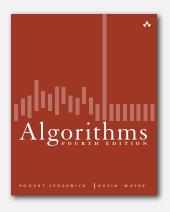
6.4 MAXIMUM FLOW



- ▶ overview
- ▶ Ford-Fulkerson algorithm
- ▶ analysis
- **▶** Java implementation
- **▶** applications

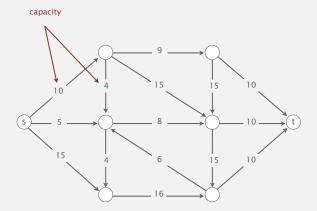
Algorithms, 4th Edition · Robert Sedgewick and Kevin Wayne · Copyright © 2002–2011 · November 17, 2011 9:54:22 AM

▶ overview

Mincut problem

Input. A weighted digraph, source vertex s, and target vertex t.

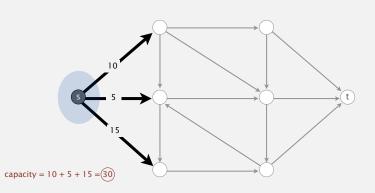
each edge has a positive capacity



Mincut problem

Def. A st-cut (cut) is a partition of the vertices into two disjoint sets, with s in one set A and t in the other set B.

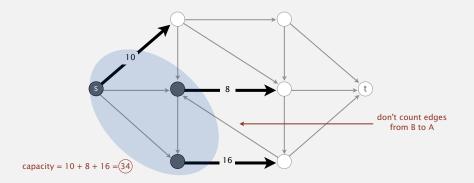
Def. Its capacity is the sum of the capacities of the edges from A to B.



Mincut problem

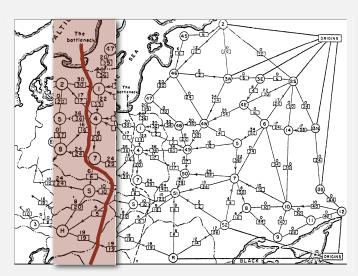
Def. A st-cut (cut) is a partition of the vertices into two disjoint sets, with s in one set A and t in the other set B.

Def. Its capacity is the sum of the capacities of the edges from A to B.



Mincut application (1950s)

"Free world" goal. Cut supplies (if cold war turns into real war).



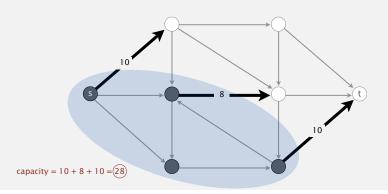
rail network connecting Soviet Union with Eastern European countries (map declassified by Pentagon in 1999)

Mincut problem

Def. A st-cut (cut) is a partition of the vertices into two disjoint sets, with s in one set A and t in the other set B.

Def. Its capacity is the sum of the capacities of the edges from A to B.

Minimum st-cut (mincut) problem. Find a cut of minimum capacity.



Potential mincut application (2010s)

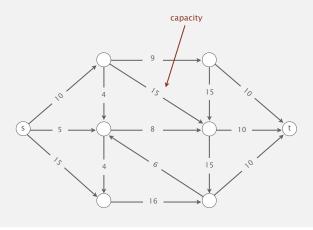
Government-in-power's goal. Cut off communication to specified set of people.



Maxflow problem

Input. A weighted digraph, source vertex s, and target vertex t.

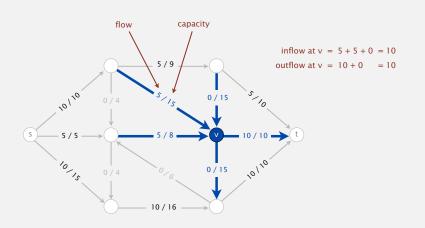
each edge has a positive capacity



Maxflow problem

Def. An st-flow (flow) is an assignment of values to the edges such that:

- Capacity constraint: $0 \le \text{edge's flow} \le \text{edge's capacity}$.
- Local equilibrium: inflow = outflow at every vertex (except s and t).



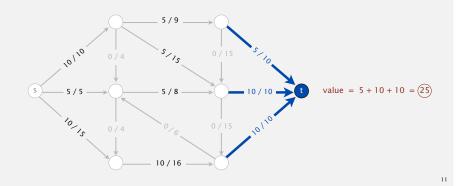
Maxflow problem

Def. An st-flow (flow) is an assignment of values to the edges such that:

- Capacity constraint: $0 \le \text{edge's flow} \le \text{edge's capacity}$.
- Local equilibrium: inflow = outflow at every vertex (except s and t).

Def. The value of a flow is the inflow at t.

we assume no edges pointing from s or to t



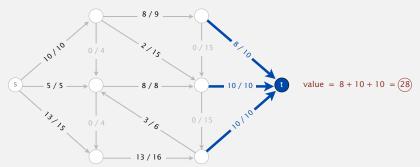
Maxflow problem

Def. An st-flow (flow) is an assignment of values to the edges such that:

- Capacity constraint: $0 \le \text{edge's flow} \le \text{edge's capacity}$.
- Local equilibrium: inflow = outflow at every vertex (except s and t).

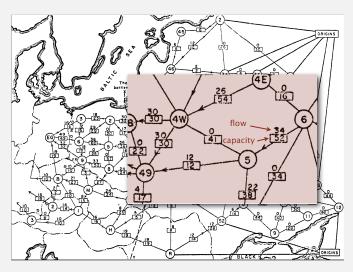
Def. The value of a flow is the inflow at t.

Maximum st-flow (maxflow) problem. Find a flow of maximum value.



Maxflow application (1950s)

Soviet Union goal. Maximize flow of supplies to Eastern Europe.



rail network connecting Soviet Union with Eastern European countries (map declassified by Pentagon in 1999)

Potential maxflow application (2010s)

"Free world" goal. Maximize flow of information to specified set of people.



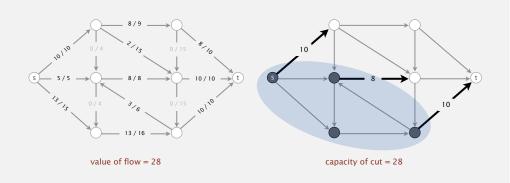
facebook graph

Summary

Input. A weighted digraph, source vertex s, and target vertex t.

Mincut problem. Find a cut of minimum capacity.

Maxflow problem. Find a flow of maximum value.



Remarkable fact. These two problems are dual!

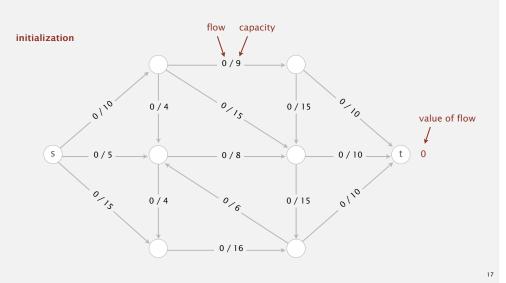
overview

→ Ford-Fulkerson algorithm

- analysis
- ▶ Java implementation
- applications

Ford-Fulkerson algorithm

Initialization. Start with 0 flow.

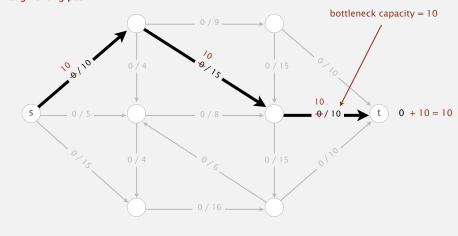


Idea: increase flow along augmenting paths

Augmenting path. Find an undirected path from s to t such that:

- Can increase flow on forward edges (not full).
- Can decrease flow on backward edge (not empty).

1st augmenting path

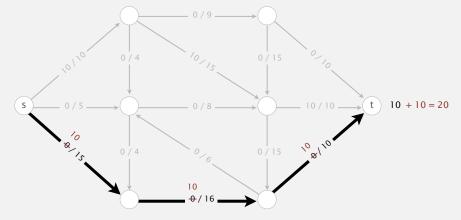


Idea: increase flow along augmenting paths

Augmenting path. Find an undirected path from s to t such that:

- Can increase flow on forward edges (not full).
- Can decrease flow on backward edge (not empty).

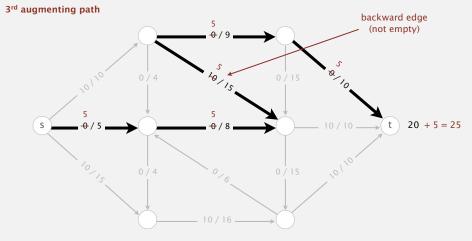
2nd augmenting path



Idea: increase flow along augmenting paths

Augmenting path. Find an undirected path from s to t such that:

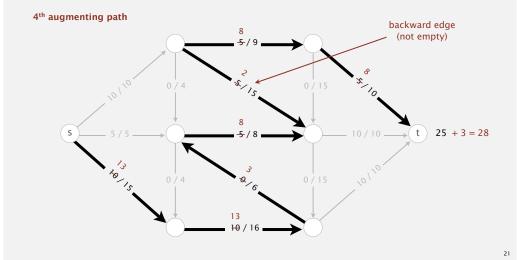
- Can increase flow on forward edges (not full).
- Can decrease flow on backward edge (not empty).



Idea: increase flow along augmenting paths

Augmenting path. Find an undirected path from s to t such that:

- Can increase flow on forward edges (not full).
- Can decrease flow on backward edge (not empty).

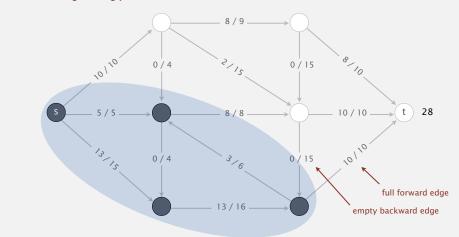


Idea: increase flow along augmenting paths

Termination. All paths from s to t are blocked by either a

- Full forward edge.
- Empty backward edge.

no more augmenting paths



Ford-Fulkerson algorithm

Ford-Fulkerson algorithm

Start with 0 flow.

While there exists an augmenting path:

- find an augmenting path
- compute bottleneck capacity
- increase flow on that path by bottleneck capacity

Questions.

- How to compute a mincut?
- How to find an augmenting path?
- If FF terminates, does it always compute a maxflow?
- Does FF always terminate? If so, after how many augmentations?

overview

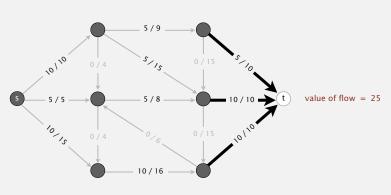
- ▶ Ford-Fulkerson algorithm
- ▶ analysis
- lava implementation
- ▶ applications

Relationship between flows and cuts

Def. The net flow across a cut (A, B) is the sum of the flows on its edges from A to B minus the sum of the flows on its edges from from B to A.

Proposition. Let f be any flow and let (A, B) be any cut. Then, the net flow across (A, B) equals the value of f.



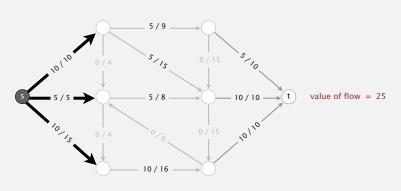


Relationship between flows and cuts

Def. The net flow across a cut (A, B) is the sum of the flows on its edges from A to B minus the sum of the flows on its edges from from B to A.

Proposition. Let f be any flow and let (A, B) be any cut. Then, the net flow across (A, B) equals the value of f.

net flow across cut = 10 + 5 + 10 = 25

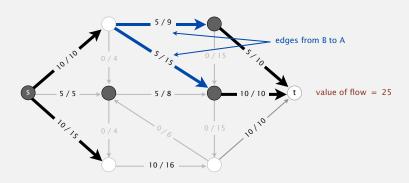


Relationship between flows and cuts

Def. The net flow across a cut (A, B) is the sum of the flows on its edges from A to B minus the sum of the flows on its edges from from B to A.

Proposition. Let f be any flow and let (A, B) be any cut. Then, the net flow across (A, B) equals the value of f.

net flow across cut = (10 + 10 + 5 + 10) - (5 + 5) = 25



Relationship between flows and cuts

Def. The net flow across a cut (A, B) is the sum of the flows on its edges from A to B minus the sum of the flows on its edges from from B to A.

Proposition [flow-value lemma]. Let f be any flow and let (A, B) be any cut. Then, the net flow across (A, B) equals the value of f.

Pf. By induction on the size of B.

- Base case: $B = \{t\}$.
- Induction step: remains true by local equilibrium when moving any vertex from A to B.

Corollary. Outflow from s = inflow to t = value of flow.

Relationship between flows and cuts

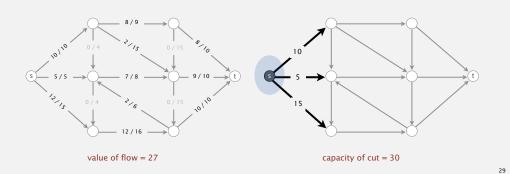
Weak duality. Let f be any flow and let (A, B) be any cut.

Then, the value of the flow \leq the capacity of the cut.

Pf. Value of flow $f = \text{net flow across cut } (A, B) \leq \text{capacity of cut } (A, B)$.







Maxflow-mincut theorem

Augmenting path theorem. A flow f is a maxflow iff no augmenting paths. Maxflow-mincut theorem. Value of the maxflow = capacity of mincut.

Pf. The following three conditions are equivalent for any flow f:

i. There exists a cut whose capacity equals the value of the flow f.

ii. f is a maxflow.

iii. There is no augmenting path with respect to f.

 $[i \Rightarrow ii]$

- Suppose that (A, B) is a cut with capacity equal to the value of f.
- Then, the value of any flow $f' \leq \text{capacity of } (A, B) = \text{value of } f$.
- Thus, f is a maxflow.

Maxflow-mincut theorem

Augmenting path theorem. A flow f is a maxflow iff no augmenting paths. Maxflow-mincut theorem. Value of the maxflow = capacity of mincut.

Pf. The following three conditions are equivalent for any flow f:

i. There exists a cut whose capacity equals the value of the flow f.

ii. f is a maxflow.

iii. There is no augmenting path with respect to f.

[$ii \Rightarrow iii$] We prove contrapositive: $\sim iii \Rightarrow \sim ii$.

- Suppose that there is an augmenting path with respect to f.
- \bullet Can improve flow f by sending flow along this path.
- ullet Thus, f is not a maxflow.

Maxflow-mincut theorem

Augmenting path theorem. A flow f is a maxflow iff no augmenting paths. Maxflow-mincut theorem. Value of the maxflow = capacity of mincut.

Pf. The following three conditions are equivalent for any flow f:

i. There exists a cut whose capacity equals the value of the flow f.

ii. f is a maxflow.

iii. There is no augmenting path with respect to f.

[$iii \Rightarrow i$]

Suppose that there is no augmenting path with respect to f.

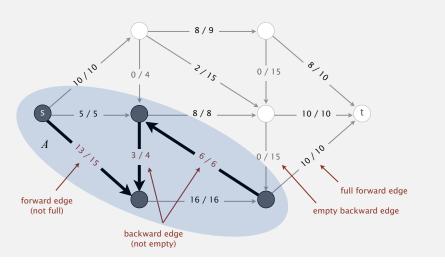
- Let (A,B) be a cut where A is the set of vertices connected to s by an undirected path with no full forward or empty backward edges.
- By definition, s is in A; since no augmenting path, t is in B.
- Capacity of cut = net flow across cut forward edges full; backward edges empty = value of flow f.

,

Computing a mincut from a maxflow

To compute mincut (A, B) from maxflow f:

- By augmenting path theorem, no augmenting paths with respect to f.
- Compute A =set of vertices connected to s by an undirected path with no full forward or empty backward edges.



Ford-Fulkerson algorithm

Ford-Fulkerson algorithm

Start with 0 flow.

While there exists an augmenting path:

- find an augmenting path
- compute bottleneck capacity
- increase flow on that path by bottleneck capacity

Questions.

33

- How to compute a mincut? Easy. ✓
- How to find an augmenting path? BFS works well.
- If FF terminates, does it always compute a maxflow? Yes. 🗸
- Does FF always terminate? If so, after how many augmentations?

yes, provided edge capacities are integers (or augmenting paths are chosen carefully) requires clever analysis (see COS 423)

Ford-Fulkerson algorithm with integer capacities

Important special case. Edge capacities are integers between 1 and U.

flow on each edge is an integer

Invariant. The flow is integer-valued throughout FF.

Pf. [by induction]

- Bottleneck capacity is an integer.
- Flow on an edge increases/decreases by bottleneck capacity.

Proposition. Number of augmentations \leq the value of the maxflow.

Pf. Each augmentation increases the value by at least 1.

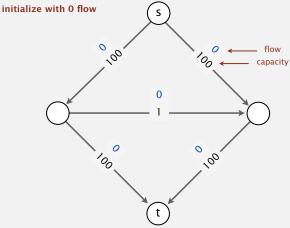
important for some applications (stay tuned) and FF finds one!

Integrality theorem. There exists an integer-valued maxflow.

Pf. FF terminates; invariant ensures maxflow that FF finds is integer-valued.

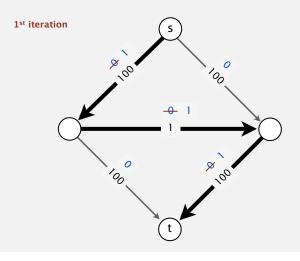
Bad case for Ford-Fulkerson

Bad news. Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.



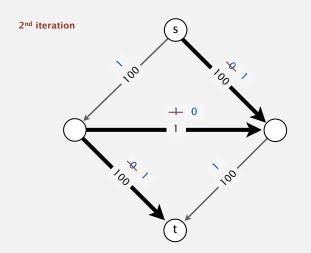
Bad case for Ford-Fulkerson

Bad news. Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.



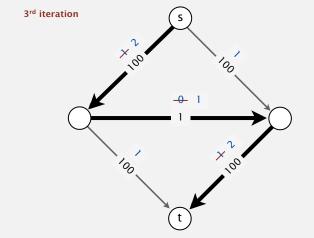
Bad case for Ford-Fulkerson

Bad news. Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.



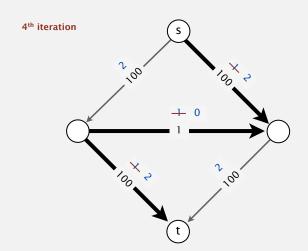
Bad case for Ford-Fulkerson

Bad news. Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.



Bad case for Ford-Fulkerson

Bad news. Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.

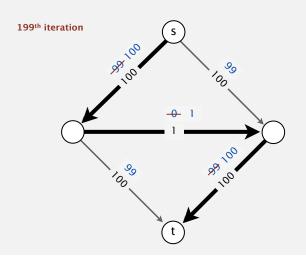


Bad case for Ford-Fulkerson

Bad news. Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.

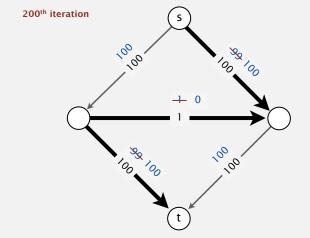
Bad case for Ford-Fulkerson

Bad news. Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.



Bad case for Ford-Fulkerson

Bad news. Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.

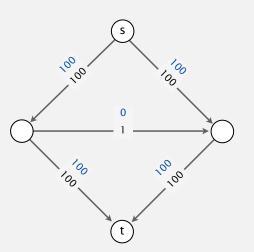


Bad case for Ford-Fulkerson

Bad news. Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.

can be exponential in input size

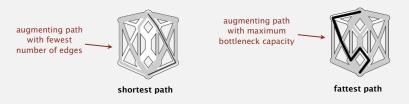
Good news. This case is easily avoided. [use shortest/fattest augmenting path]



How to choose augmenting paths?

FF performance depends on choice of augmenting paths.

augmenting path	number of paths	implementation
shortest path	≤ ½ E V	queue (BFS)
fattest path	≤ E In(E U)	priority queue
random path	≤ E U	randomized queue
DFS path	≤ E U	stack (DFS)



overview

→ Ford-Fulkerson algorithm

analysis

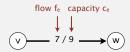
▶ Java implementation

applications

46

Flow network representation

Flow edge data type. Associate flow f_e and capacity c_e with edge $e = v \rightarrow w$.



Flow network data type. Need to process edge $e = v \rightarrow w$ in either direction: Include e in both v and w's adjacency lists.

Residual capacity.

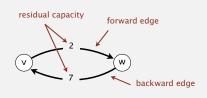
• Forward edge: residual capacity = $c_e - f_e$.

• Backward edge: residual capacity $= f_e$.

Augment flow.

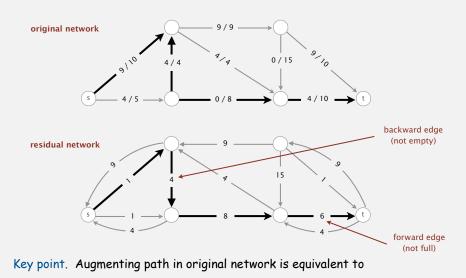
• Forward edge: add Δ .

• Backward edge: subtract Δ .



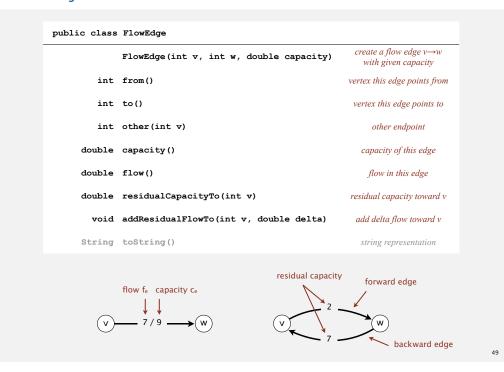
Flow network representation

Residual network. A useful view of a flow network.

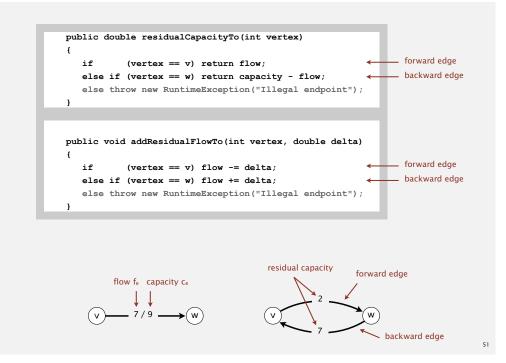


directed path in residual network.

Flow edge API



Flow edge: Java implementation



Flow edge: Java implementation

```
public class FlowEdge
    private final int v, w;
                                     // from and to
    private final double capacity;
                                     // capacity
                                                                           flow variable
   private double flow;
    public FlowEdge(int v, int w, double capacity)
       this v
       this.w
                     = w:
       this.capacity = capacity;
    public int from()
                               { return v;
    public int to()
                               { return w;
    public double capacity() { return capacity; }
    public double flow()
                              { return flow;
    public int other(int vertex)
               (vertex == v) return w;
       else if (vertex == w) return v;
       else throw new RuntimeException("Illegal endpoint");
    public double residualCapacityTo(int vertex)
                                                              {...}
    public void addResidualFlowTo(int vertex, double delta) {...}
```

Flow network API

```
public class FlowNetwork
                        FlowNetwork(int V)
                                                             create an empty flow network with V vertices
                        FlowNetwork (In in)
                                                                 construct flow network input stream
                 void addEdge(FlowEdge e)
                                                                 add flow edge e to this flow network
Iterable<FlowEdge> adj(int v)
                                                              forward and backward edges incident to v
Iterable<FlowEdge> edges()
                                                                   all edges in this flow network
                  int V()
                                                                        number of vertices
                  int E()
                                                                        number of edges
               String toString()
                                                                       string representation
```

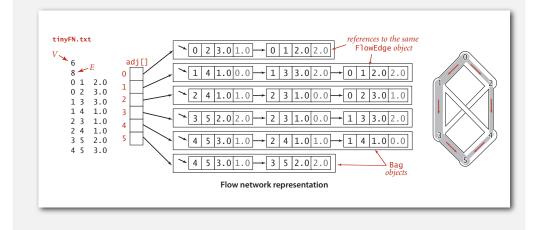
Conventions. Allow self-loops and parallel edges.

Flow network: Java implementation

```
public class FlowNetwork
                                                      same as EdgeWeightedGraph,
    private final int V;
                                                      but adjacency lists of
   private Bag<FlowEdge>[] adj;
                                                      FlowEdges instead of Edges
   public FlowNetwork(int V)
       this.V = V;
       adj = (Bag<FlowEdge>[]) new Bag[V];
       for (int v = 0; v < V; v++)
          adj[v] = new Bag<FlowEdge>();
   public void addEdge(FlowEdge e)
       int v = e.from();
       int w = e.to();
                                                     add forward edge
       adj[v].add(e);
                                                     add backward edge
       adj[w].add(e);
   public Iterable<FlowEdge> adj(int v)
    { return adj[v]; }
```

Flow network: adjacency-lists representation

Maintain vertex-indexed array of FlowEdge lists (use Bag abstraction).



Ford-Fulkerson: Java implementation

```
public class FordFulkerson
   private boolean[] marked; // true if s->v path in residual network
   private FlowEdge[] edgeTo; // last edge on s->v path
   private double value;
                               // value of flow
   public FordFulkerson(FlowNetwork G, int s, int t)
      value = 0:
      while (hasAugmentingPath(G, s, t))
                                                         bottleneck capacity
         double bottle = Double.POSITIVE INFINITY;
         for (int v = t; v != s; v = edgeTo[v].other(v))
            bottle = Math.min(bottle, edgeTo[v].residualCapacityTo(v));
         for (int v = t; v != s; v = edgeTo[v].other(v))
            edgeTo[v].addResidualFlowTo(v, bottle);
         value += bottle;
                                                         augment flow
   public double hasAugmentingPath(FlowNetwork G, int s, int t)
   { /* See next slide. */ }
   public double value()
   { return value; }
   public boolean inCut(int v)
                                      — is v reachable from s in residual graph?
     return marked[v]; }
```

Finding a shortest augmenting path (cf. breadth-first search)

```
private boolean hasAugmentingPath(FlowNetwork G, int s, int t)
   edgeTo = new FlowEdge[G.V()];
   marked = new boolean[G.V()];
   Queue<Integer> q = new Queue<Integer>();
   q.enqueue(s);
   marked[s] = true;
   while (!q.isEmpty())
       int v = q.dequeue();
       for (FlowEdge e : G.adj(v))
                                        found path from s to w
                                       in the residual network?
           int w = e.other(v);
           if (e.residualCapacityTo(w) > 0 && !marked[w])
                                     save last edge on path to w;
              edgeTo[w] = e;
                                    mark w:
              marked[w] = true; ←
                                     add w to the queue
              q.enqueue(w);
```

-

Maxflow and mincut applications

Maxflow/mincut is a widely applicable problem-solving model.

- · Data mining.
- · Open-pit mining.
- · Bipartite matching.
- · Network reliability.
- Image segmentation.
- Network connectivity.
- Distributed computing.
- Egalitarian stable matching.
- Security of statistical data.
- Multi-camera scene reconstruction.
- Sensor placement for homeland security.
- Baseball elimination. ← see next programming assignment
- Many, many, more.

liver and hepatic vascularization segmentation

Bipartite matching problem

N students apply for N jobs.



Each gets several offers.

→ applications



Is there a way to match all students to jobs?



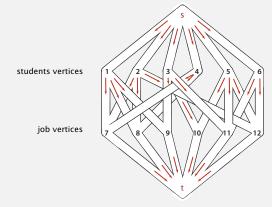
bipartite matching problem

57

1	Alice	7	Adobe
	Adobe		Alice
	Amazon		Bob
	Facebook		Dave
2	Bob	8	Amazon
	Adobe		Alice
	Amazon		Bob
	Yahoo		Dave
3	Carol	9	Facebook
	Facebook		Alice
	Google		Carol
	IBM	10	Google
4	Dave		Carol
	Adobe		Eliza
	Amazon	11	IBM
5	Eliza		Carol
	Google		Eliza
	IBM		Frank
	Yahoo	12	Yahoo
6	Frank		Bob
	IBM		Eliza
	Yahoo		Frank

Network flow formulation of bipartite matching

- Create s, t, one vertex for each student, and one vertex for each job.
- Add edge from s to each student (capacity 1).
- Add edge from each job to t (capacity 1).
- Add edge from student to each job offered (infinite capacity).



bipartite matching problem

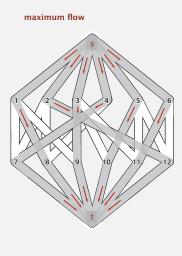
	partite mater	9	problem
1	Alice Adobe Amazon Facebook	7	Adobe Alice Bob Dave
2	Bob	8	
	Adobe		Alice
	Amazon		Bob
	Yahoo		Dave
3	Carol	9	Facebook
	Facebook		Alice
	Google		Carol
	IBM	10	Google
4	Dave		Carol
	Adobe		Eliza
	Amazon	11	IBM
5	Eliza		Carol
	Google		Eliza
	IBM		Frank
	Yahoo	12	Yahoo
6	Frank		Bob
	IBM		Eliza
	Yahoo		Frank

Network flow formulation of bipartite matching

1-1 correspondence between integer-valued maxflow solution of value N and perfect matchings.

perfect matching (solution)

Alice — Amazon
Bob — Yahoo
Carol — Facebook
Dave — Adobe
Eliza — Google
Frank — IBM



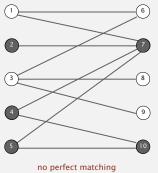
bipartite matching problem

1 Alice 7 Adobe Adobe Alice Amazon Bob Facebook Dave 2 Bob 8 Amazon Adobe Alice Amazon Bob Yahoo Dave 3 Carol 9 Facebook Facebook Alice Google Carol IBM 10 Google 4 Dave Adobe Eliza Amazon 11 IBM 5 Eliza Carol Google Eliza IBM Frank 12 Yahoo Yahoo 6 Frank Bob IBM Eliza Yahoo Frank

What the mincut tells us

Goal. When no perfect matching, explain why.

S = { 2, 4, 5 } T = { 7, 10 } student in S can be matched only to companies in T |S|>T

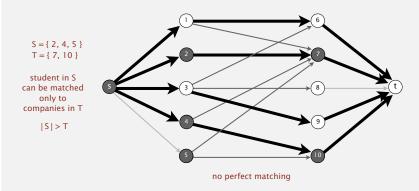


...

What the mincut tells us

Mincut. Consider mincut (A, B).

- Let S =students on sside of cut.
- Let T =companies on s side of cut.
- Fact: |S| > |T| and students in S can only be matched to companies in T.



Bottom line. When no perfect matching, mincut explains why.

Baseball elimination

Q. Which teams have a chance of finishing the season with the most wins?

i		team	wins	losses	to play	ATL	PHI	NYM	MON
0	A	Atlanta	83	71	8	-	1	6	1
1		Philly	80	79	3	1	_	0	2
2		New York	78	78	6	6	0	-	0
3		Montreal	77	82	3	1	2	0	-

Montreal is mathematically eliminated.

- Montreal finishes with ≤ 80 wins.
- Atlanta already has 83 wins.



Baseball elimination

Q. Which teams have a chance of finishing the season with the most wins?

i		team	wins	losses	to play	ATL	PHI	NYM	MON
0	A	Atlanta	83	71	8	-	1	6	1
1		Philly	80	79	3	1	-	0	2
2		New York	78	78	6	6	0	_	0
3		Montreal	77	82	3	1	2	0	-

Philadelphia is mathematically eliminated.

- Philadelphia finishes with ≤ 83 wins.
- Either New York or Atlanta will finish with ≥ 84 wins.



Observation. Answer depends not only on how many games already won and left to play, but on whom they're against.

Baseball elimination

Q. Which teams have a chance of finishing the season with the most wins?

i		team	wins	losses	to play	NYY	BAL	BOS	TOR	DET
0		New York	75	59	28	-	3	8	7	3
1	(4)	Baltimore	71	63	28	3	_	2	7	4
2		Boston	69	66	27	8	2	_	0	0
3		Toronto	63	72	27	7	7	0	-	0
4		Detroit	49	86	27	3	4	0	0	_

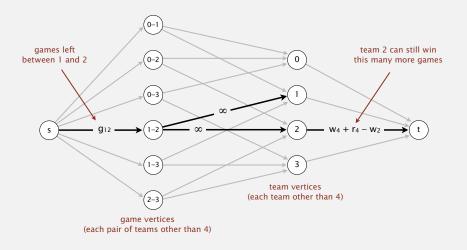
AL East (August 30, 1996)

Detroit is mathematically eliminated.

- Detroit finishes with ≤ 76 wins.
- Wins for $R = \{ NYY, BAL, BOS, TOR \} = 278.$
- Remaining games among $\{ NYY, BAL, BOS, TOR \} = 3 + 8 + 7 + 2 + 7 = 27.$
- Average team in R wins 305/4 = 76.25 games.

Baseball elimination: maxflow formulation

Intuition. Remaining games flow from s to t.



Fact. Team 4 not eliminated iff all edges pointing from s are full in maxflow.

Maximum flow algorithms: theory

(Yet another) holy grail for theoretical computer scientists.

year	method	worst case order of growth	discovered by
1951	simplex	E³ U	Dantzig
1955	augmenting path	E² U	Ford-Fulkerson
1970	shortest augmenting path	E ³	Dinitz, Edmonds-Karp
1972	fattest augmenting path	E ² log E log(EU)	Dinitz, Edmonds-Karp
1983	dynamic trees	E ² log E	Sleator-Tarjan
1985	capacity scaling	E² log U	Gabow
1997	length function	E ^{3/2} log E log U	Goldberg-Rao
?	?	Е	?

maxflow algorithms for sparse digraphs with E edges, integer capacities (max U)

Maximum flow algorithms: practice

Warning. Worst-case order-of-growth is generally not useful for predicting or comparing maxflow algorithm performance in practice.

Best in practice. Push-relabel method with gap relabeling: $E^{3/2}$.

On Implementing Push-Relabel Method for the Maximum Flow Problem

Boris V. Cherkassky¹ and Andrew V. Goldberg²

- ¹ Central Institute for Economics and Mathematics, Krasikova St. 32, 117418, Moscow, Russia cher@cemi.msk.su
- cher@cemi.msk.su

 Computer Science Department, Stanford University
 Stanford, CA 94305, USA
 goldberg@cs.stanford.edu

Abstract. We study efficient implementations of the push-relabel method for the maximum flow problem. The resulting codes are faster than the previous codes, and much faster on some problem families. The speedup is due to the combination of heurastics used in our implementations, we also exhibit a family of problems for which the running time of all known methods seem to have a roughly quadratic growth rate.



Summary

Mincut problem. Find an st-cut of minimum capacity.

Maxflow problem. Find an st-flow of maximum value.

Duality. Value of the maxflow = capacity of mincut.

Proven successful approaches.

- Ford-Fulkerson (various augmenting-path strategies).
- Preflow-push (various versions).

Open research challenges.

- Practice: solve real-word maxflow/mincut problems in linear time.
- Theory: prove it for worst-case inputs.
- · Still much to be learned!