

$$\begin{aligned}
h(z) &= (f * g)(z) \\
&= \int_{-\infty}^{\infty} f(z-x) g(x) dx \\
&= \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-\frac{(z-x)^2}{2}} e^{-\frac{x^2}{2}} dx \\
&= \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-\frac{(z-x)^2}{2} - \frac{x^2}{2}} dx \\
&= \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-\frac{z^2}{4} - (x - \frac{z}{2})^2} dx \\
&= \frac{1}{2\pi} e^{-\frac{z^2}{4}} \int_{-\infty}^{\infty} e^{-(x - \frac{z}{2})^2} dx
\end{aligned}$$

Let $\frac{w}{\sqrt{2}} = x - \frac{z}{2} \Rightarrow \frac{1}{\sqrt{2}} dw = dx$, and $\frac{w^2}{2} = (x - \frac{z}{2})^2$

$$\begin{aligned}
h(z) &= \frac{1}{2\pi} e^{-\frac{z^2}{4}} \int_{-\infty}^{\infty} e^{-\frac{w^2}{2}} \cdot \frac{1}{\sqrt{2}} dw \\
&= \frac{1}{2\pi} e^{-\frac{z^2}{4}} \cdot \sqrt{\pi} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{w^2}{2}} \\
&= \frac{1}{2\sqrt{\pi}} e^{-\frac{z^2}{4}} = \frac{1}{\sqrt{2\pi} \sqrt{2}} e^{-\frac{1}{2} \left(\frac{z-0}{\sqrt{2}} \right)^2}
\end{aligned}$$

Hence, sum of two standard Normal r.v's is again a normal r.v. with mean zero and variance 2.

Example (50)

what is the probability density of the sum of two independent r.v's, each of which is Cauchy?

Solution: Let $Z = X + Y$, where $X \sim \text{Cauchy}(0, 1)$ and $Y \sim \text{Cauchy}(0, 1)$. Hence the density function $f(x)$ of the r.v. X and Y is given by

$$f(x) = \frac{1}{\pi(1+x^2)} \quad \text{and} \quad g(y) = \frac{1}{\pi(1+y^2)}$$

since X and Y are independent, the density function of Z can be obtained by the method of convolution. The sum $Z = x + y$ is between $(-\infty)$ and $(+\infty)$. Hence, the density function of Z is:

$$\begin{aligned}
 h(z) &= (f * g)(z) \\
 &= \int_{-\infty}^{\infty} f(z-x) g(x) dx \\
 &= \int_{-\infty}^{\infty} \frac{1}{\pi[1+(z-x)^2]} \cdot \frac{1}{\pi(1+x^2)} dx \\
 &= \frac{1}{\pi^2} \int_{-\infty}^{\infty} \frac{1}{1+(z-x)^2} \cdot \frac{1}{1+x^2} dx
 \end{aligned}$$

we decompose the above integral using partial fraction decomposition. Hence

$$\frac{1}{1+(z-x)^2} \cdot \frac{1}{1+x^2} = \frac{2Ax+B}{1+x^2} + \frac{2C(z-x)+D}{1+(z-x)^2}$$

where; $A = \frac{1}{z(4+z^2)} = C$ and $B = \frac{1}{4+z^2} = D$

Then;

$$\begin{aligned}
 h(z) &= \frac{1}{\pi^2 z^2 (4+z^2)} \left[z \ln \left(\frac{1+x^2}{1+(z-x)^2} \right) + \frac{z^2}{2} \tan^{-1} x - \frac{z^2}{2} \tan^{-1} (z-x) \right]_{-\infty}^{\infty} \\
 &= \frac{1}{\pi^2 z^2 (4+z^2)} [0 + z^2 \pi + z^2 \pi] \\
 &= \frac{2}{\pi(4+z^2)}
 \end{aligned}$$

Hence, the sum of two independent Cauchy r.v.'s is not a Cauchy r.v.

4. Moment Generating Function Method: $\frac{1}{2}$

We know that if X and Y are independent r.v.'s, then

$$M_{X+Y}(t) = M_X(t) M_Y(t)$$

Example (51) Let $X \sim P(\lambda_1)$ and $Y \sim P(\lambda_2)$. What is the probability function of $X+Y$ if X and Y are independent.

Solution: Since $X \sim P(\lambda_1)$ and $Y \sim P(\lambda_2)$, we get:

$$M_X(t) = e^{\lambda_1(e^t - 1)} \text{ and } M_Y(t) = e^{\lambda_2(e^t - 1)}$$

$$M_{X+Y}(t) = \frac{\lambda_1 (e^t - 1)}{e} \frac{\lambda_2 (e^t - 1)}{e}$$

$$= e^{-(\lambda_1 + \lambda_2)} (e^t - 1)$$

that is $X+Y \sim p(\lambda_1 + \lambda_2)$. Hence the density Function of $Z = X+Y$ is given by:

$$h(z) = \begin{cases} \frac{e^{-(\lambda_1 + \lambda_2)}}{z!} (\lambda_1 + \lambda_2)^z, & z = 0, 1, 2, \dots \\ 0 & \text{o.w} \end{cases}$$

Ex

Example (52) what is the p.d.f. of the sum of two independent r.v's, each of which is gamma distribution with θ and α ?

Solution:

Let X and Y be two independent gamma r.v's with parameters θ and α , that is:

$$X \sim \text{Gam}(\theta, \alpha) \text{ and } Y \sim \text{Gam}(\theta, \alpha).$$

Hence;

$$M_X(t) = (1 - \theta)^{-\alpha}$$

$$M_Y(t) = (1 - \theta)^{-\alpha}$$

$$M_{X+Y}(t) = M_X(t) M_Y(t)$$

$$= (1 - \theta)^{-\alpha} (1 - \theta)^{-\alpha}$$

$$= (1 - \theta)^{-2\alpha}$$

Thus $X+Y$ has a m.g.f. of Gamma r.v. with parameters θ and 2α . Therefore

$$X+Y \sim \text{Gam}(\theta, 2\alpha).$$

M.Sc. Test. Advanced Mathematical Statistics
Prof. Dr. Jawad Al. Mosawi

1. Let $X \sim f(x) = \frac{2c}{3^x}$, $x=1, 2, \dots$ for some constant c . what is the probability that X is odd?

2. Let $X \sim f(x) = \frac{2}{x^2}$, $1 < x < 2$, find the p.d.f. of $Y = \sqrt{x}$.

3. Let $(X, Y) \sim p(x, y) = \frac{x-y}{20}$, $x = y, 1, 2, 3, 4$
 $y = 0, 1, \dots, x$ $\frac{1}{20}$

Find the marginal Conditional probability of X given $Y=y$.

4. Let $X \sim f(x) = [\pi(1 + \{x-0\}^2)]^{-1}$, $x \in \mathbb{R}^1$.
Find the c.d.f. of X .

5. Let $(x, y) \sim f(x, y) = 4x$, $0 < x < \sqrt{y} < 1$, what is the density function of $\frac{x}{y}$?

6. Let X and Y distributed as bivariate Bernoulli dist.
i.e. $(X, Y) \sim \text{Ber}(\theta, \lambda)$.
Find $E(X/Y)$.

- Good Luck -


د. جواد الموسوي
c.c. / 11 / 3.

M.Sc. Course Advanced Mathematical Statistics
Prof. Dr. Jawad Al-Mosawi

1. A continuous r.v. X has the following density function:

$$f(x) = \begin{cases} \frac{3x^2}{\theta^3} & , 0 < x < \theta, \theta > 0 \\ 0 & , o.w \end{cases}$$

what is the probability of X less than the ratio of the mode to the median, so that the mode of this distⁿ is 0? $P(X \leq \frac{0}{\sqrt{2}}) = \left(\frac{3\sqrt{2}}{2} \right)^2$

$$P\left(X < \frac{\theta}{3\sqrt{2}}\right) = \int_0^{\frac{\theta}{3\sqrt{2}}} \frac{3x^2}{\theta^3} dx$$

2. Let $(X, Y) \sim \text{Ber}(\alpha, \beta)$. Use the joint moment generating function to find the $\text{Var}(XY=y)$.

3. Let $f(x) = a e^{-ax}$, $x > 0$. Find $M_x(-3a)$.

4. Let $X \sim U(0, 10)$. Compute $\Pr[X + \frac{10}{X} > 7]$.

5. Let $f(x, y) = \begin{cases} 24xy & , x > 0, y > 0, 0 < x+y < 1. \\ 0 & \text{elsewhere} \end{cases}$

what is the conditional probability $P(X < \frac{1}{2} / Y = \frac{1}{4})$?

6. Let $f(x, y) = 4x$, $0 < x < \sqrt{y} < 1$.

what is the density function of $X-Y$?

- Good Luck -

ا. د. پرواد کاظم حقیر

5.55 / 1 / 4

Prof.Dr.

First attempt

Department of Statistics

Jawad kadhim khudhier

advanced mathematical statistics exam

Master science degree

2/4/2023

1. Let X and Y be two jointly random variables with j.p.d.f. :

$$f(x, y) = \begin{cases} 4x & , 0 < x < \sqrt{y} < 1 \\ 0 & , o. w. \end{cases}$$

What is the distribution of $\frac{X}{Y}$?

2. If the random variable X is distributed as Normal with mean (1) and standard deviation (2), then what is $P(X^2 - 2X \leq 8)$.

3. Let

$$f(x) = \begin{cases} \frac{x}{4a} & , 0 < x \leq a \\ 0 & , o. w. \end{cases}$$

Find the 25th percentile of the distribution.

4. Let X_1 and X_2 be two random variables with the following joint pmf:

$$p(x_1, x_2) = \begin{cases} \binom{x_2}{x_1} \left(\frac{1}{2}\right)^{x_1} \left(\frac{x_1}{15}\right) & , x_1 = 1, \dots, 5 \\ 0 & , x_2 = 0, \dots, x_1 \\ & , o. w. \end{cases}$$

Find $E(X_2)$.

5. Let X and Y be two jointly random variables with j.p.d.f. :

$$f(x, y) = \begin{cases} 2 & , 0 < y < x - 1 ; \text{ and } 0 < x < 1 \\ 0 & , o. w. \end{cases}$$

What is the conditional mean of Y given $X = x$?

Good luck

سر (< < < - < < <)

Non-Central Distributions

1- Non-central T-dist:

Let $T = \frac{Z + \delta}{\sqrt{\frac{W}{r}}}$ $\Rightarrow$ if $\delta = 0 \Rightarrow$ we get Central T-dist.

$Z \sim N(0, 1)$ and $Z + \delta \sim N(\delta, 1)$ indep. of
 $W \sim \chi^2_{(r)}$

$$\therefore T = \frac{u}{\sqrt{w/r}}$$

$$\begin{aligned} g(z, w) &= g(z) \cdot g(w) \\ &= \frac{1}{\sqrt{2\pi}} e^{-z^2/2} \cdot \frac{1}{\Gamma(r/2) 2^{r/2}} w^{r/2-1} e^{-w/2} \end{aligned}$$

Let $t = \frac{z + \delta}{\sqrt{\frac{w}{r}}}$ and $u = w$

$$\therefore z = \left(\frac{u}{r}\right)^{1/2} t - \delta$$

$$|J| = \begin{vmatrix} \frac{dz}{dt} & \frac{dz}{du} \\ \frac{dw}{dt} & \frac{dw}{du} \end{vmatrix} = \begin{vmatrix} \sqrt{\frac{u}{r}} & \frac{t}{2\sqrt{ur}} \\ 0 & 1 \end{vmatrix} = \sqrt{\frac{u}{r}}$$

$\therefore$ The j.p.d.f. of t and u is:-

$$\begin{aligned} h(t, u) &= \frac{1}{\sqrt{2\pi} \Gamma(r/2) 2^{r/2}} u^{r/2-1} e^{-\frac{1}{2} \left[\left(\frac{u}{r}\right)^{1/2} t - \delta \right]^2 + u} \cdot \sqrt{\frac{u}{r}} \\ &= \frac{u^{r/2-1}}{\sqrt{2\pi r} \Gamma(r/2) 2^{r/2}} e^{-\frac{1}{2} \left[\frac{u}{r} t^2 - 2t\delta\sqrt{\frac{u}{r}} + \delta^2 + u \right]} \end{aligned}$$

but, $f(t, \delta) = \int_0^{\infty} h(t, u) du \dots (1)$

X

$$\text{Let } u\left(\frac{t^2}{r} + 1\right) = v^2 \Rightarrow u = \frac{v^2}{\left(\frac{t^2}{r} + 1\right)}$$

$$\Rightarrow \left| \frac{du}{dv} \right| = \frac{2vr}{t^2 + r}$$

$$\therefore h(v, t) = \frac{\left(\frac{v^2}{\frac{t^2}{r} + 1}\right)^{\frac{r-1}{2}}}{\sqrt{2\pi r} \cdot 2^{\frac{r-1}{2}} \Gamma\left(\frac{r}{2}\right)} e^{-\frac{1}{2}\left[v^2 - 2t\delta \frac{v}{\sqrt{t^2+r}} + \delta^2\right]} \cdot \frac{2vr}{t^2 + r}$$

So, by adding $-\frac{t^2\delta^2}{t^2+r}$ to exponential term:

$$\therefore h(v, t) = \frac{v^{r-1} \left[\frac{r}{t^2+r}\right]^{\frac{r-1}{2}} \cdot \frac{2vr}{t^2+r}}{\sqrt{2\pi r} \cdot 2^{\frac{r-1}{2}} \Gamma\left(\frac{r}{2}\right)} \exp\left[-\frac{1}{2}\left\{v^2 - 2t\delta \frac{v}{\sqrt{t^2+r}} + \frac{t^2\delta^2}{t^2+r} - \frac{t^2\delta^2}{t^2+r} + \delta^2\right\}\right]$$

$$h(v, t) = \frac{v^r \left[\frac{r}{t^2+r}\right]^{\frac{r+1}{2}}}{\sqrt{\pi r} \cdot 2^{\frac{r-1}{2}} \Gamma\left(\frac{r}{2}\right)} e^{-\frac{1}{2}\left[\frac{-t^2\delta^2}{t^2+r} + \delta^2\right] - \frac{1}{2}\left[v - \frac{t\delta}{\sqrt{t^2+r}}\right]^2}$$

$$\therefore f(t, \delta) = \int_0^\infty \frac{v^r \left[\frac{r}{t^2+r}\right]^{\frac{r+1}{2}}}{\sqrt{\pi r} \cdot 2^{\frac{r-1}{2}} \Gamma\left(\frac{r}{2}\right)} e^{-\frac{1}{2}\left[\frac{-t^2\delta^2}{t^2+r} + \delta^2\right] - \frac{1}{2}\left[v - \frac{t\delta}{\sqrt{t^2+r}}\right]^2} dv$$

$$\therefore f(t, \delta) = \frac{\left[\frac{r}{t^2+r}\right]^{\frac{r+1}{2}}}{2^{\frac{r-1}{2}} \sqrt{\pi r} \Gamma\left(\frac{r}{2}\right)} e^{-\frac{1}{2}\left(\frac{\delta^2 r}{t^2+r}\right)} \int_0^\infty v^r e^{-\frac{1}{2}\left[v - \frac{t\delta}{\sqrt{t^2+r}}\right]^2} dv$$

Home work: in equation (1-1), Let expand $e^{\frac{t\delta\sqrt{r}}{v}}$ in a series
Show that:

$$f(t, \delta) = \frac{e^{-\delta^2/2}}{\sqrt{\pi} \Gamma\left(\frac{r}{2}\right)} \sum_{i=0}^{\infty} \frac{(\delta\sqrt{r})^i}{i!} \cdot \frac{\Gamma\left(\frac{r+i+1}{2}\right)}{r^{\frac{i+1}{2}}} \cdot \frac{t^i}{(1+\frac{t^2}{r})\left(\frac{r+i+1}{2}\right)}$$

Solution: the Eq. (1-1)

$$\begin{aligned}
 f(t, s) &= \int_0^{\infty} h(u, t) du = \int_0^{\infty} \frac{u^{\frac{r-1}{2}}}{\sqrt{2\pi r} z^{r/2} \Gamma_{\frac{r}{2}}} e^{-\frac{1}{2} \left[\frac{u}{r} t^2 - 2t\delta \sqrt{\frac{u}{r}} + \delta^2 + u \right]} du \\
 &= \frac{e^{-\frac{\delta^2}{2}}}{\sqrt{2\pi r} z^{r/2} \Gamma_{\frac{r}{2}}} \int_0^{\infty} u^{\frac{r-1}{2}} e^{-\frac{1}{2} u \left(\frac{t^2}{r} + 1 \right) + t\delta \sqrt{\frac{u}{r}}} du \\
 &= K \int_0^{\infty} u^{\frac{r-1}{2}} e^{-\frac{1}{2} u \left(\frac{t^2}{r} + 1 \right)} e^{t\delta \sqrt{\frac{u}{r}}} du \\
 &= K \int_0^{\infty} u^{\frac{r-1}{2}} e^{-\frac{1}{2} u \left(\frac{t^2}{r} + 1 \right)} \left[1 + t\delta \sqrt{\frac{u}{r}} + t^2 \delta^2 \cdot \frac{u}{r} \cdot \frac{1}{2!} + \dots \right] du \\
 &= K \left[\int_0^{\infty} u^{\frac{r-1}{2}} e^{-\frac{1}{2} u \left(\frac{t^2}{r} + 1 \right)} du + \frac{t\delta}{\sqrt{r}} \int_0^{\infty} u^{\frac{r}{2}} e^{-\frac{1}{2} u \left(\frac{t^2}{r} + 1 \right)} du \right. \\
 &\quad \left. + \frac{t^2 \delta^2}{r 2!} \int_0^{\infty} u^{\frac{r}{2} + \frac{1}{2}} e^{-\frac{1}{2} u \left(\frac{t^2}{r} + 1 \right)} du + \dots \right]
 \end{aligned}$$

$$\text{Let } x = \frac{u}{2} \left(\frac{t^2}{r} + 1 \right) \Rightarrow u = \frac{2x}{\left(\frac{t^2}{r} + 1 \right)} \Rightarrow du = \frac{2dx}{\left(\frac{t^2}{r} + 1 \right)}$$

$$\begin{aligned}
 \therefore f(t, s) &= K \left[\int_0^{\infty} \left[\frac{2x}{\left(\frac{t^2}{r} + 1 \right)} \right]^{\frac{r-1}{2}} e^{-x} \cdot \frac{2}{\left(\frac{t^2}{r} + 1 \right)} dx + \dots \right] \\
 &= K \left[\Gamma_{\frac{r+1}{2}} \left\{ 2 \left(\frac{t^2}{r} + 1 \right) \right\}^{-\left(\frac{r+1}{2} \right)} + \frac{t\delta}{\sqrt{r}} \Gamma_{\frac{r+2}{2}} \left\{ 2 \left(\frac{t^2}{r} + 1 \right) \right\}^{-\left(\frac{r+2}{2} \right)} \right. \\
 &\quad \left. + \dots \right] \\
 &= K \left[\Gamma_{\frac{r+1}{2}} 2^{\frac{r+1}{2}} \left\{ \left(\frac{t^2}{r} + 1 \right) \right\}^{-\left(\frac{r+1}{2} \right)} + r^{-\frac{1}{2}} \delta t \Gamma_{\frac{r+2}{2}} 2^{\frac{r+2}{2}} \left(\frac{t^2}{r} + 1 \right)^{-\left(\frac{r+2}{2} \right)} \right. \\
 &\quad \left. + \dots \right]
 \end{aligned}$$

$$f(t, \delta) = \frac{e^{-\delta^2/2}}{\sqrt{\pi} \sqrt{\frac{r}{2}}} \left[\sqrt{\frac{r+1}{2}} \left(\frac{t^2}{r} + 1 \right)^{-\frac{(r+1)}{2}} r^{-\frac{1}{2}} + t \delta r^{-\frac{3}{2}} \sqrt{\frac{r+2}{2}} z^{\frac{1}{2}} \right. \\ \left. + \left(\frac{t^2}{r} + 1 \right)^{-\frac{(r+3)}{2}} r^{-\frac{3}{2}} t \delta^2 \sqrt{\frac{r+3}{2}} z^{\frac{1}{2}} \left(\frac{t^2}{r} + 1 \right)^{-\frac{(r+3)}{2}} + \dots \right]$$

$$= \frac{e^{-\delta^2/2}}{\sqrt{\pi} \sqrt{\frac{r}{2}}} \left[\frac{\sqrt{\frac{r+1}{2}}}{r^{\frac{1}{2}} \left(\frac{t^2}{r} + 1 \right)^{\frac{r+1}{2}}} + \frac{(\delta t \sqrt{2}) \sqrt{\frac{r+2}{2}}}{r^{\frac{3}{2}} \left(\frac{t^2}{r} + 1 \right)^{\frac{r+2}{2}}} \right. \\ \left. + \frac{(t \delta \sqrt{2})^2}{r^{\frac{5}{2}} \left(\frac{t^2}{r} + 1 \right)^{\frac{r+3}{2}}} + \dots \right]$$

$$\therefore f(t, \delta) = \frac{e^{-\delta^2/2}}{\sqrt{\pi} \sqrt{\frac{r}{2}}} \sum_{i=0}^{\infty} \frac{(t \delta \sqrt{2})^i \sqrt{\frac{r+i+1}{2}}}{r^{\frac{i+1}{2}} \left(1 + \frac{t^2}{r} \right)^{\frac{r+i+1}{2}}}$$

Home work!

Show that

$$E[\omega^{-K/2}] = \frac{2^{-\frac{K}{2}} \sqrt{\frac{r-K}{2}}}{\sqrt{\frac{r}{2}}}, \quad r > K.$$

Solution

$$T = \frac{z + \delta}{\sqrt{\frac{w}{r}}} = (z + \delta) \left(\frac{w}{r} \right)^{-1/2}$$

$$T = (z + \delta) r^{\frac{1}{2}} w^{-1/2} \\ = z r^{1/2} w^{-1/2} + \delta r^{1/2} w^{-1/2}$$

$$E[T] = E[z r^{1/2} w^{-1/2}] + E[\delta r^{1/2} w^{-1/2}]$$

$$z \sim N(0, 1)$$

$$w \sim \chi^2_{(r)}$$

$$\Rightarrow E(z) = 0$$

$$\therefore E[T] = 0 + \delta r^{1/2} E(w^{-1/2}) = \delta r^{1/2} E(w^{-1/2}) \dots (1)$$

$$T^3 = (Z + S)^3 = (Z + S)(Z^2 + S^2 + 2SZ)$$

$$ET^3 = (0 + S)(1 + S^2 + 0) = S(1 + S^2)$$

$$\begin{aligned} E(W^{-K/2}) &= \int_0^\infty W^{-K/2} f(W) dW \\ &= \int_0^\infty W^{-K/2} \frac{W^{r/2-1} e^{-W/2}}{\sqrt{\frac{r}{2}} 2^{r/2}} dW \\ &= \int_0^\infty \frac{W^{\frac{r-K}{2}-1} e^{-W/2}}{\sqrt{\frac{r}{2}} 2^{r/2}} dW \\ &= \frac{\sqrt{\frac{r-K}{2}} 2^{\frac{r-K}{2}}}{\sqrt{\frac{r}{2}} 2^{r/2}} \left[\int_0^\infty \frac{W^{\frac{r-K}{2}-1} e^{-W/2}}{\sqrt{\frac{r-K}{2}} 2^{\frac{r-K}{2}}} dW \right] \\ &= \frac{\sqrt{\frac{r-K}{2}} 2^{\frac{r-K}{2}}}{\sqrt{\frac{r}{2}} 2^{r/2}} \cdot 1 = \frac{2^{-K/2} \sqrt{\frac{r-K}{2}}}{\sqrt{\frac{r}{2}}} \end{aligned}$$

$$\begin{aligned} \therefore E(T) &= S r^{1/2} E(W^{-1/2}) \\ &= S r^{1/2} \frac{2^{-1/2} \sqrt{\frac{r-1}{2}}}{\sqrt{\frac{r}{2}}} = \frac{\sqrt{r/2} \sqrt{\frac{r-1}{2}}}{\sqrt{\frac{r}{2}}} S \end{aligned}$$

To Find ET^2 ?

$$T = \frac{Z + S}{\sqrt{W/r}} \Rightarrow T^2 = \frac{(Z + S)^2}{W/r}$$

$$T^2 = (Z + S)^2 \left(\frac{W}{r}\right)^{-1}$$

$$T^2 = (Z^2 + 2ZS + S^2) \left(\frac{W}{r}\right)^{-1}$$

$$ET^2 = [EZ^2 + 2SEZ + ES^2] \cdot r EW^{-1}$$

we have $EZ^2 = 1$, $EZ = 0$, $ES^2 = S^2$ and

$$\begin{aligned} E(W^{-1}) &= \frac{2^{-2/2} \sqrt{\frac{r-2}{2}}}{\sqrt{\frac{r}{2}}} \\ &= E(W^{-2/2}) \frac{1}{\sqrt{\frac{r}{2}}} \end{aligned}$$

$$(Z + S)^3 \frac{W^{-1}}{r} = \frac{W^{-1}}{r} \frac{1}{r^{3/2}}$$

$$T = \frac{(Z + S)^3}{\sqrt{\frac{W}{r}}^3}$$

- 62 -

$$\sqrt{\frac{W}{r}} \sqrt{\frac{W}{r}} \sqrt{\frac{W}{r}} = \frac{W}{r} \sqrt{\frac{W}{r}}$$

$$\begin{aligned}
 \therefore ET^2 &= (1 + 0 + \delta^2) r \left[\frac{\frac{r-2}{2}}{2 \sqrt{\frac{r}{2}}} \right] \\
 &= (1 + \delta^2) r \frac{\left[\frac{r-2}{2} - 1 \right]!}{2 \left[\frac{r}{2} - 1 \right]!} \\
 &= (1 + \delta^2) r \frac{\left[\frac{r-2}{2} \right]! \left[\frac{r-2}{2} \right]}{2 \left[\frac{r}{2} - 1 \right]! \left[\frac{r}{2} - 2 \right]!} = (1 + \delta^2) r \cdot \frac{1}{2 \left(\frac{r-2}{2} \right)}
 \end{aligned}$$

$$\therefore ET^2 = \frac{(1 + \delta^2) r}{r-2}$$

2. Non-Central χ^2 dist:

$$\text{Let } g(u) = \sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!} \cdot \frac{u^{m+\frac{n}{2}-1} e^{-u/2}}{2^{\frac{m+n}{2}} \Gamma(m+\frac{n}{2})}, \quad u > 0$$

- (i) Show that $g(u)$ is a density function
 (ii) Find the m.g.f. of U .

Sol

$$(i) \int_0^{\infty} g(u) du \stackrel{?}{=} 1$$

$$\therefore \int_0^{\infty} \sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!} \cdot \frac{u^{m+\frac{n}{2}-1} e^{-u/2}}{2^{\frac{m+n}{2}} \Gamma(m+\frac{n}{2})} du$$

$$\underbrace{\sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!}}_{\text{Poisson dist.}} \int_0^{\infty} \frac{u^{m+\frac{n}{2}-1} e^{-u/2}}{2^{\frac{m+n}{2}} \Gamma(m+\frac{n}{2})} du$$

$\sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!} = 1$

$\int_0^{\infty} \frac{u^{m+\frac{n}{2}-1} e^{-u/2}}{2^{\frac{m+n}{2}} \Gamma(m+\frac{n}{2})} du = 1$

$\therefore g(u)$ is a density function

(iii) $M_x(t) = \bar{C} e^{xt}$
 $\therefore M_u(t) = \bar{C} e^{ut} = \int_0^\infty e^{ut} g(u) du$

$$= \sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!} \int_0^\infty \frac{u^{m+\frac{n}{2}-1}}{\frac{2^{m+\frac{n}{2}}}{\Gamma(m+\frac{n}{2})}} e^{-u(\frac{1}{2}-t)} du$$

Let $y = u(\frac{1}{2}-t) \Rightarrow u = \frac{y}{\frac{1}{2}-t} \Rightarrow du = \frac{dy}{\frac{1}{2}-t}$

$$M_u(t) = \sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!} \int_0^\infty \left(\frac{y}{\frac{1}{2}-t}\right)^{m+\frac{n}{2}-1} e^{-y} \cdot \frac{dy}{\frac{1}{2}-t}$$

$$= \sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!} \cdot \frac{1}{\frac{2^{m+\frac{n}{2}}}{\Gamma(m+\frac{n}{2})}} \cdot \frac{1}{(\frac{1}{2}-t)^{m+\frac{n}{2}}} \int_0^\infty y^{m+\frac{n}{2}-1} e^{-y} dy$$

$$= \sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!} \cdot \frac{1}{\frac{2^{m+\frac{n}{2}}}{\Gamma(m+\frac{n}{2})}} \cdot \frac{1}{(\frac{1}{2}-t)^{m+\frac{n}{2}}} \cdot \Gamma(m+\frac{n}{2})$$

$$= \sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!} \cdot \frac{1}{2^{m+\frac{n}{2}}} \cdot \frac{\Gamma(m+\frac{n}{2})}{(\frac{1}{2}-t)^{m+\frac{n}{2}}}$$

$$= \sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2}\right)^m}{m!} \cdot \frac{1}{(1-2t)^{m+\frac{n}{2}}}$$

$$= \frac{e^{-\theta/2}}{(1-2t)^{n/2}} \sum_{m=0}^{\infty} \frac{\left(\frac{\theta}{2}\right)^m}{m!} \left(\frac{1}{1-2t}\right)^m$$

$$= \frac{e^{-\theta/2}}{(1-2t)^{n/2}} \sum_{m=0}^{\infty} \frac{\left(\frac{\theta}{2-4t}\right)^m}{m!} \cdot \frac{e^{-\frac{\theta}{2-4t}}}{e^{-\frac{\theta}{2-4t}}}$$

$$= \frac{e^{-\theta/2}}{(1-2t)^{n/2}} \cdot \frac{1}{e^{-\theta/(2-4t)}} \cdot \sum_{m=0}^{\infty} \frac{e^{-\theta/(2-4t)} \left[\frac{\theta}{2-4t}\right]^m}{m!}$$

$$= \frac{e^{-\theta/2 + \theta/(2-4t)}}{(1-2t)^{n/2}} = \frac{e^{\frac{\theta t}{1-2t}}}{(1-2t)^{n/2}}$$

Handwritten notes and calculations on the right side of the page, including:

$$\frac{(\frac{1}{2}-t)^{m+\frac{n}{2}}}{\frac{1}{2^{m+\frac{n}{2}}}} \sim \frac{1}{2^{m+\frac{n}{2}}}$$

$$\sim \frac{1}{(1-2t)^{m+\frac{n}{2}}}$$

3- Non-Central F-dist:

If u and v are indep. r.v's

$$u \sim \chi^2_{(r_1)}(\theta)$$

$$v \sim \chi^2_{(r_2)}(\theta) \Rightarrow F = \frac{u/r_1}{v/r_2} \sim F'_{(r_1, r_2)}(\theta)$$

we seek prob. density function of j.p.d.f. of u and v .

$$h_1(u, v) = \frac{\frac{r_2}{2} - 1}{v^{\frac{r_2}{2}}} \frac{e^{-v/2}}{\sqrt{\frac{r_2}{2}}} \sum_{m=0}^{\infty} \frac{e^{-\theta/2} (\frac{\theta}{2})^m}{m!} \cdot \frac{u^{\frac{m+r_1}{2}-1} e^{-u/2}}{\sqrt{\frac{m+r_1}{2}}}$$

$$f = \frac{u/r_1}{v/r_2} = \frac{r_2 u}{r_1 v} \Rightarrow \left\{ u = \frac{f r_1 v}{r_2} \right\} \text{ and } \left\{ w = v \right\}$$

$$\Rightarrow u = \frac{f r_1 w}{r_2} \Rightarrow |J| = \frac{r_1 w}{r_2}, \text{ then:}$$

$$\begin{aligned} h_2(w, f) &= \frac{\frac{r_2}{2} - 1}{w^{\frac{r_2}{2}}} \frac{e^{-w/2}}{\sqrt{\frac{r_2}{2}}} \sum_{m=0}^{\infty} \frac{e^{-\theta/2} (\frac{\theta}{2})^m}{m!} \left[\frac{f r_1 w}{r_2} \right]^{\frac{m+r_1}{2}-1} \frac{e^{-\frac{1}{2} \left[\frac{f r_1 w}{r_2} \right]}}{\sqrt{\frac{m+r_1}{2}}} \\ &= \frac{\frac{r_2}{2} - 1}{w^{\frac{r_2}{2}}} \frac{e^{-w/2}}{\sqrt{\frac{r_2}{2}}} \sum_{m=0}^{\infty} \frac{e^{-\theta/2} (\frac{\theta}{2})^m}{m!} \frac{(\frac{r_1}{r_2})^{m+\frac{r_1}{2}-1} e^{-\frac{r_1 w f}{2}}}{\sqrt{\frac{m+\frac{r_1}{2}}{2}}} \\ &\quad * f^{m+\frac{r_1}{2}-1} w^{m+\frac{r_1}{2}-1} \end{aligned}$$

$$\begin{aligned} h_2(w, f) &= \sum_{m=0}^{\infty} \frac{e^{-\theta/2} (\frac{\theta}{2})^m}{m!} \cdot \frac{1}{\frac{m+\frac{r_1}{2}}{2} \sqrt{\frac{m+\frac{r_1}{2}}{2}}} \cdot \frac{(\frac{r_1}{r_2})^{m+\frac{r_1}{2}-1}}{\frac{m+\frac{r_1}{2}}{2} \sqrt{\frac{m+\frac{r_1}{2}}{2}}} \\ &\quad * f^{m+\frac{r_1}{2}-1} w^{m+\frac{r_1}{2}-1+\frac{r_2}{2}} e^{-\frac{w}{2} - \frac{r_1 f w}{2 r_2}} \end{aligned}$$

$$\text{Let } \left(\frac{\omega}{2} + \frac{r_1 f \omega}{2 r_2} \right) = t$$

$$\Rightarrow \frac{\omega}{2} \left[1 + \frac{r_1}{r_2} f \right] = t$$

$$\Rightarrow \omega = \frac{2t}{\left[1 + \frac{r_1}{r_2} f \right]}, \quad d\omega = \frac{2dt}{\left[1 + \frac{r_1}{r_2} f \right]}$$

$$h_3(t, f) = \sum_{m=0}^{\infty} \frac{e^{-\theta/2} \left(\frac{\theta}{2} \right)^m}{m!} \cdot \frac{\left(\frac{r_1}{r_2} \right)^{m+\frac{r_1}{2}-1}}{\sqrt{\frac{r_1}{2}} \sqrt{m+\frac{r_1}{2}}} f^{m+\frac{r_1}{2}-1} \\ \times \frac{1}{2^{\frac{r_1+r_2}{2}+m}} \cdot e^{-t} \cdot t^{\frac{r_1+r_2}{2}+m-1} \cdot \frac{1}{\left[1 + \frac{r_1}{r_2} f \right]^{\frac{r_1+r_2}{2}+m}}$$

Finally; to find $g(f)$

$$g(f) = \int_0^{\infty} h(t, f) dt \\ = \sum_{m=0}^{\infty} \frac{e^{-\frac{\theta}{2}} \left(\frac{\theta}{2} \right)^m}{m!} \cdot \frac{\left(\frac{r_1}{r_2} \right)^{m+\frac{r_1}{2}}}{\beta(m+\frac{r_1}{2}, \frac{r_2}{2})} \cdot \frac{f^{m+\frac{r_1}{2}-1}}{\left(1 + \frac{r_1}{r_2} f \right)^{\frac{r_1+r_2}{2}+m}} \\ , f > 0$$

where $\beta(a, b)$ is a Beta function with $a, b > 0$

Some Special Discrete Bivariate Distributions

1. Bivariate Bernoulli Distribution:

Definition (42)

A discrete bivariate random variable (X, Y) is said to have the bivariate Bernoulli distribution if its joint probability density is of the form:

$$P(x, y) = \begin{cases} \frac{1}{x! y! (1-x-y)!} p_1^x p_2^y (1-p_1-p_2)^{1-x-y} & , x, y = 0, 1 \\ 0 & , \text{o.w} \end{cases}$$

where $0 < p_1, p_2, p_1 + p_2 < 1$ and $x + y \leq 1$. we denote a bivariate Bernoulli random variable by writing:

$$(X, Y) \sim \text{Ber}(p_1, p_2)$$

Recall that; the joint moment generating function of X and Y is defined as

$$M(s, t) = E(e^{sX + tY})$$

Theorem (26)

Let $(X, Y) \sim \text{Ber}(p_1, p_2)$, where p_1 and p_2 are parameters.

Then:

$$E(X) = p_1$$

$$E(Y) = p_2$$

$$\text{Var}(X) = p_1(1-p_1)$$

$$\text{Var}(Y) = p_2(1-p_2)$$

$$\text{Cov}(X, Y) = -p_1 p_2$$

$$M(s, t) = 1 - p_1 - p_2 + p_1 e^s + p_2 e^t.$$

proof

The joint m.g.f. of X and Y is given by:

$$\begin{aligned}
 M(s, t) &= E(e^{sx+ty}) \\
 &= \sum_{x=0}^1 \sum_{y=0}^1 p(x, y) e^{sx+ty} \\
 &= p(0,0) + p(1,0)e^s + p(0,1)e^t + p(1,1)e^{s+t} \\
 &= 1 - p_1 - p_2 + p_1 e^s + p_2 e^t + 0 e^{s+t} \\
 &= 1 - p_1 - p_2 + p_1 e^s + p_2 e^t.
 \end{aligned}$$

The expected value of X is given by:

$$\begin{aligned}
 E(X) &= \left. \frac{\partial M(s, t)}{\partial s} \right|_{(0,0)} \\
 &= \left. \frac{\partial}{\partial s} (1 - p_1 - p_2 + p_1 e^s + p_2 e^t) \right|_{(0,0)} \\
 &= p_1 e^s \big|_{(0,0)}
 \end{aligned}$$

$$\therefore E(X) = p_1$$

Similarly, the expected value of Y is given by:

$$\begin{aligned}
 E(Y) &= \left. \frac{\partial M(s, t)}{\partial t} \right|_{(0,0)} \\
 &= \left. \frac{\partial}{\partial t} (1 - p_1 - p_2 + p_1 e^s + p_2 e^t) \right|_{(0,0)} \\
 &= p_2 e^t \big|_{(0,0)}
 \end{aligned}$$

$$\therefore E(Y) = p_2$$

Therefore the covariance of X and Y is

$$\text{cov}(X, Y) = E(XY) - E(X)E(Y)$$

The product moment of X and Y is:

$$\begin{aligned}
 E(XY) &= \frac{\partial^2 M(s,t)}{\partial t \partial s} \Big|_{(0,0)} \\
 &= \frac{\partial^2}{\partial t \partial s} (1 - p_1 - p_2 + p_1 e^s + p_2 e^t) \Big|_{(0,0)} \\
 &= \frac{\partial}{\partial t} (p_1 e^s) \Big|_{(0,0)} \\
 &= 0
 \end{aligned}$$

Then: $\text{Cov}(X,Y) = 0 - p_1 p_2 = \boxed{-p_1 p_2}$

Similarly, it can be shown that:

$\boxed{E(X^2) = p_1}$ and $\boxed{E(Y^2) = p_2}$

Thus, we have:

$$\begin{aligned}
 \text{Var}(X) &= E(X^2) - [E(X)]^2 = p_1 - p_1^2 = p_1(1-p_1) \\
 \text{Var}(Y) &= E(Y^2) - [E(Y)]^2 = p_2 - p_2^2 = p_2(1-p_2)
 \end{aligned}$$

Theorem (27) Let $(X,Y) \sim B(p_1, p_2)$, where p_1 and p_2 are parameters. Then the Conditional distribution $P(Y/X)$ and $p(X/Y)$ are also Bernoulli and

$$E(Y/X) = \frac{p_2(1-x)}{1-p_1} = E(Y^c/X)$$

$$E(X/Y) = \frac{p_1(1-y)}{1-p_2} = E(X^c/Y)$$

$$\text{Var}(Y/X) = \frac{p_2(1-p_1-p_2)(1-x)}{(1-p_1)^2}$$

$$\text{Var}(X/Y) = \frac{p_1(1-p_1-p_2)(1-y)}{(1-p_2)^2}$$

Proof:

$$\begin{aligned}
 P(Y/X) &= \frac{P(X,Y)}{P(X)} \\
 &= \frac{P(X,Y)}{\sum_{y=0}^1 P(X,Y)}
 \end{aligned}$$

$$P(Y/X=x) = \frac{P(x,y)}{P(x)} = \frac{P(x,y)}{\sum_{y=0}^1 P(x,y)} = \frac{P(x,y)}{P(x,0) + P(x,1)}$$

Hence; $P(1/0) = \frac{P(0,1)}{P(0,0) + P(0,1)} = \frac{P_2}{1 - P_1 - P_2 + P_2}$

$$= \frac{P_2}{1 - P_1}$$

and; $P(1/1) = \frac{P(1,1)}{P(1,0) + P(1,1)} = \frac{0}{P_1 + 0} = 0$

$$\Rightarrow E(Y/X=0) = \sum_{y=0}^1 y P(y/0) = P(1/0) = \frac{P_2}{1 - P_1}$$

$$\text{and } E(Y/X=1) = \sum_{y=0}^1 y P(y/1) = P(1/1) = 0$$

Merging these together, we have

$$E(Y/X) = \frac{P_2}{1 - P_1} \neq 0$$

Similarly, we compute:

$$E(Y^2/X=0) = \sum_{y=0}^1 y^2 P(y/0) = P(1/0)$$

$$= \frac{P_2}{1 - P_1}$$

and: $E(Y^2/X=1) = P(1/1) = 0$

Therefore;

$$\begin{aligned} \text{var}(Y/X=0) &= E(Y^2/X=0) - [E(Y/X=0)]^2 \\ &= \frac{P_2}{1 - P_1} - \left(\frac{P_2}{1 - P_1}\right)^2 \\ &= \frac{P_2(1 - P_1 - P_2)}{(1 - P_1)^2} \end{aligned}$$

and $\text{var}(Y/X=1) = 0$

Merging these together, we have

$$\text{var}(Y/X) = \frac{P_2(1 - P_1 - P_2)(1 - x)}{(1 - P_1)^2}, x=0,1$$

The $E(X/Y)$ and $\text{var}(X/Y)$ can be obtained in a similar manner.

2. Bivariate Binomial Distribution:

Definition (43)

A discrete bivariate random variable (X, Y) is said to have the bivariate Binomial distribution with parameters n, p_1, p_2 if its joint probability density is of the form:

$$p(x, y) = \begin{cases} \frac{n!}{x! y! (n-x-y)!} p_1^x p_2^y (1-p_1-p_2)^{n-x-y}, & x, y = 0, 1, \dots, n \\ 0 & \text{o.w.} \end{cases}$$

where $0 < p_1, p_2, p_1 + p_2 < 1, x + y \leq n$ and n is a positive integer. We denote a bivariate binomial random variable by writing:

$$(X, Y) \sim \text{Bin}(n, p_1, p_2).$$

Example (53) A certain game involves rolling a fair die and watching the numbers of rolls of 4 and 5. What is the probability that in 10 rolls of the die one 4 and three 5 will be observed?

Solution:

Let X denote the number of 4 and Y denote the number of 5. Then the joint distribution of X and Y is bivariate binomial with $n=10, p_1 = \frac{1}{6}, p_2 = \frac{1}{6}$ and $1-p_1-p_2 = \frac{4}{6}$. Hence the probability that in 10 rolls of the die one 4 and three 5 will be observed is:

$$\begin{aligned} P(X=1, Y=3) &= P(1, 3) \\ &= \frac{n!}{x! y! (n-x-y)!} p_1^x p_2^y (1-p_1-p_2)^{n-x-y} \\ &= \frac{10!}{1! 3! (10-1-3)!} \left(\frac{1}{6}\right)^1 \left(\frac{1}{6}\right)^3 \left(\frac{4}{6}\right)^{10-1-3} \\ &= 0.0569 \end{aligned}$$

Hint: The following result generalizes the bivariate theorem and can be called trinomial theorem. Similar to the proof of binomial theorem, one can establish:

$$(a+b+c)^n = \sum_{x=0}^n \sum_{y=0}^n C_{x,y}^n a^x b^y c^{n-x-y}$$

where $0 \leq x+y \leq n$ and

$$C_{x,y}^n = \frac{n!}{x! y! (n-x-y)!}$$

Theorem (28)

Let $(X, Y) \sim \text{Bin}(n, p_1, p_2)$, where n, p_1 and p_2 are parameters. Then

$$E(X) = np_1$$

$$E(Y) = np_2$$

$$\text{Var}(X) = np_1(1-p_1)$$

$$\text{Var}(Y) = np_2(1-p_2)$$

$$\text{Cov}(X, Y) = -np_1p_2$$

$$M(s, t) = (1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^n$$

Proof: First, we find the joint moment generating function of X and Y . The $M(s, t)$ is given by:

$$M(s, t) = E(e^{sx+ty})$$

$$= \sum_{x=0}^n \sum_{y=0}^n e^{sx+ty} p(x, y)$$

$$= \sum_{x=0}^n \sum_{y=0}^n e^{sx+ty} \frac{n!}{x! y! (n-x-y)!} p_1^x p_2^y (1-p_1-p_2)^{n-x-y}$$

$$= \sum_{x=0}^n \sum_{y=0}^n \frac{n!}{x! y! (n-x-y)!} (e^s p_1)^x (e^t p_2)^y (1-p_1-p_2)^{n-x-y}$$

$$= (1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^n, \text{ by theorem.}$$

$$\underline{E(X)} = \frac{\partial M(s, t)}{\partial s} \Big|_{(0,0)}$$

$$= \frac{\partial}{\partial s} (1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^n \Big|_{(0,0)}$$

$$= n(1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^{n-1} p_1 e^s \Big|_{(0,0)}$$

$$= n p_1$$

$$\underline{E(Y)} = \frac{\partial M(s, t)}{\partial t} \Big|_{(0,0)}$$

$$= \frac{\partial}{\partial t} (1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^n \Big|_{(0,0)}$$

$$= n(1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^{n-1} p_2 e^t \Big|_{(0,0)}$$

$$= n p_2$$

The product moment of X and Y is:

$$\underline{E(XY)} = \frac{\partial^2 M(s, t)}{\partial t \partial s} \Big|_{(0,0)}$$

$$= \frac{\partial^2}{\partial t \partial s} (1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^n \Big|_{(0,0)}$$

$$= \frac{\partial}{\partial t} [n(1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^{n-1} p_1 e^s \Big|_{(0,0)}]$$

$$= n(n-1) p_1 p_2$$

$$\frac{\partial^2 M(s, t)}{\partial t \partial s} = n p_2 \left[(1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^{n-1} + e^t (n-1) (1 - p_1 - p_2 + p_1 e^s + p_2 e^t)^{n-2} p_2 e^t \right] \Big|_{(0,0)}$$

Therefore the covariance of X and Y is

$$\underline{\text{Cov}(X, Y)} = E(XY) - E(X)E(Y)$$

$$= n(n-1) p_1 p_2 - (n p_1)(n p_2)$$

$$= n(n-1) p_1 p_2 - n^2 p_1 p_2$$

$$= -n p_1 p_2$$

$$= n p_2 + (n-1) p_2 \cdot n p_2 \frac{e^t}{e^t} \Big|_{(0,0)}$$

$$\text{var}(X) = E(X^2) - [E(X)]^2$$

$$E(X^2) = n(n-1)P_1^2 + nP_1$$

$$\begin{aligned} \Rightarrow \text{var}(X) &= n(n-1)P_1^2 + nP_1 - n^2P_1^2 \\ &= n^2P_1^2 - nP_1^2 + nP_1 - n^2P_1^2 \\ &= nP_1(1-P_1) \end{aligned}$$

Similarly;

$$E(Y^2) = n(n-1)P_2^2 + nP_2$$

$$\begin{aligned} \text{var}(Y) &= n(n-1)P_2^2 + nP_2 - n^2P_2^2 \\ &= nP_2(1-P_2) \end{aligned}$$

Remark: For any real numbers a and b , we have

$$\sum_{y=0}^m y \binom{m}{y} a^y b^{m-y} = ma(a+b)^{m-1}$$

and;

$$\sum_{y=0}^m y^2 \binom{m}{y} a^y b^{m-y} = ma(ma+b)(a+b)^{m-2}$$

where m is a positive integer.

Example (54)

If X equals the number of ones and Y equals the number of twos and threes when a pair of fair dice are rolled, then what the correlation coefficient of X and Y .

Solution:

The joint density of X and Y is bivariate Binomial and is given by:

$$P(x, y) = \frac{2!}{x! y! (2-x-y)!} \left(\frac{1}{6}\right)^x \left(\frac{2}{6}\right)^y \left(\frac{3}{6}\right)^{2-x-y}, 0 \leq x+y \leq 2$$

By theorem

$$\text{var}(X) = nP_1(1-P_1) = 2\left(\frac{1}{6}\right)\left(1-\frac{1}{6}\right) = \frac{10}{36}$$

$$\text{var}(Y) = nP_2(1-P_2) = 2\left(\frac{2}{6}\right)\left(1-\frac{2}{6}\right) = \frac{16}{36}$$

$$\begin{aligned}\text{cov}(X, Y) &= -nP_1P_2 \\ &= -2\left(\frac{1}{6}\right)\left(\frac{2}{6}\right) = -\frac{4}{36}\end{aligned}$$

Therefore;

$$\begin{aligned}\text{corr}(X, Y) &= \frac{\text{cov}(X, Y)}{\sqrt{\text{var}(X)\text{var}(Y)}} \\ &= \frac{-4/36}{\sqrt{\frac{10}{36} \cdot \frac{16}{36}}} = \frac{-4}{4\sqrt{10}} \\ &= -0.3162\end{aligned}$$

Theorem (29)

Let $(X, Y) \sim \text{Bin}(n, P_1, P_2)$, where n, P_1 and P_2 are parameters. Then the conditional distributions $P(Y/X)$ and $P(X/Y)$ are also binomial and:

$$E(Y/X) = \frac{P_2(n-X)}{1-P_1}$$

$$E(X/Y) = \frac{P_1(n-Y)}{1-P_2}$$

$$\text{var}(Y/X) = \frac{P_2(1-P_1-P_2)(n-X)}{(1-P_1)^2}$$

$$\text{var}(X/Y) = \frac{P_1(1-P_1-P_2)(n-Y)}{(1-P_2)^2}$$

Proof:

2. The marginal density $P_1(x)$ of X is given by:

$$P_1(x) = \sum_{y=0}^{n-x} \frac{n!}{x! y! (n-x-y)!} P_1^x P_2^y (1-P_1-P_2)^{n-x-y} \quad \left\} x \cdot \frac{(n-x)!}{(n-x)!}\right.$$

$$= \frac{n! P_1^x}{x! (n-x)!} \sum_{y=0}^{n-x} \frac{(n-x)!}{y! (n-x-y)!} P_2^y (1-P_1-P_2)^{n-x-y}$$

$$= C_x^n P_1^x (1-P_1-P_2 + P_2)^{n-x}$$

$$= C_x^n P_1^x (1-P_1)^{n-x}$$

, $x = 0, 1, \dots, n$
by binomial Theorem.

The conditional density of Y given $X=x$ is

$$\begin{aligned}
 P(Y/x) &= \frac{P(x,y)}{P_1(x)} = \frac{P(x,y)}{C_x^n P_1^x (1-P_1)^{n-x}} \\
 &= \frac{(n-x)!}{(n-x-y)! y!} P_2^y (1-P_1-P_2)^{n-x-y} (1-P_1)^{x-n} \\
 &= (1-P_1)^{x-n} C_y^{n-x} P_2^y (1-P_1-P_2)^{n-x-y}
 \end{aligned}$$

Hence;

$$\begin{aligned}
 E(Y/x) &= \sum_{y=0}^{n-x} y P(Y/x) \\
 &= \sum_{y=0}^{n-x} y (1-P_1)^{x-n} C_y^{n-x} P_2^y (1-P_1-P_2)^{n-x-y} \\
 &= (1-P_1)^{x-n} \sum_{y=0}^{n-x} y C_y^{n-x} P_2^y (1-P_1-P_2)^{n-x-y} \\
 &= (1-P_1)^{x-n} P_2 (n-x) (1-P_1-P_2)^{n-x-1} \\
 &= \frac{P_2 (n-x)}{1-P_1}
 \end{aligned}$$

Next, we find the Conditional variance of Y given $X=x$.

$$\begin{aligned}
 E(Y^2/x) &= \sum_{y=0}^{n-x} y^2 P(Y/x) \\
 &= \sum_{y=0}^{n-x} y^2 (1-P_1)^{x-n} C_y^{n-x} P_2^y (1-P_1-P_2)^{n-x-y} \\
 &= (1-P_1)^{x-n} \sum_{y=0}^{n-x} y^2 C_y^{n-x} P_2^y (1-P_1-P_2)^{n-x-y} \\
 &= (1-P_1)^{x-n} P_2 (n-x) (1-P_1-P_2)^{n-x-1} [(n-x)P_2 + 1-P_1-P_2] \\
 &= \frac{P_2 (n-x) [(n-x)P_2 + 1-P_1-P_2]}{(1-P_1)^2}
 \end{aligned}$$

Hence, the Conditional variance of Y given $X=x$ is:

$$\begin{aligned}\text{Var}(Y/x) &= E(Y^2/x) - [E(Y/x)]^2 \\ &= \frac{P_2(n-x)[(n-x)P_2 + 1 - P_1 - P_2]}{(1-P_1)^2} - \left(\frac{P_2(n-x)}{1-P_1}\right)^2 \\ &= \frac{P_2(1-P_1-P_2)(n-x)}{(1-P_1)^2}\end{aligned}$$

similarly, one can establish:

$$E(X/y) = \frac{P_1(n-y)}{1-P_2} \quad \text{and}$$

$$\text{Var}(X/y) = \frac{P_1(1-P_1-P_2)(n-y)}{(1-P_2)^2}.$$

Note: $P(Y/x)$ in the previous theorem is a univariate binomial probability density function. To see this observe that:

$$(1-P_1)^{x-n} \binom{n-x}{y} P_2^y (1-P_1-P_2)^{n-x-y} = \binom{n-x}{y} \left(\frac{P_2}{1-P_1}\right)^y \left(1 - \frac{P_2}{1-P_1}\right)^{n-x-y}.$$

Hence,

$$Y/x \sim \text{Bin}(n-x, \frac{P_2}{1-P_1}).$$

Example (55)

Let W equal the weight of soap in a 1-Kilogram box. Suppose $P(W < 1) = 0.02$ and $P(W > 1.072) = 0.08$. Call a box of soap light, good, or heavy depending on whether on $W < 1$, $1 \leq W \leq 1.072$ or $W > 1.072$, respectively. In a random Sample of 50 boxes, let X equal the number of light boxes and Y the number of good boxes.

Compute:

1. the regression curve $[E(Y/x)]$ of Y on X .
2. the Scedastic curve $[\text{Var}(Y/x)]$ of Y on X .

$\binom{n-y}{0}$