

Mathematical Statistics 1

Ch.2 Discrete Distributions

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Ch.2.6 The Poisson Distribution

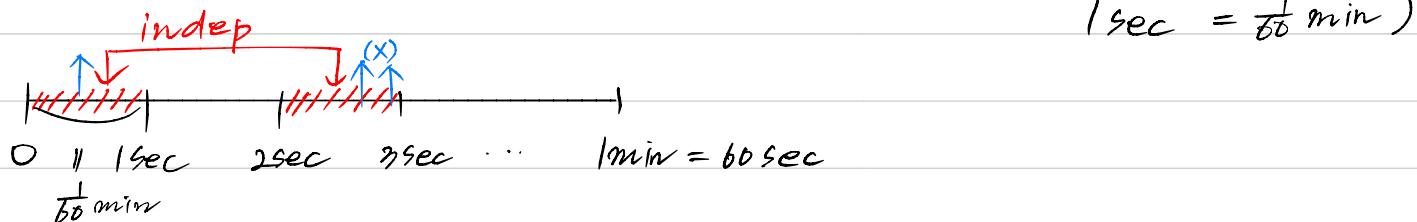
6.1 Approximate Poisson process

Let the number of occurrences of some event in a given continuous interval be counted. Then we have an **approximate Poisson process** with parameter $\lambda > 0$ if the following conditions are satisfied:

- The number of occurrences in non-overlapping subintervals are independent.
- The probability of exactly one occurrence in a sufficiently short subinterval of length h is approximately λh .
- The probability of two or more occurrences in a sufficiently short subinterval is essentially zero.

$X \sim$ Approximate Poisson process with " λ "
 (where λ means the average counts in given time)

Let's consider 1 min as the time unit ($1 \text{ min} = 60 \text{ sec}$)



- ① # of outcomes in $[0, 1](\text{sec})$ is indep of
 # of outcomes in $[2, 3](\text{sec})$
- ② $P(\text{one outcome appears in short sub-interval of } h (= \frac{1}{60} \text{ min})) \approx \lambda h = \frac{1}{60} \lambda$
- ③ $P(\text{outcomes over two appear in short sub-interval}) \approx 0$

$$X = \# \text{ of outcomes in time interval}$$

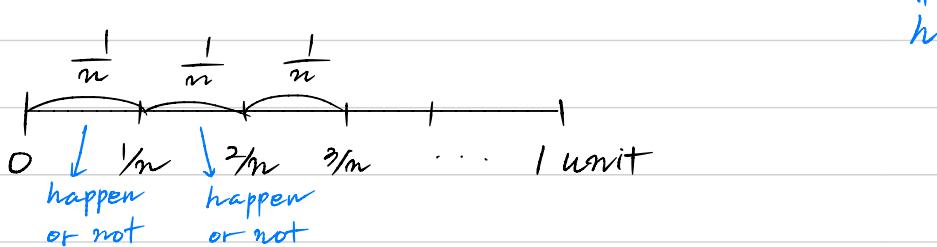
$$\Rightarrow f_X(x) = P(X=x)$$

Derivation of Poisson process

$X = \# \text{ of outcomes in time interval "1"}$

Goal: We should like to find an approximation for $P(X=x)$

i) Partition the time unit interval into large "n" sub-intervals of equal length " $\frac{1}{n}$ "



$$\begin{aligned} P(X=x) &= P(x \text{ outcomes in "1" time unit}) \\ &= P(x \text{ outcomes in "n" sub-intervals}) \end{aligned}$$

ii) By approx. Poisson process (b),

$$\begin{aligned} P(\text{exactly one outcome appears in the sub-interval}) \\ \approx \lambda h = \frac{\lambda}{n} \end{aligned}$$

By approx. Poisson process (c),

two situations - outcome appears or not in the sub-interval

\Rightarrow Sequence of "n" Bernoulli trials with $p = \frac{\lambda}{n}$
(prob. of occurrence)

$$P(X=x) \approx \binom{n}{x} p^x (1-p)^{n-x} = \binom{n}{x} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x}$$

As n increases without bound,

$$P(X=x) = \lim_{n \rightarrow \infty} \binom{n}{x} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x} =$$

$$\begin{aligned}
 &= \lim_{n \rightarrow \infty} \frac{n(n-1) \cdots (n-x+1)}{x!} \frac{\lambda^x}{n^x} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x} \\
 &= \lim_{n \rightarrow \infty} \frac{n(n-1) \cdots (n-x+1)}{x!} \frac{\lambda^x}{n^x} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x} = \frac{\lambda^x e^{-\lambda}}{x!}
 \end{aligned}$$

↓ ↓ ↓ ↑
 / e^{-λ} / prob of Poisson
 distribution

X = # of outcomes in a given time interval

$$f_X(x) = P(X=x) = \frac{\lambda^x \cdot e^{-\lambda}}{x!} \quad x = 0, 1, 2, \dots$$

$$f_X(x) = P(X=x) = \frac{e^{-\lambda} \cdot \cancel{\lambda}^x}{x!} \quad x = 0, 1, 2, \dots$$

$$X \sim \text{Poi}(\lambda)$$

$$\sum_{x=0}^{\infty} \frac{e^{-\lambda} \cdot \lambda^x}{x!} = e^{-\lambda} \cdot \boxed{\sum_{x=0}^{\infty} \frac{\lambda^x}{x!}} = e^{-\lambda} \cdot e^{\lambda} = 1$$

6.2 Poisson Distribution, $\text{Poi}(\lambda)$

- Let r.v X be the number of occurrences in a given time interval.
- pmf

$$f(x) = \frac{e^{-\lambda} \lambda^x}{x!}, \quad x = 0, 1, \dots ; \quad 0 < \lambda < \infty$$

- Hint:* $\sum_{x=0}^{\infty} \frac{\lambda^x}{x!} = e^{\lambda}$ (Taylor series expansion)
- $P(X = x)$: `dpois(x,lambda)` in R
- $P(X \leq x)$: `ppois(x,lambda)` in R

- Mean: $E(X) = \lambda$
- Variance: $\text{Var}(X) = \lambda$
- mgf: $M(t) = \exp\{\lambda(e^t - 1)\}$
- λ : average events within a given time interval (unit)

$$\sum_{x=0}^{\infty} \frac{\lambda^x}{x!} = e^\lambda$$

$$E(X) = \sum_{x=0}^{\infty} x f_X(x) = \sum_{x=1}^{\infty} x \cdot \frac{e^{-\lambda} \lambda^x}{x!} = e^{-\lambda} \sum_{x=1}^{\infty} \frac{\lambda^x}{(x-1)!}$$

$y=x-1$

$$= e^{-\lambda} \sum_{y=0}^{\infty} \frac{\lambda^y}{y!} = e^{-\lambda} \cdot \lambda \cdot \sum_{y=0}^{\infty} \frac{\lambda^y}{y!} = e^{-\lambda} \cdot \lambda \cdot e^\lambda = \lambda$$

$$E(X(X-1)) = \sum_{x=0}^{\infty} x(x-1) f_X(x) = \sum_{x=2}^{\infty} x(x-1) \frac{e^{-\lambda} \lambda^x}{x!} = e^{-\lambda} \sum_{x=2}^{\infty} \frac{\lambda^x}{(x-2)!}$$

$y=x-2$

$$= e^{-\lambda} \cdot \sum_{y=0}^{\infty} \frac{\lambda^y}{y!} = e^{-\lambda} \cdot \lambda^2 \cdot e^\lambda = \lambda^2$$

$$\begin{aligned} \text{Var}(X) &= E(X(X-1)) + E(X) - [E(X)]^2 \\ &= \lambda^2 + \lambda - \lambda^2 = \lambda \end{aligned}$$

mgf

1) $E(X) = \lambda \Rightarrow \lambda$: average outcomes in time interval.

$\text{Var}(X) = \lambda$ Data 1). 0. 1. 2. 3. 0. 2. 3. $\sim \text{Poi}$
 average = $\lambda = 1.5 \approx \text{variance}$.

Data 2) 20. 10. 15. 0. 5.

\Rightarrow average < variance $\Rightarrow \begin{cases} E(X) = \lambda \\ \text{Var}(X) = \lambda \end{cases}$ not satisfied
 units \rightarrow $E(X) \approx \text{Var}(X)$. Poi is not good distribution.

Data: 0 0 0 0 0 1 2 3 0 0

$\sim \text{Poi}(?)$

not generally Poi

\rightarrow Zero-inflated Poi

0-truncated Poi $x=1, 2, 3, \dots$

- Mean

$$\begin{aligned}
 E(X) &= \sum_{x=0}^{\infty} x \cdot f(x) = e^{-\lambda} \sum_{x=1}^{\infty} \frac{\lambda^x}{(x-1)!} = \lambda \cdot e^{-\lambda} \sum_{x=1}^{\infty} \frac{\lambda^{x-1}}{(x-1)!} \\
 &= \lambda \cdot e^{-\lambda} \sum_{y=0}^{\infty} \frac{\lambda^y}{y!} = \lambda \cdot e^{-\lambda} \cdot e^{\lambda} = \lambda
 \end{aligned}$$

- Variance

$$\begin{aligned}
 E[X(X-1)] &= \sum_{x=0}^{\infty} x(x-1) \cdot f(x) = e^{-\lambda} \sum_{x=2}^{\infty} \frac{\lambda^x}{(x-2)!} \\
 &= \lambda^2 \cdot e^{-\lambda} \sum_{x=2}^{\infty} \frac{\lambda^{x-2}}{(x-2)!} = \lambda^2 \cdot e^{-\lambda} \sum_{y=0}^{\infty} \frac{\lambda^y}{y!} = \lambda^2 \cdot e^{-\lambda} \cdot e^{\lambda} = \lambda^2
 \end{aligned}$$

$$\text{Var}(X) = E[X(X-1)] - \{E(X)\}^2 + E(X) = \lambda^2 - \lambda^2 + \lambda = \lambda$$

- mgf

$$M(t) = E(e^{tX}) = \sum_{x=0}^{\infty} e^{tx} f(x) = e^{-\lambda} \sum_{x=0}^{\infty} \frac{(\lambda e^t)^x}{x!} = \exp \{ \lambda(e^t - 1) \}$$

$$M^{(1)}(0) = \left. \frac{d}{dt} M(t) \right|_{t=0} = \left. \frac{d}{dt} \exp \{ \lambda(e^t - 1) \} \right|_{t=0} = \lambda e^t \exp \{ \lambda(e^t - 1) \} \Big|_{t=0} = \lambda$$

$$M^{(2)}(0) = \left. \frac{d^2}{dt^2} M(t) \right|_{t=0} = \left. \frac{d^2}{dt^2} \exp \{ \lambda(e^t - 1) \} \right|_{t=0}$$

$$= (\lambda e^t)^2 \exp \{ \lambda(e^t - 1) \} + \lambda e^t \exp \{ \lambda(e^t - 1) \} \Big|_{t=0} = \lambda^2 + \lambda$$

$$\text{Var}(X) = M^{(2)}(0) - \{ M^{(1)}(0) \}^2 = \lambda$$

Example 2.6-1

Let X have a Poisson distribution with a mean of $\lambda = 5$. Compute $P(X \leq 6)$, $P(X > 5)$, and $P(X = 6)$.

Ans R.

$$\begin{aligned} f_X(x) &= \frac{e^{-\lambda} \cdot \lambda^x}{x!} \\ &= \frac{e^{-5} \cdot 5^6}{6!} \end{aligned}$$

Example 2.6-4

In a large city, telephone calls to 911 come on the average of two every 3 minutes. If one assumes an approximate Poisson process, what is the probability of five or more calls arriving in a 9-minute period?

$$2/3 \text{ min.} \quad 5 \uparrow / 9 \text{ min.}$$

average. 2. calls / 3 min
" 2x3=6. calls / 9 min

$X = \# \text{ of calls in 9 min.}$
 $\sim \text{Poi}(\lambda = 6)$

$$P(X \geq 5).$$

6.3 Poisson approximation

- The Poisson r.v may be used as an approximation for a binomial r.v. with parameter (n, p) when n is large and p is small.
- $X \sim B(n, p)$ when $\lambda = np \ll \infty$, $n \rightarrow \infty$, and p is small.

$$\begin{aligned}
 f(x) &= \binom{n}{x} p^x (1-p)^{n-x} = \frac{n!}{x!(n-x)!} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x} \\
 &= \frac{n(n-1)\cdots(n-x+1)}{n^x} \frac{\lambda^x}{x!} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x} \\
 &\rightarrow \frac{e^{-\lambda} \lambda^x}{x!}, \quad \text{since } \lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^n = e^{-\lambda}.
 \end{aligned}$$

$$X \sim \text{Bin}(n, p)$$

$$f_X(x) = P(X=x) = \binom{n}{x} p^x (1-p)^{n-x}$$

$$n=100, \quad p=0.01$$

$$P(X=2) = \binom{100}{2} (0.01)^2 (0.99)^{98}$$

$$P = \frac{\lambda}{n}, \quad \lambda = np = 1$$

$\lambda \uparrow \infty \Rightarrow \frac{e^{-\lambda} \cdot \lambda^x}{x!}$

$P \downarrow \text{small}$

예상 결과가 1인 빈번한 경우에 1을 따른다.

$$P(X=2) \approx \frac{e^{-1} \cdot 1^2}{2!} = \frac{e^{-1}}{2} = \frac{1}{2e}$$

$$\textcircled{1} \quad n \geq 20 \quad \& \quad p \leq 0.05$$

$$\textcircled{2} \quad n \geq 100 \quad \& \quad p \leq 0.01$$

Example 2.6-5

A manufacturer of Christmas tree light bulbs knows that 2% of its bulbs are defective. Assuming independence, we have a binomial distribution with parameters $p = 0.02$ and $n = 100$. Find the approximate probability that a box of 100 of these bulbs contains at most three defective bulbs.

$$X \sim f_X(x) = P(X=x) : \text{pmf}$$

$$F_X(x) = P(X \leq x) : \text{cdf}$$

$$E[g(x)] = \sum_{\forall x} g(x) f_X(x)$$

$$\hookrightarrow E(x) = \mu_x$$

$$\text{Var}(x) = E[(x - \mu_x)^2]$$

$$M_X(t) = E[e^{tx}]$$

Ber(p)

Bin(n,p)

NB(r,p) - 2 r.v.

Geo(p)

Poi(λ)