# Lecture 7

### The Five Basic Discrete Random Variables

- Binomial
- 2 Hypergeometric
- 3 Geometric
- Megative Binomial
- 5 Poisson

#### Remark

On the handout "The basic probability distributions" there are six distributions. I did not list the Bernoulli distribution above because it is too simple.

In this lecture we will do 1, and 2, above.

### The Binomial Distribution

Suppose we have a Bernoulli experiment with P(S) = P, for example, a weighted coin with P(H) = P. As usual we put q = 1 - p. Repeat the experiment (flip the coin). Let  $X = \sharp$  of successes ( $\sharp$  of heads). We want to compute the probability distribution of X. Note, we did the special case n = 3 in Lecture 6, pages 4 and 5.

Clearly the set of possible values for X is  $0, 1, 2, 3, \ldots, n$ . Also

$$P(X = 0) = P(TT \ T) = qq \dots q = q$$

# Explanation

Here we assume the outcomes of each of the repeated experiments are independent so

$$P((T \text{ on } 1^{\text{st}}) \cap (T \text{ on } 2^{\text{nd}}) \cap \dots \cap (T \text{ on } n\text{-th})$$

$$P(T \text{ on } 1^{\text{st}})P(T \text{ on } 2^{\text{rd}})\dots P(T \text{ on } n\text{-th})$$

$$q q \dots q = q^n$$

Note T on  $2^{nd}$  mean) T on  $2^{nd}$  with no other information so

$$P(T \text{ on } 2^{\text{nd}}) = q.$$

Also

$$P(X = n) = P(HH ... H) = P^n$$

Now we have to work

What is P(X = 1)?

### Another standard mistake

The events (X = 1) and  $\underbrace{HTT \dots T}_{n-1}$  are NOT equal.

# Why - the head doesn't have to come on the first toss

So in fact

$$(X = 1) = HTT \dots T \cup THT \dots T \cup \dots \cup TTT \dots TH$$

All of the n events on the right have the same probability namely  $pq^{n-1}$  and they are mutually exclusive. There are n of them so

$$P(X=1) = npq^{n-1}$$

Similarly

$$P(X = n - 1) = npq^{n-1}$$

(exchange *H* and *T* above)

# The general formula

Now we want P(X = k)First we note

$$P(\underbrace{H \dots H}_{k} \underbrace{TT \dots T}) = p^{k} q^{n-k}$$

But again the heads don't have to come first. So we need to

- (1) Count all the words of length n in H and T that involve k. It's and n k T's.
- (2) Multiply the number in (1) by  $p^k q^{n-k}$ .

So how do we solve 1. Think of filling n slot's with the H's and n - k T's

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### Main Point

Once you decide where the kH's go you have no choice with the T's. They have to go in the remaining n-k slots.

So choose the k-slots when the heads go. So we here to make a choose of k things from n things so  $\binom{n}{k}$ .

So,

$$P(X=k) = \binom{n}{k} P^k q^{n-k}$$

So we have motivated the following definition.

#### Definition

A discrete random variable X is said to have binomial distribution with parameters n and p (abbreviated  $X \sim Bin(n, p)$ )

If X takes values 0, 1, 2, ..., n and

$$P(X=k) = \binom{n}{k} p^k q^{n-k}, 0 \le k \le n.$$
 (\*)

### Remark

The text uses x instead of k for the independent (i.e., input) variable. So this would be written

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

I like to save x for the case of continuous random variables.

Finally we may write

$$P(k) = \binom{n}{k} p^k q^{n-k}, \quad 0 \le k \le n$$
 (\*\*)

The text uses  $b(\cdot, n, p)$  for  $p(\cdot)$  so would write for (\*\*)

$$b(k,n,p) = \binom{n}{k} p^k q^{n-k}$$

# The Expected Value and Variance of a Binomial Random Variable

### Proposition

Suppose  $X \sim \text{Bin}(n, p)$ . Then E(X) = np and V(X) = npq so  $\sigma = \text{standard}$  deviation  $= \sqrt{npq}$ .

### Remark

The formula for E(X) is what you might expect. If you toss a fair coin 100 times the E(X) = expected number of heads  $np = (100) \left(\frac{1}{2}\right) = 50$ .

However if you toss it 51 times then  $E(X) = \frac{51}{2}$  - not what you "expect".

# Using the binomial tables

Table A1 in the text pg. 664-666 tabulates the cdf B(x, n, p) for n = 5, 10, 15, 20, 25 and selected values of p.

# Example (3.32)

Suppose that 20% of all copies of a particular text book fail a certain binding strength text. Let X denote the number among 15 randomly selected copies that fail the test. Find

$$P(4 \le X \le 7)$$
.

#### Solution

 $X \sim \text{Bin}(15, .2)$ . We wont to compute  $P(4 \le X \le 7)$  using the table on page 664. So how to we write  $P(4 = \text{leq}X \le 7)$  in terms of terms of the form  $P(X \le a)$ 



#### Answer

$$(\sharp)P(4 \le X \le 7) = P(X \le 7) - P(X \le 3)$$

So

$$P(4 \le X \le 7) = B(7, .15, .2) - B(3, .15, .2)$$

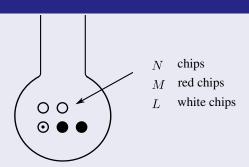
from table

$$= .996 - .648$$

**N.B.** Understand  $(\sharp)$ . This the key using computers and statistical calculators to compute.

# The hypergeometric distribution

### Example



Consider an urn containing N

Chips of which M are red and L = N - M are white. Suppose we remove n chips without replacement so  $n \le N$ .

Define a random variable X by  $X = \sharp$  of red chips we get.

Find the probability distribution of *X*.

### Proposition

$$P(X=k) = \frac{\binom{M}{k}\binom{L}{n-k}}{\binom{N}{n}} \tag{*}$$

if

(b) 
$$\max(0, n-L) \le k \ne \min(n, M)$$

This means  $k \le both n$  and M and both 0 and  $n - L \le k$ . These are the possible values of k, that is, if k doesn't satisfy (b) then

$$P(X=k)=0.$$

# Proof of the formula (\*)

Suppose we first consider the special case where all the chips are red so

$$P(X = n)$$
.

This is the same problem as the one of finding all hearts in bridge.

 $\begin{tabular}{ll} \begin{tabular}{ll} \be$ 

So we use the principle of restricted choise

$$P(X=n) = \frac{\binom{M}{n}}{\binom{N}{n}}$$

This agrees with (\*).

But (\*) is harder because we have to consider the case where there are kMn red chip. So we have to choose n - k white chips as well.

So choose k red chips  $-\binom{M}{k}$  ways, then for each such choice, choose n-k white chipts  $\binom{L}{n-k}$  ways.

$$\sharp \left\{ \begin{array}{l} \text{choices of } \underline{\text{exceptly}} \\ k \text{ red chips} \\ \text{in the } n \text{ chips} \end{array} \right\} = \binom{M}{k} \binom{L}{n-k}$$

Clearly there are  $\binom{N}{n}$  ways of choosing n chips from N chips so (\*) follows.

### Definition

If X is a discrete random variable with pmf defined by page 14 then X is said to have hyper geometric distribution with parameters n, M, N. In the text the pmf is denoted

h(x; n, M, N).

#### What about the conditions

$$\max(0, n - L) \le k \le \min(n, M) \tag{b}$$

This really means

$$k \le \text{ both } n \text{ and } M$$
 (b<sub>1</sub>)

and

both 0 and 
$$n - L \le k$$
 (b<sub>2</sub>)

(b<sub>1</sub>) says

 $k \le n \longleftrightarrow \text{we can't choose more then } n$ red chips because we are only choosing n chips in to to

 $k \le M \longleftrightarrow \text{because there are only } M \text{ red}$  chips to choose from

 $(b_2)$  k > 0 is obvious.

So the above three inequalities are necessary. At first glance they look sufficient because if k satisfies the above three inequalities you can certainly go ahead and choose k red chips.

But what about the white chips? We aren't tone yet, you have to choose n - k white chips and there are only L white chips available so if n - k > L we are sun k.

So we must have

$$n - k \le L \Leftrightarrow k \ge n - L$$

This is the second inequality of  $(b_2)$ . If it is satisfied we can go ahead and choose the n-b white chips to the inequalities in (b) are necessary and sufficient.

### Proposition

Suppose X has hypergeometric distribution with parameters n, M, N. Then

(i) 
$$E(X) = n\frac{M}{N}$$

(ii) 
$$V(X) = \left(\frac{N-n}{N-1}\right)n\frac{M}{N}\left(1-\frac{M}{N}\right)$$

If you put

$$P_1 = \frac{M}{N} = \frac{\text{the probability of getting}}{\text{a red disk on the first draw}}$$

then we may rewrite the above formulas as

$$E(X) = nP$$

$$V(X) = \left(\frac{N-n}{N-1}\right) npq$$

$$\begin{cases}
reminiscent \\
of the \\
binomial \\
distribution
\end{cases}$$

# Another way to Derive (\*)

There is another way to derive (\*) - the way we derived the binomial distribution. It is way harder.

### Example

Take n=2

$$P(X = 0) = \frac{L}{N} \frac{L-1}{N+1}$$

$$P(X = 2) = \frac{M}{N} \frac{M-1}{N-1}$$

$$P(X = 1) = P(RW) + P(WR)$$

$$= \frac{M}{N} \frac{L}{N-1} + \frac{L}{N} \frac{M}{N-1}$$

$$= 2\frac{M}{N} \frac{L}{N-1}$$

$$= 2\frac{M}{N} \frac{L}{N-1}$$

In general, we claim that all the words with kR's and n-k W's have the some probability. Indeed each of these probabilities are fractions with the same denominator

$$N(N-1)...(N-n-1)$$

and they have the same factors in the numerator scrambled up

$$M(M-1)(M-L+1)$$
 and  $L(L-1),...,(L-n-k+i)$ 

But the order of the factors doesn't matter so

$$P(X = k) = \binom{n}{k} P(\overline{R \dots R} W \dots W)$$
$$= \binom{n}{k} \frac{M(M-1) \dots (M-k+1)L(L-1) \dots (L-n-k+1)}{N(N-1) \dots N(-n+1)}$$

Why is (\*) equal to this?

$$(*) = \frac{\binom{M}{k}\binom{L}{n-k}}{\binom{N}{n}} \underbrace{\text{cancelling}}_{(M} (M-1)...(L-n-k+1)}$$

$$= \frac{M(M-1)...(M-k+1)}{\binom{k!}{k!}} \underbrace{\frac{L(L-1)...(L-n-k+1)}{(n-k)!}}_{n!} \underbrace{\frac{N(N-1)...(N-n+1)}{n!}}_{\text{goes on top}}$$

$$(*) = \frac{\binom{M}{k} \binom{L}{n-k}}{\binom{N}{n}}$$

$$= \frac{\frac{M(M-1)...(M-k+1)}{k!} \frac{L(L-1)...(L-n-k+1)}{(n-k)!}}{\frac{N(N-1)...(N-n+1)}{n!}}$$

exercise in fractions

$$= \frac{n!}{k!(n-k)!} \frac{M(M-1)\dots(M-k+1)L(L-1)\dots(L-n-k+1)}{N(N-1)\dots(N-n+1)}$$

$$= \binom{n}{k} \frac{M(M-1)\dots(M-k+1)L(L-1)\dots(L-n-k+1)}{N(N-1)\dots(N-n+1)}$$

Obviously, the first way (\*) is easier so if you are doing a reel-world problem and you start getting things that look like (\*\*) step back and see if you can use the first method instead. You will tend to try the second method first. I will test you on this later.

<u>Prediction</u> (I was wrong before)

Most of you will use the second (wrong) method.

# An Important General Problem

Suppose you draw n chips with replacement and let X be the number of red chips you get. What distribution does X have?

This explains (a little) the formulas on page 21. Note that if N is far bigger than n then it is almost like drawing with replacement. "The urn doesn't notice that any chaps have been removed because so few (relatively) have been removed."

In this case

$$\frac{N-n}{N-1} = \frac{N\left(1-\frac{n}{N}\right)}{N\left(1-\frac{1}{N}\right)} \approx \frac{N}{N} = 1$$

(because N is huge  $\frac{1}{N}$  and  $\frac{n}{N} = 0$ )

So  $V(X) \approx npq$  (see the bottom at pg. 21)

This is what is going on in page 118 of the text.

The number  $\frac{N-n}{N-1}$  is called the "finite population correction factor".