

2 Martingales – theory

2.1 Definitions

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space, let (E, \mathcal{E}) be a measurable space and let I be a countable subset of \mathbb{R} . A *process in E* is a family $X = (X_t)_{t \in I}$ of random variables in E . A *filtration* $(\mathcal{F}_t)_{t \in I}$ is an increasing family of sub- σ -algebras of \mathcal{F} : thus, $\mathcal{F}_s \subseteq \mathcal{F}_t$ whenever $s \leq t$. We set $\mathcal{F}_{-\infty} \stackrel{\text{def}}{=} \cap_{t \in I} \mathcal{F}_t$ and $\mathcal{F}_{\infty} = \sigma(\mathcal{F}_t : t \in I)$. Every process has a *natural filtration* $(\mathcal{F}_t^X)_{t \in I}$, given by

$$\mathcal{F}_t^X \stackrel{\text{def}}{=} \sigma(X_s : s \leq t).$$

We will always assume some filtration $(\mathcal{F}_t)_{t \in I}$ to be given. The σ -algebra \mathcal{F}_t is interpreted as modelling the state of our knowledge at time t . In particular, \mathcal{F}_t^X contains all the events which depend (measurably) only on X_s , $s \leq t$, that is, everything we know about the process X by time t . We say that X is *adapted* (to $(\mathcal{F}_t)_{t \in I}$) if X_t is \mathcal{F}_t -measurable for all t . Of course every process is adapted to its natural filtration. We say that X is *integrable* if X_t is integrable for all t .

Unless otherwise indicated, it is to be understood from now on that $E = \mathbb{R}$.

Definition 2.1.1 A *martingale* X is an adapted integrable process such that, for all $s, t \in I$ with $s \leq t$,

$$\mathbb{E}(X_t | \mathcal{F}_s) = X_s \quad a.s..$$

On replacing the equality in this condition by \leq or \geq , we get the notions of *supermartingale* and *submartingale*, respectively. Note that every process which is a martingale with respect to the given filtration is also a martingale with respect to its natural filtration.

2.2 Optional stopping

We say that a random variable $T : \Omega \rightarrow I \cup \infty$ is a *stopping time* if $\{T \leq t\} \in \mathcal{F}_t$ for all t . For a stopping time T , we set

$$\mathcal{F}_T \stackrel{\text{def}}{=} \{A \in \mathcal{F} : A \cap \{T \leq t\} \in \mathcal{F}_t \text{ for all } t\}.$$

It is easy to check that, if $T \equiv t$, then T is a stopping time and $\mathcal{F}_T = \mathcal{F}_t$. Given a process X , we set

$$X_T(\omega) \stackrel{\text{def}}{=} X_{T(\omega)}(\omega) \quad \text{whenever } T(\omega) < \infty.$$

we also define the *stopped process* X^T by $X_t^T \stackrel{\text{def}}{=} X_{T \wedge t}$.

We assume in the following two results that $I = \{0, 1, 2, \dots\}$. In this context, we will write n , m or k for elements of I , rather than t or s .

Proposition 2.2.2 Let S and T be stopping times and let $X = (X_n)_{n \geq 0}$ be an adapted process. Then

- (a) $S \wedge T$ is a stopping time;
- (b) if $S \leq T$, then $\mathcal{F}_S \subseteq \mathcal{F}_T$;
- (c) $X_T \mathbb{1}_{\{T < \infty\}}$ is an \mathcal{F}_T -measurable random variable;
- (d) X^T is adapted;
- (e) if X is integrable, then X^T is integrable.

Theorem 2.2.3 (Optional stopping theorem) Let $X = (X_n)_{n \geq 0}$ be an adapted integrable process. Then the following are equivalent:

- (a) X is supermartingale;
- (b) for all bounded stopping times T and all stopping times S ,

$$\mathbb{E}(X_T | \mathcal{F}_S) \leq X_{S \wedge T} \quad \text{a.s.};$$

- (c) for all stopping times T , X^T is a supermartingale;
- (d) for all bounded stopping times S and T with $S \leq T$,

$$\mathbb{E}(X_S) \geq \mathbb{E}(X_T).$$

Proof. For $S \geq 0$ and $T \leq n$, we have

$$\begin{aligned} X_T &= X_{S \wedge T} + \sum_{S \leq k < T} (X_{k+1} - X_k) \\ &= X_{S \wedge T} + \sum_{k=0}^n (X_{k+1} - X_k) \mathbb{1}_{\{S \leq k < T\}}. \end{aligned} \tag{2.1}$$

Suppose that X is a supermartingale and that S and T are stopping times, with $T \leq n$. If $A \in \mathcal{F}_S$, then $A \cap \{S \leq k\}$, $\{T > k\} \in \mathcal{F}_k$, so

$$\mathbb{E}((X_{k+1} - X_k) \mathbb{1}_{\{S \leq k < T\}} \mathbb{1}_A) \leq 0.$$

Hence, on multiplying (2.1) by $\mathbb{1}_A$ and taking expectations, we obtain

$$\mathbb{E}(X_T \mathbb{1}_A) \leq \mathbb{E}(X_{S \wedge T} \mathbb{1}_A).$$

We have shown that (a) \implies (b).

It is obvious that (b) \implies (c), (b) \implies (d), and (c) \implies (a).

Let $m \leq n$ and $A \in \mathcal{F}_m$. Set $T = m \mathbb{1}_A + n \mathbb{1}_{A^c}$, then T is a stopping time and $T \leq n$. We note that

$$\mathbb{E}(X_n \mathbb{1}_A) - \mathbb{E}(X_m \mathbb{1}_A) = \mathbb{E}(X_n) - \mathbb{E}(X_T).$$

It follows that (d) \implies (a). □

2.3 Doob's inequalities

Let X be a process and let $a, b \in \mathbb{R}$ with $a < b$. For $J \subseteq I$, set

$$U([a, b], J) \stackrel{\text{def}}{=} \sup \{n : X_{s_1} < a, X_{t_1} > b, \dots, X_{s_n} < a, X_{t_n} > b \\ \text{for some } s_1 < t_1 < \dots < s_n < t_n \text{ in } J\}.$$

Then $U[a, b] \equiv U([a, b], I)$ is the number of *upcrossings* of $[a, b]$ by X .

Theorem 2.3.4 (Doob's upcrossing inequality) *Let X be a supermartingale. Then*

$$(b - a)\mathbb{E}(U[a, b]) \leq \sup_{t \in I} \mathbb{E}((X_t - a)^-).$$

Proof. Since $U([a, b], I) = \lim_{J \uparrow I, J \text{ finite}} U([a, b], J)$, it suffices, by monotone convergence, to consider the case where I is finite. Let us assume that $I = \{0, 1, \dots, n\}$.

Write $U = U[a, b]$ and note that $U \leq n$. Set $T_0 = 0$ and define inductively for $k \geq 0$ (with the usual convention $\inf \emptyset = \infty$):

$$S_{k+1} \stackrel{\text{def}}{=} \inf\{m \geq T_k : X_m < a\}, \quad T_{k+1} \stackrel{\text{def}}{=} \inf\{m \geq S_{k+1} : X_m > b\}.$$

Then $U = \max\{k : T_k < \infty\}$. For $k \leq U$, set $G_k \stackrel{\text{def}}{=} X_{T_k} - X_{S_k}$ and note that $G_k \geq b - a$. Observe that $T_U \leq n$ and $T_{U+1} = \infty$. Set

$$R = \begin{cases} X_n - X_{S_{U+1}} & \text{if } S_{U+1} < \infty, \\ 0 & \text{if } S_{U+1} = \infty \end{cases}$$

and note that $R \geq -(X_n - a)^-$.

Then we have

$$\sum_{k=1}^n (X_{T_k \wedge n} - X_{S_k \wedge n}) = \sum_{k=1}^U G_k + R \geq (b - a)U - (X_n - a)^-. \quad (2.2)$$

Now X is a supermartingale and $S_k \wedge n$ and $T_k \wedge n$ are bounded stopping times, with $S_k \wedge n \leq T_k \wedge n$. Hence, by optional stopping, $\mathbb{E}(X_{T_k \wedge n}) \leq \mathbb{E}(X_{S_k \wedge n})$ and the desired inequality results on taking expectations in (2.2). \square

For any process X , for $J \subseteq I$, we set

$$X^*(J) \stackrel{\text{def}}{=} \sup_{t \in J} |X_t|, \quad X^* \stackrel{\text{def}}{=} X^*(I).$$

Theorem 2.3.5 (Doob's maximal inequality) *Let X be a martingale or a non-negative submartingale. Then, for all $\lambda \geq 0$,*

$$\lambda \mathbb{P}(X^* \geq \lambda) \leq \sup_{t \in I} \mathbb{E}(|X_t|).$$

Proof. Note that

$$\lambda \mathbb{P}(X^* \geq \lambda) = \lim_{\nu \uparrow \lambda} \nu \mathbb{P}(X^* > \nu) \leq \lim_{\nu \uparrow \lambda} \left(\lim_{J \uparrow I, J \text{ finite}} \nu \mathbb{P}(X^*(J) \geq \nu) \right).$$

It therefore suffices to consider the case where I is finite. Let us assume then that $I = \{0, 1, \dots, n\}$. If X is a martingale, then $|X|$ is a non-negative submartingale. It therefore suffices to consider the case where X is non-negative.

Set $T = \inf\{m \geq 0 : X_m \geq \lambda\} \wedge n$. Then T is a stopping time and $T \leq n$ so, by optional stopping,

$$\begin{aligned} \mathbb{E}(X_n) &\geq \mathbb{E}(X_T) = \mathbb{E}(X_T \mathbb{1}_{\{X^* \geq \lambda\}}) + \mathbb{E}(X_T \mathbb{1}_{\{X^* < \lambda\}}) \\ &\geq \lambda \mathbb{P}(X^* \geq \lambda) + \mathbb{E}(X_n \mathbb{1}_{\{X^* < \lambda\}}). \end{aligned}$$

Hence,

$$\lambda \mathbb{P}(X^* \geq \lambda) \leq \mathbb{E}(X_n \mathbb{1}_{\{X^* \geq \lambda\}}) \leq \mathbb{E}(X_n). \quad (2.3)$$

□

Theorem 2.3.6 (Doob's L^p -inequality) *Let X be a martingale or non-negative submartingale. Then, for all $p > 1$ and $q = p/(p-1)$,*

$$\|X^*\|_p \leq q \sup_{t \in I} \|X_t\|_p.$$

Proof. Since $X^* = \lim_{J \uparrow I, J \text{ finite}} X^*(J)$, it suffices, by monotone convergence, to consider the case where I is finite. Let us assume that $I = \{0, 1, \dots, n\}$. If X is a martingale, then $|X|$ is a non-negative submartingale. So it suffices to consider the case where X is non-negative.

Fix $k < \infty$. By Fubini's theorem, eqn. (2.3) and Hölder's inequality,

$$\begin{aligned} \mathbb{E}[(X^* \wedge k)^p] &= \mathbb{E} \int_0^k p \lambda^{p-1} \mathbb{1}_{\{X^* \geq \lambda\}} d\lambda = \int_0^k p \lambda^{p-1} \mathbb{P}(X^* \geq \lambda) d\lambda \\ &\leq \int_0^k p \lambda^{p-2} \mathbb{E}(X_n \mathbb{1}_{\{X^* \geq \lambda\}}) d\lambda = q \mathbb{E}(X_n (X^* \wedge k)^{p-1}) \\ &\leq q \|X_n\|_p \cdot \|X^* \wedge k\|_p^{p-1}. \end{aligned}$$

Hence $\|X^* \wedge k\|_p \leq q \|X_n\|_p$ and the result follows by monotone convergence on letting $k \rightarrow \infty$. □

2.4 Convergence theorems

Recall that, for $p \geq 1$, a process X is said to be *bounded in L^p* if

$$\sup_{t \in I} \|X_t\|_p < \infty.$$

Also X is *uniformly integrable* if

$$\sup_{t \in I} \mathbb{E}(|X_t| \mathbb{I}_{\{|X_t| > k\}}) \rightarrow 0 \quad \text{as } k \rightarrow \infty.$$

Recall that, if X is bounded in L^p for some $p > 1$, then X is uniformly integrable. Also if X is uniformly integrable then X is bounded in L^1 .

Theorem 2.4.7 (Almost sure martingale convergence theorem) *Let X be a supermartingale which is bounded in L^1 . Then $X_t \rightarrow X_\infty$ a.s. for some $X_\infty \in L^1(\mathcal{F}_\infty)$.*

Note that for $I \subseteq [0, \infty)$, a non-negative supermartingale is automatically bounded in L^1 .

Proof. By Doob's upcrossing inequality, for all $a < b$,

$$\mathbb{E}(U[a, b]) \leq (b - a)^{-1} \sup_{t \in I} \mathbb{E}(|X_t| + |a|) < \infty.$$

Consider for $a < b$ the sets

$$\begin{aligned} \Omega_{a,b} &\stackrel{\text{def}}{=} \left\{ \liminf_{t \rightarrow \infty} X_t < a < b < \limsup_{t \rightarrow \infty} X_t \right\}, \\ \Omega_0 &\stackrel{\text{def}}{=} \left\{ X_t \text{ converges in } [-\infty, \infty] \text{ as } t \rightarrow \infty \right\}. \end{aligned}$$

Since $U[a, b] = \infty$ on $\Omega_{a,b}$, we must have $\mathbb{P}(\Omega_{a,b}) = 0$; consequently, the equality

$$\Omega_0 \cup \left(\bigcup_{a,b \in \mathbb{Q}, a < b} \Omega_{a,b} \right) = \Omega$$

implies $\mathbb{P}(\Omega_0) = 1$. Define

$$X_\infty = \begin{cases} \lim_{t \rightarrow \infty} X_t & \text{on } \Omega_0, \\ 0 & \text{on } \Omega \setminus \Omega_0. \end{cases}$$

Then X_∞ is \mathcal{F}_∞ -measurable and, by Fatou's lemma,

$$\mathbb{E}(|X_\infty|) \leq \liminf_{t \rightarrow \infty} \mathbb{E}(|X_t|) < \infty.$$

So $X_\infty \in L^1$ as required. \square

Let us denote by \mathcal{M}^1 the set of uniformly integrable martingales and, for $p > 1$, by \mathcal{M}^p the set of martingales bounded in L^p .

Theorem 2.4.8 (L^p martingale convergence theorem) *Let $p \in [1, \infty)$.*

- (a) *Suppose $X \in \mathcal{M}^p$. Then $X_t \rightarrow X_\infty$ as $t \rightarrow \infty$, a.s. and in L^p , for some $X_\infty \in L^p(\mathcal{F}_\infty)$. Moreover, $X_t = \mathbb{E}(X_\infty | \mathcal{F}_t)$ a.s. for all t .*
- (b) *Suppose $Y \in L^p(\mathcal{F}_\infty)$ and set $X_t = \mathbb{E}(Y | \mathcal{F}_t)$. Then $X = (X_t)_{t \in I} \in \mathcal{M}^p$ and $X_t \rightarrow Y$ as $t \rightarrow \infty$, a.s. and in L^p .*

Thus the map $X \mapsto X_\infty$ is a one-to-one correspondence between \mathcal{M}^p and $L^p(\mathcal{F}_\infty)$.

Proof for $p = 1$. Let X be a uniformly integrable martingale. Then $X_t \rightarrow X_\infty$ a.s. by the almost sure martingale convergence theorem. Since X is uniformly integrable, it follows that $X_t \rightarrow X_\infty$ in L^1 . Next, for $s \geq t$,

$$\|X_t - \mathbb{E}(X_\infty | \mathcal{F}_t)\|_1 = \|\mathbb{E}(X_s - X_\infty | \mathcal{F}_t)\|_1 \leq \|X_s - X_\infty\|_1.$$

Let $s \rightarrow \infty$ to deduce $X_t = \mathbb{E}(X_\infty | \mathcal{F}_t)$ a.s..

Suppose now that $Y \in L^1(\mathcal{F}_\infty)$ and set $X_t \stackrel{\text{def}}{=} \mathbb{E}(Y | \mathcal{F}_t)$. Then $X = (X_t)_{t \in I}$ is a martingale by the tower property and is uniformly integrable by Lemma 1.5.2. Hence X_t converges a.s. and in L^1 , with limit X_∞ , say. For all t and all $A \in \mathcal{F}_t$ we have

$$\mathbb{E}(X_\infty \mathbb{1}_A) = \lim_{s \rightarrow \infty} \mathbb{E}(X_s \mathbb{1}_A) = \mathbb{E}(Y \mathbb{1}_A).$$

Now $X_\infty, Y \in L^1(\mathcal{F}_\infty)$ and $\cup_t \mathcal{F}_t$ is a π -system generating \mathcal{F}_∞ . Hence, $X_\infty = Y$ a.s.. \square

Proof for $p > 1$. Let X be a martingale bounded in L^p for some $p > 1$. Then $X_t \rightarrow X_\infty$ a.s. by the almost sure martingale convergence theorem. By Doob's L^p -inequality,

$$\|X^*\|_p \leq q \sup_{t \in I} \|X_t\|_p < \infty.$$

Since $|X_t - X_\infty|^p \leq (2X^*)^p$ for all t , we can use dominated convergence to deduce that $X_t \rightarrow X_\infty$ in L^p . It follows that $X_t = \mathbb{E}(X_\infty | \mathcal{F}_t)$ a.s., as in the case $p = 1$.

Suppose now that $Y \in L^p(\mathcal{F}_\infty)$ and set $X_t = \mathbb{E}(Y | \mathcal{F}_t)$. Then $X = (X_t)_{t \in I}$ is a martingale by the tower property and

$$\|X_t\|_p = \|\mathbb{E}(X_\infty | \mathcal{F}_t)\|_p \leq \|Y\|_p$$

for all t , so X is bounded in L^p . Hence X_t converges a.s. and in L^p , with limit X_∞ , say, and we can show that $X_\infty = Y$ a.s., as in the case $p = 1$. \square

Theorem 2.4.9 (Backward martingale convergence theorem) *Let $p \in [1, \infty)$ and let $Y \in L^p$. Set $X_t = \mathbb{E}(Y | \mathcal{F}_t)$. Then $X_t \rightarrow \mathbb{E}(Y | \mathcal{F}_{-\infty})$ as $t \rightarrow -\infty$, a.s. and in L^p .*

Proof. The argument is a minor modification of that used in Theorems 2.3.4, 2.4.7, and 2.4.8. The process X is automatically uniformly integrable, by Lemma 1.5.2 and is bounded in L^p because $\|X_t\|_p = \|\mathbb{E}(Y | \mathcal{F}_t)\|_p \leq \|Y\|_p$ for all t . We leave the details to the reader. \square

In the following result we take $I = \{0, 1, 2, \dots\}$.

Theorem 2.4.10 (Optional stopping theorem – 2) *Let X be a uniformly integrable martingale and let S and T be stopping times. Then*

$$\mathbb{E}(X_T | \mathcal{F}_S) = X_{S \wedge T} \quad \text{a.s.}$$

Proof. We have already proved the result when T is bounded. If T is unbounded, then $T \wedge n$ is a bounded stopping time, so

$$\mathbb{E}(X_n^T | \mathcal{F}_S) = \mathbb{E}(X_{T \wedge n} | \mathcal{F}_S) = X_{S \wedge T \wedge n} = X_{S \wedge n}^T \quad \text{a.s.} \quad (2.4)$$

Now

$$\|\mathbb{E}(X_n^T | \mathcal{F}_S) - \mathbb{E}(X_T | \mathcal{F}_S)\|_1 \leq \|X_n^T - X_\infty^T\|_1. \quad (2.5)$$

We have $X_n \rightarrow X_\infty$ in L^1 . So, in the case $T \equiv \infty$, we can pass to the limit in (2.4) to obtain

$$\mathbb{E}(X_\infty | \mathcal{F}_S) = X_S \quad \text{a.s.}$$

Then, returning to (2.5), for general T , we have

$$\|X_n^T - X_\infty^T\|_1 = \|\mathbb{E}(X_n - X_\infty | \mathcal{F}_T)\|_1 \leq \|X_n - X_\infty\|_1$$

and the result follows on passing to the limit in (2.4). \square

Theorem 2.4.11 (Optional stopping theorem – 3) *Let T be a stopping time with $\mathbb{E}T < \infty$ and let X_n be a supermartingale with uniformly bounded increments, i.e., there exists a finite constant $K > 0$ such that*

$$|X_n(\omega) - X_{n-1}(\omega)| \leq K \quad \forall (n, \omega).$$

Then X_T is integrable and $\mathbb{E}(X_T) \leq \mathbb{E}(X_0)$.