HW 1

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1

Let event A be the event that in a group of n people, no one shares a birthday. Let event B be the complement of event A, such that in a group of n people, at least 2 people share a birthday. Thus, P(B) = 1 - P(A). We can calculate P(A) by starting with the first person, who can have any birthday. The next person must have a different birthday, thus have a $\frac{364}{365}$ chance of not sharing a birthday with the first person. This goes on until the nth person, who has a $\frac{365-n+1}{365}$ chance of not sharing a birthday with any of the previous n-1 people. Thus, we have:

$$P(A) = \frac{365}{365} \cdot \frac{364}{365} \cdot \frac{363}{365} \cdot \dots \frac{365 - n + 1}{365} = \frac{1}{365^n} \cdot \frac{365 \cdot 364 \cdot \dots (365 - n + 1)}{1}$$
Recognize that
$$\frac{365 \cdot 364 \cdot \dots (365 - n + 1)}{1} = \frac{365!}{(365 - n)!} \cdot \dots$$

$$P(A) = \frac{365!}{365^n (365 - n)!}$$

We want to find a minimum n such that $P(B) \ge 0.5$. P(A) < 0.5. Solving P(A) < 0.5 for n, we find that the minimum n is 23, where $P(A) \approx 0.493$ and $P(B) \approx 0.507$.

Thus, in a group of n unrelated individuals, the probability that at least 2 people share a birthday is $P(B) = 1 - \frac{365!}{365^n(365-n)!}$. This probability exceeds 0.5 when $n \ge 23$.

 $\mathbf{2}$

2.1

Each bag has a probability of $\frac{1}{3}$ of being chosen. The probability of drawing a blue marble at random is thus the sum of the probabilities of drawing a

blue marble from each bag multiplied by the probability of choosing that bag. Thus, we have:

$$P(\text{blue marble}) = \frac{1}{3} \cdot \frac{25}{100} + \frac{1}{3} \cdot \frac{40}{100} + \frac{1}{3} \cdot \frac{55}{100} = \frac{2}{5} = 0.4$$

2.2

If the first bag has the probability of 0.5 of being chosen, the other two bags then have a probability of 0.25 of being chosen. Thus, we now have:

$$P(\text{blue marble}) = 0.5 \cdot \frac{25}{100} + 0.25 \cdot \frac{40}{100} + 0.25 \cdot \frac{55}{100} = \frac{29}{80} = 0.3625$$

3

For the gambler's current fortune of k dollars, they can either win or lose on the next gamble. Thus by the law of total probability, we have:

$$q_k = \frac{1}{2} \cdot q_{k+1} + \frac{1}{2} \cdot q_{k-1}$$
$$2q_k = q_{k+1} + q_{k-1}$$
$$q_{k+1} - q_k = q_k - q_{k-1}$$

This shows that the difference between consecutive q values is constant, thus q_k is linear and can express q_k as $q_k = A + Bk$ for some constants A, B. We solve for A, B using the boundary conditions where $q_0 = 1$ and $q_N = 0$:

$$q_0 = A + B \cdot 0 : A = 1$$

$$q_N = 1 + BN = 0 : B = -\frac{1}{N}$$

$$\therefore q_k = 1 - \frac{k}{N}$$

4

For a given element x in Ω , the probability of x being in either set A or B is equally likely because A and B are independently selected subsets of Ω . Thus, for any element x to satisfy $A \subseteq B$, element x must either be:

- $x \notin A \& x \notin B$
- $x \notin A \& x \in B$
- $x \in A \& x \in B$

The only case that does not satisfy $A \subseteq B$ is if $x \in A \& x \notin B$. Since each of the 4 cases are equally likely, the probability of x satisfying $A \subseteq B$ is $\frac{3}{4}$. Since this must be true for all n elements in Ω , the probability of $A \subseteq B$ is consequently:

$$P(A \subseteq B) = \left(\frac{3}{4}\right)^n$$

5

5.1

Each coin is equally likely to be chosen with a probability of $\frac{1}{3}$. The probability of getting tails when a coin is flipped is thus the sum of the probabilities of getting tails with each coin multiplied by the probability of choosing that coin. Thus, we have:

$$P(T) = \frac{1}{3} \cdot \frac{1}{2} + \frac{1}{3} \cdot \frac{1}{2} + \frac{1}{3} \cdot 0 = \frac{1}{3}$$

5.2

$$\begin{split} P(\text{fake coin}|\mathbf{H}) &= \frac{P(\mathbf{H}|\text{fake coin}) \cdot P(\text{fake coin})}{P(\mathbf{H})} \\ &= \frac{1 \cdot \frac{1}{3}}{\frac{1}{3} \cdot \frac{1}{2} + \frac{1}{3} \cdot \frac{1}{2} + \frac{1}{3} \cdot 1} = \frac{1}{2} \end{split}$$

6

Any joint distribution function must satisfy:

- 1. $F(-\infty, x_2) = F(x_1, -\infty) = 0$
- 2. $F(\infty,\infty)=0$
- 3. $F(x_1, \infty) = F(x_1) \& F(\infty, x_2) = F(x_2)$
- 4. monotonicity: $F(a_1, a_2) \leq F(b_1, b_2)$ if $a_1 \leq b_1 \& a_2 \leq b_2$
- 5. $P(a_1 < X \le b_1, a_2 < Y \le b_2) = F(b_1, b_2) F(a_1, b_2) F(b_1, a_2) + F(a_1, a_2) \ge 0$

6.1

For
$$F(x_1, x_2) = 1 - e^{-x_1 - x_2}$$
 if $x_1, x_2 \ge 0$:

1.
$$F(-\infty, x_2) = F(x_1, -\infty) = 0$$
 is satisfied as $x_1, x_2 \ge 0$

2.
$$F(\infty, \infty) = 1 - e^{-\infty - \infty} = 1 - 0 = 1$$
 is satisfied

3.
$$F(x_1, \infty) = 1 - e^{-x_1 - \infty} = 1 - 0 = 1$$
 and $F(\infty, x_2) = 1 - e^{-\infty - x_2} = 1 - 0 = 1$ is satisfied

- 4. monotonicity is satisfied as e^{-x} is a decreasing function
- 5. $P(a_1 < X \le b_1, a_2 < Y \le b_2) = F(b_1, b_2) F(a_1, b_2) F(b_1, a_2) + F(a_1, a_2)$ $= (1 - e^{-b_1 - b_2}) - (1 - e^{-a_1 - b_2}) - (1 - e^{-b_1 - a_2}) + (1 - e^{-a_1 - a_2})$ Let $a_1 = 0, a_2 = 0, b_1 = t_1 > 0, b_2 = t_2 > 0$, then: $P(0 < X \le t_1, 0 < Y \le t_2) = (1 - e^{-t_1 - t_2}) - (1 - e^{0 - t_2}) - (1 - e^{-t_1 + 0}) + (1 - e^{0 + 0})$ $= 1 - e^{-t_1 - t_2} - 1 + e^{-t_2} - 1 + e^{-t_1} = e^{-t_2} + e^{-t_1} - e^{-t_1 - t_2} - 1$ $= (e^{-t_1} - 1)(1 - e^{-t_2})$ $(e^{-t_1} - 1) < 0$ and $(1 - e^{-t_2}) > 0$, therefore $P(0 < X \le t_1, 0 < Y \le t_2) < 0$ which violates condition 5.

Thus, $F(x_1, x_2) = 1 - e^{-x_1 - x_2}$ if $x_1, x_2 \ge 0$ is not a valid joint distribution function.

6.2

For $F(x_1, x_2) = 1 - e^{-\min(x_1, x_2)} - \min(x_1, x_2)e^{-\min(x_1, x_2)}$ if $x_1, x_2 > 0$:

1.
$$F(-\infty, x_2) = F(x_1, -\infty) = 0$$
 is satisfied as $x_1, x_2 > 0$

2.
$$F(\infty, \infty) = 1 - e^{-\infty} - \infty \cdot e^{-\infty} = 1 - 0 - 0 = 1$$
 is satisfied

3.
$$F(x_1, \infty) = 1 - e^{-x_1} - x_1 e^{-x_1} = F(x_1)$$
 and $F(\infty, x_2) = 1 - e^{-x_2} - x_2 e^{-x_2} = F(x_2)$ is satisfied

- 4. monotonicity is satisfied as e^{-x} is a decreasing function and min (x_1, x_2) is increasing
- 5. $P(a_1 < X \le b_1, a_2 < Y \le b_2)$ Again, let $a_1 = 0, a_2 = 0, b_1 = t, b_2 = t$ where t > 0, then:

$$P(0 < X \le t, 0 < Y \le t) = F(t,t) - F(0,t) - F(t,0) + F(0,0)$$

= $(1 - e^{-t} - te^{-t}) - (1 - e^{0} - 0 \cdot e^{0}) - (1 - e^{0} - 0 \cdot e^{0}) + (1 - e^{0} - 0 \cdot e^{0})$
= $1 - e^{-t} - te^{-t}$, which has a minimum of 0 when $t = 0$ and is positive for all $t > 0$ as know from condition 4.

Thus, $F(x_1, x_2) = 1 - e^{-\min(x_1, x_2)} - \min(x_1, x_2)e^{-\min(x_1, x_2)}$ if $x_1, x_2 \ge 0$ is a valid joint distribution function. Marginal distribution functions $F_1(x_1)$ and $F_2(x_2)$ are determined in condition 3.

For $\max(X, Y)$, both X and Y must be less than or equal to some value t for $\max(X, Y)$ to be possible. Thus:

$$P(\max(X,Y) \leq t) = P(X \leq t, Y \leq t)$$
 By independence,
$$P(X \leq t, Y \leq t) = P(X \leq t)P(Y \leq t)$$

$$\therefore F_{\max}(t) = F(t)G(t)$$

For $\min(X, Y)$, either X or Y must be greater than to some value t for $\min(X, Y)$ to be possible. Thus:

$$P(\min(X,Y)>t) = P(X>t,Y>t)$$
 By independence,
$$P(X>t,Y>t) = P(X>t)P(Y>t)$$
 By law of total probability,
$$P(X>t) = 1 - P(X \le t) = 1 - F(t)$$
 Likewise,
$$P(Y>t) = 1 - G(t)$$

$$\therefore P(\min(X,Y)>t) = (1-F(t))(1-G(t))$$
 By law of total probability,
$$P(\min(X,Y) \le t) = 1 - P(\min(X,Y) > t)$$