

Introduction to Probability Models 11th Edition Ross Solutions M

2.
$$-n$$
, $-n + 2$, $-n + 4$, . . . , $n - 2$, n

3.
$$P{X = -2} = \frac{1}{4} = P{X = 2}$$

 $P{X = 0} = \frac{1}{2}$

- 4. (a) 1, 2, 3, 4, 5, 6
 - (b) 1, 2, 3, 4, 5, 6
 - (c) 2,3,...,11,12
 - (d) -5, -4, ..., 4, 5

11

5.
$$P\{\max = 6\} = \overline{36} = P\{\min = 1\}$$

 $P\{\max = 5\} = \frac{1}{-} P\{\min = 2\}$
 $4 = \frac{1}{36} = P\{\min = 2\}$
 $P\{\max = 4\} = \frac{7}{36} = P\{\min = 3\}$
 $P\{\max = 3\} = \frac{5}{36} = P\{\min = 4\}$
 $P\{\max = 2\} = \frac{1}{12} = P\{\min = 5\}$
 $P\{\max = 1\} = \frac{1}{36} = P\{\min = 6\}$

6.
$$(H, H, H, H, H)$$
, p^5 if $p = P\{\text{heads}\}$

7.
$$p(0) = (.3)^3 = .027$$

 $p(1) = 3(.3)^2(.7) = .189$
 $p(2) = 3(.3)(.7)^2 = .441$
 $p(3) = (.7)^3 = .343$

8.
$$p(0) = \frac{1}{2}, p(1) = 2^{-1}$$

9. $p(0) = \frac{1}{2}, p(1) = \frac{1}{10}, p(2) = 5^{-1}$
 $p(3) = T0, p(3.5) = \frac{1}{10}$

$$10.1 - \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \end{bmatrix}^{2} \begin{bmatrix} 1 & 5 \\ 6 & 6 \end{bmatrix} - \begin{bmatrix} 1 & 1 \end{bmatrix}^{3} = \begin{bmatrix} 200 \\ 216 \end{bmatrix}$$

11.
$$\frac{3}{8}$$
12. $\frac{[5][1]^4}{4}[2] + \frac{[5][1]^5}{5} = \frac{10+1}{243} = \frac{11}{243}$

$$\sum (10)[1]10$$

13.
$$i = 7$$
14.
$$P\{X = 0\} = P\{X = 6\} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \frac{1}{64}$$

$$P\{X = 1\} = P\{X = 5\} = 6 - \frac{6}{64}$$

$$P\{X = 2\} = P\{X = 4\} = \begin{bmatrix} 6 \\ 1 \end{bmatrix} = \begin{bmatrix} 6 \\ 64 \end{bmatrix}$$

$$P\{X = 3\} = \begin{bmatrix} 6 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$$
15.
$$P\{X = k\} = \begin{bmatrix} 6 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 64 \end{bmatrix} = \begin{bmatrix} 1 \\ 64$$

Hence,

$$\frac{P\{X = k\}}{P\{X = k - 1\}} \ge 1 \leftrightarrow (n - k + 1)p > k(1 - p)$$
$$\leftrightarrow (n + 1) p \ge k$$

The result follows.

- 16. $1 (.95)^{52} 52(.95)^{51}(.05)$
- 17. Follows since there are $n!_{x_1! \cdots x_r ! p}$ ermutations of nobjects of which x_1 are alike, x_2 are alike,..., x_r are alike.

are anke,...,
$$x_r$$
 are anke.
18. (a) $P(X_i = x_i, i = 1, ..., r - 1 | X_r = j)$

$$= P(X_i = x_i, i = 1, ..., r - 1, X_r = j)$$

$$= P(X_r = j)$$

$$= \frac{x_1! ... x_{r-1}! j!}{n!} \frac{p_1}{j} ... p^{x_{r-1}} r - 1} r$$

$$= \frac{n!}{n!} \frac{x_1}{j!} \frac{p_1}{(n-j)!} \frac{p_1}{p_r} (1 - p_r)^n$$

$$= \frac{(n-j)!}{x_1! ... x_{r-1}!} \frac{p_i}{j!} \frac{p_i}{1 - p_r}$$

- (b) The conditional distribution of X_1, \ldots, X_{r-1} given that $X_r = j$ is multinomial with parameters $n - j_{r}^{p_{i}}, i = 1, ..., r - 1$.
- (c) The preceding is true because given that $X_r = j$, each of the n j trials that did not result in outcome r resulted in outcome i with probability $p_{i_1-p_r}$, $i = 1, \ldots, r - 1.$

19.
$$P\{X_1 + \cdots + X_k = m\}$$

$$= {n \choose m} {p_1 \choose r} + \cdots + p_k {m \choose p_{k+1}} + \cdots + p_r {n-m \choose r}$$

20.
$$\frac{5!}{2!1!2!} \frac{\begin{bmatrix} 1 \\ 1 \end{bmatrix}^{2}}{\frac{5}{5}} \frac{3}{10} \frac{1}{2} = .054$$

$$21. 1 - \frac{3}{10} - 5 \frac{3}{10} \frac{1^{4}}{10} \frac{7}{10} - \frac{5!}{2} \frac{1}{10} \frac{1^{3}}{10} \frac{7}{10}^{2}$$

$$22. \frac{1}{32}$$

23. In order for X to equal n, the first n-1 flips must have r-1 heads, and then the nth flip must land heads. By independence the desired probability is thus

$$\begin{bmatrix} n-1 \end{bmatrix} p^{r-1} (1-p)^{n-r} xp$$

- 24. It is the number of tails before heads appears for the r th time.
- 25. A total of 7 games will be played if the first 6 result in 3 wins and 3 losses. Thus,

$$P{7 \text{ games}} = {6 \choose 3} p^3 (1 - p^3)$$

Differentiation yields

$$\frac{d}{dp}P\{7\} = 20 \, 3p^2 (1-p)^3 - p^3 3(1-p)^2$$

$$= 60 \, p^2 (1-p)^2 [1-2 \, p]$$

Thus, the derivative is zero when p = 1/2. Taking the second derivative shows that the maximum is attained at this value.

26. Let *X* denote the number of games played.

(a)
$$P{X = 2} = p^2 + (1 - p)^2$$

 $P{X = 3} = 2p(1 - p)$
 $E[X]=2$

$$p^2 + (1 - p)^2 + 6p(1 - p)$$

$$= 2 + 2p(1 - p)$$

Since p(1-p) is maximized when p = 1/2, we see that E[X] is maximized at that value of p.

(b)
$$P\{X = 3\} = p^3 + (1 - p)^3$$

 $P\{X = 4\}$
 $= P\{X = 4, \text{I has 2 wins in first 3 games}\}$
 $+ P\{X = 4, \text{II has 2 wins in first 3 games}\}$
 $= 3p^2(1 - p)p + 3p(1 - p)^2(1 - p)$
 $P\{X = 5\}$
 $= P\{\text{each player has 2 wins in the first 4 games}\}$
 $= 6p^2(1 - p)^2$

$$E[X]=3 \begin{bmatrix} p^3 + (1-p)^3 + 12 p(1-p) \\ p^2 + (1-p)^2 + 30 p^2(1-p)^2 \end{bmatrix}$$

Differentiating and setting equal to 0 shows that the maximum is attained when p = 1/2.

27.
$$P\{\text{same number of heads}\} = \sum_{P\{A=i, B=i\}}$$

$$= \sum_{i} {\binom{n}{k}} {\binom{n-k}{n-k}}^{n-k} {\binom{1/2}{2}}^{n-k}$$

$$= \sum_{i} {\binom{k}{n-k}} {\binom{n-k}{i}}^{n}$$

$$= \sum_{i} {\binom{k}{k-i}} {\binom{n-k}{i}} {\binom{1/2}{2}}^{n}$$

$$= {\binom{n}{k}} {\binom{n-k}{k}}^{n}$$

Another argument is as follows:

$$P$$
{# heads of A = # heads of B }
= P {# tails of A = # heads of B }

since coin is fair

=
$$P\{k - \# \text{ heads of } A = \# \text{ heads of } B\} = P\{k = \text{ total } \# \text{ heads}\}$$

28. (a) Consider the first time that the two coins give different results. Then

$$P\{X = 0\} = P\{(t,h)|(t,h) \text{ or } (h,t)\}$$
$$= p\frac{(1-p)}{2p(1-p)} = \frac{1}{2}$$

(b) No, with this procedure

$$P{X = 0} = P$$
 {first flip is a tail} = 1 - p

29. Each flip after the first will, independently, result in a changeover with probability 1/2. Therefore,

$$P\{k \text{ changeovers}\} = {\binom{n-1}{k}} (1/2)^{n-1}$$

30.
$$\frac{P\{X=i\}}{P\{X=i-1\}} = \frac{\lambda \lambda^{i} \lambda^{i} \lambda^{i}}{e^{-\lambda} \lambda^{i-1} / (i-1)!} = \lambda / i$$
Hence, $P\{X=i\}$ is increasing for $\lambda \geq i$ and decreasing for $\lambda < i$.

32. (a)
$$_{1}^{394}$$
 (b) $_{2}^{303}$ (c) $_{1}^{391}$ (d) $_{2}^{391}$ (e) $_{3}^{392}$ (f) $_{2}^{392}$ (g) $_{3}^{392}$ (

40. Let X denote the number of games played.

$$P\{X = 4\} = p^{4} + (1 - p)^{4}$$

$$P\{X = 5\} = P\{X = 5, \text{ I wins 3 of first 4}\}$$

$$+ P\{X = 5, \text{ II wins 3 of first 4}\}$$

$$= 4p^{3}(1 - p)p + 4(1 - p)^{3}p(1 - p)$$

$$P\{X = 6\} = P\{X = 6, \text{ I wins 3 of first 5}\}$$

$$+ P\{X = 6, \text{ II wins 3 of first 5}\}$$

$$= 10 p^{3}(1 - p)^{2} p + 10 p^{2}(1 - p)^{3}(1 - p)$$

$$P\{X = 7\} = P\{\text{first 6 games are split}\}$$

$$= 20 p^{3}(1 - p)^{3}$$

$$E[X] = \sum_{i=4}^{4} iP\{X = i\}$$

When p = 1/2, E[X] = 93/16 = 5.8125

41. Let X_i equal 1 if a changeover results from the i th flip and let it be 0 otherwise. Then

number of changeovers =
$$\sum_{i=2}^{\infty} X_i$$

As,

$$E[X_i] = P\{X_i = 1\} = P\{\text{flip } i - 1 = \text{flip } i\}$$

= $2p(1 - p)$

we see that

$$E [number of changeovers] = \sum_{i=2}^{n} E[X_i]$$
$$= 2(n-1)p(1-p)$$

42. Suppose the coupon collector has i different types. Let X_i denote the number of additional coupons collected until the collector has i + 1 types. It is easy to see that the X_i are independent geometric random variables with respective parameters (n - i)/n, $i = 0, 1, \ldots, n - 1$. Therefore,

$$\sum_{i=0}^{[n]} \frac{\sum_{i=0}^{n-1} \sum_{i=0}^{n} \sum_{i=0}^{n} \sum_{j=0}^{n/(n-i)} \frac{\sum_{i=0}^{n} n/(n-i)}{\sum_{i=1}^{n} 1/j}$$

43. (a)
$$X = \sum_{i=1}^{n} X_i$$

(b)
$$E[X_i] = P\{X_i = 1\}$$

= $P\{\text{red ball } i \text{ is chosen before all } n \text{ black balls}\}$

= 1/(n + 1) since each of these n + 1 balls is equally likely to be the one chosen earliest

Therefore,

$$E[X] = \sum_{i=1}^{m} E[X_i] = n/(n+1)$$

44. (a) Let Y_i equal 1 if red ball i is chosen after the first but before the second black ball, $i = 1, \ldots, n$. Then

$$Y = \sum_{i=1}^{n} Y_i$$

(b)
$$E[Y_i] = P\{Y_i = 1\}$$

= $P\{\text{red ball } i \text{ is the second chosen from a set of } n+1 \text{ balls}\} = 1/(n+1) \text{ since each of the } n+1 \text{ is equally likely to be}$
the second one chosen.

Therefore,

$$E[Y] = n/(n+1)$$

- (c) Answer is the same as in Problem 41.
- (d) We can let the outcome of this experiment be the vector (R_1, R_2, \ldots, R_n) where R_i is the number of red balls chosen after the (i-1)st but before the ith black ball. Since all orderings of the n+m balls are equally likely it follows that all different orderings of R_1, \ldots, R_n will have the same probability distribution. For instance,

$$P\{R_1 = a, R_2 = b\} = P\{R_2 = a, R_1 = b\}$$

From this it follows that all the R_i have the same distribution and thus the same mean.

45. Let N_i denote the number of keys in box $i, i = 1, \ldots, k$. Then, with X equal to the number of collisions we have that $X = \sum_{i=1}^{k_i} {N_i - 1} + \sum_{i=1}^{k_i} {N_i - 1} + \sum_{i=1}^{k_i} {N_i - 1} + i$ $\{N_i = 0\}$) where $I\{N_i = 0\}$ is equal to 1 if $N_i = 0$ and is equal to 0 otherwise.

Hence,

$$E[X] = \sum_{i=1}^{\infty} (r p_i - 1 + (1 - p_i)^r) = r - k$$

$$+ \sum_{i=1}^{\infty} (1 - p_i)^r$$

Another way to solve this problem is to let Y denote the number of boxes having at least one key, and then use the identity X = r - Y, which is true since only the first key put in each box does not result in a collision. Writing $Y = {}_{i=1} I^{\{N_i\}} > 0$ and taking expectations yields

$$E[X] = r - E[Y] = r - \sum_{i=1}^{\infty} [1 - (1 - p_i)^r]$$

$$= r - k + \sum_{i=1}^{\infty} (1 - p_i)^r$$

46. Using that $X = \sum_{n=1}^{\infty} I_n$, we obtain

$$E[X] = \sum_{n=1}^{\infty} E[I_n] = \sum_{n=1}^{\infty} P\{X \ge n\}$$

Making the change of variables m = n - 1 gives

$$E[X] = \sum_{m=0}^{\infty} P\{X \ge m + 1\} = \sum_{m=0}^{\infty} P\{X \ge m\}$$

(b) Let

$$I_n = \begin{cases} 1, & \text{if } n \leq X \\ 0, & \text{if } n > X \end{cases}$$
$$J_m = \begin{cases} 1, & \text{if } m \leq Y \\ 0, & \text{if } m > Y \end{cases}$$

Then

$$XY = \sum_{n=1}^{\infty} \tilde{I}_n^{\circ} \sum_{m=1}^{\infty} J_m = \sum_{n=1}^{\infty} \sum_{m=1}^{\infty} I_n J_m$$

Taking expectations now yields the result

$$E[XY] = \sum_{n=1}^{\infty} \sum_{m=1}^{\infty} E[I_n J_m]$$

$$= \sum_{n=1}^{\infty} \sum_{m=1}^{\infty} P(X \ge n, Y \ge m)$$

- 47. Let X_i be 1 if trial i is a success and 0 otherwise.
 - (a) The largest value is .6. If $X_1 = X_2 = X_3$, then $1.8 = E[X] = 3E[X_1] = 3P\{X_1 = 1\}$ and so

$$P{X = 3} = P{X1 = 1} = .6$$

That this is the largest value is seen by Markov's inequality, which yields

$$P\{X \ge 3\} \le E[X]/3 = .6$$

(b) The smallest value is 0. To construct a probability scenario for which $P\{X = 3\} = 0$ let U be a uniform random variable on (0, 1), and define

$$X_1 = \begin{cases} 1 & \text{if } U \le .6 \\ 0 & \text{otherwise} \end{cases}$$

$$X_2 = \begin{cases} 1 & \text{if } U \ge .4 \\ 0 & \text{otherwise} \end{cases}$$

$$X_3 = \begin{cases} 1 & \text{if either } U \le .3 \text{ or } U \ge .7 \\ 0 & \text{otherwise} \end{cases}$$

It is easy to see that

$$P{X_1 = X_2 = X_3 = 1} = 0$$

49. $E[X^2] - (E[X])^2 = V \ ar \ (X) = E \ (X - E \ [X])^2 \ge 0$. Equality when $V \ ar \ (X) = 0$, that is, when X is constant.

50.
$$Var(cX) = E[(cX - E[cX])^2]$$

 $= E[c^2(X - E(X))^2]$
 $= c^2Var(X)$
 $Var(c + X) = E[(c + X - E[c + X])^2]$
 $= E[(X - E[X])^2]$
 $= Var(X)$

51. $N_{e\overline{n}ce, X_i}^{\sum r} \frac{\sum_{i=1}^{r} X_i \text{ where } X_i \text{ is geometric with mean } 1/p. \text{ Thus,}$

$$E[N] = \sum_{i=1}^{r} E[X_i] = \frac{r}{p}$$

52. (a)
$$\frac{n}{n+1}$$

- (b) 0
- (c) 1

Instructor's Manual to Accompany
$$\begin{array}{c|c}
\hline
 & 1 \\
\hline
 & 1 \\
\hline
 & 1
\end{array}$$
53.
$$\frac{1}{n+1}, \frac{1}{2n+1} - \frac{1}{n+1}$$

- 54. (a) Using the fact that E[X + Y] = 0 we see that 0 = 2p(1, 1) 2p(-1, -1), which gives the result. (b) This follows since

$$0 = E[X - Y] = 2p(1, -1) - 2p(-1, 1)$$

- (c) $V ar(X) = E[X^2] = 1$ (d) $V ar(Y) = E[Y^2] = 1$
- (e) Since

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

= $2p(1,1) + 2p(1,-1)$

we see that if p = 2 p(1, 1) then 1 - p = 2 p(1, -1)Now,

$$Cov(X,Y) = E[XY]$$

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

$$= p - (1 - p) = 2p - 1$$

$$= j) = \sum_{i=0}^{\sum_{j}} (j) e^{-3\lambda} \lambda^{j} / j!$$

$$= e^{-2\lambda} \lambda \int_{j}^{j} (j) e^{-2\lambda} \lambda \int_{j}^{j-1} (j) e^{-2\lambda} \lambda \int_{j}^{j-1}$$

55. (a)
$$P(Y = j) = \sum_{i=0}^{j} (j) e^{-3\lambda} \lambda^{j} / j!$$

$$= e^{-2\lambda} \lambda^{j} \int_{i=0}^{j} 1i \int_{i=0}^{j-1} 1i \int_{i=0}^{j-1} e^{-2\lambda} \lambda^{j} / j!$$

$$= e^{-2\lambda} \left(\sum_{i=0}^{2\lambda} \frac{1}{j!} \right)^{j}$$
(b)
$$P(X = i) = i e^{-2\lambda} \lambda^{j} / j!$$

$$= \frac{1}{i!} e^{-2\lambda} \sum_{j=i}^{j} \frac{1}{(j-i)!} \lambda_{j}$$

$$= \lambda^{i} e^{-2\lambda} \sum_{k=0}^{k} \lambda^{k} / k!$$

$$= e^{-\lambda} \lambda^{i} \frac{1}{i!}$$

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