Approximation algorithms

An algorithm has approximation ratio r if it outputs solutions with cost such that

 $c/c^* \le r$ and $c^*/c \le r$

where c* is the optimal cost.

We focus on ratio (as opposed to difference) because that appears to be more natural for problems of interest

• Problem: Cover edges by vertexes

Input: Graph Output: A minimal set of nodes that touches every edge

Algorithm: While there is an edge (u, v) Add both u and v to your cover. Erase all edges adjacent to either u or v.

• Claim: This is a 2 approximation

• Proof:

Consider the set A of edges picked by the algorithm. Note any cover must have at least one node for each edge, and so size at least |A|. • Problem: Cover edges by weighted vertexes

Input: Graph, weights for vertexes Output: A minimal-cost set of nodes that touches every edge

Formulate problem as integer program: min $\sum x(v) w(v)$: $x(u) + x(v) \ge 1 \forall (u,v) \in E$, $x(u) \in \{0,1\} \forall u \in V$

Integer programs should not be solvable efficiently

• Problem: Cover edges by weighted vertexes

Input: Graph, weights for vertexes Output: A minimal-cost set of nodes that touches every edge

Relax to linear programming min $\sum x(v) w(v)$: $x(u) + x(v) \ge 1 \forall (u,v) \in E$, $x(u) \in [0,1] \forall u \in V$

• Algorithm:

Solve relaxation

Round: Take nodes with $x(u) \ge 1/2$.

Claim: This is a cover. Proof: Because $x(u) + x(v) \ge 1/2$ for every edge (u,v)

Claim: This is a 2 approximation Proof: Let C* be an optimal solution. z be cost of relaxed linear program C be cost of output of algorithm

Obviously, $z \le C^*$ since solution space is bigger

Now note
$$z = \sum x(v) w(v) \ge \sum_{v : x(v) \ge 1/2} w(v) / 2 = C/2.$$

So C/2 $\leq z \leq C^*$

Paradigm:

Believed infeasible Feasible Relaxation

Integer program \rightarrow linear program

Quadratic program \rightarrow vector program

Integral solution

Rounding

←

2-approximation:

How?

2-approximation:

Pick the cut at random. You expect to cut 1/2 of the edges

Possible to do deterministically

We now improve 2 to 1 / 0.87... < 2

Maximize 1/2 $\sum_{(i,j) \in E} 1 - y_i y_j : y_i \in \{-1,1\}$

Relax to vector program:

 $\begin{array}{ll} y_i & \rightarrow \text{vector } v_i \in \mathsf{R}^d \quad (\text{where } d = \text{polynomial in } |V|) \\ y_i \, y_j & \rightarrow \text{inner product} < v_i \,, \, v_j > \\ y_i \in \{-1,1\} \quad \rightarrow |v_i| = 1 \end{array}$

Algorithm:

Solve vector program

Round: Take random vector r of length 1. One side of the cut is $\{i : < v_i, r > \ge 0\}$

Analysis:

Expected size of cut is $\sum_{(i,j)} \Pr[v_i \text{ and } v_j \text{ are separated}]$ = $\sum_{i,j} \theta_{i,j} / \pi$ (lemma) $\geq \alpha \sum_{i,j} (1 - \cos_{i,j}) / 2$ ($\exists \alpha = 0.87...:$ this is true $\forall \theta$) $\geq \alpha \sum_{i,j} (1 - \langle v_i, v_j \rangle) / 2$ ($\langle v_i, v_j \rangle = \cos_{i,j}$)

= α cost of vector program

 $\geq \alpha$ optimal cost

Problem: Cover points by sets

Input: A family of sets over n points. Output: A minimal number of sets that covers every point.

Algorithm:

Greedily pick a set that covers as much as possible of what's left.

Claim: This is a log(n) approximation Proof:

Fix an execution of the algorithm: $(S_1, S_2, ...,)$

S_i is the i-th set picked by algorithm.

Given this, for each element x, define cost

 $c_x := 1/\#$ of new elements covered by set that covers x first

= (if S_i covers x first) 1/| $S_i - U_{j < i} S_j$ |

Note cost of algorithm $|C| = \sum_{x} c_{x}$

Also, let C* be optimal.

Have $|C| \leq \sum_{|S| \in |C^*} \sum_{|x| \in |S|} c_x$, since every point is covered

We will show $\forall S, \sum_{x \in S} c_x \leq O(\log n)$,

yielding $|C| \leq O(|C^*| \log n)$.

Claim: $\forall S, \sum_{x \in S} c_x \leq O(\log n)$,

Proof: Fix S. $u_i := #$ elements in S uncovered after i-th iteration of algorithm = |S - $U_{j \le i} S_j$ |

u₀ = |S|

Let k be the first such that $u_k = 0$.

Note u is decreasing, $u_{i\mbox{-}1}$ - u_i is # elements in S covered first time by S_i .

$$\begin{split} \sum_{x \in S} c_x &= \sum_{1 \le i \le k} (u_{i-1} - u_i) / |S_i - U_{j < i} S_j| \\ &\leq \sum_{1 \le i \le k} (u_{i-1} - u_i) / |S_i - U_{j < i} S_j| \quad (\text{greedy choice}) \\ &= \sum_{1 \le i \le k} (u_{i-1} - u_i) / u_{i-1} \\ &= \sum_{1 \le i \le k, \ 1 + ui \le j \le u(i-1)} \frac{1}{u_j} \\ &= \sum_{1 + uk \le i \le u0} 1 / i = O(H(u_0)) = O(H(|S|)) = O(\log |S|) \end{split}$$

Problem: Given n numbers x_1 , x_2 , ..., x_n integer t, compute maximum size of subset of numbers not exceeding t

This problem has fully polynomial-time approximation algorithm: in time poly(n,1/ ϵ) finds a sum that does not exceed t and is within 1+ ϵ of largest not exceeding t.

Naive approach: $L_0 = \emptyset$ For every i: $L_{i+1} = L_i + x_i$; Remove elements bigger than t Return Max in L_n

Problem ?

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Naive approach:

$$L_0 = \emptyset$$

For every i: $L_{i+1} = L_i + x_i$; Remove elements bigger than t Return Max in L_n

Problem, list gets too big.

For approximation, don't keep elements close to each other.

Trim(L, δ) : Go through elements in L in sorted order. Add element y in L $\leftarrow \rightarrow$ bigger than 1 + δ of what you have already

Approximation algorithm($x_1, ..., x_n, t, \epsilon$)

 $L_0 = \emptyset$

For every i: $L_{i+1} = L_i + x_i$

Trim(L_{i+1}, ϵ /2n)

Remove elements bigger than t Return Max in L_n

• Correctness:

Claim: Let P_i be set of possible sums of first i elements $\forall y \in P_i \ \exists z \in L_i : y/(1 + \epsilon/2n)^i \le z \le y$

i.e., $\forall y \exists a close lower bound z$

Proof by induction. Won't see

Given claim, easy to see algorithm gives an ε approximation.

• Running time: We bound length of lists. Let $\delta = \epsilon / 2n$ By construction $|L_i| \le \log_{1+\delta} t$ $= O(\log t / \delta)$ $= O(n / \epsilon) \log t$