

TT = proportion of non-smokers (women) who get lung cancer

So,  

$$P_{r}(S|L) = \frac{P_{r}(L|S)P_{r}(S)}{P_{r}(L)} = \frac{(0.183)(13\pi)}{(0.183)(13\pi) + (0.817)(\pi)}$$

$$= \frac{(0.183)(13)}{(0.183)(13) + 0.817} = 0.744.$$

$$P_{r}(S|L) = \frac{P_{r}(L|S)P_{r}(S)}{P_{r}(L)}$$

$$= \frac{(0.231)(23\pi)}{(0.231)(23\pi) + (0.769)(\pi)}$$

$$= \frac{(0.231)(23)}{(0.231)(23) + 0.769} = 0.874.$$

2.51 
$$\times \sim Geom(\pi) \implies P_r(\chi = x) = (1-\pi)^x \pi$$

(a)  $P_r(\chi \ge k) = \left[ (1-\pi)^k + (1-\pi)^{k+1} + \dots \right] \pi$ 

$$= (1-\pi)^k \pi \left[ 1 + (1-\pi)^2 + \dots \right] = \frac{(1-\pi)^k \pi}{1-(1-\pi)} = (1-\pi)^k$$

$$P_{r}(X = x \mid X \geq k) = \frac{P_{r}(X \geq k \text{ and } X = x)}{P_{r}(X \geq k)}$$

$$= \begin{cases} 0, & x < k \\ \frac{\pi(1-\pi)^{x}}{(1-\pi)^{u}}, & x \geq k \end{cases} = \begin{cases} 0, & x < k \\ \frac{\pi(1-\pi)^{x-k}}{(1-\pi)^{u}}, & x \geq k \end{cases} = P_{r}(X = x-k).$$

(c) Saying X ≥ k is like starting over.

2.62 (a) 
$$X = 2$$
:  $\binom{4}{2} \binom{26}{5} - 2\binom{13}{5} \binom{52}{5} = 0.1459$ 

$$X = 4$$
:  $\binom{4}{1} \binom{13}{2} \binom{13}{1} \binom{13}{5} \binom{52}{5} = 0.2637$ 

$$X = 7r(X = x)$$

$$0.00198$$

$$0.1459$$

$$0.5884$$

$$X = 3$$
:  $1 - (P_r(X = 1) + P_r(X = 2) + P(X = 4)) = 0.5884$ 

(b) 
$$E(x) = (0.00198) + (2)(0.1459) + (3)(0.5884) + (4)(0.2637) = 3.114$$

3.5 For 
$$X \sim \exp(\lambda)$$
, we have  $c \le f = \begin{cases} 0 & , & x < 0 \\ 1 - e^{-\lambda x}, & x \ge 0 \end{cases}$ 

For the median, we solve  $0.5 = 1 - e^{-\lambda x} \Rightarrow x = \frac{1}{\lambda} \ln 2$ .

The first quartile  $\times$  satisfies  $0.25 = 1 - e^{-\lambda x} \Rightarrow \times = -\frac{1}{\lambda} \ln(3/4)$ .

The third quartile  $\times$  satisfies  $0.75 = 1 - e^{-\lambda x} \Rightarrow x = \frac{2}{\lambda} \ln 2$ .

3.22 For 
$$X \sim Geom(\pi)$$
,  $f_{\chi(x)} = (1-\pi)^x \pi$   

$$\Rightarrow M_{\chi}(t) = E(e^{tX}) = \sum_{x=0}^{\infty} e^{tx} (1-\pi)^x \pi = \pi \sum_{x=0}^{\infty} \left[ e^t (1-\pi) \right]^x$$

$$= \pi \cdot \frac{1}{1-e^t (1-\pi)} \qquad -\frac{1}{t} \int_0^t e^{ty} dy + \frac{1}{t} \int_1^2 e^{ty} dy = -\frac{1}{t^2} \left[ e^{ty} \right]_0^1 + \frac{1}{t^2} \left[ e^{ty} \right]_1^2$$

$$= \frac{1}{t^2} (1-e^t) + \frac{1}{t^2} \left[ e^{2t} - e^t \right]$$

3.23 
$$M_{\gamma}(t) = \int_{-\infty}^{\infty} e^{ty} f(y) dy = \int_{0}^{1} y e^{ty} dy + \int_{0}^{2} (2-y) e^{ty} dy$$

$$= \frac{1}{t} y e^{ty} \Big|_{0}^{1} - \frac{1}{t} \int_{0}^{1} e^{ty} dy + \frac{1}{t} (2-y) e^{ty} \Big|_{1}^{2} + \frac{1}{t} \int_{0}^{2} e^{ty} dy$$

$$= \frac{1}{t} e^{t} - \frac{1}{t^{2}} \Big[ e^{ty} \Big]_{0}^{1} - \frac{1}{t} e^{t} + \frac{1}{t^{2}} \Big[ e^{ty} \Big]_{1}^{2}$$

$$= \frac{1}{t^{2}} \Big( 1 - e^{t} \Big) + \frac{1}{t^{2}} \Big( e^{2t} - e^{t} \Big) = \frac{1}{t^{2}} \Big( 1 - 2e^{t} + e^{2t} \Big) = \frac{1}{t^{2}} \Big( 1 - e^{t} \Big)^{2}.$$

3.31 
$$M'_{x}(t) = 2e^{2t}(1-t^{2})^{-1} + 2te^{2t}(1-t^{2})^{-2} \longrightarrow E(x) = M'_{x}(0) = 2$$
 $M''_{x}(t) = 4e^{2t}(1-t^{2})^{-1} + 8te^{2t}(1-t^{2})^{-2} + 2e^{2t}(1-t^{2})^{-2} + 8t^{2}e^{2t}(1-t^{2})^{3}$ 
 $\longrightarrow E(x^{2}) = M''_{x}(0) = 6$ 

Thus,  $V_{ar}(x) = 6 - 7^{2} = 2$ .

3.33 
$$M'_{X}(t) = \frac{18}{(3-t)^3} \longrightarrow E(X) = M'_{X}(0) = \frac{2}{3}$$
  
 $M''_{X}(t) = \frac{54}{(3-t)^4} \longrightarrow E(X^2) = M''_{X}(0) = \frac{2}{3}$   
So,  $Var(X) = \frac{2}{3} - \left(\frac{2}{3}\right)^2 = \frac{2}{9}$ .

3.37 (a) Since 
$$X \sim Binom(n, \pi)$$
 has  $MGF$   $M(t) = (1 - \pi + \pi e^t)^n$ , when  $M(t) = (\frac{1}{2}(e^t + 1))^{10}$ ,  $X \sim Binom(10, \frac{1}{2})$ .

(b) Since 
$$X \sim Norm(\mu, \sigma)$$
 has  $MGF$   $M_{\chi}(t) = e^{\mu t + \sigma^2 t^2/2}$ , when  $M_{\chi}(t) = e^{t + t^2/2}$ ,  $X \sim Norm(1, 1)$ .

(c) Since 
$$X \sim \text{Exp}(\lambda)$$
 has MGF  $M(t) = \frac{1}{1 - t/\lambda}$ , when  $M(t) = \frac{1}{1 - 2t}$ ,  $X \sim \text{Exp}(Y_2)$ .

(d) Since 
$$X \sim Gamma(\alpha, \lambda)$$
 has  $MGF M(t) = \frac{1}{(1 - t/\lambda)^{\alpha}}$   
when  $M(t) = (1 - 2t)^{-3}$ ,  $X \sim Gamma(\alpha = 3, \lambda = 1/2)$ , or  $Gamma(\alpha = 3, \beta = 2)$ .

 $X \sim Gamma(\alpha, \lambda)$ , so  $M_{\chi}(t) = \frac{1}{(1-t/\lambda)^{\alpha}}$ . Setting Y = 3X, we have 3.38  $M_{y}(t) = E(e^{ty}) = E(e^{t(3X)}) = E(e^{(3t)x}) = M_{x}(3t) = \frac{1}{(1-3t/x)^{\alpha}}$ ⇒ Y ~ Gamma (x, 1/3).

$$3.39$$
 (a)  $pexp(2) - pexp(0) = 0.865$ 

(a) 
$$pexp(2) - pexp(0) = 0.865$$
 (d)  $E(X) = \frac{1}{3}$ ,  $Var(X) = \frac{2}{63}$   
(b)  $pexp(1,2) - pexp(0,2) = 0.865$  diff(pbeta(\frac{1}{3} + c(-1,1) \* sqrt(2/63), 2,4))
$$= 0.6522$$

(c) 
$$\frac{2}{2\sqrt{3}} \cdot (b-a) \cdot \frac{1}{b-a} = \frac{1}{\sqrt{3}} = 0.5774$$
.

3.62 (a) Because R~ Norm(100, 20), his obtaining 150 would correspond to a Z-score  $Z_{R} = \frac{150 - 100}{20} = 2.5$ 

For 
$$C \sim Norm(110, 15)$$
,  $Z_c = \frac{150 - 110}{15} = 2.667$ 

A higher Z-score corresponds to a rarer event. Thus, Ralph should reach scores of 150 (or higher) more often than Claudia.

(b) By normality and independence, R-C ~ Norm (-10, \(\sigma 15^2 + 20^2\) = Norm (-10, 25) Pr(R>C) = 1 - pnorm (0, -10, 25) = 0.345

- (C) Let R, C be their averages over three games. R~ Norm (100, 20/13), C~ Norm (110, 15/13) and R-C~ Norm (-10, 25/13).  $Pr(R > C) = 1 - pnorm(0, -10, \frac{25}{\sqrt{3}}) = 0.244.$
- (d) Let X = # of games won by Ralph. Assuming independence, X ~ Binom (3, 0.345) Pr(X = 2) = 1 - phinom (1, 3, 0.345) = 0.275.

$$C.4 \quad (a) \quad \text{proj}(\vec{u} \rightarrow \vec{v}) = \frac{\langle 1,0 \rangle \cdot \langle 1,1 \rangle}{\langle 1,1 \rangle \cdot \langle 1,1 \rangle} \langle 1,1 \rangle = \frac{1}{2} \langle 1,1 \rangle = \langle \frac{1}{2}, \frac{1}{2} \rangle.$$

(b) 
$$\operatorname{proj}(\vec{u} \rightarrow \vec{v}) = \frac{\langle 1,0\rangle \cdot \langle 1,-1\rangle}{\langle 1,-1\rangle \cdot \langle 1,-1\rangle} \langle 1,-1\rangle = \frac{1}{2} \langle 1,-1\rangle = \langle \frac{1}{2},-\frac{1}{2}\rangle.$$

(c) 
$$\operatorname{proj}(\vec{u} \rightarrow \vec{v}) = \frac{\langle 1,0 \rangle \cdot \langle 1,2 \rangle}{\langle 1,2 \rangle \cdot \langle 1,2 \rangle} \langle 1,2 \rangle = \frac{1}{5} \langle 1,2 \rangle = \langle \frac{1}{5}, \frac{2}{5} \rangle$$

(d) 
$$\text{Proj}(\vec{u} \rightarrow \vec{v}) = \frac{\langle 1,2,3\rangle \cdot \langle 1,1,1\rangle}{\langle 1,1,1\rangle \cdot \langle 1,1,1\rangle} \langle 1,1,1\rangle = \frac{6}{3} \langle 1,1,1\rangle = \langle 2,2,2\rangle.$$

(e) 
$$\text{proj}(\vec{u} \rightarrow \vec{r}) = \frac{\langle 1, 1, 1 \rangle \cdot \langle 1, 2, 3 \rangle}{\langle 1, 2, 3 \rangle \cdot \langle 1, 2, 3 \rangle} \langle 1, 2, 3 \rangle = \frac{6}{14} \langle 1, 2, 3 \rangle = \langle \frac{3}{7}, \frac{6}{7}, \frac{9}{7} \rangle$$

$$(f) \quad \text{proj}(\vec{u} \to \vec{v}) = \frac{\langle 1, 2, 3 \rangle \cdot \langle 1, -1, 0 \rangle}{\langle 1, -1, 0 \rangle \cdot \langle 1, -1, 0 \rangle} \langle 1, -1, 0 \rangle = -\frac{1}{2} \langle 1, -1, 0 \rangle = \langle -\frac{1}{2}, \frac{1}{2}, 0 \rangle$$

C.21 This statement is true. To demonstrate it, let  $B = (A^T)^T$ . Then  $I = BA^T$ . Taking transposes of both sides and noting  $I^T = I$ , we have  $I = (BA^T)^T = AB^T$ . Showing that  $B^T = A^{-1}$ . Transposing again gives  $B = (A^{-1})^T$ .

(b) It is evident that 
$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}$$
, when multiplied by the now-rescaled  $\frac{1}{ad-bc}\begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$ , is I

(c) 
$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}^{-1} = -\frac{1}{2} \begin{bmatrix} 4 & -2 \\ -3 & 1 \end{bmatrix} = \begin{bmatrix} -2 & 1 \\ 3/2 & -1/2 \end{bmatrix}.$$