

Algorithm Design and Implementation

Principle of Algorithms XI

Network Flow II

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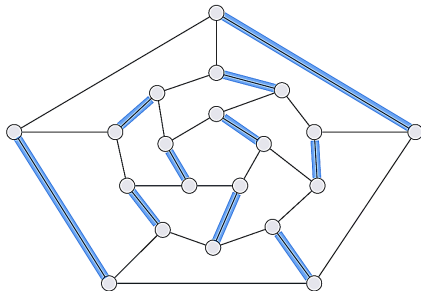
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Bipartite Matching

Definition

Given an undirected graph $G = (V, E)$, subset of edges $M \subseteq E$ is a **matching** if each node appears in at most one edge in M .

Max matching. Given a graph G , find a max-cardinality matching.

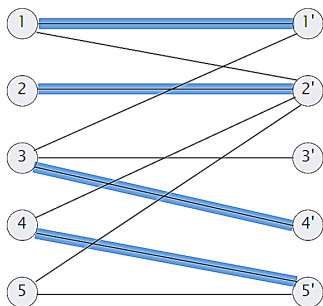


Bipartite matching

Definition

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 with a node in R .

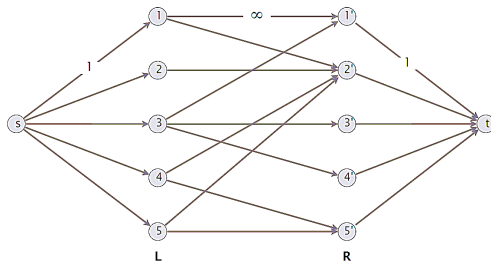
Bipartite matching. Given a bipartite graph $G = (L \cup R, E)$, find a max-cardinality matching.



Bipartite matching: max-flow formulation

Formulation.

- Create digraph $G' = (L \cup R \cup \{s, t\}, E')$.
- Direct all edges from L to R , and assign infinite (or unit) capacity.
- Add unit-capacity edges from s to each node in L .
- Add unit-capacity edges from each node in R to t .



Theorem

1-1 correspondence between matchings of cardinality k in G and integral flows of value k in G' .

Proof. \Rightarrow

- Let M be a matching in G of cardinality k .
- Consider flow f that sends 1 unit on each of the k corresponding paths.
- f is a flow of value k .

Theorem

1-1 correspondence between matchings of cardinality k in G and integral flows of value k in G' .

Proof. \Leftarrow

- Let f be an integral flow in G' of value k .
- Consider $M = \text{set of edges from } L \text{ to } R \text{ with } f(e) = 1$.
 - each node in L and R participates in at most one edge in M .
 - $|M| = k$: apply flow-value lemma to cut $(L \cup \{s\}, R \cup \{t\})$.

Max-flow formulation: proof of correctness

Theorem

1-1 correspondence between matchings of cardinality k in G and integral flows of value k in G' .

Corollary

Can solve bipartite matching problem via max-flow formulation.

Proof.

- Integrality theorem \Rightarrow there exists a max flow f^* in G' that is integral.
- 1-1 correspondence $\Rightarrow f^*$ corresponds to max-cardinality matching.

Quiz 1

What is running time of Ford–Fulkerson algorithms to find a max-cardinality matching in a bipartite graph with $|L| = |R| = n$?

- A. $O(m + n)$
- B. $O(mn)$
- C. $O(mn^2)$
- D. $O(m^2 n)$

Perfect matchings in bipartite graphs

Definition

Given a graph $G = (V, E)$, a subset of edges $M \subseteq E$ is a **perfect matching** if each node appears in exactly one edge in M .

Q. When does a bipartite graph have a perfect matching?

Structure of bipartite graphs with perfect matchings.

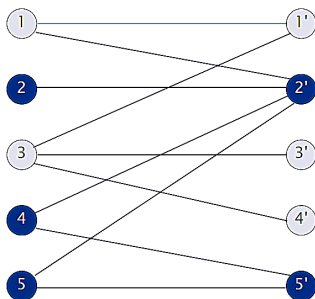
- Clearly, we must have $|L| = |R|$.
- Which other conditions are necessary?
- Which other conditions are sufficient?

Perfect matchings in bipartite graphs

Notation. Let S be a subset of nodes, and let $N(S)$ be the set of nodes adjacent to nodes in S .

Observation. If a bipartite graph $G = (L \cup R, E)$ has a perfect matching, then $|N(S)| \geq |S|$ for all subsets $S \subseteq L$.

Proof. Each node in S has to be matched to a different node in $N(S)$.



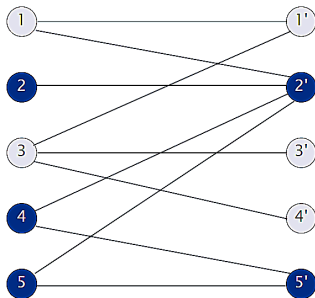
Hall's marriage theorem

Theorem (Frobenius 1917, Hall 1935)

Let $G = (L \cup R, E)$ be a bipartite graph with $|L| = |R|$. Then, graph G has a perfect matching iff $|N(S)| \geq |S|$ for all subsets $S \subseteq L$.

Proof. \Rightarrow

This was the previous observation.

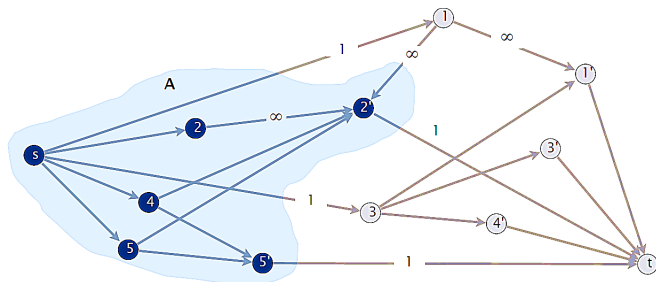


Hall's marriage theorem

Proof. \Leftarrow

Suppose G does not have a perfect matching.

- Formulate as a max-flow problem and let (A, B) be a min cut in G' .
- By max-flow min-cut theorem, $\text{cap}(A, B) < |L|$.
- Define $L_A = L \cap A, L_B = L \cap B, R_A = R \cap A$.
- $\text{cap}(A, B) = |L_B| + |R_A| \Rightarrow |R_A| < |L_A|$
- Min cut can't use ∞ edges $\Rightarrow N(L_A) \subseteq R_A$.
- $|N(L_A)| \leq |R_A| < |L_A|$.
- Choose $S = L_A$.



Bipartite matching

Problem. Given a bipartite graph, find a max-cardinality matching.

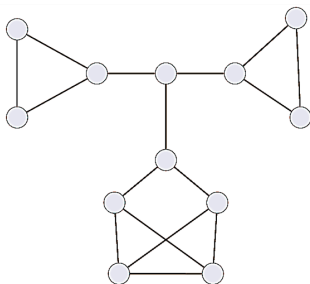
year	worst case	technique	discovered by
1955	$O(mn)$	augmenting path	Ford–Fulkerson
1973	$O(mn^{1/2})$	blocking flow	Hopcroft–Karp, Karzanov
2004	$O(n^{2.378})$	fast matrix multiplication	Mucha–Sankowski
2013	$\tilde{O}(m^{10/7})$	electrical flow	Madry
20xx	???		

running time for finding a max-cardinality matching in a bipartite graph with n nodes and m edges

Quiz 2

Which of the following are properties of the graph $G = (V, E)$?

- A. G has a perfect matching.
- B. Hall's condition is satisfied: $|N(S)| \geq |S|$ for all subsets $S \subseteq V$.
- C. Both A and B.
- D. Neither A nor B.



Problem. Given an undirected graph, find a max-cardinality matching.

- Structure of nonbipartite graphs is more complicated.
- But well understood. [Tutte–Berge formula, Edmonds–Gallai]
- Blossom algorithm: $O(n^4)$. [Edmonds 1965]
- Best known: $O(mn^{1/2})$. [Micali–Vazirani 1980, Vazirani 1994]

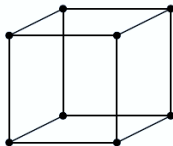
Hackathon problem

Hackathon problem.

- Hackathon attended by n Harvard students and n Princeton students.
- Each Harvard student is friends with exactly $k > 0$ Princeton students; each Princeton student is friends with exactly k Harvard students.
- Is it possible to arrange the hackathon so that each Princeton student pair programs with a different friend from Harvard?

Mathematical reformulation. Does every k -regular bipartite graph have a perfect matching?

Ex. Boolean hypercube.



Hackathon problem

Theorem

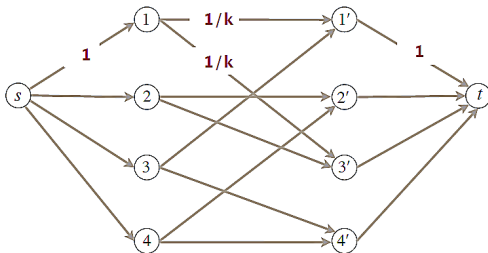
Every k -regular bipartite graph G has a perfect matching.

Proof.

- Size of max matching = value of max flow in G' .
- Consider flow

$$f(u, v) = \begin{cases} 1 & \text{if } u = s \text{ or } v = t \\ 1/k & \text{otherwise} \end{cases}$$

- The value of flow f is $n \Rightarrow G'$ has a perfect matching.



Disjoint Paths

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

Edge-disjoint paths problem. Given a digraph $G = (V, E)$ and two nodes s and t , find the max number of edge-disjoint $s \rightsquigarrow t$ paths.

Max-flow formulation. Assign unit capacity to every edge.

Theorem

1-1 correspondence between k edge-disjoint $s \rightsquigarrow t$ paths in G and integral flows of value k in G' .

Proof. \Rightarrow

- Let P_1, \dots, P_k be k edge-disjoint $s \rightsquigarrow t$ paths in G .
- Set $f(e) = \begin{cases} 1 & \text{edge } e \text{ participates in some path } P_j \\ 0 & \text{otherwise} \end{cases}$
- Since paths are edge-disjoint, f is a flow of value k .

Theorem

1-1 correspondence between k edge-disjoint $s \rightsquigarrow t$ paths in G and integral flows of value k in G' .

Proof. \Leftarrow

- Let f be an integral flow in G' of value k .
- Consider edge (s, u) with $f(s, u) = 1$.
 - by flow conservation, there exists an edge (u, v) with $f(u, v) = 1$.
 - continue until reach t , always choosing a new edge
- Produces k (not necessarily simple) edge-disjoint paths.

Edge-disjoint paths

Theorem

1-1 correspondence between k edge-disjoint $s \rightsquigarrow t$ paths in G and integral flows of value k in G' .

Corollary

Can solve edge-disjoint paths problem via max-flow formulation.

Proof.

- Integrality theorem \Rightarrow there exists a max flow f^* in G' that is integral.
- 1-1 correspondence $\Rightarrow f^*$ corresponds to max number of edge-disjoint $s \rightsquigarrow t$ paths in G .

Definition. A set of edges $F \subseteq E$ disconnects t from s if every $s \rightsquigarrow t$ path uses at least one edge in F .

Network connectivity. Given a digraph $G = (V, E)$ and two nodes s and t , find minimal number of edges whose removal disconnects t from s .

Theorem (Menger 1927)

The max number of *edge-disjoint* $s \rightsquigarrow t$ paths equals the min number of edges whose removal *disconnects* t from s .

Proof. \leq

- Suppose the removal of $F \subseteq E$ disconnects t from s , and $|F| = k$.
- Every $s \rightsquigarrow t$ path uses at least one edge in F .
- Hence, the number of edge-disjoint paths is $\leq k$.

Theorem (Menger 1927)

The max number of *edge-disjoint* $s \rightsquigarrow t$ paths equals the min number of edges whose removal *disconnects* t from s .

Proof. \geq

- Suppose max number of edge-disjoint $s \rightsquigarrow t$ paths is k .
- Then value of max flow = k .
- Max-flow min-cut theorem \Rightarrow there exists a cut (A, B) of capacity k .
- Let F be set of edges going from A to B .
- $|F| = k$ and disconnects t from s .

How to find the max number of edge-disjoint paths in an undirected graph?

- A. Solve the edge-disjoint paths problem in a digraph
(by replacing each undirected edge with two antiparallel edges).
- B. Solve a max flow problem in an undirected graph.
- C. Both A and B.
- D. Neither A nor B.

Definition

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

Edge-disjoint paths problem in undirected graphs. Given a graph $G = (V, E)$ and two nodes s and t , find the max number of edge-disjoint s - t paths.

Edge-disjoint paths in undirected graphs

Max-flow formulation. Replace each edge with two antiparallel edges and assign unit capacity to every edge.

Observation. Two paths P_1 and P_2 may be edge-disjoint in the digraph but not edge-disjoint in the undirected graph.

Edge-disjoint paths in undirected graphs

Lemma

In any flow network, there exists a maximum flow f in which for each pair of antiparallel edges e and e' : either $f(e) = 0$ or $f(e') = 0$ or both. Moreover, integrality theorem still holds.

Proof. [by induction on number of such pairs]

- Suppose $f(e) > 0$ and $f(e') > 0$ for a pair of antiparallel edges e and e' .
- Set $f(e) = f(e) - \delta$ and $f(e') = f(e') - \delta$, where $\delta = \min\{f(e), f(e')\}$.
- f is still a flow of the same value but has one fewer such pair.

Edge-disjoint paths in undirected graphs

Lemma

In any flow network, there exists a maximum flow f in which for each pair of antiparallel edges e and e' : either $f(e) = 0$ or $f(e') = 0$ or both. Moreover, integrality theorem still holds.

Theorem

Max number of edge-disjoint $s \rightsquigarrow t$ paths = value of max flow.

Proof. Similar to proof in digraphs; use lemma.

More Menger theorems

Theorem

Given an *undirected* graph and two nodes s and t , the max number of *edge-disjoint* s - t paths equals the min number of edges whose removal disconnects s and t .

Theorem

Given an *undirected* graph and two nodes s and t , the max number of internally *node-disjoint* s - t paths equals the min number of internal nodes whose removal disconnects s and t .

Theorem

Given an *directed* graph with two nonadjacent nodes s and t , the max number of internally *node-disjoint* $s \rightsquigarrow t$ paths equals the min number of internal nodes whose removal disconnects t and s .

Extensions to Max Flow

Which extensions to max flow can be easily modeled?

- A. Multiple sources and multiple sinks.
- B. Undirected graphs.
- C. Lower bounds on edge flows.
- D. All of the above.

Definition

Given a digraph $G = (V, E)$ with edge capacities $c(e) \geq 0$ and multiple source nodes and multiple sink nodes, find max flow that can be sent from the source nodes to the sink nodes.

Multiple sources and sinks: max-flow formulation

- Add a new source node s and sink node t .
- For each original source node s_i add edge (s, s_i) with capacity ∞ .
- For each original sink node t_j , add edge (t_j, t) with capacity ∞ .

Claim

1-1 correspondence between flows in G and G' .

Definition

Given a digraph $G = (V, E)$ with edge capacities $c(e) \geq 0$ and node demands $d(v)$, a **circulation** is a function $f(e)$ that satisfies:

- For each $e \in E$: $0 \leq f(e) \leq c(e)$ **capacity**
- For each $v \in V$: $\sum_{e \text{ into } v} f(e) - \sum_{e \text{ out of } v} f(e) = d(v)$ **flow conservation**

Circulation with supplies and demands: max-flow formulation

- Add new source s and sink t .
- For each v with $d(v) < 0$, add edge (s, v) with capacity $-d(v)$.
- For each v with $d(v) > 0$, add edge (v, t) with capacity $d(v)$.

Claim

G has circulation iff G' has max flow of value

$$D = \sum_{v:d(v)>0} d(v) = \sum_{v:d(v)<0} -d(v)$$

Integrality theorem. If all capacities and demands are integers, and there exists a circulation, then there exists one that is integer-valued.

Proof. Follows from max-flow formulation + integrality theorem for max flow.

Theorem

Given (V, E, c, d) , there does *not* exist a circulation iff there exists a node partition (A, B) such that $\sum_{v \in B} d(v) > \text{cap}(A, B)$.

Proof sketch. Look at min cut in G' .

Definition

Given a digraph $G = (V, E)$ with edge capacities $c(e) \geq 0$, lower bounds $\ell(e) \geq 0$, and node demands $d(v)$, a circulation $f(e)$ is a function that satisfies:

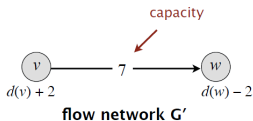
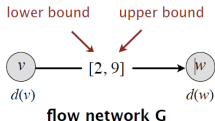
- For each $e \in E : \ell(e) \leq f(e) \leq c(e)$ capacity
- For each $v \in V : \sum_{e \text{ into } v} f(e) - \sum_{e \text{ out of } v} f(e) = d(v)$
flow conservation

Circulation problem with lower bounds. Given (V, E, ℓ, c, d) , does there exist a feasible circulation?

Circulation with supplies, demands, and lower bounds

Max-flow formulation. Model lower bounds as circulation with demands.

- Send $\ell(e)$ units of flow along edge e .
- Update demands of both endpoints.



Theorem

There exists a circulation in G iff there exists a circulation in G' . Moreover, if all demands, capacities, and lower bounds in G are integers, then there exists a circulation in G that is integer-valued.

Proof sketch. $f(e)$ is a circulation in G iff $f'(e) = f(e) - \ell(e)$ is a circulation in G' .

Survey Design

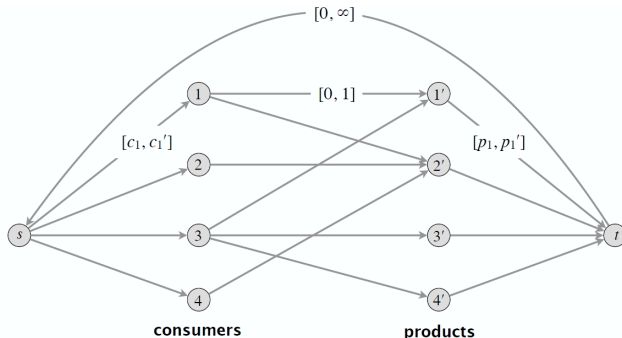
- Design **survey** asking n_1 consumers about n_2 products.
- Can survey consumer i about product j only if they own it
- Ask consumer i between c_i and c'_i questions.
- Ask between p_j and p'_j consumers about product j .

Goal. Design a survey that meets these specs, if possible.

Bipartite perfect matching. Special case when $c_i = c'_i = p_j = p'_j = 1$.

Max-flow formulation. Model as a circulation problem with lower bounds.

- Add edge (i, j) if consumer j owns product i .
- Add edge from s to consumer j .
- Add edge from product i to t .
- Add edge from t to s .
- All demands = 0.
- Integer circulation \Leftrightarrow feasible survey design.



Airline Scheduling

Airline scheduling.

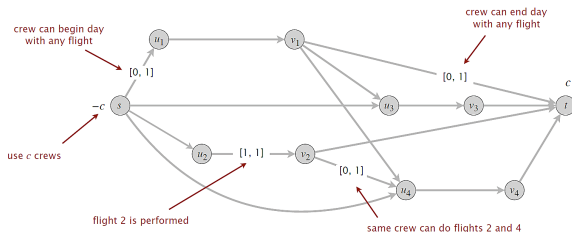
- Complex computational problem faced by airline carriers.
- Must produce schedules that are efficient in terms of equipment usage, crew allocation, and customer satisfaction.
- One of largest consumers of high-powered algorithmic techniques.

"Toy problem."

- Manage flight crews by reusing them over multiple flights.
- Input: set of k flights for a given day.
- Flight i leaves origin o_i at time s_i and arrives at destination d_i at time f_i .
- Minimize number of flight crews.

Circulation formulation. [to see if c crews suffice]

- For each flight i , include two nodes u_i and v_i .
- Add source s with demand $-c$, and edges (s, u_i) with capacity 1.
- Add sink t with demand c , and edges (v_i, t) with capacity 1.
- For each i , add edge (u_i, v_i) with lower bound and capacity 1.
- If flight j reachable from i , add edge (v_i, u_j) with capacity 1.



Theorem

The airline scheduling problem can be solved in $O(k^3 \log k)$ time.

Proof.

- k = number of flights.
- c = number of crews (unknown).
- $O(k)$ nodes, $O(k^2)$ edges.
- At most k crews needed.
 - ⇒ solve \log_2 circulation problems. ← binary search for min value c^*
- Value of any flow is between 0 and k
 - ⇒ at most k augmentations per circulation problem.
- Overall time = $O(k^3 \log k)$.

Remark

*Can solve in $O(k^3)$ time by formulating as **minimum-flow problem**.*

Remark. We solved a toy version of a real problem.

Real-world problem models countless other factors:

- Union regulations: e.g., flight crews can fly only a certain number of hours in a given time window.
- Need optimal schedule over planning horizon, not just one day.
- Deadheading has a cost.
- Flights don't always leave or arrive on schedule.
- Simultaneously optimize both flight schedule and fare structure

Message.

- Our solution is a generally useful technique for efficient reuse of limited resources but trivializes real airline scheduling problem.
- Flow techniques useful for solving airline scheduling problems and are widely used in practice.
- Running an airline efficiently is a very difficult problem.