Course Logistics

- CLRS Chapter 34

- Homework due Friday

1 Approximation Algorithms for Optimization Problems

There are so many hard problems out there without known polynomial time solutions! Is all hope lost? What do we do?

Approximation algorithms "quickly" find an answer that is "close" to the solution.

We are now focused on optimization problems, rather than just their decision versions.

Definition. Let Q be a computational minimization problem (e.g., find a minimum vertex cover, find a minimum s-t cut) and assume that C^* is the optimal (minimum) solution to the problem. An _____ for Q with approximation factor p is an algorithm that:

- Runs in

- Outputs a solution value C that is guaranteed to satisfy

Question 1. What is a lower bound we must assume holds for the value of p when defining an approximation algorithm?

A No lower bound is needed, p can be any real number.

B $p \geq -1$.

C $p \geq 0$.

D $p \geq 1$.

E $p \geq C^*$.

We can also have approximation algorithms for maximization problems, but in this case C^* represents the maximum value (i.e., the solution) for the problem and _____

Wait, is this even possible? We want to guarantee that $\frac{C}{C^*}$ is small in polynomial time, but finding C^* is NP-hard. How can we do that?

To design an approximation algorithm, we need two pieces:

1. _____ A procedure for finding a value $\hat{C}$ that satisfies _____

   - It should take polynomial time.

   - This method _____ solve the original problem, but is often re-lated.

2. _____ An algorithm for the original problem that returns a suboptimal solution _____

   - It should take polynomial time.

   - This method often:

We then must prove that _____ and this provides a p-approximation!

Caveat. Often, the lower bounding procedure is *explicit*: there is an actual algorithm that computes the lower bound, and then the upper bounding algorithm explicitly uses the lower bound. However, in some cases no explicit lower bound is computed, and instead one shows implicitly that the upper bounding algorithm must provide a solution that is better than some lower bound, even though the lower bound isn't computed explicitly. This can be tricky, but it is often possible!

2   Matchings and Vertex Covers

Let $G = (V, E)$ be an undirected and unweighted graph.

A **matching** $M \subseteq E$ is a set of edges such that no two edges share the same node.

A **vertex cover** $C \subseteq V$ is a set of nodes such that each edge $(u, v) \in E$ has at least one node in C.

Lemma 2.1. *Let M be a matching of G and C be a vertex cover. Then $|M| \leq |C|$.*

3   The approximation algorithm for vertex cover

A matching $M \subseteq E$ is a **maximal matching** if for every edge $e \in E \setminus M$, $M \cup \{e\}$ is no longer a matching.

The algorithm

`VertexCoverApprox`$(G = (V, E))$

1. Compute a maximal matching M of G:

    - Set $F = E$, $M = \emptyset$.
    - While $|F| > 0$:
        - Add any edge $e \in F$ to M.
        - For each remaining $f \in F$, if $|e \cap f| > 0$, remove f from F.

2. Let S be the set of nodes adjacent to an edge in M.

3. Return S.

Theorem 3.1. *Let C^* be the minimum sized vertex cover of G. The algorithm* `VertexCoverApprox` *runs in polynomial time in terms of the size of G and outputs a vertex cover $S \subseteq V$ satisfying $|S| \leq 2C^*$. Thus, this is a 2-approximation algorithm for vertex cover.*

4   Linear Programming

A **linear program** is a mathematical optimization problem with:

- A linear objective function

- Linear constraints

We will use x_i to denote variables—unknowns that we need to find—to make the objective function as large as possible, and such that the constraints hold.

Examples

5   Types of Linear Programs

Linear programs have many variations:

- The objective function can be a maximization or a minimization problem.

- The constraints can be equalities or inequalities.

- Often, the variables will be constrained to be greater than or equal to zero.

Actually, we can perform different conversions to:

- Turn a maximization problem into a minimization problem.

- Turn an equality constraint into inequality constraints.

- Turn an inequality constraint into an equality constraint.

- Turn an unconstrained variable into a pair of positive variables.

Important Fact: A linear program can be solved in polynomial time.

6 Graph Problems as Mathematical Optimization Problems

We can write the maximum s-t flow problem as the following optimization problem:

The weighted vertex cover problem can be written as the following optimization problem:

Question 2. Which of the above two problems is a linear program?

A The first.

B The second.

C Both.

D Neither.

7 Linear Programming Relaxation for Weighted Vertex Cover

This is the integer program for the weighted vertex cover problem:

Question 3. Let $\hat{C}$ be the solution to the linear programming relaxation of the weighted vertex cover integer program, and let C^* be the minimum weighted vertex cover solution. Which of the following is always true?

A $\quad \hat{C} \leq C^*$

B $\quad \hat{C} \geq C^*$

C $\quad \hat{C} < C^*$

D $\quad \hat{C} > C^*$

E $\quad \hat{C} = C^*$

8 The Algorithm

`WeightedVertexCoverApprox`$(G = (V, E))$

1. Solve the linear programming relaxation of the weighted vertex cover problem.

2. For each node $v \in V$, if $x_v \geq \frac{1}{2}$, add v to a node set S.
   In other words, define:
   $$S = \{v \in V : x_v \geq \frac{1}{2}\}$$

3. Return S as a vertex cover.

Theorem 8.1. *Let C^* be the minimum weighted vertex cover of G. The algorithm* `WeightedVertexCoverApprox` *runs in polynomial time in terms of the size of G and outputs a vertex cover $S \subseteq V$ satisfying:*

$$\sum_{v \in S} w_v x_v \leq 2C^*.$$

Thus, this is a 2-approximation algorithm for weighted vertex cover.