If 
$$P(A|c) \% P(B|A) \Rightarrow \frac{P(A\cap c)}{P(c)} \% \frac{P(B\cap c)}{P(c)}$$

$$\Rightarrow P(A\cap B) \% P(B\cap c) \qquad (1)$$

$$\Rightarrow P(A) \rangle P(B)$$

a) Set 
$$\int_{0}^{\infty} Kx^{2} e^{-x^{2}/10} dx = 1$$
 and solve  $t$ .
$$\int_{0}^{\infty} Kx^{2} e^{-x^{2}/10} dx = -\frac{10}{3}K \int_{0}^{\infty} \frac{-3x^{2}}{10} e^{-x^{2}/10} dx$$

$$= -\frac{10}{3}K \left(e^{-x^{2}/10}\right)\Big|_{0}^{\infty} = 0 + \frac{10}{3}K = 1$$

b) 
$$F(x) = P(x \le x) = \int_{0}^{x} 0.3 y^{2} e^{-y^{3}/10} dy$$

$$= -\frac{y^{3}}{10} | x - \frac{y^{3}}{10} | 0 = 1 - e^{-y^{3}/10} for x/0$$

$$F(x) = \begin{cases} 0 & x(0) \\ -\frac{3}{1-e} & x(0) \end{cases}$$

(c) 
$$1 - e^{-\frac{x^3}{40}}$$
 gives  $--\Rightarrow x = (-10 \ln (0.5))^3 = 1.907 \text{ years}$ 

a) 
$$P(Monday)$$
 and  $good Service) = P(good Service | Monday)  $P(Monday)$$ 

$$= 0.72 * 0.26 = 0.1872$$

b) 
$$P(good Service) = P(good Service | Monday) P(M) + P(good Service | F) P(F)$$

$$= 0.72 \pm 0.26 \pm 0.13 \pm 0.74 = 0.2834$$

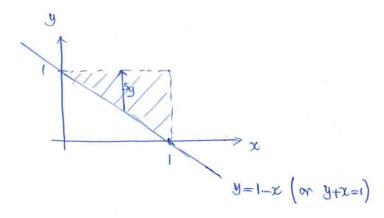
c) 
$$P(M \mid good Service) = \frac{P(M \text{ and good Service})}{P(good Service)} = \frac{0.72 \pm 0.26}{0.2834}$$

$$E\left(\frac{1}{1+\lambda}\right) = \sum_{\chi=0}^{\infty} \frac{1}{1+\chi} \frac{\lambda}{\chi} \frac{e^{-\lambda}}{\chi!}$$

$$= \frac{1}{\lambda} \sum_{\chi=0}^{\infty} \frac{\lambda}{\chi} \frac{e^{-\lambda}}{\chi!}$$

$$= \frac{1}{\lambda} \sum_{y=1}^{\infty} \frac{\lambda}{y!} \frac{e^{-\lambda}}{y!}$$

$$= \frac{1}{\lambda} \left(1 - e^{-\lambda}\right)$$



$$f_{\chi}(x) = \int_{Sy} f(x,y) dy = \int_{1-x}^{1} 2 dy = \frac{\partial y}{\partial x} \Big|_{1-x}^{1} = 2\left(1-(1-x)\right)$$

$$= \frac{\partial x}{\partial x} = \frac{\partial x}{\partial x}$$

$$f(x,y) = 2 \qquad x(y), \quad x(1, y)$$

$$f(x) = 2x$$
  $0\langle x\langle 1$ 

$$\Rightarrow f_{\gamma|\chi}(y|x) = \frac{2}{2x} = \frac{1}{x}$$

$$\begin{cases} y(1) \\ x+y(1) \Rightarrow y(1-x) \\ y(1) \end{cases}$$

$$\begin{cases} y(1) \\ x+y(1) \Rightarrow y(1-x) \end{cases}$$

$$\Rightarrow f_{y|x}(y|x=3/4) = \frac{4}{3} ; \frac{1}{4}(y(1)$$

(c) 
$$f_{Y|X}(y|x) = \frac{1}{x}$$
  $1-x < y < 1$ 

$$f_{y|x}(y|x=\frac{3}{4}) = \frac{4}{3} \qquad \frac{1}{4} \langle y \langle 1 \rangle$$

$$P(y) = \int_{2}^{1} |x=3y| = \int_{2}^{1} f_{y|x}(y|x=3y|) dy$$

$$= \int_{2}^{1} y|x dy = \frac{y}{3} \cdot y|_{2}^{1}$$

$$= \frac{4}{3} \left( 1 - \frac{1}{2} \right) = \frac{34}{3} * \frac{1}{2} = \frac{2}{3}$$



$$f(y;a,\beta) = \alpha \beta y - (\alpha+1)$$

we introduce indicator function 
$$I_{\beta}(y) = \begin{cases} 1 & \text{if } y \text{if }$$

Since of is known; we have

Explore 
$$f(y;\beta) = d\beta y$$
  $I_{\beta}(y)$ 

$$L(\beta) = f_1(y_1) \times \dots \times f_n(y_n) = \prod_{i=1}^n f(y_i) = \prod_{i=1}^n \alpha \beta y_i^{d_i} - (\alpha + i) \prod_{j=1}^n (y_j)$$

$$= d \beta \left( \prod_{i=1}^{n} y_{i} \right) \prod_{i=1}^{n} I_{\beta}(y_{i})$$

$$= \alpha \beta \left( \prod_{i=1}^{n} y_{i} \right) \qquad I_{\beta}(y_{(i)}) = g(y_{(i)}, \beta) h(y_{(i-1)}y_{n})$$

where 
$$y_{(1)} = \min(y_1, y_2, -1y_n)$$

$$g(y_{(i)},\beta) = \beta \operatorname{I}_{\beta}(y_{(i)})$$

$$h(y_{(i+1)}y_n) = \alpha \left( \prod_{i=1}^{n} y_i \right)$$

$$E(y) = \int_{0}^{3} dy \frac{d^{3} - a}{dy} dy = d \frac{3}{3} \frac{y^{d+1}}{d+1} \Big|_{0}^{3} = \frac{3d}{d+1}$$

$$\mu_{i} = \frac{3d}{E(y)} = \frac{3d}{d+1}$$

$$\mu_{i} = m_{i} \Rightarrow \frac{3d}{d+1} = \overline{y}$$

$$M_{i} = \frac{1}{n} \sum_{i=1}^{n} y_{i} = \overline{y}$$

$$= d\overline{y} + \overline{y} = 3d$$

$$\Rightarrow d(3-\bar{y}) = \bar{y} \Rightarrow d = \frac{\bar{y}}{3-\bar{y}}$$

0 < 4, < 3

$$\hat{\lambda}_{MM} = \frac{\bar{Y}}{3 - \bar{Y}}$$

Let  $F_Z(z)$  and  $f_Z(z)$  denote the standard normal distribution and density functions respectively.

$$F_{U}(u) = P(U \leqslant u) = P(Z \leqslant u) = P(-\sqrt{u} \leqslant Z \leqslant \sqrt{u}) = F_{Z}(\sqrt{u}) - F_{Z}(-\sqrt{u})$$

The density function for V is then

$$f_{U}(u) = F_{U}(u) = \frac{1}{2\sqrt{u}} f_{Z}(\sqrt{u}) + \frac{1}{2\sqrt{u}} f_{Z}(-\sqrt{u}) - \frac{1}{\sqrt{u}} f_{Z}(\sqrt{u}) ; \quad u_{X}o$$

Evaluating, we find 
$$f_{\nu}(u) = \frac{1}{\sqrt{R}} \frac{-1/2}{\sqrt{2}} = \frac{1}{\sqrt{R}} \frac{-1/2}{\sqrt{2}}$$

If 
$$V=5-\left(\frac{1}{2}\right) \Rightarrow \gamma=2\left(5-U\right) \Rightarrow \vec{h}(u)=y=2\left(5-U\right)$$

Thus 
$$|\dot{j}| = \left| \frac{\partial y}{\partial u} \right| = 2$$

50 
$$f_{u}(u) = f_{y}(y) |j| = u f_{xx}$$
$$= 2 \cdot f_{y}(\overline{h}(u)) = 2 \cdot (\frac{3}{2})(\overline{h}(u)) + (\overline{h}(u))$$

$$\begin{vmatrix} y: 0 \longrightarrow 1 \\ u=5-\left(\frac{y}{2}\right): \quad y.5 \longrightarrow 5$$

a)  $Z_1^2 + Z_2 + Z_3$ 

Squared standard normal distribution ~ X(1)

 $Z_1^2 \sim \chi_{(1)}^2$   $Z_2^2 \sim \chi_{(1)}^2$   $Z_3^2 \sim \chi_{(1)}^2$ 

The sum of independent  $\chi_{(1)}^2$  and  $\chi_{(1)}^2$  and  $\chi_{(1)}^2$  and  $\chi_{(1)}^2$  has a  $\chi_{(1)}^2$  distribution with degrees of freedom 3.

b)  $v = \frac{(z_1 + z_2)/2}{z_3^2}$ 

The numerator is a X distribution with df=2 divided by its df degree of freedom; the denominator is independent of the numerator and also a X in divided by df=1

Hence ; U ~ F(2,1)

Let p= proportion of overweight children and adolecents. Then, Ho: p=.15 Ha: p<0.15 and the Computed large sample test statistic for a proportion is z=-0.56. This doesn't not lead to a rejection at the d=0.05 level.

## Question 8.65 of text book

$$0.18 - 0.12 \pm 2.326$$
  $\sqrt{\frac{0.18(0.82) + 0.12(0.88)}{100}}$ 

or 
$$0.06\pm0.117$$
 or  $\left(-0.057, 0.177\right)$ 

b) Since the interval contains zero, it is likely that the two assembly lines produce the Same proportion of defectives.

$$L(\alpha) = f_1(y_1) * f_2(y_2) * --- * f_n(y_n)$$

$$= \left(\frac{d+1}{dy_1}\right) \left(\frac{dy_2}{dy_2}\right) - - * \left(\frac{dy_n}{dy_n}\right)$$

$$= \prod_{i=1}^n dy_i = a \left(\prod_{i=1}^n y_i\right)$$

(1) 
$$l(\alpha) = \log L(\alpha) = n \ln \alpha + (\alpha + 1) \sum_{i=1}^{n} \ln y_i$$

(2) 
$$\frac{\partial}{\partial a} l(a) = \frac{n}{a} + \sum_{i=1}^{n} ln y_i = 0 \Rightarrow \frac{n}{a} = -\sum_{i=1}^{n} lg y_i$$

$$\Rightarrow d = -\frac{1}{n} \sum_{i=1}^{n} l_{i} y_{i}$$

From (2) 
$$\frac{\partial^2}{\partial x^2} \mathcal{L}(x) = -\frac{n}{\alpha^2} \langle 0 \rangle \qquad \text{So} \qquad \frac{\lambda}{\lambda} = -\frac{1}{n} \sum_{i=1}^n \ln y_i^2$$

is maximum likelihood estimate for a