

# Claims Frequency Distribution Models

## Chapter 6

Stat 346 - Short-term Actuarial Math

# Introduction

- Here we introduce a large class of counting distributions, which are discrete distributions with support consisting of non-negative integers.
- Generally used for modeling number of events, but in an insurance context, the number of claims within a certain period, e.g. one year.
- We call these **claims frequency** models.
- Let  $N$  denote the number of events (or claims). Its probability mass function (pmf),  $p_k = \Pr(N = k)$ , for  $k = 0, 1, 2, \dots$ , gives the probability that exactly  $k$  events (or claims) occur.

# Some familiar discrete distributions

Some of the most commonly used distributions for number of claims:

- Binomial (with Bernoulli as special case)
- Poisson
- Geometric
- Negative Binomial
- The  $(a, b, 0)$  class
- The  $(a, b, 1)$  class

# Bernoulli random variables

- $N$  is **Bernoulli** if it takes only one of two possible outcomes:

$$N = \begin{cases} 1, & \text{if a claim occurs} \\ 0, & \text{otherwise} \end{cases}.$$

- $q$  is the standard symbol for the probability of a claim, i.e.  
 $\Pr(N = 1) = q$ .
- We write  $N \sim \text{Bernoulli}(q)$ .
- Mean  $E(N) = q$  and variance  $\text{Var}(N) = q(1 - q)$

# Binomial random variables

- We write  $N \sim \text{Binomial}(m, q)$  if  $N$  has a **Binomial** distribution with pmf:

$$p_k = \Pr(N = k) = \binom{m}{k} q^k (1 - q)^{m-k} = \frac{m!}{k!(m-k)!} q^k (1 - q)^{m-k},$$

for  $k = 0, \dots, m$ .

- Binomial r.v. is also the sum of independent Bernoulli's with  $N = \sum_{k=1}^m N_k$  where each  $N_k \sim \text{Bernoulli}(q)$ .
- Mean  $E(N) = mq$  and variance  $\text{Var}(N) = mq(1 - q)$

# Poisson random variables

- $N \sim \text{Poisson}(\lambda)$  if pmf is

$$p_k = P(X = k) = e^{-\lambda} \frac{\lambda^k}{k!}, \quad \text{for } k = 0, 1, 2, \dots$$

- Mean and variance are equal:  $E(N) = \text{Var}(N) = \lambda$
- Sums of independent Poissons: If  $N_1, \dots, N_n$  be  $n$  independent Poisson variables with parameters  $\lambda_1, \dots, \lambda_n$ , then the sum

$$N = N_1 + \dots + N_n$$

has a Poisson distribution with parameter  $\lambda = \lambda_1 + \dots + \lambda_n$ .

# Negative binomial random variable

- $N$  has a **Negative Binomial** distribution, written  $N \sim \text{NB}(\beta, r)$ , if its pmf can be expressed as

$$p_k = \Pr(N = k) = \binom{k+r-1}{k} \left( \frac{1}{1+\beta} \right)^r \left( \frac{\beta}{1+\beta} \right)^k,$$

for  $k = 0, 1, 2, \dots$  where  $r > 0, \beta > 0$ .

- Mean:  $E(N) = r\beta$
- Variance:  $\text{Var}(N) = r\beta(1 + \beta)$ .
- Clearly, since  $\beta > 0$ , the variance of the NB exceeds the mean.

## Geometric random variable

- The **Geometric** distribution is a special case of the Negative Binomial with  $r = 1$ .
- $N$  is said to be a **Geometric** r.v. and written as  $N \sim \text{Geometric}(p)$  if its pmf is therefore expressed as

$$p_k = \Pr(N = k) = \frac{1}{1 + \beta} \left( \frac{\beta}{1 + \beta} \right)^k, \quad \text{for } k = 0, 1, 2, \dots$$

- Mean is  $E(N) = \beta$  and variance is  $\text{Var}(N) = \beta(1 + \beta)$ .



# Special class of distributions

- The  $(a, b, 0)$  class of distributions satisfies the recursion equations of the general form:

$$\frac{p_k}{p_{k-1}} = a + \frac{b}{k}, \quad \text{for } k = 1, 2, \dots$$

- The three distributions (including Geometric as special case of NB) are the only distributions that belong to this class: Binomial, Poisson, and Negative Binomial.
- It can be shown that the applicable parameters  $a$  and  $b$  are:

Distribution	Values of $a$ and $b$
Binomial( $m, q$ )	$a = -\frac{q}{1-q}, b = (m+1)\frac{q}{1-q}$
Poisson( $\lambda$ )	$a = 0, b = \lambda$
NB( $\beta, r$ )	$a = \frac{\beta}{1+\beta}, b = (r-1)\frac{\beta}{1+\beta}$

## Example

Suppose  $N$  is a counting distribution satisfying the recursive probabilities:

$$\frac{p_k}{p_{k-1}} = \frac{4}{k} - \frac{1}{3},$$

for  $k = 1, 2, \dots$

Identify the distribution of  $N$ .

# Solution to the Recursive Probability Example

- Given recursive probabilities:

$$\frac{p_k}{p_{k-1}} = \frac{4}{k} - \frac{1}{3},$$

for  $k = 1, 2, \dots$

- Comparing with the general form of  $(a, b, 0)$  class:

$$\frac{p_k}{p_{k-1}} = a + \frac{b}{k}$$

we have  $a = -\frac{1}{3}$  and  $b = 4$ .

- Comparing with the known distributions, only the binomial distribution has a negative  $a$ . Set  $a = -\frac{1}{3} = -\frac{q}{1-q}$  and  $b = 4 = (m+1)\frac{q}{1-q}$

## SOA question

The distribution of accidents for 84 randomly selected policies is as follows:

Number of Accidents	Number of Policies
0	32
1	26
2	12
3	7
4	4
5	2
6	1

Identify the frequency model that best represents these data.

# Analyzing the Accident Data

- Given accident data for 84 policies:

Number of Accidents (k)	Number of Policies	Ratio $\frac{p_k}{p_{k-1}}$
0	32	-
1	26	$\frac{26}{32} \approx 0.81$
2	12	$\frac{12}{26} \approx 0.46$
3	7	$\frac{7}{12} \approx 0.58$
4	4	$\frac{4}{7} \approx 0.57$
5	2	$\frac{2}{4} = 0.50$
6	1	$\frac{1}{2} = 0.50$

- Analyzing the ratio  $\frac{p_k}{p_{k-1}}$  for each  $k$  to identify a possible distribution from the  $(a, b, 0)$  class.

# Truncation and modification at zero

- The  $(a, b, 1)$  class of distributions satisfies the recursion equations of the general form:

$$\frac{p_k}{p_{k-1}} = a + \frac{b}{k}, \quad \text{for } k = 2, 3, \dots$$

- Only difference with the  $(a, b, 0)$  class is the recursion here begins at  $p_1$  instead of  $p_0$ . The values from  $k = 1$  to  $k = \infty$  are the same up to a constant of proportionality. For the class to be a distribution, the remaining probability must be set for  $k = 0$ .
  - **zero-truncated** distributions: the case when  $p_0 = 0$
  - **zero-modified** distributions: the case when  $p_0 > 0$
- The distributions in the second subclass is indeed a mixture of an  $(a, b, 0)$  and a degenerate distribution. A zero-modified distribution can be viewed as a zero-truncated by setting  $p_0 = 0$ .

# Expectation and Variance

## Zero-Modified and Zero-Truncated Distributions

- For a **zero-truncated** distribution, the expected value (mean) and variance are given by:

$$E[X_{\text{trunc}}] = \frac{E[X]}{1 - p_0},$$

$$\text{Var}[X_{\text{trunc}}] = \frac{\text{Var}[X]}{1 - p_0} - \frac{p_0 \cdot [E[X]]^2}{(1 - p_0)^2},$$

where  $X_{\text{trunc}}$  is the zero-truncated version of the random variable  $X$ .

- For a **zero-modified** distribution, assuming that  $X_{\text{mod}}$  is the modified variable and  $p'_0$  is the modified probability at zero:

$$E[X_{\text{mod}}] = (1 - p'_0) \cdot E[X_{\text{trunc}}],$$

$$\text{Var}[X_{\text{mod}}] = (1 - p'_0) \cdot \text{Var}[X_{\text{trunc}}] + p'_0 \cdot (1 - p'_0) \cdot [E[X_{\text{trunc}}]]^2.$$

# Zero-Modified and Zero-Truncated Distributions

## • Zero-Modified Distributions:

- In zero-modified distributions, the probability at zero,  $p_0$ , is artificially altered.
- This modification changes the probabilities  $p_k$  for  $k \geq 1$ .
- The adjusted probabilities for  $k \geq 1$  are:

$$p'_k = \frac{(1 - p'_0) \cdot p_k}{1 - p_0} \quad \text{for } k \geq 1,$$

where  $p_k$  are the original probabilities,  $p_0$  is the original probability at zero, and  $p'_0$  is the modified probability at zero.

## • Zero-Truncated Distributions:

- In a zero-truncated distribution, occurrences at zero are removed (i.e.,  $p'_0 = 0$ ).
- The probabilities  $p_k$  for  $k \geq 1$  are scaled up so that the distribution sums to 1.
- The adjusted probabilities are:

$$p'_k = \frac{p_k}{1 - p_0} \quad \text{for } k \geq 1,$$



# The Role of $C = \frac{1-p'_0}{1-p_0}$

- Define the constant:

$$C = \frac{1 - p'_0}{1 - p_0}$$

- This constant establishes a proportionality between the probabilities of the zero-modified and zero-truncated distributions:

$$p'_k = C \cdot p_k, \quad \text{for } k \geq 1.$$

- Additionally, this proportionality extends to the moments of the distribution:

$$E[X_{\text{mod}}^k] = C \cdot E[X^k].$$

- $C$  plays a key role in linking the expectations and variances of zero-modified and zero-truncated distributions.

## Zero-Modified Poisson Example ( $p'_0 > 0$ )

- Consider a Poisson distribution with rate parameter  $\lambda = 3$ :

$$p_k = \frac{\lambda^k e^{-\lambda}}{k!}, \quad k = 0, 1, 2, \dots$$

- Let the original  $p_0 = e^{-\lambda} \approx 0.0498$ , and modify it to  $p'_0 = 0.1$ .
- Compute  $C$ :

$$C = \frac{1 - p'_0}{1 - p_0} = \frac{1 - 0.1}{1 - 0.0498} \approx 0.947.$$

- Adjusted probabilities:

$$p'_k = C \cdot p_k, \quad \text{for } k \geq 1.$$

- Example for  $k = 1, 2$ :

$$p'_1 = 0.947 \cdot 0.1494 \approx 0.1415, \quad p'_2 = 0.947 \cdot 0.224 \approx 0.212.$$

- Adjusted expectation:

$$E[X_{mod}] = C \cdot E[X] = 0.947 \cdot 3 \approx 2.841.$$

## Variance of the Zero-Modified Poisson Distribution

- To compute the variance of the zero-modified Poisson distribution, we use:

$$\text{Var}[X_{\text{mod}}] = E[X_{\text{mod}}^2] - (E[X_{\text{mod}}])^2.$$

- Start by finding  $E[X_{\text{mod}}^2]$ :

$$E[X_{\text{mod}}^2] = C \cdot E[X^2],$$

- For a Poisson distribution with  $\lambda = 3$ :

$$\text{Var}[X] = \lambda = 3, \quad E[X] = \lambda = 3.$$

- Thus:

$$E[X^2] = 3 + 3^2 = 12.$$

- Compute  $E[X_{\text{mod}}^2]$ :

$$E[X_{\text{mod}}^2] = C \cdot E[X^2] = 0.947 \cdot 12 \approx 11.364.$$

- Finally, compute  $\text{Var}[X_{\text{mod}}]$ :

$$\text{Var}[X_{\text{mod}}] = 11.364 - (2.841)^2 \approx 11.364 - 8.070 = 3.294.$$

## Zero-Truncated Poisson Example ( $p'_0 = 0$ )

- For a zero-truncated Poisson distribution ( $p'_0 = 0$ ):

$$p'_k = \frac{p_k}{1 - p_0}, \quad \text{for } k \geq 1,$$
$$p_k = \frac{\lambda^k e^{-\lambda}}{k!}.$$

- Let  $\lambda = 3$ , so  $p_0 = e^{-\lambda} \approx 0.0498$  and  $1 - p_0 \approx 0.9502$ .
- Example for  $k = 1, 2$ :

$$p'_1 = \frac{p_1}{1 - p_0} = \frac{0.1494}{0.9502} \approx 0.1572, \quad p'_2 = \frac{p_2}{1 - p_0} = \frac{0.2240}{0.9502} \approx 0.2357.$$

- Expected value for the zero-truncated Poisson:

$$E[X_{\text{trunc}}] = \frac{E[X]}{1 - p_0} = \frac{3}{0.9502} \approx 3.157.$$

## Illustrative example

Consider the zero-modified Geometric distribution with probabilities

$$\begin{aligned}p_0 &= \frac{1}{2} \\ p_k &= \frac{1}{6} \left( \frac{2}{3} \right)^{k-1}, \quad \text{for } k = 1, 2, 3, \dots\end{aligned}$$

Derive the mean and the variance of this distribution.