

## Set Cover Primal

$$\begin{aligned} \min \quad & \sum_{j=1}^m x_j w_j \\ \text{s.t.} \quad & \sum_{j: e_i \in S_j} x_j \geq 1, \quad i=1, 2, \dots, n \\ & x_j \geq 0, \quad j=1, 2, \dots, m \end{aligned}$$

Covering  
problem

## Set Cover Dual

$$\begin{aligned} \max \quad & \sum_{i=1}^n y_i \\ \text{s.t.} \quad & \sum_{e_i \in S_j} y_i \leq w_j, \quad j=1, 2, \dots, m \\ & y_i \geq 0, \quad i=1, 2, \dots, n \end{aligned}$$

Packing  
problem

The primal problem is a **covering** problem:  
Each element has to be covered by at least one set.

The dual problem can be viewed as a **packing** problem:

Each set  $S_j$  has a capacity of  $w_j$ .

We interpret  $y_i$  as the weight of  $e_i$ ,  
and the total weight of elements in  $S_j$  must not exceed  $w_j$ .

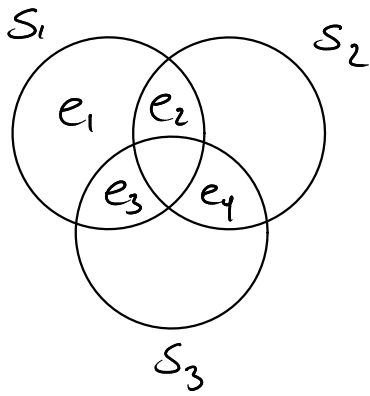
Recall that the dual is constructed such that the value of any solution to the dual is a lower bound on the value of any solution to the primal:

$$Z_{\text{Primal}} \geq Z_{\text{Dual}} \quad (\text{weak duality property})$$

In fact,

$$Z_{\text{Primal}}^* = Z_{\text{Dual}}^* \quad (\text{strong duality property})$$

Ex:



$$\begin{aligned}w_1 &= 1 \\w_2 &= 2 \\w_3 &= 3\end{aligned}$$

Primal:

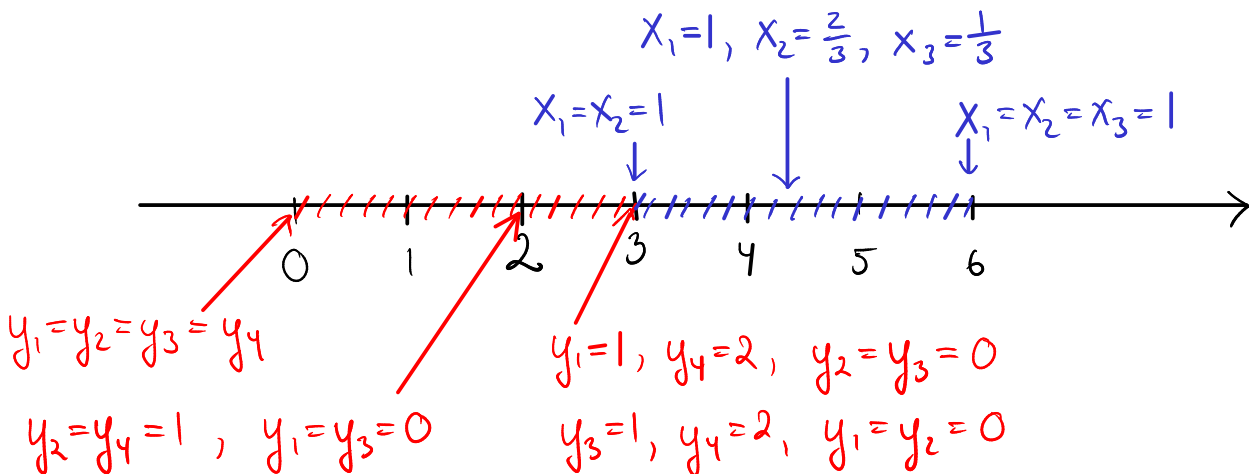
$$\begin{aligned}\min \quad & x_1 + 2x_2 + 3x_3 \\ \text{s.t.} \quad & x_1 \geq 1 \\ & x_1 + x_2 \geq 1 \\ & x_1 + x_3 \geq 1 \\ & x_2 + x_3 \geq 1 \\ & x_1, x_2, x_3 \geq 0\end{aligned}$$

$$\begin{aligned}\text{OPT} &= 3 : \\ x_1 &= x_2 = 1\end{aligned}$$

Dual:

$$\begin{aligned}\max \quad & y_1 + y_2 + y_3 + y_4 \\ \text{s.t.} \quad & y_1 + y_2 + y_3 \leq 1 \\ & y_2 + y_4 \leq 2 \\ & y_3 + y_4 \leq 3 \\ & y_1, y_2, y_3, y_4 \geq 0\end{aligned}$$

$$\begin{aligned}\text{OPT} &= 3 : \\ y_1 &= 1 \quad \text{or} \quad y_3 = 1 \\ y_4 &= 2 \quad \quad \quad y_4 = 2\end{aligned}$$



Now back to approximation algorithms for Set Cover.

We have seen an alg. (Alg.1) which solves the LP relaxation and then does deterministic rounding.

We will now look at an alg. (Alg.2) which solves the dual of the LP relaxation.

Then each set corresponding to a tight constraint is selected:

### Alg.2 for Set Cover

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

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

Let's see how the alg. would work on the example from before. Recall that there were two optimal solutions to the dual:

$y_1=1, y_2=y_3=0, y_4=2$  and  $y_1=y_2=0, y_3=1, y_4=2$ .

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. on this example:

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,

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

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

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

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

Before giving the general proof that Alg. 2 is an  $f$ -approx. alg., we show that it always produces a valid set cover:

### 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  $y_4$  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

- $\sum_{j: e_i \in S_j} x_j \leq f$ , for all  $i=1, 2, \dots, n$ , by the def. of  $f$ .
- $x_j = 1 \Rightarrow \sum_{e_i \in S_j} y_i = w_j$ , by the def. of the alg.

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.

- $\{S_j \mid j \in I''\}$  is a set 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.)