

## HITTING TIMES FOR GAUSSIAN PROCESSES

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We establish a general formula for the Laplace transform of the hitting times of a Gaussian process. Some consequences are derived, and particular cases like the fractional Brownian motion are discussed.

**1. Introduction.** Consider a zero mean continuous Gaussian process  $(X_t, t \geq 0)$ , and for any  $a > 0$ , we denote by  $\tau_a$  the hitting time of the level  $a$  defined by

$$(1.1) \quad \tau_a = \inf\{t \geq 0 : X_t = a\} = \inf\{t \geq 0 : X_t \geq a\}.$$

Thus, the map  $(a \mapsto \tau_a)$  is left-continuous and increasing, hence, with right limits. The map  $(a \mapsto \tau_{a+})$  is right continuous where

$$\tau_{a+} = \lim_{b \downarrow a, b > a} \tau_a = \inf\{t \geq 0 : X_t > a\}.$$

Little is known about the distribution of  $\tau_a$ . It is explicitly known in particular cases like the Brownian motion. If  $X$  is a fractional Brownian motion with Hurst parameter  $H$ , there is a result by Molchan [5] which stands that

$$P(\tau_a > t) = t^{-(1-H)+o(1)}$$

as  $t$  goes to infinity.

When  $X$  is a standard Brownian motion, it is well known that

$$(1.2) \quad E(\exp(-\alpha\tau_a)) = \exp(-a\sqrt{2\alpha})$$

for all  $\alpha > 0$ . This result is easily proved using the exponential martingale

$$M_t = \exp(\lambda B_t - \frac{1}{2}\lambda^2 t).$$

By Doob's optional stopping theorem applied at time  $t \wedge \tau_a$  and letting  $t \rightarrow \infty$ , one gets  $1 = E(M_{\tau_a}) = E(\exp(\lambda B_{\tau_a} - \lambda^2 \tau_a / 2))$ . Since  $B_{\tau_a} = a$ , we thus obtain (1.2). If we consider a general Gaussian process  $X_t$ , the exponential process

$$M_t = \exp(\lambda X_t - \frac{1}{2}\lambda^2 V_t),$$

where  $V_t = E(X_t^2)$  is no longer a martingale. However, it is equal to 1 plus a divergence integral in the sense of Malliavin calculus. The aim of this paper is to

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take advantage of this fact in order to derive a formula for  $E(\exp(-\frac{1}{2}\lambda^2 V_{\tau_a}))$ . We derive an equation involving this expectation in Theorem 3.4, under rather general conditions on the covariance of the process. As a consequence, we show that if the partial derivative of the covariance is nonnegative, then  $E(\exp(-\frac{1}{2}\lambda^2 V_{\tau_a})) \leq 1$ , which implies that  $V_{\tau_a}$  has infinite moments of order  $p$  for all  $p \geq \frac{1}{2}$  and finite negative moments of all orders. In particular, for the fractional Brownian motion with Hurst parameter  $H > \frac{1}{2}$ , we have the inequality

$$E(\exp -\alpha \tau_a^{2H}) \leq \exp(-a\sqrt{2\alpha})$$

for all  $\alpha, a > 0$ .

The paper is organized as follows. In Section 2 we present some preliminaries on Malliavin calculus, and the main results are proved in Section 3.

**2. Preliminaries on Malliavin calculus.** Let  $(X_t, t \geq 0)$  be a zero mean Gaussian process such that  $X_0 = 0$  and with covariance function

$$R(s, t) = E(X_t X_s).$$

We denote by  $\mathcal{E}$  the set of step functions on  $[0, +\infty)$ . Let  $\mathcal{H}$  be the Hilbert space defined as the closure of  $\mathcal{E}$  with respect to the scalar product

$$\langle \mathbf{1}_{[0,t]}, \mathbf{1}_{[0,s]} \rangle_{\mathcal{H}} = R(t, s).$$

The mapping  $\mathbf{1}_{[0,t]} \rightarrow X_t$  can be extended to an isometry between  $\mathcal{H}$  and the Gaussian space  $H_1(X)$  associated with  $X$ . We will denote this isometry by  $\varphi \rightarrow X(\varphi)$ .

Let  $\mathcal{F}$  be the set of smooth and cylindrical random variables of the form

$$(2.1) \quad F = f(X(\phi_1), \dots, X(\phi_n)),$$

where  $n \geq 1, f \in \mathcal{C}_b^\infty(\mathbb{R}^n)$  ( $f$  and all its partial derivatives are bounded), and  $\phi_i \in \mathcal{H}$ .

The *derivative operator*  $D$  of a smooth and cylindrical random variable  $F$  of the form (2.1) is defined as the  $\mathcal{H}$ -valued random variable

$$DF = \sum_{i=1}^n \frac{\partial f}{\partial x_i}(X(\phi_1), \dots, X(\phi_n))\phi_i.$$

The derivative operator  $D$  is then a closable operator from  $L^2(\Omega)$  into  $L^2(\Omega; \mathcal{H})$ . The Sobolev space  $\mathbb{D}^{1,2}$  is the closure of  $\mathcal{F}$  with respect to the norm

$$\|F\|_{1,2}^2 = E(F^2) + E(\|DF\|_{\mathcal{H}}^2).$$

The *divergence operator*  $\delta$  is the adjoint of the derivative operator. We say that a random variable  $u$  in  $L^2(\Omega; \mathcal{H})$  belongs to the domain of the divergence operator, denoted by  $\text{Dom } \delta$ , if

$$|E(\langle DF, u \rangle_{\mathcal{H}})| \leq c_u \|F\|_{L^2(\Omega)}$$

for any  $F \in \mathcal{F}$ . In this case  $\delta(u)$  is defined by the duality relationship

$$(2.2) \quad E(F\delta(u)) = E(\langle DF, u \rangle_{\mathcal{H}}),$$

for any  $F \in \mathbb{D}^{1,2}$ .

Set  $V_t = R(t, t)$ . For any  $\lambda > 0$ , we define

$$M_t = \exp(\lambda X_t - \frac{1}{2}\lambda^2 V_t).$$

Formally, the Itô formula for the divergence integral, proved, for instance, in [1], implies that

$$(2.3) \quad M_t = 1 + \lambda\delta(M\mathbf{1}_{[0,t]}),$$

where  $M\mathbf{1}_{[0,t]}$  represents the process  $(s \mapsto M_s\mathbf{1}_{[0,t]}(s), s \geq 0)$ . However, the process  $M\mathbf{1}_{[0,t]}$  does not belong, in general, to the domain of the divergence operator. This happens, for instance, in the following basic example.

EXAMPLE 1. Fractional Brownian motion with Hurst parameter  $H \in (0, 1)$  is a zero mean Gaussian process  $(B_t^H, t \geq 0)$  with the covariance

$$(2.4) \quad R_H(t, s) = \frac{1}{2}(t^{2H} + s^{2H} - |t - s|^{2H}).$$

In this case, the processes  $(B_s^H\mathbf{1}_{[0,t]}(s), s \geq 0)$  and  $(\exp(\lambda B_s^H - \frac{1}{2}\lambda^2 s^{2H})\mathbf{1}_{[0,t]}(s), s \geq 0)$  do not belong to  $L^2(\Omega; \mathcal{H})$  if  $H \leq \frac{1}{4}$  (see [2]).

In order to define the divergence of  $M\mathbf{1}_{[0,t]}$  and to establish formula (2.3), we introduce the following additional property on the covariance function of the process  $X$ .

(H0) The covariance function  $R(t, s)$  is continuous, the partial derivative  $\frac{\partial R}{\partial s}(s, t)$  exists in the region  $\{0 < s, t, s \neq t\}$ , and for all  $T > 0$ ,

$$\sup_{t \in [0, T]} \int_0^T \left| \frac{\partial R}{\partial s}(s, t) \right| ds < \infty.$$

Notice that this property is satisfied by the covariance (2.4) for all  $H \in (0, 1)$ .

Define

$$(2.5) \quad \delta_t M = \frac{1}{\lambda}(M_t - 1).$$

The following proposition asserts that  $\delta_t M$  satisfies an integration by parts formula, and in this sense, it coincides with an extension of the divergence of  $M\mathbf{1}_{[0,t]}$ .

PROPOSITION 2.1. Suppose that (H0) holds. Then, for any  $t > 0$ , and for any smooth and cylindrical random variable of the form  $F = f(X_{t_1}, \dots, X_{t_n})$ , we have

$$(2.6) \quad E(F\delta_t M) = E\left(\sum_{i=1}^n \frac{\partial f}{\partial x_i}(X_{t_1}, \dots, X_{t_n}) \int_0^t M_s \frac{\partial R}{\partial s}(s, t_i) ds\right).$$

PROOF. First notice that condition (H0) implies that the right-hand side of equation (2.6) is well defined. Then, it suffices to show equation (2.6) for a function of the form

$$f(x_1, \dots, x_n) = \exp\left(\sum_{i=1}^n \lambda_i x_i\right),$$

where  $\lambda_i \in \mathbb{R}$ . In this case we have, for all  $0 < t_1 < \dots < t_n$ ,

$$\begin{aligned} & \frac{1}{\lambda} E(F(M_t - 1)) \\ &= \frac{1}{\lambda} \exp\left\{\frac{1}{2} \sum_{i=1}^n \lambda_i \lambda_j R(t_i, t_j)\right\} \left(\exp\left\{\sum_{i=1}^n \lambda \lambda_i R(t, t_i)\right\} - 1\right) \\ &= \sum_{i=1}^n \int_0^t \exp\left\{\frac{1}{2} \sum_{i=1}^n \lambda_i \lambda_j R(t_i, t_j) + \lambda \sum_{i=1}^n \lambda_i R(s, t_i)\right\} \lambda_i \frac{\partial R}{\partial s}(s, t_i) ds \\ &= \int_0^t E\left(\sum_{i=1}^n \frac{\partial f}{\partial x_i}(X_{t_1}, \dots, X_{t_n}) M_s \frac{\partial R}{\partial s}(s, t_i)\right) ds, \end{aligned}$$

which completes the proof of the proposition.  $\square$

In many cases like in Example 1 with  $H > \frac{1}{4}$ , the process  $M\mathbf{1}_{[0,t]}$  belongs to the space  $L^2(\Omega; \mathcal{H})$ , and then the right-hand side of equation (2.6) equals

$$E\langle DF, M\mathbf{1}_{[0,t]} \rangle_{\mathcal{H}}.$$

In this situation, taking into account the duality formula (2.2), equation (2.6) says that  $M\mathbf{1}_{[0,t]}$  belongs to the domain of the divergence and  $\delta(M\mathbf{1}_{[0,t]}) = \delta_t M$ .

**3. Hitting times.** In this section we will assume the following conditions:

- (H1) The partial derivative  $\frac{\partial R}{\partial s}(s, t)$  exists and it is continuous in  $[0, +\infty)^2$ .
- (H2)  $\limsup_{t \rightarrow \infty} X_t = +\infty$  almost surely.
- (H3) For any  $0 \leq s < t$ , we have  $E(|X_t - X_s|^2) > 0$ .

Under these conditions, the process  $X$  has a continuous version because

$$\begin{aligned} E(|X_t - X_s|^2) &= R(t, t) + R(s, s) - 2R(s, t) \\ &= \int_s^t \left[ \frac{\partial R}{\partial u}(u, t) - \frac{\partial R}{\partial u}(u, s) \right] du \\ &\leq 2|t - s| \sup_{s \leq u \leq t} \left| \frac{\partial R}{\partial u}(u, t) \right|. \end{aligned}$$

For any  $a > 0$ , we define the hitting time  $\tau_a$  by (1.1). We know that  $P(\tau_a < \infty) = 1$  by condition (H2). Set

$$(3.1) \quad S_t = \sup_{s \in [0,t]} X_s.$$

From the results of [6], it follows that, for all  $t > 0$ , the random variable  $S_t$  belongs to the space  $\mathbb{D}^{1,2}$ . Furthermore, condition (H3) allows us to compute the derivative of this random variable.

LEMMA 3.1. *For all  $t > 0$ , with probability one, the maximum of the process  $X$  in the interval  $[0, t]$  is attained in a unique point, that is,  $\tau_{S_t} = \tau_{S_t}^+$  and  $DS_t = \mathbf{1}_{[0, \tau_{S_t}]}$ .*

PROOF. The fact that the maximum is attained in a unique point follows from condition (H3) and Lemma 2.6 in Kim and Pollard [4]. The formula for the derivative of  $S_t$  follows easily by an approximation argument.  $\square$

We need the following regularization of the stopping time  $\tau_a$ . Suppose that  $\varphi$  is a nonnegative smooth function with compact support in  $(0, +\infty)$  and define for any  $T > 0$

$$(3.2) \quad Y = \int_0^\infty \varphi(a)(\tau_a \wedge T) da.$$

The next result states the differentiability of the random variable  $Y$  in the sense of Malliavin calculus and provides an explicit formula for its derivative.

LEMMA 3.2. *The random variable  $Y$  defined in (3.2) belongs to the space  $\mathbb{D}^{1,2}$ , and*

$$(3.3) \quad D_r Y = - \int_0^{S_T} \varphi(y) \mathbf{1}_{[0, \tau_y]}(r) d\tau_y.$$

PROOF. Clearly,  $Y$  is bounded. On the other hand, for any  $r > 0$ , we have

$$\{\tau_a > r\} = \{S_r < a\}.$$

Therefore, we can write using Fubini's theorem

$$Y = \int_0^\infty \varphi(a) \left( \int_0^{\tau_a \wedge T} d\theta \right) da = \int_0^T \left( \int_{S_\theta}^\infty \varphi(a) da \right) d\theta,$$

which implies that  $Y \in \mathbb{D}^{1,2}$  because  $S_\theta \in \mathbb{D}^{1,2}$ , and

$$D_r Y = - \int_0^T \varphi(S_\theta) D_r S_\theta d\theta = - \int_0^T \varphi(S_\theta) \mathbf{1}_{[0, \tau_{S_\theta}]}(r) d\theta.$$

Finally, making the change of variable  $S_\theta = y$  yields

$$D_r Y = - \int_0^{S_T} \varphi(y) \mathbf{1}_{[0, \tau_y]}(r) d\tau_y. \quad \square$$

Notice that  $M_Y = \exp(\lambda X_Y - \frac{1}{2} \lambda^2 V_Y)$ . Hence, letting  $t = Y$  in equation (2.5) and taking the mathematical expectation of both members of the equality yields

$$(3.4) \quad E(M_Y) = 1 + \lambda E(\delta_t M|_{t=Y}).$$

We are going to show the following result which provides a formula for the left-hand side of equation (3.4).

LEMMA 3.3. *Assume conditions (H1), (H2) and (H3). Then, we have*

$$(3.5) \quad E(M_Y) = 1 - \lambda E\left(M_Y \int_0^{S_T} \varphi(y) \frac{\partial R}{\partial s}(Y, \tau_y) d\tau_y\right).$$

PROOF. The proof will be done in two steps.

Step 1. We claim that for any function  $p(x)$  in  $\mathcal{C}_0^\infty(\mathbb{R})$  we have

$$(3.6) \quad E(\delta_t M p(Y)) = -E\left(\int_0^t M_s p'(Y) \int_0^{S_T} \varphi(y) \frac{\partial R}{\partial s}(s, \tau_y) d\tau_y ds\right).$$

We can write  $Y = \int_0^T \psi(S_\theta) d\theta$ , where  $\psi(x) = \int_x^\infty \varphi(a) da$ . Consider an increasing sequence  $D_n$  of finite subsets of  $[0, T]$  such that their union is dense in  $[0, T]$ . Set  $Y_n = \int_0^T \psi(S_\theta^n) d\theta$ , and  $S_\theta^n = \max\{X_t, t \in D_n \cap [0, \theta]\}$ . Then,  $Y_n$  is a Lipschitz function of  $\{X_t, t \in D_n\}$ . Hence, formula (2.6), which holds for Lipschitz functions, implies that

$$E(\delta_t M p(Y_n)) = -E\left(p'(Y_n) \int_0^T \varphi(S_\theta^n) \left(\int_0^t M_s \frac{\partial R}{\partial s}(s, \tau_{S_\theta^n}) ds\right) d\theta\right).$$

The function  $r \rightarrow \int_0^t M_s \frac{\partial R}{\partial s}(s, r) ds$  is continuous and bounded by condition (H1). As a consequence, we can take the limit of the above expression as  $n$  tends to infinity and we get

$$E(\delta_t M p(Y)) = -E\left(p'(Y) \int_0^T \varphi(S_\theta) \left(\int_0^t M_s \frac{\partial R}{\partial s}(s, \tau_{S_\theta}) ds\right) d\theta\right).$$

Finally, making the change of variable  $S_\theta = y$  yields (3.6).

Step 2. We write

$$E(\delta_t M|_{t=Y}) = E\left(\lim_{\varepsilon \rightarrow 0} \int_{-\infty}^\infty \delta_t M p_\varepsilon(Y - t) dt\right),$$

where  $p_\varepsilon(x)$  is an approximation of the identity, and by convention, we assume that  $\delta_t M = 0$  if  $t$  is negative. We can commute the expectation with the above

limit by the dominated convergence theorem because

$$\begin{aligned} \int_{-\infty}^{\infty} |\delta_t M| p_\varepsilon(Y - t) dt &= \int_{-\infty}^{\infty} \frac{1}{\lambda} |M_t - 1| p_\varepsilon(Y - t) dt \\ &\leq \frac{1}{\lambda} \sup_{0 \leq t \leq T+1} (|M_t| + 1), \end{aligned}$$

if the support of  $p_\varepsilon(x)$  is included in  $[-\varepsilon, \varepsilon]$ , and  $\varepsilon \leq 1$ . Hence,

$$(3.7) \quad E(\delta_t M|_{t=Y}) = \lim_{\varepsilon \rightarrow 0} \int_{-\infty}^{\infty} E(\delta_t M p_\varepsilon(Y - t)) dt.$$

Using formula (3.6) yields

$$(3.8) \quad \begin{aligned} &E(\delta_t M p_\varepsilon(Y - t)) \\ &= - \int_0^t E\left(p'_\varepsilon(Y - t) M_s \left(\int_0^{S_T} \varphi(y) \frac{\partial R}{\partial s}(s, \tau_y) d\tau_y\right)\right) ds. \end{aligned}$$

Hence, substituting (3.8) into (3.7) and integrating by parts, we obtain

$$\begin{aligned} &E(\delta_t M|_{t=Y}) \\ &= - \lim_{\varepsilon \rightarrow 0} E\left(\int_{-\infty}^{\infty} p'_\varepsilon(Y - t) \left(\int_0^t M_s \left(\int_0^{S_T} \varphi(y) \frac{\partial R}{\partial s}(s, \tau_y) d\tau_y\right) ds\right) dt\right) \\ &= - \lim_{\varepsilon \rightarrow 0} E\left(\int_{-\infty}^{\infty} p_\varepsilon(Y - t) \left(M_t \int_0^{S_T} \varphi(y) \frac{\partial R}{\partial t}(t, \tau_y) d\tau_y\right) dt\right). \end{aligned}$$

Notice that

$$\left| \int_0^{S_T} \varphi(y) \frac{\partial R}{\partial s}(s, \tau_y) d\tau_y \right| \leq T \sup_{0 \leq s, u \leq T} \left| \frac{\partial R}{\partial s}(s, u) \right| \|\varphi\|_\infty.$$

Hence, applying the dominated convergence theorem, we get

$$\begin{aligned} E(M_Y) &= 1 + \lambda E(\delta_t M|_{t=Y}) \\ &= 1 - \lambda \lim_{\varepsilon \rightarrow 0} E\left(\int_{-\infty}^{\infty} p_\varepsilon(Y - t) \left(M_t \int_0^{S_T} \varphi(y) \frac{\partial R}{\partial s}(t, \tau_y) d\tau_y\right) dt\right) \\ &= 1 - \lambda E\left(M_Y \int_0^{S_T} \varphi(y) \frac{\partial R}{\partial s}(Y, \tau_y) d\tau_y\right). \end{aligned} \quad \square$$

The next step will be to replace the function  $\varphi(x)$  by an approximation of the identity and let  $T$  tend to infinity. Notice that (3.5) still holds for  $\varphi(x) = \mathbf{1}_{[0,b]}(x)$  for any  $b \geq 0$ . In this way we can establish the following result.

**THEOREM 3.4.** *Assume conditions (H1), (H2) and (H3). For any  $a > 0$  and  $\lambda \in \mathbb{R}$ , we have*

$$(3.9) \quad \int_0^a E(M_{\tau_y}) dy = a - \lambda E \left( \int_0^a \int_0^1 M_{z\tau_y + (1-z)\tau_y} \frac{\partial R}{\partial s}(z\tau_y + (1-z)\tau_y, \tau_y) dz d\tau_y \right).$$

Notice that we are not able to differentiate with respect to  $a$ , the integral in the rightmost expectation of (3.9), because the (random) measure  $d\tau_y$ , in general, is not absolutely continuous with respect to the Lebesgue measure.

**PROOF OF THEOREM 3.4.** Fix  $a > 0$ . We first replace the function  $\varphi(x)$  by an approximation of the identity of the form  $\varphi_\varepsilon(x) = \varepsilon^{-1} \mathbf{1}_{[0,1]}(x/\varepsilon)$  in formula (3.5). We will make use of the following notation:

$$Y_{\varepsilon,a} = \int_0^\infty \varphi_\varepsilon(x - a)(\tau_x \wedge T) dx.$$

At the same time we fix a nonnegative smooth function  $\psi(x)$  with compact support such that  $\int_{\mathbb{R}} \psi(a) da = c$  and we set

$$\int_{\mathbb{R}} E(M_{Y_{\varepsilon,a}}) \psi(a) da = c - \lambda \int_{\mathbb{R}} E \left( M_{Y_{\varepsilon,a}} \int_0^{S_T} \varphi_\varepsilon(y - a) \frac{\partial R}{\partial s}(Y_{\varepsilon,a}, \tau_y) d\tau_y \right) \psi(a) da.$$

The increasing property of the function  $x \rightarrow \tau_x$  implies that  $\tau_{a^+} \wedge T \leq Y_{\varepsilon,a} \leq \tau_{a+\varepsilon} \wedge T$ . Hence,  $Y_\varepsilon$  converges to  $\tau_{a^+} \wedge T$  as  $\varepsilon$  tends to zero. Thus, almost surely, we have

$$\lim_{\varepsilon \rightarrow 0} M_{Y_{\varepsilon,a}} = \exp(\lambda X_{\tau_{a^+} \wedge T} - \frac{1}{2} \lambda^2 V_{\tau_{a^+} \wedge T}).$$

By the dominated convergence theorem,

$$\lim_{\varepsilon \rightarrow 0} \int_{\mathbb{R}} E(M_{Y_{\varepsilon,a}}) \psi(a) da = \int_{\mathbb{R}} E(\exp(\lambda X_{\tau_{a^+} \wedge T} - \frac{1}{2} \lambda^2 V_{\tau_{a^+} \wedge T})) \psi(a) da.$$

Now, set  $F(t) = M_t \frac{\partial R}{\partial s}(t, \tau_y)$ . Then, assuming that  $\varphi_\varepsilon(x) = \varepsilon^{-1} \mathbf{1}_{[0,1]}(x/\varepsilon)$ , we have

$$\begin{aligned} & \int_{y-\varepsilon}^y \varphi_\varepsilon(y - a) M_{Y_{\varepsilon,a}} \frac{\partial R}{\partial s}(Y_{\varepsilon,a}, \tau_y) \psi(a) da \\ &= \frac{1}{\varepsilon^2} \int_{y-\varepsilon}^y \mathbf{1}_{[0,1]} \left( \frac{y - a}{\varepsilon} \right) F \left( \int_a^{a+\varepsilon} \mathbf{1}_{[0,1]} \left( \frac{x - a}{\varepsilon} \right) (\tau_x \wedge T) dx \right) \psi(a) da \\ &= \int_0^1 F \left( \int_0^1 (\tau_{y+\varepsilon\xi - \varepsilon\eta} \wedge T) d\xi \right) \psi(y - \varepsilon\eta) d\eta \end{aligned}$$

$$= \int_0^1 F\left(\int_0^\eta (\tau_{y+\varepsilon\xi-\varepsilon\eta} \wedge T) d\xi + \int_\eta^1 (\tau_{y+\varepsilon\xi-\varepsilon\eta} \wedge T) d\xi\right) \psi(y - \varepsilon\eta) d\eta.$$

As  $\varepsilon$  tends to zero, this expression clearly converges to

$$\psi(y) \int_0^1 F(\eta(\tau_y \wedge T) + (1 - \eta)(\tau_{y^+} \wedge T)) d\eta.$$

So, we have proved that

$$(3.10) \quad \begin{aligned} & \lim_{\varepsilon \rightarrow 0} \int_{\mathbb{R}} M_{Y_{\varepsilon,a}} \varphi_\varepsilon(y - a) \frac{\partial R}{\partial s}(Y_{\varepsilon,a}, \tau_y) \psi(a) da \\ &= \psi(y) \int_0^1 M_{z\tau_{y^+} + (1-z)\tau_y} \frac{\partial R}{\partial s}(z\tau_{y^+} + (1 - z)\tau_y, \tau_y) dz. \end{aligned}$$

In order to complete the proof of the theorem, we will apply the dominated convergence theorem. We have the following estimate for  $y \leq S_T$ :

$$\left| \int_{\mathbb{R}} M_{Y_{\varepsilon,a}} \varphi_\varepsilon(y - a) \frac{\partial R}{\partial s}(Y_{\varepsilon,a}, \tau_y) \psi(a) da \right| \leq \|\psi\|_\infty \sup_{s,t \leq T} \left| \frac{\partial R}{\partial s}(s, t) \right| \sup_{t \leq T} |M_t|,$$

which allows us to commute the limit (3.10) with the integral with respect to the measure  $P \times d\tau_y$  on the set  $\{(\omega, y) : y \leq S_T(\omega)\}$ . In this way we get

$$\begin{aligned} & \int_{\mathbb{R}} E(M_{\tau_y}) \psi(y) dy \\ &= \int_{\mathbb{R}} \psi(y) dy \\ & \quad - \lambda E\left(\int_0^{S_T} \psi(y) \int_0^1 M_{z\tau_{y^+} + (1-z)\tau_y} \frac{\partial R}{\partial s}(z\tau_{y^+} + (1 - z)\tau_y, \tau_y) dz d\tau_y\right). \end{aligned}$$

Approximating  $\mathbf{1}_{[0,a]}$  by a sequence of smooth functions  $(\psi_n, n \geq 1)$  and letting  $T$  tend to infinity completes the proof.  $\square$

If we assume that the partial derivative  $\frac{\partial R}{\partial t}(t, s)$  is nonnegative, then we can derive the following result.

**PROPOSITION 3.5.** *Assume that  $X$  satisfies hypotheses (H1), (H2) and (H3). If  $\frac{\partial R}{\partial s}(s, t) \geq 0$ , then, for all  $\alpha, a > 0$ , we have*

$$(3.11) \quad E(\exp(-\alpha V_{\tau_a})) \leq e^{-a\sqrt{2\alpha}}.$$

**PROOF.** Since  $\frac{\partial R}{\partial t}(t, s) \geq 0$ , we obtain

$$E(M_{\tau_a}) \leq 1,$$

that is,

$$E(\exp(\lambda a - \frac{1}{2}\lambda^2 V_{\tau_a})) \leq 1,$$

or

$$E(\exp(-\alpha V_{\tau_a})) \leq e^{-a\sqrt{2\alpha}}.$$

The result follows.  $\square$

The above proposition means that the Laplace transform of the random variable  $V_{\tau_a}$  is dominated by the Laplace transform of  $\tau_a$ , where  $\tau_a$  is the hitting time of the level  $a$  for the ordinary Brownian motion. This domination implies some consequences on the moments of  $V_{\tau_a}$ . In fact, for any  $r > 0$ , we have, multiplying (3.11) by  $\alpha^r$ ,

$$\begin{aligned} E(V_{\tau_a}^{-r}) &= \frac{1}{\Gamma(r)} \int_0^\infty E(e^{-\alpha V_{\tau_a}}) \alpha^{r-1} d\alpha \\ (3.12) \quad &\leq \frac{1}{\Gamma(r)} \int_0^\infty e^{-a\sqrt{2\alpha}} \alpha^{r-1} d\alpha \\ &= \frac{2^r \Gamma(r + 1/2)}{\sqrt{\pi}} a^{-2r}. \end{aligned}$$

On the other hand, for  $0 < r < 1$ ,

$$\begin{aligned} E(V_{\tau_a}^r) &= \frac{r}{\Gamma(1-r)} \int_0^\infty (1 - E(e^{-\alpha V_{\tau_a}})) \alpha^{-r-1} d\alpha \\ (3.13) \quad &\geq \frac{r}{\Gamma(1-r)} \int_0^\infty (1 - e^{-a\sqrt{2\alpha}}) \alpha^{-r-1} d\alpha. \end{aligned}$$

In particular, for  $r \in [1/2, 1)$ ,  $E(V_{\tau_a}^r) = +\infty$ .

REMARK 3.6. If  $X$  is the standard Brownian motion, its covariance  $s \wedge t$  does not satisfy condition (H1), but we still can apply our approach. It is known from [3] that  $d\tau_a$  has no absolutely continuous part and that  $\{a, \tau_a = \tau_a^+\}$  is a Cantor set, hence, of zero Lebesgue measure. It follows from this observation and from (3.10) that

$$\int E(M_{\tau_y}) \psi(y) dy = \int \psi(y) dy.$$

Choosing  $\psi = \mathbf{1}_{[0,a]}$  yields to the expected result:

$$E\left(\int_0^a e^{\lambda y - (\lambda^2/2)V(\tau_y)} dy\right) = a.$$

If  $X$  has independent increments and satisfies (H3), then

$$E(e^{-(\lambda^2/2)V(\tau_a)}) = e^{-\lambda a}.$$

This follows easily from the fact that  $X$  can be written as a time-changed Brownian motion.

REMARK 3.7. Consider that  $X$  is a fractional Brownian motion of Hurst index  $H = 1$ . Then  $R(s, t) = st$ , and consequently,  $X_t = Yt$ , where  $Y$  is a one-dimensional standard Gaussian random variable. Then,  $\tau_a = \tau_{a^+} = a/Y^+$ . It is then easy to compute the Laplace transform of  $\tau_a$  and we obtain

$$(3.14) \quad E(\exp(-\alpha\tau_a^2)) = \frac{1}{2}e^{-a\sqrt{2\alpha}}.$$

We show now that our formula also yields to the right answer. We just note that  $(y \mapsto \tau_y)$  is continuous. This entails that

$$\frac{\partial R}{\partial s}(z\tau_{y^+} + (1-z)\tau_y, \tau_y) = \frac{\partial R}{\partial s}(\tau_y, \tau_y) = \frac{1}{2}V'(\tau_y)$$

and

$$(3.15) \quad \int_0^a E\left(\exp\left(\lambda y - \frac{\lambda^2}{2}V(\tau_y)\right)\right) dy \\ = a - \frac{\lambda}{2}E\left(\int_0^a \exp\left(\lambda y - \frac{\lambda^2}{2}V(\tau_y)\right)V'(\tau_y) d\tau_y\right).$$

Set

$$\Psi(a, \lambda) = E\left(\exp\left(\lambda a - \frac{\lambda^2}{2}V(\tau_a)\right)\right),$$

then

$$(3.16) \quad \frac{\partial \Psi}{\partial a}(a, \lambda) = \lambda\Psi(a, \lambda) - \frac{\lambda^2}{2}E\left(M_{\tau_a} \frac{\partial V(\tau_a)}{\partial a}\right).$$

Substitute (3.15) into (3.16) to obtain

$$\frac{\partial \Psi}{\partial a} = 2\lambda\Psi - \lambda.$$

Then, there exists a function  $C(\lambda)$  such that

$$\Psi(a, \lambda) = \frac{1}{2} + C(\lambda)e^{2\lambda a} \quad \text{so that } E\left(\exp\left(-\frac{\lambda^2}{2}\tau_a^2\right)\right) = \frac{1}{2}e^{-\lambda a} + C(\lambda)e^{\lambda a}.$$

By dominated convergence, it is clear that, for any  $\lambda$ ,

$$E\left(\exp\left(-\frac{\lambda^2}{2}\tau_a^2\right)\right) \xrightarrow{a \rightarrow \infty} 0,$$

thus,  $C(\lambda) = 0$  and

$$E\left(\exp\left(-\frac{\lambda^2}{2}\tau_a^2\right)\right) = \frac{1}{2}e^{-\lambda a}.$$

Changing  $\lambda^2/2$  into  $\alpha$  gives (3.14).

REMARK 3.8. Consider the case of a fractional Brownian motion with Hurst parameter  $H > \frac{1}{2}$ . Conditions (H1), (H2) and (H3) are satisfied and we obtain

$$\begin{aligned} & \int_0^a E(M_{\tau_y}) dy \\ &= a - \lambda H E \left( \int_0^a \int_0^1 M_{z\tau_{y^+} + (1-z)\tau_y} ([z\tau_{y^+} + (1-z)\tau_y]^{2H-1} \right. \\ & \quad \left. - |z(\tau_{y^+} - \tau_y)|^{2H-1}) dz d\tau_y \right). \end{aligned}$$

Moreover,  $E(e^{-\alpha\tau_a^{2H}}) \leq e^{-a\sqrt{2\alpha}}$ , and therefore,  $E(\tau_a^p) < \infty$  if  $p < H$ . According to (3.13),  $E(\tau_a^p)$  is infinite if  $pH > 1/4$  and (3.12) entails that  $\tau_a$  has finite negative moments of all orders.

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