

CSE 421

Image Segmentation, Polynomial Time Reductions

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Image Segmentation

Given an image we want to separate foreground from background

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



Foreground / background segmentation

Label each pixel as foreground/background.

- V = set of pixels, E = pairs of neighboring pixels.
- $a_i \ge 0$ is likelihood pixel i in foreground.
- $b_i \ge 0$ is likelihood pixel i in background.
- *p_{i,j}* ≥ 0 is separation penalty for labeling one of i and j as foreground, and the other as background.
 Goals.

Accuracy: if $a_i > b_i$ in isolation, prefer to label i in foreground.

Smoothness: if many neighbors of i are labeled foreground, we should be inclined to label i as foreground.

Find partition (A, B) that maximizes:

Foreground
$$\sum_{i \in A} a_i + \sum_{j \in B} b_j - \sum_{\substack{(i,j) \in E \\ i \in A, j \in B}} p_{i,j}$$

Background



Image Seg: Min Cut Formulation

Difficulties:

- Maximization (as opposed to minimization)
- No source or sink
- Undirected graph
- Step 1: Turn into Minimization

Maximizing
$$\sum_{i \in A} a_i + \sum_{j \in B} b_j - \sum_{\substack{(i,j) \in E \\ i \in A, j \in B}} p_{i,j}$$

Equivalent to minimizing

$$+\sum_{i\in V}a_i + \sum_{j\in V}b_j - \sum_{i\in A}a_i - \sum_{j\in B}b_j + \sum_{\substack{(i,j)\in E\\i\in A,j\in B}}p_{i,j}$$

Equivalent to minimizing

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

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Min cut Formulation (cont'd)

G' = (V', E').
Add s to correspond to foreground;
Add t to correspond to background
Use two anti-parallel edges instead of undirected edge.





Min cut Formulation (cont'd)

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

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

Precisely the quantity we want to minimize.



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Reductions & NP-Completeness

Computational Complexity

Goal: Classify problems according to the amount of computational resources used by the best algorithms that solve them

Here we focus on time complexity

Recall: worst-case running time of an algorithm

• **max** # steps algorithm takes on any input of size **n**

Computational Complexity and Reduction

In most cases, we cannot characterize the true hardness of a computational problem

So?

We only reduce the number of problems

Want to be able to make statements of the form

- "If we could solve problem B in polynomial time then we can solve problem A in polynomial time"
- "Problem B is at least as hard as problem A"

Polynomial Time Reduction

Def A \leq_P B: if there is an algorithm for problem A using a 'black box' (subroutine) that solve problem B s.t.,

- Algorithm uses only a polynomial number of steps
- Makes only a polynomial number of calls to a subroutine for B



In words, B is as hard as A (it can be even harder)

 \leq_p^1 Reductions

In this lecture we see a restricted form of polynomial-time reduction often called Karp or many-to-one reduction

 $A \leq_p^1 B$: if and only if there is an algorithm for A given a black box solving B that on input **x**

- Runs for polynomial time computing an input f(x) of B
- Makes one call to the black box for B for input f(x)
- Returns the answer that the black box gave

We say that the function f(.) is the reduction

Example 1: Indep Set \leq_p Clique

Indep Set: Given G=(V,E) and an integer k, is there $S \subseteq V$ s.t. $|S| \ge k$ an no two vertices in S are joined by an edge?

Clique: Given a graph G=(V,E) and an integer k, is there $S \subseteq V$, $|U| \ge k$ s.t., every pair of vertices in S is joined by an edge?

Claim: Indep Set \leq_p Clique

Pf: Given G = (V, E) and instance of indep Set. Construct a new graph G' = (V, E') where $\{u, v\} \in E'$ if and only if $\{u, v\} \notin E$.



Example 2: Vertex Cover \leq_p Indep Set

Vertex Cover: Given a graph G=(V,E) and an integer k, is there a vertex cover of size at most k?

Claim: For any graph G = (V, E), S is an independent set iff V - S is a vertex cover Pf: => Let S be a independent set of G Then, S has at most one endpoint of every edge of G So, V - S has at least one endpoint of every edge of G So, V - S is a vertex cover.

 \leq Suppose *V* – *S* is a vertex cover

Then, there is no edge between vertices of S (otherwise, V - S is not a vertex cover)

So, *S* is an independent set.

Example 3: Vertex Cover \leq_p Set Cover

Set Cover: Given a set U, collection of subsets $S_1, ..., S_m$ of U and an integer k, is there a collection of k sets that contain all elements of U?

Claim: Vertex Cover \leq_p Set Cover

Pf:

Given (G = (V, E), k) of vertex cover we construct a set cover input f(G, k)

- U = E
- For each $v \in V$ we create a set S_v of all edges connected to v

This clearly is a polynomial-time reduction

So, we need to prove it gives the right answer

Example 3: Vertex Cover \leq_p Set Cover

Claim: Vertex Cover \leq_p Set Cover Pf: Given (G = (V, E), k) of vertex cover we construct a set cover input f(G, k)

- U = E
- For each $v \in V$ we create a set S_v of all edges connected to v

Vertex-Cover (G,k) is yes => Set-Cover f(G,k) is yes

If a set $W \subseteq V$ covers all edges, just choose S_v for all $v \in W$, it covers all U.

Set-Cover f(G,k) is yes => Vertex-Cover (G,k) is yes If $(S_{v_1}, ..., S_{v_k})$ covers all U, the set $\{v_1, ..., v_k\}$ covers all edges of G.

Decision Problems

A decision problem is a computational problem where the answer is just yes/no

Here, we study computational complexity of decision Problems.

Why?

- much simpler to deal with
- Decision version is not harder than Search version, so it is easier to lower bound Decision version
- Less important, usually, you can use decider multiple times to find an answer .