## DM545/DM871 Linear and Integer Programming

## Lecture 11 Network Flows

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Outline

1. Duality in Network Flow Problems

	X <sub>e1</sub>	X <sub>e2</sub>	 Xij	 $X_{e_m}$		
	C <sub>e1</sub>	Ce <sub>2</sub>	 Cij	 Cem		
1	-1	<del>-</del>	 		=	$b_{1}$
2			 		=	$b_2$
:	:	1.			=	:
i	1		 -1		=	$b_i$
:	:	100			=	:
j			 1		=	$b_j$
:	:	100			=	:
n					=	$b_n$
$e_1$	1		 	 	≤ ≤	$u_1$
$e_2$	l I	1			$\leq$	$u_2$
:	:	100			<	:
(i,j)			1		≤ ≤	u <sub>ij</sub>
:	:	100			< <	:
e <sub>m</sub>				1	$\leq$	$u_m$

Outline

1. Duality in Network Flow Problems

## Shortest Path - Dual LP

$$z = \min \sum_{ij \in A} c_{ij} x_{ij}$$

$$\sum_{i} x_{ji} - \sum_{i} x_{ij} = 1 \qquad \qquad \text{for } i = s \qquad (\pi_s)$$

$$\sum_{i:ii\in A} x_{ij} - \sum_{i:ij\in A} x_{ji} = 0 \qquad \forall i\in V\setminus\{s,t\}$$
 (\pi\_i)

$$\sum_{i:ii\in A} \mathsf{x}_{ji} - \sum_{i:ij\in A} \mathsf{x}_{ij} = -1 \qquad \qquad \text{for } i = t \tag{$\pi_t$}$$

$$x_{ij} \geq 0$$
  $\forall ij \in A$ 

 $(\pi_s)$ 

 $(\pi_i)$ 

 $(\pi_t)$ 

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$$\sum_{j:ji\in A} x_{ji} - \sum_{j:ij\in A} x_{ij} = -1$$

$$x_{ij} \geq 0$$

for 
$$i = s$$

$$\forall i \in V \setminus \{s, t\}$$

for 
$$i = t$$

$$\forall ij \in A$$

$$g^{LP} = \max \pi_s - \pi_t$$
$$\pi_j - \pi_i \le c_{ij}$$

$$\forall ij \in A$$

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# Shortest Path - Dual LP

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$$x_{ij} \geq 0$$

$$\forall i \in V \setminus \{s, t\}$$

$$(\pi_i)$$

$$(\pi_t)$$

$$\forall i \in A$$

#### Dual problem:

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

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

# Maximum (s, t)-Flow

## Adding a backward arc from t to s:

$$z = \max_{j:ji \in A} x_{ts}$$

$$\sum_{j:ji \in A} x_{ij} - \sum_{j:ij \in A} x_{ji} = 0 \qquad \forall i \in V \qquad (\pi_i)$$

$$x_{ij} \leq u_{ij} \qquad \forall ij \in A \qquad (w_{ij})$$

$$x_{ij} \geq 0 \qquad \forall ij \in A$$

# Maximum (s, t)-Flow

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

$$z = \max_{j:ji \in A} x_{ij} - \sum_{j:ij \in A} x_{ji} = 0$$
  $\forall i \in V$   $(\pi_i)$ 
 $x_{ij} \leq u_{ij}$   $\forall ij \in A$   $(w_{ij})$ 
 $x_{ij} \geq 0$   $\forall ij \in A$ 

### Dual problem:

$$g^{LP} = \min \sum_{ij \in A} u_{ij} w_{ij}$$

$$\pi_i - \pi_j + w_{ij} \ge 0 \qquad \forall ij \in A$$

$$\pi_t - \pi_s \ge 1$$

$$w_{ij} \ge 0 \qquad \forall ij \in A$$

	X <sub>e</sub> 1	$X_{e_2}$	 $x_{ij}$	 $X_{e_m}$		
	C <sub>e</sub> 1	Ce <sub>2</sub>	 Cij	 Cem		
1	-1			 	=	$b_1$
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:	:	100			=	:
i	1		 -1		=	$b_i$
:	:	14.			_	:
j			 1		=	b <sub>j</sub>
:	:	14.			=	:
n					=	$\dot{b}_n$
$e_1$	1		 	 	$\leq$	$u_1$
<i>e</i> <sub>2</sub>		1			$\leq$	$u_2$
:		100			<	:
(i,j)	 		1		≤ ≤	u <sub>ij</sub>
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$$\pi_i - \pi_j + w_{ij} \ge 0 \qquad \forall ij \in A$$

$$\pi_t - \pi_s \ge 1 \qquad (3)$$

$$w_{ij} \ge 0 \qquad \forall ij \in A \qquad (4)$$

• Without (3) all potentials would go to 0.

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- Keep w low because of objective function

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$$\pi_i - \pi_j + w_{ij} \ge 0$$

$$\pi_t - \pi_s \ge 1$$

$$w_{ii} \ge 0$$

$$\forall ij \in A$$

$$(2)$$

$$(3)$$

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$$\pi_{i} - \pi_{j} + w_{ij} \ge 0 \qquad \forall ij \in A$$

$$\pi_{t} - \pi_{s} \ge 1 \qquad (3)$$

$$w_{ii} > 0 \qquad \forall ij \in A \qquad (4)$$

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- Vars w identify the cut  $\rightsquigarrow \pi_j \pi_i + w_{ij} \geq 0 \rightsquigarrow w_{ij} = 1$

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

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

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$$\forall ij \in A$$

$$(2)$$

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- Keep all potentials low  $\leadsto$  (3)  $\pi_s=0, \pi_t=1$
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• 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:

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

# Max Flow Algorithms

#### **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  $O(n^2m)$

# Min Cost Flow - Dual LP

$$\min \sum_{ij \in A} c_{ij} x_{ij}$$

$$\sum_{j: ji \in A} x_{ij} - \sum_{j: ij \in A} x_{ji} = b_{i} \qquad \forall i \in V \qquad (\pi_{i})$$

$$x_{ij} \leq u_{ij} \qquad \forall ij \in A \qquad (w_{ij})$$

$$x_{ij} \geq 0 \qquad \forall ij \in A$$

#### Dual problem:

$$\max \sum_{i \in V} b_i \pi_i - \sum_{ij \in E} u_{ij} w_{ij}$$

$$-c_{ij} - \pi_i + \pi_j \le w_{ij} \qquad \forall ij \in E$$

$$w_{ii} \ge 0 \qquad \forall ij \in A$$

$$(1)$$

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- $u_e < \infty$  then  $w_e \ge 0$  and  $w_e \ge -\bar{c}_{ij}$  then  $w_e = \max\{0, -\bar{c}_{ij}\}$ , hence  $w_e$  is determined by others and irrelevant

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  - $x_e > 0$  then  $-\bar{c}_e = w_e = \max\{0, -\bar{c}_e\}$ ,  $x_e > 0 \implies -\bar{c}_e \ge 0$  or equivalently (by negation)  $\bar{c}_e > 0 \implies x_e = 0$  each dual variable  $\times$  the corresponding primal slack must be equal 0, ie,  $w_e(x_e u_e) = 0$ ;

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 then  $x_e = u_e$   
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#### Hence:

$$ar{c}_e > 0$$
 then  $x_e = 0$   
 $ar{c}_e < 0$  then  $x_e = u_e \neq \infty$ 

# Min Cost Flow Algorithms

## Theorem (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  $O(n^2 mM)$ ,  $M = \max |b(v)|$