M, 11/18/19

Fall'19 CSCE 629

Analysis of Algorithms

Fang Song Texas A&M U

Lecture 31

- Approximating vertex cover
- Linear programming relaxation

Coping with NP-Completeness



"I can't find an efficient algorithm, I guess I'm just too dumb."



"I can't find an efficient algorithm, but neither can all these famous people."

https://i.stack.imgur.com/EkpIV.jpg

- Better (constructive) answers: sacrifice one of three desired features
 - I. Solve arbitrary instances2. Solve problems in poly-time
 - 3. Solve problems to optimality

Techniques

- Identifying structured special cases
- Local search heuristics (e.g., gradient descent)
- Approximation algorithms

Input. Graph G = (V, E)

• Vertex cover $S \subseteq V$: subset of vertices that touches all edges

Goal. Find a vertex cover S of minimum size

First attempt: greedily pick the vertex that covers most edges

Finding near-optimal vertex cover

```
APP-VC: on input G = (V, E)
For v \in V (in descending order of degrees)
Add v in S
Delete v and its neighbors from G
```

- Claim. Suppose the minimum vertex cover has size OPT. Then the output of APP-VC has size at most $O(\log n \cdot OPT)$
- Pf. Exercise (Hint on board)

Recall: $M \subseteq E$ is a matching in G = (V, E) if each node appears in at most one edge in M.

2-approximation vertex cover

Observation: For any matching M and any vertex cover S, $|M| \le |S|$. In particular, $|M| \le OPT$ (size of min vertex cover).

Ind attempt: find a MAX matching

2-APP-VC: on input G = (V, E)Find a maximal matching $M \subseteq E$ **Return** $S = \{$ all end points of edges in $M \}$



Claim. The output of 2-APP-VC has size at most 2 · OPT

• Pf. $|S| = 2|M| \le 2 \cdot 0$ PT. Why does S have to be a vertex cover?

• Exercise. Is this tight, i.e., 2-APP-VC's output = $2 \cdot 0PT$ on some graph?

Input. Graph G = (V, E)

• Vertex cover $S \subseteq V$: subset of vertices that touches all edges

Goal. Find a vertex cover S of minimum size

Formulating vertex cover as an integral linear program

For each $i \in V$, introduce $x_i \in \{0,1\}$ Min $\sum_{i=1}^{n} x_i$ Subject to: $x_i + x_j \ge 1$ for each $(i,j) \in E$ [i.e., Pick *i* in vertex cover iff. $x_i = 1$]

Integer linear programming (ILP)

⁽³⁾ We don't know (expect) a poly-time algorithm (ILP)

• Without integrality (LP), we do know poly-time algorithms

$$\begin{split} \mathbf{L}\mathbf{P} \Pi \mathbf{Min} \sum_{i=1}^{n} x_i \\ \text{bject to:} \\ x_i + x_j \ge 1, \quad \forall (i,j) \in E \\ x_i \in \{0,1\}, \quad \forall i \in V \end{split} \\ \mathbf{x}_i \coloneqq [x_i^*] = \begin{cases} 1, & \text{if } x_i^* \ge \frac{1}{2} \\ 0, & \text{otherwise} \end{cases} \end{split}$$

(LP Σ) Min $\sum_{i=1}^{n} x_i$ Subject to: $x_i + x_j \ge 1$, $\forall (i, j) \in E$ $0 \le x_i \le 1$, $\forall i \in V$

Let x^* be an optimal soln. for LP Σ & optimal value OPT = $\sum_i x_i^*$

(Threshold) Rounding:

Sı

i. $\{x_i\}$ is a feasible integral solution: $\forall (i,j) \in E, x_i^* \ge \frac{1}{2}$ or $x_j^* \ge \frac{1}{2}$ or both

Putting aside the integral constraint

ii. $\sum_{i} x_i \le \sum_{i} 2 \cdot x_i^* = 2 \cdot \text{OPT} \le 2 \cdot \text{OPT}_{\text{Int}}$

[optimal value of ILP П, i.e. size of min vertex cover]



Input. Set U of n elements, $S_1, ..., S_m$ of subsets of U Goal. Find $I \subseteq \{1, ..., m\}$ of minimum size such that $\bigcup_{i \in I} S_i = U$

Approximating set cover

(ILP Π for Set cover)

For each $i \in \{1, ..., m\}$, introduce $x_i \in \{0, 1\}$ Min $\sum_{i=1}^{m} x_i$ Subject to: $\sum_{i:u \in S_i} x_i \ge 1$, $\forall u \in U$

(Set cover ILP II)

$$\begin{array}{l}
\text{Min } \sum_{i=1}^{m} x_{i} \\
\text{Subject to:} \\
\sum_{i:u \in S_{i}} x_{i} \geq 1, \quad \forall u \in U \\
x_{i} \in \{0,1\}, \quad \forall i \in \{1, \dots, m\}
\end{array}$$
(Set cover Σ)

$$\begin{array}{l}
\text{Min } \sum_{i=1}^{m} x_{i} \\
\text{Subject to:} \\
\sum_{i:u \in S_{i}} x_{i} \geq 1, \quad \forall u \in U \\
0 \leq x_{i} \leq 1, \forall i \in \{1, \dots, m\}
\end{array}$$

$$\begin{array}{l}
\text{Let } x^{*} \text{ be an optimal soln. for LP } \Sigma \\
\text{Subject LP } X_{i} = X_{i} \times X_{i} \\
\text{Subject to:} \\
\text{Subject to:} \\
\text{Min } \sum_{i=1}^{m} x_{i} \\
\text{Subject to:} \\
\text{Subject to:}$$

Threshold rounding: does it cover all elements?

• Ex. $u \in S_1, ..., S_{100}; x_1^*, ..., x_{100}^* = \frac{1}{100} \Rightarrow x_1 = \cdots = x_{100} = 0. u$ is missed!

LP relaxation for set cover

Randomized rounding! [Stay tuned]

 $\forall u \in U$

& optimal value OPT = $\sum_{i} x_{i}^{*}$

Hardness of approximation



Theorem. It is NP-Hard to approximate Vertex Cover to with any factor below 1.36067. [i.e., otherwise, you can solve 3-SAT in poly-time]

Theorem'. It is NP-Hard to approximate Vertex Cover to with any factor below 2, assuming the unique games conjecture (UGC).

Want to read more?

<u>https://cs.nyu.edu/~khot/papers/UGCSurvey.pdf</u> <u>https://cs.stanford.edu/people/trevisan/pubs/inapprox.pdf</u>