

7.12) Since X_1, X_2 are independent, the joint probability distribution is:

$$\Rightarrow f(x_1, x_2) = f(x_1)f(x_2) = (e^{-x_1})(e^{-x_2}) = e^{-(x_1+x_2)}, x_1 > 0, x_2 > 0$$

$$\Rightarrow y_1 = x_1 + x_2, y_2 = \frac{x_1}{x_1 + x_2}$$

$$\Rightarrow x_1 = y_2(x_1 + x_2), \therefore x_2 = y_1(1 - y_2)$$

$$\therefore x_1 = y_1 y_2, \text{ for } y_1 > 0, 0 < y_2 < 1$$

$$\Rightarrow J = \begin{vmatrix} \frac{\partial x_1}{\partial y_1} & \frac{\partial x_1}{\partial y_2} \\ \frac{\partial x_2}{\partial y_1} & \frac{\partial x_2}{\partial y_2} \end{vmatrix} = \begin{vmatrix} y_2 & y_1 \\ 1 - y_2 & -y_1 \end{vmatrix} = -y_1 y_2 - y_1 + y_1 y_2 = -y_1$$

$$\Rightarrow g(y_1, y_2) = f(x_1, x_2) |J| = e^{-(y_1 y_2 + y_1(1 - y_2))} |-y_1| = y_1 e^{-y_1}, \text{ for } y_1 > 0, 0 < y_2 < 1$$

$$\Rightarrow g(y_1) = \int_0^1 y_1 e^{-y_1} dy_2 = y_1 e^{-y_1}, y_1 > 0$$

$$\Rightarrow g(y_2) = \int_0^\infty y_1 e^{-y_1} dy_1 = \Gamma(2) = 1, 0 < y_2 < 1$$

$$\Gamma(\alpha) = \int_0^\infty x^{\alpha-1} e^{-x} dx, \text{ for } \alpha > 0$$

$$\Rightarrow g(y_1, y_2) = y_1 e^{-y_1} = g(y_1)g(y_2) \quad \#$$

\therefore Since $g(y_1, y_2) = g(y_1)g(y_2)$, \therefore random variables Y_1 and Y_2 are independent.

7.14) $\because y = x^2$
 $\therefore x_1 = \sqrt{y}, x_2 = -\sqrt{y} \Rightarrow f(x) = \begin{cases} \frac{1+x}{2}, & -1 < x < 1 \\ 0, & \text{elsewhere} \end{cases}$

$$\Rightarrow J_1 = \frac{1}{2\sqrt{y}}, J_2 = -\frac{1}{2\sqrt{y}}$$

$$\Rightarrow g(y) = f(\sqrt{y})|J_1| + f(-\sqrt{y})|J_2| = \frac{1+\sqrt{y}}{2} \left(\frac{1}{2\sqrt{y}}\right) + \frac{1-\sqrt{y}}{2} \left(\frac{1}{2\sqrt{y}}\right)$$

$$= \frac{2}{4\sqrt{y}} = \frac{1}{2\sqrt{y}}, 0 < y < 1$$

$$\therefore g(y) = \begin{cases} \frac{1}{2\sqrt{y}}, & 0 < y < 1 \\ 0, & \text{elsewhere} \end{cases} \quad \#$$

7.18) \Rightarrow moment-generating function of X is

$$\Rightarrow M_X(t) = E(e^{tx}) = \sum_{x=1}^{\infty} e^{tx} f(x) = \sum_{x=1}^{\infty} e^{tx} p q^{x-1}$$

$$\hookrightarrow \text{if } \sum_{x=0}^{\infty} (qe^t)^x \neq \infty,$$

$$\Rightarrow |qe^t| < 1$$

$$e^t < \frac{1}{q}$$

$$\therefore t < \ln q^{-1} = -\ln q \quad \therefore t < -\ln q$$

$$= \frac{p}{q} \sum_{x=1}^{\infty} e^{tx} q^x$$

$$= \frac{p}{q} \sum_{x=1}^{\infty} (e^t q)^x \quad (\text{geometric series})$$

$$= \frac{p}{q} \left(\frac{qe^t}{1-qe^t} \right)$$

$$\therefore M_X(t) = \frac{pe^t}{1-qe^t}, \quad t < -\ln q, \quad \because 1-qe^t \neq 0, \quad \therefore t \neq \ln \frac{1}{q}$$

$$\Rightarrow \mu = \left. \frac{dM_X(t)}{dt} \right|_{t=0}$$

$$= \left. \frac{(1-qe^t)(pe^t) - (pe^t)(-qe^t)}{(1-qe^t)^2} \right|_{t=0} = \frac{(1-q)p + pq}{(1-q)^2} = \frac{p-pq+pq}{(1-q)^2} = \frac{p}{(1-q)^2}$$

$$\because p=1-q \quad \therefore \mu = \frac{1}{p} \quad \#$$

$$\Rightarrow \mu'_2 = \left. \frac{d^2 M_X(t)}{dt^2} \right|_{t=0} = \frac{(1-qe^t)^2 (pe^t) - (pe^t)(2)(1-qe^t)(-qe^t)}{(1-qe^t)^4} \Big|_{t=0}$$

$$= \frac{(1-q)^2 p + 2pq(1-q)}{(1-q)^4}$$

$$= \frac{p-pq+2pq}{(1-q)^3} = \frac{p+pq}{(1-q)^3} = \frac{p(1+q)}{p^3} = \frac{1+q}{p^2}$$

$$\therefore \sigma^2 = \mu'_2 - \mu^2 = \frac{1+q}{p^2} - \frac{1}{p^2} = \frac{q}{p^2} \quad \#$$

7.22) $\Rightarrow M_X(t) = (1-2t)^{-\frac{v}{2}}$ (moment-generating function)

$$\Rightarrow \mu = \left. \frac{dM_X(t)}{dt} \right|_{t=0} = -\frac{v}{2} (1-2t)^{-\frac{v}{2}-1} \cdot (-2) \Big|_{t=0}$$

$$= -\frac{v}{2} \cdot (-2) = v \quad \#$$

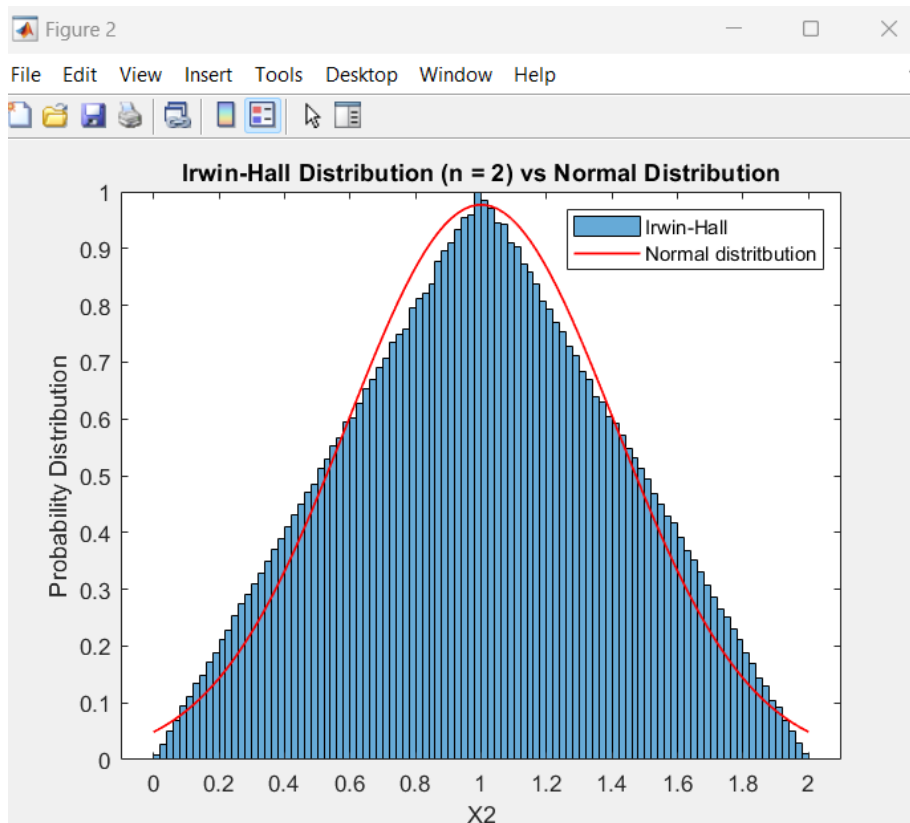
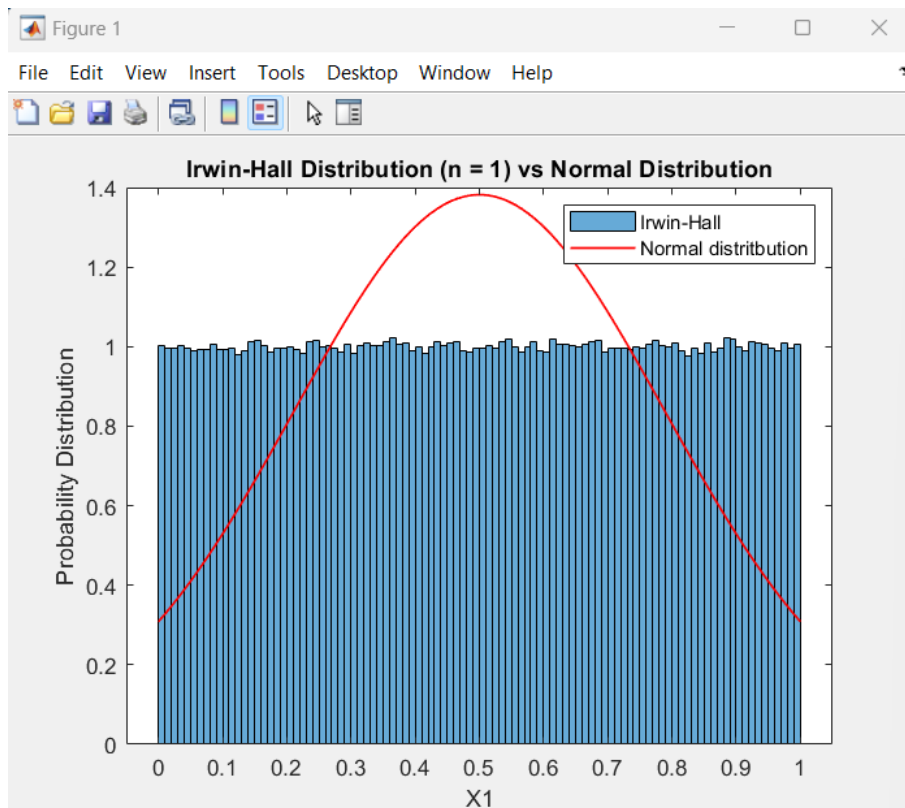
$$\Rightarrow \mu'_2 = \left. \frac{d^2 M_X(t)}{dt^2} \right|_{t=0} = \left(-\frac{v}{2} - 1 \right) (v) (1-2t)^{-\frac{v}{2}-2} \cdot (-2) \Big|_{t=0}$$

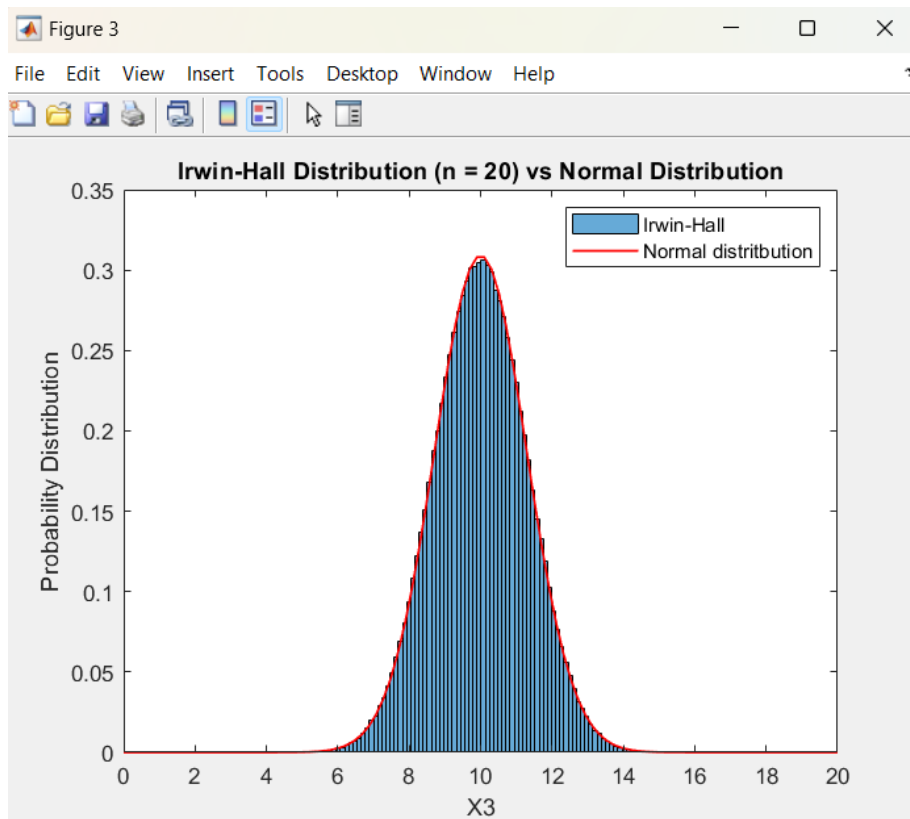
$$= (v+2)(v) = v^2 + 2v$$

$$\Rightarrow \sigma^2 = \mu'_2 - \mu^2 = v^2 + 2v - v^2 = 2v \quad \#$$

1)

b)





We can observe that as n increases, the Irwin-Hall distribution becomes more approximate and less errors to a normal distribution. For $n = 1$, the Irwin-Hall distribution is a uniform distribution, and it deviates significantly from the normal distribution. However, as n increases to $n = 2$ and $n = 20$, the Irwin-Hall distribution becomes more bell-shaped and the approximation and the error rate to the normal distribution improves. The errors in approximating a normal distribution with the Irwin-Hall distribution decrease as n gets larger.