

Last Class

LP Relaxation for Set Cover

$$\min \sum_{S_i \in S} x_i c(S_i)$$

$$\text{s.t. } \sum_{\substack{i: j \in S_i}} x_i \geq 1 \quad \forall j \in U$$

$$0 \leq x_i \leq 1$$

Dual LP

$$\max \sum_{j \in U} y_j$$

$$\text{s.t., } \sum_{j: j \in S_i} y_j \leq c(S_i) \quad \forall S_i \in S$$

$$y_j \geq 0 \quad \forall j \in U$$

10/27/16

① A3 due next Fri

② Update to A3 posted
(bonus + hints added)

Relaxed primal C.S. :

Let $\alpha \geq 1$.

$\forall j$: either $x_j = 0$ or

$$\frac{c_j}{\alpha} \leq \sum_{i: j \in S_i} y_i \leq c_j$$

Relaxed dual C.S. :

Let $\beta \geq 1$.

$\forall i$: either $y_i = 0$ or

$$b_i \leq \sum_{j=1}^n a_{ij} x_j \leq \beta b_i$$

Algorithm: Choose $\alpha = 1$, $\beta = f$, where f frequency of most common element $j \in U$ in input.

\Rightarrow Primal CS for set cover LP :

$$\forall i, x_i \neq 0 \Rightarrow \sum_{j \in S_i} y_j = c(S_i)$$

\Rightarrow Relaxed CS for dual:

$$\forall j, y_j \neq 0 \Rightarrow 1 \leq \sum_{i: j \in S_i} x_i \leq f$$

Algorithm 15.2 (Set Cover via Primal-dual Schema) [Bar-Yehuda, Even, 1981]

① Set all $x_i = 0$ and $y_j = 0$. (not primal feasible!)

② Until all elements j covered:

Pick an uncovered element j .

Raise y_j until some set goes "tight".

Pick all tight sets in the cover and update $\{x_i\}$ accordingly.

Declare all elements in the tight sets covered.

③ Output set cover $\{x_i\}$.

Analysis: Runtime: each time run loop, we cover at least 1 new element.
∴ # iterations is $\text{poly}(|U|)$.

- Correct:
- Alg returns valid set cover since continues looping so long as \exists uncovered elements.
 - Note $x_i \in \{0, 1\}$ & i is integral sol'n!
 - Also dual feasible sol'n since stop increasing any y_j as soon as a set goes tight.

Approximation achieved:

Claim: Algorithm achieves f -approximation.

PF/ By step ②, we satisfy Primal C.S. condition.
Dual C.S. condition satisfied trivially by any valid set cover.

Suppose alg returns primal/dual solns (x, y) .

By Prop 15.1, $\sum_{S_i \in S} x_i \cdot c(S_i) \leq f \cdot \sum_j y_j \leq f \cdot \sum_j y_j^* = f \cdot \text{OPT}$

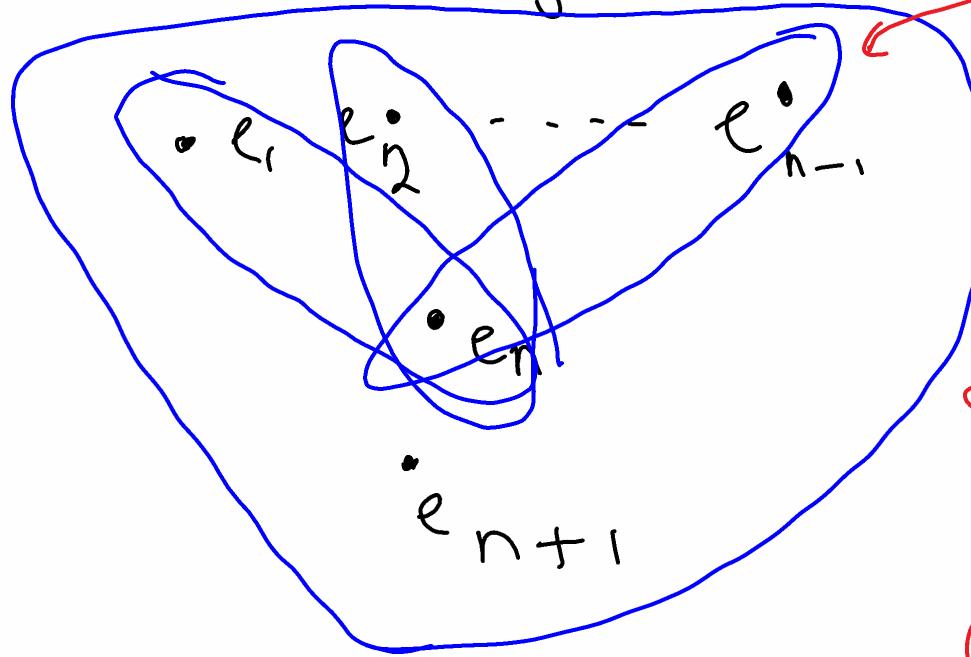
Prop 15.1 Primal value Strong duality for LPs

y_j^* dual optimal

Q: Is our analysis tight?

A: Yes!

Consider set system:



n-1 2-element sets, each cost of 1.

1 (n+1)-element set, cost $1+\epsilon$
for any $\epsilon > 0$.

$OPT = \boxed{1+\epsilon}$ (pick biggest set!)

Our alg: If it picks e_n is first iteration and raises y_{e_n} , and all (α) -element sets will go tight.
∴ it picks all 2-element sets.

$$\boxed{n+\epsilon \approx f \cdot OPT = f(1+\epsilon)}$$

⇒ Total cost: $(n-1) \cdot 1 + (1) \cdot (1+\epsilon) = \boxed{n+\epsilon}$
Since $f=n$ here, this means we saturate f-approximation ratio.

What about rounding-based alg for Set Cover?

Alg 14.1 (Set Cover via LP-rounding)

1. Find optimal soln to set cover LP.

2. Pick all sets S_i for which
 $x_i \geq \frac{1}{F}$ in this soln.

(Disadvantage: Have to use LP solver!)

Claim: Alg 14.1 is an f -approx alg.

PF / Correctness: Why get valid set cover?

↳ Recall from primal $\sum_{S_i \text{ s.t. } j \in S_i} x_i \geq 1 \quad \forall j \in U$.

\exists at most f terms in sum

$\therefore \exists i \text{ s.t. } x_i \geq \frac{1}{F} \quad \forall j \in U$

\therefore all elements of U are covered.

Why do we get approx ratio of f ?

↳ In optimal sol'n to LP, any $x_i^* > \frac{1}{f}$ is rounded to 1
 $x_i^* < \frac{1}{f}$ is rounded to 0

$$(\text{obj fcn: } \sum_i x_i \cdot c(s_i))$$

∴ obj fcn value can increase by $\leq f$.

Let $\{x'_i\}$ be rounded sol'n.
due to this

∴ value of $\{x'_i\} \leq f \cdot (\text{value of } \{x_i^*\}) \leq f \cdot (\text{opt value for set cover}) = f_{\text{OPT}}$

↑
for LP relaxation

↑
I.P. version

Chapter 16 Maximum Satisfiability

Problem 16.1 (MAX-SAT)

Input: ① CNF formula $f: \{0,1\}^n \rightarrow \{0,1\}$, i.e.

$$f = \bigwedge_{c \in C} c \quad \text{where each } c \text{ is OR of literals.}$$

↑
set of clauses

- ② non-negative weights w_c for each $c \in C$.

Output: What is max total weight of satisfied clauses over all assignments $x \in \{0,1\}^n$.

e.g. $f = (\underbrace{x_1 \vee x_2 \vee x_3}_c) \wedge (\underbrace{\bar{x}_1 \vee x_3 \vee \bar{x}_5 \vee x_6}_{c_2}) \wedge (\underbrace{\bar{x}_1 \vee \bar{x}_2 \vee \bar{x}_3}_c)$

$$w_{c_1} = 2, w_{c_2} = 3, w_{c_3} = 9.$$

$$\text{OPT} = 2 + 3 + 9 = 14.$$

NP-complete!

Each x_i is a variable.
Each x_i or \bar{x}_i is a literal.

Notation: Random variable W - total weight of satisfied clauses
 w_c - weight contributed to W by clause c .

$$\therefore W = \sum_{c \in C} w_c.$$

$$E[W] = E\left[\sum_{c \in C} w_c\right] = \sum_{c \in C} E[w_c] = \sum_{c \in C} w_c \cdot \Pr[\text{satisfying clause } c]$$

- Alg strategy: Design 2 algs, and combine them.
- ① Random assignment \leftarrow good for large clauses
 - ② LP relaxation + randomized rounding \leftarrow good for small clauses

Alg 1: Random Assignment (for large clauses)

1. Set each $x_i = 1$ w.p. $\frac{1}{2}$ independently & uniformly at random.

How well does this do? ie what is $E[W]$?

Lemma 16.2 If $c \in C$ has k literals, then $E[w_c] = (1 - \frac{1}{2^k}) \cdot w_c$.

$$\text{Pf} | \quad E[w_c] = \Pr[\text{satisfy } c] \cdot w_c$$

$$\begin{aligned} \Pr[\text{satisfy } c] &= 1 - \Pr[\text{fail to satisfy } c] \\ &= 1 - \frac{1}{2^k} \leftarrow \begin{array}{l} \text{eq } c = (x_1 \vee \overline{x}_2 \vee \dots \vee x_k) \\ \text{only 1 assignment out of } 2^k \text{ fails!} \end{array} \end{aligned}$$

\therefore For whole alg:

$$E[W] = \sum_{c \in C} E[w_c] = \sum_{c \in C} \left(1 - \frac{1}{2^{k_c}}\right) w_c \stackrel{\text{size of clause } c}{\geq} \left(1 - \frac{1}{2^{k_{\min}}}\right) \sum_{c \in C} w_c \stackrel{\text{OPT.}}{\geq} \left(1 - \frac{1}{2^{k_{\min}}}\right) \sum_{c \in C} w_c$$

$k_{\min} \triangleright$
smallest
clause size

$\sum_{c \in C} w_c$ means
satisfy
all clauses.