## Analysis of Algorithms, I CSOR W4231.002

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#### Outline

- 1 Recap
- 2 Flow networks
  - Applications
- 3 The residual graph and augmenting paths
- 4 The Ford-Fulkerson algorithm for max flow
- 5 Faster algorithms for max flow

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#### Review of the last lecture

A union-find data structure for maintaining disjoint sets.

- ► Implementation: maintain sets as directed rooted trees;
  - ▶ Makeset: worst-case time is O(1)
  - ▶ Find: worst-case time is  $O(\log n)$
  - ▶ Union: worst-case time is  $O(\log n)$
- ▶ Improved implementation by using **path compression**; amortized time for a sequence of 2m Find operations:  $O((m+n)\log^* n)$ 
  - this is not the tightest possible analysis but it is already fairly subtle

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### Modeling transportation networks



Source: Communications of the ACM, Vol. 57, No. 8

Can model a fluid network or a highway system by a graph: edges carry *traffic*, nodes are *switches* where traffic gets diverted.

#### Flow networks

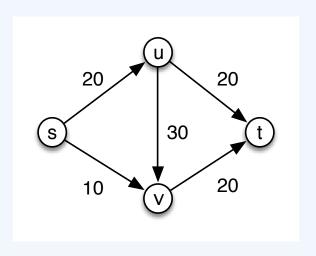
A flow network G = (V, E) is a directed graph such that

- 1. Every edge has a capacity  $c(e) \ge 0$ . A1: integer capacities
- 2. There is a single source  $s \in V$ . A2: no edge enters s
- 3. There is a single sink  $t \in V$ . A3: no edge leaves t

Two more assumptions for the purposes of the analysis

- $A4: if (u, v) \in E then (v, u) \notin E$ .
- ▶ A5: Every  $v \in V \{s,t\}$  is on some s-t path. Hence G has  $m \ge n - 1$  edges.

# An example flow network



#### Flows

Given a flow network G, an s-t flow f in G is a function

$$f: E \to R^+$$

Intuitively, the flow f(e) on edge e is the amount of traffic that edge e carries.

# Two kinds of constraints that every flow must satisfy

- 1. Capacity constraints: for all  $e \in E$ ,  $0 \le f(e) \le c(e)$ .
- 2. Flow conservation: for all  $v \in V \{s, t\}$ ,

$$\sum_{(u,v)\in E} f(u,v) = \sum_{(v,w)\in E} f(v,w)$$
 (1)

In words, the flow into node v equals the flow out of v, or

$$\sum_{e \text{ into } v} f(e) = \sum_{e \text{ out of } v} f(e)$$

### A cleaner equation for flow conservation constraints

Define

1. 
$$f^{\text{out}}(v) = \sum_{e \text{ out of } v} f(e)$$

2. 
$$f^{\text{in}}(v) = \sum_{e \text{ into } v} f(e)$$

So we can rewrite equation (1) as: for all  $v \in V - \{s, t\}$ 

$$f^{\text{in}}(v) = f^{\text{out}}(v) \tag{2}$$

#### The value of a flow

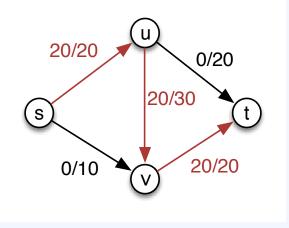
#### Definition 1.

The value of a flow f, denoted by |f|, is

$$|f| = \sum_{e \text{ out of } s} f(e) = f^{\text{out}}(s)$$

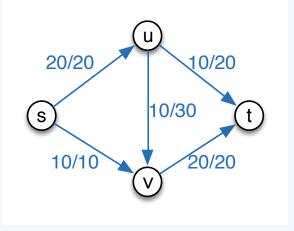
Can show that  $|f| = f^{in}(t)$ . (exercise)

### An example flow of value 20



A flow f of value 20.

#### Another flow, of value 30



A (max) flow of value 30.

### Max flow problem

**Input:** (G, s, t, c) such that

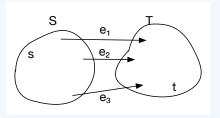
- G = (V, E) is a flow network;
- ▶  $s, t \in V$  are the source and sink respectively;
- ightharpoonup c is the (integer-valued) capacity function.

Output: a flow of maximum possible value

#### s-t cuts in flow networks

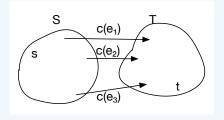
#### Definition 2.

An s-t cut (S,T) in G is a partition of the vertices into two sets S and T, such that  $s \in S$  and  $t \in T$ .



#### A natural upper bound for the max value of a flow

- ▶ Flow f must cross (S,T) to go from source s to sink t.
- ► So it uses some (at most all) of the capacity of the edges crossing this cut.



▶ So, intuitively, the value of the flow cannot exceed

$$\sum_{e \text{ out of } S} c(e)$$

#### Max flow and min cut

#### Definition 3.

The capacity c(S,T) of an s-t cut (S,T) is defined as

$$c(S,T) = \sum_{e \text{ out of } S} c(e).$$

 $\triangle$  Note asymmetry in the definition of c(S,T)!

So, *intuitively*, the value of the max flow is upper bounded by the capacity of *every* cut in the flow network, that is,

$$\max_{f} |f| \le \min_{(S,T) \text{ cut in } G} c(S,T) \tag{3}$$

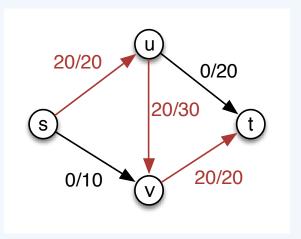
# Applications of max-flow and min-cut

- ► Min-cut
  - find a set edges of smallest capacity whose deletion disconnects the network
- ► Max-flow
  - ▶ Bipartite matching (next lecture)
  - Airline scheduling
  - ▶ Baseball elimination
  - Distribution of goods to cities
  - ► Image segmentation
  - ► Survey design
  - **.**..

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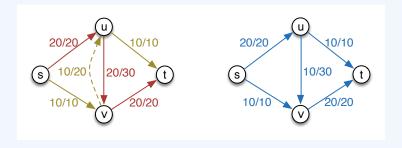
# "Undoing" flow



A flow f of value 20.

Would like to undo 10 units of flow along (u, v) and divert it along (u, t).

### Pushing flow back



- ▶ Push back 10 units of flow along (v, u).
- Send 10 more units from s to t along edges (s, v), (v, u), (u, t).
- ▶ New flow f' (on the right) with value 30.

## Pushing flow forward and backward

By pushing flow back on (v, u) we created an s-t path on which we are pushing flow

- **Forward**, on edges with leftover capacity (e.g., on (s, v))
- ▶ Backward, on edges that are already carrying flow so as to divert it to a different direction (e.g., on (u, v)).

# The residual graph $G_f$

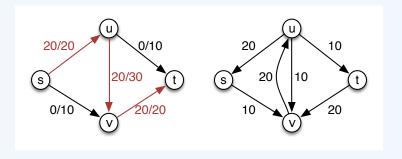
#### Definition 4.

Given flow network G and flow f, the residual graph  $G_f$  has

- $\blacktriangleright$  the same vertices as G;
- ▶ for every edge  $e = (u, v) \in E$  such that f(e) < c(e), an edge e = (u, v) with capacity  $c_f(e) = c(e) f(e)$  (forward edge);
- ▶ for every edge  $e = (u, v) \in E$  such that f(e) > 0, an edge  $e^r = (v, u)$  with capacity  $c_f(e^r) = f(e)$  (backward edge).

So  $G_f$  has  $\leq 2m$  edges.

# Example residual graph



Left: a flow f of value 20.

Right: the residual graph  $G_f$  for this flow.

### Augmenting paths

The residual graph  $G_f$  provides a roadmap for augmenting f.

- 1. Let P be a simple s-t path in  $G_f$ .
- 2. Augment f by pushing extra flow on P.

How much extra flow can we push on P without violating capacity constraints in  $G_f$ ?

Let c(P) be the capacity of path P defined as the minimum residual capacity of **any** edge of P.

$$c(P) = \min_{e \in P} c_f(e)$$

▶ The maximum amount of flow we can safely push on **every** edge of P is c(P).

# The augmented flow f'

Let P be an augmenting path in the residual graph  $G_f$ . Augmented flow f' is as follows:

1. For a **forward** edge  $e \in P$ 

$$f'(e) = f(e) + c(P)$$

2. For a **backward** edge  $e^r = (u, v) \in P$ , let  $e = (v, u) \in G$ 

$$f'(e) = f(e) - c(P)$$

3. For  $e \in E$  but not in P, f'(e) = f(e).

#### Fact 5 (1).

f' is a flow.

#### Pseudocode

```
\begin{aligned} \operatorname{Augment}(f,P) & \quad \text{for each edge } (u,v) \in P \text{ do} \\ & \quad \text{if } e = (u,v) \text{ is a forward edge then} \\ & \quad f'(e) = f(e) + c(P) \\ & \quad \text{else} \\ & \quad f'(v,u) = f(v,u) - c(P) \\ & \quad \text{end if} \\ & \quad \text{end for} \\ & \quad \operatorname{Return} f' \end{aligned}
```

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# The Ford-Fulkerson algorithm

```
Ford-Fulkerson( G = (V, E, c), s, t)
for all e \in E do f(e) = 0
end for
while there is an s-t path in G_f do
   Let P be a simple s-t path in G_f
   f' = Augment(f, P)
   Update f = f'
   Update G_f = G_{f'}
end while
Return f'
```

### Running time analysis

The algorithm terminates if the following claims are both true

- 1. **Claim 1:** every iteration of the while loop returns a flow increased by an integer amount; and
- 2. Claim 2: there is a finite upper bound to the flow.

#### Proof of Claim 2.

Let U be the largest edge capacity. Then

$$|f| \le \sum_{e \text{ out of } s} c(e) \le nU$$

### f increases by an integer amount after Augment(f, P)

#### Proof of Claim 1.

It follows from the following facts.

#### Fact 6 (2).

During execution of the Ford-Fulkerson algorithm, the flow values  $\{f(e)\}$  and the residual capacities in  $G_f$  are all integers.

#### Fact 7 (3).

Let f be a flow in G and P a simple s-t path in  $G_f$  with residual capacity c(P) > 0. Then after Augment(f, P)

$$|f'| = |f| + c(P) \ge |f| + 1.$$

# f increases by an integer amount after Augment(f, P)

#### Proof of Fact 3.

Recall that  $|f| = f^{\text{out}}(s)$ .

- 1. Since P is an s-t path, it contains an edge out of s, say (s, u).
- 2. Since P is simple, it does not contain any edge entering s (P is in  $G_f$ , where there are edges entering s!): otherwise, s would be visited again.
- 3. Since no edge enters s in G, (s, u) is a forward edge in  $G_f$ , thus the flow on this edge is updated to f(s, u) + c(P) > f(s, u) + 1.
- 4. Since no other edge going out of s is updated, it follows that the value of f' is  $|f'| = |f| + c(P) \ge |f| + 1$ .

#### Running time of Ford-Fulkerson

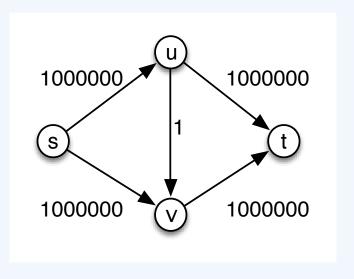
- 1. Fact 3 guarantees at most nU iterations.
- 2. The running time of each iteration is bounded as follows:
  - ightharpoonup O(m+n) to create  $G_f$  using adjacency list representation
  - ightharpoonup O(m+n) to run BFS or DFS to find the augmenting path
  - ightharpoonup O(n) for Augment(f, P) since P has at most n-1 edges
  - $\Rightarrow$  Hence one iteration requires O(m) time

The running time of Ford-Fulkerson is O(mnU).

#### Remark 1.

This is a pseudo-polynomial time algorithm: it is not polynomial in the description of the input U

### Problems with pseudo-polynomial running times



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### Algorithms

- ▶ Can be made polynomial: use BFS instead of DFS
  - ▶ Edmonds-Karp:  $O(nm^2)$
- Unit capacities:  $O(\min(\sqrt{m}, n^{2/3})m)$
- ▶ Integral capacities:  $O(\min(\sqrt{m}, n^{2/3}) m \log(n^2/m) \log U)$  [GoldbergRao1998]
- ▶ Real capacities:  $O(nm \log (n^2/m))$ 
  - ▶ Improved: O(nm) [Orlin2013]