

# CSE 3500 Algorithms and Complexity – Fall 2016

## Lecture 9: September 27, 2016

### Hints on Homework 1

- Problem 1: When analyzing nested loops start from the innermost loop and progress outward. As an example, consider:

```
Result = 0;
for i = 1 to n do
    for j = 1 to i do
        Result++;
```

The instruction in the second **for** loop takes one unit of time per execution. As a result, the second **for** loop takes  $i$  time. This in turn means that the run time of the first **for** loop is  $\sum_{i=1}^n i = \frac{n(n+1)}{2} = \Theta(n^2)$ .

- Problem 2c: Let  $f(n)$  and  $g(n)$  be two non-negative functions of  $n$ . We say  $f(n) = o(g(n))$  if  $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$ . As an example, let  $f(n) = n^2$  and  $g(n) = n^2 \log \log n$ . Then  $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \lim_{n \rightarrow \infty} \frac{1}{\log \log n} = 0$ . Thus, in this case  $f(n) = o(g(n))$ .
- A general hint: When comparing two functions, it might help to express both of them as 2 to the power of some functions and compare the exponents. For example consider the two functions  $f_1(n) = \sqrt{n}$  and  $f_2(n) = 2\sqrt{\log n}$ . We note that  $\sqrt{n} = 2^{(1/2)\log n}$ . When we compare the exponents of these two functions we realize that  $f_1(n)$  is larger than  $f_2(n)$ .
- Here is a (small) list of functions in increasing order (In this list  $\epsilon$  is any constant such that  $0 < \epsilon < 1$  and  $\alpha$  is a constant  $> 0$ ):

$$\Theta(1), (\log n)^\epsilon, \log n, (\log n)^{1+\alpha}, 2^{(\log n)^\epsilon}, n^\epsilon, n, n^{1+\alpha}, 2^{n^\epsilon}, 2^n, 2^{n^{1+\alpha}}$$

- Problem 3b:  $\max\{f(n), g(n)\}$  is nothing but the pointwise maximum between  $f(n)$  and  $g(n)$ .
- We can easily see that  $\max\{f(n), g(n)\} = O(f(n) + g(n))$  since  $\max\{f(n), g(n)\} \leq (f(n) + g(n))$ .

## Master Theorem

- We studied the Master theorem in the last lecture.
- There are recurrence relations that may not be solvable using the Master theorem.
- As an example, consider the recurrence relation:  $T(n) = 27T(n/3) + n^3 2^{\sqrt{\log n}}$ . Here  $a = 27, b = 3$ , and  $f(n) = n^3 2^{\sqrt{\log n}}$ .  $n^{\log_b a} = n^3$ . It seems like case 3 may be applicable. For case 3 to apply, it should be the case that there exists a constant  $\epsilon > 0$  such that  $f(n) = \Omega(n^{\log_b a} n^\epsilon)$ . I.e.,  $n^3 2^{\sqrt{\log n}} = \Omega(n^3 n^\epsilon)$ . In other words,  $2^{\sqrt{\log n}} = \Omega(n^\epsilon)$ . However, there is no such constant since  $2^{\sqrt{\log n}} = o(n^\epsilon)$  for any constant  $\epsilon > 0$ .

## Solving a recurrence relation by induction

- Another technique for solving a recurrence relation uses a guess and a proof by induction.
- Steps involved in this technique are:
  1. Guess a solution to the recurrence relation;
  2. Attempt to prove the guess by induction;
  3. Iterate if there is a need.
- An Example: Consider the recurrence relation:  $T(n) = 2T(n/2) + n$ . We make the following guess:  $T(n) \leq cn \log n$  for some constant  $c$ .
- Now a proof by induction is attempted. The base case is easy.
- Induction step: Assume the hypothesis for all inputs of size up to  $n - 1$ . We'll prove it for  $n$ .
- $T(n) = 2T(n/2) + n$ . Applying the induction hypothesis to  $T(n/2)$  we realize that  $T(n) \leq 2c \frac{n}{2} \log \left(\frac{n}{2}\right) + n = cn \log n - (c - 1)n$ . The RHS will be  $\leq cn \log n$  when  $c \geq 1$ . Thus we infer that  $T(n) \leq n \log n$ .

## Quick sort

- We have revisited the quick sort algorithm and analyzed its worst case and best case run times.
- **Lemma:** The average run time of quick sort is  $O(n \log n)$  on any sequence of  $n$  elements.

- **Proof:** Let  $X = k_1, k_2, \dots, k_n$  be the input sequence and let  $\pi_1, \pi_2, \dots, \pi_n$  be the sorted order of  $X$ .

Let

$$X_{ij} = \begin{cases} 1 & \text{if } \pi_i \text{ and } \pi_j \text{ will be compared} \\ 0 & \text{otherwise.} \end{cases}$$

Total number of comparisons made in the algorithm is  $\sum_{j=i+1}^n \sum_{i=1}^n X_{ij}$ . We are interested in computing the expected value of this.

The average number of comparisons made in the algorithm  $A(n)$  is given by

$$A(n) = E\left[\sum_{j=i+1}^n \sum_{i=1}^n X_{ij}\right] = \sum_{j=i+1}^n \sum_{i=1}^n E[X_{ij}] \quad (1)$$

using the fact that  $E[X_1 + X_2] = E[X_1] + E[X_2]$ .

Let  $p_{ij}$  be the probability that  $\pi_i$  and  $\pi_j$  will be compared in the quick sort algorithm. Then,  $E[X_{ij}] = 1 \times p_{ij} + 0 \times (1 - p_{ij}) = p_{ij}$ .

Substituting this in equation 1, we get:

$$A(n) = \sum_{j=i+1}^n \sum_{i=1}^n p_{ij}. \quad (2)$$

The proof will be completed in the next lecture.