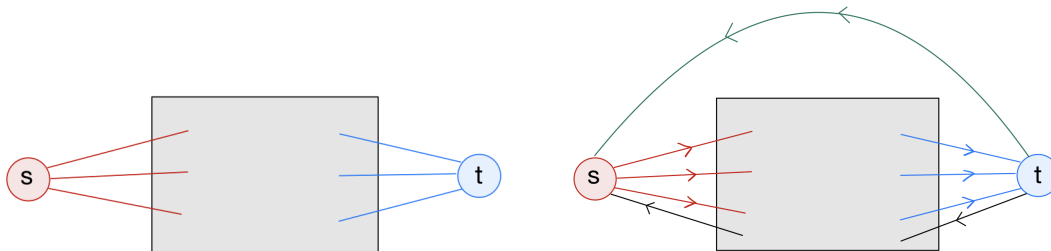


Figure 13.2: Sending flow from  $s$  to  $t$

Although this problem seems different from the minimum cost flow problem, we may formulate the problem as a min cost flow model. The common trick is to add a dummy arc from the sink node  $t$  to the source node  $s$ . This dummy arc  $(t, s)$  sends back all the flows coming from  $s$  to  $t$ . Basically, we impose that

$$x_{ts} = \underbrace{\sum_{k \in N: (k,t) \in A} x_{kt} - \sum_{j \in N: (t,j) \in A} x_{tj}}_{\text{the net amount of flows into } t}.$$

Moreover,  $A' = A \cup \{(t, s)\}$  is the arc set of the new network obtained after adding the dummy arc  $(t, s)$ . Then

$$\begin{aligned} 0 &= x_{ts} + \sum_{j \in N: (t,j) \in A} x_{tj} - \sum_{k \in N: (k,t) \in A} x_{kt} \\ &= \underbrace{\sum_{j \in N: (t,j) \in A'} x_{tj} - \sum_{k \in N: (k,t) \in A'} x_{kt}}_{\text{the net amount of flows into } t \text{ in the new network}}. \end{aligned}$$

Furthermore, the amount of flows that the sink node  $t$  receives is equal to the amount of flows that the source node  $s$  sends out. Hence, we have

$$\underbrace{\sum_{j \in N: (s,j) \in A} x_{sj} - \sum_{k \in N: (k,s) \in A} x_{ks}}_{\text{the net amount of flows out of } s} = \sum_{k \in N: (k,t) \in A} x_{kt} - \sum_{j \in N: (t,j) \in A} x_{tj} = x_{ts}$$

Then it follows that

$$\begin{aligned} 0 &= \sum_{j \in N: (s,j) \in A} x_{sj} - \sum_{k \in N: (k,s) \in A} x_{ks} - x_{ts} \\ &= \underbrace{\sum_{j \in N: (s,j) \in A'} x_{sj} - \sum_{k \in N: (k,s) \in A'} x_{ks}}_{\text{the net amount of flows out of } s \text{ in the new network}}. \end{aligned}$$

The other nodes in the network are transshipment nodes and are not connected to the dummy arc  $(t, s)$ , so we have

$$\sum_{j \in N: (i,j) \in A'} x_{ij} - \sum_{k \in N: (k,i) \in A'} x_{ki} = 0, \quad i \in N \setminus \{s, t\}.$$

Then the problem can be formulated as

$$\begin{aligned} \max \quad & x_{ts} \\ \text{s.t.} \quad & \sum_{j \in N: (i,j) \in A \cup \{(t,s)\}} x_{ij} - \sum_{k \in N: (k,i) \in A \cup \{(t,s)\}} x_{ki} = 0, \quad \forall i \in N \\ & 0 \leq x_{ij} \leq c_{ij}, \quad \forall (i, j) \in A. \end{aligned}$$

Observe that the dummy arc  $x_{ts}$  is a free variable, which is equivalent to  $-\infty \leq x_{ts} \leq +\infty$ . As this formulation is an instance of the minimum cost flow model, it returns an integer flow as long as the capacities  $c_{ij}$  for  $(i, j) \in A$  are integers.

## 4 Minimum $st$ -cut

A **directed  $st$ -cut** is a set of arcs of the form

$$\delta^+(S) = \{(u, v) \in A : u \in S, v \notin S\}$$

where  $S \subseteq N$  contains  $s$  but not  $t$ . In words,  $\delta^+(S)$  is the set of arcs going out of the node set  $S$ . Given arc weights  $c_{uv}$  for  $(u, v) \in A$ , the weight of an  $st$ -cut  $\delta^+(S)$  is given by

$$\sum_{(u,v) \in \delta^+(S)} c_{uv}.$$

Then the minimum  $st$ -cut problem is to find an  $st$ -cut whose weight sum is minimized.

We may formulate the minimum  $st$ -cut problem as an integer program.

- For nodes  $s$  and  $t$ , we assign integer variables  $y_s, y_t \in \mathbb{Z}_+$  that satisfy

$$y_s = 0 \quad \text{and} \quad y_t = 1.$$

Then  $y_s$  and  $y_t$  satisfy

$$y_t - y_s = 1.$$

- For  $i \in N \setminus \{s, t\}$ , we assign an integer variable  $y_i \in \mathbb{Z}_+$ . We set

$$y_i = \begin{cases} 0, & \text{if } i \in S, \\ 1, & \text{if } i \notin S. \end{cases}$$

- For each arc  $(i, j) \in A$ , we assign variable  $z_{ij}$  to indicate whether arc  $(i, j)$  is part of the  $st$ -cut. We can model this by adding

$$z_{ij} \geq y_j - y_i.$$

Here, if  $y_j = 1$  and  $y_i = 0$ , then  $i \in S$  and  $j \notin S$ , which implies that  $(i, j)$  is part of  $\delta^+(S)$ .

Then we deduce

$$\begin{aligned} \min \quad & \sum_{(i,j) \in A} c_{ij} z_{ij} \\ \text{s.t.} \quad & z_{ij} \geq y_j - y_i, \quad (i, j) \in A \\ & y_t - y_s = 1 \\ & z_{ij} \in \mathbb{Z}_+, \quad (i, j) \in A. \end{aligned}$$

Although  $z_{ij}$  can take an arbitrary nonnegative integer, if the arc lengths  $c_{ij}$  for  $(i, j) \in A$  are all nonnegative, then an optimal solution would have  $z_{ij} \in \{0, 1\}$  for  $(i, j) \in A$ .

## 5 The max-flow min-cut theorem

Let us rewrite the LP relaxation of the integer program for solving the minimum  $st$ -cut problem.

$$\begin{aligned}
 \min \quad & \sum_{(i,j) \in A} c_{ij} z_{ij} \\
 \text{s.t.} \quad & y_i - y_j + z_{ij} \geq 0, \quad (i,j) \in A \\
 & y_t - y_s = 1 \\
 & z_{ij} \geq 0, \quad (i,j) \in A.
 \end{aligned}$$

In fact, the dual of this linear program is precisely

$$\begin{aligned}
 \max \quad & x_{ts} \\
 \text{s.t.} \quad & \sum_{j \in N: (i,j) \in A \cup \{(t,s)\}} x_{ij} - \sum_{k \in N: (k,i) \in A \cup \{(t,s)\}} x_{ki} = 0, \quad \forall i \in N \\
 & 0 \leq x_{ij} \leq c_{ij}, \quad \forall (i,j) \in A,
 \end{aligned}$$

which is the linear program for solving the maximum  $st$ -flow problem. Based on the duality relationship, we may observe the following.

1. The constraint matrix for the maximum  $st$ -flow formulation is totally unimodular. Then the transpose of it is the constraint matrix for the minimum  $st$ -cut formulation. This implies that the LP relaxation of the minimum  $st$ -cut formulation has an optimal solution that has integer entries only, which corresponds to a minimum  $st$ -cut.
2. The primal optimum is the minimum weight of an  $st$ -cut, and the dual optimum is the maximum amount of an  $st$ -flow. By strong LP duality, the two values are equal.

**Theorem 13.2** (The max-flow min-cut theorem). *Let  $D = (N, A)$  be a directed graph with arc weights  $c_{ij} \in \mathbb{R}_+$  for  $(i, j) \in A$ , and let  $s, t \in N$  be two distinct nodes. Then the maximum amount of an  $st$ -flow is equal to the minimum weight of an  $st$ -cut.*