### Design and Analysis of Algorithms 6.046J/18.401J



### LECTURE 14

#### Network Flow & Applications

- Review
- Max-flow min-cut theorem
- Edmonds Karp algorithm
- Flow Integrality
- Part II: Applications



### **Recall from Lecture 13**

- *Flow value:* |f| = f(s, V).
- *Cut*: Any partition (S, T) of V such that  $s \in S$  and  $t \in T$ .
- Lemma. |f| = f(S, T) for any cut (S, T).
- **Corollary.**  $|f| \le c(S, T)$  for any cut (S, T).
- *Residual graph:* The graph  $G_{f=}(V, E_f)$  with strictly positive *residual capacities*  $c_f(u, v) = c(u, v) f(u, v) > 0$ .
- *Augmenting path:* Any path from s to t in  $G_f$ .
- *Residual capacity* of an augmenting path:

$$c_f(p) = \min_{(u,v) \in p} \{c_f(u,v)\}.$$

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#### **Algorithm:**

 $f[u, v] \leftarrow 0 \text{ for all } u, v \in V$ while an augmenting path p in G wrt f exists do augment f by  $c_f(p)$ 



### Max-flow, min-cut theorem

**Theorem.** The following are equivalent: 1. |f| = c(S, T) for some cut (S, T). 2. *f* is a maximum flow. 3. *f* admits no augmenting paths.

#### Proof.

(1)  $\Rightarrow$  (2): Since  $|f| \le c(S, T)$  for any cut (S, T), the assumption that |f| = c(S, T) implies that f is a maximum flow.

(2)  $\Rightarrow$  (3): If there were an augmenting path, the flow value could be increased, contradicting the maximality of *f*.



### **Proof (continued)**

(3)  $\Rightarrow$  (1): Suppose that *f* admits no augmenting paths. Define  $S = \{v \in V : \text{there exists a path in } G_f \text{ from } s \text{ to } v\}$ , and let T = V - S. Observe that  $s \in S$  and  $t \in T$ , and thus (*S*, *T*) is a cut. Consider any vertices  $u \in S$  and  $v \in T$ .

$$s \xrightarrow{path in G_f} s \xrightarrow{v}_{S} T$$

We must have  $c_f(u, v) = 0$ , since if  $c_f(u, v) > 0$ , then  $v \in S$ , not  $v \in T$  as assumed. Thus, f(u, v) = c(u, v), since  $c_f(u, v)$ = c(u, v) - f(u, v). Summing over all  $u \in S$  and  $v \in T$ yields f(S, T) = c(S, T), and since |f| = f(S, T), the theorem follows.

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### **Algorithm:**

 $f[u, v] \leftarrow 0 \text{ for all } u, v \in V$ while an augmenting path p in G wrt f exists
do augment f by  $c_f(p)$ Can be slow:  $G: \qquad 5 \qquad 10^9 \qquad 10^9 \qquad t$ 

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10

10<sup>9</sup>



### **Algorithm:**

 $f[u, v] \leftarrow 0 \text{ for all } u, v \in V$ while an augmenting path p in G wrt f exists do augment f by  $c_f(p)$ Can be slow:  $0:10^9 \qquad 0:10^9$   $G: \qquad 0:10^9 \qquad 0:10^9$   $0:10^9 \qquad 0:10^9$ 



#### **Algorithm:**

 $f[u, v] \leftarrow 0$  for all  $u, v \in V$ while an augmenting path p in G wrt f exists **do** augment f by  $c_f(p)$ Can be slow: 0:10<sup>9</sup> 0:10<sup>9</sup> *G*: 0:1 0:10<sup>9</sup> 0:10



### **Algorithm:**

 $f[u, v] \leftarrow 0$  for all  $u, v \in V$ while an augmenting path p in G wrt f exists **do** augment f by  $c_f(p)$ Can be slow: 1:10<sup>9</sup> 0:10<sup>9</sup> *G*: 1:1 1:10<sup>9</sup> 0:10



### **Algorithm:**

 $f[u, v] \leftarrow 0 \text{ for all } u, v \in V$ while an augmenting path p in G wrt f exists do augment f by  $c_f(p)$ Can be slow:  $f(p) = \frac{1:10^9}{0:10^9} + \frac{1:10^9}{1:10^9} + \frac{1:10^9}{1:10^9}$ 



### **Algorithm:**

 $f[u, v] \leftarrow 0 \text{ for all } u, v \in V$ while an augmenting path p in G wrt f exists do augment f by  $c_f(p)$ Can be slow:  $f(p) = \frac{1:10^9}{0:1 + 1:10^9}$ 



### **Algorithm:**

 $f[u, v] \leftarrow 0$  for all  $u, v \in V$ while an augmenting path p in G wrt f exists **do** augment f by  $c_f(p)$ Can be slow: 1:10<sup>9</sup> 1:10<sup>9</sup> *G*: 0:1 1:10<sup>9</sup> 1:10



### **Algorithm:**

 $f[u, v] \leftarrow 0$  for all  $u, v \in V$ while an augmenting path p in G wrt f exists **do** augment f by  $c_f(p)$ Can be slow: 2:10<sup>9</sup> 1:10<sup>9</sup> *G*: 1:1 2:10<sup>9</sup> 1:10



### **Algorithm:**

 $f[u, v] \leftarrow 0 \text{ for all } u, v \in V$ while an augmenting path p in G wrt f exists
do augment f by  $c_f(p)$ Can be slow:  $2:10^9$   $1:10^9$  1:1

### 2 billion iterations on a graph with 4 vertices!

1:10

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2:10<sup>9</sup>



### **Edmonds-Karp algorithm**

Edmonds and Karp noticed that many people's implementations of Ford-Fulkerson augment along a *breadth-first augmenting path*: a shortest path in  $G_f$  from *s* to *t* where each edge has weight 1. These implementations would always run relatively fast.

Since a breadth-first augmenting path can be found in O(E) time, their analysis, which provided the first polynomial-time bound on maximum flow, focuses on bounding the number of flow augmentations.

(In independent work, Dinic also gave polynomialtime bounds.)

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### Best to date

- The Edmonds-Karp maximum-flow algorithm runs in  $O(VE^2)$  time.
  - Breadth-first search takes O(E) time
  - O(VE) augmentations in worst case
- The asymptotically fastest algorithm through 2011 for maximum flow, due to King, Rao, and Tarjan, runs in  $O(VE \log_{E/(V \lg V)} V)$  time.
- Recently Orlin came up with an O(VE) time algorithm!
  - One variant uses fast matrix multiplication

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### **Flow Integrality**

• Claim: Suppose the flow network has integer capacities. Then, the maximum flow will be integer-valued.

*Proof*: Start with a flow of 0 on all edges. Use Ford-Fulkerson. Initially, and at each step, Ford-Fulkerson will find an augmenting path with residual capacity that is an integer. Therefore, all flow values on edges always remain integral throughout the algorithm.



- Baseball Elimination
- Bipartite Matching
- Flow integrality important to reducing these problems to max flow!
- See additional notes for L14 for Baseball Elimination

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