# Inf 2B: Hash Tables

## Lecture 4 of ADS thread

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# Dictionaries

A *Dictionary* stores key–element pairs, called *items*. Several elements might have the same key. Provides three methods:

- findElement($k$): If the dictionary contains an item with key $k$, then return its element; otherwise return the special element NO_SUCH_KEY.
- insertItem($k, e$): Insert an item with key $k$ and element $e$.
- removeItem($k$): If the dictionary contains an item with key $k$, then delete it and return its element; otherwise return NO_SUCH_KEY.

# List Dictionaries

- Items are stored in a *singly linked list* (in any order).
- Algorithms for all methods are straightforward.
- Running Time:

$$\begin{aligned} \textbf{insertItem} : &\quad \Theta(1) \\ \textbf{findElement} : &\quad \Theta(n) \\ \textbf{removeItem} : &\quad \Theta(n) \end{aligned}$$

($n$ always denotes the number of items stored in the dictionary)

# Direct Addressing

Suppose:

- Keys are integers in the range $0, \ldots, N - 1$.
- All elements have distinct keys.

A data structure realising *Dictionary* (sometimes called a *direct address table*):

- Elements are stored in array $B$ of length $N$.
- The element with key $k$ is stored in $B[k]$.
- Running Time: $\Theta(1)$ for all methods.

# Bucket Arrays

Suppose:

- Keys are integers in the range $0, \ldots, N-1$.
- Several elements might have the same key, so collisions may occur.

What do we do about these collisions?

Store them all together in a *List* pointed to by $B[k]$ (sometimes called *chaining*).

# Bucket Arrays

*Bucket array* implementation of *Dictionary*:

- Bucket array $B$ of length $N$ holding *Lists*
- Element with key $k$ is stored in the *List* $B[k]$.
- Methods of *Dictionary* are implemented using insertFirst(), first(), and remove($p$) of *List*

Running Time: $\Theta(1)$ for all methods (with linked list implementation of *List* - $p$ is always the first pointer, so we can easily keep track of it).

► Works because findElement($k$) and removeItem($k$) only need 1 item with key $k$.

A good solution if $N$ is not much larger than the number of keys (a small constant multiple).

# Hash Tables

*Dictionary* implementation for arbitrary keys (not necessarily all distinct).

Two components:

- *Hash function* $h$ mapping keys to integers in the range $0, ..., N - 1$ (for some suitable $N \in \mathbb{N}$).
- *Bucket array* $B$ of length $N$ to hold the items.

Item (key–element pair) with key $k$ is stored in the bucket $B[h(k)]$.

# Issues for Hash Tables

- Need to consider collision handling. (Here we might have $h(k_1) = h(k_2)$ even for $k_1 \neq k_2$, so *List* implementation is more complicated.
- Analyse the running time.
- Find good hash functions.
- Choose appropriate $N$.

# Implementation

**Problem:** Elements with distinct keys might go into the same bucket.
**Solution:** Let buckets be *list dictionaries* storing the items (key-element pairs).

**The methods:**

**Algorithm** findElement($k$)

1. Compute $h(k)$
2. **return** $B[h(k)]$.findElement($k$)

# Implementation

**Algorithm** InsertItem($k, e$)

1. Compute $h(k)$
2. $B[h(k)]$.insertItem($k, e$)

**Algorithm** removeItem($k$)

1. Compute $h(k)$
2. **return** $B[h(k)]$.removeItem($k$)

# Implementation

Running time?

Depends on the list methods

- $B[h(k)]$.findElement($k$),
- $B[h(k)]$.insertItem($k, e$), and
- $B[h(k)]$.removeItem($k$).

Assume we Insert at front (or end):

- $\Theta(1)$ time for $B[h(k)]$.insertItem($k, e$).

# Analysis

- Let $T_h$ be the running time required for computing $h$ (more precisely: $T_h(n_{\text{key}})$, where $n_{\text{key}}$ is the size of the key)
- Let $m$ be the maximum size of a bucket. Then the running time of the hash table methods is:

$$\begin{array}{ll} \textbf{insertItem}: & T_h + \Theta(1) \\ \textbf{findElement}: & T_h + \Theta(m) \\ \textbf{removeItem}: & T_h + \Theta(m) \end{array}$$

  Worst case:

$$m = n.$$

- $m$ depends on hash function *and* on input distribution of keys.

# Hash functions

Hash function $h$ maps keys to $\{0, \ldots, N-1\}$.

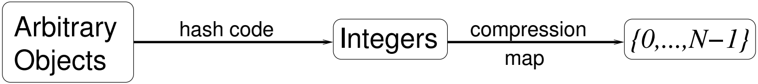
Criteria for a good hash function:

(H1) $h$ evenly distributes the keys over the range of buckets (hope input keys are well distributed originally) .

(H2) $h$ is easy to compute.

# Hash functions

- Simpler if we start with keys that are already integers.
- Trickier if the original key is not Integer type (eg `string`).
  **One approach:** Split hash function into:
  - hash code and
  - compression map.



# Hash Codes

- Keys (of *any* type) are just sequences of bits in memory.
- *Basic idea:* Convert bit representation of key to a binary integer, giving the hash code of the key.
- *But* computer integers have bounded length (say 32 bits).
  - consider bit representation of key as *sequence* of 32-bit integers $a_0, \ldots, a_{\ell-1}$
- *Summation method:* Hash code is

$$a_0 + \cdots + a_{\ell-1} \bmod N$$

- *Polynomial method:* Hash code is

$$a_0 + a_1 \cdot x + a_2 \cdot x^2 + \cdots + a_{\ell-1} \cdot x^{\ell-1} \bmod N$$

  (for some integer $x$).

Sometimes $N = 2^{32}$.

# Evaluating Polynomials

*Horner's Rule*:

$$a_0 + a_1 \cdot x + a_2 \cdot x^2 + \cdots + a_{\ell-1} \cdot x^{\ell-1}$$

$$=$$

$$a_0 + a_1 \cdot x + a_2 \cdot x \cdot x + \cdots + a_{\ell-1} \cdot x \cdot x \cdots x \qquad [\Theta(\ell^2) \text{ operations}$$

$$=$$

$$a_0 + x(a_1 + x(a_2 + \cdots + x(a_{\ell-2} + x \cdot + a_{\ell-1}) \cdots)) \quad [\Theta(\ell) \text{ operations}]$$

Has been *proved* to be best possible.

**Note:** Sensible to reduce mod $N$ after each operation.

**Warning:** Deciding what is a "good hash function" is something of a "black art".

Polynomials look good because it is harder to *see* regularities (many keys mapping to the same hash value).
**Warning:** we haven't proved anything! For some situations there are bad regularities, usually due to a bad choice of $N$.

# Hash functions for character strings

Characters are 7-bit numbers $(0, \ldots, 127)$.

- $x = 128, N = 96$. Bad for small words.
  (because $gcd(96, 128) = 32$. NOT coprime)
- $x = 128, N = 97$, good.
- $x = 127, N = 96$, good.

# Compression Map

Integer $k$ is mapped to

$$|ak + b| \bmod N,$$

where $a$, $b$ are randomly chosen integers.

Whole point of hashing is to “Compress” (evenly).

Works particularly *well* if $a$, $N$ are coprime (*experimental observation only*).

# Quick quiz question

Consider the hash function

$$h(k) = 3k \bmod 9.$$

Suppose we use $h$ to hash exactly one item for every key $k = 0, \ldots, 9M - 1$ (for some big $M$) into a bucket array with 9 buckets $B[0], B[1], \ldots, B[8]$. How many items end up in bucket $B[5]$?

1. 0.
2. $M$.
3. $2M$.
4. $4M$.

Answer is 0.

# Load Factors and Re-hashing

- Number of items: $n$
  Length of bucket array: $N$

  *Load factor*: $\dfrac{n}{N}$
- High load factor (**definitely**) causes many collisions (large buckets).
  Low load factor - waste of memory space.
  *Good compromise:* Load factor around $3/4$.
- Choose $N$ to be a prime number around $(4/3)n$.
- If load factor gets too high or too low, **re-hash** (amortised analysis similar to *dynamic arrays*).

# JVC and `HashMap`

- No duplicate keys.
- will hash many different types of key.
- User can specify - `initial capacity` (def. N=16), `load factor` (def. 3/4).
- *Dynamic* Hash table - “re-hash” takes place frequently behind scenes.
- Different hash functions for different key domains. For `String`, uses polynomial hash code with $a = 31$.
- `Hashtable` is more-or-less identical.

# Reading and Resources

- If you have [GT]: The “Maps and Dictionaries” chapter.
- If you have [CLRS]: The “Hash tables” chapter.
  **Nicest:** “Algorithms in Java”, by Robert Sedgewick (3rd ed), chapter 14.
- Two nice exercises on Lecture Note 4 (handed out).