Chapter 9

Hashing

Up to this point, all the data structures we saw in that course assume nothing but comparability about the inputs keys they expected to mange. Hash functions, in general, are functions¹ from the keys space into lower diminsional space, that in the most basic version can be thought as our memory storage.

For example, assume the keys are taken form the integers, and the hash function $h:\mathbb{N}\to [10]$ sends numbers to theirs residue module 10, namely $h(x)=x \mod 10$. Now one might using h and 10-length array A to store numbers. Any time he would like to insert a number x into the structure he would set $A[h(x)]\leftarrow x$. That is not a "good" method thouw, since two different keys with the same residual module 10 are mapped into the same cell in A, we count such pair as collision:

Definition 9.0.1 (collison.). Fix a function $h:U\to \star$. We name any pair of different keys $x\neq y\in U$ that are mapped by h to the same value, namely h(x)=h(y), a collision.

9.1 Universal Hashing.

Definition 9.1.1. Let $\mathcal{H} = \{h_i : U \to [m]\}$ be a family of function from the domain U into $[m] = \{0, 1, ... m - 1\}$. \mathcal{H} will be said universial if for any $x \neq y \in U$:

$$\mathbf{Pr}_{h \sim \mathcal{H}} \left[h(x) = h(y) \right] \le \frac{1}{m}$$

Question. For x = y what is the probability that h(x) = h(y)?

Example 9.1.1. \mathcal{H} is the set of all function from $U \to [m]$.

Example 9.1.2. \mathcal{H} is the set of all binary matrices from $U = \mathbb{F}_2^n \to \mathbb{F}_2^k$.

Example 9.1.3. \mathcal{H} is the set of all function from $U \to [m]$.

Exercise 9.1.1. U is the set of all matrices $\mathbb{F}_2^{n \times n}$ and $h_x(A) = x^{\top} A x$.

9.2 Perfect Hashing.

In the past week, we have seen how to store keys in hash tables so that the number of mapped keys in a specific cell is O(1) in expectation. The table is constructed

¹projections.

using a hashing function h: key space $\to m$ cells, randomly chosen from a universal hash function family. This function maps keys to cells, and in each cell, the keys are stored using a linked list. The cost of supported subroutines depends on the length of the list. We named any pair of different keys $x \neq y$ that are mapped to the same cell in the table, namely h(x) = h(y), a collision.

Perfect hashing is a method to ensure that no collision occurs, it works only if all keys are given in advance and they are unique, meaning that the table doesn't support insertion. The idea is as follows, we sample an hash function, and then check if, for all x,y in the input, it holds that $h(x) \neq h(y)$. If so then we continue. Otherwise we repeat.

```
1 let collision \leftarrow True
   while collision do
        collision \leftarrow False
        let T be array at length m
 4
        h \leftarrow sample uniformly random from universal hash family \mathcal{H}
 5
       for x \in x_1, x_2..x_n do
 6
            if T_{h(x)} is not empty then
 7
                collision \leftarrow True
 8
                break the for-loop
            end
10
            else
11
             T_{h(x)} \leftarrow x
13
       end
15 end
16 return T, h
```

Algorithm 1: perfect-hashing $(x_1, x_2, ...x_n)$

Question. What is the probability of choosing h with no collisions on the first trial? Notice that the answer depends on m. (To see this, imagine the case where m=1. In this case, there must be collisions.) Therefore, the correct question is: for what values of m do we succeed in finding a hash function with no collisions on the first trial? Let $X_{x,y}$ be the indicator of the event h(x)=h(y). The expected number of collisions is then:

$$\mathbf{E}\left[\sum_{x\neq y} X_{x,y}\right] = \sum_{x\neq y} \mathbf{E}\left[X_{x,y}\right] = \binom{n}{2} \frac{1}{m}$$

Now, we would like to answer for what value of m there is no collision. Therefore, if we take $m=n^2$, then the expected number of collisions is less than 1/2. By the Markov inequality, the probability of having more than one collision is less than:

$$P\left(\sum_{x \neq y} X_{x,y} > 1\right) \leq \mathbf{E}\left[\sum_{x \neq y} X_{x,y}\right] = \frac{1}{2}$$

And therefore the expected number of rounds is less than:

$$\mathbf{E}\,[\text{ rounds }] = \sum_{t=0}^{\infty} t P(t \text{ rounds }) \leq \sum_{t=0}^{\infty} t \frac{1}{2^{t-1}} = O(1)$$

Question. What is the space complexities? We have to allocate an array at length m which is $\Theta(n^2)$ memory. Is that good? So remember that in standard hash tables, the expected number of elements that were hashed into the same cell as the key x is

$$1 + \frac{n-1}{m}$$

Taking $m = \Theta(n)$ is enough to ensure that the expected running time of in insertion/deletion/access is O(1). This raises the question of whether the space complexity of perfect hashing can be reduced to linear.

9.2.1 Perfect Hashing in Linear Space.

The idea is as follows: we will use a two-stage hashing process. In the first stage, keys will be mapped to hash tables instead of cells. Each hash table will be constructed using perfect hashing and may require a space that is quadratic in the number of elements stored in it (which were mapped to it in the first stage). Therefore, if we denote by n_i the number of elements mapped to the ith hash table, the space cost will be $\sum_i n_i^2$. Instead of starting over when a collision occurs, we will do so when $\sum_i n_i^2 > 4n$. So, now it's left to show that we expect $\sum_i n_i^2$ to be

```
1 let toomanycollisions \leftarrow True
2 while toomanycollisions do
       toomanycollisions ← False
3
       let T be array at length m
4
       initialize any T_i to be an empty linked list.
5
       h \leftarrow sample uniformly random from universal hash family \mathcal{H}
       for x \in x_1, x_2..x_n do
7
            T_{h(x)}.insert(x)
8
            T_{h(x)}.size = T_{h(x)}.size +1
9
10
       if \sum_{i} T_{h(i)}.size^2 \ge \mu then
        toomanycollisions \leftarrow True
12
       end
13
14 end
15 let H be an array at length m
  for i \in [m] do
       T_i, h_i \leftarrow \text{hash the elements in } T_i \text{ using }
           perfect hashing
18
19 end
20 return T, h
           Algorithm 2: perfect-hashing-linear-space(x_1, x_2, ...x_n)
```

linear, which implies that the expected number of rounds is constant.

$$n_i^2 = 2\binom{n_i}{2} + n_i$$

On the other hand, $\sum_i \binom{n_i}{2}$ is exactly the number of collisions, as for any i, $\binom{n_i}{2}$ counts the number of distinct pairs in the ith table, which is equivalent to counting the number of $x \neq y$ such that h(x) = h(y) = i. Thus,

$$\mathbf{E}\left[\sum_{i} n_{i}^{2}\right] = \mathbf{E}\left[\sum_{i} 2\binom{n_{i}}{2} + n_{i}\right] = 2 \cdot \mathbf{E}\left[\text{collisions}\right] + \mathbf{E}\left[\sum_{i} n_{i}\right]$$
$$= 2 \cdot \binom{n}{2} \frac{1}{m} + n$$

Therefore, by choosing m=4n for the first stage, the probability of failing to choose a proper hash function is less than 1/2.