

DM545  
Linear and Integer Programming

Lecture 11  
More on Network Flows

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1. Duality in Network Flow Problems

2. More on Network Flows

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# Shortest Path - Dual LP

$$z = \min \sum_{ij \in A} c_{ij} x_{ij}$$
$$\sum_{j:ij \in A} x_{ij} - \sum_{j:ji \in A} x_{ji} = -1 \quad \text{for } i = s \quad (\pi_s)$$
$$\sum_{j:ij \in A} x_{ij} - \sum_{j:ji \in A} x_{ji} = 0 \quad \forall i \in V \setminus \{s, t\} \quad (\pi_i)$$
$$\sum_{j:ij \in A} x_{ij} - \sum_{j:ji \in A} x_{ji} = 1 \quad \text{for } i = t \quad (\pi_t)$$
$$x_{ij} \geq 0 \quad \forall ij \in A$$

Dual problem:

$$g^{LP} = \max \pi_t - \pi_s$$
$$\pi_j - \pi_i \leq c_{ij} \quad \forall ij \in A$$

Hence, the shortest path can be found by potential values  $\pi_i$  on nodes such that  $\pi_s = 0, \pi_t = z$  and  $\pi_j - \pi_i \leq c_{ij}$  for  $ij \in A$

# Maximum $(s, t)$ -Flow

Adding a backward arc from  $t$  to  $s$ :

$$\begin{aligned} z &= \max x_{ts} \\ \sum_{j:ij \in A} x_{ij} - \sum_{j:ji \in A} x_{ji} &= 0 & \forall i \in V & \quad (\pi_i) \\ x_{ij} &\leq u_{ij} & \forall ij \in A & \quad (w_{ij}) \\ x_{ij} &\geq 0 & \forall ij \in A & \end{aligned}$$

Dual problem:

$$\begin{aligned} g^{LP} &= \min \sum_{ij \in A} u_{ij} w_{ij} \\ \pi_i - \pi_j + w_{ij} &\geq 0 & \forall ij \in A \\ \pi_t - \pi_s &\geq 1 \\ w_{ij} &\geq 0 & \forall ij \in A \end{aligned}$$

$$g^{LP} = \min \sum_{ij \in A} u_{ij} w_{ij} \quad (1)$$

$$\pi_i - \pi_j + w_{ij} \geq 0 \quad \forall ij \in A \quad (2)$$

$$\pi_t - \pi_s \geq 1 \quad (3)$$

$$w_{ij} \geq 0 \quad \forall ij \in A \quad (4)$$

- ▶ Without (3) all potentials would go to 0.
- ▶ Keep  $w$  low because of objective function
- ▶ Keep all potentials low  $\rightsquigarrow$  (3)  $\pi_s = 1, \pi_t = 0$
- ▶ Cut  $C$ : on left =1 on right =0. Where is the transition?
- ▶ Vars  $w$  identify the cut  $\rightsquigarrow \pi_j - \pi_i + w_{ij} \geq 0 \rightsquigarrow w_{ij} = 1$

$$w_{ij} = \begin{cases} 1 & \text{if } ij \in C \\ 0 & \text{otherwise} \end{cases}$$

for those arcs that minimize the cut capacity  $\sum_{ij \in A} u_{ij} w_{ij}$

- ▶ Complementary slackness:  $w_{ij} = 1 \implies x_{ij} = u_{ij}$

## Theorem

A strong dual to the max  $(st)$ -flow is the minimum  $(st)$ -cut problem:

$$\min_X \left\{ \sum_{ij \in A: i \in X, j \notin X} u_{ij} : s \in X \subset V \setminus \{t\} \right\}$$

# Min Cost Flow - Dual LP

$$\begin{aligned} & \min \sum_{ij \in A} c_{ij} x_{ij} \\ & \sum_{j:ij \in A} x_{ij} - \sum_{j:ji \in A} x_{ji} = b_i \quad \forall i \in V \quad (\pi_i) \\ & \quad \quad \quad x_{ij} \leq u_{ij} \quad \forall ij \in A \quad (w_{ij}) \\ & \quad \quad \quad x_{ij} \geq 0 \quad \forall ij \in A \end{aligned}$$

Dual problem:

$$\begin{aligned} (1) \quad & \max \sum_{i \in V} b_i \pi_i - \sum_{ij \in E} u_{ij} w_{ij} \\ (2) \quad & \quad \quad \quad -c_{ij} - \pi_i + \pi_j \leq w_{ij} \quad \forall ij \in E \\ (3) \quad & \quad \quad \quad w_{ij} \geq 0 \quad \forall ij \in A \end{aligned}$$



- ▶ define reduced costs  $\bar{c}_{ij} = c_{ij} + \pi_j - \pi_i$ , hence (2) becomes  $-\bar{c}_{ij} \leq w_{ij}$
- ▶  $u_e = \infty$  then  $w_e = 0$  (from obj. func) and  $\bar{c}_{ij} \geq 0$  (optimality condition)
- ▶  $u_e < \infty$  then  $w_e \geq 0$  and  $w_e \geq -\bar{c}_{ij}$  then  $w_e = \max\{0, -\bar{c}_{ij}\}$ , hence  $w_e$  is determined by others and may be skipped
- ▶ Complementary slackness  
(at optimality: each primal variable  $\times$  the corresponding dual slack must be equal 0, ie,  $x_e(\bar{c}_e + w_e) = 0$ ;  
each dual variable  $\times$  the corresponding primal slack must be equal 0, ie,  $w_e(x_e - u_e) = 0$ )
  - ▶  $x_e > 0$  then  $-\bar{c}_e = w_e = \max\{0, \bar{c}_e\}$ ,  
 $x_e > 0 \implies -\bar{c}_e > 0$  or equivalently (by negation)  $\bar{c}_e < 0 \implies x_e = 0$
  - ▶  $w_e > 0$  then  $x_e = u_e$   
 $-\bar{c} > 0 \implies x_e = u_e$  or equivalently  $-\bar{c} > 0 \implies x_e = u_e$

Hence:

$$\bar{c}_e < 0 \text{ then } x_e = u_e \neq \infty$$

$$\bar{c}_e > 0 \text{ then } x_e = 0$$

1. Duality in Network Flow Problems

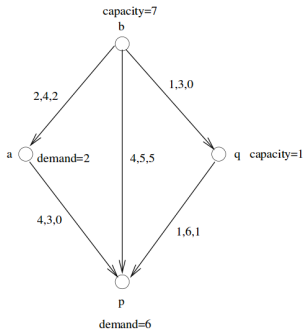
2. More on Network Flows

# Residual Network

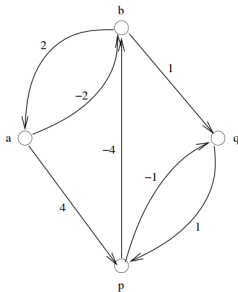
## Residual Network $N(x)$ :

how flow excess can be moved in  $G$  given that a flow  $x$  already exists  
replace arc  $ij \in N$  with arcs:

	residual capacity	cost
$ij$ :	$r_{ij} = u_{ij} - x_{ij}$	$c_{ij}$
$ji$ :	$r_{ji} = x_{ij}$	$-c_{ij}$



$(N, c, u, x)$



$(N(x), c')$

# Min Cost Flow Algorithms

**Optimality conditions:** Let  $x$  be feasible flow in  $N(V, A, l, u, b)$  then  $x$  is min cost flow in  $N$  iff  $N(x)$  contains no directed cycle of negative cost.

- ▶ Cycle canceling algorithm with Bellman Ford Moore for negative cycles  $O(nm^2UC)$ ,  $U = \max |u_e|$ ,  $C = \max |c_e|$
- ▶ Build up algorithms  $O(n^2mM)$ ,  $M = \max |b(v)|$

# Minimum spanning tree

## Definition

Given a graph  $G = (V, E)$

- ▶ a **forest** is a subgraph  $G' = (V, E')$  containing no cycles
- ▶ a **tree** is a subgraph  $G' = (V, E')$  that is a forest and is **connected** ( $\exists$  a  $(uv)$ -path  $\forall u, v \in V$ )

## Proposition

A graph  $G = (V, E)$  is a tree iff

- ▶ it is a forest containing exactly  $n - 1$  edges
- ▶ it is an edge minimal connected graph spanning  $V$
- ▶ it contains a unique path between every pair of nodes of  $V$
- ▶ the addition of an edge not in  $E$  creates a unique cycle.

Solvable via greedy algorithm (Kruskall)

$$\min \sum_{e \in E} c_e x_e \quad (5)$$

$$\sum_{e \in E} x_e = n - 1 \quad (6)$$

$$\sum_{e \in E(S)} x_e \leq |S| - 1 \quad \text{for } 2 \leq |S| \leq n - 1 \quad (7)$$

$$x_e \geq 0 \quad \text{for } e \in E \quad (8)$$

$$x \in \mathbb{Z}^{|E|} \quad (9)$$

### Theorem

*The convex hull of the incidence vectors of the forests in a graph is given by the constraints (2)-(3)*

- ▶ Improved version of the simplex method for network flows (still not polynomial but performs well in practice)
- ▶ it goes through same basic steps at each iteration:  
finding basic variable + determining leaving variable + solving for the new basis
- ▶ executes these steps exploiting network structure without needing a simplex tableau
- ▶ Key idea: network representation of basic feasible solutions

- ▶ in min cost flow formulation one of the node constraints is **redundant** (summing all these constraints yields zero on both sides -  $\sum_i b_i = 0$ )
- ▶ with  $n - 1$  non redundant node constraints, we have just  $n - 1$  basic variables for a basic solution  
each basic variable  $x_{ij}$  represents the flow through arc  $ij$ : **basic arcs**
- ▶ basic arcs never form **undirected cycles**...
- ▶ hence they form a **spanning tree**



## Optimality Condition

- ▶ Ford Fulkerson augmenting path algorithm  $O(m|x^*|)$
- ▶ Edmonds-Karp algorithm (augment by shortest path) in  $O(nm^2)$
- ▶ Dinic algorithm in layered networks  $O(n^2m)$
- ▶ Karzanov's push relabel  $O(n^2m)$

Matching:  $M \subseteq E$  of pairwise non adjacent edges

- ▶ bipartite graphs
- ▶ arbitrary graphs
- ▶ cardinality (max or perfect)
- ▶ weighted

Assignment problem  $\equiv$  min weighted perfect bipartite matching  $\equiv$  special case of min cost flow

## bipartite cardinality

### Theorem

The cardinality of a max matching in a bipartite graph equals the value of a maximum  $(s, t)$ -flow in  $N_{st}$ .

↪ Dinic  $O(\sqrt{nm})$

### Theorem (Optimality condition (Berge))

A matching  $M$  in a graph  $G$  is a maximum matching iff  $G$  contains no  $M$ -augmenting path.

↪ augmenting path  $O(\min(|U|, |V|), m)$

## bipartite weighted

build up algorithm  $O(n^3)$

bipartite weighted: Hungarian method  $O(n^3)$

## minimum weight perfect matching

Edmonds  $O(n^3)$

### Theorem (Hall's (marriage) theorem)

A bipartite graph  $B = (X, Y, E)$  has a matching covering  $X$  iff:

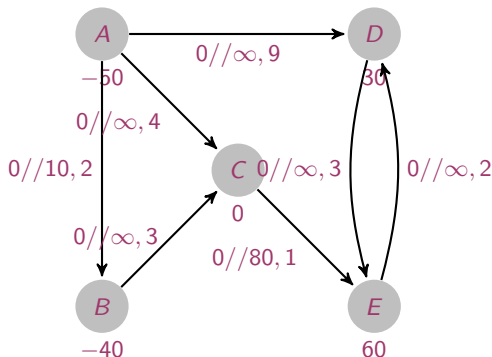
$$|N(U)| \geq |U| \quad \forall U \subseteq X$$

### Theorem (König, Egeavary theorem)

Let  $B = (X, Y, E)$  be a bipartite graph. Let  $M^*$  be the maximum matching and  $V^*$  the minimum vertex cover:

$$|M^*| = |V^*|$$

A company produces the same product at two different factories ( $A$  and  $B$ ) and then the product must be shipped to two warehouses, where either factory can supply either warehouse. The distribution network is shown below where  $C$  is a distribution center. There are costs and bounds on the amount of product to ship through the connections



What problem is it? Transshipment problem (ie, min cost flow)

See file mincost.xlsx

If Solver is not there, click **Tools**, select **Add-Ins**, **Solver Add-in** and OK.

Then **Tools**, **Solve**

What if  $\sum b(v) \neq 0$ ?

# Summary

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