

$$1) f_X(x) = \lambda e^{-\lambda x} u(x), \quad E[X] = \frac{1}{\lambda}, \quad \text{var}(X) = \frac{1}{\lambda^2}, \quad P[X \geq a] = 1 - P[X < a] = \begin{cases} 1 & a \leq 0 \\ e^{-\lambda a} & a > 0 \end{cases}$$

$$i) \text{ Markov: } P[X \geq a] \leq \frac{E[X]}{a} = \boxed{\frac{1}{\lambda a}} \quad \text{for } \boxed{a \geq 0}$$

$$ii) \text{ Chebyshev: } P[|X - \mu| \geq b] = P[X - \mu \geq b] + P[X - \mu \leq -b] \\ = P[X \geq \mu + b] + \underbrace{P[X \leq \mu - b]}_{=0 \text{ when } b \geq \mu}$$

$$a = \mu + b \Rightarrow P[X \geq a] = P[X - \mu \geq a - \mu] \quad \text{when } \boxed{a \geq \mu} \\ \leq \frac{\text{var}(X)}{(a - \mu)^2} = \boxed{\frac{1}{\lambda^2 (a - \mu)^2}} \quad \text{by Chebyshev inequality}$$

$$iii) \text{ Chernoff: } \Theta_X(t) = \int_0^\infty f_X(x) e^{tx} dx = \lambda \int_0^\infty e^{(t-\lambda)x} dx = \frac{\lambda}{t-\lambda} \left[ e^{(t-\lambda)\infty} - 1 \right] = \frac{\lambda}{\lambda - t} \quad \text{for } t < \lambda$$

$$e^{-at} \Theta_X(t) = e^{-at} \frac{\lambda}{\lambda - t} \quad \text{for } t < \lambda$$

Differentiate to find  $\min_{t > 0} e^{-at} \Theta_X(t)$

$$\frac{d}{dt} [e^{-at} \Theta_X(t)] = e^{-at} \left( \frac{-a\lambda(\lambda - t) + \lambda}{(\lambda - t)^2} \right) = 0 \quad \text{when } t = \lambda - \frac{1}{a}$$

Recall that we need  $t > 0$  &  $t < \lambda$ . When  $t = \lambda - \frac{1}{a}$ , implies  $\boxed{a > \frac{1}{\lambda}}$

$$P[X \geq a] \leq \min_{\substack{t > 0 \\ t < \lambda}} e^{-at} \Theta_X(t) = e^{-a(\lambda - \frac{1}{a})} \frac{\lambda}{\lambda - (\lambda - \frac{1}{a})} = \boxed{e^{-a\lambda + 1} a \lambda}$$

See plots on following pages...

$$2) E[Y] = E[atX + b] = a^t E[X] + b = a^t \mu + b = (2, -1, 2) \begin{pmatrix} 5 \\ -5 \\ 6 \end{pmatrix} + 5 = 32$$

$$\text{var}(Y) = E[(Y - E[Y])^2] = E[(atX + b - a^t \mu + b)^2] = E[(a^t(X - \mu))^2] \\ = a^t E[(X - \mu)(X - \mu)^t] a = a^t K a = 25$$

$$3) a) \text{ Must have } K_{XX} = K_{XX}^t$$

b) Diagonal elements of  $K_{XX}$  are variances, which must be positive

c) Off-diagonal elements of  $K_{XX}$  are  $\text{cov}(X_i, X_j)$ , which are  $\mathbb{R}$  when  $X \in \mathbb{R}^3$

d) By the Schwarz inequality

$$\text{cov}^2(X_1, X_2) = E^2[(X_1 - \mu_1)(X_2 - \mu_2)] \leq E[(X_1 - \mu_1)^2] E[(X_2 - \mu_2)^2] = \text{var}(X_1) \text{var}(X_2)$$

but in this example,  $\text{cov}^2(X_1, X_2) = (-4)^2 = 16$  and  $\text{var}(X_1) \text{var}(X_2) = 2 \cdot 3 = 6$

e) Again, (a) fails since now  $K_{XX} \neq K_{XX}^H$  and (b) & (d) fail as before

But (c) satisfies  $K_{XX} = K_{XX}^H$  and is positive definite, thus a valid cov matrix