- TA Office Hours on Course Page
- Homework adventur Continues (HW#1 is still with me ~)
- Homework #2 is in progress
  - Thursday morning office hours gone forever

# STAT 430: Lecture 8<sup>-</sup> More Discrete Random Variables

Continuing Chapter 2

Course page: imouzon.github.io/stat430

# Random Variables

Idea: ux defint a probability
on sets of outromes
for a given experiment

Problem: - The sample spaces

are context specific

- not much can be

learned from studying

a specific sample space

- Sample spaces had to

describe and possibly

hard to work with.

Whet Random variables do i)
connect the samples space to
the set of seel mumbers i.e. Random Variables or Conctions that han the sample space as their donair Since the actual outcome, com, is unknown, the the value that will result from a function of W Balsounhnown. We say that the probability on Q is induced on X.

We use capital letters from the end of the alphabet to to represent random variables (i.e., X, Y, Z)

We use lower case to represent specific values the random Variable can take

For discrete random veriables, we describe the probability on X using a "probability mass function"

pmf: p(x) means P(X=x)

Properties of ponts

1. p(x) ≥ 0 for all x

$$2. \sum_{n \mid x} \rho(x) = 1$$

Expectation and Variance

notation for Experted Value:

E(X) or M

other terms: "mean of the RV"
"expectation of the RV"

notation for Variance:

Var(X) or V(X) or o

Variance: flow spread out are the possible values of X.

What these two properties do is:

Expectation: "What is the center of the possible values of X if we take the probability into account"

$$\frac{2 \text{ groups}}{4}$$
 $\frac{2 \text{ groups}}{4}$ 
 $\frac{1}{4}$ 
 $\frac{1}{4}$ 

$$\frac{0.5 + 0 + 0.5}{3} = \frac{1}{3}$$

Notice: Mean = 
$$\frac{\text{Sum of all valves}}{\text{# of valves}} = \frac{1}{n} \cdot \text{Valve#} + \frac{1}{n} \cdot \text{Valve} # \Omega$$

Expected value =  $\sum_{\text{all } x} x \cdot p(x)$ .

What about spread?

First idea: absolute devistion

if M is my mean

then  $\sum_{a \in X} |x-M| \cdot p(x)$ 

Variance: 
$$\sum_{\text{all } x} (x-M)^2 p(x)$$

Ex: if X is a random variable with  $P(X=-2)=P(X=-1)=P(X=0)=P(X=1)=P(X=2)=\frac{1}{5}$ 

$$\frac{1}{1} = \frac{1}{5}(-2) + \frac{1}{5}(-1) + \frac{1}{5}(0) + \frac{1}{5}(-1) + \frac{1}{5}(0)$$

$$= -\frac{2}{3} - \frac{1}{3} + \frac{1$$

The absolute deviation of X from the mean:

$$\sum_{G||X} |X-M| \cdot \rho(X) = |-2-6| \cdot \frac{1}{5} + |-1-0| \cdot \frac{1}{5} + |0-0| \cdot \frac{1}{5} + |1-0| \cdot \frac{1}{5} + |2-6| \cdot \frac{1}{5}$$

$$= \frac{2}{5} + \frac{1}{5} + 0 \cdot \frac{1}{5} + \frac{1}{5} + \frac{2}{5}$$

$$= \frac{6}{5}$$

$$Var(X) = \sum_{G||X} (X-M)^{G} \rho(X) = \frac{1}{5} + (1-6)^{G} \cdot \frac{1}{5} + (1-6)^{G} \cdot \frac{1}{5} + (2-0)^{G} \cdot \frac{1}{5}$$

$$= \frac{1}{5} + \frac{1}{5}$$

### Binomial Random Variables

### A Sequence of Bernoulli Experiments

Suppose that we don't just have a single Bernoulli expirement, but instead have a sequence of *n* independent Bernoulli experiments and are interested in the *total* number of successful outcomes.

**Example** We roll a fair six-sided die 5 times and record a success if we observer a 6.

- What is the sample space for  $\Omega$ ?
- If we define  $X_1, X_2, \ldots, X_5$  as a Bernoulli random variables for each single experiment, what is the "sample space" related to the vector  $(X_1, X_2, \ldots, X_5)$ ?
- If we define *Y* as the total number of successful outcomes, what is the "sample space" for *Y*?
- How can we write the probability function for *Y*?

$$P(X_{1}=1,X_{2}=1,X_{3}=1,X_{4}=1,X_{5}=6)=P(X_{1}=1)\cdot P(X_{2}=1)\cdot P(X_{3}=1)\cdot P(X_{3}=1)\cdot P(X_{5}=6)$$

$$= \left(\frac{1}{6}\right)\left(\frac{1}{6}\right)\left(\frac{1}{6}\right)\left(\frac{1}{6}\right)\left(\frac{5}{6}\right)$$

By continuing, we get = 
$$(\frac{1}{6})^4 (\frac{5}{6})^1$$

$$P(Y=3) = \frac{5!}{3! \, a!} \left(\frac{1}{6}\right)^3 \left(\frac{3}{6}\right)^2$$

$$P(Y=2) = \frac{5!}{2!3!} (-\frac{1}{6})^2 (-\frac{5}{6})^3$$

$$P(Y=0) = \frac{5!}{6! 5!} (\frac{1}{6})^0 (\frac{5}{6})^5$$

### Binomial Random Variables

### **Binomial Random Variables**

#### def: Binomial Random Variable

Suppose that n experiments, or trials, are performed and that

- 1. The trials have two possible outcomes ("success" and "failure")
- 2. Each trial has the same probability of success, *p*, and
- 3. The trials are independent of each other. Then the random variable defined as the sum of the number of successful experiments is a Binomial Random Variable and we say that it follows a Binomial Distribution

### **Binomial Random Variables**

## Binomial Random

*Motivation:* 

If I try this experiment n times, how do I figure out how many outcomes will be what I want?

# Random Statement: Variables Lat V has the

Let X be the number of successful outcomes observed by repeating n independent Bernoulli experiments, each with probability of success p.

Notation: 
$$X \sim \text{Binom}(n, p)$$

$$\left(\bigcap_{k}\right) = \frac{\bigcap_{k!(n-k)!}^{!}}{k!(n-k)!}$$

pmf:

The probability of seeing k successful outcomes is given using p robability of  $f_{qi}$  or  $f_{qi}$ 

$$p(k) = \binom{n}{k} p^k (1 - p)^{n-k}, k = 0, 1, 2, \dots, n$$

### Binomial Random Variables

### **Binomial Random Variables**

Cumulative Probability:

$$P(X \le k) = p(0) + p(1) + \dots + p(k), k = 0, 1, 2, \dots, n$$

Expectation:

$$\mu = n \cdot p$$

Variance:

$$\sigma^2 = n \cdot p \cdot (1 - p)$$

Geometric Random Variables

### Binomial Random Variables

### Geometric Random Variables

### **Geometric Random Variables**

#### Motivation:

How many times do I have to repeat this experiment until the outcome is what I want it to be?

#### Statement:

Let X be the trial upon which the first successful outcome is observed in a sequence of independent Bernoulli experiments, each with probability of success p.

*Notation:* 

$$X \sim \text{Geometric}(p)$$

pmf:

The probability of seeing the first successful outcome on trial k

$$p(k) = p(1-p)^{k-1}, k = 1, 2, ...$$

### **Geometric Random Variables**

Cumulative Probability:

$$P(X \le k) = p(1) + p(2) + \dots + p(k), k = 1, 2, \dots$$

Binomial Random Variables

Expectation:

$$\mu = \frac{1}{p}$$

Geometric Random Variables

Variance:

$$\sigma^2 = \frac{1 - p}{p^2}$$

**Geometric Random Variables** 

### **Geometric Random Variables**

*Motivation:* 

Binomial Random Variables What if I have an almost infinite number of independent Bernoulli trials and a very small probability of success?

Notation:

 $X \sim \text{Poisson}(p)$ 

Geometric Random Variables

pmf:

The probability of seeing k successes is:

Poisson Random Variables

$$p(k) = \frac{\lambda^k}{k!} e^{-\lambda}$$

### **Geometric Random Variables**

Cumulative Probability:

$$P(X \le k) = p(0) + p(1) + p(2) + \dots + p(k), k = 0, 1, 2, \dots$$

Binomial Random Variables

Expectation:

$$\mu = \lambda$$

Variance:

Geometric Random Variables

$$\sigma^2 = \lambda$$

Poisson Random Variables