

# CSE 312: Foundations of Computing II

## Quiz Section #6: 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  $\text{Var}(X) = \mathbb{E}[(X - \mu)^2]$ . Notice that since this is an expectation of a random variable  $((X - \mu)^2)$ , variance is always non-negative. With some algebra, we can simplify this to  $\text{Var}(X) = \mathbb{E}[X^2] - \mathbb{E}^2[X]$ .

**Property of Variance:** Let  $a, b \in \mathbb{R}$  and let  $X$  be a random variable. Then,  $\text{Var}(aX + b) = \text{Var}(X)$ .

**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  $X_1, \dots, X_n$  are i.i.d. (or iid) iff they are independent and have the same distribution.

**Variance of Independent Variables:** If  $X \perp Y$ ,  $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$ . 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  $X \perp Y$ ,  $\text{Var}(aX + bY + c) = \text{Var}(aX) + \text{Var}(bY)$ .

### Zoo of Discrete Random Variables

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

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

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

**Bernoulli (or indicator):**  $X \sim \text{Ber}(p)$  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  $\text{Var}(X) = p(1 - p)$ . An example of a Bernoulli r.v. is one flip of a coin with  $P(\text{head}) = p$ . By a clever trick, we can write

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

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

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

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

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

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

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

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

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

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

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

**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  $\text{Bin}\left(n, \frac{K}{N}\right)$ .

## Exercises

- 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)):
  - how many of the next 10 fish I catch are blue, if I catch and release
  - how many fish I had to catch until my first green fish, if I catch and release
  - how many red fish I catch in the next five minutes, if I catch on average  $r$  red fish per minute
  - whether or not my next fish is blue
  - 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
  - how many fish I have to catch until I catch three red fish, if I catch and release
- Suppose  $Y_1, \dots, Y_n$  are iid with  $\mathbb{E}[Y_i] = \mu$  and  $\text{Var}(Y_i) = \sigma^2$ , and let  $Y = \frac{1}{n} \sum_{i=1}^n i Y_i$ . What is  $\mathbb{E}[Y]$  and  $\text{Var}(Y)$ ? 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  $V = \max\{X, Y\}$ .
- Find  $p_U(u)$ .
  - Find  $\mathbb{E}[U]$ .
  - Find  $\mathbb{E}[Z]$ .
  - Find  $\mathbb{E}[UV]$ .
  - Find  $\text{Var}(U + V)$ .
5. Suppose  $X$  has the following probability mass function:

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

- Suppose  $Y_1 = \sin(X)$ . Find  $\mathbb{E}[Y_1^2]$ .
  - Suppose  $Y_2 = \cos(X)$ . Find  $\mathbb{E}[Y_2^2]$ .
  - 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  $x$ ,  $\sin^2(x) + \cos^2(x) = 1$ ).
  - 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?