

Math 562 Fall 2020

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Outline

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- 1 **General Info**
- 2 4.4 The Bracket of Two Continuous Local Martingales
- 3 4.5 Continuous Semimartingales
- 4 5.1 The Construction of Stochastic Integrals

I will post HW4 later today. HW4 is due 10/16.

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Proposition 4.18 (Kunita-Watanabe)

Let M and N be two continuous local martingales and let H and K be two measurable processes. Then, a.s.,

$$\int_0^\infty |H_s| |K_s| |d\langle M, N \rangle_s| \leq \left(\int_0^\infty H_s^2 d\langle M, M \rangle_s \right)^{1/2} \left(\int_0^\infty K_s^2 d\langle N, N \rangle_s \right)^{1/2}$$

Proof of Proposition 4.18

In this proof, we use the special notation $\langle M, N \rangle_s^t = \langle M, N \rangle_t - \langle M, N \rangle_s$ for $0 \leq s \leq t$. The first step of the proof is to observe that we have a.s. for every choice of the rationals $s < t$ (and also by continuity for every reals $s < t$)

$$|\langle M, N \rangle_s^t| \leq \sqrt{\langle M, M \rangle_s^t} \sqrt{\langle N, N \rangle_s^t}.$$

Indeed, this follows from the approximations of $\langle M, M \rangle$ and $\langle M, N \rangle$ in Theorem 4.9 and in Proposition 4.15 respectively, together with the Cauchy-Schwarz inequality.

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Proof of Proposition 4.18 (cont)

From now on, we fix ω such that the inequality of the last display holds for every $s < t$, and we argue with this value ω .

Note also that, for every $s < t$,

$$\int_s^t |d\langle M, N \rangle_u| \leq \sqrt{\langle M, M \rangle_s^t} \sqrt{\langle N, N \rangle_s^t}. \quad (1)$$

Indeed, we use Proposition 4.2, noting that, for any partition $s = t_0 < t_1 < \dots < t_p = t$, we have

$$\begin{aligned} \sum_{j=1}^p |\langle M, N \rangle_{t_{j-1}}^{t_j}| &\leq \sum_{j=1}^p \sqrt{\langle M, M \rangle_{t_{j-1}}^{t_j}} \sqrt{\langle N, N \rangle_{t_{j-1}}^{t_j}} \\ &\leq \left(\sum_{j=1}^p \langle M, M \rangle_{t_{j-1}}^{t_j} \right)^{1/2} \left(\sum_{j=1}^p \langle N, N \rangle_{t_{j-1}}^{t_j} \right)^{1/2} \\ &= \sqrt{\langle M, M \rangle_s^t} \sqrt{\langle N, N \rangle_s^t}. \end{aligned}$$

Proof of Proposition 4.18 (cont)

We then get that, for every bounded Borel subset A of \mathbb{R}_+ ,

$$\int_A |d\langle M, N \rangle_u| \leq \sqrt{\int_A d\langle M, M \rangle_u} \sqrt{\int_A d\langle N, N \rangle_u}.$$

When $A = [s, t]$, this is (1). If A is a finite union of intervals, this follows from (1) and another application of the Cauchy-Schwarz inequality. A monotone class argument shows that the inequality of the last display remains valid for any bounded Borel set A .

Next let $h = \sum_{j=1}^p \lambda_j 1_{A_j}$ and $k = \sum_{j=1}^p \mu_j 1_{A_j}$ be two non-negative simple functions on \mathbb{R}_+ with bounded support contained in $[0, K]$, for some $K > 0$. Here A_1, \dots, A_p is a measurable partition of $[0, K]$, and $\lambda_1, \dots, \lambda_p, \mu_1, \dots, \mu_p$ are non-negative reals. Then

Proof of Proposition 4.18 (cont)

$$\begin{aligned}\int h(s)k(s)|d\langle M, N\rangle_s| &= \sum_{j=1}^p \lambda_j \mu_j \int_{A_j} |d\langle M, N\rangle_s| \\ &\leq \left(\sum_{j=1}^p \lambda_j^2 d\langle M, M\rangle_s \right)^{1/2} \left(\sum_{j=1}^p \mu_j^2 d\langle N, N\rangle_s \right)^{1/2} \\ &= \left(\int h^2(s) d\langle M, M\rangle_s \right)^{1/2} \left(\int k^2(s) d\langle N, N\rangle_s \right)^{1/2},\end{aligned}$$

which gives the desired inequality for simple functions. Since every non-negative Borel function is a monotone increasing limit of simple functions with bounded support, an application of the monotone convergence theorem completes the proof.

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Definition 4.19

A process $X = (X_t)_{t \geq 0}$ is a continuous semimartingale if it can be written in the form

$$X_t = M_t + A_t$$

where M is a continuous local martingale and A is a finite variation process.

The decomposition $X = M + A$ is then unique up to indistinguishability thanks to Theorem 4.8. We say that this is the canonical decomposition of X .

By construction, continuous semimartingales have continuous sample paths. It is possible to define a notion of semimartingale with cadlag sample paths, but we will only deal with continuous semimartingales, and for this reason we sometimes omit the word continuous.

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Definition 4.20

Let $X = M + A$ and $Y = M' + A'$ be the canonical decompositions of two continuous semimartingales X and Y . The bracket $\langle X, Y \rangle$ is the finite variation process defined by

$$\langle X, Y \rangle_t = \langle M, M' \rangle_t.$$

In particular, we have $\langle X, X \rangle_t = \langle M, M \rangle_t$.

Proposition 4.21

Let $0 = t_0^n < t_1^n < \dots < t_p^n = t$ be an increasing sequence of partitions of $[0, t]$ whose mesh tends to 0. Then,

$$\lim_{n \rightarrow \infty} \sum_{j=1}^{p_n} (X_{t_j^n} - X_{t_{j-1}^n})(Y_{t_j^n} - Y_{t_{j-1}^n}) = \langle X, Y \rangle_t$$

in probability.

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in probability.

Proof Proposition 4.21

We treat the case $X = Y$ only. The general case is similar. We have

$$\begin{aligned}\sum_{j=1}^{p_n} (X_{t_j^n} - X_{t_{j-1}^n})^2 &= \sum_{j=1}^{p_n} (M_{t_j^n} - M_{t_{j-1}^n})^2 + \sum_{j=1}^{p_n} (A_{t_j^n} - A_{t_{j-1}^n})^2 \\ &\quad + 2 \sum_{j=1}^{p_n} (M_{t_j^n} - M_{t_{j-1}^n})(A_{t_j^n} - A_{t_{j-1}^n}).\end{aligned}$$

By Theorem 4.9,

$$\lim_{n \rightarrow \infty} \sum_{j=1}^{p_n} (M_{t_j^n} - M_{t_{j-1}^n})^2 = \langle M, M \rangle_t = \langle X, X \rangle_t$$

in probability. On the other hand,

Proof Proposition 4.21 (cont)

$$\begin{aligned} \sum_{j=1}^{p_n} (A_{t_j^n} - A_{t_{j-1}^n})^2 &\leq \left(\sup_{1 \leq j \leq p_n} |A_{t_j^n} - A_{t_{j-1}^n}| \right) \sum_{j=1}^{p_n} |A_{t_j^n} - A_{t_{j-1}^n}| \\ &\leq \left(\int_0^t |dA_s| \right) \sup_{1 \leq j \leq p_n} |A_{t_j^n} - A_{t_{j-1}^n}| \end{aligned}$$

which tends to 0 a.s. when $n \rightarrow \infty$ by the continuity of sample paths of A . The same argument shows that

$$\left| \sum_{j=1}^{p_n} (M_{t_j^n} - M_{t_{j-1}^n})(A_{t_j^n} - A_{t_{j-1}^n}) \right| \leq \left(\int_0^t |dA_s| \right) \sup_{1 \leq j \leq p_n} |M_{t_j^n} - M_{t_{j-1}^n}|$$

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Throughout this chapter, we fix a filtered probability space $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$ and assume that $(\mathcal{F}_t)_{t \geq 0}$ is complete. We write \mathbb{H}^2 for the space of all continuous martingales M which are bounded in L^2 such that $M_0 = 0$, with the usual convention that two indistinguishable processes are identified. Equivalently, $M \in \mathbb{H}^2$ if and only if M is a continuous local martingale such that $M_0 = 0$ and $\mathbb{E}[\langle M, M \rangle_\infty] < \infty$.

If $M \in \mathbb{H}^2$, then $M_t = \mathbb{E}[M_\infty | \mathcal{F}_t]$ where $M_\infty \in L^2$ is the almost sure limit of M_t as $t \rightarrow \infty$. If $M, N \in \mathbb{H}^2$, then $\langle M, N \rangle_\infty$ is well defined and $\mathbb{E}[|\langle M, N \rangle_\infty|] < \infty$. This allows us to define a symmetric bilinear form on \mathbb{H}^2 via the formula

$$(M, N)_{\mathbb{H}^2} = \mathbb{E}[\langle M, N \rangle_\infty] = \mathbb{E}[M_\infty N_\infty],$$

where the second equality comes from Proposition 4.15 (v).

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where the second equality comes from Proposition 4.15 (v).

Clearly $(M, M)_{\mathbb{H}^2} = 0$ if and only if $M = 0$. The scalar product $(M, N)_{\mathbb{H}^2}$ thus yields a norm on \mathbb{H}^2 given by

$$\|M\|_{\mathbb{H}^2} = (M, M)_{\mathbb{H}^2}^{1/2} = \mathbb{E}[\langle M, M \rangle_{\infty}]^{1/2} = \mathbb{E}[M_{\infty} M_{\infty}]^{1/2}.$$

Proposition 5.1

The space \mathbb{H}^2 equipped with the scalar product $(M, N)_{\mathbb{H}^2}$ is a Hilbert space.

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Proof of Proposition 5.1

We need to verify that the vector space \mathbb{H}^2 is complete for the norm $\|M\|_{\mathbb{H}^2}$. Let $(M^n)_{n \geq 1}$ be a Cauchy sequence in \mathbb{H}^2 . We have then

$$\lim_{m, n \rightarrow \infty} \mathbb{E}[(M_\infty^n - M_\infty^m)^2] = \lim_{m, n \rightarrow \infty} (M^n - M^m, M^n - M^m)_{\mathbb{H}^2} = 0.$$

Consequently, the sequence $(M_\infty^n)_{n \geq 1}$ converges in L^2 to a limit, which we denote by Z . On the other hand, Doob's L^2 -inequality and a straightforward passage to the limit show that, for every m, n ,

$$\mathbb{E}[\sup_{t \geq 0} (M_t^n - M_t^m)^2] \leq 4\mathbb{E}[(M_\infty^n - M_\infty^m)^2].$$

Thus

$$\lim_{m, n \rightarrow \infty} \mathbb{E}[\sup_{t \geq 0} (M_t^n - M_t^m)^2] = 0. \quad (2)$$

Hence, for every $t > 0$, M_t^n converges in L^2 , and we want to argue that the limit yields a process with continuous sample paths.

Proof of Proposition 5.1 (cont)

To this end, we use (2) to find an increasing sequence $n_k \uparrow \infty$ such that

$$\mathbb{E} \left[\sum_{k=1}^{\infty} \sup_{t \geq 0} |M_t^{n_k} - M_t^{n_{k+1}}| \right] \leq \sum_{k=1}^{\infty} \mathbb{E} \left[\sup_{t \geq 0} (M_t^{n_k} - M_t^{n_{k+1}})^2 \right]^{1/2} < \infty.$$

Thus we have a.s.

$$\sum_{k=1}^{\infty} \sup_{t \geq 0} |M_t^{n_k} - M_t^{n_{k+1}}| < \infty.$$

and thus the sequence $(M_t^{n_k})_{t \geq 0}$ converges uniformly on \mathbb{R}_+ a.s., to a limit denoted by $(M_t)_{t \geq 0}$. On the zero probability set where the uniform convergence does not hold, we take $M_t = 0$ for every $t > 0$.

Proof of Proposition 5.1 (cont)

Clearly the limiting process M has continuous sample paths and is adapted. Furthermore, from the L^2 -convergence of (M_∞^n) to Z we immediately get by passing to the limit in the identity $M_t^{n_k} = \mathbb{E}[M_\infty^{n_k} | \mathcal{F}_t]$ that $M_t = \mathbb{E}[Z | \mathcal{F}_t]$. Hence $(M_t)_{t \geq 0}$ is a continuous martingale and is bounded in L^2 , so that $M \in \mathbb{H}^2$. The a.s. uniform convergence of $(M_t^{n_k})_{t \geq 0}$ to $(M_t)_{t \geq 0}$ then ensures that $M_\infty = \lim_{k \rightarrow \infty} M_\infty^{n_k} = Z$ a.s. Finally, the L^2 -convergence of (M_∞^n) to $Z = M_\infty$ shows that the sequence (M^n) converges to M in \mathbb{H}^2 .