<span id="page-0-0"></span>Center for Statistics and the Social Sciences Math Camp 2020 Discrete Distributions

### Jessica Godwin & Emily Finchum-Mason

Department of Statistics & Evans School University of Washington

## **Motivation**

A discrete random variable could have a number of different probability distributions. Today we will focus on the following discrete distributions.

- **•** Bernoulli
- Binomial
- **•** Multinomial
- **Geometric**
- **•** Hypergeometric
- **•** Poisson

Some experiments only have two possible outcomes. Often the experiments can be phrased as either having a characteristic or not having a characteristic. Examples

- Flipping a coin Heads or Tails
- Disease Incidence Either you have it or you don't
- A test pass or fail

## Bernoulli

Any experiment that can be described as either a 'success' or a 'failure' has a **Bernoulli** distribution. For  $0 \leq p \leq 1$ 

$$
X = \left\{ \begin{array}{ll} 1 \text{ with prob} & p \\ 0 \text{ with prob} & 1 - p \end{array} \right.
$$

You need to first determine what outcome is your 'success' usually the outcome of interest. For example, if you were trying to determine the probability of a specific birth defect, each child born would be assigned a 1 if he/she had the defect, 0 if not.

The probability distribution for a Bernoulli random variable:

$$
P(X = x | p) = p^{x}(1-p)^{1-x}
$$
 for  $x = 0, 1$ 

**Bernoulli** Mean & Variance

The expected value of X when  $X \sim \text{Bern}(p)$  (has a Bernoulli)

$$
E[X] = \sum_{i=1}^{n} x_i \cdot P(X = x_i | p)
$$
  
=  $0 \cdot p^0 (1 - p)^{1-0} + 1 \cdot p^1 (1 - p)^{1-1}$   
=  $0 + 1 \cdot p^1 (1 - p)^0$   
=  $p$ 

Bernoulli Mean & Variance

The variance of X when  $X \sim \text{Bern}(p)$ 

$$
V[X] = \sum_{i=1}^{n} (x_i - E[X])^2 \cdot P(X = x_i | p)
$$
  
=  $(0-p)^2 \cdot p^0 (1-p)^{1-0} + (1-p)^2 \cdot p^1 (1-p)^{1-1}$   
=  $p^2 (1-p)^1 + (1-p)^2 \cdot p^1$   
=  $p(1-p)[p + (1-p)] = p(1-p) \cdot 1$   
=  $p(1-p)$ 

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## **Bernoulli**

pmf for *Bernoulli*( $p = 0.3$ ).



X

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What if we were looking at several Bernoulli random variables? We could ask the question: how many of the random variables or experiments were successes?

Examples

- A person draws 3 cards from a deck. What is the probability two are red?
- We flip 10 coins. What is the probability we get 6 heads?
- We survey 25 teenage drivers. What is the probability that 15 of them have had car accidents?

The binomial distribution requires n independent experiments. We have to randomly sample people, flip fair coins, etc. The success of one random variable or experiment must have no effect on the success of any other random variable or experiment.

We say that we have  $n$  experiments, each with success probability p and failure probability  $1 - p$ .

Probability Distribution:

$$
P(X = x | n, p) = {n \choose x} p^{x} (1-p)^{n-x} \text{ for } x = 0, 1, 2, 3..., n
$$

This looks very similar to a Bernoulli except for the constant  $\binom{n}{x}$  $\binom{n}{x}$ .

 $\binom{n}{x}$  $n \choose x$  counts how many combinations of x successes and  $n - x$ failures we can have out of  $n$  experiments.

$$
\binom{n}{x} = \frac{n!}{x!(n-x)!}
$$

Example: To find the probability of 2 red cards out of 3 cards drawn, we need to consider all combinations of cards that result in 2 red cards.

Out of the 8 possible combinations  $S = \{RRR, RRB, BRR, BRR, RBB, BRB, BBR, BBB\}$  three of the combinations (RRB, RBR, BRR) result in 2 red cards.

$$
\binom{3}{2} = \frac{3!}{2!(3-2)!} = \frac{3 \cdot 2 \cdot 1}{(2 \cdot 1)(1)} = \frac{6}{2} = 3
$$

Mean & Variance

The easiest way to find the mean and variance of  $X$  when  $X \sim Bin(n, p)$  is to think of  $X = Y_1 + ... + Y_n$  where  $Y_i \sim Bern(p)$ .

$$
E[X] = E\left[\sum_{i=1}^{n} Y_i\right] = \sum_{i=1}^{n} E[Y_i]
$$

$$
= \sum_{i=1}^{n} p = np
$$

$$
V[X] = V\left[\sum_{i=1}^{n} Y_i\right] = \sum_{i=1}^{n} V[Y_i]
$$

$$
= \sum_{i=1}^{n} p(1-p) = np(1-p)
$$

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Distribution when  $p = 0.3$  and  $n = 10$ .





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## Multinomial

We can expand the binomial to the case where we have n experiments each with 3 possible outcomes.

For example: asking someone to check a box from (agree, couldn't care less, do not agree). We can ask 'out of ten people, what is the probability that three check agree, four checked couldn't care less, and three checked do not agree.' Here our outcome is a vector with a multinomial distribution. This can be expanded to include m outcomes.

$$
P(X = x_1, ..., x_m | n, p_1, ..., p_m) = \frac{n!}{x_1! \cdots x_m!} p_1^{x_1} \cdots p_m^{x_m}
$$

where  $p_1 + ... + p_m = 1$  and  $x_1 + ... x_m = n$ .

Multinomial Mean, Variance, & Covariance

> The expected number of times we observe outcome  $i$ , where  $i = 1, ..., m$ .

$$
E[X_i] = np_i
$$

The variance of  $X_i$ 

$$
V[X_i] = np_i(1-p_i)
$$

The covariance of  $X_i$  and  $X_i$ 

$$
cov[X_i,X_j]=-np_ip_j.
$$

## Geometric

Sometimes we are interested in how long it will take us to get to a success. That is, how many failure do we have to see before we get a success?

Examples

- What is the probability that we have to lose 10 games before a win?
- What is the probability that you flip four tails before a head?
- What is the probability that you will take your qualifying exams 3 times before you pass?

Geometric Mean & Variance

> These are examples of a **geometric** distribution. If  $X$ =the number of the trial with the first success, the probability distribution of X is:

$$
P(X = x | p) = p(1 - p)^{x-1} \text{ for } x = 1, 2, ....
$$

The expected value of  $X$  is

$$
E[X]=1/p
$$

The variance of  $X$ 

$$
V[X] = \frac{1-p}{p^2}.
$$

## Geometric

Distribution when  $p = 0.3$ ,  $X =$ the trial with the first success.



### Geometric Alternative parameterization

The geometric distribution can also be parameterized such that  $X =$  the number of failures before a success

$$
P(X = x | p) = p(1 - p)^{x} \text{ for } x = 0, 1, 2, \dots
$$

The expected value of  $X$  is

$$
E[X] = \frac{1-p}{p}
$$

The variance of X

$$
V[X] = \frac{1-p}{p^2}
$$

## Geometric

Distribution when  $p = 0.3$ ,  $X =$ the number of failures before the first success.



Often we are taking samples without replacement and are interested in the probability of  $k$  success out of the  $n$  samples taken from a population of size N. When you draw from a finite population, the probability of success changes on each draw.

For example, if you are in a room with 5 men and 5 women. Initially the probability of selecting a woman is  $1/2$ . If the first draw is a man, then there are 9 people left with 5 women. The probability of drawing a woman after a man has been selected is 5/9.

### Hyper Geometric Mean & Variance

The hypergeometric distribution requires three parameters. The size of the population N, the total number of 'successes'  $K \in \{0, 1, ..., N\}$  and the size of the sample  $n \in \{0, 1, ..., N\}$ .

The probability distribution of  $X$  is

$$
\frac{\binom{K}{x}\binom{N-x}{n-x}}{\binom{N}{n}}
$$

The expected value is

$$
E[X] = n\frac{K}{N}
$$

The variance is

$$
V[X] = n\frac{K(N-K)}{N} \frac{(N-n)}{(N-1)}.
$$

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## Hyper Geometric

Distribution when  $N = 100$ ,  $K = 30$ , and  $n = 10$ .



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## Poisson

The Poisson distribution is often used when we are interested in how many events will occur in fixed amount of time or space. Examples

- What is the probability that three people walk into a store in the next ten minutes?
- What is the probability that we get two phone calls in the next hour?
- What is the probability that 10 people are diagnosed with leukemia in King county in 2015?

Poisson Mean & Variance

The probability distribution for  $X \sim Poisson(\lambda)$  is

$$
P(X = x | \lambda) = \frac{e^{-\lambda} \cdot \lambda^x}{x!}
$$
 for  $x = 0, 1, 2, 3...$ 

where  $\lambda$  is the average number of occurrences in a given time period (rate).

The mean and variance of a  $X \sim Poisson(\lambda)$ 

$$
E[X] = \lambda, \ V[X] = \lambda
$$

# Poisson

Distribution when  $\lambda = 3$ .



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# Which Distribution?

A couple has four pets in a world where dogs and cats are the only pets.

 $X =$  number of dogs  $X \sim Bin(n = 4, p = 1/2)$ 

A student takes a qualifying exam until they pass, with a 0.9 probability of passing each time.

 $X =$  the number of times they take the exam

 $X \sim$  Geometric( $p = 0.9$ )

A summer camp is trying to buy sunblock for summer camps and needs to know if there will be more than 10 redheads in the camp if the incidence of red hair is about 1 in 100 people (in the Pacific Northwest).

 $X =$  number of redheads  $X \sim Poisson(\lambda = 0.01)$ 

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## Which Distribution?

There are 200 people with diabetes in a census block of 600 people. We randomly sample 50 people.

 $X =$  number sampled with diabetes

$$
X \sim HyperGeometric(N = 600, K = 200, n = 50)
$$

• Kids play 10 rounds of twister.

 $X_1$  = number spinner on green,  $X_2$  = number spinner on red,

 $X_3$  = number spinner on yellow,  $X_4$  = number spinner on blue,  $X \sim$  Multinomial( $p_1 = 1/4$ ,  $p_2 = 1/4$ ,  $p_3 = 1/4$ ,  $p_4 = 1/4$ ,  $n = 10$ )