

Assignment 1: Randomized Algorithm Design

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January 18, 2019

Assume that the random variables are either continuous or discrete if not explicitly mentioned.

1. Let $\mathcal{X}_i, i \in [n]$ be n random variables each with finite support. Then prove the following.

$$\text{var} \left(\sum_{i=1}^n \mathcal{X}_i \right) = \sum_{i=1}^n \text{var}(\mathcal{X}_i) + 2 \sum_{1 \leq i < j \leq n} \text{cov}(\mathcal{X}_i, \mathcal{X}_j)$$

where for any two random variables \mathcal{X} and \mathcal{Y} , we define $\text{cov}(\mathcal{X}, \mathcal{Y}) = \mathbb{E}[\mathcal{X}\mathcal{Y}] - \mathbb{E}[\mathcal{X}]\mathbb{E}[\mathcal{Y}]$.

2. Let \mathcal{X} and \mathcal{Y} be two independent random variables. Then prove that $\mathbb{E}[\mathcal{X}\mathcal{Y}] = \mathbb{E}[\mathcal{X}]\mathbb{E}[\mathcal{Y}]$. From this conclude that, for n pairwise random variables $\mathcal{X}_i, i \in [n]$, we have the following.

$$\text{var} \left(\sum_{i=1}^n \mathcal{X}_i \right) = \sum_{i=1}^n \text{var}(\mathcal{X}_i)$$

3. Fix any input sequence of n integers to the quick sort algorithm. Let \mathcal{X} be the random variable denoting the number of comparisons the quick sort algorithm makes on the input sequence. Then prove that $\text{var}(\mathcal{X}) = \mathcal{O}(n^2)$.
4. Let $A_i, i \in [n]$ be n objects each having two attributes A_i^x and A_i^y . The attribute y is 0 for every A_i . Suppose we have a deterministic quick sort algorithm that can sort $A_i, i \in [n]$ on the attribute x or on the attribute y . Can you use this deterministic quick sort algorithm to design a randomized algorithm to sort $A_i, i \in [n]$ on the attribute x which makes an expected $\mathcal{O}(n \log n)$ comparisons? Please prove that your algorithm indeed makes $\mathcal{O}(n \log n)$ comparisons on expectation.
5. Let $\mathcal{X}_i, i \in [n]$ be n pairwise independent random variables each taking values in $\{0, 1\}$ with expectation μ and $\mathcal{S} = \sum_{i=1}^n \mathcal{X}_i$. Then for any positive real number δ we have the following.

$$\Pr \left[\mathcal{S} \leq (1 - \delta)\mu \right] \leq \left(\frac{e^{-\delta}}{(1 - \delta)^{1-\delta}} \right)^\mu$$

6. Show that the expected number of balls one needs to throw randomly into m bins to have every bin at least one ball is $\mathcal{O}(m \log m)$.
7. Give an example of a random variable whose expectation exists but variance does not exist.
8. Find the expectation and variance of the number of swaps that the bubble sort algorithm performs on a uniformly random permutation of n distinct integers.

9. Prove the weak law of large numbers using Chebyshev inequality. The weak law of large number states that, for random variables $X_i, i \in \mathbb{N}$ which are distributed independently and identically with mean μ and variance σ^2 , we have the following for any constant $\varepsilon > 0$

$$\lim_{n \rightarrow \infty} \Pr \left[\left| \frac{X_1 + X_2 + \dots + X_n}{n} - \mu \right| > \varepsilon \right] = 0$$

10. Let $X_i, i \in [n]$ be n pairwise independent random variables each taking values in $\{0, 2\}$ with expectation μ and $S = \sum_{i=1}^n X_i$. Use standard Chernoff bound proved in class to upper bound the probability that S takes value more than $(1 + \delta)\mu$.
11. Let X be a random variable with expectation μ and variance σ^2 . Then for any $t \in \mathbb{R}_{\geq 0}$, prove the following.

$$\Pr [X - \mu \geq t\sigma] \leq \frac{1}{1 + t^2} \text{ and } \Pr [X - \mu \leq -t\sigma] \leq \frac{1}{1 + t^2}$$

12. Let X be a non-negative integer valued random variable with positive expectation. Then prove the following.

$$\Pr [X = 0] \leq \frac{\mathbb{E}[X^2] - \mathbb{E}[X]^2}{\mathbb{E}[X]^2} \text{ and } \frac{\mathbb{E}[X]^2}{\mathbb{E}[X^2]} \leq \Pr[X \neq 0] \leq \mathbb{E}[X]$$