

# Distribution of the number of trials $n$ needed to obtain $s$ successes, each with probability $p$

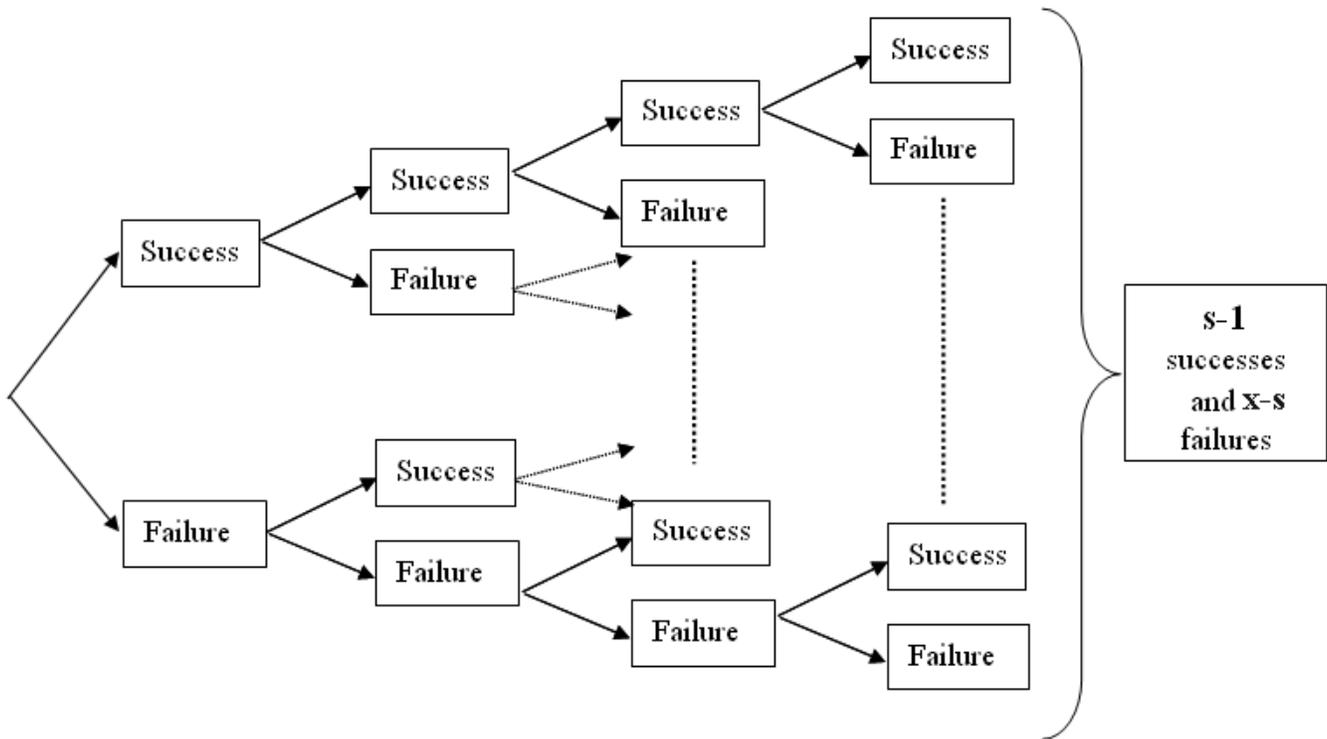
## The idea

We have [seen](#) how the Binomial distribution allows us to model the number of successes that will occur in  $n$  trials where we know the probability of success  $p$ . Sometimes, however, we know the target number of successes ( $s$ ), we know the probability  $p$ , but we wish to estimate the number of trials that we will need to complete in order to achieve these  $s$  successes, assuming we stop once the  $s^{\text{th}}$  success has occurred.

For example, imagine you have to interview ten people ( $s$ ) that have completed a marathon at some time in their life, knowing that 20% ( $p$ ) of all people have ever ran a marathon. If you would go out on the street and randomly ask people, how many people ( $n$ ) would you have to ask (estimate  $n$ )? In this case,  $n$  is the random variable.

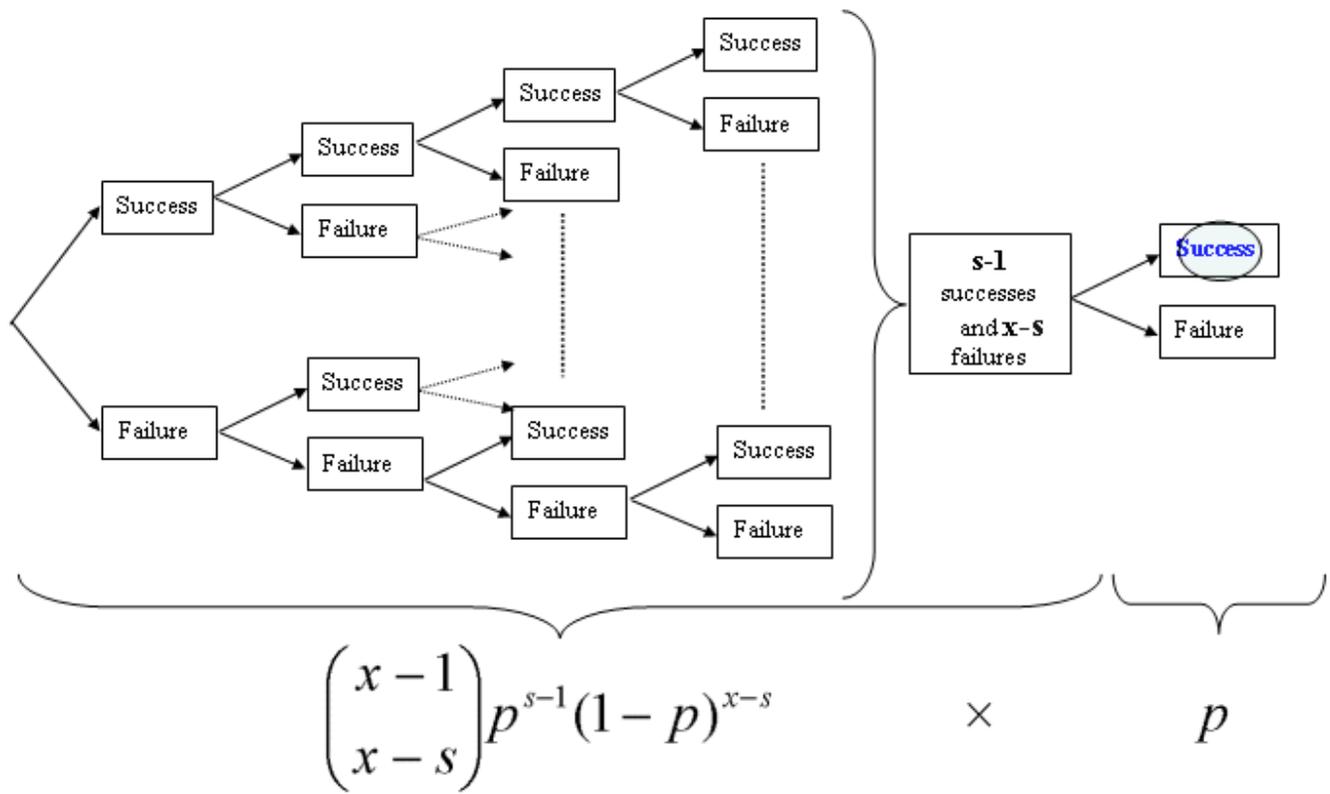
## Derivation of the Negative Binomial distribution

Now that we have the binomial distribution, we can readily determine the distribution for  $n$ . Let  $x$  be defined as the total number of trials needed to obtain  $s$  successes. Since the very last trial is by definition a success, by the  $(x-1)$ th trial we must have observed  $(s-1)$  successes and  $(x-1) - (s-1) = x-s$  failures. You can see this in the figure below.



The probability of  $(s-1)$  successes in  $(s+x-1)$  trials is given immediately by the binomial distribution as  $\binom{x-1}{s-1} p^{s-1} (1-p)^{x-s}$ . The probability of this being followed by a success is the same equation multiplied by  $p$ , i.e.:

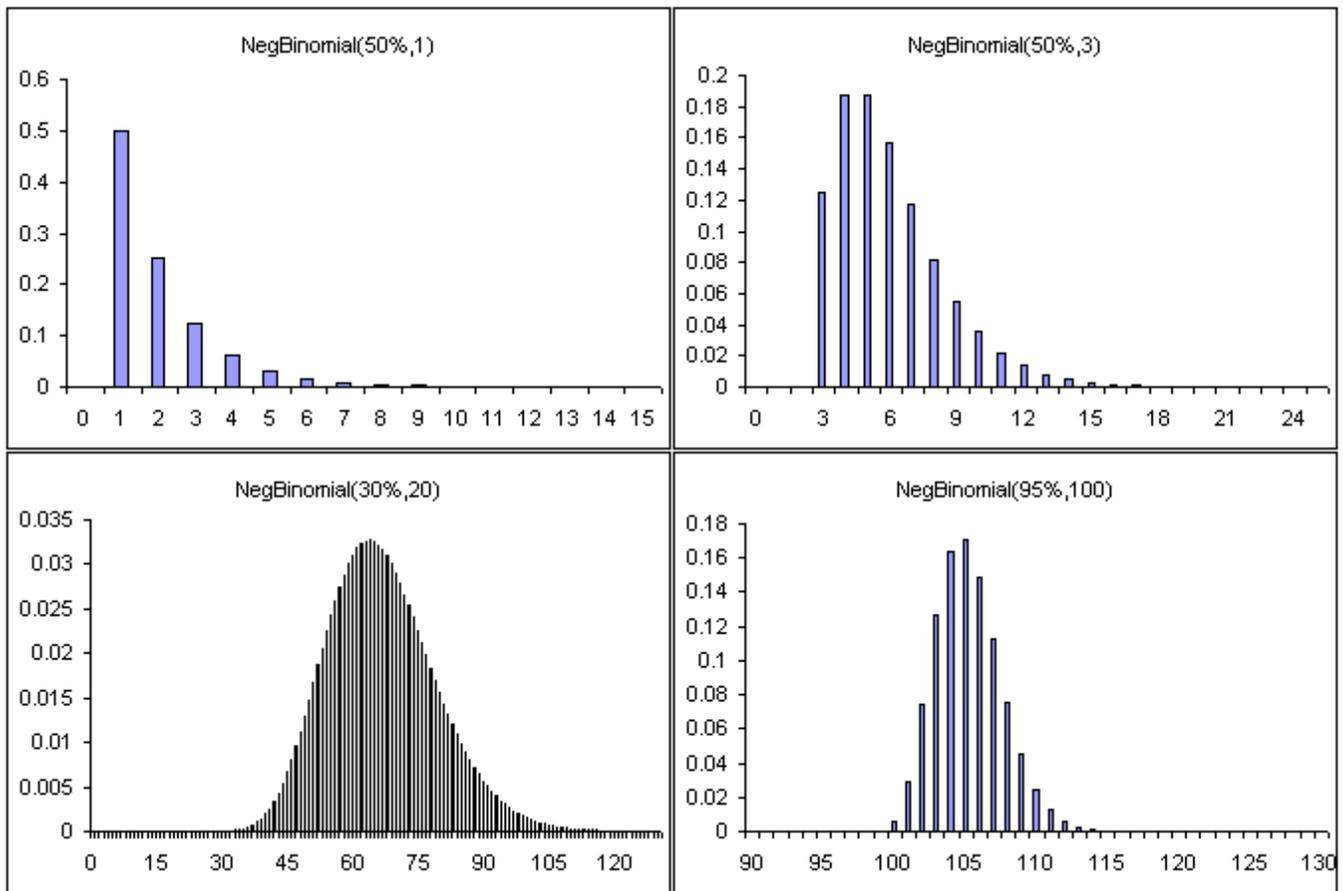
$$p(x) = \binom{x-1}{s-1} p^{s-1} (1-p)^{x-s} \times p$$



which is the probability mass function of the [Negative Binomial](#) distribution  $\text{NegBinomial}(p,s)$ . In other words, the  $\text{NegBinomial}(p,s)$  distribution returns the number of *trials* one will need to do, to observe  $s$  successes as given by:

$$n = \text{NegBinomial}(p,s)$$

The four figures below show Negative Binomial distributions for different situations.



If  $s = 1$  (see top left figure above), then the distribution (known as the [Geometric](#) distribution) is very right skewed and  $p(1) = p$ , i.e. the probability that there will be one trial (e.g. zero failures) equals  $p$ , the probability that the first trial is a success. We can also see that, as  $s$  gets larger, the distribution looks more like a [Normal](#) distribution. In fact, it is common to [approximate the Negative Binomial distribution with a Normal distribution](#) under certain circumstances where  $n$  is large, in order to avoid calculating the large factorials for  $p(x)$  above. The Negative Binomial distribution is sometimes called a *Binomial Waiting Time* distribution, or a *Pascal* distribution.