

第六次機率與統計作業

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一、手寫部分：

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7.12

$$1^{\circ} \begin{cases} Y_1 = X_1 + X_2 \\ Y_2 = \frac{X_1}{(X_1 + X_2)} \end{cases} \Rightarrow \begin{cases} Y_1 = X_1 + X_2 \\ Y_2(X_1 + X_2) = X_1 \end{cases} \Rightarrow \begin{cases} Y_1 = X_1 + X_2 \\ 0 = (1 - Y_2)X_1 - X_2 \end{cases} \Rightarrow \begin{cases} Y_1 Y_2 = X_1 Y_2 + X_2 Y_2 \quad -\textcircled{1} \\ 0 = (1 - Y_2)X_1 - X_2 Y_2 \quad -\textcircled{2} \end{cases}$$

$\textcircled{1} + \textcircled{2}$

$$\Rightarrow Y_1 Y_2 = X_1 - X_1 Y_2 + X_1 Y_2 \Rightarrow X_1 = Y_1 Y_2 \quad -\textcircled{1}$$

Substitute $\textcircled{1}$ to $\textcircled{2}$

$$\Rightarrow Y_1 Y_2 = Y_1 Y_2 + X_2 Y_2 \Rightarrow X_2 = Y_1 - Y_1 Y_2$$

$$(e^{Y_1 Y_2})(e^{-(Y_1 - Y_1 Y_2)}) = e^{Y_1}$$

By Jacobian,

$$J = \begin{vmatrix} \frac{\partial Y_1}{\partial X_1} & \frac{\partial Y_1}{\partial X_2} \\ \frac{\partial Y_2}{\partial X_1} & \frac{\partial Y_2}{\partial X_2} \end{vmatrix} = \begin{vmatrix} 1 & 1 \\ -\frac{Y_2}{X_1 + X_2} & \frac{X_1 - X_2}{(X_1 + X_2)^2} \end{vmatrix} = -Y_2$$

Therefore, we can write

$$g(Y_1, Y_2) = f(X_1, X_2) \cdot |J| = (*)$$

Since X_1 & X_2 are independent

$$\therefore (*) = f(X_1) \cdot f(X_2) \cdot |J| = e^{-X_1} \cdot e^{-X_2} \cdot Y_2 = Y_2 e^{-Y_1} = g(Y_1, Y_2), \text{ for } Y_1 > 0 \text{ \& } 0 < Y_2 < 1;$$

also, by the property of $f(x)$, then for otherwise, $g(Y_1, Y_2) = 0$.

2 $^{\circ}$ To prove Y_1, Y_2 independent, apply $g(Y_1, Y_2) = h(Y_1) \cdot h(Y_2)$:

$$h(Y_1) = \int_0^1 g(Y_1, Y_2) dY_2 = \int_0^1 Y_2 e^{-Y_1} dY_2 = Y_1 e^{-Y_1}$$

$$h(Y_2) = \int_0^{\infty} g(Y_1, Y_2) dY_1 = \int_0^{\infty} Y_2 e^{-Y_1} dY_1 = \Gamma(2) = 1$$

$$\therefore \text{we can get } h(Y_1) \times h(Y_2) = Y_1 e^{-Y_1} \cdot 1 = Y_1 e^{-Y_1} = g(Y_1, Y_2)$$

Therefore, Y_1, Y_2 are independent #

7.14

By the probability distribution of the random variable $Y = X^2$

\Rightarrow change to $X = \pm\sqrt{Y}$, for $0 < Y < 1$

And by the property of $f(x)$, we can get

$$\text{for } 0 < X_1 < 1: Y = X_1^2 \Rightarrow X_1 = \sqrt{Y}$$

$$\text{for } -1 < X_2 < 0: Y = X_2^2 \Rightarrow X_2 = -\sqrt{Y}$$

Then,

$$J_1 = X_1 \frac{dX_1}{dY} = \frac{1}{2\sqrt{Y}}$$

$$J_2 = X_2 \frac{dX_2}{dY} = -\frac{1}{2\sqrt{Y}}$$

Therefore, we can get the probability distribution

$$\begin{aligned} g(y) &= f(X_1) |J_1| + f(X_2) |J_2| \\ &= \frac{1+\sqrt{y}}{2} \times \frac{1}{2\sqrt{y}} + \frac{1-\sqrt{y}}{2} \times \frac{1}{2\sqrt{y}} \\ &= \frac{1}{2\sqrt{y}} \left(\frac{(1+\sqrt{y}) + (1-\sqrt{y})}{2} \right) = \frac{1}{2\sqrt{y}} \times 1 \\ &= \frac{1}{2\sqrt{y}}, \text{ for } 0 < y < 1 \end{aligned}$$

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

7.18

1° By moment generating function,

$$M_X(t) = E(e^{tx}) = \sum_{x=1}^{\infty} e^{tx} g(x) = \sum_{x=1}^{\infty} e^{tx} p q^{x-1} = p e^t \sum_{x=1}^{\infty} e^{t(x-1)} q^{x-1} = p e^t \sum_{x=0}^{\infty} e^{tx} q^x = p e^t \sum_{x=0}^{\infty} (q e^t)^x$$

$$= p e^t \cdot \frac{1}{1 - q e^t} = \frac{p e^t}{1 - q e^t}$$

Also, if $\sum_{x=0}^{\infty} (q e^t)^x$ has value (not ∞), then

$$|q e^t| < 1 \Rightarrow e^t < \frac{1}{q} \Rightarrow t < \ln(q^{-1}) \Rightarrow t < -\ln q$$

2° Find the mean of X :

$$\mu_X = E(X) = \left. \frac{dM_X(t)}{dt} \right|_{t=0} = \left. \frac{d}{dt} \left(\frac{p e^t}{1 - q e^t} \right) \right|_{t=0} = \left. \frac{p e^t (1 - q e^t) - p e^t (-q e^t)}{(1 - q e^t)^2} \right|_{t=0}$$

$$= \left. \frac{p e^t}{(1 - q e^t)^2} \right|_{t=0} = \frac{p}{(1 - q)^2} \quad \text{by } q = 1 - p \quad \frac{p}{p^2} = \frac{1}{p} \quad \#$$

(p-1)²
= p² - 2p + 1

3° Find the variance of X :

$$\text{Var}(X) = E(X^2) - E(X)^2 = \left. \frac{d^2 M_X(t)}{dt^2} \right|_{t=0} - \left(\frac{1}{p} \right)^2 = \left. \frac{d}{dt} \left(\frac{p e^t}{(1 - q e^t)^2} \right) \right|_{t=0} - \frac{1}{p^2}$$

$$= \left. \frac{p e^t (1 - q e^t)^{-2} - p e^t \cdot 2(1 - q e^t)^{-3} \cdot (-q e^t)}{(1 - q e^t)^4} \right|_{t=0} - \frac{1}{p^2} \quad p - p + 2p^2 - p^2$$

$$= \left. \frac{p e^t - p q^2 e^t}{(1 - q e^t)^4} \right|_{t=0} - \frac{1}{p^2} = \frac{p - p q^2}{(1 - q)^4} - \frac{1}{p^2} \quad \text{by } q = 1 - p \quad \frac{p - p(1-p)^2}{p^4} - \frac{1}{p^2} = \frac{7p^3 - p^3 - p^2}{p^4}$$

$$= \frac{1-p}{p} \quad \text{by } p = 1 - q \quad \frac{q}{p} \quad \#$$

7.22

From Exercise 7.21, we know the random variable X having chi-squared distribution with ν degrees of freedom is $M_X(t) = (1 - 2t)^{-\nu/2}$

1° The mean of random variable X :

$$\mu_X = \left. \frac{dM_X(t)}{dt} \right|_{t=0} = \left. \frac{d}{dt} (1 - 2t)^{-\nu/2} \right|_{t=0} = \left(-\frac{\nu}{2} \cdot (1 - 2t)^{-\nu/2 - 1} \cdot (-2) \right) \Big|_{t=0} = \nu$$

2° The variance of random variable X :

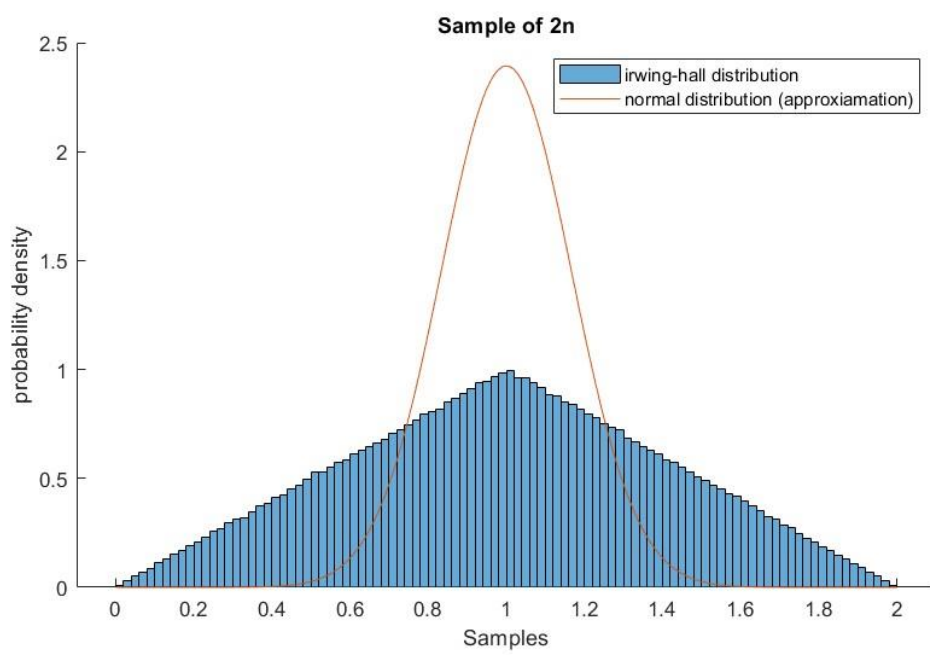
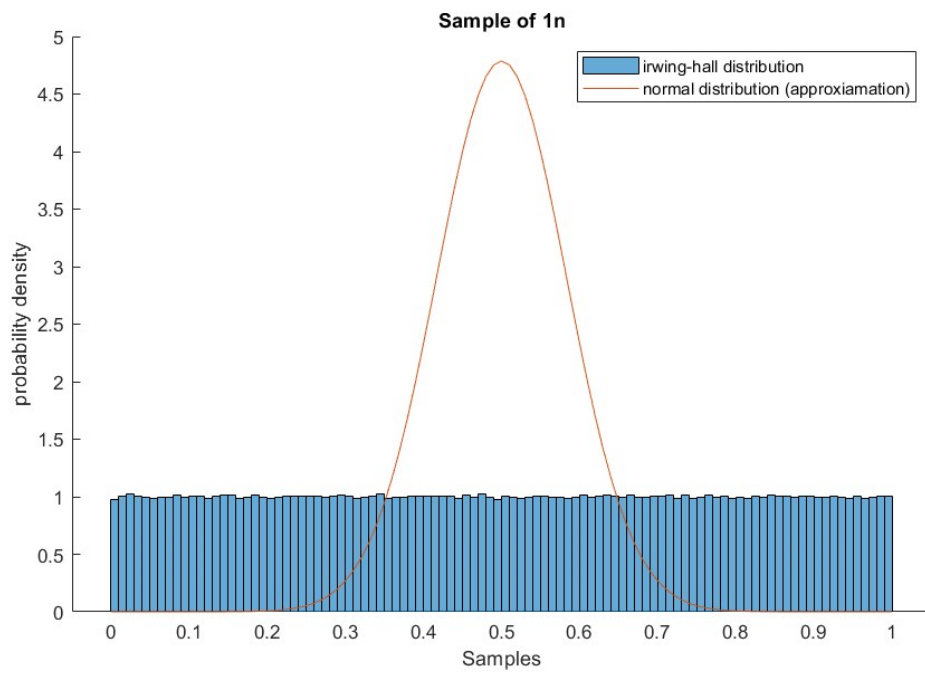
$$\sigma_X^2 = E(X^2) - E(X)^2 = \left. \frac{d^2 M_X(t)}{dt^2} \right|_{t=0} - \nu^2 = \left. \frac{d}{dt} (\nu (1 - 2t)^{-\nu/2 - 1}) \right|_{t=0} - \nu^2$$

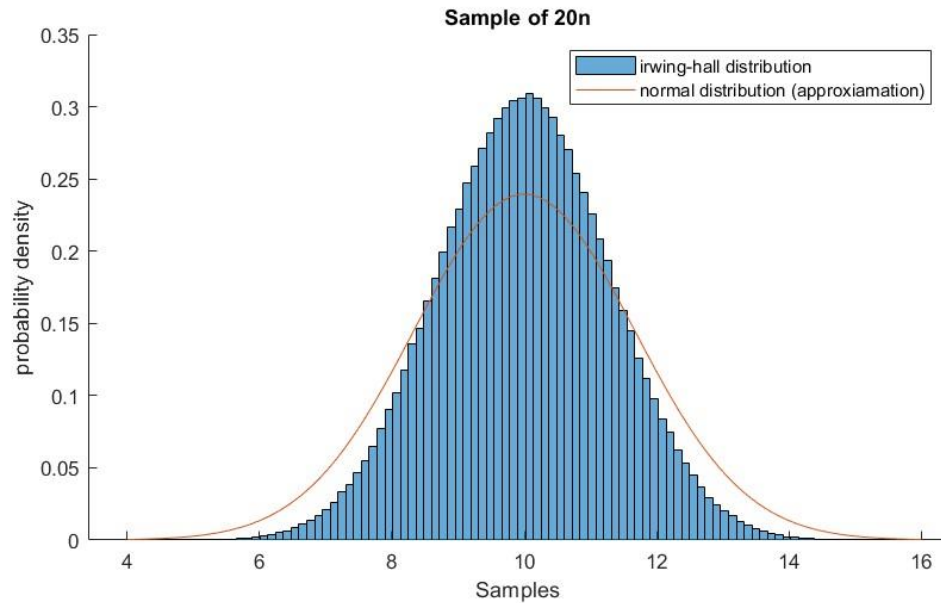
$$= \left(\nu \left(-\frac{\nu}{2} - 1 \right) (1 - 2t)^{-\nu/2 - 2} \cdot (-2) \right) \Big|_{t=0} - \nu^2 = \nu^2 + 2\nu - \nu^2 = 2\nu$$

\therefore we show that the mean and variance of the chi-squared distribution with ν degrees of freedom are, respectively, ν and 2ν #

二、matlab 部分：

HW7_1b：





從三個結果圖來說，irwing-hall distribution 在 $n = 1$ 時確實是 uniform distribution； $n = 2$ 時也是題目說的 triangular distribution；而 $n = 20$ 的 case 確實有點像 normal distribution。其中，在經過 normal distribution approximation 後，可以發現到，在 irwing-hall distribution 在 $n = 1$ 及 $n = 2$ 的時候的 irwing-hall distribution histogram 和 normal distribution approximation plot 的波形相差很多，代表著失真率大，其中尤以 $n = 1$ 的 case 最為嚴重。相反的，而在 $n = 20$ 時，由於 n 有著一定的大小，因此和 normal distribution approximation plot 相比時，他們波形比率相差的並不大，也就是說這個 case 是可以被 normal distribution 做近似的，和作業 pdf 中的說法完全吻合