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If the number of trials n is fixed in advance, the number of successes X has a **binomial** distribution

If trials continue indefinitely until the first success is obtained, the number of failures obtained X has a **geometric** distribution.

If trials continue indefinitely until the  $r^{th}$  success is obtained, the number of failures obtained X has a **negative binomial** distribution.

The negative binomial experiment consists of:

- independent Bernoulli trials are performed until r successes are obtained
- The random variable X is the number of **failures** that occur before the  $r^{th}$  success.
- The probability of success p is the same for all trials

The probability mass function is:

$$f(x) = nb(x; r, p) = \binom{x+r-1}{r-1} p^r (1-p)^x, \quad x = 0, 1, 2, 3, \dots$$

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Computation:

Value R  $P(X \le x) \quad pnbinom(x, n, p)$ 

Spreadsheet P(X = x) dnbinom(x, n, p) = NEGBINOMDIST(x, r, p)

Note that the expected value and variance for the negative binomial distribution:

$$E(X) = \frac{r(1-p)}{p}$$
  $V(X) = \frac{r(1-p)}{p^2}$ 

is r times the expected value and variance of the *geometric* distribution:

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This reflects the fact that the negative binomial can be viewed as the sum of r independent geometric experiments:

Conduct Bernoulli trials until the first success, and repeat this r times.

Now we will perform some numerical experiments.

First generate a sample of 1,000,000 observations for a negative binomial experiment with r = 3 and a probability of success p = 0.4 at each trial:

*x*<*-rnbinom*(100000,3,0.4)

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Now plot a histogram of the results: *hist(x)* 

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Now plot a histogram of the results: *hist(x)* 

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The results through X = 6 should look something like:

0 1 2 3 4 5 63784 115545 138570 138259 124481 103801 8349

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The result should be something like [1] 0.064

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The result should be something like

[1] 0.064

To get the probability that X = 1 enter *dnbinom(1,3,0.4)* 

0 1 2 3 4 5 63784 115545 138570 138259 124481 103801 8349 Now compare the frequencies to the probabilities. First compute the probability that X = 0: dnbinom(0,3,0.4)

The result should be something like

[1] 0.064

To get the probability that X = 1 enter *dnbinom(1,3,0.4)* 

This time the results should look something like:

[1] 0.1152

0 1 2 3 4 5 63784 115545 138570 138259 124481 103801 8349 Next compute the probability that X = 2: dnbinom(2,3,0.4)

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[1] 0.13824

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[1] 0.13824

To get the probability that X = 5 enter *dnbinom(5,3,0.4)* 

0 1 2 3 4 5 63784 115545 138570 138259 124481 103801 8349 Next compute the probability that X = 2: *dnbinom(2,3,0.4)* The result should be something like

[1] 0.13824

To get the probability that X = 5 enter *dnbinom*(5,3,0.4)

This time the results should look something like: [1] 0.10451

The expected value E(X) in this case is:

$$E(X) = \frac{r(1-p)}{p} = \frac{3(.6)}{.4} = 4.5$$

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To compute the sample mean  $\overline{x}$ , enter mean(x)

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To compute the sample mean  $\overline{x}$ , enter *mean(x)* The result should be something like [1] 4.499121

The variance V(X) in this case is:

$$V(X) = \frac{r(1-p)}{p^2} = \frac{3(.6)}{.4^2} = 11.25$$

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To compute the sample variance  $s^2$ , enter *var(x)* 

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$$V(X) = \frac{r(1-p)}{p^2} = \frac{3(.6)}{.4^2} = 11.25$$

To compute the sample variance  $s^2$ , enter *var(x)* The result should be something like [1] 11.2477

A fair coin is tossed until the second heads comes up.

Find the probability that the second heads comes up on the fifth toss (x=3).

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Solution: 0.125

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Solution: 0.125

dnbinom(3, 2, 0.5)

#### **The Geometric Distribution**

A fair coin is tossed until the fourth heads comes up.

Find the probability that the fourth heads comes up on the seventh toss or sooner  $x \leq 3$ .

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pnbinom(3, 4, 0.5)

A fair coin is tossed until the fifth heads comes up.

Find the probability that this takes more than 8 tosses (x > 3)

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Solution: 0.63672

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1 - pnbinom(3, 5, 0.5)

A fair coin is tossed until the third heads comes up.

Find the probability that this takes 9 or more tosses (x > 5)

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1 - pnbinom(4, 3, 0.5) or = 1 - GEOMDIST(7, 0.5, TRUE)

(Spreadsheet function is for GNUMERIC. EXCEL does not have this function)

A baseball player has a .300 batting average.

Find the probability that their second hit in a game occurs on the  $5^{th}$  time at bat. (x = 3)

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dnbinom(3, 2, 0.3)