

**Exercises****11.4-1**

Consider inserting the keys 10, 22, 31, 4, 15, 28, 17, 88, 59 into a hash table of length  $m = 11$  using open addressing with the auxiliary hash function  $h'(k) = k$ . Illustrate the result of inserting these keys using linear probing, using quadratic probing with  $c_1 = 1$  and  $c_2 = 3$ , and using double hashing with  $h_1(k) = k$  and  $h_2(k) = 1 + (k \bmod (m - 1))$ .

**11.4-2**

Write pseudocode for HASH-DELETE as outlined in the text, and modify HASH-INSERT to handle the special value DELETED.

**11.4-3**

Consider an open-address hash table with uniform hashing. Give upper bounds on the expected number of probes in an unsuccessful search and on the expected number of probes in a successful search when the load factor is  $3/4$  and when it is  $7/8$ .

**11.4-4 ★**

Suppose that we use double hashing to resolve collisions—that is, we use the hash function  $h(k, i) = (h_1(k) + ih_2(k)) \bmod m$ . Show that if  $m$  and  $h_2(k)$  have greatest common divisor  $d \geq 1$  for some key  $k$ , then an unsuccessful search for key  $k$  examines  $(1/d)$ th of the hash table before returning to slot  $h_1(k)$ . Thus, when  $d = 1$ , so that  $m$  and  $h_2(k)$  are relatively prime, the search may examine the entire hash table. (*Hint:* See Chapter 31.)

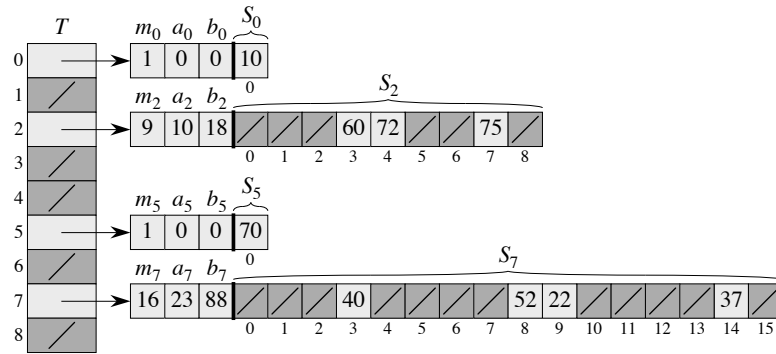
**11.4-5 ★**

Consider an open-address hash table with a load factor  $\alpha$ . Find the nonzero value  $\alpha$  for which the expected number of probes in an unsuccessful search equals twice the expected number of probes in a successful search. Use the upper bounds given by Theorems 11.6 and 11.8 for these expected numbers of probes.

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**★ 11.5 Perfect hashing**

Although hashing is often a good choice for its excellent average-case performance, hashing can also provide excellent *worst-case* performance when the set of keys is *static*: once the keys are stored in the table, the set of keys never changes. Some applications naturally have static sets of keys: consider the set of reserved words in a programming language, or the set of file names on a CD-ROM. We



**Figure 11.6** Using perfect hashing to store the set  $K = \{10, 22, 37, 40, 52, 60, 70, 72, 75\}$ . The outer hash function is  $h(k) = ((ak + b) \bmod p) \bmod m$ , where  $a = 3$ ,  $b = 42$ ,  $p = 101$ , and  $m = 9$ . For example,  $h(75) = 2$ , and so key 75 hashes to slot 2 of table  $T$ . A secondary hash table  $S_j$  stores all keys hashing to slot  $j$ . The size of hash table  $S_j$  is  $m_j = n_j^2$ , and the associated hash function is  $h_j(k) = ((a_j k + b_j) \bmod p) \bmod m_j$ . Since  $h_2(75) = 7$ , key 75 is stored in slot 7 of secondary hash table  $S_2$ . No collisions occur in any of the secondary hash tables, and so searching takes constant time in the worst case.

call a hashing technique **perfect hashing** if  $O(1)$  memory accesses are required to perform a search in the worst case.

To create a perfect hashing scheme, we use two levels of hashing, with universal hashing at each level. Figure 11.6 illustrates the approach.

The first level is essentially the same as for hashing with chaining: we hash the  $n$  keys into  $m$  slots using a hash function  $h$  carefully selected from a family of universal hash functions.

Instead of making a linked list of the keys hashing to slot  $j$ , however, we use a small **secondary hash table**  $S_j$  with an associated hash function  $h_j$ . By choosing the hash functions  $h_j$  carefully, we can guarantee that there are no collisions at the secondary level.

In order to guarantee that there are no collisions at the secondary level, however, we will need to let the size  $m_j$  of hash table  $S_j$  be the square of the number  $n_j$  of keys hashing to slot  $j$ . Although you might think that the quadratic dependence of  $m_j$  on  $n_j$  may seem likely to cause the overall storage requirement to be excessive, we shall show that by choosing the first-level hash function well, we can limit the expected total amount of space used to  $O(n)$ .

We use hash functions chosen from the universal classes of hash functions of Section 11.3.3. The first-level hash function comes from the class  $\mathcal{H}_{pm}$ , where as in Section 11.3.3,  $p$  is a prime number greater than any key value. Those keys

hashing to slot  $j$  are re-hashed into a secondary hash table  $S_j$  of size  $m_j$  using a hash function  $h_j$  chosen from the class  $\mathcal{H}_{p,m_j}$ .<sup>1</sup>

We shall proceed in two steps. First, we shall determine how to ensure that the secondary tables have no collisions. Second, we shall show that the expected amount of memory used overall—for the primary hash table and all the secondary hash tables—is  $O(n)$ .

**Theorem 11.9**

Suppose that we store  $n$  keys in a hash table of size  $m = n^2$  using a hash function  $h$  randomly chosen from a universal class of hash functions. Then, the probability is less than  $1/2$  that there are any collisions.

**Proof** There are  $\binom{n}{2}$  pairs of keys that may collide; each pair collides with probability  $1/m$  if  $h$  is chosen at random from a universal family  $\mathcal{H}$  of hash functions. Let  $X$  be a random variable that counts the number of collisions. When  $m = n^2$ , the expected number of collisions is

$$\begin{aligned} E[X] &= \binom{n}{2} \cdot \frac{1}{n^2} \\ &= \frac{n^2 - n}{2} \cdot \frac{1}{n^2} \\ &< 1/2. \end{aligned}$$

(This analysis is similar to the analysis of the birthday paradox in Section 5.4.1.) Applying Markov's inequality (C.30),  $\Pr\{X \geq t\} \leq E[X]/t$ , with  $t = 1$ , completes the proof. ■

In the situation described in Theorem 11.9, where  $m = n^2$ , it follows that a hash function  $h$  chosen at random from  $\mathcal{H}$  is more likely than not to have *no* collisions. Given the set  $K$  of  $n$  keys to be hashed (remember that  $K$  is static), it is thus easy to find a collision-free hash function  $h$  with a few random trials.

When  $n$  is large, however, a hash table of size  $m = n^2$  is excessive. Therefore, we adopt the two-level hashing approach, and we use the approach of Theorem 11.9 only to hash the entries within each slot. We use an outer, or first-level, hash function  $h$  to hash the keys into  $m = n$  slots. Then, if  $n_j$  keys hash to slot  $j$ , we use a secondary hash table  $S_j$  of size  $m_j = n_j^2$  to provide collision-free constant-time lookup.

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<sup>1</sup>When  $n_j = m_j = 1$ , we don't really need a hash function for slot  $j$ ; when we choose a hash function  $h_{ab}(k) = ((ak + b) \bmod p) \bmod m_j$  for such a slot, we just use  $a = b = 0$ .

We now turn to the issue of ensuring that the overall memory used is  $O(n)$ . Since the size  $m_j$  of the  $j$ th secondary hash table grows quadratically with the number  $n_j$  of keys stored, we run the risk that the overall amount of storage could be excessive.

If the first-level table size is  $m = n$ , then the amount of memory used is  $O(n)$  for the primary hash table, for the storage of the sizes  $m_j$  of the secondary hash tables, and for the storage of the parameters  $a_j$  and  $b_j$  defining the secondary hash functions  $h_j$  drawn from the class  $\mathcal{H}_{p,m_j}$  of Section 11.3.3 (except when  $n_j = 1$  and we use  $a = b = 0$ ). The following theorem and a corollary provide a bound on the expected combined sizes of all the secondary hash tables. A second corollary bounds the probability that the combined size of all the secondary hash tables is superlinear (actually, that it equals or exceeds  $4n$ ).

**Theorem 11.10**

Suppose that we store  $n$  keys in a hash table of size  $m = n$  using a hash function  $h$  randomly chosen from a universal class of hash functions. Then, we have

$$\mathbb{E} \left[ \sum_{j=0}^{m-1} n_j^2 \right] < 2n ,$$

where  $n_j$  is the number of keys hashing to slot  $j$ .

**Proof** We start with the following identity, which holds for any nonnegative integer  $a$ :

$$a^2 = a + 2 \binom{a}{2} . \quad (11.6)$$

We have

$$\begin{aligned} \mathbb{E} \left[ \sum_{j=0}^{m-1} n_j^2 \right] &= \mathbb{E} \left[ \sum_{j=0}^{m-1} \left( n_j + 2 \binom{n_j}{2} \right) \right] && \text{(by equation (11.6))} \\ &= \mathbb{E} \left[ \sum_{j=0}^{m-1} n_j \right] + 2 \mathbb{E} \left[ \sum_{j=0}^{m-1} \binom{n_j}{2} \right] && \text{(by linearity of expectation)} \\ &= \mathbb{E} [n] + 2 \mathbb{E} \left[ \sum_{j=0}^{m-1} \binom{n_j}{2} \right] && \text{(by equation (11.1))} \end{aligned}$$

$$= n + 2 \mathbb{E} \left[ \sum_{j=0}^{m-1} \binom{n_j}{2} \right] \quad (\text{since } n \text{ is not a random variable}) .$$

To evaluate the summation  $\sum_{j=0}^{m-1} \binom{n_j}{2}$ , we observe that it is just the total number of pairs of keys in the hash table that collide. By the properties of universal hashing, the expected value of this summation is at most

$$\begin{aligned} \binom{n}{2} \frac{1}{m} &= \frac{n(n-1)}{2m} \\ &= \frac{n-1}{2} , \end{aligned}$$

since  $m = n$ . Thus,

$$\begin{aligned} \mathbb{E} \left[ \sum_{j=0}^{m-1} n_j^2 \right] &\leq n + 2 \frac{n-1}{2} \\ &= 2n - 1 \\ &< 2n . \end{aligned} \quad \blacksquare$$

### Corollary 11.11

Suppose that we store  $n$  keys in a hash table of size  $m = n$  using a hash function  $h$  randomly chosen from a universal class of hash functions, and we set the size of each secondary hash table to  $m_j = n_j^2$  for  $j = 0, 1, \dots, m-1$ . Then, the expected amount of storage required for all secondary hash tables in a perfect hashing scheme is less than  $2n$ .

**Proof** Since  $m_j = n_j^2$  for  $j = 0, 1, \dots, m-1$ , Theorem 11.10 gives

$$\begin{aligned} \mathbb{E} \left[ \sum_{j=0}^{m-1} m_j \right] &= \mathbb{E} \left[ \sum_{j=0}^{m-1} n_j^2 \right] \\ &< 2n , \end{aligned} \quad (11.7)$$

which completes the proof. ■

### Corollary 11.12

Suppose that we store  $n$  keys in a hash table of size  $m = n$  using a hash function  $h$  randomly chosen from a universal class of hash functions, and we set the size of each secondary hash table to  $m_j = n_j^2$  for  $j = 0, 1, \dots, m-1$ . Then, the probability is less than  $1/2$  that the total storage used for secondary hash tables equals or exceeds  $4n$ .

**Proof** Again we apply Markov's inequality (C.30),  $\Pr\{X \geq t\} \leq E[X]/t$ , this time to inequality (11.7), with  $X = \sum_{j=0}^{m-1} m_j$  and  $t = 4n$ :

$$\begin{aligned} \Pr\left\{\sum_{j=0}^{m-1} m_j \geq 4n\right\} &\leq \frac{E\left[\sum_{j=0}^{m-1} m_j\right]}{4n} \\ &< \frac{2n}{4n} \\ &= 1/2. \end{aligned}$$

■

From Corollary 11.12, we see that if we test a few randomly chosen hash functions from the universal family, we will quickly find one that uses a reasonable amount of storage.

### Exercises

#### 11.5-1 ★

Suppose that we insert  $n$  keys into a hash table of size  $m$  using open addressing and uniform hashing. Let  $p(n, m)$  be the probability that no collisions occur. Show that  $p(n, m) \leq e^{-n(n-1)/2m}$ . (Hint: See equation (3.12).) Argue that when  $n$  exceeds  $\sqrt{m}$ , the probability of avoiding collisions goes rapidly to zero.

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### Problems

#### 11-1 Longest-probe bound for hashing

Suppose that we use an open-addressed hash table of size  $m$  to store  $n \leq m/2$  items.

- Assuming uniform hashing, show that for  $i = 1, 2, \dots, n$ , the probability is at most  $2^{-k}$  that the  $i$ th insertion requires strictly more than  $k$  probes.
- Show that for  $i = 1, 2, \dots, n$ , the probability is  $O(1/n^2)$  that the  $i$ th insertion requires more than  $2 \lg n$  probes.

Let the random variable  $X_i$  denote the number of probes required by the  $i$ th insertion. You have shown in part (b) that  $\Pr\{X_i > 2 \lg n\} = O(1/n^2)$ . Let the random variable  $X = \max_{1 \leq i \leq n} X_i$  denote the maximum number of probes required by any of the  $n$  insertions.

- Show that  $\Pr\{X > 2 \lg n\} = O(1/n)$ .
- Show that the expected length  $E[X]$  of the longest probe sequence is  $O(\lg n)$ .