CS 630 – Fall 2024 – Lab 7 Oct 30, 2024

1. Hashing

Suppose we have a hash function defined as $h(x) = (3x + 5) \mod n$ where n equals to the hash table size, and we have following elements [42, 56, 21, 17, 29, 18, 55, 80, 20, 1].

1. Supporse we have a hash table of size 10, what is the resulting hash table.

Solution:

The hash-table will look like this: [[55], [42], [29], [56], [], [80, 20], [17], [], [21, 1], [18]]

2. What is the ideal hash table size if we want to have a collision rate for each cell that is lower than 10% with 10 elements to insert?

Solution:

This is the same calculation as the birthday paradox. There exists $10 \times 9/2 = 45$ possible parings between the 10 elements (suppose they are unique). Then the probability that each pair has $\frac{n-1}{n}$ chance of not colliding with uniform hashing. Therefore we get:

$$\Pr(\text{at least one pair share the same hashing}) = 1 - (\frac{n-1}{n})^{45} = 10\%$$

$$n \approx 427.61$$

2. Markov v.s. Chernoff bounds

Let X be a random variable representing the number of heads obtained when flipping n independent coins, each with a probability p of showing heads. We know that X follows a binomial distribution, $X \sim \text{Binomial}(n, p)$, with E[X] = np and Var[X] = np(1 - p)

1. Calculate an upper bound for $P(X \ge a)$ where a > E[X] using Markov's Inequality.

Solution:

$$P(X \ge a) \le \frac{np}{a}$$

2. Calculate an upper bound for $P(X \ge a)$ where a > E[X] using Chebyshev's Inequality.

Solution:

$$P(X \ge a) = P(X - np \ge a - np)$$

$$\le P(|X - np| \ge a - np)$$

$$= P\left(|X - np| \ge \frac{a - np}{\sqrt{(np(1 - p))}}\sqrt{(np(1 - p))}\right)$$

$$\le \frac{np(1 - p)}{(a - np)^2}$$

3. Calculate an upper bound for $P(X \ge a)$ where a > E[X] using Chernoff Bound.

Solution:

$$P(X \ge a) = P\left(X \ge \frac{a}{np}np\right)$$
$$= P\left(X \ge \left(1 + \frac{a - np}{np}\right)np\right)$$
$$\le \left(\frac{e^{\frac{a - np}{np}}}{(a/np)^{a/np}}\right)^{np}$$

3. Better chance

Given a function with randomized output that is correct with probability of 0.6 with fixed time c, write an algorithm that can produce a correct result with probability higher than 0.9. Create the algorithm and analyze how much time the algorithm is expected to take.

Solution:

We repeat the function n time and take the answer that is the majority. Let X be number of wrong answers. so $X \sim \text{Binomial}(n, 0.4)$, so we have E[X] = 0.4n and Var[X] = 0.24n. So the probability of our algorithm returns an error can be bounded by Chebyshev's Inequality.

$$P(X \ge 0.5n) = P(X - 0.4n \ge 0.1n)$$

$$\le P(|X - 0.4n| \ge 0.1n)$$

$$= P(|X - 0.4n| \ge \frac{0.1\sqrt{n}}{\sqrt{0.24}}\sqrt{0.24n})$$

$$\le \frac{24}{n}$$

To get success rate of 0.9 we can set n=240 so that the fail rate is ≤ 0.1 so we get $P(correct) \geq 0.9$

4. Las Vegas and Monte Carlo Algorithm

A Monte Carlo algorithm is a type of randomized algorithm that may produce incorrect results with a small probability but typically runs in deterministic time. In contrast, a Las Vegas algorithm always produces a correct result, but its running time may vary.

1. Given a Monte Carlo algorithm MC(x) that has a probability p of outputting the correct answer with O(1) time complexity and a verifier V(x) that verifies the output in O(1) time. Write a Las Vegas Algorithm LV(x) and give a expected time complexity analysis.

Solution:

Repeat calling MC(x) and verifying the result with V(x), stop when V(x) returns true. The number of repetition is a geometric distribution with probability of success p. The expected runtime is O(1/p)

2. Given a Las Vegas algorithm LV(x) with a runtime of T(x). Write a Monte Carlo algorithm MC(x) with a constant time complexity.

Solution:

Specify some constant time T, run LV(x), early stop if LV(x) returns an output by T, return the output, else return nothing. This will have O(1) time complexity.