



## Exercise Set X

These exercises are for your own benefit. Feel free to collaborate and share your answers with other students. Solve as many problems as you can and ask for help if you get stuck for too long. Problems marked \* are more difficult but also more fun :).

These problems are taken from various sources at EPFL and on the Internet, too numerous to cite individually.

- 1 **LSH for Jaccard similarity.** Suppose we have a universe  $U$ . For non-empty sets  $A, B \subseteq U$ , the Jaccard index is defined as

$$J(A, B) = \frac{|A \cap B|}{|A \cup B|}.$$

Design a locality sensitive hash (LSH) family  $\mathcal{H}$  of functions  $h : 2^U \rightarrow [0, 1]$  such that for any non-empty sets  $A, B \subseteq U$ ,

$$\Pr_{h \sim \mathcal{H}}[h(A) \neq h(B)] \begin{cases} \leq 0.01 & \text{if } J(A, B) \geq 0.99, \\ \geq 0.1 & \text{if } J(A, B) \leq 0.9. \end{cases}$$

(In this problem you are asked to explain the hash family and argue that it satisfies the above properties. Recall that you are allowed to refer to material covered in the course.)

**Solution:** Let us describe  $\mathcal{H}$  by giving a procedure to sample an element  $h \in \mathcal{H}$ :

- for each  $u \in U$ , sample  $h_u$  uniformly at random from  $[0, 1]$ .
- set  $h(A) = \min_{u \in A} h_u$  for any non-empty  $A \subseteq U$  (i.e., MinHashing).

In Exercise Set 10, we showed that  $\Pr[h(A) = h(B)] = J(A, B)$ . So  $\Pr[h(A) \neq h(B)] = 1 - J(A, B)$  and the claimed bounds follow immediately.

- 2 In this problem we design an LSH for points in  $\mathbb{R}^d$  with the  $\ell_1$  distance, i.e.

$$d(p, q) = \sum_{i=1}^d |p_i - q_i|.$$

Define a class of hash functions as follows: Fix a positive number  $w$ . Each hash function is defined via a choice of  $d$  independently selected random real numbers  $s_1, s_2, \dots, s_d$ , each uniform in  $[0, w)$ . The hash function associated with this random set of choices is

$$h(x_1, \dots, x_d) = \left( \left\lfloor \frac{x_1 - s_1}{w} \right\rfloor, \left\lfloor \frac{x_2 - s_2}{w} \right\rfloor, \dots, \left\lfloor \frac{x_d - s_d}{w} \right\rfloor \right).$$

Let  $\alpha_i = |p_i - q_i|$ . What is the probability that  $h(p) = h(q)$ , in terms of the  $\alpha_i$  values? It may be easier to first think of the case when  $w = 1$ . Try to also simplify your expression if  $w$  is much larger than  $\alpha_i$ 's, using that  $(1 - x) \approx e^{-x}$  for small values of  $x \geq 0$ .

**Solution:** Let us try to picture what the hashing function does. On the  $i$ -th coordinate, it partitions  $\mathbb{R}$  into buckets of the form  $\dots, [s_i - w, s_i), [s_i, s_i + w), [s_i + w, s_i + 2w), \dots$ , each of length  $w$ , with a random “offset”. Given two numbers  $p_i$  and  $q_i$ , the probability that they fall into the same bucket is  $1 - \frac{|p_i - q_i|}{w}$  (unless they are farther away than  $w$ , in which case it is 0).<sup>1</sup> Therefore:

- if  $|p_i - q_i| > w$  for some  $i$ , then  $\Pr[h(p) = h(q)] = 0$ ,
- otherwise

$$\Pr[h(p) = h(q)] = \prod_{i=1}^d \left(1 - \frac{|p_i - q_i|}{w}\right) \approx \prod_{i=1}^d e^{-\frac{|p_i - q_i|}{w}} = e^{-\frac{\sum_{i=1}^d |p_i - q_i|}{w}} = e^{-\frac{\|p - q\|_1}{w}}.$$

- 3** Consider two LSH hash families  $\mathcal{H}_1$  and  $\mathcal{H}_2$  designed for a distance function  $\text{dist} : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$ . For  $r = 0.1$  and  $c = 2$ ,  $\mathcal{H}_1$  satisfies

$$\text{dist}(p, q) \leq r \implies \Pr_{h \sim \mathcal{H}_1} [h(p) = h(q)] \geq 1/2$$

$$\text{dist}(p, q) \geq c \cdot r \implies \Pr_{h \sim \mathcal{H}_1} [h(p) = h(q)] \leq 1/8$$

and  $\mathcal{H}_2$  satisfies

$$\text{dist}(p, q) \leq r \implies \Pr_{h \sim \mathcal{H}_2} [h(p) = h(q)] \geq 1/8$$

$$\text{dist}(p, q) \geq c \cdot r \implies \Pr_{h \sim \mathcal{H}_2} [h(p) = h(q)] \leq 1/200$$

- 3a** Which Hash family would you choose to build the data structure  $\text{ANNS}(r, c)$  explained in class? What would the space requirement and query time be (logs are not so important)?

- 3b** On query  $q \in \mathbb{R}^d$ , asymptotically how many hash function computations are done?

**Solution: Deferred to Exercise Set 11.**

- 4** Suppose you have a database with a set  $P \subseteq \mathbb{R}^d$  of  $n$  items that are equipped with a distance function  $\text{dist} : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$  satisfying the following sparsity condition:

$$|\{p \in P : \text{dist}(p, q) \leq 2\}| \leq 10.$$

Further assume that you have a  $(r, c \cdot r, p_1, p_2)$ -LSH hash family  $\mathcal{H}$  for the considered distance function with parameters  $r = 1$ ,  $c = 2$ ,  $p_1 = 1/2$  and  $p_2 = 1/8$ . That is,

$$\text{dist}(p, q) \leq 1 \implies \Pr[h(p) = h(q)] \geq 1/2$$

$$\text{dist}(p, q) \geq 2 \implies \Pr[h(p) = h(q)] \leq 1/8$$

where the probabilities are over  $h \sim \mathcal{H}$ .

Exploit the sparsity condition to modify the  $\text{ANNS}(c, r)$  construction seen in class so as to obtain a structure with the *same* asymptotic preprocessing and query times, but with the following improved guarantee:

On query  $q \in \mathbb{R}^d$ , if  $\min_{p \in P} \text{dist}(p, q) \leq 1$ , then we return  $\arg \min_{p \in P} \text{dist}(p, q)$  with probability close to 1.

(Notice that this is stronger than the guarantee seen in class as in that case one is only guaranteed to return a point  $p'$  such that  $\text{dist}(p', q) \leq c \cdot r$  with probability close to 1.)

What is the preprocessing time, query time, and space requirement of your solution?

<sup>1</sup>To see this, assume wlog that  $p_i < q_i < p_i + w$ ; there will be exactly one bucket-beginning in the interval  $(p_i, p_i + w]$ , the position of that bucket-beginning is distributed uniformly on that interval, and  $p_i$  and  $q_i$  will go into different buckets if and only if that bucket-beginning falls into  $(p_i, q_i]$ . The probability of this happening is  $\frac{|p_i - q_i|}{w}$ .

**Solution: Deferred to Exercise Set 11.**