

By *random variables* or *discrete random variables* we mean random variables taking either finitely many values or countably infinite values.

1. Given a positive integer k , describe a non-negative random variable X such that

$$\Pr[X \geq k\mathbb{E}[X]] = \frac{1}{k}.$$

2. Let X be a non-negative integer-valued random variable such that $X \leq m$, and $\mathbb{E}[X] \geq 2m^{1-t\delta/2}$. Prove that

$$\Pr[X \geq m^{1-t\delta/2}] \geq m^{-t\delta/2}.$$

3. Let the random variable X be given by $X = \sum_{i=1}^n X_i$. Show that if $\mathbb{E}[X_i X_j] = \mathbb{E}[X_i]\mathbb{E}[X_j]$ for every pair of i and j with $1 \leq i < j \leq n$, then $\text{Var}[X] = \sum_{i=1}^n \text{Var}[X_i]$.
4. Give an example of a random variable with finite expectation, and unbounded variance.
5. (Probability amplification) Let a and b be chosen independently and randomly from $\mathbb{Z}_n = \{0, 1, 2, \dots, n-1\}$, where n is a prime. Let $f : \mathbb{Z}_n \rightarrow \{0, 1\}$ be an unknown but fixed function, such that $f(x) = 1$ for a random subset $W \subset \mathbb{Z}_n$, which is called the *witness set*.

- (i) Compute the probability that none of a, b belong to the witness set. How many random bits did you need to generate a and b ? If you select t random numbers a_1, \dots, a_t , such that the probability that none of the selected numbers lies in the witness set is at most $1/t$, how many random bits do you need (here $0 \leq t < n$)?

- (ii) A set of random variables X_1, \dots, X_k is said to be *pairwise independent* if for all $i \neq j$, for all $x, y \in \mathbb{R}$, we have $\Pr[X_i = x | X_j = y] = \Pr[X_i = x]$.

Suppose we generate t pseudo-random numbers from \mathbb{Z}_n by choosing $r_i = a \cdot i + b \pmod n$, for $1 \leq i \leq t$. Let $|W| = n/2$. Show that (a) the r_i 's are pairwise independent. (b) The probability that none of the r_i 's belong to the witness set is at most $1/t$. How many random bits were needed using this method?

6. (Chernoff Bounds: Upper Tail) Let X be the sum of n independent indicator random variables, each equal to 1 with probability p , and zero otherwise. Let μ denote $\mathbb{E}[X]$.
 - (a) Apply the substitution $Y = e^{tX}$. Given $\delta > 0$, express the event $X > (1 + \delta)\mu$ in terms of Y .
 - (b) Obtain an upper bound on the expression obtained in (a), by applying Markov's inequality to Y .
 - (c) Obtain an upper bound on the moment generating function of X , i.e. $\mathbb{E}[Y]$, in terms of n, t and p .
 - (d) Substitute the bound obtained in (c), to the expression obtained in (b).
 - (e) Differentiate the expression obtained in (c) w.r.t. t and optimize to get the tightest possible upper bound.
7. (Chernoff Bounds: Lower Tail) Redo the previous exercise, but with the event $X < (1 - \delta)\mu$, to get an upper bound on its probability of occurrence.
8. Let $\mathcal{G}_p(n)$ be the random graph model having vertices $V = 1, 2, \dots, n$, and each pair of vertices joined by an edge with probability $p = p(n)$ independently of the others.

- (a) The degree of a vertex $v \in V$ is the number of edges incident on v . Compute the expected degree of a vertex in $\mathcal{G}_p(n)$ in terms of n, p .
- (b) Let $p = n^{-\epsilon}$, where $\epsilon > 0$. Find the maximum degree of the random graph $\mathcal{G}_p(n)$, with probability tending to 1 as $n \rightarrow \infty$.
9. Suppose we have n jobs to distribute among m processors. [Assume m divides n]. A job requires one unit of time with probability p , and $k > 1$ units of time with probability $1 - p$. Use Chernoff bounds, to derive upper and lower bounds on the time required (with high probability) for all jobs to be completed, if we randomly assign n/m jobs to each processor. (Notice the indicator variables are not 0 – 1 variables here!)