

CSC 226

Algorithms and Data Structures: II

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Analysis of Hash table Access

- Given a hash table of size N containing n elements
 - We saw that the running time for a `put()` with chaining is $O(1)$
 - We also saw the runtime for `get()` is $O(\text{size of list})$
 - This is $O(n)$ in the worst-case.
- Some degree of clustering is inevitable with any hashing scheme.
 - If the hash function is chosen properly we can get away with this

Load Factor

- Assume our hash function, h , maps n keys to independent uniform random values in range $[0, N - 1]$.
- Let X be the number of items that map to index i in array A .
- Then, the expected value of X

$$E(X) = \frac{n}{N}$$

- The **load factor**, α , of a hash table is the ratio of occupied slots to total slots (which is $E(X)$). ³

Expected Run Time

- **Theorem:** Under the assumption of uniformity, the *expected* size of the linked list in each index of a hash table of size N storing n keys is which is equal to the load factor.

$$\alpha = n/N$$

- **Corollary:** The expected run time of `get()` is $O(\text{expected size of list})$ which is

$$O(\alpha)$$

by above theorem.

Load Factor

- If the number of collisions is small, then searching, inserting, and deleting elements in a hash table take $O(1)$ time
- To reduce the number of collisions, in addition to using a good hash function, we should make sure that the table does not get too full
- If it gets higher, we should extend the hash table and **rehash** all of its elements (i.e., remove all elements and re-insert the elements)
 - Look at this more when talking about open addressing

Advice

- Choosing a high quality hash function for a particular application usually requires some knowledge of the expected input data.
 - Simulations using sample data can be helpful in choosing a good function for particular inputs.
- ***Random linear hash function:*** For a hash table with n integer keys, the following rules of thumb often produce good results
 - Choose the table size to be a prime number p such that $1.5n < p$.
 - Choose values $0 < a < p, 0 \leq b < p$ and randomly

$$h(k) = ak + b \bmod p$$

Open Addressing

- **Open Addressing** collision resolution schemes store every key in a table index, using a *probing scheme* to find an available index when a collision occurs.
- **Linear probing** starts at the hash value $h(k)$, then checks successive indices until an empty space is found. The probe sequence is

$$h(k), \quad h(k) + 1, \quad h(k) + 2, \quad \dots$$

➤ where all values are taken mod the table size, N .

- Thus, at probe i the index checked is

$$(h(k) + i) \bmod N$$

Open Addressing

- **Quadratic probing** uses the sequence

$$h(k) + 0^2, \quad h(k) + 1^2, \quad h(k) + 2^2,$$

- Thus, at probe i the index checked is

$$(h(k) + i^2) \bmod N$$

- **Double hashing** uses two hash functions $h_1(k)$ and $h_2(k)$, where $h_2(k) \neq 0$ for any k . The probe sequence is

$$h_1(k), \quad h_1(k) + h_2(k), \quad h_1(k) + 2h_2(k), \quad \dots$$

- Thus, at probe i the index checked is

$$(h_1(k) + ih_2(k)) \bmod N$$

Linear Probing Example

Index	Value
0	
1	
2	
3	
4	
5	
6	

$$h(k) = 2k + 4 \bmod 7$$

- **Exercise:** Insert the keys 1,3,4,10,18 into the table using the given hash function
- Resolve collisions with linear probing

Linear Probing Example

Index	Value
0	18
1	
2	
3	3
4	10
5	4
6	1

$$h(k) = 2k + 4 \bmod 7$$

- **Exercise:** Search for key 10 in the resulting table. (i.e. perform `get(10)`).
- **Exercise:** `get(17)`.

Linear Probing Example

Index	Value
0	18
1	
2	
3	3
4	10
5	4
6	1

$$h(k) = 2k + 4 \bmod 7$$

- **Exercise:** `get(17)`.
- To search for a key, probe successive indices starting at the initial hash value until the key is found or an empty space is reached.

Linear Probing Example

Index	Value
0	18
1	
2	
3	3
4	10
5	4
6	1

$$h(k) = 2k + 4 \bmod 7$$

- **Exercise:** `remove(1)`.
- **Exercise:** `get(18)`.

Linear Probing Example

Index	Value
0	18
1	
2	
3	3
4	10
5	4
6	

$$h(k) = 2k + 4 \bmod 7$$

- **Exercise:** `remove(1)`.
- **Exercise:** `get(18)`.

Remove function

- For separate chaining, delete element in the linked list.
- For probing, mark element as *deleted* in the hash table since there might be elements following the deleted element in the open addressing probing chain.
- The usual protocol is to replace the key with a sentinel 'invalid element' marker, which is treated as an empty space during `put()` operations but treated as an element during `get()` operations.

Linear Probing Example

Index	Value
0	18
1	
2	
3	3
4	10
5	4
6	×

$$h(k) = 2k + 4 \bmod 7$$

- **Exercise:** `remove(1)`.
- **Exercise:** `get(18)`.

Linear Probing

- Note that when we did the `get(17)` operation we scanned every key in the table.
- This was due to clustering.
- One disadvantage of linear probing is that if a region of the table becomes clustered, every index in that region will suffer from long probing sequences.
- One way to alleviate this problem is to make the ‘step size’ of a probing sequence vary in some way based on the starting point.

Quadratic Probing Example

Index	Value
0	
1	
2	
3	
4	
5	
6	

$$h(k) = 2k + 4 \bmod 7$$

- **Exercise:** Insert the keys 1,3,4,10,18 into the table using the given hash function
- Resolve collisions with quadratic probing

Quadratic Probing Example

Index	Value
0	
1	
2	18
3	3
4	10
5	4
6	1

$$h(k) = 2k + 4 \bmod 7$$

- The table elements are all still clustered in a group of consecutive indices.
- In this case, the clustering is mainly caused by the table size being too small.
- Quadratic probing does result in shorter probe sequences for the `get()` operation, though.

Quadratic Probing Example

Index	Value
0	
1	
2	18
3	3
4	10
5	4
6	1

$$h(k) = 2k + 4 \bmod 7$$

- **Exercise:** get (17) .

Double Hashing Example

Index	Value
0	
1	
2	
3	
4	
5	
6	

$$h_1(k) = 2k + 4 \bmod 7$$

$$h_2(k) = 1 + (k \bmod 6)$$

- **Exercise:** Insert the keys 1,3,4,10,18 into the table using the given hash function
- Resolve collisions with double hashing

Double Hashing Example

Index	Value
0	18
1	10
2	
3	3
4	
5	4
6	1

$$h_1(k) = 2k + 4 \bmod 7$$

$$h_2(k) = 1 + (k \bmod 6)$$

- **Exercise:** `get(17)`.

Load Factor

- In separate chaining, α is the average length of a chain.
 - can be > 1 and still efficient
- for open addressing, α is the % of occupied cells
 - has to be < 1
- Proposition: Linear probing with n elements in a table of size N , the average number of probes is

$$\leq \frac{1}{2} \left(1 + \frac{1}{1-\alpha} \right) \text{ for search hits}$$

$$\leq \frac{1}{2} \left(1 + \frac{1}{(1-\alpha)^2} \right) \text{ for search misses}$$

Load Factor

- So, if $\alpha = 1/2$, then the expected number of probes for a search hit is

$$\leq \frac{1}{2} \left(1 + \frac{1}{1 - \frac{1}{2}} \right) = \frac{1}{2} (1 + 2) = \frac{3}{2}$$

- The expected number of probes for a search miss is

$$\leq \frac{1}{2} \left(1 + \frac{1}{\left(1 - \frac{1}{2}\right)^2} \right) = \frac{1}{2} (1 + 4) = \frac{5}{2}$$

- **Theorem:** The expected running time for doing a `get()`, `put()` or `remove()` in a hash table of size N , containing n items, with a load factor of $\alpha \leq 1/2$, is $O(1)$.

Rehashing

- As we have seen, the load factor, $\alpha = n/N$, has a big impact on the performance of a hash table.
- For a table using chaining, as long as the load factor is a small constant (near 1), our methods run in $O(1)$ time.
- For tables using probing, it needs to be sufficiently smaller than 1 (near $1/2$).
- In both cases as the load factor exceeds these limits the performance drops off quickly.
- It turns out that it's worth it to **rehash** when this happens.

Rehashing

- **Rehashing** – Double the size of the table and apply a new hash function to every element.
- It may seem that by doing this we are now running in $O(n)$ time for each of our methods but this is actually not the case.
- It turns out that the **amortized** running time for each of the hash table methods is still $O(1)$.
- Why? The cost of each regular `put()` is $O(1)$. After n regular `puts` there is a `put` that costs $O(n)$ time.
- So, all together you have n `puts` that cost $O(n)$ time in total. Thus, each costs $O(1)$ time.

Other Collision Strategies

- Separate chaining with binary trees
- Coalesced hashing – separate chaining in place
- Cuckoo hashing – new keys “push” old keys elsewhere
- Hopscotch hashing – constant number of neighbor buckets, move empty buckets closer
- Robin Hood hashing – displace key based on probe counts