CSE 312: Foundations of Computing II Quiz Section #6: Important Discrete Random Variables

Review: Main Theorems and Concepts

Variance: Let X be a random variable and $\mu = \mathbb{E}[X]$. The variance of X is defined to be $\operatorname{Var}(X) = \underbrace{((X - \mu)^2)}_{,}$ variance is always ______. Notice that since this is an expectation of a ______ random variable $((X - \mu)^2)$, variance is always ______. With some algebra, we can simplify this to $\operatorname{Var}(X) = \mathbb{E}[X^2] - \mathbb{E}^2[X]$. **Property of Variance:** Let $a, b \in \mathbb{R}$ and let X be a random variable. Then, $\operatorname{Var}(aX + b) = ______.$ **Independence:** Random variables X and Y are independent, written $X \perp Y$, iff

In this case, we have $\mathbb{E}[XY] = \mathbb{E}[X]\mathbb{E}[Y]$ (the converse is not necessarily true).

i.i.d. (independent and identically distributed): Random variables X_1, \ldots, X_n are i.i.d. (or iid) iff they are ______ and have the same ______.

Variance of Independent Variables: If $X \perp Y$, Var(X + Y) = Var(X) + Var(Y). This depends on independence, whereas linearity of expectation always holds. Note that this combined with the above shows that $\forall a, b, c \in \mathbb{R}$ and if $X \perp Y$, Var(aX + bY + c) =______.

Zoo of Discrete Random Variables

Uniform: $X \sim Unif(a, b)$, for integers $a \leq b$, iff X has the following probability mass function:

$$p_X(k) = \frac{1}{b-a+1}, \ k = a, a+1, \dots, b$$

 $\mathbb{E}[X] = \frac{a+b}{2}$ and $\operatorname{Var}(X) = \frac{(b-a)(b-a+2)}{12}$. This represents each integer from [a, b] to be equally likely. For example, a single roll of a fair die is Unif(1, 6).

Bernoulli (or indicator): $X \sim Ber(p)$ iff X has the following probability mass function:

$$p_X(k) = \begin{cases} p, & k=1\\ 1-p, & k=0 \end{cases}$$

 $\mathbb{E}[X] = p$ and Var(X) = p(1 - p). An example of a Bernoulli r.v. is one flip of a coin with P (head) = p. By a clever trick, we can write

$$p_X(k) = p^k (1-p)^{1-k}, \ k = 0, 1$$

Binomial: $X \sim Bin(n, p)$ iff X is the sum of n iid Ber(p) random variables. X has probability mass function

$$p_X(k) = {n \choose k} p^k (1-p)^{n-k}, \ k = 0, 1, \dots, n$$

 $\mathbb{E}[X] = np$ and Var(X) = np(1-p). An example of a Binomial r.v. is the number of heads in *n* independent flips of a coin with P (head) = *p*. Note that $Bin(1, p) \equiv Ber(p)$. As $n \to \infty$ and $p \to 0$, with $np = \lambda$, then $Bin(n, p) \to Poi(\lambda)$. If X_1, \ldots, X_n are independent Binomial r.v.'s, where $X_i \sim Bin(N_i, p)$, then $X = X_1 + \ldots + X_n \sim Bin(N_1 + \ldots + N_n, p)$.

Geometric: $X \sim Geo(p)$ iff X has the following probability mass function:

$$p_X(k) = (1-p)^{k-1} p, \ k = 1, 2, \dots$$

 $\mathbb{E}[X] = \frac{1}{p}$ and $\operatorname{Var}(X) = \frac{1-p}{p^2}$. An example of a Geometric r.v. is the number of independent coin flips up to and including the first head, where P(head) = p.

Negative Binomial: $X \sim NegBin(r, p)$ iff X is the sum of r iid Geo(p) random variables. X has probability mass function

$$p_X(k) = \binom{k-1}{r-1} p^r (1-p)^{k-r}, \ k = r, r+1, \dots$$

 $\mathbb{E}[X] = \frac{r}{p}$ and $\operatorname{Var}(X) = \frac{r(1-p)}{p^2}$. An example of a Negative Binomial r.v. is the number of independent coin flips up to and including the r^{th} head, where P(head) = p. If X_1, \ldots, X_n are independent Negative Binomial r.v.'s, where $X_i \sim NegBin(r_i, p)$, then $X = X_1 + \ldots + X_n \sim NegBin(r_1 + \ldots + r_n, p)$.

Poisson: $X \sim Poi(\lambda)$ iff X has the following probability mass function:

$$p_X(k) = e^{-\lambda} \frac{\lambda^k}{k!}, \quad k = 0, 1, \dots$$

 $\mathbb{E}[X] = \lambda$ and $\operatorname{Var}(X) = \lambda$. An example of a Poisson r.v. is the number of people born during a particular minute, where λ is the average birth rate per minute. If X_1, \ldots, X_n are independent Poisson r.v.'s, where $X_i \sim Poi(\lambda_i)$, then $X = X_1 + \ldots + X_n \sim Poi(\lambda_1 + \ldots + \lambda_n)$.

Hypergeometric: $X \sim HypGeo(N, K, n)$ iff X has the following probability mass function:

$$p_X(k) = \frac{\binom{K}{k}\binom{N-K}{n-k}}{\binom{N}{n}}, \quad k = max\{0, n+K-N\}, \dots, \min\{K, n\}$$

 $\mathbb{E}[X] = n\frac{K}{N}$. This represents the number of successes drawn, when *n* items are drawn from a bag with *N* items (*K* of which are successes, and *N* – *K* failures) without replacement. If we did this with replacement, then this scenario would be represented as Bin $(n, \frac{K}{N})$.

Exercises

- 1. Suppose I am fishing in a pond with *B* blue fish, *R* red fish, and *G* green fish, where B + R + G = N. For each of the following scenarios, identify the most appropriate distribution (with parameter(s)):
 - (a) how many of the next 10 fish I catch are blue, if I catch and release
 - (b) how many fish I had to catch until my first green fish, if I catch and release
 - (c) how many red fish I catch in the next five minutes, if I catch on average r red fish per minute
 - (d) whether or not my next fish is blue
 - (e) how many of the next 10 fish I catch are blue, if I do not release the fish back to the pond after each catch
 - (f) how many fish I have to catch until I catch three red fish, if I catch and release
- 2. Suppose Y_1, \ldots, Y_n are iid with $\mathbb{E}[Y_i] = \mu$ and $\operatorname{Var}(Y_i) = \sigma^2$, and let $Y = \frac{1}{n} \sum_{i=1}^n iY_i$. What is $\mathbb{E}[Y]$ and $\operatorname{Var}(Y)$? Recall that $\sum_{i=1}^n i = \frac{n(n+1)}{2}$ and $\sum_{i=1}^n i^2 = \frac{n(n+1)(2n+1)}{6}$.

- 3. Is the following statement true or false? If $\mathbb{E}[XY] = \mathbb{E}[X]\mathbb{E}[Y]$, then $X \perp Y$. If it is true, prove it. If not, provide a counterexample.
- 4. Suppose we roll two fair 5-sided dice independently. Let X be the value of the first die, Y be the value of the second die, Z = X + Y be their sum, $U = \min \{X, Y\}$ and $V = \max \{X, Y\}$.
 - (a) Find $p_U(u)$.
 - (b) Find $\mathbb{E}[U]$.
 - (c) Find $\mathbb{E}[Z]$.
 - (d) Find $\mathbb{E}[UV]$.
 - (e) Find Var(U + V).
- 5. Suppose *X* has the following probability mass function:

$$p_X(x) = \begin{cases} c, & x = 0\\ 2c, & x = \frac{\pi}{2}\\ c, & x = \pi\\ 0, & \text{otherwise} \end{cases}$$

- (a) Suppose $Y_1 = \sin(X)$. Find $\mathbb{E}[Y_1^2]$.
- (b) Suppose $Y_2 = \cos(X)$. Find $\mathbb{E}[Y_2^2]$.
- (c) Suppose $Y = Y_1^2 + Y_2^2 = \sin^2(X) + \cos^2(X)$. Before any calculation, what do you think $\mathbb{E}[Y]$ should be? Find $\mathbb{E}[Y]$, and see if your hypothesis was correct. (Recall for any real number x, $\sin^2(x) + \cos^2(x) = 1$).
- (d) Let W be any discrete random variable with probability mass function $p_W(w)$. Then, $\mathbb{E}[\sin^2(W) + \cos^2(W)] = 1$. Is this statement always true? If so, prove it. If not, give a counterexample by giving a probability mass function for a discrete random variable W for which the statement is false.
- 6. If electricity power failures occur according to a Poisson distribution with an average of 3 failures every twenty weeks, calculate the probability that there will be more than one failure during a particular week.
- 7. An average page in a book contains one typo. What is the probability that there are exactly 8 typos in a given 10-page chapter, using the Poisson model?
- 8. A company makes electric motors. The probability an electric motor is defective is 0.01, independent of other motors made. What is the probability that a sample of 300 electric motors will contain exactly 5 defective motors? Do it first exactly, then approximate it with a Poisson distribution. How good is the approximation?
- 9. Suppose that $X_1 \sim Poi(\lambda_1)$ and $X_2 \sim Poi(\lambda_2)$, and X_1 and X_2 are independent. Prove that $Y = X_1 + X_2 \sim Poi(\lambda_1 + \lambda_2)$.