

Lecture 7

The Five Basic Discrete Random Variables

1. Binomial
2. Hypergeometric
3. Geometric
4. Negative 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 2

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 $n = \#$ of trials.
 $X = \#$ of successes ($\#$ 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 $0, 1, 2, 3, \dots, n$.

Also the

$$P(X=0) = P(TT \dots T) = 2 \cdot 2 \cdot 2 \dots 2 = 2^n$$

Explanation

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

$$\begin{aligned} &P((T \text{ on } 1^{\text{st}}) \cap (T \text{ on } 2^{\text{nd}}) \cap \dots \cap (T \text{ on } n\text{-th})) \\ &= P(T \text{ on } 1^{\text{st}}) P(T \text{ on } 2^{\text{nd}}) \dots P(T \text{ on } n\text{-th}) \\ &= 2 \cdot 2 \dots 2 = 2^n \end{aligned}$$

Note $T \text{ on } 2^{\text{nd}}$ means $T \text{ on } 2^{\text{nd}}$ with no other information so

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

Also

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$$P(X=n) = P(HH \dots H) = p^n$$

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

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

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

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

Similarly

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

(exchange H and T above)

The general formula

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Now we want $P(X=k)$

First we note

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

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 H '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 slots
with k H's and $n-k$ T's



Main Point Once you decide
where the k H'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
where the heads go. So we
have to make a choice 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 $X \sim \text{Bin}(n, p)$) if X takes values $0, 1, 2, \dots, n$

and

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

Remark

The text uses x instead of k for the independent (ie 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

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

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

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

The Expected Value and Variance 10 of a Binomial Random Variable

Proposition

Suppose $X \sim \text{Bin}(n, p)$. Then

$$E(X) = np \text{ and } V(X) = npq$$

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 $E(X) = \text{expected number of heads}$ $np = (100)\left(\frac{1}{2}\right) = 50$

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 test.

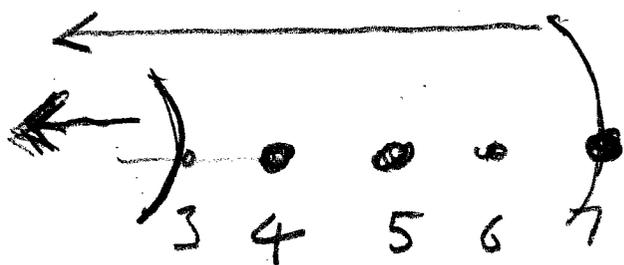
Let X denote the number among 15 randomly selected copies that fail the test. Find

$$P(4 \leq X \leq 7)$$

Solution

$X \sim \text{Bin}(15, .2)$. We want to compute $P(4 \leq X \leq 7)$ using the table on page 664.

So how do we write $P(4 \leq X \leq 7)$ in terms of terms of the form $P(X \leq a)$



Answer (#) $P(4 \leq X \leq 7) = P(X \leq 7) - P(X \leq 3)$

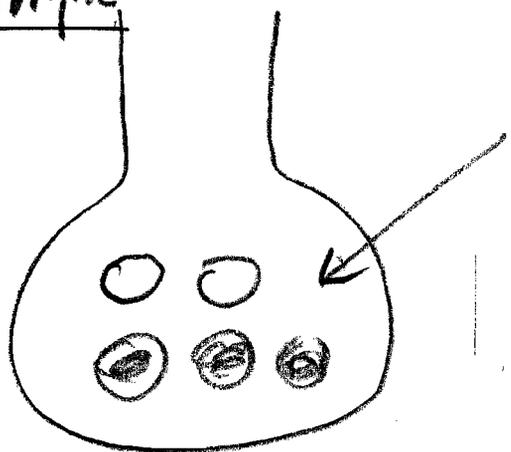
$$\begin{aligned}
 \text{So } P(4 \leq X \leq 7) &= B(7; 15, .2) - B(3; 15, .2) \\
 &\quad \text{from table} \\
 &= .996 - .648 \\
 &= .348
 \end{aligned}$$

N.B. Understand (#)! This is the key using computers and statistical calculators to compute.

2. The hypergeometric distribution

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Example



N chips

M red chips

L white chips

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 $n \leq N$.

Define a random variable

X by $X = \#$ of red chips we get

Find the probability distribution of X .

Proposition

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

if

$$(b) \quad \underbrace{\max(0, n-L) \leq k \leq \min(n, M)}$$

This means $k \leq$ both n and M
and

both 0 and $n-L \leq k$.

These are the possible values of k ,
that is, if k doesn't satisfy (b) then

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

Proof of the formula (*) 15

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

red chip \leftrightarrow heart

white chip \leftrightarrow non heart

So we use the principle of restricted choice

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

This agrees with (*)

But (*) is harder

because we have to consider
the case where there are

$k < n$ red chips. 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 chips $\binom{L}{n-k}$ ways.

So

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

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 19 then X is said

to have hypergeometric distribution with parameters n, M, N .

In the text the pmf is denoted

$h(x; n, M, N)$.

What about the conditions

$$\max(0, n-L) \leq k \leq \min(n, M) \quad (b)$$

This really means

$$k \leq \text{both } n \text{ and } M \quad (b_1)$$

and

$$\text{both } 0 \text{ and } n-L \leq k \quad (b_2)$$

(b₁) says

$$k \leq n \quad \longleftrightarrow$$

we can't choose more than n red chips because we are only choosing n chips in total

$$k \leq M \quad \longleftrightarrow$$

because there are only M red chips to choose from

(b₂)

$$k \geq 0 \quad \text{is obvious.}$$

So the above three inequalities 19 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 done yet, you have to choose $n-k$ white chips and there are only L white chips available so if $n-k > L$ we are sunk so we must have

$$n-k \leq L \iff k \geq n-L$$

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

Proposition

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

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

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

If you put

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

then we may rewrite the above formulas as

$$\left. \begin{aligned} E(X) &= np \\ V(X) &= \left(\frac{N-n}{N-1} \right) npq \end{aligned} \right\} \begin{array}{l} \text{reminiscent} \\ \text{of the} \\ \text{binomial} \\ \text{distribution} \end{array}$$

Another Way to Derive (*)

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

Example Take $n=2$

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

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

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

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

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

In general, we claim that all the words with k R's and $n-k$ W's have the same probability. Indeed each of these probabilities are fractions with the same denominator $N(N-1)\dots(N-n+1)$

and they have the same factors in the numerator scrambled up $M(M-1)\dots(M-k+1)$ and $L(L-1)\dots(L-n-k+1)$. But the order of the factors doesn't matter so

$$P(X=k) = \binom{n}{k} P(\underbrace{R\dots R}_k W\dots W)$$

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

Why is (*) equal to this? 23

$$\begin{aligned}
 (*) &= \frac{\binom{M}{k} \binom{L}{n-k}}{\binom{N}{n}} \\
 &= \frac{M(M-1)\dots(M-k+1)}{k!} \cdot \frac{L(L-1)\dots(L-n-k+1)}{(n-k)!} \\
 &= \frac{N(N-1)\dots(N-n+1)}{n!}
 \end{aligned}$$

Annotations:

- Arrows from $(M-k)!$ and $(L-(n-k))!$ point to the $k!$ and $(n-k)!$ terms respectively, with the label "cancelling".
- An arrow from "goes on top" points to the $n!$ term in the denominator of the final fraction.

exercise in fractions

$$\begin{aligned}
 &= \frac{n!}{k! (n-k)!} \cdot \frac{M(M-1)\dots(M-k+1) L(L-1)\dots(L-n-k+1)}{N(N-1)\dots(N-n+1)} \\
 &= \binom{n}{k} \frac{M(M-1)\dots(M-k+1) L(L-1)\dots(L-n-k+1)}{N(N-1)\dots(N-n+1)}
 \end{aligned}$$

Obviously, the first way (*) is easier so if you are doing a real-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.

Prediction (I was wrong before)

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

An Important General Problem

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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 chips
have been removed because so few
(relatively) have been removed."

In this case T

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

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

So $V(X) \approx npq$ (see the
bottom of 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".