

Approximation Algorithms for Set Cover

→ Greedy algorithm & analysis

$$U = \{e_1, e_2, \dots, e_n\}$$

$$\mathcal{S} = \{S_1, S_2, \dots, S_m\}$$

Goal: pick min # subsets from \mathcal{S} to cover U .

Greedy Algo for Set Cover

Initially $Y = U$

While $Y \neq \emptyset$

choose $S \in \mathcal{S}$ with maximum $|S \cap Y|$
 set $Y \leftarrow Y \setminus S$

Theorem [Johnson 1971]: Greedy is $(1 + \ln n)$ -approximation for set cover.

Independent of # of sets

$$n_0 = n$$

$$n_1 = \# \text{ uncovered elements after 1st set}$$

$$n_2 = \# \text{ " " " 2nd set}$$

$$n_3 = \# \text{ " " " 3rd set}$$

$$n_i = \text{" " " } i^{\text{th}} \text{ set}$$

Let $k = |\mathcal{OPT}|$, i.e. \exists k sets to cover all elements.

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1st set we pick covers $\geq \frac{n}{k}$ elements.

$S_1^x, S_2^x, \dots, S_k^x \subseteq S$ cover V together
at least $\geq \frac{n}{k}$ elements from V

$$\Rightarrow n_1 \leq n \left(1 - \frac{1}{k}\right)$$

Using same argument @ time 2, greedy set covers $\geq \frac{n_1}{k}$ elements from uncovered elems.

$$\Rightarrow n_2 \leq n_1 \left(1 - \frac{1}{k}\right) \leq n \left(1 - \frac{1}{k}\right)^2$$

After 't' steps,

$$n_t \leq n \left(1 - \frac{1}{k}\right)^t$$

At what value of t will $n_t < 1$

$$n \left(1 - \frac{1}{k}\right)^t < 1$$

$$\left(1 - \frac{1}{k}\right)^t < \frac{1}{n}$$

$$t = k \ln n + 1$$

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

not = k

}

n... p.k. = ln n + 1

$$\left. \begin{array}{l} \text{Opt} = k \\ \text{Us} = k \ln n + 1 \end{array} \right\}$$

$$\text{Apx Ratio} \approx \ln n + 1$$

To summarize

- LP based f -approximation \leftarrow deterministic
- LP based $O(\ln n)$ -approximation \leftarrow randomized
- Greedy $\ln n$ -approx (deterministic)

Q: Is there a 2-approx for set cover?

A: No! In fact, any polytime $\ln n (1-\epsilon)$ approx $\Rightarrow P = NP$ \leftarrow Hardness of Approx.

Related question :-

Max k Coverage

{ Given (U, \mathcal{L}) & a parameter 'k',
choose k sets to maximize # elements covered }

Can implement same greedy algorithm
just stop after k rounds



Proof (can use same ideas) to show that we will cover at least $\text{OPT} \left(1 - \left(1 - \frac{1}{k}\right)^k \right)$
 $\approx \text{OPT} \left(1 - \frac{1}{e} \right)$

(0.63 approx for Max k Coverage)

$$\approx 0.63 \cdot \text{OPT}$$

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$\Rightarrow P = NP$

Hardness of

Max k coverage

Theorem Any $(0.63 - \epsilon)$ -approx $\Rightarrow P = NP$

Hardness of approx!!

Detail

Submodular functions

$$f: 2^S \rightarrow \mathbb{R}$$

eg

$$f(\{s_1\}) = 3$$

$$f(\{s_1, s_2\}) = 4$$

$$f(\{s_1, s_5\}) = 5$$

\approx

$$f(\{s_1, s_2\}) \leq f(\{s_1, s_5\})$$

\hookrightarrow monotone function

$$U = \{1, 2, 3, 4, 5\}$$

$$\mathcal{I} = \{s_1 = \{1, 2, 3\},$$

$$s_2 = \{1, 4\}$$

$$s_3 = \{1, 3, 5\}$$

$$s_4 = \{4\}$$

$$s_5 = \{4, 5\}\}$$

$$f(X) \geq f(Y) \text{ if } X \supseteq Y$$

Diminishing returns property

$$X \subseteq 2^S$$

$$Y \subseteq 2^S$$

$$X \subseteq Y$$

$$s_i \notin X \text{ \& } s_i \notin Y$$

$$f(X \cup s_i) - f(X) \geq f(Y \cup s_i) - f(Y)$$

Any function over a domain 2^D which satisfies diminishing returns property is called a Submodular function

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Game theory, ML, economics, Computer vision, etc.

If f satisfies both (P_1) & (P_2) then it's called MONOTONE SUBMODULAR fn.

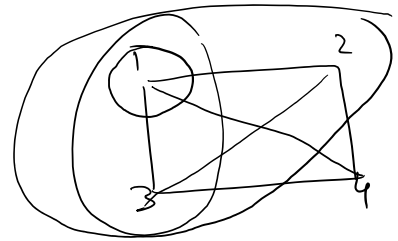
Coverage fn \rightarrow Monotone Δ Submodular

$$G = (V, E)$$

$$f: 2^V \rightarrow \mathbb{N}$$

$$f(S) = \# \text{ edges crossing } S$$

HW: show that f is submodular !!



$$f(\{1\}) = 3$$

$$f(\{1, 3\}) = 4$$

$$f(\{1, 2, 3\}) = 3$$

$$f(\{1, 2, 3, 4\}) = 0$$

For monotone submodular fn,

Greedy Algo = $(1 - \frac{1}{e})$ - approximation

Given Monotone Submodular fn f & a target k , pick set X of $|X| = k$ to maximize $f(X)$

\downarrow
also the best-possible algorithm !!

MIN SUM SET COVER PROBLEM

given U, \mathcal{S}

Find an ordering of sets to minimize

$$\sum_{e \in U} \text{cover}(e)$$

$$U = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10\}$$

$$\mathcal{S} = \left\{ \{1, 2, 3, 4\}, \{5, 6, 7, 8, 9, 10\}, \{1, 2, 3, 4, 5, 6, 7, 8\}, \{9\}, \{10\} \right\}$$

$$\text{Set Cover} = \{S_1, S_2\} \quad \text{size } 2$$

$$O = S_1, S_2, S_3, S_4, S_5$$

$$\text{Acc / O, Cover}(i) = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}$$

$$\begin{matrix} \text{Cover}(5) = 2 \\ \text{Cover}(6) = 2 \\ \vdots \\ 10 = 2 \end{matrix}$$

$$\text{Obj Value} = 4 \times 1 + 6 \times 2 = 16$$

$$O = S_2, S_1, S_3, S_4, S_5$$

$$6 \times 1 + 4 \times 2 = 14$$

$$S_3, S_4, S_5, S_1, S_2$$

$$= 8 \times 1 + 2 + 3 = 13$$

$$S_3, S_2, S_1, S_4, S_5$$

$$8 \times 1 + 2 \times 2 = 12$$

Turns out,

greedy is very good for MSSC

↳ 4-approximation !!

Recall

Find an ordering to minimize

$$\max_e \text{cover}(e)$$

} = Set cover

Any $(4-\epsilon)$ -approx in polytime $\Rightarrow P = NP$

Proof

(will show $O(1)$ -approx, 4 is much more clever proof)

$A_1 = \# \text{ uncovered elts after time 1}$

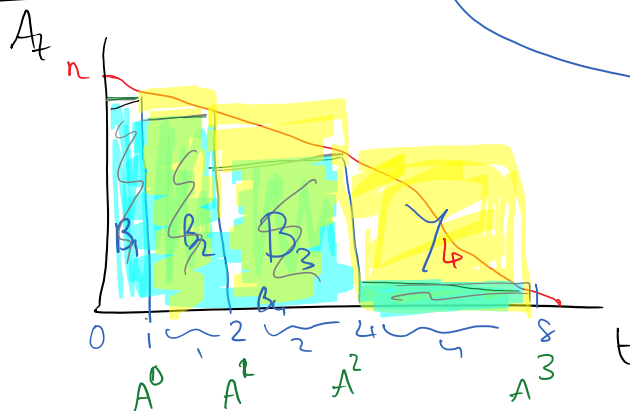
$A_2 = \dots \dots \dots \text{time } 2$

$A_t = \dots \dots \dots \text{time } t$

$O_t = \# \text{ uncovered elts in opt after time } t$

Claim:

$$\sum_e \text{cover}(e) = \sum_t A_t \quad (\text{for algo soln}) \quad \left| \quad \sum_e \text{cover}^*(e) = \sum_t O_t$$



$\sum A_t = \text{area under this curve.}$

$$\frac{1}{2} \sum_{t \geq 1} A_t \leq \sum_{j \geq 0} \frac{A^j (2^j - 2^{j-1})}{2} \leq \sum_{t \geq 1} A_t$$

$2B_3 \geq \frac{1}{4} \geq \text{Area under red curve}$

ans: 14

$$\sum_{j \geq 0} A^j \cdot 2^j$$

($A^j = \# \text{ uncovered pts @ time } 2^j$)

Goal

$$\sum_{j \geq 0} A^j \cdot 2^j \leq O(1) \sum_{j \geq 0} O^j \cdot 2^j$$

One attempt

$$\text{s.t. } \forall j, A^j \leq O(1) \cdot O^j \quad ??$$

We'll show

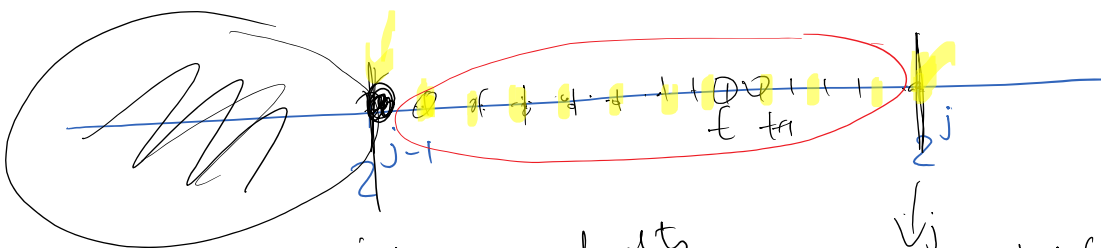
$$\left. \begin{array}{l} \text{either } A^j \leq 10 \cdot O^j \\ \text{or } A^j \leq \frac{1}{3} A^{j-1} \end{array} \right\} \text{ (or)}$$

$$A^j \leq \frac{1}{3} A^{j-1} + 10 \cdot O^j$$

$$\sum_j 2^j A^j \leq \frac{2}{3} \sum_j 2^{j-1} A^{j-1} + 10 \sum_j 2^j O^j$$

$$Alg \leq \frac{2}{3} Alg + 10 \cdot OPT$$

$$\Rightarrow Alg \leq 30 \cdot OPT$$



$$X = \frac{9}{10} A^j$$

etc

$2^{j-1} = \text{Uncovered pts in OPT @}$

$2^j = \text{uncovered pts @ } 2^j$

V_j^{j-1} Uncovered els
in OPT @ 2^{j-1}

V_j^j = uncovered els
@ 2^j

either OPT soln covers $\geq \frac{9}{10}$ of els of A^j by 2^{j-1} (OK)
it leaves $\geq \frac{1}{10}$ of A^j uncovered

in case ②

$$A^j \leq 10 V_j^{j-1}$$

$$\frac{x}{2^{j-1}} \text{ coverage}$$

Case ①

opt in time 2^{j-1} has covered $\geq \frac{9}{10}$ of my uncovered els @ time 2^j

Main Claim

$$\forall 2^{j-1} \leq t \leq 2^j,$$

$$A_{t-1} - A_t \geq \frac{\frac{9}{10} \cdot A^j}{2^{j-1}}$$

Greedy rule

Focus on $X = V_j^{j-1} \cap A^j$

= set of els opt covers in 2^{j-1} time which Alg still has uncovered at 2^j time

at time $t = 2^j$

$$\exists \text{ set in OPT } \geq \frac{|X|}{2^{j-1}} \text{ els.} \geq \frac{\frac{9}{10} \cdot A^j}{2^{j-1}}$$

$$A^{j-1} - A^j \geq \frac{9}{10} \cdot A^j$$

$$\frac{19}{10} A^j \leq A^{j+1}$$

$$\Rightarrow A^j \leq \left(\frac{10}{19}\right) A^{j+1}$$

different parameters will give me
 $A^j \leq \frac{1}{3} A^{j+1}$, completing the proof

Recap

Greedy \rightarrow $1 - \frac{1}{e}$ for Max k Coverage
 \rightarrow it extends to Submodular Max
 \rightarrow O(1) approx for MSSC problem

Primal Dual Technique

\downarrow
 algs are combinatorial
 analysis will use LP extensively

rakri.github.io \rightarrow
 www.cs.cmu.edu/~ravishan

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