

**CSE 5500 Algorithms**  
Fall 2018; Exam I – Helpsheet

1. **Preliminaries.** We say  $f(n) = O(g(n))$  if  $f(n) \leq cg(n)$  for all  $n \geq n_0$  for some constants  $c$  and  $n_0$ . We say  $f(n) = \Omega(g(n))$  if and only if  $g(n) = O(f(n))$ . Also,  $f(n) = \Theta(g(n))$  if  $f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$ .

A partial list of functions in increasing order is:  $O(1)$ ,  $(\log n)^\epsilon$ ,  $\log n$ ,  $(\log n)^{1+\mu}$ ,  $n^\epsilon$ ,  $n$ ,  $n^{1+\mu}$ ,  $2^{n^\epsilon}$ ,  $2^n$ ,  $2^{n^{1+\mu}}$  where  $0 < \epsilon < 1$  and  $\mu > 0$  are constants.

Stirling's approximation:  $n! \approx (n/e)^n \sqrt{2\pi n}$ .

$$\sum_{i=1}^n i = n(n+1)/2. \quad \sum_{i=1}^n i^2 = n(n+1)(2n+1)/6. \quad \sum_{i=1}^n i^3 = n^2(n+1)^2/4.$$

2. **Master theorem.** Consider the recurrence relation:  $T(n) = aT(n/b) + f(n)$ , where  $a \geq 1$  and  $b > 1$  are constants.  
**Case 1:** If  $f(n) = O(n^{\log_b a - \epsilon})$  for some constant  $\epsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$ . **Case 2:** If  $n^{\log_b a} = \Theta(f(n))$ , then  $T(n) = \Theta(f(n) \log n)$ . **Case 3:** If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some constant  $\epsilon > 0$  and  $af(n/b) \leq cf(n)$  for some constant  $c < 1$ , then,  $T(n) = \Theta(f(n))$ .

3. **Randomized algorithms.** A Monte Carlo algorithm runs for a prespecified amount of time and its output is correct with high probability. By high probability we mean a probability of  $\geq (1 - n^{-\alpha})$ , for any constant  $\alpha$ . A Las Vegas algorithm always outputs the correct answer and its run time is a random variable. We say the run time of a Las Vegas algorithm is  $\tilde{O}(f(n))$  if the run time is  $\leq cf(n)$  for all  $n \geq n_0$  with probability  $\geq (1 - n^{-\alpha})$ , for some constants  $c$  and  $n_0$ .

**Chernoff bounds:** If  $X$  is a random variable with a binomial distribution  $B(n, p)$ , then  $Pr[X \geq (1 + \epsilon)np] \leq \exp(-\epsilon^2 np/3)$  and  $Pr[X \leq (1 - \epsilon)np] \leq \exp(-\epsilon^2 np/2)$ , for any  $0 < \epsilon < 1$ .

4. **Sorting.** The mergesort algorithm has a worst case run time of  $O(n \log n)$ . The run time of quicksort is  $O(n^2)$  in the worst case and  $O(n \log n)$  on the average.

Frazer-McKellar's randomized algorithm does sorting using  $n \log n + \tilde{O}(n \log \log n)$  comparisons.

**Radix or Integer Sorting:** We can sort  $n$  keys in  $O(n)$  time if the keys are integers in the range  $[1, n^c]$  (for any constant  $c$ ).

5. **Selection:** This problem takes as input a sequence  $X = k_1, k_2, \dots, k_n$  of arbitrary real numbers and an integer  $i$ ,  $1 \leq i \leq n$ . The problem is to identify the  $i$ th smallest element of  $X$ . The quickselect algorithm runs in  $O(n^2)$  time in the worst case and  $O(n)$  time on the average. BFPRT algorithm runs in  $O(n)$  time in the worst case. The randomized algorithm of Floyd and Rivest makes only  $n + \min\{i, n - i\} + \tilde{O}(n)$  comparisons. In this context we used the following lemma: Let  $X$  be any set of arbitrary real numbers and let  $S$  be a random sample of  $X$  with  $|S| = s$ . Let  $q$  be an element of  $S$  such that  $\text{rank}(q, S) = j$ . If  $r_j$  is the rank of  $q$  in  $X$ , then the following holds:  
 $\text{Prob.} \left[ |r_j - j \frac{n}{s}| > \sqrt{4\alpha} \frac{n}{\sqrt{s}} \sqrt{\log n} \right] \leq n^{-\alpha}$ , for any  $\alpha > 0$ .

6. **Fingerprinting:** The technique of fingerprinting can be applied to check if two given objects are the same. Instead of comparing the objects directly, we apply a function on each and compare the images. Some examples we have seen are: 1) For three given  $n \times n$  matrices  $A, B$ , and  $C$ , the problem is to check if  $AB = C$ . We pick a random vector  $r$  from  $\{0, 1\}^n$  and check if  $A(Br) = Cr$ . If so, we output: " $AB = C$ "; else we output " $AB \neq C$ ". If  $AB \neq C$ , we showed that  $Pr[A(Br) = Cr] \leq 1/2$ ; 2) Given three polynomials  $f(x), g(x)$ , and  $h(x)$ , the problem is to check if  $h(x) = f(x) \times g(x)$ . The idea of fingerprinting can be used as follows: we pick a random element  $r$  from  $\mathcal{S}$  (which is a subset of the field  $\mathcal{F}$ ) and check if  $f(r) \times g(r) = h(r)$ . If so, we output: " $h(x) = f(x) \times g(x)$ "; else we output: " $h(x) \neq f(x) \times g(x)$ ". We can show that the probability of an incorrect answer is no more than  $\frac{d}{|\mathcal{S}|}$  where  $d$  is the degree of  $h(x) - f(x) \times g(x)$ ; 3) Schwartz-Zippel theorem extends the above technique to check if a given multivariate polynomial is identically zero; 4) Edmonds' theorem in conjunction with Schwartz-Zippel theorem can be used to check for the existence of perfect matchings in graphs; 5) Karp-Rabin's algorithm applies fingerprinting to the string matching problem. To check if two given  $n$ -bit integers  $A$  and  $B$  are the same, the basic idea of this algorithm is to pick a random prime  $p \leq t$  and check if  $A \bmod p = B \bmod p$ . Probability of an incorrect answer is  $O\left(\frac{n}{t/(\log t)}\right)$ .