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Problem Set

Chapter 1

Analysis of algorithms

The analysis of algorithms is to determine the amount of resource (time/space) necessary to execute algorithms. By analyzing the resources used in algorithms, we can compare different algorithms theoretically.

The amount of resource used in an algorithm is usually represented by a function $T(n)$, where n is the length of the input. The goal of the analysis of algorithms is to find out the asymptotic growth rate of $T(n)$ in terms of n. In computer science, since n denotes the length of the input, we only care about positive values of n. Similarly, since $T(n)$ denotes the amount of resource, we usually assume that $T(n)$ is positive. This is not a real limitation, but can simplify the discussion.

Each textbook usually discusses the analysis of algorithms in the first several chapters. For those who want to know further about the analysis of algorithms, you can watch the videos of Analysis of Algorithms on Coursera and read books [3, 5, 6].

1.1 Recurrence relations

During the exam, we are often given a recurrence relation $T(n) = a(n)T(b(n)) + f(n)$, and we need to give to a tight bound of $T(n)$. If the problem statements do not explicitly specify the base care, we usually assume that $T(n) = \Theta(1)$ when n is small (less than some constant). In general, we can just focus on some particular values of n as long as these values approach infinity[1]. For example, we can only consider $n = 2^i$ for all positive integer i. Making assumptions can make the analysis easier.

Exercise 1 (*YZU CSIE 90*)

1. Prove or disprove $n^{2+\frac{\sin n}{\lg n}} = O(n^2)$.
2. If $T(n) = 49T(n/7) + n^2\cos(\sqrt{n})$, please find the tight asymptotic order of $T(n)$.

Answer of exercise 1

According to the definition of big O notation, we say $f(n) = O(g(n))$ if there exists constants c and n', such that $f(n) \leq cg(n)$ for all $n > n'$.

For the first problem, we know that the sin function will fluctuate between 1 and -1. Although $\frac{\sin n}{\lg n}$ will become very small when n approaches infinity, we still cannot get constants c and n' such that $T(n) \leq cn^2$ for all $n > n'$.

For the second problem, we know that the cos function will fluctuate between 1 and -1. So, the equation $n^2\cos(\sqrt{n}) = O(n^2)$ holds, and we can apply the master theorem to get $T(n) = O(n^2 \lg n)$.

1.1.1 Master theorem

Master theorem is the most powerful technique in solving divide-and-conquer recurrence relation. Several forms of the master theorem have been proven. Verma gives the following master theorem [7]:

[1] There should be a condition specifying what assumptions are applicable.

Theorem 1. *Let $T(n) = aT(n/b) + f(n)$ for all $n > 1$ and $T(1) = c$ for some constants $a \geq 1$, $b > 1$, $c > 0$, and non-negative function $f(n)$.*

1. *if $f(n) = O(n^{\lg_b a}/\lg n)(1+\epsilon)$ for some constant $\epsilon > 0$, then $T(n) = \Theta(n^{\lg_b a})$.*
2. *if $f(n) = \Theta(n^{\lg_b a} \lg^k n)$ for some $k \geq 0$, then $T(n) = \Theta(n^{\lg_b a} \lg^{k+1} n)$.*
3. *if $f(n) = \Omega(n^{\epsilon+\lg_b a})$ for some constant $\epsilon > 0$, and if $af(n/b) \leq kf(n)$ for some constant $k < 1$ and all sufficiently large n, then $T(n) = \Omega(f(n))$.*
4. *if $f(n) = \Theta(n^{\lg_b a}/\lg n)$, then $T(n) = \Theta(f(n) \lg n \lg\lg n)$.*

Remark. When you apply the master theorem, please pay attention to the following:

1. When the recursion involves floor or ceiling function, the master theorem does not apply. For example, $T(n) = T(\lfloor \frac{n}{2} \rfloor) + T(\lceil \frac{n}{2} \rceil) + n$.
2. In order to apply the case 2 (the extended master theorem), k must be non-negative. For example, case 2 does not apply in the case of $T(n) = 2T(\frac{n}{2}) + n/\lg n$.
3. In order to apply the case 3, the regularity condition must be satisfied. For example, case 3 does not apply in the case of $T(n) = T(\frac{n}{2}) + n(2 - \cos n)$.

Exercise 2 (*NCTU CSIE 104*)

Let $T(n) = \Theta(f(n))$. Assume that $T(n)$ is a constant for sufficiently small n. Derive $f(n)$ in the simplest formula for each of the following $T(n)$.

1. $T(n) = 2T(\frac{n}{2}) + \frac{n}{\lg^2 n}$.

Answer of exercise 2

Problem 1 By case 1, we know $T(n) = \Theta(n)$.

Problem 1 alternative solution Suppose that $n = 2^k$. We have

$$\begin{aligned} & T(n) = 2T(\frac{n}{2}) + \frac{n}{\lg^2 n} \\ \equiv \quad & T(n) = 4T(\frac{n}{4}) + \frac{n}{\lg^2 n} + \frac{n}{(\lg(n)-1)^2} \\ \equiv \quad & T(n) = 2^k T(1) + n \sum_{i=}^{k} i^{-2} \\ \equiv \quad & T(n) = n + n \sum_{i=}^{k} i^{-2} \end{aligned}$$

Since $\sum_{i=}^{k} i^{-2}$ is lower bounded by one and is upper bounded by $\sum_{i=}^{\infty} i^{-2} = \zeta(2) = \frac{\pi^2}{6}$[2], we have $T(n) = \Theta(n)$.

1.1.2 Akra-Bazzi method

Akra–Bazzi method provides a more general way to solve divide-and-conquer recurrence relations [1]. The following version is from [4].

[2] This is the solution for the Basel problem.

Theorem 2. *Let*

$$T(n) = \begin{cases} \Theta(1) & 1 \le n \le n_0 \\ \sum_{i=1}^{k} a_i T(b_i n + h_i(n)) + f(n) & \forall n > n_0 \end{cases}$$

where

1. $k \ge 1$ *is a constant and for all* i, $a_i > 0$ *and* $b_i \in (0,1)$ *are constants.*
2. $|f(n)|$ *is polynomially-bounded.*
3. *for all* i, $|h_i(n)| = O(x/\lg^2 x)$.
4. n_0 *is large enough.*

Let p *be the unique solution for* $\sum_{i=1}^{k} a_i b_i^p = 1$. *Then*

1. *if* $f(n) = O(n^{p-\epsilon})$ *for some constant* $\epsilon > 0$, *then* $T(n) = \Theta(n^p)$.
2. *if* $f(n) = \Theta(n^p)$, *then* $T(n) = \Theta(n^p \lg n)$.
3. *if* $f(n) = \Omega(n^{p+\epsilon})$ *and* $f(n)/x^{p+\epsilon}$ *is non-decreasing for some constant* $\epsilon > 0$, *then* $T(n) = \Theta(f(n))$.
4. $T(n) = \Theta\left(n^p \left(1 + \int_1^n \frac{f(u)}{u^{p+1}} du\right)\right)$.

Remark. This version of the Akra-Bazzi method can deal with floor and ceil function by picking h_i.

General version Some of the requirements can be relaxed [4].

1. The second condition is called *polynomial-growth condition* and can be replaced by the following weaker requirement: $g(n)$ is nonnegative and exist constants c_1 and c_2 such that for all i and for all $u \in [b_i n + h_i(n), n]$, we have $c_1 f(n) \le f(u) \le c_2 f(n)$.
2. The third condition can be replaced by the following weaker requirement: there exists a constant $\epsilon > 0$, $|h_i(n)| \le n/(\lg^{1+\epsilon} n)$ for all i and $n \ge n_0$.
3. The fourth condition is pretty technical and you can find the complete version in the original paper.

Drmota and Szpankowski prove a more general theorem that can deal with floor and ceil functions directly [2].

Exercise 3

Let $T(n) = \Theta(f(n))$. Assume that $T(n)$ is a constant for sufficiently small n. Derive $f(n)$ in the simplest formula for each of the following $T(n)$.

1. $T(n) = 5T(\frac{n}{5}) + n/\lg n$. [NCTU CSIE 93]
2. $T(n) = T(\frac{n}{2} + \sqrt{n}) + n$. [NTU CSIE 103]
3. $T(n) = 4T(\frac{n}{5}) + T(\frac{n}{4}) + n$. [NCTU BIOINFO 93]
4. $T(n) = T(\frac{n}{10}) + T(\frac{9n}{10}) + n$. [NTU CSIE 106]

Answer of exercise 3

Problem 1 Suppose that $n = 5^k$. We have

$$\begin{aligned} & T(n) = 5(5T(\frac{n}{25}) + n/(5(\lg_5 n - \lg_5 5))) + n/\lg n \\ \equiv \quad & T(n) = 5^k T(1) + n/(\lg n + \lg(n-1) + \cdots + 1) \\ \equiv \quad & T(n) = \Theta(n \lg\lg n) \end{aligned}$$

Problem 1 another solution apply the Akra-Bazzi method. Set $k = 1$, $a_1 = 5$, $b_1 = 1/5$, and solve $p = 1$. We get

$$T(n) = \Theta(n(1 + \int_1^n (x/\lg x)x^{-2}dx)) = \Theta(n \lg\lg n).$$

Remark. Similarly, for any constant c, we have $T(n) = cT(\frac{n}{c}) + n/\lg n = \Theta(n \lg\lg n)$.

Problem 2 apply the Akra-Bazzi method. Set $k = 1$, $a_1 = 1$, $b_1 = 1/2$, and solve $p = 0$. Then, we get

$$T(n) = \Theta(1(1 + \int_1^n (x/x)dx)) = \Theta(n).$$

Problem 3 apply the Akra-Bazzi method. Set $k = 2$, $a_1 = 4$, $b_1 = 1/5$, $a_2 = 1$, $b_2 = 1/4$, and solve $p \approx 1.03$. Then, we get

$$T(n) = \Theta(n^p), \text{where } p \approx 1.03.$$

Problem 4 apply the Akra-Bazzi method. Set $k = 2$, $a_1 = 1$, $b_1 = 1/10$, $a_2 = 1$, $b_2 = 9/10$, and solve $p = 1$. Then, we get

$$T(n) = \Theta(n \lg n).$$

Exercise 4 (*NCTU CSIE 92*)

Given positive constants c', $c_1, c_2, \ldots, c_k$, assume that $T(n) \le c'n + \sum_{i=1}^{k} T(c_i n)$ and $\sum_{i=1}^{k} c_i < 1$. Prove $T(n) = O(n)$.

Answer of exercise 4

Apply the Akra-Bazzi method.

1.1.3 Full-history recurrence

Exercise 5 (*NTU CSIE 90*)

Let $T(n) = \Theta(f(n))$. Assume that $T(n)$ is a constant for sufficiently small n. Derive $f(n)$ in the simplest formula for each of the following $T(n)$.

1. $T(n) = n + \frac{4}{n}\sum_{i=1}^{n-1} T(i)$.

Answer of exercise 5

Problem 1

$$\begin{aligned}
T(n) &= n + \frac{4}{n}\sum_{i=1}^{n-1} T(i) \\
\equiv \quad nT(n) &= n^2 + 4\sum_{i=1}^{n-1} T(i) \\
\equiv \quad (n+1)T(n+1) &= n^2 + 4\sum_{i=1}^{n} T(i) \\
\equiv \quad (n+1)T(n+1) - nT(n) &= 2n + 1 + 4T(n) \\
\equiv \quad (n+1)T(n+1) &= (n+4)T(n) + 2n + 1 \\
\equiv \quad \frac{T(n+1)}{(n+2)(n+3)(n+4)} &= \frac{T(n)}{(n+1)(n+2)(n+3)} + \frac{2n+1}{(n+1)(n+2)(n+3)(n+4)}
\end{aligned}$$

Let $S(n) = \frac{T(n)}{((n+1)(n+2)(n+3))}$.

$$\begin{aligned} & S(n+1) = S(n) + \frac{2n+1}{(n+1)(n+2)(n+3)(n+4)} \\ \equiv \quad & S(n) = \sum_{i=0}^{n-1} \frac{2i+1}{(i+1)(i+2)(i+3)(i+4)} \\ \equiv \quad & S(n) = \Theta(1) \\ \equiv \quad & T(n) = (n+1)(n+2)(n+3)S(n) \\ \equiv \quad & T(n) = \Theta(n^3) \end{aligned}$$

1.1.4 Range transformation

Exercise 6 (*NCTU CSIE 93*)

Let $T(n) = \Theta(f(n))$. Assume that $T(n)$ is a constant for sufficiently small n. Derive $f(n)$ in the simplest formula for each of the following $T(n)$.

1. $T(n) = \sqrt{n}T(\sqrt{n}) + \sqrt{n}$.

Answer of exercise 6

Problem 1 The trick is to divide n by both side and then expand. $\frac{T(n)}{n} = \frac{T(\sqrt{n})}{\sqrt{n}} + n^{-0.5}$. Suppose that $n = 2^{2^k}$. Let $S(k) = T(2^{2^k})/2^{2^k}$.

$$\begin{aligned} & T(n) = S(k) = S(k-1) + 2^{-2^{k-1}} \\ \equiv \quad & S(k) = \Theta(1) + \sum_{i=1}^{k-1} 2^{-2^{k-1}} = \Theta(1) \\ \equiv \quad & T(n) = n \cdot S(k) = n \cdot \Theta(1) = \Theta(n) \end{aligned}$$

1.1.5 Recursion trees

Exercise 7

Let $T(n) = \Theta(f(n))$. Assume that $T(n)$ is a constant for sufficiently small n. Derive $f(n)$ in the simplest formula for each of the following $T(n)$.

1. $T(n) = 2T(\sqrt{n}) + \lg n$. [NCKU CSIE 95]
2. $T(n) = nT(\sqrt{n}) + n^2 \lg n$. [NTU CSIE 98]

Answer of exercise 7

Problem 1 Suppose that $n = 2^{2^k}$. We have

$$\begin{aligned} & T(n) = 2^2 T(\sqrt[4]{n}) + 2(\lg n - 1) + \lg n \\ \equiv \quad & T(n) = 2^k T(1) + \lg n + 2(\lg n)/2 + 4(\lg n)/4 + ... + 2^k(\lg n)/2^k \\ \equiv \quad & T(n) = 2^k T(1) + \lg n \sum_{i=1}^{k} 1 \\ \equiv \quad & T(n) = \Theta(\lg n \lg\lg n) \end{aligned}$$

Problem 2 Suppose that $n = 2^{2^k}$. We have

$$\begin{aligned} & T(n) = n^{1.5} T(\sqrt[4]{n}) + \frac{n^2 \lg n}{2} + n^2 \lg n \\ \equiv \quad & T(n) = n^2 T(1) + n^2 \lg n \sum_{i=0}^{k} 2^{-i} \\ \equiv \quad & T(n) = \Theta(n^2 \lg n) \end{aligned}$$

1.1.6 Comparison

*** Exercise 8 (*NCTU BIOINFO 93*)**

Let $T(n) = \Theta(f(n))$. Assume that $T(n)$ is a constant for sufficiently small n. Derive $f(n)$ in the simplest formula for each of the following $T(n)$.

1. $T(n) = T(\frac{n}{2}) + T(\sqrt{n}) + n$. [NCTU BIOINFO 93]

Answer of exercise 8

Problem 1 By the recurrence relation, we have $T(n) = \Omega(n)$. Let $F(n) = F(n/2) + F(n/3) + n$. Since $n/3 > \sqrt{n}$ for all $n > 9$, we have $F(n) \geq T(n)$. By using the Akra-Bazzi method, we get $F(n) = \Theta(n)$. Thus, we have $T(n) = \Theta(n)$.

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Chapter 2

Problems on sequences

Arrays are the simplest data structures in computers. Although the structure of an array is very simple, there are still some hard problems. In this chapter, we will go through several hard problems of arrays.

2.1 Searching

2.1.1 k-sum problem

Exercise 9

1. Given three sets of integers X, Y, and Z, and another integer k, we want to know if there exists three numbers x, y, and z, such that $x \in X$, $y \in Y$, and $z \in Z$, and $x + y + z = k$. Design an algorithm to solve this problem. A naive method by checking all possible sums of triples $(x + y + z)$ will take $\Theta(|X||Y||Z|)$ time. Your algorithm must be more efficient than it. Analyze the time complexity of your algorithm. [NCU CSIE 99]
2. Given an array of n numbers, and a number s, determine whether the array contains 4 elements whose sum is s. Analyze the time efficiency of your algorithm. Your algorithm should be more efficient than $O(n^4)$. [NCU CSIE 93]

Answer of exercise 9

Problem 1 This problem is called the 3SUM problem. Construct a set $Z' = \{k - z \mid z \in Z\}$. The problem is equal to finding $x \in X$, $y \in Y$, and $z \in Z'$, such that $x + y = z$. Sort the sets Y and Z'. For each element $x \in X$, construct a set $Y_x = \{y + x \mid y \in Y\}$. If one element a in both Y_x and Z' exists, then we find an answer. When Y_x is fixed, finding common elements in sorted Y_x and Z' can be done in $O(|Y| + |Z|)$ time. Since we apply this operation for each element $x \in X$, the time complexity is $\Theta(|X||Y||Z|)$.

Remark. This problem can be solved in $o(n^2)$ time [5].

Problem 2 Let a denote the input array. Let S_{ij} be the sum of $a[i]$ and $[j]$. Sort all S_{ij} in increasing order and store them in an array A of length L. The problem is equal to finding two elements in A sum to s, which can be solved in time linear in L. Since the number of pairs is $O(n^2)$, the sort takes $O(n^2 \lg n)$ time. Moreover, the search takes $O(n^2)$ time. Thus, the time complexity is $O(n^2 \lg n)$.

2.1.2 Binary search

Exercise 10

1. Let $A[n]$ be an array with n elements sorted in ascending order. It is simple to construct an $O(\lg n)$ algorithm to find the position k in $A[n]$ for a given value v. Assume that k is much less than n (i.e.,

$k \ll n$). Write an $O(\lg k)$ time algorithm to search for k. (Note: you do not know the value of k in advance, only v is known) [YZU CSIE 93]

2. Suppose you are given an array of n sorted numbers that has been circularly shifted m positions to the right. For example $\{36, 45, 5, 18, 26, 29\}$ is a sorted array that has been circularly shifted $m = 2$ positions to the right. Give an $O(\lg n)$ algorithm to find the largest number in this array. Note that you don't know what m is. [CYCU CSIE 89]

   Give an efficient algorithm to determine if there exists an integer i such that $A_i = i$ in an array of integers $A_1 < A_2 < \cdots < A_n$. What is the running time of your algorithm? [NDHU CSIE 96, CYCU CSIE90]

Answer of exercise 10

Problem 1 We use linear search to find an index i such that $2^i \leq k < 2^{i+1}$, which takes $O(\lg k)$ time. Then, we apply binary search for find $x = A[k]$ for $k \in [1 \ldots 2^{i+1}]$, which takes $O(\lg k)$ time as well.

Remark. This technique is called exponential search [1].

Problem 2 Let $m = \frac{n}{2}$. We have three cases:

1. $A[m] > A[m+1]$: $A[m]$ is the maximum.
2. $A[m] > A[1]$: the maximum locates in $A[m+1] \sim A[n]$. Recurse.
3. $A[m] < A[1]$: the maximum locates in $A[1] \sim A[m-1]$. Recurse.

Since the problem size is halved in each recursion, the time complexity is $O(\lg n)$.

Problem 3 Let $m = \frac{n}{2}$. We have three cases:

1. $A[m] = m$: $A[m]$ is the fixed point.
2. $A[m] > m$: the fixed point locates in $A[1] \sim A[m-1]$. Recurse.
3. $A[m] < m$: the fixed point locates in $A[m+1] \sim A[n]$. Recurse.

Since the problem size is halved in each recursion, the time complexity is $O(\lg n)$.

2.1.3 Majority problem

Exercise 11

1. An array $a[1, \ldots, n]$ is said to contain a majority element if there is some element that appears more than $\frac{n}{2}$ times in the array. The task is to determine if a has a majority element and, if so, to find the element. We do not assume that the elements of the array come from some ordered domain such as the integers, so we cannot sort the array or perform comparisons such as $a[i] < a[j]$. However, we assume we are able to test if $a[i] = a[j]$ in constant time. Can you come up with a simple $O(n)$-time algorithm for this problem? [CCU CSIE 95]
2. Given an n-element array A of real numbers, design an $O(n)$ time algorithm which determines whether any value occurs more than $\frac{n}{7}$ times in A. [NTUT CSIE 102]

Answer of exercise 11

Problem 1 The algorithm is based on the following observation: Let x and y be two distinct elements in a. Discarding x and y from a will keep the majority unchanged.

We choose the first element as a candidate and set M as one. If the next element differs from the candidate, then decrease M by one. If M becomes zero, then choose the next element as a candidate. Repeat this process until all elements are processed. Finally, we need one more pass to test whether the candidate is indeed a majority. Since this is a two pass algorithm, the algorithm runs in linear time.

Remark. More information can be found in [7].

Problem 2 If one element occurs more than $\frac{n}{7}$ times in A, then the element must be one of $\frac{kn}{7}$-th largest element in A, where $1 \leq k \leq 7$. Thus, we can use the selection algorithm to find all $\frac{kn}{7}$-th largest elements in A for all k in linear time. For each of them, determine whether this element occurs more than $\frac{n}{7}$ times in A can also be done in linear time. Thus, the time complexity is linear.

Remark. For any threshold value t, finding element with frequency larger than $\frac{n}{t}$ can be done in linear time (independent from t) [6].

2.1.4 Saddleback search

Exercise 12 (*NTU CSIE 93*)

M is an $n \times n$ integer matrix in which the entries of each row are in increasing order (reading left to right) and the entries in each column are in increasing order (reading top to bottom). Give an efficient algorithm to find the position of an integer x in M, or determine that x is not there. Tell how many comparisons of x with matrix entries your algorithm does in the worst case.

Answer of exercise 12

For any pair of indices , i and j, if $M[i][j] > x$, then we know $M[i][k] \neq x$ for all $j \leq k \leq n$ and $M[k][j] \neq x$ for all $i \leq k \leq n$. Similarly, if $M[i][j] < x$, then we know $M[i][k] \neq x$ for all $1 \leq k \leq j$ and $M[k][j] \neq x$ for all $1 \leq k \leq i$.

The idea is as follows: in each iteration, we compare an entry with x and the eliminate either a row or a column.

We always compare x of the top-right element of the remaining submatrix. If this entry equals x, then we found the answer. If this entry is smaller x, then the top row can be discarded. If this entry is larger than x, the the rightmost column can be discarded. Since the matrix has $O(n)$ rows and columns, the time complexity is $O(n)$.

Remark. This technique is called *saddleback search* [2].

2.2 Sorting

* Exercise 13 Pancake sorting problem (*NCKU CSIE 102*)

Jeremy is a waiter working in a restaurant. The chef there is sloppy; when he prepares a stack of pancakes, they come out all different sizes. When Jeremy delivers the pancakes to the customer, he wants to rearrange them by grabbing several from the top and flipping them over on the way. After repeating this for several times, the smallest pancake is on top, and so on, down to the largest at the bottom. If there are n pancakes, how many flips are required? Design an algorithm to help Jeremy, and analyze its time complexity.

Answer of exercise 13

This problem is called the pancake sorting problem. For a stack of pancakes, first locate the largest pancake. Then flip the largest pancake to the top by using one flip and use another flip to move the largest pancake to the bottom. Sort the top $n-1$ pancakes recursively. Since every pancake except the smallest one needs at most two flips to move to the correct position, the number of flips is $2n-2$.

Remark. One algorithm with at most $\frac{18}{11}n$ flips exists [3].

Exercise 14 (*NTUT CSIE 95*)

Let A be an array of n arbitrary and distinct numbers. A has the following property: If we imagine B as being sorted version of A, then any element that is at position i in array A would, in B, be at a position j such that $|i-j| \leq k$. In other words, each element in A is not farther than k positions away from where it belongs in the sorted version of A. Suppose you are given such an array A, and you are told that A has this property for a particular value k (that value of k is also given to you). Design an $O(n \lg k)$ time algorithm for sorting A.

Answer of exercise 14

Initially, insert the first $2k$ elements into a min heap; then, repeat the following process $n-2k$ times. In iteration i, delete the minimum value in the heap and output it; then, insert the $(2k+i)$-th element into the heap. After loops terminates, sort the elements in the heap by the heap sort and output it.

For any position i in the sorted array, the possible positions in the unsorted array are from $i-k$ to $i+k$, so using a min heap guarantees the correctness. Since the size of heap is $O(k)$, the insertion and deletion takes $O(\lg k)$ time. Since we insert and delete $O(n)$ times, the time complexity is $O(n \lg k)$.

Exercise 15 (*NCU CSIE 98*)

The input is a sequence of n integers with many duplications, such that the number of distinct integers in the sequence is $O(\lg n)$. Design a sorting algorithm to sort such sequences using at most $O(n \lg \lg n)$ comparisons in the worst case.

Answer of exercise 15

Augment an AVL-tree by creating a count field in each tree node. First process all elements in the input sequentially. For each integer x in input, search for x in the tree; if x has been inserted, then increment the x's count; otherwise, insert x into the tree and initialize the count to 1. Then, in-order traverse the data structure and output the elements with its multiplicity. Since the distinct integers is $O(\lg n)$, both of search and insertion take $O(\lg \lg n)$ time. Since the number of insertion and search is $O(n)$, the time complexity $O(n \lg \lg n)$.

2.2.1 Bucket sort

Exercise 16 (*NDHU CSIE 97*)

We are given n points in an unit circle, $p_i = (x_i, y_i)$, such that $0 < \sqrt{x_i^2 + y_i^2} < 1$, for $i = 1, \ldots, n$. Suppose that the points are uniformly distributed; that is, the probability of finding a point in any region of the circle is proportional to the area of that region. Design and prove a $\Theta(n)$ expected-time algorithm to sort the n points by their distance $d_i = \sqrt{x_i^2 + y_i^2}$ from the origin. (Hints: Design the bucket sizes in Bucket-Sort to reflect the uniform distribution of the points in the unit circle.)

Answer of exercise 16

Divide the circle into n concentric circles with radii $\sqrt{\frac{1}{n}}, \sqrt{\frac{2}{n}}, \ldots, 1$. The area of the ith level is $\pi\sqrt{\frac{i}{n}}^2 - \pi\sqrt{\frac{i-1}{n}}^2 = \pi/n$. Since the areas are equal-size, the probability of point locates in any level is $1/n$. So we can partition the points by their levels and use the bucket sort to sort these points.

2.3 Order statistics

* Exercise 17 (*NTUST CSIE 98*)

It is trivial to find the median of the integers in the sorted array a with median $= a[\lfloor \frac{n}{2} \rfloor]$. Suppose we have $3n$ distinct integers that are randomly stored in arrays $a[0 \ldots n-1]$, $b[0 \ldots n-1]$, and $c[0 \ldots n-1]$, and each array is sorted independently. Write an algorithm to find the median of these $3n$ distinct integers. Please note that you are not allowed to merge arrays a, b, and c into a $3n$ integer array and then perform sorting.

Answer of exercise 17

The idea is as follows: in each iteration, we eliminate some elements that cannot be the answer. Repeat this process until only one element remains.

Let N be the number of all remaining elements and our goal is to find the k-th smallest element. Initially, we have $N = 3n$ and $k = \frac{N}{2}$. In each iteration, we collect the middle elements of all arrays as a set P. If $k < \frac{N}{2}$, then we pick the array with the maximum middle element. In this array, the elements that are larger than the middle element must be larger than $\frac{N}{2}$ elements in all elements. Thus, we can remove these elements

safely. The case in which $k > \frac{N}{2}$ is symmetric. We adjust the k value and recurse on the remaining elements. Since one array is halved in each iteration, the number of iterations is $O(\lg n)$. Since each iteration takes constant time, the time complexity is $O(\lg n)$.

Exercise 18 (*NCU CSIE 96*)

Suppose that k workers are given the task of scanning through a shelf of books in search of a given piece of information. To get the job done efficiently, the books are to be partitioned among k workers. To avoid the need to rearrange the books, it would be simplest to divide the shelf into k regions and assign each region to one worker. Each book can only be scanned by one worker. you are asked to find the fairest way to divide the shelf up. For example, if a shelf has 9 books of sizes 100, 200, 300, 400, 500, 600, 700, 800 and 900 pages, and $k = 3$, the fairest possible partition for the shelf would be

100 200 300 400 500 | 600 700 | 800 900

where the largest job is 1700 pages and the smallest job 1300. In general, we have the following problem: Given an arrangement S of n nonnegative numbers, and an integer k, partition S into k regions so as to minimize the difference between the largest and the smallest sum over all regions. Design algorithm to find an optimal solution for any given $k \leq n$.

Answer of exercise 18

For a number d, we can verify whether these n books can be partitions into k parts with the maximum load at most d in linear time. Moreover, the minimum feasible d value must be $\sum_{k=i}^{j} S_k$ for some i and j. Create a matrix M, such that $M[i][j] = \sum_{k=i}^{j} S_k$; note that all rows and columns in M are in increasing order. Our goal is to find a pair of indices, i and j, such that $M[i][j]$ is the minimum feasible d value.

The idea is to eliminate a constant fraction of elements of M iteratively until only one element remains; the remaining candidate is the answer. Initially, all elements in all columns are candidates. Since the algorithm will remove elements gradually, for each column, we maintain a sub-column of remaining elements.

In each iteration, first pick all median elements in all sub-columns. Then, find the median, m, among all medians. If m leads to a feasible partition, then we know that the true answer is no greater than m; otherwise, the true answer is no more than m. We apply saddleback search to prune all candidates that cannot be the answer.

Collecting all median elements and find the median among all medians takes $O(n)$ time. Since saddleback search also takes $O(n)$ time, each iteration takes $O(n)$ time as well. Because we eliminate at least a quarter of candidates in each iteration, the time complexity is $O(n \lg n)$.

Remark. This problem can be solved in linear time [4].

2.4 String

* Exercise 19 Longest common substring problem (*NTU CSIE 97*)

Given two length-n binary strings A and B, consider the problem of computing a longest string C that is a substring of both A and B. You are asked to prove or disprove that this so-called *longest common substring problem* can be solved in $O(n \lg n)$ time.

Answer of exercise 19

This problem is called the longest common substring problem. Build a generalized suffix tree from A and B. Find the deepest internal node that has leaves from two strings. The string corresponding to the path from the root to the deepest internal node is the longest common substring. Since building a suffix tree takes linear time, the time complexity is linear.

2.5 Others

Exercise 20 Cycle detection problem (*MCU CSIE 95*)

Design an algorithm which detects whether there exists a cycle within a singly linked list. (Suppose that each node in this list contains data and next fields for storing data and the pointer to the next node respectively.) Please analyze the time and space complexities of your algorithm.

Answer of exercise 20

This problem is identical to the cycle detection problem. Maintain two pointers, fast and slow, and traverse the linked list from head in parallel. The fast pointer traverse the list two nodes per iteration, while the slow pointer traverses one node. If the list contains a cycle, then these two pointers will point to the same node in some iteration. Otherwise, the fast pointer will achieve the end of the list. The time complexity is linear and the space usage is constant.

Remark. More information can be found in [8].

Exercise 21 (*NCTU CSIE 91*)

Given a finite set A and a mapping function f from A to itself, describe an algorithm to find a subset S of A with maximum size such that f is one-to-one when restricted to S.

Answer of exercise 21

The idea is as follows: we repeatedly remove one element $x \in S$ such that $f(y) \neq x$ for all $y \in S$ until no such an element exists.

In implementation, we maintain an hashtable h, where $h[x]$ stores how many elements y satisfy $f(y) = x$. In each iteration, eliminate an element x with $h[x] = 0$, and decrement $h[f(x)]$, until $h[x] \neq 0$ for all x. The time complexity is $O(n^2)$.

2.5.1 Query support

Exercise 22 (*NTHU CSIE 100*)

Suppose we have n ranges $[a_1, b_1], [a_2, b_2], \ldots, [a_n, b_n]$, where all a_i's are negative and all b_i's are positive. We are asked to preprocess these ranges so that for any input value x, we can efficiently count the number of the ranges containing x. Design an efficient representation of the ranges so that the desired counting can be done in $O(\lg n)$ time. Your representation must take $O(n)$ space. Describe you representation and how to answer a query in $O(\lg n)$ time in details.

Answer of exercise 22

Suppose that x is a positive number. Since a_i is negative and b_i is positive, we only need to count the number of intervals whose b_i is larger than x. If we preprocess the ranges in order by b_i, counting the can be done in $O(\lg n)$ time by using binary search. The case in which x is a negative number is symmetric.

* Exercise 23 Range minimum query problem (*NCTU BIOINFO 100*)

Suppose that we are given a sequence of n unsorted values, say $x_1, x_2, \ldots, x_n$, and at the same time, we are asked to quickly answer repeated queries defined as follows: Given i and j, where $1 \leq i \leq j \leq n$, find the smallest value in $x_i, x_{i+1}, \ldots, x_j$. Please design a data structure that uses $O(n)$ space and answers the queries in $O(\lg n)$ time.

Answer of exercise 23

This problem is called the range minimum query problem. We first divide n values into blocks of size $\lg n$. For a given query (i, j), since each block has $O(\lg n)$ elements, the minimum among all elements in i's block that are after i can be found in $O(\lg n)$ time. Similarly, the minimum among all elements in j's block that are before j can be found in $O(\lg n)$ time as well.

The problem becomes how to find the minimum of all full blocks between i and j in $O(\lg n)$ time. We can precompute a matrix M, such that $M[s][k]$ is the minimum in blocks $s, s+1, \ldots, s+2^k$. The minimum of all full blocks between i and j can be found in $O(\lg n)$ time by reading M. Since the number of blocks is $O(\frac{n}{\lg n})$, the size of M is $O(\frac{n}{\lg n} \lg \frac{n}{\lg n}) = O(n)$.

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Chapter 3

Problems on graphs

3.1 Tree

Exercise 24 Vertex cover on trees (*NTUT CSIE 101*)

Let T be an n-node tree rooted at some node r. We want to place as few guards as possible on nodes in T, such that every edge of T is guarded: an edge between a parent node v and its child w is guarded if one places a guard on at least one of these two nodes v, w. Give an $O(n)$ time algorithm for finding an optimal solution to the problem. Please show the analysis on the time and correctness of your algorithm

Answer of exercise 24

For any edge incident with a leaf, since the edge should be guarded, either the leaf node or its parent node should be chosen. Choosing the parent node not only can guard this edge but only guard the edge incident with the parent node. Thus, choosing the parent node is no worse than choosing the leaf node. Hence, we can choose all nodes that connect to leaf nodes. After removing all edges that are guarded, we get a smaller tree. We can recursively apply this procedure to get a minimum vertex cover of the tree. The time complexity is $O(n)$.

3.1.1 Lowest common ancestor

Exercise 25

1. Let T be a binary tree rooted at r with vertex set V and edge set E. Suppose it is represented using adjacency list format. If node u is an ancestor of v, there is a path from r to v passing through u. Consider the function ancestor(u, v) which returns TRUE if u is a ancestor of v and FALSE otherwise. In order to have this function run in $O(1)$ time, we are asked to design an algorithm to preprocess the tree. Please provide a linear-time, i.e., $O(|V| + |E|)$ time algorithm for this preprocess. [NTUT CSIE 100]
2. a) Let T be a binary search tree, where each vertex contains a pointer to its parent and pointers to its children, and also a field named temp, which is of type integer. You are given two pointers q_1, q_2, pointing to two vertices v_1, v_2 in T. Find in time $O(k)$ what is the length of the shortest path in T connecting v_1 to v_2, where k is the length of this path. You may assume that before the execution of your program, the value of all the "temp" fields is zero, and you can use these fields for your algorithm.
   b) Same as above, but this time the "temp" fields do not exist, you cannot write on the tree (so its information is "read-only" and the expected running time of your algorithm should be $O(k)$. Hint: Assume that T is stored in the memory of your computer and you can find, in time $O(1)$, what is the address in which each node is stored. [CCU CSIE 93]

Answer of exercise 25

Problem 1 We can use the pre-order traversal and post-order traversal on the tree to get each node's pre-order and post-order numbers. A node u is an ancestor of a node v if and only if u's pre-order number is smaller than v's pre-order number and u's post-order number is larger than v's post-order number. Since pre-order traversal and post-order traversal take linear time, the preprocess takes linear time.

Problem 2 Question a: Initially, let two pointers a and b point to v_1 and v_2 respectively and set their temp fields to 1. In each iteration, move a and b to their parent nodes respectively. If their temp fields are zero, then set them to one and repeat; otherwise, we find the solution. Suppose that a's temp field is one without loss of generality, then a is the lowest common ancestor of v_1 and v_2. If a path between v_1 and v_2 of length k exists, then the process will terminate in k iterations. Thus, the time complexity is $O(k)$.

Question b: If the temp field does not exist, then we can use a hashtable instead. Thus, the time complexity is $O(k)$ in expectation assuming universal hash function is used.

3.2 Traversal

3.2.1 DFS

Exercise 26

1. An undirected graph $G = (V, E)$ is stored in a text file with the following format: The first line contains two integer numbers n and m that denote the numbers of vertices and edges of G respectively. Then, the first line is followed by m lines. Each line contains two distinct integers, say i and j, indicating that there is an edge between vertices i and j. Given such a file, design an $O(n)$ time algorithm to test if the undirected graph represented by the file is a tree. You should specify the data structure used to store the graph in the memory and how you construct such a data structure. [NCU CSIE 96, NTU CSIE 99]
2. For an undirected graph $G = (V, E)$, a vertex $v \in V$, and an edge $(x, y) \in E$, let $G \setminus v$ denote the subgraph of G obtained by removing v and all the edges incident to v from G; and let $G \setminus (x, y)$ denote the subgraph of G obtained by removing the edge (x, y) from G. If G is connected, then $G \setminus v$ can be disconnected or connected.
   a) Given a connected graph G, design an $O(|V|)$ time algorithm to find a vertex $v \in G$ such that $G \setminus v$ is connected.
   b) Given a connected graph G, design an $O(|V|)$ time algorithm to either find an edge $(x, y) \in G$ such that $G \setminus (x, y)$ is connected or report that no such an edge exists. [NCU CSIE 102]

Answer of exercise 26

Problem 1 Without loss of generality, suppose that $m = n - 1$, otherwise it is not a tree obviously. Since a tree is a connected graph without a cycle, we can use DFS to check whether the graph contains a cycle. Because DFS takes $O(|V| + |E|)$ time, the time complexity is $O(n)$.

Problem 2 For Question a, no such an algorithm exists.[1] For Question b, use DFS to find a cycle and any edge on this cycle can be removed without disconnecting the graph. If no cycle exists, this graph is a tree and every edge cannot be removed. The time complexity is $O(|V|)$.

3.2.2 BFS

Exercise 27

1. Given is a directed graph $G = (V, E)$ represented via adjacency lists and a vertex $v_a \in V$. Design an algorithm that outputs the length of the shortest cycle containing v_a in G. your algorithm should solve the problem in $O(|V| + |E|)$ time. [NTHU CSIE 95]

[1] http://cstheory.stackexchange.com/a/33703/28542

2. We have a directed graph $G = (V, E)$ represented using adjacency lists. The edge costs are integers in the range $\{1, 2, 3, 4, 5\}$. Assume that G has no self-loops or multiple edges. Design an algorithm that solves the single-source shortest path problem on G in $O(|V| + |E|)$. [NTHU CSIE 95]

Answer of exercise 27

Problem 1 Use BFS to traverse from v_a. Find the first back edge from vertex u back to v_a. This cycle is the shortest cycle containing v_a. The time complexity is $O(|V| + |E|)$.

Problem 2 The idea is to transform the problem to an unweighted graph and use BFS to find the shortest path. Construct a graph $G' = (V \cup V', E')$ as follows:. For every edge $(u, v) \in E$ of weight k, if $k = 1$, then add an edge (u, v) to E'. Otherwise, add $k - 1$ vertices $uv_1 \sim uv_{k-1}$ to V' and add edges (u, uv_1), (uv_{k-1}, v), and (uv_i, uv_{i+1}) for all $1 \leq i \leq k - 2$ to E'. This transformation does not create path between any vertex in V. Moreover, if one path from u to v of cost C exists in G, then one path from u to v of C edges also exists in G'. Since k is at most 5, the size of V' and E' is $O(|E|)$. Thus, the time complexity is $O(|E|)$.

3.2.3 Topological sort

Exercise 28 (*CYCU CSIE 92*)

Professor Lee wants to construct the tallest tower possible out of building blocks. She has n types of blocks, and an unlimited supply of blocks of each type. Each type-i block is rectangular solid with linear dimension (x_i, y_i, z_i). A block can be oriented so that any two of its three dimensions determine the dimensions of a base and the other dimension is the height. In building a tower, one block may be placed on top of another block as long as the two dimensions of the lower block. (Thus, for example, blocks oriented to have equal-sized bases cannot be stacked.) Use graph model to design an efficient algorithm to determine the tallest tower that the professor can build. Analyze the run time complexity.

Answer of exercise 28

For each type i block, construct three nodes, v_{i1}, v_{i2}, and v_{i3}, corresponding to three faces, $x_i \times y_i$, $y_i \times z_i$, and $x_i \times z_i$. If one type j block can be placed on top of a type i block, then create the edges between the corresponding faces and the edge's weight is the height of type j block. The longest path in this graph is the tallest tower can be built. Since the graph is a directed acyclic graph, we can find the longest path in $O(|V| + |E|)$ time by using topological sort. Bacause the graph has $3n$ vertices and $O(n^2)$ edges, the time complexity is $O(n^2)$.

3.3 Path

Exercise 29 Johnson's algorithm (*NTPU CSIE 100*)

Given a graph $G = (V, E)$ and a weight function $w : E \to \mathbb{R}$, describe a method to decide whether there is a function $h : V \to \mathbb{R}$ such that the new weight function w_h defined by $w_h = w(u, v) + h(u) - h(v)$ is non-negative.

Answer of exercise 29

Pick a vertex s and solve shortest path problem from s. If no negative cycle exists in the graph, the shortest distance between s and vertex v is a candidate of $h(v)$, since we know that $h(v) \leq h(u) + w(u, v)$ by the property of the shortest paths. On the other hand, if a negative cycle exists in G, the function h can not exist.

Exercise 30 Arbitrage (*NCTU CSIE 96*)

Given an N by N positive matrix R (i.e., each entry $R[I, J]$ is positive) design an efficient algorithm to determine whether or not there exists a sequence of distinct indices: $I_1, I_2, \ldots, I_k$, where $1 \leq k \leq N$, such

that $R[I_1, I_2] \times R[I_2, I_3] \times \cdots \times R[I_{k-1}, I_k] \times R[I_k, I_1] > 1$. State your algorithm precisely and analyze the running time of your algorithm. [NCTU CSIE 96]

Answer of exercise 30

The idea is to reduce this problem to finding a negative weight cycle in graph and use Bellman-Ford algorithm to solve it. Let $A[i, j] = -\lg R[i, j]$. We know that one sequence satisfies the problem's criterion, if and only if, one negative weight cycle exists in the graph represented by A. Since the transformation take $O(|V|^2)$ time and the time complexity of Bellman-Ford algorithm is $O(|V||E|) = O(|V|^3)$, the time complexity is $O(|V|^3)$.

3.4 Spanning tree

Exercise 31

1. Consider the following variation of the Minimum Spanning Tree problem: Given a graph G of n vertices and m edges AND a minimum spanning tree T of graph G, we wish to add new edge e with weight w_e to G forming a new graph G' and construct the new minimum spanning tree of the new graph G'. Give an algorithm which constructs the minimum spanning tree of G' in $O(n)$ time. [NCU CSIE 102]
2. Suppose that a graph G has a minimum spanning tree already computed. How quickly can the minimum spanning be updated if a new vertex and incident edges are added to G? Please justify your answer. [NTUT CSIE 98]

Answer of exercise 31

Problem 1 Add e to T to form a cycle. Remove the largest weight edge in this cycle then we get the new minimum spanning tree.

Problem 2 Let T be the original MST. When we add a new vertex v to T, we need to check all cycles and then remove the edge with the largest weight from each cycle. This can be done by using DFS on T as follows: during the recursive call of DFS, we maintain the edge with the largest weight, m, from the current node, r, to v through all explored descents. For a new explored descent d, we find the edge with the largest weight, t, from the descent to v through its subtree. Since m, t, and (r, d) will form a cycle, we only keep the smaller two among these three edges. The total complexity is $O(|T|)$.

Remark. After deleting edges/vertices, new MST can also be found efficiently [1, 2].

3.5 Matching

Exercise 32 Edge cover problem (*NTHU CSIE 101*)

Let $X = \{1, \ldots, n\}$. For a subset of X, we say that it covers its elements. Given a set $\mathcal{S} = \{S_1, S_2, \ldots, S_m\}$ of m subsets of X such that $\cup_{i=1}^{m} S_i = X$, the set cover problem is to find the smallest subset T of S whose union is equal to X, that is, $\cup_{S_i \in T} S_i = X$. Suppose that each subset $S_i \in \mathcal{S}$ contains only two elements. Can the set cover problem then be solved in polynomial time? If yes, please also design a polynomial-time algorithm to solve this set cover problem and analyze its time complexity.

Answer of exercise 32

Since each subset S_i has only two elements, we can create a graph $G = (X, S)$. Finding the minimum set cover problem is equal to finding the minimum edge cover on G, which can be solved in $O(n^4)$ time.

3.6 Network flow

Exercise 33 (*NTU CSIE 100*)

The escape problem is defined as the following. An $n \times n$ grid is an undirected graph consisting of n rows and n columns of vertices. We denote the vertex in the i-th row and j-th column by (i, j). All vertices in a grid have exactly four neighbors, except for the boundary vertices, which are the vertices (i, j) for which $i = 1$, $i = n$, $j = 1$, or $j = n$. Given $m \leq n^2$ starting vertices in the grid, the escape problem is to determine whether or not there are m vertex-disjoint paths from the starting vertices to any m different vertices on the boundary. Vertex-disjoint paths mean that each vertex can be used at most once in the escape. Show how to convert the escape problem into the maximum flow problem. It is enough to give the conversion procedure. It is not require to show the correctness of your procedure.

Answer of exercise 33

We can create a graph $G = (V, E)$, where V consists of all vertices on the grid and two new vertices s and t. For every adjacent pair of vertices, we create an edge on G. We connect s to every starting vertices and connect all boundary vertices to t. Set the node capacity for each node be one. If the maximum flow from s to t equals m, then m vertex-disjoint path exist.

Although traditional network has only edge capacity, we can transform node-capacitated network into edge-capacitated graph. For each vertex v, we create two vertices v' and v'', such that all the incoming edges of v connect to v' and all outgoing edges of v connect to v''. Moreover, we connect v' to v'' with edge capacity equals to v's node capacity. It is easy to see that a feasible flow in the transformed network can be transformed to a feasible flow in the original network, and *vice versa*.

3.7 Others

Exercise 34 Strong orientation (*NCKU IM 99*)

Suppose you are asked to assign direction for each edge in the graph to make it a digraph such that each vertex can connect to each other vertex by some directed graph (i.e. strongly connected). How do you know whether such strongly connected orientation exists for an undirected graph G of n vertices and m edges. Explain your method and discuss its complexity.

Answer of exercise 34

We can prove that a graph $G = (V, E)$ has a strong orientation if and only if G has no bridge. If G has a bridge, then it is impossible to assign the direction to become strongly connected. On the other hand, suppose that G has no bridge but the orientation does not exist. One maximum orientable subgraph H of G must exist. For a vertex $v \in V - H$, since G is 2-edge-connected, two paths from v to some vertex $u \in H$ must exist as well. Let the two paths be P and Q and let the first vertex of P and Q that enter H be p and q. Then, we can construct orientation of paths from v to p and from q to v. Thus, we construct a larger subgraph with orientation, which is impossible. Finally, testing for 2-edge-connectivity can be done in $O(n + m)$ time.

Remark. This theorem is called Robbins theorem.

Exercise 35 (*NCKU CSIE 100*)

A tournament $T = (V, E)$ is a simple digraph of $|V| = n$ vertices and $|E| = \frac{n(n-1)}{2}$ edges, suppose you already know outdeg[i], the outdegree for each vertex i. A tournament is transitive, whenever edge $(u, v) \in E$ and $(v, w) \in E$ implies $(u, w) \in E$. In other words, if there exists any 3 vertices i, j, k in T with edges (i, j), (j, k), and (k, i), then T is NOT transitive. Now you want to check whether T is transitive or not.

Answer of exercise 35

A naive solution is to generate all 3-tuple (i, j, k) and check. This method needs $O(n^3)$ time. We know that a tournament is transitive if and only if T is acyclic. Thus, we can use DFS to test whether the graph is acyclic in $O(n^2)$ time. Moreover, we know that a tournament is transitive if and only if all values in outdegree

are distinct. Since the range of outdegree is from 0 to $n-1$, we can use an array to test whether all elements are distinct in $O(n)$ time.

Exercise 36 (*NCU CSIE 95*)

Given an undirected graph $G = (V, E)$ with $n = |V|$ vertices, four vertices of G, say, u, v, x, and y, are said to form a 4-cycle if (u, v), (v, x), (x, y) and (y, u) are in E. Consider the problem of determining whether G contains a 4-cycle. A naive method by checking all possible 4-combinations of the vertex set will need $\Omega(n^4)$ time to complete the job. Design a more efficient algorithm (i.e., the time complexity of your algorithm should be $O(n^k)$ with $k < 4$) to solve the problem. Analysis the execution time of your algorithm.

Answer of exercise 36

Finding a 4-cycle is the same as finding two vertices with two common neighbors. For each vertex v, enumerate all pairs of v's pair. During the enumeration, if a pair is enumerated twice, then a 4-cycle must exist. Since there are at most $O(n^2)$ pairs of vertices, the time complexity is $O(n^2)$.

Remark. For any even number k, finding a cycle of length k can be done in $O(n^2)$ time [4]. Finding a path of length at least $k > 0$ can be done in $O^*(k^2)$ [3].

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Chapter 4

Problems on computational geometry

4.1 Sweep line

** Exercise 37 Power diagrams (*NTU CSIE 103*)

Suppose that you have n circles on a 2D plane. The radius and the center coordinate of each circle can be retrieved in $O(1)$ time. A closed region is defined as a non-empty set of connected 2-D points, and each point is covered by at least one circle.

1. Your task is to find the number of closed regions. Describe your algorithm and data structure in detail. What is the time complexity of your algorithm.

2. Now we start to add more circles one-by-one to the plane. After each addition, we want to keep track of the number of closed regions. Describe an algorithm and data structure to do so. What is the time complexity of your algorithm for each addition.

Answer of exercise 37

This problems can be solved by using power diagrams, a generalization of Voronoi diagrams. For the first problem, we build a power diagram and then construct the dual graph of the diagram, where each vertex represents a circle. Then, remove each edge of two non-intersecting circles. The number of components of the resulting graph is the answer. Since power diagram can be built in $O(n \lg n)$ time and the size of the graph is $O(n)$, the time complexity is $O(n \lg n)$. For the second problem, we need to build the power diagram incrementally and this problem can also be done in $O(n \lg n)$ [1].

Exercise 38 (*NCTU CSIE 93*)

The input is a set of n rectangles all of whose edges are parallel to the axes. Design an $O(n \lg n)$ algorithm to mark all the rectangles that are contained in other rectangles.

Answer of exercise 38

Use the sweep line algorithm. Let the position of left-top point and the right-bottom point of i-th rectangle be (L_i^x, L_i^y) and (R_i^x, R_i^y) respectively. Sort all L's and R's by increasing x-coordinate. If some L's or R's have the same x-coordinate, then sort them by decreasing y-coordinate. According to this ordering, insert or delete all vertical edges of rectangles into an interval tree I one by one.

When encountering a point (L_i^x, L_i^y), we can insert the left edge of rectangle i into I and test whether it is contained by some other left edge of rectangle j. If so, we can determine whether rectangle j contains rectangle i by their right edge. When encountering a point (R_i^x, R_i^y), we delete the left edge of rectangle i from I. The sorting only takes $O(n \lg n)$ time. Since each insertion and deletion takes $O(\lg n)$ time and the algorithm inserts and deletes $O(n)$ times, the time complexity is $O(n \lg n)$.

4.2 Convex hull

Exercise 39 Farthest pair (*NTU CSIE 97*)

Let the input set X consists of n points on the 2-dimensional plane with integral coordinates. The *farthest pair problem* is to identify two points in X whose Euclidean distance is maximum over all pairs X. You are also asked to prove or disprove that the farthest pair problem can be solved in $O(n \lg n)$ time. [NTU CSIE 97]

Answer of exercise 39

Problem 39 The first step is to find the convex hull, since the farthest pair must be on the convex hull. Then, apply the technique rotating calipers and find the farthest pair of points on the convex hull. Since finding the convex hull takes $O(n \lg n)$ time and rotating calipers takes linear time, the time complexity is $O(n \lg n)$.

4.3 Duality

Exercise 40

Show how to determine in $O(n^2 \lg n)$ time whether any three points in the set $S = \{(x_1, y_1), (x_2, y_2), \ldots, (x_n, y_n)\}$ are collinear. [NTU CSIE 92]

Answer of exercise 40

First, transform the points to dual lines. If there are three lines intersect at the same point on the dual plane, then the three points in the original plane are collinear. We can construct the arrangement of lines to find whether there are three lines intersect at the same point. The time complexity is $O(n^2)$.

Remark. Building the arrangement of lines can be done in $O(n^2)$ time in linear space [2].

4.4 Others

Exercise 41 Maximal points (*NCU CSIE 98*)

In a 2D plane, we say that a point (x_1, y_1) dominates (x_2, y_2) if $x_1 > x_2$ and $y_1 > y_2$. A point is called a *maximal point* if no other point dominates it. Given a set of n points, the maxima finding problem is to find all of the maximal points.

1. Write a divide and conquer algorithm to solve the maxima finding problem with the time complexity $O(n \lg n)$.
2. Show that your algorithm is indeed of the time complexity $O(n \lg n)$. [NCU CSIE 98]

Answer of exercise 41

The idea is based on divide and marriage before conquest method. First, sort the points by increasing x-coordinate. Then, we recursively find maximal points. If there are only two points, then finding the maximal points is easy.

When the number of points is more than two, divide the points into two parts, L and R, by the x-coordinate, such that $||L| - |R|| \leq 1$. We first find R's maximal points R_m. Let y_m be the maximum y-coorinate in R_m. Since each point in L has smaller x-coordinate than any points in R_m, we remove all points in L with y-coordinate smaller than y_m. Then, we recursively find the maximal points in the remaining points in L.

Sorting takes $O(n \lg n)$ time. Since merging two solutions takes linear time, the recurrence relation of the time complexity is $T(n) = 2T(\frac{n}{2}) + O(n)$. By using the master theorem, we prove that the time complexity is $O(n \lg n)$.

Remark. In each recursive call, we will find at least one maxima in R, and we either find one maxima in L or all points in L will be removed. The time complexity actually is $O(n \lg h)$, where h is the number of maximal points [3].

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Chapter 5

Algorithm design problems

5.1 Greedy

Exercise 42 (*NTU CSIE 97, NTNU CSIE 97*)

Consider the following scheduling problem. Suppose a man has several jobs waiting for his treatments. Each job takes one unit of time to finish and has a deadline and a profit. He can only do one job at any time. If a job starts before or at its deadline, its profit is obtained. The goal is to schedule the jobs so as to maximize the total profit. But not all jobs have to be scheduled. Please design an efficient algorithm to find a schedule that maximizes the total profit.

Answer of exercise 42

The idea is based on the greedy method. Sort the jobs by increasing deadline and process jobs in the sorted order. Let the schedule be empty initially. When considering the job j, if assigning job j to the last slot will not violate the constraint, then assign job j to the latest time slot; otherwise if there is a job k in schedule that has profit smaller than job j, then replace job k by job j. We prove the correctness of this algorithm by contradiction. Suppose that there exists an optimal solution that is different from the solution found by the greedy algorithm, then it has some jobs that we did not select. The most profitable job among these unselected jobs is assigned into an optimal schedule but unassigned in our schedule; however, it is impossible. The time complexity is $O(n \lg n)$.

5.2 Dynamic programming

Exercise 43 (*CYCU CSIE 90*)

A one way railway has n stops. Suppose that for all $i < j$, the price of the ticket fro the i-th stop to j-th stop is know, denoted cost(i, j). (There is no traffic in the reverse direction since the railway is one-way.) Apply the dynamic programming technique to design your algorithm that outputs the minimum travel cost from stop 1 to stop n, and all the intermediate stops that the travel takes. What is the time complexity of your algorithm.

Answer of exercise 43

Let $F(i, j)$ be the minimum cost of traveling from stop i to stop j. The recurrence relation of F is

$$F(i,j) = \begin{cases} 0 & \text{if } i = j \\ \min(\min_k F(i,k) + F(k,j), C_{ij}) & \text{if } i < j. \end{cases}$$

The time complexity is $O(n^3)$.

Exercise 44 (*NCNU CSIE 93*)

Suppose that we cut a stick of length L (a positive integer) with the probability P at each position such that its distance from the left end is a positive integer. Design an efficient dynamic programming algorithm for calculating the probability that a stick of length at least n remains.

Answer of exercise 44

Let $F(k, n)$ be the probability of cutting stick of length k such that at least length n remains. The recurrence relation of F is

$$F(k,n) = \begin{cases} 0 & \text{for all } 1 \le k \le n-1 \\ (1-P)^{n-1} + \sum_{i=1}^{n-1} P(1-P)^{i-1} F(k-i-1,n) + & \text{for all } k \ge n. \end{cases}$$

We can use this recurrence relation to calculate the probability.

Chapter 6

Complexity theory

6.1 NP-completeness

6.2 Approximation algorithms

Exercise 45 Christofides algorithm (*NCU CSIE 100*)

Let V be a set of n points in the plane. Let $G = (V, E)$ be the complete graph over V, and the weight of each edge $e \in E$ is the length of this edge. The Euclidean Traveling Salesman Problem (ETSP) of V is to find the cycle C^* such that C^* visits each node exactly once, and it has the minimum weight among all such cycles. Let T^* be a minimum spanning tree of G.

1. Show that $w(T^*) \leq w(C^*)$, where $w(X)$ is the weight of a subgraph X of G.
2. Given T^*, design an algorithm to compute a cycle C that is a 2-approximation of the optimal ETSP cycle C^*. Namely, $w(C^*) \leq w(C) \leq 2w(C^*)$. (Hint: first show that $w(T^8) \leq w(C^*) \leq 2w(T^*)$.)

[NCU CSIE 100]

Answer of exercise 45

This algorithm is called Christofides algorithm.

Part II

Further Information

Chapter 7

Binary search tree & related data structures

7.1 Binary search tree (BST)

Binary search tree is a common implementation of associative arrays that store elements from an ordered set while supporting the following operations:

1. Find: look up an element in the data structure.
2. Insert: insert an element.
3. Delete: delete an element. There are two ways to define the deletion: remove a given element or a given handle (i.e. tree node). When the type of deletion is not specified, deletion refers to the former in this chapter.

Although all of the above operations can be supported by a hash table efficiently, there are some operations that cannot be supported by hash table efficiently but can be supported by binary search tree efficiently.

1. Upper bound: find the smallest element in the data structure that is larger than a given key.
2. Lower bound: find the larger element in the data structure that is smaller than a given key.
3. Rank: for a given positive integer k, find the k-th smallest element in the data structure.

Since the tree height of an ordinary binary search tree is $O(n)$ in the worst case, many self-balancing binary search trees are designed.

Self-balancing binary search tree's nodes usually contain the following information:

1. Data: the data must come from an ordered set.
2. Left, Right: the pointers pointing to the left and right subtrees respectively.
3. Parent: the pointer pointing to the parent tree node, but this field is optional.
4. Auxiliary information: the additional information needs to be maintained in order to maintain balanced. Some self-balancing binary search trees (like splay tree and scapegoat tree) do not need auxiliary information.

When evaluating the performance for self-balancing binary search trees, we usually analyze from the following aspects:

1. Type of guarantee: the guarantee can be worst case, amortized, or expected.
2. Time complexity: the running time for each operation.

3. Space usage per element: the additional space required to maintain auxiliary information.

Since many self-balancing binary search trees provide the same time complexity with the same guarantee, we can further compare the performance of these trees from the following aspects:

1. Space complexity: the space usage to perform the operation. If the algorithm is recursive, then the space usage will be $\Omega(\lg n)$. Thus, in order to save space, iterative algorithm is preferable.

2. Number of passes: the maximum number of time a node can be 'traversed'. Traditionally, top-down algorithm means that each node is traversed only once from root to the leaf, while bottom-up algorithm means that each node is traversed twice from root to the leaf and from leaf to the root.

3. Parent pointer: using parent pointer makes implementation easier since a recursive algorithm, which requires $\Omega(\lg n)$ space, can be easily transformed into an iterative algorithm, which uses $O(1)$ space, although each node consumes more space. Most standard libraries use parent pointers, since without pointer, iterator cannot be implemented efficiently.

4. Cost of rebalance: the number of operations need to be done after inserting or deleting a node. In other words, when the insert or delete position is known, the time complexity of insert or delete operations. This is important for C++'s `std::set`, since the `insert` and `erase` methods of `std::set` can take hint, which is a guess of insert or delete position. Since the standard requires that when the hint is correct, the running time must be amortized $O(1)$, `std::set` can only be implemented by red-black tree or WAVL tree.

5. Duplication/non-existence detection: when inserting an element already in the tree or deleting an element not in the tree, whether the algorithm can report it. For certain trees, one more pass is necessary to detect or recover from such a situation.

7.1.1 Properties

The tree height of BST is $O(n)$ in the worst case. However, when the elements are inserted from a random permutation, the following properties can be shown:

1. The average height is $\alpha \ln n - \beta \ln \ln n + O(1)$, where $\alpha \approx 4.311$ and $\beta \approx 1.953$ [16].

2. The variance of the average height is $O(1)$ [7].

7.1.2 Rebalancing of a BST

Transforming an arbitrary BST into a perfectly balanced BST can be done in linear time using Day-Stout-Warren algorithm, which uses $2n - \lceil \lg n \rceil$ comparisons. The number of comparisons can be reduced to $2n - 2\lceil \lg n \rceil$ [11].

7.2 AVL tree

AVL tree is a self-balancing binary search tree that the difference of heights of left and right subtrees is bounded by one.

The balancing information stored in each node can be one of the following:

1. Height: each node maintains the height of the subtree rooted at this node. This approach is easier to implement, but each insert and delete inherently requires two passes since all heights of nodes along the search path need to be updated.

2. Balance factor: since the difference of the heights of subtrees is bounded by one, each node can use two bits to store the difference, which is called *balance factor*.

3. Height difference between node and it's parent: given children's height difference, the balance factor can be computed easily. For a node without two children, since the balance factor can be computed in $O(1)$ time, no information needs to be stored. Since height difference is either 1 or 2, the height difference can be represented by one bit [4]. One way to reduce the space usage is to swap left and right links when the bit is 1.

Both insert and delete operations can be performed in $O(\lg n)$ time. Moreover, the insertion requires at most 2 rotations, while the deletion might need $O(\lg n)$ rotations.

When balance factor is stored in the tree node, the insertion can be done in $O(1)$ space in the following manner. During insertion, a node is called the *safe node* if it is the lowest node along the search path from root to the inserted node with non-zero balance factor. Note that if rotation occurs after insertion, the rotation can only occurs at the safe node. Moreover, if one node's balance factor is changed after insertion, then this node must be between the safe node and the inserted node along the search path. Thus, we can first find the safe node using one traversal. Then, rotate and update balance factors accordingly from the safe node to the inserted node.

7.2.1 Properties

1. For an arbitrary AVL tree, a sequence of n deletions takes $\approx 1.618m = O(n)$ time [19]. From an empty AVL tree, a sequence of n insertions takes $\approx 2.26n = O(n)$ time [12]. Alternating insertions and deletions in an n-node AVL tree can cause $O(\lg n)$ rotations in each deletion [1].

2. It is possible to enforce the balance factor to be either 0 or 1, i.e. -1 is not allowed. Both insertion and deletion can be done in $O(\lg n)$ time, although the algorithm is impractical [20, 15].

3. If elements are inserted into an AVL in the sorted order, the resulting AVL tree is a perfectly balanced binary search tree [1].

4. The deletion can be done in $O(1)$ space using link reversal technique [5].

7.3 Red-black tree

Red-black tree is a self-balancing binary search tree that can represent a 2-3-4 tree as a binary tree.

1. Each node is either red or black.

2. Red node cannot have a red child.

3. External nodes are black.

4. All paths from root to the external nodes have equal number of black nodes.

Each node needs one bit to store the color of the node. One way to reduce the space usage is to swap left and right links when the bit is 1.

7.3.1 History

1. 1972: A binary representation of B-tree is invented [2]. The author designed bottom-up insert and delete methods that require $O(\lg n)$ rotations.

2. 1978: Red-black tree is invented [8]. The authors designed top-down insert and delete methods that require $O(\lg n)$ rotations.

3. 1982: Half-balanced binary search tree is invented [14]. The author designed bottom-up insert and delete methods, where insertion takes at most 2 rotations and deletion takes at most 3 rotations.

[1] https://11011110.github.io/blog/2015/07/25/when-avl-trees.html

4. 1983: The equivalence between red-black tree and half-balanced binary search tree is proved [18]. Bottom-up insert and delete methods for red-black tree is explained, where insertion takes at most 2 rotations and deletion takes at most 3 rotations. Moreover, the rebalance process (including rotations, recoloring, and backtracking) is shown to be amortized $O(1)$ time due to the equivalence between red-black tree and 2-3-4 tree [2].

5. 1985: Top-down insert and delete methods with $O(1)$ amortized rebalancing time are designed [17], although the number of rotations is not $O(1)$. This method also implies $O(1)$ space implementation of bottom-up insert and delete, which needs additional amortized $O(1)$ key comparisons.

7.3.2 Properties

1. Tree height is at most $2 \lg n$.

2. Both insert and delete operations can be done in $O(1)$ space using link reversal technique [6].

7.3.3 Left-leaning red-black tree (LLRB)

Left-leaning red-black tree is a variant of red-black tree that right link cannot point to a red node.

Left-leaning red-black tree can represent 2-3-4 tree or 2-3 tree. However, the insert and delete methods require $O(\lg n)$ rotations, which is different from red-black tree. In addition, the deletion requires rotations on the way up and on the way down, so the performance is worse than red-black tree.

7.3.4 WAVL tree

WAVL tree is a self-balancing binary search tree based on the following rank function:

1. Every external node has rank 0.

2. Every leaf node has rank 1.

3. The difference of rank between parent and children is either 1 or 2.

WAVL tree has all good properties that red-black tree has: amortized $O(1)$ time rebalancing, constant number of rotations in insert and delete operations. Moreover, the deletion of WAVL tree only requires 2 rotations, while red-black tree requires 3 rotations.

In fact, AVL tree can be described by using another rank function – tree height, and Red-black can also be described by using a different rank function – black height. Thus, rank-based search is a pretty generic framework. WAVL tree stands for weak AVL tree, whose constraints are weaker than AVL but stronger than red-black tree. Every red-black trees is a WAVL trees, and every WAVL is an AVL tree.

7.4 Weight-balanced BST (WB)

Weight-balanced tree uses the sizes of each subtree as the balancing criterion. Let the weight of a tree be the size plus one. A binary search is *weight-balanced* if for each node, the weight of the smaller subtree times Δ is greater than the weight of the larger subtree, where Δ is the balancing parameter.

When inserting or deleting an element, since the resulting tree might be imbalanced, a rotation must be performed in order to rebalance the tree.

Consider a subtree rooted at node x, where all subtrees of x are balanced except for x itself. Suppose that the left subtree of x is too small so that the balancing constraint is violated. In this case, a left rotation will be performed at x.

However, a single left rotation may not be sufficient. Let r be the right subtree of x. The issue is that the left subtree of r, rl, may be larger than the right subtree of r, rr. After a single left rotation is done at x and r is promoted to be the root. The tree rooted at r may still imbalanced.

[2] https://cs.stackexchange.com/questions/52660/reb-black-tree-amortized-cost-of-the-rebalancing

In order to remedy this issue, a single left rotation is performed when the weight of rl is less than Γ times the weight of rr, otherwise a double rotation is performed, where Γ is another balancing parameter. The details of top-down insert and delete methods are described in [10].

7.4.1 Properties

Since the weight of a tree is at least $\frac{\Delta}{\Delta-1}$ larger than any subtree, the height of the tree is $\lg_{\frac{\Delta}{\Delta-1}}(n+1) = \frac{-1}{\lg\frac{\Delta-1}{\Delta}}\lg(n+1)$. Furthermore, rebalancing takes amortized $O(1)$ time [3]. The only valid integer pair of (Δ, Γ) is $(3, 2)$ [9].

7.5 AA tree

AA tree is a self-balancing binary search tree that is another representation of 2-3 tree. Insert and delete operations take $O(\lg n)$ time. The code is the simplest among all binary search tree with $O(\lg n)$ worst case guarantee.

7.6 Splay tree

Splay tree is a self-balancing binary search tree. Insert and delete operations take amortized $O(\lg n)$ time. The idea is to rotate the accessed node to the root.

7.7 Scapegoat tree

Scapegoat tree is a self-balancing binary search tree. Insert and delete operations take amortized $O(\lg n)$ time. The idea is to rebuild part of the tree (rooted at *scapegoat* when the tree becomes imbalance with respect to certain criterion. In the original paper, the rebuilding phase takes $O(\lg n)$ space, but the rebuilding can be done by using Day-Stout-Warren algorithm to achieve $O(1)$ space.

7.8 Treap

Treap is a self-balancing binary search tree. Insert and delete operations take expected $O(\lg n)$ time. The idea is to assign each node a random real number between 0 and 1. In addition to the binary search tree constraints, the assigned random numbers of nodes also needs to satisfy heap property. Thus, a treap is always a random binary search tree if no collision of random number occurs. However, since real number is only represented by finite number of floating point numbers, there is no guarantee that collision won't occur. The original insert and delete methods are based on rotation.

7.8.1 Randomized binary search tree (RBST)

Randomized binary search tree is a self-balancing binary search tree. Insert and delete operations take expected $O(\lg n)$ time. The idea is to always maintain the tree a random binary search tree. Unlike treap, which requires random number from an infinity set, RBST only needs random integers from a finite set. However, the number of random numbers required by RBST is more than treap. The original insert and delete methods are based on split and join, which can also be used in treap.

7.9 Skip list

Skip list is a list like data structure. The idea is to build index structure probabilistically. Insert and delete operations take expected $O(\lg n)$ time. Unlike BST that will compare a given key with an element in the tree at most once, skip list might compare the given key with an element in the skip list multiple time that leads to more number of comparisons.

Name	Variant	Guarantee	Time	Space	# Pass	Rebalancing	Error detection
Red-black tree	Bottom-up	Worst case	$O(\lg n)$	$O(1)$	2	$O(\lg n)$[1]	Y
		Amortized				$O(1)$	
	Top-down	Worst case	$O(\lg n)$	$O(1)$	1	$O(\lg n)$	Y
		Amortized				$O(1)$[2]	
AVL	Bottom-up	Worst case	$O(\lg n)$	$O(1)$	2	$O(\lg n)$[3]	Y
WAVL	Bottom-up	Worst case	$O(\lg n)$	$O(1)$	2	$O(\lg n)$[4]	Y
		Amortized				$O(1)$	
LLRB	Bottom-up	Worst case	$O(\lg n)$	$O(\lg n)$	2	$O(\lg n)$	Y
WB	Top-down	Worst case	$O(\lg n)$	$O(1)$	1	$O(\lg n)$	N[5]
		Amortized				$O(\lg n)$[6]	
AA-tree	Bottom-up	Worst case	$O(\lg n)$	$O(\lg n)$	2	$O(\lg n)$	Y
Splay tree	Bottom-up	Worst case	$O(n)$	$O(\lg n)$	2	$O(n)$	Y
		Amortized				$O(\lg n)$	
	Top-down	Worst case	$O(n)$	$O(1)$	1	$O(n)$	Y
		Amortized				$O(\lg n)$	
Scapegoat tree	Bottom-up	Worst case	$O(n)$	$O(1)$[7]	1	$O(n)$	Y
		Amortized	$O(\lg n)$			$O(\lg n)$	
Treap	Rotation	Worst case	$O(n)$	$O(n)$	2	$O(n)$	Y
		Expected	$O(\lg n)$			$O(\lg n)$	
	Split-join	Worst case	$O(n)$	$O(1)$	1	$O(n)$	N
		Expected	$O(\lg n)$			$O(\lg n)$	
RBST	Split-join	Worst case	$O(n)$	$O(1)$	1	$O(n)$	N
		Expected	$O(\lg n)$			$O(\lg n)$	
Skip List		Worst case	$O(n)$	$O(1)$	1	$O(n)$	Y
		Expected	$O(\lg n)$				
DSL		Worst case	$O(\lg n)$	$O(1)$	1	$O(\lg n)$	Y

[1] Although the insertion rotates at most twice, the number of color flips is $O(\lg n)$.
[2] Tarjan's version [17]. The original red-black top-down insertion does not have amortized $O(1)$ guarantee.
[3] Although the insertion rotates at most twice, the number of changing balance factor is $O(\lg n)$.
[4] Although the insertion rotates at most twice, the number of changing rank is $O(\lg n)$.
[5] When a duplicated is inserted, one more pass is needed.
[6] Although the number of rotations is amortized $O(1)$, maintaining the size of subtrees requires
[7] Using Day-Stout-Warren algorithm to rebuild the tree.

Table 7.1: Summary of insert operation of various data structures.

7.9.1 Deterministic skip list (DSL)

Deterministic skip list is a derandomization version of skip list [13]. Insert and delete operations take $O(\lg n)$ time.

7.10 Summary

Tables 7.1 and 7.2 are summaries of the insertion/deletion operation of data structures assuming no parent pointers or link reversal technique are used. When parent pointers or link reversal technique is used, the space complexity of all operations can be reduced to $O(1)$.

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Name	Variant	Guarantee	Time	Space	# Pass	Rebalancing	Error detection
Red-black tree	Bottom-up	Worst case	$O(\lg n)$	$O(1)$	2	$O(\lg n)$[1]	Y
		Amortized				$O(1)$	
	Top-down	Worst case	$O(\lg n)$	$O(1)$	1	$O(\lg n)$	Y
		Amortized				$O(1)$[2]	
AVL	Bottom-up	Worst case	$O(\lg n)$	$O(1)$	2	$O(\lg n)$[3]	Y
WAVL	Bottom-up	Worst case	$O(\lg n)$	$O(1)$	2	$O(\lg n)$[4]	Y
		Amortized				$O(1)$	
LLRB	Bottom-up	Worst case	$O(\lg n)$	$O(\lg n)$	2	$O(\lg n)$	Y
WB	Top-down	Worst case	$O(\lg n)$	$O(1)$	1	$O(\lg n)$	N[5]
		Amortized				$O(\lg n)$[6]	
AA-tree	Bottom-up	Worst case	$O(\lg n)$	$O(\lg n)$	2	$O(\lg n)$	Y
Splay tree	Bottom-up	Worst case	$O(n)$	$O(\lg n)$	2	$O(n)$	Y
		Amortized				$O(\lg n)$	
	Top-down	Worst case	$O(n)$	$O(1)$	1	$O(n)$	Y
		Amortized				$O(\lg n)$	
Scapegoat tree	Bottom-up	Worst case	$O(n)$	$O(1)$[7]	1	$O(n)$	Y
		Amortized	$O(\lg n)$			$O(\lg n)$	
Treap	Rotation	Worst case	$O(n)$	$O(1)$	1	$O(n)$	Y
		Expected	$O(\lg n)$			$O(\lg n)$	
	Split-join	Worst case	$O(n)$	$O(1)$	1	$O(n)$	Y
		Expected	$O(\lg n)$			$O(\lg n)$	
RBST	Split-join	Worst case	$O(n)$	$O(1)$	1	$O(n)$	Y
		Expected	$O(\lg n)$			$O(\lg n)$	
Skip List		Worst case	$O(n)$	$O(1)$	1	$O(n)$	Y
		Expected	$O(\lg n)$				
DSL		Worst case	$O(\lg n)$	$O(1)$	1	$O(\lg n)$	Y

[1] Although the deletion rotates at most three times, the number of color flips is $O(\lg n)$.
[2] Tarjan's version [17]. The original red-black top-down insertion does not have amortized $O(1)$ guarantee.
[3] The number of rotations is $O(\lg n)$.
[4] Although the insertion rotates at most twice, the number of changing rank is $O(\lg n)$.
[5] When a duplicated is inserted, one more pass is needed.
[6] Although the number of rotations is amortized $O(1)$, maintaining the size of subtrees requires $O(\lg n)$.
[7] Using Day-Stout-Warren algorithm to rebuild the tree.

Table 7.2: Summary of delete operation of various data structures.

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