

Lecture 8 (May 18): Parametric Pruning for Metric k -Center

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8.1 Introduction

Definition 1 (metric) Metric on set X is a function $d : X \times X \rightarrow \mathbb{R}_{\geq 0}$ for which all of the following holds for all $x, y, z \in X$:

1. $d(x, x) = 0$
2. $d(x, y) = d(y, x)$
3. $d(x, z) \leq d(x, y) + d(y, z)$

Example 1 (Metric k -center Problem)

- *Input:*
 - a complete graph $G = (V, E)$
 - a metric cost function $c : E \rightarrow \mathbb{Q}_{\geq 0}$
 - a parameter $k \in \mathbb{N}_{>0}$
- *Output:*
 - A set $S^* \subset V$ with $|S^*| \leq k$ and that minimizes $\max_{v \in V} c(v, S^*)$

In the above definition of the problem, $c(v, S) \triangleq \min_{u \in S} c(v, u)$.

8.2 Metric k -center

Target of this section is to produce a 2-approximation algorithm for the METRIC k -CENTER problem.

Start the analysis of the problem by noting that since $OPT = c(v, u)$ for some distinct $v, u \in V$, we have that $OPT \in \{c(u, v) \mid (u, v) \in E\}$. Because G is a complete graph, i.e. $|E| = \binom{|V|}{2}$, this then shows that OPT is one of $|E| = \binom{|V|}{2}$ values.

Label the edges of G by $1, 2, \dots, |E| = m$ and order them such that

$$c(e_1) \leq c(e_2) \leq \dots \leq c(e_m)$$

and let j be such that $c(e_j) = OPT$. For any $i \in \{1, 2, \dots, m\}$, let $G_i = \{V, E_i\}$ be a graph with $E_i = \{e \in E : c(e) \leq c(e_j)\}$.

Definition 2 (dominating set) For graph $G = (V, E)$, set $D \subset V$ is a dominating set iff for all $v \in V \setminus D$, there is a $u \in D$ such that $(v, u) \in E$.

Note that for G_j , with $c(e_j) = OPT$, any k -element dominating set D is a solution to the METRIC k -CENTER problem in G . This turns the problem into the Dominating Set problem:

Example 2 (Dominating Set Problem)

- *Input:*
 - a graph $G = (V, E)$
 - a parameter $k \in \mathbb{N}_{>0}$
- *Output:*
 - A dominating set $D \subset V$ with $|D| \leq k$.

Unfortunately, this problem is NP-hard. Assuming $P \neq NP$, it is also known to be hard to approximate with ratio better than $\log(|V|)$. Therefore we want to "relax" our definition of dominating set. This is done through the concept of a square graph.

Definition 3 (square graph) Square of graph $G = (V, E)$ is the graph $G^2 = (V, E')$, where $E' = \{(u, v) \in V \times V : d(u, v) \leq 2\}$.

Definition 4 (independent set) For graph $G = (V, E)$, a set $I \subset V$ is independent iff for all distinct $u, v \in I$, $(u, v) \notin E$.

A simple observation about independent sets is that any maximal independent set is also a dominating set. This is because if for independent set I there are no vertices that can be added to I , then every vertex outside the independent set must be adjacent to some vertex in I .

Maximal independent set can also be found greedily by the following algorithm:

Algorithm 1 Greedy for maximal independent set

Input: Graph $G = (V, E)$

Output: A maximal independent set $I \subset V$

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 $I \leftarrow \emptyset$ 
while  $V \neq \emptyset$  do
  choose  $v \in V$ 
   $I \leftarrow I \cup \{v\}$ 
  remove all neighbors of  $v$  from  $V$ 
end while
return  $I$ 

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Algorithm 2 A 2-approximation for METRIC k -CENTER PROBLEM via maximal matching

Input: Graph $G = (V, E)$, metric $c : V \rightarrow \mathbb{Q}_{\geq 0}$ **Output:** A set $S^* \subseteq V$ sort edges such that $c(e_1) \leq c(e_2) \leq \dots \leq c(e_m)$ **for** $i = 1$ to m **do** create G_i^2 $I_i \leftarrow$ maximal independent set of G_i^2 **if** $|I_i| \leq k$ **then** **return** $S^* \leftarrow I_i$ **end if****end for**

Claim 1 For maximal independent set I_i of G_i^2 , and minimum size dominating set D_i of G_i , $|I_i| \leq |D_i|$

Proof. For the sake of contradiction, assume $|I_i| > |D_i|$. Let $D_i = \{v_1, v_2, \dots, v_d\}$. for each $v_j \in D_i$, consider its closed neighborhood $N_{G_i}[v_j]$. D_i is a dominating set so vertex in V is in some $N_{G_i}[v_j]$. By the pigeonhole principle, because $|I_i| \geq |D_i|$, there are two vertices $u, v \in I_i$ that are in the same closed neighborhood $N_{G_i}[v_j]$. uv_jv is a path on G , so $d(u, v) \leq 2$. Then u, v are adjacent in G_i^2 , which contradicts I_i being an independent set. ■

For rest of this section, let $i, j \in \mathbb{N}$ be such that I_i is the output of algorithm 2 and $c(e_j) = OPT$.

Claim 2 $c(e_i) \leq c(e_j) = OPT$

Proof. Proof by contradiction. Assume $c(e_i) > c(e_j)$, i.e. $j < i$. Then, $|I_j| > k$. The optimal solution S_{OPT}^* is a dominating set of G_j . By claim 1 this then implies that $|S_{OPT}^*| \geq |I_j| \geq k$, which contradicts the feasibility of S_{OPT}^* . ■

Combining these claims gives us the following theorem.

Theorem 1 Algorithm 2 is 2-approximation for METRIC k -CENTER.

Proof. I_i is a dominating set of G_i^2 and therefore for all $v \in V \setminus I_i$ and some $u \in I_i$, in G_i we have that $d(u, v) \leq 2$. If $d(u, v) = 1$, $c(u, v) \leq c(e_i) \leq OPT$. If $d(u, v) = 2$, then there is some vertex $w \in W$ such that uwv is a path in G_i , i.e. $c(u, w) \leq OPT$ and $c(w, v) \leq OPT$. c is a metric and therefore $c(u, v) \leq c(u, w) + c(w, v) \leq 2 \cdot OPT$. This shows that

$$\max_{v \in V} c(v, I_i) \leq 2 \cdot OPT.$$

■

This bound is tight as can be seen from the following example.

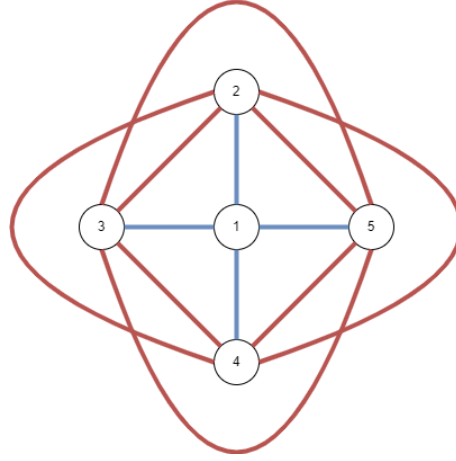


Figure 8.1: Blue edges have weight 1 and red edges weight 2

Optimal solution for the graph in figure 8.1 and any $k \geq 1$ is $S^* = \{v_1\}$ with $OPT = 1$. However, during the first iteration of the algorithm, $I_1 = \{v_5\}$ would be a possible output with $d(v_2, v_5) = 2$, i.e. $ALG = 2$.

Theorem 2 *Assuming $NP \neq P$, for any $\epsilon > 0$, there does not exist a $(2 - \epsilon)$ -approximation for the metric k -center problem.*

Proof. We use the fact that the dominating set problem does not have a $(2 - \epsilon)$ -approximation. Assume that there is a algorithm that produces a $(2 - \epsilon)$ -approximation for the metric k -center problem.

Let $G = (V, E)$ be some graph and let $G' = (V, E \cup E')$ be a complete graph. Further let $c(u, v) = 1$ if $(u, v) \in E$ and $c(u, v) = 2$ otherwise. c satisfies the triangle equality, since $c(u, w) + c(w, v) \geq 2$ for all distinct $u, w, v \in V$. If G has a dominating set of size k , the algorithm will return a solution for G' with cost smaller than 2. Otherwise the algorithm will produce a solution with cost larger or equal to 2. Thus the algorithm can be used to deduce whether G has a dominating set of size k , which is known to be not possible in polynomial time. ■

8.3 Metric Weighted-Center

In this chapter we consider a generalisation of the metric k -center problem and produce a 3-approximation algorithm for the generalisation.

Example 3 (Metric weighted-center Problem)

- *Input:*
 - a complete graph $G = (V, E)$
 - a metric cost function $c : E \rightarrow \mathbb{Q}_{\geq 0}$
 - a vertex weight function $w : V \rightarrow \mathbb{Q}_{\geq 0}$
 - a parameter $W \in \mathbb{Q}_{> 0}$

- *Output:*

- A set $S \subset V$ that satisfies $\sum_{v \in S} w(v) \leq W$ and that minimizes $\max_{v \in V} c(v, S)$

Definition 5 (closed neighborhood) In graph $G = (V, E)$, the closed neighborhood of $u \in V$ is $N_G[u] = N_G(u) \cup \{u\}$.

Definition 6 For graph $G_i = (V, E)$, let $s_i(u)$ denote the lightest vertex in $N_{G_i}[u]$.

Algorithm 3 A 3-approximation for METRIC WEIGHTED-CENTER PROBLEM

Input: Graph $G = (V, E)$, metric $c : V \rightarrow \mathbb{Q}_{\geq 0}$, weight function $w : V \rightarrow \mathbb{Q}_{\geq 0}$, parameter $W \in \mathbb{Q}_{\geq 0}$.

Output: A set $S \subseteq V$

sort edges such that $c(e_1) \leq c(e_2) \leq \dots \leq c(e_m)$

for $i = 1$ to m **do**

 create G_i^2

$I_i \leftarrow$ maximal independent set in G_i^2

$S_j \rightarrow \{s_i(u) : u \in I_i\}$

if $w(S_i) \leq W$ **then**

return $S \leftarrow S_j$

end if

end for

Theorem 3 Algorithm 3 is 3-approximation.

Proof. Let j be such that $c(e_j) = OPT$ and i be such that S_i is the output of the algorithm. Let D_j be the optimal solution to the problem. As we had in the previous section, for all $v \in V$, $c(v, I_i) \leq 2 \cdot OPT$. Also, $c(s_i(u), u) \leq c(e_i) \leq c(e_j)$. Therefore $c(v, s_i(u)) \leq 3 \cdot OPT$ and since this holds for all $v \in V$, $ALG \leq 3 \cdot OPT$. ■