

# LOWER TAIL INDEPENDENCE OF HITTING TIMES OF TWO-DIMENSIONAL DIFFUSIONS

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The coefficient of tail dependence is a quantity that measures how extreme events in one component of a bivariate distribution depend on extreme events in the other component. It is well known that the Gaussian copula has zero tail dependence, a shortcoming for its application in credit risk modeling and quantitative risk management in general. We show that this property is shared by the joint distributions of hitting times of bivariate (uniformly elliptic) diffusion processes.

**Keywords:** hitting times, large deviations, Schilder's theorem, small time

## 1. INTRODUCTION

Let  $Y$  and  $Z$  be continuous random variables, with distribution functions  $F_Y$  and  $F_Z$ , respectively. The coefficients of lower tail dependence,  $\lambda_L$  and upper tail dependence  $\lambda_U$ , of  $Y$  and  $Z$  are defined to be:

$$\lambda_L = \lim_{\alpha \downarrow 0} \mathbb{P}(Y \leq F_Y^-(\alpha) | Z \leq F_Z^-(\alpha)),$$

and

$$\lambda_U = \lim_{\alpha \uparrow 1} \mathbb{P}(Y \geq F_Y^-(\alpha) | Z \geq F_Z^-(\alpha))$$

where  $F^-$  denotes the generalized inverse of  $F$ :  $F^-(y) = \inf\{x \in \mathbb{R} : F(x) \geq y\}$ .

The coefficients of tail dependence can also be expressed in terms of the copula  $C$  of  $Y$  and  $Z$  as (McNeil, Frey, and Embrechts [24], p. 209):

$$\lambda_L = \lim_{\alpha \downarrow 0} \frac{C(\alpha, \alpha)}{\alpha}, \quad \lambda_U = \lim_{\alpha \uparrow 1} \frac{1 - 2\alpha + C(\alpha, \alpha)}{\alpha}$$

It is well known that when  $Y$  and  $Z$  have a bivariate Gaussian distribution with correlation  $\rho$ ,  $|\rho| < 1$ , or more generally, when  $Y$  and  $Z$  have a Gaussian copula with this correlation,  $\lambda_L = \lambda_U = 0$  (McNeil *et al.* [24], p. 211).

In this paper, we show that the property of zero lower tail dependence of the Gaussian copula is shared by the hitting times of the components of two-dimensional diffusion processes. In particular, we show that if  $X$  is a two-dimensional diffusion process with generator  $L$ , acting on smooth functions  $f$  as

$$Lf = \sum_i b_i(x) \frac{\partial f}{\partial x_i} + \frac{1}{2} \sum_{i,j} a_{ij}(x) \frac{\partial^2 f}{\partial x_i \partial x_j}$$

where  $L$  is uniformly elliptic with smooth and bounded coefficients, and if the hitting times  $\tau_i$  are defined as  $\tau_i = \inf\{t > 0 | X^i(t) = c_i\}$ , then the coefficient of lower tail dependence between  $\tau_1$  and  $\tau_2$  is  $\lambda_L = 0$ . For upper tail independence, the fact that  $\lambda_U = 0$  was proved for the case when  $b$  and  $a$  are constant in Metzler [25].

The main technical result employed in our proof of the lower tail independence ( $\lambda_L = 0$ ) for the hitting times of a non-degenerate two-dimensional diffusion is the large deviations principle for small time for such processes due to Varadhan [28]. More refined results are available; see, for example, Azencott [1]. There is a large literature on the application of large deviations to the problem of the exit of a diffusion from a domain; see, for example, Freidlin and Wentzell [14], Fleming and James [13], and references therein. Exact asymptotics of hitting times of Gaussian processes are presented in Debicki *et al.* [8].

We have not aimed for maximum generality with respect to the coefficients in presenting our results. Large deviation principles generalizing the result in Varadhan [28] are known (see, e.g. Baldi and Chaleyat-Maurel [2], or the aforementioned work of Azencott [1]), and under further conditions, these results can be extended (with the same rate function) to inhomogeneous diffusion processes (e.g. Herrmann, Imkeller, and Peithmann [18]). Smoothness of the coefficients is also used to obtain smoothness of minimizing geodesics based on classical results in the calculus of variations (used in Theorem 4.11), and also leads to increased regularity of the distributions of hitting times (see Remark 4.5). It is straightforward to extend the results of this paper to show that the hitting times of any two components of a uniformly elliptic  $n$ -dimensional diffusion process with smooth bounded coefficients have zero tail dependence.

There is a large literature on the extremes of multivariate distributions. Resnick [26] provides a motivated introduction to multivariate extreme value analysis, with a number of applications. Balkema and Embrechts [3] provide mathematical and statistical techniques for finding the asymptotic behavior of multivariate extremes. Multivariate notions of tail dependence have been studied by some authors (e.g. Li [23]). Furman *et al.* [17] study generalized notions of tail dependence. The tail order of Ledford and Tawn [21] (see also Hua and Joe [19]) provides a more refined measure of dependence in the tail. Using the large deviation results employed in this paper (or perhaps the extensions of Azencott [1]), it may be possible to evaluate these tail dependence measures for diffusion hitting times. Consideration of extensions of the results in this paper to such notions is left for future research.

It is possible to construct models with positive lower tail dependence, using auxiliary variables (e.g. latent variables or subordinators) that affect both components of the diffusion simultaneously. As a simple example, let  $W$  be a standard two-dimensional Brownian motion, and  $N_t$  a standard Poisson process. Then a straightforward calculation shows that the hitting times of  $X_t = W(N_t)$  exhibit positive tail dependence. The situation is somewhat analogous to the positive coefficient of tail dependence for multivariate Student- $t$  random vectors, which are variance mixtures of multivariate Gaussians.

The analysis of the asymptotic behavior of joint hitting times is of interest in a number of applied fields. Bivariate hitting time densities arise in applications in neuroscience, where they may be used to model the joint distributions of firing times of action potentials of coupled neurons (Iyengar [20]). Sacerdote, Tamborrino, and Zucca [27] provide a numerical method for the computation of the joint density of  $\tau_1$  and  $\tau_2$  based on the solution of a system of integral equations, and prove convergence of their method. Hitting times of multidimensional diffusion processes often arise in credit risk modelling in mathematical finance. See, for example, Bielecki and Rutkowski [5]. In this context, our results may be interpreted as showing that the tail independence property of the Gaussian copula, which has received so much criticism in the context of the application of the model of Li [22] to the pricing of collateralized debt obligations, is shared by multivariate versions of the credit risk model of Black and Cox [6].

The remainder of this paper is structured as follows. The second section sets notation and reviews some results from large deviations theory needed in subsequent sections. The third section considers the case of a Brownian motion, that is, when  $b \equiv 0$  and  $a$  is a constant matrix, and proves lower tail independence of the hitting times using Schilder’s Theorem on large deviations of the Brownian sample paths. This special case is included here as its proof is particularly simple, and contains all the main ideas behind the proof of the general case. The fourth section shows lower tail independence of  $\tau_1$  and  $\tau_2$  in the general case, using results on the small-time behavior of diffusion processes due to Varadhan [28].

## 2. NOTATION AND BACKGROUND

Denote by  $C([0, 1], \mathbb{R}^2)$  the set of all continuous functions  $\omega$  from  $[0, 1]$  to  $\mathbb{R}^2$ , with  $\|\omega\| = \sup_{t \in [0,1]} |\omega(t)|$ , and  $C_x([0, 1], \mathbb{R}^2)$  the subset with  $\omega(0) = x$ . Let  $b : \mathbb{R}^2 \rightarrow \mathbb{R}$  and  $\sigma : \mathbb{R}^{2 \times 2} \rightarrow \mathbb{R}$  be bounded,  $C^2$  functions with bounded derivatives, and suppose that there exist  $\kappa, K > 0$  such that  $\kappa \|\xi\|^2 \leq \xi' a(x) \xi \leq K \|\xi\|^2$  for all  $x, \xi \in \mathbb{R}^2$ , where  $a = \sigma \sigma'$ . Let  $L_\varepsilon$  be the operator:

$$L_\varepsilon f = \varepsilon \sum_i b_i(x) \frac{\partial f}{\partial x_i} + \frac{\varepsilon}{2} \sum_{i,j} a_{ij}(x) \frac{\partial^2 f}{\partial x_i \partial x_j} \tag{1}$$

acting on smooth functions  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ , and let  $L = L_1$ . Let  $X^\varepsilon$  (with  $X = X^1$ ) be the Markov process associated with  $L_\varepsilon$ , and let  $\mathbb{P}_x^\varepsilon$  be the measure on  $C([0, 1], \mathbb{R}^2)$  giving the distribution of the process  $X^\varepsilon$  started at the point  $x$  at time 0. This may be realized as the distribution of the unique strong solution to the stochastic differential equation:

$$dX^\varepsilon(t) = \varepsilon b(X^\varepsilon(t))dt + \sqrt{\varepsilon} \cdot \sigma(X^\varepsilon(t)) dW_t, \quad X^\varepsilon(0) = x_0, \tag{2}$$

where  $W_t$  is a two-dimensional standard Brownian motion defined on a filtered probability space  $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \in [0,1]}, \mathbb{P})$  satisfying the usual conditions. We assume, without loss of generality, that  $x_0 = 0$ . For  $i = 1, 2$ , let  $c_i > 0$ ,  $\tau_i = \inf\{t > 0 | X_i(t) \geq c_i\}$ , and let  $F_i$  be the

distribution of  $\tau_i$ , and note the time-scaling property:

$$\begin{aligned} \mathbb{P}(\tau_i \leq \varepsilon) &= \mathbb{P}\left(\sup_{t \in [0, \varepsilon]} X^i(t) \geq c_i\right) \\ &= \mathbb{P}\left(\sup_{t \in [0, 1]} X^i(\varepsilon t) \geq c_i\right) = \mathbb{P}\left(\sup_{t \in [0, 1]} X^{\varepsilon, i}(t) \geq c_i\right). \end{aligned} \tag{3}$$

Let  $H^1$  be the subset of  $C([0, 1], \mathbb{R}^2)$  consisting of absolutely continuous functions  $\omega$  with square-integrable derivative, and  $H_x^1$  the subset of  $H^1$  with  $\omega(0) = x$ . Similarly, for any  $A \subseteq C([0, 1], \mathbb{R}^2)$ , let  $A_x = A \cap C_x([0, 1], \mathbb{R}^2)$ . Let  $g = a^{-1}$ , and define  $I : C([0, 1], \mathbb{R}^2) \rightarrow [0, \infty]$  by

$$I(\omega) = \begin{cases} \frac{1}{2} \int_0^1 \dot{\omega}(t)' g(\omega(t)) \dot{\omega}(t) dt & \omega \in H^1 \\ \infty & \omega \notin H^1 \end{cases} \tag{4}$$

and recall that  $I$  is lower-semi-continuous, with compact lower level sets (see, e.g. Friedman [16], pp. 326–332).

Recall the distance  $d(x, y)$  on  $\mathbb{R}^2$ , defined through the length function  $l : H^1 \rightarrow \mathbb{R}_+$ :

$$l(\omega) = \int_0^1 \sqrt{\dot{\omega}(t)' g(\omega(t)) \dot{\omega}(t)} dt \tag{5}$$

by:

$$d(x, y) = \inf\{l(\omega) \mid \omega(0) = x, \omega(1) = y\}. \tag{6}$$

If  $\omega \in H^1$  and  $\omega_{\alpha, \beta}(s) : [\alpha, \beta] \rightarrow \mathbb{R}^2$  is defined by  $\omega_{\alpha, \beta}(s) = \omega((\beta - \alpha)^{-1}(s - \alpha))$  then

$$\frac{1}{2} \int_{\alpha}^{\beta} \dot{\omega}_{\alpha, \beta}(s)' g(\omega_{\alpha, \beta}(s)) \dot{\omega}_{\alpha, \beta}(s) ds = (\beta - \alpha)^{-1} I(\omega). \tag{7}$$

This leads to the following two facts stated in Varadhan [28]:

LEMMA 2.1: *Let  $0 \leq \alpha \leq \beta$ ,*

$$\inf\{I(\omega) \mid \omega(\alpha) = x, \omega(\beta) = y\} = \frac{d^2(x, y)}{2(\beta - \alpha)}. \tag{8}$$

LEMMA 2.2: *For  $0 \leq t_1 < t_2 < t_3 \cdots < t_n \leq 1$ ,*

$$\inf\{I(\omega) \mid \omega(t_j) = x_j, j = 1, \dots, n\} = \frac{1}{2} \sum_{j=1}^{n-1} \frac{d^2(x_{j+1}, x_j)}{t_{j+1} - t_j}. \tag{9}$$

The following large deviation principle is the main result required to derive the tail independence of the hitting times of  $X$ , and is due to Varadhan [28], which generalized the result for the special case of Brownian motion, often referred to as Schilder’s Theorem:

THEOREM 2.3: *For  $G \subseteq C([0, 1], \mathbb{R}^2)$  open and  $F \subseteq C([0, 1], \mathbb{R}^2)$  closed:*

$$\liminf_{\varepsilon \downarrow 0, y \rightarrow x} \varepsilon \log \mathbb{P}_y^\varepsilon(G) \geq - \inf_{\omega \in G_x} I(\omega), \tag{10}$$

$$\limsup_{\varepsilon \downarrow 0, y \rightarrow x} \varepsilon \log \mathbb{P}_y^\varepsilon(F) \leq - \inf_{\omega \in F_x} I(\omega). \tag{11}$$

For  $k_1, k_2 \in [0, \infty)$ ,  $t_1, t_2 \in [0, 1]$ , define:

$$B(k_1, t_1, k_2, t_2) = \{\omega \in C_0([0, 1], \mathbb{R}^2) \mid \sup_{t \in [0, t_1]} \omega_1(t) \geq k_1, \sup_{t \in [0, t_2]} \omega_2(t) \geq k_2\}. \tag{12}$$

That is,  $B(k_1, t_1, k_2, t_2)$  is the set of paths that start at 0, cross  $k_1$  by time  $t_1$  and cross  $k_2$  by  $t_2$ . For  $k_1, k_2 \in [0, \infty)$  define:

$$B_{k_1, k_2} = \{\omega \in C_0([0, 1], \mathbb{R}^2) \mid \sup_{t \in [0, 1]} \omega_1(t) \geq k_1, \sup_{t \in [0, 1]} \omega_2(t) \geq k_2\}, \tag{13}$$

so that  $B_{k_1, k_2} = B(k_1, 1, k_2, 1)$ .

Define the constants:

$$J_1 = \inf\{I(\omega), \omega \in B_{c_1, 0}\}, \tag{14}$$

$$J_2 = \inf\{I(\omega), \omega \in B_{0, c_2}\}, \tag{15}$$

$$J_{1,2} = \inf\{I(\omega), \omega \in B_{c_1, c_2}\}. \tag{16}$$

Since  $B_{c_1, c_2} \subseteq B_{c_1, 0}$ , and  $B_{c_1, c_2} \subseteq B_{0, c_2}$ , we have  $J_{1,2} \geq J_1$  and  $J_{1,2} \geq J_2$ . Also, notice that the time-scaling property (7) implies that with  $t < 1$ ,

$$\inf\{I(\omega) \mid \omega \in B(c_1, t, c_2, t)\} = t^{-1} J_{1,2}. \tag{17}$$

### 3. BROWNIAN MOTION CASE

In this section, we will consider tail independence in the special case where  $b \equiv 0$  and  $a$  is a constant matrix with  $a_{ii} = \sigma_i^2$ , and  $a_{12} = a_{21} = \sigma_1 \sigma_2 \rho$  for  $\sigma_i > 0$ ,  $i = 1, 2$ , and  $\rho \in (-1, 1)$ . In this case,  $X_t$  may be taken as the solution of the stochastic differential equation (2) with  $b \equiv 0$  and

$$\sigma = \begin{pmatrix} \sigma_1 \sqrt{1 - \rho^2} & \rho \sigma_1 \\ 0 & \sigma_2 \end{pmatrix}$$

with  $\sigma_i > 0, i = 1, 2$  and  $\rho \in (-1, 1)$ . Thus,  $X_t^1$  is  $\sigma_1$  multiplied by a Brownian motion  $Z_t^1 = \sqrt{1 - \rho^2} W_t^1 + \rho W_t^2$  and  $X_t^2$  is  $\sigma_2$  multiplied by a Brownian motion  $Z_t^2 = W_t^2$ , and  $Z_t^1$  and  $Z_t^2$  have correlation  $\rho$ . Denote  $\tilde{c}_i = \frac{c_i}{\sigma_i}$  for  $i = 1, 2$ . Without loss of generality, we may assume that  $\tilde{c}_2 \leq \tilde{c}_1$ , so that  $\kappa = (\tilde{c}_2 / \tilde{c}_1)^2 \leq 1$ .

Here we summarize a proof that the coefficient of lower tail dependence is equal to zero in the Brownian motion case, using Schilder’s Theorem. The extension to the case of variable coefficients based on Theorem 2.3 from Varadhan [28] is presented in the next section. We may assume  $\rho \neq 0$ , since zero tail dependence follows immediately from the independence of  $\tau_1$  and  $\tau_2$  in the case of zero correlation. The argument proceeds as follows:

1. Use properties of marginal hitting time distributions to remove inverse CDFs from the ratio in the definition of the coefficient of tail dependence: Let  $\beta_t$  be a standard Brownian motion and let  $\kappa = (\tilde{c}_2 / \tilde{c}_1)^2 \leq 1$ . By Brownian scaling:

$$F_1(t) = \mathbb{P}\left(\sup_{s \leq t} \beta_s \geq \tilde{c}_1\right) = \mathbb{P}\left(\sup_{s \leq \kappa t} \beta_s \geq \tilde{c}_2\right) = F_2(\kappa t), \tag{18}$$

which can also be seen from the explicit formula for the hitting time distribution of Brownian motion. Recall the definition of  $\lambda_L$ , and make the substitution  $F_1(\varepsilon) = \alpha$

$$\lambda_L = \lim_{\alpha \downarrow 0} \frac{\mathbb{P}(\tau_1 \leq F_1^{-1}(\alpha), \tau_2 \leq F_2^{-1}(\alpha))}{\mathbb{P}(\tau_1 \leq F_1^{-1}(\alpha))} = \lim_{\varepsilon \downarrow 0} \frac{\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \kappa \varepsilon)}{\mathbb{P}(\tau_1 \leq \varepsilon)}. \tag{19}$$

2. *Apply time scaling and large deviation results to approximate probabilities:* Using Brownian scaling

$$\begin{aligned} \mathbb{P}(\tau_1 \leq \varepsilon) &= \mathbb{P}\left(\sup_{s \leq \varepsilon} \beta_s \geq \tilde{c}_1\right) = \mathbb{P}\left(\sup_{s \leq 1} \varepsilon^{-1/2} \beta_{\varepsilon s} \geq \varepsilon^{-1/2} \tilde{c}_1\right) \\ &= \mathbb{P}(\sup_{s \leq 1} \sqrt{\varepsilon} \beta_s \geq \tilde{c}_1) = \mathbb{P}_0^\varepsilon(B_{\tilde{c}_1, 0}) \end{aligned} \tag{20}$$

and similarly:

$$\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \kappa\varepsilon) = \mathbb{P}_0^\varepsilon(B(\tilde{c}_1, 1, \tilde{c}_2, \kappa)). \tag{21}$$

Applying Theorem 2.3 yields:<sup>1</sup>

$$-\lim_{\varepsilon \downarrow 0} \varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon)) = \inf_{\omega \in B_{\tilde{c}_1, 0}} I(\omega) = \tilde{J}_1 \tag{22}$$

$$-\lim_{\varepsilon \downarrow 0} \varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \kappa\varepsilon)) = \inf_{\omega \in B(\tilde{c}_1, 1, \tilde{c}_2, \kappa)} I(\omega) = \tilde{J}_{1,2}. \tag{23}$$

Heuristically,  $\mathbb{P}(\tau_1 \leq \varepsilon) \approx \exp(-\varepsilon^{-1} \tilde{J}_1)$  and  $\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \kappa\varepsilon) \approx \exp(-\varepsilon^{-1} \tilde{J}_{1,2})$ .

3. *Show that  $\tilde{J}_{1,2} > \tilde{J}_1$ , and conclude that  $\lambda_L = 0$ .*

Let  $\omega \in H_0^1$  be any path such that  $\sup_{t \in [0,1]} \omega_1(t) \geq \tilde{c}_1$ , that is,  $\omega \in B_{\tilde{c}_1, 0}$ , with  $\omega^1(t') = \tilde{c}_1$ . Then, by the Cauchy–Schwartz inequality:

$$I(\omega) \geq \frac{1}{2(1 - \rho^2)} \int_0^{t'} (\rho \cdot \dot{\omega}_1(t) - \dot{\omega}_2(t))^2 dt + \frac{1}{2} \int_0^{t'} \dot{\omega}_1(t)^2 dt \geq \frac{\tilde{c}_1^2}{2t'}$$

with equality if and only if  $\omega = \omega^*$  with  $\omega_1^*(t) = \tilde{c}_1 t$ ,  $\omega_2^*(t) \equiv \rho \tilde{c}_1 t < \tilde{c}_2$  for  $t \leq \kappa$ . So  $\tilde{J}_1 = \frac{\tilde{c}_1^2}{2}$ . Since  $I$  is lower-semi-continuous with compact level sets, and  $F = B(\tilde{c}_1, 1, \tilde{c}_2, \kappa)$  is closed,  $I$  attains its minimum on  $F$  at some  $\tilde{\omega}$ . The above argument shows that  $\tilde{J}_{1,2} = I(\tilde{\omega}) > \tilde{J}_1$  (since  $\omega^* \notin F$ , so  $\tilde{\omega} \neq \omega^*$ ). Thus:

$$0 > \tilde{J}_1 - \tilde{J}_{1,2} = \lim_{\varepsilon \downarrow 0} \varepsilon \log \left( \frac{\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \kappa\varepsilon)}{\mathbb{P}(\tau_1 \leq \varepsilon)} \right) \tag{24}$$

implying that  $\lambda_L = 0$ .

#### 4. GENERAL CASE

In this section, we present a proof of the tail independence of the hitting times of diffusions in the general two-dimensional case. Throughout, we assume without loss of generality that  $J_1 \geq J_2$  where  $J_1$  and  $J_2$  are defined in (14) and (15), respectively.

##### 4.1. Properties of the Hitting Time Distributions

In this section, we derive properties of the distributions of the hitting times  $\tau_1$  and  $\tau_2$  that are used later in the paper. While many of the results are likely well-known, we include proofs when we are unaware of a precise reference.

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<sup>1</sup> Here we use the fact that  $B_{\tilde{c}_1, 0}$  and  $B(\tilde{c}_1, 1, \tilde{c}_2, \kappa)$  are both *continuity sets*, meaning that the upper and lower bounds given by Theorem 2.3 are equal. This is shown below in Proposition 4.6.

First of all, we have that  $F_i(0) = 0$  since  $X$  has continuous sample paths and  $c_i > x_0^i = 0$ . Also, as a distribution function,  $F_i$  is right-continuous with left-hand limits, and is increasing (not necessarily strictly increasing). We would like to show that  $F_i$  are continuous and strictly increasing, as this justifies the use of expressions such as  $F_1^{-1}(\alpha)$  and  $F_2^{-1}(F_1(\varepsilon))$  (as well as the continuity of these inverses), below. The fact that the distribution  $F_i$  is strictly increasing can be derived as a consequence of the following result from Bass [4], for  $d = 2$ , and referred to there as the *support theorem* for  $X_t$ .

**THEOREM 4.1:** *Suppose  $\sigma$  and  $b$  are bounded,  $\sigma^{-1}$  is bounded,  $x \in \mathbb{R}^d$ , and  $X_t$  satisfies (2) with  $X_0 = x$ ,  $\varepsilon = 1$ . Suppose  $\psi : [0, t] \rightarrow \mathbb{R}^d$  is continuous, with  $\psi(0) = x$  and  $\delta > 0$ . There exists  $k > 0$ , depending only on  $\delta, t$ , the modulus of continuity of  $\psi$ , and the bounds on  $b$  and  $\sigma$  such that*

$$\mathbb{P} \left( \sup_{s \leq t} |X_s - \psi(s)| < \delta \right) \geq k. \tag{25}$$

**PROOF:** See Bass [4] (pp. 25–27). ■

Quoting Bass [4] (p. 26) (substituting our notation) “This can be phrased as saying the graph of  $X_s$  stays inside an  $\delta$ -tube about  $\psi$ . By this we mean, if  $G_\psi^\delta = \{(s, y) : |y - \psi(s)| < \delta, s \leq t\}$ , then  $\{(s, X_s) : s \leq t\}$  is contained in  $G_\psi^\delta$  with positive probability.”

**PROPOSITION 4.2:**  *$F_i$  is strictly increasing,  $i = 1, 2$ .*

**PROOF:** By reordering the indices, the result only needs to be proved for  $i = 1$ . Let  $\delta > 0$ , and  $r > 0$ . Applying the above theorem with  $\psi_1(s) = (c_1 + 2\delta)s/r$  gives that  $F_1(r) > 0 = F(0)$ . Now let  $t > r$ , and consider a smooth  $\psi$  such that  $\psi_1(s) \leq c_1 - 2\delta$  for  $s \leq r$  and  $\psi_1(t) > c_1 + 2\delta$ . Then

$$F_1(t) - F_1(r) = \mathbb{P}(r < \tau_1 \leq t) \geq \mathbb{P} \left( \sup_{s \leq t} |X_s - \psi(s)| < \delta \right) > 0. \tag{26}$$

■

Next, we consider whether  $F_i(t)$  are continuous, that is, whether  $F_i(t-) = F_i(t)$ . To prove this, we use two results. The first is another result from Bass [4]. Throughout,  $K_j$  will be strictly positive constants.

**THEOREM 4.3:** *Suppose  $X_t$  solves (2) with  $\varepsilon = 1$ ,  $\sigma$  and  $b$  bounded. There exist  $K_1$  and  $K_2$  depending only on  $|\sigma|$  such that:*

$$\mathbb{P} \left( \sup_{s \leq t} |X_s - X_0| > \lambda + \|b\|_\infty t \right) \leq K_1 \exp(-K_2 \lambda^2 / t). \tag{27}$$

**PROOF:** See Bass [4] (p. 23). ■

Let  $\Gamma(t, x; y)$  denote the density of  $X_t$  given that  $X_0 = y$ , which is the fundamental solution of the operator  $L - \partial_t$ . Bounds on the fundamental solution, given by Friedman [15] (p. 24), imply that

$$0 \leq \Gamma(t, x; y) \leq \frac{K_3}{t} \exp \left( -K_4 \frac{\|x - y\|^2}{t} \right) \tag{28}$$

(where we have specialized the results in the reference to the case  $d = 2$ ).

PROPOSITION 4.4: For all  $t > 0$ ,  $F_i(t-) = F_i(t)$ .

PROOF: Let  $t > 0$ , and suppose to the contrary that  $F_1(t) - F_1(t-) = \xi > 0$ . Let  $\delta_n = n^{-1}\|b\|_\infty$ , and  $t_n = t - \frac{1}{n}$ , and note that:

$$F_1(t) - F_1(t_n) = \mathbb{P}(t_n < \tau \leq t) > \xi. \tag{29}$$

But

$$\begin{aligned} \mathbb{P}(t_n < \tau \leq t) &\leq \mathbb{P}(X_1(t_n) \in [c_1 - \delta_n, c_1]) \\ &\quad + \mathbb{P}\left(X_1(t_n) < c_1 - \delta_n, \sup_{s \in (t_n, t]} X_1(s) \geq c_1\right) = P_1(n) + P_2(n). \end{aligned} \tag{30}$$

We will control each term separately.

$$P_1(n) = \int_{c_1 - \delta_n}^{c_1} \int_{-\infty}^{\infty} \Gamma(t_n, 0, y) dy_2 dy_1 \tag{31}$$

$$\leq \frac{K_3}{t_n} \int_{c_1 - \delta_n}^{c_1} \int_{-\infty}^{\infty} \exp\left(-K_4 \frac{(y_1^2 + y_2^2)}{t_n}\right) dy_2 dy_1 \tag{32}$$

$$\leq \frac{K_5 \delta_n}{t_n} \leq K_6 \delta_n. \tag{33}$$

Using the Markov property, and the bound (27)

$$\begin{aligned} P_2(n) &\leq \int_{-\infty}^{c_1 - \delta_n} \int_{-\infty}^{\infty} \Gamma(t_n, 0, y) \mathbb{P}\left(\sup_{0 < s \leq n^{-1}} |X_s - y| \geq c_1 - y_1\right) dy_2 dy_1 \tag{34} \\ &= \int_{-\infty}^{c_1 - \delta_n} \int_{-\infty}^{\infty} \Gamma(t_n, 0, y) \mathbb{P}\left(\sup_{0 < s \leq n^{-1}} |X_s - y| \geq c_1 - y_1 - \delta_n + n^{-1}\|b\|_\infty\right) dy_2 dy_1 \\ &\leq K_1 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mathbf{1}_{\{y_1 < c_1 - \delta_n\}} \Gamma(t_n, 0, y) \exp(-K_2 n (c_1 - y_1 - n^{-1}\|b\|_\infty)^2) dy_2 dy_1 \\ &\leq K_1 K_3 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mathbf{1}_{\{y_1 < c_1 - \delta_n\}} \exp\left(-K_4 \frac{\|y\|^2}{t_n}\right) \\ &\quad \times \exp(-K_2 n (c_1 - y_1 - n^{-1}\|b\|_\infty)^2) dy_2 dy_1. \end{aligned}$$

$P_1(n)$  clearly tends to zero, and  $P_2(n)$  tends to zero by the Dominated Convergence Theorem. But then  $F_1(t) - F_1(t_n)$  tends to zero, contradicting (29). ■

*Remark 4.5:* According to Elliott, Siu, and Yang [10],  $u(x, t) = \mathbb{P}(\tau_1 > t | X_0 = x)$  solves the partial differential equation  $\partial_t u = Lu$ . where It should be noted that the boundary data is not continuous (the initial condition is  $u(x, 0) = \mathbf{1}_{\{x_1 < c_1\}}$ , but the boundary condition is  $u(c_1, x_2, t) = 0$ ). Nonetheless, a suitable notion of weak solution exists that is smooth on the interior of the domain. Let  $v(x, t) = 1 - u(x, t) = \mathbb{P}(\tau_1 \leq t | X_0 = x)$ , so that  $F_1(t) = v(0, t)$ .  $v$  is as smooth as  $u$ ; so  $F_1$  should inherit the maximum smoothness (in time) of the solution to the partial differential equation (PDE). One can also argue that the density of the hitting time is strictly positive for  $t > 0$  as follows. Let  $f_1(t) = F_1'(t) = \partial_t v(0, t) > 0$ . Let  $w(x, t) = \partial_t v(x, t)$ , so that  $f_1(t) = w(0, t)$ , and note that, by differentiating the PDE:



$\partial_t u = Lu \Rightarrow \partial_t v = Lv \Rightarrow \partial_t w = Lw$ . Since  $v$  is increasing in  $t$ ,  $w \geq 0$ . Now, suppose that  $f_i(t_0) = 0$  for some  $0 < t_0 < T$ . Then  $w(0, t) = 0$ . But  $w \geq 0$ , so this means that  $w$  attains a minimum at  $(0, t_0)$ , and by the Strong Maximum Principle for parabolic PDEs (Evans [11], pp. 396–397),  $w$  is constant on  $U_{t_0} = U \times (0, t_0]$ . In particular, this implies that  $f_1(t) = w(0, t) = w(0, t_0) = 0$  for all  $0 < t < t_0$ , contradicting the fact that  $F_1$  is strictly increasing. (In order to meet the hypotheses in the reference, we can consider  $w$  over  $U \times (0, T]$  where  $U = (-\varepsilon, \varepsilon) \times (-\varepsilon, \varepsilon)$  to meet the requirements that  $U$  be bounded, and  $w$  smooth on  $U_T$  and continuous on  $\bar{U}_T$ .)

It is clear that in general we cannot expect a result as precise as  $F_1(\varepsilon) = F_2(\kappa\varepsilon)$ , as held in the Brownian motion case. Nonetheless, when  $J_1 > J_2$  we can show that for  $\kappa > \frac{J_2}{J_1}$  and  $\varepsilon$  small enough,  $F_1(\varepsilon) \leq F_2(\kappa\varepsilon)$ , and when  $J_1 = J_2$ , the logarithmic asymptotics of  $\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq F_2^{-1}(F_1(\varepsilon)))$  and  $\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \varepsilon)$  are the same. These results are given in Proposition (4.7), which requires the following consequences of Theorem (2.3).

PROPOSITION 4.6: For  $J_i$  defined as in (14) and (15):

$$\lim_{\varepsilon \downarrow 0} -\varepsilon \log(F_i(\varepsilon)) = J_i. \tag{35}$$

Furthermore, for any  $\gamma \leq 1$ :

$$\lim_{\varepsilon \downarrow 0} -\varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \gamma\varepsilon)) = \inf\{I(\omega) | \omega \in B(c_1, 1, c_2, \gamma)\} \tag{36}$$

$$\lim_{\varepsilon \downarrow 0} -\varepsilon \log(\mathbb{P}(\tau_1 \leq \gamma\varepsilon, \tau_2 \leq \varepsilon)) = \inf\{I(\omega) | \omega \in B(c_1, \gamma, c_2, 1)\} \tag{37}$$

PROOF: We will consider the result for  $F_1$ . The proof of the other results is similar. By (3):

$$F_1(\varepsilon) = \mathbb{P}_x^1(B(c_1, \varepsilon, 0, \varepsilon)) = \mathbb{P}_x^c(B_{c_1, 0}).$$

Noting that  $\mathring{B}_{c_1, 0}$  consists of all  $\omega \in C([0, 1], \mathbb{R}^2)$  such that the supremum of the first component is strictly greater than  $c_1$  we have by Theorem 2.3:

$$\begin{aligned} -\inf\{I(\omega) | \omega \in \mathring{B}_{c_1, 0}\} &\leq \liminf_{\varepsilon \downarrow 0} \varepsilon \log(F_1(\varepsilon)) \\ &\leq \limsup_{\varepsilon \downarrow 0} \varepsilon \log F_1(\varepsilon) \leq -\inf\{I(\omega) | \omega \in B_{c_1, 0}\} = -J_1. \end{aligned} \tag{38}$$

Now let  $\omega^* \in B_{c_1, 0}$  be such that  $I(\omega^*) = J_1$  (existence of  $\omega^*$  follows from the fact that  $B_{c_1, 0}$  is closed, and  $I$  has compact lower level sets). Take  $\omega^n(s) = \omega^*(s) + (s/n)$  for  $n \geq 1$ . Then  $\omega^n \in \mathring{B}_{c_1, 0}$  (the interior of  $B_{c_1, 0}$ ),  $\omega^n \rightarrow \omega^*$ , and  $I(\omega_n) \rightarrow I(\omega^*)$  by Dominated Convergence, so the two bounds in (38) coincide and the result follows. ■

Note that a special case of the last limit in the above proposition is that:

$$\lim_{\varepsilon \downarrow 0} -\varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \varepsilon)) = \inf\{I(\omega) | \omega \in B_{c_1, c_2}\} = J_{1,2}. \tag{39}$$

PROPOSITION 4.7:

- (i) Suppose that  $J_1 > J_2$ . Then for any  $\gamma > J_2/J_1$  there exists  $t_0(\gamma) > 0$  such that for all  $t \leq t_0$ ,  $F_1(t) \leq F_2(\gamma t)$ . In particular, for  $t$  small enough,  $F_2^{-1}(F_1(t)) \leq \gamma t$ .
- (ii) Suppose that  $J_1 = J_2$ . Then

$$-\lim_{\varepsilon \downarrow 0} \varepsilon \log \mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq F_2^{-1}(F_1(\varepsilon))) = J_{1,2}. \tag{40}$$

PROOF: (i) We may assume  $\gamma < 1$ . Let  $c < 1$  be such that  $c^2 \geq J_2/(J_1\gamma)$ . By Proposition 4.6, there is a  $t_0$  small enough so that for all  $t \leq t_0$ ,

$$F_1(t) \leq \exp\left(-\frac{J_1 c}{t}\right) \leq \exp\left(-\frac{J_2}{c(\gamma t)}\right) \leq F_2(\gamma t). \tag{41}$$

(ii) Let  $J = J_1 = J_2$  and  $\eta > 0$ . By Proposition 4.6 for  $r < r(\eta)$ , and  $i = 1, 2$ , we have that

$$\exp\left(-\frac{J\sqrt{1+\eta}}{r}\right) \leq F_i(r) \leq \exp\left(-\frac{J}{r\sqrt{1+\eta}}\right) \tag{42}$$

so for  $\varepsilon$  small enough:

$$F_1(\varepsilon) \leq \exp\left(-\frac{J}{\varepsilon\sqrt{1+\eta}}\right) = \exp\left(-\frac{J\sqrt{1+\eta}}{(1+\eta)\varepsilon}\right) \leq F_2((1+\eta)\varepsilon) \tag{43}$$

and we have that  $F_2^{-1}(F_1(\varepsilon)) \leq (1+\eta)\varepsilon$ . A similar argument shows that for  $\varepsilon$  small enough,  $F_2^{-1}(F_1(\varepsilon)) \geq (1+\eta)^{-1}\varepsilon$ . Using (17) then implies

$$\begin{aligned} (1+\eta)J_{1,2} &= \inf\{I(\omega) \mid \omega \in B(c_1, (1+\eta)^{-1}, c_2, (1+\eta)^{-1})\} \\ &\geq \inf\{I(\omega) \mid \omega \in B(c_1, 1, c_2, (1+\eta)^{-1})\} \\ &= \lim_{\varepsilon \downarrow 0} [-\varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq (1+\eta)^{-1}\varepsilon))] \\ &\geq \limsup_{t \downarrow 0} [-\varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq F_2^{-1}(F_1(\varepsilon))))] \\ &\geq \liminf_{\varepsilon \downarrow 0} [-\varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq F_2^{-1}(F_1(\varepsilon))))] \\ &\geq \lim_{\varepsilon \downarrow 0} [-\varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq (1+\eta)\varepsilon))] \\ &= (1+\eta)^{-1} \inf\{I(\omega) \mid \omega \in B(c_1, (1+\eta)^{-1}, c_2, 1)\} \\ &\geq (1+\eta)^{-1} J_{1,2}. \end{aligned} \tag{44}$$

The result now follows by letting  $\eta \downarrow 0$ . ■

### 4.2. Properties of the Variational Problems

LEMMA 4.8: *Suppose that  $\omega \in B_{c_1,0}$  is such that  $I(\omega) = J_1$ . Then  $\omega_1(1) = c_1$ , and  $\omega_1(t) < c_1$  for  $t < 1$ . Similarly, if  $\omega \in B_{0,c_2}$  is such that  $I(\omega) = J_2$ , then  $\omega_2(1) = c_2$  and  $\omega_2(t) < c_2$  for  $t < 1$ .*

PROOF: Suppose to the contrary that  $\omega \in B_{c_1,0}$ ,  $I(\omega) = J_1$ ,  $\omega_1(\tilde{t}) = c_1$  for some  $\tilde{t} < 1$ . Using Lemma 2.1:

$$J_1 \geq \int_0^{\tilde{t}} \dot{\omega}(s)' g(\omega(s)) \dot{\omega}(s) ds \geq \frac{d^2(0, \omega(\tilde{t}))}{2\tilde{t}} > \frac{d^2(0, \omega(\tilde{t}))}{2} \geq J_1 \tag{45}$$

since a path that reaches the point  $\omega(\tilde{t})$  at time 1 is in  $B_{c_1,0}$ . This is a contradiction, so we must have  $\omega_1(t) < c_1$  for all  $t < 1$ . The proof when  $I(\omega) = J_2$  is similar. ■

LEMMA 4.9: *If  $J_1 > J_2$  then there exists  $\gamma \in (\frac{J_2}{J_1}, 1)$  such that*

$$\inf\{I(\omega) \mid \omega \in B(c_1, 1, c_2, \gamma)\} > J_1. \tag{46}$$

PROOF: Since  $B(c_1, 1, c_2, \gamma) \subseteq B_{c_1, 0}$ , weak inequality in (46) is immediate for any  $\gamma \geq 0$ . Suppose to the contrary that equality holds for all  $\gamma \in (J_2/J_1, 1)$ . Let  $\gamma_n \downarrow J_2/J_1$ , and let  $\omega^n \in B(c_1, 1, c_2, \gamma_n)$  and  $t_n \in (0, \gamma_n]$  be such that  $I(\omega^n) = J_1$ , and  $\omega_2^n(t_n) = c_2$ . Passing to a subsequence if necessary, and using the fact that  $I$  is lower semi-continuous with compact level sets, we obtain  $\omega^n \rightarrow \omega^*$ ,  $t_n \rightarrow t^* \leq J_2/J_1$ , and  $\omega_2^*(t^*) = c_2$ , with  $I(\omega^*) = J_1$ . Now:

$$\begin{aligned} J_1 = I(\omega^*) &= \frac{1}{2} \int_0^{J_2/J_1} \dot{\omega}^*(t)' g(\omega^*(t)) \dot{\omega}^*(t) dt + \frac{1}{2} \int_{J_2/J_1}^1 \dot{\omega}^*(t)' g(\omega^*(t)) \dot{\omega}^*(t) dt \\ &\geq J_1 + \frac{1}{2} \int_{J_2/J_1}^1 \dot{\omega}^*(t)' g(\omega^*(t)) \dot{\omega}^*(t) dt \\ &\geq J_1 + \frac{d^2(\omega^*(J_2/J_1), \omega^*(1))}{2(1 - J_2/J_1)}, \end{aligned}$$

where the first line follows by applying (7), and the second line from Lemma 2.1. Using Lemma 4.8 and the fact that  $I(\omega^*) = I(\omega^n) = J_1$ ,  $\omega_1^*(1) = \lim_{n \rightarrow \infty} \omega_1^n(1) = c_1$ , and we must have that  $\omega_1^*(1) < c_1$  for  $t < 1$ , so  $\omega^* \in B(c_1, 1, c_2, J_2/J_1)$ , and  $d^2(\omega^*(J_2/J_1), \omega^*(1)) > 0$ , yielding a contradiction. ■

### 4.3. Tail Independence

In this section, we prove that  $\tau_1$  and  $\tau_2$  have zero lower tail dependence. In order to do so, we need to recall some facts about the (scaled) squared-distance function:

$$u(z) = \frac{1}{2} d^2(0, z) : \mathbb{R}^2 \rightarrow \mathbb{R}_+. \tag{47}$$

We begin with some standard definitions. Here  $G \subseteq \mathbb{R}^2$  is an open set.

DEFINITION 4.10: *Let  $u \in C(G)$ ,  $x \in G$ . The sets  $D^+u(x)$  and  $D^-u(x)$  are defined to be:*

$$D^+u(x) = \left\{ p \in \mathbb{R}^2 : \limsup_{y \rightarrow x, y \in \Omega} \frac{u(y) - u(x) - p \cdot (y - x)}{|x - y|} \leq 0 \right\}, \tag{48}$$

$$D^-u(x) = \left\{ p \in \mathbb{R}^2 : \liminf_{y \rightarrow x, y \in \Omega} \frac{u(y) - u(x) - p \cdot (y - x)}{|x - y|} \geq 0 \right\}. \tag{49}$$

$D^+u(x)$  is referred to as the (viscosity) super-differential of  $u$  at  $x$  and  $D^-(x)$  is the (viscosity) sub-differential of  $u$  at  $x$ .

As noted earlier,  $u(z)$  is the optimal value of the optimization problem:

$$u(z) = \inf_{\omega \in C_0[0,1], \omega(1)=z} I(\omega) = \inf_{\omega \in H_0^1, \omega(1)=z} \frac{1}{2} \int_0^1 \dot{\omega}(t)' g(\omega(t)) \dot{\omega}(t) dt. \tag{50}$$

Results in the calculus of variations can be used to show that minimizers  $\omega$  of (50) are Lipschitz continuous (Clarke [7], Theorem 16.18, pp. 330–332), and indeed since  $\lambda(t, x, p) = (1/2)p'g(x)p$  is smooth, any Lipschitz minimizer  $\omega^*$  satisfies the integral Euler equation

(the Theorem of du-Bois-Reymond, Clarke [7], Theorem 15.2, pp. 308–309). The positive definiteness of  $\Lambda_{pp}$  then implies higher order regularity of  $\omega^*$  given higher regularity of  $\Lambda$ , by a Theorem of Hilbert and Weierstrass (Clarke [7], Theorem 15.7, p. 313), which in turn implies that  $\omega^*$  is a smooth classical solution of the Euler equation for the problem.

Optimal solutions of (50) are geodesics connecting 0 and  $z$  in the Riemannian metric on  $\mathbb{R}^2$  defined by the distance (6). As such, they have constant, nonzero speed, that is,  $\dot{\omega}^*(t)'g(\omega^*(t))\omega^*(t) = k > 0$  (this may also be seen from the point of view of the calculus of variations as a consequence of the Erdmann condition, Clarke [7], Proposition 14.4, pp. 290–291). Using the formula for the first variation of the energy  $I$  (see do Carmo [9], pp. 192–196), it can be shown that  $u$  is super-differentiable, and  $g(z)\dot{\omega}^*(1) \in D^+(u(z))$  (for a sketch of the proof, see Figalli and Villani [12], p. 178).

**THEOREM 4.11:** *The hitting times  $\tau_1$  and  $\tau_2$  have zero lower tail dependence:*

$$\lambda_L = \lim_{\alpha \downarrow 0} \frac{\mathbb{P}(\tau_1 \leq F_1^{-1}(\alpha), \tau_2 \leq F_2^{-1}(\alpha))}{\alpha} = 0 \tag{51}$$

**PROOF:** Let  $\varepsilon = F_1^{-1}(\alpha)$ , so that  $\alpha = F_1(\varepsilon)$  and

$$\lambda_L = \lim_{\varepsilon \downarrow 0} \frac{\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq F_2^{-1}(F_1(\varepsilon)))}{F_1(\varepsilon)} = \lim_{\varepsilon \downarrow 0} L(\varepsilon). \tag{52}$$

It is enough to show that  $\limsup_{\varepsilon \downarrow 0} \Lambda(\varepsilon) < 0$  where:

$$\Lambda(\varepsilon) = \varepsilon \log(L(\varepsilon)) = \varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq F_2^{-1}(F_1(\varepsilon))) - \varepsilon \log(F_1(\varepsilon))). \tag{53}$$

(i) Suppose  $J_1 > J_2$ . Using Proposition 4.6 we need to show that:

$$\liminf_{\varepsilon \downarrow 0} -\varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq F_2^{-1}(F_1(\varepsilon)))) > J_1. \tag{54}$$

Let  $\gamma$  be as in Lemma 4.9. Then, applying Propositions 4.7 and 4.6:

$$\begin{aligned} \liminf_{\varepsilon \downarrow 0} -\varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq F_2^{-1}(F_1(\varepsilon)))) &\geq \lim_{\varepsilon \downarrow 0} -\varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \gamma\varepsilon)) \\ &= \inf\{I(\omega) \mid \omega \in B(c_1, 1, c_2, \gamma)\} \\ &> J_1. \end{aligned} \tag{55}$$

(ii) Suppose  $J_1 = J_2$ . Then

$$\begin{aligned} \Lambda(\varepsilon) &= \varepsilon (\log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq F_2^{-1}(F_1(\varepsilon)))) - \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \varepsilon))) \\ &\quad + \varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon, \tau_2 \leq \varepsilon)) - \varepsilon \log(\mathbb{P}(\tau_1 \leq \varepsilon)). \end{aligned} \tag{56}$$

Using Propositions 4.7 and 4.6, the result will follow if we can show that  $J_{12} > J_1$ . Since  $B_{c_1, c_2} \subseteq B_{c_1, 0}$ , we immediately have  $J_{12} \geq J_1$ . Suppose that  $J_{12} = J_1 = J_2$ . Let  $\omega^* \in B_{c_1, c_2}$  be such that  $I(\omega^*) = J_{12} = u(c)$ , where  $c = (c_1, c_2)'$  (under the assumptions,  $J_{12} = u(c)$  follows immediately from Lemma 4.8). By Lemma 4.8,  $\omega^*(1) = (c_1, c_2)'$  and  $\omega_1^*(t) < c_1$ ,  $\omega_2^*(t) < c_2$  for all  $t < 1$ , and in particular,  $\dot{\omega}^*(1) \geq 0$ . Let  $p \in D^+u(c)$ . Considering the sequence  $y_n = (c_1 - \frac{1}{n}, c_2)$  and using the fact that  $u(y_n) \geq J_2 = u(c)$  yields:

$$0 \geq \limsup_{y_n \rightarrow c} \frac{u(y_n) - u(c) - p \cdot (y_n - c)}{|y_n - c|} \geq \limsup_{y_n \rightarrow c} \frac{-p \cdot (y_n - c)}{|y_n - c|} = p_1. \tag{57}$$

Similarly, we have  $p_2 \leq 0$ . Since  $g(c)\dot{\omega}^*(1) \in D^+u(c)$ ,  $g(c)\dot{\omega}^*(1) \leq 0$ . But then  $\dot{\omega}^*(1)'g(c)\dot{\omega}^*(1) \leq 0$ . ■

## 5. CONCLUSION

By utilizing the large deviations results of Varadhan [28], we have shown that the hitting times  $\tau_1$  and  $\tau_2$  of the components of a two-dimensional uniformly elliptic diffusion process have coefficient of lower tail dependence equal to zero.

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