

15854(B): (Advanced) Approx Algos

(1)

- HWS. • 10 weeks. Xinyu / Praveen / Anupam.
- Piazza / Google / etc.

Optimization Problems. ~~Search vs. Decision~~ vs. Decision Problems.

- ~~Max~~ 3SAT. vs. Max 3SAT, min 3-unsat etc.
- 2SAT (in P) vs. Max 2SAT, min 2-unsat -
- 3 color (NP-comp) vs. Min-coloring, Max 3 cut etc.
- 2 color (in P) vs. Max 2 cut, Min 2 uncut.

~~landscape~~ is Approx Problem

- Input: instance I .
- ~~valid~~ valid solution $S \in \text{sols}(I)$.

assume: checking \uparrow is in P. Also that there exists a soln

notion of value for each solⁿ S . $\text{val}(S) : \text{sols}(I) \mapsto \mathbb{R}_{\geq 0}$.

- minimization or max.

Want: algo: maps instances to solutions st. $\text{val}(S(I))$ is as high/low as possible.

Example: Max 3SAT. Instances are 3CNF formulas. (m clauses, n vars)

solutions: any truth assignment, assignment of T/F to each var.

value: $\frac{\# \text{ satisfied clauses}}{\text{total \# clauses}} \in [0, 1]$.

Algo: assign random T/F assignment.

Produces solution that satisfies $\frac{7}{8}$ clauses in expectation

Approx ratio = ~~max~~ worst case value $\frac{\text{Alg}(I)}{\text{OPT}(I)}$ Can we do better?

Finer grained Landscape than just NP-hardness

- ① Min-TSP (\mathbb{R}^2) is NP hard. but has PTAS. — later
- ② Min-Set cover is NP-hard but has $O(\lg n)$ apx. — soon
- ③ Max 3SAT — but $\nexists 7/8$ apx is best possible
- ④ Max Clique — but $n^{1-\epsilon}$ -apx $\Rightarrow P=NP$.
for any $\epsilon > 0$

So goal of course: apx algos (show positive results)
 hardness of apx (negative results)
 or integrality gaps / algorithmic gaps (limitations of our algos).

Sometimes talk about c -vs- s search problems. $s \leq c$

[if $OPT(I) \geq c$ produce soln of value $\geq s$]

or c -vs- s decision problems.

[if $OPT(I) \geq c$, answer YES, else if $OPT(I) < s$, answer NO]
 (anything in the middle is OK)

Fact: Algo for c -vs- s SEARCH \Rightarrow c -vs- s DECISION

Pf: ~~run algo, if $ALG(I) \geq s$, then output YES, else if $ALG(I) < s$, then output NO.~~
 ~~$OPT(I) \geq c$ then $ALG(I) \geq s$ (so output YES if $ALG(I) \geq s$)~~
~~if $ALG(I) < s \Rightarrow OPT(I) < c$ too so answer NO is OK.~~

Run algo. if $ALG(I) \geq s$ say YES, else say NO.



$OPT(I) \geq c \Rightarrow$ YES is fine



$OPT(I) < c \Rightarrow$ NO is fine.



min Set Cover: General problem, captures many others. (HW?)

Input: Set system $\mathcal{F} = (U, \{S_1, S_2, \dots, S_m\})$ $S_i \subseteq U$.

Assume $\bigcup_{i=1}^m S_i = U$. $|U| = n$.

sets have cost $c_1, \dots, c_m \geq 0$ (sometimes unit cost is interesting as well).

Solution: sub collection $I \subseteq [m]$

st $\bigcup_{i \in I} S_i = U$.

Value: ~~cost~~ cost / cardinality.

minimization

↑
wtd S.C.

↑
unweighted S.C.

$$\text{cost}(I) = \sum_{i \in I} c_i$$

$$|I|$$

Also Greedy: (unweighted)

[while not all elements covered
 ↳ Pick set S_i that covers most uncovered elements.]

Thm: Greedy is a $(\ln n)$ -apx algorithm for unweighted set cover.

Pf: sps $\text{OPT}(I) = K$. then sps $n_t = \# \text{uncovered elts after } t \text{ sets}$.
 $n_0 = n$.

fact: \exists a set that covers $\geq \frac{1}{K}$ frac of current uncovered elts.

Pf: OPT covers the n_t elts. so \exists set in OPT that covers $\geq \frac{n_t}{K}$

$\Rightarrow n_{t+1} \leq n_t (1 - \frac{1}{K})$. $\Rightarrow n_T \leq n_0 (1 - \frac{1}{K})^T < n_0 e^{-T/K} = 1$ elts. \boxtimes
if $T = K \ln n$. \boxtimes

Fact: Greedy solves the 1-vs-(1/e) for Max-k-coverage.

[Oh: Max-k-coverage. Solution is = k sets in set system
Val = #elements covered by picked sets.]

Pf: After k picked, $n_e \leq n_0 (1 - \frac{1}{k})^k < n_0 e^{-1}$

\Rightarrow coverage fraction $\geq \frac{n(1-1/e)}{n} = 1-1/e$ \square

Fact: if \exists algo for Max-k-coverage that is 1-vs-(1- δ) search algo.

$\Rightarrow \exists (\log_{1/\delta})^{n+1}$ apx algo for min Set Cover.

Pf: "Guess" DPT = k.

Use algo A. Covers 1- δ . repeat on remainder.
min k

\Rightarrow # uncovered after T rounds = $\delta^T \cdot n < 1$
if $T = \log_{1/\delta} n + 1$.

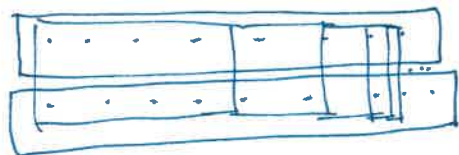
Corollary: 1 vs (1-1/e) for Max k Cov \Rightarrow ln n for min Set Cover

~~if~~: if $\ln n(1-\epsilon)$ for SC hard \Rightarrow $(1-1/e+\delta)$ for Max k Cov is hard.
a little more complicated, maybe see. see this in a future lecture.

Algorithmic Gap: Does greedy do better than $\ln n$?

Fact: Greedy no better than $\ln n (1-\epsilon)$. (even for unweighted)

Pf:



$OPT = 2$.

$Alg = \log_2(n/2)$

$\Rightarrow \text{gap} = \frac{\log_2 n - 1}{2}$

$\approx \frac{\ln n}{2 \ln 2}$

to get $\epsilon \ln n$, use sets whose $OPT = k$ vertically

But each other set covers $1/k$ of remainder.

$\Rightarrow \#sets = \log_{(1-1/k)} n \Rightarrow \text{gap} = \frac{\log_{(1-1/k)} n}{k} \approx \frac{\ln n}{k \ln(1-1/k)}$

but $\ln(1+\epsilon) = \epsilon - \Theta(\epsilon^2)$
for ϵ small.

$\approx (1 - \Theta(1/k)) \cdot \ln n$

So another algorithm?

Before that: are greedy algo for weighted case.

at each step, pick set that $\max \left(\frac{\text{coverage}}{\text{cost}} \right)$.

Thm: Greedy is $\Theta(\ln n)$ -apx for weighted set cover.

Pf: (sketch) Same idea as before. Show that if costs c_1, c_2, \dots, c_t

$n_t \leq n \left(1 - \frac{c_1}{OPT}\right) \left(1 - \frac{c_2}{OPT}\right) \dots \left(1 - \frac{c_t}{OPT}\right)$

for sets in steps
1, 2, ..., t

$\leq n \exp\left(-\frac{\sum c_i}{OPT}\right)$

etc



Linear Program - based Algos:

Idea: Relax-and-Round

- ① write an IP for Set Cover. (IP = Integer (Linear) Program).
- ② "Relax" it to an LP (LP = Linear Program).
- ③ solve this LP. Fact: can solve LPs in poly time.
- ④ "Round" the fractional solution to Integers.

Usually: ① $IP(I) = Opt(I)$.

② $LP(I) \leq IP(I)$.

③ $Alg(I) \leq \alpha \cdot LP(I) \Rightarrow Alg(I) \leq \alpha \cdot OPT(I)$.

Set Cover: variable $x_S \in \{0, 1\}$ for each set $S \in \{S_1, S_2, \dots, S_m\}$.

IP.

$$\begin{aligned} \min \quad & \sum_S c_S x_S \\ \text{st} \quad & \sum_{S: e \in S} x_S \geq 1 \quad \forall e \in U. \\ & x_S \in \{0, 1\}. \end{aligned}$$

LP

$x_S \geq 0$

Round: Imagine each x_S as a prob. value. (Fact: $x_S \in [0, 1]$, no reason for x_S to be larger).

Algo: [For $T = \underline{\hspace{2cm}}$ times
 $\forall S \in \mathcal{F}$
 select S independently w.p x_S .]

T rounds of sampling

What if this is not a feasible solution?

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Clean-up: \forall element e , pick cheapest set covering e if e not covered by sampling.

Lemma: $E[\text{cost of solution}] \leq T \cdot LP(I) + \left[\sum_{e \in E} (\text{cheapest set covering } e) \right] e^{-T}$.

Pf: $E[\text{cost of each round}] = \sum_s c_s \cdot \Pr[S \text{ picked}] = \sum_s c_s x_s = LP(I)$.
 (Linear exp.)

Now: $\text{prob}(e \text{ not covered in one round}) = \prod_{s: e \in S} (1 - x_s) \leq e^{-\sum_{s: e \in S} x_s} \leq e^{-1}$

$\Rightarrow \Pr(e \text{ not covered in } T \text{ rounds}) \leq e^{-T}$

Now use linearity of expectation again. ▣

Hence set $T = \lceil \ln n \rceil$.

$$E[\text{cost}] \leq (\ln n) \cdot LP + \frac{1}{n} \cdot n \cdot LP$$

$$= (\ln n + 1) LP$$

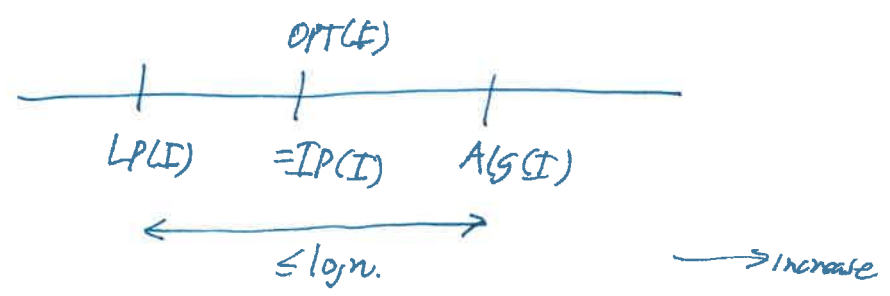
(b/c LP value \geq cheapest set covering for any e)

————— x —————

HW: Show that if sets are of size B , then ~~the~~ LP roundly gives $O(\ln B)$ apx.

Greedy too (but see more later).

Picture



Ask 2 questions:

using this ^{LP} roudy, cannot beat $\lg n$.

① Algorithmic gap: does \exists instance where $\frac{Alg(I)}{OPT(I)} = \Omega(\lg n)$.

② Integrality gap: does \exists instance st $\frac{OPT(I)}{LP(I)} = \Omega(\lg n)$.

shows that using this ~~approach~~ ^{LP} cannot beat the \lg -apx. no matter what roudy we do.

[as long as we relate ourselves to the LP value, of course!]

Algo gap: see in HW.

Integrality gaps:

• Take $U = \{x \in \{0,1\}^d \mid \|x\|_1 = d/2\}$ $d \cong \log_2 n$.

$n = |U| = \binom{d}{d/2} \cong \theta\left(\frac{2^d}{\sqrt{d}}\right)$.

• Sets: all "dictator" sets $S_i = \{x \in U \mid x_i = 1\}$. $\text{cost}_i = 1$.

• OPT $\geq d/2 + 1$ Else \exists element not covered

• LP value: set $\frac{1}{2}$ on each set S . (ie. $x_S = 1 \forall S$)

\Rightarrow total LP value = $d \cdot \frac{1}{2} = \frac{d}{2}$.

\Rightarrow Integrality gap $\geq \frac{d/2 + 1}{d/2} = \Omega(d) = \Omega(\lg n)$.

Fact: Can do better, get $\ln n$ for integrality gap as well.