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## THE DISTRIBUTION OF INTERVALS BETWEEN ZEROS FOR A STATIONARY GAUSSIAN PROCESS\*

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**1. Introduction.** This note supplements an earlier paper [1], where the distribution for the interval between two successive zeros was found for a real zero-mean Gaussian process with the following covariance function:

$$(1) \quad \rho(\tau) = Ex(t)x(t + \tau) = \frac{3}{2}e^{-|\tau|/\sqrt{3}}\left(1 - \frac{1}{3}e^{-(2/\sqrt{3})|\tau|}\right).$$

This earlier work was based on a time change for the process after which a formula of McKean [2] was used to derive the desired result. A second look at McKean's paper has revealed that the distribution of the interval between any two zeros (not only successive ones) can be found in a similar way. However, except for the case of two successive zeros, I have not been able to carry out a final integration to reduce the distribution to a closed-form expression.

**2. Computation of the inter-zero interval distribution.** Let  $x(t)$  be a zero-mean Gaussian process with a covariance function given by (1). All processes in this paper are assumed to be separable and real-valued. It is easy to verify by direct computation that  $x(t)$  can be represented in terms of a standard Brownian motion  $W(t)$  as

$$(2) \quad x(t) = \sqrt{3}e^{-\sqrt{3}t} \int_0^{\exp(2/\sqrt{3})t} W(s) ds,$$

a standard Brownian motion process being defined as a zero-mean Gaussian process with  $EW(t)W(s) = \min(t, s)$ . It follows that for  $t \geq t_0$  we can write

$$(3) \quad \begin{aligned} x(t) = & x(t_0)e^{-\sqrt{3}(t-t_0)}\left[1 + \frac{3}{2}g(t-t_0)\right] \\ & + \sqrt{3}e^{-\sqrt{3}(t-t_0)} \cdot \left[ \frac{1}{2}\dot{x}(t_0)g(t-t_0) + \int_0^{g(t-t_0)} W(s) ds \right], \end{aligned}$$

where  $W(s)$  is again a standard Brownian motion (but not the same one as in (2)) and is independent of both  $x(t_0)$  and  $\dot{x}(t_0)$ . We have also set

$$(4) \quad g(t) = \exp(2/\sqrt{3})t - 1.$$

Let  $\tau_n(t_0)$ ,  $n = 0, 1, \dots$ , be defined as follows:

$$(5) \quad \begin{aligned} \tau_0(t_0) &= t_0, \\ \tau_{n+1}(t_0) &= \min \{t : t > \tau_n(t_0), x(t) = 0\}. \end{aligned}$$

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Because of the stationarity of  $x(t)$ , the probability

$$(6) \quad \begin{aligned} & \text{Prob}(\tau_{m+1}(t_0) < t_0 + t | x(t_0) = 0, \text{h.w.}) \\ &= \lim_{\alpha \downarrow 0} \text{Prob} \left( \tau_{m+1}(t_0) < t_0 + t | 0 \leq \frac{x(t_0)}{\dot{x}(t_0)} \leq \alpha \right) \end{aligned}$$

depends only on  $t$  and not on  $t_0$  and will be denoted by  $F_m(t)$ . In (6) the notation “ $x(t_0) = 0, \text{h.w.}$ ” stands for “ $x(t_0) = 0$  in the horizontal-window sense” [3] and is defined by the limit on the right-hand side. It turns out that for the process being considered here the limit in (6) is the same as

$$\lim_{\alpha \downarrow 0} \text{Prob}(\tau_{m+1}(t_0) < t_0 + t | x(s) \text{ has at least one zero for } s \text{ in } [t_0 - \alpha, t_0]).$$

However, we do not make use of this fact in this paper.

Now, let  $\text{Prob}(\tau_{m+1}(0) < t | x(0) = x_0, \dot{x}(0) = y_0)$  be the conditional probability defined in the ordinary sense. Because of the symmetry of the process  $x(t)$  about zero, we have

$$\begin{aligned} & \text{Prob}(\tau_{m+1}(0) < t | x(0) = -x_0, \dot{x}(0) = -y_0) \\ &= \text{Prob}(\tau_{m+1}(0) < t | x(0) = -\dot{x}(0) = -y_0). \end{aligned}$$

It is also easy to verify that  $x(0)$  and  $\dot{x}(0)$  have a joint density

$$p(x_0, y_0) = \frac{1}{2\pi} e^{-(x_0^2 + y_0^2)/2}.$$

Using these facts, we can evaluate  $F_m(t)$  as follows:

$$(7) \quad \begin{aligned} F_m(t) &= \lim_{\alpha \downarrow 0} \text{Prob} \left( \tau_{m+1}(0) < t | 0 \leq \frac{x(0)}{\dot{x}(0)} \leq \alpha \right) \\ &= \lim_{\alpha \downarrow 0} \left[ \frac{\int_0^\infty dy_0 \int_0^{\alpha y_0} dx_0 e^{-(x_0^2 + y_0^2)/2} \text{Prob}(\tau_{m+1}(0) < t | x(0) = x_0, \dot{x}(0) = y_0)}{\int_0^\infty dy_0 \int_0^{\alpha y_0} dx_0 e^{-(x_0^2 + y_0^2)/2}} \right] \\ &= \lim_{\alpha \downarrow 0} \int_0^\infty d\eta_0 \eta_0 e^{-\eta_0^2/2} \\ & \quad \cdot \left[ \frac{1}{\tan^{-1} \alpha} \int_0^{\tan^{-1} \alpha} d\theta \text{Prob}(\tau_{m+1}(0) < t | x(0) = \eta_0 \sin \theta, \dot{x}(0) = \eta_0 \cos \theta) \right] \\ &= \int_0^\infty d\eta_0 \eta_0 e^{-\eta_0^2/2} \text{Prob}(\tau_{m+1}(0) < t | x(0) = 0, \dot{x}(0) = \eta_0), \end{aligned}$$

where we have made use of dominated convergence and the continuity of  $\text{Prob}(\tau_{m+1}(0) < t | x(0) = x_0, \dot{x}(0) = y_0)$  in  $x_0$  and  $y_0$ . We note that the conditional probability in the last expression of (7) is in the ordinary sense and not in the horizontal-window sense. The principal concern of this paper is the computation of the density function  $P_m(t) = \dot{F}_m(t)$ .

Let  $\{W(t), t \geq 0\}$  be a standard Brownian motion, and let  $\sigma_n(\eta_0), n = 0, 1, \dots$ , be defined as follows:

$$(8) \quad \begin{aligned} \sigma_0(\eta_0) &= 0, \quad \eta_0 \geq 0, \\ \sigma_{n+1}(\eta_0) &= \min \left\{ t : t > \sigma_n(\eta_0), t\eta_0 + \int_0^t W(s) ds = 0 \right\}, \quad \eta_0 \geq 0. \end{aligned}$$

From (3) and a comparison of (5) and (8), it is seen that

$$(9) \quad \begin{aligned} &\text{Prob}(\tau_{m+1}(0) < t | x(0) = 0, \dot{x}(0) = \eta_0) \\ &= \text{Prob} \left( \sigma_{m+1} \left( \frac{\eta_0}{2} \right) < g(t) \right), \quad \eta_0 \geq 0. \end{aligned}$$

Since  $\text{Prob}(\tau_{m+1}(0) < t | x(0) = 0, \dot{x}(0) = \eta_0)$  is symmetric in  $\eta_0$ , (7) becomes

$$(10) \quad F_m(t) = \int_0^\infty \eta_0 e^{-\eta_0^2/2} \text{Prob} \left( \sigma_{m+1} \left( \frac{\eta_0}{2} \right) < g(t) \right) d\eta_0.$$

Thus, the problem of evaluating  $F_m(t)$  and its derivative  $P_m(t)$  is reduced to the problem of finding the distribution of  $\sigma_{m+1}$ . Let  $\sigma_n(\eta_0)$  be as defined and let

$$(11) \quad h_n(\eta_0) = (-1)^n [\eta_0 + W(\sigma_n(\eta_0))], \quad \eta_0 \geq 0.$$

The quantity  $h_n(\eta_0)$  is the magnitude of the slope of  $\eta_0 t + \int_0^t W(s) ds$  at its  $n$ th zero.

The joint distribution of  $(\sigma_1(\eta_0), h_n(\eta_0))$  was found by McKean in [2]. Let  $f(t, a)$  denote the density function for the distribution of  $(\sigma_1(1), h_1(1))$ , i.e.,

$$(12) \quad \text{Prob} \{ \sigma_1(1) \in dt, h_1(1) \in da \} = f(t, a) dt da.$$

McKean derived the formula

$$(13) \quad \int_0^\infty f(t, a) e^{-at} dt = \int_0^\infty \frac{K_{i\gamma}(\sqrt{8\alpha}) K_{i\gamma}(\sqrt{8\alpha a})}{2 \cosh(\pi\gamma/3)} d0,$$

where  $K_\nu$  is the modified Bessel function and  $d0 = 2\pi^{-2}\gamma \sinh \pi\gamma$ , and, upon inverting the Laplace transform, obtained

$$(14) \quad f(t, a) = \frac{3a}{\pi\sqrt{2t^2}} e^{-(2/t)(1-a+a^2)} \int_0^{4a/t} \frac{e^{-3\theta/2}}{\sqrt{\pi\theta}} d\theta.$$

For a standard Brownian motion  $W(t)$ ,  $CW(t/C^2)$  is again a standard Brownian motion. From this scaling property we can show that  $(\sigma_n(\eta_0), h_n(\eta_0))$  have the same probability law as  $(\eta_0^2\sigma_n(1), \eta_0 h_n(1))$ . Therefore, if we denote the joint density function of  $(\sigma_n(\eta_0), h_n(\eta_0))$  by  $\pi_n(t, \eta | \eta_0)$ , we find

$$(15) \quad \begin{aligned} \pi_1(t, \eta | \eta_0) dt d\eta &= \text{Prob}(\sigma_1(\eta_0) \in dt, h_1(\eta_0) \in d\eta) \\ &= \frac{dt d\eta}{\eta_0^2 \eta_0} f \left( \frac{t}{\eta_0^2}, \frac{\eta}{\eta_0} \right), \end{aligned}$$

where  $f$  is given by (14).

For a more general  $n$ , the fact that  $\eta_0 t + \int_0^t W(s)$  and its derivative are jointly Markovian leads to the recursive relationship

$$(16) \quad \pi_{n+1}(t, \eta | \eta_0) = \int_0^t \int_0^\infty \pi_1(t-s, \eta | \zeta) \pi_n(s, \zeta | \eta_0) ds d\zeta.$$

Letting

$$(17) \quad \hat{\pi}_n(\alpha, \eta | \eta_0) = \int_0^\infty e^{-\alpha t} \pi_n(t, \eta | \eta_0) dt,$$

we can transform (16) into

$$(18) \quad \hat{\pi}_{n+1}(\alpha, \eta | \eta_0) = \int_0^\infty \hat{\pi}_1(\alpha, \eta | \zeta) \hat{\pi}_n(\alpha, \zeta | \eta_0) d\zeta.$$

The function  $\hat{\pi}_1(\alpha, \eta | \zeta)$  can be found from (15) and (13) as follows:

$$(19) \quad \begin{aligned} \hat{\pi}_1(\alpha, \eta | \eta_0) &= \int_0^\infty e^{-\alpha t} \pi_1(t, \eta | \eta_0) dt \\ &= \int_0^\infty \frac{1}{\eta_0^3} f\left(\frac{t}{\eta_0^2}, \frac{\eta}{\eta_0}\right) e^{-\alpha t} dt \\ &= \int_0^\infty \frac{1}{\eta_0} f\left(\tau, \frac{\eta}{\eta_0}\right) e^{-\alpha \eta_0^2 \tau} d\tau \\ &= \frac{1}{\eta_0} \int_0^\infty \frac{K_{i\gamma}(\sqrt{8\alpha\eta_0}) K_{i\gamma}(\sqrt{8\alpha\eta})}{2 \cosh(\pi\gamma/3)} d\theta. \end{aligned}$$

From the Lebedev transform pair [4, vol. 2, p. 173]

$$\begin{aligned} g(y) &= \int_0^\infty f(x) K_{ix}(y) 2\pi^{-2} x \sinh \pi x dx, \\ f(x) &= \int_0^\infty g(y) K_{ix}(y) y^{-1} dy, \end{aligned}$$

we conclude that

$$(20) \quad 2\pi^{-1} x \sinh \pi x \int_0^\infty K_{ix}(y) K_{ix}(y) y^{-1} dy = \delta(x - x').$$

Hence, by using (19) repeatedly in (18), we get

$$(21) \quad \hat{\pi}_n(\alpha, \eta | \eta_0) = \frac{1}{\eta_0} \int_0^\infty \frac{K_{i\gamma}(\sqrt{8\alpha\eta_0}) K_{i\gamma}(\sqrt{8\alpha\eta})}{[2 \cosh(\pi\gamma/3)]^n} d\theta,$$

where  $d0 = 2\pi^{-2}\gamma \sinh \pi\gamma$ , which is a surprisingly simple formula. From (10) we have

$$\begin{aligned}
 F_m(t) &= \int_0^\infty \eta_0 e^{-\eta_0^2/2} \text{Prob} \left( \sigma_{m+1} \left( \frac{\eta_0}{2} \right) < g(t) \right) d\eta_0 \\
 (22) \quad &= \int_0^\infty \eta_0 e^{-\eta_0^2/2} \text{Prob} \left( \frac{\eta_0^2}{4} \sigma_{m+1}(1) < g(t) \right) d\eta_0 \\
 &= \int_0^\infty d\eta_0 e^{-\eta_0^2/2} \int_0^{4g(t)/\eta_0^2} ds \int_0^\infty d\eta \pi_{m+1}(s, \eta|1).
 \end{aligned}$$

Thus, the density  $P_m(t) = \dot{F}_m(t)$  can be expressed as

$$(23) \quad P_m(t) = 4\dot{g}(t) \int_0^\infty \int_0^\infty \frac{1}{\eta_0} e^{-\eta_0^2/2} \pi_{m+1} \left( \frac{4g(t)}{\eta_0^2}, \eta|1 \right) d\eta_0 d\eta.$$

The Laplace transform (21) can now be inverted to yield

$$(24) \quad \pi_{m+1}(t, \eta|1) = \frac{1}{2t} e^{-2(1+\eta^2)t} \int_0^\infty \frac{K_{i\gamma}(4\eta/t)}{(2 \cosh \pi\gamma/3)^{m+1}} d0.$$

Using (24) in (23) yields

$$\begin{aligned}
 P_m(t) &= \frac{1}{2}\dot{g}(t) \int_0^\infty \int_0^\infty \int_0^\infty \frac{\eta_0}{g(t)} e^{-\eta_0^2/2} e^{-\eta_0^2(1+\eta^2)/(2g(t))} \\
 (25) \quad &\quad \cdot \frac{K_{i\gamma}(\eta\eta_0^2/g(t))}{(2 \cosh \pi\gamma/3)^{m+1}} d\eta_0 d\eta d0 \\
 &= \frac{1}{2}\dot{g}(t) \int_0^\infty \int_0^\infty \int_0^\infty \eta_0 e^{-\eta_0^2(1+\eta_0^2+g(t))/2} \\
 &\quad \cdot \frac{K_{i\gamma}(\eta\eta_0^2)}{(2 \cosh \pi\gamma/3)^{m+1}} d\eta_0 d\eta d0.
 \end{aligned}$$

Now, if we use the formula

$$(26) \quad K_{i\gamma}(a) = \int_0^\infty e^{-a \cosh u} \cos \gamma u du$$

in (25), we get a fourfold integral with variables of integration  $\eta_0, \eta, \gamma, u$ . Integrating first with respect to  $\eta_0$ , then  $u$ , we find

$$\begin{aligned}
 P_m(t) &= \frac{1}{2}\dot{g}(t) \int_0^\infty d0 \int_0^\infty d\eta \frac{1}{(2 \cosh \pi\gamma/3)^{m+1}} \\
 (27) \quad &\cdot \frac{\pi \sin(\gamma \cosh^{-1}(1 + \eta^2 + g(t))/(2\eta))}{\sinh \pi\gamma \sqrt{[1 + \eta^2 + g(t)]^2 - 4\eta^2}} \quad \left( d0 \equiv \frac{2}{\pi^2} \gamma \sinh \pi\gamma \right) \\
 &= \frac{1}{\pi} \dot{g}(t) \int_0^\infty d\gamma \int_{\cosh^{-1}\sqrt{1+g(t)}}^\infty dx \frac{\gamma \sin \gamma x}{\sqrt{\sinh^2 x - g(t)} (2 \cosh \pi\gamma/3)^{m+1}}.
 \end{aligned}$$

The expression (27) can be integrated once more with respect to  $\gamma$  to yield

$$(28) \quad P_m(t) = \frac{1}{\pi} \dot{g}(t) \int_{\cosh^{-1} \sqrt{1+g(t)}}^{\infty} \frac{1}{\sqrt{\sinh^2 x - g(t)}} (-1) \frac{d}{dx} f_m(x) dx,$$

where  $f_m(x)$  is given by

$$(29) \quad \begin{aligned} f_0(x) &= \frac{3}{4} \frac{1}{\cosh \frac{3}{2}x}, \\ f_1(x) &= \frac{9}{8\pi} \frac{x}{\sinh \frac{3}{2}x}, \\ f_m &= \frac{1}{m} \left[ \left( \frac{3x}{2\pi} \right)^2 + \left( \frac{m-1}{2} \right)^2 \right] f_{m-2}(x), \quad m \geq 2. \end{aligned}$$

I have not been able to carry out the integration in (28), except for  $m = 0$ . In [1],  $P_0(t)$  was obtained in terms of complete elliptic integrals [1, (25)].

**3. Computation of  $P_m(0)$ .** From (4), we have  $g(0) = 0$  and  $\dot{g}(0) = 2/\sqrt{3}$ . Therefore, for  $t = 0$ , (27) becomes

$$(30) \quad \begin{aligned} P_m(0) &= \frac{2}{\sqrt{3}\pi} \int_0^{\infty} d\gamma \int_0^{\infty} dx \frac{\gamma \sin \gamma x}{\sinh x (2 \cosh \pi\gamma/3)^{m+1}} \\ &= \frac{1}{\sqrt{3}} \int_0^{\infty} \left( \frac{e^{\pi\gamma} - 1}{e^{\pi\gamma} + 1} \right) \frac{\gamma d\gamma}{(2 \cosh \pi\gamma/3)^{m+1}} \\ &= \frac{1}{\sqrt{3}} \left( \frac{3}{\pi} \right)^2 \int_1^{\infty} \left( \frac{x^3 - 1}{x^3 + 1} \right) \frac{\ln x}{x(x + 1/x)^{m+1}} dx \\ &= \frac{1}{2\sqrt{3}} \left( \frac{3}{\pi} \right)^2 \int_0^{\infty} \left( \frac{x^3 - 1}{x^3 + 1} \right) \frac{x^m \ln x}{(x^2 + 1)^{m+1}} dx, \end{aligned}$$

which can be evaluated by contour integration to yield

$$(31) \quad P_m(0) = \frac{1}{4\sqrt{3}} \left( \frac{3}{\pi} \right)^2 (2\pi i) \sum \text{Res} \left\{ \frac{(z^3 - 1)z^m \ln z(1 - \ln z/(2\pi i))}{(z^3 + 1)(z^2 + 1)^{m+1}} \right\},$$

where the summation is taken over the residues at the five poles  $z = e^{\pi i/2}, e^{3\pi i/2}, e^{\pi i/3}, e^{5\pi i/3}$ . The expression (31) can be further elaborated to give

$$(32) \quad \begin{aligned} P_m(0) &= \frac{9}{2\sqrt{3}} \left\{ \frac{(-1)^m}{3 \cdot 2^{m+1}} - \frac{10}{27} + \frac{i}{\pi m!} \frac{d^m}{dz^m} \left[ \frac{z^m}{(z+i)^{m+1}} H(z) \right] \right\} \Bigg|_{z=e^{\pi i/2}} \\ &\quad + \frac{1}{\pi} \frac{i}{m!} \frac{d^m}{dz^m} \left[ \frac{z^m}{(z-i)^{m+1}} H(z) \right] \Bigg|_{z=e^{3\pi i/2}} \end{aligned}$$

with

$$(33) \quad H(z) = \frac{z^3 - 1}{z^3 + 1} \ln z \left[ 1 - \frac{\ln z}{2\pi i} \right].$$

I have not been able to reduce (32) further.

For a stationary zero-mean Gaussian process with covariance function of the form

$$(34) \quad \rho(\tau) = 1 - \frac{\tau^2}{2} + \alpha|\tau|^3 + o(|\tau|^3),$$

it is not hard to show that  $P_m(0)$  is proportional to  $\alpha$ . Longuet-Higgins [5] has obtained bounds for  $P_m(0)/\alpha$  for  $m$  up to 7. For  $m = 0, 1, 2$ , these bounds read

$$(35) \quad \begin{aligned} 1.1556 &< \frac{1}{\alpha} P_0(0) < 1.158, \\ 0.1971 &< \frac{1}{\alpha} P_1(0) < 0.198, \\ 0.0491 &< \frac{1}{\alpha} P_2(0) < 0.0556. \end{aligned}$$

Now, the covariance function (1) under consideration in this paper has the form of (34) with  $\alpha = \frac{2}{3}\sqrt{3}$ . Hence, the true values for  $P_m(0)/\alpha$  can be evaluated and compared against the bounds in (35). For  $m = 0, 1, 2$ , (32) can be evaluated to yield

$$(36) \quad \begin{aligned} P_0(0) &= \frac{2}{3\sqrt{3}} \left( \frac{37}{32} \right) = \frac{2}{3\sqrt{3}} (1.15625), \\ P_1(0) &= \frac{2}{3\sqrt{3}} \left( \frac{47}{64} - \frac{108}{64\pi} \right) \cong \frac{2}{3\sqrt{3}} (0.1972), \\ P_2(0) &= \frac{2}{3\sqrt{3}} \left( \frac{121}{128} - \frac{81}{32\pi} - \frac{27}{32\pi^2} \right) \cong \frac{2}{3\sqrt{3}} (0.0541), \end{aligned}$$

which are in agreement with (35).

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#### REFERENCES

- [1] E. WONG, *Some results concerning the zero-crossings of Gaussian noise*, this Journal, 14 (1966), pp. 1246–1254.
- [2] H. P. MCKEAN, JR., *A winding problem for a resonator driven by a white noise*, J. Math. Kyoto Univ., 2 (1963), pp. 227–235.
- [3] M. KAC AND D. SLEPIAN, *Large excursions of Gaussian processes*, Ann. Math. Statist., 30 (1959), pp. 1215–1228.
- [4] A. ERDÉLYI, W. MAGNUS, F. OBERHETTINGER AND F. G. TRICOMI, *Tables of Integral Transforms*, McGraw-Hill, New York, 1954.
- [5] M. S. LONGUET-HIGGINS, *Bounding approximations to the distribution of intervals between zeros of a stationary Gaussian process*, Time Series Analysis, M. Rosenblatt, ed., John Wiley, New York, 1963, pp. 104–115.