

PART I

1 Given a set Ω , a collection \mathcal{F} of subsets of Ω is a σ -algebra on Ω if

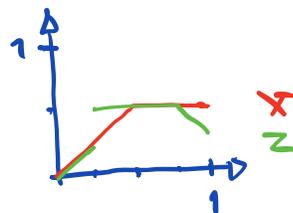
- $\Omega \in \mathcal{F}$
 - $A \in \mathcal{F} \Rightarrow A^c \in \mathcal{F}$
 - $A_1, A_2, \dots \in \mathcal{F} \Rightarrow \bigcup_{k=1}^{\infty} A_k \in \mathcal{F}$.
- 2p

\mathcal{F}_2 and \mathcal{F}_4 both satisfy these conditions, and are σ -algebras on $\Omega = \{1, 2, 3, 4, 5, 6\}$.

\mathcal{F}_1 does not contain Ω and \mathcal{F}_3 does not contain e.g. $\{1\} \cup \{2\} = \{1, 2\}$. These are then not σ -algebras on Ω . 3p

2 We first draw a figure.

It is clear from the figure and the definition that



$$\{X=Z\} = [0, 1/4) \cup [1/2, 3/4].$$

$$\mathbb{P}(X=Z) = \mathbb{P}([0, 1/4)) + \mathbb{P}([1/2, 3/4]) = \frac{1}{4} + \frac{1}{4} = \frac{1}{2} \neq 1.$$

2p

We also have

$$\{X \leq x\} = \begin{cases} [0, x] & x < 1/2 \\ [0, 1] & x \geq 1/2 \end{cases}$$

$$\mathbb{P}(X \leq x) = \begin{cases} x & 0 \leq x < 1/2 \\ 1 & x \geq 1/2 \end{cases}$$

and

$$\{Z \leq x\} = \begin{cases} [0, x] & x < 1/4 \\ [0, x] \cup [5/4-x, 1] & 1/4 \leq x < 1/2 \\ [0, 1] & x \geq 1/2 \end{cases}$$

$$\mathbb{P}(Z \leq x) = \begin{cases} x & x < 1/4 \\ \frac{1}{4} + (x - \frac{1}{4}) = x & 1/4 \leq x < 1/2 \\ 1 & x \geq 1/2 \end{cases}$$

Hence $F_X = F_Z$ so X and Z have the same distribution. 3p

3

Let X_1, X_2 be iid zero mean, variance one.
 Set $S_n = X_1 + X_2 + \dots + X_n$. Then $\mathbb{E}[S_n] = 0$, so
 by Chebyshev's inequality

$$\mathbb{P}(|S_n| \geq \varepsilon \sqrt{n \log \log n}) \leq \frac{\text{Var}(S_n)}{\varepsilon^2 n \log \log n}. \quad 2p$$

This holds $\forall \varepsilon > 0$. Since iid variance one,

$$\text{Var}(S_n) = \sum_{k=1}^n \text{Var}(X_k) = n.$$

This gives

$$\mathbb{P}\left(\frac{|S_n|}{\sqrt{n \log \log n}} \geq \varepsilon\right) \leq \frac{1}{\varepsilon^2 \log \log n} \rightarrow 0 \quad 3p$$

as $n \rightarrow \infty$. Since true for all $\varepsilon > 0$, this shows conv. in probab.

4

Let X_1, X_2, \dots be iid zero mean, variance one.
 Set $S_n = X_1 + \dots + X_n$ and $\mathcal{F}_n = \sigma(X_1, \dots, X_n)$.

Clearly both S_n and $S_n^2 - n$ are \mathcal{F}_n -measurable since
 functions of X_1, \dots, X_n . Also

$$\mathbb{E}[|S_n|] \leq \sum_{k=1}^n \mathbb{E}[|X_k|] \leq \sum_{k=1}^n \mathbb{E}[X_k^2]^{1/2} = n.$$

$$\mathbb{E}[|S_n^2 - n|] \leq \mathbb{E}[S_n^2] + n = \sum_{k=1}^n \mathbb{E}[X_k^2] + n = 2n. \quad 2p$$

Finally we verify that the two sequences satisfy the
 martingale property.

$$\mathbb{E}[S_{n+1} | \mathcal{F}_n] = S_n + \mathbb{E}[X_{n+1} | \mathcal{F}_n] = S_n.$$

$$\begin{aligned} \mathbb{E}[S_{n+1}^2 - (n+1) | \mathcal{F}_n] &= \mathbb{E}[(S_n + X_{n+1})^2 - (n+1) | \mathcal{F}_n] \\ &= \mathbb{E}[S_n^2 + 2S_n \cdot X_{n+1} + X_{n+1}^2 | \mathcal{F}_n] - (n+1) \\ &= S_n^2 + 2S_n \underbrace{\mathbb{E}[X_{n+1} | \mathcal{F}_n]}_{=0} + \underbrace{\mathbb{E}[X_{n+1}^2 | \mathcal{F}_n]}_{=1} - (n+1) \\ &= S_n^2 - n. \end{aligned}$$

so $(S_n)_{n \geq 0}$ and $(S_n^2 - n)_{n \geq 0}$ are martingales
 w.r.t $(\mathcal{F}_n)_{n \geq 0}$. 3p

PART II

5 Let $(X_n)_{n \geq 1}$ be a square-integrable martingale w.r.t $(\mathcal{F}_n)_{n \geq 1}$ where $\mathcal{F}_n = \sigma(X_1, \dots, X_n)$.

(a) We have for $n \geq 1$ that

$$\mathbb{E}[X_{n+1} - X_n] = \mathbb{E}[\underbrace{\mathbb{E}[X_{n+1} - X_n | \mathcal{F}_n]}_{=0 \text{ since martingale}}] = 0. \quad 3p$$

(b) Suppose $i > j$. Then

$$\begin{aligned} & \text{Cov}(X_{i+1} - X_i, X_{j+1} - X_j) \\ & \mathbb{E}[X_{i+1} - X_i] = 0 \quad \text{from (a)} \\ & \mathbb{E}[(X_{i+1} - X_i)(X_{j+1} - X_j)] \\ & = \mathbb{E}[(X_{j+1} - X_j) \underbrace{\mathbb{E}[X_{i+1} - X_i | \mathcal{F}_i]}_{=0}] \\ & \stackrel{i \geq j+1}{\rightarrow} = 0 \quad 3p \end{aligned}$$

(c) For $n \geq 1$ we have

$$\begin{aligned} X_{n+1}^2 &= (X_n + (X_{n+1} - X_n))^2 \\ &= X_n^2 + \underbrace{(X_{n+1} - X_n)^2}_{\geq 0} + 2X_n(X_{n+1} - X_n). \end{aligned}$$

This gives

$$\begin{aligned} \mathbb{E}[X_{n+1}^2] &\geq \mathbb{E}[X_n^2] + 2\mathbb{E}[X_n(X_{n+1} - X_n)] \\ &= \mathbb{E}[X_n^2] + 2\mathbb{E}[X_n \underbrace{\mathbb{E}[X_{n+1} - X_n | \mathcal{F}_n]}_{=0}] \\ &= \mathbb{E}[X_n^2] \end{aligned}$$

Hence $\mathbb{E}[X_{n+1}^2] \geq \mathbb{E}[X_n^2]$ for all $n \geq 1$. 4p

6 Let X_1, X_2, \dots be independent $\exp(1)$ -distributed, and set $A_k = \{X_k > \log k\}$. Note that

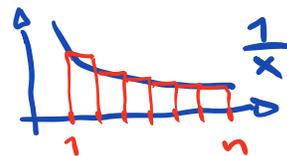
$$N_n = \sum_{k=1}^n \mathbb{1}_{A_k} \text{ and } N = \sum_{k=1}^{\infty} \mathbb{1}_{A_k}.$$

(a) Clearly $N_n \leq N$ for all $n \geq 1$. Hence, for all $n \geq 1$

$$\mathbb{E}[N] \geq \mathbb{E}[N_n] = \sum_{k=1}^n \mathbb{P}(A_k) = \sum_{k=1}^n \frac{1}{k}.$$

Since

$$\sum_{k=1}^n \frac{1}{k} \geq \int_1^n \frac{1}{x} dx = \log n \nearrow \infty$$



as $n \rightarrow \infty$, the claim follows. 3p

(b) The above computation and Borel-Cantelli's second lemma give (events independent)

$$\mathbb{P}(N = \infty) = \mathbb{P}(A_k \text{ i.o.}) = 1. \quad \text{3p}$$

(c) Chebyshev's inequality gives

$$\mathbb{P}(|N_n - \mathbb{E}[N_n]| > \varepsilon \mathbb{E}[N_n]) \leq \frac{\text{Var}(N_n)}{\varepsilon^2 \mathbb{E}[N_n]^2}. \quad \text{2p}$$

Moreover, due to independence

$$\text{Var}(N_n) = \sum_{k=1}^n \text{Var}(\mathbb{1}_{A_k}) \leq \sum_{k=1}^n \mathbb{P}(A_k) = \mathbb{E}[N_n].$$

From (a) we conclude that, as $n \rightarrow \infty$,

$$\mathbb{P}(|N_n - \mathbb{E}[N_n]| > \varepsilon \mathbb{E}[N_n]) \leq \frac{1}{\varepsilon^2 \mathbb{E}[N_n]} \leq \frac{1}{\varepsilon^2 \log n} \rightarrow 0.$$

This shows that $\frac{N_n}{\mathbb{E}[N_n]} \xrightarrow{P} 1$. 2p

7

$\Omega = \{0,1\}^4$, \mathbb{F} the power set of Ω and \mathbb{P} the uniform measure on Ω .

(a) The random variable X takes the values 0, 1, 2, 3, 4. For $k=0,1,\dots,4$ let

$$A_k = \{\omega : X(\omega) = k\}.$$

Then $\mathcal{F} = \{A_0, A_1, A_2, A_3, A_4\}$ is a partition of Ω . $\sigma(X)$ contains all sets that can be written as a union of elements in \mathcal{F} . There are $2^5 = 32$ sets in $\sigma(X)$. 5p

(b) X is a simple rv since it only takes finitely many values. By definition we have

$$\mathbb{E}[X] = \sum_{k=0}^4 k \cdot \mathbb{P}(A_k).$$

Since \mathbb{P} is the uniform measure, we have

$$\mathbb{P}(A_k) = \frac{|A_k|}{|\Omega|} = \frac{\binom{4}{k}}{2^4}.$$

Hence

$$\begin{aligned} \mathbb{E}[X] &= \frac{1}{2^4} (0 + 1 \cdot 4 + 2 \cdot 6 + 3 \cdot 6 + 4 \cdot 1) \\ &= \frac{38}{16} = \frac{19}{8}. \end{aligned}$$

8

(a) Clearly S_n is \mathcal{F}_n -measurable, since a function of X_1, \dots, X_n . Hence T and T' are stopping times since

$$\{T = n\} = \{S_1 \neq \pm m\} \cap \dots \cap \{S_{n-1} \neq \pm m\} \cap \{S_n = \pm m\}.$$

$$\{T' \leq n\} = \{S_1 \neq m\} \cap \dots \cap \{S_{n-1} \neq m\} \cap \{S_n = m\},$$

and these are events in \mathcal{F}_n . 3p

(b) Let us first show that $\mathbb{E}[T] < \infty$. For $k \geq 1$ let

$$A_k = \{X_k = X_{k+1} = \dots = X_{k+2m} = 1\},$$

and note that A_k implies that starting at step k , the sequence $(S_n)_{n \geq 1}$ takes $2m+1$ steps up in a row. Hence if $|S_{k-1}| < m$ and A_k occurs, then $S_{k+2m} \geq m$. Let $N = \min\{k \geq 1 : A_k \text{ occurs}\}$. Then

$$T \leq N + 2m.$$

Next note that A_k and A_{k+l} are independent if $l \geq 2m+1$. Hence let $N' = \min\{k \geq 1 : A_{(2m+1)k} \text{ occurs}\}$ and note that

$$N \leq (2m+1)N'. \quad 4p$$

Moreover, N' is geometrically distributed with parameter $p = 1/2^{2m+1}$, and hence that $\mathbb{E}[N'] < \infty$. So,

$$\mathbb{E}[T] \leq \mathbb{E}[N] + 2m \leq (2m+1)\mathbb{E}[N'] + 2m < \infty.$$

We next show that $\mathbb{E}[T'] = \infty$. Suppose the contrary. Then since $(S_n)_{n \geq 0}$ is a martingale with bounded increments, it follows from Optional stopping III that

$$\mathbb{E}[S_{T'}] = \mathbb{E}[S_0] = 0. \quad 3p$$

However $\mathbb{E}[S_{T'}] = m$, which contradicts the assumption that $\mathbb{E}[T'] < \infty$. \downarrow