

1.  $X(t)$  is wide-sense stationary if and only if

$$E[X(t)] \text{ is constant wrt } t, \quad \text{Cov}(X(t), X(t+h)) = \text{Cov}(X(0), X(h)) \quad \forall t, h \in \mathbb{R}_+$$

Observe that

$$E[X(t)] = E[X_1] \cos(2\pi f_c t) + E[X_2] \sin(2\pi f_c t)$$

In particular,

$$E[X(0)] = E[X_1], \quad E[X(\frac{1}{f_c})] = E[X_2].$$

It then follows that

$$E[X(t)] \text{ is constant wrt } t \iff E[X_1] = 0 = E[X_2]. \rightarrow \textcircled{1}$$

Furthermore, for any  $h \in \mathbb{R}_+$ ,  $\underbrace{0}_{\text{mean}} \quad \underbrace{0}_{\text{mean}}$

$$\begin{aligned} \text{Cov}(X(0), X(h)) &= \text{Cov}(X_1, X_1 \cos(2\pi f_c h) + X_2 \sin(2\pi f_c h)) \\ &= E[X_1 \cdot (X_1 \cos(2\pi f_c h) + X_2 \sin(2\pi f_c h))] \\ &= E[X_1^2] \cos(2\pi f_c h) + E[X_1 X_2] \sin(2\pi f_c h), \end{aligned}$$

whereas for  $t, h \in \mathbb{R}_+$ ,

$$\begin{aligned} \text{Cov}(X(t), X(t+h)) &= E[X(t) X(t+h)] \quad (\text{because } E[X(t)] = 0 = E[X(t+h)]) \\ &= E[X_1^2] \cos(2\pi f_c t) \cos(2\pi f_c t + 2\pi f_c h) \\ &\quad + E[X_1 X_2] (\cos(2\pi f_c t) \sin(2\pi f_c t + 2\pi f_c h) \\ &\quad \quad \quad + \sin(2\pi f_c t) \cos(2\pi f_c t + 2\pi f_c h)) \\ &\quad + E[X_2^2] \sin(2\pi f_c t) \sin(2\pi f_c t + 2\pi f_c h) \end{aligned}$$

$$\begin{aligned} &= \underbrace{E[X_1^2] \cos(2\pi f_c h)}_{\text{function of } h \text{ only}} \\ &\quad \text{add and subtract } E[X_1 X_2] \sin(4\pi f_c t + 2\pi f_c h) \\ E[X_1^2] \sin(2\pi f_c t) \sin(2\pi f_c(t+h)); &\quad + (E[X_2^2] - E[X_1^2]) \sin(2\pi f_c t) \sin(2\pi f_c t + 2\pi f_c h) \\ &\quad \text{use identities} \\ \cos(A-B) &= \cos A \cos B + \sin A \sin B \\ \sin(A+B) &= \sin A \cos B + \cos A \sin B \end{aligned}$$

We then see that  $E[X(t) X(t+h)]$  will be independent of  $t$

if and only if

$$\mathbb{E}[X_1 X_2] = 0, \quad \mathbb{E}[X_1^2] = \mathbb{E}[X_2^2]. \quad \rightarrow (2)$$

Thus,  $\{X(t) : t \geq 0\}$  is wide-sense stationary if and only if the conditions in (1) and (2) are satisfied, i.e.,

- a)  $\mathbb{E}[X_1] = 0 = \mathbb{E}[X_2]$ .
- b)  $\mathbb{E}[X_1 X_2] = 0$ .
- c)  $\mathbb{E}[X_1^2] = \mathbb{E}[X_2^2]$ .

2. We note that

$$\mathbb{E}[X_n] = \begin{cases} \mathbb{E}[U_n] = 0, & n \text{ odd} \\ \frac{1}{\sqrt{2}} (\mathbb{E}[U_n^2] - 1) = 0, & n \text{ even.} \end{cases}$$

Thus,  $\mathbb{E}[X_n] = 0 \quad \forall n \in \mathbb{N}$ . Furthermore, because  $\{U_n\}_{n=1}^{2^\infty}$  is i.i.d.,

it follows that  $\mathbb{E}[X_m X_n] = 1_{\{m=n\}}$ ,

thus proving that  $\{X_n\}_{n=1}^{2^\infty}$  is wide-sense stationary.

To show that  $\{X_n\}_{n=1}^{2^\infty}$  is not stationary, it suffices to show

that the CDFs of  $X_1$  and  $X_2$  are different. Indeed,

$$\mathbb{P}(X_1 \geq 0) = \mathbb{P}(U_1 \geq 0) = \frac{1}{2}, \quad \text{while}$$

$$\begin{aligned} \mathbb{P}(X_2 \geq 0) &= \mathbb{P}(U_2^2 \geq \sqrt{2}) = \mathbb{P}(U_1^2 \geq \sqrt{2}) \\ &\neq \mathbb{P}(U_1^2 \geq 0). \end{aligned}$$

3.

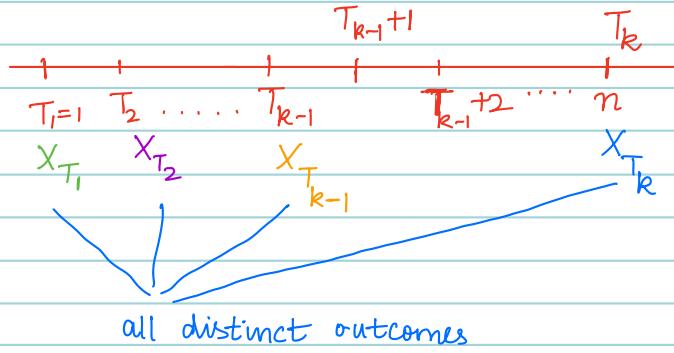
a)  $T_k$  represents the first instant of observing the  $k^{\text{th}}$  distinct outcome.

b) Clearly,  $T_1 = 1$  is a stopping time, as

$$\{T_1 = n\} = \begin{cases} \Omega, & n=1, \\ \emptyset, & n \neq 1, \end{cases}$$

and therefore  $\{T_1 = n\} \in \sigma(X_1, \dots, X_n) \quad \forall n \in \mathbb{N}$ .

Assume, by induction, that  $T_{k-1}$  is a stopping time for some  $k \geq 1$ .



$$\begin{aligned} \{T_k = n\} &= \left\{ X_{T_{k-1}+1} \in \{X_{T_1}, \dots, X_{T_{k-1}}\}, \dots, X_{n-1} \in \{X_{T_1}, \dots, X_{T_{k-1}}\}, \right. \\ &\quad \left. X_n \notin \{X_{T_1}, \dots, X_{T_{k-1}}\} \right\} \\ &= \bigcup_{m \leq n} \left\{ T_{k-1} = m, X_{m+1} \in \{X_1, \dots, X_m\}, \dots, X_{n-1} \in \{X_1, \dots, X_m\}, \right. \\ &\quad \left. X_n \notin \{X_1, \dots, X_m\} \right\} \end{aligned}$$

$$\in \sigma(X_1, \dots, X_n) \quad \forall n \in \mathbb{N}.$$

Furthermore, we have  $\mathbb{P}(T_1 < +\infty) = 1$  by definition of  $T_1$ .

Assume, by induction, that  $\mathbb{P}(T_{k-1} < +\infty) = 1$  for some  $k \geq 1$ .

Then,

$$T_k = T_k - T_{k-1} + T_{k-1} + T_{k-2} + \dots + T_2 - T_1 + T_1$$

all finite

Thus,  $T_k < +\infty$  will occur if and only if  $T_k - T_{k-1} < +\infty$  occurs.

$$\Rightarrow \{T_k < +\infty\} = \{T_k - T_{k-1} < +\infty\}.$$

$$\Rightarrow \mathbb{P}(T_k < +\infty) = \mathbb{P}(T_k - T_{k-1} < +\infty)$$

$$= \sum_{n \in \mathbb{N}} \mathbb{P}(T_k - T_{k-1} = n)$$

$$= \sum_{n \in \mathbb{N}} \left( \frac{k-1}{K} \right)^{n-1} \left( 1 - \frac{k-1}{K} \right)$$

for  $(n-1)$  time instants, one of the previously seen  $(k-1)$  outcomes must occur, & at  $n^{\text{th}}$  instant, the  $k^{\text{th}}$  distinct outcome must occur

$$= 1.$$

Thus,  $T_k$  is a stopping time wrt  $\{X_n\}_{n=1}^{+\infty}$  for all  $k \in \{1, \dots, 6\}$ .

c) As already seen above,

$$S_k \sim \text{Geometric} \left( 1 - \frac{k-1}{K} \right), \quad k \in \{2, \dots, 6\}.$$

d) We have

$$\begin{aligned}
 E[T_k] &= E[T_k - T_{k-1}] + E[T_{k-1} - T_{k-2}] + \dots + E[T_2 - T_1] + E[T_1] \\
 &= 1 + \sum_{i=2}^k E[S_i] \\
 &= 1 + \sum_{i=2}^k \frac{1}{1 - \frac{i-1}{K}} \quad (E[X] = \frac{1}{P} \text{ if } X \sim \text{Geo}(P)) \\
 &= 1 + \sum_{i=1}^{k-1} \frac{1}{1-i} \frac{1}{K} \\
 &= 1 + \sum_{i=1}^{k-1} \frac{K}{K-i}, \quad k \in \{2, \dots, 6\}.
 \end{aligned}$$

4.

a) Observe that

$$\Omega = X^{-1}(\mathbb{R}), \text{ and } \mathbb{R} \in \mathcal{B}(\mathbb{R}).$$

$$\Rightarrow \Omega \in \sigma(X).$$

Next, suppose that  $A \in \sigma(X)$ .

$$\Rightarrow \exists B \in \mathcal{B}(\mathbb{R}) \text{ such that } A = X^{-1}(B).$$

$$\Rightarrow A = \{w \in \Omega : X(w) \in B\}$$

We then have

$$\begin{aligned}
 A^c &= \Omega \setminus A \\
 &= \{w \in \Omega : X(w) \notin B\} \\
 &= \{w \in \Omega : X(w) \in \mathbb{R} \setminus B\} \\
 &= X^{-1}(\mathbb{R} \setminus B), \text{ and } \mathbb{R} \setminus B \in \mathcal{B}(\mathbb{R}) \text{ as } \mathbb{R}, B \in \mathcal{B}(\mathbb{R}).
 \end{aligned}$$



$$\Rightarrow A^c \in \sigma(X).$$

Finally, if  $A_1, A_2, \dots \in \sigma(X)$ , then  $\exists B_1, B_2, \dots \in \mathcal{B}(\mathbb{R})$  such that

$$A_i = X^{-1}(B_i) \quad \forall i \in \mathbb{N}$$

$$\Rightarrow \bigcup_{i \in \mathbb{N}} A_i = \bigcup_{i \in \mathbb{N}} X^{-1}(B_i)$$

$$= X^{-1}\left(\bigcup_{i \in \mathbb{N}} B_i\right), \text{ and } \bigcup_{i \in \mathbb{N}} B_i \in \mathcal{B}(\mathbb{R})$$

$$\Rightarrow \bigcup_{i \in \mathbb{N}} A_i \in \sigma(X).$$

Thus,  $\sigma(X)$  is a  $\sigma$ -algebra of subsets of  $\Omega$ .

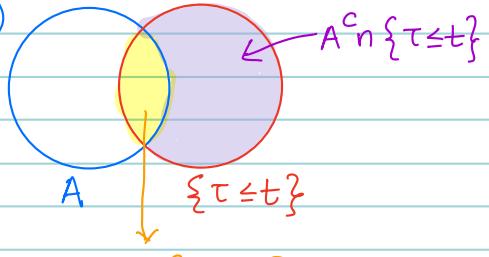
b) First, we observe that  $\Omega \in \mathcal{F}_\tau$ , as

$$\Omega \cap \{\tau \leq t\} = \{\tau \leq t\} \in \mathcal{F}_t \quad \forall t \in \mathbb{T}, \text{ by the fact that } \tau \text{ is a stopping time.}$$

Next, suppose that  $A \in \mathcal{F}_\tau$ . That is,

$$A \cap \{\tau \leq t\} \in \mathcal{F}_t \quad \forall t \in \mathbb{T}.$$

Then, for any  $t \in \mathbb{T}$ ,

$$A^c \cap \{\tau \leq t\} = \underbrace{\{\tau \leq t\}}_{\in \mathcal{F}_t} \setminus \underbrace{(A \cap \{\tau \leq t\})}_{\in \mathcal{F}_t} \in \mathcal{F}_t.$$


$$\Rightarrow A^c \cap \{\tau \leq t\} \in \mathcal{F}_t \quad \forall t \in \mathbb{T}$$

$$\Rightarrow A^c \in \mathcal{F}_\tau.$$

Finally, suppose that  $A_1, A_2, \dots \in \mathcal{F}_\tau$

$$\Rightarrow A_1 \cap \{\tau \leq t\} \in \mathcal{F}_t \quad \forall t \in \mathbb{T}$$

$$A_2 \cap \{\tau \leq t\} \in \mathcal{F}_t \quad \forall t \in \mathbb{T}$$

$\vdots$

Then, for any  $t \in \mathbb{T}$ ,

$$\left( \bigcup_{i \in \mathbb{N}} A_i \right) \cap \{\tau \leq t\} = \bigcup_{i \in \mathbb{N}} (A_i \cap \{\tau \leq t\}) \in \mathcal{F}_t.$$

de-Morgan's law

$$\Rightarrow \left( \bigcup_{i \in \mathbb{N}} A_i \right) \cap \{\tau \leq t\} \in \mathcal{F}_t \quad \forall t \in \mathbb{T}$$

$$\Rightarrow \bigcup_{i \in \mathbb{N}} A_i \in \mathcal{F}_\tau.$$

Thus,  $\mathcal{F}_\tau$  is a  $\sigma$ -algebra of subsets of  $\Omega$ .

5.

a) Observe that

$$\begin{aligned} \{N > n\} &= \{ \text{no two consecutive ones are observed in } X_1, \dots, X_n \} \\ &= \{X_1 X_2 = 0, X_2 X_3 = 0, \dots, X_{n-1} X_n = 0\} \quad (\text{each } X_i \in \{0, 1\}) \\ &= \{X_1 X_2 + X_2 X_3 + \dots + X_{n-1} X_n = 0\} \\ &\in \sigma(X_1, \dots, X_n). \quad \forall n \in \mathbb{N} \\ \Rightarrow \{N \leq n\} &\in \sigma(X_1, \dots, X_n) \quad \forall n \in \mathbb{N}. \end{aligned}$$

Furthermore, let

$$a_n := P(N > n) = P(X_1 X_2 + \dots + X_{n-1} X_n = 0).$$

Now,

$$\begin{aligned} \{N > n\} &\subseteq \{N > n-1\} \quad \forall n \in \mathbb{N} \\ \Rightarrow \bigcap_{n \in \mathbb{N}} \{N > n\} &= \lim_{n \rightarrow \infty} \{N > n\} = \{N = +\infty\}. \end{aligned}$$

$$\Rightarrow P(N = +\infty) = \lim_{n \rightarrow \infty} P(N > n) = \lim_{n \rightarrow \infty} a_n.$$

*Continuity  
of probability*

We now show  $\lim_{n \rightarrow \infty} a_n = 0$ , thereby proving that  $P(N < +\infty) = 1$ .

We have

$$\begin{aligned} a_n &= P(X_1 X_2 + \dots + X_{n-1} X_n = 0) \\ &= P(X_1 X_2 + \dots + X_{n-1} X_n = 0, X_1 = 0) + \\ &\quad \text{Law of total prob.} \quad P(X_1 X_2 + \dots + X_{n-1} X_n = 0, X_1 = 1) \\ &= P(X_2 X_3 + \dots + X_{n-1} X_n = 0, X_1 = 0) \end{aligned}$$

$$+ \mathbb{P}(X_2 + X_2 X_3 + \dots + X_{n-1} X_n = 0, X_1 = 1)$$

independence  
of  $X_i$ 's

$$= \mathbb{P}(X_1 = 0) \cdot \mathbb{P}(X_2 X_3 + \dots + X_{n-1} X_n = 0)$$

$$+ \mathbb{P}(X_1 = 1, X_2 = 0, X_3 X_4 + \dots + X_{n-1} X_n = 0)$$

independence  
of  $X_i$ 's

$$= \frac{1}{2} \cdot \mathbb{P}(\text{no 2 consecutive ones in } X_2, \dots, X_n)$$

$$+ \mathbb{P}(X_1 = 1) \cdot \mathbb{P}(X_2 = 0) \cdot \mathbb{P}(\text{no 2 consecutive ones in } X_3, \dots, X_n)$$

Noting that

identically distributed as

$$(X_2, \dots, X_n) \stackrel{d}{=} (X_1, \dots, X_{n-1}), \quad \left. \begin{array}{l} \text{stationarity of } \{X_i\}_{i \in \mathbb{N}} \\ (X_3, \dots, X_n) \stackrel{d}{=} (X_1, \dots, X_{n-2}). \end{array} \right\} \text{courtesy of iid nature of } X_i \text{'s.}$$

it follows that

$$\mathbb{P}(X_2 X_3 + \dots + X_{n-1} X_n = 0) = \mathbb{P}(X_1 X_2 + \dots + X_{n-2} X_{n-1} = 0) = a_{n-1},$$

$$\mathbb{P}(X_3 X_4 + \dots + X_{n-1} X_n = 0) = \mathbb{P}(X_1 X_2 + \dots + X_{n-3} X_{n-2} = 0) = a_{n-2}.$$

Thus, we get

$$a_n = \frac{1}{2} a_{n-1} + \frac{1}{4} a_{n-2}.$$

Taking limits as  $n \rightarrow \infty$  on both sides, we get

$$\lim_{n \rightarrow \infty} a_n = \frac{1}{2} \lim_{n \rightarrow \infty} a_{n-1} + \frac{1}{4} \lim_{n \rightarrow \infty} a_{n-2}$$

$$\Rightarrow \lim_{n \rightarrow \infty} a_n = 0.$$

This proves that  $N$  is a stopping time w.r.t  $\{X_n\}_{n \in \mathbb{N}}$ .

b) We have

$$\mathbb{P}(X_{N+1} = 0, X_{N+2} = 0) \stackrel{\text{because } \mathbb{P}(N < \infty) = 1}{=} \mathbb{P}(X_{N+1} = 0, X_{N+2} = 0, N < +\infty)$$

Law of  
total  
prob.

$$\stackrel{=}{\leftarrow} \sum_{n \in \mathbb{N}} \mathbb{P}(X_{n+1} = 0, X_{n+2} = 0, \underbrace{N=n}_{\in \sigma(X_1, \dots, X_n)})$$

N is a stopping  
time and  $X_i$ 's are indep.

$$\stackrel{=}{\leftarrow} \sum_{n \in \mathbb{N}} \mathbb{P}(X_{n+1} = 0) \cdot \mathbb{P}(X_{n+2} = 0) \cdot \mathbb{P}(N = n)$$

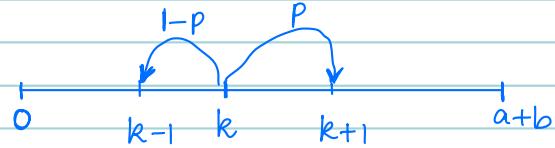
$x_i$ 's are i.d.

$$= \frac{1}{4} \sum_{n \in \mathbb{N}} \mathbb{P}(N=n)$$

$$= \frac{1}{4} \mathbb{P}(N < +\infty) = \frac{1}{4}.$$

6.

a)  $a[0] = 0, a[a+b] = 1.$



(b) Suppose A has \$k at some time n.

Let  $X_n^A$  denote this value, i.e.,  $X_n^A = k$ .

$\xleftarrow{\text{direction of B winning}}$   $\xrightarrow{\text{direction of A winning}}$

$$a[k] = \mathbb{P}(A \text{ wins} \mid X_n^A = k)$$

$$= \mathbb{P}(A \text{ wins}, X_{n+1}^A = k+1 \mid X_n^A = k)$$

$$+ \mathbb{P}(A \text{ wins}, X_{n+1}^A = k-1 \mid X_n^A = k)$$

$$= \mathbb{P}(X_{n+1}^A = k+1 \mid X_n^A = k) \cdot \mathbb{P}(A \text{ wins} \mid X_{n+1}^A = k+1, X_n^A = k)$$

$$+ \mathbb{P}(X_{n+1}^A = k-1 \mid X_n^A = k) - \mathbb{P}(A \text{ wins} \mid X_{n+1}^A = k-1, X_n^A = k)$$

$$= p \cdot \mathbb{P}(A \text{ wins} \mid X_{n+1}^A = k+1) + (1-p) \mathbb{P}(A \text{ wins} \mid X_{n+1}^A = k-1)$$

once we know  $X_{n+1}^A$ ,  
we can discard  $X_n^A$

$$= p a[k+1] + (1-p) a[k-1]$$

(c) Rearranging terms, we get

$$p(a[k+1] - a[k]) = (1-p)(a[k] - a[k-1])$$

$$\Rightarrow a[k+1] - a[k] = \left(\frac{1-p}{p}\right) (a[k] - a[k-1])$$

$$= \left(\frac{1-p}{p}\right)^2 (a[k-1] - a[k-2])$$

$$= \left(\frac{1-p}{p}\right)^k (a[1] - a[0])$$

Using the results of part (a), we get

$$a[2] - a[1] = \left(\frac{1-p}{p}\right) (a[1] - a[0])$$

$$a[3] - a[2] = \left(\frac{1-p}{p}\right)^2 (a[1] - a[0])$$

$$\begin{aligned}
 a[k] &= (a[k] - a[k-1]) + a[k-1] - a[k-2] + \dots + a[2] - a[1] \\
 &\quad + a[1] - a[0] \\
 &+ a[0] \\
 &= \sum_{i=1}^k (a[i] - a[i-1]) \\
 &= \sum_{i=1}^k \left(\frac{1-p}{p}\right)^{i-1} (a[1] - a[0]) \\
 &= a[1] \cdot \sum_{i=0}^{k-1} \left(\frac{1-p}{p}\right)^i
 \end{aligned}$$

i) If  $p = \frac{1}{2}$ , then

$$a[k] = k a[1].$$

Using  $a[a+b] = 1$ , we get  $a[1] = \frac{1}{a+b}$ .

$$\Rightarrow a[k] = \frac{k}{a+b}, \quad k \in \{0, \dots, a+b\}.$$

ii) If  $p > \frac{1}{2}$ , then  $\frac{1-p}{p} < 1$ , and

$$\begin{aligned}
 a[k] &= a[1] \cdot \frac{1 - \left(\frac{1-p}{p}\right)^k}{1 - \left(\frac{1-p}{p}\right)}.
 \end{aligned}$$

Using  $a[a+b] = 1$ , we get

$$a[1] = \frac{1 - \left(\frac{1-p}{p}\right)}{1 - \left(\frac{1-p}{p}\right)^{a+b}}, \quad \text{and hence}$$

$$\begin{aligned}
 a[k] &= \frac{1 - \left(\frac{1-p}{p}\right)^k}{1 - \left(\frac{1-p}{p}\right)^{a+b}}, \quad k \in \{0, \dots, a+b\}.
 \end{aligned}$$

(iii) If  $p < \frac{1}{2}$ , then  $\frac{1-p}{p} > 1$ , and it is easy to show that

$$\begin{aligned}
 a[k] &= \frac{\left(\frac{1-p}{p}\right)^k - 1}{\left(\frac{1-p}{p}\right)^{a+b} - 1}, \quad k \in \{0, \dots, a+b\}.
 \end{aligned}$$

$$d) \quad P(A \text{ ruins } B) = a[a] = \begin{cases} \frac{a}{a+b}, & p = \frac{1}{2}, \\ \frac{1 - \left(\frac{1-p}{p}\right)^a}{1 - \left(\frac{1-p}{p}\right)^{a+b}}, & p > \frac{1}{2}, \\ \frac{\left(\frac{1-p}{p}\right)^a - 1}{\left(\frac{1-p}{p}\right)^{a+b} - 1}, & p < \frac{1}{2}. \end{cases}$$

7.

a) Observe that for any  $n \in \mathbb{N}$ ,

$$\{N > n\} = \{X_1 \geq X_2 \geq \dots \geq X_n\} \in \sigma(X_1, \dots, X_n).$$

Furthermore,

$$\begin{aligned} P(N > n) &= P(X_1 \geq X_2 \geq \dots \geq X_n) \\ &= \int_{\substack{1 \\ 0}}^{\infty} \int_{\substack{x_1 \\ 0}}^{\infty} \int_{\substack{x_2 \\ 0}}^{\infty} \dots \int_{\substack{x_{n-1} \\ 0}}^{\infty} dx_n \dots dx_2 dx_1, \\ &\quad \text{limits for } x_1 \quad \text{limits for } x_2 \quad \text{limits for } x_n \\ &= \frac{1}{n!}. \end{aligned}$$

Thus, we have  $\lim_{n \rightarrow \infty} P(N > n) = 0$ , from which it follows that

$$\begin{aligned} P(N < +\infty) &= 1 - P(N = +\infty) \\ &= 1 - P\left(\bigcap_{n \in \mathbb{N}} \{N > n\}\right) \end{aligned}$$

$$\xleftarrow{\text{Continuity of probability, observing that}} = 1 - \lim_{n \rightarrow \infty} P(N > n)$$

observing that  $= 1$ .

$$\{N > n\} \subseteq \{N > n-1\}$$

$\forall n \in \mathbb{N}$

Thus,  $N$  is a stopping time wrt  $\{X_n\}_{n \in \mathbb{N}}$ .

b) We have

$$\begin{aligned}\mathbb{E}[N] &= \sum_{n \in \mathbb{N}} n \cdot \mathbb{P}(N=n) + (+\infty) \underbrace{\mathbb{P}(N=+\infty)}_{=0} \\ &= \sum_{n=2}^{\infty} n \cdot \mathbb{P}(N=n) \quad (\text{because } \mathbb{P}(N=1)=0 \text{ by defn}) \\ &= \sum_{n=2}^{\infty} n \left( \mathbb{P}(N>n-1) - \mathbb{P}(N>n) \right) \\ &= \sum_{n=2}^{\infty} n \left( \frac{1}{(n-1)!} - \frac{1}{n!} \right) \\ &= \sum_{n=2}^{\infty} \frac{1}{(n-2)!} \\ &= \sum_{m=0}^{\infty} \frac{1}{m!} = e. \quad \left( e^x = 1+x+\frac{x^2}{2!}+\frac{x^3}{3!}+\dots \right)\end{aligned}$$

Using Wald's Lemma,

$$\begin{aligned}\mathbb{E}\left[\sum_{i=1}^N X_i\right] &= \mathbb{E}[N] \cdot \mathbb{E}[X_1] \\ &= \frac{e}{2}.\end{aligned}$$