

COMPSCI 311: Introduction to Algorithms

Lecture 18: Network Flow Applications

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slides credit: Dan Sheldon (adapted)

Review: Network Flow

- ▶ Residual graph G_f .
Capacities: $c(e) - f(e)$ forward, $f(e)$ backward.
- ▶ Ford-Fulkerson
Initialize flow f to all zeros
▷ Augment flow as long as it is possible
while there exists an s - t path P in G_f **do**
 $f = \text{Augment}(f, P)$
 update residual graph G_f
end while
- ▶ Analysis
 - ▶ Always maintain a flow: use facts of residual graph and augment operation, verify that definition of flow still holds
 - ▶ Termination and running time: flow increases at least one in each iteration, and cannot exceed total capacity leaving s
 - ▶ Correctness: Max-Flow Min-Cut Theorem

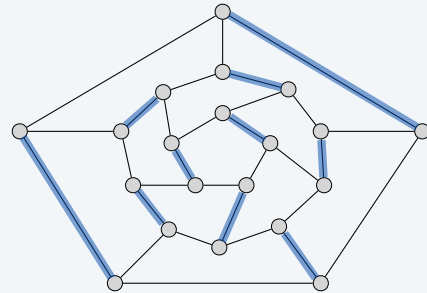
Max-Flow Min-Cut Theorem

- ▶ $v(f) \leq c(A, B)$ for any flow f and any s - t cut $c(A, B)$
- ▶ On termination, Ford-Fulkerson produces a max-flow f and min-cut (A, B)
- ▶ A = set of nodes reachable from s in residual graph to find the min-cut
- ▶ Complexity
 - ▶ $O(mnC_{\max})$ for basic version
 - ▶ $O(m^2 \log C_{\max})$ for capacity scaling
 - ▶ Capacity-independent for choosing shortest augmenting paths
 $O(m^2n)$ Edmonds-Karp, $O(n^2m)$ Dinitz

Matching

Def. 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.

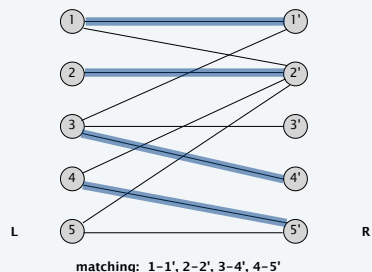


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Bipartite matching

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 with a node in R .

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



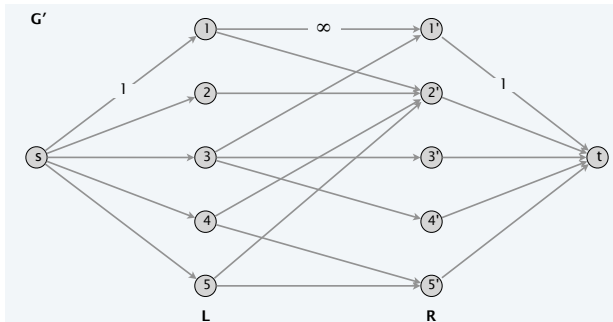
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Formulating Matching as Network Flow problem

- ▶ **Goal:** given matching instance $G = (L \cup R, E)$:
 - ▶ create a flow network G' ,
 - ▶ find a maximum flow f in G'
 - ▶ use f to construct a maximum matching M in G .
- ▶ Standard approach for reducing a problem to network flow
- ▶ Intuition?
Connect left set L to source, right set R to sink.
Larger matchings should have larger flows.
How to select capacities?

Maximal Matching as Network Flow

- ▶ Add a source s and sink t
- ▶ For each edge $(u, v) \in E$, add $u \rightarrow v$ (directed), capacity 1
- ▶ Add an edge with capacity 1 from s to each node $u \in L$
- ▶ Add an edge with capacity 1 from each node $v \in R$ to t .



- ▶ Does it matter if we have unit or infinite capacities from L to

Clicker Question 1

Suppose I want to work with ∞ values (for shortest path, minimum cost, maximum flow, etc.) What properties would I need to add and compare values?

- A: $x \neq \infty \rightarrow x < \infty$
- B: $\infty + \infty = \infty$
- C: Both A and B
- D: A and something stronger than B

Maximal Matching: Analysis

- ▶ Run F-F to get an **integral** max-flow f
- ▶ Set M to the set of edges from L to R with flow $f(e) = 1$
- ▶ **Claim:** The set M is a maximum matching.

Let's prove that:

- Integer flow f in $G' \implies$ matching M in G with $|M| = v(f)$
- Matching M in $G \implies$ flow f in G' with $v(f) = |M|$

Therefore, max-flow f in $G' \iff$ maximum matching M in G

Proof of 1: given f , construct M

- ▶ $M =$ edges from L to R carrying one unit of flow
- ▶ Capacity constraints \implies at most 1 unit of flow leaving $u \in L$
- ▶ Edge flows are 0 or 1 $\implies M$ has at most one edge incident to u .
- ▶ Similar argument for $v \in R$

Maximal Matching: Analysis

Proof of 2: given M , construct f

- ▶ Set $f(e) = 1$ if $e \in M$
- ▶ Send one unit of flow from s to $u \in L$ if u is matched
- ▶ Send one unit of flow from $v \in R$ to t if t is matched
- ▶ All other edge flow values are zero
- ▶ Verify that capacity and flow conservation constraints are satisfied, and that $v(f) = |M|$.

Network flow II: quiz 1



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

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

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Perfect matchings in bipartite graphs

Def. 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?

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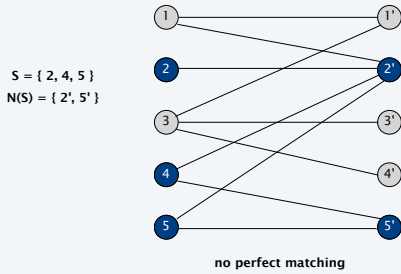
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Perfect matchings in bipartite graphs

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 $S \subseteq L$.

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



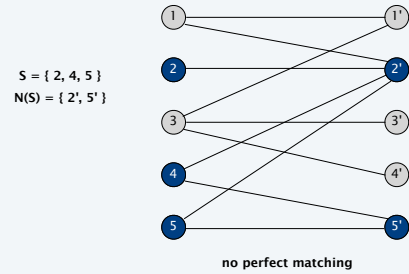
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Hall's marriage theorem

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 $S \subseteq L$.



Pf. → This was the previous observation.

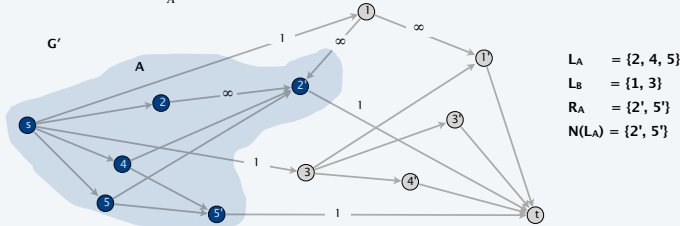


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Hall's marriage theorem

Pf. ← 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, $cap(A, B) < |L|$.
- Define $L_A = L \cap A$, $L_B = L \cap B$, $R_A = R \cap A$.
- $cap(A, B) = |L_B| + |R_A| \Rightarrow |R_A| < |L_A|$.
- Min cut can't use ∞ edges $\Rightarrow N(L_A) \subseteq R_A$.
- $|N(L_A)| \leq |R_A| < |L_A|$.
- Choose $S = L_A$. ■

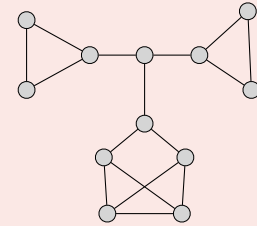


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Network flow II: quiz 2

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.



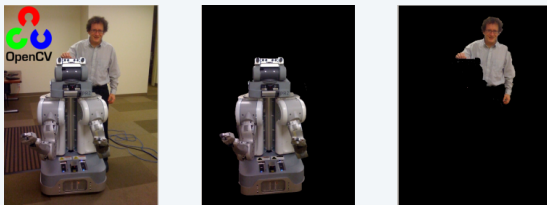
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Image segmentation

Image segmentation.

- Divide image into coherent regions.
- Central problem in image processing.

Ex. Separate human and robot from background scene.



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Grabcut image segmentation

Grabcut. [Rother-Kolmogorov-Blake 2004]

"GrabCut" — Interactive Foreground Extraction using Iterated Graph Cuts

Carsten Rother¹ Vladimir Kolmogorov¹ Andrew Blake²
 Microsoft Research Cambridge, UK



Figure 1: Three examples of GrabCut. The user drags a rectangle loosely around an object. The object is then extracted automatically.

slide credit: Kevin Wayne / Pearson

Bokeh Effect: Blurring Background

- ▶ Using an expensive camera and appropriate lenses, you can get a "bokeh" effect on portrait photos: the background is blurred and the foreground is in focus.



- ▶ Can fake effect using cheap phone cameras and appropriate software

Formulating the Problem

Given set V of pixels, classify each as foreground or background. Assume you have:

- ▶ Likelihood that a pixel is in foreground (a_i) / background (b_i)
- ▶ Numeric penalty p_{ij} for assigning neighboring pixels i and j to different classes

Graph edges E : for each pixel, edge to neighbors (4? 8? other?)

Criteria:

- ▶ Accuracy if $a_i > b_i$, would prefer to label pixel i as foreground
- ▶ Smoothness: if many neighbors are labeled the same (foreground), would like to label pixel i as foreground (minimize penalties)

Image Segmentation as Network Flow

Maximize correct labeling scores, minimize penalties

Let A : set of pixels labeled foreground, B : pixels in background

$$\text{Maximize: } \sum_{i \in A} a_i + \sum_{j \in B} b_j - \sum_{(i,j) \in E, i \in A, j \in B} p_{ij}$$

Insight: (A, B) is a partition \Rightarrow forms a **cut**

First sum is $\sum_{i \in V} (a_i + b_i) - \sum_{i \in A} b_i - \sum_{j \in B} a_j$
(constant minus "penalties" for mislabeling)

$$\text{Must minimize } \sum_{i \in A} b_i + \sum_{j \in B} a_j + \sum_{(i,j) \in E, i \in A, j \in B} p_{ij}$$

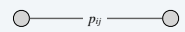
\Rightarrow find **minimum cut**

Image segmentation

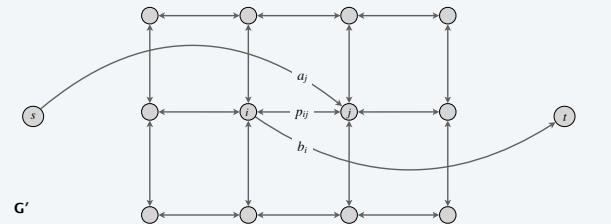
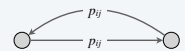
Formulate as min-cut problem $G' = (V', E')$.

- Include node for each pixel.
- Use two antiparallel edges instead of undirected edge.
- Add source s to correspond to foreground.
- Add sink t to correspond to background.

edge in G



two antiparallel edges in G'



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Image segmentation

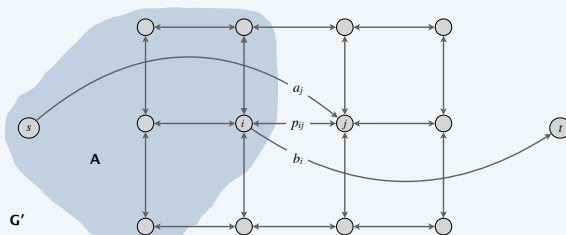
Consider min cut (A, B) in G' .

- A = foreground.

$$\text{cap}(A, B) = \sum_{j \in B} a_j + \sum_{i \in A} b_i + \sum_{\substack{(i,j) \in E \\ i \in A, j \in B}} p_{ij}$$

if i and j on different sides, p_{ij} counted exactly once

- Precisely the quantity we want to minimize.



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More Network Flows

- ▶ Extensions

- ▶ Multiple sources and sinks
- ▶ Circulations with supplies and demands
- ▶ Flows with lower bounds

- ▶ Improved Algorithms: Preflow-push $O(n^3)$

- ▶ Applications

- ▶ Network connectivity
- ▶ Data mining: survey design
- ▶ Airline scheduling
- ▶ Baseball elimination
- ▶ Multi-camera placement / scene reconstruction