## Some Notes on the Negative Binomial Distribution

The negative binomial distribution can be defined in terms of the random variable Y = number of failures in independent Bernoulli trials (with probability of "success, p) before the r<sup>th</sup> success.

The probability mass function for *Y* is:

$$\Pr[Y = y] = {\binom{r+y-1}{y}} p^r (1-p)^y; \qquad y = 0, 1, 2, \dots; r = 1, 2, \dots; 0 (1)$$

The negative binomial distribution gets its name from the following relationship:

$$\binom{r+y-1}{y} = (-1)^{y} \binom{-r}{y} = (-1)^{y} \frac{(-r)(-r-1)\dots(-r-y+1)}{(y)(y-1)(y-2)\dots(2)(1)},$$
(2)

which defines the usual binomial coefficients in the case of *negative* integers.

Then, using the usual binomial expansion for a negative power (remember this from high school?), namely:

$$(1+x)^{-r} = \sum_{i=0}^{\infty} {\binom{-r}{i}} x^i = \sum_{i=0}^{\infty} (-1)^i {\binom{r+i-1}{i}} x^i , \qquad (3)$$

it follows immediately that

$$\sum_{y=0}^{\infty} \Pr[Y=y] = 1.$$
 (4)

The mean of the distribution can be obtained as follows:

$$E(Y) = \sum_{y=0}^{\infty} y \binom{r+y-1}{y} p^{y} (1-p)^{y} = \sum_{y=1}^{\infty} \frac{(r+y-1)!}{(y-1)!(r-1)!} p^{r} (1-p)^{y}$$
  
$$= \frac{r(1-p)}{p} \sum_{y=1}^{\infty} \binom{r+y-1}{y-1} p^{r+1} (1-p)^{y-1} = \frac{r(1-p)}{p} \sum_{k=0}^{\infty} \binom{r+1+k-1}{k} p^{r+1} (1-p)^{k}$$
  
$$= \frac{r(1-p)}{p} = r[(1/p)-1]$$
  
(5)

In the same way, we can show that

$$E(Y^2) = \frac{r(1-p)[1+r(1-p)]}{p^2},$$
(6)

so that

$$Var.(Y) = E(Y^{2}) - [E(Y)]^{2} = \frac{r(1-p)}{p^{2}}.$$
(7)

Some other characteristics of this distribution are as follows:

$$Skew(Y) = \frac{(2-p)}{\sqrt{r(1-p)}}$$
; which is always positive (8)

Excess Kurtosis 
$$(Y) = \frac{6(1-p)+p^2}{r(1-p)}$$
; which is always positive (9)

$$Mode(Y) = \frac{(r-1)(1-p)}{p} ; \qquad \text{if } r > 1 \quad (\text{otherwise the mode} = 0), \tag{10}$$

and the characteristic function for the negative binomial distribution is

$$\phi_Y(t) = \left(\frac{p}{1 - (1 - p)e^{it}}\right)^r.$$
(11)

You know that the Poisson is a limiting case of the Negative Binomial distribution. This comes about by re-parameterizing the latter distribution in terms of the mean,  $\mu = r[(1/p) - 1]$ , derived above. Then the probability mass function for the Negative Binomial distribution becomes

$$\Pr[Y = y] = \frac{\lambda^{y}}{y!} \frac{(y+r-1)!}{(r-1)!(r+\lambda)^{y}} \frac{1}{(1+\lambda/r)^{r}}.$$
(12)

Taking the limit as  $r \rightarrow \infty$ , this p.m.f collapses to that for a Poisson-distributed random variable:

$$\Pr[Y = y] = \frac{\lambda^{y}}{y!} \exp(-\lambda).$$
(13)

The Geometric distribution is also a special case of the Negative Binomial distribution. In equation (1), set r = 1:

$$\Pr[Y = y] = p(1-p)^{y} ; \qquad y = 0, 1, 2, \dots; \qquad 0 
(13)$$