

Algorithms - 2 Data Structures

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Data Structure

Basic Abstract Data Types (ADTs)

stack

queue

Common data structures

1. Heap

2. Hash

dynamic Set operations:

Hashing and Hash Tables

Universal Hashing:

Perfect Hashing

Consistent Hashing

Bloom-filer

3. Tree

Red-Black Tree

Data Structure

DS is a way to store and organize data in order to facilitate access and modifications

Basic Abstract Data Types (ADTs)

A mathematical model of the data methods used to modify and access the data, we don't care the implementation

Including: list, stack, queue, set, dictionary

stack

PUSH(S,x), POP(S)

queue

ENQUEUE(Q,x), DEQUEUE(Q)

Q.head, Q.tail, Q.length

Priority queue

data with a key

INSERT(P,x), EXTRACT-MAX(P), MAX(P), INCREASE-KEY(P,x,k)

Applications: OS(manage jobs), Graph Algorithms(Dijkstra), arrange events, compression(Huffman encoding), heapsort...

How should we implement the priority queue:
binary heap,...

Common data structures

1. Heap

Binary Heap

structure

array structure, visualized as nearly complete tree(only the left of last layer may be not full)

root = A[1], PARENT(i) = $\frac{i}{2}$, LEFT(i) = $2i$, RIGHT(i) = $2i+1$

attributes: A.length

Theorem: A nearly complete binary tree of n nodes has height $\theta(\log n)$

prove:

a complete binary tree with height h has $\sum_{i=0}^h 2^i = \frac{1-2^{h+1}}{1-2} = 2^{h+1} - 1$ nodes
so for a nearly complete tree $1 + \sum_{i=0}^{h-1} 2^i = 2^h \leq n \leq 2^{h+1} - 1$

Some other properties about the tree:

<https://cs.stackexchange.com/questions/841/proving-a-binary-heap-has-lceil-n-2-rceil-leaves>

Attribute

min binary heap PARENT(A[i]) ≤ A[i]

Operations - e.g. Min-heap

```
HEAP-MINIMUM(A)
    return A[1]

HEAP-EXTRACT-MIN(A)
    if A.heap-size < 1
        error 'heap underflow'
    min = A[1]
    A[1] = A[A.heap-size]
    A.heap-size = A.heap-size - 1
    MIN-HEAPIFY(A, 1)
    return min

MIN-HEAP-INSERT(A, key)
    A.heap-size = A.heap-size + 1
    A[A.heap-size] = inf
    HEAP-DECREASE-KEY(A, A.heap-size, key)

HEAP-DECREASE-KEY(A, i, key)
    if key > A[i]
        error 'new key is larger'
    A[i] = key
    while i > 1 and A[PARENT(i)] > A[i]
        exchange A[i] with A[PARENT(i)]
        i = PARENT(i)
```

```

BUID-MIN-HEAP(A)
    A.heap-size = A.length
    for i = floor(A.length/2) downto 1
        MAX-HEAPIFY(A,i)

MIN_HEAPIFY(A,i)
    l = LEFT(i)
    r = RIGHT(i)
    if l <= A.heap-size and A[l] < A[i]
        smallest = l
    else smallest = i
    if r <= A.heap-size and A[r] < A[smallest]
        smallest = r
    if smallest != i:
        exchange A[i] and A[smallest]
        MIN-HEAPIFY(A, smallest)

HEAPSORT(A) #O(nlogn)
    BUILD-MAX-HEAP(A)
    for i = A.length downto 2
        exchange A[1] with A[i]
        A.heap-size = A.heap-size - 1
        MAX-HEAPIFY(A, 1)

```

Time complexity of build a heap: $O(n)$

Amortized analysis: determine worst-case of a sequence of data structure operations

layer	height	depth	nodes	time = relate node*c
1	3	0	$1 = 2^0$	$c \cdot 3 \cdot 2^0$
2	2	1	$2 = 2^1$	$c \cdot 2 \cdot 2^1$
3	1	2	$4 = 2^2$	$c \cdot 1 \cdot 2^0$
4	0	3	$\leq 8 = 2^3$	

h is the height of the tree

$$\begin{aligned}
 T(n) &\leq c \sum_{i=1}^h 2^{h-i} \cdot i \leq c \cdot s^h \sum_{i=1}^h s^{h-i} \cdot i \leq c \cdot 2^{\log n} \sum_{i=1}^h 2^{-i} \cdot i \leq cn \sum_{i=1}^{\inf} 2^{-i} \cdot i \\
 &= cn \sum_{i=1}^{\inf} \left(\frac{1}{2}\right)^i \cdot i = \frac{cn}{2} \sum_{i=1}^{\inf} i \cdot \left(\frac{1}{2}\right)^{i-1} = \frac{cn}{2} \frac{1}{(1-1/2)^2} = O(n)
 \end{aligned}$$

notice that:

$$\frac{\partial \sum_{i=0}^{\inf} x^i}{\partial x} = \sum_{i=0}^{\inf} i \cdot x^{i-1}$$

$$\frac{\partial \frac{1}{1-x}}{\partial x} = \frac{1}{(1-x)^2}$$

$$\sum_{i=0}^{\inf} x^i = \frac{1}{1-x}$$

2. Hash

dynamic Set operations:

key(x) = k(satellite data)
SEARCH(S,k)
MINIMUM(S)/MAXIMUM(S)
SUCCESSOR(S,x)
PREDECESSOR(S,x)

dictionary operation

SEARCH(S,k) - returns a pointer x, where x.key = k or nil if not found
INSERT(S,x)
DELETE(S,x)

dictionary data structure

universe of keys U
a data structure that stores a subset $S \subseteq U$
operations:
SEARCH(S,k): given $k \in U$
INSERT(S,x): given x, a pointer to $k \notin S$
DELETE(S,x): given x, a pointer to $k \in S$

Hashing and Hash Tables

a hash table is an array T of size m
a hash function creates an index in the array from an $k \in U, h : U \rightarrow \{0, \dots, m-1\}$ (hashes to slot h(k) in T)

Collision solution - chaining:

Operations:

CHAINED-HASH-INSERT(S,x): insert x at the head of list T[x.key]
CHAINED-HASH-DELETE(S,x): delete x from the list T[x.key]

Assumption:

Simple uniform hashing: any key is equally likely to hash to any of the m slots
for all $i \neq j, \Pr[h(k_i) = h(k_j)] = \frac{1}{m}$

Analysis for unsuccessful search

Let S be the items already in the table where $x \notin S, |S| = n$
$$C_{x,y} = \begin{cases} 1 & h(x) = h(y) \\ 0 & \text{otherwise} \end{cases}$$
$$E[C_{x,y}] = 1 \cdot P[h(x) = h(y)] + 0 \cdot P[h(x) \neq h(y)] = \frac{1}{m}$$
$$C_x = \sum_{y \in S, y \neq x} C_{x,y} = \# \text{ of items that x has a collision with}$$
$$E[C_x] = \sum_{y \neq x} E[C_{x,y}] = \frac{n}{m} = \# \text{ items in the table divided by the size of the table}$$

So, if $n = O(m)$ then $\alpha = n/m = O(m)/m = O(1)$, for insertion, deletion, find

Universal Hashing:

choosing a hash function

using universal Hashing - randomness
Solve:
1. Uniform random
2. fundamental weakness

for any hash function, we can find a set of keys that are hashed to the same spot

Universal Class of Hash Functions:

$H = h_1, h_2, \dots, h_w$ is a universal hash function family if for all $k, k' \in U$, $Pr_{h \in H}[h(k) = h(k')] \leq \frac{1}{m}$, m is the table size

How to construct a universal Hash functions family:

$H_{pm} = \{h_{a,b}(x) = ((ax + b) \bmod p) \mid 1 \leq a \leq p-1, 0 \leq b \leq p-1, \text{ and } p \geq \text{all keys}\}$

Proof:

Let $k, l \in U$ are two keys, WLOG $k > l, k \neq l$.

1. If $h_{ab}(k) = h_{ab}(l)$ it is because they collided after mod p or mod m - Property of prime numbers
p is a prime larger than any key, and $p > m$

2. If $h_{ab}(k) = h_{ab}(l)$ it is because they collided after mod m - Modulo arithmetic

let $r = (ak+b) \bmod p$, $s = (al+b) \bmod p$

since $a < p$, $k, l < p \Rightarrow a$ doesn't divide p , $(k-l)$ doesn't divide p

the collision cannot be mod p since $0 \neq s - r \equiv a(k - l) \bmod p$

3. There is a 1-1 correspondence between the (r,s) pairs and the (a,b) pairs

$a = ((r-s)((k-l)^{-1} \bmod p)) \bmod p$

$b = (r-ak) \bmod p$

There are $p(p-1)$ possible pairs (r,s) that $r \neq s$, thus there is a 1-1 correspondence between pairs (a,b) and pairs (r,s). Thus if a, b is chosen randomly the pair (r,s) is equally likely to be any pair of distinct values modulo p

4. If we randomly choose an (r,s) pair, the probability that they collided mod m is $\leq 1/m$. Thus if I randomly choose an (a, b) pair the probability that $h_{a,b}(k) = h_{a,b}(l)$ is at most $\frac{1}{m}$.

of choice for s such that $r \neq s$ and $s \equiv r$ is at most $\lceil \frac{p}{m} \rceil - 1$

(just list all the possible s that collide with r: $r' = r \bmod m$

$r', r'+m, r'+2m, r'+3m, r'+4m, \dots, r' + (\lfloor \frac{p}{m} \rfloor - 1)m$) are all less than p, $r' + (\lceil \frac{p}{m} \rceil - 1)m$ may be less than p

Theorem

If for all $k, k' \in U$, $Pr_{h \in H}[h(k) = h(k')] \leq \frac{1}{m}$, then for any $S \subseteq U$, for any $x \in U - S$ the expected number of collisions between x and the other elements in S is at most n/m , when $|S|=n$

Proof:

Corollary 1: If for all $k, k' \in U$, $Pr_{h \in H}[h(k) = h(k')] \leq \frac{1}{m}$, then for any $S \subseteq U$, for any $x \in U - S$ the expected number of comparisons for a successful search is $O(1+n/m)$ and for an unsuccessful search is $O(n/m)$ where $|S| = n$.

Corollary 2: Using universal hashing and collision resolution by chaining in an initially empty table with m slots, it takes expected time $\theta(n)$ to handle any sequence of n INSERT, SEARCH, and DELETE operations containing $O(m)$ INSERT operations where m is the table size

Perfect Hashing

Static

We can do better when keys are static:

Given a fixed set of n keys we can construct a static hash table of size $m = \theta(n)$ such that search takes $\theta(1)$ time in the worst case

Theorem:

Hashing n keys into $m = n^2$ slots using $h \in_R H$ then $E(\# \text{ collisions}) < 1/2$

Proof:

probability that two keys collide is $1/m = 1/n^2$, and # pairs of key = $\binom{n}{2}$

$$E(\# \text{ collisions}) \leq \binom{n}{2} \cdot \frac{1}{n^2} = \frac{n(n-1)}{2} \cdot \frac{1}{n^2} < \frac{1}{2}$$

Corollary:

Probability of no collisions $> 1/2$

Proof: X be a R.V. holding the number of collisions, according to Markov inequality:

$$Pr[X \geq 1] \leq \frac{E[X]}{1} < \frac{1}{2}$$

(FYI: Markov's inequality can refer to Math.md)

To reduce the space: two levels**Theorem:**

Let $m=n$ be the size of the hash table at level 1

Let n_j be the size of the hash table at index j

Let h be randomly chosen from a universal set H

Then $Pr[\sum_{j=0}^{m-1} (n_j)^2 \geq 4n] < 1/2$

Proof:

the number of pairs that collide at the first layer:

$$\sum_{x \in S} \sum_{y \in S} C_{x,y} = \sum_{j=0}^{m-1} \sum_{x \in S_j} \sum_{y \in S_j} C_{x,y} = \sum_{j=0}^{m-1} (n_j)^2$$

$$E[\sum_{j=0}^{m-1} n_j^2] = E[\sum_{x \in S} \sum_{y \in S} C_{x,y}] = n + \sum_{x \in S} \sum_{y \in S-x} E[C_{x,y}] \leq n + n(n-1)/m < 2n$$

Markov's inequality: $Pr[\sum_{i=1}^{m-1} (n_i)^2 \geq 4n] < 2n/4n = 1/2$

dynamic

Consistent Hashing**Bloom-filter**

is a space-efficient probabilistic data structure for test membership

property:

- False positive - increase with n
- never generate false negative result
- can't delete element
- can always adding element with the prize that FP increaser

Operations:

- insert(x), lookup(x)

Probability of FP:

- m : size of bit array,
- k : number of hash functions
- n : number of expected elements to be inserted in the filter
- $P = (1 - (1 - \frac{1}{m})^{kn})^k$

3. Tree

we may need more operations than dictionary
each node: .key, .satellite_data, .p, .left, .right

Self Balancing Tree

Balanced: height is $O(\log n)$

2-3, 2-3-4, Red-Black Trees, Augmenting Data

B-trees

2-3 Tree

at most 3 children, all external nodes have same depth

2-node: 1 key 2 links

3-node: 2 keys 3 links

$\lfloor \log_3 n \rfloor \leq h \leq \lfloor \log_2 n \rfloor$

How to insert a key: local transformation preserve global property: 1. tree is ordered, 2. perfectly balanced

always insert into an existing node

always maintain depth condition

split node if a 4-node:

v, is created

split into two

Binary Search Tree

the left tree are all smaller, and the right trees are all greater

Red-Black Tree

is a binary search tree that obeys **5 properties**:

1. Every node is colored red or black
2. The root is black
3. Every leaf (T.nil) is black and doesn't contain any data
4. Both Children of a red node are black
5. For every node in the tree, all paths from that node down to a leaf (nil) have the same number of black nodes along the path

Creating a Red-Black BST from a 2-3-4 tree Cont.

Creating a 2-3-4 tree from Red-Black BST

Claim: The subtree rooted at a node, x, contains at least $2^{bh(x)} - 1$ internal nodes (non nil)

Basis: if x has height 0, it is a leaf (nil node) and $bh(x) = 0$, the subtree rooted at x has 0 internal nodes $2^0 - 1 = 0$

Inductive hypothesis: A node y of height at most h-1 contains at least $2^{bh(y)} - 1$.

Lemma: A red-black tree with n internal nodes has height at most $2\lg(n+1)$

Proof: Let h be the height, and $b = bh(x)$ be the black height of the root node x . By the previous two claims $n \geq 2^b - 1 \geq 2^{h/2} - 1$. Thus $n + 1 \geq 2^b \geq 2^{h/2} \implies \lg(n + 1) \geq h/2 \implies h \leq 2\lg(n + 1)$

Corollary: On a red-black tree with n nodes, we can implement dynamic-set operations SEARCH, MINIMUM, MAXIMUM, PREDECESSOR.... In $(\lg n)$

Modify the Tree:

Insert it as a red node, and then handle violation

There are three possible violations:

Case1: sibling of parent is red

Case2:

Case3:

```

LEFT-ROTATE(T,x)
    y = x.right
    x.righ = y.left
    if y.left != T.nil:
        y.left.p = x

RB-INSERT(T,z)
    // find where to insert
    y = T.nil
    x = T.root
    while x != T.nil
        y = x
        if z.key < x.key
            x = x.left
        else: x = x.right
    // create node and attach it
    z.p = y
    if y == T.nil: // tree T was empty
        T.root = z
    else if z.key < y.key
        y.left = z
    else y.right = z
    // solve color violation
    z.left = T.nil
    z.right = T.nil
    z.color = RED
    RB-INSERT-FIXUP(T,z)

RB-INSERT-FIXUP(T,z)
    while z.p.color == RED
        if z.p == z.p.p.right: // parent is a left child
            y = z.p.p.left // y is uncle
            if y.color == RED // case 1: just change the color
                z.p.color = BLACK
                y.color = BLACK
                z.p.p.color = RED
                z = z.p.p
            else // case 2 or case 3
                if z == z.p.left // z is a left child case 2
                    z = z.p
                    RIGHT-ROTATE(T,z)
                // handle case 3
                z.p.color = BLACK
                z.p.p.color = RED
                LEFT-ROTATE(T,z.p.p)
        else (same as then clause with right and left exchanged)

```


`T.root = Black`

Proof we maintain the Red-Black tree properties:

Loop invariant:

At the start of each iteration of the while loop, node z is red.

There is at most one red-black tree violation either:

- a. Z is a red root
- b. z and z.p are both red

Initialization:

before we add the new red node z,

Termination: loop terminates when z.p is black. Thus, there is no red-red violation.

Maintenance:

WLOG we only consider the 3 cases, where z.p is the right child of z.p.p let $y = z.p.p.left$ // y is z's uncle. z.p.p is black, since z and z.p are red and there is only one red-red violation

Case1: y is red and z.p.p is black, by making y and z.p black and z.p.p red. Thus black height property is maintained, but we might have created a red-red violation between z.p.p and its parent. Assign $z = z.p.p$

Case2: y is black and z is a left child. Set $z = z.p$ right rotate around z. Now z is now a right child. Both z and z.p are red. Only one case 3 red-red violation.

Case3: y is black z is a right child, make z.p black and z.p.p red left rotate on z.p.p