

Stochastic calculus with Hermite polynomials and stochastic differential equations

1 Introduction

The proof of the existence and uniqueness theorem for stochastic differential equations can be found in chapter 5 of [1].

2 Stochastic calculus with Hermite polynomials

This section expands example D.3 of chapter 4 of [1].

Proposition 2.1. *the transition probability of the Wiener process admits the expansion*

$$\frac{e^{-\frac{(x-y)^2}{2t}}}{\sqrt{2\pi t}} = \sum_{n=0}^{\infty} \frac{e^{-\frac{x^2}{2t}}}{\sqrt{2\pi t}} \left(\frac{y}{t}\right)^n h_n(x, t)$$

where the h_n 's are Hermite polynomials

$$h_n(x, t) = \frac{(-t)^n}{\Gamma(n+1)} e^{\frac{x^2}{2t}} \frac{d^n}{dx^n} e^{-\frac{x^2}{2t}} \quad (2.1)$$

Proof. Let us postulate for the n -th order of the Taylor expansion the form

$$\frac{y^n}{\Gamma(n+1)} \frac{d^n}{dz^n} \Big|_{z=0} \frac{e^{-\frac{(x-z)^2}{2t}}}{\sqrt{2\pi t}} := \frac{e^{-\frac{x^2}{2t}}}{\sqrt{2\pi t}} \left(\frac{y}{t}\right)^n h_n(x, t)$$

we can then calculate the explicit form of the polynomial h_n . Namely

$$\begin{aligned} \frac{e^{-\frac{x^2}{2t}}}{\sqrt{2\pi t}} \left(\frac{y}{t}\right)^n h_n(x, t) &= \\ \frac{y^n}{\Gamma(n+1)} \frac{d^n}{dz^n} \Big|_{z=0} \sum_{k=0}^{\infty} \int_{\mathbb{R}} \frac{d^d p}{(2\pi)^d} e^{i p x - \frac{t p^2}{2}} \frac{(-i p z)^k}{\Gamma(k+1)} &= \frac{y^n}{\Gamma(n+1)} \int_{\mathbb{R}} \frac{d^d p}{(2\pi)^d} e^{i p x - \frac{t p^2}{2}} (-i p)^n \end{aligned}$$

Observing that powers of p are generated by taking derivatives with respect to x we get into

$$\frac{e^{-\frac{x^2}{2t}}}{\sqrt{2\pi t}} \left(\frac{y}{t}\right)^n h_n(x, t) = \frac{(-y)^n}{\Gamma(n+1)} \frac{d^n}{dx^n} \int_{\mathbb{R}} e^{i p x - \frac{t p^2}{2}}$$

Performing the integral and contrasting the left to the right hand side yields the claim. □

The polynomials h_n defined by (2.1) are called the **Hermite polynomials**. It is readily checked that they enjoy the scaling property

$$h_n(\lambda x, \lambda^2 t) = \lambda^n h_n(x, t) \quad \Rightarrow \quad (x \partial_x + 2t \partial_t) h_n(x, t) = n h_n(x, t)$$

Furthermore

Proposition 2.2. *The expected value of an Hermite polynomial having for argument a Wiener process starting at x is*

$$\mathbb{E} h_n(w_t + x, t) = \frac{x^n}{\Gamma(n+1)} = h_n(x, 0) \quad (2.2)$$

Proof.

$$\mathbb{E} h_n(w_t + x, t) := \int_{\mathbb{R}} dy h_n(y, t) \frac{e^{-\frac{(y-x)^2}{2t}}}{\sqrt{2\pi t}} = \frac{(-t)^n}{\Gamma(n+1)} \int_{\mathbb{R}} dy \frac{e^{-\frac{x^2}{2t} + \frac{xy}{t}}}{\sqrt{2\pi t}} \frac{d^n}{dy^n} e^{-\frac{y^2}{2t}}$$

integrating by parts yields the claim. \square

The reason why the expectation value is preserved is that the differential of Hermite along realizations of the Wiener process takes the form.

Proposition 2.3. *The differential of Hermite polynomials of the Wiener process is*

$$dh_n(w_t, t) = dw_t \partial_{w_t} h_n(w_t, t)$$

Proof. By Ito lemma we have

$$dh_n(w_t, t) = dt \left(\partial_t + \frac{1}{2} \partial_{w_t}^2 \right) h_n(w_t, t) + dw_t \partial_{w_t} h_n(w_t, t)$$

In order to prove the claim we need to show that

$$\left(\partial_t + \frac{1}{2} \partial_{w_t}^2 \right) h_n(w_t, t) = 0$$

Such result can be achieved by direct calculation. It is instructive to proceed in a slightly indirect way. For any $t > 0$

$$\begin{aligned} 0 &= \left(\partial_t - \frac{1}{2} \partial_x^2 \right) \frac{e^{-\frac{(x-y)^2}{2t}}}{\sqrt{2\pi t}} \\ &= \sum_{n=0}^{\infty} \frac{e^{-\frac{x^2}{2t}}}{\sqrt{2\pi t}} \left(\frac{y}{t} \right)^n \frac{1}{t} \left(-n + t \partial_t - \frac{t}{2} \partial_x^2 + x \partial_x \right) h_n = - \sum_{n=0}^{\infty} \frac{e^{-\frac{x^2}{2t}}}{\sqrt{2\pi t}} \left(\frac{y}{t} \right)^n \left(\partial_t + \frac{1}{2} \partial_x^2 \right) h_n \end{aligned}$$

which implies

$$\left(\partial_t + \frac{1}{2} \partial_x^2 \right) h_n = 0$$

as each of these multiply positive definite terms of different order in y . \square

We have therefore a probabilistic interpretation of the statistical conservation law

$$h_n(w_t + x, t) = h_n(x, 0) + \int_0^t dw_t \partial_{w_t} h_n(w_t, t)$$

From the property of the Ito integral

$$\mathbb{E} h_n(w_t + x, t) = h_n(x, 0) = \frac{x^n}{\Gamma(n+1)}$$

2.1 Recursion relation and multiple integrals over the Wiener process

Proposition 2.4. *Stochastic integrals over Hermite polynomials satisfy the simple recursion relation*

$$\int_0^t dw_s h_n(w_s, s) = h_{n+1}(w_t, t)$$

Proof. Consider the exponential process:

$$\xi_t = e^{\lambda w_t - \frac{\lambda^2}{2} t} \quad (2.3)$$

Applying to it Ito lemma yields

$$d\xi_t = \lambda \xi_t dt + \lambda dw_t \quad (2.4)$$

or equivalently

$$e^{\lambda w_t - \frac{\lambda^2}{2} t} = 1 + \lambda \int_0^t dw_s e^{\lambda w_s - \frac{\lambda^2}{2} s}$$

If we differentiate an arbitrary number of times with respect to λ we then obtain

$$\left. \frac{d^n}{d\lambda^n} \right|_{\lambda=0} e^{\lambda w_t - \frac{\lambda^2}{2} t} = \left. \frac{d^n}{d\lambda^n} \right|_{\lambda=0} \lambda \int_0^t dw_s e^{\lambda w_s - \frac{\lambda^2}{2} s}$$

Contrasting the left-hand side with the definition of Hermite polynomials we conclude

$$\left. \frac{d^n}{d\lambda^n} \right|_{\lambda=0} e^{\lambda w_t - \frac{\lambda^2}{2} t} = t^n \left. \frac{d^n}{dz^n} \right|_{z=0} e^{\frac{z}{t} w_t - \frac{z^2}{2t}} = \Gamma(n+1) h_n(w_t, t)$$

The right hand side is

$$\left. \frac{d^n}{d\lambda^n} \right|_{\lambda=0} \lambda \int_0^t dw_s e^{\lambda w_s - \frac{\lambda^2}{2} s} = n \left. \frac{d^{n-1}}{d\lambda^{n-1}} \right|_{\lambda=0} \int_0^t dw_s e^{\lambda w_s - \frac{\lambda^2}{2} s} = \Gamma(n+1) \int_0^t dw_s h_{n-1}(w_s, s)$$

We have therefore proved that

$$h_n(w_t, t) = \int_0^t dw_s h_{n-1}(w_s, s)$$

□

An important consequence is the following. Since

$$h_0(w_t, t) = 1$$

we have that

$$\int_0^t dw_s = \int_0^t dw_s h_0(w_s, s) = h_1(w_t, t)$$

and

$$\int_0^t dw_{s_1} \int_0^{s_1} dw_{s_0} = h_2(w_t, t)$$

or in full generality

$$\int_0^t dw_{s_1} \prod_{i=1}^{n-1} \int_0^{s_{i-1}} dw_{s_{i-1}} = h_n(w_t, t)$$

3 Existence and uniqueness theorem

Theorem 3.1 (*Existence and uniqueness*). Suppose that for some $T \in \mathbb{R}_+$

$$\mathbf{b} : \mathbb{R}^d \times [0, T] \rightarrow \mathbb{R}^d$$

and

$$\mathbf{A} : \mathbb{R}^{d \times d} \times [0, T] \rightarrow \mathbb{R}^{d \times m}$$

are continuous and satisfy the following conditions in the Euclidean norm

$$\|\mathbf{b}(\mathbf{x}, t) - \mathbf{b}(\mathbf{y}, t)\| < C \|\mathbf{x} - \mathbf{y}\| \quad \& \quad \|\mathbf{A}(\mathbf{x}, t) - \mathbf{A}(\mathbf{y}, t)\| < C \|\mathbf{x} - \mathbf{y}\|$$

and

$$\|\mathbf{b}(\mathbf{x}, t)\| < C(1 + \|\mathbf{x}\|) \quad \& \quad \|\mathbf{A}(\mathbf{x}, t)\| < C(1 + \|\mathbf{x}\|)$$

for all $0 \leq t \leq T$ and some positive constant C . Let also ξ_o

$$\xi_o : \Omega \rightarrow \mathbb{R}^d$$

a random variable such that

$$\mathbb{E} \|\xi_o\|^2 < \infty$$

Furthermore ξ_o is independent of the σ -algebra \mathcal{W} generated by a given m -dimensional Wiener process for $t \geq 0$. Then, there exists a unique solution of

$$d\xi_t = \mathbf{b}(\xi_t, t) dt + \mathbf{A}(\xi_t, t) \cdot d\mathbf{w}_t \tag{3.1a}$$

$$\xi_0 = \xi_o \tag{3.1b}$$

Uniqueness here means that any square integrable ξ_t and $\tilde{\xi}_t$ with continuous paths, satisfying (3.1a), (3.1b) then for all $0 \leq t \leq T$

$$\xi_t = \tilde{\xi}_t \quad a.s.$$

3.1 Example: absence of Lipschitz continuity

Consider the ordinary differential equation:

$$\dot{\xi} = C\xi^{1/3}$$

The field

$$f = C x^{1/3}$$

is *not* differentiable in zero therefore not Lipschitz continuous there. As a consequence the equation has multiple solutions

$$\xi_t = \begin{cases} 0 & t < t_o \\ \tilde{C} t^{3/2} & t \geq t_o \end{cases}$$

for arbitrary t_o .

4 Solution by iteration

If \mathbf{b} and \mathbf{A} are smooth

$$\begin{aligned}
\xi_t &= \xi_o + \int_0^t ds \mathbf{b}(\xi_s, s) + \int_0^t \mathbf{A}(\xi_s, s) \cdot d\mathbf{w}_s \\
&= \xi_o + \mathbf{b}(\xi_o, 0)t + \mathbf{A}(\xi_o, 0) \cdot d\mathbf{w}_t + \int_0^t ds \int_0^s d\mathbf{b}(\xi_u, u) + \int_0^t \int_0^s d\mathbf{A}(\mathbf{x}_u, u) \cdot d\mathbf{w}_s \\
&= \xi_o + \mathbf{b}(\xi_o, 0)t + \mathbf{A}(\xi_o, 0) \cdot d\mathbf{w}_t + \int_0^t d\mathbf{b}(\xi_s, s)(t-s) + \int_0^t d\mathbf{A}(\mathbf{x}_s, s)[\mathbf{w}_t - \mathbf{w}_s]
\end{aligned} \tag{4.1}$$

We then apply Ito lemma to \mathbf{b} and \mathbf{A} and iterate. In such a way the solution is constructed as a power series in t and \mathbf{w}_t .

Example 4.1 (1d-linear case). Consider the **Ito** SDE

$$d\xi_t = \frac{\xi_t}{\tau} dt + \sigma \xi_t d\mathbf{w}_t \tag{4.2}$$

we can remove the drift by setting

$$\xi_t = \tilde{\xi}_t e^{\frac{t}{\tau}}$$

The new process $\tilde{\xi}_t$ is related to the original by a function independent of the Wiener process. Hence, Ito calculus lemma

$$d(\tilde{\xi}_t \eta_t) = (d\tilde{\xi}_t) \eta_t + \tilde{\xi}_t d\eta_t + \langle d\tilde{\xi}_t, d\eta_t \rangle$$

($\langle \bullet, \bullet \rangle$ is the quadratic co-variation) reduces for

$$\eta_t = e^{\frac{t}{\tau}}$$

to the standard Leibniz rule. We find

$$d(\tilde{\xi}_t e^{\frac{t}{\tau}}) = (d\tilde{\xi}_t) e^{\frac{t}{\tau}} + \tilde{\xi}_t \frac{e^{\frac{t}{\tau}}}{\tau}$$

The new Ito stochastic differential equation is

$$d\tilde{\xi}_t = \sigma \tilde{\xi}_t d\mathbf{w}_t$$

If we apply the recursion equations (4.1) we get into

$$\begin{aligned}
\tilde{\xi}_t &= \tilde{\xi}_o + \sigma \tilde{\xi}_o \mathbf{w}_t + \sigma \int_0^t d\mathbf{w}_s \int_0^s d\tilde{\xi}_{s_1} \\
&= \tilde{\xi}_o + \sigma \tilde{\xi}_o \mathbf{w}_t + \sigma^2 \tilde{\xi}_o \int_0^t d\mathbf{w}_s \int_0^s d\mathbf{w}_{s_1} + \sigma^2 \int_0^t d\mathbf{w}_s \int_0^s d\mathbf{w}_{s_1} \int_0^{s_2} d\tilde{\xi}_{s_2}
\end{aligned}$$

Repeating for an arbitrary number of steps

$$\tilde{\xi}_t = \tilde{\xi}_o + \tilde{\xi}_o \sum_{i=1}^{\infty} \sigma^i \int_0^t d\mathbf{w}_{s_1} \prod_{j=1}^{i-1} \int_0^{s_j} d\mathbf{w}_{s_j} \tag{4.3}$$

We already know that

$$\int_0^t dw_{s_1} \prod_{j=1}^{i-1} \int_0^{s_j} dw_{s_j} = h_i(w_t, t)$$

with h_i the Hermite polynomial

$$h_i(x, t) = \frac{t^n}{\Gamma(i+1)} \frac{d^n}{dz^n} \Big|_{z=0} e^{\frac{zx}{t} - \frac{z^2}{2t}} = \frac{1}{\Gamma(i+1)} \frac{d^n}{d\lambda^n} \Big|_{z=0} e^{\lambda x - \frac{\lambda^2 t}{2}}$$

Upon inserting in (4.3), we get into

$$\tilde{\xi}_t = \tilde{\xi}_o \left\{ 1 + \sum_{i=1}^{\infty} \frac{\sigma^i h_i(w_t, t)}{\Gamma(i+1)} \right\} = \xi_o e^{\sigma w_t - \frac{\sigma^2 t}{2}}$$

and consequently

$$\xi_t = \xi_o e^{\frac{t}{\tau} + \sigma w_t - \frac{\sigma^2 t}{2}}$$

The same result is straightforwardly obtained by converting (4.2) to *Stratonovich form*

$$d\xi_t = \left(1 - \frac{\sigma^2 \tau}{2} \right) \xi_t \frac{dt}{\tau} + \sigma \xi_t dw_t$$

and by integrating it according to the usual rules of calculus

$$\xi_t = \xi_o e^{\left(1 - \frac{\sigma^2 \tau}{2} \right) \frac{t}{\tau} + \sigma w_t}$$

References

- [1] L. C. Evans. An introduction to stochastic differential equations. UC Berkeley, Lecture Notes, 2006.