

1. (a) If Monty opens a door and reveals a goat then we are left with two events in the sample space. One with chosen door has car and remaining door has computer and second with chosen door has computer and remaining door has computer. Hence chances are fifty-fifty that contestant will get car if he or she switches.

So if switched probability of

$$\text{getting car} = \frac{1}{2}$$

(b) Let,  $C_i = i^{\text{th}}$  door has car for  $i=1,2,3$   
 $D_i = i^{\text{th}}$  door has computer for  $i=1,2,3$

Let contestant chooses door 1 and Monty reveals door 2 has computer.

So if contestant switches, we need to calculate  $P(C_3|D_2)$

From Baye's theorem,

$$P(C_3|D_2) = \frac{P(D_2|C_3) P(C_3)}{P(D_2)}$$

$$P(c_i) = \frac{1}{3} \quad \text{for } i=1,2,3$$

$$P(D_2|c_3) = P \quad (\text{because door 2 is less preferred choice given } c_3)$$

$$P(D_2|c_2)=0 \quad (\text{because door 2 has computer from the given criteria})$$

$$P(D_2|c_1)=q \quad (\text{Now door 2 is more preferred choice because door 3 has goat } \cancel{\text{sheep}} \text{ given } c_1)$$

$$\Rightarrow \frac{P(D_2|c_3)P(c_3)}{P(D_2)} = \frac{\cancel{\frac{1}{3} \times P}}{\frac{P(D_2)}{P(D_2)}}$$

From LOTP,

$$\frac{\frac{1}{3} \times P}{P(D_2|c_1)P(c_1) + P(D_2|c_2)P(c_2) + P(D_2|c_3)P(c_3)}$$

$$= \frac{\frac{P}{3}}{\frac{P}{3} + \frac{q}{3}} = \frac{\frac{P}{3}}{\frac{P+q}{3}} = \frac{\frac{P}{3}}{\frac{1}{3}} = P$$

Hence contestant should switch if  $P > \frac{1}{2}$

(2) The recurrence relation for  $P_i$  (Probability of reaching  $N$  dollars starting with  $i$  dollars) is:

$$P_i = \frac{2}{3} P_{i-1} + \frac{1}{3} P_{i+1}$$

Solution to this difference equation is

$$\frac{1 - \left(\frac{q}{p}\right)^i}{1 - \left(\frac{q}{p}\right)^N}, \text{ where } q=2/3 \text{ and } p=\frac{1}{3}$$

Here,  $q/p$  is 2 and  $N = i+2$

$$\Rightarrow P_i = \frac{1 - 2^i}{1 - 2^{i+2}} = \frac{2^i - 1}{4 \times 2^i - 1}$$

$$= \frac{1}{4} \left( \frac{2^i - 1}{2^i - \frac{1}{4}} \right)$$

$$\text{Now, } 2^i - \frac{1}{4} > 2^i - 1$$

$$\Rightarrow \frac{2^i - 1}{2^i - \frac{1}{4}} < 1$$

$$\Rightarrow \frac{1}{4} \left( \frac{2^i - 1}{2^i - \frac{1}{4}} \right) < \frac{1}{4}$$

$$\Rightarrow P_i < \frac{1}{4}$$

3.(a)

	$C_1$	$C_2$
green	141	19
red	39	1
green %.	78.33%	95%

	$M_1$	$M_2$
green	10	158
red	10	22
green %.	50%	87.7%

overall % of green in  $C_1$  and  $C_2$  =  $\frac{141+19}{141+39+1} = \frac{160}{200} = 80\%$ .

overall % of green in  $M_1$  and  $M_2$  =  $\frac{10+158}{10+158+10+22} = \frac{168}{200} = 84\%$ .

- (b) A is percentage of green gummi bears  
 B is Mauve jar and  $B^c$  is crimson jar  
 C is Jar 1 and  $C^c$  is Jar 2

Hence,

$$P(A|B,C) < P(A|B^c,C)$$

$$P(A|B,C^c) < P(A|B^c,C^c)$$

$$\text{but, } P(A|B) > P(A|B^c)$$

Intuitively, ratio of individuals jars are compared first and then overall ratio is being compared here and results are opposite in the two cases. Hence relates to Simpson's paradox.

4. (a) We need to find  $P(D | T_1, T_2, T_3, \dots, T_n)$

From Baye's theorem,

$$P(D | T_1, T_2, T_3, \dots, T_n) = \frac{P(T_1, T_2, T_3, \dots, T_n | D) P(D)}{P(T_1, T_2, T_3, \dots, T_n)}$$

Given  $D$  or  $D^c$   $T_1, T_2, T_3, \dots, T_n$   
are independant.

$$\Rightarrow P(T_1, T_2, T_3, \dots, T_n | D) = P(T_1 | D) P(T_2 | D) \dots P(T_n | D)$$

$$\text{and } P(T_1, T_2, T_3, \dots, T_n | D^c) = P(T_1 | D^c) P(T_2 | D^c) \dots P(T_n | D^c)$$

$$\Rightarrow P(T_1, T_2, T_3, \dots, T_n | D) = a^n$$

$$\text{And, } P(T_1, T_2, T_3, \dots, T_n | D^c) = b^n$$

$$\Rightarrow P(D | T_1, T_2, \dots, T_n) = \frac{P a^n}{P(T_1, T_2, \dots, T_n)}$$

From LOTP,

$$P(T_1, T_2, \dots, T_n) = P(T_1, T_2, \dots, T_n | D) P(D) + P(T_1, T_2, \dots, T_n | D^c) P(D^c)$$

$$= P a^n + q b^n$$

~~$\Rightarrow P(T_1, T_2, T_3, \dots, T_n | D)$~~ 

$$\Rightarrow P(D | T_1, T_2, \dots, T_n) = \frac{P a^n}{P a^n + q b^n}$$

(b) From the question,

$$P(T_j | G) = 1 \quad \text{for } j=1, 2, 3, \dots, n$$

$$\text{Now, } P(D | T_1, T_2, \dots, T_n) = \frac{P(T_1, T_2, \dots, T_n | D) P(D)}{P(T_1, T_2, \dots, T_n)} \quad \text{--- (i)}$$

$$P(T_1, T_2, \dots, T_n | D) = \frac{P(T_1, T_2, \dots, T_n | D, G) P(G) + P(T_1, T_2, \dots, T_n | D, G^c) P(G^c)}{P(G^c)}$$

$$= 1 \times \frac{1}{2} + P(T_1 | D, G^c) P(T_2 | D, G^c) \dots P(T_n | D, G^c) \frac{1}{2}$$

$$= \frac{1}{2} (1 + a_0^n)$$

$$\text{So numerator of (i)} = \frac{P}{2} (1 + a_0^n)$$

Using LOTP,

$$P(T_1, T_2, \dots, T_n) = P(T_1, T_2, \dots, T_n | D) P(D) + P(T_1, T_2, \dots, T_n | D^c) P(D^c)$$

$$P(T_1, T_2, \dots, T_n | D) P(D) = \frac{P}{2} (1 + a_0^n)$$

$$P(T_1, T_2, \dots, T_n | D^c) P(D^c) = q \left( P(T_1, T_2, \dots, T_n | D^c, G) P(G) + P(T_1, T_2, \dots, T_n | D^c, G^c) P(G^c) \right)$$

$$= \frac{q}{2} (1 + b_0^n)$$

$$\text{Hence denominator of (i)} = \frac{P}{2} (1 + a_0^n) + \frac{q}{2} (1 + b_0^n)$$

$$\begin{aligned} \Rightarrow P(D | T_1, T_2, \dots, T_n) &= \frac{\frac{P}{2} (1 + a_0^n)}{\frac{P}{2} (1 + a_0^n) + \frac{q}{2} (1 + b_0^n)} \\ &= \frac{P (1 + a_0^n)}{P (1 + a_0^n) + q (1 + b_0^n)} \end{aligned}$$

5. (a) Let  $M$  = mother has disease

$c_1$  = child one has disease

$c_2$  = child two has disease

We need to find  $P(c_1^c, c_2^c)$

Using LOTP,

$$P(c_1^c, c_2^c) = P(c_1^c, c_2^c | M) P(M) + P(c_1^c, c_2^c | M^c) P(M^c)$$

$$P(M) = \frac{1}{3}, P(M^c) = \frac{2}{3}$$

$$P(c_1^c, c_2^c | M^c) = 1$$

$$\begin{aligned} P(c_1^c, c_2^c | M) &= P(c_1^c | M) P(c_2^c | M) \\ &= \frac{1}{2} \times \frac{1}{2} = \frac{1}{4} \end{aligned}$$

$$\begin{aligned} \Rightarrow P(c_1^c, c_2^c) &= \frac{1}{4} \times \frac{1}{3} + 1 \times \frac{2}{3} = \frac{1}{12} + \frac{2}{3} \\ &= \frac{9}{12} = \frac{3}{4} \end{aligned}$$

(b) If any child has disease implies mother has disease and thus other child may or may not have disease. Thus if one child has disease, it predicts something about disease status of second child. Hence they are not independent.

$$\text{(c)} \quad P(M | C_1^c, C_2^c) = \frac{P(C_1^c, C_2^c | M) P(M)}{P(C_1^c, C_2^c)}$$
$$= \frac{\frac{1}{2} \times \frac{1}{2} \times \frac{1}{3}}{\frac{3}{4}}$$
$$= \frac{1}{12} \times \frac{4}{3} = \frac{1}{9}$$

6. (a)  $P(X=k) = 1 - \text{Probability of no success in } k \text{ trials}$

$$= 1 - \left(\frac{1}{2}\right)^k$$

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

(b) If  $k$  trials are performed all trials will have at least one success and at least one failure except these two cases

- (i) All trials result in failure
- (ii) All trials result in success

And there  $2^k$  possible results

$$\text{Hence, } P(X=k) = \frac{2^k - 2}{2^k} \quad \forall k \geq 2$$

$$= 0 \quad \forall 0 < k < 2$$

(7) In the pebble world, given that  $x$  is in  $B$  we remove all the pebbles that are not in  $B$ . Now for every remaining pebbles, probability of ~~any~~<sup>every</sup> pebble is same because distribution is uniform.

Hence,  $P(x=x | x \in B) = \frac{1}{|B|}$

$$(x | x \in B) \sim D\text{Unif}(B)$$

8. (a) r.v.  $X$  = number of winning tickets

$$X \sim \text{Bin}(3, P)$$

$$P(X=k) = \binom{3}{k} P^k (1-P)^{3-k}$$

(b)

~~Probabilistically~~

(b) Using Inclusion-exclusion,

Probability of atleast one winning ticket

$$\begin{aligned} &= P(X=1) + P(X=2) + P(X=3) \\ &= \binom{3}{1} P(1-P)^2 + \binom{3}{2} P^2(1-P) + \binom{3}{3} P^3 \\ &= P^3 + 3P(1+P^2 - 2P) + 3P^2(1-P) \\ &= P^3 + 3P + 3P^3 - 6P^2 + 3P^2 - 3P^3 \\ &= P^3 + 3P - 3P^2 \\ &= 3P - 3P^2 + P^3 \end{aligned}$$

(c)

Using Compliment,

$$\text{Required probability} = 1 - P(X=0)$$

$$= 1 - \binom{3}{0} P^0 (1-P)^3$$

$$\begin{aligned} &= 1 - (1-P)^3 &= 1 - 1 + P^3 - 3P^2 + 3P \\ &= 3P - 3P^2 + P^3 \end{aligned}$$

(c)

$$3P - 3P^2 + P^3 = 3P - \{P^2(3-P)\}$$

$$\text{Now, } P^2(3-P) > 0$$

$$\text{Hence, } 3P - 3P^2 + P^3 < 3P$$

If  $P$  is very small then  $P^2(3-P)$  tends to zero.  
In that case required probability is close to  $3P$ .

9. (a) From LOTP,

$$P(X=x) = P(X=x|C_1)P(C_1) + P(X=x|C_2)P(C_2)$$

where,  $C_1$  = coin one is chosen

$C_2$  = coin two is chosen

$$P(C_1) = P(C_2) = \frac{1}{2}$$

$$\Rightarrow P(X=x) = \frac{1}{2} (P(X=x|C_1) + P(X=x|C_2))$$

$$\text{Now, } P(X=x|C_1) = \binom{n}{x} p_1^x (1-p_1)^{n-x}$$

$$\text{and } P(X=x|C_2) = \binom{n}{x} p_2^x (1-p_2)^{n-x}$$

$$\text{so, } P(X=x) = \frac{1}{2} (\binom{n}{x} p_1^x (1-p_1)^{n-x} + \binom{n}{x} p_2^x (1-p_2)^{n-x})$$

(b) Let  $P_1 = P_2 = P$

$$\Rightarrow P(X=x) = \frac{1}{2} (\binom{n}{x} P^x (1-P)^{n-x} + \binom{n}{x} P^x (1-P)^{n-x})$$

$$= \binom{n}{x} P^x (1-P)^{n-x}$$

Hence,  $X \sim \text{Bin}(n, P)$

where  $P = P_1 = P_2$

(c) If  $P_1 \neq P_2$  then each coin toss is not a independent Bernoulli trial because previous trials will predict outcomes of future coin tosses as it will give us hint about which coin is chosen. Hence each trial is not an independent Bernoulli trials which is a requirement of Binomial distribution.

(10) Any given person will vote for kodos with probability  $P = P_1 P_2 P_3$

Each person will either vote for kodos or will not vote for kodos and each person's vote is independent of each other.

r.v.  $X$ : number of votes for kodos

$X$  has Binomial distribution

$$X \sim \text{Bin}(n, P)$$

$$P(X=x) = \binom{n}{x} P^x (1-P)^{n-x}$$

$$\text{where } P = P_1 P_2 P_3$$

11.(a) If first two tosses are heads out of  $x=n$  heads then remaining  $(n-2)$  tosses will result in  $n-2$  heads and distribution will be  $\text{Bin}(n-2, \frac{1}{2})$

Let  $F = \text{first two heads}$

$$\begin{aligned} P(X=n | F) &= \binom{n-2}{n-2} \left(\frac{1}{2}\right)^{n-2} \left(\frac{1}{2}\right)^{n-2-n-2} \\ &= \binom{n-2}{n-2} \left(\frac{1}{2}\right)^{n-2} = \binom{8}{n-2} \left(\frac{1}{2}\right)^8 \end{aligned}$$

(b)  $P(X=x) = 0 \quad \text{for } 0 \leq x \leq 2$

otherwise,  $P(X=x) = \binom{n}{x} \left(\frac{1}{2}\right)^n$  (because if number of heads are at least two then for  $x \geq 2$  its true anyways)

12. (a) From the question

$$P(X=x) = \binom{n}{x} p^x (1-p)^{n-x}$$

Let  $Y$  is a r.v.

$Y$  = Number of statistics major  
in sample size  $m$

For a given  $x=x$  if we have to chose  $y=y$  statistics major then its a hypergeometric distribution.

$$\cancel{\Rightarrow P(Y=y | X=x) =}$$

$$(Y | X=x) \sim HGeom(x, n-x, m)$$

And from Law of total probability,

$$\begin{aligned} P(Y=y) &= \sum_{x=y}^n P(Y=y | X=x) P(X=x) \\ &= \sum_{x=y}^n \frac{\binom{x}{y} \binom{n-x}{m-y}}{\binom{n}{m}} \binom{n}{x} p^x (1-p)^{n-x} \end{aligned}$$

(b) Once we have sample drawn of size  $m$ , then amongst those  $m$  students each student has same probability of being a statistics major student. And they are independantly either a statistics students or not. Hence a Binomial distribution.

$$Y \sim Bin(m, p)$$

$$\Rightarrow P(Y=y) = \binom{m}{y} p^y (1-p)^{m-y}$$