

Binomial and Poisson distributions: rationale

The Binomial distribution is used to model the number of successes occurring in n independent Bernoulli trials. The Poisson distribution is used to model a phenomenon that represents the count of events in a fixed time interval (or space unit). The r.v. X represents then the number of successes in this interval.

Suppose that λ denotes the number of independent observations of an event over a given time interval. If we divide this interval into n smaller intervals, then we find $p = \frac{\lambda}{n} \Leftrightarrow \lambda = np$, the probability of observing a success over a time interval n times smaller. Conditions for a Poisson approximation to the Binomial distribution:

- (i) n is large
- (ii) $n * p$ is small enough

Poisson distribution as an asymptotic case of the Binomial distribution

$$\begin{aligned}P(X = x) &= \binom{n}{x} p^x (1-p)^{n-x} \\&= \binom{n}{x} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x} \\&= \binom{n}{x} \frac{\lambda^x}{n^x} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x} \\&= \frac{n(n-1)\dots(n-x+1)}{x!} \frac{\lambda^x}{n^x} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x} \\&= \frac{n(n-1)\dots(n-x+1)}{n^x} \frac{\lambda^x}{x!} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x} \\&= \underbrace{1 \left(1 - \frac{1}{n}\right) \left(1 - \frac{2}{n}\right) \dots \left(1 - \frac{x-1}{n}\right)}_{\approx 1} \underbrace{\frac{\lambda^x}{x!} \left(1 - \frac{\lambda}{n}\right)^n}_{\approx e^{-\lambda}} \underbrace{\left(1 - \frac{\lambda}{n}\right)^{-x}}_{\approx 0} \\&\approx \frac{\lambda^x}{x!} e^{-\lambda} \quad \text{as } n \rightarrow \infty\end{aligned}$$

Example of application

A machine tool produces 20,000 pieces a day. The probability of one piece being defective is 0.1%. The probability that this machine produces exactly 20 defective pieces is given by the Binomial distribution and equals $f_X(20) = P(X = 20) = \binom{20000}{20} 0.001^{20} 0.999^{20000-20} = 0.0889$.

In the R programming language:

```
1 dbinom(x = 20, size = 20000, prob = 0.001)
2 [1] 0.08887977
```

In the Python language:

```
1 from scipy.stats import binom
2 binom.pmf(20, n = 20000, p = 0.001)
3 0.08887976802110753
```

Poisson approximation

A machine tool produces 20,000 pieces a day. The probability of one piece being defective is 0.1%. The probability that this machine produces exactly 20 defective pieces can be approximated by a Poisson distribution. If we let $\lambda = np$, that is, in our example $\lambda = 20000 * 0.001 = 20$, we can thus write $f_X(20) = P(X = 20) = \frac{20^{20}e^{-20}}{20!} = 0.0888$.

In the R programming language:

```
1 dpois(x = 20, lambda = 20000*0.001)
2 [1] 0.08883532
```

In the Python language:

```
1 from scipy.stats import poisson
2 poisson.pmf(20, mu = 20000*0.001)
3 0.0888353173920848
```

References

Rizzo, M.L. (2019). Statistical Computing with R, Second Edition (2nd ed.). Chapman and Hall/CRC.

<https://doi.org/10.1201/9780429192760>

The R Project for Statistical Computing:

<https://www.r-project.org/>

Python:

<https://www.python.org/>

course notes