

Theorem If X_1, X_2, \dots, X_n are mutually independent exponential random variables each with mean $\alpha > 0$, then $X = \frac{2}{\alpha} \sum_{i=1}^n X_i$ is a $\chi^2(2n)$ random variable.

Proof Since the random variables X_1, X_2, \dots, X_n have the exponential distribution with mean α , they each have probability density function

$$f_{X_i}(x) = \frac{1}{\alpha} e^{-x/\alpha} \quad x > 0,$$

for $i = 1, 2, \dots, n$. The moment generation function of X_i is

$$M_{X_i}(t) = (1 - \alpha t)^{-1} \quad t < \frac{1}{\alpha},$$

for $i = 1, 2, \dots, n$. Since the random variables X_1, X_2, \dots, X_n are mutually independent, the moment generation function of $X = \frac{2}{\alpha} \sum_{i=1}^n X_i$ is

$$\begin{aligned} M_X(t) &= E[e^{tX}] \\ &= E[e^{(2t/\alpha) \sum_{i=1}^n X_i}] \\ &= E[e^{t \sum_{i=1}^n 2X_i/\alpha}] \\ &= E[e^{t \cdot 2X_1/\alpha} e^{t \cdot 2X_2/\alpha} \dots e^{t \cdot 2X_n/\alpha}] \\ &= E[e^{t \cdot 2X_1/\alpha}] E[e^{t \cdot 2X_2/\alpha}] \dots E[e^{t \cdot 2X_n/\alpha}] \\ &= M_{X_1}(2t/\alpha) M_{X_2}(2t/\alpha) \dots M_{X_n}(2t/\alpha) \\ &= (1 - 2t)^{-1} (1 - 2t)^{-1} \dots (1 - 2t)^{-1} \\ &= (1 - 2t)^{-n} \quad t < 1/2, \end{aligned}$$

which is the moment generation function of a $\chi^2(2n)$ random variable.

APPL verification: The following APPL statements verify a special case ($n = 3$) of this result.

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assume(alpha > 0);
X1 := [[x -> exp(-x / alpha) / alpha], [0, infinity], ["Continuous", "PDF"]];
X2 := [[x -> exp(-x / alpha) / alpha], [0, infinity], ["Continuous", "PDF"]];
X3 := [[x -> exp(-x / alpha) / alpha], [0, infinity], ["Continuous", "PDF"]];
Y := Convolution(X1, X2, X3);
g := [[x -> 2 * x / alpha], [0, infinity]];
X := Transform(Y, g);
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In this case X has a $\chi^2(6)$ distribution.