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SOME PROPERTIES OF HERMITE POLYNOMIALS OF A WIENER PROCESS

We investigate a stochastic process $h_k(W_t, t) = t^{k/2} H_k(W_t/\sqrt{t})$, where $\{W_t, t \geq 0\}$ is a Wiener process, H_k - Hermite polynomial of degree k . We give formulas for moments, conditional moments and limit distribution of h_k when $t \rightarrow \infty$.

1. Introduction and formulation of the results

Let $\{W_t, t \geq 0\}$ be a Wiener process, H_k - Hermite polynomial defined by formula

$$\frac{d^k}{dx^k} e^{-x^2/2} = (-1)^k e^{-x^2/2} H_k(x).$$

Let

$$h_k(W_t, t) = t^{k/2} H_k(W_t/\sqrt{t}).$$

It is evident that $h_1 = W_t$, $h_2 = W_t^2 - t$, $h_3 = W_t^3 - 3tW_t$ and so on. We shall use the typical denotations for σ -fields:

$$\mathfrak{F}_{\leq t} = \sigma(W_s, s \leq t), \quad \mathfrak{F}_{\geq t} = \sigma(W_s, s \geq t).$$

It is well known (see e.g. [3], [6]) that

$$(1) \quad E\{h_k(W_t, t)\} = 0, \quad E\{h_k(W_T - W_t, T - t)\} = 0,$$

$$(2) \quad E\{h_k^2(W_t, t)\} = k! t^k,$$

$$(3) \quad h_k(W_t, t) = I_k(1_{[0, t]}^{\otimes k}) = \int_{\{t_1 < \dots < t_k < t\}} dW(t_1) \dots dW(t_k),$$

$\{h_k(W_t, t), \mathfrak{F}_{\leq t}\}$ is a martingale and a Markov process.

The aim of this paper is to give formulas for moments, conditional moments and limit distribution of the process

$h_k(W_t, t)$. More precisely, we are going to prove that for every natural k , every $t < T$ the following properties hold

$$(4) \quad h_k(W_T, T) = \sum_{r=0}^k \binom{k}{r} h_r(W_t, t) h_{k-r}(W_T - W_t, T - t),$$

$$(5) \quad E\left(h_k(W_t, t) \mid \mathfrak{F}_{\geq T}\right) = \left(\frac{t}{T}\right)^k h_k(W_T, T)$$

(Linear regression),

$$(6) \quad E\left(h_k^2(W_T, T) \mid \mathfrak{F}_{\leq t}\right) = \sum_{r=0}^k \binom{k}{r}^2 (k-r)! (T-t)^{k-r} h_r^2(W_t, t),$$

$$(7) \quad E(h_k^4(W_t, t)) = (k!)^2 t^{2k} \sum_{r=0}^k \binom{k}{r}^2 \binom{2r}{r},$$

$$(8) \quad E(h_k(W_T, T) - h_k(W_t, t))^4 = (T-t)^2 \varphi_{2(k-1)}(t, T),$$

where $\varphi_{2(k-1)}$ is a polynomial of degree $2(k-1)$.

Next we shall prove the following limit property:
for every $w \neq 0$

$$(9) \quad \lim_{\substack{t \rightarrow \infty \\ t/T \rightarrow 1}} E\left(\frac{h_k(W_T, T) - E\{h_k(W_T, T) \mid \mathfrak{F}_{\leq t}\}}{\sigma_k(t, T, w)} \mid \frac{W_t}{\sqrt{t}} = w\right) = N(0, 1),$$

where $\sigma_k^2(t, T, w)$ is the conditional variance given by formula

$$\sigma_k^2(t, T, w) = \text{Var}\{h_k(W_T, T) \mid W_t = w\sqrt{t}\}, \quad t < T.$$

It follows from the martingale property and (9) that for $0 < T - t < t$, for given $w = W_t/\sqrt{t}$, we can write

$$E\{h_k(W_T, T)\} \approx N\{h_k(w\sqrt{t}, t), \sigma_k(t, T, w)\},$$

$$(10) \quad h_k(W_T, T) \approx h_k(w\sqrt{t}, t) + Y\sigma_k(t, T, w),$$

where Y has a normal distribution $N(0, 1)$.

Formula (10) can be interpreted as a forecasting at the moment T when the state of the process at the moment t is known. This is of course a linear forecasting. Formula (10) can be treated as a basis for linear forecasting. Of course a

basis of another kind than in ARIMA models (see [1]). Unconditional limit distributions (similar to (10)) were considered in the earlier papers of the authors [8], [9].

The addition formula (4) is the main tool in the proofs of (5)-(9). (Relations (1)-(3) also follow immediately from (4)). This formula looks like binomial formula. On the other hand (4) can be treated as a version of Cameron-Martin formula given in Kallianpur book [4].

Different connections and applications of Hermite polynomials in random theory were considered by Ito [3], Wiener [11], McKean [6], Dym and McKean [2]. Especially homogenous chaos is described by Kallianpur [4].

Hermite polynomials of a Wiener process are a nice example of different known class of stochastic processes (martingale, Markov property, the linearity of regression).

2. Auxiliary results

By the definition of Hermite polynomial and formulas given in [11] we get the following relations

$$(11) \quad H_k(x) = \sum_{r=0}^{[k/2]} \frac{(-1)^r k!}{r!(k-2r)! 2^r} x^{k-2r},$$

$$(11') \quad h_k(x, t) = \sum_{r=0}^{[k/2]} \frac{(-1)^r k!}{r!(k-2r)! 2^r} t^r x^{k-2r},$$

$$(12) \quad (c^2+1)^{k/2} H_k \left\{ \frac{cx+y}{\sqrt{c^2+1}} \right\} = \sum_{r=0}^k \binom{k}{r} c^r H_r(x) H_{k-r}(y),$$

$$(13) \quad \frac{1}{\sqrt{2\pi}} \int H_k(x) H_r(x) e^{-x^2/2} dx = \begin{cases} k! & \text{for } k=r, \\ 0 & \text{for } k \neq r, \end{cases}$$

$$(14) \quad \frac{1}{\sqrt{2\pi}} \int e^{-x^2/2} H_k^4(x) dx = (k!)^2 \sum_{r=0}^k \binom{k}{r}^2 \binom{2r}{r},$$

$$(15) \quad \frac{1}{\sqrt{2\pi}} \int \exp\left\{-\frac{(x-y)^2}{2}\right\} H_k(ax) dx = (1-a^2)^{k/2} H_k \left(\frac{ay}{\sqrt{1-a^2}} \right),$$

$$(16) \quad H_{k+1}(x) = xH_k(x) - kH_{k-1}(x),$$

$$(16') \quad h_{k+1}(x,t) = xh_k(x,t) - kth_{k-1}(x,t).$$

The proof of (9) will be based on a limit theorem for sum of dependent random variables.

Let $\{X_{nr}\}$ be a sequence of random variables, $Y_{nr} = g(X_{nr})$, $E(Y_{nr}) = 0$, where $g \in C^0$. We denote $\varepsilon_0(\cdot) = E\{(\cdot) | X_{n0} = x_0\}$, $s_n^2 = \sum_{r=1}^n \varepsilon_0(Y_{nr}^2)$.

Lemma 1. (see [8]). If

$$(17) \quad \varepsilon_0(Y_{nr} | X_{n0}, \dots, X_{n,r-1}) = 0,$$

$$(18) \quad \lim_{n \rightarrow \infty} \frac{1}{s_n^2} \sum_{r=1}^n \varepsilon_0 |\varepsilon_0(Y_{nr}^2 | X_{n0}, \dots, X_{n,r-1}) - \varepsilon_0(Y_{nr}^2)| = 0,$$

there exists $\delta > 0$ such that

$$(19) \quad \lim_{n \rightarrow \infty} \frac{1}{s_n^{2+\delta}} \sum_{r=1}^n \varepsilon_0(|Y_{nr}^{2+\delta}|) = 0$$

then for every x_0

$$(20) \quad \lim_{n \rightarrow \infty} \mathcal{L} \left(\frac{1}{s_n} \sum_{r=1}^n Y_{nr} | X_{n0} = x_0 \right) = N(0,1).$$

Lemma 1 is a modified version of comparison theorem (see [5]).

3. Proofs of formulas (4) - (9)

We can get formula (4) by elementary methods. Namely we substitute in (12):

$$x = W_t / \sqrt{t}, \quad y = \frac{W_T - W_t}{\sqrt{T-t}}, \quad c = \sqrt{t/(T-t)}.$$

Then after some easy transformations we get (4). Formula (4) can be also deduced by non elementary methods based on representation (3). In virtue of (15) we have

$$\begin{aligned} E\left\{h_k(W_t, t) \mid \tilde{y}_{\geq T}\right\} &= \sqrt{\frac{T}{2\pi t(T-t)}} \int \exp\left\{-\frac{\left(x - W_T \frac{t}{T}\right)^2}{2(T-t)t/T}\right\} h_k(x, t) dx = \\ &= \left(\frac{t}{T}\right)^k h_k(W_T, t). \end{aligned}$$

Therefore (5) is proved.

Formula (6) follows from (2), (4) and (13). Formula (7) from (14).

It is evident that (4) can be written in the following form

$$(21) \quad h_k(W_T, t) - h_k(W_t, t) = \sum_{r=0}^{k-1} \binom{k}{r} h_r(W_T, T) h_{k-r}(W_T - W_t, T-t).$$

In virtue of (21) and the formulas for moments of the Wiener process we obtain (8).

Now we are going to show (9) using Lemma 1.

We introduce the following partition of the interval (t, T) : $t_r = t + \frac{r}{n}(T-t)$, $r=0, \dots, n$. We put $X_{nr} = W_{t_r}$,

$$Y_{knr} = h_k(W_{t_r}, t_r) - h_k(W_{t_{r-1}}, t_{r-1}),$$

$$\varepsilon_0(\cdot) = \varepsilon_w(\cdot) = E\left\{(\cdot) \mid \frac{W_t}{\sqrt{t}} = w\right\}.$$

It is evident that

$$h_k(W_T, t) - h_k(W_t, t) = \sum_{r=1}^n \left[h_k(W_{t_r}, t_r) - h_k(W_{t_{r-1}}, t_{r-1}) \right] = \sum_{r=1}^n Y_{knr}.$$

We will prove that random variables X_{nr} , Y_{knr} satisfy the assumptions of Lemma 1.

Condition (17) follows from (1). In virtue of (2) and (6) we have

$$\begin{aligned} (22) \quad & \varepsilon_0\{|\varepsilon_0(Y_{knr}^2 | X_{n0}, \dots, X_{n, r-1}) - \varepsilon_0(Y_{knr}^2)\}| = \\ &= \sum_{s=0}^{k-1} \binom{k}{s}^2 (k-s)! (t_r - t_{r-1})^{k-s} \varepsilon_w | h_s^2(W_{t_{r-1}}, t_{r-1}) - \end{aligned}$$

$$\begin{aligned}
 & - \sum_{l=1}^s \binom{s}{l}^2 (s-1)! (t_{r-1} - t)^{s-1} h_1(w_t, t) \leq \\
 & \leq \sum_{s=1}^{k-1} \binom{k}{s}^2 (k-s)! \left(\frac{T-t}{n}\right) \sqrt{(t_{r-1} - t) p_{2s-1}(t, t_{r-1})}
 \end{aligned}$$

where p_{2s-1} is a polynomial of degree $2s-1$.

Taking into account (1) we get

$$\begin{aligned}
 (23) \quad s_n^2 &= \sum_{r=1}^n \varepsilon_w(Y_{knr}^2) = \sum_{r=1}^n \left[\varepsilon_w \left\{ h_k^2(w_{t_r}, t_r) \right\} - \right. \\
 & \quad \left. - \varepsilon_w \left\{ h_k^2(w_{t_{r-1}}, t_{r-1}) \right\} \right] = \\
 &= \varepsilon_w \{ h_k^2(w_T, T) \} - \varepsilon_w \{ h_k^2(w_t, t) \} = \varepsilon_w \{ h_k(w_T, T) - h_k(w_t, t) \}^2.
 \end{aligned}$$

By (22) and (23) we have

$$\begin{aligned}
 & \frac{1}{s_n^2} \sum_{r=1}^n \varepsilon_w | \varepsilon_w(Y_{nkr}^2 | X_{n0}, \dots, X_{n,r-1}) - \varepsilon_w(Y_{nkr}^2) | \leq \\
 & \leq \frac{\sum_{r=1}^n \sum_{s=0}^{k-1} \binom{k}{s}^2 (k-s)! \left(\frac{T-t}{n}\right)^{k-s} \sqrt{(t_{r-1} - t) p_{2s-1}(t, t_{r-1})}}{\sum_{s=0}^{k-1} \binom{k}{s}^2 (k-s)! (T-t)^{k-s} t^s H_s(w)} \leq \\
 & \leq \sqrt{\frac{T-t}{t}} \left[\sum_{s=0}^{k-1} c_{ks} \left(\frac{T-t}{n}\right)^{k-s-1} T^{s-1/2} \right] t^{\frac{3}{2}-k} \longrightarrow 0
 \end{aligned}$$

as $\lim_{t \rightarrow \infty} \frac{T-t}{t} = 0$, c_{ks} - depend only on their subscripts. Thus

(18) holds.

Condition (19) for $\delta = 2$ follows from (8).

Therefore all the assumptions of Lemma 1 are satisfied and (20) is true. Thus by Lemma 1 formula (9) holds.

4. Remark

Properties similar to (1) - (10) are true for Hermite polynomials of optional gaussian Markov processes.

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