

Math 3501 - Probability and Statistics I

2.6 - The negative binomial distribution

Negative binomial distribution

Observe a sequence of independent Bernoulli trials until exactly r successes occur, where r is a fixed positive integer.

Let X denote the number of trials needed to observe the r th success.

The pmf of X is

$$g(x) = \binom{x-1}{r-1} p^r (1-p)^{x-r}, \quad x = r, r+1, \dots$$

we need to repeat the experiment at least r times to observe r successes

\uparrow \uparrow \uparrow

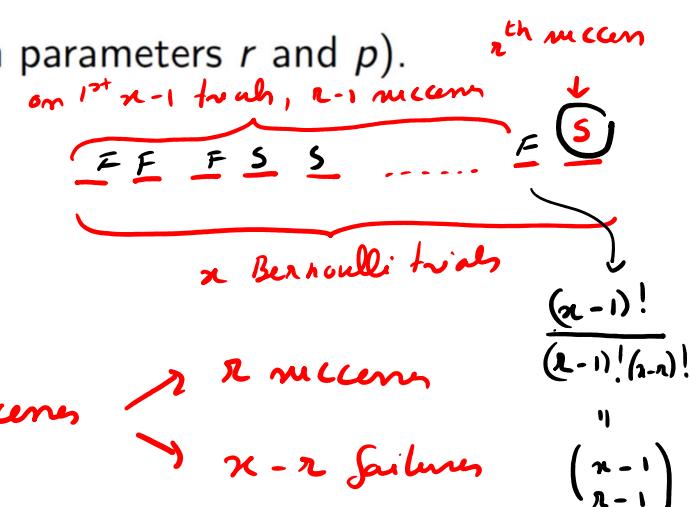
r success x-r failures

and we say that X has a *negative binomial distribution* (with parameters r and p).

ways in which $r-1$ successes
may occur among $x-1$ trials

$$g(x) = P(X=x) = \binom{x-1}{r-1} p^r \cdot (1-p)^{x-r}$$

we need x repetitions to observe r successes



Calculus II

Recall: Recall that the Binomial series, i.e., the Maclaurin's series expansion for the function

$$h(w) = (1 - w)^{-r} = \frac{1}{(1 - w)^r}, \quad r \in \mathbb{R},$$

is given by

$$(1 - w)^{-r} = \sum_{k=0}^{\infty} \frac{h^{(k)}(0)}{k!} w^k = \sum_{k=0}^{\infty} \binom{r+k-1}{r-1} w^k, \quad |w| < 1,$$

where

$$\binom{r+k-1}{r-1} = \frac{(r+k-1)(r+k-2)(r+k-3)\cdots r}{k!}$$

are (generalized) binomial coefficients.

Setting $x = k + r$ (or, equivalently, $k = x - r$), we obtain

new general term is
very similar to the
negative binomial pmf

$$(1 - w)^{-r} = \sum_{x=r}^{\infty} \binom{x-1}{r-1} w^{x-r}, \quad |w| < 1.$$

alternative representation for
Binomial series

generalizes the binomial expansion
 $(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$
 n positive integer $n+1$ terms

Consequence: the pmf of the negative binomial distribution satisfies

$$\begin{aligned}
 \sum_{x=r}^{\infty} g(x) &= \sum_{x=r}^{\infty} \binom{x-1}{r-1} p^r q^{x-r} \\
 &= p^r \sum_{x=r}^{\infty} \binom{x-1}{r-1} q^{x-r} = p^r (1-q)^{-r} = \frac{p^r}{p^r \cdot p^{-r}} = 1
 \end{aligned}$$

*Binomial series from
previous slide with
n replaced by r*

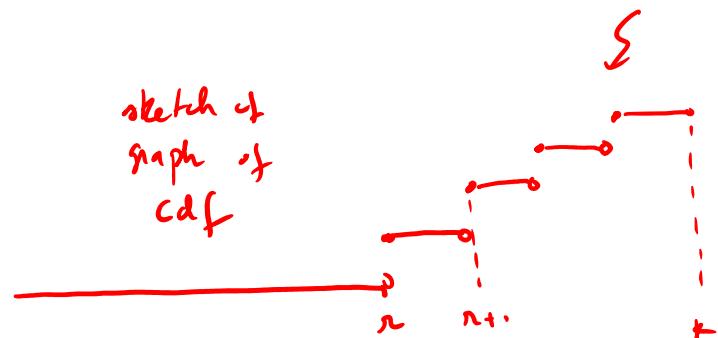
now equals $(1-q)^{-r}$

Note that
 $g(x) = \binom{x-1}{r-1} p^r (1-p)^{x-r} > 0$
 for all $x=r, r+1, \dots$

$$q = 1 - p$$

$$1 - q = p$$

sketch of
graph of
cdf



$g(x)$ is indeed a pmf !!

Geometric distribution → Special case of negative binomial distribution with $r = 1$ } 1st success

If $r = 1$ in the negative binomial distribution, we say that X has a *geometric distribution*, with pmf given by

$$g(x) = p(1-p)^{x-1}, \quad x = 1, 2, 3, \dots \quad \left. \right\} \begin{array}{l} X \text{ gives # of trials until} \\ \text{1st success is observed!} \end{array}$$

Note: The pmf of a geometric distribution consists of terms of a geometric series

Recall: the sum of a geometric series is given by

Calculus II

$$\left\{ \sum_{k=0}^{\infty} ar^k = \sum_{k=1}^{\infty} ar^{k-1} = \frac{a}{1-r} \quad \begin{array}{l} \text{ratio} \\ \text{1st term} \end{array} \quad \text{provided } |r| < 1 . \right.$$

For the geometric distribution, we have

$$\sum_{x=1}^{\infty} g(x) = \sum_{x=1}^{\infty} (1-p)^{x-1} p = \frac{p}{1 - (1-p)} = 1 .$$

ratio in $q = (1-p) \in (0, 1)$

Still from the sum of a geometric series, we get that whenever k is a ~~an~~^{positive} integer

$$\rightarrow P(X > k) = \sum_{x=k+1}^{\infty} g(x) = \frac{(1-p)^k p}{1 - (1-p)} = (1-p)^k = q^k .$$

{ Need to know the process to sum geometric series }

Thus, the value of the cdf of the geometric distribution at a positive integer k is

$$P(X \leq k) = 1 - P(X > k) = 1 - (1-p)^k = 1 - q^k .$$

{ } }

cdf $F(x) = P(X \leq x) = \sum_{x=1}^k g(x) = \dots$

$$P(\bar{A}) = 1 - P(A)$$

Example

Biology students were checking eye color in a large number of fruit flies. For an individual fly, suppose that the probability of white eyes is $1/4$ and the probability of red eyes is $3/4$, and that we may treat these observations as independent Bernoulli trials.

Find the probability that:

- a) • at least four flies have to be checked for eye color to observe a white-eyed fly.
- b) • the first fly with white eyes is the fourth fly.

Define the r.v. X to be the number of flies that need to be checked until the first fly with white eyes is observed

Recall that the probability of observing white eyes in one of the flies is $1/4$

Then, X has a geometric distribution with prob of success $P = \frac{1}{4}$

that is, the pmf of X is
$$g(x) = \underbrace{\frac{1}{4} \cdot \left(\frac{3}{4}\right)^{x-1}}_{}, x=1,2,3,\dots$$

a) $P(X \geq 4) = \sum_{x=4}^{\infty} g(x) = \sum_{x=4}^{\infty} \frac{1}{4} \left(\frac{3}{4}\right)^{x-1} =$

$\frac{\frac{1}{4} \left(\frac{3}{4}\right)^{4-1}}{1 - \frac{3}{4}}$

geometric series with
1st term equal to $\frac{1}{4} \left(\frac{3}{4}\right)^{4-1}$
and ratio $\frac{3}{4}$

b) $P(X=4) = g(4) = \frac{1}{4} \cdot \left(\frac{3}{4}\right)^3 \quad \left\{ \begin{array}{l} \\ \end{array} \right.$

$= \frac{\frac{1}{4} \left(\frac{3}{4}\right)^3}{\frac{1}{4}} = \left(\frac{3}{4}\right)^3$

additional questions

$$\left\{ \begin{array}{l} P(X > 4) = 1 - P(X \leq 4) = 1 - P(X \leq 3) = 1 - \sum_{x=1}^3 \frac{1}{4} \left(\frac{3}{4}\right)^{x-1} \\ P(X \leq 6) = 1 - P(X > 6) = 1 - P(X \geq 7) = 1 - \sum_{x=7}^{\infty} \frac{1}{4} \left(\frac{3}{4}\right)^{x-1} \\ = 1 - \frac{\frac{1}{4} \left(\frac{3}{4}\right)^6}{1 - \frac{3}{4}} = 1 - \left(\frac{3}{4}\right)^6 \quad \checkmark \end{array} \right.$$

What if we really want to evaluate $P(X \leq 6)$ as

$$\sum_{x=1}^6 \frac{1}{4} \cdot \left(\frac{3}{4}\right)^{x-1}$$

6 terms to sum ...

$$\frac{1}{4} \cdot \frac{1 - \left(\frac{3}{4}\right)^6}{1 - \frac{3}{4}}$$

$$= 1 - \left(\frac{3}{4}\right)^6$$

OR we partial sum for a geometric series:

$$S = \sum_{k=1}^N ar^{k-1} \Rightarrow S = a + (ar) + (ar^2) + (ar^3) + \dots + (ar^{N-2}) + (ar^{N-1})$$

$$\Rightarrow rS = (ar) + (ar^2) + (ar^3) + (ar^4) + \dots + (ar^{N-1}) + (ar^N)$$

$$\Rightarrow S - rS = a - ar^N \Rightarrow (1-r)S = a - ar^N$$

named after
Gauss

$$S = \frac{a(1-r^N)}{1-r}$$

1st term *ratio* *$N = \# \text{ terms}$*

Moment generating function for negative binomial

Suppose X has a negative binomial distribution (with parameters r and p).

The mgf of X is given by

$$\text{Recall that } M(t) = E[e^{tx}]$$

$$\begin{aligned}
 M(t) &= \sum_{x=r}^{\infty} e^{tx} \binom{x-1}{r-1} p^r (1-p)^{x-r} \\
 &= (pe^t)^r \sum_{x=r}^{\infty} \binom{x-1}{r-1} [(1-p)e^t]^{x-r} \\
 &= (pe^t)^r [1 - (1-p)e^t]^r
 \end{aligned}$$

pmf of negative binomial
 $e^{tx} = e^{t(x-r+r)} = e^{t(x-r)} \cdot e^{tr}$
Binomial series
 $(1-w)^{-r} = \sum_{x=r}^{\infty} \binom{x-1}{r-1} w^{x-r}$
 with $w = (1-p)e^t$

where $(1-p)e^t < 1$ or, equivalently, $t < -\ln(1-p)$.

\uparrow
 condition for convergence of binomial series

Negative binomial distribution mean and variance

Differentiating

$$M(t) = \frac{(pe^t)^r}{[1 - (1-p)e^t]^r} \quad \left. \begin{array}{l} \\ \end{array} \right\} \text{from previous slide}$$

we obtain

$$\begin{aligned} M'(t) &= r (pe^t)^r \left[1 - (1-p)e^t \right]^{-r-1} && \xrightarrow{\text{use quotient rule + chain rule and simplify}} \\ M''(t) &= r (pe^t)^r (-r-1) \left[1 - (1-p)e^t \right]^{-r-2} [-(1-p)e^t] \\ &\quad + r^2 (pe^t)^{r-1} (pe^t) \left[1 - (1-p)e^t \right]^{-r-1} && \xrightarrow{\text{use product rule + chain rule and simplify}} \end{aligned}$$

Evaluating at $t = 0$, we find that

$$\mu = E(X) = M'(0) = \left(\frac{r}{p} \right)$$

and

$$\begin{aligned} \sigma^2 &= E(X^2) - [E(X)]^2 = M''(0) - [M'(0)]^2 \\ &= \left(\frac{r(r+1-p)}{p^2} \right) - \frac{r^2}{p^2} = \left(\frac{r(1-p)}{p^2} \right) \\ &= \frac{r(n+1)(1-p)}{p^2} \end{aligned}$$

Conclusion: If X has a negative binomial distribution with parameters $r \in \mathbb{N}$ and $p \in (0, 1)$:

- $\mu = E[X] = \frac{r}{p}$ ✓
- $\sigma^2 = \text{Var}(X) = \frac{rq}{p^2}$ ✓

Intuition: when p is the probability of success on each trial, the expected number of trials required to observe a single success is $1/p$.

obtained
in earlier
examples

→ **Special case:** when $r = 1$, X has a geometric distribution and we have

$$\left. \begin{array}{l} \bullet M(t) = \frac{pe^t}{1 - (1 - p)e^t} \\ \bullet \mu = E[X] = \frac{1}{p} \\ \bullet \sigma^2 = \text{Var}(X) = \frac{q}{p^2} \end{array} \right\} \quad \leftarrow \text{not } r=1 \text{ in the formulas for the negative binomial!}$$

Example

Suppose that during practice a basketball player can make a free throw 80% of the time. Furthermore, assume that a sequence of free-throw shooting can be thought of as independent Bernoulli trials.

Let X equal the minimum number of free throws that this player must attempt to make a total of 10 shots.

Determine the mean and variance of X .

$X = \text{r.v. counting the } \# \text{ of trials until making}$
↑
"free-throw shot"

first 10 successes
10 shots
↑
success is a shot
and probability of success is $p=0.8$

We know that X has a negative binomial distribution with parameters:

$n=10$ ← X counts # of trials until 10 successes are observed

$$p=0.8$$

$$\text{mean is } \mu = E[X] = \frac{\pi}{P} = \frac{10}{0.8}$$

$$\text{Variance is } \sigma^2 = \text{Var}(X) = \frac{\pi(1-\pi)}{P^2} = \frac{10 \cdot (1-0.8)}{(0.8)^2} = \frac{2}{(0.8)^2}$$

Example

A fair six-sided die is rolled until each face is observed at least once.

On the average, how many rolls of the die are needed?

↳ to observe each face at least once!

Roll die once → observe the first face (whatever it is) in 1 roll.

To observe the next distinct face: probability of success is $\frac{5}{6}$ ← one face already observed
5 yet to be observed

times that we need to roll die until observing one of the 5 missing faces is

a geometric r.v. with $p = \frac{5}{6}$ ⇒ mean is $\frac{1}{p} = \frac{6}{5}$

To observe the 3rd distinct face: probability of success is $\frac{4}{6}$ } 2 faces were already
observed

⇒ geometric r.v. with $p = \frac{4}{6}$ ⇒ mean is $\frac{1}{p} = \frac{6}{4}$

i^{th} distinct face: prob of success is $\frac{3}{5}$

\Rightarrow geometric r.v. with $p = \frac{3}{6}$ \Rightarrow mean is $\frac{6}{3}$

j^{th} distinct face : prob. of success is $\frac{2}{5}$

\Rightarrow geometric r.v. with $p = \frac{2}{6}$ \Rightarrow mean is $\frac{6}{2}$

Finally, for the last distinct face : prob success is $\frac{1}{6}$

\Rightarrow geometric r.v. with $p = \frac{1}{6}$ \Rightarrow mean is $\frac{6}{1}$

In total we need roll the die, on average, $1 + \frac{6}{5} + \frac{6}{4} + \frac{6}{3} + \frac{6}{2} + 6 = \frac{147}{10}$
(nearly 15 rolls to see all faces)