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 AMATH 561

PROBLEM SET 6

1. Let $X \sim \text{Binomial}(n, U)$, where $U \sim \text{Uniform}((0, 1))$. What is the probability generating function $G_X(s)$ of X ? What is $P(X = k)$ for $k \in \{0, 1, 2, \dots, n\}$?

Solution: The probability mass function for $X \sim \text{Binomial}(n, U)$, is given by

$$f_X(x) = \binom{n}{x} U^x (1 - U)^{n-x} \quad \text{for } x = 0, 1, 2, \dots, n$$

And the density for the Uniform distribution is

$$g_X(x) = \frac{1}{b-a} = \frac{1}{1-0} = 1 \quad \text{for } x \in (0, 1)$$

Then $G_X(s)$ is

$$\begin{aligned} G_X(s) &= E(s^X) \\ &= E(E(s^X | U)) \\ &= E\left(\sum_{x=0}^n \binom{n}{x} U^x (1 - U)^{n-x} s^x\right) \\ &= E\left(\sum_{x=0}^n \binom{n}{x} (Us)^x (1 - U)^{n-x}\right) \\ &= E((Us + 1 - U)^n) \\ &= \int_0^1 (us + 1 - u)^n du \\ &= \left. \frac{(us + 1 - u)^{n+1}}{(n+1)(s-1)} \right|_0^1 \\ &= \frac{(1s + 1 - 1)^{n+1}}{(n+1)(s-1)} - \frac{(0s + 1 - 0)^{n+1}}{(n+1)(s-1)} \\ &= \frac{s^{n+1}}{(n+1)(s-1)} - \frac{1^{n+1}}{(n+1)(s-1)} \\ &= \frac{s^{n+1} - 1}{(n+1)(s-1)} \end{aligned}$$

Now to calculate $P(X = k)$, notice

$$\begin{aligned}
 G_X(s) &= \frac{s^{n+1} - 1}{(n+1)(s-1)} = \frac{1}{n+1} \frac{s^{n+1} - 1}{s-1} \\
 &= \frac{1}{n+1} \frac{1 - s^{n+1}}{1-s} \\
 &= \sum_{k=0}^n \frac{1}{n+1} s^k \\
 &= \sum_{k=0}^n P(X = k) s^k \\
 &= G_X(s)
 \end{aligned}$$

Therefore,

$$P(X = k) = \frac{1}{n+1} \quad \text{for all } k \in \{0, 1, 2, \dots, n\}.$$

□

2. Consider a branching process with immigration

$$Z_0 = 1, \quad Z_{n+1} = \sum_{i=1}^{Z_n} \xi_i^{n+1} + Y_{n+1},$$

where the (ξ_i^{n+1}) are iid with common distribution ξ , the (Y_n) are iid with common distribution Y and the (ξ_i^{n+1}) and (Y_{n+1}) are independent. What is $G_{Z_{n+1}}(s)$ in terms of $G_{Z_n}(s)$, $G_\xi(s)$ and $G_Y(s)$? Write $G_{Z_2}(s)$ explicitly in terms of $G_\xi(s)$ and $G_Y(s)$.

Solution:

We can write the generating function $G_{Z_{n+1}}(s)$ as follows

$$\begin{aligned} G_{Z_{n+1}}(s) &= G_{\sum_{i=1}^{Z_n} \xi_i^{n+1} + Y_{n+1}}(s) \\ &= G_{\xi_1^{n+1} + \xi_2^{n+1} + \xi_3^{n+1} + \dots + \xi_{Z_n}^{n+1} + Y_{n+1}}(s) \\ &= G_{\xi_1^{n+1} + \xi_2^{n+1} + \xi_3^{n+1} + \dots + \xi_{Z_n}^{n+1}} G_{Y_{n+1}}(s) \\ &= G_{Z_n}(G_\xi(s)) G_{Y_{n+1}}(s). \end{aligned}$$

The second to third equality comes from the fact that the Y_{n+1} and ξ_i^{n+1} are independent. Finally, the last equality comes from an application of the Theorem 3 from Lecture 15.

Next to calculate $G_{Z_2}(s)$ explicitly we get

$$\begin{aligned} G_{Z_2}(s) &= G_{Z_1}(G_\xi(s)) G_Y(s) \\ &= \left(G_\xi(G_\xi(s)) G_Y(G_\xi(s)) \right) G_Y(s). \end{aligned}$$

□

3. (a) Let X be exponentially distributed with parameter λ . Show by elementary integration (not complex integration) that $E(e^{itX}) = \lambda/(\lambda - it)$.

Solution:

We can begin by looking directly at the expectation we want to calculate

$$\begin{aligned} E(e^{itX}) &= \int_{\Omega} e^{itX} dP = \int_{\mathbb{R}} e^{itx} \lambda e^{-\lambda x} dx \\ &= \int_0^{\infty} e^{itx} \lambda e^{-\lambda x} dx \\ &= \int_0^{\infty} \lambda e^{(it-\lambda)x} dx \\ &= \int_0^{\infty} \lambda e^{-(\lambda-it)x} dx \end{aligned}$$

Notice this integral is off by a scale factor to the density of an exponentially distributed random variable with parameter $\lambda - it$. Additionally, we know the integral of a probability density function is equal to 1, therefore,

$$\begin{aligned} \int_0^{\infty} (\lambda - it) e^{-(\lambda-it)x} dx &= 1 \\ \frac{\lambda}{\lambda - it} \int_0^{\infty} (\lambda - it) e^{-(\lambda-it)x} dx &= \frac{\lambda}{\lambda - it} \\ \int_0^{\infty} \frac{\lambda}{\lambda - it} (\lambda - it) e^{-(\lambda-it)x} dx &= \frac{\lambda}{\lambda - it} \\ \int_0^{\infty} \lambda e^{-(\lambda-it)x} dx &= \frac{\lambda}{\lambda - it}. \end{aligned}$$

Which is indeed the integral we wanted to compute. □

(b) Find the characteristic function of the density function $f(x) = \frac{1}{2}e^{-|x|}$ for $x \in \mathbb{R}$.

Solution: The characteristic function is (skipping directly to the change of variable form of the expectation)

$$\begin{aligned} \phi_X(t) &= E(e^{itX}) = \int_{-\infty}^{\infty} e^{itx} \frac{1}{2} e^{-|x|} dx \\ &= \frac{1}{2} \int_{-\infty}^{\infty} e^{itx - |x|} dx. \end{aligned}$$

Let's split up the integral into cases in order to handle the absolute value. Then we have

$$\begin{aligned}
 \phi_X(t) &= \frac{1}{2} \int_{-\infty}^{\infty} e^{itx-|x|} dx = \frac{1}{2} \left[\int_{-\infty}^0 e^{itx-|x|} dx + \int_0^{\infty} e^{itx-|x|} dx \right] \\
 &= \frac{1}{2} \left[\int_{-\infty}^0 e^{itx+x} dx + \int_0^{\infty} e^{itx-x} dx \right] \\
 &= \frac{1}{2} \left[\int_{-\infty}^0 e^{(it+1)x} dx + \int_0^{\infty} e^{-(1-it)x} dx \right] \\
 &= \frac{1}{2} \left[\left(\frac{1}{it+1} e^{(it+1)x} \right) \Big|_{-\infty}^0 + \left(-\frac{1}{1-it} e^{-(1-it)x} \right) \Big|_0^{\infty} \right].
 \end{aligned}$$

Now, as we evaluate these expressions at their respective bounds of integration, notice the terms evaluated at $-\infty$ and at ∞ in the left and right integrals both go to 0. Then we have

$$\begin{aligned}
 \phi_X(t) &= \frac{1}{2} \left[\left(\frac{1}{it+1} e^{(it+1)0} - \frac{1}{it+1} e^{(it+1)(-\infty)} \right) + \left(-\frac{1}{1-it} e^{-(1-it)\infty} + \frac{1}{1-it} e^{-(1-it)0} \right) \right] \\
 &= \frac{1}{2} \left[\frac{1}{it+1} + \frac{1}{1-it} \right] \\
 &= \frac{1}{2} \left(\frac{1-it+it+1}{(it+1)(1-it)} \right) \\
 &= \frac{1}{2} \left(\frac{2}{1-i^2 t^2} \right) \\
 &= \frac{1}{1+t^2}.
 \end{aligned}$$

Hence, the characteristic function for a random variable with its density given by

$f(x) = \frac{1}{2}e^{-|x|}$ for $x \in \mathbb{R}$ is

$$\phi_X(t) = \frac{1}{1+t^2}.$$

4. A coin is tossed repeatedly, with heads turning up with probability p on each toss. Let N be the minimum number of tosses required to obtain k heads. Show that, as $p \rightarrow 0$, the distribution function of $2Np$ converges to that of a gamma distribution. Note that, if $X \sim \Gamma(\lambda, r)$ then

$$f_X(x) = \frac{1}{\Gamma(r)} \lambda^r x^{r-1} e^{-\lambda x} 1_{x \geq 0}.$$

Solution:

Notice our scenario of N being the minimum number of tosses required to obtain k heads (call a heads a “success”) is related to the negative binomial distribution. This distribution describes the probability of ℓ failures occurring before k successes, the p.m.f. is given by

$$P(N = n) = \binom{n-1}{k-1} p^k (1-p)^{n-k}.$$

To further connect this distribution to our scenario $N = \ell + k$. I think...

$$\begin{aligned} \lim_{p \rightarrow 0} \binom{n-1}{k-1} p^k (1-p)^{n-k} &= \lim_{p \rightarrow 0} \frac{(n-1)!}{(n-k)!(k-1)!} p^k (1-p)^{n-k} \\ &= \lim_{p \rightarrow 0} \frac{(n-1)!}{(n-k)!(k-1)!} p^k \left[(1-p)^{1/p} \right]^{(n-k)p} \\ &= \frac{(n-1)!}{(n-k)!(k-1)!} p^k e^{(n-k)p} \\ &= \frac{\frac{(n-1)!}{(k-1)!}}{\Gamma(n-k+1)} p^k e^{(n-k)p} \dots \end{aligned}$$

TODO: Come back to this, you’re so close We wish to show that as $p \rightarrow 0$ the distribution of $2Np$ converges to that of a gamma distribution. Recall that the density of the gamma distribution is