

12 Feb 2025

Hashing

(Typeset lecture notes coming soon!)

Balls \rightsquigarrow "keys"

Bins \rightsquigarrow "buckets"

Hashing: assign keys to buckets
randomly but reproducibly.

Dictionary a.k.a. Associative array, Key-value store.

Data structure that stores a set of key-value pairs
 $(x, v) \in X \times V$, with at most one value per key.

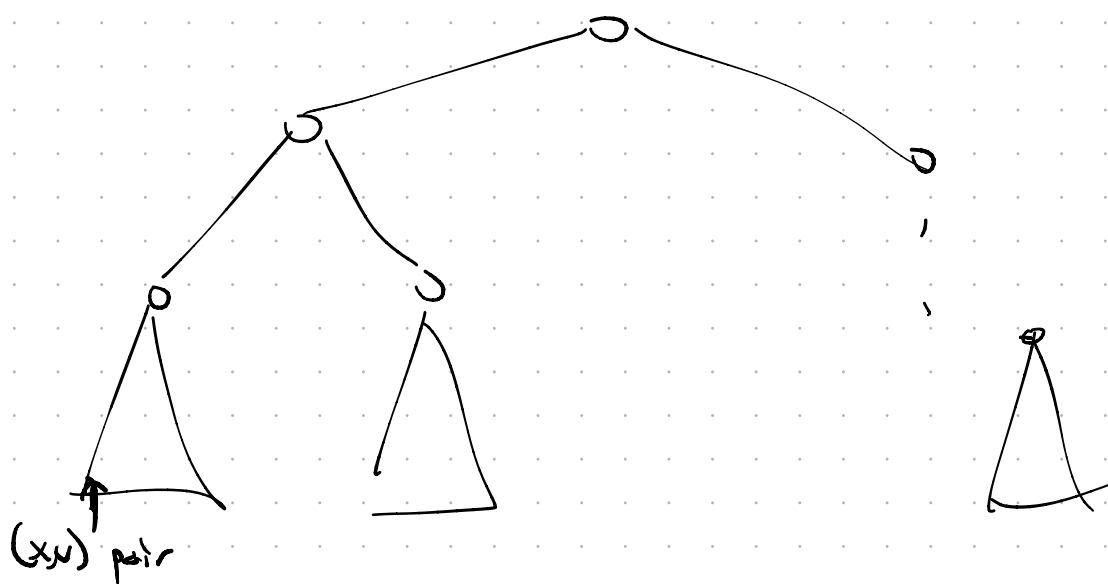
Supports: LOOKUP(x): returns value v stored with x
or NOT FOUND.

INSERT(x, v): insert (x, v) , overwriting (x, v') if necessary

DELETE(x): remove (x, v) from data structure
if any such pair exists.

a.k.a. Python dictionaries, Java/C++ Hash Maps.

Deterministic implementation. ^{Balanced} \wedge Binary search tree.



Tree structured sit.

- a. Every node has equal # Leaves (within ± 1) in Left, Right subtrees.
- b. All keys in left subtree are earlier in the ordering of X than the keys in the right subtree.

Assume m key-value pairs organized this way

\Rightarrow tree has depth $\lceil \log_2(m) \rceil$

\Rightarrow LOOKUP(x) takes $O(\log_2 m)$ time.

Space complexity: $O(m)$ tree nodes

so $O(m)$ space if each key and value occupies $O(1)$ space.

More generally, the data structure occupies space $O(S)$ where S is the min amt. of storage space needed to store all keys and values.

Hash Table: an array of n buckets.

(Typically $n = O(m)$.)

Each bucket contains a linked list of key-value pairs. ("chain hashing")

A hash function maps keys (randomly

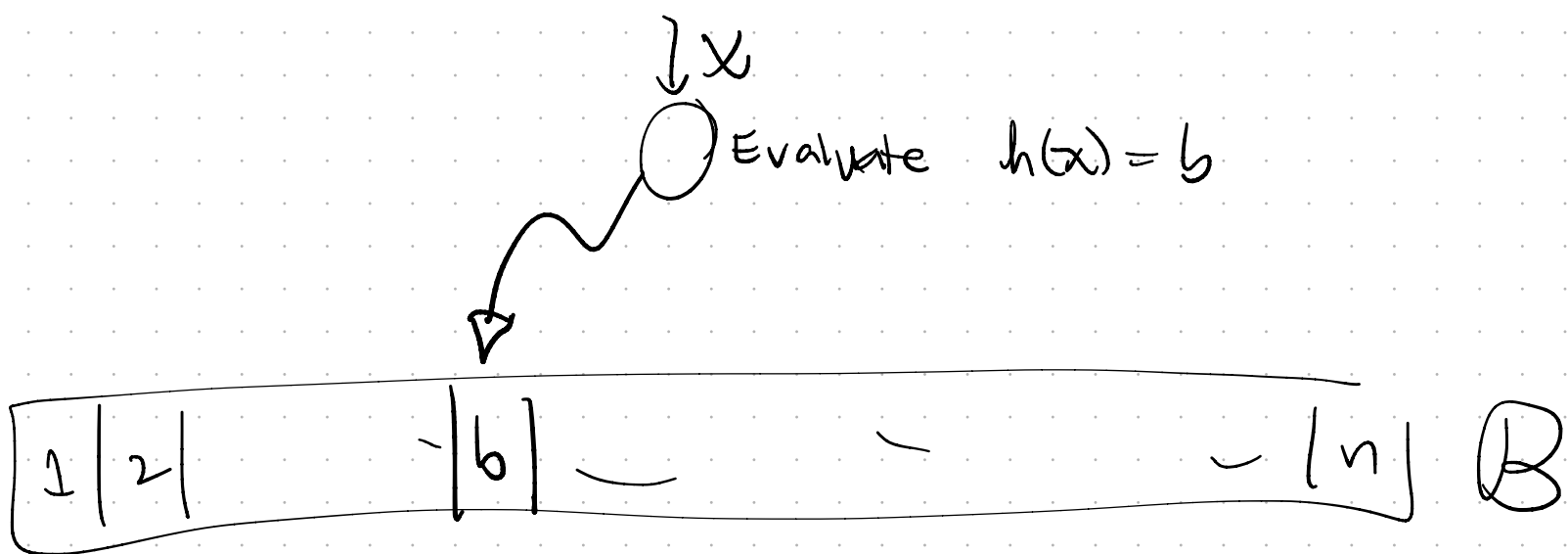
or pseudorandomly) to buckets.

Let B denote {buckets}.

For now assume ^(unrealistically) the hash function is

a uniformly random $h: X \rightarrow B$,

and that evaluating $h(x)$ takes $O(1)$ time.



LOOKUP(x): iterates through linked list
in bucket $b = h(x)$.

IF (x, v) found, return v .

Else return NOT FOUND.

INSERT(x, v): iterates through linked list b .

IF (x, v') found, replace with (x, v) .

Else append (x, v) to linked list

DELETE(x): iterates through linked list b .

if (x, v) found delete it

else do nothing.

Actual running time is

$O(1 + \text{length of linked list stored at } h(x))$.

random variable depending on h
and on the data stored in the table.

Linearity of expectation:

$$\begin{aligned} & \mathbb{E}[\text{length of linked list stored at } h(x)] \\ &= \mathbb{E}[\# y \in X \text{ s.t. a pair } (y, w) \text{ is} \\ & \quad \text{stored in the list at } h(x)] \\ &= \sum_{b \in B} \sum_{y \in X} \Pr(h(x) = h(y) = b \text{ and } (y, w) \text{ is a pair} \\ & \quad \text{stored in the table}) \end{aligned}$$

$\leq m$ distinct pairs (y, w) stored in table.

At most one satisfies $y = x$.

For that one $\Pr(h(x) = h(y) = b)$ is $\frac{1}{n}$.

For the other $\leq m$ pairs (y, w) with $y \neq x$

$\Pr(h(x) = h(y) = b)$ is $\frac{1}{n^2}$.

So,

$$\sum_{b \in B} \sum_{y \in X} \Pr(h(x) = h(y) = b \text{ and } (y, w) \text{ is a pair} \\ \text{stored in the table})$$

$$\leq \sum_{b \in B} \left[\frac{1}{n} + \frac{m}{n^2} \right]$$

$$= 1 + \frac{m}{n} \leftarrow \text{"load factor"}$$

Typical implementations use $n \geq m$

\Rightarrow load factor ≤ 1 .

Def. A pairwise independent (aka. 2-universal) hash family is a probability distribution on functions $h: X \rightarrow B$ satisfying

$\forall x_1 \neq x_2$ in X the pair $(h(x_1), h(x_2))$ is uniformly distributed over B^2 .

Example. Suppose $X = B = \{0, 1, 2, \dots, p-1\}$ and p is prime.

Sample random h by sampling coefficients $a, b \in \{0, \dots, p-1\}$ at random and defining $h(x) \equiv ax + b \pmod{p}$