

## AVL

1. 结点个数:  $n(h) = n(h-1) + n(h-2) + 1$  -> 保证树高  $\log n$
2. insert: 插入后, 最多只有  $\log n$  个节点不平衡, 最多只需两次旋转恢复平衡
3. delete: 刚删除完毕, 最多只有一个节点不平衡, 最多需要  $O(\log n)$  次旋转恢复平衡
4. The balance factor  $BF(\text{node}) = hL - hR$

## 红黑树

- 结点个数: 算上 NIL: 奇数个
- 黑高: 不包括自身, 包括 NIL。到叶子的 path 上的黑节点之和  
->  $h(T) \leq 2bh(T)$  (每个 node 到 leaf 的 path 上红点数不超过黑点)
- 每个结点到叶子的 path 的长度之比不超过 2
- A red-black tree with  $N$  internal nodes has height at most  $2\ln(N+1)$ .
  - $\text{size}(Tu) := \# \text{ internal nodes in } Tu$
  - $\text{size}(Tu) \geq 2^{bh(u)-1}$
  - $bh(T) \leq \log_2(n+1)$
  - $h(T) \leq 2\log_2(n+1)$

插入:

1. 插入点标为红
2. 出现红红冲突后看父节点的兄弟节点
  1. 红: 将其全部标记为黑, 将父节点标记为红, 向上推
  2. 黑或无: zigzig: 将中间节点上拎; zigzag: 将下面的节点上拎, 然后转化为 zigzig

删除:

1. 只交换 key 不交换颜色
2. 看 double black 的兄弟节点:
  1. 黑:
    1. 孩子全为黑, 左右减一个黑, double black 上移
    2. 右孩子为红 zigzig: 把中间点拎起 (全部标黑)
    3. 左孩子为红 zigzag: 把最后点拎起 (全部标黑)
  2. 红:  
把此节点拎起标黑, 原先的父节点标红

delete 旋转最多 3 次,  $O(1)$

## 均摊分析

- aggregation method:  $T(m)/m$
- accounting method: 借钱还钱
- potential function: 势函数

*In general, a good potential function should always assume its minimum at the start of the sequence.*

$$\begin{aligned}\sum_{i=1}^n \hat{c}_i &= \sum_{i=1}^n (c_i + \Phi(D_i) - \Phi(D_{i-1})) \\ &= \left( \sum_{i=1}^n c_i \right) + \underbrace{\Phi(D_n) - \Phi(D_0)}_{\geq 0}\end{aligned}$$

## Splay树

- w-v-u
- zig-zag: 拎u两次
- zig-zig: 先拎v, 再拎u

## B+树

- order  $B=3$  (2-3树) order = 4 (2-3-4 tree)
- internal node =  $\lceil B/2 \rceil$  (上整) - B
- root = 2-B
- leaf =  $\lceil B/2 \rceil$  (上整) - B
- # leafs  $\leq N / \lceil B/2 \rceil$  (上整)
- space =  $2N$
- height  $\leq O(\log BN)$

## Leftist heap

- unbalanced
- The null path length,  $Npl(X)$ , of any node X is the length of the shortest path from X to a node without two children.
  - $Npl(NULL) = -1$
  - $Npl(X) = \min \{ Npl(C) + 1 \text{ for all } C \text{ as children of } X \}$
- A leftist tree with r nodes on the right path must have at least  $2^r - 1$  nodes.

## Skewed tree

- 均摊分析：
  - $D_i$  = the root of the resulting tree
  - $\Phi(D_i)$  = number of heavy nodes
  - A node  $p$  is heavy if the number of descendants of  $p$ 's right subtree is at least half of the number of descendants of  $p$ , and light otherwise. Note that the number of descendants of a node **includes the node itself**.
  - 在merge过程中, The only nodes whose heavy/light status can change are nodes that are **initially on the right path**.
  - 交换子树的过程中, 原来重的都变轻了, 而我们假设轻的都变重了以此获得 $\Phi(D_i)$ 的上界
  - the right path上的light nodes最多为 $\lg n$ 个。

## Binomial Queue

- A binomial tree of height 0 is a one-node tree.
- $B_k$  consists of a root with  $k$  children, which are  $B_0, B_1, B_2, \dots$ .  $B_k$  has exactly  $2^k$  nodes. The number of nodes at depth  $d$  is  $C(d, k)$ .
- A priority queue of any size can be uniquely represented by a collection of binomial trees.
- If the smallest nonexistent binomial tree is  $B_i$ , then  $T_p = \text{Const} \cdot (i + 1)$ .
- Performing  $N$  Inserts on an initially empty binomial queue will take  **$O(N)$  worst-case time**. Hence the average time is constant.
- A binomial queue of  $N$  elements can be **built by  $N$  successive insertions in  $O(N)$  time**.

operation	binary heap	leftist heap	skewed heap	binomial heap
make-heap	$O(n)$	$O(n)$	$O(n \lg n)$	$O(n)$
find-min	$O(1)$	$O(1)$	$O(1)$	$O(1)$
insert	$O(\lg n)$	$O(\lg n)$	$O(\lg n)$	$O(\lg n)$
delete-min	$O(\lg n)$	$O(\lg n)$	$O(\lg n)$	$O(\lg n)$
merge	$O(n)$	$O(\lg n)$	$O(\lg n)$	$O(\lg n)$

## Inverted File Index

- Index is a mechanism for locating a given term in a text.
- Inverted file contains a list of pointers (e.g. the number of a page) to all occurrences of that term in the text.
- Word Stemming: 把派生词都变回原型
- Stop Words: such as "a" "the" "it"
- Distributed indexing: Each node contains index of a subset of collection
  - Term-partitioned index (A~C)
  - Document-partitioned index (1~10000)
- Dynamic indexing
- a search engine: How fast does it index/How fast does it search/Expressiveness of query language
- User happiness:
  - Data Retrieval Performance Evaluation > Response time > Index space
  - Information Retrieval Performance Evaluation > + How relevant is the answer set?

## Backtracking

- N皇后问题——**NP-Hard**
  - For the problem with n queens, there are n! candidates in the solution space.
- The Turnpike Reconstruction Problem
- Tic-tac-toe
- $\alpha$ - $\beta$  pruning: when both techniques are combined. In practice, it limits the searching to only  $O(\sqrt{N})$  nodes, where N is the size of the full game tree.
  - $\alpha$  pruning: max在上, min在下
  - $\beta$  pruning: min在上, max在下

## Greedy Algorithms

- Greedy algorithm works only if the local optimum is equal to the global optimum
- Beneath every greedy algorithm, there is almost always a more cumbersome dynamic-programming solution
- Activity Selection Problem
  - Given a set of activities  $S = \{a_1, a_2, \dots, a_n\}$  that wish to use a resource (e.g. a classroom). Each  $a_i$  takes place during a time interval  $[s_i, f_i]$ .
  - 按最早结束的排序—— $O(n \log n)$
- Huffman Codes—— $O(n \log n)$ 
  - optimal prefix tree: full tree

## Divide and Conquer

- Cases solved by divide and conquer
  - The maximum subsequence sum – the  $O(N \log N)$  solution
  - Tree traversals –  $O(N)$
  - Mergesort and quicksort –  $O(N \log N)$
- Closest Points Problem
  - Given  $N$  points in a plane. Find the closest pair of points.
  - Sort according to x-coordinates and divide;
  - Conquer by forming a solution from left, right, and cross.
- 主定理

## Dynamic Programming

- ww

## NP-Completeness

- 停机问题——incomputable (undecidable)
- computable——complexity class/P、NP、ENP
- 图灵机
  - A Deterministic Turing Machine executes one instruction at each point in time. Then depending on the instruction, it goes to the next unique instruction.
  - A Nondeterministic Turing Machine is free to choose its next step from a finite set. And if one of these steps leads to a solution, it will always choose the correct one.
- **NP—— Nondeterministic polynomial-time**
  - The problem is NP if we can prove any solution is true in polynomial time.
- **NP-complete**——An NP-complete problem has the property that any problem in NP can be polynomially reduced to it.

*If we can solve any NP-complete problem in polynomial time, then we will be able to solve, in polynomial time, all the problems in NP*

- 经典NPC问题
  - The Hamilton Cycle Problem
  - the Travelling Salesman Problem
  - the Clique Problem
  - the vertex cover problem
  - **The first problem that was proven to be NP-complete was the Satisfiability problem (Circuit-SAT)**  
*proved it by solving this problem on a nondeterministic Turing machine in polynomial time.*
- language

- $P = \{ L \subseteq \{0, 1\}^* : \text{there exists an algorithm } A \text{ that decides } L \text{ in polynomial time} \}$
- A language  $L$  belongs to NP iff there exist a two-input polynomial-time algorithm  $A$  and a constant  $c$  such that  $L = \{ x \in \{0, 1\}^* : \text{there exists a certificate } y \text{ with } |y| = O(|x|^c) \text{ such that } A(x, y) = 1 \}$ . We say that algorithm  $A$  verifies language  $L$  in polynomial time.
- complexity class co-NP = the set of languages  $L$  such that  $\bar{L} \in NP$
- 规约
  - **Hamiltonian cycle problem vs Traveling salesman problem**
    - Hamiltonian cycle problem: Given a graph  $G=(V, E)$ , is there a simple cycle that visits all vertices? 【已知NPC】
    - Traveling salesman problem: Given a complete graph  $G=(V, E)$ , with edge costs, and an integer  $K$ , is there a simple cycle that visits all vertices and has total cost  $\leq K$ ?
    - TSP is obviously in NP, as its answer can be verified polynomially.
    - **$G$  has a Hamilton cycle iff  $G'$  has a traveling salesman tour of total weight  $|V|$ .**
    - HCP  $\rightarrow$  TSP
  - **Clique problem vs Vertex cover problem**
    - Clique problem: Given an undirected graph  $G = (V, E)$  and an integer  $K$ , does  $G$  contain a complete subgraph (clique) of (at least)  $K$  vertices? 最大团 (最大子图) 【已知NPC】
    - Vertex cover problem: Given an undirected graph  $G = (V, E)$  and an integer  $K$ , does  $G$  contain a subset  $V' \subseteq V$  such that  $|V'|$  is (at most)  $K$  and every edge in  $G$  has a vertex in  $V'$  (vertex cover)? (每条边至少有一个端点被选中)
    - Clique  $\rightarrow$  vertex cover
    - **$G$  has a clique of size  $K$  iff  $G$ 的补图 has a vertex cover of size  $|V| - K$ .**

## Approximation

- **【Definition】** An *approximation scheme* for an optimization problem is an approximation algorithm that takes as input not only an instance of the problem, but also a value  $\epsilon > 0$  such that for any fixed  $\epsilon$ , the scheme is a  **$(1 + \epsilon)$ -approximation algorithm**. We say that an approximation scheme is a *polynomial-time approximation scheme (PTAS)* if for any fixed  $\epsilon > 0$ , the scheme runs in time polynomial in the size  $n$  of its input instance.
- PTAS 只要求对输入实例的规模  $n$  为多项式时间复杂度，而不考虑  $\epsilon$ 。因此当算法运行时间复杂度为  $O(n^{1/\epsilon})$  甚至  $O(n^{\exp(1/\epsilon)})$  时，仍是 PTAS 算法。
- **fully polynomial-time approximation scheme (FPTAS)** —— 要求算法对问题规模  $n$  和  $1/\epsilon$  都是多项式时间复杂度的。
- **Bin Packing —— NP-hard**
  - On-line Algorithms
    - Next Fit

- Let  $M$  be the optimal number of bins required to pack a list  $I$  of items. Then next fit never uses more than  $2M - 1$  bins. There exist sequences such that next fit uses  $2M - 1$  bins.
- First Fit
  - Let  $M$  be the optimal number of bins required to pack a list  $I$  of items. Then first fit never uses more than  $1.7M$  bins. There exist sequences such that first fit uses  $1.7(M - 1)$  bins.
- Best Fit
  - Place a new item in the tightest spot among all bins.
  - $T = O(N \log N)$  and bin no.  $\leq 1.7M$
- There are inputs that force any on-line bin-packing algorithm to use at least **5/3** the optimal number of bins.
- Off-line Algorithms
  - FFD/BFD
    - Let  $M$  be the optimal number of bins required to pack a list  $I$  of items. Then first fit decreasing never uses more than  **$11M / 9 + 6/9$**  bins. There exist sequences such that first fit decreasing uses  $11M / 9 + 6/9$  bins.
    - $3/2$
- The Knapsack Problem — **NP-hard**
- The K-center Problem — **NP-hard**
  - **Unless  $P = NP$ , there is no  $p$ -approximation for center-selection problem for any  $p < 2$ .**

## Local Search

- Vertex cover problem: Given an undirected graph  $G = (V, E)$ . Find a minimum subset  $S$  of  $V$  such that for each edge  $(u, v)$  in  $E$ , either  $u$  or  $v$  is in  $S$ .
- Hopfield Neural Networks:
  - Graph  $G = (V, E)$  with integer edge weights  $w$  (positive or negative).
  - 好边:  $w_{uv} < 0$ 
    - If  $w_e < 0$ , where  $e = (u, v)$ , then  $u$  and  $v$  want to have the same state;
    - if  $w_e > 0$  then  $u$  and  $v$  want different states.
  - The absolute value  $|w_e|$  indicates the strength of this requirement.
- **Maximum Cut Problem — NP-hard**
  - Given an undirected graph  $G = (V, E)$  with positive integer edge weights  $w_e$ , find a node partition  $(A, B)$  such that the total weight of edges crossing the cut is maximized.
  - **Unless  $P = NP$ , no  $17/16$  approximation algorithm for MAX-CUT.**
- K-L heuristic

## Randomized Algorithms

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## Parallel Algorithms

- Parallel Random Access Machine (PRAM)
- Work-Depth (WD)

- - 👉 **Exclusive-Read Exclusive-Write (EREW)**
  - 👉 **Concurrent-Read Exclusive-Write (CREW)**
  - 👉 **Concurrent-Read Concurrent-Write (CRCW)**

### Measuring the performance

👉 **Work load – total number of operations:  $W(n)$**

👉 **Worst-case running time:  $T(n)$**

- $W(n)$  operations and  $T(n)$  time
- $P(n) = W(n)/T(n)$  processors and  $T(n)$  time (on a PRAM)
- $W(n)/p$  time using any number of  $p \leq W(n)/T(n)$  processors (on a PRAM)
- $W(n)/p + T(n)$  time using any number of  $p$  processors (on a PRAM)

*All asymptotically equivalent*

- **The summation problem**
  - 串行:  $W=O(n)$   $D=O(n)$
  - 树:  $W=O(n)$   $D=O(\log n)$
- **Prefix sum: 前1、2、3.....i个数的和**
  - serial:  $W=O(n)$   $D=O(n)$
  - native:  $W=O(n^2)$  每个都算一次  $D=(\log n)$
  - 树:  $W=O(n)$   $D=O(\log n)$
- **Parallel merge sort**
  - 第i层有 $2^i$ 个节点, 每个节点有 $n/2^i$ 个元素
  - $W_v=O(n/2^i)$   $W_i=O(n)$   $W=O(n \log n)$
  - $D_v=O(n/2^i)$   $D_i=O(n/2^i)$   $D=O(n)$
  - **Merge 加快**



- input: sorted array A and B
  - output: sorted C
  - serial :  $W = O(n)$   $D = O(n)$
  - 假设已经知道  $\text{rank}(i, B)$  和  $\text{rank}(i, A)$  ——  $W = O(n)$   $D = O(1)$
  - **Ranking**
    - serial ranking ——  $W = O(n)$   $D = O(n)$
    - binary search ——  $W = O(n \log n)$   $D = O(\log n)$
    - parallel ranking
      - 把A和B每间隔k划分
      - 在组与组之间用binary search ranking ——  $W_1 = O(2n/k * \log n)$   $D_1 = O(\log n)$  【最多  $2n/k$  组】
      - 在一组中用serial ranking ——  $W_2 = O(n)$   $D_2 = O(k)$  【一组最多  $2k$  个数】
      - $W = O(n)$   $D = O(\log n)$
  - **$D = O(\log n^2)$**
  - **Maximum finding**
    - serial :  $W = O(n)$   $D = O(n)$
    - use summation alg. 【每两个选其中大的那个】 :  $W = O(n)$   $D = O(\log n)$
    - 并行比较每两个数:  $W = O(n^2)$   $D = O(1)$
    - Divide - and -conquer:
      - 把n分成根号n个子问题
      - 用并行比较每两个数来解决根号n个数
- $$W(n) = \sqrt{n}W(\sqrt{n}) + O(\sqrt{n}^2)$$
- $$D(n) = D(\sqrt{n}) + O(1)$$
- $W = O(n \log \log n)$   $D = O(\log \log n)$
- 分为k组, 用D&C找到k组中最大的数
    - $W = O(n)$   $D = O(\log \log n)$
  - random sampling
    - $W = O(n)$   $D = O(1)$  with high probability  $1 - 1/n^c$  return maximum

## External Sorting

- k-way merge
  - In general, for a k-way merge we need  $2k$  input buffers and 2 output buffers for parallel operations.
- polyphase merge
  - $k+1$  tapes for k-way merge