

**15. procedure** *Cantor permutation*( $n, i$ : integers with  $n \geq 1$  and  $0 \leq i < n!$ )  
 $x := n$   
**for**  $j := 1$  **to**  $n$   
 $p_j := 0$   
**for**  $k := 1$  **to**  $n - 1$   
**begin**  
 $c := \lfloor x/(n - k)! \rfloor$ ;  $x := x - c(n - k)!$ ;  $h := n$   
**while**  $p_h \neq 0$   
 $h := h - 1$   
**for**  $j := 1$  **to**  $c$   
**begin**  
 $h := h - 1$   
**while**  $p_h \neq 0$   
 $h := h - 1$   
**end**  
 $p_h := n - k + 1$   
**end**  
 $h := 1$   
**while**  $p_h \neq 0$   
 $h := h + 1$   
 $p_h := 1$   
 $\{p_1 p_2 \dots p_n$  is the permutation corresponding to  $i\}$

### Supplementary Exercises

1. a) 151,200 b) 1,000,000 c) 210 d) 5005 3.  $3^{100}$   
5. 24,600 7. a) 4,060 b) 2688 c) 25,009,600 9. a) 192  
b) 301 c) 300 d) 300 11. 639 13. The maximum possible sum is 240, and the minimum possible sum is 15. So the number of possible sums is 226. Because there are 252 subsets with five elements of a set with 10 elements, by the pigeonhole principle it follows that at least two have the same sum. 15. a) 50 b) 50 c) 14 d) 5 17. Let  $a_1, a_2, \dots, a_m$  be the integers, and let  $d_i = \sum_{j=1}^i a_j$ . If  $d_i \equiv 0 \pmod{m}$  for some  $i$ , we are done. Otherwise  $d_1 \bmod m, d_2 \bmod m, \dots, d_m \bmod m$  are  $m$  integers with values in  $\{1, 2, \dots, m - 1\}$ . By the pigeonhole principle  $d_k = d_l$  for some  $1 \leq k < l \leq m$ . Then  $\sum_{j=k+1}^l a_j = d_l - d_k \equiv 0 \pmod{m}$ . 19. The decimal expansion of the rational number  $a/b$  can be obtained by division of  $b$  into  $a$ , where  $a$  is written with a decimal point and an arbitrarily long string of 0s following it. The basic step is finding the next digit of the quotient, namely,  $\lfloor r/b \rfloor$ , where  $r$  is the remainder with the next digit of the dividend brought down. The current remainder is obtained from the previous remainder by subtracting  $b$  times the previous digit of the quotient. Eventually the dividend has nothing but 0s to bring down. Furthermore, there are only  $b$  possible remainders. Thus, at some point, by the pigeonhole principle, we will have the same situation as had previously arisen. From that point onward, the calculation must follow the same pattern. In particular, the quotient will repeat. 21. a) 125,970 b) 20 c) 141,120,525 d) 141,120,505 e) 177,100 f) 141,078,021 23. a) 10 b) 8 c) 7 25.  $3^n$   
27.  $C(n + 2, r + 1) = C(n + 1, r + 1) + C(n + 1, r) = 2C(n + 1, r + 1) - C(n + 1, r + 1) + C(n + 1, r) =$

$2C(n + 1, r + 1) - (C(n, r + 1) + C(n, r)) + (C(n, r) + C(n, r - 1)) = 2C(n + 1, r + 1) - C(n, r + 1) + C(n, r - 1)$  29. Substitute  $x = 1$  and  $y = 3$  into the Binomial Theorem. 31.  $C(n + 1, 5)$  33. 3,491,888,400  
35.  $5^{24}$  37. a) 45 b) 57 c) 12 39. a) 386 b) 56 c) 512  
41. 0 if  $n < m$ ;  $C(n - 1, n - m)$  if  $n \geq m$  43. a) 15,625  
b) 202 c) 210 d) 10

**45. procedure** *next permutation* ( $n$ : positive integer,  
 $a_1, a_2, \dots, a_r$ : positive integers not exceeding  
 $n$  with  $a_1 a_2 \dots a_r \neq nn \dots n$ )  
 $i := r$   
**while**  $a_i = n$   
**begin**  
 $a_i := 1$   
 $i := i - 1$   
**end**  
 $a_i := a_i + 1$   
 $\{a_1 a_2 \dots a_r$  is the next permutation in lexicographic order}

47. We must show that if there are  $R(m, n - 1) + R(m - 1, n)$  people at a party, then there must be at least  $m$  mutual friends or  $n$  mutual enemies. Consider one person; let's call him Jerry. Then there are  $R(m - 1, n) + R(m, n - 1) - 1$  other people at the party, and by the pigeonhole principle there must be at least  $R(m - 1, n)$  friends of Jerry or  $R(m, n - 1)$  enemies of Jerry among these people. First let's suppose there are  $R(m - 1, n)$  friends of Jerry. By the definition of  $R$ , among these people we are guaranteed to find either  $m - 1$  mutual friends or  $n$  mutual enemies. In the former case, these  $m - 1$  mutual friends together with Jerry are a set of  $m$  mutual friends; and in the latter case, we have the desired set of  $n$  mutual enemies. The other situation is similar: Suppose there are  $R(m, n - 1)$  enemies of Jerry; we are guaranteed to find among them either  $m$  mutual friends or  $n - 1$  mutual enemies. In the former case, we have the desired set of  $m$  mutual friends, and in the latter case, these  $n - 1$  mutual enemies together with Jerry are a set of  $n$  mutual enemies.

## CHAPTER 6

### Section 6.1

1.  $1/13$  3.  $1/2$  5.  $1/2$  7.  $1/64$  9.  $47/52$   
11.  $1/C(52, 5)$  13.  $1 - [C(48, 5)/C(52, 5)]$  15.  $C(13, 2)$   
 $C(4, 2)C(4, 2)C(44, 1)/C(52, 5)$  17.  $10,240/C(52, 5)$   
19.  $1,302,540/C(52, 5)$  21.  $1/64$  23.  $8/25$  25. a)  $1/C(50, 6) = 1/15,890,700$  b)  $1/C(52, 6) = 1/20,358,520$   
c)  $1/C(56, 6) = 1/32,468,436$  d)  $1/C(60, 6) = 1/50,063,860$  27. a) 139,128/319,865 b) 212,667/511,313  
c) 151,340/386,529 d) 163,647/446,276 29.  $1/C(100, 8)$   
31.  $3/100$  33. a)  $1/7,880,400$  b)  $1/8,000,000$  35. a)  $9/19$   
b)  $81/361$  c)  $1/19$  d)  $1,889,568/2,476,099$  e)  $48/361$   
37. Three dice 39. The door the contestant chooses is chosen at random without knowing where the prize is, but the door chosen by the host is not chosen at random, because he always avoids opening the door with the prize. This makes

any argument based on symmetry invalid. 41. a) 671/1296  
b)  $1 - 35^{24}/36^{24}$ ; no c) Yes

## Section 6.2

1.  $p(T) = 1/4$ ,  $p(H) = 3/4$  3.  $p(1) = p(3) = p(5) = p(6) = 1/16$ ;  $p(2) = p(4) = 3/8$  5. 9/49 7. a) 1/2  
b) 1/2 c) 1/3 d) 1/4 e) 1/4 9. a) 1/26! b) 1/26  
c) 1/2 d) 1/26 e) 1/650 f) 1/15,600 11. Clearly,  
 $p(E \cup F) \geq p(E) = 0.7$ . Also,  $p(E \cup F) \leq 1$ . If we apply  
Theorem 2 from Section 6.1, we can rewrite this as  $p(E) +$   
 $p(F) - p(E \cap F) \leq 1$ , or  $0.7 + 0.5 - p(E \cap F) \leq 1$ . Solv-  
ing for  $p(E \cap F)$  gives  $p(E \cap F) \geq 0.2$ . 13. Because  
 $p(E \cup F) = p(E) + p(F) - p(E \cap F)$  and  $p(E \cup F) \leq 1$ ,  
it follows that  $1 \geq p(E) + p(F) - p(E \cap F)$ . From this in-  
equality we conclude that  $p(E) + p(F) \leq 1 + p(E \cap F)$ .  
15. We will use mathematical induction to prove that the  
inequality holds for  $n \geq 2$ . Let  $P(n)$  be the statement that  
 $p(\bigcup_{j=1}^n E_j) \leq \sum_{j=1}^n p(E_j)$ . *Basis step:*  $P(2)$  is true because  
 $p(E_1 \cup E_2) = p(E_1) + p(E_2) - p(E_1 \cap E_2) \leq p(E_1) +$   
 $p(E_2)$ . *Inductive step:* Assume that  $P(k)$  is true. Using  
the basis case and the inductive hypothesis, it follows that  
 $p(\bigcup_{j=1}^{k+1} E_j) \leq p(\bigcup_{j=1}^k E_j) + p(E_{k+1}) \leq \sum_{j=1}^{k+1} p(E_j)$ . This  
shows that  $P(k+1)$  is true, completing the proof by mathe-  
matical induction. 17. Because  $E \cup \bar{E}$  is the entire  
sample space  $S$ , the event  $F$  can be split into two disjoint  
events:  $F = S \cap F = (E \cup \bar{E}) \cap F = (E \cap F) \cup (\bar{E} \cap F)$ ,  
using the distributive law. Therefore,  $p(F) = p((E \cap F) \cup$   
 $(\bar{E} \cap F)) = p(E \cap F) + p(\bar{E} \cap F)$ , because these two events  
are disjoint. Subtracting  $p(E \cap F)$  from both sides, using the  
fact that  $p(\bar{E} \cap F) = p(\bar{E}) \cdot p(F)$  (the hypothesis that  $E$   
and  $F$  are independent), and factoring, we have  $p(F)[1 -$   
 $p(E)] = p(\bar{E} \cap F)$ . Because  $1 - p(E) = p(\bar{E})$ , this says  
that  $p(\bar{E} \cap F) = p(\bar{E}) \cdot p(F)$ , as desired. 19. a) 1/12  
b)  $1 - \frac{11}{12} \cdot \frac{10}{12} \cdot \dots \cdot \frac{13-n}{12}$  c) 5 21. 614 23. 1/4 25. 3/8  
27. a) Not independent b) Not independent c) Not inde-  
pendent 29. 3/16 31. a) 1/32 = 0.03125 b)  $0.49^5 \approx$   
 $0.02825$  c) 0.03795012 33. a) 5/8 b) 0.627649  
c) 0.6431 35. a)  $p^n$  b)  $1 - p^n$  c)  $p^n + n \cdot p^{n-1} \cdot (1 - p)$   
d)  $1 - [p^n + n \cdot p^{n-1} \cdot (1 - p)]$  37.  $p(\bigcup_{i=1}^{\infty} E_i)$  is the sum  
of  $p(s)$  for each outcome  $s$  in  $\bigcup_{i=1}^{\infty} E_i$ . Because the  $E_i$ s  
are pairwise disjoint, this is the sum of the probabili-  
ties of all the outcomes in any of the  $E_i$ s, which is what  
 $\sum_{i=1}^{\infty} p(E_i)$  is. (We can rearrange the summands and still get  
the same answer because this series converges absolutely.)  
39. a)  $\bar{E} = \bigcup_{j=1}^n \bar{F}_j$ , so the given inequality now follows  
from Boole's Inequality (Exercise 15). b) The probability  
that a particular player not in the  $j$ th set beats all  $k$  of the play-  
ers in the  $j$ th set is  $(1/2)^k = 2^{-k}$ . Therefore, the probability  
that this player does not do so is  $1 - 2^{-k}$ , so the probability  
that all  $m - k$  of the players not in the  $j$ th set are unable to  
boast of a perfect record against everyone in the  $j$ th set is  
 $(1 - 2^{-k})^{m-k}$ . That is precisely  $p(F_j)$ . c) The first ineq-  
uality follows immediately, because all the summands are the  
same and there are  $\binom{m}{k}$  of them. If this probability is less than  
1, then it must be possible that  $\bar{E}$  fails, i.e., that  $E$  happens.  
So there is a tournament that meets the conditions of the

problem as long as the second inequality holds. d)  $m \geq 21$   
for  $k = 2$ , and  $m \geq 91$  for  $k = 3$

41. **procedure** probabilistic prime ( $n, k$ )

```

composite := false
i := 0
while composite = false and i < k
begin
    i := i + 1
    choose b uniformly at random with  $1 < b < n$ 
    apply Miller's test to base b
    if n fails the test then composite := true
end
if composite = true then print ("composite")
else print ("probably prime")

```

## Section 6.3

NOTE: In the answers for Section 6.3, all probabilities given  
in decimal form are rounded to three decimal places. 1. 3/5  
3. 3/4 5. 0.481 7. a) 0.999 b) 0.324 9. a) 0.740  
b) 0.260 c) 0.002 d) 0.998 11. 0.724 13. 3/17  
15. a) 1/3 b)  $p(M = j | W = k) = 1$  if  $i, j$ , and  $k$  are dis-  
tinct;  $p(M = j | W = k) = 0$  if  $j = k$  or  $j = i$ ;  $p(M = j |$   
 $W = k) = 1/2$  if  $i = k$  and  $j \neq i$  c) 2/3 d) You should  
change doors, because you now have a 2/3 chance to win  
by switching. 17. The definition of conditional probability  
tells us that  $p(F_j | E) = p(E \cap F_j)/p(E)$ . For the numerator,  
again using the definition of conditional probability, we have  
 $p(E \cap F_j) = p(E | F_j)p(F_j)$ , as desired. For the denomina-  
tor, we show that  $p(E) = \sum_{i=1}^n p(E | F_i)p(F_i)$ . The events  
 $E \cap F_i$  partition the event  $E$ ; that is,  $(E \cap F_{i_1}) \cap (E \cap F_{i_2}) =$   
 $\emptyset$  when  $i_1 \neq i_2$  (because the  $F_i$ 's are mutually exclusive), and  
 $\bigcup_{i=1}^n (E \cap F_i) = E$  (because the  $\bigcup_{i=1}^n F_i = S$ ). Therefore,  
 $p(E) = \sum_{i=1}^n p(E \cap F_i) = \sum_{i=1}^n p(E | F_i)p(F_i)$ . 19. No  
21. Yes 23. By Bayes' Theorem,  $p(S | E_1 \cap E_2) = p(E_1 \cap$   
 $E_2 | S)p(S)/[p(E_1 \cap E_2 | S)p(S) + p(E_1 \cap E_2 | \bar{S})p(\bar{S})]$ .  
Because we are assuming no prior knowledge about whether  
a message is or is not spam, we set  $p(S) = p(\bar{S}) = 0.5$ , and  
so the equation above simplifies to  $p(S | E_1 \cap E_2) = p(E_1 \cap$   
 $E_2 | S)/[p(E_1 \cap E_2 | S) + p(E_1 \cap E_2 | \bar{S})]$ . Because of  
the assumed independence of  $E_1, E_2$ , and  $S$ , we have  $p(E_1 \cap$   
 $E_2 | S) = p(E_1 | S) \cdot p(E_2 | S)$ , and similarly for  $\bar{S}$ .

## Section 6.4

1. 2.5 3. 5/3 5. 336/49 7. 170 9.  $(4n + 6)/3$   
11. 50,700,551/10,077,696  $\approx 5.03$  13. 6 15.  $p(X \geq$   
 $j) = \sum_{k=j}^{\infty} p(X = k) = \sum_{k=j}^{\infty} (1 - p)^{k-1} p = p(1 - p)^{j-1}$   
 $\sum_{k=0}^{\infty} (1 - p)^k = p(1 - p)^{j-1}/(1 - (1 - p)) = (1 - p)^{j-1}$   
17. 2302 19.  $(7/2) \cdot 7 \neq 329/12$  21.  $p + (n - 1)p(1 - p)$   
23. 5/2 25. a) 0 b)  $n$  27. a) We are told that  $X_1$  and  
 $X_2$  are independent. To see that  $X_1$  and  $X_3$  are independent,  
we enumerate the eight possibilities for  $(X_1, X_2, X_3)$  and find  
that  $(0, 0, 0), (1, 0, 1), (0, 1, 1), (1, 1, 0)$  each have probability  
1/4 and the others have probability 0 (because of the definition  
of  $X_3$ ). Thus,  $p(X_1 = 0 \wedge X_3 = 0) = 1/4, p(X_1 = 0) = 1/2,$

and  $p(X_3 = 0) = 1/2$ , so it is true that  $p(X_1 = 0 \wedge X_3 = 0) = p(X_1 = 0)p(X_3 = 0)$ . Essentially the same calculation shows that  $p(X_1 = 0 \wedge X_3 = 1) = p(X_1 = 0)p(X_3 = 1)$ ,  $p(X_1 = 1 \wedge X_3 = 0) = p(X_1 = 1)p(X_3 = 0)$ , and  $p(X_1 = 1 \wedge X_3 = 1) = p(X_1 = 1)p(X_3 = 1)$ . Therefore by definition,  $X_1$  and  $X_3$  are independent. The same reasoning shows that  $X_2$  and  $X_3$  are independent. To see that  $X_3$  and  $X_1 + X_2$  are not independent, we observe that  $p(X_3 = 1 \wedge X_1 + X_2 = 2) = 0$ . But  $p(X_3 = 1)p(X_1 + X_2 = 2) = (1/2)(1/4) = 1/8$ . **b)** We see from the calculation in part (a) that  $X_1$ ,  $X_2$ , and  $X_3$  are all Bernoulli random variables, so the variance of each is  $(1/2)(1/2) = 1/4$ . Therefore,  $V(X_1) + V(X_2) + V(X_3) = 3/4$ . We use the calculations in part (a) to see that  $E(X_1 + X_2 + X_3) = 3/2$ , and then  $V(X_1 + X_2 + X_3) = 3/4$ . **c)** In order to use the first part of Theorem 7 to show that  $V((X_1 + X_2 + \cdots + X_k) + X_{k+1}) = V(X_1 + X_2 + \cdots + X_k) + V(X_{k+1})$  in the inductive step of a proof by mathematical induction, we would have to know that  $X_1 + X_2 + \cdots + X_k$  and  $X_{k+1}$  are independent, but we see from part (a) that this is not necessarily true. **29.**  $1/100$  **31.**  $E(X)/a = \sum_r (r/a) \cdot p(X = r) \geq \sum_{r \geq a} 1 \cdot p(X = r) = p(X \geq a)$  **33. a)**  $10/11$  **b)**  $0.9984$  **35. a)** Each of the  $n!$  permutations occurs with probability  $1/n!$ , so  $E(X)$  is the number of comparisons, averaged over all these permutations. **b)** Even if the algorithm continues  $n - 1$  rounds,  $X$  will be at most  $n(n - 1)/2$ . It follows from the formula for expectation that  $E(X) \leq n(n - 1)/2$ . **c)** The algorithm proceeds by comparing adjacent elements and then swapping them if necessary. Thus, the only way that inverted elements can become uninverted is for them to be compared and swapped. **d)** Because  $X(P) \geq I(P)$  for all  $P$ , it follows from the definition of expectation that  $E(X) \geq E(I)$ . **e)** This summation counts 1 for every instance of an inversion. **f)** This follows from Theorem 3. **g)** By Theorem 2 with  $n = 1$ , the expectation of  $I_{j,k}$  is the probability that  $a_k$  precedes  $a_j$  in the permutation. This is clearly  $1/2$  by symmetry. **h)** The summation in part (f) consists of  $C(n, 2) = n(n - 1)/2$  terms, each equal to  $1/2$ , so the sum is  $n(n - 1)/4$ . **i)** From part (a) and part (b) we know that  $E(X)$ , the object of interest, is at most  $n(n - 1)/2$ , and from part (d) and part (h) we know that  $E(X)$  is at least  $n(n - 1)/4$ , both of which are  $\Theta(n^2)$ . **37. 1** **39.**  $V(X + Y) = E((X + Y)^2) - E(X + Y)^2 = E(X^2 + 2XY + Y^2) - [E(X) + E(Y)]^2 = E(X^2) + 2E(XY) + E(Y^2) - E(X)^2 - 2E(X)E(Y) - E(Y)^2 = E(X^2) - E(X)^2 + 2[E(XY) - E(X)E(Y)] + E(Y^2) - E(Y)^2 = V(X) + 2\text{Cov}(X, Y) + V(Y)$  **41.**  $[(n - 1)/n]^m$  **43.**  $(n - 1)^m/n^{m-1}$

## Supplementary Exercises

1.  $1/109,668$  **3. a)**  $1/C(52, 13)$  **b)**  $4/C(52, 13)$   
**c)**  $2,944,656/C(52, 13)$  **d)**  $35,335,872/C(52, 13)$   
**5. a)**  $9/2$  **b)**  $21/4$  **7. a)**  $9$  **b)**  $21/2$  **9. a)**  $8$  **b)**  $49/6$   
**11. a)**  $n/2^{n-1}$  **b)**  $p(1 - p)^{k-1}$ , where  $p = n/2^{n-1}$   
**c)**  $2^{n-1}/n$  **13.**  $\frac{(m-1)(n-1) + \gcd(m, n) - 1}{mn-1}$  **15. a)**  $2/3$  **b)**  $2/3$   
**17.**  $1/32$  **19. a)** The probability that one wins  $2^n$  dollars is

$1/2^n$ , because that happens precisely when the player gets  $n - 1$  tails followed by a head. The expected value of the winnings is therefore the sum of  $2^n$  times  $1/2^n$  as  $n$  goes from 1 to infinity. Because each of these terms is 1, the sum is infinite. In other words, one should be willing to wager any amount of money and expect to come out ahead in the long run. **b)** \$9, \$9 **21. a)**  $1/3$  when  $S = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12\}$ ,  $A = \{1, 2, 3, 4, 5, 6, 7, 8, 9\}$ , and  $B = \{1, 2, 3, 4\}$ ;  $1/12$  when  $S = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12\}$ ,  $A = \{4, 5, 6, 7, 8, 9, 10, 11, 12\}$ , and  $B = \{1, 2, 3, 4\}$  **b)** 1 when  $S = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12\}$ ,  $A = \{4, 5, 6, 7, 8, 9, 10, 11, 12\}$ , and  $B = \{1, 2, 3, 4\}$ ;  $3/4$  when  $S = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12\}$ ,  $A = \{1, 2, 3, 4, 5, 6, 7, 8, 9\}$ , and  $B = \{1, 2, 3, 4\}$  **23. a)**  $p(E_1 \cap E_2) = p(E_1)p(E_2)$ ,  $p(E_1 \cap E_3) = p(E_1)p(E_3)$ ,  $p(E_2 \cap E_3) = p(E_2)p(E_3)$ ,  $p(E_1 \cap E_2 \cap E_3) = p(E_1)p(E_2)p(E_3)$  **b)** Yes **c)** Yes; yes **d)** Yes; no **e)**  $2^n - n - 1$  **25.**  $V(aX + b) = E((aX + b)^2) - E(aX + b)^2 = E(a^2X^2 + 2abX + b^2) - [aE(X) + b]^2 = E(a^2X^2) + E(2abX) + E(b^2) - [a^2E(X)^2 + 2abE(X) + b^2] = a^2E(X^2) + 2abE(X) + b^2 - a^2E(X)^2 - 2abE(X) - b^2 = a^2[E(X^2) - E(X)^2] = a^2V(X)$  **27.** To count every element in the sample space exactly once, we must include every element in each of the sets and then take away the double counting of the elements in the intersections. Thus  $p(E_1 \cup E_2 \cup \cdots \cup E_m) = p(E_1) + p(E_2) + \cdots + p(E_m) - p(E_1 \cap E_2) - p(E_1 \cap E_3) - \cdots - p(E_1 \cap E_m) - p(E_2 \cap E_3) - p(E_2 \cap E_4) - \cdots - p(E_2 \cap E_m) - \cdots - p(E_{m-1} \cap E_m) = qm - (m(m - 1)/2)r$ , because  $C(m, 2)$  terms are being subtracted. But  $p(E_1 \cup E_2 \cup \cdots \cup E_m) = 1$ , so we have  $qm - [m(m - 1)/2]r = 1$ . Because  $r \geq 0$ , this equation tells us that  $qm \geq 1$ , so  $q \geq 1/m$ . Because  $q \leq 1$ , this equation also implies that  $[m(m - 1)/2]r = qm - 1 \leq m - 1$ , from which it follows that  $r \leq 2/m$ . **29. a)** We purchase the cards until we have gotten one of each type. That means we have purchased  $X$  cards in all. On the other hand, that also means that we purchased  $X_0$  cards until we got the first type we got, and then purchased  $X_1$  more cards until we got the second type we got, and so on. Thus,  $X$  is the sum of the  $X_j$ 's. **b)** Once  $j$  distinct types have been obtained, there are  $n - j$  new types available out of a total of  $n$  types available. Because it is equally likely that we get each type, the probability of success on the next purchase (getting a new type) is  $(n - j)/n$ . **c)** This follows immediately from the definition of geometric distribution, the definition of  $X_j$ , and part (b). **d)** From part (c) it follows that  $E(X_j) = n/(n - j)$ . Thus by the linearity of expectation and part (a), we have  $E(X) = E(X_0) + E(X_1) + \cdots + E(X_{n-1}) = \frac{n}{n} + \frac{n}{n-1} + \cdots + \frac{n}{1} = n \left( \frac{1}{n} + \frac{1}{n-1} + \cdots + \frac{1}{1} \right)$ . **e)** About 224.46 **31.**  $24 \cdot 13^4 / (52 \cdot 51 \cdot 50 \cdot 49)$

## CHAPTER 7

### Section 7.1

- 1. a)** 2, 12, 72, 432, 2592 **b)** 2, 4, 16, 256, 65,536  
**c)** 1, 2, 5, 11, 26 **d)** 1, 1, 6, 27, 204 **e)** 1, 2, 0, 1, 3