

STAT 410 HW #4 Solution

$$\begin{aligned}
 1. \quad E(X^k) &= \int_0^{\infty} x^k \cdot \frac{1}{\Gamma(\alpha)} x^{\alpha-1} e^{-x} dx \\
 &= \frac{1}{\Gamma(\alpha)} \int_0^{\infty} x^{k+\alpha-1} e^{-x} dx \\
 &= \frac{\Gamma(\alpha+k)}{\Gamma(\alpha)} \\
 &= \frac{(\alpha+k-1)\Gamma(\alpha+k-1)}{\Gamma(\alpha)} \\
 &= \dots \\
 &= \frac{(\alpha+k-1)(\alpha+k-2)\dots(\alpha+1)\alpha\Gamma(\alpha)}{\Gamma(\alpha)} \\
 &= (\alpha+k-1)(\alpha+k-2)\dots(\alpha+1)\alpha
 \end{aligned}$$

$$2. \quad X \sim \text{Gamma}(2, 2) \sim \chi^2(4)$$

$$P(X > 7.779) = 1 - P(X \leq 7.779) \doteq 0.1$$

$$3. \quad a. \quad P(\text{Gamma}(7, \frac{1}{3}) > 3) \doteq 0.20$$

$$b. \quad P(1 < \text{Gamma}(7, \frac{1}{3}) < 2) = P(\text{Gamma}(7, \frac{1}{3}) < 2) - P(\text{Gamma}(7, \frac{1}{3}) < 1) \doteq 0.360$$

$$4. \quad Y = (1, -1, 2) \begin{pmatrix} X_1 \\ X_2 \\ X_3 \end{pmatrix}$$

$$a. \quad EY = (1, -1, 2) \vec{\mu} = (1, -1, 2) \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = 5 \quad \text{Var}(Y) = (1, -1, 2) \Sigma \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix} = 5$$

$$b. \quad P(X_1 > X_2 + X_3 - 4) = P(-X_1 + X_2 + X_3 < 4)$$

$$-X_1 + X_2 + X_3 \sim N(4, 2)$$

$$P(N(4, 2) < 4) = P(N(0, 1) < 0) = 0.5$$

$$5. \quad P(Y=1) = P(X > 0)$$

$$= 1 - P(X \leq 0)$$

$$= 1 - P\left(\frac{X-\eta}{1} \leq -\eta\right)$$

$$= 1 - \Phi(-\eta)$$

$$= \Phi(\eta)$$

$$\begin{aligned}
 b. a. M_A(t) &= E(e^{At}) \\
 &= E(e^{SZt}) \\
 &= E(E(e^{SZt} | S)) \\
 &= \frac{1}{2} E(e^{Zt}) + \frac{1}{2} E(e^{-Zt}) \\
 &= \frac{1}{2} e^{\frac{1}{2}t^2} + \frac{1}{2} e^{\frac{1}{2}t^2} \\
 &= e^{\frac{1}{2}t^2}
 \end{aligned}$$

$$\Rightarrow A = SZ \sim N(0, 1)$$

$$\begin{aligned}
 b. P(Z+A \leq x) &= P(Z+A \leq x, S=1) + P(Z+A \leq x, S=-1) \\
 &= \frac{1}{2} P(Z \leq x) + \frac{1}{2} P(0 \leq x) \\
 &= \begin{cases} \frac{1}{2} \Phi(\frac{x}{\sigma}) & x < 0 \\ \frac{1}{2} \Phi(\frac{x}{\sigma}) + \frac{1}{2} & x \geq 0 \end{cases}
 \end{aligned}$$

the CDF has a jump at $x=0 \Rightarrow Z+A$ is not normally distributed

$$\begin{aligned}
 7. E(e^{Yt}) &= e^{\mu t + \frac{1}{2}\sigma^2 t^2} = M_Y(t) \\
 E(X) &= E(e^Y) = M_Y(t) \Big|_{t=1} = e^{\mu + \frac{1}{2}\sigma^2} \\
 E(X^2) &= E(e^{2Y}) = M_Y(t) \Big|_{t=2} = e^{2\mu + 2\sigma^2} \\
 \text{Var}(X) &= E(X^2) - [E(X)]^2 = e^{2\mu + 2\sigma^2} - e^{2\mu + \sigma^2}
 \end{aligned}$$

$$8. E(\bar{X}) = E(\vec{a}' \underline{X} / n) = \frac{1}{n} E(\vec{a}' \underline{X}) = \frac{1}{n} \vec{a}' \vec{\mu} = \frac{1}{n} (1, 1, \dots, 1) \begin{pmatrix} \mu_1 \\ \mu_2 \\ \vdots \\ \mu_n \end{pmatrix} = \frac{1}{n} \sum_{i=1}^n \mu_i$$

$$\text{Var}(\bar{X}) = \frac{1}{n^2} \vec{a}' I_n \vec{a} = \frac{1}{n}$$

$$\bar{X} \sim N\left(\frac{1}{n} \sum_{i=1}^n \mu_i, \frac{1}{n}\right)$$

$$9. \Sigma = \begin{pmatrix} 1-\rho & & & 0 \\ & 1-\rho & & \\ & & \ddots & \\ 0 & & & 1-\rho \end{pmatrix}_{n \times n} + \begin{pmatrix} \rho & \rho & \dots & \rho \\ \rho & \rho & & \vdots \\ & & \ddots & \\ \rho & & & \rho \end{pmatrix}_{n \times n} = \begin{pmatrix} 1 & \rho & \rho & \dots & \rho \\ \rho & 1 & & & \\ \vdots & & \ddots & & \rho \\ \rho & \rho & \dots & \rho & 1 \end{pmatrix}_{n \times n}$$

$$a. \text{Var}(X_i) = (\Sigma)_{ii} = 1 \quad i=1, 2, \dots, n$$

$$b. \text{Cov}(X_i, X_j) = (\Sigma)_{ij} = \rho \quad 1 \leq i < j \leq n$$

$$\text{Corr}(X_i, X_j) = \frac{\rho}{1 \cdot 1} = \rho \quad 1 \leq i < j \leq n$$

$$10. E(\bar{X}) = \frac{1}{n} \sum_{i=1}^n \mu_i$$

$$\text{Var}(\bar{X}) = \frac{1}{n^2} (1, 1, \dots, 1) \sum \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} = \frac{1}{n^2} \cdot n \cdot (1 + (n-1)\rho) = \frac{1}{n} + \frac{n-1}{n}\rho$$

$$\bar{X} \sim N\left(\frac{1}{n} \sum_{i=1}^n \mu_i, \frac{1}{n} + \frac{n-1}{n}\rho\right)$$

$$11. E(\Phi(a-bX)) = \int_{-\infty}^{+\infty} \Phi(a-bx) f_X(x) dx$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{a-bx} f_Z(z) dz f_X(x) dx \quad Z, X \sim N(0, 1)$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{a-bx} f_{ZX}(z, x) dz dx$$

$$= P_r(Z + bX \leq a) \quad Z + bX \sim N(0, 1+b^2)$$

$$= P_r\left(\frac{Z + bX}{\sqrt{1+b^2}} \leq \frac{a}{\sqrt{1+b^2}}\right)$$

$$= \Phi\left(\frac{a}{\sqrt{1+b^2}}\right)$$

$$12. f(x) = C \cdot e^{-\ln(2)x^2} \sim N(0, \frac{1}{2\ln 2})$$

$$C = \frac{1}{\sqrt{2\pi} \cdot \sigma} = \frac{1}{\sqrt{2\pi} \sqrt{\frac{1}{2\ln 2}}} = \sqrt{\frac{\ln 2}{\pi}}$$

$$a. P(0.4 < X < 1.3) = P\left(\frac{0.4-0}{\sqrt{\frac{1}{2\ln 2}}} < \frac{X-0}{\sqrt{\frac{1}{2\ln 2}}} < \frac{1.3-0}{\sqrt{\frac{1}{2\ln 2}}}\right)$$

$$= P(0.4\sqrt{2\ln 2} < N(0, 1) < 1.3\sqrt{2\ln 2})$$

$$\approx 0.256$$

$$b. \int_{0.5}^{1.8} 2^{-x^2} dx = \frac{\sqrt{\pi}}{\sqrt{\ln 2}} \int_{0.5}^{1.8} f(x) dx$$

$$= \frac{\sqrt{\pi}}{\sqrt{\ln 2}} P(0.5 < X < 1.8)$$

$$= \frac{\sqrt{\pi}}{\sqrt{\ln 2}} P(0.5\sqrt{2\ln 2} < N(0, 1) < 1.8\sqrt{2\ln 2}) \approx 0.556$$

