

- Hash Table

- we have a universe of n keys and want to store them to a Hash Table with m slots
- we create a hash value $h(k_i)$ for each key k_i
- Running Time: the time to compute hash value + one time step to look up in the hash table.

- Collision

- when two keys map to the same hash value \rightarrow leading to store to the same slot.

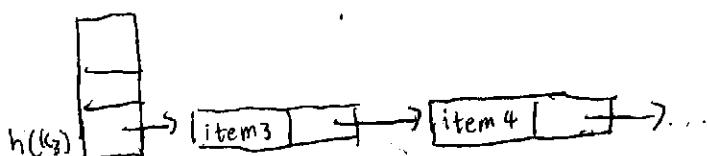
• for ex.

item1	$h(k_1)$
item2	$h(k_2)$
item3	$h(k_3)$

Now if $h(k_4) = h(k_1)$, we have to store item 4 in the same slot as item 1

- collision resolution strategy:

Chaining: store collided item in linked list. ex.



if $h(k_3) = h(k_4)$

pro: can store as many keys as possible.

cons: if we have a long collision, lookup time may not be constant.

* Collision resolution strategy (cont'd)

Linear probing: resolve collision by sequentially searching the hash table for free location.

$$\text{Ex. } h(x, i) = (H(x) + i) \pmod{m}$$

where $H(x)$ is an ordinary hash function, and i is the i^{th} stepsize (all previous $i-1$ slots occupied)

Quadratic probing: similar to linear probing but search for open slots as a quadratic function

$$\text{Ex. } h(x, i) = (H(x) + i^2) \pmod{m}$$

pro of linear/quadratic probing: fast insert, lookup.

cons: we are limited in the number of elements can be stored, and we can not empty the slot when deleting. (we have to put a dummy element in deleted slots)

• simple uniform hashing assumption.

• each key $k \in K$ is equally likely to be hashed to any slot of table T , independent of where other keys are hashed.

• Load factor = $\alpha = \frac{n}{m} = \text{average number of keys per slot.}$

Example : key a_1 and a_2 would have $\frac{1}{m}$ chance of collision under simple uniform hashing assumption so under SUHA, both a_1 and a_2 don't get hashed to slot 1 is $\left(\frac{m-1}{m}\right)^2$.

-Rolling Hash.

useful in string matching. consider 2 strings A and B, we want to find all matching subsequence of length k.

$$h(A[i:i+k-1]) = A[i] \cdot 2^{k-1} + A[i+1] \cdot 2^{k-2} + \dots + A[i+k-1]$$

$$\text{so } h(A[i:i+k]) = A[i+k] + 2^k \cdot h(A[i:i+k-1]) - A[i] \cdot 2^k$$

so for string length k, total running time to calculate all hash value is $O(k) + O(n-k) = O(n)$

when attempting to find the longest common substring, use binary search. Hash all substring length k of A into a hash-table and look up all substring length k of B.

Another example would be looking up words in a dictionary of DNA bank and trying to find patterns.