3.10 $E(Y) = \sum yp(y) = 1(.4) + 2(.3) + 3(.2) + 4(.1) = 2.0$ $E\left(\frac{1}{Y}\right) = \sum \frac{1}{y} p(y) = 1(.4) + \frac{1}{2}(.3) + \frac{1}{3}(.2) + \frac{1}{4}(.1) = .6417$ $E(Y^2 - 1) = E(Y^2) - 1 = [1(.4) + 4(.3) + 9(.2) + 16(.1)] - 1 = 5 - 1 = 4$ Using Theorem 3.6,

$$V(Y) = E(Y^2) - [E(Y)]^2 = 5 - (2)^2 = 1$$

3.16 Since the die is fair, the probability distribution for Y is

$$p(y) = \frac{1}{6}$$
 $y = 1, 2, 3, 4, 5, 6$

Then

$$E(Y) = \sum yp(y) = \frac{1}{6}(1+2+\ldots+6) = \frac{21}{6} = 3.5$$

$$E(Y^2) = \sum y^2p(y) = \frac{1}{6}(1+4+9+\ldots+36) = \frac{91}{6} = 15.1667$$

$$V(Y) = E(Y^2) - [E(Y)]^2 = 15.1667 - (3.5)^2 = 2.9167$$

3.17 Define G to be the gain to a person in drawing one card. G can take on only three values, \$15, \$5, or \$-4, with probabilities as shown in the accompanying table.

$$\begin{array}{ccc}
G & p(G) \\
\hline
15 & \frac{2}{13} \\
5 & \frac{2}{13} \\
-4 & \frac{9}{13}
\end{array}$$

Then $E(G) = \sum G p(G) = 15\left(\frac{2}{13}\right) + 5\left(\frac{2}{13}\right) - 4\left(\frac{9}{13}\right) = \frac{4}{13} = .31$

- The expected gain is \$.31.
- 3.21 Let Y be a random variable representing the payout on an individual policy.

 $P(Y=85,000)=P(\text{total loss})=.001,\ P(Y=45,000)=P(50\% \text{ loss})=.01,\ \text{and}\ P(Y=0)=1-.001-.01=.989.$ Let C represent the premium the insurance company charges. Then the company's net gain or loss for this policy is given by C-y. To yield a long term average loss of 0 the company should choose C such that:

E(C-Y)=0, or C=E(Y). Then we have:

$$E(Y) = \sum_{y} yp(y)$$
= (85,000)(.001) + (42,500)(.01) + (0)(.989)
= 510 = C.

3.115 Using the binomial theorem, $(a+b)^n = \sum_{y=0}^n \binom{n}{y} a^y b^{n-y}$, we have:

$$m(t) = E(e^{ty}) = \sum_{y=0}^{n} {n \choose y} (pe^{t})^{y} q^{n-y} = (pe^{t} + q)^{n}.$$

3.116 Refer to Exercise 3.115.

$$E(Y) = \frac{d}{dt} m(t) \Big|_{t=0} = (pe^t + q)^{n-1} n p e^t \Big|_{t=0} = n(p+q)^{n-1} p = n p.$$

$$E(Y^2) = \frac{d^2}{dt^2} m(t) \Big|_{t=0}$$

$$E(Y) = \frac{1}{dt^2} m(t) \Big|_{t=0}$$

$$= npe^t (n-1)pe^t (pe^t + q)^{n-2} + n(pe^t + q)^{n-1}pe^t \Big|_{t=0}$$

$$= np^{2}(n-1) + np V(Y) = E(Y^{2}) - n^{2}p^{2} = n^{2}p^{2} - np^{2} + np - n^{2}p^{2} = np(1-p) = npq$$

3.117 Recall that $\sum_{y=1}^{\infty} q^{y-1}p = 1$ or equivalently $\sum_{y=1}^{\infty} q^{y-1} = \frac{1}{p}$ for 0 < q < 1. Now we have

$$m(t) = E(e^{ty}) = \sum_{y=1}^{\infty} pe^{ty}q^{y-1} = pe^{t} \sum_{y=1}^{\infty} e^{t(y-1)}q^{y-1} = pe^{t} \sum_{y=1}^{\infty} (qe^{t})^{y-1} = \frac{pe^{t}}{1-qe^{t}}$$

if $qe^t < 1$ or equivalently, $t < -\ln q$.

3.118 Refer to Exercise 3.117.

$$\begin{split} E(Y) &= \frac{d}{dt} \, m(t) \big|_{t=0} = \frac{(1-qe^t)(pe^t) - pe^t(-qe^t)}{(1-qe^t)^2} \Big|_{t=0} = \frac{pe^t}{(1-qe^t)^2} \Big|_{t=0} \\ &= \frac{p}{(1-q)^2} = \frac{1}{p} \\ E(Y^2) &= \frac{d}{dt} \frac{pe^t}{(1-qe^t)^2} \Big|_{t=0} = \frac{(1-qe^t)^2 pe^t - 2pe^t(-qe^t)(1-qe^t)}{(1-qe^t)^4} \Big|_{t=0} \\ &= \frac{p^3 + 2pq^2}{p^4} = \frac{p + 2q}{p^2} = \frac{1+q}{p^2} \end{split}$$

Finally,

$$V(Y) = \frac{1+q}{p^2} - \frac{1}{p^2} = \frac{q}{p^2}$$

3.123 a. Differentiate m(t) to find the necessary moments.

$$E(Y) = \frac{d}{dt} m(t) \Big|_{t=0} = \frac{1}{6} e^t + \frac{4}{6} e^{2t} + \frac{9}{6} e^{3t} \Big|_{t=0} = \frac{14}{6} = \frac{7}{3}$$

b.
$$E(Y^2) = \frac{d^2}{dt^2} m(t) \Big|_{t=0} = \frac{1}{6} + \frac{8}{6} + \frac{27}{6} = 6, \ V(Y) = 6 - \left(\frac{7}{3}\right)^2 = \frac{5}{9}.$$

- c. Since $m(t) = E(e^{ty})$, Y must take only the values Y = 1, 2, and 3, with
- 3.132 The random variable Y has a binomial distribution with p = .35 and n = 2300.

a.
$$E(Y) = np = (2300)(.35) = 805$$

b.
$$V(Y) = npq = (2300)(.35)(.65) = 523.25 = \sigma^2$$

 $\sigma = \sqrt{523.25} = 22.875$

- The interval is (805 2(22.875), 805 + 2(22.875)) = (759.25, 850.75)
- **d.** The observation Y = 249 is 24.3 standard deviations below the mean value 805 $(\frac{805-249}{27.875} = 24.3)$. This value (Y = 249) is not consistent with a rate of 35%

as Tchebyscheff's theorem tells us $P(|Y - 805| > (24.3)(22.875)) \le \frac{1}{24.3^2} \approx .002$.

3.133 a.
$$E(Y) = (-1)p(-1) + (0)p(0) + (1)p(1)$$

= $-1\left(\frac{1}{18}\right) + 0\left(\frac{16}{18}\right) + 1\left(\frac{1}{18}\right)$
= 0.

$$E(Y^{2}) = (-1)^{2}p(-1) + (0)^{2}p(0) + (1)^{2}p(1)$$

$$= (-1)^{2}(\frac{1}{18}) + (0)^{2}(\frac{16}{18}) + (1)^{2}(\frac{1}{18})$$

$$= \frac{1}{9}$$

$$V(Y) = E(Y^{2}) - (E(Y))^{2}$$
$$= \frac{1}{9} - 0 = \frac{1}{9}$$

b.
$$\sigma = \sqrt{V(Y)} = \sqrt{\frac{1}{9}} = \frac{1}{3}$$
.

By Tchebysheff's theorem,

$$P(|y-\mu|\geq 3\sigma)\leq \frac{1}{3^2}=\frac{1}{9}.$$

According to the probability distribution of Y,

$$P(|y - \mu| \ge 3\sigma) = P(|y| \ge 1)$$

$$= p(-1) + p(1)$$

$$= \frac{1}{18} + \frac{1}{18} = \frac{1}{9}$$
so that the bound is attained when $k = 3$.

Let x have the probability distribution

$$p(-1) = \frac{1}{8}$$
 $p(0) = \frac{6}{8}$ and $p(1) = \frac{1}{8}$

 $p(-1) = \frac{1}{8}$ $p(0) = \frac{6}{8}$ so that E[x] = 0 and $V(x) = E[x^2] = \frac{1}{4}$.

It follows that
$$P(|X - \mu_x| \ge 2\sigma_x) = P(|x| \ge 1) = p(-1) + p(1) = \frac{1}{4}$$
,

as desired.

Letting all the probability mass be on values
$$-1$$
, 0 , 1 , $E(W) = 0$ if $p(-1) = p$ $p(0) = 1 - 2p$ and $p(1) = p$ for some probability p . We want $k\sigma_W = 1$ so that $\sigma_W = \frac{1}{k}$ and $\sigma_W^2 = \frac{1}{k^2}$. With $E(W) = 0$, the $V(W) = E(W^2) = 2p$. Setting $2p = \frac{1}{k^2}$ gives $p = \frac{1}{2k^2}$. Therefore for any specified $k > 1$, $P(|W - \mu_W| \ge k\sigma_W) = \frac{1}{k^2}$ if

 $p(-1) = \frac{1}{2k^2}$ $p(0) = 1 - \frac{1}{k^2}$ and $p(1) = \frac{1}{2k^2}$ Alternatively, we can show the same result using complements. We want, with $k\sigma_W=1$,

$$P(W = 0) = P(|W - \mu_W| < k\sigma_W) = 1 - \frac{1}{k^2}$$

Then, in order for E(W) = 0, we must have the same distribution as above.

- **3.134** Similar to Exercise 3.131 a. The interval .48 to .52 is the interval $|Y .50| \le 2\sigma$. Hence the lower bound is $1 - \left(\frac{1}{k^2}\right) = 1 - \left(\frac{1}{4}\right) = \frac{3}{4}$. The expected number of coins is then at least $(\frac{3}{4})(400) = 300$.
- **3.139** Let Y be the number of fatalities. Then Y is binomial with p = .0006 and n = 40,000.

a.
$$E(Y) = np = 40,000(.0006) = 24$$

a.
$$E(Y) = np = 40,000(.0000)$$

b. $V(Y) = npq = 24(.9994) = 23.9856 = \sigma^2$
 $\sigma = \sqrt{23.9856} = 4.898$

- $\sigma = \sqrt{23.9856} = 4.898$ No. The value 40 is $\frac{40-24}{4.898} = 3.26$ standard deviations above the mean, and both Tchebyscheff's theorem and the empirical rule suggest this is unlikely. (Tchebyscheff's theorem is probably more relevant in this situation, see 3.120).
- **3.141** The mean of C is E(C) = \$50 + \$3 E(Y) = \$50 + \$3(10) = \$80. The variance is V(C) = V(50 + 3Y) = 9V(Y) = 9(10) = 90, so that $\sigma = \sqrt{90} = 9.487$. Using Tchebysheff's theorem with k=2, we have $P(|Y-80|<2(9.487))\geq .75$ so that the required interval is (80 - 2(9.487), 80 + 2(9.487)) or (61.03, 98.97).
- 3.160 Write

Write
$$(q + pe^t)^n \left[q + p \left(1 + t + \frac{t^2}{2!} + \frac{t^3}{3!} + \dots \right) \right]^n = \left(q + p + pt + p \frac{t^2}{2!} + p \frac{t^3}{3!} + \dots \right)^n$$

$$= \left(1 + pt + p \frac{t^2}{2!} + p \frac{t^3}{3!} + \dots \right)^n$$

$$= \left(1 + pt + p \frac{t^2}{2!} + p \frac{t^3}{3!} + \dots \right)^n$$

The terms that are of interest in this exercise are only those terms that contain either tor t^2 , since we are interested in obtaining only μ_1' and μ_2' , the coefficients of t and $\frac{t^2}{2!}$. Hence we need only to expand the above multinomial to show the first few terms. Then

$$(q + pe^t)^n = \left[1^n + n(pt)(1)^{n-1} + n\left(p\frac{t^2}{2!}\right)(1)^{n-1} + \frac{n(n-1)}{2}(pt)^2(1)^{n-2} + (\text{terms involving } t^3 \text{ and higher powers}) \right].$$

Recall that the multinomial coefficient was given in Chapter 2 as $\frac{n!}{n_1! \, n_2! \cdots n_k!}$

$$\frac{n!}{n_1! \, n_2! \cdots n_k!}$$

where n_i represents the exponent given in the i^{th} member of the multinomial sum in the particular term we wish to evaluate. For example, the fourth term in the above expansion is actually

$$\frac{\frac{n!}{2!(n-2)!0!0!\cdots}(1)^{n-2}(pt)^2\left(p\,\frac{t^2}{2!}\right)^0\left(p\,\frac{t^3}{3!}\right)^0\cdots=\frac{n(n-1)}{2}\,p^2t^2}{\frac{n!}{2!(n-2)!0!0!\cdots}(1)^{n-1}-nn}$$
 The term involving t^2 is

Thus the coefficient of t is $np(1)^{n-1}=np$. The term involving t^2 is $np(1)^{n-1}\frac{t^2}{2}+n(n-1)p^2\frac{t^2}{2}$

$$np(1)^{n-1}\frac{t^2}{2} + n(n-1)p^2\frac{t^2}{2}$$

so that the coefficient of $\frac{t^2}{2}$ is $np + n(n-1)p^2$, which agrees with the results of Exercise 3.100.

3.162 Let W = # of drivers who wish to park, and W' = # of cars, which is Poisson with mean λ . We consider



$$P(W = k) = \sum_{n=k}^{\infty} P(W = k \cap W' = n) = \sum_{n=k}^{\infty} P(W = k|W' = n) P(W' = n)$$

$$= \sum_{n=k}^{\infty} \left\{ \left[\frac{n!}{k! (n-k)!} \right] p^k (1-p)^{n-k} \right\} \left[e^{-\lambda} \left(\frac{\lambda^n}{n!} \right) \right]$$

$$= \lambda^k e^{-\lambda} \left(\frac{p^k}{k!} \right) \sum_{n=k}^{\infty} \left[\frac{(1-p)^{n-k}}{(n-k)!} \right] \lambda^{n-k}$$

$$= \left(\frac{\lambda^k p^k e^{-\lambda}}{k!} \right) \sum_{j=0}^{\infty} \frac{\left[(1-p)\lambda \right]^j}{j!} = \left[\frac{(\lambda p)^k}{k!} \right] e^{-\lambda} e^{(1-p)\lambda}$$

$$= \left[\frac{(\lambda p)^k}{k!} \right] e^{-\lambda p}.$$

Where j = n - k, in line 4 of the previous equation.

a. If W = # of drivers who wish to park, then the probability that a space will still be available when you reach the lot = P(W = 0). Using the information that was derived above,

$$P(W=0) = \left\lceil \frac{(\lambda p)^0}{0!} \right\rceil e^{-\lambda p} = e^{-\lambda p}.$$

b. Using the information derived above, we see that the probability distribution for W is Poisson with mean λp .