

Exercise 5.7:

Derandomize the rounding alg. from Section 1.7, using the method of conditional expectations.

Hint: Use the following obj. fct. with random variables X_j , $1 \leq j \leq m$, and Z .

$$f = \sum_{j=1}^m X_j w_j + \lambda Z$$

$\left| \begin{array}{l} n \cdot \ln n \cdot Z_{LP}^* \\ \end{array} \right.$

{ 0, if set cover
1, otherwise

{ 1, if S_j incl.
0, otherwise

With this obj. fct.,

any infeasible sol. has $f \geq \lambda = n \cdot \ln n \cdot Z_{LP}^*$ (*)

For AlgRR₂,

$$\begin{aligned} E[f] &= E\left[\sum_{j=1}^m X_j w_j\right] + \lambda E[Z], \text{ by lin. of exp.} \\ &\leq 2 \cdot \ln n \cdot Z_{LP}^* + n \cdot \ln n \cdot Z_{LP}^* \cdot n^{1/2}, \text{ by the analysis in Sec. 1.7} \\ &= 3 \cdot \ln n \cdot Z_{LP}^* \end{aligned}$$

Thus, using the method of cond. exp., we can find a sol with $f \leq E[f] = 3 \cdot \ln n \cdot Z_{LP}^*$, and by (*), such a sol is a set cover.

To derandomize the alg. we must be able to calculate conditional exp values, i.e., calculate $E[f]$, given that decisions about S_1, \dots, S_ℓ have already been made:

$$E[f | \vec{X}_\ell] = \sum_{j=1}^{\ell} X_j w_j + \sum_{j=\ell+1}^m x_j w_j + \lambda E[z | \vec{X}_\ell]$$

where $\vec{X}_\ell = (X_1, X_2, \dots, X_\ell)$, and $E[z | \vec{X}_\ell]$ can be calculated in the following way.

For each element e_i

$$\Pr[e_i \text{ covered} | \vec{X}_\ell]$$

$$= \begin{cases} 1, & \text{if } e_i \text{ is contained in a set } S_j \\ & \text{s.t. } j \leq \ell \text{ and } X_j = 1 \text{ (i.e., } e_i \text{ is} \\ & \text{covered by one of the sets } S_1, \dots, S_\ell) \\ 1 - \underbrace{\prod_{\substack{j: e_i \in S_j \\ \wedge j > \ell}} (1 - x_j)}_{\text{prob. that } e_i \text{ will not be covered by any}} & \text{otherwise} \end{cases}$$

$$E[z | \vec{X}_\ell] = 1 - \prod_{i=1}^n \Pr[e_i \text{ covered} | \vec{X}_\ell]$$

DeRR₂

Solve LP optimally

For $\ell \leftarrow 1$ to m

If $E[f | (x_1, x_2, \dots, x_{\ell-1}, 0)] \leq E[f | (x_1, x_2, \dots, x_{\ell-1}, 1)]$

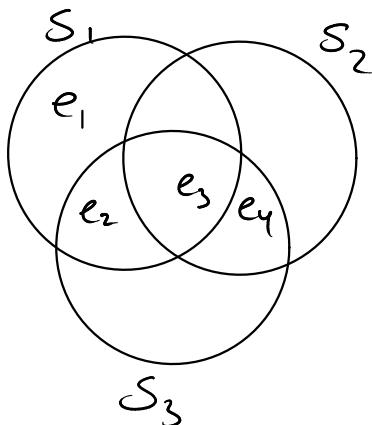
$x_\ell \leftarrow 0$

Else

$x_\ell \leftarrow 1$

Greedy recap.

Ex:



Price per element
1. it. 2. it.

$w_1 = 12$	4	6
$w_2 = 4$	2	-
$w_3 = 9$	3	9

1. Pick $S_2 \rightarrow \text{price}(e_3) = \text{price}(e_4) = 2$
2. Pick $S_1 \rightarrow \text{price}(e_1) = \text{price}(e_2) = 6$

Total weight

$$\begin{aligned}
 &= w_2 + w_1 \\
 &= (\text{price}(e_3) + \text{price}(e_4)) + (\text{price}(e_1) + \text{price}(e_2)) \\
 &= (2+2) + (6+6) \\
 &= 16
 \end{aligned}$$

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

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} and a feasible solution \vec{y}' such that

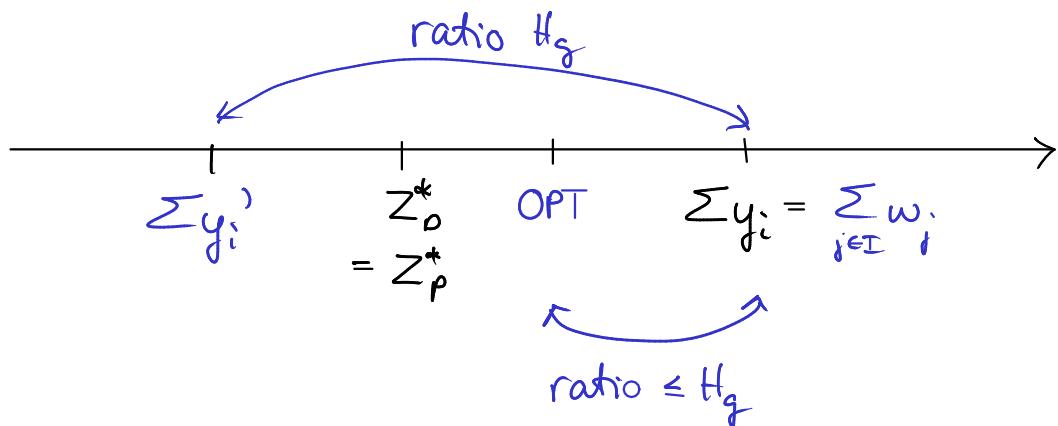
- $\sum_{i=1}^n y_i = \sum_{j \in I} w_j$

- $y'_i = \frac{y_i}{H_g}, \quad 1 \leq i \leq n$

Then,

$$\sum_{j \in I} w_j = \sum_{i=1}^n y_i = H_g \sum_{i=1}^n y'_i \leq H_g Z_D^* \leq H_g \cdot OPT,$$

proving the claimed approximation factor.



For $1 \leq i \leq n$, let $y_i = \text{price}(e_i)$. Then,

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

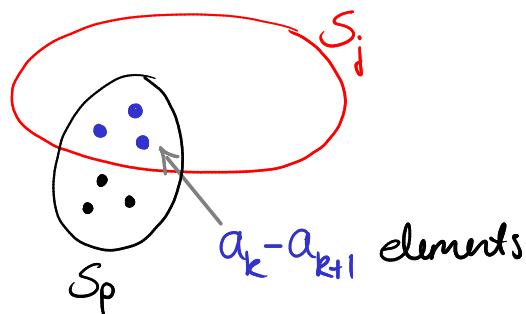
Hence, we just need to show that \vec{y} is feasible:

Consider an arbitrary set S_j .

Let a_k be #uncovered elements in S_j at the beginning of the k 'th iteration.

Let S_p be the set chosen by Greedy in the k 'th iteration.

S_p covers $a_k - a_{k+1}$ previously uncovered elements in S_j



The price per element in S_j covered in the k 'th iteration is at most

$$\frac{w_p}{|S_p|} \leq \frac{w_j}{a_k}$$

since otherwise S_j would be a more greedy choice. ↴

Thus,

Total #terms = $|S_j|$, since $a_i = |S_j|$ and $a_{r+1} = 0$

$$\begin{aligned}\sum_{e_i \in S_j} y_i &\leq \underbrace{\sum_{k=1}^r (a_k - a_{k+1})}_{|S_j|} \frac{w_j}{a_k} \\ &\leq w_j \sum_{i=1}^{|S_j|} \frac{1}{i}, \text{ by the same arguments as in} \\ &\quad \text{the proof of Thm 1.12.} \\ &\leq w_j \sum_{i=1}^g \frac{1}{i} \\ &= w_j \cdot H_g\end{aligned}$$

Hence,

$$\sum_{e_i \in S_j} y'_i = \frac{1}{H_g} \sum_{e_i \in S_j} y_i \leq w_j$$

□

Compare the proof of Thm 1.12 (dual fitting) to the proof of Thm 1.11:

- Simpler: Compare priors to w_j instead of OPT
- Stronger result: H_g instead of H_n
(could also have been obtained with the technique of the proof of Thm 1.11)

Ex from before:

$$y_3 = y_4 = 2$$

$$y_1 = y_2 = 6$$

$$H_3 = 1 + \frac{1}{2} + \frac{1}{3} = \frac{11}{6}$$

$$y'_3 = y'_4 = \frac{1}{H_3} \cdot 2 = \frac{6}{11} \cdot 2 = \frac{12}{11}$$

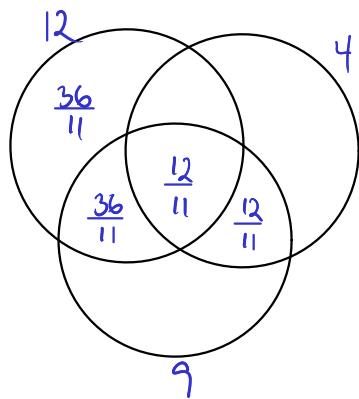
$$y'_1 = y'_2 = \frac{6}{11} \cdot 6 = \frac{36}{11}$$

\vec{y}' is feasible:

$$y'_1 + y'_2 + y'_3 = 2 \cdot \frac{36}{11} + \frac{12}{11} < 8 \leq w_1$$

$$y'_3 + y'_4 = 2 \cdot \frac{12}{11} < 3 \leq w_2$$

$$y'_2 + y'_3 + y'_4 = \frac{36}{11} + 2 \cdot \frac{12}{11} < 6 \leq w_3$$



Is the upper bound of H_n tight?

If it is, the matching lower bound must come from an instance with

- one set containing all elements
(follows from the upper bound of H_g)
- only one additional element covered in each it.
(otherwise, some of the terms in $\frac{1}{n} + \frac{1}{n-1} + \dots + 1$ would be replaced by smaller terms.)

Ex:

