

Computer algorithms test review

Format

True and false
Pseudo code

Topic

- _____ Fibonacci Sequence
 - _____ Recursive method
 - _____ Case
 - _____ Pseudo-code
 - _____ Polynomial method
 - _____ Case
 - _____ Pseudocode
 - _____ Compare the two methods
- _____ Basic knowledge of graphs
- _____ Basic knowledge of modular
- _____ Divide and conquer
 - _____ General idea
 - _____ Steps
- _____ Binary search
 - _____ Pseudocode
 - _____ Merge sort
 - _____ Pseudocode
- _____ Greedy algorithm
 - _____ General idea
 - _____ Pseudocode
 - _____ Case
- _____ Kruskal's algorithm for MST

BOOK

- Divide and conquer
 - Chapter 2
- Greedy algorithm
 - Chapter 5
- Algorithm with numbers
 - Chapter 1

Resources

Merge sort

<http://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-006-introduction-to-algorithms-fall-2011/lecture-videos/lecture-3-insertion-sort-merge-sort/>

Binary search

<http://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-006-introduction-to-algorithms-fall-2011/lecture-videos/lecture-5-binary-search-trees-bst-sort/>

List of algorithms that you already know (not really)

General

- Recursive algorithm

- Polynomial algorithm

Searching algorithms

- Linear search

- Binary search

- Search with hashing

Sorting algorithms

- Quick sort

- Selection sort

- Bubble sort

- Insertion sort

- Heap sort

- Merge sort

Graph related algorithms

- Bread-first Search/traversal

- Depth first search/traversal

- Single-source shortest path

Binary tree related algorithm

- Bread-first traversal

- Depth-first traversal

 - In-order

 - Pre-order

 - Post-order

Algorithm

Sequence of well-defined computational steps that transform the input into the output

Steps are precise, unambiguous, mechanical, efficient and correct

A tool for solving well-specified computational problems

3 questions to ask about any algorithm

- Is it correct?

- How much time does it take, as a function of n ?

- And can we do better?

Fibonacci numbers

Generated by a simple rule

$$F_n = F_{n-1} + F_{n-2}$$

Grows almost as fast as the power of 2

$$F_n = 2^{0.694n}$$

Recursive algorithm

Function fib1(n)

If $n == 0$: return 0;

If $n == 1$: return 1;

else

return fib1(n-1) + fib1(n-2)

Efficiency

$T(n)$ is the number of computer steps to compute fib1(n)

$$T(n) \leq 2 \text{ for } n \leq 1$$

$$T(n) = T(n-1) + T(n-2) + C \text{ for } n > 1$$

$$T(n) \geq F_n \text{ (exponential though } F_n = 2^{0.694n})$$

Recursive algorithm is too slow except for very small values of n

Why?

Many computations are repeated

None are stored

Polynomial algorithm

Function fib2(n)

If $n == 0$ return 0

Create array $f[0 \dots n]$

$F[0] = 0, f[1] = 1$

for $i = 2 \dots n$ {

$F[i] = f[i-1] + f[i-2]$ }

Return $f[n]$

Array is used to store intermediate results

Inner loop consists of a single computer step and is executed $n-1$ time

Thus the algorithm is linear in n

Evaluating running time

Processor's instruction set has a variety of basic primitives

Branching

Storing to memory

Comparing numbers

Simple arithmetic

More-

Estimating running time of algorithm

Considered as one category of basic computer steps

Architecture-specific and hardware-specific details are ignored

Big-o Notation

Running time (number of basic computer steps) of an algorithm is expressed as a function of the size of input n

$$\text{EXP: } T(n) = 5n^3 + 4n + 3$$

Lower-order terms and details of coefficient in the leading term are ignored

(because as n is generally rather large they become insignificant quickly)

$$\text{EXP: } T(n) = O(n^3)$$

$f(n)$ and $g(n)$ are functions from positive integers to positive real numbers

$f = O(g)$ (f grows no faster than g)

If there is a constant $c > 0$ such that $f(n) \leq c * g(n)$

$$\text{EXP: } 10n = O(n)$$

When comparing two algorithms, we care about the rate of growth of n

$$f_1(n) = n^2, f_2(n) = 2n + 20$$

$$f_2(n) / f_1(n) =$$

$$(2n + 20) / n^2$$

$$\leq 22$$

For all n $f_2 = O(f_1)$

Common rules

Multiplicative constants can be omitted

$$\text{EXP: } 14n^2 \rightarrow n^2$$

n^a dominates n^b if $a > b$

$$\text{EXP: } n^2 \text{ dominates } n$$

Any exponential dominates any polynomial

$$\text{EXP: } 3^n \text{ dominates } n^5; 3^n \text{ dominates } 2^n$$

Any polynomial dominates any logarithm

$$\text{EXP: } n \text{ dominates } (\log n)^3; n^2 \text{ dominates } n(\log n)$$

What is a logarithm

Answers question

How many of one number do we multiply to get another number

$$\log_{(\text{Base})}(\text{integer}) = (\text{number of times Base must be multiplies to get to integer})$$

$$\log_{(2)}(8) = 3$$

Insertion sort

Insertion-sort(A)

for j <- 2 to length[a]

do key <- a[j]

Insert A[j] into the sorted sequence a[1..j-1]

i <- j - 1

while i > 0 and A[i] > key

do A[i+1] <- A[i]

i <- i-1

A[i+1] <- key

Input: A sequence of n numbers (a_1, a_2, \dots, a_n)

Output: a permutation (reordering) (a_1', a_2', \dots, a_n') of the input sequence such that

$a_1' \leq a_2' \leq \dots \leq a_n'$

Efficiency

$T(n) = c_1(n-1)[c_2 + c_4 + (t_j - 1)(c_5 + c_6 + c_7) + c_5 + c_8] -- t_j \text{ epsilon } [1, n-1]$

c1 cost n = 1

c2 cost n-1

c4 n - 1

c5 $\sum_{j=2}^n t_j$

c6 $\sum_{j=2}^n 2^{(t_j - 1)}$

c7 $\sum_{j=2}^n 2^{(t_j - 1)}$

c8 n - 1

Best case : array already sorted, $t_j = 1$:

$T(n) = c_1(n-1) [c_2 + c_4 + c_5 + c_8]$

= an + b

Worst case: array is in reverse sorted order, $t_j = n-1$

$T(n) = an^2 + bn + c$

Comparison of sorting algorithms

Sort	Best case	average case	worst case
Selection sort	$O(N^2)$	$O(N^2)$	$O(N^2)$
Bubble sort	$O(N^2)$	$O(N^2)$	$O(N^2)$
Shortbubble	$O(N)(*)$	$O(N^2)$	$O(N^2)$
Insertion sort	$O(N)(*)$	$O(N^2)$	$O(N^2)$
Merge sort	$O(N \log_2 N)$	$O(N \log_2 N)$	$O(N \log_2 N)$
Quick sort	$O(N \log_2 N)$	$O(N \log_2 N)$	$O(N^2)$ (depends on split)
Heap sort	$O(N \log_2 N)$	$O(N \log_2 N)$	$O(N \log_2 N)$

Addition

Basic property of decimal numbers (binary number)

Sum of any three single-digit numbers is at most two digits long

Add two numbers in any base

Align their right hand ends

Compute the sum of digit by digit in a single right to left pass

Maintain overflow as a carry

Since each individual sum is a two-digit number

The carry is always a single digit

At any given step three single digit numbers are added

Efficiency of adding

Two n-bit long binary numbers x and y

How long to add them expressed as a function of the size of the input n

Sum of x and y is n+1 bits at most

Each individual bit of this sum gets computed in a fixed amount of time

$T(n) = c_0 + c_1n$ where c_0 and c_1 are some constants

Running time is linear $O(n)$

Multiplication

Multiplying two binary numbers x and y

Creates an array of intermediate sums,

Each representing the product of x by a single digit of y

Values left shifted then added up

Efficiency of multiplication

X and y are both n bits

There are n intermediate rows

With lengths of up to $2n$ bits (cause of shift)

Total time take to add up these rows two numbers at a time

n-1 times of $O(n)$

Which is $O(n^2)$ and thus quadratic in the size of the input n

Division also $O(n^2)$

Modular Arithmetic

A system for dealing with restricted ranges of integers

X modulo N

defined to be the remainder when x is divided by N;

If $x = qN + r : 0 \leq r < N$

Then x modulo N is equal to r

It limits numbers to a pre-defined range

$\{0, 1, \dots, N-1\}$ wraps around whenever the number leaves this range

It deals with all integers but divides them into N equivalence classes

Of the form : $\{i + kN : k \in \mathbb{Z}\}$ for some i between 0 and N-1

Equivalence between numbers

X and y are congruent modulo N if they differ by a multiple of N

X congruent y (mod N) if N divides (x-y);

Example

253 minutes is 4 hours and 13 minutes

253 congruent 13(mod 60)

Can also be negative

59 congruent 1 (mod 60): when it is 59 minutes past the hour it is also one minute short the hour

Divide and conquer

Top-down technique for designing algorithms

Divide

Split the problem into a number of sub problems that are similar to the original problem but smaller in size

Conquer

Solve the subproblem recursively (successively and independently)

Combine

Merge the solution to the subproblems into the solution for the original problem

Binary search

Uses Divide and conquer

Searching problem

MUST BE SORTED FIRST

Given $A[1 \dots n]$ an array of non decreasing sorted order

$A[i] \leq A[j]$ whenever $1 \leq i \leq j \leq n$

Let q be the query point

The problem consist of finding q in the array A

Divide

Split A into two segments from the middle

$A[0]$ to $A[A.length/2]$ -> Array B

$A[A.length/2]$ to $A[A.length]$ -> Array C

Conquer (recursively)

Compare q with the middle element ($A[A.length/2]$)

If q is less than $A[A.length/2]$

Continue to Array B

If q is more than $A[A.length/2]$

Continue to Array C

Repeat the Divide step in the appropriate Array and then repeat the Compare step in the appropriate array until the we can safely conclude that q is or is not in the Array

Combine

The conclusion is that q is found / not found

Binary search can be accomplished in logarithmic time

$T(n) = O(\log_2 n)$

Merge Sort

Closely follows the divide-and-conquer paradigm

Sorting problem

Given $A = \{a_1, a_2, a_3, \dots, a_n\}$ a sequence of n numbers

Reorder elements of A such that

$A' = \{a_1', a_2', a_3', \dots, a_n'\} : a_1' \leq a_2' \leq a_3' \leq \dots \leq a_n'$

Divide

Split A into two subsequences of $n/2$ elements each

$A[0]$ to $A[A.length/2]$ -> Array B

$A[A.length/2]$ to $A[A.length]$ -> Array C

Conquer

Sort the two subsequences recursively using the merge sort

Combine

Merge the two sorted subsequences to produce A'

Merge sort implementation

```
Merge-sort (Array, Starting_point, Ending_point)
  If Starting_point < Ending_point
    Mid = ((Starting_point + Ending_point)/2)
    Merge-sort ( Array, Starting_point, Mid)
    Merge-sort ( Array, Mid + 1, Ending_point)
    Merge ( Array, Starting_point, q, Ending_point)
```

Recurrence

An equation describes the overall running time on a problem of size n in terms of the running time on smaller inputs

If the problem of size n is divided into c subproblems

The size of each subproblem is n/b and it takes $f(n) = O(nd)$ time to divide and combine the solutions, then recurrence is

$$T(n) = cT(n/b) + O(n^d), \text{ while } n \text{ is large} \\ (c > 0, b > 1 \text{ and } d \geq 0)$$

Master theorem

For $T(n) = cT(n/b) + f(n)$, when $f(n) = O(n^d)$

If $d > \log_b c$, then $T(n) = O(n^d)$

If $d = \log_b c$, then $T(n) = O(n^d \log_2 n)$

If $d < \log_b c$, then $T(n) = O(n^{\log_b c})$

Analysis of merge sort

We assume the original problem size n is a power of 2

Divide

Each divide step yields 2 subsequences of size $n/2$

A sequence of n will be divided $\log_2 n$ times to reach 1 element sequence

Combine

Merge n elements into a sorted array

$$T(n) = cT(n/b) + f(n)$$

Divide

It takes constant time to divide a sequence -- $O(1)$

Conquer

We recursively solve two subproblems of size $n/2$

Contributes $2T(n/2)$ to the running time ($c = 2$ and $b = 2$)

Combine

Merge n element subarray takes time $O(n)$ ($d=1$)

Master theorem case No. 2

$$(d = \log_b c) : T(n) = O(n^d \log_2 n) = O(n \log_2 n)$$

Median, min. & Max

The i th order statistic of a set of n elements is the i th smallest element

The minimum of a set is the first order statistic

$$i = 1$$

The maximum is the n th order statistic

$$i = n$$

A median is the halfway point of the set

$$i = (n+1)/2 \text{ when } n \text{ is odd}$$

$$i = n/2 \text{ and } i = n/2 + 1 \text{ when } n \text{ is even}$$

Matrix multiplication

Product of two $n \times n$ matrices X and Y is a third $n \times n$ matrix $Z = XY$: with (i, j) th entry

$Z_{ij} = \text{summation}_{k=1}^n X_{ik} Y_{kj}$

Square matrix-multiply

$O(n^3)$

Input A and B $n \times n$ matrices

C is a new $n \times n$ matrix

For $i = 1$ to n

For $j = 1$ to n

$C_{ij} = 0$

For $k = 1$ to n

$C_{ij} = C_{ij} + A_{ik} * B_{kj}$

Return C

Partition matrix

If n is the exact power of 2, then each of A , B and C matrices could be divided into four $n/2 \times n/2$ matrices

$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \quad B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix} \quad C = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$$

We could write $C = A * B$ as

$$\begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix} = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} * \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}$$

Where

$$C_{11} = A_{11} * B_{11} + A_{12} * B_{21}$$

$$C_{12} = A_{11} * B_{12} + A_{12} * B_{22}$$

$$C_{21} = A_{21} * B_{11} + A_{22} * B_{21}$$

$$C_{22} = A_{21} * B_{12} + A_{22} * B_{22}$$

Greedy algorithms

They build up a solution piece by piece

Always choosing the next piece that offers the most obvious immediate benefit

Local optimization

Making a move seems best at the moment

Not worrying too much about future consequences

Graph

Data structure that consists of a set of nodes and a set of edges that relate the nodes to each other

Vertex

Node in a graph

Edge (arc)

Connection between two nodes in a graph, represented by a pair of vertices

Undirected graph

All edges are bidirectional

Directed graph

All edges point in a direction

Formally a graph G is defined as $G = (V, E)$

$V(G)$ is a finite nonempty set of vertices

$E(G)$ is a set of edges (written as pairs of vertices)

Adjacent vertices

Two vertices in a graph that are connected by an edge

Path

A sequence of vertices that connects two nodes in a graph

Cycle

In a directed graph it is a circular path where $V_0 \rightarrow V_1 \rightarrow \dots \rightarrow V_k \rightarrow V_0$

Acyclic

Graph with no cycles

(DAG) directed acyclic Graph (DAG)

A directed graph with no cycles

Complete graph

A graph in which every vertex is directly connected to every other vertex;

For a graph with N nodes

$N*(N-1)$ edges in a complete directed graph

$N*(N-1)/2$ edges in a complete undirected graph

Weighted graph

A graph in which each edge carries a value

Matrix representation of graph

Adjacency Matrix

For a graph with N nodes, an N by N table that shows the existence (and weights) of all edges in the graph

Mapping array

An array that maps vertex names into array indexes

Marking

Associate a boolean variable with each vertex: true if visited, false if not yet visited

Linked-list representation of Graph

Adjacency list

A linked list that identifies all the vertices to which a particular vertex is connected

Each vertex has its own adjacency list

Graph traversal

Visit all the vertices, beginning with a specified start vertex

No vertex is visited more than once and vertices are only visited if they can be reached

If there is a path from the start vertex

Breadth first search algorithm

Visit all the nodes on one level before going to the next level

Depth first search algorithm

Visit all the nodes in a branch to its deepest point before moving up

Breadth first traversal

“neighbors-first”

Queue data structure is needed

Holds a list of vertices which have not been visited yet but which should be visited soon

While visit a vertex involves adding its neighbors to the queue.

Neighbors are not added to the queue if they are already in the queue, or have already been visited

Queue FIFO

Depth first traversal

A stack data structure is needed

Holds a list of vertices which have not been visited yet but should be soon

While visit a vertex involves adding its neighbors to the stack

Neighbors are not added to the stack if they are already in the stack or have already been visited

Stack LIFO

(MST) Minimum Spanning Tree (MST)

Given an undirected graph $G = (V, E)$

Weight of edge e is W_e , to find an undirected acyclic graph (tree) $T = (V, E')$

With E' member of E that minimizes

Weight (T) = summation of e is a member of E' for W_e

Application

Build cheapest network for a group of computers

Go through all vertices in the undirected graph while discarding all edges that would create a cycle

Making the lowest weighted edges graphed first

Kruskal's algorithm

Is a greedy algorithm and MST

Min spanning tree is acyclic

Property 1

Removing a cycle edge cannot disconnect a graph

The algorithm starts with the empty graph and then selects edges from E according to the following rule

Repeatedly add the next lightest edge that doesn't produce a cycle

Property 2

A tree on n nodes has $n-1$ edges

Property 3

Any connected undirected graph $G = (V, E)$ with $|E| = |V| - 1$ is a tree

Property 4

An undirected graph is a tree if and only if there is a unique path between any pair of nodes

Cut property

A cut is any partition of the vertices into two groups, S and $V-S$

Supposed edges X are part of a minimum spanning tree of $G = (V, E)$

Pick any subset of nodes S for which X does not cross between S and $V-S$ and let e be the lightest edge across this partition

Then $X \cup \{e\}$ is part of some MST

It is always safe to add the lightest edge across any cut

Provided X has no edges across the cut

Kruskals algorithm

Implementation

Procedure kruskal (G, w)

Input : a connected undirected graph $G = (V, E)$ with edge weights w_e

Output: a minimum spanning tree defined by the edges X

For all u member of V

 Makeset(u) // makeset(u) : create a singleton set containing just u

$X = \{\}$

Sort the edges E by weight

For all edges $\{u, v\}$ member of E , in increasing order of weight

 If find(u) does not equal find(v):

 Add edge $\{u, v\}$ to X // find (v) to which set does v belong

 Union(u, v) // union(u, v) merge the sets containing u and v

Implementation of kruskal's algorithm

The algorithm's state is modeled as a collection of disjoint sets, each of which contains the nodes of a particular component (a cut)

Initially each node is in a component by itself

 Makeset(x) : create singleton set containing just x

Pairs of nodes are tested repeatedly to see if they belong to the same set

 Find(x) : to which set does x belong to

When an edge is added two components are merged

 Union(x, y) : merge the sets containing x and y

The algorithm uses $|V|$ makeset, $2|E|$ find, and $|V|-1$ union operations

Time complexity

Kruskal's algorithm

 Sorting the edges: $O(|E| \log_2 |V|)$ (remember $\log_2 |E| = \log_2 |V|$)

Data structure for disjoint sets

A set could be stored as a directed tree

 Nodes are elements of set, not in order

 Root is a representative, or name, for the set

Each node has a parent pointer TT that eventually leads up to the root of the tree

 The parent pointer of root points to itself

Each node also has a rank - the height of the subtree hanging from that node

 Create a singleton set containing just x

 Procedure makeset(x)

 Pie(x) = x

 Rank(x) = 0

 To which set does x belong? It returns the root of its set

 While x does not equal pie (x) : $x = \text{pie}(x)$

 Return x

Minimum spanning tree

Property 1

Removing a cycle edge cannot disconnect a graph