

151[25] Suppose that the probability of an insect laying n eggs is given by the Poisson distribution with mean $\mu > 0$, that is, by the probability distribution given over all the nonnegative integers defined by $e^{-\mu} \mu^{n}/n!$, $n \in D = \{0, 1, 2, 3, \ldots\}$. Suppose further, that the probability of an egg developing is p. Assuming mutual independence of the eggs, show that the probability distribution $f_Y(y)$ for the probability that there are y survivors is also of Poisson type, and find the mean.

We are given two random variables, X , the number of eggs laid and Y , the number of eggs that survive. Both X and Y take values in D. The probability that $n \in D$ eggs are laid is

$$
f_X(n) = \mathbf{P}(\{X = n\}) = \frac{e^{-\mu}\mu^n}{n!}
$$

Given that n eggs are laid, then since each of the eggs survive independently, the number of these that survive is a binomial variable, so that the conditional probability that y survive is given by

$$
f_Y(y|\{X=n\}) = \mathbf{P}(\{Y=y\}|\{X=n\}) = \begin{cases} {n \choose y} p^y q^{n-y}, & \text{if } y \in \{0,1,2,\ldots,n\}; \\ 0, & \text{otherwise.} \end{cases}
$$

Note that the number that survive can't be more than the number of eggs laid. Now condition on the number of eggs laid. That is, the sets $\{X = n\}$ for $n \in D$ partition Ω , that is they are mutually disjoint and exhaustive. We use the fact that the conditional mass function is zero if $y > n$ because more can't survive than are laid, $P({Y = y} | {X = n}) = 0$ if $y > n$. By the partitioning formula, if $y \in D$,

$$
f_Y(y) = \mathbf{P}(\{Y = y\}) = \mathbf{P}\left(\bigcup_{n=0}^{\infty} (\{Y = y\} \cap \{X = n\})\right)
$$

=
$$
\sum_{n=0}^{\infty} \mathbf{P} (\{Y = y\} \cap \{X = n\})
$$

=
$$
\sum_{n=0}^{\infty} \mathbf{P} (\{Y = y\} | \{X = n\}) \mathbf{P}(\{X = n\})
$$

=
$$
\sum_{n=y}^{\infty} {n \choose y} p^y q^{n-y} \frac{e^{-\mu} \mu^n}{n!}
$$

=
$$
\frac{e^{-\mu} p^y \mu^y}{y!} \sum_{n=y}^{\infty} \frac{q^{n-y} \mu^{n-y}}{(n-y)!}
$$

=
$$
\frac{e^{-\mu} p^y \mu^y}{y!} e^{q\mu} = \frac{e^{-(1-q)\mu} p^y \mu^y}{y!} = \frac{e^{-p\mu} (p\mu)^y}{y!}.
$$

Thus we see that the distribution is also Poisson, but this time the parameter is $p\mu$ instead of μ . Since the mean of a Poisson distribution is the parameter we get

$$
\mathbf{E}(Y) = p\mu.
$$

This should not come as a surprise. It says that if an insect lays on average $\mathbf{E}(X) = \mu$ eggs in a given period and p is the survival rate, then there should be on average $p\mu$ eggs surviving in the same period.

- [A.] Suppose that $X \sim \text{Geom}(p)$ is a geometric random variable with parameter p. Find
	- (a) $\mathbf{P}(X \text{ is odd});$
	- (b) $\mathbf{P}(X \text{ is even});$
	- (c) $\mathbf{P}(X > k);$
	- (d) Let k be an integer such that $1 \leq k \leq n$. Find $P(X = k | X \leq k)$;
	- (e) $P(2 \le X \le 9 | X \ge 4);$
	- (f) Let $k \in \mathbb{N}$. Let $g(x) = \min(x, k)$ and $Y = g(X)$. Find the pmf $f_Y(y)$ and the expectation $\mathbf{E}(Y)$;
	- (g) $\mathbf{E}(1/X)$.

The standard picture of a geometric variable is a sequence of independent coin flips where the probability of head is p and X is the number of flips to get the first head. It takes values in the natural numbers $D = \mathbb{N} = \{1, 2, 3, ...\}$, and its pmf for $x \in D$ is

$$
f_X(x) = \mathbf{P}(X = x) = p q^{x-1}.
$$

(a.) The event that X is odd is given by

$$
\{X \text{ is odd}\} = \{X = 1\} \cup \{X = 3\} \cup \{X = 5\} \cup \dots = \bigcup_{k=0}^{\infty} \{X = 2k + 1\}
$$

Since these are disjoint events, we may add using the geometric sum $\sum_{k=0}^{\infty} r^k = (1-r)^{-1}$ with $r = q^2$,

$$
\mathbf{P}(X \text{ is odd}) = \sum_{k=0}^{\infty} \mathbf{P}(X = 2k+1) = \sum_{k=0}^{\infty} p q^{2k} = \frac{p}{1-q^2} = \frac{1}{1+q}.
$$

(b.) The complementary event is

$$
\mathbf{P}(X \text{ is even}) = \mathbf{P}(\{X \text{ is odd}\}^c) = 1 - \mathbf{P}(X \text{ is odd}) = 1 - \frac{1}{1+q} = \frac{q}{1+q}.
$$

(c.) The event that X is greater than k is

$$
\{X > k\} = \bigcup_{x \in D \text{ and } x > k} \{X = x\}
$$

Since these are disjoint events, we may add using the geometric sum. If $k + 1 \in D$,

$$
\mathbf{P}(X > k) = \sum_{x=k+1}^{\infty} p q^{x-1} = p (q^k + q^{k+1} + q^{k+2} + \cdots)
$$

$$
= p q^k (1 + q + q^2 + \cdots) = \frac{p q^k}{1 - q} = q^k.
$$

Note that this implies for $k \in D$, $P(X \ge k) = P(X > k - 1) = q^{k-1}$. Also the cumulative distribution function for $x \in D$,

$$
F_X(x) = \mathbf{P}(X \le x) = \mathbf{P}(\{X > x\}^c) = 1 - \mathbf{P}(X > x) = 1 - q^x.
$$
 (1)

(d.) If k is an integer so that $1 \leq k \leq n$, then the event

$$
\{X = k\} \subset \{X \le n\}.
$$

Compute conditional probabilities as usual using (1)

$$
\mathbf{P}(X = k | X \le n) = \frac{\mathbf{P}(\{X = k\} \cap \{X \le n\})}{\mathbf{P}(X \le n)} = \frac{\mathbf{P}(X = k)}{\mathbf{P}(X \le n)} = \frac{pq^{k-1}}{1 - q^n}.
$$

(e.) Using the fact that the latter events are disjoint

$$
\{X > 3\} = \{4 \le X \le 9\} \cup \{X > 9\}
$$

we get the probability using (c.),

$$
\mathbf{P}(4 \le X \le 9) = \mathbf{P}(X > 3) - \mathbf{P}(X > 9) = q^3 - q^9.
$$

Equivalently, $P(4 \le X \le 9) = P(X \le 9) - P(X \le 3) = F_X(9) - F_X(3)$. We compute conditional probabilities as usual:

$$
\mathbf{P}(2 \le X \le 9 | X \ge 4) = \frac{\mathbf{P}(\{2 \le X \le 9\} \cap \{X \ge 4\})}{\mathbf{P}(X \ge 4)} = \frac{\mathbf{P}(4 \le X \le 9)}{\mathbf{P}(X > 3)} = \frac{q^3 - q^9}{q^3} = 1 - q^6.
$$

 $(f.)$ For the natural number k, the function

$$
g(x) = \min(x, k) = \begin{cases} x, & \text{if } x < k; \\ k, & \text{if } x \ge k. \end{cases}
$$

g maps D to $D' = \{1, 2, 3, ..., k\}$. Thus if $y \in D'$ and $y < k$ then there is exactly one $x \in D$ such that $y = g(x)$, namely, $x = y$. If $y = k$ then the set of x's that map to k is $\{k, k+1, k+2, \ldots\} = \{X \geq k\}.$ This set is also called the preimage $g^{-1}(\{k\})$. Using the formula for the pmf of the new random variable $Y = g(X)$ we have

$$
f_Y(y) = \sum_{x \in D \text{ such that } g(x) = y} f_X(x).
$$

For this $g(X)$, using (c.), it becomes for $y \in D'$,

$$
f_Y(y) = \begin{cases} f_X(y) = p q^{y-1}, & \text{if } y < k; \\ \sum_{x=k}^{\infty} f_X(x) = \mathbf{P}(X \ge k) = q^{k-1}, & \text{if } x = k. \end{cases}
$$

Of course, $f_Y(y) = 0$ if $y \notin D'$.

To find the expectation we may use the definition or Theorem 4.3.4, which give the same expression. Thus if $k > 1$,

$$
\mathbf{E}(Y) = \sum_{y \in D'} y f_Y(y) = \left(\sum_{y=1}^{k-1} y p q^{y-1}\right) + k q^{k-1} = p \frac{d}{dq} \left(\sum_{y=0}^{k-1} q^y\right) + k q^{k-1}
$$

= $p \frac{d}{dq} \left(\frac{1-q^k}{1-q}\right) + k q^{k-1} = p \left(\frac{-k q^{k-1}}{1-q} + \frac{1-q^k}{(1-q)^2}\right) + k q^{k-1} = \frac{1-q^k}{p}.$

If $k = 1$ there is one term and $E(Y) = 1$ so the formula works for $k \ge 1$ as well. Since $g(x) \leq x$, it is no surprise that this is close to but less than $\mathbf{E}(X) = 1/p$.

(g.) To find the expectation of $Z = h(X)$ where $h(x) = 1/x$, we use Theorem 4.3.4.

$$
\mathbf{E}(Z) = \sum_{x \in D} h(x) f_X(x) = \sum_{x=1}^{\infty} \frac{p q^{x-1}}{x} = \frac{p}{q} \sum_{x=1}^{\infty} \frac{q^x}{x} = -\frac{p \log(1-q)}{q} = -\frac{p \log p}{q}.
$$

See problem 151[18] and page 23.

[B.] Suppose that an unfair coin is tossed repeatedly. Suppose that the tosses are independent and the probability of each head is p . Let X denote the number of tosses it takes to get three heads. Derive the formulas for $\mathbf{E}(X)$ and $\mathbf{Var}(X)$.

X is distributed according to the negative binomial distribution with parameters $k = 3$ and p. The variable takes values in $D = \{k, k+1, k+2, \ldots\}$ since one must toss the coin k times at least in order to have k heads. In order that the k-th head occur at the x-th toss, there must be $k-1$ heads in the first $x-1$ tosses and the x-th toss has to be a head. Thus the negative binomial pmf is for $x \in D$,

$$
f_X(x) = \binom{x-1}{k-1} p^k q^{x-k}.
$$

The expectation is given by the sum

$$
\mathbf{E}(X) = \sum_{x \in D} x f_X(x) = \sum_{x=k}^{\infty} x {x-1 \choose k-1} p^k q^{x-k}.
$$

Using the formula for binomial coefficients,

$$
x\binom{x-1}{k-1} = \frac{x(x-1)!}{(k-1)!(x-k)!} = \frac{k x!}{k (k-1)!(x-k)!} = \frac{k x!}{k! (x-k)!} = k \binom{x}{x-k}.
$$

Inserting and changing dummy index by $j = x - k$,

$$
\mathbf{E}(X) = \sum_{x=k}^{\infty} k \binom{x}{x-k} p^k q^{x-k} = k p^k \sum_{j=0}^{\infty} \binom{k+j}{j} q^j.
$$

From page 22, the negative binomial series is

$$
\sum_{j=0}^{\infty} {m+j-1 \choose j} z^j = (1-z)^{-m}
$$

which makes sense even if $m > 0$ is a real number. In our series, the role of m is played by $m = k + 1$. Thus

$$
\mathbf{E}(X) = k p^{k} (1 - q)^{-(k+1)} = \frac{k}{p}.
$$

To compute the variance, we use the computation formula

$$
\mathbf{Var}(X) = \mathbf{E}(X^2) - \mathbf{E}(X)^2.
$$

Using the equation $x^2 = (x+1)x - x$, the expectation of the square is

$$
\mathbf{E}(X^2) = \sum_{x \in D} x^2 f_X(x) = \sum_{x=k}^{\infty} x^2 {x-1 \choose k-1} p^k q^{x-k}
$$

=
$$
\sum_{x=k}^{\infty} (x+1) x {x-1 \choose k-1} p^k q^{x-k} - \sum_{x=k}^{\infty} x {x-1 \choose k-1} p^k q^{x-k}.
$$

The second sum is $-\mathbf{E}(X)$. We have another binomial coefficient identity

$$
(x+1)x\binom{x-1}{k-1} = \frac{(x+1)x(x-1)!}{(k-1)!(x-k)!} = \frac{(k+1)k(x+1)!}{(k+1)k(k-1)!(x-k)!}
$$

$$
= \frac{(k+1)k(x+1)!}{(k+1)!(x-k)!} = (k+1)k\binom{x+1}{x-k}.
$$

Inserting this, changing dummy index to $j = x - k$ and using $m = k + 2$ in the negative binomial series, the first sum in $E(X^2)$ yields

$$
\sum_{x=k}^{\infty} (x+1)x {x-1 \choose k-1} p^k q^{x-k} = \sum_{x=k}^{\infty} (k+1)k {x+1 \choose x-k} p^k q^{x-k}
$$

= $(k+1)kp^k \sum_{j=0}^{\infty} {k+j+1 \choose j} q^j = (k+1)kp^k (1-q)^{-(k+2)} = \frac{(k+1)k}{p^2}.$

Assembling,

$$
\mathbf{Var}(X) = \frac{(k+1)k}{p^2} - \frac{k}{p} - \frac{k^2}{p^2} = \frac{(k+1)k - pk - k^2}{p^2} = \frac{kq}{p^2}.
$$