

(7.12)

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

$$y_2 = \frac{x_1}{x_1+x_2} = \frac{x_1}{y_1} \Rightarrow x_1 = y_1 y_2 \Rightarrow \begin{cases} x_1 = y_1 y_2 \\ x_2 = y_1(1-y_2) \end{cases}, \text{ for } y_1 > 0, 0 < y_2 < 1$$

$$x_2 = y_1 - x_1 = y_1 - y_1 y_2 = y_1(1-y_2)$$

$$J = \begin{vmatrix} y_2 & y_1 \\ 1-y_2 & -y_1 \end{vmatrix} = -y_1 y_2 - y_1 + y_2 y_1 = -y_1$$

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

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

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

Since $g(y_1, y_2) = g(y_1)g(y_2)$, the random variables Y_1 and Y_2 are independent #

(7.14)

$$x_1 = \sqrt{y}, J_1 = \frac{1}{2\sqrt{y}}$$

$$x_2 = -\sqrt{y}, J_2 = -\frac{1}{2\sqrt{y}}$$

$$|g(y)| = f(\sqrt{y})|J_1| + f(-\sqrt{y})|J_2| = \frac{1+\sqrt{y}}{2} \frac{1}{2\sqrt{y}} + \frac{1-\sqrt{y}}{2} \frac{1}{2\sqrt{y}} = \frac{2}{4\sqrt{y}} = \frac{1}{2\sqrt{y}}, \text{ for } 0 < y < 1$$

$$\Rightarrow g(y) = \frac{1}{2\sqrt{y}}, \text{ for } 0 < y < 1 \#$$

(7.18)

$$M_X(t) = E(e^{tx}) = p \sum_{x=1}^{\infty} e^{tx} x^{-1} = \frac{p}{1} \sum_{x=1}^{\infty} (e^t q)^x = \frac{pe^t}{1-qe^t}, t < \ln q \#$$

$$M' = M'_X(0) = \frac{pe^t(1-qe^t) + pe^t qe^t}{(1-qe^t)^2} \Big|_{t=0} = \frac{p(1-q) + pq}{(1-q)^2} = \frac{1}{p}$$

$$M''_2 = M''_X(0) = \frac{(1-qe^t)^2 pe^t + 2pqe^{2t}(1-qe^t)}{(1-qe^t)^4} \Big|_{t=0} = \frac{(1-q)^2 p + 2pq(1-q)}{(1-q)^4} = \frac{2-p}{p^2}$$

$$\sigma^2 = \frac{2-p}{p^2} - \frac{1}{p^2} = \frac{1-p}{p^2} = \frac{q}{p^2}, M = \frac{1}{p} \#$$

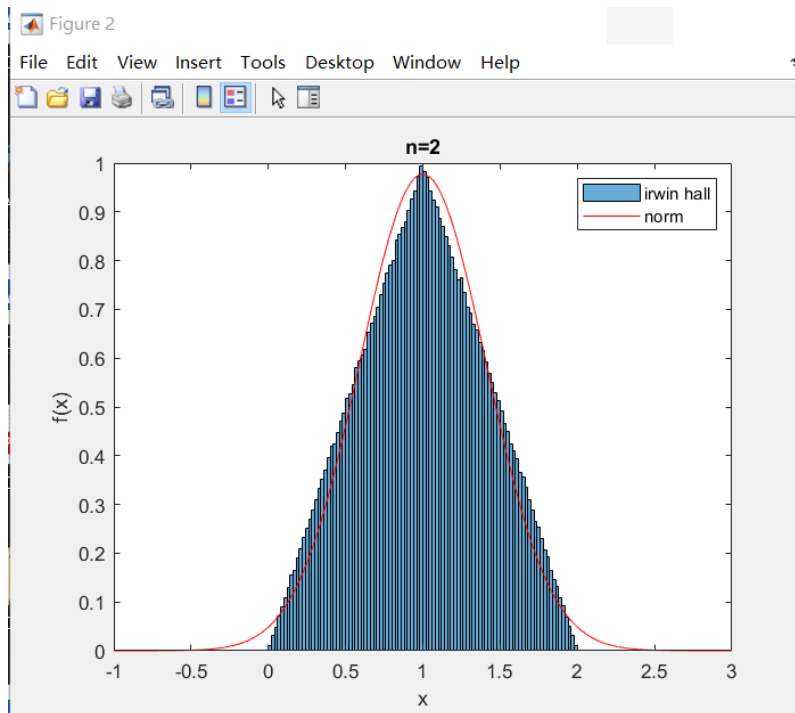
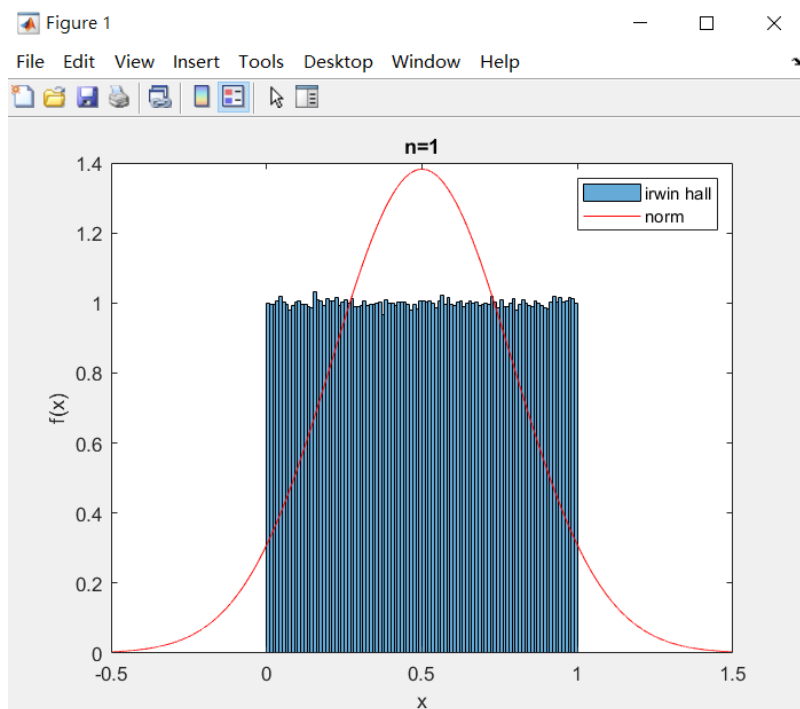
$$(7.22) M_X(t) = E(e^{tx}) = \frac{1}{\Gamma(\frac{V}{2}) 2^{\frac{V}{2}}} \int_0^\infty x^{\frac{V}{2}-1} e^{-(\frac{1}{2}-t)x} dx = \frac{(\frac{1}{2}-t)^{-\frac{V}{2}}}{2^{\frac{V}{2}}} \int_0^\infty x^{\frac{V}{2}-1} e^{-\frac{1}{2}tx} dx = (1-t)^{-\frac{V}{2}}$$

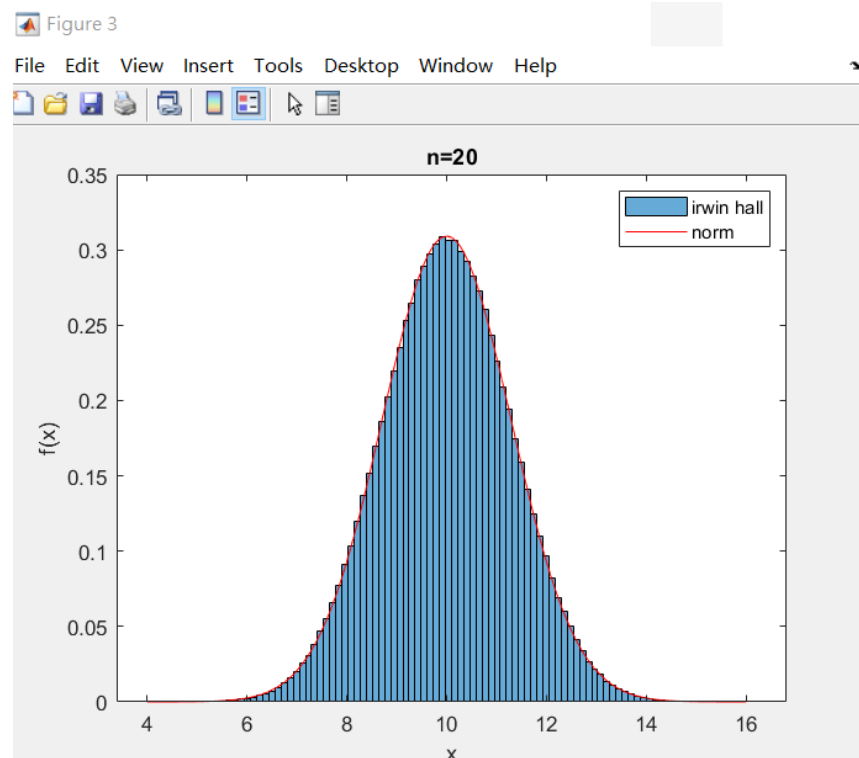
$$M = -\frac{V}{2} (1-t)^{-\frac{V}{2}-1} (-1) \Big|_{t=0} = V$$

$$M'_2 = M''_X(0) = \left(V(1-t)^{-\frac{V}{2}-1} \right)' = \left(-\frac{V}{2}-1 \right) V(1-t)^{-\frac{V}{2}-2} (-1) \Big|_{t=0} = V(V+2)$$

$$M = V \#, \sigma^2 = V(V+2) - V^2 = 2V \#$$

1.(b)





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