

# CS7800: Advanced Algorithms

## Class 17: Approximation Algorithms II

- Max Coverage
- Vertex and Set Cover

Jonathan Ullman

November 4, 2025

# Maximum Coverage

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

Outputs/Objective: Choose sets  $\{A_1, \dots, A_k\} \subseteq \{S_1, \dots, S_m\}$   
maximizing  $\left| \underbrace{\bigcup_{i=1}^k A_i}_{\substack{\text{covers as many} \\ \text{elements as possible}}} \right|$

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

• Problem is NP-hard to solve exactly  $\leftarrow$  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  $|A_i \setminus (A_1 \cup \dots \cup A_{i-1})|$

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

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

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

$(\frac{1}{2} + \epsilon) \times 1$

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

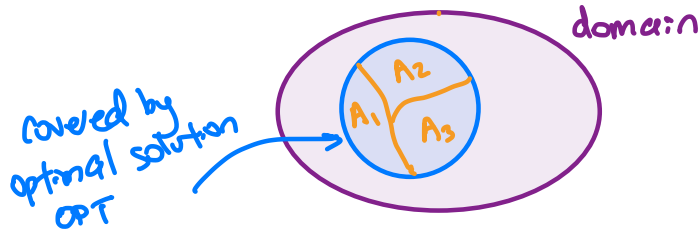
# Greedy Max Coverage Analysis

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

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

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

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



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

# Greedy Max Coverage Analysis

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

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

greedy picks sets  $A_1, \dots, A_k$

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

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

greedy chooses  $A_i$  to maximize this quantity

# Greedy Max Coverage Analysis

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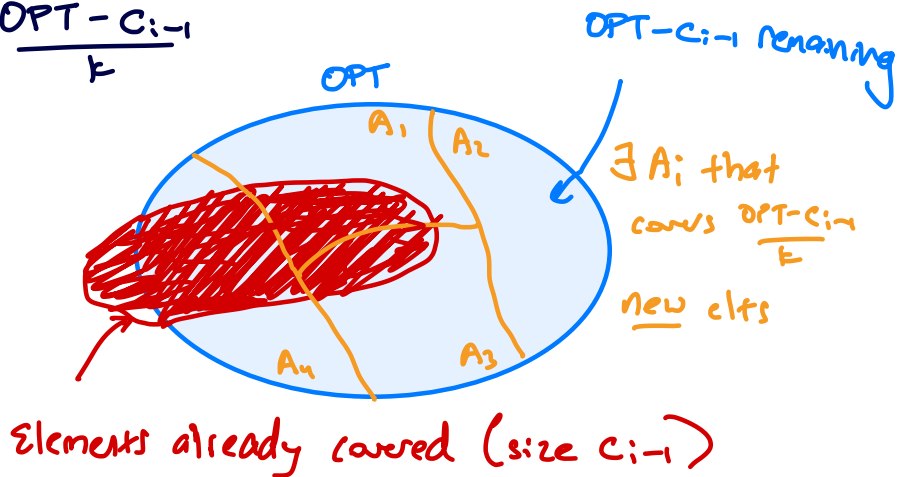
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# Greedy Max Coverage Analysis

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

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

value of greedy

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

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

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

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

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

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

# 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  $(1 - 1/e)$ -approximation in time  $O(n^3)$

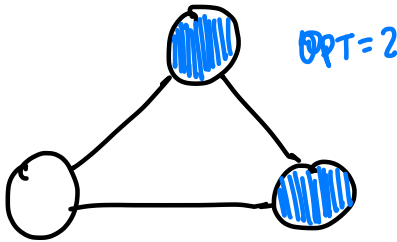
I didn't  
think hard  
about this



# Vertex Cover

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

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



NP-hard to solve exactly

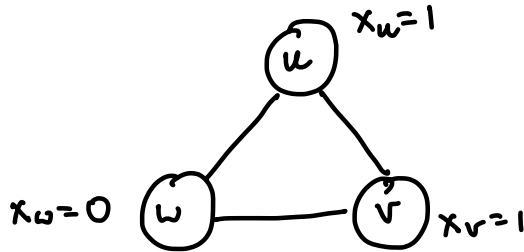
# LP / ILP for Vertex Cover

ILP

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

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

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



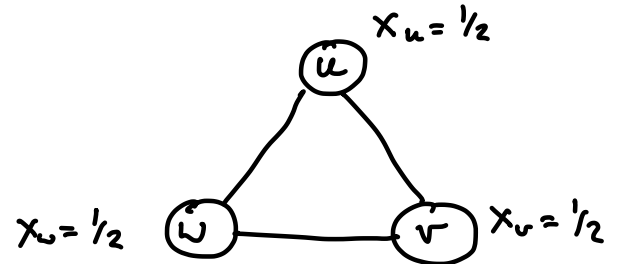
$$\text{OPT}_{\text{ILP}} = 2$$

LP

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

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

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



$$\text{OPT}_{\text{LP}} = 3/2$$

# LP / ILP for Vertex Cover

## ILP

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

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

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## LP

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

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

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

- Turn it into  $x_{IP}$  that is feasible for the ILP  
and also  $OBJ(x_{IP}) \leq c \cdot OBJ(x_{LP}^*)$

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

# LP / ILP for Vertex Cover

ILP

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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: 
$$x_u = \begin{cases} 1 & \text{if } x_u^* \geq 1/2 \\ 0 & \text{if } x_u^* < 1/2 \end{cases}$$

①  $x$  is feasible for ILP

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

$$\text{OPT}_{\text{ILP}} \leq \text{OPT}_{\text{LP}} = \text{OBJ}(x^*) \leq \frac{1}{2} \cdot \text{OBJ}(x)$$

$$\leadsto \text{OBJ}(x) \leq 2 \cdot \text{OPT}_{\text{ILP}}$$

# LP Relaxation for Set Cover

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

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

ILP

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

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

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

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

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

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 elt  $j$

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