

Ch2

21. If X is a geometric random variable, show that

$$P(X > n + k - 1 | X > n - 1) = P(X > k)$$

In light of the construction of a geometric distribution from a sequence of independent Bernoulli trials, how can this be interpreted so that it is "obvious"?

If $X \sim \text{Geo}(p)$

$$\begin{aligned} \Rightarrow P(X > n-1) &= \sum_{k=n}^{\infty} P(X=k) \\ &= \sum_{k=n}^{\infty} p(1-p)^{k-1} = \frac{p(1-p)^{n-1}}{1-(1-p)} = (1-p)^{n-1} \\ P(X > n+k-1 | X > n-1) &= \frac{P((X > n+k-1) \cap (X > n-1))}{P(X > n-1)} \quad \because k \geq 0 \\ &= \frac{P(X > n+k-1)}{P(X > n-1)} \\ &= \frac{(1-p)^{n+k-1}}{(1-p)^{n-1}} \\ &= (1-p)^k \\ &= P(X > k) \end{aligned}$$

If you have failed $n-1$ times and succeeded $n+k-1$ th times, then you are now in the same situation as if you had started the experiment at the beginning and succeeded at time k . No matter how many times you fail, the next experiment is like starting from scratch.

31. Phone calls are received at a certain residence as a Poisson process with parameter $\lambda = 2$ per hour.
- If Diane takes a 10-min. shower, what is the probability that the phone rings during that time?
 - How long can her shower be if she wishes the probability of receiving no phone calls to be at most .5?

(a) Define $X \equiv$ The number of calls received per 10 mins at a residence.

$$X \sim \text{Poi}(\lambda = 2 \times \frac{1}{6} = \frac{1}{3})$$

$$P(X > 0) = 1 - P(X = 0) = 1 - \frac{e^{-\frac{1}{3}} \frac{1}{3}^0}{0!} = 1 - e^{-\frac{1}{3}}$$

(b) Define $N(t) \equiv$ Number of calls received every t hours at a residence.

$$N(t) \sim \text{Poi}(\lambda = 2t)$$

$$P(N(t) = k) = \frac{(2t)^k e^{-2t}}{k!}$$

$$P(N(t) = 0) = e^{-2t} \leq 0.5$$

$$\Leftrightarrow -2t \leq \ln(0.5)$$

$$\Leftrightarrow t \geq \frac{\ln(0.5)}{-2} \approx 0.3466$$

The maximum shower time is 0.3466 hours.

#48 T is an exponential random variable, and $P(T < 1) = .05$. What is λ ?

$$T \sim \text{Exp}(\lambda) \Rightarrow F(t) = \begin{cases} 1 - e^{-\lambda t}, & t \geq 0 \\ 0, & t < 0 \end{cases}$$

$$0.05 = P(T < 1)$$

$$= F(1)$$

$$= 1 - e^{-\lambda}$$

$$\Rightarrow e^{-\lambda} = 0.95$$

$$\Rightarrow \lambda = -\ln(0.95) \approx 0.0513$$

61. Find the density of cX when X follows a gamma distribution. Show that only λ is affected by such a transformation, which justifies calling λ a scale parameter.

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$$X \sim T(\alpha, \lambda)$$

$$\begin{aligned} \Rightarrow M_X(t) &= E(e^{tX}) \\ &= \left(\frac{\lambda}{\lambda - t} \right)^\alpha, t < \lambda \end{aligned}$$

$$\begin{aligned} M_{cX}(t) &= E(e^{t cX}) \\ &= M_X(ct) \\ &= \left(\frac{\lambda}{\lambda - ct} \right)^\alpha \\ &= \left(\frac{\frac{\lambda}{c}}{\frac{\lambda}{c} - t} \right)^\alpha, ct < \lambda \end{aligned}$$

By uniqueness theorem of moment generating function, $cX \sim T(\alpha, \frac{\lambda}{c})$

Compare the distributions of X and cX , we note that only λ is affected.

Ch 3 22. Consider a Poisson process on the real line, and denote by $N(t_1, t_2)$ the number of events in the interval (t_1, t_2) . If $t_0 < t_1 < t_2$, find the conditional distribution of $N(t_0, t_1)$ given that $N(t_0, t_2) = n$. (Hint: Use the fact that the numbers of events in disjoint subsets are independent.)

Claim: $X_1 \sim P(\lambda_1)$, $X_2 \sim P(\lambda_2)$, $X_1 \perp\!\!\!\perp X_2$

$$\Rightarrow X_1 | X_1 + X_2 = n \sim \text{Bin}(n, \frac{\lambda_1}{\lambda_1 + \lambda_2}) \quad \dots \dots <1>$$

$\therefore X_1 \perp\!\!\!\perp X_2$

$$\begin{aligned} \Rightarrow P(X_1 = k, X_2 = m) &= P(X_1 = k)P(X_2 = m) \\ &= \frac{\lambda_1^k e^{-\lambda_1}}{k!} \frac{\lambda_2^m e^{-\lambda_2}}{m!} \end{aligned}$$

$\therefore X_1 \perp\!\!\!\perp X_2 \quad \therefore X_1 + X_2 \sim \text{Poi}(\lambda_1 + \lambda_2)$ (by LN Ch 1-6, P2-68 note)

$$P(X_1 + X_2 = n) = \frac{(\lambda_1 + \lambda_2)^n e^{-(\lambda_1 + \lambda_2)}}{n!}$$

$$P(X_1 = k, X_1 + X_2 = n) = P(X_1 = k)P(X_2 = n - k) = \frac{\lambda_1^k e^{-\lambda_1}}{k!} \frac{\lambda_2^{n-k} e^{-\lambda_2}}{(n-k)!}$$

$$\begin{aligned} P(X_1 = k | X_1 + X_2 = n) &= \frac{P(X_1 = k, X_1 + X_2 = n)}{P(X_1 + X_2 = n)} \\ &= \frac{\frac{\lambda_1^k e^{-\lambda_1}}{k!} \frac{\lambda_2^{n-k} e^{-\lambda_2}}{(n-k)!}}{\frac{(\lambda_1 + \lambda_2)^n e^{-(\lambda_1 + \lambda_2)}}{n!}} \\ &= \frac{n!}{k!(n-k)!} \frac{\lambda_1^k \lambda_2^{n-k}}{(\lambda_1 + \lambda_2)^n} = \binom{n}{k} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^k \left(\frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{n-k} \end{aligned}$$

$$P(X_1=k | X_1+X_2=n) = \binom{n}{k} \left(\frac{\lambda_1}{\lambda_1+\lambda_2}\right)^k \left(\frac{\lambda_2}{\lambda_1+\lambda_2}\right)^{n-k} \text{ is pdf of } \text{Bin}(n, \frac{\lambda_1}{\lambda_1+\lambda_2})$$

$$\text{Thus } X_1 | X_1+X_2=n \sim \text{Bin}(n, \frac{\lambda_1}{\lambda_1+\lambda_2})$$

$$N(t_1, t_2) \sim \text{Poi}(\lambda(t_2 - t_1))$$

$$N(t_0, t_1) \sim \text{Poi}(\lambda(t_1 - t_0))$$

$$\therefore (t_1, t_2) \cap (t_0, t_1) = \emptyset \quad \therefore N(t_1, t_2) \perp N(t_0, t_1)$$

$$\text{By <1> we have } N(t_0, t_1) | N(t_0, t_2)=n \sim \text{Bin}(n, \frac{t_1 - t_0}{t_2 - t_0})$$

#26 Let P have a uniform distribution on $[0, 1]$, and, conditional on $P = p$, let X have a Bernoulli distribution with parameter p . Find the conditional distribution of P given X .

$$P \sim U(0, 1), \quad X|P=p \sim \text{Ber}(p)$$

$$\begin{aligned} f_{P,X}(p, x) &= f_P(p) P(X=x|P=p) \\ &= 1 \times p^x (1-p)^{1-x}, \quad p \in (0, 1), \quad x=0 \text{ or } 1 \end{aligned}$$

$$\begin{aligned} f_X(x) &= \int_0^1 f_{P,X}(p, x) dp \\ &= \int_0^1 p^x (1-p)^{1-x} dp \\ &= \frac{\Gamma(x+1) \Gamma(2-x)}{\Gamma(3)} \\ &= \frac{x! (1-x)!}{2!} \\ &= \frac{1}{2}, \quad x=0 \text{ or } 1 \end{aligned}$$

$$\begin{aligned} f_{P|X}(p|x) &= \frac{f_{P,X}(p, x)}{f_X(x)} \\ &= \frac{p^x (1-p)^{1-x}}{\frac{1}{2}} \\ &= 2 p^x (1-p)^{1-x}, \quad p \in (0, 1) \end{aligned}$$

Thus $P|X=x \sim \text{Beta}(x+1, 2-x)$, for $x=0$ or 1 .

CH4

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Find $E[1/(X+1)]$, where X is a Poisson random variable.

$$\begin{aligned}
 \text{pf. } E\left[\frac{1}{X+1}\right] &= \sum_{k=0}^{\infty} \frac{1}{k+1} \cdot \frac{\lambda^k e^{-\lambda}}{k!} = 1 \quad \text{because prob. add up to 1} \\
 &= \sum_{k=0}^{\infty} \frac{\lambda^{k+1} e^{-\lambda}}{(k+1)!} \cdot \frac{1}{\lambda} \\
 &= \frac{1}{\lambda} \left(\sum_{k=1}^{\infty} \frac{\lambda^k e^{-\lambda}}{k!} + \frac{\lambda^0 e^{-\lambda}}{0!} \right) - \frac{1}{\lambda} \cdot \frac{\lambda^0 e^{-\lambda}}{0!} \\
 &= \frac{1}{\lambda} - \frac{1}{\lambda} (e^{-\lambda}) = \frac{e^{-\lambda}}{\lambda} (e^{\lambda} - 1)
 \end{aligned}$$

42. Let X be an exponential random variable with standard deviation σ . Find $P(|X - E(X)| > k\sigma)$ for $k = 2, 3, 4$, and compare the results to the bounds from Chebyshev's inequality.

$$\text{pf. } P(|X - E(X)| > k\sigma) = P(X - E(X) < -k\sigma) + P(X - E(X) > k\sigma)$$

and since for exponential v.v., $\mu = \sigma$

$$\begin{aligned}
 \text{we have } P(X - E(X) < -k\sigma) &= P(X < -(k+1)\sigma) = 0 \quad \text{for } k=2,3,4 \\
 P(|X - E(X)| > k\sigma) &= P(X > (1+k)\sigma) \\
 &= e^{-\frac{(1+k)\sigma}{\sigma}} = e^{-(1+k)}
 \end{aligned}$$

using Chebyshev's inequality

$$P(|X - E(X)| > k\sigma) \leq \frac{1}{k^2}$$

	$k=2$	$k=3$	$k=4$
Chebyshev's	$< \frac{1}{4}$	$< \frac{1}{9}$	$< \frac{1}{16}$
Real Value	$e^{-3} \approx 0.05$	$e^{-4} \approx 0.018$	$e^{-5} \approx 0.007$

76. Let the point (X, Y) be uniformly distributed over the half disk $x^2 + y^2 \leq 1$, where $y \geq 0$. If you observe X , what is the best prediction for Y ? If you observe Y , what is the best prediction for X ? For both questions, "best" means having the minimum mean squared error.

pf. The best guess will be $E[Y|X]$
(see example 33 in course note)

Since unif. dist. and area of half disk be $\frac{\pi}{2}$,

$$f(x, y) = \frac{2}{\pi} \text{ for } x^2 + y^2 \leq 1 \text{ and } y \geq 0$$

$$f_X(x) = \int_0^{\sqrt{1-x^2}} \frac{2}{\pi} dy = \frac{2}{\pi} \sqrt{1-x^2} \quad -1 \leq x \leq 1$$

$$f_Y(y) = \int_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} \frac{2}{\pi} dx = \frac{4}{\pi} \sqrt{1-y^2} \quad 0 \leq y \leq 1$$

$$E[Y|X] = \int y f(y|x) dy = \int_0^{\sqrt{1-x^2}} y \cdot \frac{f(x, y)}{f_X(x)} dy$$

$$= \frac{\pi}{2\sqrt{1-x^2}} \cdot \frac{2}{\pi} \cdot \left(\frac{1}{2} y^2 \Big|_0^{\sqrt{1-x^2}} \right)$$

$$= \frac{1}{\sqrt{1-x^2}} \cdot \frac{1-x^2}{2} = \frac{\sqrt{1-x^2}}{2}$$

$$E[X|Y] = \int x f(x|y) dx = \int_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} x \cdot \frac{f(x, y)}{f_Y(y)} dx$$

$$= \int_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} x \cdot \frac{2}{\pi} \cdot \frac{\pi}{4\sqrt{1-y^2}} dx$$

$$= \frac{1}{2\sqrt{1-y^2}} \left(\frac{1}{2} x^2 \right) \Big|_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}}$$

$$= 0$$

80. Let X be a continuous random variable with density function $f(x) = 2x$, $0 \leq x \leq 1$. Find the moment-generating function of X , $M(t)$, and verify that $E(X) = M'(0)$ and that $E(X^2) = M''(0)$.

pf.

$$M(t) = E[e^{tx}] = \int_0^1 e^{tx} \cdot 2x \, dx$$

$$= \frac{1}{t} \int_0^1 2x \, de^{tx} = \frac{1}{t} \left(2xe^{tx} \Big|_0^1 - 2 \int_0^1 e^{tx} \, dx \right)$$

$$= \frac{1}{t} \left(2e^t - \frac{2}{t}(e^t - 1) \right)$$

$$= \frac{2te^t - 2e^t + 2}{t^2}$$

$$M'(t) = \frac{2t^2e^t - 4te^t + 4e^t - 4}{t^3}$$

Since $M'(t)$ is $\frac{0}{0}$ when $t=0$, we use L'Hospital

$$\lim_{t \rightarrow 0} M'(t) = \lim_{t \rightarrow 0} \frac{2t^2e^t - 4te^t + 4e^t - 4}{t^3} \quad \left(\frac{0}{0} \right)$$

$$\stackrel{\text{LH}}{=} \lim_{t \rightarrow 0} \frac{4te^t + 2t^2e^t - 4e^t - 4te^t + 4e^t}{3t^2}$$

$$= \lim_{t \rightarrow 0} \frac{2t^2e^t}{3t^2} = \frac{2}{3}$$

Same

$$E[X] = \int_0^1 x \cdot 2x \, dx = \frac{2}{3}$$

$$M''(t) = \frac{2t^3e^t - 6t^2e^t + 12te^t - 12e^t + 12}{t^4}$$

Since $M''(t)$ is $\frac{0}{0}$ when $t=0$, use L'Hos

$$\lim_{t \rightarrow 0} M''(t) = \lim_{t \rightarrow 0} \frac{2t^3e^t - 6t^2e^t + 12te^t - 12e^t + 12}{t^4} \quad \left(\frac{0}{0} \right)$$

$$\stackrel{\text{LH}}{=} \lim_{t \rightarrow 0} \frac{6t^2e^t + 2t^2e^t - 12te^t - 6t^2e^t + 12e^t + 12te^t - 12e^t}{4t^3}$$

$$= \lim_{t \rightarrow 0} \frac{2t^3 e^t}{4t^3} = \frac{1}{2}$$

$$E[X^2] = \int_0^1 x^2 \cdot 2x \, dx = \frac{2}{4} x^4 \Big|_0^1 = \frac{1}{2}$$

Same

□

92. Suppose that Θ is a random variable that follows a gamma distribution with parameters λ and α , where α is an integer, and suppose that, conditional on Θ , X follows a Poisson distribution with parameter Θ . Find the unconditional distribution of $\alpha + X$. (Hint: Find the mgf by using iterated conditional expectations.)

$$\text{pf. } P(X|\Theta=\theta) = \frac{\theta^x}{x!} e^{-\theta}$$

$$E[e^{tx} | \Theta=\theta] = \sum_{x=0}^{\infty} e^{tx} \cdot \frac{\theta^x e^{-\theta}}{x!}$$

$$= e^{-\theta} \sum_{x=0}^{\infty} \frac{(e^t \theta)^x}{x!}$$

$$= e^{-\theta} \cdot e^{+(e^t \theta)} \cdot \underbrace{\sum_{x=0}^{\infty} \frac{(e^t \theta)^x e^{-(e^t \theta)}}{x!}}_{=1}$$

$$= e^{\theta(e^t - 1)}$$

$$\begin{aligned} \mu_X(t) &= E[e^{tx}] = E[E[e^{tx} | \Theta=\theta]] = E[e^{\theta(e^t - 1)}] \\ &= \mu_{\Theta}(e^t - 1) \end{aligned}$$

$$\mu_{\Theta}(t) = \left(\frac{\lambda}{\lambda - t} \right)^{\alpha}, \quad t < \lambda \quad (\text{gamma dist. mgt.})$$

$$\Rightarrow \mu_{\Theta}(e^t - 1) = \left(\frac{\lambda}{\lambda - e^t + 1} \right)^{\alpha} = \mu_X(t)$$

$$\mu_{\alpha+X}(t) = e^{t\alpha} \cdot \left(\frac{\lambda}{\lambda - e^t + 1} \right)^{\alpha} = \left(\frac{\lambda e^t}{\lambda - e^t + 1} \right)^{\alpha}$$

$$= \left(\frac{\frac{\lambda}{\lambda+1} e^t}{1 - \frac{e^t}{\lambda+1}} \right)^\alpha$$

by uniqueness thm we have $\alpha + X \sim NB(\alpha, \frac{\lambda}{\lambda+1})$

→ Negative Binomial MGF

$$\left(\frac{p e^t}{1 - (1-p)e^t} \right)^r \quad \text{for } t < -\ln(1-p) \quad \text{see p 2-63}$$

so we have $p = \frac{\lambda}{\lambda+1} \quad r = \alpha$

100. If X is uniform on $[10, 20]$, find the approximate and exact mean and variance of $Y = 1/X$, and compare them.

pf.

Exact mean and variance

$$E[Y] = \int_{10}^{20} \frac{1}{x} \cdot \frac{1}{10} dx = \frac{1}{10} \ln 2 \approx 0.069$$

$$\begin{aligned} \text{Var}(Y) &= E[Y^2] - (E[Y])^2 = \int_{10}^{20} \frac{1}{x^2} \cdot \frac{1}{10} dx - \left(\frac{\ln 2}{10} \right)^2 \\ &= \frac{1}{10} \left(-\frac{1}{x} \right) \Big|_{10}^{20} - \frac{(\ln 2)^2}{100} \\ &= \frac{1}{20} - \frac{(\ln 2)^2}{100} \approx 0.0002 \end{aligned}$$

Approx. method

$$E[X] = 15 \quad \text{Var}(X) = \frac{10^2}{12} = \frac{25}{3}$$

$$\text{let } g(x) = \frac{1}{x} \quad g'(x) = -\frac{1}{x^2} \quad g''(x) = \frac{2}{x^3}$$

by Thm 3.9 (2-56)

$$E\left[\frac{1}{x}\right] \approx g(E(X)) = \frac{1}{15} \approx 0.06\bar{6}$$

$$\text{Var}\left(\frac{1}{x}\right) \approx \text{Var}(X) \cdot [g'(E[X])]^2$$

$$= \frac{25}{3} \cdot \frac{1}{(15)^4} \approx 0.000165$$

Another more accurate approximate method

$$E[g(x)] \approx g(\mu_x) + \frac{1}{2} \sigma_x^2 g''(\mu_x)$$
$$= \frac{1}{15} + \frac{1}{2} \cdot \frac{25}{3} \cdot \frac{2}{15^3} \approx 0.06914$$