

1 Geg  $P(M) = 0,9$   $P(C|M) = 0,8$   $P(C|\bar{M}) = 0,1$

$$a. P(\bar{M}|\bar{C}) = P(\bar{C}|\bar{M})P(\bar{M}) = 0,9 \cdot 0,1 = 0,09$$

$$b. P(\bar{M}|C) = P(C|\bar{M})P(\bar{M}) / P(C)$$

$$= \frac{P(C|\bar{M})P(\bar{M})}{P(C|\bar{M})P(\bar{M}) + P(C|M)P(M)} = \frac{1}{73}$$

2 a

$$f_x(x) = \int_{-\infty}^{\infty} f(x,y) dy = e^{-x} \int_0^{\infty} xe^{-xy} dy = e^{-x}, x \geq 0$$

b.

$$f_{y|x}(y|x) = \frac{f(x,y)}{f_x(x)} = xe^{-xy}, y \geq 0$$

$$c. F_Z(z) = P(Z \leq z) = P(XY \leq z)$$

$$= \iint_{\{(x,y): xy \leq z\}} f(x,y) dy dx$$

$$= \int_0^{\infty} \int_0^{z/x} xe^{-x(y+1)} dy dx$$

$$= \int_0^{\infty} e^{-x} \int_0^{z/x} xe^{-xy} dy dx$$

$$= \int_0^{\infty} e^{-x} (1 - e^{-z}) dx = 1 - e^{-z}, z \geq 0$$

$$f_Z(z) = e^{-z}, z \geq 0$$

$$3 \quad \text{and} \quad P(X > 1) = 1 - (-p)^n - np(1-p)^{n-1}$$

$\dots, x_n), \quad y_2 =$   
in, say,  $x_1 =$   
the joint den-

and so

$$\phi''(0) = E[X^2]$$

In general, the  $n$ th derivative of  $\phi(t)$  evaluated at  $t = 0$  equals  $E[X^n]$ , that is,

$$\phi^n(0) = E[X^n], \quad n \geq 1$$

We now compute  $\phi(t)$  for some common distributions.

**Example 2.39** (The Binomial Distribution with Parameters  $n$  and  $p$ )

defined for all

$$\begin{aligned} b \quad \phi(t) &= E[e^{tX}] \\ &= \sum_{k=0}^n e^{tk} \binom{n}{k} p^k (1-p)^{n-k} \\ &= \sum_{k=0}^n \binom{n}{k} (pe^t)^k (1-p)^{n-k} \\ &= (pe^t + 1 - p)^n \end{aligned}$$

moments of  $X$

Hence,

$$C \quad \phi'(t) = n(pe^t + 1 - p)^{n-1} pe^t$$

and so

$$E[X] = \phi'(0) = np$$

which checks with the result obtained in Example 2.17. Differentiating a second time yields

$$\phi''(t) = n(n-1)(pe^t + 1 - p)^{n-2}(pe^t)^2 + n(pe^t + 1 - p)^{n-1} pe^t$$

and so

$$E[X^2] = \phi''(0) = n(n-1)p^2 + np$$

Thus, the variance of  $X$  is given

$$\begin{aligned} \text{Var}(X) &= E[X^2] - (E[X])^2 \\ &= n(n-1)p^2 + np - n^2 p^2 \\ &= np(1-p) \blacksquare \end{aligned}$$

4 a  $X_i \sim N(75, 100)$

$$E\bar{X}_n = \frac{1}{n} \sum_{i=1}^n EX_i = 75$$

2  $\text{var } \bar{X}_n = \frac{1}{n^2} \sum_{i=1}^n \text{var } X_i = \frac{100}{n}$

$$\therefore \bar{X}_n \sim N(75, \frac{100}{n})$$

b  $P\left(\sum_{i=1}^{12} X_i > 1000\right) = P\left(\bar{X}_{12} > \frac{1000}{12}\right)$

$$= P\left(\frac{\bar{X}_{12} - 75}{10/\sqrt{12}} > \frac{1000/12 - 75}{10/\sqrt{12}}\right)$$

2  $= 1 - \Phi\left(\frac{5}{\sqrt{3}}\right) = 1 - \Phi(2.886) = 0,0019$

c.  $EX_1 = \frac{1}{2}(a+b) = 75$

$$\text{var } X_1 = \frac{1}{12}(b-a)^2 = 100$$

2  $\therefore a = 75 - 10\sqrt{3} \quad b = 75 + 10\sqrt{3}$

d  $P\left(\sum_{i=1}^n X_i > 1000\right) < 0,001$

$$\Leftrightarrow P\left(\frac{\sum_{i=1}^n X_i - 75n}{10\sqrt{n}} > \frac{1000 - 75n}{10\sqrt{n}}\right) < 0,001$$

2 c.l.s.  $\Leftrightarrow 1 - \Phi\left(\frac{1000 - 75n}{10\sqrt{n}}\right) < 0,001$

$$\Leftrightarrow \Phi(\dots) > 0,999$$

$$\Leftrightarrow \frac{1000 - 75n}{10\sqrt{n}} \geq 3,10$$

$$\Leftrightarrow 31\sqrt{n} + 75n \leq 1000 \Leftrightarrow n \leq 11$$

$$5. \quad a \quad P(Y=k) = \int_0^{\infty} P(Y=k|X=x) \lambda e^{-\lambda x} dx \quad 2$$

$$= \lambda \int_0^{\infty} \frac{(\mu x)^k}{k!} e^{-(\lambda+\mu)x} dx$$

$$3 \quad P(Y=0) = \frac{\lambda}{\lambda+\mu}$$

$$P(Y=k) = \frac{\mu}{\lambda+\mu} P(Y=k-1)$$

$$\therefore P(Y=k) = \left(\frac{\mu}{\lambda+\mu}\right)^k \frac{\lambda}{\lambda+\mu} \quad k=0,1,\dots$$

$$1. \quad b \quad E(Y|X=x) = \sum_{k=0}^{\infty} k P(Y=k|X=x) = \mu x$$

$$2. \quad c \quad EY = E(E(Y|X)) = E(\mu X) = \mu EX = \frac{\mu}{\lambda}$$

$$d \quad P(X \leq x | Y=0)$$

$$= \frac{1}{P(Y=0)} P(X \leq x, Y=0) = \frac{1}{P(Y=0)} \int_0^x P(Y=0|X=u) \lambda e^{-\lambda u} du$$

$$= \frac{1}{P(Y=0)} \int_0^x \lambda e^{-(\lambda+u)u} du = \frac{1}{P(Y=0)} \frac{\lambda}{\lambda+u} \left(1 - e^{-\frac{(\lambda+u)x}{\lambda+u}}\right), x \geq 0$$

$$= 1 - e^{-\frac{\lambda x}{\lambda+x}}, x \geq 0$$

$$f_{X|Y}(x|0) =$$

$$= (\lambda+u) e^{-(\lambda+u)x}, x \geq 0$$

6 a  $P(V > t) = P(X > t, Y > t) = e^{-(\frac{1}{a} + \frac{1}{b})t}, t \geq 0$

$$\therefore V \sim E(\frac{1}{a} + \frac{1}{b})$$

$$\therefore EV = \frac{ab}{a+b} \quad \text{var } V = \left(\frac{ab}{a+b}\right)^2$$

b  $P(W=X) = P(X > Y) = \int_0^{\infty} P(X > Y | Y=y) \frac{1}{b} e^{-\frac{1}{b}y} dy$

$$= \dots = \frac{a}{a+b}$$

c  $V+W = X+Y$

$$\therefore EV + EW = EX + EY = a+b$$

$$\therefore EW = a+b - \frac{ab}{a+b}$$