Algorithm Design Manual Notes

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Contents

| 1 | Intro | duction To Algorithm Design | 3 |
|---|-------|-----------------------------------|---|
| | 1.1 | Robot Tour Optimization | 4 |
| | 1.2 | Selecting the Right Jobs | 4 |
| | 1.3 | Reasoning about Correctness | 4 |
| | | 1.3.1 Expressing Algorithms | 4 |
| | | 1.3.2 Problems and Properties | 4 |
| | | 1.3.3 Demonstrating Incorrectness | 4 |
| | | 1.3.4 Induction and Recursion | 5 |
| | | 1.3.5 Summations | 5 |
| | 1.4 | Modeling The Problem | 5 |
| | | 1.4.1 Combinatorial Objects | 5 |
| | | 1.4.2 Recursive Objects | 5 |
| | 1.5 | About the War Stories | 6 |
| | 1.6 | War Story: Psychic Modeling | 6 |
| 2 | Algo | rithm Analysis | 6 |
| | 2.1 | The RAM Model of Computation | 6 |

| | 2.1.1 | Best, Worst, and Average-Case Complexity | 6 | | | | |
|------|----------------------|--|----|--|--|--|--|
| 2.2 | The Bi | g Oh Notation | 6 | | | | |
| 2.3 | Growth | Rates and Dominance Relations | 7 | | | | |
| | 2.3.1 | Dominance Relations | 7 | | | | |
| 2.4 | Workin | ng with the Big Oh | 8 | | | | |
| | 2.4.1 | Adding Functions | 8 | | | | |
| | 2.4.2 | Multiplying Functions | 8 | | | | |
| 2.5 | Reason | ing About Efficiency | 9 | | | | |
| | 2.5.1 | Selection Sort | 9 | | | | |
| | 2.5.2 | Insertion Sort | 9 | | | | |
| | 2.5.3 | String Pattern Matching | 9 | | | | |
| | 2.5.4 | Matrix Multiplication | 10 | | | | |
| 2.6 | Logarit | hms and Their Applications | 10 | | | | |
| | 2.6.1 | Logarithms and Binary Search | 10 | | | | |
| | 2.6.2 | Logarithms Trees | 11 | | | | |
| | 2.6.3 | Logarithms and Bits | 11 | | | | |
| | 2.6.4 | Logarithms and Multiplication | 11 | | | | |
| | 2.6.5 | Fast Exponentiation | 11 | | | | |
| | 2.6.6 | Logarithms and Summations | 11 | | | | |
| | 2.6.7 | Logarithms and Criminal Justice | 11 | | | | |
| 2.7 | Propert | ties of Logarithms | 12 | | | | |
| 2.8 | War Sto | ory: Mystery of the Pyramids | 12 | | | | |
| 2.9 | Advano | eed Aanalysis (*) | 12 | | | | |
| 2.10 | 0 Esoteric Functions | | | | | | |
| 2 11 | Limite | and Dominance Relations | 12 | | | | |

1 Introduction To Algorithm Design

the algorithmic *problem* known as *sorting* is defined as follows:

Problem: Sorting

Input: A sequence of n keys $a_1, ..., a_n$.

Output: The permutation (reordering) of the input sequence such that $a_1' \leq a_2' \leq ... \leq a_{n-1}' \leq a_n'$

```
I N S E R T I O N S O R T I N S E R T I O N S O R T I N S E R T I O N S O R T E I N S R T I O N S O R T E I N R S T I O N S O R T E I I N R S T O N S O R T E I I N N O R S T N S O R T E I I N N O R S T T E I I N N O R S T T T E I I N N O R S S T R T E I I N N O O R R S S T R T E I I N N O O R R S S T T T E I I N N O O R R S S T T T E I I N N O O R R S S T T T E I I N N O O R R S S T T T E I I N N O O R R S S T T T E I I N N O O R R S S T T T E I I N N O O R R S S T T T E I I N N O O R R S S T T T E I I N N O O R R S S T T T E I I N N O O R R S S T T T E I I N N O O R R S S T T
```

Figure 1: Animation of insertion sort in action (time flows down)

```
insertion_sort(item s[], int n)
{
    int i,j; /* counters */

    for (i=1; i<n; i++) {
        j=i;
        while ((j>0) && (s[j] < s[j-1])) {
            swap(&s[j], &s[j-1]);
            j = j-1;
        }
    }
}</pre>
```

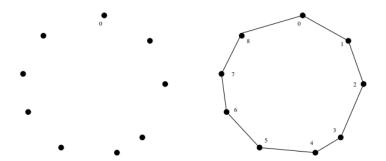


Figure 2: A good instance for the nearest neighbor heuristic

1.1 Robot Tour Optimization

Problem: Robot Tour Optimization *Input:* A set S of N points in the plane.

Output: What is the shortest cycle tour that visits each point in the set S?

1.2 Selecting the Right Jobs

Problem: Movie Scheduling Problem *Input:* A set I of n intervals on the line.

Output: What is the largest subset of mutually non-overlapping intervals which can be selected from I?

Take-Home Lesson: Reasonable-looking algorithms can easil beincorrect. Algorithm correctnes is a property that must be carefully demonstrated.

1.3 Reasoning about Correctness

1.3.1 Expressing Algorithms

Take-Home Lesson: The heart of any algorithm is an *idea*. If your idea is not clearly revealed when you express an algorithm, then you are using too low-level a notation to describe it.

1.3.2 Problems and Properties

Take-Home Lesson: An important and honorable tecnique in algorithm design is to narrow the set of llowable instances until there *is* a correct and efficient algorithm. For example, we can restrit a graph problem from general graphs down to trees, or a geometric problem from two dimensions down to one.

1.3.3 Demonstrating Incorrectness

- Verifiability
- Simplicity
- Think small
- Think exhaustively
- Hunt for the weakness
- · Seek extremes

Take-Home Lesson: Searching for conterexamples is the best way to disprove the correctness of a heuristic.

1.3.4 Induction and Recursion

Take-Home Lesson: Mathematical induction is usually the right wy to verify the correctness of a recursive or incremental insertion algorithm.

1.3.5 Summations

- Arithmetic progressions
- Geometric series

1.4 Modeling The Problem

1.4.1 Combinatorial Objects

- Permutations
- Subsets
- Trees
- Graphs
- Points
- Polygons
- Strings

Take-Home Lesson: Modeling your application in terms of well-defined structures and algorithms is the most important single step towards a solution.

1.4.2 Recursive Objects

- Permutations
- Subsets
- Trees
- Graphs
- Points
- Polygons
- Strings

1.5 About the War Stories

1.6 War Story: Psychic Modeling

2 Algorithm Analysis

Our two most important tools are

- 1. The RAM model of computation
- 2. The asymptotic analysis of worst-case complexity

2.1 The RAM Model of Computation

Take-Home Lesson: Algorithms can be understood and studied in a language and machine-independant manner..

2.1.1 Best, Worst, and Average-Case Complexity

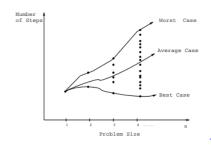


Figure 3: Best, Worst and average case complexity

2.2 The Big Oh Notation

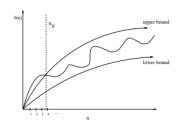


Figure 4: Upper and lower bounds valid for n>n0 smooth out the behavior of complex functions

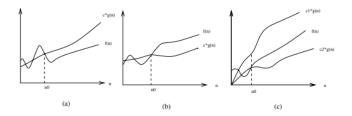


Figure 5: Illustrating the big (a) O, (b) Ω , and (c) Θ notations

Stop and Think: Hip to the Squares

Problem: Is
$$(x + y)^2 = O(x^2 + y^2)$$

Stop and Think: Back to the Definition

Problem: Is
$$2^{n+1} = \Theta(2^n)$$
?

2.3 Growth Rates and Dominance Relations

| n f(n) | $\lg n$ | n | $n \lg n$ | n^2 | 2 ⁿ | n! |
|---------------|-----------------|----------------------|---------------|-------------|--------------------------------|----------------------------------|
| 10 | $0.003~\mu s$ | $0.01~\mu s$ | $0.033~\mu s$ | $0.1~\mu s$ | 1 μs | 3.63 ms |
| 20 | $0.004~\mu s$ | $0.02~\mu s$ | $0.086~\mu s$ | $0.4~\mu s$ | 1 ms | 77.1 years |
| 30 | $0.005~\mu s$ | $0.03~\mu s$ | $0.147~\mu s$ | $0.9~\mu s$ | 1 sec | $8.4 \times 10^{15} \text{ yrs}$ |
| 40 | $0.005~\mu s$ | $0.04~\mu s$ | $0.213~\mu s$ | 1.6 µs | 18.3 min | · |
| 50 | $0.006~\mu s$ | $0.05~\mu s$ | $0.282~\mu s$ | $2.5~\mu s$ | 13 days | |
| 100 | $0.007~\mu s$ | 0.1 μs | 0.644 μs | 10 μs | $4 \times 10^{13} \text{ yrs}$ | |
| 1,000 | $0.010 \ \mu s$ | $1.00 \ \mu s$ | $9.966~\mu s$ | 1 ms | | |
| 10,000 | $0.013~\mu s$ | $10 \mu s$ | 130 μs | 100 ms | | |
| 100,000 | $0.017~\mu s$ | 0.10 ms | 1.67 ms | 10 sec | | |
| 1,000,000 | $0.020~\mu s$ | 1 ms | 19.93 ms | 16.7 min | | |
| 10,000,000 | $0.023~\mu s$ | $0.01 \mathrm{sec}$ | 0.23 sec | 1.16 days | | |
| 100,000,000 | $0.027~\mu s$ | $0.10 \mathrm{sec}$ | 2.66 sec | 115.7 days | | |
| 1,000,000,000 | $0.030~\mu s$ | 1 sec | 29.90 sec | 31.7 years | | |

Figure 6: Growth rates of common functions measured in nanoseconds

2.3.1 Dominance Relations

Take-Home Lesson: Although esoteric functions arise in advanced algorithm analysis, a small variety of time complexities suffice and account for most algorithms that are widely used in practice.

2.4 Working with the Big Oh

2.4.1 Adding Functions

$$O(f(n)) + O(g(n)) \longrightarrow O(max(f(n), g(n)))$$

$$\Omega(f(n)) + \Omega(g(n)) \longrightarrow \Omega(max(f(n), g(n)))$$

$$\Theta(f(n)) + \Theta(g(n)) \longrightarrow \Theta(max(f(n), g(n)))$$

2.4.2 Multiplying Functions

$$O(c * f(n))) \longrightarrow O(f(n))$$

$$\Omega(c * f(n))) \longrightarrow \Omega(f(n))$$

$$\Theta(c * f(n))) \longrightarrow \Theta(f(n))$$

$$O(f(n)) * O(g(n)) \longrightarrow O(f(n) * g(n))$$

$$\Omega(f(n)) * \Omega(g(n)) \longrightarrow \Omega(f(n) * g(n))$$

$$\Theta(f(n)) * \Theta(g(n)) \longrightarrow \Theta(f(n) * g(n))$$

Stop and Think: Hip to the SquaresTransitive Experience

Show that Big Oh relationships are transitive. That is, if f(n) = O(g(n)) and g(n) = O(h(n)), then f(n) = O(h(n))

2.5 Reasoning About Efficiency

2.5.1 Selection Sort

Figure 7: Animation of selection sort in action.

2.5.2 Insertion Sort

```
for (i=1; i<n; i++) {
    j=i;
    while((j>0) && (s[j] < s[j-1])) {
        swap(&s[j], &s[j-1]);
        j = j-1;
    }
}</pre>
```

2.5.3 String Pattern Matching

Problem: Substring Pattern Matching *Input:* A text string t and a pattern string p

Output: Does t contain the pattern p as a substring, and if so where?

```
a b a b b a a a b a b b a
```

Figure 8: Searching for the substring abba in the text aababba

2.5.4 Matrix Multiplication

Problem: Matrix Multiplication

Input: Two matrices, A (of dimesion $x \times y$) and B (dimension $y \times z$).

Output: An $x \times z$ matrix C where C[i][j] is the dot product of the ith row of A and the jth column of B.

2.6 Logarithms and Their Applications

$$b^x = y \leftrightarrow x = log_b y$$
$$b^{log_b y} = y$$

2.6.1 Logarithms and Binary Search



Figure 9: A height h tree with d children per node as d^h leaves. Here h=2 and d=3

2.6.2 Logarithms Trees

2.6.3 Logarithms and Bits

2.6.4 Logarithms and Multiplication

$$log_a(xy) = log_a(x) + log_a(y)$$

$$log_a n^b = b \cdot log_a n$$

$$a^b = e^{(ln(a^b))} = e^{(b(ln(a)))}$$

2.6.5 Fast Exponentiation

2.6.6 Logarithms and Summations

Harmonic Numbers:

$$H(n) = \sum_{i=1}^{n} \frac{1}{i} \sim ln(n)$$

2.6.7 Logarithms and Criminal Justice

| Loss (apply the greatest) | Increase in level |
|----------------------------|-------------------|
| (A) \$2,000 or less | no increase |
| (B) More than \$2,000 | add 1 |
| (C) More than \$5,000 | add 2 |
| (D) More than \$10,000 | add 3 |
| (E) More than \$20,000 | add 4 |
| (F) More than \$40,000 | add 5 |
| (G) More than \$70,000 | add 6 |
| (H) More than \$120,000 | add 7 |
| (I) More than \$200,000 | add 8 |
| (J) More than \$350,000 | add 9 |
| (K) More than \$500,000 | add 10 |
| (L) More than \$800,000 | add 11 |
| (M) More than \$1,500,000 | add 12 |
| (N) More than \$2,500,000 | add 13 |
| (O) More than \$5,000,000 | add 14 |
| (P) More than \$10,000,000 | add 15 |
| (Q) More than \$20,000,000 | add 16 |
| (R) More than \$40,000,000 | add 17 |
| (Q) More than \$80,000,000 | add 18 |

Figure 10: The Federal Sentencing Guidelines for fraud

Take-Home Lesson: Logarithms arise whenever things are repeatedly halved or doubled

2.7 Properties of Logarithms

$$log_a b = \frac{log_c b}{log_c a}$$

Stop and Think: Importnce of an Even Split

How many queries does binary search take on the million-name Manhattan phone boo if each split was 1/3 to 2/3 instead of 1/2 to 1/2?

- 2.8 War Story: Mystery of the Pyramids
- 2.9 Advanced Aanalysis (*)
- 2.10 Esoteric Functions
- 2.11 Limits and Dominance Relations

Take-Home Lesson: By interleaving the functions here with those of Section 2.3.1 we see where everything fits into the dominnce pecking order:

$$n! \gg c^n \gg n^3 \gg n^2 \gg n^{1+\epsilon} \gg n \cdot log(n) \gg n \gg \sqrt{n}$$

$$\gg log^2n \gg log(n) \gg \frac{log(n)}{log(log(n))} \gg log(log(n)) \gg \alpha(n) \gg 1$$

3 Data Structures

3.1 Contigous vs. Linked Data Structures