Lecture 6

Bernoulli and Binomial Distributions

STAT 330 - Iowa State University

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Discrete Distributions

Common distributions for discrete random variables

• Bernoulli distribution

$$X \sim Bern(p)$$

• Binomial distribution

$$X \sim Bin(n, p)$$

• Geometric distribution

$$X \sim Geo(p)$$

• Poisson distribution

$$X \sim Pois(\lambda)$$

We will also discuss *joint distributions* for 2 or more discrete random variables

Bernoulli Distribution

Bernoulli Distribution

Bernoulli Experiment: Random experiment with only 2 outcomes:

- Success (S)
- Failure (F)

where P(Success) = P(S) = p for $p \in [0, 1]$

Example 1: (Bernoulli experiments):

- 1. Flip a coin: S = heads, F = tails
- 2. Watch stock prices: S = increase, F = decrease
- 3. Cancer screening: S = cancer, F = no cancer

Working with Bernoulli Random Variable

Suppose we have a Bernoulli experiment (only 2 outcomes: "success", "failure").

We obtain the outcome "success" with probability p

When random variable X follows a Bernoulli Distribution, we write

$$X \sim Bern(p)$$

• Define a random variable X

$$X = \begin{cases} 1 & \text{Success (S)} \\ 0 & \text{Failure (F)} \end{cases}$$

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Bernoulli Random Variable Cont.

- Probability Mass Function (pmf)
 - 1. $Im(X) = \{0, 1\}$
 - 2. P(X = 1) = P(S) = pP(X = 0) = P(F) = 1 - p

The pmf can be written in tabular form:

$$\begin{array}{c|cccc} x & 0 & 1 \\ \hline p_X(x) & 1-p & p \end{array}$$

The pmf can be written as a function:

$$p_X(x) = \begin{cases} p^x (1-p)^{1-x} & x \in \{0,1\} \\ 0 & \text{otherwise} \end{cases}$$

Typically, we use the above functional form to describe the *probability mass function (pmf)* of Bernoulli random variable.

Bernoulli Random Variable Cont.

• Cumulative distribution function (cdf)

$$F_X(t) = P(X \le t) = \left\{ egin{array}{ll} 0 & t < 0 \ 1 - p & 0 \le t < 1 \ 1 & t \ge 1 \end{array}
ight.$$

• Expected Value: E(X) = p

$$E(X) = \sum_{x \in \{0,1\}} xP(X = x) = 0(1 - p) + 1(p) = p$$

• Variance: Var(X) = p(1-p)

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Binomial Distribution

Binomial Distribution

Set up: Conduct multiple trials of *identical* and *independent* Bernoulli experiments

- Each trial is independent of the other trials
- P(Success) = p for each trial

We are interested in the number of success after n trials. The random variable X is

$$X =$$
 " # of successes in n trials"

This random variable X follows a Binomial Distribution

$$X \sim Bin(n, p)$$

where n is the number of trials, and p is the probability of success for each trial.

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Binomial Distribution Cont.

Example 2: Flip a coin 10 times, and record the number of heads.

Success = "heads";
$$P(Success) = p = 0.5$$

Define the random variable X

$$X =$$
 " # of heads in $n = 10$ trials"

• The distribution of *X* is . . .

$$X \sim Bin(10, 0.5)$$

Derivation of Binomial pmf

• Probability Mass Function (pmf)

1.
$$Im(X) = \{0, 1, 2, 3, 4, \dots, n\}$$

2.
$$P(X = x) = ?$$

Recall
$$P(Success) = P(S) = p$$
, $P(Failure) = P(F) = 1 - p$

Case:
$$X = 0$$
 \underline{F} \underline{F} $\underline{F} \cdots \underline{F}$

$$P(X=0)=(1-p)^n$$

Case:
$$X = 1$$

$$P(X = 1) = \binom{n}{1} p^{1} (1 - p)^{n-1}$$

Case:
$$X=2$$

$$P(X = 2) = \binom{n}{2} p^2 (1 - p)^{n-2}$$

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Binomial Random Variables

In general, the *probability mass function (pmf)* of a Binomial R.V can be written as:

$$p_X(x) = \begin{cases} \binom{n}{x} p^x (1-p)^{n-x} & x = 0, 1, 2, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

• Cumulative distribution function (cdf)

$$F_X(t) = P(X \le t) = \sum_{x=0}^{\lfloor t \rfloor} \binom{n}{x} p^x (1-p)^{n-x}$$

(Add up the pmfs to obtain the cdf)

• Expected Value: E(X) = np

• Variance: Var(X) = np(1-p)

IID Random Variables

Properties of IID Random Variables

Independent and identically distributed (iid) random variables have properties that simplify calculations

Suppose Y_1, \ldots, Y_n are iid random variables

- Since they are *identically* distributed,
 - $E(Y_1) = E(Y_2) = \dots = E(Y_n)$ $\rightarrow E(\sum Y_i) = \sum E(Y_i) = nE(Y_1)$
 - $Var(Y_1) = Var(Y_2) = \ldots = Var(Y_n)$
- Since they are also independent,
 - ullet o $Var(\sum Y_i) = \sum Var(Y_i) = nVar(Y_1)$

Working with IID Random Variables

A Binomial random variable, X, is the sum of n independent and identically distributed (iid) Bernoulli random variables, Y_i .

Let Y_i be a sequence of iid Bernoulli R.V. For $i=1,\ldots,n$,

$$Y_i \stackrel{iid}{\sim} Bern(p)$$

with $E(Y_i) = p$ and $Var(Y_i) = p(1 - p)$. Then,

$$X = \sum_{i=1}^{n} Y_i \sim Bin(n, p)$$

Then, we obtain E(X) and Var(X) using properties of iid R.V.s

$$E(X) = nE(Y_1) = np$$

$$Var(X) = nVar(Y_1) = np(1 - p)$$

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Examples

Binomial Distribution Examples

Example 3: A box contains 15 components that each have a defective rate of 5%. What is the probability that . . .

- 1. exactly 2 out of 15 components are defective?
- 2. at most 2 components are defective?
- 3. more than 3 components are defective?
- 4. more than 1 but less than 4 components are defective?

How to approach solving these types of problems?

- 1. Define the random variable
- 2. Determine the R.V's distribution (and values for the parameters)
- 3. Use appropriate pmf/cdf/E(X)/Var(X) formulas to solve

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Binomial Distribution Examples Cont.

<u>Define the R.V:</u> X = # defective out of n = 15 components <u>State the Distribution of X:</u> $X \sim Bin(15, 0.05)$

$$n = 15, p = 0.05$$

1. What is the probability that exactly 2 out of 15 components are defective?

$$P(X = 2) =$$

Binomial Distribution Examples Cont.

2. What is the probability that at most 2 components are defective?

$$P(X \leq 2) =$$

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How to Use Binomial CDF Table (Appendix A)

Suppose we have random variable $X \sim Bin(n = 15, p = 0.05)$.

$$P(X \le 2) = ?$$

- Find the n = 15 sub-table
- $P(X \le 2)$ is found inside the table corresponding to p = 0.05 (column) and x = 2 (row).

		I		$P(X \le 2) = 0$	$P(X \le 2) = 0.9637998$		
n	1 = 15	p=0.01	0.05	0.1	0.15	1/6	
	x=0	0.8600584	0.4632912	2 0.2058911	0.08735422	0.06490547	
	1	0.9903702	0.8290475	0.5490430	0.31858598	0.25962189	
	2	0.9995842	0.9637998	0.8159389	0.60422520	0.53222487	
	3	0.9999875	0.9945327	0.9444444	0.82265520	0.76848078	
	4	0.9999997	0.9993853	0.9872795	0.93829461	0.91023433	
	5	1.0000000	0.9999472	0.9977503	0.98318991	0.97260589	
	6	1.0000000	0.9999965	0.9996894	0.99639441	0.99339642	
	—	1 0000000	0 000000	0 0000004	0.00000000	0 00054055	

Binomial Distribution Examples Cont.

3. What is the probability that more than 3 components are defective?

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Binomial Distribution Examples Cont.

4. What is the probability that more than 1 but less than 4 components are defective?