

Cálculo de Probabilidades II
Respuestas Tema 2

1. Si

$$X_i \sim Ber(p) \quad y \quad Y = \sum_{i=1}^n x_i$$

entonces,

$$\begin{aligned} M_Y(t) &= E(E^{ty}) \\ &= E(e^{tx_1} e^{tx_2} \dots e^{tx_n}) \\ &= \prod_{i=1}^n E(e^{tx_i}) \\ &= (e^t p + (1-p)^n) \end{aligned}$$

$$Y \sim Bin(n, p)$$

2. Sea

$$X \sim \chi_k^2$$

Y

$$Y = Z_1^2 + Z_2^2 + \dots + Z_n^2$$

$$\begin{aligned} \rightarrow M_Y(t) &= E(e^{ty}) \\ &= \prod_{i=1}^n e^{tk(z_i)^2} \\ &= (1 - 2t)^{nk/2} \end{aligned}$$

$$\rightarrow Y \sim \chi_{\sum n_i}^2$$

3. Sea $Y = aX + b$

$$\begin{aligned} \rightarrow M_Y t &= E(e^{ty}) \\ &= E(e^{t(aX+b)}) \\ &= e^{tb} M_X(at) \end{aligned}$$

4. Sea

$$\begin{aligned} f(x; \mu, \sigma) &= \frac{1}{x\sigma\sqrt{2\pi}} e^{-(\ln x - \mu)^2 / 2\sigma^2} \\ E(X) &= e^{\mu + \sigma^2/2} \\ Var(X) &= e^{2\mu + 2\sigma^2} - e^{2\mu + \sigma^2} \end{aligned}$$

5. Sea $X \sim N(0, 1)$, p.d.

$$Z = X^2 \sim \chi_{(1)}^2$$

Dem.

$$\begin{aligned} M_Z(t) &= \int_0^\infty e^{-x^2(t-\frac{1}{2})} \frac{1}{\sqrt{2\pi}} dx \\ &= \frac{1}{t - \frac{1}{2}} \\ &\rightarrow Z = X^2 \sim \chi_{(1)}^2 \end{aligned}$$

6. Sean

$$\begin{aligned} X_i &\sim Ga(\alpha, 1) \\ Y &= \sum_{i=1}^n X_i \end{aligned}$$

(a) Con $n = 2$

$$\begin{aligned} f_{X_1, X_2}(y) &= \int_{-\infty}^\infty f_{X_1}(x_1) f_{X_2}(y - x_1) dx_1 \\ &= \int_0^y f_{X_1}(x_1) f_{X_2}(y - x_1) dx_1 \\ &= \int_0^y \frac{x_1^{\alpha-1} e^{-x_1}}{\Gamma(\alpha)} \frac{(y - x_1)^{\alpha-1} e^{-(y-x_1)}}{\Gamma(\alpha)} dx_1 \end{aligned}$$

Sea

$$x_1 = yt \quad dx_1 = y dt$$

y sustituyendo en la integral anterior,

$$\begin{aligned} f_{X_1, X_2}(y) &= e^{-y} y^{\alpha+\alpha-1} \int_0^1 yt^{\alpha-1} (1-t)^{\alpha-1} \frac{1}{\Gamma(\alpha)\Gamma(\alpha)} dt \\ &= \frac{e^{-y} y^{2\alpha-1}}{\Gamma(2\alpha)} \sim Ga(2\alpha, 1) \end{aligned}$$

(b) Sea

$$Y \sim Ga(\alpha, n) \quad y \quad W = 2nY.$$

$$\begin{aligned} P(W \leq w) &= P(2nY \leq w) \\ &= P\left(Y \leq \frac{w}{2n}\right) \\ &= F_Y\left(\frac{w}{2n}\right) \end{aligned}$$

$$\begin{aligned} \rightarrow f_W(w) &= \frac{1}{2n} f_Y\left(\frac{w}{2n}\right) \\ &= \frac{w^{\alpha-1}}{\Gamma(\alpha)} e^{-w/2} \frac{1}{2^\alpha} \sim \chi_{2\alpha}^2 \end{aligned}$$

(c) Sea

$$X \sim Ga(\alpha, 1) \quad y \quad Y = \frac{X}{n}$$

$$\begin{aligned} P(Y \leq y) &= P\left(\frac{X}{n} \leq y\right) \\ &= P(X \leq ny) \\ &= F_X(ny) \end{aligned}$$

$$\begin{aligned} \rightarrow f_Y(y) &= n * f_X(ny) \\ &= \frac{1}{\Gamma(\alpha)} e^{-ny} (ny)^{\alpha-1} n^\alpha \sim Ga(\alpha, n) \end{aligned}$$

7. Sea

$$Y = \sum_{i=1}^n a_i X_i$$

P.D.

$$Y \sim N\left(\sum_{i=1}^n a_i \mu_i, \sum_{i=1}^n (a_i \sigma_i)^2\right)$$

Sea

$$u_i = a_i x_i$$

$$\begin{aligned} M_U(t) &= E(e^{ta_i x_i}) \\ &= \exp\left(ta_i \mu_i + \frac{(ta_i \sigma_i)^2}{2}\right) \end{aligned}$$

Sea

$$Y = \sum_{i=1}^n u_i$$

$$\begin{aligned} M_Y(t) &= E(e^{\sum_{i=1}^n t u_i}) \\ &= \prod_{i=1}^n u_i E(e^{t u_i}) \\ &= \exp\left(\sum_{i=1}^n ta_i \mu_i + \sum_{i=1}^n \frac{(ta_i \sigma_i)^2}{2}\right) \end{aligned}$$

$$\rightarrow Y \sim N\left(\sum_{i=1}^n a_i \mu_i, \sqrt{\sum_{i=1}^n (a_i \sigma_i)^2}\right)$$

8. Sea $f_X(x)$ y $Y = aX + b$

$$\begin{aligned} F_Y &= P(Y \leq y) \\ &= P(aX + b \leq y) \\ &= P\left(X \leq \frac{y-b}{a}\right) \\ \rightarrow f_Y &= \frac{d}{dy} F_X\left(\frac{y-b}{a}\right) \end{aligned}$$

9. Sea

$$Y = e^x \quad X \sim N(\mu, \sigma^2)$$

$$\begin{aligned} P(Y \leq y) &= P(e^x \leq y) \\ &= P(X \leq \ln(y)) \\ &= F_X(\ln(y)) \\ &= P\left(Z \leq \frac{\ln(y) - \mu}{\sigma}\right) \\ &= \Phi\left(\frac{\ln(y) - \mu}{\sigma}\right) \end{aligned}$$

Derivando obtenemos

$$\varphi_Y(y)$$

10. Sean

(a)

$$Y = \left(\frac{x-b}{a}\right)^\beta \quad y \quad X \sim Weibull(b, a, \beta)$$

$$\begin{aligned} P(Y \leq y) &= P\left(\left(\frac{x-b}{a}\right)^\beta \leq y\right) \\ &= P(X \leq ay^{\frac{1}{\beta}} + b) \\ &= \int_0^{ay^{\frac{1}{\beta}} + b} \frac{\beta}{a} \left(\frac{x}{a}\right)^{\beta-1} e^{-(x/a)^\beta} dx, \quad \text{Sea } \beta = 1 \quad yb = 0 \\ &= 1 - e^{-y} \end{aligned}$$

$$\rightarrow Y \sim Exp(1)$$

(b) Sean

$$Y = \left(\frac{x-b}{a}\right)^\beta \quad yX = ay^{\frac{1}{\beta}} + b$$

$$\begin{aligned} P(X \leq x) &= P(ay^{1/\beta} + b \leq x) \\ &= P(Y \leq \left(\frac{x-b}{a}\right)^\beta) \\ &= 1 - \exp\left(-\left(\frac{x-b}{a}\right)^\beta\right) \end{aligned}$$

$$\begin{aligned} \rightarrow f_X &= \frac{\beta}{a} \left(\frac{x-b}{a}\right)^{\beta-1} \exp\left(-\left(\frac{x-b}{a}\right)^\beta\right) \\ X &\sim \text{Weibull}(a, b, \beta) \end{aligned}$$

11. Sean

$$X \sim \text{CauchyEstandar} \quad W = \frac{1}{X}$$

$$\begin{aligned} P(W \leq w) &= P\left(\frac{1}{X} \leq w\right) \\ &= P\left(X \geq \frac{1}{w}\right) \\ &= 1 - \left(\frac{1}{2} + \frac{\arctan\frac{1}{w}}{\pi}\right) \end{aligned}$$

$$f_W = \frac{1}{\pi(w^2 + 1)}$$

$$W \sim \text{CauchyEstandar}$$

12. Sean $X \sim U(-1, 1)$ $W = |X|$

$$P(W \leq w) = P(-w \leq X \leq w) = w$$

$$\rightarrow f_W = 1, \quad 0 \leq w \leq 1$$

$$\text{Sea } Y = X^2 + 1$$

$$P(Y \leq y) = P(-\sqrt{y-1} \leq X \leq \sqrt{y-1}) = \sqrt{y-1}$$

$$f_Y = \frac{1}{2} \left(\frac{1}{\sqrt{y-1}}\right)$$

$$\text{Sea } Z = \frac{1}{X+1}$$

$$P(Z \leq z) = P\left(X \geq \frac{1}{z} - 1\right) = \frac{z-1}{2z}$$

13. Sean

$$X \sim \text{exp}(1) \quad Y = \ln(X)$$

$$P(Y \leq y) = P(X \leq e^y) = 1 - e^{-e^y}$$
$$f_Y = \text{exp}(-e^y + y)$$

14. Sean

$$X \sim U(0, 1) \quad Y = e^X$$

$$P(Y \leq y) = P(X \leq \ln(y)) = \ln(y)$$
$$\rightarrow f_Y = \frac{1}{y} I_{[1, e]}(y)$$

15. Sea

$$\theta \sim U\left(\frac{-\pi}{2}, \frac{\pi}{2}\right) \quad R = A \sin(\theta)$$

$$P(R \leq r) = P(\theta \leq \arcsin\left(\frac{r}{A}\right)) = \frac{\arcsin\left(\frac{r}{A}\right)}{\pi}$$
$$\rightarrow f_R(r) = \frac{1}{\pi A} \left(\frac{1}{\sqrt{1 - \left(\frac{r}{A}\right)^2}} \right)$$

16. Sean

$$X \sim U(0, 1) \quad Y = -\ln(X)$$

$$P(Y \leq y) = P(X \geq \frac{1}{e^y}) = 1 - \frac{1}{e^y}$$

$$\rightarrow f_Y(y) = e^{-y}$$
$$\rightarrow Y \sim \text{Exp}(1)$$

17. Sea

$$Long \sim N(3.25, 0.05^2) \quad L_1 + L_2 \sim N(6.50, .005)$$

$$\rightarrow P(L_1 + L_2 \leq 6.60) = P(Z \leq \frac{6.60 - 6.50}{\sqrt{.005}})$$
$$= \phi(1.4142)$$

18. Si $X_1, X_2, X_3, \dots, X_n$ son v.a.i.i.d.

$$Y_n = \max(X_1, X_2, \dots, X_n)$$
$$p.d.f_{Y_n}(y) = n(F_X(y))^{n-1} f_X(y)$$

Dem.

$$\begin{aligned}F_Y &= P(Y_n \leq y) \\&= P(\max(X_1, X_2, \dots, X_n) \leq y) \\&= \prod_{i=1}^n P(X_i \leq y) \\&= (F_X(y))^n\end{aligned}$$

$$\rightarrow f_Y(y) = n(F_X(y))^{n-1} f_X(y)$$

19. Si

$$X \sim U(0, 1) \quad Y_1 = \min X_i$$

$$\begin{aligned}F_Y(y) &= P(Y_1 \leq y) \\&= P(\min X_i \leq y) \\&= 1 - (1 - F_X(y))^n\end{aligned}$$

$$\rightarrow P(Y_1 \leq \frac{1}{4}) = 1 - (\frac{3}{4})^n$$

20. Si X_1 y X_2 son v.a.i.i.d. $N(0,1)$ y sea

$$Y = \frac{(X_2 - X_1)^2}{2}$$

p.d.

$$Y \sim Ga(\frac{1}{2}, \frac{1}{2})$$

$$\begin{aligned}M_Y(t) &= E(e^{yt}) \\&= \int_0^\infty \int_0^\infty \exp\left(\frac{(x_2 - x_1)^2 t}{2}\right) \exp\left(-\frac{1}{2}(x_1^2 + x_2^2)\right) dx_1 dx_2 \\&= \left(\frac{\frac{1}{2}}{\frac{1}{2} - t}\right)^{1/2} \quad \text{para } t < \frac{1}{2}\end{aligned}$$

21. Si

$$X_1, X_2 \text{ v.a.i.i.d. } Normal(0, 1) \quad Y_1 = X_1^2 + X_2^2 \quad Y_2 = X_2$$

$$f(y_1, y_2) = \frac{1}{2\sqrt{y_1 - y_2^2}} \frac{1}{2\pi} e^{-y_1/2} I_{y_2^2, \infty}(y_1) I_{-\infty, \infty}(y_2)$$

$$Y_1 \sim \chi_2^2$$

22. (a) Sea X_1, X_2 v.a.i.i.d. Normal(0,1)

$$Y_1 = (X_1^2 + X_2^2)^{\frac{1}{2}}$$

$$Y_2 = \arctan\left(\frac{X_2}{X_1}\right)$$

$$|\mathbf{J}| = \frac{y_1 \sec^2(y_2)}{1 + \tan^2(y_2)}$$

$$\rightarrow f(y_1, y_2) = \frac{y_1}{2\pi} e^{-y_1^2/2} I_{\mathbb{R}}(y_1) I_{(0,\pi/2)}(y_2)$$

No son independientes.

(b) Sea $X_i \sim U(0, 1)$

$$Y_1 = X_2 + X_1$$

$$Y_2 = \frac{X_1}{X_2}$$

$$|\mathbf{J}| = \frac{y_1}{(1 - y_2)^2}$$

$$\rightarrow f(y_1, y_2) = \frac{y_1}{(1 - y_2)^2} I_{(0,1)}(y_2) I_{(0,1-y_2)}(y_1)$$

No son independientes.

(c) Sea $X_i \sim \text{Exp}(1)$, $i = 1, 2, 3$

$$Y_1 = \frac{X_1}{X_1 + X_2}$$

$$Y_2 = \frac{X_1 + X_2}{X_1 + X_2 + X_3}$$

$$Y_3 = X_1 + X_2 + X_3$$

$$|\mathbf{J}| = y_2 y_3^2$$

$$\rightarrow f(y_1, y_2, y_3) = y_2 y_3^2 e^{-y_3} I_{(0,1)}(y_1, y_2, y_3)$$

Si son independientes.

23. Sea

$$X_i \sim N(\mu_i, \sigma_i^2)$$

$$M_W(t) = E(e^{tw})$$

$$= \prod_{i=1}^n M_{X_i}(t)$$

$$= \prod_{i=1}^n \exp(\mu_i t + \sigma_i^2 t^2)$$

$$\rightarrow Y \sim N\left(\sum \mu_i, \sum \sigma_i^2\right)$$

24. (a) Sea

$$X \sim U(0, 2) \quad Y \sim U(0, 1)$$

$$Z = X + Y$$

$$f_Z = \frac{1}{2}z \quad I_{(0,1)}(z) + \frac{1}{2}I_{(1,2)}(z)$$

(b)

$$X, Y \sim N(0, 1)$$

$$Z = \frac{Y}{X}$$

$$W = X$$

$$\rightarrow |\mathbf{J}| = W$$

$$f_{(Z,W)} = \frac{w}{2\pi} e^{-w^2(z^2+1)} \quad I_{(-\infty, \infty)}(w, z)$$

$$\begin{aligned} f_Z(z) &= \int_{-\infty}^{\infty} \frac{w}{2\pi} e^{-w^2(z^2+1)} dw \quad I_{(-\infty, \infty)}(z) \\ &= \frac{1}{\pi(1+z^2)} \quad I_{\mathbb{R}}(z) \end{aligned}$$

(c) X, Y tal que

$$f_T(t) = bt^{b-1} \quad I_{(0,1)}(t)I_{(0,\infty)}(b)$$

$$Z = YX$$

$$W = Y$$

$$\rightarrow |\mathbf{J}| = \frac{1}{W}$$

$$\begin{aligned} f_{(Z,W)} &= \frac{b^2}{w} z^{b-1} \quad I_{(0,1)}(z)I_{(z,1)}(w)I_{(0,\infty)}(b) \\ \rightarrow f_Z &= b^2 z^{b-1} \ln(Z) \quad I_{(0,1)}(z)I_{(0,\infty)}(b) \end{aligned}$$

25. Sea

$$f(x_i) = 2x_i \quad I_{(0,1)}(x) \quad Z = \min(X_i)$$

$$F_Z(z) = 1 - (1 - x^2)^3$$

$$\rightarrow f_Z = 6x(1 - x^2)^2$$

$$\text{mediana} = \frac{1}{\sqrt{2}}$$

$$P(Z > \frac{1}{\sqrt{2}}) = 1 - F_z(1/\sqrt{2}) = (1 - 1/2)^3 = \frac{1}{8}$$

26. (a) Sean X_1, X_2, \dots, X_n v.a.i.i.d. con $F_X(x)$ y la función de densidad de la j -ésima estadística de orden Y_j esta dada por:

$$f(y_j) = \frac{n!}{(j-1)!(n-j)!} f_X(y) (F_X(y))^{j-1} (1 - F_X(y))^{n-j}$$

Para variables continuas.

- (b) Sea $Y_1 = \min(X_i)$

$$\begin{aligned} F_{Y_1} &= P(Y_1 \leq y_1) \\ &= P(\min(X_i) \leq y_1) \\ &= 1 - P(\min(X_i) > y_1) \\ &= 1 - \prod_{i=1}^n P(X_i > y_1) \\ &= 1 - (1 - F_X(y_1))^n \\ &\rightarrow f_{Y_1}(y_1) = n(1 - F_X(y_1))^{n-1} f_X(y_1) \end{aligned}$$

- (c) Sea $Y_n = \max(X_i)$

$$\begin{aligned} F_{Y_n} &= P(Y_n \leq y_n) \\ &= P(\max(X_i) \leq y_n) \\ &= \prod_{i=1}^n P(X_i \leq y_n) \\ &= (F_X(y_n))^n \\ &\rightarrow f_{Y_n}(y_n) = n(F_X(y_n))^{n-1} f_X(y_n) \end{aligned}$$

27. (a) Sea X_1, X_2, \dots, X_n una muestra aleatoria, en donde

$$f_{X_i}(x_i) \sim U(\mu - \sqrt{3}\sigma, \mu + \sqrt{3}\sigma)$$

Obtener el Rango muestral:

$$\begin{aligned} Z &= Y_n - Y_1 \\ W &= Y_1 \\ &\rightarrow |\mathbf{J}| = 1 \end{aligned}$$

$$f_{(Z,W)}(z, w) = n^2 \left(1 - \frac{w}{2\sqrt{3}\sigma}\right)^{n-1} \frac{1}{12\sigma} \left(\frac{z+w}{2\sqrt{3}\sigma}\right)^{n-1} I_{(\mu-\sqrt{3}\sigma, \mu+\sqrt{3}\sigma)}(w) I_{(0, 2\sqrt{3}\sigma)}(z)$$

$$f_Z(z) = \int_{\mu-\sqrt{3}\sigma}^{\mu-\sqrt{3}\sigma} n^2 \left(1 - \frac{w}{2\sqrt{3}\sigma}\right)^{n-1} \frac{1}{12\sigma} \left(\frac{z+w}{2\sqrt{3}\sigma}\right)^{n-1} dw I_{(0, 2\sqrt{3}\sigma)}(z)$$

Obtener el rango medio muestral:

$$\begin{aligned} Z &= \frac{Y_1 - Y_n}{2} \\ W &= Y_n \\ &\rightarrow |\mathbf{J}| = 2 \end{aligned}$$

$$\begin{aligned} f_{Z,W}(z, w) &= 2n^2 \left(1 - \frac{2z - w}{2\sqrt{3}\sigma}\right)^{n-1} \frac{1}{12\sigma} \left(\frac{w}{2\sqrt{3}\sigma}\right)^{n-1} I_{(\mu - \sqrt{3}\sigma, \mu)}(z) I_{(\mu - \sqrt{3}\sigma, 2z - \mu + \sqrt{3}\sigma)}(w) \\ &\quad + I_{(2z - \mu - \sqrt{3}\sigma, \mu + \sqrt{3}\sigma)}(w) I_{(\mu, \mu + \sqrt{3}\sigma)}(z) \\ \rightarrow f_Z(z) &= \int_{\mu - \sqrt{3}\sigma}^{2z - \mu + \sqrt{3}\sigma} 2n^2 \left(1 - \frac{2z - w}{2\sqrt{3}\sigma}\right)^{n-1} \frac{1}{12\sigma} \left(\frac{w}{2\sqrt{3}\sigma}\right)^{n-1} dw I_{(\mu - \sqrt{3}\sigma, \mu)}(z) \\ &\quad + \int_{2z - \mu - \sqrt{3}\sigma}^{\mu + \sqrt{3}\sigma} 2n^2 \left(1 - \frac{2z - w}{2\sqrt{3}\sigma}\right)^{n-1} \frac{1}{12\sigma} \left(\frac{w}{2\sqrt{3}\sigma}\right)^{n-1} dw I_{(\mu, \mu + \sqrt{3}\sigma)}(z) \end{aligned}$$

(b) Sean $X_1, X_2, \dots, X_n \sim \text{Exp}(\theta)$. Obtener el rango muestral:

$$\begin{aligned} Z &= Y_n - Y_1 \\ W &= Y_1 \\ &\rightarrow |\mathbf{J}| = 1 \end{aligned}$$

$$\begin{aligned} f_{Z,W}(z, w) &= \theta^2 n^2 (1 - e^{\theta(z+w)})^{n-1} e^{\theta w(n+1)} e^{\theta z} I_{(0, \infty)}(z) I_{(z, \infty)}(w) \\ f_Z(z) &= \int_z^\infty \theta^2 n^2 (1 - e^{\theta(z+w)})^{n-1} e^{\theta w(n+1)} e^{\theta z} dw I_{(0, \infty)}(z) \end{aligned}$$

Obtener el Rango medio muestral:

$$\begin{aligned} Z &= \frac{Y_1 - Y_n}{2} \\ W &= Y_n \\ &\rightarrow |\mathbf{J}| = 2 \end{aligned}$$

$$\begin{aligned} f_{Z,W}(z, w) &= 2\theta^2 n^2 (1 - e^{\theta(z+w)})^{n-1} e^{\theta w(n+1)} e^{\theta z} I_{(0, \infty)}(z) I_{(0, 2z)}(w) \\ f_Z(z) &= \int_0^{2z} 2\theta^2 n^2 (1 - e^{\theta(z+w)})^{n-1} e^{\theta w(n+1)} e^{\theta z} dw I_{(0, \infty)}(z) \end{aligned}$$

28. Sea

$$L \sim N(\mu, 1)$$

(a)

$$P(\hat{Y} > 8) = 1 - \phi(4.5)$$

(b)

$$P(6.2 \leq \hat{Y} \leq 6.8) = 2\phi(0.9) - 1$$

(c)

$$P(\mu - 0.3 \leq \hat{Y} \leq \mu + 0.3) = 0.95$$

$$\rightarrow 2\phi(0.3\sqrt{n}) - 1 = 0.95$$

$$\rightarrow n = 42.6844$$

29. Sea $N = 10$, $\sigma = 1$

$$P(a \leq S^2 \leq b) = 0.90$$

$$\text{en donde } S^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 \sim Ga\left(\frac{n-1}{2}, \frac{n-1}{2\sigma^2}\right)$$

$$\rightarrow 0.90 = \frac{\left(\frac{9}{2}\right)^{9/2}}{\Gamma\left(\frac{9}{2}\right)} \int_0^b x^{\frac{7}{2}} e^{-\frac{7x}{2}} dx - \frac{\left(\frac{9}{2}\right)^{9/2}}{\Gamma\left(\frac{9}{2}\right)} \int_0^a x^{\frac{7}{2}} e^{-\frac{7x}{2}} dx$$

30. Sea

$$N_1 = 6 \quad N_2 = 10$$

ambas con la misma varianza de poblacional.

$$P\left(\frac{S_1^2}{S_2^2} \leq b\right) = 0.90$$

sabemos que

$$\frac{S_1^2}{S_2^2} \sim F(5, 9)$$

$$\rightarrow b = 2.611$$

31. Sea

$$P(X \leq Q_1) = 0.25, \quad X \sim N(0, 1)$$

$$Qx_1 = 1 - \phi(0.68)$$

$$Qx_2 = 0$$

$$Qx_3 = \phi(0.68)$$

Sea

$$Y \sim t - student(10)$$

$$Qy_1 = 1 - F_X(.700)$$

$$Qy_2 = 0.129$$

$$Qy_3 = 0.700$$

$$W \sim \chi^2(20)$$

$$Qw_1 = 15425$$

$$Qw_2 = 19337$$

$$Qw_3 = 23828$$

32. Sean

$$X \sim U(0, 1) \quad Y = \ln\left(\frac{x}{1-x}\right)$$

$$P(Y \leq y) = P\left(\ln\left(\frac{x}{1-x}\right) \leq y\right)$$

$$= P\left(x \leq \frac{e^y}{1+e^y}\right)$$

$$\rightarrow f_Y(y) = \frac{e^y}{(1+e^y)^2}$$

33. Sea

$$X_i \sim \chi_2^2$$

$$W = \frac{X_1 - X_2}{2}$$

$$Z = X_1$$

$$\rightarrow |\mathbf{J}| = 2$$

$$\rightarrow f_{Z,W}(z, w) = e^{-z-w} I_{(0,\infty)}(w) I_{(2w,\infty)}(z)$$

$$\rightarrow f_W(w) = e^{-3w} I_{(0,\infty)}(w)$$

34. Sean

$$X \sim N(\mu, \sigma^2) \quad Y = \frac{x - \mu}{\sigma} \sim N(0, 1) \quad W = Y^2$$

$$\rightarrow f_W(w) = \frac{1}{\sqrt{2\pi}} w^{\frac{1}{2}} e^{-\frac{1}{2w}} I_{(0,\infty)}(w) \quad \text{como } \sqrt{\pi} = \Gamma\left(\frac{1}{2}\right)$$

$$= \frac{\left(\frac{1}{2}\right)^{\frac{1}{2}}}{\Gamma\left(\frac{1}{2}\right)} w^{\frac{1}{2}} e^{-\frac{1}{2w}} I_{(0,\infty)}(w)$$

$$\rightarrow W \sim Ga\left(\frac{1}{2}, \frac{1}{2}\right) = \chi_1^2$$

35. Sea

$$X \sim Ga(\alpha, 1)$$

$$y_i = \frac{x_i}{Y_{k+1}}$$

$$Y_{k+1} = \sum_{i=1}^{k+1} x_i$$

sea

$$x_1 = y_1 * Y_{k+1}, x_2 = y_2 * Y_{k+1}, \dots, x_k = y_k * Y_{k+1}$$

Y

$$x_{k+1} = (1 - \sum_{i=1}^k y_i) * y_{k+1}$$

$$\rightarrow |\mathbf{J}| = |y_{k+1}^k - 2(y_{k+1}^k * \sum y_i)|$$

$$\rightarrow f_Y(y) = \left(\frac{1}{\Gamma(\alpha)}\right)^{k+1} \int_0^\infty (y_{k+1}^k - 2(y_{k+1}^k * \sum y_i)) * e^{-y_{k+1}} y_{k+1}^{2(\alpha-1)} (1 - \sum y_i)^{\alpha-1} \prod_{i=1}^k y_i^{\alpha-1} dy_{k+1}$$

Sea

$$\gamma = \left(\frac{1}{\Gamma(\alpha)}\right)^{k+1} (1 - 2 \sum y_i) (1 - \sum y_i)^{\alpha-1} \prod_{i=1}^k y_i^{\alpha-1}$$

$$\rightarrow \gamma \int_0^\infty y_{k+1}^k e^{-y_{k+1}} dy_{k+1} y_{k+1}^{2(\alpha-1)} = \gamma * \Gamma(2\alpha + k - 1)$$

$$\rightarrow f_Y(y) = \left(\frac{1}{\Gamma(\alpha)}\right)^{k+1} (1 - 2 \sum y_i) (1 - \sum y_i)^{\alpha-1} \prod_{i=1}^k y_i^{\alpha-1} * \Gamma(2\alpha + k - 1)$$

Con K = 1 se tiene una Beta

36. (a) Sean

Sea $X_1, X_2, \dots, X_n v.a.i. N(\mu, \sigma)$

$$YS^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

$$\rightarrow M = \sum (x_i - \mu)^2(t) = M \sum (x_i - \bar{x})^2(t) * M(n\bar{x} - n\mu)^2$$

Si son independientes,

$$\rightarrow M \sum (x_i - \bar{x})^2(t) = \frac{M \sum (x_i - \mu)^2(t)}{M(n\bar{x} - n\mu)^2}$$

$$\rightarrow M \sum \left(\frac{x_i - \bar{x}}{\sigma}\right)^2(t) = \frac{M \sum \left(\frac{x_i - \mu}{\sigma}\right)^2(t)}{M\left(\frac{n\bar{x} - n\mu}{\sigma}\right)^2}$$

En donde

$$M \sum \left(\frac{x_i - \mu}{\sigma}\right)^2(t) \sim \chi_n^2$$

$$M\left(\frac{n\bar{x} - n\mu}{\sigma}\right)^2 \sim \chi_1^2$$

$$\begin{aligned} \rightarrow M \sum \left(\frac{x_i - \bar{x}}{\sigma} \right)^2 (t) &= \frac{\left(\frac{1}{2} - t \right)^{\frac{n}{2}}}{\left(\frac{1}{2} - t \right)^{\frac{1}{2}}} \\ &= \left(\frac{1}{2} - t \right)^{\frac{n-1}{2}} \end{aligned}$$

Por Teorema de unicidad.

$$\sum \left(\frac{x_i - \bar{x}}{\sigma} \right)^2 = \frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$$

(b) Sabemos que si

$$X \sim N(0, 1) \quad Y \sim \chi_n^2$$

entonces

$$\begin{aligned} T &= \frac{X}{\sqrt{\frac{Y}{n}}} \sim t(n) \\ \rightarrow \frac{\sqrt{n}(\bar{x} - \mu)}{S_X} &= \frac{\frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0, 1)}{\sqrt{\frac{(n-1)S^2}{\sigma^2(n-1)}} \sim \sqrt{\frac{\chi_{n-1}^2}{n-1}}} \sim t(n-1) \end{aligned}$$

37. Sea $n = 3$

(a)

$$P(Y_1 > \text{median}) \quad \text{median} = \frac{1}{\sqrt{2}}$$

$$\begin{aligned} P(Y_1 > \text{median}) &= \int_0^{\frac{1}{\sqrt{2}}} 6(1-y^2)^2 y_1 dy_1 \\ &= 0.125 \end{aligned}$$

(b) Sean

$$\begin{aligned} Z_1 &= \frac{Y_1}{Y_2} \\ Z_2 &= \frac{Y_2}{Y_3} \\ Z_3 &= Y_3 \\ \rightarrow |J| &= z_2 z_3^2 \\ \rightarrow f(z_1, z_2, z_3) &= z_2 z_3^2 f_{Y_1}(z_1 z_2 z_3) f_{Y_2}(z_2 z_3) f_{Y_3}(z_3) \end{aligned}$$

Son independientes

38. Sea

$$f_{X,Y}(x, y) = \frac{12}{7}x(x+y)I_{(0,1)}(x)I_{(0,1)}(y)$$

$$U = \min(X, Y) \quad V = \min(X, Y)$$

$$\begin{aligned} F_{U,V} &= P(U \leq u, V \leq v) \\ &= P(V \leq v) - P(U > u, V \leq v) \end{aligned}$$

$$\begin{aligned} P(V \leq v) &= \int_0^v \int_0^v \frac{12}{7}x(x+y)dx dy \\ &= v^4 \end{aligned}$$

$$\begin{aligned} \rightarrow P(U > u, V \leq v) &= \int_u^v \int_u^v \frac{12}{7}x(x+y)dy dx \\ &= v^4 + u^4 - \frac{6}{7}u^2v^2 - \frac{4}{7}v^3u - \frac{4}{7}u^3v \end{aligned}$$

$$f_{U,V}(u, v) = \frac{1}{7}(4v^3u + 4u^3v - 7u^4 + 6v^2u^2) \quad I(0 < u < v < 1)$$

39. Si $X_1, X_2, X_3, \dots, X_n$ es una muestra aleatoria.

$$E(X_i) = \mu \quad Var(X_i) = \sigma^2$$

$$V(\bar{x}) = V\left(\sum_{i=1}^n \frac{1}{n}x_i\right) = \frac{\sigma^2}{n}$$

$$\begin{aligned} E(S^2) &= E\left(\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2\right) \\ &= \frac{1}{n-1} \left(\sum_{i=1}^n Var(x_i) - nVar(\bar{x})\right) \\ &\rightarrow E(S^2) = \sigma^2 \end{aligned}$$

40. Sea

$$Y = a + bx$$

$$\begin{aligned} \rho_{X,Y} &= \frac{Cov(X, Y)}{\sigma_Y \sigma_X} \\ &= \frac{bV(X)}{b\sigma_X \sigma_X} \\ &= 1 \end{aligned}$$

41. Si $X_1, X_2, X_3, \dots, X_n$ es una muestra con $V(X_i) = \sigma^2$. P.D. $Cov(x_i - \bar{x}, \bar{x}) = 0$.

$$\begin{aligned} Cov(x_i - \bar{x}, \bar{x}) &= Cov(x_i, \bar{x}) - Var(\bar{x}) \\ &= \frac{\sigma^2}{n} - \frac{\sigma^2}{n} \\ &= 0 \end{aligned}$$

42. Sea $X_1, X_2, X_3, \dots, X_n$ una sucesion de v.a.i.i.d. con media μ y varianza σ^2 .

(a)

$$Nm \sim Po(\gamma)$$

$$\begin{aligned} E(S_N) &= E(E(N|S_N)) \\ &= \frac{n\mu}{\gamma} \end{aligned}$$

$$\begin{aligned} V(S_N) &= V(E(N|S_N)) + E(V(N|S_N)) \\ &= \frac{n}{\gamma^2}(n\sigma^2 + \mu) \end{aligned}$$

(b)

$$N \sim Geo\left(\frac{1}{\gamma}\right) \quad \gamma = \frac{1-p}{p}$$

$$\begin{aligned} E(S_N) &= E(E(N|S_N)) \\ &= n\mu \frac{1}{\gamma} \end{aligned}$$

$$\begin{aligned} V(S_N) &= V(E(N|S_N)) + E(V(N|S_N)) \\ &= \left(n\sigma \frac{1}{\gamma}\right)^2 + \frac{1}{\gamma} \left(1 - \frac{1}{\gamma}\right) n\mu \end{aligned}$$

43. Sea X una v.a. con segundo momento finito.

$$\begin{aligned} E((X - a)^2) &= E(X^2) - 2aE(X) + a^2 + 2aE(X) - a^2 \\ &= E(X^2) - E^2(X) \\ &= Var(X) \end{aligned}$$

44. Sean

$$Z \sim N(0, 1) \quad Y = a + bz + cz^2$$

$$\begin{aligned} \rightarrow \rho_{Y,Z} &= \frac{Cov(Z, Y)}{\sigma_Z \sigma_Y} \\ &= \frac{b}{\sqrt{b^2 + 2c^2}} \end{aligned}$$

45.

$$\begin{aligned}
 V\left(\sum_{i=1}^k \alpha_i x_i\right) &= E\left[\left(\sum_{i=1}^k \alpha_i x_i\right)^2\right] - E^2\left[\sum_{i=1}^k \alpha_i x_i\right] \\
 &= E\left[\sum_{i=1}^k \alpha_i x_i - \sum_{i=1}^n \alpha_i \mu_i\right] \\
 &= E\left[\sum_{i=1}^k \alpha_i (x_i - \mu_i)^2\right] \\
 &= E\left[\sum_{i=1}^k \alpha_i^2 \text{Var}(x_i) + 2 \sum_{i < j} \alpha_i \alpha_j \text{cov}(x_i, x_j)\right]
 \end{aligned}$$

46. Sean

$$Z \sim N(0, 1) \quad Y = a + bz + cz^2$$

$$\rho = \frac{\text{Cov}(Z, Y)}{\sigma_Y \sigma_Z}$$

$$\begin{aligned}
 \text{Cov}(Z, Y) &= E(ZY) - E(Z)E(Y) \\
 &= E(ZY) \\
 &= E(az + bz^2 + cz^3) \\
 &= b + cE(Z^3) \\
 &= b \quad \text{por independencia}
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(Z) &= \text{Var}(a + bz + cz^2) \\
 &= b^2 + 2c^2
 \end{aligned}$$

$$\rightarrow \sigma_Y = \sqrt{b^2 + 2c^2}$$

$$\rho_{X,Y} = \frac{b}{\sqrt{b^2 + 2c^2}}$$

47. Sea $X \sim N(0, 1)$ y sea I otra v.a.i. de X tal que $P(I = 1) = P(I = 0) = \frac{1}{2}$. Y sea también $Y = X$ si $I = 1$ y $Y = -X$ si $I = 0$.

$$f_Y = \begin{cases} Y \sim N(0, 1) & \text{c.p. } \frac{1}{2} \\ Y \sim N(0, 1) & \text{c.p. } \frac{1}{2} \end{cases}$$

$$\rightarrow f_Y = \frac{1}{2\pi} e^{-\frac{x^2}{2}} \sim N(0, 1)$$

$$\text{Cov}(X, Y) = 0$$

48. Sean

$$f(Y|X) = \text{Bin}(n, x) \quad X \sim U(0, 1)$$

$$f_{X,Y}(x, y) = \binom{n}{y} x^y (1-x)^{n-y} \quad I_{(0,1)}(x) I_{0,1,2,\dots,n}(y)$$

$$f_{X,Y}(x, y) = \binom{n}{y} \int_0^1 x^y (1-x)^{n-y} dx \quad I_{0,1,2,\dots,n}(y)$$

$$= \binom{n}{y} \frac{\Gamma(y+1)\Gamma(n-y+1)}{\Gamma(n+2)} \quad I_{0,1,2,\dots,n}(y)$$

* Kernel de una beta.

49. Sea

$$ECM = E[(y - h(x))^2 | X]$$

P.D.

$$\min_f ECM(f) = E(Y|X = x)$$

$$\min_{f_X} E[(y - h(x))^2 | X] = \min_c E[(y - c)^2 | X]$$

$$= \min_c [E(y^2 | X) - 2E(y c | X = x) + E(c^2 | X = x)]$$

$$\rightarrow -2E(Y|X) + 2c = 0$$

$$\rightarrow f_X(x) = E(Y|X)$$

50. Sean

$$u = \frac{y_1 - \mu_1}{\sigma_1} \quad v = \frac{y_2 - \mu_2}{\sigma_2}$$

$$f(y_2|y_1) = \frac{\frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp\left[\frac{-1}{1-\rho^2}(u^2 - 2\rho uv + v^2)\right]}{\frac{1}{2\pi\sigma_1} e^{-u^2/2}}$$

$$= \frac{1}{\sqrt{2\pi}\sigma_2\sqrt{1-\rho^2}} \exp\left[\frac{-1}{2}\left(\frac{v - \rho u}{\sqrt{1-\rho^2}}\right)^2\right]$$

$$\rightarrow f(y_2|y_1) = \frac{1}{\sqrt{2\pi}\sigma_2\sqrt{1-\rho^2}} \exp\left[\frac{-1}{2} \frac{y_2 - (\mu_2 + \rho\frac{\sigma_2}{\sigma_1}(y_1 - \mu_1))}{\sigma_2\sqrt{1-\rho^2}}\right]$$

$$\rightarrow f(y_2|y_1) \sim N(\mu_2 + \rho\frac{\sigma_2}{\sigma_1}(y_1 - \mu_1), (\sigma_2\sqrt{1-\rho^2})^2)$$

51. Sea

$$X \sim N(0, 1)$$

$$\phi_X(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{ixt} e^{-x^2/2} dx$$

$$= e^{-t^2/2} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-(x-it)^2/2} dx$$

$$= e^{-t^2/2}$$

En general, sea

$$W = \mu + \sigma x, W \sim N(\mu, \sigma^2)$$

$$\begin{aligned}\phi_W(t) &= E(e^{i(\mu + \sigma x)t}) \\ &= e^{i\mu t} E(e^{i\sigma x t}) \\ &= e^{i\mu t} \phi_X(\sigma t) \\ &= e^{i\mu t} e^{-\sigma^2 t^2 / 2}\end{aligned}$$

52.

$$f_{p|n} = \frac{f_p f_{n|p}}{f_n}$$

$$\begin{aligned}f_n &= \int_0^1 \binom{n+m}{n} p^n (1-p)^m dp \\ &= \binom{n+m}{n} \frac{\Gamma(n+1)\Gamma(m+1)}{\Gamma(n+m+2)}\end{aligned}$$

$$f_{p|n} = \frac{p^n (1-p)^m}{\frac{\Gamma(n+1)\Gamma(m+1)}{\Gamma(n+m+2)}}$$

53.

$$f_{X,Y} = e^{-(x+y)} I_{[0,\infty)}(x) I_{[0,\infty)}(y)$$

Sea

$$\begin{aligned}Z &= \frac{X}{Y} && \leftrightarrow X = WZ \\ W &= Y && \leftrightarrow Y = W\end{aligned}$$

$$\begin{aligned}\mathbf{J} &= \begin{pmatrix} Z & W \\ 1 & 0 \end{pmatrix} \\ &\rightarrow |\mathbf{J}| = W\end{aligned}$$

$$\begin{aligned}f_{(W,Z)} &= w * f_{(X,Y)}(wz, w) \\ &= w * e^{-w(z+1)} I_{[0,\infty)}(w) I_{[0,\infty)}(z)\end{aligned}$$

$$\begin{aligned}f_Z &= \int_0^\infty w e^{-w(z+1)} dw I_{[0,\infty)}(z) \\ &= \frac{1}{z+1} I_{[0,\infty)}(z)\end{aligned}$$

54. Sea

$$\begin{aligned} X &\sim Po(\theta), \\ Y &\sim Po(\lambda), \\ Z = X + Y &\sim Po(\theta + \lambda). \end{aligned}$$

p.d.

$$f_{X|Z} \sim Bin\left(Z, \frac{\theta}{\theta + \lambda}\right)$$

$$\begin{aligned} f_{X,Z} &= f_{(X,Y)}(x, z-x) \\ &= \frac{\theta^x}{x!} e^{-(\theta)} \frac{\lambda^{z-x}}{(z-x)!} e^{-\lambda} I_{[0,1,2,\dots,z]}(x) I_{[0,1,2,\dots]}(y) \end{aligned}$$

$$f_Z = \frac{(\lambda + \theta)^z}{z!} e^{-(\lambda + \theta)} I_{[0,1,2,\dots]}(z)$$

$$\begin{aligned} \frac{f_{X,Z}}{f_Z} &= \frac{\frac{\theta^x}{x!} e^{-(\theta)} \frac{\lambda^{z-x}}{(z-x)!} e^{-\lambda} I_{(0,1,2,\dots,z)}(x) I_{(0,1,2,\dots)}(y)}{\frac{(\lambda + \theta)^z}{z!} e^{-(\lambda + \theta)} I_{(0,1,2,\dots)}(z)} \\ &= \frac{z!}{(x-z)! x!} \theta^x \lambda^{z-x} (\lambda + \theta)^{-z} I_{(0,1,2,\dots,z)}(x) \\ &= \binom{z}{x} \left(\frac{\theta}{\theta + \lambda}\right)^x \left(\frac{\lambda}{\theta + \lambda}\right)^{z-x} I_{[0,1,2,\dots,z]}(x) \end{aligned}$$

$$f_{(X|Z)} \sim Bin\left(z, \frac{\theta}{\theta + \lambda}\right)$$

55. Sean $X_1, X_2, X_3, \dots, X_n$ v.a.i.i.d

$$P(X_k = 1) = P(X_k = -1) = \frac{1}{2}$$

$$N \sim Geo(\alpha)$$

$$P(N = n) = \alpha(1 - \alpha)^n I_{1,2,3,4,\dots}(n)$$

$$Y = \sum_{i=1}^n x_i \quad x_i \sim Geo(\alpha)$$

$$\rightarrow Y \sim BinNeg(n, \alpha)$$

$$\rightarrow f_Y = \binom{r + y - 1}{y} \alpha^r (1 - \alpha)^y I_{1,2,3,4,\dots}(y)$$

56. Sea

(a)

$$f_S = k(s^2 + 1) \quad I_{(1,4)}(s)$$

(b)

$$f_T = ct \quad I_{(0,600)}(t)$$

(c)

$$1 = k \int_1^4 (s^2 + 1) ds \quad \rightarrow k = \frac{1}{24}$$

(d)

$$1 = c \int_0^6 00t dt \quad \rightarrow c = \frac{1}{180000}$$

(e)

$$f_X = \begin{cases} 100 & \text{c.p.0.11112,} \\ 200 & \text{c.p.0.33336,} \\ 300 & \text{c.p.0.55556.} \end{cases}$$

(f)

$$f_Y = \begin{cases} 1.5 & \text{c.p.0.1388,} \\ 2.5 & \text{c.p.0.3055,} \\ 3.5 & \text{c.p.0.5555} \end{cases}$$

(g)

$$f_Y = \begin{cases} x = 100 & y = 1.5 & \text{c.p. 0.022224,} \\ x = 100 & y = 2.5 & \text{c.p. 0.044448,} \\ x = 100 & y = 3.5 & \text{c.p. 0.044448,} \\ x = 200 & y = 1.5 & \text{c.p. 0.181963,} \\ x = 200 & y = 2.5 & \text{c.p. 0.060654,} \\ x = 200 & y = 3.5 & \text{c.p. 0.090981,} \\ x = 300 & y = 1.5 & \text{c.p. 0.06486,} \\ x = 300 & y = 2.5 & \text{c.p. 0.20039,} \\ x = 300 & y = 3.5 & \text{c.p. 0.42007} \end{cases}$$

57. Sean $X_1, X_2, X_3, \dots, X_n$ v.a.i.i.d.

$$f_{X_i} = 2X_i \quad I_{(0,1)}(x)$$

$$W = Y_n - Y_1$$

$$Z = Y_n$$

$$\rightarrow |J| = 1$$

$$f_{W,Z} = 4n^2(1 - (z - w)^2)^{n-1}(z - w)z^{2n-1} \quad I_{(0,1)}(w)I_{(w,1)}(z)$$

$$f_W = \int_w^1 4n^2(1 - (z - w)^2)^{n-1}(z - w)z^{2n-1} dz \quad I_{(0,1)}(x)$$

$$E(W) = \int_w^1 \int_w^1 4n^2(1 - (z - w)^2)^{n-1}(z - w)z^{2n-1} dz dw$$

58. Sean

$$f_X = \frac{1}{4} I_{(-2,-1,1,2)}(x)$$

Y sea

$$Y = X^2$$

(a)

X/Y	1	4
-2	0	0.25
-1	0.25	0
1	0.25	0
2	0	0.25

(b)

$$\begin{aligned} CovXY &= E(XY) - E(X)E(Y) \\ &= \sum_x \sum_y xyf(x, y) - \sum_x xf(x) \sum_y yf(y) \\ &= 0 \end{aligned}$$

(c)

$$\rho_{XY} = 0$$

(d) Pero no son independientes porque Y siempre va a depender de X

59. Sea

$$X \sim U(0, 1)$$

Y

$$Y = \log \frac{X}{1-X}$$

$$\begin{aligned} P(Y \leq \log \frac{X}{1-X}) &= P(X \leq \frac{e^y}{1+e^y}) \\ \rightarrow f_Y &= (1+e^y)e^y - e^{2y}(1+e^y)^{-2} I_{(-\infty, \infty)}(Y) \end{aligned}$$