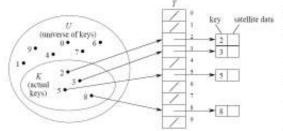
### **Dictionaries**

- Maintain a *dynamic* set S of objects
  - Each object is a (key, value) pair
- Each object x has a unique **key**, denoted by x.key, selected from universe U, associated with a **value** 
  - K is the set of keys in S. Goal: implement mapping  $key \mapsto value$
- Support *dictionary operations*:
  - Insert(x, S): stores a new (key, value) pair x
  - Delete(x, S), where  $x \in S$ : removes x from S
  - Search(k, S), where  $k \in U$ : returns the *value* associated with k
- Two efficient data structures:
  - Hashing is a generalization of simple array addressing
    - S stored in a table T of size m
  - Binary search trees generalize binary search
    - Allow efficient implementation of order-dependent operations

## Direct Addressing

- One place for each item, at most one item in each place
  - Each slot corresponds to a key in U
  - If there is an element x with key k then T[k] points to x



DIRECT-ADDRESS-SEARCH(T, k)return T[k]

Direct-Address-Insert(T, x) $T[key[x]] \leftarrow x$ 

Direct-Address-Delete(T, x) $T[key[x]] \leftarrow NIL$ 

### Hash Tables

- Direct addressing is not practical if U is much larger than K
- Would like space to be small
  - Use a table of size  $m \in O(|K|)$
- Can still get O(1) search time, but on average, not worst case.
- **Idea**: instead of storing x in slot x.key, use function h and store x in slot h(x.key)  $h: U \rightarrow \{0,1,...,m-1\}$  is the **hash function**

We say that k hashes to h(k)

**Hash Functions** 

- Many different functions used in practice
  - Interpret key k as a number
- Division method
  - $h(k) = k \mod m$
  - If k consists of many "digits" can interpret as a
     r + 1 digit number in base b:

$$h(k) = \left(\sum_{i=0}^{r} k_i b^i\right) \bmod m$$

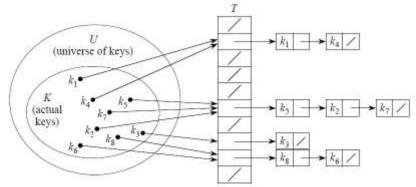
• Multiplication method:  $h(k) = \lfloor m(k \cdot A \mod 1) \rfloor$ , where A is a suitable constant

### **Collisions**

- A *collision* occurs when two or more keys hash to the same slot
- Can happen whenever |U| > 1 or with poorly designed hashing functions
  - May or may not happen if |K| ≤ m
  - Will definitely happen if |K| > m
- Can resolve by:
  - Finding a different slot (open addressing)
  - Chaining

## Chaining

• Use a linked list to store all items that hash to the same slot.



•  $n_i$  denotes the length of the *i*-th chain

#### **Operations**

· Insertion:

CHAINED-HASH-INSERT(T, x)insert x at the head of list T[h(kxy[x])]

- Worst-case running time is O(1).
- · Assumes that the element being inserted isn't already in the list.
- · It would take an additional search to check if it was already inserted.
- · Search:

CHAINED-HASH-SEARCH(T, k)search for an element with key k in list T[h(k)]

Running time is proportional to the length of the list of elements in slot h(k).

· Deletion:

CHAINED-HASH-DELETE (T, x)delete x from the list T[h(key[x])]

- Given pointer x to the element to delete, so no search is needed to find this
  element.
- Worst-case running time is O(1) time if the lists are doubly linked.

### Performance

- Analysis assumes *simple uniform hashing*: any element is equally likely to hash to any of the m slots
- In terms of *load factor*  $\alpha = n/m$

**Theorem.** With simple uniform hashing, a search (successful or not) runs in expected  $\Theta(1+\alpha)$  time.

Why is this result not useful enough?

## Universal Hashing: Motivation

- Any fixed *h* is vulnerable to "malicious adversary", i.e., there is a set of keys that makes that *h* perform poorly.
- Can improve performance by randomly choosing *h*, independently of *K*.
- No set *K* can consistently elicit worst case behavior.
- Algorithm may perform differently on different runs with same input.

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### Universal Hashing

**Definition.** A collection of hash functions  $\mathcal{H} = \{h \text{ where } h: \mathcal{U} \to \{0, ..., m-1\} \}$  is **universal** if for <u>every</u> pair of keys  $i, j \in \mathcal{U}, i \neq j$ , the number of functions  $h \in \mathcal{H}$  for which h(i) = h(j) is  $\leq |\mathcal{H}|/m$ 

**Theorem.** Let h be chosen at random from a universal set and used to hash n keys, with chaining, into a table T of size m. For any  $k \in \mathcal{U}$ , the expected length of chain T[h(k)] is at most  $\alpha$  if  $k \notin K$  and at most  $1 + \alpha$  if  $k \in K$ .

*Proof.* For every 
$$i, j, k \in U$$
, define :

$$X_{ij} = I[h(i) = h(j)]$$
 and  $Y_k = \sum_{j \in K, j \neq k} X_{kj}$ 

Then, 
$$E[Y_k] = E\left[\sum_{h \in K, h \neq k} X_{kh}\right]$$
  

$$= \sum_{h \in K, h \neq k} E[X_{kh}]$$

$$\leq \sum_{h \in K, h \neq k} \frac{1}{m}$$

Two cases:

$$\begin{aligned} &1.\, k \not\in K \Rightarrow n_{h(k)} = Y_k \Rightarrow E[n_{h(k)}] \leq n \,/\, m = \alpha \\ &2.\, k \in K \Rightarrow n_{h(k)} = 1 + Y_k \Rightarrow E[n_{h(k)}] \leq 1 + (n-1) \,/\, m < 1 + \alpha \end{aligned}$$

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## Universal Hashing...

**Theorem.** Consider a chained hash table with m slots built using universal hashing. If the number of insertions is O(m) then the expected time to process any sequence of n Insert, Search, and Delete operations is O(n).

#### Fields

A field is a set F with two binary operations ⊕ ("addition") and
 ⊙ ("multiplication") that satisfies the following properties:

Closure:  $\forall a, b \in F : a \oplus b \in F, a \odot b \in F$ Commutativity:  $\forall a, b \in F, a \oplus b = b \oplus a, a \odot b = b \odot a$ Associativity of addition:  $\forall a, b, c \in F, (a \oplus b) \oplus c = a \oplus (b \oplus c)$ Additive Identity:  $\exists z \in F : \forall a \in F : a \oplus z = z \oplus a = a$ Additive Inverse:  $\forall a \in G : \exists b \in G : a \oplus b = b \oplus a = z$ Associativity of multiplication:  $\forall a, b, c \in F, (a \odot b) \odot c = a \odot (b \odot c)$ Multiplicative Identity:  $\exists u \in F : \forall a \in F : a \odot u = u \odot a = a$ Multiplicative Inverse:  $\forall a \in F, a \neq z : \exists b \in F : a \odot b = b \odot a = u$ Distributivity:  $\forall a, b, c \in F, a \odot (b \oplus c) = (a \odot b) \oplus (a \odot c),$ 

• A set with one operation that satisfies closure, associativity, and existence of an identity and inverses is called a *group*.

 $(b \oplus c) \odot a = (b \odot a) \oplus (c \odot a)$ 

- The set of nonzero elements of a field from a group with respect to ⊙

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# The Field $\mathbb{Z}_p$

- Consider the integers  $\{0,1,...,p-1\}$  with the usual addition and multiplication modulo p
- If p is prime, then  $(Z_p, +\text{mod } p, *\text{mod } p)$  is a field

$\mathbb{Z}_5^+$	0	1	2	3	4
0	0	1	2	3	4
1	1	2	4	4	0
2	2	3	4	0	1
3	3	4	0	1	2
4	4	0	1	2	3

$\mathbb{Z}_{5}^{*}$	1	2	3	4
1	1	2	3	4
2	2	4	1	3
3	3	1	4	2
4	4	3	2	1

We denote the multiplicative inverse of a by  $a^{-1}$ , e.g., in  $\mathbb{Z}_5$   $3^{-1} = 2$ 

# $\mathbb{Z}_q^*$ , for non prime q

• If q is not prime, then  $\mathbb{Z}_q^*$  is not a group

$\mathbb{Z}_6^*$	1	2	3	4	5
1	1	2	3	4	5
2	2	4	0	2	4
3	3	0	3	0	3
4	4	2	0	4	2
5	5	4	3	2	1

$\mathbb{Z}_{7}^{*}$	1	2	3	4	5	6
1	1	2	3	4	5	6
2	2	4	6	1	3	5
3	3	6	2	5	1	4
4	4	1	5	2	6	3
5	5	3	1	6	4	2
6	6	5	4	3	2	1

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## Constructing a Universal Set

- Consider a universe of keys  $\mathcal{U} = \{0, ..., u 1\}$  and let p be a prime such that  $u \le p \le 2u$
- Assume m < p (else use direct addressing!)
- For  $a \in \mathbb{Z}_p^*$  and  $b \in \mathbb{Z}_p$  define a hashing function  $h_{ab}(k) = ((a \cdot k + b) \mod p) \mod m$
- The family of hashing functions is defined as  $\mathcal{H}_{pm} = \{h_{ab} : a \in \mathbb{Z}_p^* \text{ and } b \in \mathbb{Z}_p\}$
- Size m of table is arbitrary provided m < p
- How big is  $\mathcal{H}_{pm}$ ? p(p-1)

### Universality

**Theorem**. The class  $\mathcal{H}_{pm}$  is universal

*Proof.* Let  $k, \ell \in \mathcal{U}, k \neq \ell$  and  $a \in \mathbb{Z}_p^*, b \in \mathbb{Z}_p$ 

Let  $r = ak + b \mod p$  and  $s = a\ell + b \mod p$ 

Then  $r \neq s$  (no collision at the mod p level). Why?

Each of the p(p-1) choices of  $h_{ab}$  yields a different pair  $(r,s) \Rightarrow$  we have a bijection between pairs (a,b) and pairs (r,s). Why?

 $a = (r - s)(k - \ell)^{-1} \mod p$  and  $b = (r - ak) \mod p$  $\Pr(h_{ab}(k) = h_{ab}(\ell)) = \Pr(r = s \mod m)$  where r and s are random distinct values in  $\mathbb{Z}_p$  (each pair equally likely)

For fixed r, how many  $s \neq r$  result in  $s = r \mod m$ ?

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## Universality Proof...

For a given r, how many  $s \neq r$  result in  $s = r \mod m$ ? Answer: at most  $\lceil p/m \rceil$  values from  $\mathbb{Z}_p$  are  $= 0 \mod p$ . Since  $s \neq r$ , at most  $\lceil p/m \rceil - 1$  of the remaining values result in  $s = r \mod p$ 

However, 
$$\lceil p/m \rceil - 1 \le \left(\frac{p+m-1}{m}\right) - 1 = \frac{p-1}{m}$$

 $Pr(collision) = Pr[h_{ab}(k) = h_{ab}(\ell)]$ 

$$\leq \frac{(p-1)/m}{p-1} = \frac{1}{m}$$

#### A Different Universal Set

- Table size *m* is a prime number.
- Each key x consists of r+1 digits, base m  $x = \langle x_r \cdots x_1 x_0 \rangle$
- Choose random (r+1)-digit base-m number a:

$$a = \langle a_r \cdots a_1 a_0 \rangle$$

$$h_a(x) = \sum_{i=0}^r a_i x_i \bmod m$$

**Theorem.**  $H = \{h_a\}$  is universal.

**Proof.** For x,y with  $x_0 \neq y_0$ , arbitrary  $a_1,a_2,...,a_r$ , how many  $a_0$  satisfy

$$a_0(x_0 - y_0) = \sum_{i=1}^r a_i(y_i - x_i) \bmod m$$

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# Number of Solutions to $a_0 b = w$

- Can assume  $b \neq 0$
- Two cases
  - 1.  $w = 0 \Rightarrow a_0 = 0$  is the only solution
  - 2.  $w \neq 0 \Rightarrow a_0 = w b^{-1}$  is the only solution
- Either way, the number of hashing functions that result in a collision of x and y is  $m^r$
- Therefore, for  $i \neq j \in \mathcal{U}$ , when h is chosen randomly from  $\mathcal{H}$ ,  $\Pr[h(i) = h(j)] = 1/m$

### Perfect Hashing

**Goal**: Given a set K of n keys, construct a *static* hash table of size m = O(n) such that Search takes O(1) time.

- Every search takes O(1) time in the worst case
- Statistical variation from list to list or key to key does not affect performance

**Definition**. A perfect hash function maps different elements of *K* to different slots, i.e., it is a *total injective function* 

**Applications**: compiler keywords, 10000 most common words in the English dictionary, files on a CD.

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## Markov's Inequality

**Theorem**. If X is a non-negative random variable then  $P[X \ge t] \le E[X] / t$ 

**Proof.** Define an indicator variable  $Y = I(X \ge t)$ . Then,  $P(X \ge t) = E(Y)$ . Since  $Y \le X/t$  for all t, then  $P(X \ge t) = E(Y) \le E(X/t) = E(X)/t$ 

### Collision Free Hashing

**Theorem**. Suppose you store n keys in a hash table of size  $m \ge n^2$  using a hash function randomly chosen from a universal set. Then, the probability of having any collisions is less than 1/2

**Proof.** Let *X* denote the number of colliding pairs.

There are n(n-1)/2 pairs that may collide.

Each pair collides with probability  $\leq 1/m$ 

 $E[X] \le n(n-1)/2m \le n(n-1)/2n^2 < 1/2$ 

By Markov's inequality (with t=1)  $P[X \ge 1] \le E[X] < 1/2$ 

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### A Better Idea

- Quadratic storage for a large set of keys is not reasonable
- Instead, use 2-level hashing with universal hashing at both levels.
  - No chaining, instead have *m* secondary hash tables built with universal hashing
  - Each secondary table has size  $m_i = (n_i)^2$  where  $n_i$  is the number of keys in slot i
- May have collisions at level 1 but no collisions at level 2

**Theorem**. Suppose we store n keys into a hash table of size  $m \ge n$ , using a hash function h chosen at random from a universal set of hash functions. Then

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

**Proof.** 
$$E\left[\sum_{j=0}^{m-1} n_j^2\right] = E\left[\sum_{j=0}^{m-1} \binom{n_j}{2} + 2\binom{n_j}{2}\right] \quad \text{(because } a^2 = a + 2\binom{a}{2}\text{)}$$

$$= E\left[\sum_{j=0}^{m-1} n_j\right] + 2E\left[\sum_{j=0}^{m-1} \binom{n_j}{2}\right] \quad \text{(linearity of expectation)}$$

$$\leq n + 2\binom{n}{2} \frac{1}{m} \quad \text{(expected number of collisions)}$$

$$\leq n + 2\frac{n-1}{2} = 2n - 1 \quad \text{(because } m \geq n\text{)}$$

*Corollary*. Suppose we store *n* keys into a hash table of size  $m \ge n$ , using a hash function h chosen at random from a universal set and we set the size of each secondary table to  $m_j = (n_j)^2$ , j = 0,...m-1. Then

- The expected total storage required by the secondary hash tables is < 2n
- The probability is < 1/2 that the total storage required by the secondary tables is  $\geq 4n$

Proof.

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

2. 
$$\Pr\left[\sum_{j=0}^{m-1} m_j \ge 4n\right] \le \frac{E\left[\sum_{j=0}^{m-1} m_j\right]}{4n} < \frac{2n}{4n} = \frac{1}{2}$$