
Algorithm Design Second Homework

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EXERCISE 1

We redefine our problem as a problem on graphs. Let $G(M, F, E)$ a fully connected graph in which each friend is a node. Each edge (u, v) has a cost $w(u, v) \geq 0$. We define the following functions:

$$\deg(v, G) := \sum_{u \in G.M \cup G.F} w(v, u);$$

$$\text{density}(G) := \frac{1}{|G.M \cup G.F|} \sum_{v, u \in G.M \cup G.F} w(v, u);$$

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1 algorithm michele_party_approx( $G$ )
2    $n := |G.M \cup G.F|$ 
3    $H_n := G$ 
4    $\text{max\_d} := -1$ 
5    $s := -1$ 
6   for  $i = \frac{n}{2}$  to 1
7      $d := \text{density}(H_i)$ 
8     if  $d > \text{max\_d}$ 
9        $\text{max\_d} = d$ 
10       $s = i$ 
11       $m := v \in H_i.M \mid \deg(v, H_i) \leq \deg(u, H_i) \forall u \in H_i.M$ 
12       $f := v \in H_i.F \mid \deg(v, H_i) \leq \deg(u, H_i) \forall u \in H_i.F$ 
13       $H_{i-1} = H_i \setminus \{m, f\}$ 
14 return  $H_s$ 
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Let OPT the optimal solution and $d_{OPT} = \frac{1}{|OPT|} \sum_{v, u \in OPT} w(v, u) = \frac{W}{N}$ its density. Let $m \in OPT \cap M$ and $f \in OPT \cap F$.

Consider that $\frac{W}{N} \geq \frac{W - \deg(m) - \deg(f)}{N-2} \Leftrightarrow \frac{\deg(m) + \deg(f)}{N-2} \geq \frac{W}{N-2} - \frac{W}{N} \Leftrightarrow \deg(m) + \deg(f) \geq W - \frac{N-2}{N} * W = 2 * \frac{W}{N}$. So in OPT the sum of the degree of a generic m and f is at least twice d_{OPT} .

Consider the iteration i of the algorithm in which for the first time two nodes $m \in OPT \cap M$ and $f \in OPT \cap F$ are removed. In the remaining graph all the pairs $m \in H_{i-1} \cap M$ and $f \in H_{i-1} \cap F$ have $\deg(m) + \deg(f) \geq 2 * d_{OPT}$ thanks to $\min(a, b) = \min(a) + \min(b)$ and the previous observation. In H_{i-1} there are $\frac{|H_{i-1}|}{2}$ pairs that can be considered in the following iterations until $i = 1$. The total cost of the edges in H_{i-1} is greater than $\frac{2 * d_{OPT} * \frac{|H_{i-1}|}{2}}{2}$ (the division by 2 is due to the fact that we want to avoid to consider edges twice). The density is greater than $\frac{2 * d_{OPT} * \frac{|H_{i-1}|}{2}}{2 * |H_{i-1}|} = \frac{d_{OPT}}{2}$. Since the algorithm returns the graph with the highest density over all the iterations we have a solution with density at least $\frac{d_{OPT}}{2}$. We proved that *michele_party_approx* is a 2-approximation.

EXERCISE 2

Our solution was inspired by the Set Cover approximation proof described in [1] but follows the path of the proof shown during the class.

Let A is the set of required skills. S is the set of all the people available, each people is represented as a set of skills $S_j \subseteq A$. Let $n = |A|$.

We can express the Set Cover with Redundancies problem using the following ILP formulation:

$$\begin{aligned} \min \quad & \sum_{S_j \in S} c_j * x_j \\ \text{s.t.} \quad & \sum_{S_j | A_i \in S_j} x_j \geq 3 \quad \forall A_i \in A \\ & x_j \in \{0, 1\} \quad \forall S_j \in S \end{aligned}$$

In order to build a randomized approximation consider the associated LP relaxation where $x_j^* \in [0, 1]$. The LP solution is a vector x^* of real values.

Consider the algorithm ALG in which each person S_j is chosen randomly with probability $p_j = \min(d * \log(n) * x_j^*, 1)$ with all choices that are independent. We denote the vector of these choices as x' .

Let $C_i = \sum_{S_j | A_i \in S_j} x'_j$ a random variable that represents the times that the skill A_i is covered. The expectation of C_i is $\mathbb{E}[C_i] = \mathbb{E} \left[\sum_{S_j | A_i \in S_j} x'_j \right] = \sum_{S_j | A_i \in S_j} p_j$. Then, $\mathbb{E}[C_i] = d * \log(n) * \sum_{S_j | A_i \in S_j} x_j^* \geq d * \log(n) * 3$ because the cover constraint in LP is satisfied.

From this, thanks to the Chernoff lower bound, follows that:

$$\begin{aligned} Pr[C_i < 3] &= Pr \left[C_i < \left(1 - \left(1 - \frac{1}{d * \log(n)} \right) \right) * d * \log(n) * 3 \right] \leq Pr \left[C_i < \left(1 - \left(1 - \frac{1}{d} \right) \right) * \right. \\ &\quad \left. d * \log(n) * 3 \right] \leq exp \left(- \frac{1}{2} * 3 * d * \log(n) * \left(1 - \frac{1}{d} \right)^2 \right) \end{aligned}$$

Note that $\left(1 - \frac{d \cdot \log(n) - 1}{d \cdot \log(n)}\right) * d * \log(n) = 1$. $0 < 1 - 1/d < 1$ must be valid in order apply Chernoff and so we have that $d > 1$.

We have that the probability that the skill A_i is covered more than 3 times is $Pr[C_i \geq 3] \geq 1 - \exp\left(-\frac{1}{2} * 3 * d * \log(n) * \left(1 - \frac{1}{d}\right)^2\right)$.

The probability that at least one A_i is not covered 3 times is $Pr\left[\bigcup_{A_i \in A} C_i < 3\right] \leq \sum_{A_i \in A} Pr[C_i < 3] = n * \exp\left(-\frac{1}{2} * 3 * d * \log(n) * \left(1 - \frac{1}{d}\right)^2\right)$.

Note that ncreasing d give us a better probability that all skills are covered but also, on the other hand, decrease the probability to have a minimal solution because over a treshhold we have that all the variables in x' are 1.

The expected cost is $\mathbb{E}\left[\sum_{S_j \in S} c_j * x'_j\right] = \sum_{S_j \in S} c_j * \mathbb{E}[x'_j] = \sum_{S_j \in S} c_j * d * \log(n) * x_j^*$ and so it is $d * \log(n)$ times the cost of the LP.

With the Markov inequality we bound with a parameter k the probability to fail on having an average cost less than $k * d * \log(n)$ times the cost of LP.

$$Pr\left[\sum_{S_j \in S} c_j * x'_j \geq k * d * \log(n) * \sum_{S_j \in S} c_j * x_j^*\right] \leq \frac{\sum_{S_j \in S} d * \log(n) * c_j * x_j^*}{\sum_{S_j \in S} k * d * \log(n) * c_j * x_j^*} = \frac{1}{k}$$

Obviously this is an event that we want to avoid and setting a big k help us in that but, on the contrary, give us also a worse approximation factor than $d * \log(n)$.

We must choose d and k in a way that maximize the probability that our solution is valid and the expected cost is in the bound and maximize the probability that such solution is minimal.

Choosing $d = 3$ we have an interesting result:

$$Pr\left[\bigcup_{A_i \in A} C_i < 3\right] = n * \exp(-2 * \log(n)) = 1/n$$

The probability that all the skills are covered, so that x' is feasible, is $1 - 1/n$ that is quite high and so we expect to need only a run to get a feasible solution.

The cost of this approximation must be at most $3 * k * \log(n) * LP$.

The probability that the solution is not feasible or that the cost exceed the bound is $Pr\left[\mathbb{E}\left[\sum_{S_j \in S} c_j * x'_j\right] \geq \sum_{S_j \in S} k * d * \log(n) * x_j^* \vee \bigcup_{A_i \in A} C_i < 3\right] \leq \frac{1}{k} + \frac{1}{n}$.

Setting $k = 2$ we get that such probability is a bit greater than $\frac{1}{2}$ and so the solution is feasible and with cost less than $6 * \log(n) * LP$ with probability a bit less than $\frac{1}{2}$ ($\frac{1}{2} - \frac{1}{2 * n}$). We conclude that, as in the normal set cover approximation, the expected number of repetitions are 2 and we choose d and k with the intention to have such number of repetitions.

EXERCISE 3

We denote as F^* the optimal solution of the problem. F^* is the minimum cost set of edges that if removed creates k connected components with each target terminal s_i inside each component.

Let $F_i^* \subset F^*$ the set of edges that if removed separated the vertex s_i from the other s_j with $i \neq j$. We have that $\cup_{i=0}^k F_i^* = F^*$. Each edge e in F^* is contained in two F_i^* because it is incident at two connected components. So we can say that:

$$\sum_{i=0}^k \sum_{e \in F_i^*} w(e) = 2 * \sum_{e \in F^*} w(e)$$

Our algorithm returns k min cuts F_i . By definition F_i is the minimum set of edges that separates s_i from the other s_j with $i \neq j$ so we have that $\sum_{e \in F_i} w(e) \leq \sum_{e \in F_i^*} w(e)$.

This implies that:

$$\sum_{i=0}^k \sum_{e \in F_i} w(e) \leq 2 * \sum_{e \in F^*} w(e)$$

And so, in the worst case, the cost of F is twice the cost of F^* . We proved that this is a 2-approximation algorithm.

EXERCISE 4

Let k an ordered multiset of genes $g \in G$ such that $g_1 || \dots || g_j || \dots || g_{|k|} = D$ with $g_j \in G$ ($||$ is the concatenation operator). K is the set of all possible k .

As instance, if $D = ACCA$ we can have $K = \{\{AC, CA\}, \{ACC, A\}\}$ if $G = \{AC, CA, ACC, A\}$. Let the variables x_i with $i = 1 \dots m$ binary variables that represents if a gene $G_i \in G$ is in D .

Let $y_k \forall k \in K$ a binary variable that tells if the multiset k is used to form D .

The ILP formulation is the following:

$$\begin{aligned} & \min \sum_{i=1}^m w_i * x_i \\ (1) \quad & \sum_{k \in K | G_i \in k} y_k - x_i \leq 0 \quad \forall i \in \{1 \dots m\} \\ (2) \quad & \sum_{k \in K} y_k \geq 1 \\ & x_i \in \{0, 1\} \quad \forall i \in \{1 \dots m\} \\ & y_k \in \{0, 1\} \quad \forall k \in K \end{aligned}$$

(1) means that if a gene G_i is not taken all the k that contains G_i must be excluded. (2) means that at least one k must be taken, and it is our solution that has minimum cost genes.

Consider now the LP-relaxion with $x_i \geq 0 \wedge y_k \geq 0$. Let's compute the dual of it. In the dual, we have that the number of variables is the same of the number of constraints in the primal ($m + 1$) and vice-versa the number of constraints is the number of variables

in the primal ($m + |K|$). Only constraint (2) contribute to the objective function (cfr. [2]).

When G_i is used the index i is always relative to G .

$$\begin{aligned}
 & \max u \\
 & \sum_{G_i \in K} v_i - u \geq 0 \quad \forall k \in K \\
 & v_i \leq w_i \quad \forall i \in \{1 \dots m\} \\
 & u \geq 0, v_i \geq 0 \quad \forall i \in \{1 \dots m\}
 \end{aligned}$$

EXERCISE 5

The proposed problem is a finite zero-sum game theory problem.

Comet, Dasher	Head	Tail
Head	4, -4	-1, 1
Tail	-2, 2	2, -2

Comet

$\max y$

$$4 * x_1 - 2 * x_2 \geq y$$

$$-x_1 + 2 * x_2 \geq y$$

$$x_1 + x_2 = 1$$

$$x_1 \geq 0, x_2 \geq 0$$

Dasher

$\max v$

$$-4 * u_1 + u_2 + v \leq v$$

$$2 * u_1 - 2 * u_2 + v \leq v$$

$$u_1 + u_2 = 1$$

$$u_1 \geq 0, u_2 \geq 0$$

EXERCISE 6

Let h the position of Giorgio's home. Let $S(i) = Pr[Safe|x(t) = i]$ the probability that Giorgio goes to home safely when he is at position i . We know that $Pr[x(t+1) = x(t) + 1] = p$ and $Pr[x(t+1) = x(t) - 1] = 1 - p$. Follows that:

$$S(i) = \begin{cases} 0 & \text{if } i = -1 \\ 1 & \text{if } i = h \\ p * S(i-1) + (1-p) * S(i+1) & \text{otherwise} \end{cases}$$

We have that in general $S(i) = p * S(i-1) + (1-p) * S(i+1)$ and $S(i) = p * S(i) + (1-p) * S(i)$ (from $1 = p - (1-p)$) and so $S(i+1) = \frac{p}{1-p} * (S(i) - S(i-1)) + S(i)$. Follows that $S(i+2) = \frac{p}{1-p} * (S(i+1) - S(i)) + S(i+1) = \frac{p}{1-p} * \left(\frac{p}{1-p} (S(i) - S(i-1)) + S(i) - S(i) \right) + \frac{p}{1-p} * (S(i) - S(i-1)) + S(i)$.

With $i = 0$ we have $S(2) = \left(\frac{p}{1-p}\right)^2 * S(0) + \frac{p}{1-p} * S(0) + S(0)$ and so clearly:

$$S(i) = \sum_{j=0}^i \left(\frac{p}{1-p} \right)^j * S(0).$$

Consider now when $i = h$. The problem says that Giorgio makes an infinite number of steps so we can set $h = +\infty$.

$$1 = S(+\infty) = S(0) * \sum_{j=0}^{+\infty} \left(\frac{p}{1-p} \right)^j.$$

This is a geometric series [3]. By definition if $\left| \frac{p}{1-p} \right| < 1$ it converges to $\frac{1}{1 - \frac{p}{1-p}}$.

The probability that Giorgio goes to the hospital from position 0 is $Pr[Hospital|x(t) = 0] = 1 + Pr[Safe|x(t) = 0] = 1 - S(0)$.

$\left| \frac{p}{1-p} \right| < 1 \Leftrightarrow p < 1/2$ and so, finally, we have that:

$$Pr[Hospital|x(t) = 0] = 1 - \frac{1}{\sum_{j=0}^{+\infty} \left(\frac{p}{1-p} \right)^j} = \begin{cases} \frac{p}{1-p} & p < 1/2 \\ 1 & p \geq 1/2 \end{cases}$$

Giorgio goes to hospital for sure when $p \geq \frac{1}{2}$ and goes to hospital with probability at most $\frac{1}{2}$ when $\frac{p}{1-p} \leq \frac{1}{2} \Rightarrow p \leq \frac{1}{3}$.

APPENDIX

EXERCISE 5

Yes, AMPL+CPLEX should be used instead of Z3 but we do not like the AMPL syntax and in addition we use Z3 almost every day for SMT/SAT so...

Primal:

```
(declare-fun k () Real)
(declare-fun x2 () Real)
(declare-fun x1 () Real)
(assert (>= (- (* 4.0 x1) (* 2.0 x2)) k))
(assert (>= (+ (- x1) (* 2.0 x2)) k))
(assert (= (+ x1 x2) 1.0))
(assert (>= x1 0.0))
(assert (>= x2 0.0))
(maximize k)
(check-sat)
```

Dual:

REFERENCES

- [1] “Randomized rounding - Wikipedia.” https://en.wikipedia.org/wiki/Randomized_rounding#Proof. Accessed: 2019-1-3.
- [2] “Massimo Roma - Appunti dalla lezione di Ricerca Operativa, chapter 8.” www.dis.uniroma1.it/~roma/didattica/R017-18/cap8.pdf. Accessed: 2019-1-3.
- [3] “Geometric series - Wikipedia.” https://en.wikipedia.org/wiki/Geometric_series#Geometric_power_series. Accessed: 2019-1-3.