

Lecture 11

Bernoulli and Binomial Distributions

Text: Chapter 4

STAT 8010 Statistical Methods I
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Agenda

Bernoulli and
Binomial
Distributions



Bernoulli Trials

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1 Bernoulli Trials

2 Bernoulli and Binomial Distributions

Bernoulli Trials

Many problems in probability and its applications involve **independently** repeating a random experiment and observing at each repetition whether a specified event occurs. We label the occurrence of the specified event a **success** and the nonoccurrence of the specified event a **failure**.

Example:

Tossing a coin several times



Bernoulli trials:

- Each repetition of the random experiment is called a trial
- We use p to denote the probability of a success on a single trial

Properties of Bernoulli trials:

- Exactly two possible outcomes **success** and **failure**
- The outcomes of trials are **independent** of one another
- The success probability, p , and therefore the failure probability, $(1 - p)$, remains the same from trial to trial

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- The variance:

$$\text{Var}(X) = \mathbb{E}[X^2] - (\mathbb{E}[X])^2 = p - (p)^2 = p(1-p)$$

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- The variance:

$$\text{Var}(X) = np(1-p)$$

Example

To test for Extrasensory perception (ESP), we have 4 cards. They will be shuffled and one randomly selected each time, and you are to guess which card is selected. This is repeated 10 times. Suppose you do not have ESP. Let R be the number of times you guess a card correctly. What are the distribution and parameter(s) of R ? What is the expected value of R ? Furthermore, suppose that you get certified as having ESP if you score at least an 8 on the test. What is the probability that you get certified as having ESP?

Example

In Whitney's Stat 8020 class, 80% of the students passed (got a C or better) on Exam 1. If you were to pick a student at random and asked them whether or not they passed. Let X represent the number of student who passed.

- 1 What type of random variable is this? How do you know? Additionally, write down the pmf, the expected value, and the variance for X
- 2 What about if you picked 10 students with replacement and let Y be the number of student(s) who passed. What type of random variable is this? Write down the pmf, the expected value, and the variance for Y

Example

Suppose that 95% of consumers can recognize Coke in a blind taste test. Assume consumers are independent of one another. The company randomly selects 4 consumers for a taste test. Let X be the number of consumers who recognize Coke.

- 1 Write out the pmf table for X
- 2 What is the probability that X is at least 1?
- 3 What is the probability that X is at most 3?

The binomial distribution describes the probability of k successes in n trials **with replacement**.

We want a distribution to describe the probability of k successes in n trials **without replacement** from a finite population of size N containing exactly K successes.

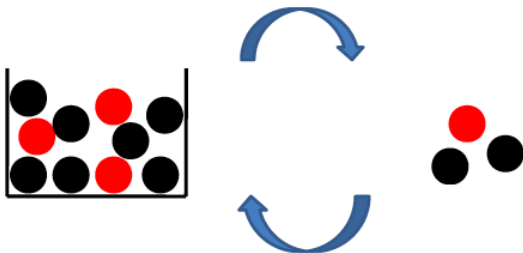
⇒ **Hypergeometric Distribution**

Important applications are **quality control** and statistical **estimation of population proportions**. The hypergeometric r.v. is the equivalent of a Binomial r.v. except that sampling is done without replacement.

An Example of Hypergeometric r.v.

Probability:

What is the probability to get 1 red and 2 black balls?



Statistics:

What percentage of balls in the box are red?

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- The expected value: $\mathbb{E}[X] = n \frac{K}{N}$
- The variance: $\text{Var}(X) = n \frac{K}{N} \frac{N-K}{N} \frac{N-n}{N-1}$

Example

There are 100 identical looking 52" TVs at Best Buy in Anderson, SC. Let 10 of them be defective. Suppose we want to buy 8 of the aforementioned TVs (at random). What is the probability that we don't get any defective TVs?

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Let D be the number of defective TVs in the sample.

$$D \sim \text{Hyp}(N = 100, n = 8, K = 10)$$

$$\mathbb{P}(D = 0) = \frac{\binom{10}{0} \binom{90}{8}}{\binom{100}{8}} = 0.4166$$