

Lecture 2

Wednesday, March 22, 2017 1:18 PM

Problems

① Set Cover

Input: • Universe of n elements $\{e_1, \dots, e_n\} = U$ } set family
• Sets $S_1, S_2, \dots, S_m \subseteq U$
• Costs $c(S_1), c(S_2), \dots, c(S_m) \geq 0$

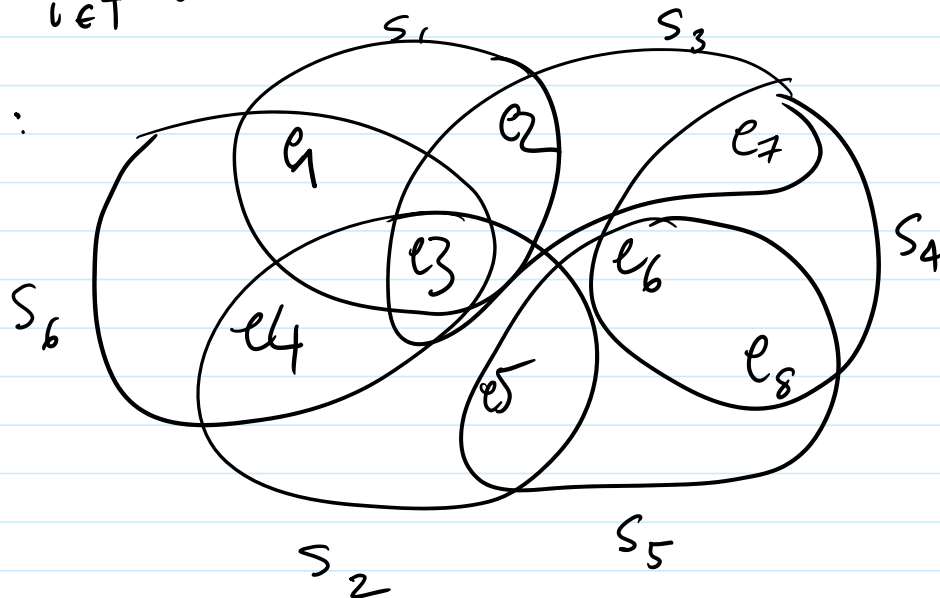
Output: - A **set cover**.

- Subcollection of the input sets whose union cover all the elements.

- $T \subseteq \{1, 2, \dots, m\}$ is a set-cover if

$$\bigcup_{i \in T} S_i = U = \{1, \dots, n\}$$

Example:



$\{1, 2, 4\}$ is a set cover

S_0 is $\{3, 5, 6\}$.

Objective: Output a set cover with

Objective: Output a set cover with minimum total cost, i.e.,
return $T \subseteq \{1, \dots, m\}$ minimizing $\sum_{i \in T} c(S_i)$

② Max k-Coverage

Input: • Set-family (Universe, Sets)

Output: • $\leq k$ -sets indexed by $T \subseteq \{1, \dots, m\}$
 $|T| \leq k$

Goal: Maximize $\left| \bigcup_{i \in T} S_i \right|$

For this problem, no costs on sets.

Algorithm for Set Cover

- $A = \emptyset$ // index of sets covered by algo.
- $C = \emptyset$ // C is the set of covered elts initialized to empty set

- At any pt, $X \stackrel{(\text{def})}{=} U \setminus C$
// X : set of uncovered elts

- While $C = U$:

- Pick set i which **minimizes**

"cost per new"
elt

$$\left\{ \frac{c(S_i) \leftarrow \text{cost of set}}{|S_i \cap X| \leftarrow \text{\# of new elts } S_i \text{ covers}} \right.$$

- $A = A \cup i$

- $C = C \cup S_i, X = X \setminus S_i$

Analysis

- Let's rename the sets s.t. $\{S_1, S_2, \dots, S_r\}$ are the sets picked by the algorithm **in that order**.

- Moreover, let C_t be the set of covered elts after set S_t has been picked.

Note: $C_0 = \emptyset, C_1 = S_1, C_2 = S_1 \cup S_2, \dots$

Similarly, we define $X_t = U \setminus C_t$.

- Finally, let the optimum set cover be indexed by $0 \leq \{1, \dots, m\}$.

Fix a "time" instant t just before S_t is picked. X_{t-1} is the set of uncovered elts.

$$\frac{c(S_t)}{|S_t \cap X_{t-1}|} \leq \frac{c(S_j)}{|S_j \cap X_{t-1}|} \quad \text{for any other set } S_j$$

In particular

$$\text{LHS} \leq \frac{c(S_j)}{|S_j \cap X_{t-1}|} \quad \forall j \in O$$

Fact: for any $a, b, c, d \geq 0$, we have

$$\min\left\{\frac{a}{b}, \frac{c}{d}\right\} \leq \frac{a+c}{b+d} \leq \max\left\{\frac{a}{b}, \frac{c}{d}\right\}$$

$$\therefore \frac{c(S_t)}{|S_t \cap X_{t-1}|} \leq \frac{\sum_{j \in O} c(S_j)}{\sum_{j \in O} |S_j \cap X_{t-1}|} \quad \left\{ = \text{opt} \right\}$$

$\left\{ \approx |X_{t-1}| \right\}$
 $\because 0$ is a set cover

$$\frac{c(S_t)}{|X_{t-1}| - |X_t|} \leq \frac{\text{OPT}}{|X_{t-1}|}$$

$$\Rightarrow c(S_t) \leq \text{OPT} \cdot \left(\frac{|X_{t-1}| - |X_t|}{|X_{t-1}|} \right)$$

$$\Rightarrow c(S_t) \leq \text{OPT} \cdot \left(\frac{1 - t^{-1}}{|X_{t-1}|} \right)$$

for all t

$$\therefore \text{ALG} = \sum_{t=1}^r c(S_t)$$

$$\leq \text{OPT} \left(\frac{|X_0| - |X_1|}{|X_0|} + \frac{|X_1| - |X_2|}{|X_1|} + \dots \dots + \frac{|X_{r-1}| - |X_r|}{|X_{r-1}|} \right)$$

$= n$

$= 0$

$$\leq \text{OPT} \left(\frac{1}{n} + \frac{1}{n-1} + \dots + 1 \right)$$

is called $H(n)$ or simply H_n and is $\approx \ln n$

Theorem: The algorithm described above is a H_n -factor approximation algorithm for the Set-Cover Problem.

A Better Analysis (Charging Argument)

- When we pick set S_t , for every element $e_j \in S_t \cap X_{t-1}$, ie, all the new elts that S_t covers, we put a "charge"
$$\alpha_j := \frac{c(S_t)}{|S_t \cap X_{t-1}|} \text{ on } e_j$$

At the end, every element gets a charge
$$\sum_{j=1}^n \alpha_j = \text{ALG}.$$

- Now, fix a set S_i in the optimal set cover, ie, $i \in \mathcal{O}$.

Order the elements of S_i in the order they are covered by the algorithm, ie, in the order these elements obtain

"charges".

- Say $|S_i| = k$, and its elements are ordered to be $\{e_1, \dots, e_k\}$

Fix elt e_j . and consider the time it gets its charge. Just before this

time $\{e_j, e_{j+1}, \dots, e_k\}$ are uncovered

by defn. Since S_i was a choice avail. to

the algo,

$$\alpha_j \leq \frac{c(S_i)}{k-j+1}$$

- $ALG = \sum_{j=1}^n \alpha_j \leq \sum_{i \in O} \sum_{j \in S_i} \alpha_j$

\uparrow
 $\because 0$ is
a valid
set cover

$$= \sum_{i \in O} \sum_{j=1}^{|S_i|} \frac{c(S_i)}{|S_i| - j + 1}$$
$$= \sum_{i \in O} c(S_i) \cdot H(|S_i|)$$

$$\leq \text{OPT} \cdot H_{\Delta}$$

where $\Delta \equiv \max_{i=1}^m |S_i|$

note, $\Delta \leq n$, and can be much smaller.

Theorem: The algorithm described above is a H_{Δ} -approximation algorithm.

Algorithm for Max k-coverage

For $t=1, 2, \dots, k$

- Pick S_t which contains the most number of uncovered elements.

Analysis

- Rename the sets s.t. S_1, \dots, S_k are the sets picked by the algorithm.
- Let $A_i := \bigcup_{j \leq i} S_j$ be the elements covered by the first i sets.
- $\text{ALG} = |A_k|$, the number of elts covered

by our algorithm.

- Let $OPT = |A^*|$ be obtained by the sets indexed by O ; $|O| = k$

What do we know?

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

since one of the set S_i , $i \in O$ must be $|S_i| \geq \frac{OPT}{k}$.

$$\begin{aligned} |A_2 \setminus A_1| &\geq \frac{|A^* \setminus A_1|}{k} && \text{(same reason)} \\ \parallel \\ |A_2| - |A_1| &\geq \frac{OPT - |A_1|}{k} \end{aligned}$$

$$\therefore |A_2| \geq \frac{OPT}{k} + \left(1 - \frac{1}{k}\right) \cdot |A_1|$$

More generally, for $i \geq 1$

$$|A_i| \geq \frac{OPT}{k} + \left(1 - \frac{1}{k}\right) |A_{i-1}|$$

$$\begin{aligned}
&\geq \frac{\text{OPT}}{k} + \left(1 - \frac{1}{k}\right) \frac{\text{OPT}}{k} \\
&\quad + \left(1 - \frac{1}{k}\right)^2 |A_{i-2}| \\
&\quad \vdots \\
&\geq \frac{\text{OPT}}{k} \left(1 + \left(1 - \frac{1}{k}\right) + \dots + \left(1 - \frac{1}{k}\right)^{i-1}\right) \\
&= \text{OPT} \cdot \left(1 - \left(1 - \frac{1}{k}\right)^i\right)
\end{aligned}$$

$$\begin{aligned}
\Rightarrow \text{ALG} &\geq \text{OPT} \cdot \left(1 - \left(1 - \frac{1}{k}\right)^k\right) \\
&\geq \text{OPT} \left(1 - \frac{1}{e}\right)
\end{aligned}$$

Fact: $\left(1 - \frac{1}{k}\right)^k \leq \frac{1}{e}$

More generally, $e^x \geq 1 + x$ for all x

and the above is obtained by setting $x = -\frac{1}{k}$.

plot & see, & then prove.

Theorem: The Greedy algorithm is a

Theorem: The Greedy algorithm is a $(1 - \frac{1}{e})$ -factor appx. algorithm for the Max k-Coverage problem.