

Poisson and Negative Binomial Distribution

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1 Introduction

This notes is about poisson and negative binomial distribution, and their application in modeling count data.

2 Poisson Distribution

2.1 Basic Information

The p.m.f of a random variable X that follows Poisson distribution is

$$P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}, \quad x = 0, 1, \dots$$

The expectation and variance of X is

$$EX = \text{Var}X = \lambda.$$

2.2 Some Additoinal Properties

1. Summation of independent poisson distributed variables is still poisson distributed. That is to say,

if

$$X_1 \sim \text{Poisson}(\lambda_1), \quad X_2 \sim \text{Poisson}(\lambda_2), \quad X_1 \perp X_2,$$

then

$$X_1 + X_2 \sim \text{Poisson}(\lambda_1 + \lambda_2).$$

2. Poisson distribution can be related to Gamma distribution, which is also mentioned in “gamma_and_beta.pdf” notes.
3. The conjugate prior for a Poisson distribution parameter λ is Gamma distribution.
4. A poisson distributed variable can be seen as counts of customer to a store or occurence of some events. And the **time interval between each count** follows exponential distribution.

3 Negative Binomial Distribution

3.1 Basic Information

Denote X the number of failing attempts just before reaching a total of r times of success and p the probability of success. If each attempt is independent of each other, then the distribution of X is called a negative binomial distribution. And the p.m.f of X is

$$P(X = x) = C_{x+r-1}^x p^r (1-p)^x = \frac{(x+r-1)!}{(r-1)!x!} p^r (1-p)^x, \quad x = 0, 1, \dots$$

In this notation, X is the number of failing attempts. In some cases, one will model the distribution of number of total attempts X' , which is just $X' = X + r$.

Previously, r is the number of required success, which is naturally an positive integer. But one can actually generalize r to positive real number using Gamma function. Then the p.m.f of X is written as

$$P(X = x) = \frac{\Gamma(x+r)}{\Gamma(r)\Gamma(x+1)} p^r (1-p)^x, \quad x = 0, 1, \dots,$$

which is also called **Polya** distribution.

The expectation of a negative binomial/Polya distribution is

$$EX = \frac{r(1-p)}{p}.$$

The variance of a negative binomial/Polya distribution is

$$\text{Var}X = \frac{r(1-p)}{p^2}.$$

3.2 Some Additional Properties

1. Summation of independent negative binomial distributed variables. That is to say, if $X_1 \sim NB(r_1, p)$, $X_2 \sim NB(r_2, p)$ and $X_1 \perp X_2$, then

$$X_1 + X_2 \sim NB(r_1 + r_2, p).$$

2. Relationship to Poisson and Gamma distribution.

- Related to poisson when $r \rightarrow \infty$.
- Related to poisson and gamma distribution in a hierarchical model:

$$\lambda \sim \text{Gamma}(\alpha, \beta), \quad f(\lambda) = \frac{1}{\Gamma(\alpha)\beta^\alpha} \lambda^{\alpha-1} \exp(-\lambda/\beta)$$

$$X|\lambda \sim \text{Poisson}(\lambda), \quad P(X = x|\lambda) = \frac{\lambda^x e^{-\lambda}}{x!}.$$

Then the marginal distribution of X is

$$\begin{aligned} P(X = x) &= \int_0^\infty P(X = x|\lambda) f(\lambda) d\lambda \\ &= \int_0^\infty \frac{\lambda^x e^{-\lambda}}{x!} \cdot \frac{1}{\Gamma(\alpha)\beta^\alpha} \lambda^{\alpha-1} \exp(-\lambda/\beta) d\lambda \\ &= \frac{1}{x! \Gamma(\alpha) \beta^\alpha} \int_0^\infty \lambda^{x+\alpha-1} \exp\left(-\frac{\lambda}{\beta/(1+\beta)}\right) d\lambda \quad \text{The kernel of Gamma}(x + \alpha, \frac{\beta}{1 + \beta}) \\ &= \frac{\Gamma(x + \alpha) \left(\frac{\beta}{1 + \beta}\right)^{x+\alpha}}{x! \Gamma(\alpha) \beta^\alpha} \\ &= \frac{\Gamma(x + \alpha)}{\Gamma(\alpha) \Gamma(x + 1)} \left(\frac{1}{1 + \beta}\right)^\alpha \left(1 - \frac{1}{1 + \beta}\right)^x, \end{aligned}$$

which is a negative binomial distribution with parameters $r = \alpha$ and $p = \frac{1}{1+\beta}$. In this way,

$$\begin{aligned} EX &= \frac{r(1-p)}{p} = \alpha\beta \\ \text{Var}X &= \frac{r(1-p)}{p^2} = \alpha\beta(1+\beta) \end{aligned}$$

Note that here β is actually the **failure odds**, which is mentioned in later properties.

3. Dispersion parameter and other parametrization of negative binomial distribution.

- In some literature, expectation and variance are used to describe negative binomial distribution:

$$\begin{aligned} EX &= \mu \\ \text{Var}X &= \sigma^2 \\ r &= \frac{\mu^2}{\sigma^2 - \mu} \\ p &= \frac{\mu}{\sigma^2} \end{aligned}$$

- In some literature, $\alpha = \frac{1}{r}$ is called the dispersion parameter while others call r the dispersion parameter. It also has other names such as “shape parameter”, “clustering coefficient”, “heterogeneity” and “aggregation parameter”. Note that the expectation and the variance of a negative binomial distributed variable satisfies

$$\sigma^2 = \mu + \alpha\mu^2 = \mu + \frac{\mu^2}{r}.$$

- Denote β the **failure odds**, then

$$\begin{aligned}\beta &= \frac{1-p}{p} \\ P(X=x) &= \frac{\Gamma(x+r)}{\Gamma(r)\Gamma(x+1)} \left(\frac{\beta}{1+\beta}\right)^r \left(\frac{1}{1+\beta}\right)^x, \quad x=0,1,\dots, \\ EX &= r\beta \\ \text{Var}X &= r\beta(1+\beta).\end{aligned}$$

4. Geometric distribution is a special case of negative binomial distribution.

4 Modeling Counts Data

4.1 Basic Modeling

4.2 Zero Inflation

References