1 Overview

In this lecture, we study hashing.

2 Hashing

We use hashing to solve the following problem: we have keys which we want to add, remove, and lookup. Keys belong to some universe. Hashing is useful when the number of keys stored is small compared to the universe, otherwise and array will work.

We now define the problem more formally:

**Definition 1.** Let $U$ be a universe. Let $S$ be the hash table, $|S| = n$, ($n \ll U$). A hash function $h$ is a function which maps from $U$ to the $n$ slots in $S$.

We first construct a simple hash function, $h_1$.

**Definition 2.** Let $U$ be a universe and $n$ be the number of keys to be stored in the hash table. Split $U$ into $n$ contiguous sections, and assign key $k$ an index in the hash table as follows:

$$h_1(x) = i, \text{ where } x \text{ belongs to the } i\text{th section of } U.$$  

Our goal is to avoid collisions.

**Definition 3.** Let $x, y \in U$ and $h$ be a hash function. $x$ and $y$ collide if $h(x) = h(y)$ and $x \neq y$.

Collisions are unavoidable, but should be minimized. For hash function $h_1$, the fraction of pairs $x, y$ which collide is

$$n \binom{|U|}{2} = \Theta\left(\frac{1}{n}\right).$$

For any way we divide up universe, $\frac{|U|^2}{n}$ is the fewest number of total possible collisions, because $\sum_{i=1}^{n} x_i^2 \geq \frac{1}{n}$, where $x_i$ is the number of elements in $U$ that are mapped to key $i$. We want to find a hash function which satisfies the following property:

**Property 1.** For all $x, y \in U$ and $x \neq y$, $\Pr[h(x) = h(y)] \leq \frac{1}{n}$.

Note that $h_1$ does not satisfy Property 1 since for any $x, y$ in the same section of $U$, $\Pr[h(x) = h(y)] = 1$. It is clear that no deterministic algorithm can satisfy Property 1 so we must use randomization. We define $h_2$ with this in mind.

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1 Some materials are from a previous note by Samuel Haney for this class in Fall 2014.
Definition 4. Let $U$ be a universe and $n$ be the number of keys to be stored in the hash table. Assign $h_2(x)$

$$h_2(x) = X,$$

where $X$ is sampled uniformly at random from $\{1, \ldots, n\}$.

Claim 1. Hash function $h_2$ satisfies Property 1.

Proof.

$$\Pr[x, y \text{ collide}] = \Pr[h(x) = h(y)] = \sum_{k \in S} \Pr[h(x) = h(y) = k]$$

$$= \sum_{k \in S} \frac{1}{n^2} = \frac{1}{n}.$$

The problem with $h_2$ is that we cannot efficiently find where each key is stored. We would need to store the location of each key in the hash table, which is exactly the problem we are trying to solve. Therefore, we would also like our hash functions to satisfy the following property.

Property 2. $h$ can be stored succinctly and $h(x)$ can be recovered efficiently.

We will propose one more hashing scheme which will satisfy both Property 1 and Property 2. First, define field $F_p$:

Definition 5. Let $F_p$ be the set $\{0, 1, \ldots, p\}$, where $p$ is prime, with operations $+$ and $\cdot$ defined as follows:

- $a + b \overset{\text{def}}{=} (a + b) \mod p$, and
- $a \cdot b \overset{\text{def}}{=} (a \cdot b) \mod p$.

Note that $F_p$ is closed under $+$ and $\cdot$.

Definition 6. Let $U$ be a universe and $n$ be the number of keys to be stored in the hash table. Assume $n = p$ is prime. For $x \in U$, let $x_1, \ldots, x_k$ be the digits of $x$ when written in base $p$ (i.e. $x_i \in F_p$, and $x = \sum_{i=1}^k x_i \cdot p^{k-1}$). Then,

$$h_3(x) = \left(\sum_{i=1}^k a_i \cdot x_i\right) \mod p,$$

where $a_i$ is chosen uniformly at random from $F_p$, for all $i$.

Claim 2. Hash function $h_3$ satisfies Property 1 and Property 2.
Proof. First, note that Property \([2]\) is satisfied. Only \(a_1, \ldots, a_k\) need to be stored, and \(h(x)\) is computed with a simple sum. Proving Property \([1]\) is more challenging.

Suppose \(x \neq y\) and \(x, y \in U\).

\[
\Pr[h(x) \neq h(y)] = \Pr\left[ \left( \sum_i a_i \cdot x_i \right) \equiv \left( \sum_i a_i \cdot y_i \right) \mod p \right]
\]

\[= \Pr\left[ \sum_i (a_i x_i - a_i y_i) \equiv 0 \mod p \right]. \tag{1}
\]

by definition of \(h\). Because \(x \neq y\), there is an index \(j\) such that \(x_j \neq y_j\) (i.e., \((x_j - y_j) \neq 0\)). Continuing from \([1]\) we have

\[
\Pr\left[ \sum_i (a_i x_i - a_i y_i) \equiv 0 \mod p \right] = \Pr\left[ \sum_{i \neq j} (a_i (x_i - y_i) + a_j (x_j - y_j)) \equiv 0 \mod p \right]. \tag{2}
\]

Next, fix all \(a_i\) for \(i \neq j\), and let

\[
\alpha = \left( -\sum_{i \neq j} a_i (x_i - y_i) \right) \mod p.
\]

\(\alpha\) is a fixed number, and \(\alpha \in F_p\). Let \(\beta = x_j - y_j\) and note that \(\beta \neq 0\). Continuing \([2]\) gives

\[
\Pr\left[ \sum_{i \neq j} (a_i (x_i - y_i) + a_j (x_j - y_j)) \equiv 0 \mod p \right] = \Pr[a_j (x_j - y_j) \equiv \alpha \mod p]
\]

\[= \Pr[a_j \beta \equiv \alpha \mod p].
\]

Multiplying all the elements of a field by some set nonzero element gives a permutation of the field. Since \(a_j\) is chosen uniformly at random from \(F_p\), we get

\[
\Pr[a_j \beta \equiv \alpha \mod p] = \frac{1}{p}.
\]

Therefore, Property \([1]\) is satisfied. \(\square\)

Definition 7. A collection of hash functions \(H\) is universal if \(\forall x, y \in U\) and \(x \neq y\), \(\Pr_{h \in H}[h(x) = h(y)] \leq \frac{1}{n}\) where \(h\) is chosen uniformly at random from \(H\).