

CS7800: Advanced Algorithms

Class 17: Approximation Algorithms II

- Max Coverage
- Vertex and Set Cover

Jonathan Ullman

November 4, 2025

Maximum Coverage

Inputs: Sets $S_1, \dots, S_m \subseteq \overbrace{\{1, \dots, n\}}$
A budget $k \geq 0$

Outputs/Objective: Choose sets $\{A_1, \dots, A_k\} \subseteq \{S_1, \dots, S_m\}$
maximizing $\left| \bigcup_{i=1}^k A_i \right|$

covers as many
elements as possible

- Can solve in time $O(\binom{m}{k}) = O(m^k)$

- Problem is NP-hard to solve exactly

Why?

Greedy Max Coverage

For $i=1, \dots, k$:

- Let A_i be the set maximizing $|A_1 \cup A_2 \cup \dots \cup A_i|$



Equivalent to
maximizing $|A_i \setminus \underbrace{(A_1 \cup \dots \cup A_{i-1})}|$

Bad Example ($k=2$):

0	0	0	0	0	0
0	0	0	0	0	0
0	0	0	0	0	0
0	0	0	0	0	0
0	0	0	0	0	0
0	0	0	0	0	0
0	0	0	0	0	0
0	0	0	0	0	0

is $1 \times \frac{1}{2}$

is $1 \times \frac{1}{2}$

is $(\frac{1}{2} + \varepsilon) \times 1$

For $k=2$, greedy is
at best a $\frac{3}{4}$ -approx

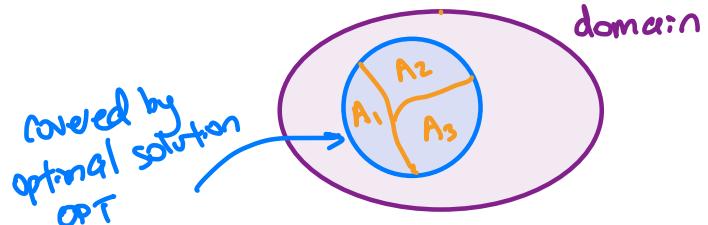
Greedy Max Coverage Analysis

Key Claim: Let c_i be the number of elts covered by the first i sets A_1, A_2, \dots, A_i . Then at iteration i there exists a set that covers at least $\frac{OPT - c_i}{k}$ new elements

$$\Rightarrow \text{for every } i, \quad c_i - c_{i-1} \geq \frac{OPT - c_{i-1}}{k}$$

Special Case: In the first iteration, greedy chooses A_1 such that

$$|A_1| \geq \frac{OPT}{k}$$



since A_1, \dots, A_k cover OPT elements, one must cover $\frac{OPT}{k}$ elements

Greedy Max Coverage Analysis

Key Claim: Let c_i be the number of elts covered by the first i sets $A_1 \cup \dots \cup A_i$. Then at iteration i there exists a set that covers at least $\frac{OPT - c_i}{k}$ new elements

$$\Rightarrow \text{for every } i, c_i - c_{i-1} > \frac{OPT - c_{i-1}}{k}$$

greedy picks sets A_1, \dots, A_k

$$c_i = |A_1 \cup A_2 \cup \dots \cup A_i|$$

$$c_i - c_{i-1} = |A_i \setminus (A_1 \cup \dots \cup A_{i-1})|$$

greedy chooses A_i to maximize this quantity

When choosing A_i , you will cover at least $\frac{1}{k}(OPT - \# \text{elts already covered})$

Greedy Max Coverage Analysis

Key Claim: Let c_i be the number of elts covered by the first i sets $A_1 \cup \dots \cup A_i$. Then at iteration i there exists a set that covers at least $\frac{OPT - c_i}{k}$ new elements

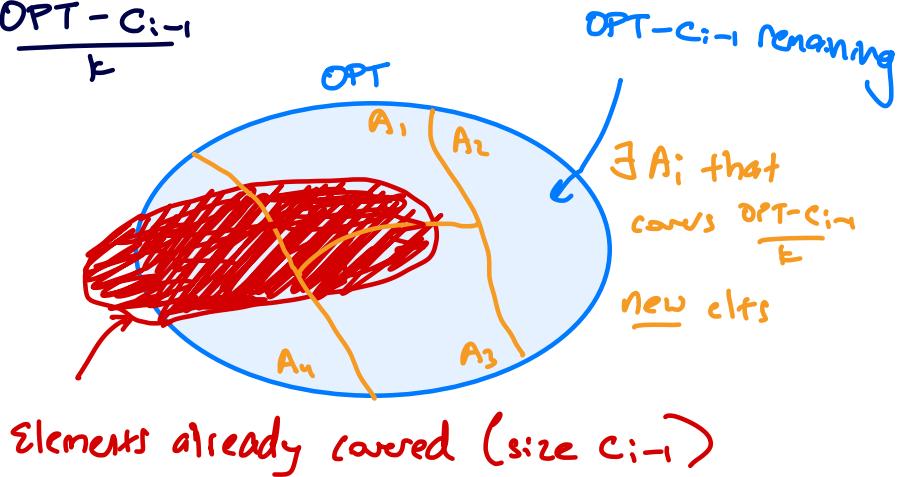
$$\Rightarrow \text{for every } i, \quad c_i - c_{i-1} > \frac{OPT - c_{i-1}}{k}$$

greedy picks sets A_1, \dots, A_k

$$c_i = |A_1 \cup A_2 \cup \dots \cup A_i|$$

$$c_i - c_{i-1} = |A_i \setminus (A_1 \cup \dots \cup A_{i-1})|$$

greedy chooses A_i to maximize this quantity



Greedy Max Coverage Analysis

$$\lim_{k \rightarrow \infty} \left(1 + \frac{c}{k}\right)^k = e^c$$

Key Clm: In iteration i , $c_i - c_{i-1} \geq \frac{OPT - c_{i-1}}{k}$

value of greedy

$$c_k = c_k - c_{k-1} + c_{k-1}$$

$$\geq \frac{OPT - c_{k-1}}{k} + c_{k-1} = \frac{OPT}{k} + \left(1 - \frac{1}{k}\right) \cdot c_{k-1}$$

$$\geq \frac{OPT}{k} + \left(1 - \frac{1}{k}\right) \left(\frac{OPT}{k} + \left(1 - \frac{1}{k}\right) \cdot c_{k-2} \right)$$

$$= \frac{OPT}{k} \left(1 + \left(1 - \frac{1}{k}\right) \right) + \left(1 - \frac{1}{k}\right)^2 \cdot c_{k-2}$$

$$= \frac{OPT}{k} \left(1 + \left(1 - \frac{1}{k}\right) + \left(1 - \frac{1}{k}\right)^2 + \dots + \left(1 - \frac{1}{k}\right)^{k-1} \right)$$

$$= \frac{OPT}{k} \cdot \frac{1 - \left(1 - \frac{1}{k}\right)^k}{1 - \left(1 - \frac{1}{k}\right)} = \frac{OPT}{k} \cdot \frac{1 - \left(1 - \frac{1}{k}\right)^k}{\frac{k}{k}} = \left(1 - \left(1 - \frac{1}{k}\right)^k\right) \cdot OPT \\ \approx \left(1 - \frac{1}{e}\right) \cdot OPT$$

Greedy Max Coverage

For $i = 1, \dots, k$:

- Let A_i be the set maximizing $|A_1 \cup A_2 \cup \dots \cup A_i|$

Equivalent to
maximizing $|A_i \setminus (A_1 \cup \dots \cup A_{i-1})|$

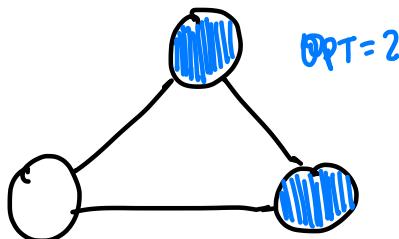
Thm: Greedy MC gives a $(1 - \frac{1}{e})$ -approximation in time $\underline{\underline{O(n^3)}}$

I didn't
think hard
about this

Vertex Cover

Input: Given an undirected, unweighted graph $G = (V, E)$

Output/Objective: Find a subset of nodes $S \subseteq V$ such that $\forall (u, v) \in E \quad u \in S \text{ or } v \in S$ and $|S|$ is as small as possible



NP-hard to solve exactly

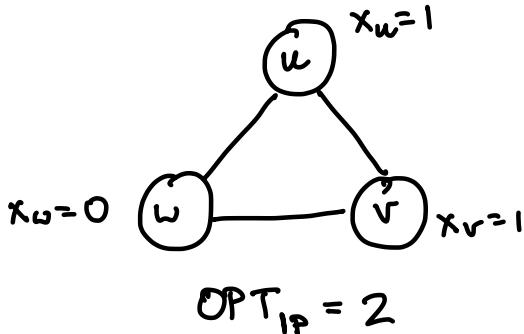
LP / ILP for Vertex Cover

ILP

$$\min \sum_{u \in V} x_u$$

$$x_u + x_v \geq 1 \text{ for } (u, v) \in E$$

$$x_u \in \{0, 1\} \text{ for } u \in V$$

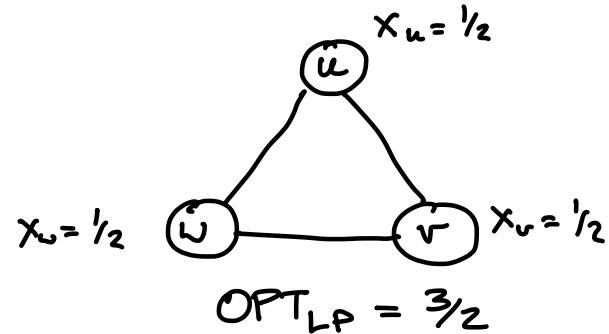


LP

$$\min \sum_{u \in V} x_u$$

$$x_u + x_v \geq 1 \text{ for } (u, v) \in E$$

$$0 \leq x_u \leq 1 \text{ for } u \in V$$



LP / ILP for Vertex Cover

ILP

$$\begin{aligned} \min & \sum_{u \in V} x_u \\ x_u + x_v & \geq 1 \text{ for } (u, v) \in E \\ x_u & \in \{0, 1\} \text{ for } u \in V \end{aligned}$$

LP

$$\begin{aligned} \min & \sum_{u \in V} x_u \\ x_u + x_v & \geq 1 \text{ for } (u, v) \in E \\ 0 \leq x_u \leq 1 & \text{ for } u \in V \end{aligned}$$

LP Rounding: - Let x_{LP}^* be the optimal LP solution

- Turn it into x_{IP} that is feasible for the ILP

and also $\text{OBJ}(x_{IP}) \leq c \cdot \text{OBJ}(x_{LP}^*)$

- $\text{OPT}_{IP} \geq \text{OPT}_{LP} = \text{OBJ}(x_{LP}^*) \geq \frac{1}{c} \cdot \text{OBJ}(x_{IP})$

LP / ILP for Vertex Cover

ILP

$$\min \sum_{u \in V} x_u$$

$$x_u + x_v \geq 1 \text{ for } (u, v) \in E$$

$$x_u \in \{0, 1\} \text{ for } u \in V$$

LP

$$\min \sum_{u \in V} x_u$$

$$x_u + x_v \geq 1 \text{ for } (u, v) \in E$$

$$0 \leq x_u \leq 1 \text{ for } u \in V$$

- Let x^* be the optimal LP solution

- Let x be as follows:

$$x_u = \begin{cases} 1 & \text{if } x_u^* > \frac{1}{2} \\ 0 & \text{if } x_u^* \leq \frac{1}{2} \end{cases}$$

① x is feasible for ILP

② $\sum_{u \in V} x_u \leq 2 \cdot \sum_{u \in V} x_u^*$

$$\begin{aligned} OPT_{LP} &\leq OPT_{ILP} = \text{OBJ}(x^*) \leq \frac{1}{2} \cdot \text{OBJ}(x) \\ \text{OBJ}(x) &\leq 2 \cdot OPT_{LP} \end{aligned}$$

LP Relaxation for Set Cover

Domain $\{1, \dots, m\}$

Sets $S_1, \dots, S_n \subseteq \{1, \dots, m\}$

ILP

$$\min_x \sum_{i=1}^n x_i$$

$$\sum_{i: j \in S_i} x_i \geq 1 \text{ for } j \in \{1, \dots, m\}$$

$$x_i \in \{0, 1\}$$

$$\text{If } x_1 + x_2 + \dots + x_t \geq 1$$

$$\text{Suppose } x_1 = x_2 = \dots = x_t = \frac{1}{t}$$

Generalization of Vertex Cover

Approx Alg

- ① Solve LP relaxation to get x^*
- ② For $i = 1, \dots, n$:
set $x_i = 1$ with probability x_i^*
- ③ Repeat r times and take the union of the covers

Claim: For every domain $i \in j$

$$\Pr(\exists i, j \in S_i \text{ and } x_i = 1) \geq 1 - \frac{1}{e}$$