

# 15-451 Assignment 07

Karan Sikka

ksikka@cmu.edu

Collaborated with Dave Cummings and Sandeep Rao.

Recitation: A

November 7, 2014

---

## 1: A Densely-Knit Community

---

(a) Consider the following assignment of values to  $x$ :

Let  $x_i = 0$  if  $v_i$  not in  $S^*$

Let  $x_i = \frac{1}{|S^*|}$  if  $v_i$  is in  $S^*$

Let  $x_{ij} = \min(x_i, x_j)$

Note that the constraints are satisfied. All values are at least 0. The sum of the  $x_i$  is 1 since it's  $|S^*| * \frac{1}{|S^*|}$ .

$$\text{OPT}_{LP} = \sum_{i,j \in E} x_{ij}$$

(b)

(c)

(d)

---

## 2: Large + Dense = Difficult

---

The problem is in NP because there exists a poly-time verifier as follows:

Given a  $(G, K, D)$  and a potential solution  $S$  of size  $K$ , compute the density which is

$$\frac{|\text{edges}(S)|}{|S|}$$

Note: Computing edges of  $S$  can be done in  $|S|^2$  time.

The problem is NP-hard because we can show a Karp reduction from K-Clique to it.

Given  $G, K$  we want to output YES if there exists a K-clique in  $G$ .

Observe that the number of edges in a K-clique is  $\frac{K(K-1)}{2}$  because you can pair each vertex with each other vertex and divide by 2 to eliminate the same pair written backwards.

Therefore the density of a K-clique is

$$\frac{\frac{K(K-1)}{2}}{K} = \frac{K-1}{2}$$

Also, a graph that is not a  $K$ -clique has a strictly smaller density because you would have to add edges to make it a  $K$ -clique, thereby increasing the density.

Therefore if a graph has density equal to  $\frac{|V|-1}{2}$  it must be a  $K$ -clique.

We construct an input to the Reasonably Sized problem as follows:

Let  $D = \frac{K(K-1)}{2}$  Let  $G$  be the same graph as in the  $K$ -clique problem. Let  $K$  be the same  $K$ .

This correctly outputs YES when there is a  $K$ -clique and NO when there isn't because of the way we set  $D$ .

Since the problem is NP and NP-hard, it's NP-Complete.

### 3: A Well-Separated Problem

(a) The problem is in NP because there exists a poly-time verifier as follows:

Define the proof of the solution as a  $K$ -element subset of  $X$ . We compute distances between each pair of distinct points and check that they are greater than or equal to  $\Delta$ . If all pairs satisfy the condition, then we verify that this is a solution. Otherwise it is not.

The Well-Separated problem is in NP-Hard because there is a Karp Reduction from the Independent Set decision problem which is NP-hard to this problem. The reduction is as follows:

#### Independent set

Given a graph  $G = (V, E)$  and integer  $k$ ,  
we want to output YES  
if there exists a set of vertices of size  $k$   
such that no two of them are adjacent.

To craft our input to the Well-Separated oracle,  
we construct a set  $X$  from the vertices of  $G$ ,  
let  $K = k$ ,  
let  $\Delta = 1.25$ ,  
for all  $i \in V$  let  $d(i, i) = 0$ ,  
for all  $i, j \in V : i \neq j \wedge (i, j) \in E$  let  $d(i, j) = 1$   
for all  $i, j \in V : i \neq j \wedge (i, j) \notin E$  let  $d(i, j) = 1.5$

We pass this input to the well separated problem, which will return YES  
if there exists a set of elements of size  $K$  where all distances are greater than 1.25.

Observe these elements in  $X$  map directly to vertices in  $V$  which are not adjacent due to the way we constructed the input to the Well-Separated problem.

Also note that the construction of  $d$  correctly obeys the triangle inequality, because two of the shortest distances ( $1 + 1$ ) is still greater than the longest distance (1.5).

(b) We will present the algorithm, prove a condition about it's correctness, and show that it runs in poly time.

#### Algorithm:

Call one set with separation at least  $\Delta^*/2$  set  $C$ .

We will maintain a vector of all points which are potentially in  $C$ , initially containing all points.

We will also maintain a vector of points which we know are in  $C$ .

1. Pick a point  $u$  potentially in  $C$ , and examine how far away all other points are from it.
2. Add  $u$  to the set of known points  $C$ .
3. Remove  $u$  from the set of points potentially in  $C$ .
4. Remove all points not at least  $\Delta^*/2$  from  $u$  from the set of points potentially in  $C$ .

Repeat for a total of  $K$  iterations,  
resulting in a set  $C$  with  $K$  points at least  $\Delta^*/2$  away from each other  
since at each step we eliminate points which could invalidate this invariant.

Now we need a proof that we can perform this  $K$  times, or in other words, we never run out of points potentially in  $C$  from which to select at each iteration.

**Proof:**

We were allowed to assume there exists some set of  $K$  points with separation  $\Delta^*$ .  
For convenience, let's call these the optimal points.

Lets examine how many optimal points are eliminated from the potentially-in- $C$  set in one iteration of the algorithm.

Claim: At most one optimal point is removed from the potentially-in- $C$  set.

If we select optimal point  $u$ , we will see that all the other optimal points are at least  $\Delta^*$  away from  $u$ , and they will not be eliminated from the potentially-in- $C$  set in this iteration.  $u$  will be the only optimal point removed.

If we select non-optimal point  $u$ , we will see that at most one optimal point is less than  $\Delta^*/2$  away from  $u$ .

This due to the triangle inequality. Consider a nonoptimal point  $u$ , the nearest optimal point  $v$ , and any other optimal point  $w$ .

$$\Delta^* \leq d(v, w) \leq d(v, u) + d(u, w)$$

Since  $v$  no farther from  $u$  than  $w$ ,  $d(v, u)$  may be less than  $\Delta^*/2$ , but  $d(u, w)$  will certainly be at least  $\Delta^*/2$ .

Therefore at most one optimal point is removed from the potentially-in- $C$  set in each iteration of the algorithm.

We know there were  $K$  optimal points to start out with in the potentially-in- $C$  set.

Therefore the algorithm can be run at least  $K$  iterations before running out of points to select.

**Runtime:**

The algorithm runs  $K$  iterations, each iteration takes  $O(|X|)$  work, so the runtime is  $O(K * |X|)$ .

(c) You could run the algorithm on all possible values of  $\Delta^*$ . More formally:

Modify the algorithm from B to return NONE if it runs out of points potentially in  $C$  before running  $K$  iterations.

Observe that the optimum separation delta may only be one of  $\binom{|X|}{2} \leq |X|^2$  values: the distances  $d(i, j) : i \neq j$ .

Sort the values from highest to lowest, and run the algorithm with these values.

Return the none-NONE answer corresponding to the input with highest  $\Delta^*$

Sorting takes  $O(|X|^2 * \log(|X|^2))$  and the rest takes  $O(K * |X| * |X|^2)$ . The limiting step is  $O(K * |X|^3)$

Note: you could do this faster by using binary search instead of linear search, but it doesn't matter for the sake of this problem.