

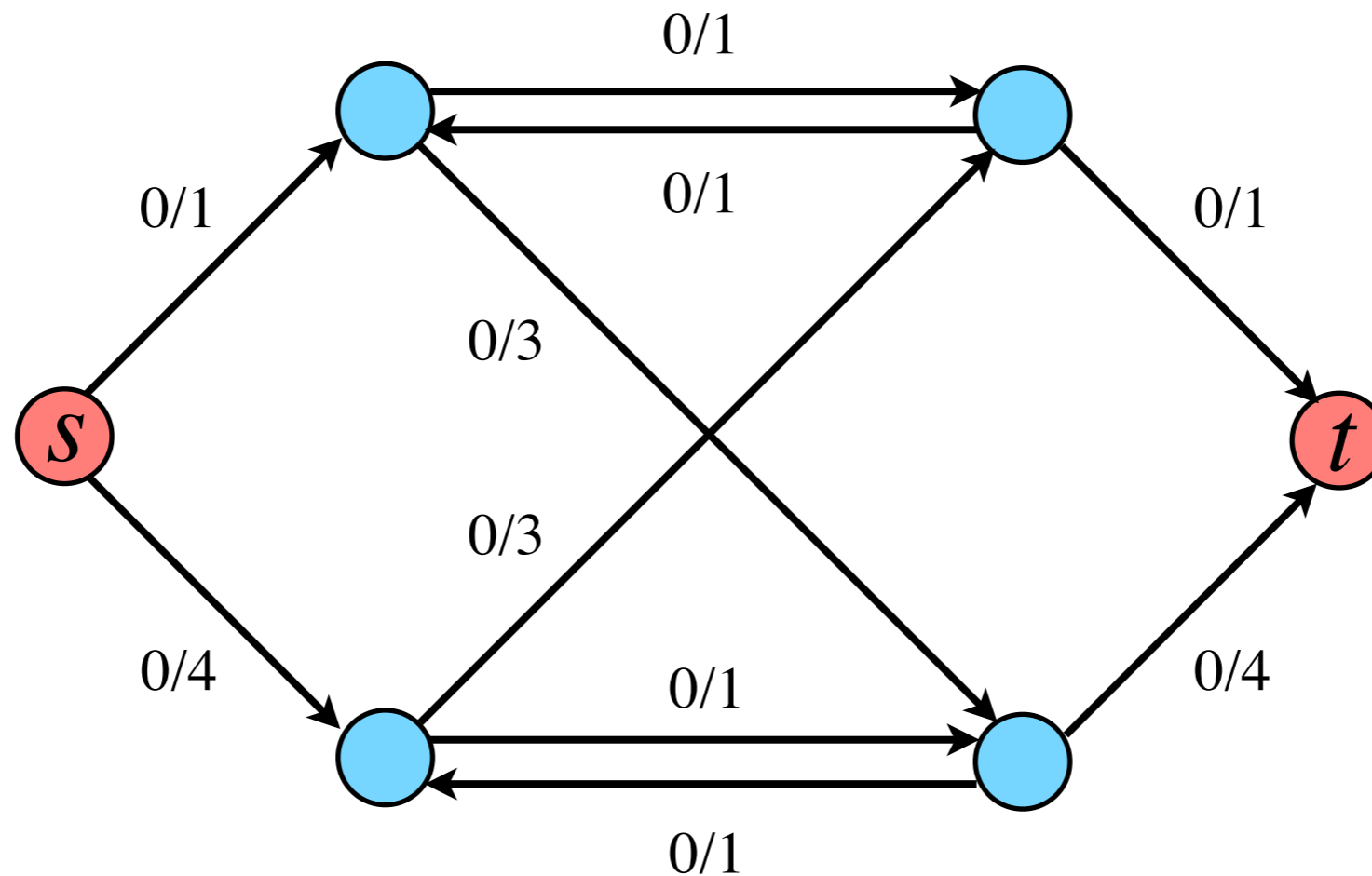
# Advanced Algorithms

## LP Duality

尹一通 Nanjing University, 2022 Fall

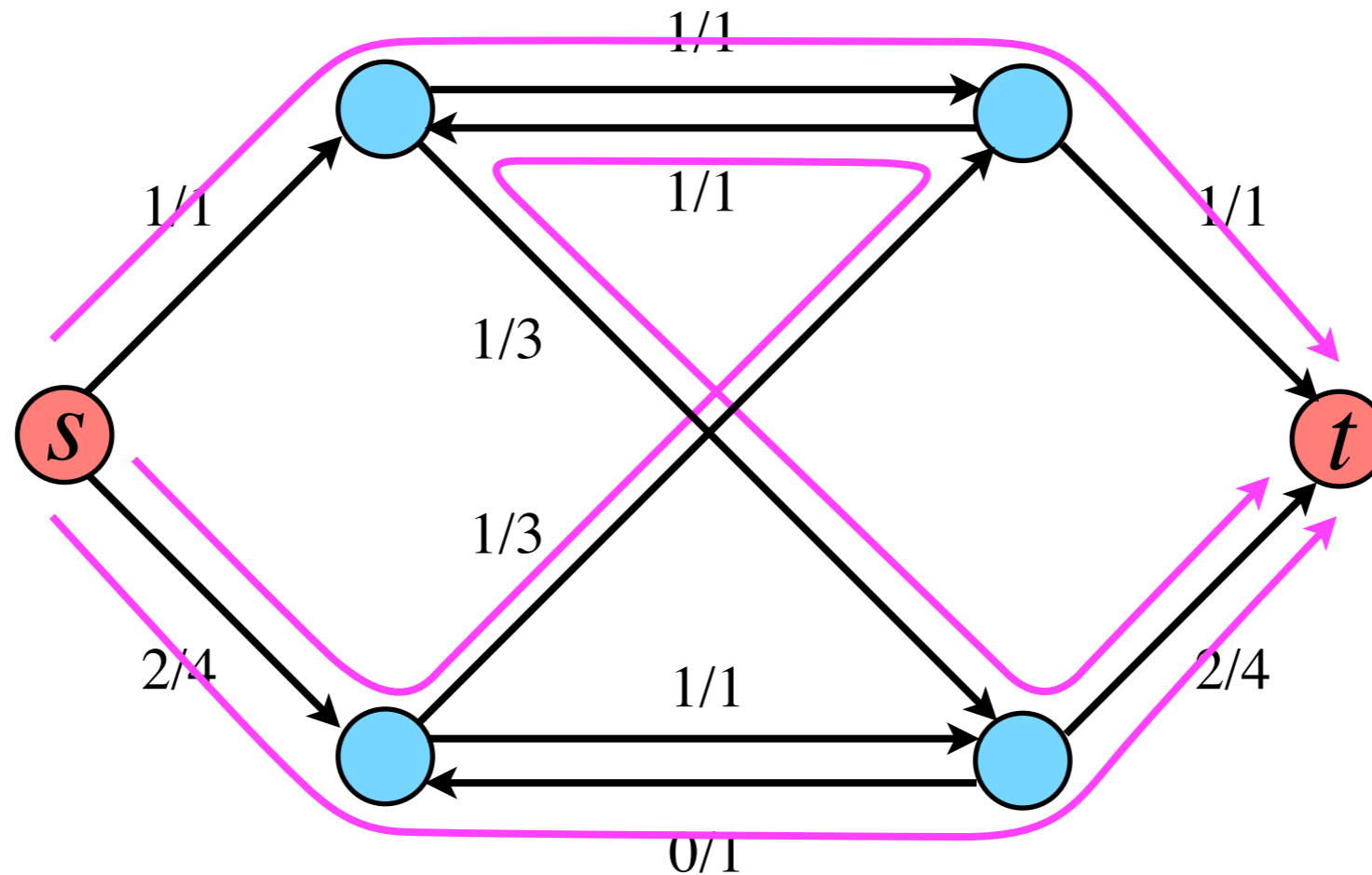
# Flow Network

- Digraph:  $D(V, E)$
- source:  $s \in V$     sink:  $t \in V$
- Capacity  $c : E \rightarrow \mathbb{R}_{\geq 0}$



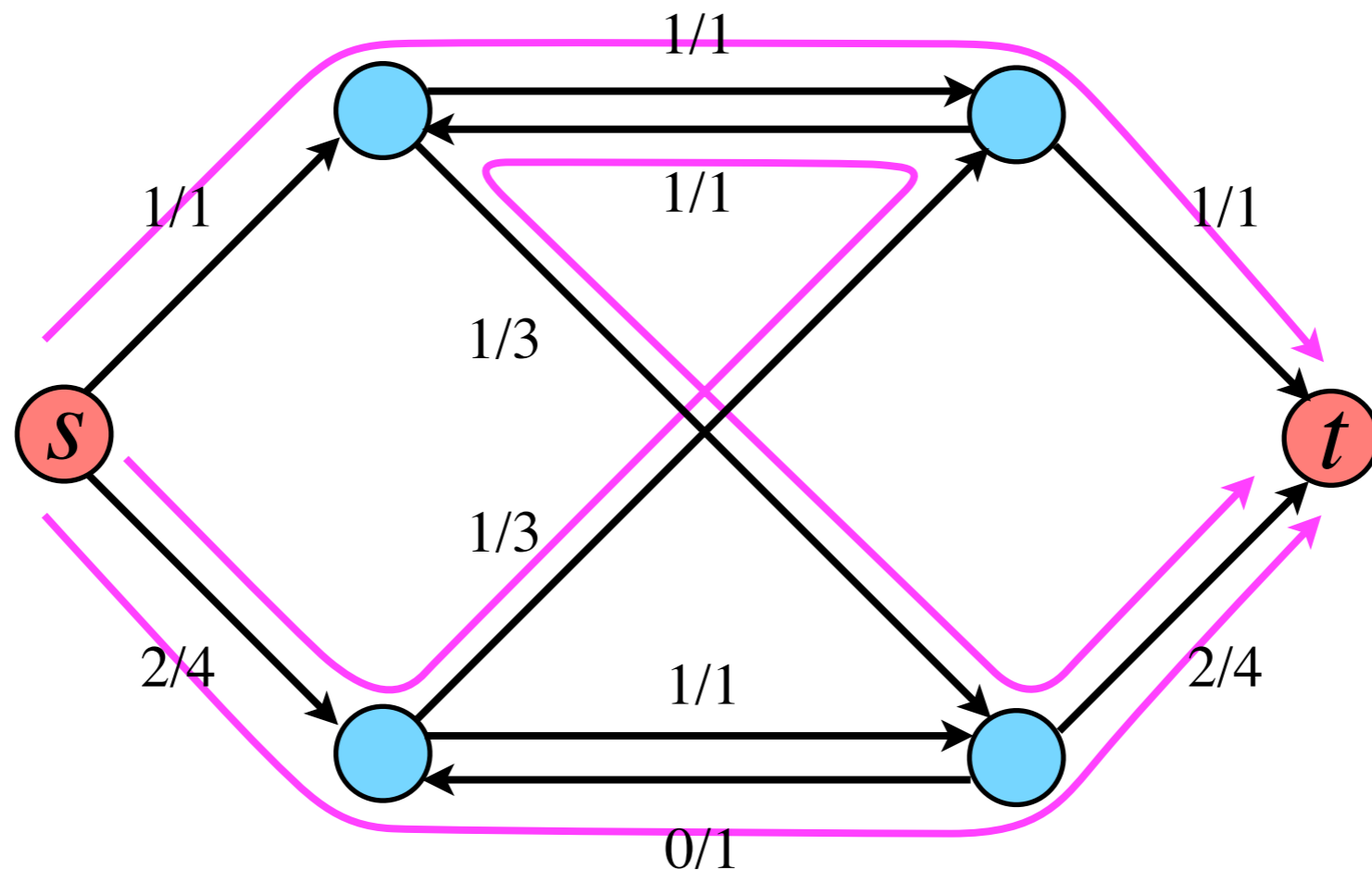
# Network Flow

- Digraph:  $D(V, E)$
- source:  $s \in V$     sink:  $t \in V$
- Capacity  $c : E \rightarrow \mathbb{R}_{\geq 0}$
- Flow  $f : E \rightarrow \mathbb{R}_{\geq 0}$



- **Capacity:**  $\forall (u, v) \in E, f_{uv} \leq c_{uv}$
- **Conservation:**  $\forall u \in V \setminus \{s, t\}, \sum_{(w,u) \in E} f_{wu} = \sum_{(u,v) \in E} f_{uv}$

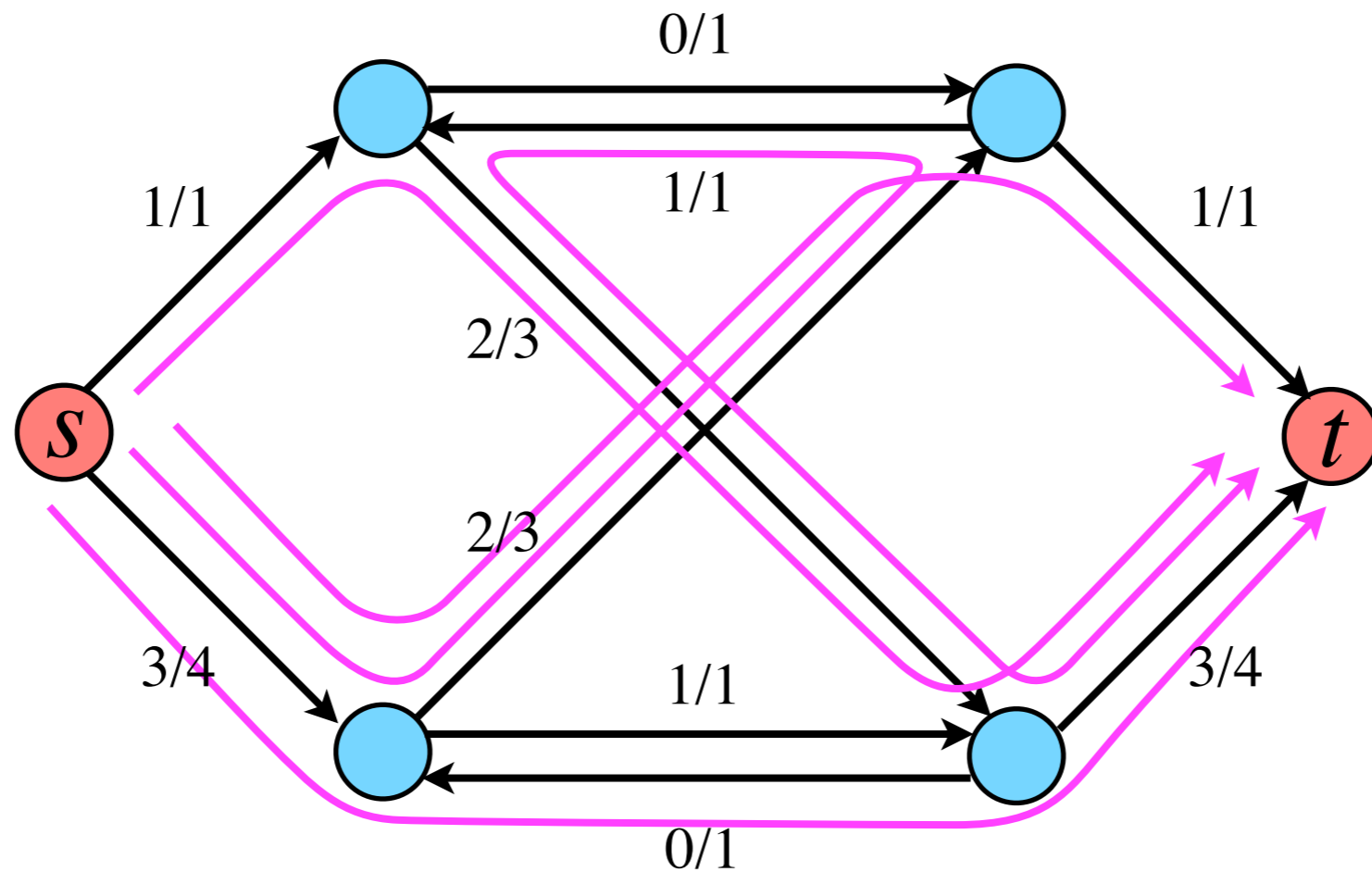
# Network Flow



- **Capacity:**  $\forall (u, v) \in E, f_{uv} \leq c_{uv}$
- **Conservation:**  $\forall u \in V \setminus \{s, t\}, \sum_{(w,u) \in E} f_{wu} = \sum_{(u,v) \in E} f_{uv}$

- **Value of flow:** 
$$\sum_{(s,u) \in E} f_{su} = \sum_{(v,t) \in E} f_{vt}$$

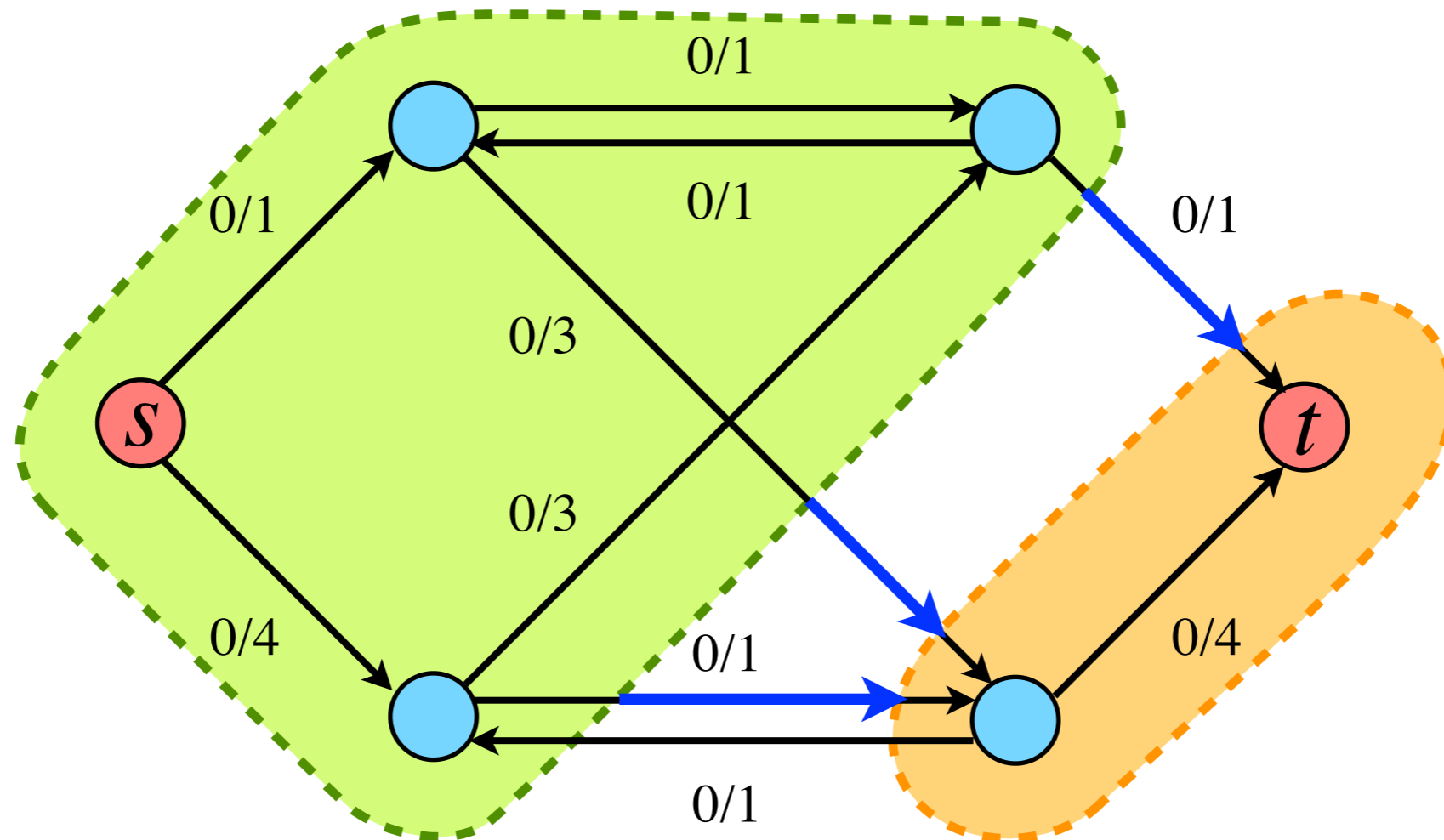
# Maximum Flow



- **Capacity:**  $\forall (u, v) \in E, f_{uv} \leq c_{uv}$
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# Cut

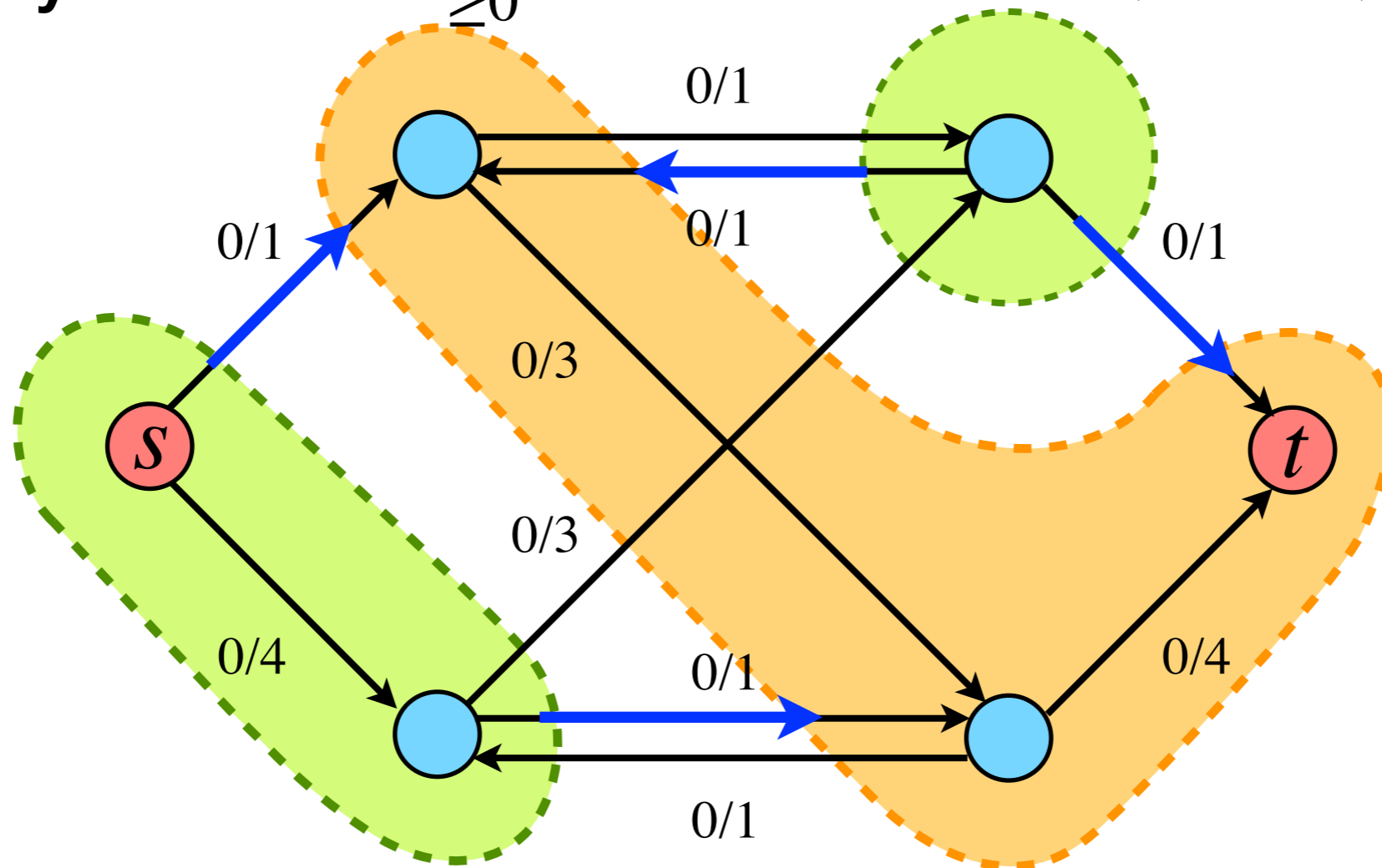
- Digraph:  $D(V, E)$
- Capacity  $c : E \rightarrow \mathbb{R}_{\geq 0}$
- source:  $s \in V$     sink:  $t \in V$
- **Cut**  $S \subset V, s \in S, t \notin S$



- **Value of cut:** 
$$\sum_{u \in S, v \notin S, (u,v) \in E} c_{uv}$$

# Minimum Cut

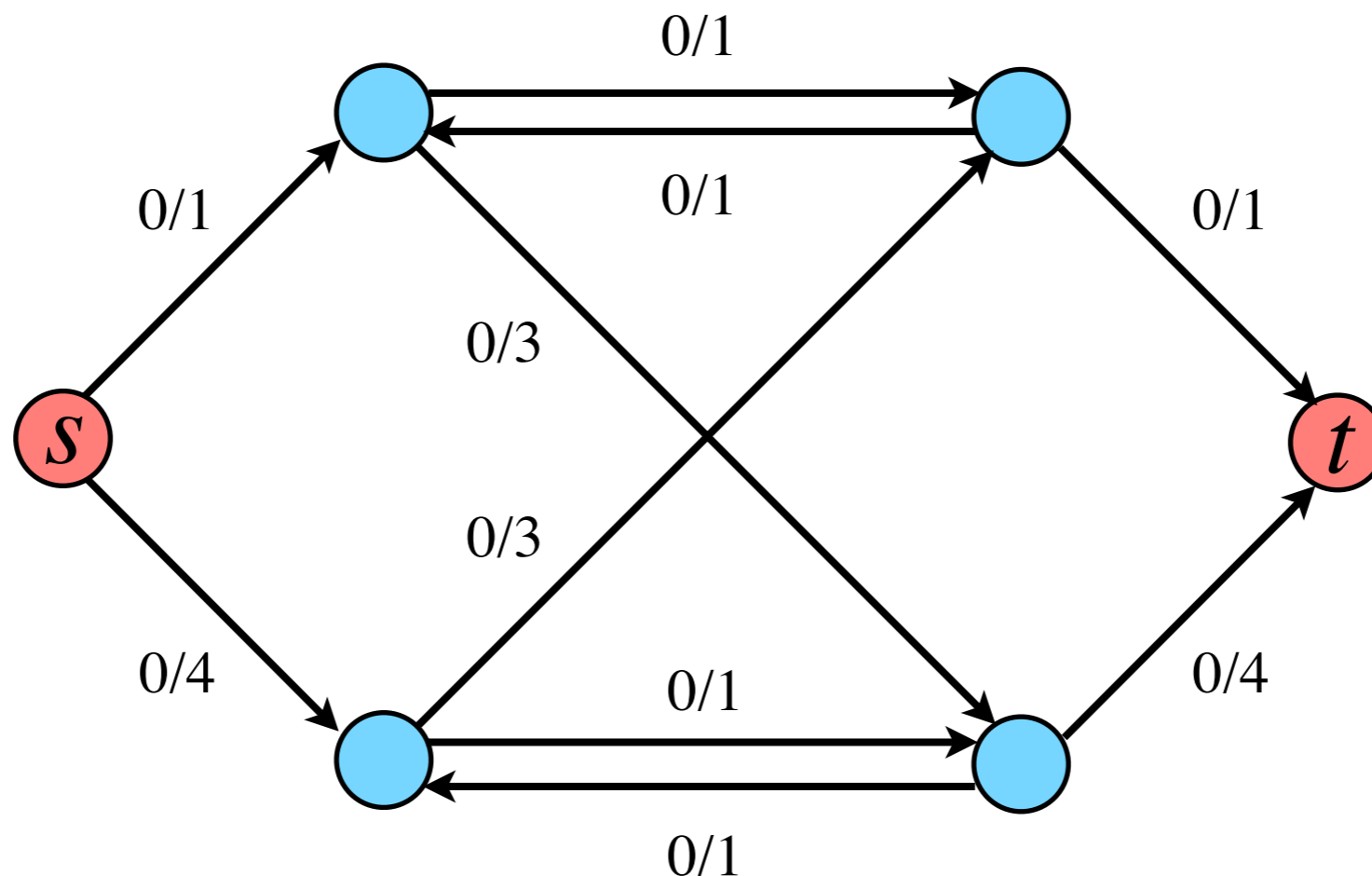
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# Fundamental Theorem of Flow

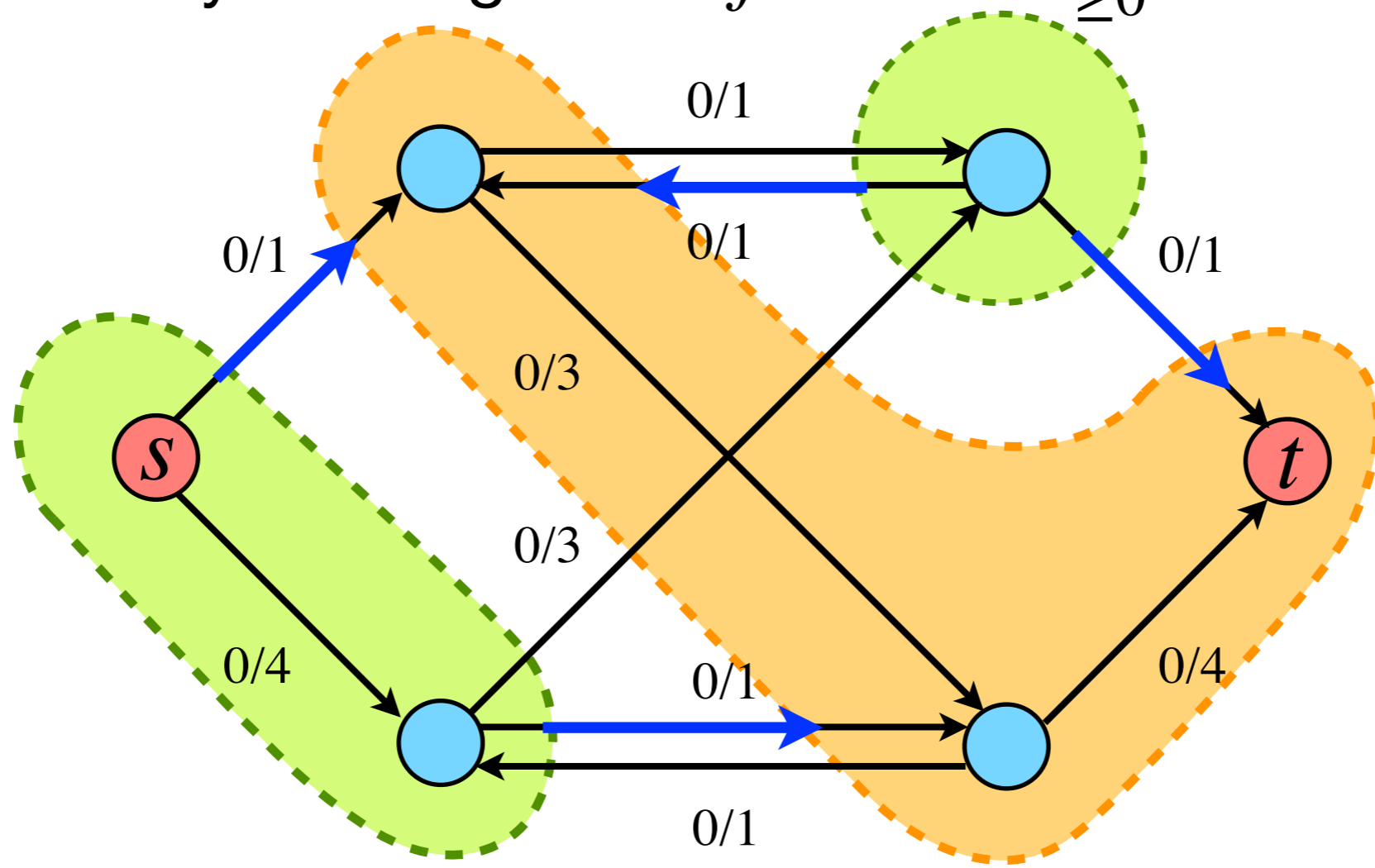
- Flow network:  $D(V, E)$ ,  $s, t \in V$ , and  $c : E \rightarrow \mathbb{R}_{\geq 0}$
- Max-flow = min-cut
- With integral capacity  $c : E \rightarrow \mathbb{Z}_{\geq 0}$ , the maximum flow is achieved by an integer flow  $f : E \rightarrow \mathbb{Z}_{\geq 0}$ .





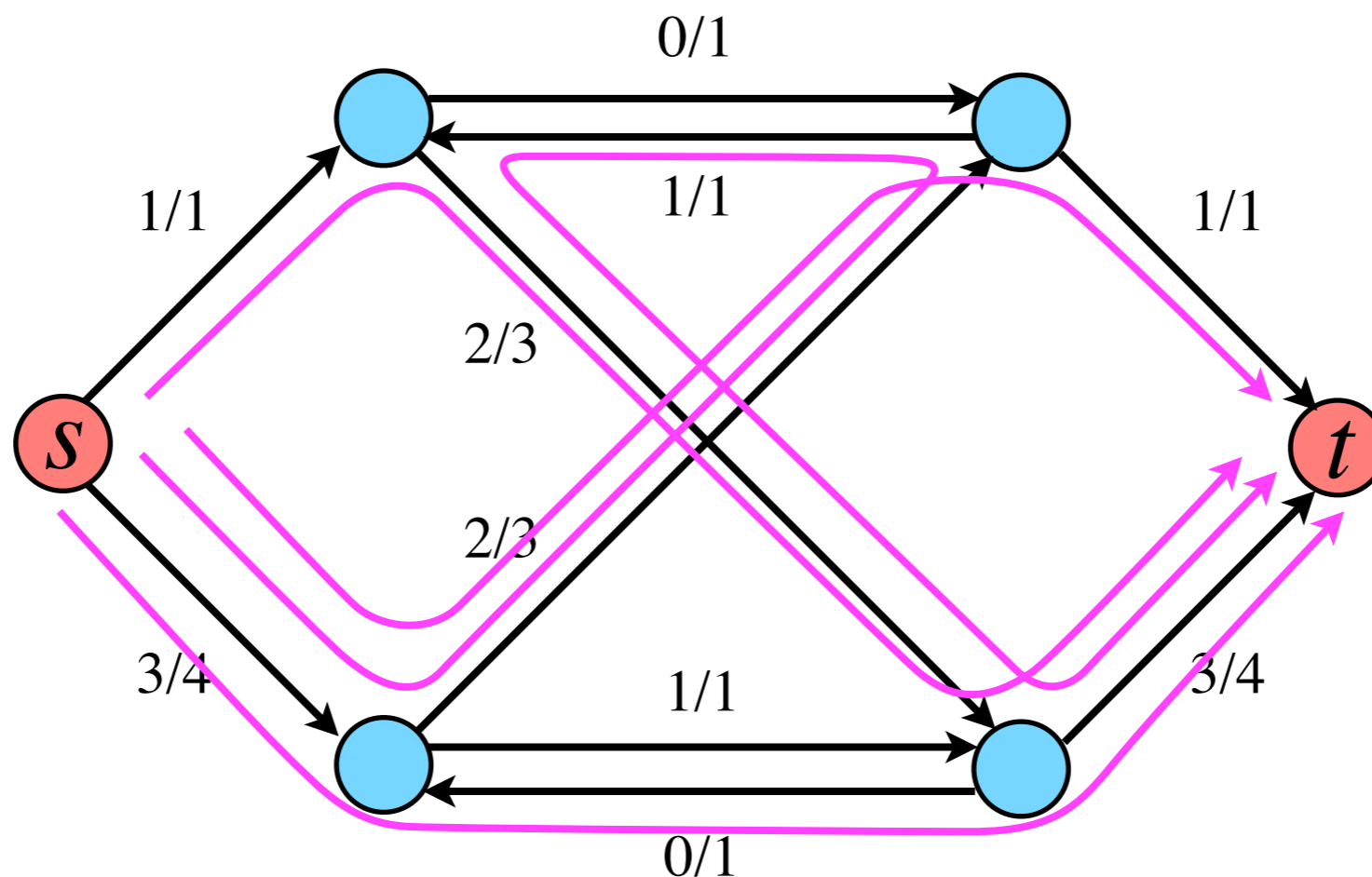
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- Flow network:  $D(V, E)$ ,  $s, t \in V$ , and  $c : E \rightarrow \mathbb{R}_{\geq 0}$ 
  - Max-flow = min-cut
  - With integral capacity  $c : E \rightarrow \mathbb{Z}_{\geq 0}$ , the maximum flow is achieved by an integer flow  $f : E \rightarrow \mathbb{Z}_{\geq 0}$ .
- *An elementary proof by augmenting path.*
- *An advanced proof by LP duality and integrality.*

# Estimate the Optima

minimize

$$7x_1 + x_2 + 5x_3$$

VI

subject to

$$x_1 - x_2 + 3x_3 \geq 10$$

+

$$5x_1 + 2x_2 - x_3 \geq 6$$

||

$$x_1, x_2, x_3 \geq 0 \quad 16$$

$$16 \leq \text{OPT} \leq \text{any feasible solution}$$

# Estimate the Optima

minimize

$$7x_1 + x_2 + 5x_3$$

VI

subject to

$$y_1 (x_1 - x_2 + 3x_3) \geq 10y_1$$

$$+ y_2 (5x_1 + 2x_2 - x_3) \geq 6y_2$$

$$x_1, x_2, x_3 \geq 0$$

$$10y_1 + 6y_2 \leq \text{OPT}$$

for any

$$\begin{array}{rcll} y_1 + 5y_2 & \leq & 7 \\ -y_1 + 2y_2 & \leq & 1 \\ 3y_1 - y_2 & \leq & 5 \end{array} \quad y_1, y_2 \geq 0$$

# Primal-Dual

Primal:

$$\begin{aligned} \min \quad & 7x_1 + x_2 + 5x_3 \\ \text{s.t.} \quad & x_1 - x_2 + 3x_3 \geq 10 \\ & 5x_1 + 2x_2 - x_3 \geq 6 \\ & x_1, x_2, x_3 \geq 0 \end{aligned}$$

Dual:

$$\begin{aligned} \max \quad & 10y_1 + 6y_2 \\ \text{s.t.} \quad & y_1 + 5y_2 \leq 7 \\ & -y_1 + 2y_2 \leq 1 \\ & 3y_1 - y_2 \leq 5 \\ & y_1, y_2 \geq 0 \end{aligned}$$

$\forall$  dual feasible  
 $\leq$  primal OPT

LP  $\in$  NP  $\cap$  coNP

# Diet Problem



calories

vitamin 1

⋮

vitamin  $m$

$c_1$	$c_2$	⋯	$c_n$
$a_{11}$	$a_{12}$	⋯	$a_{1n}$
⋮	⋮		⋮
$a_{m1}$	$a_{m2}$	⋯	$a_{mn}$

healthy

$\geq b_1$

⋮

$\geq b_m$

solution:

$x_1$

$x_2$

⋯

$x_n$

minimize the calories while keeping healthy

# Surviving Problem



price  
vitamin 1  
⋮  
vitamin  $m$

$c_1$	$c_2$	⋯	$c_n$
$a_{11}$	$a_{12}$	⋯	$a_{1n}$
⋮	⋮		⋮
$a_{m1}$	$a_{m2}$	⋯	$a_{mn}$

healthy

$$\begin{aligned} &\geq b_1 \\ &\vdots \\ &\geq b_m \end{aligned}$$

solution:  $x_1$     $x_2$    ⋯    $x_n$

minimize the total price while keeping healthy



# Surviving Problem

$$\min \mathbf{c}^T \mathbf{x}$$

$$\text{s.t. } \mathbf{Ax} \geq \mathbf{b}$$

$$\mathbf{x} \geq \mathbf{0}$$

price	$c_1$	$c_2$	$\dots$	$c_n$	healthy
vitamin 1	$a_{11}$	$a_{12}$	$\dots$	$a_{1n}$	$\geq b_1$
$\vdots$	$\vdots$	$\vdots$		$\vdots$	$\vdots$
vitamin $m$	$a_{m1}$	$a_{m2}$	$\dots$	$a_{mn}$	$\geq b_m$

solution:  $x_1$   $x_2$   $\dots$   $x_n$

minimize the total price while keeping healthy

# LP Duality

Primal:

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & Ax \geq b \\ & x \geq 0 \end{aligned}$$

Dual:

$$\begin{aligned} \max \quad & b^T y \\ \text{s.t.} \quad & A^T y \leq c \\ & y \geq 0 \end{aligned}$$

dual

solution: price

$y_1$  vitamin 1  
 $\vdots$   
 $y_m$  vitamin  $m$

$c_1$	$c_2$	$\dots$	$c_n$
$a_{11}$	$a_{12}$	$\dots$	$a_{1n}$
$\vdots$	$\vdots$		$\vdots$
$a_{m1}$	$a_{m2}$	$\dots$	$a_{mn}$

healthy

$\geq b_1$   
 $\vdots$   
 $\geq b_m$

$m$  types of vitamin pills, design a pricing system  
 competitive to  $n$  natural foods, max the total price

# LP Duality

**Primal:**

$$\min \mathbf{c}^T \mathbf{x}$$

$$\text{s.t. } \mathbf{A}\mathbf{x} \geq \mathbf{b}$$

$$\mathbf{x} \geq \mathbf{0}$$

$\geq$

**Dual:**

$$\max \mathbf{b}^T \mathbf{y}$$

$$\text{s.t. } \mathbf{A}^T \mathbf{y} \leq \mathbf{c}$$

$$\mathbf{y} \geq \mathbf{0}$$

- **Monogamy:**  $\text{dual}(\text{dual}(\text{LP})) = \text{LP}$
- **Weak duality:**
  - $\forall$  feasible primal solution  $\mathbf{x}$  and  $\forall$  feasible dual solution  $\mathbf{y}$

$$\mathbf{y}^T \mathbf{b} \leq \mathbf{y}^T \mathbf{A} \mathbf{x} \leq \mathbf{c}^T \mathbf{x}$$

# LP Duality

**Primal:**

$$\min \mathbf{c}^T \mathbf{x}$$

$$\text{s.t. } \mathbf{Ax} \geq \mathbf{b}$$

$$\mathbf{x} \geq \mathbf{0}$$

**Dual:**

$$\max \mathbf{b}^T \mathbf{y}$$

$$\text{s.t. } \mathbf{A}^T \mathbf{y} \leq \mathbf{c}$$

$$\mathbf{y} \geq \mathbf{0}$$

**Weak Duality Theorem:**

$\forall$  feasible primal solution  $\mathbf{x}$  and  $\forall$  feasible dual solution  $\mathbf{y}$ :

$$\mathbf{b}^T \mathbf{y} \leq \mathbf{c}^T \mathbf{x}$$

# LP Duality

**Primal:**

$$\min \mathbf{c}^T \mathbf{x}$$

$$\text{s.t. } \mathbf{Ax} \geq \mathbf{b}$$

$$\mathbf{x} \geq \mathbf{0}$$

**Dual:**

$$\max \mathbf{b}^T \mathbf{y}$$

$$\text{s.t. } \mathbf{A}^T \mathbf{y} \leq \mathbf{c}$$

$$\mathbf{y} \geq \mathbf{0}$$

**Strong Duality Theorem:**

Primal LP has an optimal solution  $\mathbf{x}^*$

$\iff$  Dual LP has an optimal solution  $\mathbf{y}^*$

$$\mathbf{b}^T \mathbf{y}^* = \mathbf{c}^T \mathbf{x}^*$$

# Maximum Flow

- Digraph:  $D(V, E)$
- Capacity  $c : E \rightarrow \mathbb{R}_{\geq 0}$

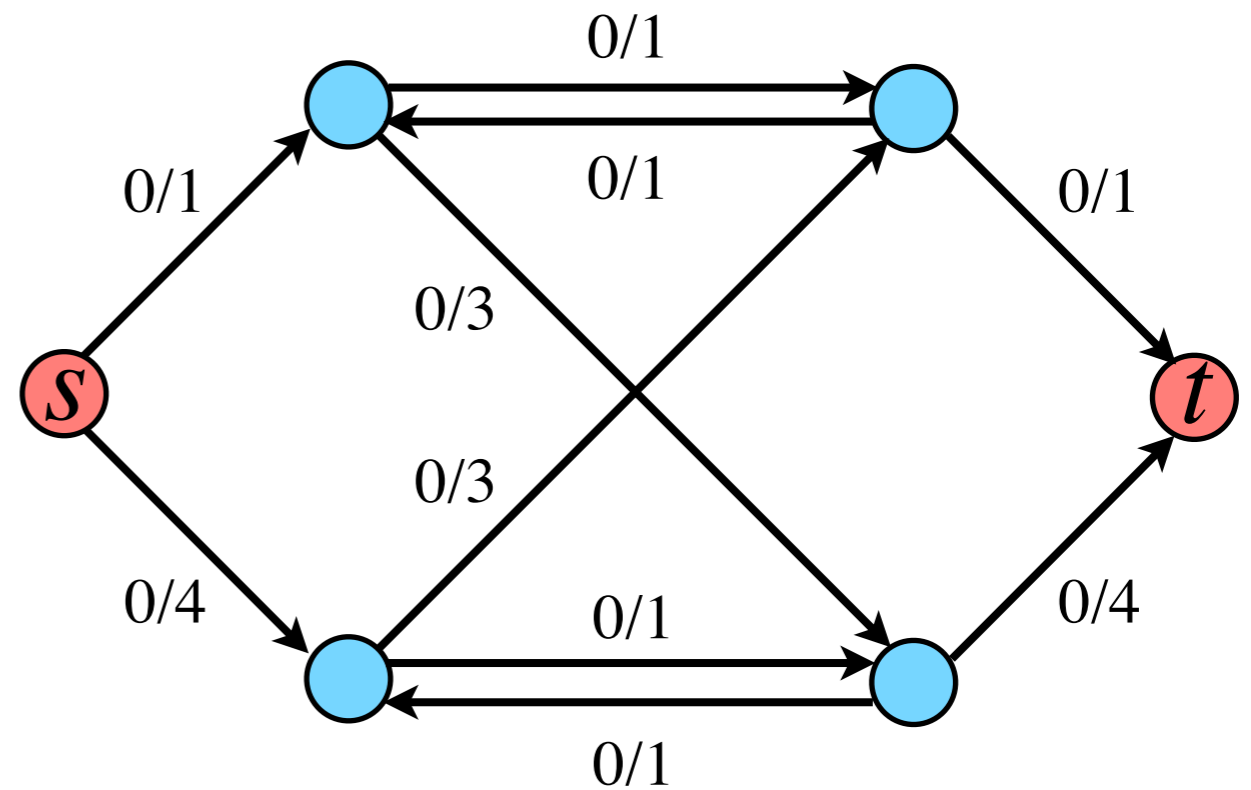
- source:  $s \in V$     sink:  $t \in V$

$$\max \sum_{u:(s,u) \in E} f_{su}$$

$$\text{s.t.} \quad f_{uv} \leq c_{uv}$$

$$\sum_{w:(w,u) \in E} f_{wu} - \sum_{v:(u,v) \in E} f_{uv} = 0 \quad \forall u \in V \setminus \{s, t\}$$

$$f_{uv} \geq 0 \quad \forall (u, v) \in E$$



$$\forall (u, v) \in E$$

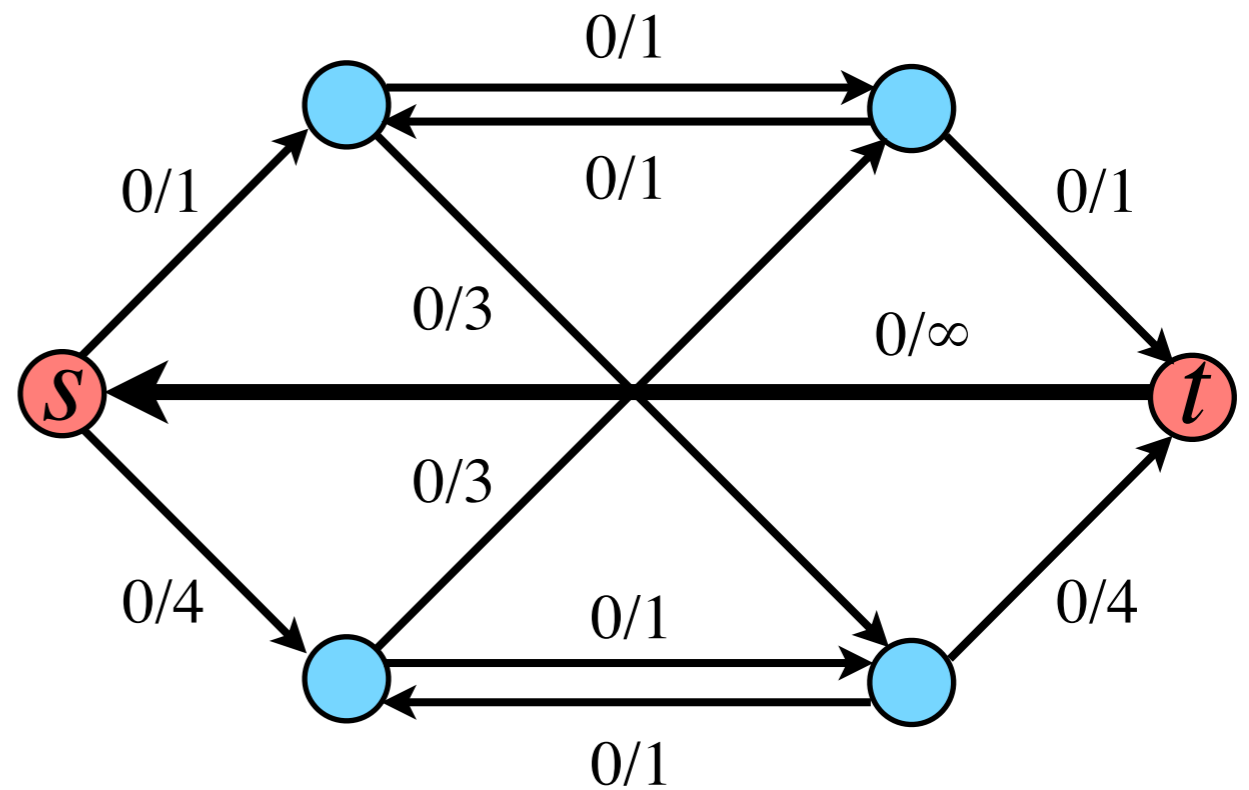
$$\forall u \in V \setminus \{s, t\}$$

$$\forall (u, v) \in E$$

# Maximum Flow

- Digraph:  $D(V, E)$
- Capacity  $c : E \rightarrow \mathbb{R}_{\geq 0}$

- source:  $s \in V$     sink:  $t \in V$



$$\max \quad f_{ts}$$

$$d_{uv} \quad \text{s.t.} \quad f_{uv} \leq c_{uv} \quad \forall (u, v) \in E$$

$$p_u \quad \sum_{w:(w,u) \in E'} f_{wu} - \sum_{v:(u,v) \in E'} f_{uv} \leq 0 \quad \forall u \in V$$

$$f_{uv} \geq 0 \quad \forall (u, v) \in E' = E \cup \{(t, s)\}$$

# Dual LP

- Digraph:  $D(V, E)$
- Capacity  $c : E \rightarrow \mathbb{R}_{\geq 0}$

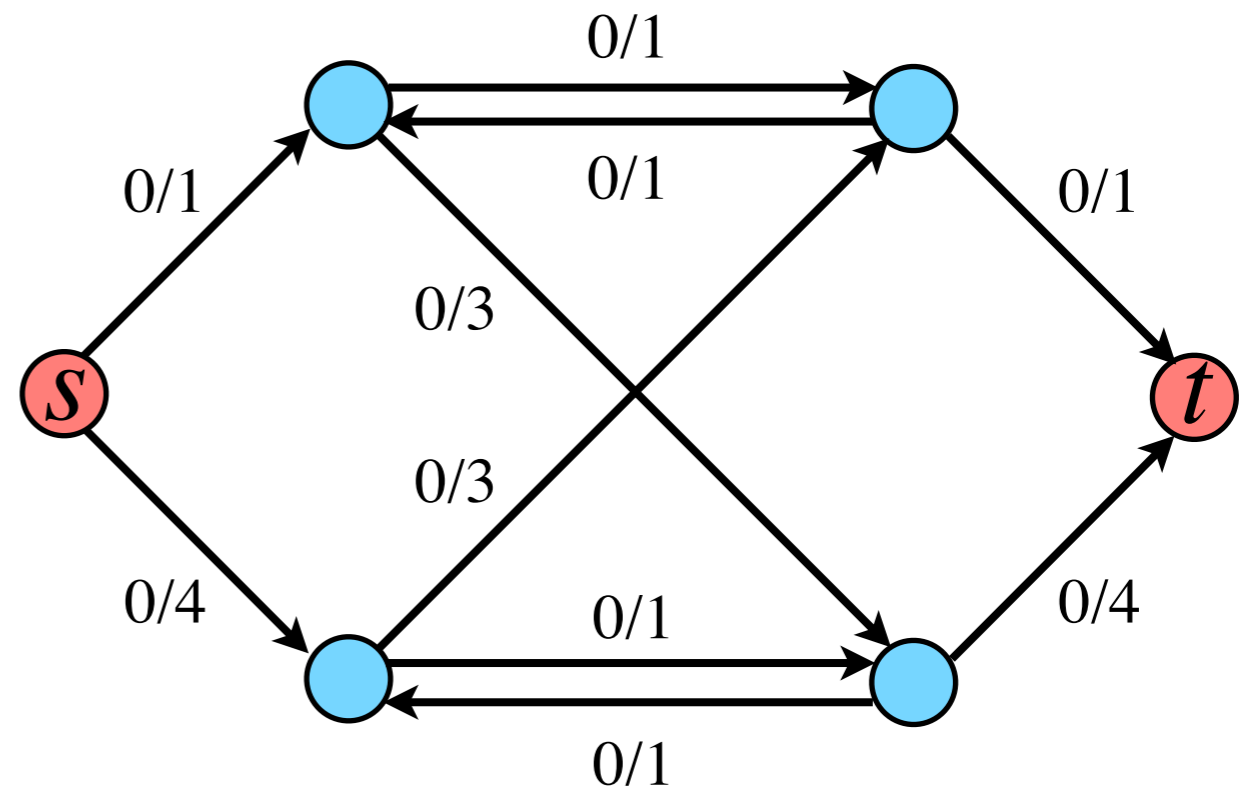
$$\min \sum_{(u,v) \in E} c_{uv} d_{uv}$$

$$\text{s.t.} \quad d_{uv} - p_u + p_v \geq 0 \quad \forall (u, v) \in E$$

$$p_s - p_t \geq 1$$

$$d_{uv} \geq 0 \quad p_u \geq 0 \quad \forall (u, v) \in E \quad \forall u \in V$$

- source:  $s \in V$     sink:  $t \in V$

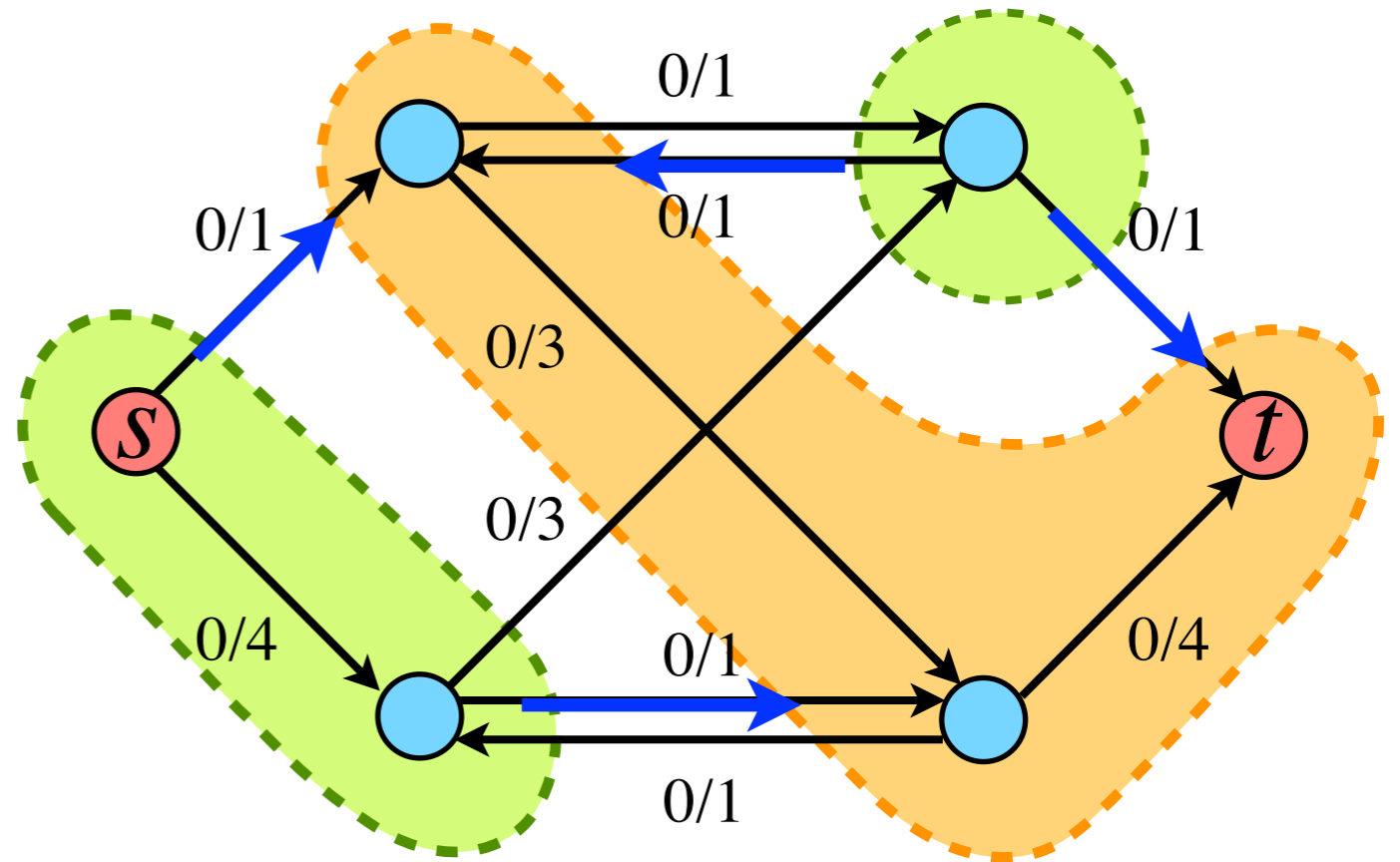




# Minimum Cut

- Digraph:  $D(V, E)$
- Capacity  $c : E \rightarrow \mathbb{R}_{\geq 0}$

- source:  $s \in V$     sink:  $t \in V$



$$\min \sum_{(u,v) \in E} c_{uv} d_{uv}$$

$$\text{s.t.} \quad d_{uv} - p_u + p_v \geq 0 \quad \forall (u, v) \in E$$

$$p_s - p_t \geq 1$$

$$d_{uv}, p_u \in \{0, 1\} \quad \forall (u, v) \in E \quad \forall u \in V$$

# Primal-Dual Schema

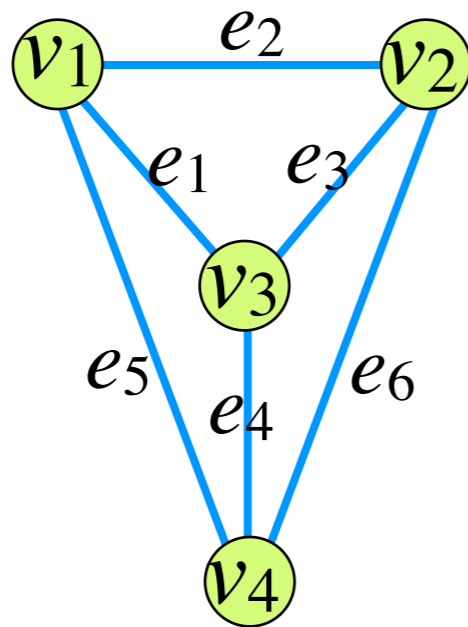
# LP-based Algorithms

- LP relaxation and rounding:
  - Relax the Integer Linear Program to an LP.
  - Round the optimal LP solution to a *feasible integral* solution.
- **Primal-dual schema:**
  - Find a pair of feasible solutions to the primal and dual programs which are close to each other.

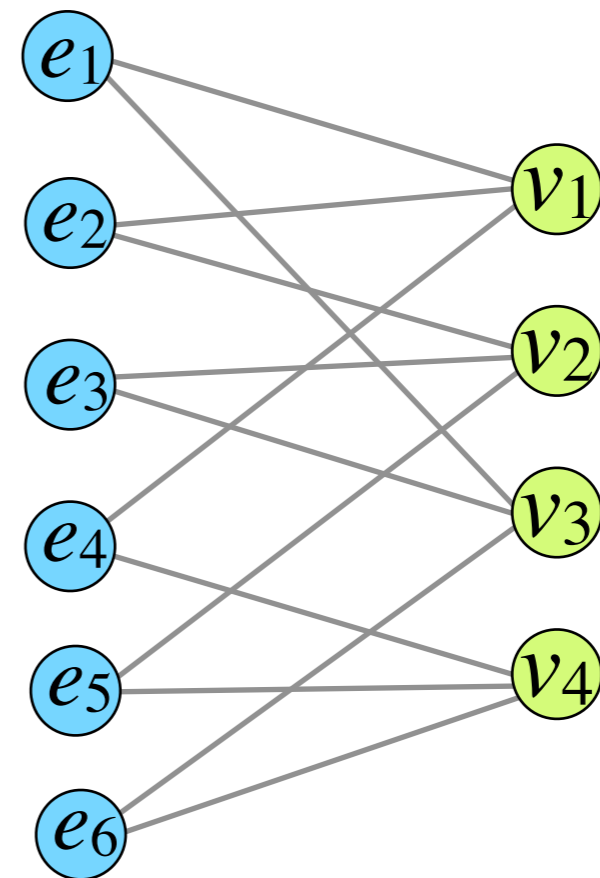
close to dual feasible solution  $\implies$  closer to OPT

# Vertex Cover

**Instance:** An undirected graph  $G(V, E)$ .  
Find the smallest  $C \subseteq V$  that intersects all edges.



incidence graph



set cover instance  
with *frequency* = 2

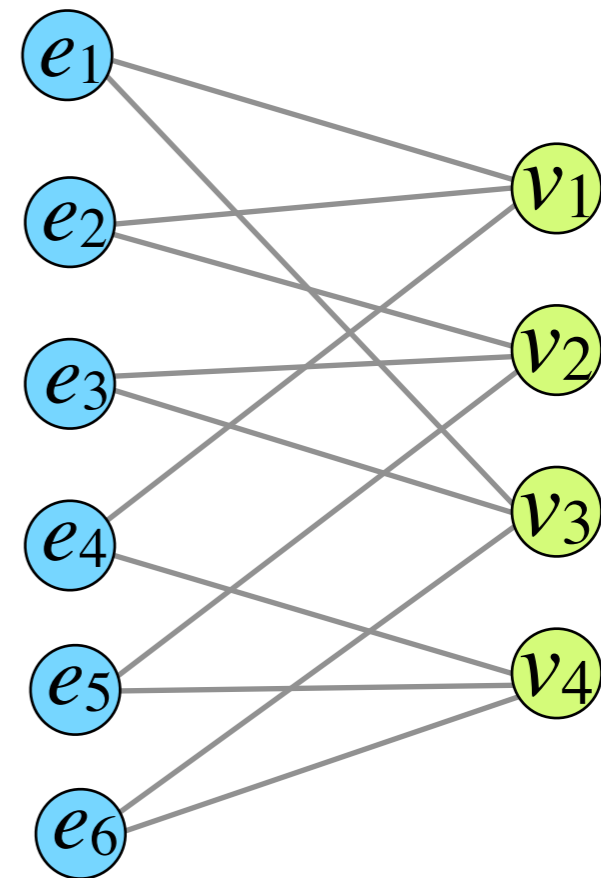
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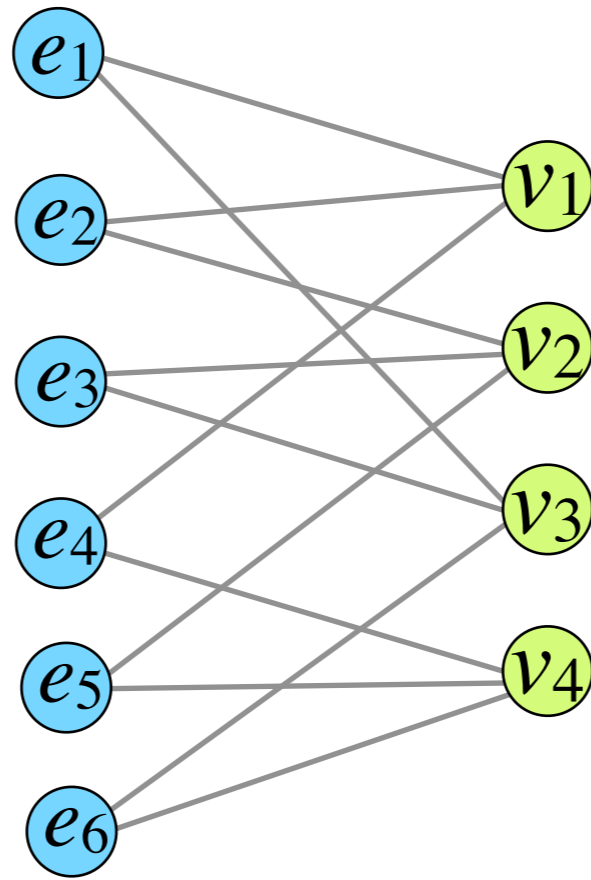
Find a *maximal matching*  $M \subseteq E$ ;  
return  $C = \{v \mid \{u, v\} \in M\}$ ;

- **Matching**  $\implies |M| \leq OPT_{VC}$   
(weak duality)
- **Maximality**  $\implies C$  is a vertex cover

$$|C| \leq 2|M| \leq 2OPT_{VC}$$



# Duality



**vertex cover:** constraints variables  
 $\sum_{v \in e} x_v \geq 1$   $x_v \in \{0,1\}$

**matching:** variables constraints  
 $y_e \in \{0,1\}$   $\sum_{e \ni v} y_e \leq 1$

# Duality

**Instance:** graph  $G(V, E)$

**primal:**  
(vertex cover)

minimize  $\sum_{v \in V} x_v$

subject to  $\sum_{v \in e} x_v \geq 1, \quad \forall e \in E$

$x_v \in \{0, 1\}, \quad \forall v \in V$

**dual:**  
(matching)

maximize  $\sum_{e \in E} y_e$

subject to  $\sum_{e \ni v} y_e \leq 1, \quad \forall v \in V$

$y_e \in \{0, 1\}, \quad \forall e \in E$

# Duality for LP-Relaxations

**Instance:** graph  $G(V, E)$

**primal:** minimize  $\sum_{v \in V} x_v$

subject to  $\sum_{v \in e} x_v \geq 1, \quad \forall e \in E$

$x_v \geq 0, \quad \forall v \in V$

**dual:** maximize  $\sum_{e \in E} y_e$

subject to  $\sum_{e \ni v} y_e \leq 1, \quad \forall v \in V$

$y_e \geq 0, \quad \forall e \in E$



# LP Duality

**Primal:**

$$\min \mathbf{c}^T \mathbf{x}$$

$$\text{s.t. } \mathbf{Ax} \geq \mathbf{b}$$

$$\mathbf{x} \geq \mathbf{0}$$

**Dual:**

$$\max \mathbf{b}^T \mathbf{y}$$

$$\text{s.t. } \mathbf{A}^T \mathbf{y} \leq \mathbf{c}$$

$$\mathbf{y} \geq \mathbf{0}$$

**Strong Duality Theorem:**

Primal LP has an optimal solution  $\mathbf{x}^*$

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# Complementary Slackness

**Primal:**  $\min \mathbf{c}^T \mathbf{x}$   
s.t.  $A\mathbf{x} \geq \mathbf{b}$   
 $\mathbf{x} \geq \mathbf{0}$

**Dual:**  $\max \mathbf{b}^T \mathbf{y}$   
s.t.  $A^T \mathbf{y} \leq \mathbf{c}$   
 $\mathbf{y} \geq \mathbf{0}$

**Theorem (Complementary Slackness Condition):**

For feasible primal solution  $\mathbf{x}$  and feasible dual solution  $\mathbf{y}$ ,  $\mathbf{x}$  and  $\mathbf{y}$  are both optimal iff:

- $\mathbf{y}^T (A\mathbf{x} - \mathbf{b}) = 0;$
- $\mathbf{x}^T (\mathbf{c} - A^T \mathbf{y}) = 0;$

$$\forall i : y_i > 0 \implies A_i \cdot \mathbf{x} = b_i$$

$$\forall j : x_j > 0 \implies A_{\cdot j}^T \mathbf{y} = c_j$$

# Complementary Slackness

$$\begin{aligned} \text{Primal: } & \min \mathbf{c}^T \mathbf{x} \\ & \text{s.t. } \mathbf{Ax} \geq \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned}$$

$$\begin{aligned} \text{Dual: } & \max \mathbf{b}^T \mathbf{y} \\ & \text{s.t. } \mathbf{A}^T \mathbf{y} \leq \mathbf{c} \\ & \mathbf{y} \geq \mathbf{0} \end{aligned}$$

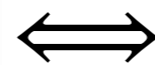
$\forall$  feasible primal solution  $\mathbf{x}$  and feasible dual solution  $\mathbf{y}$ :

$$\mathbf{y}^T \mathbf{b} \leq \mathbf{y}^T \mathbf{Ax} \leq \mathbf{c}^T \mathbf{x}$$

$$\mathbf{x} \text{ and } \mathbf{y} \text{ are both optimal} \iff \mathbf{y}^T \mathbf{b} = \mathbf{y}^T \mathbf{Ax} = \mathbf{c}^T \mathbf{x}$$



$$\begin{aligned} \forall i : y_i > 0 & \implies A_{i \cdot} \mathbf{x} = b_i \\ \forall j : x_j > 0 & \implies A_{\cdot j}^T \mathbf{y} = c_j \end{aligned}$$



$$\begin{aligned} & \bullet \mathbf{y}^T (\mathbf{Ax} - \mathbf{b}) = 0; \\ & \bullet \mathbf{x}^T (\mathbf{c} - \mathbf{A}^T \mathbf{y}) = 0; \end{aligned}$$

# Complementary Slackness

**Primal:**  $\min \mathbf{c}^T \mathbf{x}$   
s.t.  $A\mathbf{x} \geq \mathbf{b}$   
 $\mathbf{x} \geq \mathbf{0}$

**Dual:**  $\max \mathbf{b}^T \mathbf{y}$   
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$$\begin{aligned} \text{Primal: } & \min \mathbf{c}^T \mathbf{x} \\ & \text{s.t. } A\mathbf{x} \geq \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned}$$

$$\begin{aligned} \text{Dual: } & \max \mathbf{b}^T \mathbf{y} \\ & \text{s.t. } A^T \mathbf{y} \leq \mathbf{c} \\ & \mathbf{y} \geq \mathbf{0} \end{aligned}$$

## Theorem:

$\forall$  feasible primal solution  $\mathbf{x}$  and feasible dual solution  $\mathbf{y}$ ,

if for  $\alpha, \beta \geq 1$  :

$$\forall i : y_i > 0 \implies A_i \cdot \mathbf{x} \leq \alpha b_i$$

$$\forall j : x_j > 0 \implies A_j^T \mathbf{y} \geq c_j / \beta$$

$$\implies \mathbf{c}^T \mathbf{x} \leq \alpha \beta \mathbf{b}^T \mathbf{y} \leq \alpha \beta OPT_{LP}$$

$$\sum_{j=1}^n c_j x_j \leq \sum_{j=1}^n \left( \beta \sum_{i=1}^m a_{ij} y_i \right) x_j = \beta \sum_{i=1}^m \left( \sum_{j=1}^n a_{ij} x_j \right) y_i \leq \alpha \beta \sum_{i=1}^m b_i y_i$$

# Primal-Dual Schema

**Primal**     $\min \mathbf{c}^T \mathbf{x}$   
**IP:**         $\text{s.t. } \mathbf{Ax} \geq \mathbf{b}$   
               $\mathbf{x} \in \mathbb{Z}_{\geq 0}^n$

**Dual**     $\max \mathbf{b}^T \mathbf{y}$   
**LP-Relax:**  $\text{s.t. } \mathbf{A}^T \mathbf{y} \leq \mathbf{c}$   
                                   $\mathbf{y} \geq \mathbf{0}$

## Primal-Dual Schema:

Find a pair  $(\mathbf{x}, \mathbf{y})$  of feasible primal *integral* solution  $\mathbf{x}$  and feasible dual solution  $\mathbf{y}$  such that for some  $\alpha, \beta \geq 1$ :

$$\forall i : y_i > 0 \implies A_i \cdot \mathbf{x} \leq \alpha b_i$$

$$\forall j : x_j > 0 \implies A_j^T \mathbf{y} \geq c_j / \beta$$

$$\implies \mathbf{c}^T \mathbf{x} \leq \alpha \beta \mathbf{b}^T \mathbf{y} \leq \alpha \beta OPT_{LP} \leq \alpha \beta OPT_{IP}$$

# Primal-Dual Schema

$$\begin{array}{ll} \text{Primal} & \min \mathbf{c}^T \mathbf{x} \\ \text{IP:} & \text{s.t. } \mathbf{Ax} \geq \mathbf{b} \\ & \mathbf{x} \in \mathbb{Z}_{\geq 0}^n \end{array}$$

$$\begin{array}{ll} \text{Dual} & \max \mathbf{b}^T \mathbf{y} \\ \text{LP-Relax:} & \text{s.t. } \mathbf{A}^T \mathbf{y} \leq \mathbf{c} \\ & \mathbf{y} \geq \mathbf{0} \end{array}$$

## Primal-Dual Schema:

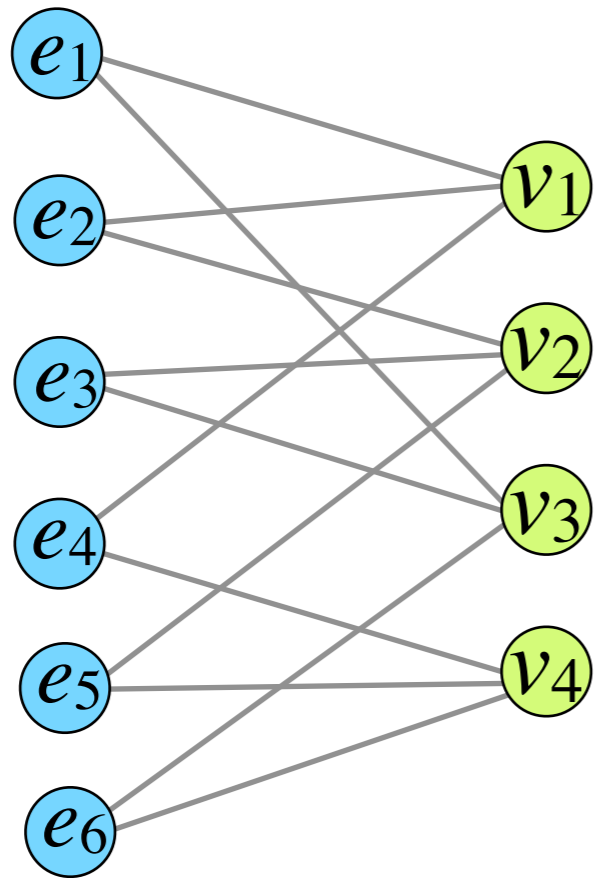
Raise a pair  $(\mathbf{x}, \mathbf{y})$  of infeasible primal *integral* solution  $\mathbf{x}$  and feasible dual solution  $\mathbf{y}$  continuously, satisfying:

$$\begin{array}{l} \forall i : y_i > 0 \implies \mathbf{A}_i \cdot \mathbf{x} \leq \alpha b_i \\ \forall j : x_j > 0 \implies \mathbf{A}_{\cdot j}^T \mathbf{y} = c_j \end{array}$$

for some  
 $\alpha \geq 1$

until  $\mathbf{x}$  becomes feasible.

$$\implies \mathbf{c}^T \mathbf{x} \leq \alpha \mathbf{b}^T \mathbf{y} \leq \alpha \text{OPT}_{LP} \leq \alpha \text{OPT}_{IP}$$



primal:

$$\begin{aligned} \min \quad & \sum_{v \in V} x_v \\ \text{s.t.} \quad & \sum_{v \in e} x_v \geq 1, \quad \forall e \in E \\ & x_v \in \{0, 1\}, \quad \forall v \in V \end{aligned}$$

dual-relax:

$$\begin{aligned} \max \quad & \sum_{e \in E} y_e \\ \text{s.t.} \quad & \sum_{e \ni v} y_e \leq 1, \quad \forall v \in V \\ & y_e \geq 0, \quad \forall e \in E \end{aligned}$$

vertex cover:

constraints

$$\sum_{v \in e} x_v \geq 1$$

variables

$$x_v \in \{0, 1\}$$

matching:

variables

$$y_e \in \{0, 1\}$$

constraints

$$\sum_{e \ni v} y_e \leq 1$$

Find feasible  $(x, y)$  such that:

$$\begin{aligned} \forall e : y_e > 0 & \implies \sum_{v \in e} x_v \leq \alpha \\ \forall v : x_v > 0 & \implies \sum_{e \ni v} y_e = 1 \end{aligned}$$



primal:

$$\begin{aligned} \min \quad & \sum_{v \in V} x_v \\ \text{s.t.} \quad & \sum_{v \in e} x_v \geq 1, \quad \forall e \in E \\ & x_v \in \{0, 1\}, \quad \forall v \in V \end{aligned}$$

dual-relax:

$$\begin{aligned} \max \quad & \sum_{e \in E} y_e \\ \text{s.t.} \quad & \sum_{e \ni v} y_e \leq 1, \quad \forall v \in V \\ & y_e \geq 0, \quad \forall e \in E \end{aligned}$$

initially  $\mathbf{x} = \mathbf{0}, \mathbf{y} = \mathbf{0}$ ;

while  $E \neq \emptyset$ : (**constraints currently violated by  $\mathbf{x}$** )

pick an  $e \in E$  and raise  $y_e$  ~~until  $\sum_{e \ni v} y_e = 1$  for some  $v \in V$~~ ; **to 1**

set  $x_v = 1$  for all such  $v \in e$  and remove all  $e' \ni v$  from  $E$ ;

$$\forall e : y_e > 0 \implies \sum_{v \in e} x_v \leq \alpha = 2$$

$$\forall v : x_v > 0 \implies \sum_{e \ni v} y_e = 1$$

**Complementary**

**slackness:**

$$\implies \sum_{v \in V} x_v \leq 2 \sum_{e \in E} y_e \leq 2 \text{OPT}$$

primal:

$$\begin{aligned} \min \quad & \sum_{v \in V} x_v \\ \text{s.t.} \quad & \sum_{v \in e} x_v \geq 1, \quad \forall e \in E \\ & x_v \in \{0, 1\}, \quad \forall v \in V \end{aligned}$$

dual-relax:

$$\begin{aligned} \max \quad & \sum_{e \in E} y_e \\ \text{s.t.} \quad & \sum_{e \ni v} y_e \leq 1, \quad \forall v \in V \\ & y_e \geq 0, \quad \forall e \in E \end{aligned}$$

initially  $\mathbf{x} = \mathbf{0}, \mathbf{y} = \mathbf{0}$ ;

while  $E \neq \emptyset$ : (**constraints violated by current  $\mathbf{x}$** )

pick an  $e \in E$  and raise  $y_e$  ~~until  $\sum_{e \ni v} y_e = 1$  for some  $v \in V$~~ ; **to 1**

set  $x_v = 1$  for all such  $v \in e$  and remove all  $e' \ni v$  from  $E$ ;  
(**constraints satisfied by current  $\mathbf{x}$** )

$$\forall e : y_e > 0 \implies \sum_{v \in e} x_v \leq \alpha = 2$$

$$\forall v : x_v > 0 \implies \sum_{e \ni v} y_e = 1$$

**Complementary**

**slackness:**

$$\implies \sum_{v \in V} x_v \leq 2 \sum_{e \in E} y_e \leq 2 \text{OPT}$$

primal:

$$\begin{aligned} \min \quad & \sum_{v \in V} x_v \\ \text{s.t.} \quad & \sum_{v \in e} x_v \geq 1, \quad \forall e \in E \\ & x_v \in \{0, 1\}, \quad \forall v \in V \end{aligned}$$

dual-relax:

$$\begin{aligned} \max \quad & \sum_{e \in E} y_e \\ \text{s.t.} \quad & \sum_{e \ni v} y_e \leq 1, \quad \forall v \in V \\ & y_e \geq 0, \quad \forall e \in E \end{aligned}$$

initially  $\mathbf{x} = \mathbf{0}, \mathbf{y} = \mathbf{0}$ ;

while  $E \neq \emptyset$ : (**constraints currently violated by  $\mathbf{x}$** )

pick an  $e \in E$  and raise  $y_e$  ~~until  $\sum_{e \ni v} y_e = 1$  for some  $v \in V$~~ ; **to 1**

set  $x_v = 1$  for all such  $v \in e$  and remove all  $e' \ni v$  from  $E$ ;

Find a **maximal matching**  $M \subseteq E$ ;

return  $C = \{v \mid \{u, v\} \in M\}$ ;

# Primal-Dual Schema

- **Modeling:** Express the optimization problem as an Integer Linear Program (ILP) and write its dual relaxed program.

$$\begin{array}{ll} \min & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{A} \mathbf{x} \geq \mathbf{b} \\ & \mathbf{x} \in \mathbb{Z}_{\geq 0} \end{array}$$

$$\begin{array}{ll} \max & \mathbf{b}^T \mathbf{y} \\ \text{s.t.} & \mathbf{y}^T \mathbf{A} \leq \mathbf{c}^T \\ & \mathbf{y} \geq \mathbf{0} \end{array}$$

- **Initialization:** Start from a primal infeasible solution  $\mathbf{x}$  and a dual feasible solution  $\mathbf{y}$  (usually  $\mathbf{x} = \mathbf{0}$ ,  $\mathbf{y} = \mathbf{0}$ ).
- Raise  $\mathbf{x}$  and  $\mathbf{y}$  until  $\mathbf{x}$  becomes feasible:
  - raise  $\mathbf{y}$  continuously until dual constraints getting tight  $A_{.j}^T \mathbf{y} = c_j$ ;
  - raise corresponding  $x_j$  integrally so that  $x_j > 0 \implies A_{.j}^T \mathbf{y} = c_j$ .
- Verify **complementary slackness condition:**

$$y_i > 0 \implies A_{i.} \mathbf{x} \leq \alpha b_i \implies \mathbf{c}^T \mathbf{x} \leq \alpha \mathbf{b}^T \mathbf{y} \leq \alpha OPT$$

# Integrality Gap

- minimum vertex cover of  $G(V, E)$ :

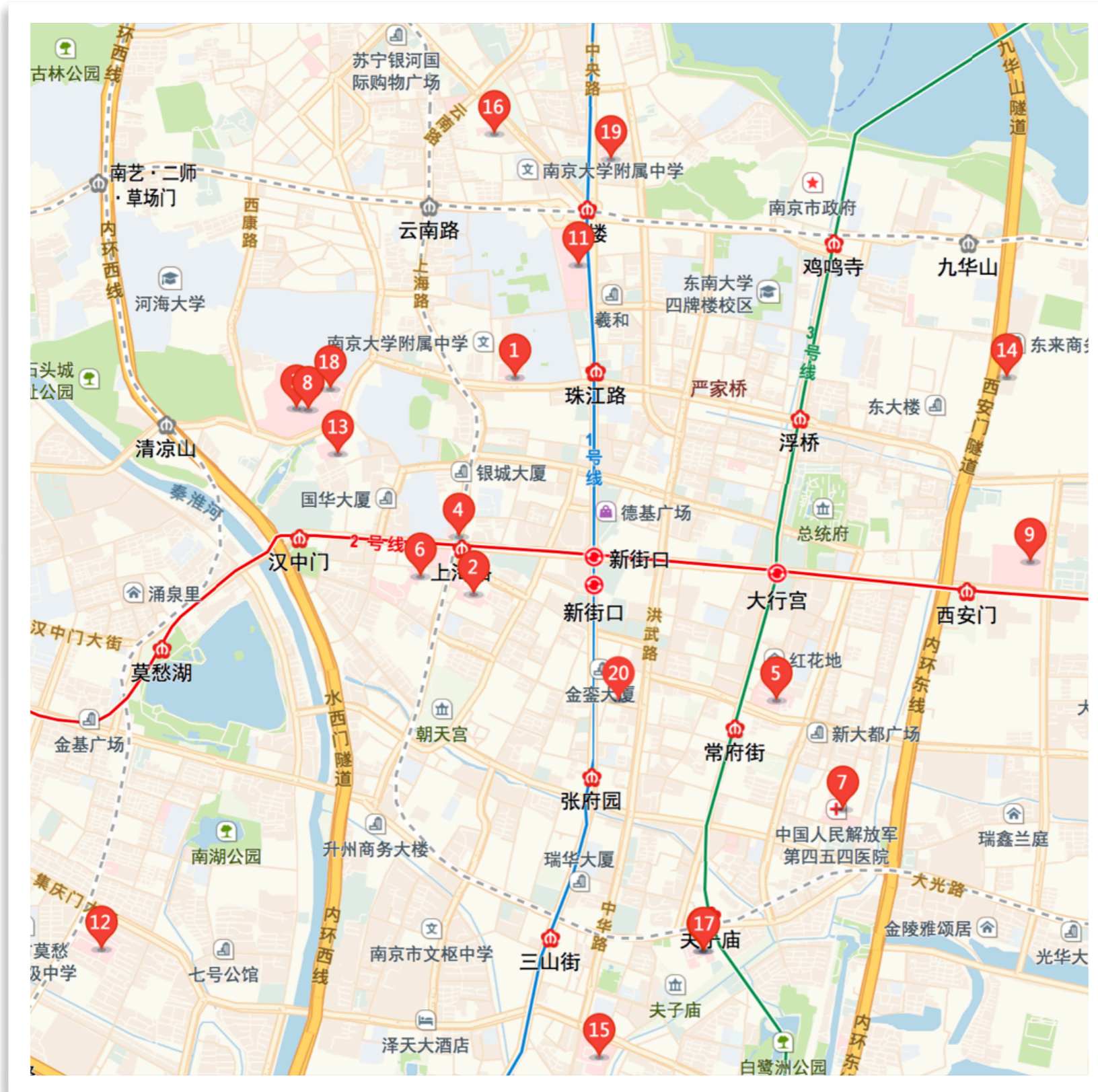
$$\begin{array}{ll} \text{minimize} & \sum_{v \in V} x_v \\ \text{subject to} & \sum_{v \in e} x_v \geq 1, \quad e \in E \\ & x_v \in \{0, 1\}, \quad v \in V \end{array}$$

$$\text{integrality gap} = \sup_G \frac{\text{OPT}(G)}{\text{OPT}_{\text{LP}}(G)}$$

- For LP relaxation of vertex cover: integrality gap = 2

# Facility Location

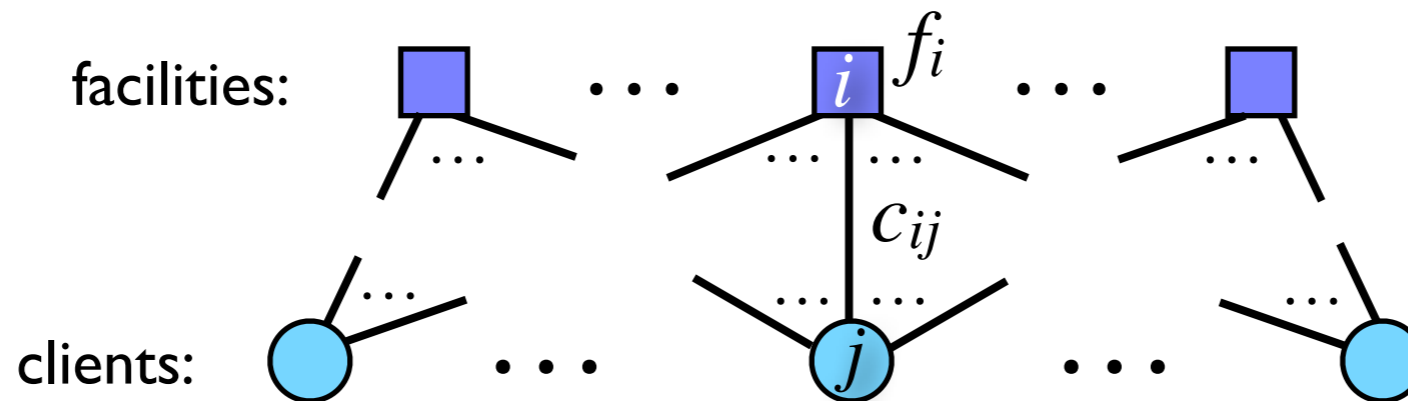
# Facility Location



hospitals  
in Nanjing



# Facility Location

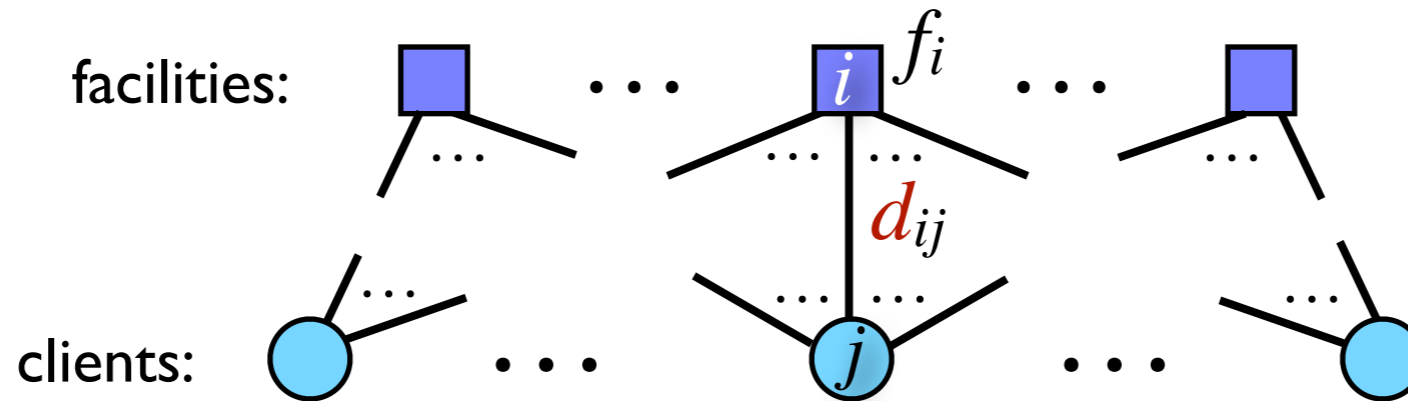


**Instance:** set  $F$  of **facilities**; set  $C$  of **clients**;  
facility opening costs  $f: F \rightarrow [0, \infty)$ ;  
connection costs  $c: F \times C \rightarrow [0, \infty)$ ;  
Find a subset  $I \subseteq F$  of **opening** facilities and a way  $\phi: C \rightarrow I$  of **connecting** all clients to them such that the total cost  $\sum_{j \in C} c_{\phi(j), j} + \sum_{i \in I} f_i$  is minimized.

- uncapacitated facility location;
- **NP**-hard; **AP**(Approximation Preserving)-reduction from Set Cover;
- [Dinur, Steuer 2014] no poly-time  $(1-o(1)) \ln n$ -approx. algorithm unless **NP** = **P**.



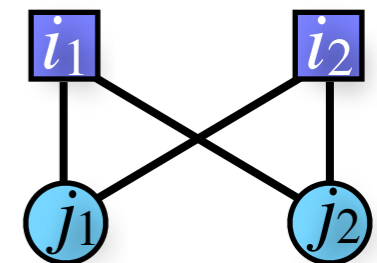
# Metric Facility Location



**Instance:** set  $F$  of **facilities**; set  $C$  of **clients**;  
facility opening costs  $f: F \rightarrow [0, \infty)$ ;  
connection **metric**  $d: F \times C \rightarrow [0, \infty)$ ;  
Find a subset  $I \subseteq F$  of **opening** facilities and a way  $\phi: C \rightarrow I$  of **connecting** all clients to them such that the total cost  $\sum_{j \in C} d_{\phi(j), j} + \sum_{i \in I} f_i$  is minimized.

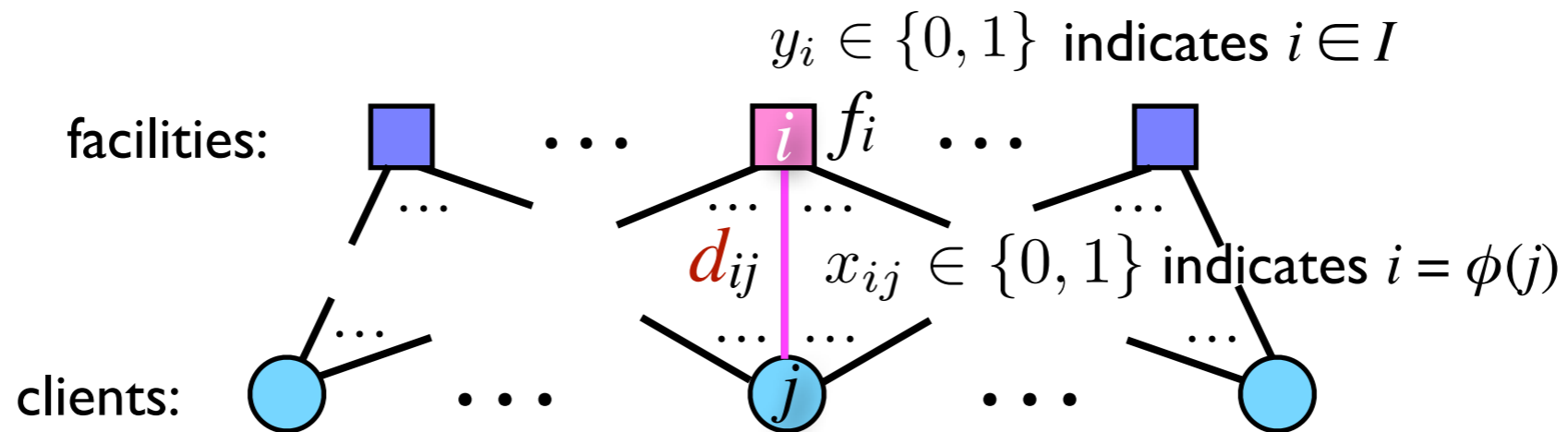
**triangle inequality:**  $\forall i_1, i_2 \in F, \forall j_1, j_2 \in C$

$$d_{i_1 j_1} + d_{i_2 j_1} + d_{i_2 j_2} \geq d_{i_1 j_2}$$



**Instance:** set  $F$  of **facilities**; set  $C$  of **clients**;  
 facility opening costs  $f: F \rightarrow [0, \infty)$ ;  
 connection **metric**  $d: F \times C \rightarrow [0, \infty)$ ;

Find  $\phi: C \rightarrow I \subseteq F$  to minimize  $\sum_{j \in C} d_{\phi(j), j} + \sum_{i \in I} f_i$



**LP-relaxation:**

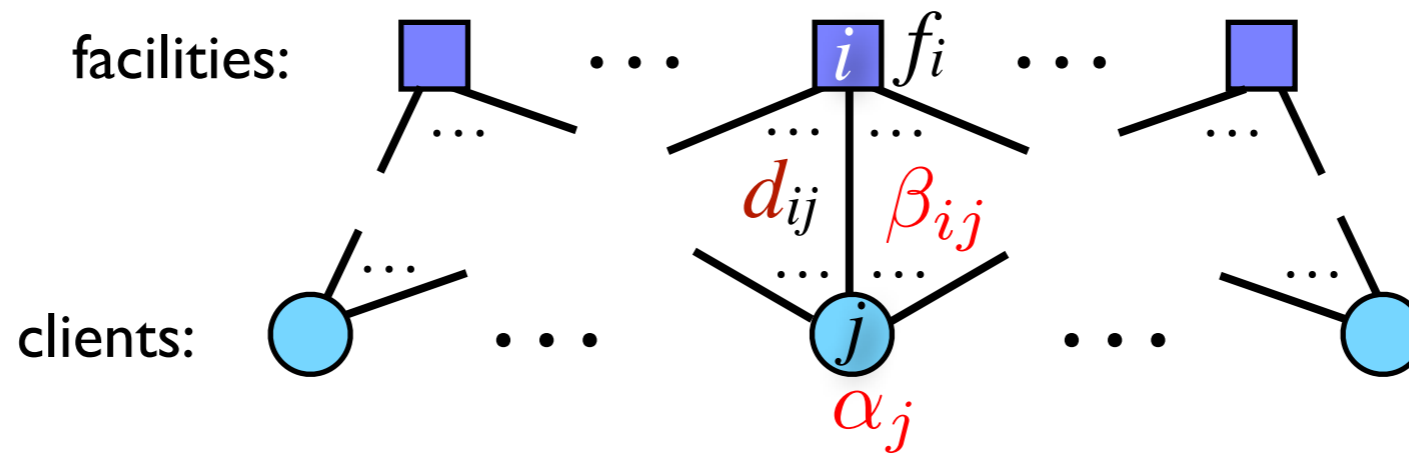
$$\min \sum_{i \in F, j \in C} d_{ij} x_{ij} + \sum_{i \in F} f_i y_i$$

**s.t.**

$$y_i \geq x_{ij}, \quad \forall i \in F, j \in C$$

$$\sum_{i \in F} x_{ij} \geq 1, \quad \forall j \in C$$

$$x_{ij}, y_i \geq 0, \quad \cancel{x_{ij}, y_i \in \{0, 1\}}, \quad \forall i \in F, j \in C$$



## Primal:

$$\begin{aligned} \min \quad & \sum_{i \in F, j \in C} d_{ij} x_{ij} + \sum_{i \in F} f_i y_i \\ \text{s.t.} \quad & y_i - x_{ij} \geq 0, \quad \forall i \in F, j \in C \\ & \sum_{i \in F} x_{ij} \geq 1, \quad \forall j \in C \\ & x_{ij}, y_i \in \{0, 1\}, \quad \forall i \in F, j \in C \end{aligned}$$

## Dual-relax:

$$\begin{aligned} \max \quad & \sum_{j \in C} \alpha_j \\ \text{s.t.} \quad & \alpha_j - \beta_{ij} \leq d_{ij}, \quad \forall i \in F, j \in C \\ & \sum_{j \in C} \beta_{ij} \leq f_i, \quad \forall i \in F \\ & \alpha_j, \beta_{ij} \geq 0, \quad \forall i \in F, j \in C \end{aligned}$$

$\alpha_j$ : amount of value paid by client  $j$  to all facilities

$\beta_{ij} \geq \alpha_j - d_{ij}$ : payment to facility  $i$  by client  $j$  (after deduction)

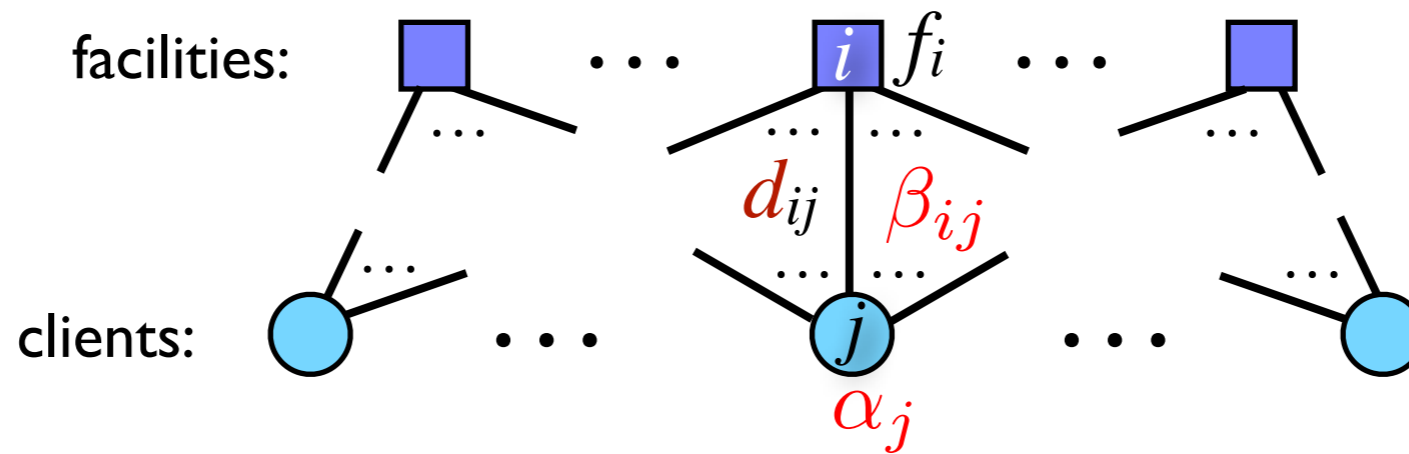
complimentary  
slackness conditions:  
(if ideally held)

$$x_{ij} = 1 \Rightarrow \alpha_j - \beta_{ij} = d_{ij};$$

$$y_i = 1 \Rightarrow \sum_{j \in C} \beta_{ij} = f_i;$$

$$\alpha_j > 0 \Rightarrow \sum_{i \in F} x_{ij} = 1;$$

$$\beta_{ij} > 0 \Rightarrow y_i = x_{ij};$$



$$\min \sum_{i \in F, j \in C} d_{ij} x_{ij} + \sum_{i \in F} f_i y_i$$

$$\text{s.t. } y_i - x_{ij} \geq 0, \quad \forall i \in F, j \in C$$

$$\sum_{i \in F} x_{ij} \geq 1, \quad \forall j \in C$$

$$x_{ij}, y_i \in \{0, 1\}, \quad \forall i \in F, j \in C$$

$$\max \sum_{j \in C} \alpha_j$$

$$\text{s.t. } \alpha_j - \beta_{ij} \leq d_{ij}, \quad \forall i \in F, j \in C$$

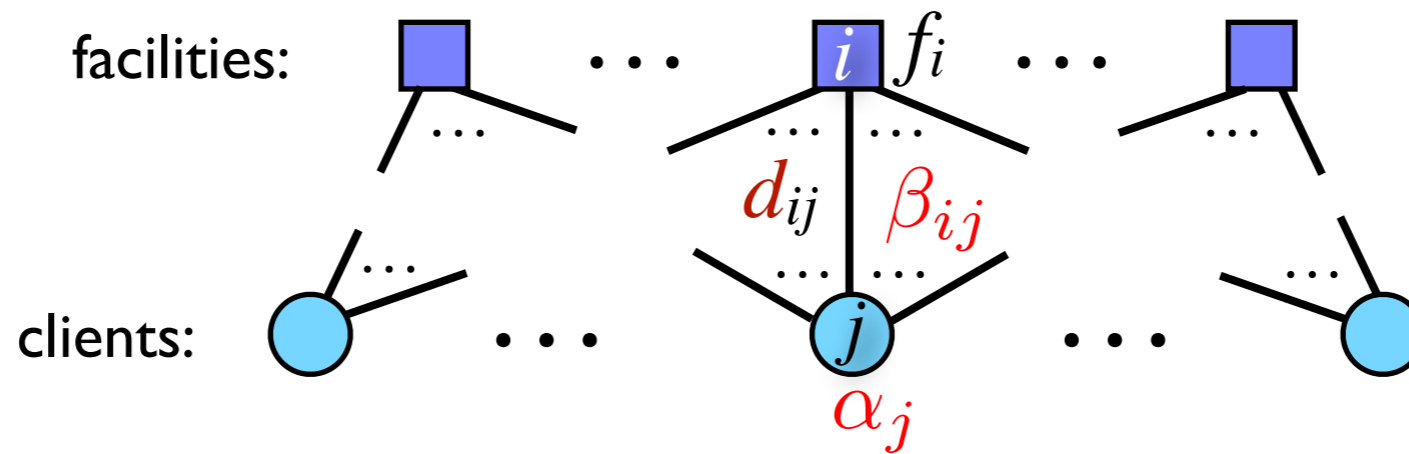
$$\sum_{j \in C} \beta_{ij} \leq f_i, \quad \forall i \in F$$

$$\alpha_j, \beta_{ij} \geq 0, \quad \forall i \in F, j \in C$$

Initially  $\alpha = \mathbf{0}$ ,  $\beta = \mathbf{0}$ , no facility is *open*, no client is *connected*;

raise  $\alpha_j$  for all client  $j$  *simultaneously* at a *uniform continuous* rate:

- upon  $\alpha_j = d_{ij}$  for a closed facility  $i$ : edge  $(i, j)$  is *paid*; fix  $\beta_{ij} = \alpha_j - d_{ij}$  as  $\alpha_j$  being raised;
- upon  $\sum_{j \in C} \beta_{ij} = f_i$ : *tentatively open* facility  $i$ ; all *unconnected* clients  $j$  with *paid* edge  $(i, j)$  to facility  $i$  are declared *connected* to facility  $i$  and stop raising  $\alpha_j$ ;
- upon  $\alpha_j = d_{ij}$  for an *unconnected* client  $j$  and *tentatively open* facility  $i$ : client  $j$  is declared *connected* to facility  $i$ : and stop raising  $\alpha_j$ ;

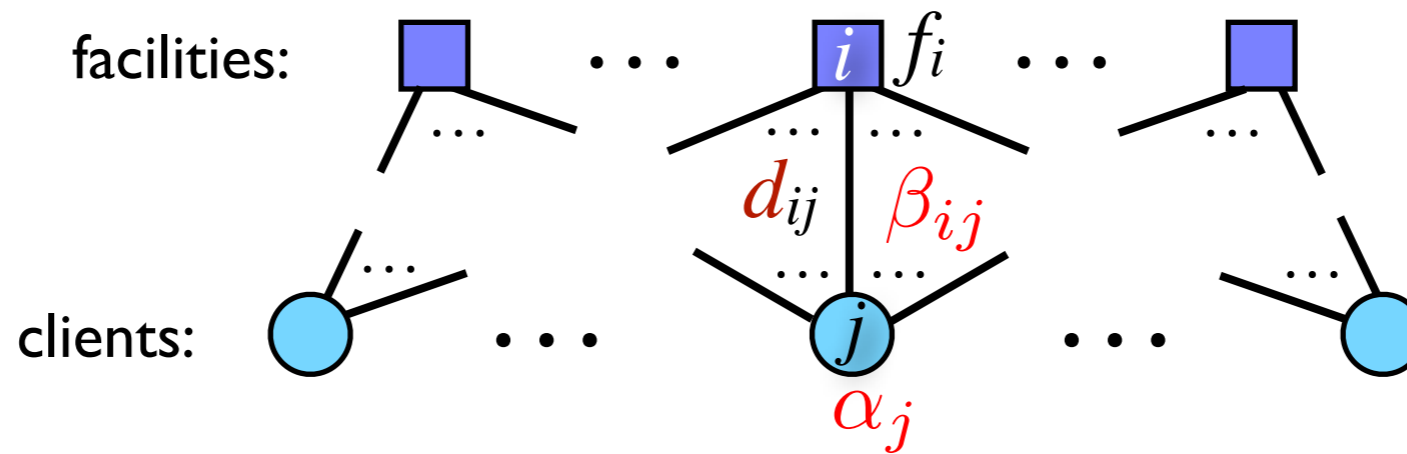


Initially  $\alpha = \mathbf{0}$ ,  $\beta = \mathbf{0}$ , no facility is *open*, no client is *connected*;  
 raise  $\alpha_j$  for all client  $j$  *simultaneously* at a *uniform continuous* rate:

- upon  $\alpha_j = d_{ij}$  for a closed facility  $i$ : edge  $(i, j)$  is *paid*; fix  $\beta_{ij} = \alpha_j - d_{ij}$  as  $\alpha_j$  being raised;
- upon  $\sum_{j \in C} \beta_{ij} = f_i$ : *tentatively open* facility  $i$ ; all *unconnected* clients  $j$  with *paid* edge  $(i, j)$  to facility  $i$  are declared *connected* to facility  $i$  and stop raising  $\alpha_j$ ;
- upon  $\alpha_j = d_{ij}$  for an *unconnected* client  $j$  and *tentatively open* facility  $i$ : client  $j$  is declared *connected* to facility  $i$ : and stop raising  $\alpha_j$ ;

- The events that occur at the same time are processed in arbitrary order.
- Fully paid facilities are tentatively open:  $\sum_{j \in C} \beta_{ij} = f_i$
- Each client is connected through a tight edge ( $\alpha_j - \beta_{ij} = d_{ij}$ ) to an open facility.
- Eventually all clients connect to tentatively opening facilities.

**A client may have tight edges to more than one facilities:  
 We might have opened more facilities than necessary!**



## Phase I:

Initially  $\alpha = \mathbf{0}$ ,  $\beta = \mathbf{0}$ , no facility is *open*, no client is *connected*;  
raise  $\alpha_j$  for all client  $j$  *simultaneously* at a *uniform continuous* rate:

- upon  $\alpha_j = d_{ij}$  for a closed facility  $i$ : edge  $(i, j)$  is *paid*; fix  $\beta_{ij} = \alpha_j - d_{ij}$  as  $\alpha_j$  being raised;
- upon  $\sum_{j \in C} \beta_{ij} = f_i$ : *tentatively open* facility  $i$ ; all *unconnected* clients  $j$  with *paid* edge  $(i, j)$  to facility  $i$  are declared *connected* to facility  $i$  and stop raising  $\alpha_j$ ;
- upon  $\alpha_j = d_{ij}$  for an *unconnected* client  $j$  and *tentatively open* facility  $i$ : client  $j$  is declared *connected* to facility  $i$  and stop raising  $\alpha_j$ ;

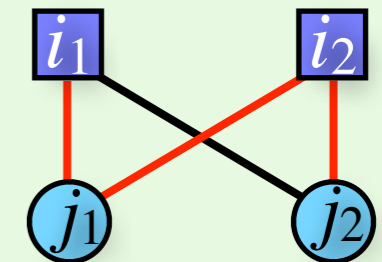
## Phase II:

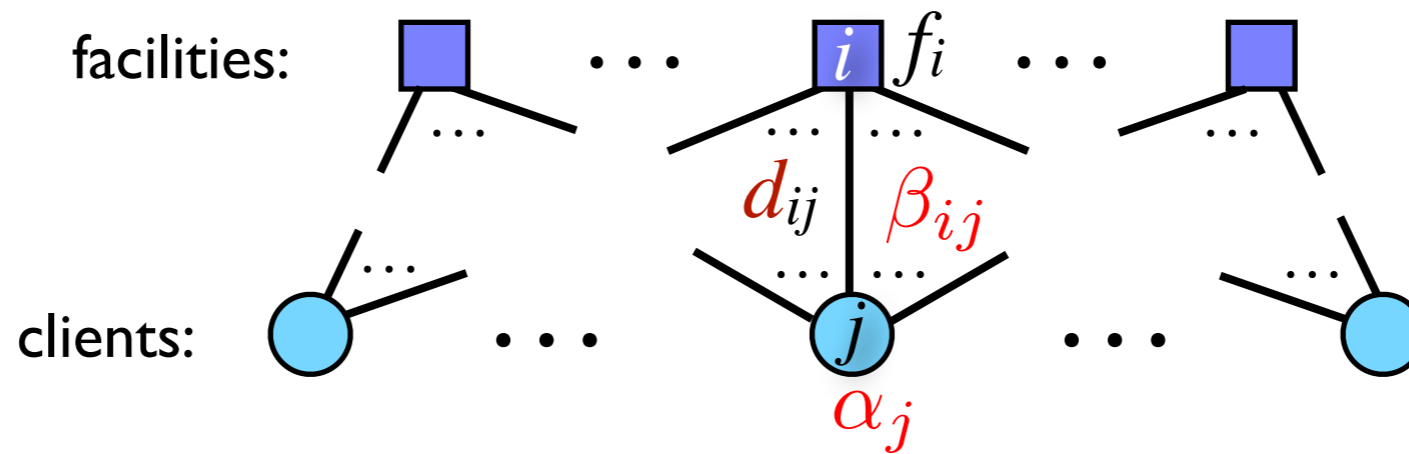
construct graph  $G(V, E)$  where  $V = \{\text{tentatively open facilities}\}$

and  $\{i_1, i_2\} \in E$  if  $\exists$  client  $j$  s.t. both  $\beta_{i_1 j} > 0$  and  $\beta_{i_2 j} > 0$  in **Phase I**;

find a *maximal independent set*  $I$  of  $G$  and *permanently open* facilities in  $I$ ;

For each client  $j$ : if the facility  $i$  with  $\beta_{ij} > 0$  has  $i \in I$  or  $j$ 's *connecting witness*  $i$  has  $i \in I$ , then  $j$  is connected to  $i$  (*directly connected*); otherwise, client  $j$  is connected to an arbitrary  $i' \in I$  that is adjacent (in  $G$ ) to  $j$ 's *connecting witness*  $i$  (*indirectly connected*);





## Primal:

$$\begin{aligned} \min \quad & \sum_{i \in F, j \in C} d_{ij} x_{ij} + \sum_{i \in F} f_i y_i \\ \text{s.t.} \quad & y_i - x_{ij} \geq 0, \quad \forall i \in F, j \in C \\ & \sum_{i \in F} x_{ij} \geq 1, \quad \forall j \in C \\ & x_{ij}, y_i \in \{0, 1\}, \quad \forall i \in F, j \in C \end{aligned}$$

## Dual-relax:

$$\begin{aligned} \max \quad & \sum_{j \in C} \alpha_j \\ \text{s.t.} \quad & \alpha_j - \beta_{ij} \leq d_{ij}, \quad \forall i \in F, j \in C \\ & \sum_{j \in C} \beta_{ij} \leq f_i, \quad \forall i \in F \\ & \alpha_j, \beta_{ij} \geq 0, \quad \forall i \in F, j \in C \end{aligned}$$

$\alpha_j$ : amount of value paid by client  $j$  to all facilities

$\beta_{ij} \geq \alpha_j - d_{ij}$ : payment to facility  $i$  by client  $j$  (after deduction)

complimentary  
slackness conditions:  
(if ideally held)

$$x_{ij} = 1 \Rightarrow \alpha_j - \beta_{ij} = d_{ij};$$

$$y_i = 1 \Rightarrow \sum_{j \in C} \beta_{ij} = f_i;$$

$$\alpha_j > 0 \Rightarrow \sum_{i \in F} x_{ij} = 1;$$

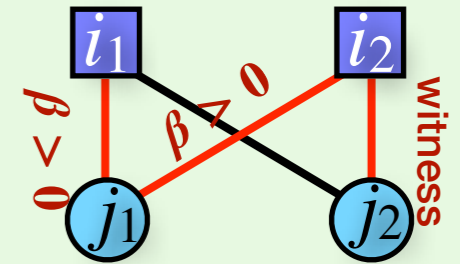
$$\beta_{ij} > 0 \Rightarrow y_i = x_{ij};$$



## Phase I:

Initially  $\alpha = \mathbf{0}$ ,  $\beta = \mathbf{0}$ , no facility is *open*, no client is *connected*;  
raise  $\alpha_j$  for all client  $j$  *simultaneously* at a *uniform continuous* rate:

- upon  $\alpha_j = d_{ij}$  for a closed facility  $i$ : edge  $(i, j)$  is *paid*; fix  $\beta_{ij} = \alpha_j - d_{ij}$  as  $\alpha_j$  being raised;
- upon  $\sum_{j \in C} \beta_{ij} = f_i$ : *tentatively open* facility  $i$ ; all *unconnected* clients  $j$  with *paid* edge  $(i, j)$  to facility  $i$  are declared *connected* to facility  $i$  and stop raising  $\alpha_j$ ;
- upon  $\alpha_j = d_{ij}$  for an *unconnected* client  $j$  and *tentatively open* facility  $i$ : client  $j$  is declared *connected* to facility  $i$ ; and stop raising  $\alpha_j$ ;



## Phase II:

construct graph  $G(V, E)$  where  $V = \{\text{tentatively open facilities}\}$

and  $\{i_1, i_2\} \in E$  if  $\exists$  client  $j$  s.t. both  $\beta_{i_1 j} > 0$  and  $\beta_{i_2 j} > 0$  in **Phase I**;

find a *maximal independent set*  $I$  of  $G$  and *permanently open* facilities in  $I$ ;

For each client  $j$ : if the facility  $i$  with  $\beta_{ij} > 0$  has  $i \in I$  or  $j$ 's *connecting witness*  $i$  has  $i \in I$ , then  $j$  is connected to  $i$  (*directly connected*); otherwise, client  $j$  is connected to an arbitrary  $i' \in I$  that is adjacent (in  $G$ ) to  $j$ 's *connecting witness*  $i$  (*indirectly connected*);

Denote by  $\phi$  the output mapping from clients to facilities.

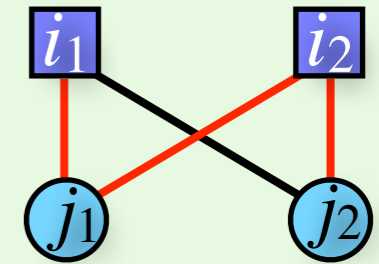
$$\begin{aligned}
 SOL &= \sum_{i \in I} f_i + \sum_{j: \text{directly connected}} d_{\phi(j)j} + \sum_{j: \text{indirectly connected}} d_{\phi(j)j} \leq 3 \sum_{j \in C} \alpha_j \leq 3 OPT \\
 &\leq \sum_{j: \text{directly connected}} \alpha_j \quad \text{triangle inequality + maximality of } I \leq 3 \sum_{j: \text{indirectly connected}} \alpha_j
 \end{aligned}$$



## Phase I:

Initially  $\alpha = \mathbf{0}$ ,  $\beta = \mathbf{0}$ , no facility is *open*, no client is *connected*;  
raise  $\alpha_j$  for all client  $j$  *simultaneously* at a *uniform continuous* rate:

- upon  $\alpha_j = d_{ij}$  for a closed facility  $i$ : edge  $(i, j)$  is *paid*; fix  $\beta_{ij} = \alpha_j - d_{ij}$  as  $\alpha_j$  being raised;
- upon  $\sum_{j \in C} \beta_{ij} = f_i$ : *tentatively open* facility  $i$ ; all *unconnected* clients  $j$  with *paid* edge  $(i, j)$  to facility  $i$  are declared *connected* to facility  $i$  and stop raising  $\alpha_j$ ;
- upon  $\alpha_j = d_{ij}$  for an *unconnected* client  $j$  and *tentatively open* facility  $i$ : client  $j$  is declared *connected* to facility  $i$ ; and stop raising  $\alpha_j$ ;



## Phase II:

construct graph  $G(V, E)$  where  $V = \{\text{tentatively open facilities}\}$

and  $\{i_1, i_2\} \in E$  if  $\exists$  client  $j$  s.t. both  $\beta_{i_1 j} > 0$  and  $\beta_{i_2 j} > 0$  in **Phase I**;

find a *maximal independent set*  $I$  of  $G$  and *permanently open* facilities in  $I$ ;

For each client  $j$ : if the facility  $i$  with  $\beta_{ij} > 0$  has  $i \in I$  or  $j$ 's *connecting witness*  $i$  has  $i \in I$ , then  $j$  is connected to  $i$  (*directly connected*); otherwise, client  $j$  is connected to an arbitrary  $i' \in I$  that is adjacent (in  $G$ ) to  $j$ 's *connecting witness*  $i$  (*indirectly connected*);

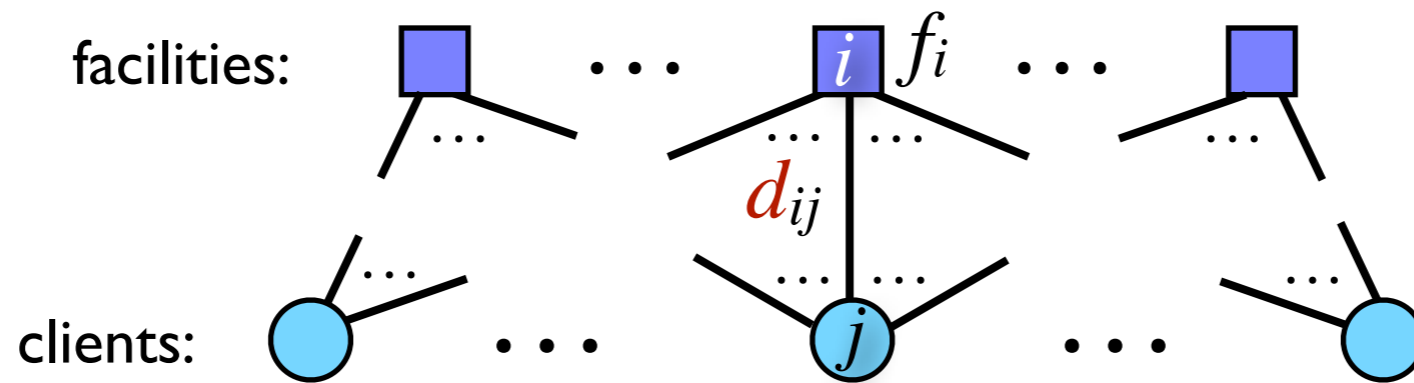
$$SOL \leq 3 OPT$$

can be implemented *discretely*: in  $O(m \log m)$  time,  $m = |F||C|$

- sort all edges  $(i, j) \in F \times C$  by non-decreasing  $d_{ij}$
- dynamically maintain the time of next event by heap

**Instance:** set  $F$  of **facilities**; set  $C$  of **clients**;  
 facility opening costs  $f: F \rightarrow [0, \infty)$ ;  
 connection **metric**  $d: F \times C \rightarrow [0, \infty)$ ;

Find  $\phi: C \rightarrow I \subseteq F$  to minimize  $\sum_{j \in C} d_{\phi(j),j} + \sum_{i \in I} f_i$



$$\min \sum_{i \in F, j \in C} d_{ij} x_{ij} + \sum_{i \in F} f_i y_i$$

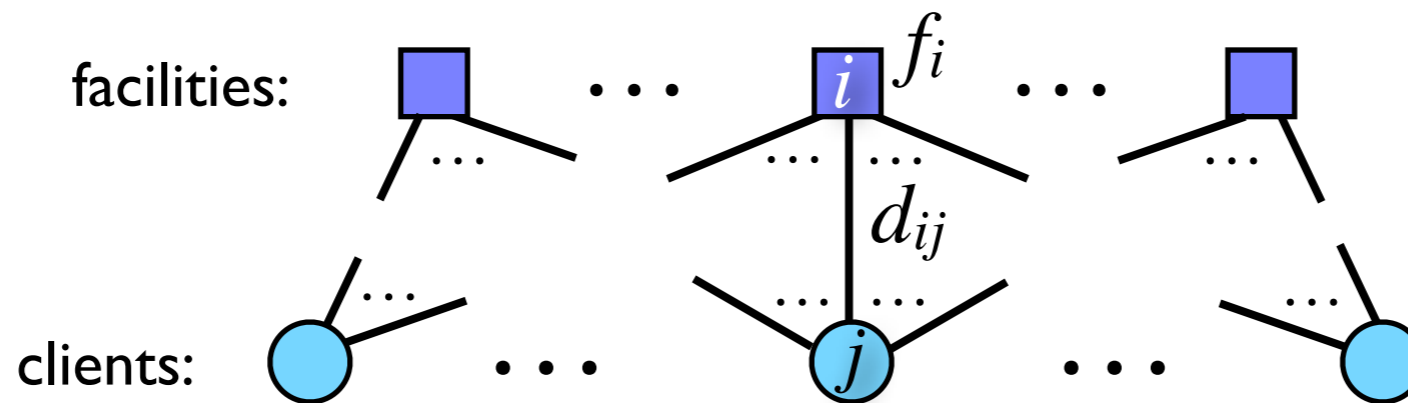
$$\text{s.t. } y_i - x_{ij} \geq 0, \quad \forall i \in F, j \in C$$

$$\sum_{i \in F} x_{ij} \geq 1, \quad \forall j \in C$$

$$x_{ij}, y_i \in \{0, 1\}, \quad \forall i \in F, j \in C$$

- **Integrality gap = 3**
- no poly-time  $< 1.463$ -approx. algorithm unless **NP=P**
- [Li 2011] 1.488-approx. algorithm

# $k$ -Median



**Instance:** set  $F$  of **facilities**; set  $C$  of **clients**;  
connection metric  $d: F \times C \rightarrow [0, \infty)$ ;

Find a subset  $I \subseteq F$  of  $\leq k$  opening facilities and a way  $\phi: C \rightarrow I$  of connecting all clients to them such that the total cost  $\sum_{j \in C} d_{\phi(j), j}$  is minimized.