

Math 3070, Applied Statistics

Section 1

September 13, 2019

Section 3.6

- Poisson Distribution

Poisson Process

Consider a process where events occur at random times, such as

- The arrival times of customers at a store
- Clicks of a Geiger counter exposed to a radioactive material
- Webpage requests on an internet server
- Incoming calls to a customer service center

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Such a process is called a *Poisson process* if the following assumptions hold:

- 1 The mean number of events which occur in a time interval of length t is λt , where λ is a constant, called the *rate* of the Poisson process.
- 2 Events occur only one at a time.
- 3 The number of events which occur in a time interval is independent of the number and timing of past events.

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- By assumption (1), the mean of X is $\mu = \lambda t$.
- Divide the interval $[0, t]$ into equal-width subintervals I_1, \dots, I_n , each of length t/n .
- Let X_k be the number of events which occur in the subinterval I_k , so $X = X_1 + X_2 + \dots + X_n$.
- By assumption (1), $E(X_k) = \lambda \cdot \frac{t}{n} = \frac{\mu}{n}$.
- By assumption (2), if n is large, then the probability of more than one event occurring in any given interval I_k is very small.
- So we may approximate X_k as a Bernoulli random variable with parameter $p = \frac{\mu}{n}$.
- By assumption (3), the random variables X_1, \dots, X_n are independent. Thus X is approximately binomial, $\text{Bin}(n, \frac{\mu}{n})$.

Poisson Distribution

Recall from calculus that

$$\lim_{n \rightarrow \infty} \left(1 + \frac{r}{n}\right)^n = e^r$$

We use this to find the pmf of a Poisson random variable X . We argued that for large n , X is approximately binomial, $\text{Bin}(n, \frac{\mu}{n})$, so

$$\begin{aligned} P(X = x) &\approx \binom{n}{x} (\mu/n)^x (1 - \mu/n)^{n-x} \\ &= \frac{n(n-1) \cdots (n-x+1)}{x!} (\mu/n)^x (1 - \mu/n)^{n-x} \\ &= \frac{n(n-1) \cdots (n-x+1)}{nn \cdots n} (1 - \mu/n)^n (1 - \mu/n)^{-x} \cdot \frac{\mu^x}{x!} \\ &\rightarrow 1 \cdot e^{-\mu} \cdot 1 \cdot \frac{\mu^x}{x!} \quad (\text{as } n \rightarrow \infty) \\ &= \frac{e^{-\mu} \mu^x}{x!} \end{aligned}$$

Poisson Distribution

Given a Poisson process with rate λ , the number X of events which occur in a time interval of length t is a **Poisson** random variable with mean $\mu = \lambda t$. $X \sim \text{Pois}(\mu)$ The possible values of X are $0, 1, 2, \dots$

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Assuming a Poisson process,

$$\begin{aligned} P(X \leq 3) &= P(X = 0) + P(X = 1) + P(X = 2) + P(X = 3) \\ &= \frac{e^{-6} 6^0}{0!} + \frac{e^{-6} 6^1}{1!} + \frac{e^{-6} 6^2}{2!} + \frac{e^{-6} 6^3}{3!} \\ &= e^{-6}(1 + 6 + 18 + 36) = 61e^{-6} \approx .151 \end{aligned}$$

Poisson Distribution, Sum of PMF

Recall from calculus that

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

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Therefore, the pmf $f(x)$ of a Poisson random variable satisfies

$$\begin{aligned}\sum_{x=0}^{\infty} f(x) &= \sum_{x=0}^{\infty} \frac{e^{-\mu} \mu^x}{x!} \\ &= e^{-\mu} \sum_{x=0}^{\infty} \frac{\mu^x}{x!} \\ &= e^{-\mu} e^{\mu} \\ &= 1\end{aligned}$$

which shows that $f(x)$ is a valid pmf.

Mean of Poisson Distribution

We may directly calculate the mean of a Poisson random variable X based on the pmf:

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

This is what we expected based on the definition.

Variance of Poisson Distribution

We can also calculate the variance:

$$\begin{aligned} E(X^2) &= \sum_{x=0}^{\infty} x^2 \cdot \frac{e^{-\mu} \mu^x}{x!} \\ &= \sum_{x=1}^{\infty} x \cdot \frac{e^{-\mu} \mu^x}{(x-1)!} \\ &= \sum_{x=0}^{\infty} (x+1) \frac{e^{-\mu} \mu^{x+1}}{x!} \\ &= \sum_{x=0}^{\infty} x \frac{e^{-\mu} \mu^{x+1}}{x!} + \sum_{x=0}^{\infty} \frac{e^{-\mu} \mu^{x+1}}{x!} \\ &= \sum_{x=1}^{\infty} \frac{e^{-\mu} \mu^{x+1}}{(x-1)!} + \mu \sum_{x=0}^{\infty} \frac{e^{-\mu} \mu^x}{x!} \\ &= \sum_{x=0}^{\infty} \frac{e^{-\mu} \mu^{x+2}}{x!} + \mu = \mu^2 + \mu \end{aligned}$$

$$\text{So } V(X) = E(X^2) - [E(X)]^2 = \mu^2 + \mu - \mu^2 = \mu.$$

Example

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$$\begin{aligned}P(X \geq 3) &= 1 - P(X \leq 2) \\&= 1 - P(X = 0) - P(X = 1) - P(X = 2) \\&= 1 - \frac{e^{-1}1^0}{0!} - \frac{e^{-1}1^1}{1!} - \frac{e^{-1}1^2}{2!} \\&= 1 - e^{-1} \left(1 + 1 + \frac{1}{2} \right) \\&= 1 - \frac{5e^{-1}}{2} \approx .080\end{aligned}$$

Approximating Binomial for Rare Events

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$$\begin{aligned}P(X = x) &= \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x} \\&= \frac{n(n-1)\cdots(n-x+1)}{x!} \left(\frac{\mu}{n}\right)^x \left(1 - \frac{\mu}{n}\right)^{n-x} \\&= \frac{1}{x!} \frac{n^x (1 - \frac{1}{n}) \cdots (1 - \frac{x-1}{n})}{n^x} \mu^x \left(1 - \frac{\mu}{n}\right)^{n-x} \\&\rightarrow \frac{e^{-\mu} \mu^x}{x!}, \quad (\text{as } n \rightarrow \infty)\end{aligned}$$

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Note: n and p do not appear as parameters of the exponential distribution, but can be related by μ .

Approximating Binomial for Rare Events

Consider $E(X)$ and $V(X)$ of $X \sim \text{bin}(n, p)$
as $np \rightarrow \mu$ while $n \rightarrow \infty$ and $p \rightarrow 0$.

$$E(X) = np \rightarrow \mu$$

$$V(X) = np(1 - p) \rightarrow \mu$$

Parameters match in the limit.

Example, Approximating Binomial for Rare Events

The Prussian army studied the likelihood of a soldier being killed by a horse kick. Over **twenty** years, it was observed that 122 soldiers from **10** corps were killed horse kick. The size of each corps is 10,000 soldiers. Approximate the probability that no Prussian soldiers from **two corps** are killed by horse kick in one year.

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Estimate p from the historical data,

$$p \approx \frac{122}{10000 \cdot \text{20} \cdot \text{10}}.$$

Compute μ .

$$\mu = np = \text{2} \cdot \text{10000} \frac{122}{10000 \cdot \text{20} \cdot \text{10}} = \frac{244}{200} = 1.22$$

Estimate with X with a $\exp(1.22)$ random variable.

$$P(X = 0) = \frac{e^{-1.22} 1.22^0}{0!} = e^{-1.22} \approx 0.29523016692$$

Poisson Distribution, Summary

- $X \sim \text{Pois}(\mu)$ mean X is a poisson random variable with rate μ and its PMF is

$$P(X = x) = \frac{e^{-\mu} \mu^x}{x!}$$

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$$E(X) = V(X) = \mu$$

- When modeling number of events in an interval, $\mu = \lambda t$. λ is the rate of occurrence per unit time. Sometimes, μ is call the rate instead of λ . Questions in this class will always assume μ is the mean.
- When approximating the number of rare events in a large population, $\mu = np$.

Midterm September 18, Information

- There is a midterm on September 18th in class. Calculator and notes allowed.
- Study the quizzes, summary slides, and homework. See the Canvas 'Files' tab for information.
- Review on September 16th. Come with questions.
- Reschedule by Wednesday, September 11, if needed. No makeup or late exams.
- One question from this week's material.
- No quiz or homework due on exam weeks. Material is shifted to the later week.