

Recap:

$\alpha$ -approx. alg.

Approx. factor / ratio

Set Cover

Vertex Cover

$f$ -approx. alg (LP-rounding)

Primal  $\longleftrightarrow$  Dual

$$\begin{array}{ll}\min & C_1 x_1 + \dots + C_n x_n \\ \text{s.t.} & a_{11} x_1 + \dots + a_{1n} x_n \geq b_1 \\ & \vdots \\ & a_{m1} x_1 + \dots + a_{mn} x_n \geq b_m \\ & x_1, \dots, x_n \geq 0\end{array}$$

$$\begin{array}{ll}\max & b_1 y_1 + \dots + b_m y_m \\ & a_{11} y_1 + \dots + a_{m1} y_m \leq C_1 \\ & \vdots \\ & a_{1n} y_1 + \dots + a_{mn} y_m \leq C_n \\ & y_1, \dots, y_m \geq 0\end{array}$$

**duality**  $\left\{ \begin{array}{l} \text{For any pair } \vec{x}, \vec{y} \text{ of solutions,} \\ C_1 x_1 + \dots + C_n x_n \leq b_1 y_1 + \dots + b_m y_m \end{array} \right\}$  Weak duality

$\left\{ \begin{array}{l} \exists \text{ a pair } \vec{x}^*, \vec{y}^* \text{ of sol. s.t.} \\ C_1 x_1^* + \dots + C_n x_n^* = b_1 y_1^* + \dots + b_m y_m^* \end{array} \right\}$

Ex:

$$\begin{aligned} \min \quad & 7x_1 + x_2 + 5x_3 \\ \text{s.t.} \quad & x_1 - x_2 + 3x_3 \geq 10 \\ & 5x_1 + 2x_2 - x_3 \geq 6 \\ & x_1, x_2, x_3 \geq 0 \end{aligned}$$

$$\begin{aligned} 7x_1 + x_2 + 5x_3 &\geq y_1(x_1 - x_2 + 3x_3) + y_2(5x_1 + 2x_2 - x_3) \\ &= (y_1 + 5y_2)x_1 + (-y_1 + 2y_2)x_2 + (3y_1 - y_2)x_3 \end{aligned}$$

$$\begin{aligned} \max \quad & 10y_1 + 6y_2 \\ \text{s.t.} \quad & y_1 + 5y_2 \leq 7 \\ & -y_1 + 2y_2 \leq 1 \\ & 3y_1 - y_2 \leq 5 \\ & y_1, y_2 \geq 0 \end{aligned}$$

$$\begin{cases} y_1 = 0 \\ 5x_1 + 2x_2 - x_3 = 6 \\ y_1 + 5y_2 = 7 \\ x_2 = 0 \\ 3y_1 - y_2 = 5 \end{cases}$$

$\Downarrow$

$$\begin{aligned} \underbrace{0}_{10y_1} + 6y_2 &= \overbrace{y_1(x_1 - x_2 + 3x_3)}^0 + \overbrace{y_2(5x_1 + 2x_2 - x_3)}^6 \\ &= \underbrace{(y_1 + 5y_2)}_7 x_1 + \underbrace{(-y_1 + 2y_2)}_0 x_2 + \underbrace{(3y_1 - y_2)}_5 x_3 \\ &= 7x_1 + \underbrace{x_2}_0 + 5x_3 \end{aligned}$$

More generally:

$$\begin{array}{l}
 \Downarrow \\
 \text{Complementary} \\
 \text{Slackness} \\
 \text{Conditions}
 \end{array}
 \left\{
 \begin{array}{l}
 7x_1 + x_2 + 5x_3 = 10y_1 + 6y_2 \\
 \begin{array}{l}
 x_1 > 0 \Rightarrow y_1 + 5y_2 = 7 \\
 x_2 > 0 \Rightarrow -y_1 + 2y_2 = 1 \\
 x_3 > 0 \Rightarrow 3y_1 - y_2 = 5
 \end{array} \\
 \begin{array}{l}
 y_1 > 0 \Rightarrow x_1 - x_2 + 3x_3 = 10 \\
 y_2 > 0 \Rightarrow 5x_1 + 2x_2 - x_3 = 6
 \end{array}
 \end{array}
 \right\}
 \begin{array}{l}
 \text{primal c.s.c.} \\
 \text{dual c.s.c.}
 \end{array}$$

By The **Strong Duality Theorem** (which we will not prove), there exist solutions fulfilling the c.s.c.

Moreover, if the c.s.c. are "close" to being satisfied, the values of the primal and dual sd. are "close":

$$\begin{array}{l}
 \text{Relaxed} \\
 \text{Complementary} \\
 \text{Slackness} \\
 \text{Conditions}
 \end{array}
 \left\{
 \begin{array}{l}
 x_1 > 0 \Rightarrow y_1 + 5y_2 \geq 7/b \\
 x_2 > 0 \Rightarrow -y_1 + 2y_2 \geq 1/b \\
 x_3 > 0 \Rightarrow 3y_1 - y_2 \geq 5/b \\
 y_1 > 0 \Rightarrow x_1 - x_2 + 3x_3 \leq 10c \\
 y_2 > 0 \Rightarrow 5x_1 + 2x_2 - x_3 \leq 6c
 \end{array}
 \right.$$

$$\Downarrow \\
 7x_1 + x_2 + 5x_3 \leq bc(10y_1 + 6y_2)$$

Ex:

$$\begin{cases} y_1 = 0 \\ 5x_1 + 2x_2 - x_3 \leq 2 \cdot 6 \\ y_1 + 5y_2 \geq 7/3 \\ x_2 = 0 \\ 3y_1 - y_2 \geq 5/3 \end{cases}$$

$$\begin{aligned} \Downarrow \quad 2 \cdot (10y_1 + 6y_2) &\geq \overbrace{y_1(x_1 - x_2 + 3x_3)}^{=0} + \overbrace{y_2(5x_1 + 2x_2 - x_3)}^{\leq 2 \cdot 6y_2} \\ &= \underbrace{(y_1 + 5y_2)x_1}_{\geq 7/3 x_1} + \underbrace{(-y_1 + 2y_2)x_2}_{=0} + \underbrace{(3y_1 - y_2)x_3}_{\geq 5/3 x_3} \\ &\geq \frac{1}{3} (7x_1 + \underbrace{x_2}_{=0} + 5x_3) \end{aligned}$$

$$\Downarrow \quad 2 \cdot 3 (10y_1 + 6y_2) \geq 7x_1 + x_2 + 5x_3$$

## Sheet 1

a) LP-formulation of unweighted Vertex Cover

$$\min \sum_{v \in V} x_v$$

$$\text{s.t. } x_u + x_v \geq 1, \quad (u, v) \in E$$
$$x_v \geq 0, \quad v \in V$$

b) Dual LP

$$\max \sum_{e \in E} y_e$$

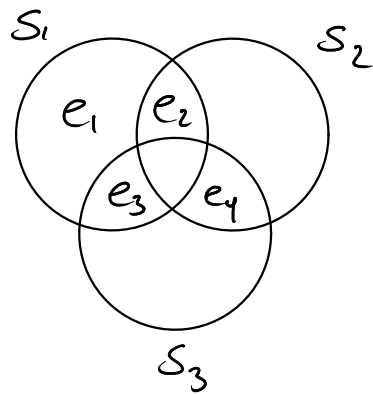
$$\text{s.t. } \sum_{(u, v) \in E} y_{(u, v)} \leq 1, \quad u \in V$$
$$y_e \geq 0, \quad e \in E$$

c) Which combinatorial problem?

Unweighted Matching (Max. Cardinality Matching)

What is the dual of the Set Cover LP?

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

$$y_3 = 1$$

$$y_4 = 2$$

### 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

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,

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

## Alg. 2 for Set Cover

Solve 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,

$$W \leq f \cdot (y_1 + y_2 + y_3 + y_4) = f \cdot \text{OPT}$$



### 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.

□

Ex:

In the ex. above, assume

$$y_1 = y_4 = 0$$

$$y_2 = y_3 = \frac{1}{2}$$

Then, only the first constraint is tight, so only  $S_1$  is picked.

$$y_1 + y_2 + y_3 = 1$$

$$y_2 + y_4 = \frac{1}{2} < 2$$

$$y_3 + y_4 = \frac{1}{2} < 3$$

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

We can increase  $y_4$  from 0 to  $\frac{3}{2}$  without violating any constraints

This increases the sol. value from 1 to  $\frac{5}{2}$ .

Thus, the sol. above was not optimal.

This illustrates the idea of the primal-dual alg of Section 1.5 (although this alg. would not start out with the sol  $y_2 = y_3 = \frac{1}{2}$ ).

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:

$$\begin{aligned}\sum_{j \in I'} w_j &= \sum_{j \in I'} \sum_{e_i \in S_j} y_i \\ &= \sum_{i=1}^n \underbrace{|\{j \in I' \mid e_i \in S_j\}|}_{\text{\#sets in the sol. containing } e_i} \cdot y_i \\ &\leq \sum_{i=1}^n \underbrace{d_i}_{\text{\#sets containing } e_i} \cdot y_i \\ &\leq \sum_{i=1}^n f \cdot y_i \\ &= f \cdot \text{OPT}\end{aligned}$$

□

Note that we could also use the relaxed c.s.c. (with  $b=1$ ,  $c=f$ ), since

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

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. In Section 1.5, we will study a more (time) efficient version of Alg. 2.

The crux is to obtain an index set  $I''$ , s.t.

- $\bigcup_{j \in I''} S_j$  is a vertex cover

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

without solving an LP.

## 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. 1.1 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,

$$\text{since } \sum_{j \in I''} w_j = \sum_{j \in I''} \sum_{e_i \in S_j} y_i$$

□

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.)

## Section 1.6: A Greedy Algorithm

### Alg 1.2 for Set Cover: Greedy

$I \leftarrow \emptyset$

For  $j \leftarrow 1$  to  $m$

$\hat{S}_j \leftarrow S_j$  („uncovered“ part of  $S_j$ )

While  $\{S_j \mid j \in I\}$  is not a set cover

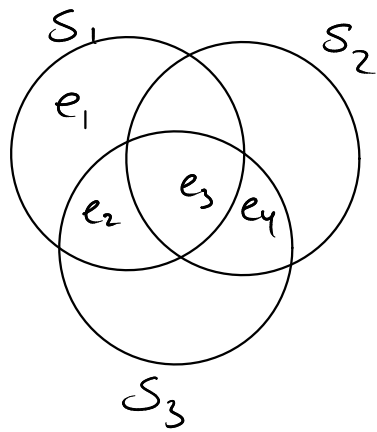
$l \leftarrow \arg \min_{j: \hat{S}_j \neq \emptyset} \frac{w_j}{|\hat{S}_j|}$  ( $S_l$ : set with smallest cost per uncovered element)

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

For  $j \leftarrow 1$  to  $m$

$\hat{S}_j \leftarrow \hat{S}_j - S_l$

Ex:



$$w_1 = 12$$

$$w_2 = 8$$

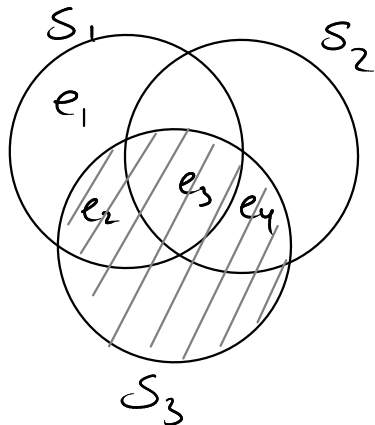
$$w_3 = 9$$

$$\text{Let } \alpha_j = \frac{w_j}{|\hat{S}_j|}$$

$$\alpha_1 = \frac{12}{3} = 4, \quad \alpha_2 = \frac{8}{2} = 4, \quad \alpha_3 = \frac{9}{3} = 3$$

Pick  $S_3$

price per element  
in first iteration



$$\alpha_1 = \frac{12}{1}$$

Pick  $S_1$

price per element in  
second iteration

Done!

$$n_1 = 5, \quad n_2 = 1, \quad n_3 = 0$$

$$l = 3$$

$$n_1 - n_2 = 4, \quad n_3 - n_2 = 1$$



### Thm 1.11

Alg. 1.2 is an  $H_n$ -approx. alg. for Set Cover

Proof:

$n_k$ : #uncovered elements at the beginning of the  $k$ 'th iteration

OPT has to cover these  $n_k$  elements using only sets in  $\mathcal{I}_k = \mathcal{I} - \{S_j \mid j \in I\}$ , since none of them are contained in  $\{S_j \mid j \in I\}$ .

Hence, there must be at least one element with a price of at most  $\text{OPT}/n_k$ . Otherwise, OPT would not be able to cover the  $n_k$  elements (and certainly not all  $n$  elements) at a cost of only OPT.

Hence, the  $n_k - n_{k+1}$  elements covered in iteration  $k$  cost at most  $(n_k - n_{k+1}) \text{OPT}/n_k$  per element.

Thus, the cost of the set cover produced by the greedy alg. is

$$\begin{aligned}
\sum_{j \in I} w_j &\leq \sum_{k=1}^{\overset{\text{\# iterations}}{\ell}} \frac{n_k - n_{k+1}}{n_k} \text{OPT} \\
&= \text{OPT} \sum_{k=1}^{\ell} \underbrace{\left( \frac{1}{n_k} + \frac{1}{n_k} + \dots + \frac{1}{n_k} \right)}_{n_k - n_{k+1} \text{ terms}} \\
&\leq \text{OPT} \sum_{k=1}^{\ell} \left( \frac{1}{n_k} + \frac{1}{n_{k-1}} + \dots + \frac{1}{n_{k+1}+1} \right) \\
&= \text{OPT} \sum_{s=1}^n \frac{1}{s} \\
&= \text{OPT} \cdot H_n \quad \square
\end{aligned}$$

Ex from before:

$$\text{OPT} = w_1 + w_2 = 12 + 8 = 20$$

The cost of the greedy alg is

$$\begin{aligned}
w_3 + w_1 &= 9 + 12 \\
&= (3+3+3) + 12 \\
&\leq \left( \frac{20}{4} + \frac{20}{4} + \frac{20}{4} \right) + \frac{20}{1} \\
&\leq \left( \frac{20}{4} + \frac{20}{3} + \frac{20}{2} \right) + \frac{20}{1} \\
&= 20 \left( \frac{1}{4} + \frac{1}{3} + \frac{1}{2} + 1 \right)
\end{aligned}$$

Let  $g = \max \{ |\delta_i| \mid \delta_i \in \mathcal{G} \}$ .

Thm 1.12

Alg. 1.2 is an  $H_g$ -approx. alg. for Set Cover

Proof: By dual fitting.

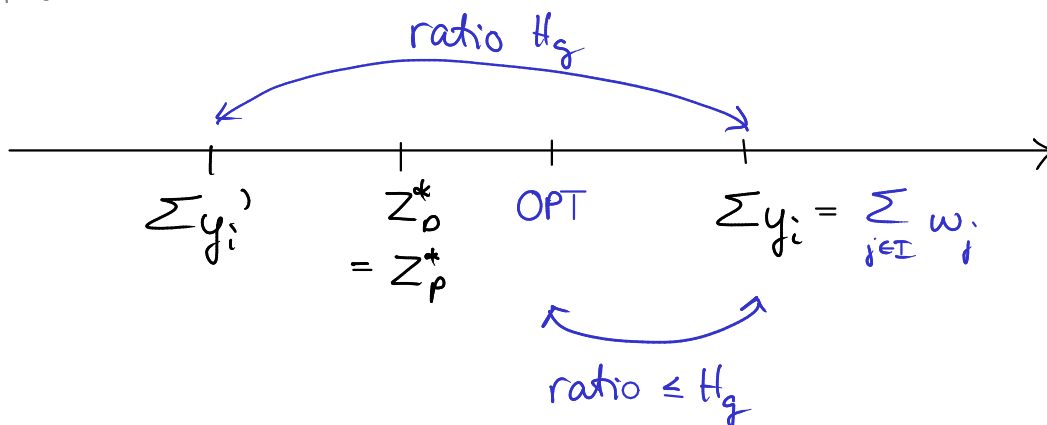
Consider the dual  $D$  of the LP for Set Cover.  
We will construct an infeasible solution  $\vec{y}$  to  $D$  s.t.

- $\sum_{j \in I} w_j = \sum_{i=1}^n y_i$
- $y_i' = \frac{y_i}{H_g}$ ,  $1 \leq i \leq n$ , is a feasible sol. to  $D$

This will imply that

$$\sum_{j \in I} w_j = \sum y_i = H_g \sum y_i' = H_g Z_D^* = H_g Z_P^* \leq H_g \cdot \text{OPT}$$

Illustration:



For each  $i$ ,  $1 \leq i \leq n$ , we let

$$y_i = \text{price}(e_i)$$

Then

$$\bullet \sum_{j \in I} w_j = \sum_{i=1}^n y_i$$

Hence, we just need to show that

$$\bullet \vec{y} \text{ is feasible,}$$

i.e., show that

$$\sum_{e_i \in S_j} y_i \leq w_j, \text{ for each set } S_j \in \mathcal{P}:$$

