

Stochastic Calculus Notes

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1 Martingales

Let (Ω, \mathcal{F}, P) be a probability space with filtration $\{\mathcal{F}_t\}$. We assume that $\{\mathcal{F}_t\}$ is complete but not right-continuous, unless so specified.

Definition 1.1. *A real-valued stochastic process $M = \{M_t : t \in \mathbb{R}_+\}$ is a submartingale adapted to $\{\mathcal{F}_t\}$ if each M_t is integrable and*

$$E[M_t | \mathcal{F}_s] \geq M_s \text{ for all } t < s$$

M is a supermartingale if $-M$ is a submartingale. Clearly, M is a martingale if it is both a submartingale and a supermartingale. Also, M is square integrable if $E[M_t^2] < \infty$ for all $t \in \mathbb{R}_+$.

Proposition 1.2.

- i) If M is a martingale and ϕ is a convex function such that $\phi(M_t)$ is integrable for all t , then $\phi(M_t)$ is a submartingale.*
- ii) If M is a submartingale and ϕ a nondecreasing convex function such that $\phi(M_t)$ is integrable for all t , then $\phi(M_t)$ is a submartingale.*

Proof.

- i) By Jensen's inequality, we have

$$E[\phi(M_t) | \mathcal{F}_s] \geq \phi(E[M_t | \mathcal{F}_s]) = \phi(M_s)$$

ii) By Jensen's inequality, we have

$$E[\phi(M_t)|\mathcal{F}_s] \geq \phi(E[M_t]|\mathcal{F}_s) \geq \phi(M_s)$$

Where in the last inequality we used that ϕ is nondecreasing.

□

We will need the notion of uniform integrability.

Definition 1.3. Let $\{X_\alpha : \alpha \in A\}$ a collection of random variables in some probability space (Ω, \mathcal{F}, P) . They are uniformly integrable if

$$\lim_{M \rightarrow \infty} \sup_{\alpha \in A} E[|X_\alpha| \cdot 1\{|X_\alpha| \geq M\}] = 0$$

This condition is equivalent to request the following two conditions:

i) $\sup_{\alpha} E[|X_\alpha|] < \infty$

ii) Given $\epsilon > 0$, there exists some $\delta > 0$ such that for every $B \in \mathcal{F}$ with $P(B) < \delta$,

$$\sup_{\alpha} \int_B |X_\alpha| \leq \epsilon$$

We will use the following two lemmas.

Lemma 1.4. Let X be an integrable random variable on (ω, \mathcal{F}, P) . Then the collection of random variables

$$\{E[X|\mathcal{A}] : \mathcal{A} \text{ is a sub-}\sigma \text{ algebra of } \mathcal{F}\}$$

is uniformly integrable

Lemma 1.5. Suppose $X_n \rightarrow X$ in L^1 on a probability space (ω, \mathcal{F}, P) . Then there exists a subsequence $\{n_j\}$ such that $E[X_{n_j}|\mathcal{A}] \rightarrow E[X|\mathcal{A}]$ almost surely.

Proof. Note that $|E[X_n|\mathcal{A}] - E[X|\mathcal{A}]| \leq E[|X_n - X||\mathcal{A}]$. Also,

$$E[E[|X_n - X||\mathcal{A}]] = E[|X_n - X|] \rightarrow 0$$

So $E[X_n|\mathcal{A}] \rightarrow E[X|\mathcal{A}]$ in L^1 , and the claim follows.

□

Proposition 1.6. *Suppose M is a right-continuous submartingale with respect to the filtration $\{\mathcal{F}_t\}$. Then M is a submartingale also with respect to $\{\mathcal{F}_{t+}\}$*

Proof. Let $s < t$ and n such that $n > (t - s)^{-1}$. $M_t \vee c$ is a submartingale, so

$$E[M_t \vee c | \mathcal{F}_{s+n-1}] \geq M_{s+n-1} \vee c$$

Applying $E[\cdot | \mathcal{F}_{s+}]$ and using that $\mathcal{F}_{s+} \subset \mathcal{F}_{s+n-1}$,

$$E[M_t \vee c | \mathcal{F}_{s+}] \geq E[M_{s+n-1} \vee c | \mathcal{F}_{s+}] \quad (1.1)$$

Using the bounds

$$c \leq M_{s+n-1} \vee c \leq E[M_t \vee c | \mathcal{F}_{s+n-1}]$$

and [Lemma 1.4](#) we deduce that $\{M_{s+n-1} \vee c\}$ is uniformly integrable, and in particular they are uniformly bounded in L^1 . The right-continuity give us for fixed c , $M_{s+n-1} \vee c \rightarrow M_s \vee c$ almost surely, so we have convergence in L^1 too. Using [Lemma 1.5](#) we can obtain a subsequence $\{n_j\}$ such that

$$E[M_{n+n_j^{-1}} \vee c | \mathcal{F}_{s+}] \rightarrow E[M_s \vee c | \mathcal{F}_{s+}]$$

This together with [\(1.1\)](#) implies

$$M_s \leq M_s \vee c = E[M_s \vee c | \mathcal{F}_{s+}] \leq E[M_t \vee c | \mathcal{F}_s]$$

So $M_s \vee c \leq E[M_t \vee c | \mathcal{F}_{s+}]$. Now, take $c \rightarrow -\infty$ and by dominated convergence finally obtain that

$$M_s \leq E[M_t | \mathcal{F}_{s+}]$$

□

There is sort of converse of this result, found in [\[2\]](#).

Proposition 1.7. *Suppose the filtration $\{\mathcal{F}_t\}$ satisfies the usual conditions, that is, (Ω, \mathcal{F}, P) is complete, \mathcal{F}_0 contains all null events, and $\mathcal{F}_{t+} = \mathcal{F}_t$. Let M be a submartingale such that $t \rightarrow EM_t$ is right-continuous. Then there exists a cadlag modification of M that is an $\{\mathcal{F}_t\}$ -submartingale.*

1.1 Optional stopping

We want to extend the submartingale property from deterministic times to stopping times.

Lemma 1.8. *Let M be a submartingale. Let σ, τ be two stopping times whose values lie in an ordered countable set $\{s_1 < s_2 < \dots\} \cup \{\infty\} \subset [0, \infty]$ where $s_j \nearrow \infty$. Then for any $T < \infty$,*

$$E[M_{\tau \wedge T} | \mathcal{F}_\sigma] \geq M_{\sigma \wedge \tau \wedge T} \quad (1.2)$$

Proof. Fix n so that $s_n \leq T < s_{n+1}$. First observe that $M_{\tau \wedge T}$ is integrable

$$|M_{\tau \wedge T}| = \sum_{i=1}^n 1_{\{\tau = s_i\}} |M_{s_i}| + 1_{\{\tau > s_n\}} |M_T| \leq \sum_{i=1}^n |M_{s_i}| + |M_T|$$

a finite sum of integrable random variables.

The second property is verify that $M_{\sigma \wedge \tau \wedge T}$ is \mathcal{F}_σ -measurable. To do this, it's enough to prove for $B \in \mathcal{B}(\mathbb{R}^d)$ that $\{M_{\sigma \wedge \tau \wedge T} \in B\} \cap \{\sigma \leq t\} \in \mathcal{F}_t$ for all $t \in \mathbb{R}_+$.

Let s_j the highest value such that $s_j \leq t$. If there is not such s_j , the $t < s_1$, so $\{\sigma \leq t\} \subset \{\sigma < s_1\} = \emptyset \in \mathcal{F}_t$, because $\sigma \geq s_1$. Otherwise,

$$\{M_{\sigma \wedge \tau \wedge T} \in B\} \cap \{\sigma \leq t\} = \cup_{i=1}^j \{\sigma \wedge \tau = s_i\} \cap \{M_{s_i \wedge T} \in B\} \cap \{\sigma \leq t\} \in \mathcal{F}_t$$

because $s_i \leq t$ and $\sigma \wedge \tau$ is a stopping time.

Because both $E[M_{\tau \wedge T} | \mathcal{F}_\sigma]$ and $M_{\sigma \wedge \tau \wedge T}$ are \mathcal{F}_σ -measurable, (1.2) follows from

$$E\{1_A E[M_{\tau \wedge T} | \mathcal{F}_\sigma]\} \geq E\{1_A M_{\sigma \wedge \tau \wedge T}\}$$

for $A \in \mathcal{F}_\sigma$. By the definition of conditional expectation, it's reduced to show that

$$E[1_A M_{\tau \wedge T}] \geq E[1_A M_{\sigma \wedge \tau \wedge T}]$$

We split $1_A = 1_{\{A \cap \sigma \leq T\}} + 1_{\{A \cap \sigma > T\}}$. If $\sigma > T$, $\sigma \wedge \tau \wedge T = \tau \wedge T$, so

$$E[1_{\{A \cap \sigma > T\}} M_{\sigma \wedge \tau \wedge T}] = E[1_{\{A \cap \sigma > T\}} M_{\tau \wedge T}]$$

For the case $\sigma \leq T$, we split into sub-cases. We want to prove

$$\begin{aligned} E[1_{\{A \cap \{\sigma = s_i\}\}} M_{\tau \wedge T}] &\geq E[1_{\{A \cap \{\sigma = s_i\}\}} M_{\sigma \wedge \tau \wedge T}] \\ &= E[1_{\{A \cap \{\sigma = s_i\}\}} M_{s_i \wedge \tau \wedge T}] \quad \text{for } 1 \leq i \leq n \end{aligned}$$

As $A \cap \{\sigma = s_i\} \in \mathcal{F}_{s_i}$, it's enough to check the following:

$$E[M_{\tau \wedge T} | \mathcal{F}_{s_i}] \geq M_{s_i \wedge \tau \wedge T} \quad \text{for } 1 \leq i \leq n \quad (1.3)$$

We prove (1.3) by reverse induction on i . We first consider the case $i = n$. First, we consider an auxiliary inequality. Recall that $M_{\sigma \wedge \tau \wedge T}$ is \mathcal{F}_σ -measurable, so in particular if $\sigma \equiv s_j$, then $M_{s_j \wedge \tau \wedge T}$ is \mathcal{F}_{s_j} -measurable. For any j ,

$$\begin{aligned} E[M_{s_{j+1} \wedge \tau \wedge T} | \mathcal{F}_{s_j}] &= E[M_{s_{j+1} \wedge \tau \wedge T} 1\{\tau > s_j\} + M_{s_j \wedge \tau \wedge T} 1\{\tau \leq s_j\} | \mathcal{F}_{s_j}] \\ &= E[M_{s_{j+1} \wedge T} | \mathcal{F}_{s_j}] \cdot 1\{\tau > s_j\} + M_{s_j \wedge \tau \wedge T} 1\{\tau \leq s_j\} \\ &\geq M_{s_j \wedge T} \cdot 1\{\tau > s_j\} + M_{s_j \wedge \tau \wedge T} 1\{\tau \leq s_j\} \\ &= M_{s_j \wedge \tau \wedge T} \end{aligned}$$

Recall that $s_n \leq T < s_{n+1}$, so if we apply this inequality to $j = n$, we conclude that $E[M_{\tau \wedge T} | \mathcal{F}_{s_n}] \geq M_{s_n \wedge \tau \wedge T}$, and this is the case $i = n$ from (1.3). Now assuming that (1.3) holds for i , we apply the auxiliary inequality again,

$$\begin{aligned} E[M_{\tau \wedge T} | \mathcal{F}_{s_{i-1}}] &= E[E[M_{\tau \wedge T} | \mathcal{F}_{s_i}] | \mathcal{F}_{s_{i-1}}] \geq E[M_{s_i \wedge \tau \wedge T} | \mathcal{F}_{s_{i-1}}] \\ &\geq M_{s_{i-1} \wedge \tau \wedge T} \end{aligned}$$

And this is (1.3) for $i - 1$. We repeat this until $i = 1$, and we are done. \square

To extend this result to general stopping times, we need a preliminary lemma, and some regularity conditions.

Lemma 1.9. *Let M be a submartingale with right-continuous paths and $T < \infty$. Then for any stopping time ρ that satisfies $P(\tau \leq T) = 1$,*

$$E[M_\rho] \leq 2E[M_T^+] - E[M_0]$$

Proof. Approximate ρ by ρ_n given by $\rho_n = T$ if $\rho = T$ and $\rho_n = 2^{-n}T([2^n \rho / T] + 1)$ if $\rho < T$. Then ρ_n is a stopping time and $\rho_n \searrow \rho$ as $n \rightarrow \infty$. Apply (1.2) to $\tau = \rho_n$, $\sigma = 0$ and taking expectations we deduce that

$$E[M_{\rho_n}] \geq E[M_0]$$

Now apply (1.2) to the submartingale $M_t^+ = M_t \vee 0$, with $\tau = T$ and $\sigma = \rho_n$ to get

$$E[M_T^+] \geq E[M_{\rho_n}^+]$$

Using both equations,

$$E[M_{\rho_n}^-] = E[M_{\rho_n}^+] - E[\rho_n] \leq E[M_T^+] - E[M_0]$$

Thus,

$$E[|M_{\rho_n}|] = E[M_{\rho_n}^+] + E[M_{\rho_n}^-] \leq 2E[M_T^+] - E[M_0]$$

Finally, take $n \rightarrow \infty$, and use Fatou's lemma to conclude

$$E|M_\rho| \leq \liminf_{n \rightarrow \infty} E|M_{\rho_n}| \leq 2E[M_T^+] - E[M_0]$$

□

Remark 1.10. The last result says that if τ is a stopping time and $T \in \mathbb{R}_+$, the stopped process $M_{\tau \wedge T}$ is integrable.

Now we extend from discrete to general stopping times.

Theorem 1.11. Let M be a submartingale with right-continuous paths, and let σ, τ be two stopping times. Then for $T < \infty$,

$$E[M_{\tau \wedge T} | \mathcal{F}_\sigma] \geq M_{\tau \wedge \sigma \wedge T} \quad (1.4)$$

Proof. Recall that the random variables $M_{\tau \wedge T}, M_{\sigma \wedge \tau \wedge T}$ are integrable, so their conditional expectation are well defined.

Approximate the stopping times defining $\sigma_n = 2^{-n}T([2^n\sigma/T] + 1)$, $\tau_n = 2^{-n}T([2^n\tau/T] + 1)$. Here, $\sigma_n = \infty$ if $\sigma = \infty$, and similarly with τ_n . Fix $c \in \mathbb{R}$. The function $x \rightarrow x \vee c$ is convex and nondecreasing, hence $M_t \vee c$ is also a submartingale. Applying Lemma 1.8 to this submartingale and stopping times σ_n, τ_n give

$$E[M_{\tau_n \wedge T} \vee c | \mathcal{F}_{\sigma_n}] \geq M_{\sigma_n \wedge \tau_n \wedge T} \vee c$$

Since $\sigma \leq \sigma_n$, $\mathcal{F}_\sigma \subset \mathcal{F}_{\sigma_n}$, and if we apply conditional expectation both sides,

$$E[M_{\tau_n \wedge T} \vee c | \mathcal{F}_\sigma] \geq E[M_{\sigma_n \wedge \tau_n \wedge T} \vee c | \mathcal{F}_\sigma] \quad (1.5)$$

Now we want to take $n \rightarrow \infty$ in (1.5) to obtain the conclusion to the process $M_t \vee c$, and then take $c \rightarrow -\infty$ to conclude. First note that $\tau_n \wedge T \rightarrow \tau \wedge T, \sigma_n \wedge T \rightarrow \sigma \wedge T$. By the right continuity of M ,

$$M_{\tau_n \wedge T} \rightarrow M_{\tau \wedge T} \text{ and } M_{\sigma_n \wedge \tau_n \wedge T} \rightarrow M_{\sigma \wedge \tau \wedge T}$$

Next, apply Lemma 1.8 to obtain

$$c \leq M_{\tau_n \wedge T} \vee c \leq E[M_T \vee c | \mathcal{F}_{\tau_n}]$$

(taking $\sigma = \tau_n, \tau = T$), and

$$c \leq M_{\sigma_n \wedge \tau_n \wedge T} \vee c \leq E[M_T \vee c | \mathcal{F}_{\tau_n \wedge \sigma_n}]$$

(taking $\sigma = \tau_n \wedge \sigma_n, \tau = T$). Recalling Lemma 1.4, the sequences $\{M_{\tau_n \wedge T} \vee c : n \in \mathbb{N}\}, \{M_{\sigma_n \wedge \tau_n \wedge T} \vee c : n \in \mathbb{N}\}$ are uniformly integrable, and the almost surely convergence of these sequence assures the L^1 convergence. Therefore, by Lemma 1.5, there exists a subsequence $\{n_j\}$ along which the conditional expectations converge almost surely

$$E[M_{\tau_{n_j} \wedge T} \vee c | \mathcal{F}_\sigma] \rightarrow E[M_{\tau \wedge T} \vee c | \mathcal{F}_\sigma]$$

and

$$E[M_{\sigma_{n_j} \wedge \tau_{n_j} \wedge T} \vee c | \mathcal{F}_\sigma] \rightarrow E[M_{\sigma \wedge \tau \wedge T} \vee c | \mathcal{F}_\sigma]$$

Now we can take limits in (1.5) to conclude that

$$E[M_{\tau \wedge T} \vee c | \mathcal{F}_\sigma] \geq E[M_{\sigma \wedge \tau \wedge T} \vee c | \mathcal{F}_\sigma] = M_{\sigma \wedge \tau \wedge T} \vee c \geq M_{\sigma \wedge \tau \wedge T}$$

where we used that $M_{\sigma \wedge \tau \wedge T}$ is $\mathcal{F}_{\sigma \wedge \tau \wedge T}$ -measurable because M is right-continuous and hence, progressively measurable (by Lemma 1.13 from first chapter), therefore is \mathcal{F}_σ -measurable. Finally, taking $c \rightarrow -\infty$ we obtain $M_{\tau \wedge T} \vee c \rightarrow M_{\tau \wedge T}$ point-wise, and for $c \leq 0, |M_{\tau \wedge T} \vee c| \leq |M_{\tau \wedge T}|$, so by dominated convergence,

$$\lim_{c \rightarrow -\infty} E[M_{\tau \wedge T} \vee c | \mathcal{F}_\sigma] = E[M_{\tau \wedge T} | \mathcal{F}_\sigma]$$

this completes the proof. □

Corollary 1.12. *Suppose M is a right-continuous submartingale and τ is a stopping time. Then the stopped process $M^\tau = \{M_{\tau \wedge t} : t \in \mathbb{R}_+\}$ is a submartingale with respect to the*

filtration $\{\mathcal{F}_t\}$. If M is also a martingale, then M^τ is a martingale. Finally, if M is an L^2 -martingale, then so is M^τ .

Proof. Take $T = t, \sigma = s < t$ in (1.4) to obtain $E[M_{\tau \wedge t} | \mathcal{F}_s] \geq M_{\tau \wedge s}$. If M is a martingale, apply the last result to both M and $-M$. Finally, if M is an L^2 -martingale, apply Lemma 1.9 to the submartingale M^2 and thus deduce the same result for M^τ . \square

Corollary 1.13. *Suppose M is a right-continuous submartingale. Let $\{\sigma(u) : u \geq 0\}$ be a nondecreasing $[0, \infty)$ -valued process such that $\sigma(u)$ is a bounded stopping time for each u . Then $\{M_{\sigma(u)} : u \geq 0\}$ is a submartingale with respect to the filtration $\{\mathcal{F}_{\sigma(u)} : u \geq 0\}$*

Proof. For $u < v$ and $T \geq \sigma(v)$, we have $\sigma(u) \wedge \sigma(v) \wedge T = \sigma(u), \sigma(v) \wedge T = \sigma(v)$, and using (1.4) give us

$$E[M_{\sigma(v)} | \mathcal{F}_{\sigma(u)}] \geq M_{\sigma(u)}.$$

As in the last corollary, if M is a martingale, apply this result to both M and $-M$, and if M is an L^2 -martingale, utilize Lemma 1.9 in the submartingale M^2 to deduce

$$E[M_{\sigma(u)}^2] \leq 2[M_T^2] + E[M_0^2]$$

\square

Remark 1.14. *The last corollary has the following implications:*

- i) Using $\sigma(t) = \tau \wedge t$, then M^τ is also a submartingale with respect to $\{\mathcal{F}_{\tau \wedge t}\}$
- ii) Using $\sigma(t) = \tau + t$ for a bounded stopping time τ , then the process $\tilde{M}_t := M_{\tau+t} - M_\tau$ is an L^2 martingale with respect to $\tilde{\mathcal{F}}_t = \mathcal{F}_{\tau+t}$ if M is an L^2 -martingale.

1.2 Inequalities and limits

Lemma 1.15. *Let M be a submartingale, $0 < T < \infty$, and H a finite subset of $[0, T]$. Then for $r > 0$,*

$$\begin{aligned} P\left(\max_{t \in H} M_t \geq r\right) &\leq \frac{1}{r} E[M_T^+] \\ P\left(\min_{t \in H} M_t \leq -r\right) &\leq \frac{1}{r} E[M_T^+ - E[M_0]] \end{aligned}$$

Proof. Let $\sigma = \min\{t \in H : M_t \geq r\}$, with the interpretation that $\sigma = \infty$ if $M_t < r$ for all $t \in H$. Now we use (1.4) with $\tau = T$,

$$E[M_T] \geq E[M_{\sigma \wedge T}] = E[M_\sigma 1\{\sigma < \infty\}] + E[M_T 1\{\sigma = \infty\}]$$

so $E[M_\sigma 1\{\sigma < \infty\}] \leq E[M_T 1\{\sigma < \infty\}]$, from which

$$\begin{aligned} rP\left(\max_{t \in H} M_t \geq r\right) &\leq rP(\sigma < \infty) \leq E(M_\sigma 1\{\sigma < \infty\}) \leq E[M_T 1\{\sigma < \infty\}] \\ &\leq E[M_T^+ 1\{\sigma < \infty\}] \leq E[M_T^+] \end{aligned}$$

And we obtain the first inequality. To probe the second one, let $\tau = \min\{t \in H : M_t \leq -r\}$ and utilize (1.4) with $\sigma = 0$ to deduce

$$E[M_0] \leq E[M_{\tau \wedge T}] = E[M_\tau 1\{\tau < \infty\}] + E[M_T 1\{\tau = \infty\}]$$

from which

$$\begin{aligned} -rP\left(\min_{t \in H} M_t \leq -r\right) &= -rP(\tau < \infty) \geq E[M_\tau 1\{\tau < \infty\}] \\ &\geq E[M_0] - E[M_T 1\{\tau = \infty\}] \geq E[M_0] - E[M_T^+] \end{aligned}$$

□

Now we generalize to uncountable suprema and infima.

Theorem 1.16. *Let M be a right-continuous submartingale and $0 < T < \infty$. Then for $r > 0$,*

$$P\left(\sup_{0 \leq t \leq T} M_t \geq r\right) \leq \frac{1}{r} E[M_T^+] \quad (1.6)$$

$$P\left(\inf_{0 \leq t \leq T} M_t \leq -r\right) \leq \frac{1}{r} E[M_T^+ - E[M_0]] \quad (1.7)$$

Proof. Let H be a countable dense subset of $[0, T]$ that contains $0, T$, and let $H_1 \subset H_2 \subset H_3 \subset \dots$ be finite sets such that $H = \cup_n H_n$. We apply the last lemma to the sets H_n 's. Let $b < r$. By right-continuity,

$$P\left(\sup_{0 \leq t \leq T} M_t > b\right) = P\left(\sup_{t \in H} M_t > b\right) = \lim_{n \rightarrow \infty} P\left(\sup_{t \in H_n} M_t > b\right) \leq \frac{1}{b} E[M_T^+]$$

Take $b \nearrow r$ and conclude the first inequality. The second one is analogous. □

Definition 1.17. Suppose that X has either left- or right-continuous paths with probability 1. In that case, define

$$X_T^*(\omega) := \sup_{0 \leq t \leq T} |X_t(\omega)| \quad (1.8)$$

We verify that X_T^* is \mathcal{F}_T measurable for each T . Define $U := \sup_{s \in R} |X_s|$, when R is a dense countable subset of $[0, T]$ that contains the endpoints. Then U is \mathcal{F}_T measurable, because it's the supremum of countable many random variables \mathcal{F}_T -measurable. On every left- or right-continuous path, $U = X_T^*$, so they are equal almost surely. Also, all the zero-probability events lie in \mathcal{F}_T by the completeness assumption. This implies that X_T^* is also \mathcal{F}_T -measurable.

Now we state the important Doob's inequality:

Theorem 1.18. (Doob's inequality) Let M be a nonnegative right-continuous submartingale and $0 < T < \infty$. Then for $1 < p < \infty$

$$E\left[\sup_{0 \leq t \leq T} M_t^p\right] \leq \left(\frac{p}{p-1}\right)^p E[M_T^p] \quad (1.9)$$

Proof. As M is nonnegative, $M_T^* = \sup_{0 \leq t \leq T} M_t$

Step 1 We show that

$$P(M_T^* > r) \leq \frac{1}{r} E[M_T 1\{M_T^* \geq r\}], \quad r > 0$$

Let $\tau = \inf\{t > 0 : M_t > r\}$, an $\{\mathcal{F}_{t+}\}$ -stopping time by [Proposition 1.6](#). By right-continuity, $M_\tau \geq r$ when $\tau < \infty$. Also, $M_T^* > r$ implies $\tau \leq T$. This says that $M_T^* > r \rightarrow \tau < \infty \rightarrow M_\tau \geq r$ and so

$$rP(M_T^* > r) \leq E[M_\tau 1\{M_T^* > r\}] \leq E[M_\tau 1\{\tau \leq T\}]$$

Since M is a submartingale with respect to $\{\mathcal{F}_{t+}\}$ by [Proposition 1.6](#) (and τ is an $\{\mathcal{F}_{t+}\}$ -stopping time), we can apply [Theorem 1.11](#) to obtain

$$\begin{aligned} E[M_\tau 1\{\tau \leq T\}] &= E[M_{\tau \wedge T}] - E[M_T 1\{\tau > T\}] \leq E[M_T] - E[M_T 1\{\tau > T\}] \\ &= E[M_T 1\{\tau \leq T\}] \leq E[M_T 1\{\tau \leq T\}] \leq E[M_T 1\{M_T^* \geq r\}] \end{aligned}$$

in the last step we used that $\tau \leq T \rightarrow M_T^* \geq r$. That concludes the first step.

Step 2 Let $0 < b < \infty$. We use the identity $E[(M_T^* \wedge b)^p] = \int_0^b pr^{p-1}P[M_T^* > r]dr$ and Hölder inequality:

$$\begin{aligned} E[(M_T^* \wedge b)^p] &= \int_0^b pr^{p-1}P[M_T^* > r]dr \leq \int_0^b pr^{p-2}E[M_T 1\{M_T^* \geq r\}]dr \\ &\stackrel{Fubini}{=} E[M_T \int_0^{b \wedge M_T^*} pr^{p-2}dr] = \frac{p}{p-1}E[M_T(b \wedge M_T^*)^{p-1}] \\ &\stackrel{Holder}{\leq} \frac{p}{p-1}E[M_T^p]^{1/p}E[(b \wedge M_T^*)^p]^{\frac{p-1}{p}} \end{aligned}$$

Now we divide by $E[(b \wedge M_T^*)^p]^{\frac{p-1}{p}}$ (which is finite by the truncation) to obtain

$$E[(M_T^* \wedge b)^p]^{1/p} \leq \frac{p}{p-1}E([M_T^p]^{1/p})$$

Raising to the $p - th$ power both sides, and taking $b \rightarrow \infty$ to conclude the claim by monotone convergence.

□

Corollary 1.19. *Let M be a nonnegative right-continuous submartingale and τ a bounded stopping time .Then for $1 < p < \infty$,*

$$E \left[\left(\sup_{0 \leq t \leq \tau} M_t \right)^p \right] \leq \left(\frac{p}{p-1} \right)^p E[M_\tau^p] \quad (1.10)$$

Proof. Consider the stopped process $M_{t \wedge \tau}$ and T with $\tau \leq T$.Then, $M_{\tau \wedge T} = M_\tau$ and $\sup_{0 \leq t \leq T} M_{\tau \wedge t} = \sup_{0 \leq t \leq \tau} M_t$.We apply the Doob's inequality no $M_{\tau \wedge t}$. □

For completeness, we state the Martingale Convergence Theorem.

Theorem 1.20. *Let M be a right-continuous submartingale such that*

$$\sup_{t \in \mathbb{R}_+} E(M_t^+) < \infty$$

Then there exists a random variable M_∞ integrable and $M_t(\omega) \rightarrow M_\infty(\omega)$ as $t \rightarrow \infty$ for almost every ω

When the limit M_∞ exist, the question is if the complete process $\{M_t : t \in [0, \infty]\}$ is a martingale,that is, if $E(M_\infty | \mathcal{F}_t) = M_t$.The following theorem characterizes this condition.

Theorem 1.21. *Let $M = \{M_t : t \in \mathbb{R}_+\}$ be a right-continuous martingale. The following are equivalent:*

- i) *The collection $\{M_t : t \in \mathbb{R}_+\}$ is uniformly integrable.*
- ii) *There exists an integrable random variable M_∞ such that*

$$\lim_{t \rightarrow \infty} E|M_t - M_\infty| = 0$$

- iii) *There exists an integrable random variable M_∞ such that*

$$\begin{aligned} M_t(\omega) &\xrightarrow{t \rightarrow \infty} M_\infty(\omega) \quad \text{almost surely} \\ E(M_\infty | \mathcal{F}_t) &= M_t \quad \text{for all } t \in \mathbb{R}_+ \end{aligned}$$

- iv) *There exists an integrable random variable Z such that $M_t = E(Z | \mathcal{F}_t)$ for all $t \in \mathbb{R}_+$.*

We state a direct corollary from the last theorem.

Corollary 1.22.

- i) *For $Z \in L^1(P)$, $E(Z | \mathcal{F}_t) \xrightarrow{t \rightarrow \infty} E(Z | \mathcal{F}_\infty)$ both almost-surely and in L^1 .*
- ii) *For $A \in \mathcal{F}_\infty$, $E(1_A | \mathcal{F}_t) \xrightarrow{t \rightarrow \infty} E(1_A | \mathcal{F}_\infty)$ both almost-surely and in L^1 .*

Proof. Part ii) follows from part i). To prove part i), define $M_t = E(Z | \mathcal{F}_t)$. Then, part iv) of [Theorem 1.21](#), there exist a limit M_∞ almost surely and a limit M'_∞ in L^1 . But as the M'_t s are uniformly integrable, then the convergence is also in L^1 , so $M'_\infty = M_\infty$. By construction, M_∞ is \mathcal{F}_∞ -measurable. We need to conclude that $M_\infty = E(Z | \mathcal{F}_\infty)$. For $A \in \mathcal{F}_s$ we have for $t > s$ and by L^1 convergence

$$E[1_A Z] = E[1_A M_t] \rightarrow E[1_A M_\infty]$$

By a standard argument, extend $E[1_A Z] = E[1_A M_\infty]$ to all $A \in \mathcal{F}_\infty$. This concludes the proof. \square

1.3 Local martingales and semimartingales

For a stopping time τ and a process $X = \{X_t : t \in \mathbb{R}_+\}$, the stopped process X^τ is defined by $X_t^\tau = X_{\tau \wedge t}$.

Definition 1.23. Let $M = \{M_t : t \in \mathbb{R}_+\}$ be a process adapted to a filtration \mathbb{F}_t . M is a local martingale if there exists a sequence of stopping times $\tau_1 \leq \tau_2 \leq \dots$ such that $P(\tau_k \nearrow \infty) = 1$ and for each k , M^{τ_k} is a martingale with respect to $\{\mathcal{F}_t\}$. M is a local square-integrable martingale (local L^2 -martingale from now) if for each k , M^{τ_k} is a square-integrable martingale. In both cases we say $\{\tau_k\}$ is a localizing sequence for M .

Remark 1.24. If we consider a local martingale $\{M_t : t \in [0, T]\}$, then it's enough to consider a nondecreasing sequence σ_n with $\sigma_n \geq T$ for large enough n almost surely. In that case, if we define $\tau_n = \sigma_n 1\{\sigma_n < T\} + \infty 1\{\sigma_n \geq T\}$, we recover the original definition.

Lemma 1.25. Suppose M is a local martingale and σ is an arbitrary stopping time. Then M^σ is also a local martingale. Similarly, if M is a local L^2 -local martingale, it's also M^σ . In both cases, if $\{\tau_k\}$ is a localizing sequence, for M , it's also for M^σ .

Proof. Let $\{\tau_k\}$ be a localizing sequence for M . Then, M^{τ_k} is a martingale for each k . By [Corollary 1.12](#), the process $M_{\sigma \wedge t}^{\tau_k} = (M^\sigma)_t^{\tau_k}$ is a martingale. Thus the stopping times τ_k also work for M^σ .

If M^{τ_k} is an L^2 -martingale, then it's also $M_{\sigma \wedge t}^{\tau_k} = (M^\sigma)_t^{\tau_k}$ by [Lemma 1.9](#) applied to the submartingale $(M^{\tau_k})^2$,

$$E[M_{\sigma \wedge \tau_k \wedge t}^2] \leq E[M_{\tau_k \wedge t}^2] + E[M_0^2]$$

□

Lemma 1.26. Suppose M is a cadlag local martingale, and there is a constant c such that $|M_t(\omega) - M_{t-}(\omega)| \leq c$ for all $t \in \mathbb{R}_+$ and $\omega \in \Omega$. Then M is a local L^2 -martingale.

Proof. Let $\tau_k \nearrow \infty$ be stopping times such that M^{τ_k} is a martingale. Let

$$\rho_k := \inf\{t \geq 0 : |M_t| \text{ or } |M_{t-}| \geq k\}$$

By the cadlag assumption, each path $t \rightarrow M_t(\omega)$ is bounded in each bounded time interval, so $\rho_k \nearrow \infty$ when $k \rightarrow \infty$. Let $\sigma_k = \rho_k \wedge \tau_k$. Then $\sigma_k \nearrow \infty$ when $k \rightarrow \infty$, and M^{σ_k} is a martingale for each k . Furthermore (make a draw if unsure about this bound),

$$|M_t^{\sigma_k}| = |M_{\tau_k \wedge \rho_k \wedge t}| \leq \sup_{0 \leq s < \rho_k} |M_s| + |M_{\rho_k} - M_{\rho_k-}| \leq k + c$$

So M^{σ_k} is a bounded process, and in particular an L^2 -process. □

Recall that the usual conditions on the filtration $\{\mathcal{F}_t\}$ meant that the filtration is complete and right-continuous.

Theorem 1.27. *(Fundamental Theorem of Local Martingales) Assume $\{\mathcal{F}_t\}$ is complete and right-continuous, M is a cadlag local martingale and $c > 0$. Then there exists cadlag local martingale \tilde{M} and A such that the jumps of \tilde{M} are bounded by c , A is a finite variation process, and $M = \tilde{M} + A$. The last lemma says that \tilde{M} is an L^2 -local martingale.*

Combining this theorem with the last lemma, we get the following corollary:

Corollary 1.28. *Assume $\{\mathcal{F}_t\}$ is complete and right-continuous. Then a cadlag local martingale M can be written as a sum $M = \tilde{M} + A$ of a cadlag local L^2 -martingale \tilde{M} and a local martingale A that is an FV process.*

Definition 1.29. *A cadlag process Y is a semimartingale if it can be written as*

$$Y_t = Y_0 + M_t + V_t$$

where M is a cadlag local martingale, V is a cadlag FV process, and $M_0 = V_0 = 0$.

1.4 Quadratic variation for semimartingales

We study the quadratic variation and covariation for semimartingales. Recall that for two independent Brownian motions B and Y , $[B]_t = t$ and $[B, Y] = 0$. For the Poisson process, if N is a homogeneous rate α Poisson process, and $M_t = N_t - \alpha t$, then $[M] = [N] = N$. If \tilde{N} is an independent rate $\tilde{\alpha}$ Poisson process with $\tilde{M}_t = \tilde{N}_t - \tilde{\alpha}t$, $[M, \tilde{M}] = 0$.

The following result is an existence theorem of quadratic variation for local martingales.

Theorem 1.30. *Let M be a right-continuous local martingale with respect to the filtration \mathcal{F}_t . Then the quadratic variation process $[M]$ exists in the sense of*

$$\lim_{\text{mesh}(\pi) \rightarrow 0} \sum_{i=0}^{m(\pi)-1} (B_{t_{i+1}} - B_{t_i})^2 = t \quad \text{in } L^2(P). \quad (1.11)$$

There is a version of $[M]$ with the following properties: $[M]$ is real-valued, right-continuous, nondecreasing adapted process such that $[M]_0 = 0$.

Suppose M is an L^2 -martingale. Then the convergence in (1.11) for $Y = M$ holds also in L^1 , namely for any $t \in \mathbb{R}_+$,

$$\lim_{n \rightarrow \infty} E \left| \sum_i (M_{t_{i+1}^n} - M_{t_i^n})^2 - [M]_t \right| = 0 \quad (1.12)$$

for any sequence of partitions $\pi^n = \{t_i^n\}_i$ of $[0, t]$ with $\text{mesh}(\pi^n) \xrightarrow{n \rightarrow \infty} 0$. Furthermore,

$$E([M]_t) = E(M_t^2 - M_0^2) \quad (1.13)$$

If M is continuous, then so is $[M]$.

The following result states that $[M]^\tau = [M^\tau]$ for local L^2 -martingales. However, the result is valid for local martingales (a proof is of the general fact together with the proof of the theorem above is found in [1])

Lemma 1.31. *Let M be a right-continuous L^2 -martingale or local L^2 -martingale. Let τ be a stopping time. Then $[M]^\tau = [M^\tau]$ in the sense that these processes are indistinguishable.*

Proof. By the last theorem, both processes are right-continuous. By Lemma 1.14 from the first chapter, it's enough that for each fixed time t , $[M_t]^\tau = [M_t^\tau]$ almost surely.

Step 1 Suppose first that τ take discrete values $u_1 < u_2 < \dots$ with $u_j \nearrow \infty$. Fix t and consider a sequence of partitions $\pi^n = \{t_i^n\}_i$ of $[0, t]$ with $\text{mesh}(\pi^n) \rightarrow 0$ as $n \rightarrow \infty$. For any u_j ,

$$\sum_i (M_{u_j \wedge t_{i+1}^n} - M_{u_j \wedge t_i^n})^2 \rightarrow [M]_{u_j \wedge t} \quad \text{in probability, as } n \rightarrow \infty$$

Observe that if $u_j > t$, the random variable above is the same of all j large enough, so they are finite many of them. Therefore we can consider a sub-partition of π^n such that for every j , the convergence above is almost surely. Denote again by π^n to this sub-partition.

Fix ω at which the convergence happens. Let $u_j = \tau(\omega)$. We have

$$\begin{aligned} [M]_t^\tau(\omega) &\stackrel{\text{Theorem 1.30}}{=} \lim_{n \rightarrow \infty} \sum_i (M_{t_{i+1}^n}^\tau(\omega) - M_{t_i^n}^\tau(\omega))^2 \\ &= \lim_{n \rightarrow \infty} \sum_i (M_{\tau \wedge t_{i+1}^n}(\omega) - M_{\tau \wedge t_i^n}(\omega))^2 \\ &= \lim_{n \rightarrow \infty} \sum_i (M_{u_j \wedge t_{i+1}^n} - M_{u_j \wedge t_i^n})^2 \\ &= [M]_{u_j \wedge t}(\omega) = [M]_{\tau \wedge t}(\omega) = [M]_t^\tau(\omega) \end{aligned}$$

Step 2 Let τ be an arbitrary stopping time, but now assume M is an L^2 -martingale. Let τ_n the stopping time approximation such that $\tau_n \searrow \tau$. We apply

$$|[X]_t - [Y]_t| \leq [X - Y]_t + 2[X - Y]_t^{1/2}[Y]_t^{1/2} \quad a.s \quad (1.14)$$

to $X = M^{\tau_n}, Y = M^\tau$, (1.13) and Cauchy Schwartz to obtain

$$\begin{aligned} E(|[M^{\tau_n}]_t - [M^\tau]_t|) &\leq E([M^{\tau_n} - M^\tau]_t) + 2E([M^{\tau_n} - M^\tau]_t^{1/2}[M^\tau]_t^{1/2}) \\ &\leq E([M^{\tau_n} - M^\tau]_t^2) + 2E([M^{\tau_n} - M^\tau]_t)^{1/2}E([M^\tau]_t)^{1/2} \\ &= E([M_{t \wedge \tau_n} - M_{t \wedge \tau}]^2) + 2\{E(M_{\tau_n \wedge t} - M_{t \wedge \tau})^2\}^{1/2}\{E(M_{t \wedge \tau})^2\}^{1/2} \\ &\leq E([M_{t \wedge \tau_n}]^2) - E([M_{t \wedge \tau}]^2) + 2(E(M_{\tau_n \wedge t})^2 - E(M_{t \wedge \tau})^2)^{1/2}\{E(M_t)^2\}^{1/2} \end{aligned}$$

In the last step was used (1.4) two times. First we used that

$$\begin{aligned} E([M_{t \wedge \tau_n} - M_{t \wedge \tau}]^2) &= E(M_{t \wedge \tau_n}^2) - 2E\{E(M_{\tau_n \wedge t}|\mathcal{F}_{\tau \wedge t})M_{\tau \wedge t}\} + E(M_{\tau \wedge t}^2) \\ &= E(M_{t \wedge \tau_n}^2) - E(M_{\tau \wedge t}^2) \end{aligned}$$

In the second sum we applied (1.4) to the submartingale M^2 to get

$$E(M_{\tau \wedge t}^2)^{1/2} \leq E(M_t^2)^{1/2}$$

Using this inequality together with M is an L^2 -martingale, we conclude that $[M^{\tau_n}]_t \rightarrow [M^\tau]_t$ in L^1 as $n \rightarrow \infty$, if we can show that

$$E(M_{\tau_n \wedge t}^2) \rightarrow E(M_{\tau \wedge t}^2)$$

But we know that $M_{\tau_n \wedge t}^2 \rightarrow M_{\tau \wedge t}^2$ almost surely by right continuity, Also by optional stopping

$$0 \leq M_{\tau_n \wedge t}^2 \leq E(M_t^2|\mathcal{F}_{\tau_n \wedge t})$$

The sequence of conditional expectations $\{E(M_t^2|\mathcal{F}_{\tau_n \wedge t})\}$ is uniformly integrable, so it is $\{M_{\tau_n \wedge t}^2\}$, and this implies the convergence in L^1 .

We have shown that $[M^{\tau_n}]_t \rightarrow [M^\tau]_t$ in L^1 as $n \rightarrow \infty$. By Step 1, $[M^{\tau_n}]_t = [M]_t^{\tau_n} = [M]_{\tau_n \wedge t} \rightarrow [M]_{\tau \wedge t}$ by the right continuity of the process $[M]$, so $[M^\tau]_t = [M]_{\tau \wedge t}$ for L^2 martingales.

Step 3 Now we consider L^2 –local martingales. Let $\{\sigma_k\}$ be stopping times with $\sigma_k \nearrow \infty$ and M^{σ_k} is an L^2 –martingale for each k . By Step 2,

$$[M^{\sigma_k \wedge \tau}]_t = [M^{\sigma_k}]_{\tau \wedge t}$$

On the event $\{\sigma_k > t\}$, throughout the time interval $[0, t]$, $M^{\sigma_k \wedge \tau} = M^\tau$, $M^{\sigma_k} = M$. Hence the corresponding square sums also agree. Taking the mesh of the partition go to zero we conclude that $[M^{\sigma_k \wedge \tau}]_t = [M^\tau]_t$ and by right continuity, $[M^{\sigma_k}]_s = [M]_s$ for all $s \in [0, t]$. If we take $s = \tau \wedge t$, we get the desired equality $[M^\tau]_t = [M]_{\tau \wedge t}$

□

Theorem 1.32.

- i) If M is a right-continuous L^2 –martingale, then $M_t^2 - [M]_t$ is a martingale.
- ii) If M is a right-continuous local L^2 –martingale, then $M_t^2 - [M]_t$ is a local martingale.

Proof.

- i) Let $s < t$ and $A \in \mathcal{F}_s$. Let $0 = t_0 < \dots < t_m = t$ be a partition of $[0, t]$, and assume that $s = t_l$ for some $l \in \{0, \dots, m-1\}$. We compute

$$\begin{aligned} E[1_A(M_t^2 - M_s^2) - [M]_t - [M]_s] &= E \left[1_A \left(\sum_{i=l}^{m-1} (M_{t_{i+1}}^2 - M_{t_i}^2) - [M]_t - [M]_s \right) \right] \\ &= E \left[1_A \left(\sum_{i=l}^{m-1} (M_{t_{i+1}} - M_{t_i})^2 - [M]_t - [M]_s \right) \right] \\ &= E \left[1_A \left(\sum_{i=0}^{m-1} (M_{t_{i+1}} - M_{t_i})^2 - [M]_t \right) \right] \\ &\quad + E \left[1_A \left([M]_s - \sum_{i=0}^{l-1} [M]_s - (M_{t_{i+1}} - M_{t_i})^2 \right) \right] \end{aligned}$$

In the second inequality we used

$$E[M_{t_{i+1}}^2 - M_{t_i}^2 | \mathcal{F}_{t_i}] = E[(M_{t_{i+1}} - M_{t_i})^2 | \mathcal{F}_{t_i}]$$

In the last equation, we take the mesh go to zero, and both sums vanish by the L^1 –convergence of the quadratic variation for L^2 –martingales ([Theorem 1.30](#)). Because $A \in \mathcal{F}_s$ is arbitrary, by the definition of conditional expectation we are done.

ii) let $X = M^2 - [M]$ for a local L^2 -martingale M . Let $\{\tau_k\}$ be a localizing sequence for M . By the last part, $(M^{\tau_k})_t^2 - [M^{\tau_k}]_t$ is a martingale. Now we use that $[M^{\tau_k}]_t = [M]_{\tau_k \wedge t}$, so $M_{\tau_k \wedge t}^2 - [M]_{\tau_k \wedge t} = X^{\tau_k}$ is a martingale. Therefore $\{\tau_k\}$ is a localizing sequence for X , and we are done. □

Remark 1.33. From [Theorem 1.30](#), the quadratic covariation $[M, N]$ of two right-continuous local martingales M and N exist. Because $[M, N]$ is the difference of two increasing processes, $[M, N]$ is a finite variation process.

Lemma 1.34. Let M and N be cadlag L^2 -martingales or local L^2 -martingales. Let τ be a stopping time. Then $[M^\tau, N] = [M^\tau, N^\tau] = [M, N]^\tau$

Proof. $[M^\tau, N^\tau] = [M, N]^\tau$ follows from

$$[X, Y] := \left[\frac{1}{2}(X + Y) \right] - \left[\frac{1}{2}(X - Y) \right] \quad (1.15)$$

and [Lemma 1.31](#). To prove the first equality, consider a partition of $[0, t]$. If $0 < \tau \leq t$, let l be the index such that $t_l < \tau \leq t_{l+1}$. Then

$$\sum_i (M_{t_{i+1}}^\tau - M_{t_i}^\tau)(N_{t_{i+1}} - N_{t_i}^\tau + N_{t_i}^\tau - N_{t_i}) = (M_\tau - M_{t_l})(N_{t_{l+1}} - N_\tau)1\{0 < \tau \leq t\} \quad (1.16)$$

This equality is true because $\tau > t$, then $\tau \wedge t_i = t_i$ for all i , and the left sum vanishes. If $0 < \tau \leq t$, then for all indexes with $i > l$, $M_{t_{i+1}}^\tau - M_{t_i}^\tau = M_\tau - M_\tau = 0$, so the sum above vanishes for $i > l$, and if $i < l$, $(N_{t_{i+1}} - N_{t_{i+1}}^\tau + N_{t_i}^\tau - N_{t_i}) = 0$, so the only non zero term in the sum in (1.16) is when $i = l$, and is precisely the right-side of the equality (1.16). Note that if $\tau = 0$, both sides in (1.16) vanish. The equation above can be written as

$$\sum_i (M_{t_{i+1}}^\tau - M_{t_i}^\tau)(N_{t_{i+1}} - N_{t_i}) = (M_\tau - M_{t_l})(N_{t_{l+1}} - N_\tau)1\{0 < \tau \leq t\} + \sum_i (M_{t_{i+1}}^\tau - M_{t_i}^\tau)(N_{t_{i+1}}^\tau - N_{t_i}^\tau)$$

If we let the mesh of the partition goes to zero, using the cadlag property, the equation above implies that

$$[M^\tau, N] = (M_\tau - M_{\tau-})(N_\tau - N_\tau)1\{0 < \tau \leq t\} + [M^\tau, N^\tau] = [M^\tau, N^\tau]$$

□

Theorem 1.35.

- i) If M, N are right-continuous L^2 -martingales, then $MN - [M, N]$ is a martingale.
- ii) If M, N are right-continuous local L^2 -martingales, then $MN - [M, N]$ is a local martingale.

Proof. Write $MN - [M, N] = \frac{1}{2}\{(M + N)^2 - [N + M]\} - \frac{1}{2}\{M^2 - [M]\} - \frac{1}{2}\{N^2 - [N]\}$, and apply the last lemma to the martingales/local martingales $M, N, M + N$ \square

We want to extend this result to semimartingales. Before we state a lemma.

Lemma 1.36. *Let f, g be real-valued cadlag functions on $[0, T]$ and assume $f \in BV([0, T])$. Then*

$$[f, g](T) = \sum_{s \in (0, T]} (f(s) - f(s-))(g(s) - g(s-))$$

and the sum above converges absolutely

Corollary 1.37. *Let M be a cadlag local martingale, V a cadlag FV process, $M_0 = V_0 = 0$, and $Y = Y_0 + M + V$ the cadlag semimartingale. Then the cadlag quadratic variation process $[Y]$ exists and satisfies*

$$\begin{aligned} [Y]_t &= [M]_t + 2[M, V]_t + [V]_t \\ &= [M]_t + 2 \sum_{s \in (0, t]} \Delta M_s \Delta V_s + \sum_{s \in (0, t]} (\Delta V_s)^2 \end{aligned} \tag{1.17}$$

Furthermore, $[Y]^\tau = [Y]^\tau$ for any stopping time τ and the covariation $[X, Y]$ exists for any pair of cadlag semimartingales.

Proof. We already know the existence properties of $[M]$. According to [Lemma 1.36](#), the two sums in (1.17) converge absolutely. It can be proved that in that case, the process given in (1.17) is a cadlag process. [Theorem 1.30](#) and [Lemma 1.36](#) together imply that (1.17) is the limit in probability of sums $\sum_i (Y_{t_{i+1}} - Y_{t_i})^2$ as $mesh(\pi) \rightarrow 0$. Denote the process in (1.17) by U_t . By definition, for $s < t$, $U_s \leq U_t$, then this happens simultaneously for all pair of rationals $s < t$. By taking limits, and using the cadlag property, we can extend the monotonicity to all times $s < t$. Thus U is an increasing process and gives a version of $[Y]$ with nonnegative paths. This proves the existence of $[Y]$. The equality $[Y]^\tau = [Y]^\tau$ follows from applying [Lemma 1.31](#)

at each term in (1.17). The covariation $[X, Y]$ exists because the existence of $[X + Y]$, $[X - Y]$ and (1.15). \square

1.5 Doob-Meyer decomposition

Throughout this section we work with a fixed probability space (Ω, \mathcal{F}, P) with a filtration $\{\mathcal{F}_t\}$ with satisfies the *usual conditions*.

Definition 1.38. *The predictable σ -algebra \mathcal{P} on the space $\mathbb{R}_+ \times \Omega$ in the sub- σ -algebra of $\mathcal{B}(\mathbb{R}^d) \otimes \mathcal{F}$ generated by left-continuous adapted processes. More precisely, \mathcal{P} is generated by events of the form $\{(t, \omega) : X_t(\omega) \in B\}$, where X is an adapted, left-continuous process and $B \in \mathcal{B}(\mathbb{R}^d)$. Recall that such processes are progressively measurable by Lemma 1.13 from the first chapter.*

Intuitively, a predictable process permit us obtain X_t if we know X_s for $s < t$. Any \mathcal{P} -measurable function $X : \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}$ is called a *predictable process*. We state a similar theorem of Theorem 1.30.

Theorem 1.39. *Assume the filtration $\{\mathcal{F}_t\}$ satisfies the usual conditions.*

- i) *Let M be a right-continuous square-integrable martingale. Then there is a unique predictable process $\langle M \rangle$ (called the predictable quadratic covariation) such that $M^2 - \langle M \rangle$ is a martingale.*
- ii) *Let M be a right-continuous local square-integrable martingale. Then there is a unique predictable process $\langle M \rangle$ such that $M^2 - \langle M \rangle$ is a local martingale.*

Uniqueness above means uniqueness up to indistinguishableness. We can define the predictable covariation by

$$\langle M, N \rangle = \frac{1}{4} \langle M + N \rangle - \frac{1}{4} \langle M - N \rangle$$

From the uniqueness from Theorem 1.39 and Theorem 1.30, if $[M]$ is predictable, then $[M] = \langle M \rangle$.

Proposition 1.40. *Assume the filtration satisfies the usual conditions.*

- i) *Suppose M is a continuous L^2 -martingale. Then $[M] = \langle M \rangle$.*

ii) Suppose M is a right-continuous L^2 -martingale with stationary and independent increments: for all $s, t \geq 0$, $M_{s+t} - M_s$ is independent of \mathcal{F}_s , and has the same distribution as $M_t - M_0$. Then $\langle M \rangle_t = t \cdot E[M_1^2 - M_0^2]$.

Proof.

i) This follows because M is continuous, hence predictable, and the uniqueness part of [Theorem 1.39](#).

ii) The deterministic, continuous function $t \rightarrow E[M_t^2 - M_0^2]$ is predictable. For any $t > 0$, and integer k ,

$$\begin{aligned} E[M_{kt}^2 - M_0^2] &= \sum_{i=1}^{k-1} E[M_{(j+1)t}^2 - M_{jt}^2] = \sum_{i=1}^{k-1} E[(M_{(j+1)t} - M_{jt})^2] \\ &= kE[(M_t - M_0)^2] = kE[M_t^2 - M_0^2] \end{aligned}$$

Using this twice, for any rational k/n ,

$$E[M_{k/n}^2 - M_0^2] = kE[M_{1/n}^2 - M_0^2] = (k/n)E[M_1^2 - M_0^2]$$

Given an irrational $t > 0$, pick rationals $q_m \searrow t$. Fix $T \geq q_m$. By right-continuity of paths, $M_{q_m} \rightarrow M_t$ almost surely. Using the bound

$$0 \leq M_{q_m}^2 \leq E[M_T^2 | \mathcal{F}_{q_m}]$$

then $\{M_{q_m}^2\}_m$ is uniformly integrable. This gives the convergence $E[M_{q_m}^2] \rightarrow E[M_t^2]$, so

$$E[M_t^2 - M_0^2] = tE[M_1^2 - M_0^2].$$

The martingale property follows from

$$\begin{aligned} E[M_t^2 | \mathcal{F}_s] &= M_s^2 + E[M_t^2 - M_s^2 | \mathcal{F}_s] = M_s^2 + E[(M_t - M_s)^2 | \mathcal{F}_s] \\ &= M_s^2 + E[(M_{t-s} - M_0)^2 | \mathcal{F}_s] = M_s^2 + (t-s)[M_1^2 - M_0^2] \end{aligned}$$

□

Example 1.41. For a standard Brownian motion, $\langle B \rangle_t = [B]_t = t$.

For a compensated Poisson process $M_t = N_t - \alpha t$,

$$\begin{aligned}\langle M \rangle_t &= tE[M_1^2] = tE[(N_1 - \alpha t)^2] = t(EN_1^2 - 2\alpha EN_1 + \alpha^2) \\ &= t(\alpha + \alpha^2 - 2\alpha^2 + \alpha^2) = \alpha t\end{aligned}$$

Definition 1.42. An increasing process A is natural if for every bounded cadlag martingale M

$$E \int_{(0,t]} M(s) dA(s) = E \int_{(0,t]} M(s-) dA(s) \quad \text{for } 0 < t < \infty. \quad (1.18)$$

Lemma 1.43. Let A be an increasing process and M a bounded cadlag martingale. If A is continuous, then (1.18) holds.

Proof. Recall that a cadlag path $\omega \rightarrow M(s, \omega)$ has at most countable many discontinuities. If A is continuous, then the Lebesgue-Stieltjes measure of every singleton is zero, because $\Lambda_A(\{s\}) = A(s) - A(s-) = 0$. Therefore, the measure of countable sets is also zero. Thus

$$\int_{(0,t]} (M(s) - M(s-)) dA(s) = 0$$

□

It can be proved that an increasing process is natural if and only if it is predictable.

Definition 1.44. for $0 < u < \infty$, let \mathcal{T}_u be the collection of stopping times τ that satisfy $\tau \leq u$. A process is of class DL if the random variables $\{X_\tau : \tau \in \mathcal{T}_u\}$ are uniformly integrable for each $0 < u < \infty$.

Lemma 1.45. A right-continuous nonnegative submartingale is of class DL.

Proof. Simply recall the inequality

$$0 \leq X_\tau \leq E[X_u | \mathcal{F}_\tau]$$

and the uniform integrability of the conditional expectations. □

Now we state the main result.

Theorem 1.46. (*Doob-Meyer Decomposition*) Assume the underlying filtration is complete and right-continuous. Let X be a right-continuous submartingale of class DL. Then there is an increasing natural process A , unique up to indistinguishableness, such that $X - A$ is a martingale.

Applying this result to a right-continuous martingale M , as M^2 is a submartingale, we can define $\langle M \rangle$ as the unique increasing, natural process such that $M_t^2 - \langle M \rangle_t$ is a martingale, given by the Doob-Meyer decomposition.

1.6 Spaces of martingales

Definition 1.47. Given a probability space (ω, \mathcal{F}, P) with filtration $\{\mathcal{F}_t\}$, let \mathcal{M}_2 the space of L^2 -cadlag martingales on this space with respect to \mathcal{F}_t . The subspace of \mathcal{M}_2 of continuous L^2 -martingales is \mathcal{M}_2^c .

For $M \in \mathcal{M}_2$, we define the quantity

$$\|M\|_{\mathcal{M}_2} := \sum_{k=1}^{\infty} 2^{-k} (1 \wedge \|M_k\|_{L^2(P)}) \quad (1.19)$$

Note that for $a, b \geq 0$, the following inequality holds :

$$1 \wedge (a, b) \leq 1 \wedge a + 1 \wedge b$$

Therefore $\|M+N\|_{\mathcal{M}_2} \leq \|M\|_{\mathcal{M}_2} + \|N\|_{\mathcal{M}_2}$. Finally, observe that as in L^p spaces, $\|M-N\| = 0$ if M and N are indistinguishable. So we will consider in this case two martingales M, N as equal if they are indistinguishable, and define

$$d_{\mathcal{M}_2}(M, N) := \|M - N\|_{\mathcal{M}_2} \quad (1.20)$$

this defines a metric on \mathcal{M}_2 .

Theorem 1.48. Assume the underlying probability space (ω, \mathcal{F}, P) and the filtration $\{\mathcal{F}_t\}$ is complete. Let indistinguishable processes be interpreted as equal. Then \mathcal{M}_2 is a complete metric space under the metric $d_{\mathcal{M}_2}$. The subspace \mathcal{M}_2^c is closed, and hence a complete metric space also.

Proof. Suppose $M \in \mathcal{M}_2$ and $\|M\|_{\mathcal{M}_2} = 0$. Then $E[M_k^2] = 0$ for each $k \in \mathbb{N}$. Since M_t^2 is a submartingale, $E[M_t^2] \leq E[M_k^2]$ for $t \leq k$, and consequently, $E[M_t^2] = 0$ for all $t \geq 0$. In particular, for each t , $P(M_t = 0) = 1$. Now we consider $\Omega_0 = \cap_{q \in \mathbb{Q}_+} \{M_q = 0\}$, so $P(\Omega_0 = 1) = 1$. By right-continuity, $M_t(\omega) = 0$ for all $t \geq 0$ almost surely. This proves that M is indistinguishable from 0, and therefore $\|M\|_{\mathcal{M}_2} = 0$ if and only if $M \equiv 0$ where \equiv is the equivalence relation of indistinguishableness. The triangle inequality was proved above, and the symmetry follows by definition. We conclude that $d_{\mathcal{M}_2}$ defines a metric in \mathcal{M}_2 .

Now we need to prove the completeness. Let $\{M^{(n)} : n \in \mathbb{N}\}$ be a Cauchy sequence in \mathcal{M}_2 . We need to show that exists some $M \in \mathcal{M}_2$ such that $d_{\mathcal{M}_2}(M^n, M) \rightarrow 0$. Given $t \leq k \in \mathbb{N}$, using that $(M_t^{(m)} - M_t^{(n)})^2$ is a submartingale, and the definition (1.19) we obtain

$$\begin{aligned} 1 \wedge E[(M_t^{(m)} - M_t^{(n)})^2]^{1/2} &\leq 1 \wedge E[(M_k^{(m)} - M_k^{(n)})^2]^{1/2} \\ &\leq 2^k \|M^{(m)} - M^{(n)}\|_{\mathcal{M}_2} \end{aligned}$$

This implies that $\{M_t^{(n)} : n \in \mathbb{N}\}$ is a Cauchy sequence in $L^2(P)$ for each $t \geq 0$. As $L^2(P)$ is complete, there exists some $Y_t \in L^2(P)$ such that $M_t^n \rightarrow Y_t$ in $L^2(P)$. In particular, $M_t^n \rightarrow Y_t$ in $L^1(P)$, so for $s < t$, $A \in \mathcal{F}_s$, from the equality $E[1_A M_t^n] = E[1_A M_t^n]$ we take $n \rightarrow \infty$ to deduce that $E[1_A Y_t] = E[1_A Y_s]$, so Y_t is a martingale for each $t \geq 0$. However, we don't know if the cadlag property holds for Y_t . To handle that, we use (1.6) to obtain the inequality

$$P\left(\sup_{0 \leq t \leq k} |M_t^{(m)} - M_t^{(n)}| \geq \epsilon\right) \leq \frac{1}{\epsilon^2} E[(M_k^{(m)} - M_k^{(n)})^2] \quad (1.21)$$

We can choose a subsequence $\{n_k\}$ such that

$$P\left(\sup_{0 \leq t \leq k} |M_t^{(n_{k+1})} - M_t^{(n_k)}| \geq 2^{-k}\right) \leq 2^{-2k} \quad (1.22)$$

To verify this, take $n_0 = 1$, and if n_{k-1} has been chose, pick $n_k > n_{k-1}$ such that

$$\|M^{(m)} - M^{(n)}\|_{\mathcal{M}_2} \leq 2^{-3k}$$

for all $n, m \geq n_k$ (such n_k exists because $M^{(n)}$ is Cauchy). Then for $m \geq n_k$,

$$1 \wedge E[(M_k^{(m)} - M_k^{(n_k)})^2]^{1/2} \leq 2^k \|M^{(m)} - M^{(n)}\|_{\mathcal{M}_2} \leq 2^{-2k}$$

So $1 \wedge E[(M_k^{(m)} - M_k^{(n_k)})^2]^{1/2} = E[(M_k^{(m)} - M_k^{(n_k)})^2]^{1/2} \leq 2^{-2k}$. Now choose $\epsilon = 2^{-k}$ in (1.21) to obtain (1.22). By the Borel-Cantelli lemma, there exists Ω_1 with $P(\Omega_1) = 1$ such that for

every $\omega \in \Omega_1$,

$$\sup_{0 \leq t \leq k} |M_t^{(n_{k+1})}(\omega) - M_t^{(n_k)}(\omega)| < 2^{-k}$$

for all but finitely many k 's. It follows that the sequence of cadlag functions $t \rightarrow M_t^{(n_k)}$ is Cauchy under the uniform metric over any bounded time interval $[0, T]$. As cadlag functions form a complete metric space under this metric, we conclude that for each $T < \infty$ there exists a cadlag process $\{N_t^{(T)}(\omega) : 0 \leq t \leq T\}$ such that $M_t^{(n_k)} \rightarrow N_t^{(T)}$ uniformly on $[0, T]$, when $k \rightarrow \infty$. $N_t^{(S)}$ and $N_t^{(T)}$ must agree if $t \in [0, S \wedge T]$, because both are the limit of the same sequence. Thus we can define one cadlag function $t \rightarrow M_t^{(\omega)}$ on \mathbb{R}_+ , and for $\omega \in \Omega_1$, $M_t^{(n_k)}(\omega) \rightarrow M_t(\omega)$ uniformly on each bounded interval $[0, T]$. If $\omega \notin \Omega_1$, we define $M_t(\omega) = 0$. As \mathcal{F}_t is complete for each t , $\Omega_1 \in \mathcal{F}_t$. In particular, $M_t(\omega)$ is \mathcal{F}_t -measurable. We know also that $M_t^{(n_k)} \rightarrow Y_t$ in $L^2(P)$, so $M_t = Y_t$ almost surely, so M is a martingale, and $M_t^{(n_k)} \rightarrow M_t$ in $L^2(P)$ when $k \rightarrow \infty$. To prove that $\|M^{(n_k)} - M\|_{\mathcal{M}_2} \rightarrow 0$, we write

$$\sum_{k=1}^{\infty} 2^{-k} (1 \wedge \|M_k^{(n_k)} - M_k\|_{L^2(P)}) \leq \sum_{k=1}^l 2^{-k} (1 \wedge \|M_k^{(n_k)} - M_k\|_{L^2(P)}) + \sum_{k=l+1}^{\infty} 2^{-k}$$

For fixed l and $\epsilon > 0$, the first sum is bounded by ϵ if k is large enough, and the second sum is $\leq c2^{-l-1}$. Taking $k \rightarrow \infty$, the left side sum is bounded by $c2^{-l-1}$. But $l \geq 1$ is arbitrary, so we are done.

If all $M^{(n)}$ are continuous, the uniform limit produces a continuous function M , so \mathcal{M}_2^c is complete under the same metric. \square

By adapting the argument above from (1.21) onwards, we get this useful consequence of convergence in \mathcal{M}_2 .

Lemma 1.49. *Suppose $\|M^{(n)} - M\|_{\mathcal{M}_2} \rightarrow 0$ as $n \rightarrow \infty$. Then for each $T < \infty$ and $\epsilon > 0$,*

$$\lim_{n \rightarrow \infty} P\left(\sup_{0 \leq t \leq T} |M_t^{(n)} - M_t| \geq \epsilon\right) = 0 \quad (1.23)$$

Furthermore, there exists a subsequence $\{M^{(n_k)}\}$ and an event Ω_0 such that $P(\Omega_0) = 1$ and for each $\omega \in \Omega_0$ and $T < \infty$,

$$\lim_{n \rightarrow \infty} \sup_{0 \leq t \leq T} |M_t^{(n_k)}(\omega) - M_t(\omega)| = 0$$

When (1.23) holds for all $T < \infty$ and each $\epsilon > 0$, it is called uniform convergence in probability on compact sets.

References

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