Review of Main Concepts

• Independence: Random variable X and event E are independent iff

$$\forall x, \quad \mathbb{P}(X = x \cap E) = \mathbb{P}(X = x)\mathbb{P}(E)$$

Random variables X and Y are independent iff

$$\forall x \forall y, \quad \mathbb{P}(X = x \cap Y = y) = \mathbb{P}(X = x)\mathbb{P}(Y = y)$$

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 mutually independent and have the same probability mass function.
- Independence of functions of a r.v.: If X and Y are independent and $g(\cdot), h(\cdot)$ are functions mapping real numbers to real numbers, then g(X) and h(Y) are independent. (See if you can prove this!)
- Variance of Independent Variables: If X is independent of 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 is independent of Y, $Var(aX + bY + c) = a^2 Var(X) + b^2 Var(Y)$.
- Review: Zoo of Discrete Random Variables
 - (a) Uniform: $X \sim \text{Uniform}(a, b)$ (Unif(a, b) for short), 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 $Var(X) = \frac{(b-a)(b-a+2)}{12}$. This represents each integer from [a, b] to be equally likely. For example, a single roll of a fair die is Uniform(1, 6).

(b) **Bernoulli (or indicator)**: $X \sim \text{Bernoulli}(p)$ (Ber(p) for short) 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 $\mathbb{P}(\text{head}) = p$.

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

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

 $\mathbb{E}[X] = np$ 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 \mathbb{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 $X = X_1 + \ldots + X_n \sim Bin(N_1 + \ldots + N_n, p)$.

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

(e) **Poisson**: $X \sim \text{Poisson}(\lambda)$ (Poi (λ) for short) 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 $Var(X) = \lambda$. 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 $X_i \sim \mathsf{Poi}(\lambda_i)$, then $X = X_1 + \ldots + X_n \sim \mathsf{Poi}(\lambda_1 + \ldots + \lambda_n)$.

(f) Negative Binomial: $X \sim \text{NegativeBinonial}(r, p)$ (NegBin(r, p) for short) iff X is the sum of r iid Geometric(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 $Var(X) = \frac{r(1-p)}{p^2}$. An example of a Negative Binomial r.v. is the number of independent coin flips up to and including the r^{th} head, where $\mathbb{P}(\text{head}) = p$. If X_1, \ldots, X_n are independent Negative Binomial r.v.'s, where $X_i \sim \text{NegBin}(r_i, p)$, then $X = X_1 + \ldots + X_n \sim \text{NegBin}(r_1 + \ldots + r_n, p)$.

(g) Hypergeometric: $X \sim \text{HyperGeometric}(N, K, n)$ (HypGeo(N, K, n) for short) iff X has the following probability mass function:

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

 $\mathbb{E}[X] = n \frac{K}{N}$. 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 $(n, \frac{K}{N})$.

1. Content Review Questions

- (a) True or false: Var(A + B) = Var(A) + Var(B)
- (b) What is Var(3A + 4)?

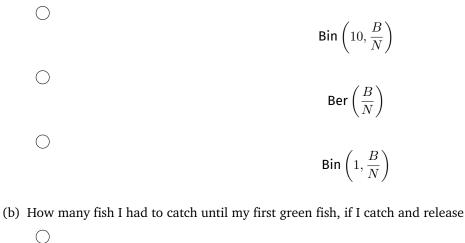
 $\bigcirc 3Var(A) + 4$

- \bigcirc 3Var(A)
- \bigcirc 9Var(A)
- $\bigcirc Var(A)$
- (c) True or false: $\mathbb{E}[A+B] = \mathbb{E}[A] + \mathbb{E}[B]$
- (d) What is $\mathbb{E}[3A+4]$?
 - $\bigcirc 3\mathbb{E}[A] + 4$
 - $\bigcirc \ 3\mathbb{E}[A]$
 - $\bigcirc \ 9\mathbb{E}[A]$
 - $\bigcirc \mathbb{E}[A]$

2. Pond Fishing

Suppose I am fishing in a pond with *B* blue fish, *R* red fish, and *G* green fish, where B + R + G = N. Each fish is equally likely to be caught. 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





- (c) How many red fish I catch in the next five minutes, if I catch on average r red fish per minute
 - $\bigcirc \qquad \qquad \mathsf{Poi}(5R) \\ \bigcirc \qquad \qquad \mathsf{Bin}\left(5,\frac{R}{N}\right) \\ \bigcirc \qquad \qquad \mathsf{Poi}(5r) \\ \end{aligned}$
- (d) Whether or not my next fish is blue

- (e) (optional) 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) (optional) How many fish I have to catch until I catch three red fish, if I catch and release

3. Balls in Bins

Note: this problem also appeared on the section 4 handout.

Let *X* be the number of bins that remain empty when *m* balls are distributed into *n* bins randomly and independently. For each ball, each bin has an equal probability of being chosen. (Notice that two bins being empty are not independent events: if one bin is empty, that decreases the probability that the second bin will also be empty. This is particularly obvious when n = 2 and m > 0.) Find $\mathbb{E}[X]$.

4. 3-sided Die

Note: a variation of this problem also appeared on the section 4 handout. Let the random variable *X* be the sum of two independent rolls of a fair 3-sided die. (If you are having trouble imagining what that looks like, you can use a 6-sided die and change the numbers on 3 of its faces.)

(a) What is the probability mass function of *X*?

(b) Find $\mathbb{E}[X]$.

(c) What is Var(X)?

5. Best Coach Ever!!

You are a hardworking boxer. Your coach tells you that the probability of your winning a boxing match is 0.2 independently of every other match.

- (a) How many matches do you expect to fight until you win 10 times and what kind of random variable is this?
- (b) You only get to play 12 matches every year. To win a spot in the Annual Boxing Championship, a boxer needs to win at least 10 matches in a year. What is the probability that you will go to the Championship this year and what kind of random variable is the number of matches you win out of the 12?
- (c) Let *p* be your answer to part (b). How many times can you expect to go to the Championship in your 20 year career?

6. Variance of a Product

Let X, Y, Z be independent random variables with means μ_X, μ_Y, μ_Z and variances $\sigma_X^2, \sigma_Y^2, \sigma_Z^2$, respectively. Find Var(XY - Z).

7. True or False?

Identify the following statements as true or false (true means always true). Justify your answer.

- (a) For any random variable X, we have $\mathbb{E}[X^2] \ge \mathbb{E}[X]^2$.
- (b) Let X, Y be random variables. Then, X and Y are independent if and only if $\mathbb{E}[XY] = \mathbb{E}[X]\mathbb{E}[Y]$.
- (c) Let $X \sim \text{Binomial}(n, p)$ and $Y \sim \text{Binomial}(m, p)$ be independent. Then, $X + Y \sim \text{Binomial}(n + m, p)$.
- (d) Let $X_1, ..., X_{n+1}$ be independent Bernoulli(p) random variables. Then, $\mathbb{E}[\sum_{i=1}^n X_i X_{i+1}] = np^2$.
- (e) Let $X_1, ..., X_{n+1}$ be independent Bernoulli(p) random variables. Then, $Y = \sum_{i=1}^n X_i X_{i+1} \sim \text{Binomial}(n, p^2)$.
- (f) If $X \sim \text{Bernoulli}(p)$, then $nX \sim \text{Binomial}(n, p)$.
- (g) If $X \sim \text{Binomial}(n, p)$, then $\frac{X}{n} \sim \text{Bernoulli}(p)$.
- (h) For any two independent random variables X, Y, we have Var(X Y) = Var(X) Var(Y).

8. Fun with Poissons

Let $X \sim Poisson(\lambda_1)$ and $Y \sim Poisson(\lambda_2)$, and X and Y are independent.

- (a) Show that $X + Y \sim Poisson(\lambda_1 + \lambda_2)$ [This was done in class.]
- (b) Show that $P(X = k \mid X + Y = n) = P(W = k)$ where $W \sim Bin(n, \frac{\lambda_1}{\lambda_1 + \lambda_2})$

9. Memorylessness

We say that a random variable X is memoryless if $\mathbb{P}(X > k + i \mid X > k) = \mathbb{P}(X > i)$ for all non-negative integers k and i. The idea is that X does not *remember* its history. Let $X \sim Geo(p)$. Show that X is memoryless.