

CSC 373: Algorithm Design and Analysis

Lecture 20

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Announcements and Outline

Announcements

- No lecture this Friday
- Question: How was the test?

Today's outline

- Continue LP relaxation of an IP and rounding to an integral solution.
- Integrality gap for the basic IP/LP for the vertex cover problem.
- Set Cover
- A ring routing problem
- The makespan problem in the unrelated machines model

Review of the IP formulation for vertex cover and its LP relaxation

- We discuss weighted vertex cover as a $\{0, 1\}$ IP.
- We will discuss its LP relaxation and “naïve” and why this provides a 2-approximation.
- The **integrality gap** of this IP/LP relaxation is $2 - \frac{1}{n}$. Adding some additional inequalities (say corresponding to odd cycles) does not help to improve the integrality gap.
- In general there will be many IP formulations for a given problem. An integrality gap pertains to one (or a class of) IP/LP relaxations.
- Despite considerable effort, there is no known polynomial time algorithm that achieves a $2 - \epsilon$ approximation for any $\epsilon > 0$ even for the unweighted case.

An IP for weighted vertex cover

IP formula for weighted vertex cover

$$\begin{array}{ll} \text{minimize} & \sum_{j=1}^n w_j \cdot x_j \\ \text{s.t.} & x_i + x_j \geq 1 \quad \text{for each edge } (i, j) \in E \\ & x_i \in \{0, 1\} \quad \text{for each } i = 1, \dots, n \end{array}$$

- The **intended meaning** is that $x_i = 1$ iff vertex v_i is in the cover.
- The **LP relaxation** is to relax the integrality condition $x_i \in \{0, 1\}$ to $x_i \geq 0$.
- In this problem it follows that an optimal LP solution also satisfies $x_i \leq 1$.

Rounding an LP optimal solution

- Suppose x^* is an LP optimum.
- We can apply a “naïve” rounding (naive in the sense that the rounding ignores the input) to the fractional solution by setting

$$x'_i = 1 \text{ iff } x_i^* \geq 1/2$$

- **Claim:** x' is an integral solution to the IP and hence $V' = \{v_i \mid x'_i = 1\}$ is a vertex cover. **Why?**

Claim

The weight of the cover V' is **at most twice** the weight of an optimal cover.

Proof.

- Because the LP is a relaxation, it also allows IP solutions. Thus, $OPT_{LP} \leq OPT_{IP}$.
- Then we have $w(V') \leq 2 \cdot OPT_{LP} \leq 2 \cdot OPT_{IP}$.

The integrality gap

- For the complete (unweighted) graph on n nodes, the optimal IP value is $n - 1$, whereas the LP optimum value is $n/2$.
- For a given IP/LP formulation of a minimization problem, the **integrality gap** is the worst case (over all input instances) ratio of an integral optimum value to a fractional (LP) optimum value.
- The approximation analysis show that this ratio is at most 2 and the previous example shows that it is at least $(2 - \frac{1}{n})$.
- For the n node cycle, the optimum IP solution is $\lceil n/2 \rceil$ and the LP OPT is $n/2$. A naive rounding would be (at best) a 2-approximation.
- For any known polynomial time approach to add additional constraints, the integrality gap essentially remains at 2.

Informal but commonly used claim

The integrality gap provides a limit to obtaining an approximation using a particular IP/LP formulation of a problem.

Set cover

The set cover problem

- Given a collection of (possibly weighted) sets $\mathcal{C} = \{S_1, \dots, S_m\}$ where $S_i \subseteq U$.
- Goal:** Find a minimal size (weight) subcollection \mathcal{C}' that covers all the elements in the universe U .
- Set cover generalizes vertex cover and turns out to be NP-hard to approximate better than H_n where $n = |U|$. Recall that

$$H_n = \sum_{k=1}^n 1/k \approx \ln n$$

- There is a natural greedy algorithm that will achieve an approximation of H_d where $d = \max_i |S_i|$.
- Unless all problems in NP can be computed in time $n^{O(\log \log n)}$, it is not possible to approximate the set cover problem to within a factor $(1 - \epsilon) \ln n$ for any $\epsilon > 0$.

Set cover and the f -frequency set cover problem

- We can express the set cover problem as an IP.

IP formula for set cover

$$\begin{array}{ll} \text{minimize} & \sum_{j=1}^n w_j \cdot x_j \\ \text{s.t.} & \sum_{i: e \in S_i} x_i \geq 1 \quad \text{for each element } e \in U \\ & x_i \in \{0, 1\} \quad \text{for each } i = 1, \dots, n \end{array}$$

- In the f -frequency set cover problem, we assume that every element occurs in at most f sets.
- Vertex cover is a 2-frequency set cover problem. **Why?** What are the elements and what are the sets?

Rounding the LP to obtain an approximation algorithm

- For the f -frequency set cover problem, we can relax it to an LP in the same way that we represented and provided an approximation for the vertex cover problem.
- We solve the LP optimally to obtain a solution \mathbf{x}^* and then round by setting

$$x'_i = 1 \text{ iff } x_i^* \geq 1/f$$

- Similar to the vertex cover approximation, this yields an f -approximation algorithm for the f -frequency set cover problem.

Ring routing: another IP/LP with naive rounding

The call routing problem

- There is an n node bi-directional ring network $G = (V, E)$ upon which calls must be routed, where

$$V = \{0, 1, \dots, n-1\}$$

$$E = \{(i, i+1 \bmod n)\} \cup \{(i, i-1 \bmod n)\}$$

- Calls c_j are pairs (s_j, f_j) originating at node s_j and terminating at node f_j .
- Each call can be routed in a clockwise or counter-clockwise direction.
- The load L_e on any directed edge is the maximum number of calls routed on this directed edge.
- Goal:** Minimize $\max_{e \in E} L_e$.

We can achieve a **2-approximation** by an LP relaxation of a natural IP followed by a naive rounding just as in the vertex cover example.

The ring routing example continued

- To formulate this problem as an IP, for each call we will introduce variables x_j and y_j that indicate the direction of call c_j .
- (You can also use just one indicator variable to represent the direction but I think it might be easier to think in terms of two such variables.)

IP formula for ring routing

minimize L

subject to:

$$\begin{array}{ll} \textcircled{1} & \sum_{\substack{j : c_j \text{ routed clockwise} \\ \text{would use edge } (i, i+1) \bmod n}} x_j \leq L & \text{for each edge } (i, i+1) \bmod n \end{array}$$

$$\begin{array}{ll} \textcircled{2} & \sum_{\substack{j : c_j \text{ routed clockwise} \\ \text{would use edge } (i, i-1) \bmod n}} y_j \leq L & \text{for each edge } (i, i-1) \bmod n \end{array}$$

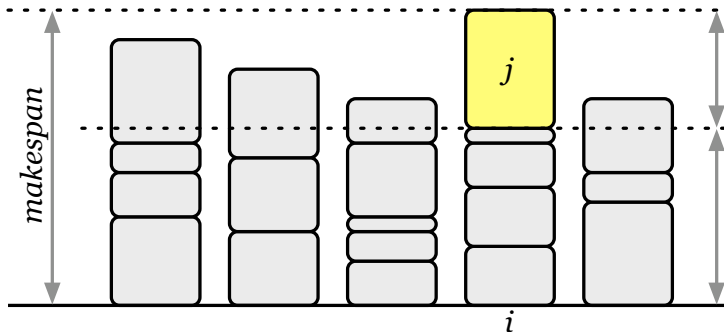
$$\textcircled{3} \quad x_j + y_j = 1 \quad \text{for each } i = 1, \dots, n$$

$$\textcircled{4} \quad x_j, y_j \in \{0, 1\} \quad \text{for each } i = 1, \dots, n$$

Recall: the makespan problem

The makespan problem for the identical machines model

- The input consists of n jobs $\mathcal{J} = \{J_1, \dots, J_n\}$ that are to be scheduled on m identical machines.
- Each job J_k is described by a processing time (or load) p_k .
- **Goal:** Minimize the latest finishing time (maximum load) over all machines.



Makespan for the unrelated machines model

The makespan problem for the unrelated machines model

- The input consists of n jobs $\mathcal{J} = \{J_1, \dots, J_n\}$ and m machines M_1, \dots, M_m .
 - Each job J_j is represented by a vector $\langle p_{1j}, p_{2j}, \dots, p_{mj} \rangle$ where p_{ij} represents the processing time of job J_j on machine i .
 - Without loss of generality, we assume $m \leq n$.
 - **Goal:** Minimize the latest finishing time (maximum load) over all machines.
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- We will sketch a 2-approximation IP/LP with non naïve rounding algorithm.
 - This is the best known poly-time approximation.
 - It is known that it is NP-hard to achieve better than $\frac{3}{2}$ -approximation even for the special case of the restrictive machines model for which every p_{ij} is either some p_j or ∞ .

- In the IP formulation, the problem is:

minimize t

$$\text{s.t. } \sum_{i=1}^m x_{ij} = 1 \quad \text{for each job } J_j$$

$$\sum_{j=1}^n p_{ij} x_{ij} \leq t \quad \text{for each machine } M_i$$

$$x_{ij} \in \{0, 1\} \quad \text{for } i = 1, \dots, m \text{ and } j = 1, \dots, n$$

- The intended meaning is that $x_{ij} = 1$ iff job J_j is scheduled on machine M_i .
- The LP relaxation is that $0 \leq x_{ij}$. The condition $x_{ij} \leq 1$ is implied.
- The integrality gap is unbounded! Why?. How do we get around this unbounded integrality gap?