We have been told

$$P_r[C|M \cap S] = 23 \cdot P_r[C|M \cap S^c],$$

 $P_r[C|W \cap S] = 13 \cdot P_r[C|W \cap S^c],$

 $Pr[S|M] = 0.231, Pr[S|W] = 0.183 \Rightarrow Pr[S^{c}|M] = 0.769, Pr[S^{c}|W] = 0.817.$

(a)
$$P_r[W|S] = \frac{\# \text{ of women who smoke}}{\# \text{ of smokers}} = \frac{21.1}{21.1 + 24.8} = 0.46$$

(b)
$$P_r[s|wnc] = \frac{P_r[snwnc]}{P_r[wnc]} = \frac{P_r[c|snw] \cdot P_r[snw]}{P_r[cnwns] + P_r[cnwns^c]}$$

$$= \frac{13 Pr[S \cap w]}{13 Pr[S \cap w] + Pr[S^c \cap w]} = \frac{13 Pr[S | w] Pr[w]}{13 Pr[S | w] Pr[w] + Pr[S^c | w] Pr[w]}$$

$$= \frac{13 \Pr[S|W]}{13 \Pr[S|W] + \Pr[S'|W]} = \frac{13(0.183)}{13(0.183) + 0.817} \approx 0.744$$

$$Pr[S|Mnc] = \frac{23(0.231)}{23(0.231) + 0.769} = 0.874$$

2.51
$$\times \sim Geom(\pi) \Rightarrow \Pr(\times = \times) = (1-\pi)^{\times} \pi$$

(a)
$$P_r(X \ge k) = [(1-\pi)^k + (1-\pi)^{k+1} + ...] \pi$$

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

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

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

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

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

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

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

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

(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 call $F_X(x) = \begin{cases} 0 & , & x < 0 \\ 1 - e^{-\lambda x}, & x \ge 0 \end{cases}$

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

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

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

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

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

$$= \pi \cdot \frac{1}{1-e^t (1-\pi)}$$

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

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

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

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

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 $MGF M_{\chi}(t) = \frac{1}{(1 - t/\lambda)^{\alpha}}$
when $M_{\chi}(t) = (1 - 2t)^{-3}$, $\chi \sim Gamma(\alpha = 3, \lambda = 1/2)$, 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)
$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}^{-1} = -\frac{1}{2} \begin{bmatrix} 4 & -2 \\ -3 & 1 \end{bmatrix} = \begin{bmatrix} -2 & 1 \\ 3/2 & -1/2 \end{bmatrix}.$$