

# 1 Introduction

We consider two-dimensional correlated Brownian motion with absorbing boundaries:

$$X(t) = x_0 + \mu_x t + \sigma_x W_x(t) \quad a_x < X(t) < b_x \quad (1)$$

$$Y(t) = y_0 + \mu_y t + \sigma_y W_y(t) \quad a_y < Y(t) < b_y \quad (2)$$

where  $W_i$  are standard Brownian motions with  $\text{Cov}(W_1(t), W_2(t)) = \rho t$  for  $0 < t' \leq t$ . In particular, we find the joint transition density function for  $(X(t), Y(t))$  under the boundary conditions:

$$p(X(t) = x, Y(t) = y | a_x < X(t') < b_x, a_y < Y(t') < b_y, 0 < t' \leq t, X(0) = x_0, Y(0) = y_0). \quad (3)$$

This density, which we shorten to  $p(x, y, t)$  from now on, is the solution to the Fokker-Planck equation [Oksendal, 2013]:

$$\frac{\partial}{\partial t} p(x, y, t') = -\mu_x \frac{\partial}{\partial x} p(x, y, t') - \mu_y \frac{\partial}{\partial y} p(x, y, t') + \frac{1}{2} \sigma_x^2 \frac{\partial^2}{\partial x^2} p(x, y, t') + \rho \sigma_x \sigma_y \frac{\partial^2}{\partial x \partial y} p(x, y, t') + \frac{1}{2} \sigma_y^2 \frac{\partial^2}{\partial y^2} p(x, y, t'), \quad (4)$$

$$p(a_x, y, t') = p(b_x, y, t') = p(x, a_y, t') = p(x, b_y, t') = 0, \quad (5)$$

$$0 < t' \leq t.$$

Differentiating  $p(x, y, t)$  with respect to the boundaries produces the transition density of a particle beginning and ending at the points  $(X_1(0), X_2(0))$  and  $(X_1(t), X_2(t))$  respectively, while attaining the minima  $a_x/a_y$  and maxima  $b_x/b_y$  in each coordinate direction:

$$\frac{\partial^4}{\partial a_x \partial b_x \partial a_y \partial b_y} p(x, y, t) =$$

$$p\left(X(t) = x, Y(t) = y \mid \min_{t'} X(t') = a_x, \max_{t'} X(t') = b_x, \min_{t'} Y(t') = a_y, \max_{t'} Y(t') = b_y, 0 < t' \leq t, X(0) = x_0, Y(0) = y_0\right). \quad (6)$$

The transition density for the considered system has been used in computing first passage times [Kou et al., 2016, Sacerdote et al., 2016], with application to structural models in credit risk and default correlations [Haworth et al., 2008, Ching et al., 2014]. He et al. [1998] use variants of the differentiated solutions with respect to some of the boundaries to price financial derivative instruments whose payoff depends on observed maxima/minima.

Closed-form solutions to (4) - (5) are available for some parameter regimes. When  $\rho = 0$ , the transition density of the process can be obtained with a Fourier expansion. When  $a_1 = -\infty$  and  $b_1 = \infty$ , the method of images can be used to enforce the remaining boundaries. For either  $a_1, a_2 = -\infty$  or  $b_1, b_2 = \infty$ , the Fokker-Planck equation is a Sturm-Liouville problem in radial coordinates. Both of these techniques are used by He et al. [1998]. However, to the best of our knowledge, there is no closed-form solution to the general problem in (4) - (5).

It is still possible to approach the general problem by proposing a Fourier expansions. However, a drawback of this out-of-the-box solution is that the system matrix for the corresponding eigenvalue problem is large and dense. An alternative is to use a finite difference scheme. However, discretization of the initial condition introduces a numerical bias in the estimation procedure.

In this paper, we propose a solution to the general problem (4) - (5) which is obtained by combining a small-time analytic solution with a finite-element method. Our application is the maximal likelihood estimation

ADD OUT APPLICATION (ESTIMATION)

## 2 Approximate Numerical Solutions

Before considering any solutions to (4) - (5), we simplify the PDE in (4) by using the fact that parameters  $(\mu_x, \mu_y, \sigma_x, \sigma_y, \rho)$  are constant and solving for the exponential decomposition

$$p(x, y, t) = \exp(\alpha x + \beta y + \gamma t) q(x, y, t).$$

We can find  $\alpha, \beta$  and  $\gamma$ , as well as a scaling transformation, such that  $q(x, y, t)$  satisfies

$$\begin{aligned} \frac{\partial}{\partial t} q(x, y, t) &= \frac{1}{2} \sigma_x^2 \frac{\partial^2}{\partial x^2} q(x, y, t) + \rho \sigma_x \sigma_y \frac{\partial^2}{\partial x \partial y} q(x, y, t) + \frac{1}{2} \sigma_y^2 \frac{\partial^2}{\partial y^2} q(x, y, t). \\ q(x, y, t) &= 0 \\ q(x, y, 0) &= \delta(x - x_0) \delta(y - y_0) \end{aligned} \quad \text{for } (x, y) \in \partial\Omega \quad (7)$$

on the unit square. We will consider the solution to this PDE without loss of generality.

### 2.1 Fourier Expansion

The formal Fourier (sinusoidal) expansion for the problem is given by

$$\begin{aligned} q(x, y, t) &= \lim_{K, L \rightarrow \infty} \sum_{k=1}^K \sum_{l=1}^L c_{k,l}(t) \sin\left(2\pi \cdot k \frac{x - a_x}{b_x - a_x}\right) \sin\left(2\pi \cdot l \frac{y - a_y}{b_y - a_y}\right) \\ \hat{q}(x, y, t) &= \sum_{k=1}^K \sum_{l=1}^L c_{k,l}(t) \sin\left(2\pi \cdot k \frac{x - a_x}{b_x - a_x}\right) \sin\left(2\pi \cdot l \frac{y - a_y}{b_y - a_y}\right) \end{aligned} \quad \text{for some } K, L$$

With  $\rho = 0$ , the sinusoidal functions are the eigenvectors for the differential operator in (7), and we would proceed by substituting  $\hat{q}$  into (7) and deriving a system of ODEs whose solution is the vector  $(c_{1,1}(t), \dots, c_{K,L}(t))$ . In this case the system matrix is diagonal so that each  $c_{k,l}(t)$  can be written down analytically.

We can proceed in the same manner in the case where  $\rho \neq 0$ . However, the mixing terms

$$\frac{\partial^2}{\partial x \partial y} \sin\left(2\pi \cdot k \frac{x - a_x}{b_x - a_x}\right) \sin\left(2\pi \cdot l \frac{y - a_y}{b_y - a_y}\right),$$

are cosines and as such have a non-sparse representation in terms of sine series. Because of this, the matrix for the system of ODEs when  $\rho \neq 0$  is dense. Moreover, the truncation values for  $K$  and  $L$  are also large. Finally, to compute

### 2.2 Finite Difference

### 2.3 Finite Element Method

The method we use relies on two pieces:

1. a small-time analytic solution  $q(x, y, t_\epsilon)$  for the IC/BC problem,
2. a family of orthonormal basis functions which represent  $q(x, y, t_\epsilon)$  parsimoniously.

By combining 1) and 2), we can efficiently find a weak solution to the PDE (7) via the finite element method [Shaidurov, 2013]. Convergence of our method to the strong solution under the  $L^2(\bar{\Omega})$  norm is guaranteed as long as the family we propose is complete in the Banach space of functions induced under  $L^2(\bar{\Omega})$ .

The small-time solution is derived by considering the fundamental solution  $G(x, y|t, x_0, y_0)$  for the unbounded problem in (7), which is the bivariate Gaussian density with mean and covariance determined by the initial condition and the diffusion parameters [Stakgold and Holst, 2011]. We can then find a small

enough  $t_\varepsilon$  such that  $G(x, y|t_\varepsilon, x_0, y_0)$  is numerically zero on three of the four boundaries of  $\bar{\Omega}$ . The zero-condition on the remaining boundary is enforced by suitably reflecting  $G(x, y|t_\varepsilon, x_0, y_0)$  about the boundary. The small-time solution therefore takes on the analytic form

$$q(x, y, t_\varepsilon) = G(x, y|t_\varepsilon, x_0, y_0) - G(x, y|t_\varepsilon, x'_0, y'_0),$$

for some known  $(x'_0, y'_0, t_\varepsilon)$ .

The construction of the orthonormal basis functions is motivated by the Green's function for the unbounded problem: before performing Gram-Schmidt orthogonalization, the finite family of basis functions are of the form

$$\tilde{\Psi}_k(x, y|x_k, y_k, \rho, \sigma) = N \left( (x, y)^T \middle| (x_k, y_k)^T, \begin{pmatrix} \sigma^2 & \rho\sigma^2 \\ \rho\sigma^2 & \sigma^2 \end{pmatrix} \right) x(1-x)y(1-y).$$

The advantage of these basis elements is that they better resolve the fundamental for the unbounded problem by taking into account  $\rho$  in the covariance of each kernel. By performing Gram-Schmidt orthogonalization under the  $L^2(\Omega)$  norm, we arrive at a family of orthonormal functions which can better resolve small-time solutions having a large correlation coefficient.

**Lemma 1.** *The maximum likelihood estimator is consistent as  $n \rightarrow \infty$  and  $k \rightarrow \infty$ :*

$$\hat{\theta}_{n,k} \rightarrow \theta$$

*Proof.* By the definition of weak convergence, given the weak solution  $q_k$  and the classical solution  $q$ , for any continuous function  $f$ ,

$$\langle q_k | f \rangle \rightarrow \langle q | f \rangle \text{ as } k \rightarrow \infty.$$

Because  $f$  can be any function in  $L^2$ , we can choose  $f$  to be  $\exp(ilx)$  for any integer  $l$ . This means that the characteristic function of  $X_k$  converges pointwise to the characteristic function of  $X$ . By Levy's continuity theorem, this means that

$$X_k \xrightarrow{d} X \text{ as } k \rightarrow \infty.$$

Next, given Theorem 4.1 in Singler [2008], we know that, for each  $k$ ,  $q_k$  satisfies the criteria in Casella and Berger [2002] to guarantee that, for data  $X_k \sim F_k(\theta)$ ,

$$\hat{\theta}_{n,k}(X_k) \xrightarrow{p} \theta$$

as  $n \rightarrow \infty$ . In other words, the MLE estimator for  $(\sigma_x, \sigma_y, \rho)$  based on the likelihood function under  $F_k$  for data sampled from  $F_k$ . Now we need to show that the same holds for data sampled from  $F$  as  $k \rightarrow \infty$ .

To do this, we will use Chebyshev's inequality:

$$\Pr_X (|\hat{\theta}_{n,k}(X_k) - \theta| \geq \varepsilon) \leq \frac{\mathbb{E}_X [(\hat{\theta}_{n,k}(X_k) - \theta)^2]}{\varepsilon^2}$$

□

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