

## Lecture 17

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## 1 Overview

In this lecture, we will study approximation algorithms for problems related to range spaces, especially the hitting set problem.

## 2 Range Spaces and the Hitting Set Problem

### 2.1 Range spaces

A range space is also called a set system or a hypergraph. Its definition is given as follows.

**Definition 1.** A range space  $\Sigma$  is a pair  $\Sigma = (X, R)$ , where  $X$  is a finite set of objects and the ranges  $R$  is a set of subsets of  $X$ .

For example, suppose we have  $X = \{1, 2, 3\}$  and  $R = \{\{1\}, \{2, 3\}, \{1, 2, 3\}\}$ , then  $\Sigma = (X, R)$  is a range space. It shall be noted that  $R$  does not need to include all the subsets of  $X$ . Yet, when  $R$  is indeed the collection of all the subsets of  $X$ , the size of  $R$  is exponential, namely

$$|R| = 2^n,$$

where we denote  $n = |X|$ .

In this lecture, we will instead study the range spaces that are *well-behaved*, which means the size of  $R$  is polynomial, namely

$$|R| \leq n^{O(1)}.$$

The linear classification for points in the  $\mathbb{R}^2$  is such an example.

**Example 1.** Let us suppose we have  $X$  as a finite set of points in  $\mathbb{R}^2$ , which can be imagined as two-dimensional data points drawn on a plane. For every line we draw in the plane, it will cut the space into two half-spaces. We can define the ranges  $R$  to be the set of points in a half-space for all possible lines, i.e.

$$R = \{\gamma \cap X \mid \gamma \text{ is a half-space}\}.$$

Then we will show  $|R|$  is polynomial instead of exponential.

To count the size of  $|R|$ , as shown in Figure 1, one can consider moving and rotating the line  $l_1$  while the classification does not change, until it is in the position of  $l_2$ , which passes through two points. That is to say, such a line will be the boundary that separates one classification from another, corresponding to a subset from another subset in  $R$ . By counting all these boundary lines, we know  $|R| = 2 \times \binom{n}{2} = O(n^2)$ . The same trick can be applied to other definitions of  $R$ . For example, if we let  $R$  to be  $R = \{\gamma \cap X \mid \gamma \text{ is a disk (circle)}\}$ , then we know  $|R| = O(n^3)$  because three points can uniquely determine a circle.

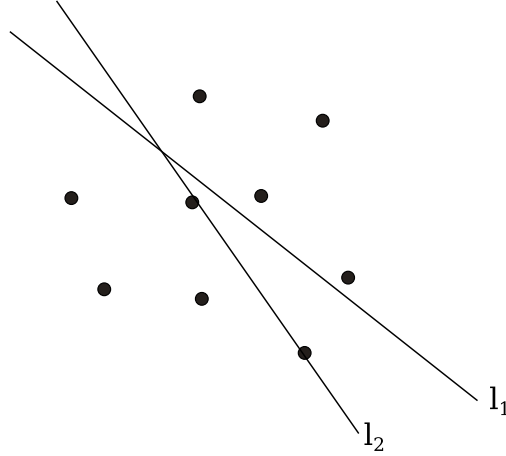


Figure 1: Moving and rotating  $l_1$  to  $l_2$  until it passes through two points.

## 2.2 The hitting set problem

**Definition 2.** For a range space  $\Sigma = (X, R)$ , a hitting set is a subset  $H \subseteq X$  such that  $\forall \gamma \in R$ , it holds that  $H \cap \gamma \neq \emptyset$ .

That is, a hitting set  $H$  “hits” every range by having at least one of its element appearing in every range. A hitting set problem can be formulated as a 0-1 linear programming. If we define  $z_i$  for every element  $x_i \in X$ , where  $z_i = 1$  denotes  $x_i$  is included by  $H$  and  $z_i = 0$  otherwise. The LP can be written as below.

$$\begin{aligned} & \min \sum_{i=1}^n z_i \\ & s.t. \sum_{x_i \in \gamma} z_i \geq 1 \quad \forall \gamma \in R, \\ & z_i \in \{0, 1\}. \end{aligned}$$

Next we introduce  $\varepsilon$ -net as a relaxed version of the hitting set. Instead of requiring the set to “hit” every range, an  $\varepsilon$ -net is only required to hit the ranges that are “heavy” enough. To measure the heaviness, we can use a weight function  $w : X \rightarrow \mathbb{R}^+$  that puts a weight for every object. And the weight of a set is simply defined as the total weights of the objects included, i.e.

$$w(\gamma) = \sum_{x \in \gamma} w(x).$$

**Definition 3.** For a range space  $\Sigma = (X, R)$ , and for  $0 < \varepsilon < 1$ , a subset  $A \subseteq X$  is an  $\varepsilon$ -net if  $A \cap \gamma \neq \emptyset$  for all  $\gamma \in R$  such that  $w(\gamma) \geq \varepsilon w(X)$ .

When we choose an uniform weight  $w(x) = 1$  for all  $x \in X$ , the condition is reduced to the condition on size  $|\gamma| \geq \varepsilon |X|$ .

## 2.3 Approximation algorithm for the hitting set

Generally speaking, it is difficult to approximate a hitting set problem. The best one can do for a general hitting set problem in polynomial time is an approximation with factor  $O(\log n)$ . Yet, an  $\varepsilon$ -net is easier to

find an approximate or randomized solution if it is defined on a well-behaved range space ( $|R|$  is polynomial). We state two facts without proof.

**Fact 1.** For a range space  $\Sigma = (X, R)$  with  $|R| \leq |X|^{O(1)}$ , an  $\varepsilon$ -net of size  $O(1/\varepsilon)$  can be computed in  $(|X|/\varepsilon)^{O(1)}$  time.

**Fact 2.** For a range space  $\Sigma = (X, R)$  with  $|R| \leq |X|^{O(1)}$ , a random subset of size  $O(\frac{1}{\varepsilon} \log \frac{1}{\varepsilon})$  formed by including each  $x \in X$  with probability  $w(x)/w(X)$  is an  $\varepsilon$ -net with probability  $p \geq 2/3$ .

Then, for a well-behaved range space, we can convert the hitting set problem to an  $\varepsilon$ -net problem and use either of the aforementioned algorithm to find an approximate or a probabilistically correct solution. As we know an  $\varepsilon$ -net only cares the ranges that are heavy, we can do the conversion by choosing  $\varepsilon$  and weights  $w$  carefully such that every range becomes heavy, that is  $w(\gamma) \geq \varepsilon w(X)$  for all  $\gamma \in R$ .

Let us suppose we have an oracle that can tell us the size  $k$  of the optimal hitting set. Then we set  $\varepsilon = \log \sqrt{2}/k$ , where the choice of the constant will be clear later. The algorithm for choosing the weights  $w$  goes as follows.

Initially, we set  $w(x) = 1$  for all  $x \in X$ . While  $\exists \gamma \in R$  such that  $w(\gamma) < \varepsilon w(X)$ , we arbitrarily pick such a  $\gamma$  and double the weights for the objects included by  $\gamma$ , i.e.

$$w(x) \leftarrow 2w(x) \quad \forall x \in \gamma.$$

Repeat it until  $w(\gamma) \geq \varepsilon w(X)$  for all  $\gamma \in R$ .

**Theorem 3.** The algorithm will terminate in  $O(k \log n)$  time.

*Proof.* Let  $w_i$  be  $w(X)$  after  $i$  iterations. Clearly,  $w_0 = |X| = n$ . In every iteration, we double the weight for the range that is lighter than  $\varepsilon w_i$ . Therefore, we have

$$w_{i+1} \leq w_i + \varepsilon w_i = (1 + \varepsilon)w_i.$$

By unrolling it, we have

$$w_i \leq (1 + \varepsilon)^i n \leq \exp(\varepsilon)^i n = \exp(\varepsilon i + \log n).$$

Supposing the optimal hitting set is  $H = \{x_1, x_2, \dots, x_k\}$ . By the definition of a hitting set, at least one of its element will appear in the range that we pick for weight-doubling in each iteration. Then, after  $i$  iterations, every  $x_i \in H$  should be weight-doubled for at least  $i/k$  times, which gives us the lower bound for  $w_i$  as

$$w_i \geq w(H) \text{ after } i \text{ iterations} \geq k2^{i/k}.$$

By combining both bounds, we have

$$k2^{i/k} \leq \exp(\varepsilon i + \log n).$$

Now we can arrive at

$$i \leq \frac{k \log n}{\log(2)/2} = O(k \log n).$$

□

Clearly, when the algorithm terminates, all ranges will be heavy and thus an  $\varepsilon$ -net will also be a hitting set. However, in practice, we do not have an oracle and therefore do not know  $k$ . We can instead start with some guess of  $k$  and then run the reweighting algorithm for at most  $\frac{k \log n}{\log(2)/2}$  iterations. If the algorithm terminates within the time limit, then we can continue to find an  $\varepsilon$ -net ; if not, we double our guess for  $k$  and repeat the process until the algorithm terminates within the limit.

### 3 Summary

In this lecture, we have introduced range spaces, the hitting set and the  $\varepsilon$ -net. We have seen that we can convert a hitting set problem on a well-behaved range space to an  $\varepsilon$ -net problem by a reweighting algorithm in polynomial time.

### 4 Further Reading Material

1. Har-Peled, Sariel. *Geometric Approximation Algorithms*. American Mathematical Soc., 2011.  
<http://sarielhp.org/book/>