## CS38 Introduction to Algorithms

Lecture 12 May 8, 2014

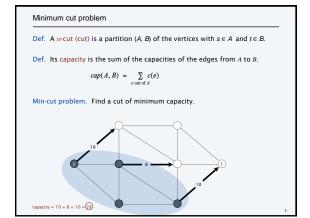
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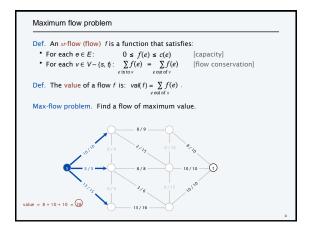
## Outline

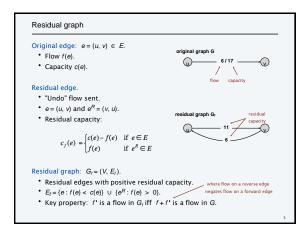
- · Network flow
  - finishing capacity-scaling analysis
  - Edmonds-Karp, blocking-flow implementation
  - unit-capacity simple graphs
  - bipartite matching
  - edge-disjoint paths
  - assignment problem

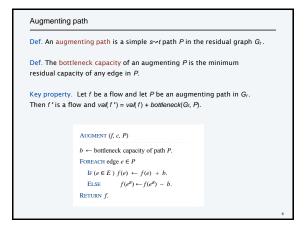
\* slides from Kevin Wayne

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## Ford-Fulkerson algorithm Ford-Fulkerson augmenting path algorithm. • Start with f(e) = 0 for all edge $e \in E$ . • Find an augmenting path P in the residual graph $G_f$ . • Augment flow along path P. • Repeat until you get stuck. FORD-FULKERSON (G, s, t, e)FOREACH edge $e \in E : f(e) \leftarrow 0$ . $G_f \leftarrow \text{residual graph}$ . WHILE (there exists an augmenting path $P \text{ in } G_f$ ) $f \leftarrow \text{AUGMENT } (f, e, P)$ . Update $G_f$ . RETURN f. }

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Capacity-scaling algorithm

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Capacity-scaling algorithm

Intuition. Choose augmenting path with highest bottleneck capacity: it increases flow by max possible amount in given iteration.

• Don't worry about finding exact highest bottleneck path.

• Maintain scaling parameter Δ.

• Let G<sub>f</sub>(Δ) be the subgraph of the residual graph consisting only of arcs with capacity ≥ Δ.

Capacity-scaling algorithm: proof of correctness 
Assumption. All edge capacities are integers between 1 and C. 
Integrality invariant. All flow and residual capacity values are integral. 
Theorem. If capacity-scaling algorithm terminates, then f is a max-flow. 
Pf. 
• By integrality invariant, when  $\Delta = 1 \Rightarrow G_r(\Delta) = G_r$ . 
• Upon termination of  $\Delta = 1$  phase, there are no augmenting paths. •

Capacity-scaling algorithm: analysis of running time

Lemma 1. The outer while loop repeats  $1+\lceil \log_2 C \rceil$  times.

Pf. Initially  $C/2 < \Delta \le C$ ;  $\Delta$  decreases by a factor of 2 in each iteration. •

Lemma 2. Let f be the flow at the end of a  $\Delta$ -scaling phase. Then, the value of the max-flow  $\le val(f) + m\Delta$ . — proof on next slide

Lemma 3. There are at most 2m augmentations per scaling phase.

Pf.

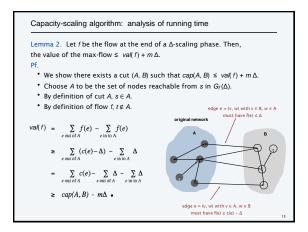
• Let f be the flow at the end of the previous scaling phase.

• LEMMA  $2 \Rightarrow val(f^*) \le val(f) + 2m\Delta$ .

• Each augmentation in a  $\Delta$ -phase increases val(f) by at least  $\Delta$ . •

Theorem. The scaling max-flow algorithm finds a max flow in  $O(m \log C)$  augmentations. It can be implemented to run in  $O(m^2 \log C)$  time.

Pf. Follows from LEMMA 1 and LEMMA 3. •





Shortest augmenting path?

Q. Which augmenting path?

A. The one with the fewest number of edges.

Can find via BFS

SHORTEST-AUGMENTING-PATH(G, s, t, c)

FOREACH  $e \in E: f(e) \leftarrow 0$ .  $G_f \leftarrow \text{residual graph}$ .

WHILE (there exists an augmenting path in  $G_f$ )  $P \leftarrow \text{BREADTH-FIRST-SEARCH}(G_f, s, t)$ .  $f \leftarrow \text{AUGMENT}(f, c, P)$ .

Update  $G_f$ .

RETURN f.

Shortest augmenting path: overview of analysis

L1. Throughout the algorithm, length of the shortest path never decreases.

L2. After at most *m* shortest path augmentations, the length of the shortest augmenting path strictly increases.

Theorem. The shortest augmenting path algorithm runs in *O*(*m*<sup>2</sup> *n*) time.

Pf.

• *O*(*m* + *n*) time to find shortest augmenting path via BFS.

• *O*(*m*) augmentations for paths of length *k*.

• If there is an augmenting path, there is a simple one.

⇒ 1 ≤ *k* < *n*⇒ *O*(*m n*) augmentations. •

Shortest augmenting path: analysis

Def. Given a digraph G = (V, E) with source s, its level graph is defined by:

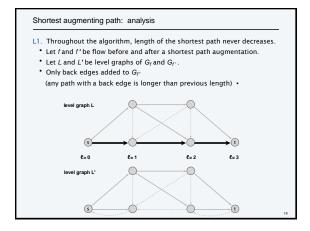
•  $\ell(v) =$  number of edges in shortest path from s to v.

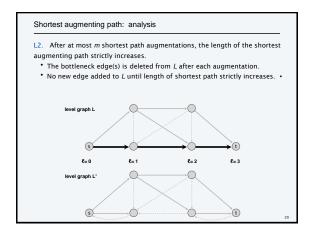
•  $L_G = (V, E_G)$  is the subgraph of G that contains only those edges  $(v, w) \in E$  with  $\ell(w) = \ell(v) + 1$ .

Property. Can compute level graph in O(m + n) time.

Pf. Run BFS; delete back and side edges.

Key property. P is a shortest  $s \sim v$  path in G iff P is an  $s \sim v$  path  $L_G$ .





Shortest augmenting path: review of analysis

L1. Throughout the algorithm, length of the shortest path never decreases.

L2. After at most *m* shortest path augmentations, the length of the shortest augmenting path strictly increases.

Theorem. The shortest augmenting path algorithm runs in  $O(m^2 n)$  time.

Pf.

O(m+n) time to find shortest augmenting path via BFS.
O(m) augmentations for paths of exactly *k* edges.
O(m n) augmentations.

Shortest augmenting path: improving the running time

Note. O(m n) augmentations necessary on some networks.

• Try to decrease time per augmentation instead.

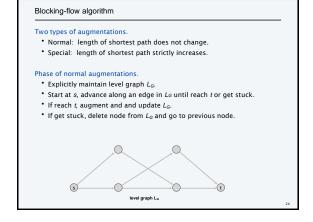
• Simple idea ⇒ O(m n²) [Dinic 1970]

• Dynamic trees ⇒ O(m n log n) [Sleator-Tarjan 1983]

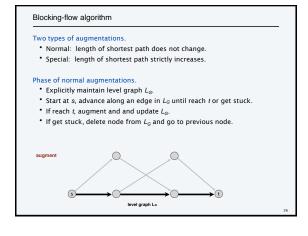
A Data Structure for Dynamic Trees

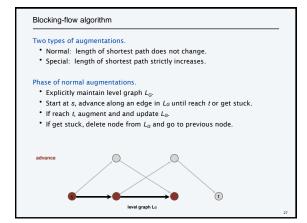
Date: D. Receive for Dynamic

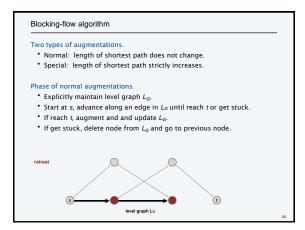
# Blocking flow May 8, 2014 CS38 Lecture 12 23

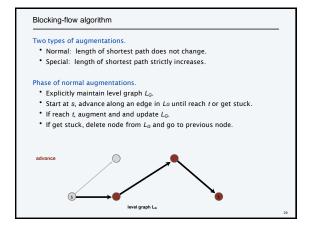


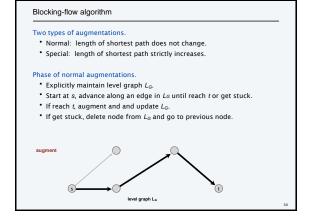
## Two types of augmentations. Normal: length of shortest path does not change. Special: length of shortest path strictly increases. Phase of normal augmentations. Explicitly maintain level graph L<sub>G</sub>. Start at s, advance along an edge in L<sub>G</sub> until reach t or get stuck. If reach t, augment and and update L<sub>G</sub>. If get stuck, delete node from L<sub>G</sub> and go to previous node.

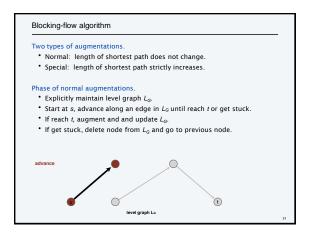


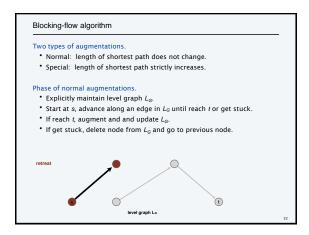












Blocking-flow algorithm

Two types of augmentations.

Normal: length of shortest path does not change.
Special: length of shortest path strictly increases.

Phase of normal augmentations.
Explicitly maintain level graph  $L_G$ .
Start at s, advance along an edge in  $L_G$  until reach t or get stuck.
If reach t, augment and and update  $L_G$ .
If get stuck, delete node from  $L_G$  and go to previous node.

Blocking-flow algorithm

Two types of augmentations.

Normal: length of shortest path does not change.
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Start at s, advance along an edge in L<sub>G</sub> until reach t or get stuck.
If reach t, augment and and update L<sub>G</sub>.
If get stuck, delete node from L<sub>G</sub> and go to previous node.

Blocking-flow algorithm INITIALIZE(G, s, t, f, c)ADVANCE(v)  $L_G \leftarrow level-graph of G_f$ . IF (v = t) $P \leftarrow \emptyset$ . GOTO ADVANCE(s). Remove saturated edges from  $L_G$ .  $P \leftarrow \emptyset$ . GOTO ADVANCE(s). RETREAT(v) IF (there exists edge  $(v, w) \in L_G$ ) IF (v = s) STOP. Add edge (v, w) to P. GOTO ADVANCE(w). Delete  $\nu$  (and all incident edges) from  $L_G$ . Remove last edge (u, v) from P. ELSE GOTO RETREAT(v). GOTO ADVANCE(u).

Blocking-flow algorithm: analysis

Lemma. A phase can be implemented in O(mn) time.

Pf.

Initialization happens once per phase. — O(m) using BFS

At most m augmentations per phase. — O(m) per phase (because an augmentation deletes at least one edge from  $L_0$ )

At most n retreats per phase. — O(m+n) per phase (because a retreat deletes one node from  $L_0$ )

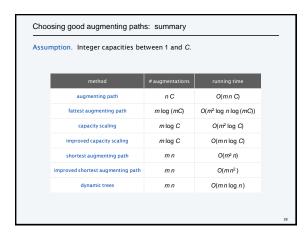
At most m advances per phase. — O(mn) per phase (because at most n advances before retreat or augmentation)

Theorem. [Dinic 1970] The blocking-flow algorithm runs in  $O(mn^2)$  time.

Pf.

By lemma, O(mn) time per phase.

At most n phases (as in shortest augment path analysis).



year	method	worst case	discovered by
1951	simplex	O(m³ C)	Dantzig
1955	augmenting path	O(m² C)	Ford-Fulkerson
1970	shortest augmenting path	O(m³)	Dinic, Edmonds-Karp
1970	fattest augmenting path	O(m² log m log( m C))	Dinic, Edmonds-Karp
1977	blocking flow	O(m <sup>5/2</sup> )	Cherkasky
1978	blocking flow	O(m <sup>7/3</sup> )	Galil
1983	dynamic trees	O(m² log m)	Sleator-Tarjan
1985	capacity scaling	O(m² log C)	Gabow
1997	length function	$O(m^{3/2} \log m \log C)$	Goldberg-Rao
2012	compact network	O(m² / log m)	Orlin
?	?	O(m)	?

## Unit capacity simple graphs

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### Bipartite matching

- Q. Which max-flow algorithm to use for bipartite matching?
- Generic augmenting path:  $O(m | f^*|) = O(mn)$ .
- Capacity scaling:  $O(m^2 \log U) = O(m^2)$ .
- Shortest augmenting path: O(mn2).
- Q. Suggests "more clever" algorithms are not as good as we first thought?
- A. No, just need more clever analysis!

Next.

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NETWORK FLOW AND TESTING GRAPH CONNECTIVITY\*

Unit-capacity simple networks

Def. A network is a unit-capacity simple network if:

• Every edge capacity is 1.

of O(mn1/2)

\* Every node (other than s or t) has either (i) at most one entering edge or (ii) at most one leaving edge.

Property. Let G be a simple unit-capacity network and let f be a 0-1 flow, then  $G_f$  is a unit-capacity simple network.



Unit-capacity simple networks

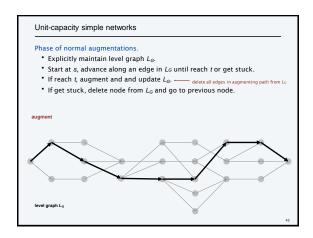
Shortest augmenting path algorithm.

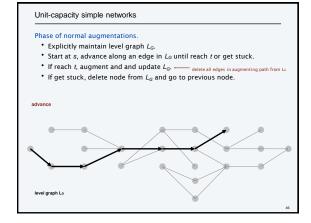
- \* Normal augmentation: length of shortest path does not change.
- Special augmentation: length of shortest path strictly increases.

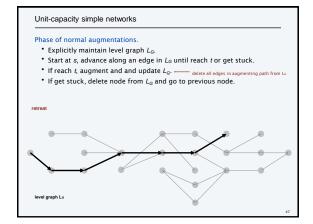
Theorem. [Even-Tarjan 1975] In unit-capacity simple networks, the shortest augmenting path algorithm computes a maximum flow in  $O(m \, n^{1/2})$  time.

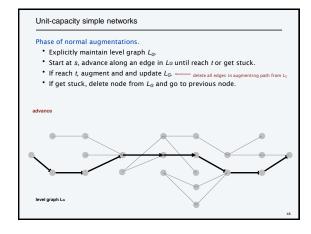
- $\bullet$  L1. Each phase of normal augmentations takes O(m) time.
- L2. After at most  $n^{1/2}$  phases,  $|f| \ge |f^*| n^{1/2}$ .
- $^{ullet}$  L3. After at most  $n^{1/2}$  additional augmentations, flow is optimal.  $^{ullet}$

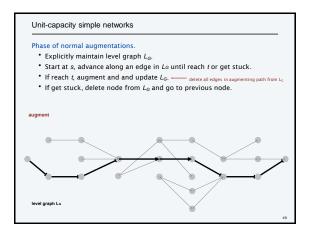
# Unit-capacity simple networks Phase of normal augmentations. Explicitly maintain level graph $L_G$ . Start at s, advance along an edge in $L_G$ until reach t or get stuck. If reach t, augment and and update $L_G$ . If get stuck, delete node from $L_G$ and go to previous node. advance

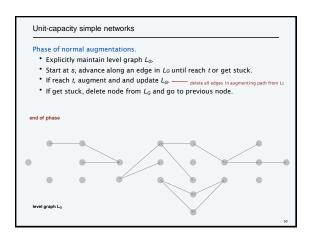












Unit-capacity simple networks: analysis

Phase of normal augmentations.

Explicitly maintain level graph L<sub>G</sub>.

Start at s, advance along an edge in L<sub>G</sub> until reach t or get stuck.

If reach t, augment and and update L<sub>G</sub>.

If get stuck, delete node from L<sub>G</sub> and go to previous node.

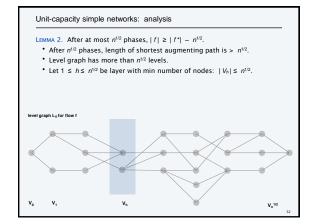
LEMMA 1. A phase of normal augmentations takes O(m) time.

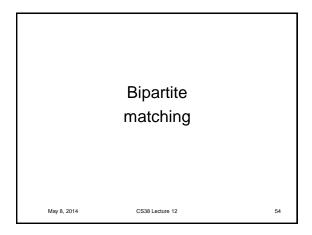
Pf.

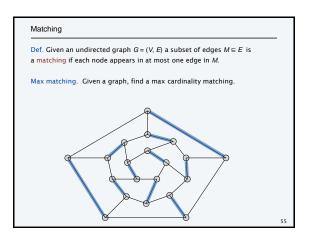
O(m) to create level graph L<sub>G</sub>.

O(1) per edge since each edge traversed and deleted at most once.

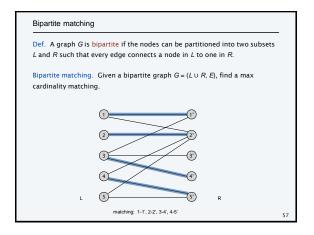
O(1) per node since each node deleted at most once.

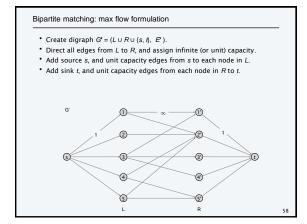


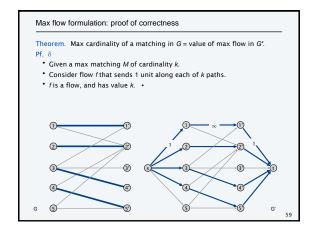


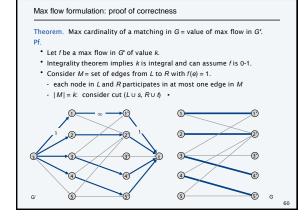


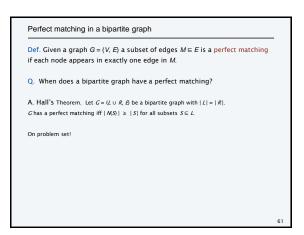
# Bipartite matching Def. A graph G is bipartite if the nodes can be partitioned into two subsets L and R such that every edge connects a node in L to one in R. Bipartite matching. Given a bipartite graph $G = (L \cup R, E)$ , find a max cardinality matching.











Bipartite matching running time

Theorem. The Ford-Fulkerson algorithm solves the bipartite matching problem in  $O(m \, n)$  time.

Theorem. [Hopcroft-Karp 1973] The bipartite matching problem can be solved in  $O(m \, n^{1/2})$  time.

BIAN J. Course.

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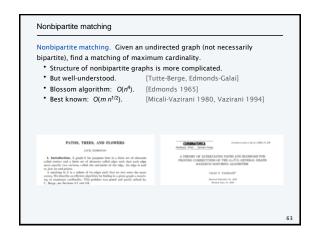
An  $\sigma^{4/2}$  ALGORITHM FOR MAXIMUM MATCHINGS

IN BIPARTITE GRAPHS\*

JOIN E. BIOROFT AND RICKARD M. KARP!

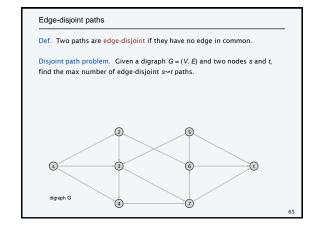
Abbret. The prosen paper show how to course a maximum matching in a biggirt graph with a vortice and or digin is a knowled or compensation step proceedited for  $\sigma = A_0/C$ ,

Ry week algorithm. algorithmic analysis, bipartite graphs, computational complexity, graphs, matching



Edge-disjoint paths

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Edge-disjoint paths

Def. Two paths are edge-disjoint if they have no edge in common.

Disjoint path problem. Given a digraph G = (V, E) and two nodes s and t, find the max number of edge-disjoint s → t paths.

Ex. Communication networks.

Edge-disjoint paths

Max flow formulation. Assign unit capacity to every edge.

Theorem. Max number edge-disjoint s v t paths equals value of max flow. Pf. ≤

• Suppose there are k edge-disjoint s v t paths P₁, ..., Pk.

• Set f(e) = 1 if e participates in some path Pj; else set f(e) = 0.

• Since paths are edge-disjoint, f is a flow of value k. •

### Edge-disjoint paths

Max flow formulation. Assign unit capacity to every edge.

Theorem. Max number edge-disjoint  $s\sim t$  paths equals value of max flow. Pf. ε
• Suppose max flow value is k.
• Suppose max flow value is the state of the

- Integrality theorem implies there exists 0-1 flow f of value k.
- Consider edge (s, u) with f(s, u) = 1.
- by conservation, there exists an edge (u, v) with f(u, v) = 1
- continue until reach t, always choosing a new edge
- ullet Produces k (not necessarily simple) edge-disjoint paths. ullet

