

1 Randomized Rounding

1.1 Set Cover

Given is $E = \{e_1, \dots, e_n\}$ set of elements, S_1, \dots, S_m with $S_i \subset E \forall 1 \leq i \leq m$ and weights $w_j \geq 0 \forall j \in [m]$.

We want to find $I \subset \{1, \dots, m\}$ such that $\sum_{j \in I} w_j$ is minimized subject to $\bigcup_{j \in I} S_j = E$.

Integer Program

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

Solve the relaxation of the IP you get the optimal solution X^* for the LP. Next we want to round every component $x_j \mapsto \begin{cases} 1 \\ 0 \end{cases}$ To gain an integer solution.

We know from the relaxed condition, that $0 \leq x_j^* \leq 1$ so we will set $P[S_j \text{ in the solution}] = x_j^*$.

Now $X_j = \begin{cases} 1 & S_j \text{ in the solution} \\ 0 & \text{otherwise} \end{cases}$ is the random variable that indicates, whether S_j is in the solution.

With this we can calculate the expected cost of the integer program.

$$\begin{aligned} E\left[\sum_{j=1}^m w_j x_j\right] &= \sum_{j=1}^m w_j P[x_j = 1] \\ &= \sum_{j=1}^m w_j x_j^* \\ &= Z_{LP}^* \leq Z_{IP}^* = OPT \end{aligned}$$

This looks too good to be true and our suspicions are fulfilled, because with this simple rounding we might not have a feasible solution.

We now look at the probability with which an element is not covered.

$$\begin{aligned}
\forall e_i : P[e_i \text{ not covered}] &= \prod_{j: e_i \in S_j} (1 - x_j^*) \\
&\leq \prod_{j: e_i \in S_j} e^{-x_j^*} \\
&= e^{-\sum_{j: e_i \in S_j} x_j^*} \\
&\stackrel{LP \text{ constraint}}{\geq} e^{-1}
\end{aligned}$$

With this error we will now construct an algorithm, that satisfies the following conditions

- The Failure probability has to be less than n^{-c} with c constant.
- We will run the algorithm many times, until it is satisfied.
- The constant has to be great enough, that the expected times of iterations is small.

Solution Idea:

Toss a coin for $S_j \quad \forall j$ more than once.

To be precise, we will toss the coin $c \cdot \log n$ times. The probabilities will be the same as in the first rounding.

Put S_j in the solution if the coin comes up heads AT LEAST once
 $P[X_j = 1] = (1 - x_j)^{c \cdot \log n}$ so the probability, that an edge was not covered is

$$\begin{aligned}
P[e_i \text{ not covered}] &= \prod_{j: e_i \in S_j} (1 - x_j)^{c \cdot \log n} \\
&\leq \prod_{j: e_i \in S_j} e^{-x_j^* (c \cdot \log n)} \\
&= e^{-(c \cdot \log n) \sum x_j^*} \\
&\leq \frac{1}{n^c}
\end{aligned}$$

Now we have to merge all the probabilities, but keep in mind, that they are not independent

$$\begin{aligned}
P[\exists e_t \text{ not covered}] &\leq \sum_{i=1}^n P[e_i \text{ not covered}] \\
&\leq n \cdot \frac{1}{n^c} \\
&\leq \frac{1}{n^{c-1}}
\end{aligned}$$

Theorem 1. \mathcal{A} is a (randomized) $O(\log n)$ -approximation algorithm and produces a set-cover with high probability.

Proof 1:

To prove this, we will first show a slightly stronger claim on the approximation factor:

If \mathcal{A} returns a set-cover, that the approximation-factor is $O(\log n)$.

$$\begin{aligned} x_j^* &\in [0, 1], \quad c \log n \geq 1 \\ p_j'(x_j) &= (c \log n)(1 - x_j^*)^{c \log n - 1} \leq c \log n \\ p_j(0) = 0 &\Rightarrow p_j(x_j^*) \leq (c \log n)x_j^* \end{aligned}$$

$$\begin{aligned} E\left[\sum_{j=1}^m w_j x_j\right] &= \sum_{j=1}^m w_j P[X_j = 1] \\ &\leq \sum_{j=1}^m w_j (c \log n) x_j^* \\ &= c \log n \cdot z_{LP}^* \\ &\leq OPT \end{aligned}$$

Next we have to proof the more general case.

A_1, \dots, A_n disjoined events that form a partition fo the sample space:

$$\begin{aligned} E[X] &= \sum_i^n P(A_i) \cdot E[X | A_i] \\ &= \sum_i^n x P(X=A(x)) \\ &= \frac{\sum_x x P(\{X=x\} \cap A)}{P(A)} \end{aligned}$$

Let F be events, when the solution is feasale and \bar{F} the complement of F in sample space.

And we conclude the probability $P[F] \geq 1 - \frac{1}{n^{c-1}}$

$$E\left[\sum_{j=1}^m w_j X_j\right] = E\left[\sum_{j=1}^m w_j X_j | F\right] \cdot P[F] + E\left[\sum_{j=1}^m w_j X_j | \bar{F}\right] \cdot P[\bar{F}]$$

$$\begin{aligned} E[\sum | F] &\geq \frac{1}{P[\bar{F}]} (E[\sum] - E[\sum | \bar{F}] P[\bar{F}]) \\ &\leq \frac{1}{P[\bar{F}]} E[\sum w_j x_j] \\ &\leq \frac{c \log n z_{LP}^*}{1 - \frac{1}{n^{c-1}}} \\ &\leq 2c(\log n) z_{LP}^* \leq 2c(\log n) OPT \end{aligned}$$

□