

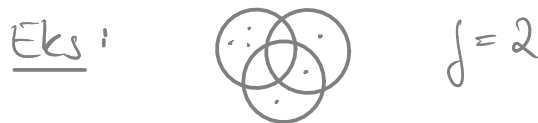
Set Cover - recap.

LP-relax: $\min \sum_{j=1}^m x_j w_j$

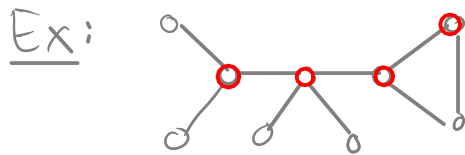
s.t. $\sum_{j: e_i \in S_j} x_j \geq 1, \quad 1 \leq i \leq n$

$x_j \geq 0, \quad 1 \leq j \leq m$

Deterministic rounding: f -approx. alg.



Vertex Cover



Exercises

Recap ctd.:

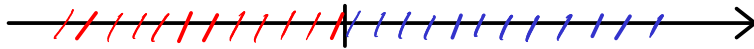
Dual LP: $\max \sum_{i=1}^n y_i$

s.t. $\sum_{e_i \in S_j} y_i \leq w_j, \quad 1 \leq j \leq m$

$y_i \geq 0, \quad 1 \leq i \leq n$

For any pair \vec{x}, \vec{y} to the primal/dual problems:

$$\sum_{i=1}^n y_i \leq \sum_{j=1}^m x_j w_j \quad (\text{weak duality})$$



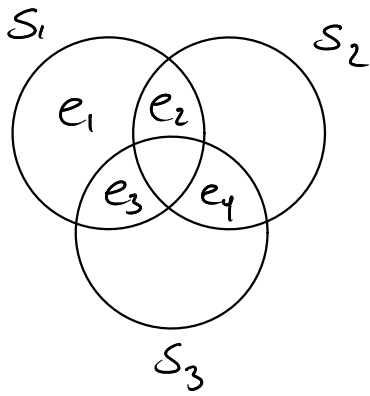
↑
optimum objective value
for both problems (strong duality)

(Relaxed) Complementary Slackness Conditions:

Primal c.s.c. $x_j > 0 \Rightarrow \sum_{i \in S_j} y_i = w_j \quad (\geq \frac{1}{b} w_j), \quad 1 \leq j \leq m$

Dual c.s.c. $y_i > 0 \Rightarrow \sum_{j \in S_i} x_j = 1 \quad (\leq c), \quad 1 \leq i \leq n$

Ex:



$$w_1 = 1$$

$$w_2 = 2$$

$$w_3 = 3$$

Primal:

$$\min \quad x_1 + 2x_2 + 3x_3$$

$$\text{s.t.} \quad x_1 \geq 1$$

$$x_1 + x_2 \geq 1$$

$$x_1 + x_3 \geq 1$$

$$x_2 + x_3 \geq 1$$

$$x_1, x_2, x_3 \geq 0$$

$$\text{OPT} = 3:$$

$$x_1 = x_2 = 1$$

Dual:

$$\max \quad y_1 + y_2 + y_3 + y_4$$

$$\text{s.t.} \quad y_1 + y_2 + y_3 \leq 1$$

$$y_2 + y_4 \leq 2$$

$$y_3 + y_4 \leq 3$$

$$y_1, y_2, y_3, y_4 \geq 0$$

$$\text{OPT} = 3:$$

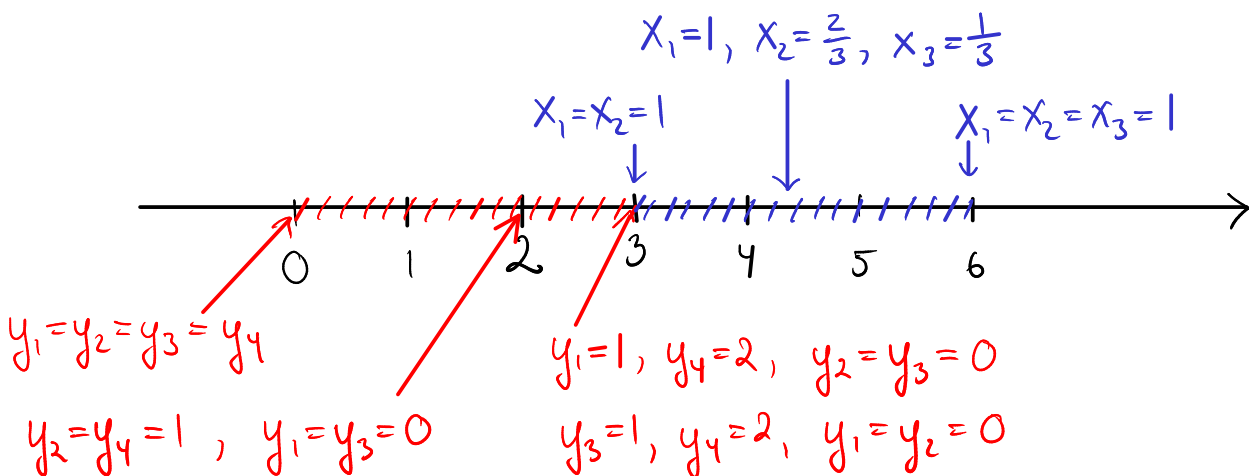
$$y_1 = 1$$

$$y_4 = 2$$

or

$$y_3 = 1$$

$$y_4 = 2$$



Alg. 2 for Set Cover

$\vec{y}^* \leftarrow$ opt. sol. to dual LP

$I' \leftarrow \{j \mid \sum_{e_i \in S_j} y_i = w_j\}$

In the ex. above:

with $y_1^* = 1$, $y_4^* = 2$, Alg. 2 would choose S_1 and S_2 with a total weight of 3.

with $y_3^* = 1$, $y_4^* = 2$, Alg. 2 would choose S_1, S_2 , and S_3 with a total weight of 6.

The first solution is optimal, and the latter is a 2-approximation (i.e., an f -approximation).

Alg. 2 is an f -approximation algo.:

If the algo. chooses S_1, S_2 , and S_3 , the total weight is $W = w_1 + w_2 + w_3$, and

$$w_1 + w_2 + w_3 = (y_1^* + y_2^* + y_3^*) + (y_2^* + y_4^*) + (y_3^* + y_4^*),$$

since the algo. chooses exactly those sets that have LHS = RHS.

Since each y_i is present in at most f constraints,

$$W \leq f \cdot (y_1^* + y_2^* + y_3^* + y_4^*)$$

$$= f \cdot Z_{\text{dual}}^*$$

$$\leq f \cdot Z_{\text{primal}}^*, \text{ by the weak duality property}$$

$$= f \cdot \text{OPT}$$

Lemma 1.7

Alg. 2 produces a set cover

Proof:

Assume for the sake of **contradiction** that some element e_k is not covered by $\{S_j \mid j \in I\}$.

Then $\sum_{e_i \in S_j} y_i < w_j$ for all S_j containing e_k .

These are exactly the constraints involving y_k . Thus, none of the constraints involving y_k are tight.

This means that y_k can be increased without violating any constraint.

Since this will increase the value $\sum_{i=1}^n y_i$ of the sol., we conclude that the solution \vec{y} was not optimal. \square

Ex:

In the ex. above, assume

$$y_1 = y_2 = y_3 = 0$$

$$y_4 = 2$$

Then, only the second constraint is tight, so only S_2 is picked:

$$y_1 + y_2 + y_3 = 0 < 1$$

$$y_2 + y_4 = 2$$

$$y_3 + y_4 = 2 < 3$$

e_4 is not covered, since none of the two constraints involving y_4 are tight.

We can increase y_3 from 0 to 1 without violating any constraints

(Then two other constraints become tight.)

This increases the sol. value from 2 to 3.

Thus, the sol. above was not optimal.

Or we could increase y_1 from 0 to 1.

Then only the first constraint becomes tight, resulting in an optimal solution.

This illustrates the idea of the primal-dual alg of Section 1.5.

We now give a more formal proof that Alg 2 is an f -approximation algo.

Thm 1.8

Alg. 2 is an f -approx. algo.

Proof:

The correctness follows from Lemma 1.7.

Approx. guarantee:

$$\begin{aligned} \sum_{j \in I'} w_j &= \sum_{j \in I'} \sum_{e_i \in S_j} y_i^* && \text{ } y_i^* \text{ appears once for each set in the sol.} \\ &= \sum_{i=1}^n \underbrace{|\{j \in I' \mid e_i \in S_j\}|}_{\text{\#sets in the sol. containing } e_i} \cdot y_i^* \\ &\leq \sum_{i=1}^n \underbrace{d_{e_i}}_{\text{\#sets containing } e_i} \cdot y_i^* \\ &\leq \sum_{i=1}^n f \cdot y_i^* \\ &= f \cdot Z_{\text{dual}}^* \\ &\leq f \cdot Z_{\text{primal}}^*, \text{ by the weak duality property} \\ &\leq f \cdot \text{OPT} \end{aligned}$$

□

Note that for proving the above theorem, we could also use the relaxed C.S.C. (with $b=1$, $c=f$), since

$$\sum_{j: e_i \in E_j} x_j \leq f, \text{ for all } i=1,2,\dots,n$$

Note that, on any instance of Set Cover, $I \subseteq I'$:

Since the LP is solved optimally,

$x_j > 0 \Rightarrow$ constraint j is tight $\Rightarrow j \in I'$.

Thus, $j \in I \Rightarrow x_j \geq \frac{1}{f} \Rightarrow j \in I'$

Thus, Alg. 1 is always at least as good as Alg. 2.

Both Alg. 1 and Alg. 2 rely on solving an LP (optimally). In Section 1.5, we will study a more time efficient alg.

The key observation is that in the proof of Thm 1.8, we did not need the fact that \vec{y}^* is optimal, since $Z_{\text{dual}} \leq Z_{\text{primal}}^*$, for any feasible dual solution.

Thus, the crux is to obtain an index set I'' s.t.

- $\bigcup_{j \in I''} S_j$ is a vertex cover
- $\sum_{j \in I''} w_j = \sum_{j \in I''} \sum_{e_i \in S_j} y_i$, for some feasible sol. \vec{y} to the dual LP

without solving an LP optimally.

Section 1.5: A Primal-Dual Alg. for Set Cover

Alg. 1.1 for Set Cover: Primal-Dual

$$I'' \leftarrow \emptyset$$

$$\vec{y} \leftarrow \vec{0}$$

While $\exists e_k \notin \bigcup_{j \in I''} S_j$

Increase y_k until some constraint, l , becomes tight, i.e., $\sum_{e_i \in S_l} y_i = w_l$

$$I'' \leftarrow I'' \cup \{l\}$$

Note that $e_k \in S_l$

Thm 1.9

Alg. 1.1 is an f -approx. alg. for Set Cover

Proof:

Alg. 3 produces a **set cover**, since as long as some element is not covered, the corresponding dual constraints are non-tight.

The **approx. guarantee** follows from the same calculations as in the proof of Thm. 1.8,

since

$$\sum_{j \in I''} w_j = \sum_{j \in I''} \sum_{e_i \in S_j} y_i \leq f \cdot Z_{\text{dual}} \leq f \cdot Z_{\text{dual}}^*$$

□

In contrast to Alg. 2 from Section 1.4, Alg. 1.1 does not necessarily produce an optimal dual solution:

In the example above, it might do the following.

$$y_2 \leftarrow 1 \quad (S_1 \text{ is picked, } e_4 \text{ still uncovered})$$

$$y_4 \leftarrow 1 \quad (S_2 \text{ is picked})$$

(This is fine, since the proof of Thm. 1.8 does not use that $\sum y_i = \text{OPT}$, only that $\sum y_i \leq \text{OPT}$, which is true for any feasible sol. to the dual.)