

CSL863: Randomized Algorithms

II semester, 2007-08

Homework # 3

Due before class on **Friday, 11th April 2008**

Instructor: Sandeep Sen

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1. Show that for any collection of hash function H . there exists x, y such that

$$\sum_{h \in H} \delta_h(x, y) \geq |H| \left(\frac{1}{m} - \frac{1}{n} \right)$$

where n and m are the sizes of universe and table respectively.

2. Let U be a universe of all possible keys of size N . Then prove that the size of a set of perfect hash functions that map n keys into a table of ($m \geq n$) is at least

$$\frac{C(N, n)}{(N/n)^n \cdot C(m, n)}$$

Here $C(a, b)$ denotes number of choices of b subsets from a set of a objects .

3. If a, b are chosen uniformly at random then show that the hash function $h(x) = (ax + b) \bmod p$ maps $x \neq 0$ uniformly at random to one of the p values , i.e. probability $(h(x) = i) = 1/p$ for $0 \leq i \leq p - 1, p$ is prime. Further show that for a pair of elements x, y , probability $(h(x) = i, h(y) = j) = \text{probability}(h(x) = i) \cdot \text{probability}(h(y) = j)$.
4. Let $|T| = p$ where p is a prime. Define a hash function from $U = p^k$ to T as follows. For a key $x = \langle s_1, s_2 \dots s_k \rangle$ $0 \leq s_i \leq p - 1$, and $a = \langle a_1, a_2 \dots, a_k \rangle$ $a_i < p$, the hash function $h_a(x) = \sum_i a_i \cdot s_i \bmod p$.
Prove that h_a defines a **strongly universal** hash family.
5. Suppose T is an ordered table of n keys x_i , $1 \leq i \leq n$ drawn uniformly from $(0, 1)$. Instead of doing the conventional binary search, we use the following approach.

Given key y , we make the first probe at the position $s_1 = \lceil y \cdot n \rceil$. If $y = x_{s_1}$, we are through. Else if $y > x_{s_1}$, we recursively search for y among the keys $(x_{s_1} \dots x_n)$
else recursively search for y among the keys $(x_1 \dots x_{s_1})$.

At any stage when we search for y in a range $(x_l \dots x_r)$, we probe the position $l + \lceil \frac{(y-x_l)(r-l)}{x_r-x_l} \rceil$. We are interested in determining the expected number of probes required by the above searching algorithm.

In order to somewhat simplify the analysis, we modify the algorithm as follows. In round i , we partition the input into $n^{1/2^i}$ sized blocks and try to locate the block that contains y and recursively search within that block. In the i -th round, if the block containing y is $(x_l \dots x_r)$, then we probe the position $s_i = l + \lceil \frac{(y-x_l)(r-l)}{x_r-x_l} \rceil$. We then try to locate the $n^{1/2^i}$ -sized block by sequentially probing every $n^{1/2^i}$ -th element starting from s_i .

Analyze the expected number of probes required. (Analyze the expected number of probes in each round using Chebychev's inequality).

6. Given a set of points on the real-line with coordinates $x_1, x_2 \dots x_n$, we want to determine if there is a subset $x_i, x_{i+1} \dots x_{i+m-1}$ with separation distances d_i $i \leq m$. Design a $O(n)$ algorithm for this problem.
7. Given a set S of n points in a plane - design a linear time algorithm to find the smallest enclosing circle of S .
8. Show how to implement the contraction algorithm (the original n contractions) in
 - (i) $O(n^2)$ time. You may want to use an adjacency matrix representation. To choose a random edge first choose a random vertex with probability proportional to its degree and then choose one of the neighbours at random. Prove that this works as required.
 - (ii) $O(m \log n)$ time.

Extend the solution of the first part to weighted graphs.