

11 (a) from $\{1, 2, \dots, (10^3)\}$, the largest number can be divisible by 3 is 999, by 5 is 10^3 , by 7 is 994, by 15 is 990, by 105 is 945, the smallest number can be divisible by 3 is 3, by 5 is 5 by 7 is 7, by 15 is 15, by 105 is 105, the number of integer N will be divisible by 3 is

$$\frac{999+3}{3} - 1 = 333, \text{ by 5 is } \frac{10^3+5}{5} - 1 = 200, \text{ by 7 is } \frac{994+7}{7} - 1 = 142, \text{ by 15 is } \frac{945+15}{15} - 1 = 66, \text{ by 105 is } \frac{945+105}{105} - 1 = 9. \therefore P(\text{divisible by 3}) = \frac{333}{1000}, P(\text{divisible by 7}) = \frac{142}{1000}, P(\text{divisible by 15}) = \frac{66}{1000}$$

$$P(\text{divisible by 5}) = \frac{200}{1000}, P(\text{divisible by 105}) = \frac{9}{1000}$$

if 10^3 is replaced by 10^k , then the largest number can be divisible by 3 is between $10^k - 3 \sim 10^k$ by 5 is between $10^k - 5 \sim 10^k$, by 7 is between $10^k - 7 \sim 10^k$, by 15 is between $10^k - 15 \sim 10^k$ by 105 is between $10^k - 105 \sim 10^k$. then the number of integer N will be divisible by 3 is

$$\text{between } \frac{10^k - 3 + 3}{3} - 1 = \frac{10^k - 3}{3} \text{ and } \frac{10^k}{3}, \text{ by 5 is between } \frac{10^k - 5}{5} \text{ and } \frac{10^k}{5}, \text{ by 7 is between } \frac{10^k - 7}{7} \text{ and } \frac{10^k}{7}, \text{ by 15 is between } \frac{10^k - 15}{15} \text{ and } \frac{10^k}{15}, \text{ by 105 is between } \frac{10^k - 105}{105} \text{ and } \frac{10^k}{105}$$

$$\text{so the limiting distribution: } P(\text{divisible by 3}) \geq \frac{10^k - 3}{3 \cdot 10^k} \rightarrow \frac{1}{3} \text{ and } P(\text{divisible by 5}) \leq \frac{10^k}{3 \cdot 10^k} = \frac{1}{3}$$

$$\therefore P(\text{divisible by 3}) = \frac{1}{3}, \text{ by the same method, we get } P(\text{divisible by 5}) = \frac{1}{5}, P(\text{divisible by 7}) = \frac{1}{7}, P(\text{divisible by 15}) = \frac{1}{15}, P(\text{divisible by 105}) = \frac{1}{105}$$

13. X could have the possible value of 0, 500, 1000, 1500, 2000.

$$P(X=0) = P(\text{no sale on first and no sale on second}) = (.7)(.4) = .28$$

$$P(X=500) = P(\text{1 sale and it is for standard}) = \frac{1}{2} \cdot P(\text{1 sale}) = \frac{1}{2} [P(\{\text{sale, no sale}\}) + P(\{\text{no sale, sale}\})] = \frac{1}{2} [(0.3)(0.4) + (0.7)(0.6)] = .27$$

$$P(X=1000) = P(\text{2 standard sales}) + P(\text{1 sale for deluxe}) = \frac{1}{4} P(\text{2 sales}) + \frac{1}{2} P(\text{1 sale})$$

$$= \frac{1}{4} \cdot (0.6)(0.3) + .27 = .315$$

$$P(X=1500) = P(\text{2 sales, one deluxe and one standard}) = \frac{1}{2} (0.3)(0.6) = .09$$

$$P(X=2000) = P(\text{2 sales, both deluxe}) = (0.3)(0.6)/4 = .045$$

20. X denote the gambler's winnings.

$$X=1 \quad \text{first bet win and (lose, win, win)} \quad P = \frac{18}{38} + \frac{20}{38} \left(\frac{18}{38}\right)^2$$

$$X=-1 \quad (\text{lose, win, lose}), (\text{lose, lose, win}) \quad P = \frac{20}{38} \left(\frac{18}{38}\right) \left(\frac{20}{38}\right) \times 2$$

$$X=-3 \quad (\text{lose, lose, lose}) \quad P = \left(\frac{20}{38}\right)^3$$

$$(a) P(X>0) = P(X=1) = \frac{18}{38} + \left(\frac{20}{38}\right)\left(\frac{18}{38}\right)^2 = .5918$$

(b) No, because if the gambler wins then he or she wins \$1, However, a lose would either be \$1 or \$3

$$(c) E(X) = 1 \left[\frac{18}{38} + \left(\frac{20}{38}\right)\left(\frac{18}{38}\right)^2 \right] - 1 \left[\left(\frac{20}{38}\right)^2 \left(\frac{18}{38}\right) \right] - 3 \left[\left(\frac{20}{38}\right)^3 \right] = -.108$$

22. (a) Let N denote the number of games played, for $i=2$.

$N=2$. if team A win all the two games probability $= p^2 + (1-p)^2$
or team B win all the two games.

$N=3$. if team A win the last game and probability $= 2p(1-p) \cdot p + 2(1-p)p(1-p)$
team B win the first or the second game.
= $2(1-p)p$.
or the same for team B.

$$\therefore E(N) = 2[p^2 + (1-p)^2] + 3[2p(1-p)] = 2 + 2p(1-p).$$

To get the maximum value, take the derivative of $E(N)$ $\frac{dE(N)}{dp} = 2 - 4p = 0 \therefore p = \frac{1}{2}$.

(b) for $i=3$

$N=3$. if team A win all the 3 games or probability $= p^3 + (1-p)^3$
the same for team B

$N=4$. if team A win the last game and probability $= 3p^3(1-p) + 3p(1-p)^3$
team B win one of the first 3 games
or the same for team B

$N=5$. if team A win the last game and probability $= 6p^3(1-p)^2 + 6p^2(1-p)^3$
team B win two of the first 4 games
or the same for team B

$$E(N) = 3[p^3 + (1-p)^3] + 4[3p^3(1-p) + 3p(1-p)^3] + 5[6p^3(1-p)^2 + 6p^2(1-p)^3]$$

$\frac{dE(N)}{dp} = 24p^3 - 36p^2 + 6p + 3$. its value at $p = \frac{1}{2}$ is easily seen to be 0.

$$30 \text{ First } E(X) = \sum_{n=1}^{\infty} 2^n \cdot \left(\frac{1}{2}\right)^n = \sum_{n=1}^{\infty} 1 = \infty$$

(a). I would probably not to pay \$1 million to play this game once. because if we want to win more than \$1 million, then $2^n > 10^6 \quad n \geq \frac{6 \log 10}{\log 2}$ the probability to win is $\frac{1}{2^n} \leq \frac{1}{10^6}$ is too small.

(b) If you could play an arbitrary large number of games., you keep on playing until you win a large number of dollars one time (2^n , $n \rightarrow \infty$) then, it is surely that you will win.

32. If T is the number of tests needed for a group of 10 people, then

$T=1$ if the test is negative, the probability is $(.9)^{10}$

$T=11$ if the test is positive, the probability is $(1-(.9)^{10})$

$$\therefore E(T) = (.9)^{10} + 11[1 - (.9)^{10}] = 11 - 10(.9)^{10}$$

54. X denote the number of cars abandoned weekly, X from poisson distribution with parameter 2.2.

$$\therefore P(X=x) = \frac{e^{-2.2}(2.2)^x}{x!}$$

$$(a) P(\text{no abandoned cars}) = P(X=0) = e^{-2.2}$$

$$(b) P(\text{at least two abandoned cars}) = P(X \geq 2) = 1 - P(X=0) - P(X=1) = 1 - 3.2e^{-2.2}$$

55. X_1 denote the number of errors when typed by the first typist, X_2 denote the number of error by the second typist, $X_1 \sim \text{poisson}(3)$, $X_2 \sim \text{poisson}(4.2)$

$$P(\text{no errors}) = P(\text{no errors} | \text{First typist}) P(\text{First typist}) + P(\text{no errors} | \text{Second typist}) P(\text{Second typist})$$
$$= P(X_1=0) \cdot \frac{1}{2} + P(X_2=0) \cdot \frac{1}{2} = \frac{1}{2} (e^{-3} + e^{-4.2})$$

76. Let E denote the event that the mathematician first discovers that the right-hand matchbox is empty and there are k matches in the left-hand box at the time. Now, the event will occur if and only if the (N_2+1) th choice of the right-hand matchbox is made at the (N_1+N_2+1-k) trial. Hence it's from negative binomial distribution with $p=\frac{1}{2}$ $r=N_2+1$, $n=N_1+N_2+1-k$

$$\therefore P(E) = \binom{N_1+N_2-k}{N_2} \left(\frac{1}{2}\right)^{N_1+N_2-k+1} \quad \text{the same for the event that first discovers the left-hand}$$

$$\text{matchbox is empty} \quad P(E') = \binom{N_1+N_2-k}{N_1} \left(\frac{1}{2}\right)^{N_1+N_2-k+1}.$$

$$\therefore P = \binom{N_1+N_2-k}{N_2} \left(\frac{1}{2}\right)^{N_1+N_2-k+1} + \binom{N_1+N_2-k}{N_1} \left(\frac{1}{2}\right)^{N_1+N_2-k+1}$$

77. Let E denote the event that ~~the mathematician~~ at the moment the right-hand matchbox is

empty, there are k matches in the left-hand box. Now the event will occur if and only if the $\frac{N}{2}$ th choice of the right-hand matchbox is made at $(\frac{2N}{2}-k)$ trial. It is also from negative binomial distribution with $p=\frac{1}{2}$ $r=\frac{N}{2}$, $n=\frac{2N}{2}-k$ $\therefore P(E) = \binom{\frac{2N}{2}-k-1}{\frac{N}{2}-1} \left(\frac{1}{2}\right)^{\frac{2N}{2}-k}$

$$\therefore P = 2 \binom{\frac{2N}{2}-k-1}{\frac{N}{2}-1} \left(\frac{1}{2}\right)^{\frac{2N}{2}-k}$$

6. the right-hand side $\sum_{i=1}^{\infty} P\{N \geq i\} = \sum_{i=1}^{\infty} \sum_{k=i}^{\infty} P\{N=k\} = \sum_{k=1}^{\infty} \sum_{i=1}^k P\{N=k\} = \sum_{k=1}^{\infty} k P\{N=k\} = E(N)$

$$\begin{aligned}
7. \text{ the left-hand side } \sum_{i=0}^{\infty} i P\{N > i\} &= \sum_{i=1}^{\infty} i \sum_{k=i+1}^{\infty} P\{N=k\} = \sum_{k=1}^{\infty} \sum_{i=0}^{k-1} i P\{N=k\} \\
&= \sum_{k=1}^{\infty} P\{N=k\} \sum_{i=0}^{k-1} i = \sum_{k=1}^{\infty} P\{N=k\} \cdot \frac{k(k-1)}{2} = \left[\sum_{k=1}^{\infty} k^2 P\{N=k\} - \sum_{k=1}^{\infty} k P\{N=k\} \right] \frac{1}{2} \\
&= \frac{1}{2} (E[N^2] - E[N])
\end{aligned}$$

15. First prove $\sum_{i=0}^{\lfloor n/2 \rfloor} \binom{n}{2i} p^{2i} q^{n-2i} = \frac{1}{2} [(p+q)^n + (q-p)^n]$

$$\begin{aligned}
\text{the right-hand side } \frac{1}{2} [(p+q)^n + (q-p)^n] &= \frac{1}{2} \left[\sum_{k=0}^n \binom{n}{k} p^k q^{n-k} + \sum_{k=0}^n \binom{n}{k} (-p)^k q^{n-k} \right] \\
&= \frac{1}{2} \left[\sum_{k=0}^{\lfloor n/2 \rfloor} \binom{n}{2k} p^{2k} q^{n-2k} + \sum_{k=0}^{\lfloor n/2 \rfloor} \binom{n}{2k+1} p^{2k+1} q^{n-2k-1} + \sum_{k=0}^{\lfloor n/2 \rfloor} \binom{n}{2k} (-p)^{2k} q^{n-2k} + \sum_{k=0}^{\lfloor n/2 \rfloor} \binom{n}{2k+1} (-p)^{2k+1} q^{n-2k-1} \right] \\
&= \frac{1}{2} \left[\sum_{k=0}^{\lfloor n/2 \rfloor} \binom{n}{2k} p^{2k} q^{n-2k} + \sum_{k=0}^{\lfloor n/2 \rfloor} \binom{n}{2k} (-p)^{2k} q^{n-2k} \right] = \sum_{k=0}^{\lfloor n/2 \rfloor} \binom{n}{2k} p^{2k} q^{n-2k}
\end{aligned}$$

Then let X denote the number of heads when toss the coin n times, then $X \sim \text{Binomial}(n, p)$

$$P(X = \text{even number}) = \sum_{k=0}^{\lfloor n/2 \rfloor} \binom{n}{2k} p^{2k} (1-p)^{n-2k} = \frac{1}{2} [(p+1-p)^n + (1-p-p)^n] = \frac{1}{2} [1 + (q-p)^n]$$

20. Let S denote the number of heads that occur when all n coins are tossed, and note that S has a distribution that is approximately that of a poisson random variable with mean λ . Then, because X is distributed as the conditional distribution of S given that $S > 0$

$$P(X=1) = P(S=1 | S > 0) = \frac{P(S=1)}{P(S > 0)} \approx \frac{\lambda e^{-\lambda}}{1 - e^{-\lambda}}$$

25. the number of events that occur in a specific time is a Poisson random variable with parameter λ
and the number of events is counted is a Binomial random variable $B(n, p)$.

$$\text{so } P(m \text{ counted}) = \sum_{n=m}^{\infty} P(m \mid n \text{ events}) P(n \text{ events})$$

$$\begin{aligned}
&= \sum_{n=m}^{\infty} \binom{n}{m} p^m (1-p)^{n-m} e^{-\lambda} \cdot \frac{\lambda^n}{n!} \\
&= e^{-\lambda p} \frac{(\lambda p)^m}{m!} \sum_{n=m}^{\infty} \frac{[\lambda(1-p)]^{n-m}}{(n-m)!} e^{-\lambda(1-p)}
\end{aligned}$$

$$= e^{-\lambda p} \frac{(\lambda p)^m}{m!} \sum_{n=0}^{\infty} \frac{[\lambda(1-p)]^n}{n!} e^{-\lambda(1-p)}$$

$$= e^{-\lambda p} \frac{(\lambda p)^m}{m!}$$

so the number of events that are counted is a Poisson random variable with parameter λp
Intuitively, the Poisson λ random variable arises as the approximate number of successes in n independent trials each having a small success probability $\frac{\lambda}{n}$. Now if each successful trial is counted with probability p , the number counted is Binomial with parameter $\frac{\lambda}{n}p$

which is approximately Poisson with parameter $\frac{1}{n} p \cdot n = \lambda p$.

if $\lambda=10$, $p=\frac{1}{50}$, $X \sim \text{Poisson}(\lambda p) = \text{Poisson}(\frac{1}{5})$

$$(a) P(X=1) = (e^{-\frac{1}{5}}) \frac{1}{5} \quad (b) P(X \geq 1) = 1 - P(X=0) = 1 - e^{-\frac{1}{5}}$$

$$(c) P(X \leq 1) = P(X=0) + P(X=1) = e^{-\frac{1}{5}} + e^{-\frac{1}{5}} \cdot \frac{1}{5} = \frac{6}{5} e^{-\frac{1}{5}}$$

2b. if $n=0$, the right-hand side $\frac{1}{0!} \int_0^\infty e^{-x} dx = [-e^{-x}] \Big|_0^\infty = e^{-\lambda} = \text{left-hand side.}$

$$\text{Assume } n=k, \text{ the equation } \sum_{i=0}^k e^{-\lambda} \frac{\lambda^i}{i!} = \frac{1}{k!} \int_0^\infty e^{-x} x^k dx$$

$$\text{Then for } n=k+1, \frac{1}{(k+1)!} \int_0^\infty e^{-x} \cdot x^{k+1} dx = \frac{1}{(k+1)!} \left[[-e^{-x} \cdot x^{k+1}] \Big|_0^\infty + \int_0^\infty e^{-x} (k+1) x^k dx \right]$$

$$= \frac{1}{(k+1)!} \left[e^{-\lambda} \cdot \lambda^{k+1} + \frac{(k+1)!}{k!} \int_0^\infty e^{-x} x^k dx \right]$$

$$= \frac{1}{(k+1)!} \left[e^{-\lambda} \cdot \lambda^{k+1} + (k+1)! \sum_{i=0}^k e^{-\lambda} \frac{\lambda^i}{i!} \right]$$

$$= \sum_{i=0}^{k+1} e^{-\lambda} \frac{\lambda^i}{i!}$$

$$\text{so we proved the equation now } \sum_{i=0}^n e^{-\lambda} \frac{\lambda^i}{i!} = \frac{1}{n!} \int_0^\infty e^{-x} x^n dx$$

27. $\because X$ is a geometric random variable, $P(X=n) = p(1-p)^{n-1}$.

$$\therefore P\{X=n+k | X > n\} = \frac{P\{X=n+k\}}{P\{X > n\}} = \frac{p(1-p)^{n+k-1}}{\sum_{k=n}^{\infty} p(1-p)^{k-1}} = \frac{p(1-p)^{n+k-1}}{p \frac{(1-p)^{n-1}}{1-(1-p)}} = p(1-p)^{k-1}$$

$$= P\{X=k\}.$$

If the first n trials are full failures, then it is as if we are beginning anew at that time.

28. The events $\{X > n\}$ and $\{Y < r\}$ are both equivalent to the event that there are fewer than r successes in the first n trials, hence, they are the same event.

31. Let Y denote the largest of the remaining m chips, the event is the same if you draw m chips from $n+m$ chips and those m chips have the largest value. $Y=j$ so the other

$m-1$ chips are drawn from $j-1$ numbers $\therefore P(Y=j) = \binom{j-1}{m-1} / \binom{n+m}{m}$, $m \leq j \leq n+m$.

Let X denote the number of chips drawn having numbers that exceed all the numbers

of those remaining, so $X = n+m-Y$.

$$P(X=i) = P(Y=n+m-i) = \binom{m+n-i-1}{m-1} / \binom{n+m}{m}, i \leq n.$$

34. Let X denote the number of elements in the chosen subset; then $P(X=k) = \frac{\binom{n}{k}}{2^n - 1}$

$$\therefore E(X) = \sum_{k=0}^n \frac{k \binom{n}{k}}{2^n - 1} \quad \text{using the result of theoretical exercise 12 of Chapter 1}$$

$$= \frac{n 2^{n-1}}{2^n - 1}$$

$$E(X^2) = \sum_{k=0}^n \frac{k^2 \binom{n}{k}}{2^n - 1} = \frac{2^{n-2} n(n+1)}{2^n - 1}$$

$$\text{Var}(X) = E[X^2] - (E[X])^2 = \frac{(2^n - 1) 2^{n-2} n(n+1) - n^2 2^{2n-2}}{(2^n - 1)^2} = \frac{n 2^{2n-2} - n(n+1) 2^{n-2}}{(2^n - 1)^2} \sim \frac{n 2^{2n-2}}{2^{2n}}$$

$$= \frac{n}{4}$$

Additional problem:

$$P(X_1 = l) = \sum_{n=l}^{\infty} P(X_1 = l | X = n) P(X = n) = \sum_{n=l}^{\infty} \binom{n}{l} p^l (1-p)^{n-l} \cdot \frac{e^{-\lambda} \lambda^n}{n!} = \frac{e^{-\lambda} p (\lambda p)^l}{l!}$$

$$P(X_2 = m) = \sum_{n=m}^{\infty} P(X_2 = m | X = n) P(X = n) = \sum_{n=m}^{\infty} \binom{n}{m} (1-p)^m p^{n-m} \frac{e^{-\lambda} \lambda^n}{n!} = \frac{e^{-\lambda} (1-p)^m (\lambda (1-p))^m}{m!}$$

$$P(X_1 = l, X_2 = m) = P(X_1 = l, X = m+l) = P(X_1 = l | X = m+l) P(X = m+l)$$

$$= \binom{m+l}{l} p^l (1-p)^m \cdot \frac{e^{-\lambda} \lambda^{m+l}}{(m+l)!} = \frac{e^{-\lambda} p (\lambda p)^l (\lambda (1-p))^m \cdot e^{-\lambda} (1-p)^m}{l! m!} = P(X_1 = l) P(X_2 = m)$$

$\therefore X_1$ and X_2 are independent random variables.