

Ito–Tanaka formula and local times of semimartingales

We want to extend Ito formula to functions which are not C^2 .

Let X be a (one-dimensional) semimartingale and $f: \mathbb{R} \rightarrow \mathbb{R}$ a convex function.

Recall that for f convex there always exists f'_- (the derivative from the left) and it is an increasing function.

Let $\rho \in C^\infty(\mathbb{R})$ which is compactly supported on $\{x < 0\}$, for example in $(-1, 0)$ and define

$$f_n(x) := n \int \rho(ny) f(x+y) dy$$

which is a smooth function such that $f_n \rightarrow f$ pointwise and for which $f'_n(x) \nearrow f'_-(x)$. By Ito formula

$$f_n(X_t) = f_n(X_0) + \int_0^t f'_n(X_s) dX_s + \frac{1}{2} A_t^{f_n}$$

with $A_t^{f_n} := \int_0^t f''_n(X_s) d[X]_s$ a continuous, increasing process. Eventually by using stopping times we can localize the problem so that f'_- is bounded, moreover we note that by Doob's inequality we have (where $X = M + V$ is the decomposition of the semimartingale)

$$\begin{aligned} \mathbb{E} \left[\sup_{t \in [0, T]} \left| \int_0^t (f'_n(X_s) - f'_-(X_s)) dM_s \right|^2 \right] &\leq \mathbb{E} \left[\left| \int_0^T (f'_n(X_s) - f'_-(X_s)) dM_s \right|^2 \right] \\ &\leq \mathbb{E} \left[\int_0^T (f'_n(X_s) - f'_-(X_s))^2 d[M]_s \right] \rightarrow 0 \end{aligned}$$

by dominated convergence (again maybe put a stopping time to guarantee boundedness). This shows that in probability and uniformly on compact sets (in t)

$$\int_0^t f'_n(X_s) dM_s \rightarrow \int_0^t f'_-(X_s) dM_s.$$

On the hand, always by dominated convergence (decomposing the finite measure dV_s into positive and negative parts)

$$\int_0^t f'_n(X_s) dV_s \rightarrow \int_0^t f'_-(X_s) dV_s.$$

We can conclude that we have the following lemma

Lemma 1. *If X is a continuous semimartingale and f a convex function, then there exists a continuous increasing process $(A_t^f)_t$ such that*

$$f(X_t) = f(X_0) + \int_0^t f'_-(X_s) dX_s + \frac{1}{2} A_t^f, \quad t \geq 0.$$

We now can take $f(x)$ nice and simple convex functions like $|x - a|$, $(x - a)_\pm$ where $(x)_+ = (x \wedge 0)$ and $(x)_- := (-x)_+$. As a corollary of the previous lemma we then have

Theorem 2. (Tanaka's formula) *For any $a \in \mathbb{R}$ there exists a continuous increasing process $(L_t^a)_{t \geq 0}$ such that*

$$\begin{aligned} |X_t - a| &= |X_0 - a| + \int_0^t \operatorname{sgn}(X_s - a) dX_s + L_t^a \\ (X_t - a)_+ &= (X_0 - a)_+ + \int_0^t \mathbb{1}_{X_s > a} dX_s + \frac{1}{2} L_t^a \\ (X_t - a)_- &= (X_0 - a)_- - \int_0^t \mathbb{1}_{X_s \leq a} dX_s + \frac{1}{2} L_t^a \end{aligned}$$

where $\text{sgn}(x) = \mathbb{1}_{x>0} - \mathbb{1}_{x\leq 0}$.

Remark 3. This proves in particular that $|X_t - a|$, $(X_t - a)_\pm$ are semimartingales. The process $(L_t^a)_{t\geq 0}$ is called the local time of X at a .

Proof. Each of the formulas derives from the previous lemma by computing the left derivative of the various convex functions. The missing point is to identify the various increasing processes $A^{\text{sgn}(x-a)}$, $A^{(x-a)_+}$, $A^{(x-a)-}$. Note that

$$X_t - a = (X_t - a)_+ - (X_t - a)_- = X_0 - a + \int_0^t \underbrace{(\mathbb{1}_{X_s > a} + \mathbb{1}_{X_s \leq a})}_{=1} dX_s + \frac{1}{2}(A_t^{(x-a)_+} - A_t^{(x-a)-})$$

so we have

$$0 = X_t - X_0 - \int_0^t dX_s = \frac{1}{2}(A_t^{(x-a)_+} - A_t^{(x-a)-}) \Rightarrow A_t^{(x-a)_+} = A_t^{(x-a)-} =: L_t^a.$$

Moreover

$$|X_t - a| = (X_t - a)_+ + (X_t - a)_- + \int_0^t \underbrace{(\mathbb{1}_{X_s > a} - \mathbb{1}_{X_s \leq a})}_{=\text{sgn}(X_s - a)} dX_s + \underbrace{\frac{1}{2}(A_t^{(x-a)_+} + A_t^{(x-a)-})}_{L_t^a}$$

□

The increasing process $(L_t^a)_{t\geq 0}$ is associated with a measure dL_t^a on \mathbb{R}_+ (times) which represents the time the process X “spent” in a up to time t . We are going to make this precise in the following.

By Ito formula wrt. the semimartingale $(|X_t - a|)_{t\geq 0}$ (with $X = M + V$)

$$\begin{aligned} (X_t - a)^2 &= (|X_t - a|)^2 = (|X_0 - a|)^2 + 2 \int_0^t |X_s - a| \text{sgn}(X_s - a) dX_s + 2 \int_0^t |X_s - a| dL_s^a + [|X - a|]_t \\ &= (X_0 - a)^2 + 2 \int_0^t (X_s - a) dX_s + 2 \int_0^t |X_s - a| dL_s^a + \int_0^t \underbrace{\text{sgn}(X_s - a)^2}_{=1} d[M]_s \end{aligned}$$

And by comparing with the standard Ito formula

$$(X_t - a)^2 = (X_0 - a)^2 + 2 \int_0^t (X_s - a) dX_s + \underbrace{[X]_t}_{=[M]_t}$$

we conclude that

$$\int_0^t |X_s - a| dL_s^a = 0, \quad t \geq 0$$

which proves that the measure $(dL_s^a)_{s\geq 0}$ is supported in the (random) set $\{s \in \mathbb{R} : X_s = a\}$ of times. The process L^a increases only when the process X visits a (in general this will be a “fractal-like” and with zero Lebesgue measure).

For Brownian motion is it true (we will not prove it) that the set $\{s \in \mathbb{R} : X_s = a\}$ is the support of the measure $(L_t^a)_{t\geq 0}$.

Theorem 4. (Ito–Tanaka formula) *If f is the difference of two convex functions and X a continuous semimartingale, then*

$$f(X_t) = f(X_0) + \int_0^t f'(X_s) dX_s + \frac{1}{2} \int_{\mathbb{R}} L_t^a f''(da)$$

and in particular $(f(X_t))_{t\geq 0}$ is a semimartingale.

In this formula $f''(da)$ denotes the measure associated to the second derivative of a convex function (and therefore of a difference of two convex functions, by linearity).

Recall that for any convex function f we have the formula

$$f(x) = \alpha + \beta x + \frac{1}{2} \int |x-a| f''(da), \quad x \in \mathbb{R}$$

and

$$f'_-(x) = \beta + \frac{1}{2} \int \operatorname{sgn}(x-a) f''(da), \quad x \in \mathbb{R}$$

for some $\alpha, \beta \in \mathbb{R}$ and $f''(da)$ is the measure associated to the increasing function $(f'_-(x))_{x \in \mathbb{R}}$ (convex functions are those functions whose second distributional derivative is a positive Radon measure).

The idea is that

$$\frac{d}{dx} |x-a| = 2\delta(x-a)$$

which justify intuitively the $1/2$ in the formula.

Proof. We can write

$$f(X_t) = \alpha + \beta X_t + \frac{1}{2} \int |X_t - a| f''(da)$$

by Tanaka's formula

$$\begin{aligned} f(X_t) &= \alpha + \beta X_0 + \beta \int_0^t dX_s + \frac{1}{2} \int |X_0 - a| f''(da) + \frac{1}{2} \int \left(\int_0^t \operatorname{sgn}(X_s - a) dX_s \right) f''(da) \\ &\quad + \frac{1}{2} \int L_t^a f''(da) \end{aligned}$$

Note that this computation makes sense since the stochastic integral $\int_0^t \operatorname{sgn}(X_s - a) dX_s$ is a measurable function of a , more precisely (see the relevant exercise in Sheet 7) the function

$$(a, t, \omega) \mapsto \left(\int_0^t \operatorname{sgn}(X_s - a) dX_s \right) (\omega)$$

is a measurable function on $\mathcal{B}(\mathbb{R}) \otimes \mathcal{P}$ (\mathcal{P} is the previsible σ field on $\mathbb{R}_+ \times \Omega$) and also a (stochastic) Fubini theorem applies so that

$$\int \left(\int_0^t \operatorname{sgn}(X_s - a) dX_s \right) f''(da) = \int_0^t \left(\int \operatorname{sgn}(X_s - a) f''(da) \right) dX_s$$

Moreover we note that

$$\beta \int_0^t dX_s + \frac{1}{2} \int_0^t \left(\int \operatorname{sgn}(X_s - a) f''(da) \right) dX_s = \int_0^t f'_-(X_s) dX_s$$

which completes the proof. □

Corollary 5. (*Occupation-time formula*) *There is a \mathbb{P} -negligible set \mathcal{N} outside which for any $t \geq 0$ and any positive Borel function $g: \mathbb{R} \rightarrow \mathbb{R}_+$ we have*

$$\int_0^t g(X_s) d[X]_s = \int_{\mathbb{R}} g(a) L_t^a da.$$

Remark 6. The measure $d[X]_s$ can be understood as some “intrinsic” time of the semimartingale. In particular, for Brownian motion X we have $d[X]_s = ds$ and if we take $g(x) = \mathbb{1}_{x \in A}$ for some set $A \in \mathcal{B}(\mathbb{R})$ we have

$$\operatorname{Leb}(\{s \in [0, t]: X_s \in A\}) = \int_0^t \mathbb{1}_{X_s \in A} ds = \int_A L_t^a da.$$

In this sense $L_t^a da$ represents the time spent by X in the infinitesimal neighborhood $a \pm da$.

Proof. For any $g: \mathbb{R} \rightarrow \mathbb{R}_+$ we can find a convex function f such that $f'' = g$, i.e. we can take $f''(da) = g(a)da$ in the formula above. By Tanaka's formula we then have

$$f(X_t) = f(X_0) + \int_0^t f'_-(X_s) dX_s + \frac{1}{2} \int_{\mathbb{R}} g(a) L_t^a da. \quad \mathbb{P} - a.s.$$

Take a countable family $(g_n)_{n \geq 0}$ of compactly supported continuous functions which is dense in $C_0(\mathbb{R})$ and consider now f_n so that $f_n'' = g_n$, note that $f_n \in C^2$ and $f_{n,-}' = f_{n,+}' = f_n'$. I have now both Ito-Tanaka's formula and Ito formula (note that f_n is the difference of two convex functions)

$$f_n(X_t) = f_n(X_0) + \int_0^t f_n'(X_s) dX_s + \frac{1}{2} \int_0^t g_n(X_s) d[X]_s. \quad \mathbb{P} - a.s.$$

So by comparing these two formulas we have

$$\ell_t(g_n) := \int_{\mathbb{R}} g_n(a) L_t^a da = \int_0^t g_n(X_s) d[X]_s. \quad \mathbb{P} - a.s.$$

This equality holds a.s. for any g_n and one can choose a \mathbb{P} -negligible set \mathcal{N} such that the equalities holds simultaneously for all n and all $t \geq 0$ (since the quantity $\ell_t(g_n)$ is continuous in time and therefore can be determined by looking to a dense set of times $(t_k)_k$).

One note now that for any $t \geq 0$, the functional ℓ_t is a positive linear functional on $C_0(\mathbb{R})$ which is continuous in the uniform norm on $C_0(\mathbb{R})$ so can be extended by continuity to all functions in $C_0(\mathbb{R})$ and by a monotone class argument to all Borel positive functions. \square

Thursday no lecture.

Next lecture on tuesday: we prove that $a \mapsto L_t^a$ is cadlag and that there is a formula of the form

$$L_t^a = \lim_{\varepsilon \downarrow 0} \frac{1}{\varepsilon} \int_0^t \mathbb{1}_{X_s \in (a, a+\varepsilon)} d[X]_s$$

and for X a martingale

$$L_t^a = \lim_{\varepsilon \downarrow 0} \frac{1}{2\varepsilon} \int_0^t \mathbb{1}_{X_s \in (a-\varepsilon, a+\varepsilon)} d[X]_s.$$

We continue to discuss some properties of local time of Brownian motion and reflected Brownian motion.

For Exercise Sheet 7, exercise 2, the invariant measure should be e^{-2V} and not e^{-V} .
