

AN OCCUPATION TIME THEOREM FOR THE ANGULAR COMPONENT OF PLANE BROWNIAN MOTION¹

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1. Summary. Let $Z(t) = (X(t), Y(t))$, $t \geq 0$, be the standard plane Brownian motion process. Let $(R(t), \theta_1(t))$ be the polar coordinates of $Z(t)$. Suppose that $R(0) = r > 0$ with probability 1, that is $Z(\cdot)$ starts away from the origin. We shall define the process $\theta(t)$, $t \geq 0$, as the total algebraic angle traveled on the continuous path $Z(s)$, $0 \leq s \leq t$; we take $\theta(0) = \theta_1(0)$; we have $\theta(\cdot) \equiv \theta_1(\cdot) \pmod{2\pi}$.

Let $f = f(\theta)$, $0 \leq \theta \leq 2\pi$, be a bounded measurable function such that $f(0) = f(2\pi)$. For $r_1 > r$ let τ be the first passage time of $R(\cdot)$ to the point r_1 . This note is devoted to the computation of the functional

$$(1.1) \quad L(f) = E_{r,\theta} \left\{ \int_0^\tau f(\theta_1(t)) dt \right\},$$

where $E_{r,\theta}\{\cdot\}$ is the expectation operator under the condition $R(0) = r$, $\theta_1(0) = \theta$. This is interpreted as the expected occupation time of a measurable subset of $[0, 2\pi]$ if f is the indicator function of the subset. We find an explicit formula for $L(f)$ as a linear functional on the Hilbert space $L_2[0, 2\pi]$.

A preliminary result of interest is presented in Section 2: the random variable $[\theta(\tau) - \theta(0)]/|\log(r_1/r)|$ has a Cauchy distribution for any positive numbers r, r_1 with $r \neq r_1$. This recalls the independent result of Spitzer that $[\theta(t) - \theta(0)]/\frac{1}{2} \log t$ has a limiting Cauchy distribution for $t \rightarrow \infty$ [4].

I thank the referee for his constructive remarks and for the alternate proof of Theorem 2.1 given in Section 6.

2. Distribution of $\theta(\tau)$. The process $\theta(t)$, $t \geq 0$, has the following representation. Let $U(t)$, $t \geq 0$, be a standard one-dimensional Brownian motion process independent of $Z(\cdot)$; then,

$$(2.1) \quad \theta(t) - \theta(0) = U\left(\int_0^t ds/R^2(s)\right), \quad t \geq 0,$$

according to Ito and McKean [2], p. 272.

We use this representation in the proof of

THEOREM 2.1. *The random variable*

$$[\theta(\tau) - \theta(0)]/|\log(r_1/r)|$$

has the Cauchy distribution for any positive $r \neq r_1$.

PROOF. By the representation (2.1) the characteristic function of $\theta(\tau) - \theta(0)$

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is

$$(2.2) \quad E\{\exp [iuU(\int_0^\tau ds/R^2(s))]\} = E\{\exp [(-u^2/2)\int_0^\tau ds/R^2(s)]\}.$$

Except for the change of variable $u \rightarrow u^2/2$, the latter expression in (2.2) is the Laplace-Stieltjis transform of the distribution function of

$$(2.3) \quad \int_0^\tau ds/R^2(s).$$

We shall find this distribution function and then its transform.

The two-dimensional process

$$(R(t), \int_0^t ds/R^2(s))$$

is a diffusion process with the following local characteristics: the local mean vector is $(1/2r, 1/r^2)$ and the local covariance matrix is the singular matrix

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}.$$

The distribution of the random variable (2.3) will be derived as a certain absorption probability for the diffusion. For fixed $t > 0$ let the variable s satisfy $0 \leq s \leq t$. Let us denote by $p(r, s; r_1, t)$ the probability, under the condition $R(0) = r$, that the sample path $R(\cdot)$ hits r_1 for the first time after the sample path $\int_0^\tau dv/R^2(v)$ hits $t - s$. This probability is the same as

$$(2.4) \quad P\{\int_0^\tau dv/R^2(v) > t - s\}$$

under the condition $R(0) = r$. According to classical diffusion theory [3], p. 48, this absorption probability is the solution of the partial differential equation

$$r^2 \partial^2 p / \partial r^2 + r \partial p / \partial r + 2 \partial p / \partial s = 0$$

$$p(r_1, s; r_1, t) = 0, \quad p(r, t; r_1, t) = 1.$$

We solve this by the method of separation of variables: the solution is

$$p(r, s; r_1, t) = (2/\pi) \int_0^\infty e^{-\frac{1}{2}(t-s)y^2} \sin [y|\log (r_1/r)|] dy/y.$$

From (2.4) we see that $1 - p(r, 0; r_1, t)$ is the distribution function of the random (2.3); hence, the density function is

$$(1/\pi) \int_0^\infty y e^{-\frac{1}{2}ty^2} \sin [y|\log (r_1/r)|] dy.$$

Integration by parts changes this integral to

$$(1/\pi t) |\log (r_1/r)| \int_0^\infty e^{-\frac{1}{2}ty^2} \cos [y|\log (r_1/r)|] dy,$$

which, in accordance with the formula

$$(2/\pi)^{-\frac{1}{2}} \int_0^\infty e^{-y^2/2} \cos uy dy = e^{-u^2/2},$$

has the value

$$(2\pi t)^{-\frac{1}{2}} |\log (r_1/r)| \exp [-|\log (r_1/r)|^2/2t].$$

The Laplace transform with respect to t is well known to be

$$\exp(-|\log(r_1/r)|(2u)^{\frac{1}{2}}), \quad u \geq 0;$$

hence, by the remark following equation (2.2), the characteristic function of $\theta(\tau) - \theta(0)$ is $\exp(-|u \log(r_1/r)|)$ which is the characteristic function of the Cauchy distribution with the scale factor $|\log(r_1/r)|$.

We remark that this characteristic function has the form

$$(2.5) \quad (r/r_1)^{|u|}$$

for $r < r_1$.

3. Application of Dynkin's formula. We shall compute $L(f)$ as

$$L(f) = \lim_{\lambda \rightarrow 0} E_{r,\theta}\{\int_0^\tau e^{-\lambda t} f(\theta(t)) dt\},$$

where f is a periodic bounded measurable function of θ of period 2π . First we do it for the special case $f(\theta) = e^{im\theta}$ ($m = \text{integer}$) and then extend it to the more general case.

Assume that $r < r_1$; then the expectation operator (1.1) is well defined for a bounded function f because $E\tau < \infty$. We apply Dynkin's fundamental identity for strong Markov processes [1]:

$$(3.1a) \quad E_{r,\theta}\{\int_0^\tau e^{-\lambda t + iu\theta(t)} dt\} = \phi(\lambda, u; r, \theta) - E_{r,\theta}\{e^{-\lambda\tau} \phi(\lambda, u; r_1, \theta(\tau))\},$$

$$-\infty < u < \infty, \lambda \geq 0,$$

where we have put $\phi(\lambda, u; r, \theta) = \int_0^\infty e^{-\lambda t} E_{r,\theta}(e^{iu\theta(t)}) dt$. The process $Z(\cdot)$ is well known to be stochastically invariant under rotations of the plane around the origin; hence, the conditional joint distribution of the random variables

$$\{\tau, \theta(\tau) - \theta(0), \theta(t) - \theta(0), t \geq 0\},$$

given $R(0)$ and $\theta(0)$, is independent of $\theta(0)$; therefore, the following relations hold:

$$\begin{aligned} \phi(\lambda, u; r, \theta) &= e^{iu\theta} \phi(\lambda, u; r, 0); \\ E_{r,\theta}\{\int_0^\tau e^{-\lambda t + iu\theta(t)} dt\} &= e^{iu\theta} E_{r,0}\{\int_0^\tau e^{-\lambda t + iu\theta(t)} dt\}; \\ E_{r,\theta}\{e^{-\lambda\tau} \phi(\lambda, u; r_1, \theta(\tau))\} &= e^{iu\theta} E_{r,0}\{e^{-\lambda\tau + iu\theta(\tau)}\} \phi(\lambda, u; r_1, 0). \end{aligned}$$

We put $E_r\{\cdot\} = E_{r,0}\{\cdot\}$ and $\phi(\lambda, u; r) = \phi(\lambda, u; r, 0)$; then equation (3.1a) takes the form

$$(3.1) \quad E_r\{\int_0^\tau e^{-\lambda t + iu\theta(t)} dt\} = \phi(\lambda, u; r) - \phi(\lambda, u; r_1) E_{r_1}\{e^{-\lambda\tau + iu\theta(\tau)}\}.$$

Ito and McKean [2], p. 271, have computed ϕ :

$$(3.2) \quad \begin{aligned} \phi(\lambda, u; r) &= K_{|u|}((2\lambda)^{\frac{1}{2}}r) \int_0^\tau I_{|u|}((2\lambda)^{\frac{1}{2}}x) 2x dx \\ &\quad + I_{|u|}((2\lambda)^{\frac{1}{2}}r) \int_r^\infty K_{|u|}((2\lambda)^{\frac{1}{2}}x) 2x dx, \end{aligned}$$

where $K_\nu(z)$ and $I_\nu(z)$ are the Bessel functions as defined in [5], p. 77–78. We substitute the expression on the right hand side of (3.2) in (3.1) and let $\lambda \rightarrow 0$; then we find $E_r\{\int_0^r e^{iu\theta(t)} dt\}$ by a careful analysis of the right hand side.

4. Asymptotic estimates of the function $\phi(\lambda, u; r)$ for $\lambda \rightarrow 0$.

LEMMA 4.1. For $u \neq 0$,

$$\lim_{\lambda \rightarrow 0} K_{|u|}((2\lambda)^{\frac{1}{2}}r) \int_0^r I_{|u|}((2\lambda)^{\frac{1}{2}}x) 2x dx = r^2/|u|(|u| + 2).$$

PROOF. Recall the formula [5], p. 183:

$$(4.1) \quad K_\nu(z) = \frac{1}{2}(z/2)^\nu \int_0^\infty \exp[-t - z^2/4t] t^{-\nu-1} dt.$$

It shows that (by the change of variable $t \rightarrow z^2/4t$)

$$K_{|u|}((2\lambda)^{\frac{1}{2}}r) \sim 2^{|u|-1} \Gamma(|u|) r^{-|u|} (2\lambda)^{-|u|/2}, \quad \lambda \rightarrow 0,$$

for $|u| \neq 0$. The formula [5], p. 79,

$$(4.2) \quad I_\nu(z) = [(z/2)^\nu / \Gamma(\nu + \frac{1}{2}) \Gamma(\frac{1}{2})] \int_0^\pi e^{z \cos \theta} \sin^{2\nu} \theta d\theta$$

shows that

$$\int_0^r I_{|u|}((2\lambda)^{\frac{1}{2}}x) 2x dx \sim (2\lambda)^{|u|/2} 2^{1-|u|} r^{|u|+2} / (|u| + 2) \Gamma(|u| + 1), \quad \lambda \rightarrow 0.$$

The lemma follows from these two estimates.

LEMMA 4.2

$$\lim_{\lambda \rightarrow 0} I_{|u|}((2\lambda)^{\frac{1}{2}}r) \int_r^\infty K_{|u|}((2\lambda)^{\frac{1}{2}}x) 2x dx = r^2/|u|(|u| - 2), \quad |u| > 2.$$

PROOF. Formula (4.2) implies that

$$I_{|u|}((2\lambda)^{\frac{1}{2}}r) \sim (2\lambda)^{|u|/2} r^{|u|} / 2^{|u|} \Gamma(|u| + 1), \quad \lambda \rightarrow 0.$$

Formula (4.1) implies that

$$\int_r^\infty K_{|u|}(zx) 2x dx = (2/z^2) \int_0^\infty e^{-t} t^{-(|u|/2)} dt \int_{r^2/z^2}^\infty y^{|u|/2} e^{-y} dy,$$

which, by the change of variable $t \rightarrow tz^2$, is equal to

$$2z^{-|u|} \int_0^\infty e^{-tz^2} t^{-(|u|/2)} [\int_{r^2/4t}^\infty y^{|u|/2} e^{-y} dy] dt \\ \sim 2z^{-|u|} \int_0^\infty t^{-(|u|/2)} [\int_{r^2/4t}^\infty y^{|u|/2} e^{-y} dy] dt, \quad (z \rightarrow 0).$$

Integration by parts transforms the latter expression into $z^{-|u|} r^{2-|u|} 2^{|u|} \Gamma(|u|) / (|u| - 2)$. Putting $z = (2\lambda)^{\frac{1}{2}}$, we obtain the result of the lemma.

LEMMA 4.3. $I_1((2\lambda)^{\frac{1}{2}}r) = r(\lambda/2)^{\frac{1}{2}} + O(\lambda^{\frac{3}{2}})$, $\lambda \rightarrow 0$.

PROOF. Put $z = (2\lambda)^{\frac{1}{2}}$, $\nu = 1$ in formula (4.2); then expand the exponential in the integrand and integrate. Note that the coefficient of λ vanishes because of the relation $\int_0^\pi \cos \theta \sin^2 \theta d\theta = 0$.

LEMMA 4.4.

$$\int_r^\infty K_1((2\lambda)^{\frac{1}{2}}x) 2x dx = (1/\lambda) [\int_0^\infty K_1(x) x dx - r(2\lambda)^{\frac{1}{2}} y K_1(y)],$$

where y is some number satisfying $0 \leq y \leq r(2\lambda)^{\frac{1}{2}}$.

PROOF. By a change of variable, the integral on the left hand side of the above equation is equal to $(1/\lambda) \int_{r(2\lambda)^{\frac{1}{2}}}^{\infty} K_1(x)x dx$. Apply the law of the mean to this integral considered as a function of the lower limit of integration.

LEMMA 4.5.

$$I_2((2\lambda)^{\frac{1}{2}}r) = \lambda r^2/4 + O(\lambda^2).$$

PROOF. The calculation is similar to that for Lemma 4.3.

LEMMA 4.6.

$$\int_r^{\infty} K_2((2\lambda)^{\frac{1}{2}}x)2x dx = \lambda^{-1}[r(2\lambda)^{\frac{1}{2}}K_1(r(2\lambda)^{\frac{1}{2}}) + 2K_0(r(2\lambda)^{\frac{1}{2}})].$$

PROOF. The expressions on each side of the equation tend to 0 as $r \rightarrow \infty$. The validity of the equation then follows from the equality of the derivatives with respect r , implied by the formulae [5], p. 79:

$$K_0'(x) = -K_1(x), \quad -xK_2(x) = xK_1'(x) - K_1(x).$$

LEMMA 4.7.

$$\lim_{\lambda \rightarrow 0} \phi(\lambda, u; r) = 2r^2/(|u| + 2)(|u| - 2), \quad |u| > 2.$$

PROOF. We apply Lemmas 4.1 and 4.2 and formula (3.2)

LEMMA 4.8.

$$\phi(\lambda, 1; r) = -\frac{2}{3}r^2 + r(2\lambda)^{-\frac{1}{2}} \int_0^{\infty} K_1(x)x dx + o(1), \quad \lambda \rightarrow 0.$$

PROOF. Apply Lemmas 4.1, 4.3, and 4.4, and formula (3.2). We also use the fact, deducible from (4.1), that

$$(4.3) \quad \lim_{y \rightarrow 0} yK_1(y) = 1.$$

LEMMA 4.9.

$$\phi(\lambda, 2; r) = 3r^2/8 + (r^2/2)[- \log r - \log (2\lambda)^{\frac{1}{2}} + C] + o(1), \quad \lambda \rightarrow 0,$$

where C is a numerical constant independent of r .

PROOF. Recall the formula [5], p. 80,

$$(4.4) \quad K_0(z) = -\log z + o(z) + C, \quad z \rightarrow 0,$$

where C is a constant; Lemma 4.6 and formula (4.3) imply

$$I_2((2\lambda)^{\frac{1}{2}}r) \int_r^{\infty} K_2((2\lambda)^{\frac{1}{2}}x)2x dx = r^2/4 + (r^2/2) K_0(r(2\lambda)^{\frac{1}{2}}) + o(1),$$

$\lambda \rightarrow 0.$

The lemma follows from the formulae (4.4) and (3.2), and Lemma 4.1

5. Computation of $L(f)$. We shall now find the explicit form of the functional $L(f)$ given in (1.1). In accordance with the remarks in Section 3, we omit the subscript θ from the expectation operator.

LEMMA 5.1. *The functional $L(e^{iu\theta})$ is equal to*

$$\begin{aligned} \frac{1}{2}(r_1^2 - r^2), & \quad u = 0, \\ \frac{2}{3}r(r_1 - r), & \quad |u| = 1, \\ 2r^2 \log(r_1/r), & \quad |u| = 2, \\ [2/(|u| + 2) (|u| - 2)] [r^2 - (r/r_1)^{|u|} r_1^2], & \quad |u| > 2. \end{aligned}$$

PROOF. The case $|u| > 2$ is a direct result of equation (3.1), Lemma 4.7, and formula (2.5). The cases $|u| = 1, 2$ follow from equation (3.1), Lemmas 4.8 and 4.9, the formula (2.5), and the asymptotic relation for fixed u and r :

$$E_r[e^{-\lambda r + iu\theta(r)}] = E_r[e^{iu\theta(r)}] + O(\lambda), \quad \lambda \rightarrow 0.$$

The expression for $u = 0$ is apparently well known; it can be verified by calculations similar to those for the other cases, or from the infinitesimal generator of the process $R(t)$.

THEOREM. *Let $f(\theta)$, $0 \leq \theta \leq 2\pi$, be a bounded measurable function such that $f(0) = f(2\pi)$. Let us denote by $\{f_n\}$ its sequence of Fourier coefficients:*

$$f_n = (1/2\pi) \int_0^{2\pi} e^{in\theta} f(\theta) d\theta, \quad n = 0, \pm 1, \dots;$$

put

$$d_n = E_r \left\{ \int_0^r e^{in\theta(t)} dt \right\}, \quad n = 0, \pm 1, \dots;$$

then,

$$(5.1) \quad L(f) = \sum_{n=-\infty}^{\infty} f_n d_n.$$

PROOF. The sequence $\{d_n\}$ is square-summable by Lemma 5.1; hence, the functional $M(f) = \sum_{n=-\infty}^{\infty} f_n d_n$ is defined for any f in $L_2[0, 2\pi]$, and is (linear and) continuous over $L_2[0, 2\pi]$. Let us denote by $N(f)$ the restriction of the functional (1.1) to the space $C[0, 2\pi]$ of continuous functions f on $[0, 2\pi]$; $N(f)$ is continuous on $C[0, 2\pi]$. The two functionals M and N are identical for functions f of the form $f(\theta) = e^{im\theta}$, ($m = \text{integer}$); by linearity, they are also identical for linear combinations of such functions. The class of linear combinations of complex exponential functions is dense in $C[0, 2\pi]$ (Weierstrass' theorem), and convergence in $C[0, 2\pi]$ implies convergence in $L_2[0, 2\pi]$; therefore, $N(f) = M(f)$ for all f in $C[0, 2\pi]$ by the continuity of these functionals.

We have shown that $M(f) = L(f)$ for all continuous functions f . The indicator function of an interval is the limit of a monotone sequence of continuous functions, where the convergence is both pointwise and in $L_2[0, 2\pi]$; therefore, M and L coincide over indicator functions since they are positive functionals, and since L is continuous on convergent monotone sequences. The equality of M and L for all bounded measurable functions now follows from conventional approximation arguments.

We point out an alternative form of the functional $L(f)^2$. By the Riesz-Fischer theorem there exists a function g in $L_2[0, 2\pi]$ whose Fourier coefficients are the

elements of the sequence $\{d_n\}$. The function g is real-valued because d_n is real and $d_n = d_{-n}$, $n = 0, 1, 2, \dots$. Parseval's theorem implies that

$$L(f) = (1/2\pi) \int_0^{2\pi} f(\theta)g(\theta)d\theta, \quad f \text{ in } L_2[0, 2\pi].$$

6. Another proof of Theorem 2.1. The referee has presented another proof of Theorem 2.1. The distribution of the random variable considered in that theorem depends on the pair (r, r_1) only through r_1/r because $\{Z(t)\}$ is stochastically equivalent to $\{c^{-\frac{1}{2}}Z(ct)\}$ for any fixed $c > 0$. Define \hat{r} by the equation $\log \hat{r} = \frac{1}{2} \log (r_1 r)$; let $\hat{\theta}$ correspond to \hat{r} as $\theta(\tau)$ does to r_1 ; then $\hat{\theta} - \theta(0)$ and $\theta(\tau) - \hat{\theta}$ are independent, identically distributed random variables. By a defining property of the Cauchy distribution it suffices to prove that $\theta(\tau) - \theta(0)$ ($= \theta(\tau) - \hat{\theta} + \hat{\theta} - \theta(0)$) is distributed as $2(\hat{\theta} - \theta(0))$. Now $\log Z(t) = \log R(t) + i\theta_1(t)$ has real and imaginary parts which define a diffusion in R^2 which, since $\log z$ is analytic, differs from $Z(t)$ only by a random time change [2], p. 280. The same is true of $2 \log Z(t)$, or more generally, of $c \log Z(t)$, $c \neq 0$. These processes have identical hitting probabilities; hence the assertion about $\theta(\tau) - \theta(0)$ follows.

² ADDED IN PROOF: The functional L has a representation as an integral operator whose kernel is the Green's function for the circle; however, I have not been able to deduce (5.1) directly from it. I would like to thank Zbigniew Ciesielski and Walter Rosenkrantz for their informative comments about this representation.

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