AI 5030: Probability and Stochastic Processes

Instructor: Dr. Karthik P. N.

Homework 7

Topics: Abstract Integrals, Expectations of Discrete Random Variables

1. Fix $n \in \mathbb{N}$. Consider the measure space $(\mathbb{R}, \mathscr{B}(\mathbb{R}), \lambda)$, where λ denotes the Lebesgue measure. Compute $\int_{\mathbb{R}} f \, \mathsf{d}\lambda$ for each of the following cases.

   (a) $f : \mathbb{R} \to \mathbb{R}$ is defined as
   $$f(\omega) = \begin{cases} \omega, & \omega \in \{0, 1, \ldots, n\}, \\ 0, & \text{otherwise.} \end{cases}$$

   (b) $f : \mathbb{R} \to \mathbb{R}$ is defined as
   $$f(\omega) = \begin{cases} 1, & \omega \in \mathbb{Q}^c \cap [0, 1], \\ 0, & \text{otherwise.} \end{cases}$$

   (c) $f : \mathbb{R} \to \mathbb{R}$ is defined as
   $$f(\omega) = \begin{cases} n, & \omega \in \mathbb{Q}^c \cap [0, n], \\ 0, & \text{otherwise.} \end{cases}$$

   Solution: We present the solution to each of the parts below.

   (a) Notice that f is a simple function which may be expressed in canonical form as
   $$f = 0\,\mathbf{1}_{\{0\}} + 1\,\mathbf{1}_{\{1\}} + \cdots + n\,\mathbf{1}_{\{n\}} + 0\,\mathbf{1}_{\mathbb{R}\setminus\{0,1,\ldots,n\}} = 0\,\mathbf{1}_{\mathbb{R}\setminus\{1,\ldots,n\}} + 1\,\mathbf{1}_{\{1\}} + \cdots + n\,\mathbf{1}_{\{n\}}.$$
   It then follows that
   $$\int_{\mathbb{R}} f \, \mathsf{d}\lambda = 0\,\lambda(\mathbb{R} \setminus \{1, \ldots, n\}) + 1\,\lambda(\{1\}) + \cdots + n\,\lambda(\{n\}) = 0,$$
   where the last equality follows by noting that $\lambda(\{i\}) = 0$ for all $i \in \{1, \ldots, n\}$.

   (b) Notice that f may be expressed as
   $$f = \mathbf{1}_{\mathbb{Q}^c \cap [0,1]}.$$
   Thus, f is simple, and
   $$\int_{\mathbb{R}} f \, \mathsf{d}\lambda = \lambda(\mathbb{Q}^c \cap [0, 1]) = \lambda([0, 1]) - \lambda(\mathbb{Q} \cap [0, 1]) \overset{(*)}{=} 1 - 0 = 1,$$
   where $(*)$ above follows by noting that $\lambda(\mathbb{Q} \cap [0, 1]) \leq \lambda(\mathbb{Q}) = 0$.

   (c) Notice that f may be expressed as
   $$f = n\mathbf{1}_{\mathbb{Q}^c \cap [0,n]}.$$
   Therefore, f is simple, and
   $$\int_{\mathbb{R}} f \, \mathsf{d}\lambda = n\,\lambda(\mathbb{Q}^c \cap [0, n]) = n\Big(\lambda([0, n]) - \lambda(\mathbb{Q} \cap [0, n])\Big) = n(n - 0) = n^2,$$
   where the last equality follows by noting that $\lambda(\mathbb{Q} \cap [0, n]) \leq \lambda(\mathbb{Q}) = 0$.

2. Fix $n \in \mathbb{N}$. Let $\Omega = \{\omega_1, \ldots, \omega_n\}$, $\mathscr{F} = 2^{\Omega}$, and $\mathbb{P}(\{\omega_i\}) = \frac{1}{n}$ for all $i \in \{1, \ldots, n\}$.
   Let $X : \Omega \to \mathbb{R}$ be a random variable defined with respect to $\mathscr{F}$.
   Compute $\mathbb{E}[X] = \int_{\Omega} X \, \mathsf{d}\mathbb{P}$ for the following cases.

   (a) $X = \mathbf{1}_A$, where $A = \{\omega_1, \ldots, \omega_m\}$, with $1 \leq m \leq n$.

(b) X is defined as

$$X(\omega) = \begin{cases} i, & \omega = \omega_i,\, \omega_i \in A, \\ 0, & \text{otherwise}. \end{cases}$$

Solution: We present the solution to each of the parts below.

(a) Notice that X is a simple random variable, and

$$\mathbb{E}[X] = \mathbb{P}(A) = \mathbb{P}(\{\omega_1, \ldots, \omega_m\}) = \sum_{i=1}^{m} \mathbb{P}(\{\omega_i\}) = \frac{m}{n}.$$

(b) Notice that X may be expressed as

$$X = \sum_{i=1}^{m} i\, \mathbf{1}_{\{\omega_i\}} + 0 \cdot \mathbf{1}_{A^c}.$$

Thus, it follows that X is a simple random variable, and

$$\mathbb{E}[X] = \sum_{i=1}^{m} i\, \mathbb{P}(\{\omega_i\}) + 0\, \mathbb{P}(A^c) = \sum_{i=1}^{m} \frac{i}{n} = \frac{m(m+1)}{2n}.$$

3. Let $(\Omega, \mathscr{F}) = (\mathbb{R}, \mathscr{B}(\mathbb{R}))$. For a fixed $c \in \mathbb{R}$, define $\delta_c : \mathscr{F} \to [0,1]$ as

$$\delta_c(A) = \begin{cases} 1, & c \in A, \\ 0, & c \notin A. \end{cases}$$

(a) Show that δ_c is a probability measure on $(\Omega, \mathscr{F})$.
Remark: δ_c is called the Dirac measure at c.
It is referred to as "unit impulse" in the engineering literature, and sometimes (incorrectly) called a Dirac delta "function".

(b) For any simple function $g : \Omega \to \mathbb{R}$, show that $\int_\Omega g \, \mathrm{d}\delta_c = g(c)$.

(c) Extend the result in part (b) above to the case when g is non-negative.

(d) Let $\mu : \mathscr{F} \to [0, +\infty]$ be defined as

$$\mu(A) = \sum_{n=1}^{\infty} \delta_n(A), \qquad A \in \mathscr{F}.$$

Show that for any simple function $g : \Omega \to \mathbb{R}$,

$$\int_\Omega g \, \mathrm{d}\mu = \sum_{n=1}^{\infty} g(n).$$

Extend the above result to the case when g is non-negative.
Remark: Here, μ is a measure on $(\Omega, \mathscr{F})$, and is called the "counting" measure.
For any given $A \in \mathscr{F}$, $\mu(A)$ is equal to the count of the number of positive integers present in the set A.

The above exercise shows that every summation is simply an integral with respect to the counting measure.

Solution: We present the solution to each of the parts below.

(a) Note that $\delta_c(\mathbb{R}) = 1$ as $c \in \mathbb{R}$. Suppose that $A_1, A_2, \ldots \in \mathscr{B}(\mathbb{R})$ are disjoint. Then, by definition,

$$\delta_c\left(\bigcup_{n=1}^{\infty} A_n\right) = \begin{cases} 1, & c \in \bigcup_{n=1}^{\infty} A_n, \\ 0, & c \notin \bigcup_{n=1}^{\infty} A_n. \end{cases}$$

Because $A_i \cap A_j = \emptyset$ for all $i \neq j$, it follows that c belongs to exactly one of the A_i or to none of them. In other words, if $\delta_c\left(\bigcup_{n=1}^{\infty} A_n\right) = 1$, then there exists a unique $N \in \mathbb{N}$ such that $c \in A_N$, in which case

$$1 = \delta_c\left(\bigcup_{n=1}^{\infty} A_n\right) = \delta_c(A_N) = \sum_{n=1}^{\infty} \delta_c(A_n).$$

On the other hand, if $\delta_c\left(\bigcup_{n=1}^{\infty} A_n\right) = 0$, then $c \notin \bigcup_{n=1}^{\infty} A_n$, in which case

$$0 = \delta_c\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} \delta_c(A_n).$$

In either case, it follows that

$$\delta_c\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} \delta_c(A_n),$$

thereby establishing that δ_c is a probability measure on $(\mathbb{R}, \mathscr{B}(\mathbb{R}))$.

(b) Suppose that g has the canonical representation

$$g = \sum_{i=1}^{n} g_i \, \mathbf{1}_{G_i},$$

where $g_1, \ldots, g_n \geq 0$ are distinct, and $G_1, \ldots, G_n \in \mathscr{B}(\mathbb{R})$ form a partition of $\mathbb{R}$. Here,

$$G_i = \{x \in \mathbb{R} : g(x) = i\}, \qquad i \in \{1, \ldots, n\}.$$

Clearly, there exists a unique $m \in \{1, \ldots, n\}$ such that $c \in G_m$ (the uniqueness of m follows from the fact that $G_m \cap G_{m'} = \emptyset$ for all $m \neq m'$). Furthermore, $g(c) = g_m$. It then follows that

$$\int_{\mathbb{R}} g \, \mathsf{d}\delta_c = \sum_{i=1}^{n} g_i \, \delta_c(G_i) = g_m \, \delta_c(G_m) = g_m = g(c).$$

The desired result is thus proved.

(c) Suppose that $g : \mathbb{R} \to [0, +\infty]$ is non-negative. For each $n \in \mathbb{N}$, let

$$g_n(x) = \begin{cases} \dfrac{\lfloor 2^n g(x) \rfloor}{2^n}, & g(x) < n, \\ n, & g(x) \geq n. \end{cases}$$

Then, we have

$$\forall x \in \mathbb{R}, \qquad 0 \leq g_1(x) \leq g_2(x) \leq \cdots, \qquad \lim_{n \to \infty} g_n(x) = g(x). \tag{1}$$

Using the Monotone Convergence Theorem, we have

$$\int_{\mathbb{R}} g \, \mathsf{d}\delta_c = \lim_{n \to \infty} \int_{\mathbb{R}} g_n \, \mathsf{d}\delta_c \overset{(\dagger)}{=} \lim_{n \to \infty} g_n(c) = g(c),$$

where $(\dagger)$ follows from part (b) (noting that g_n is simple for each n), and the last equality follows from the pointwise convergence in (1). The desired result is thus established.

(d) We first verify that μ as defined in the question is indeed a measure on $(\mathbb{R}, \mathscr{B}(\mathbb{R}))$. First, we note that

$$\mu(\emptyset) = \sum_{n=1}^{\infty} \delta_n(\emptyset) = \sum_{n=1}^{\infty} 0 = 0.$$

Next, for any $A_1, A_2, \ldots \in \mathscr{B}(\mathbb{R})$ such that $A_i \cap A_j = \emptyset$ for all $i \neq j$, we have

$$\mu\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{n=1}^{\infty} \delta_n\left(\bigcup_{i=1}^{\infty} A_i\right) \overset{(a)}{=} \sum_{n=1}^{\infty} \sum_{i=1}^{\infty} \delta_n(A_i) \overset{(b)}{=} \sum_{i=1}^{\infty} \sum_{n=1}^{\infty} \delta_n(A_i) = \sum_{i=1}^{\infty} \mu(A_i),$$

where (a) above follows from the fact that δ_n is a probability measure on $(\mathbb{R}, \mathscr{B}(\mathbb{R}))$ (see part (a) of the question), and (b) follows by noting that the two infinite summations may be changed because $\delta_n(A_i) \geq 0$ for all i, n (when summing up non-negative real numbers, the order of summation does not matter). The above facts prove that μ is a measure.

Suppose that g is simple and has the canonical representation

$$g = \sum_{i=1}^{m} g_i \, \mathbf{1}_{G_i},$$

where $g_1, \ldots, g_m \geq 0$ are distinct, and $G_1, \ldots, G_m \in \mathscr{B}(\mathbb{R})$ form a partition of $\mathbb{R}$. Here,

$$G_i = \{x \in \mathbb{R} : g(x) = i\}, \qquad i \in \{1, \ldots, m\}.$$

We then have

$$\int_{\mathbb{R}} g \, \mathrm{d}\mu = \sum_{i=1}^{m} g_i \, \mu(G_i) = \sum_{i=1}^{m} g_i \sum_{n=1}^{\infty} \delta_n(G_i) = \sum_{i=1}^{m} \sum_{n=1}^{\infty} g_i \, \delta_n(G_i) = \sum_{n=1}^{\infty} \sum_{i=1}^{m} g_i \, \delta_n(G_i), \tag{2}$$

where in writing the last equality in (2), we interchange the order of summation noting that $g_i \, \delta_n(G_i) \geq 0$ for all i, n. Because $G_1, \ldots, G_m$ constitute a partiton of $\mathbb{R}$, it follows that for each n, there exists a unique $i_n^\star \in \{1, \ldots, m\}$ such that $n \in G_{i_n^\star}$. Furthermore, $g(n) = g_{i_n^\star}$. Using this in (2), we get

$$\int_{\mathbb{R}} g \, \mathrm{d}\mu = \sum_{n=1}^{\infty} g_{i_n^\star} \, \delta_n(G_{i_n^\star}) = \sum_{i=1}^{\infty} g(n).$$

The desired result is thus established.

4. Suppose that N is a discrete random variable taking values in $\mathbb{N}$. Prove that

$$\mathbb{E}[N] = \sum_{n=0}^{\infty} \mathbb{P}(\{N > n\}).$$

Hint: Notice that $N = \sum_{n=0}^{N-1} 1 = \sum_{n=0}^{\infty} \mathbf{1}_{\{N>n\}}$.
Apply expectations on both sides and use MCT to justify passing the expectation inside the infinite summation.

Solution: For each $m \in \mathbb{N}$, let

$$N_m \coloneqq \sum_{n=0}^{m} \mathbf{1}_{\{N>n\}}.$$

Then, we have

$$\forall \omega \in \Omega, \qquad 0 \leq N_1(\omega) \leq N_2(\omega) \leq \cdots, \qquad \lim_{m \to \infty} N_m(\omega) = \lim_{m \to \infty} \sum_{n=1}^{m} \mathbf{1}_{\{N>n\}}(\omega) = \sum_{n=0}^{\infty} \mathbf{1}_{\{N>n\}}(\omega) = N(\omega).$$

Thus, applying the Monotone Convergence Theorem to the sequence $\{N_m\}_{m=1}^{\infty}$, we get

$$\mathbb{E}[N] = \lim_{m \to \infty} \mathbb{E}[N_m] = \lim_{m \to \infty} \mathbb{E}\left[\sum_{n=0}^{m} \mathbf{1}_{\{N>n\}}\right] = \lim_{m \to \infty} \sum_{n=0}^{m} \mathbb{P}(\{N > n\}) = \sum_{n=0}^{\infty} \mathbb{P}(\{N > n\}).$$

This establishes the desired result.

5. Fix a probability space $(\Omega, \mathscr{F}, \mathbb{P})$.
Let $X : \Omega \to \mathbb{R}$ be a non-negative random variable with respect to $\mathscr{F}$.

(a) Suppose that $\mathbb{E}[X] < +\infty$. Then, show that

$$\lim_{n \to \infty} n \, \mathbb{P}(\{X > n\}) = 0. \tag{3}$$

Hint: Write $X = X\mathbf{1}_{\{X\leq n\}} + X\mathbf{1}_{\{X>n\}}$.
For each $n \in \mathbb{N}$, let $X_n = X\mathbf{1}_{\{X\leq n\}}$.
Show that $0 \leq X_n \leq X_{n+1}$ for all n, and $X_n \overset{\text{pointwise}}{\longrightarrow} X$. Using MCT, compute $\lim_{n\to\infty} \mathbb{E}[X_n]$.
Show that $\lim_{n\to\infty} \mathbb{E}[X\mathbf{1}_{\{X>n\}}] = 0$.
Finally, argue that $0 \leq \lim_{n\to\infty} n\,\mathbb{P}(\{X > n\}) \leq \lim_{n\to\infty} \mathbb{E}[X\mathbf{1}_{\{X>n\}}] = 0$.

(b) Produce an example of a random variable X for which $\mathbb{E}[X] = +\infty$, and

$$\lim_{n\to\infty} n\,\mathbb{P}(\{X > n\}) > 0.$$

This exercise shows that (3) holds only when $\mathbb{E}[X] < +\infty$.

Solution: We present the solution to each part below.

(a) For each $n \in \mathbb{N}$, let $X_n = X\mathbf{1}_{\{X\leq n\}}$. That is,

$$\forall \omega \in \Omega, \qquad X_n(\omega) = \begin{cases} X(\omega), & X(\omega) \leq n, \\ 0, & \text{otherwise}. \end{cases}$$

Noting that $\{X \leq n\} \subseteq \{X \leq n+1\}$ for all n, we have $\mathbf{1}_{\{X\leq n\}} \leq \mathbf{1}_{\{X\leq n+1\}}$. Using the fact that X is a non-negative random variable, it follows that

$$0 \leq X_n = X\,\mathbf{1}_{\{X\leq n\}} \leq X\,\mathbf{1}_{\{X\leq n+1\}} = X_{n+1} \qquad \forall n \in \mathbb{N}.$$

Furthermore, for each $\omega \in \Omega$, we have

$$X_n(\omega) = X(\omega) \qquad \forall n \geq \lceil X(\omega) \rceil,$$

thus implying that

$$\forall \omega \in \Omega, \qquad \lim_{n\to\infty} X_n(\omega) = X(\omega).$$

Applying the Monotone Convergence Theorem, we get

$$\mathbb{E}[X] = \lim_{n\to\infty} \mathbb{E}[X_n]. \tag{4}$$

We then have

$$\begin{aligned} \mathbb{E}[X] &= \mathbb{E}[X\mathbf{1}_{\{X\leq n\}} + X\mathbf{1}_{\{X>n\}}] \\ &= \mathbb{E}[X\mathbf{1}_{\{X\leq n\}}] + \mathbb{E}[X\mathbf{1}_{\{X>n\}}]. \end{aligned}$$

Because $\mathbb{E}[X] < +\infty$ (given in the question), it follows that each of the terms on the right-hand side of the above equation is finite. Hence, it follows that

$$\lim_{n\to\infty} \mathbb{E}[X\mathbf{1}_{\{X>n\}}] = \lim_{n\to\infty} \mathbb{E}[X] - \lim_{n\to\infty} \mathbb{E}[X\mathbf{1}_{\{X\leq n\}}] = \mathbb{E}[X] - \lim_{n\to\infty} \mathbb{E}[X_n] \overset{(**)}{=} \mathbb{E}[X] - \mathbb{E}[X] = 0,$$

where $(**)$ follows from (4). Finally, we note that

$$n\,\mathbb{P}(\{X > n\}) = n\,\mathbb{E}[\mathbf{1}_{\{X>n\}}] = \mathbb{E}[n\,\mathbf{1}_{\{X>n\}}] \leq \mathbb{E}[X\,\mathbf{1}_{\{X>n\}}],$$

where the last inequality above follows by noting that on the set $\{X > n\}$, we have $n < X$. We then have

$$0 \leq \lim_{n\to\infty} n\,\mathbb{P}(\{N > n\}) \leq \lim_{n\to\infty} \mathbb{E}[X\,\mathbf{1}_{\{X>n\}}] = 0,$$

thus proving the desired result.

(b) Suppose that X has the PMF

$$p_X(n) = \begin{cases} \dfrac{1}{n(n+1)}, & n \in \mathbb{N}, \\ 0, & \text{otherwise.} \end{cases}$$

Then, it follows that

$$\mathbb{P}(\{X > n\}) = \frac{1}{n+1}, \qquad n \in \{0, 1, \ldots\}.$$

Also, using the result in Question 4, we have

$$\mathbb{E}[X] = \sum_{n=0}^{\infty} \mathbb{P}(\{X > n\}) = \sum_{n=0}^{\infty} \frac{1}{n+1} = +\infty,$$

while at the same time, we have

$$\lim_{n \to \infty} n\, \mathbb{P}(\{X > n\}) = \lim_{n \to \infty} \frac{n}{n+1} = 1 > 0.$$

6. Fix a probability space $(\Omega, \mathscr{F}, \mathbb{P})$.
Let $X : \Omega \to [0, +\infty]$ be a non-negative, extended real-valued random variable with respect to $\mathscr{F}$.
(Here, X is allowed take the value $+\infty$.)

(a) Show that $\{X = +\infty\} = \{\omega \in \Omega : X(\omega) = +\infty\} \in \mathscr{F}$.
Hint: If $X(\omega) = +\infty$, then $X(\omega) > N$ for all $N \in \mathbb{N}$.

(b) Show that $\mathbb{E}[X] < +\infty$ implies that

$$\mathbb{P}(\{X < +\infty\}) = 1.$$

Hint: We have to show that $\mathbb{P}(\{X = +\infty\}) = 0$. We will do this by contradiction.
Let $L = \mathbb{E}[X]$. Suppose that $\mathbb{P}(\{X = +\infty\}) = p > 0$.
Let $C = \{X > 2L/p\}$. Using the reasoning of part (a), argue that $\mathbb{P}(C) \geq p$.
From class, we know that there exists a sequence of simple random variables $\{X_n\}_{n=1}^{\infty}$ such that $X_n \overset{\text{pointwise}}{\longrightarrow} X$.
Using the pointwise convergence property and MCT, argue that

$$\mathbb{E}[X] = \lim_{n \to \infty} \mathbb{E}[X_n] \geq \lim_{n \to \infty} \mathbb{E}[X_n \mathbf{1}_C] \geq \frac{2L}{p}\, \mathbb{P}(C) \geq 2L,$$

thereby leading to a contradiction.

(c) Construct an example of a non-negative random variable for which $\mathbb{P}(\{X < +\infty\}) = 1$, yet $\mathbb{E}[X] = +\infty$.
This exercise shows that $\mathbb{P}(\{X < +\infty\}) = 1$ does not imply $\mathbb{E}[X] < +\infty$.

Solution: We provide the solution to each of the parts below.

(a) We note that

$$\{X = +\infty\} = \bigcap_{n=1}^{\infty} \{X > n\}.$$

Because X is a random variable with respect to $\mathscr{F}$, it follows that $\{X > n\} \in \mathscr{F}$ for all $n \in \mathbb{N}$, and hence by the property that $\mathscr{F}$ is closed under countable intersections, it follows that $\{X = +\infty\} \in \mathscr{F}$.

(b) From part (a), we know that

$$\{X = +\infty\} = \bigcap_{n=1}^{\infty} \{X > n\}.$$

In particular, we note that $\{X = +\infty\} \subseteq \{X > n\}$ for all $n \in \mathbb{N}$, and therefore $\mathbb{P}(\{X > n\}) \geq \mathbb{P}(\{X = +\infty\})$ for all $n \in \mathbb{N}$. We now have

$$\mathbb{P}(\{X > 2L/p\}) \overset{(***)}{\geq} \mathbb{P}(\{X > \lceil 2L/p \rceil\}) \geq \mathbb{P}(\{X = +\infty\}) = p,$$

where the inequality in $(***)$ follows by noting that

$$\{X > \lceil 2L/p \rceil\} \subseteq \{X > 2L/p\}.$$

The rest of the arguments are already given in the question.

(c) Suppose that X has the PMF

$$p_X(n) = \begin{cases} \frac{1}{n(n+1)}, & n \in \mathbb{N}, \\ 0, & \text{otherwise}. \end{cases}$$

It is then clear that $\mathbb{P}(\{X < +\infty\}) = 1$. Furthermore,

$$\mathbb{E}[X] = \sum_{n=1}^{\infty} n\, p_X(n) = \sum_{n=1}^{\infty} \frac{1}{n+1} = +\infty.$$

7. A biased coin with heads probability $p \in (0, 1)$ is tossed repeatedly.
Let $X_n \in \{0, 1\}$ denote the outcome of the nth toss, $n \in \mathbb{N}$.
Let N be defined as the random variable

$$N := \min\{n \geq 2 : X_n = 1 - X_1\}.$$

That is, N is the first time index $n \geq 2$ for which the outcome X_n is the complement of the first outcome.

(a) Compute the PMF of N.

(b) Show that

$$\mathbb{E}[N] = \frac{p}{q} + \frac{q}{p},$$

where $q = 1 - p$.

Solution: We solve each of the parts below.

(a) Observe that

$$\{N = n\} = \begin{cases} \{X_2 = 1 - X_1\}, & n = 2, \\ \left(\bigcap_{m=2}^{n-1} \{X_m = X_1\}\right) \cap \{X_n = 1 - X_1\}, & n > 2. \end{cases}$$

Thus, it follows that

$$\mathbb{P}(\{N = n\}) = \begin{cases} \mathbb{P}(\{X_2 = 1 - X_1\}), & n = 2, \\ \mathbb{P}\left(\left(\bigcap_{m=2}^{n-1} \{X_m = X_1\}\right) \cap \{X_n = 1 - X_1\}\right), & n > 2. \end{cases}$$

We now note that

$$\begin{aligned} \mathbb{P}(\{X_2 = 1 - X_1\}) &= \mathbb{P}(\{X_2 = 1 - X_1\} \cap \{X_1 = 0\}) + \mathbb{P}(\{X_2 = 1 - X_1\} \cap \{X_1 = 1\}) \\ &= \mathbb{P}(\{X_2 = 1\} \cap \{X_1 = 0\}) + \mathbb{P}(\{X_2 = 0\} \cap \{X_1 = 1\}) \\ &\overset{(a)}{=} \mathbb{P}(\{X_2 = 1\}) \cdot \mathbb{P}(\{X_1 = 0\}) + \mathbb{P}(\{X_2 = 0\}) \cdot \mathbb{P}(\{X_1 = 1\}) \\ &= pq + qp \\ &= 2pq, \end{aligned}$$

where (a) above follows from the fact that $X_1 \perp\!\!\!\perp X_2$. Similarly, for any $n > 2$, we have

$$\begin{aligned} &\mathbb{P}\left(\left(\bigcap_{m=2}^{n-1} \{X_m = X_1\}\right) \cap \{X_n = 1 - X_1\}\right) \\ &= \mathbb{P}\left(\left(\bigcap_{m=2}^{n-1} \{X_m = X_1\}\right) \cap \{X_n = 1 - X_1\} \cap \{X_1 = 0\}\right) \\ &\qquad\qquad + \mathbb{P}\left(\left(\bigcap_{m=2}^{n-1} \{X_m = X_1\}\right) \cap \{X_n = 1 - X_1\} \cap \{X_1 = 1\}\right) \\ &= \mathbb{P}\left(\left(\bigcap_{m=2}^{n-1} \{X_m = 0\}\right) \cap \{X_n = 1\} \cap \{X_1 = 0\}\right) \end{aligned}$$

$$
\begin{aligned}
&\qquad\qquad\qquad\qquad + \mathbb{P}\left(\left(\bigcap_{m=2}^{n-1}\{X_m = 1\}\right) \cap \{X_n = 0\} \cap \{X_1 = 1\}\right) \\
&= \prod_{m=2}^{n-1} \mathbb{P}(\{X_m = 0\}) \cdot \mathbb{P}(\{X_n = 1\}) \cdot \mathbb{P}(X_1 = 0\}) + \prod_{m=2}^{n-1} \mathbb{P}(\{X_m = 1\}) \cdot \mathbb{P}(\{X_n = 0\}) \cdot \mathbb{P}(X_1 = 1\}) \\
&= q^{n-1}p + p^{n-1}q,
\end{aligned}
$$

where the penultimate line above follows from the fact that $X_1, \ldots, X_n \overset{\text{i.i.d.}}{\sim}$ Ber(p).
Combining the above results, we see that

$$
p_N(n) = \begin{cases} q^{n-1}p + p^{n-1}q, & n \in \mathbb{N},\ n \geq 2, \\ 0, & \text{otherwise.} \end{cases}
$$

(b) We have

$$
\mathbb{E}[N] = \sum_{n=2}^{\infty} n\left(q^{n-1}p + p^{n-1}q\right) = \frac{p}{q} + \frac{q}{p},
$$

where in writing the last equality, we make use of the fact that the mean of a Geometric distribution with parameter $p \in (0, 1)$ is $1/p$.