

18) remember
K-set covering problem defined in ex: 15.

Formulation:

$$\text{Let } X_S \begin{cases} 1 & \text{if set } S \text{ belongs to the cover} \\ 0 & \text{otherwise} \end{cases}$$

$$\text{minimize } \sum_{S \in \mathcal{S}} X_S \cdot c(S)$$

$$\text{restrictions: for all } e \in U: \sum_{S: e \in S} X_S \geq 1$$

$$\text{in IP} \rightarrow X_S \in \{0, 1\}$$

$$\text{in LP} \rightarrow X_S \in [0, 1]$$

Notation:

X_β^i : LP optimum solution in iteration β

$X_{\beta i}^i$: Value of the variable indicating if set i is in the set cover or not, in X_β^i .
(LP)
(iteration β)

OUT: output of the algorithm

IPOPT: optimum solution in IP

algorithm

while $U \neq \emptyset$ do

compute optimum basic LP solution x^i

Choose i with $X_i^i \geq \frac{1}{K}$

Buy set S_i , delete elements in S_i from instance

end while

output bought sets

* In each of the iterations:

$$c(X_\beta^i) - c(X_{\beta+1}^i) \geq c(S_{\beta i}) \cdot \frac{1}{K},$$

because a solution with the same values can always be constructed, since we have removed all elements in $S_{\beta i}$. The new solution $X_{\beta+1}^i$ will have at most cost $c(X_\beta^i) - \frac{c(S_{\beta i})}{K}$.

* At the end of the algorithm with m iterations:

$$\sum_{\beta=0}^m \underbrace{(c(X_\beta^i) - c(X_{\beta+1}^i))}_{\text{telescoping sum}} \geq \frac{\sum c(S_{\beta i})}{K} \Rightarrow c(X_1^i) - c(X_m^i) \geq \frac{c(\text{OUT})}{K} \Rightarrow$$

$$\Rightarrow c(\text{IPOPT}) \geq c(X_1^i) \geq c(X_1^i) - c(X_m^i) \geq \frac{c(\text{OUT})}{K} \Rightarrow \boxed{c(\text{IPOPT}) \geq \frac{c(\text{OUT})}{K}}$$

Therefore, this algorithm gives a K -approximation. $\ddot{\smile}$