

Approximation Algorithms

- NP-hard optimization problems (under the assumption $P \neq NP$) cannot be solved optimally in poly-time
- Relax optimality: solve with a guaranteed approximation

Definitions: An NP optimization problem Π consists of the following:

valid instances: each instance I is like an input

Feasible solution: For each I , $S_{\Pi}(I)$ is the set of all solutions for instance I

objective function: A poly-time computable func. $f(s, I)$ that gives a non-neg value for each solution s for I .

Definition: For a minimization problem Π , an algorithm A is an α -approximation if it is polytime & for each instance I produces a solution $s \in S_{\Pi}(I)$ s.t.

$f(s, I) \leq \alpha(|I|) \cdot \text{OPT}(I)$, where $\text{OPT}(I)$ is the smallest objective func. value among all feasible solutions of I .

if we use $A(I)$ to denote the value of solution returned by A then $A(I) \leq \alpha(|I|) \cdot \text{OPT}(I)$

We define approximation for maximization problems similarly

Asymptotic α -approx: if $\lim_{|I| \rightarrow \infty} \frac{A(I)}{\text{OPT}(I)} \leq \alpha$.

PTAS (polynomial time approximation scheme): is a series of algorithms s.t. for any given $\epsilon > 0$ the algorithm A_{ϵ} (that is dependent on ϵ) runs in time polynomial in n and returns a $(1+\epsilon)$ -approximate solution.

(e.g. the time can be $O(n^{\frac{1}{\epsilon^2}})$ or $O(2^{\frac{1}{\epsilon}} \cdot n^{\frac{1}{\epsilon^{10}}})$)

FPTAS (Fully PTAS): Same as PTAS but the time is polynomial in both n & $\frac{1}{\epsilon}$ (e.g. $O(n^3/\epsilon^5)$).

Set Cover & Max-Coverage

Set Cover universe of n elements

$$U = \{e_1, e_2, \dots, e_n\}$$

$S = \{s_1, \dots, s_m\}$ collection of subsets of U

$$c: S \rightarrow \mathbb{Q}^+$$

Goal: Find a min-cost subcollection of S that covers U .

$$\text{i.e. Find } I \subseteq \{1, \dots, m\} \quad \bigcup_{i \in I} S_i = U \text{ \& \; min } \sum_{i \in I} c(s_i)$$

Def: Frequency of an element e_i is # of sets S_j that contain e_i .

Max-Coverage: input is U, S , and int k

Goal: Find a collection of k sets that max # of elements covered.

Natural Greedy for both S.C. & Max-Coverage

— At each step pick a set that covers the max # of uncovered elements at lowest cost.

— repeat until done: $\left\{ \begin{array}{l} \text{Set Cover: all elements are covered} \\ \text{Max-Coverage: } k \text{ iterations} \end{array} \right.$

Theorem: Greedy is a $(\underbrace{\ln n + 1}_{H_n})$ -approx for Set Cover

and a $1 - (1 - \frac{1}{k})^k \geq 1 - \frac{1}{e}$ approx for max-coverage.

Analysis for max-Coverage

Let opt^* be an opt sol with value opt .

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- let a_i be # of new elements covered at iteration i

and $b_i = \sum_{j=1}^i a_j$ and $\underbrace{C_i = \text{opt} - b_i}_{\text{how far from opt}}$

$$a_0 = b_0 = 0 \quad \& \quad C_0 = \text{opt}$$

Lemma 1: For all $0 \leq i < k$: $a_{i+1} \geq \frac{C_i}{k}$

Proof: Consider iteration $i+1$. Since b_i elements are covered until end of iteration i , C_i elements of opt^* are left uncovered. let Z^* be the elements of opt^* that are not covered yet. $|Z^*| \geq C_i$

Since opt uses k sets, all elements of Z^* are covered (in opt) using k sets. $\rightarrow \exists$ a set that covers $\geq \frac{|Z^*|}{k}$ elements of Z^*

$\rightarrow \exists$ a set in iteration $i+1$ that can be chosen to cover $\geq \frac{|Z^*|}{k} \geq \frac{C_i}{k} \rightarrow a_{i+1} \geq \frac{C_i}{k}$ □

Lemma 2: For all $i \geq 0$: $C_i \leq (1 - \frac{1}{k})^i \cdot \text{opt}$
(so $C_k \leq (1 - \frac{1}{k})^k \cdot \text{opt}$)

proof: by induction on i

- $i=0$ base

- Ind steps

$$\rightarrow b_k \geq \left(1 - \left(1 - \frac{1}{k}\right)^k\right) \text{opt}$$

$$C_{i+1} = C_i - \underbrace{a_{i+1}}$$

$$\leq C_i - \frac{C_i}{k}$$

$$= C_i \left(1 - \frac{1}{k}\right)$$

by lemma 1

$$\leq (1 - \frac{1}{k})^i \cdot \text{opt} \cdot (1 - \frac{1}{k})$$

Ind. Hyp.

$$= (1 - \frac{1}{k})^{i+1} \cdot \text{opt}$$



Greedy SetCover SC1

$C \leftarrow \emptyset$

$S' \leftarrow \emptyset$

while $C \neq U$

select $S_i \in S$ with min cost effectiveness $\alpha = \frac{c(S_i)}{|S_i - C|}$

$S' \leftarrow S' \cup \{S_i\}$

$C \leftarrow C \cup S_i$

[for each newly covered $e \in S_i$ price(e) as α]

$|S_i - C|$
of new elements covered

Lemma : $\sum_{e \in U} \text{price}(e) \leq H_n \cdot \text{opt}$ $H_n = 1 + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \approx \ln n + 1$

order elements of U by the time they are covered by SC1
 $e_1, e_2, \dots, e_n$ (break ties arbit).

Consider the time just before e_k is covered.

So $n - k + 1$ (uncovered) elements can be covered at a total price of opt . So we can pick some sets of total cost opt to cover $n - k + 1$ elements.

→ $\exists$ a set which can cover elements at a price of $\leq \frac{\text{opt}}{n - k + 1}$ So e_k pays no more than this

to be covered.

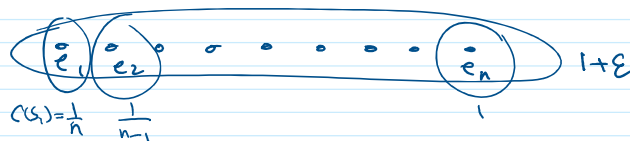
$$\sum_{i=1}^n \text{Price}(e_i) \leq \sum_{i=1}^n \frac{\text{opt}}{n - i + 1} = \text{opt} \cdot \sum_{i=1}^n \frac{1}{n - i + 1} = H_n \cdot \text{opt}$$

Unless $P=NP$ there is no $(1-\epsilon)\ln n$ -approx for



Unless $P=NP$ there is no $(1-\epsilon)\ln n$ -approx for Set cover for any $\epsilon > 0$.

Tight example:



$$\text{opt} = 1+\epsilon$$

$$\text{Greedy} \approx H_n$$

Randomized Rounding: Set Cover

Rounding an LP solution to an integer one while bounding the change in obj value is a common way of designing approximation algorithms.

IP/LP formulation

$$\min \sum c_S \cdot x_S$$

$$\forall e \in U: \sum_{e \in S} x_S \geq 1$$

$$x_S \geq 0$$

LP rounding for sc

Let x^* be an LP opt. solution

let $C \leftarrow \emptyset$

for each set S_i : set $\tilde{x}_{S_i} = 1$ with prob. $x_{S_i}^*$

let $C = \{S_i \mid \tilde{x}_{S_i} = 1\}$

return C

First let's compute the expected cost of C

$$E[\text{cost of } C] = \sum_{S_i \in \mathcal{S}} \Pr[S_i \text{ is selected}] \cdot c_{S_i} = \sum_{S_i} x_{S_i}^* \cdot c_{S_i} = \text{opt}_{LP}$$

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let α be large enough s.t. $(\frac{1}{e})^{\alpha \ln n} \leq \frac{1}{4n}$ and suppose we repeat this algorithm α times.

let C_i be the solution for repetition i and return

$$C = \bigcup_{i=1}^{\alpha \ln n} C_i$$

Easy to see: $E[\text{cost of } C] = \sum_{i=1}^{\alpha} E[\text{cost}(C_i)] = (\alpha \ln n) \text{opt}_{LP}$

We prove that w.h.p. C is a feasible solution.

- Consider some element e_j and suppose it belongs to sets $S_1, \dots, S_k$. So $x_{S_1}^* + x_{S_2}^* + \dots + x_{S_k}^* \geq 1$

Lemma: Given $x_{S_1}^* + x_{S_2}^* + \dots + x_{S_k}^* = 1$ if we select each S_i with prob. $x_{S_i}^*$ the prob. that none is selected is maximized when all $x_{S_i}^*$'s are equal to $\frac{1}{k}$

Proof: Calculus.

In this case: $\Pr[e_j \text{ is not covered}] \leq (1 - \frac{1}{k})^k \leq 1/e$
in one iteration

So over all iterations: $\Pr[e_j \notin \bigcup_{i=1}^{\alpha \ln n} C_i] \leq (\frac{1}{e})^{\alpha \ln n} = \frac{1}{4n}$

Using union bound: $\Pr[C \text{ is not a S.C.}] \leq n \cdot \frac{1}{4n} = \frac{1}{4}$ event E_1

Using Markov's inequality: $\Pr[\text{cost}(C) > 4 \ln n \text{opt}_{LP}] \leq \frac{1}{4}$ event E_2

→ with prob $\geq \frac{1}{2}$ E_1 & E_2 don't happen i.e.

With prob $\geq \frac{1}{2}$ (C is a S.C.) and ($\text{cost}(C) \leq 4 \ln n \cdot \text{opt}_{LP}$).