

**In the name of God**

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# **Network Flows**

## **5. Minimum Cost Flow Problem**

### **5.1 Introduction**

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# Outline

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- Introduction
- Cycle Canceling Algorithm
- Successive Shortest Path Algorithm

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# **Introduction**

# Introduction

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- In the previous chapters, we have considered two special cases of *minimum cost flow problem*
  - *the shortest path problem*
  - *the maximum flow problem*
- These problems address different components of the overall minimum cost flow problem
  - *Shortest path problems* consider arc flow costs but no flow capacities;
  - *Maximum flow problems* consider capacities but only the simplest cost structure.
- The *minimum cost flow problem* combines these problem ingredients

# Introduction

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- It is easy to understand the basic nature of *shortest path* and *maximum flow problems* and to develop core algorithms for solving them;
  - nevertheless, designing and analyzing efficient algorithms is a very challenging task, requiring considerable ingenuity and considerable insight concerning both basic algorithmic strategies and their implementations.
- The algorithms for solving the *minimum cost flow problem* are not as efficient as those for the *shortest path* and *maximum flow problems*

# Notation and Assumptions

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- $G = (N, A)$  : a directed network with a cost  $c_{ij}$  and a capacity  $u_{ij}$  associated with every arc  $(i, j) \in A$ .
- $b(i)$  : a number  $b(i)$ ,  $i \in N$ , indicates its supply or demand depending on whether  $b(i) > 0$  or  $b(i) < 0$
- $x_{ij}$  : amount shipped on arc  $(i, j)$
- $C$  : the largest magnitude of any arc cost.
- $U$  : the largest magnitude of any supply/demand or finite arc capacity.
- $l_{ij}$  : the lower bounds on arc flows which are all zero.

# Notation and Assumptions

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- The *minimum cost flow problem* can be stated as follows:

$$\text{Minimize } z(x) = \sum_{(i,j) \in A} c_{ij} x_{ij}$$

subject to

$$\sum_{\{j:(i,j) \in A\}} x_{ij} - \sum_{\{j:(j,i) \in A\}} x_{ji} = b(i) \quad \text{for all } i \in N,$$

$$0 \leq x_{ij} \leq u_{ij} \quad \text{for all } (i, j) \in A.$$

# Notation and Assumptions

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- We further make the following assumptions:
- *Assumption 1. All data (cost, supply/demand, and capacity) are integral.*
  - this assumption is not really restrictive in practice because computers work with rational numbers which we can convert to integer numbers by multiplying by a suitably large number.



# Notation and Assumptions

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- *Assumption 2. The network is directed.*
  - we can always fulfill this assumption by transforming any undirected network into a directed network.

- *Assumption 3. The supplies/demands at the nodes satisfy the condition*

$$\sum_{i \in N} b(i) = 0$$

- *and the minimum cost flow problem has a feasible solution.*

# Notation and Assumptions

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- *Assumption 4. We assume that the network  $G$  contains an uncapacitated directed path (i.e., each arc in the path has infinite capacity) between every pair of nodes.*
  - We impose this condition, if necessary, by adding artificial arcs  $(1, j)$  and  $(j, 1)$  for each  $j \in N$  and assigning a large cost and infinite capacity to each of these arcs.
  - No such arc would appear in a minimum cost solution unless the problem contains no feasible solution without artificial arcs.

# Notation and Assumptions

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- *Assumption 5. All arc costs are nonnegative.*
  - This assumption imposes no loss of generality since the arc reversal transformation

# Introduction

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- ***Residual Network***

- The residual network  $G(x)$  corresponding to a flow  $x$  is defined as follows.
- We replace each arc  $(i, j) \in A$  by two arcs  $(i, j)$  and  $(j, i)$ .
- The arc  $(i, j)$  has cost  $c_{ij}$  and residual capacity  $r_{ij} = u_{ij} - x_{ij}$
- the arc  $(j, i)$  has cost  $c_{ji} = -c_{ij}$  and residual capacity  $r_{ji} = x_{ij}$
- The residual network consists only of arcs with positive residual capacity.

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# **Cycle Canceling Algorithm**

# Cycle Canceling Algorithm

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- *Cycle Canceling Algorithm*

- This algorithm maintains a feasible solution and at every iteration attempts to improve its objective function value.
- The algorithm first establishes a feasible flow  $x$  in the network by solving a maximum flow problem
- Then it iteratively finds *negative cost-directed cycles* in the residual network and augments flows on these cycles.
- The algorithm terminates when the residual network contains no negative cost-directed cycle.
- When the algorithm terminates, it has found a minimum cost flow.

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- ***Theorem (Negative Cycle Optimality Conditions)***

- A feasible solution  $x^*$  is an optimal solution of the minimum cost flow problem if and only if it satisfies the negative cycle optimality conditions: namely, the residual network  $G(x^*)$  contains no negative cost (directed) cycle.

# Cycle Canceling Algorithm

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- *Cycle Canceling Algorithm*

**algorithm** *cycle-canceling*;

**begin**

    establish a feasible flow  $x$  in the network;

**while**  $G(x)$  contains a negative cycle **do**

**begin**

        use some algorithm to identify a negative cycle  $W$ ;

$\delta := \min\{r_{ij} : (i, j) \in W\}$ ;

        augment  $\delta$  units of flow in the cycle  $W$  and update  $G(x)$ ;

**end;**

**end;**



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# **Successive Shortest Path Algorithm**

# Successive Shortest Path Algorithm

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- The cycle-canceling algorithm maintains feasibility of the solution at every step and attempts to achieve optimality.
- In contrast, the *successive shortest path algorithm* maintains optimality of the solution at every step and strives to attain feasibility.
- It maintains a solution  $x$  that satisfies the nonnegativity and capacity constraints, but violates the mass balance constraints of the nodes.

# Successive Shortest Path Algorithm

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- At each step, the algorithm selects a node  $s$  with excess supply (i.e., supply not yet sent to some demand node) and a node  $t$  with unfulfilled demand and sends flow from  $s$  to  $t$  along a shortest path in the residual network.
- The algorithm terminates when the current solution satisfies all the mass balance constraints.

# Successive Shortest Path Algorithm

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- To describe this algorithm as well as several later developments, we first introduce the concept of *pseudoflows*.
- A pseudoflow is a function  $x: A \rightarrow \mathbb{R}^+$  satisfying only the capacity and nonnegativity constraints; it need not satisfy the mass balance constraints.
- For any pseudoflow  $x$ , we define the imbalance of node  $i$  as

$$e(i) = b(i) + \sum_{\{j:(j,i) \in A\}} x_{ji} - \sum_{\{j:(i,j) \in A\}} x_{ij} \quad \text{for all } i \in N.$$

# Successive Shortest Path Algorithm

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- For some node  $i$ :
  - If  $e(i) > 0$ , we refer to  $e(i)$  as the *excess* of node  $i$ ;
  - if  $e(i) < 0$ , we call  $-e(i)$  the node's *deficit*.
  - If  $e(i) = 0$ , we refer to a node  $i$  as *balanced*.
- Let  $E$  and  $D$  denote the sets of excess and deficit nodes in the network.
- Notice that

$$\sum_{i \in N} e(i) = \sum_{i \in N} b(i) = 0,$$

- And hence

$$\sum_{i \in E} e(i) = -\sum_{i \in D} e(i).$$

# Successive Shortest Path Algorithm

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- Consequently, if the network contains an excess node, it must also contain a deficit node.
- The residual network corresponding to a pseudoflow is defined in the same way that we define the residual network for a flow.



The End