



# Chapter 6 Graph Algorithms

# Algorithm Theory WS 2016/17



**Given:** Directed network with positive edge capacities

**Sources & Sinks:** Instead of one source and one destination, several sources that generate flow and several sinks that absorb flow.

Supply & Demand: sources have supply values, sinks demand values

**Goal:** Compute a flow such that source supplies and sink demands are exactly satisfied

• The circulation problem is a feasibility rather than a maximization problem

# Circulations with Demands: Formally



**Given:** Directed network G = (V, E) with

- Edge capacities  $c_e > 0$  for all  $e \in E$
- Node demands  $d_v \in \mathbb{R}$  for all  $v \in V$ 
  - $d_{v} > 0$ : node needs flow and therefore is a sink
  - $d_{v} < 0$ : node has a supply of  $-d_{v}$  and is therefore a source
  - $d_v = 0$ : node is neither a source nor a sink

**Flow:** Function  $f: E \to \mathbb{R}_{\geq 0}$  satisfying

- Capacity Conditions:  $\forall e \in E: 0 \leq f(e) \leq c_e$
- Demand Conditions:  $\forall v \in V$ :  $f^{in}(v) f^{out}(v) = d_v$

### **Objective:** Does a flow f satisfying all conditions exist? If yes, find such a flow f.

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### Example





### **Condition on Demands**

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**Claim:** If there exists a feasible circulation with demands  $d_v$  for  $v \in V$ , then



**Proof:** 

• 
$$\sum_{v} d_{v} = \sum_{v} \left( f^{\text{in}}(v) - f^{\text{out}}(v) \right)$$

f(e) of each edge e appears twice in the above sum with different signs → overall sum is 0

#### Total supply = total demand:

Define 
$$D \coloneqq \sum_{v:d_v>0} d_v = \sum_{v:d_v<0} -d_v$$

### **Reduction to Maximum Flow**







### Example





### Formally...



**Reduction:** Get graph G' from graph as follows

- Node set of G' is  $V \cup \{s^*, t^*\}$
- Edge set is *E* and edges
  - $-(s^*, v)$  for all v with  $d_v < 0$ , capacity of edge is  $d_v$
  - (v,  $t^*$ ) for all v with  $d_v > 0$ , capacity of edge is  $d_v$

### **Observations:**

- Capacity of min  $s^*-t^*$  cut is at most D (e.g., the cut  $(s^*, V \cup \{t^*\})$
- A feasible circulation on G can be turned into a feasible flow of value D of G' by saturating all (s\*, v) and (v, t\*) edges.
- Any flow of G' of value D induces a feasible circulation on G
  - $(s^*, v)$  and  $(v, t^*)$  edges are saturated
  - By removing these edges, we get exactly the demand constraints

## **Circulation with Demands**



**Theorem:** There is a feasible circulation with demands  $d_v, v \in V$  on graph G if and only if there is a flow of value D on G'.

If all capacities and demands are integers, there is an integer circulation

The max flow min cut theorem also implies the following:

**Theorem:** The graph G has a feasible circulation with demands  $d_v, v \in V$  if and only if for all cuts (A, B),

$$\sum_{\nu\in B}d_{\nu}\leq c(A,B).$$

# **Circulation: Demands and Lower Bounds**



**Given:** Directed network G = (V, E) with

- Edge capacities  $c_e > 0$  and lower bounds  $0 \le \ell_e \le c_e$  for  $e \in E$
- Node demands  $d_v \in \mathbb{R}$  for all  $v \in V$ 
  - $d_{v} > 0$ : node needs flow and therefore is a sink
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  - $d_v = 0$ : node is neither a source nor a sink

**Flow:** Function  $f: E \to \mathbb{R}_{\geq 0}$  satisfying

- Capacity Conditions:  $\forall e \in E$ :  $\ell_e \leq f(e) \leq c_e$
- Demand Conditions:  $\forall v \in V$ :  $f^{in}(v) f^{out}(v) = d_v$

### **Objective:** Does a flow f satisfying all conditions exist? If yes, find such a flow f.

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### Solution Idea



- Define initial circulation  $f_0(e) = \ell_e$ Satisfies capacity constraints:  $\forall e \in E : \ell_e \leq f_0(e) \leq c_e$
- Define

$$L_{v} \coloneqq f_{0}^{\mathrm{in}}(v) - f_{0}^{\mathrm{out}}(v) = \sum_{e \mathrm{into} v} \ell_{e} - \sum_{e \mathrm{out} \mathrm{of} v} \ell_{e}$$

• If  $L_v = d_v$ , demand condition is satisfied at v by  $f_0$ , otherwise, we need to superimpose another circulation  $f_1$  such that

$$d'_{\nu} \coloneqq f_1^{\text{in}}(\nu) - f_1^{\text{out}}(\nu) = d_{\nu} - L_{\nu}$$

- Remaining capacity of edge  $e: c'_e \coloneqq c_e \ell_e$
- We get a circulation problem with new demands  $d'_{v}$ , new capacities  $c'_{e}$ , and no lower bounds

### Eliminating a Lower Bound: Example



# Reduce to Problem Without Lower Bounds

Graph G = (V, E):

- Capacity: For each edge  $e \in E$ :  $\ell_e \leq f(e) \leq c_e$
- Demand: For each node  $v \in V$ :  $f^{in}(v) f^{out}(v) = d_v$

Model lower bounds with supplies & demands:



#### Create Network G' (without lower bounds):

- For each edge  $e \in E: c'_e = c_e \ell_e$
- For each node  $v \in V: d'_v = d_v L_v$

# **Circulation: Demands and Lower Bounds**



**Theorem:** There is a feasible circulation in G (with lower bounds) if and only if there is feasible circulation in G' (without lower bounds).

- Given circulation f' in G',  $f(e) = f'(e) + \ell_e$  is circulation in G
  - The capacity constraints are satisfied because  $f'(e) \leq c_e \ell_e$
  - Demand conditions:

$$f^{\text{in}}(v) - f^{\text{out}}(v) = \sum_{e \text{ into } v} \left(\ell_e + f'(e)\right) - \sum_{e \text{ out of } v} \left(\ell_e + f'(e)\right)$$
$$= L_v + \left(d_v - L_v\right) = d_v$$

- Given circulation f in G,  $f'(e) = f(e) \ell_e$  is circulation in G'
  - The capacity constraints are satisfied because  $\ell_e \leq f(e) \leq c_e$
  - Demand conditions:

$$f^{\prime \text{in}}(v) - f^{\prime \text{out}}(v) = \sum_{e \text{ into } v} (f(e) - \ell_e) - \sum_{e \text{ out of } v} (f(e) - \ell_e)$$
$$= d_v - L_v$$

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# Integrality



**Theorem:** Consider a circulation problem with integral capacities, flow lower bounds, and node demands. If the problem is feasible, then it also has an integral solution.

#### **Proof:**

- Graph G' has only integral capacities and demands
- Thus, the flow network used in the reduction to solve circulation with demands and no lower bounds has only integral capacities
- The theorem now follows because a max flow problem with integral capacities also has an optimal integral solution
- It also follows that with the max flow algorithms we studied, we get an integral feasible circulation solution.

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### Matrix Rounding



- **Given:**  $p \times q$  matrix  $D = \{d_{i,j}\}$  of real numbers
- row *i* sum:  $a_i = \sum_j d_{i,j}$ , column *j* sum:  $b_j = \sum_i d_{i,j}$
- Goal: Round each d<sub>i,j</sub>, as well as a<sub>i</sub> and b<sub>j</sub> up or down to the next integer so that the sum of rounded elements in each row (column) equals the rounded row (column) sum
- Original application: publishing census data

#### Example:

3.14	6.80	7.30	17.24
9.60	2.40	0.70	12.70
3.60	1.20	6.50	11.30
16.34	10.40	14.50	



3	7	7	17
10	2	1	13
3	1	7	11
16	10	15	

### possible rounding

original data

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**Theorem:** For any matrix, there exists a feasible rounding.

**Remark:** Just rounding to the nearest integer doesn't work

0.35	0.35	0.35	1.05
0.55	0.55	0.55	1.65
0.90	0.90	0.90	

original data

0	0	0	0
1	1	1	3
1	1	1	

#### rounding to nearest integer

0	0	1	1
1	1	0	2
1	1	1	

feasible rounding

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### **Reduction to Circulation**



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3.60	1.20	6.50	11.30
16.34	10.40	14.50	

Matrix elements and row/column sums give a feasible circulation that satisfies all lower bound, capacity, and demand constraints

columns:

rows:





**Theorem:** For any matrix, there exists a feasible rounding.

#### **Proof:**

- The matrix entries  $d_{i,j}$  and the row and column sums  $a_i$  and  $b_j$  give a feasible circulation for the constructed network
- Every feasible circulation gives matrix entries with corresponding row and column sums (follows from demand constraints)
- Because all demands, capacities, and flow lower bounds are integral, there is an integral solution to the circulation problem

### → gives a feasible rounding!



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### Gifts-Children Graph

• Which child likes which gift can be represented by a graph





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### Matching



#### Matching: Set of pairwise non-incident edges



Maximal Matching: A matching s.t. no more edges can be added

Maximum Matching: A matching of maximum possible size



**Perfect Matching:** Matching of size n/2 (every node is matched)

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### Bipartite Graph



**Definition:** A graph G = (V, E) is called bipartite iff its node set can be partitioned into two parts  $V = V_1 \cup V_2$  such that for each edge  $\{u, v\} \in E$ ,

 $|\{u, v\} \cap V_1| = 1.$ 

• Thus, edges are only between the two parts



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### Santa's Problem

#### **Maximum Matching in Bipartite Graphs:**

Every child can get a gift iff there is a matching of size #children

Clearly, every matching is at most as big

If #children = #gifts, there is a solution iff there is a perfect matching



### **Reducing to Maximum Flow**





### all capacities are 1

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# **Reducing to Maximum Flow**



**Theorem:** Every integer solution to the max flow problem on the constructed graph induces a maximum bipartite matching of *G*.

### **Proof:**

- 1. An integer flow f of value |f| induces a matching of size |f|
  - Left nodes (gifts) have incoming capacity 1
  - Right nodes (children) have outgoing capacity 1
  - Left and right nodes are incident to  $\leq 1$  edge e of G with f(e) = 1
- 2. A matching of size k implies a flow f of value |f| = k
  - For each edge  $\{u, v\}$  of the matching:

$$f((s,u)) = f((u,v)) = f((v,t)) = 1$$

All other flow values are 0

# Running Time of Max. Bipartite Matching



**Theorem:** A maximum matching of a bipartite graph can be computed in time  $O(m \cdot n)$ .

## Perfect Matching?



- There can only be a perfect matching if both sides of the partition have size n/2.
- There is no perfect matching, iff there is an *s*-*t* cut of size < <sup>n</sup>/<sub>2</sub> in the flow network.



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### *s*-*t* Cuts





Partition (A, B) of node set such that  $s \in A$  and  $t \in B$ 

- If  $v_i \in A$ : edge  $(v_i, t)$  is in cut (A, B)
- If  $u_i \in B$ : edge  $(s, u_i)$  is in cut (A, B)
- Otherwise (if u<sub>i</sub> ∈ A, v<sub>i</sub> ∈ B), all edges from u<sub>i</sub> to some v<sub>j</sub> ∈ B are in cut (A, B)

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### Hall's Marriage Theorem



**Theorem:** A bipartite graph  $G = (U \cup V, E)$  for which |U| = |V|has a perfect matching if and only if  $\forall U' \subseteq U: |N(U')| \ge |U'|$ ,

where  $N(U') \subseteq V$  is the set of neighbors of nodes in U'.

**Proof:** No perfect matching  $\Leftrightarrow$  some *s*-*t* cut has capacity < n/2

1. Assume there is U' for which |N(U')| < |U'|:



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2. Assume that there is a cut (A, B) of capacity < n/2



### Hall's Marriage Theorem



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**Proof:** No perfect matching  $\Leftrightarrow$  some *s*-*t* cut has capacity < n/2

2. Assume that there is a cut (A, B) of capacity < n/2

$$|U'| = \frac{n}{2} - x$$
$$|N(U')| \le y + z$$
$$x + y + z < \frac{n}{2}$$