

Christiane Fuchs

Inference for Diffusion Processes

With Applications in Life Sciences

扩散过程推断理论
及其在生命科学中的应用

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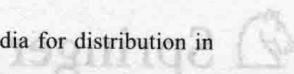
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by Christiane Fuchs

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Inference for Diffusion Processes

Inference for Diffusion Processes

With Applications in Life Sciences



Foreword

Sampling with Brownian *To Florian* – modifications, different processes and stochastic differential equations have a long tradition as mathematical models for stochastic phenomena evolving in continuous time, with applications in diverse substantive fields such as engineering, finance and life sciences. While many textbooks, in particular in financial mathematics, include concise and rigorous introductions to diffusion processes and stochastic calculus, they often provide very limited information on modern statistical inference and usually neglect the question of how to adequately approximate original phenomena through diffusion processes at all. These two issues are, however, of fundamental importance in a number of applied fields, in particular in modern life sciences. First, the original processes typically have a huge but discrete state space, and approximations through diffusion processes can be challenging. Second, data are usually observed with low frequency and often at non-equidistant time points, involve measurement error etc. This requires modern tools such as simulation-based Bayesian inference. And through applications in epidemiology and molecular biology, this book covers the existing gap. For the first time, it provides a unified presentation of the approximation techniques, previously often developed for special cases only, and a thorough account of modern statistical inference, including a powerful Bayesian approach. Together with the two application chapters, this book will be of high value to theoretical and applied work. For me, it was a pleasure to watch it grow, to be able to give some advice, to read the final version, and, at last, to see it on my shelf!

Bielefeld, Germany

Ludwig Fahrmeir

Foreword

Beginning with Brownian motion and its modifications, diffusion processes and stochastic differential equations have a long tradition as mathematical models for stochastic phenomena evolving in continuous time, with applications in diverse substantive fields such as engineering, finance and life sciences. While many textbooks, in particular in financial mathematics, include concise and rigorous introductions to diffusion processes and stochastic calculus, they often provide only limited information on modern statistical inference and usually neglect the question of how to adequately approximate original phenomena through a diffusion process at all. These two issues are, however, of fundamental importance in a number of applied fields, in particular in modern life sciences: First, the original processes typically have a large but discrete state space, and approximations through diffusion processes can be challenging. Second, data are usually observed with low frequency and often at non-equidistant time points, involve measurement error etc. This requires modern tools such as simulation-based Bayesian inference. Motivated through applications in epidemiology and molecular biology, this book closes the existing gap. For the first time, it provides a unified presentation of the approximation techniques, previously often developed for special cases only, and a thorough account of modern statistical inference, including a powerful Bayesian approach. Together with the two application chapters, this book will be of high value for theoretical and applied work. For me, it was a pleasure to watch it grow, to be able to give some advice, to read the final version, and, at last, to see it on my bookshelf!

Munich, Germany

Ludwig Fahrmeir

In addition to existing literature, this book emphasizes building an application. Importance is attached to presenting the methods both comprehensively and mathematically well-founded. As such, the book addresses both theoreticians (mathematicians and statisticians), as well as practitioners, like biostatisticians and biologists. The reader is required to have basic knowledge about deterministic differential equations, probability theory and statistics. An introduction to stochastic calculus, in particular to diffusions, is provided in this book.

Preface

This book originates from my doctoral thesis. One of the first problems of my work was to describe the spread of an infectious disease by a diffusion process and to statistically estimate the involved model parameters. At that time, I did not expect such a seemingly straightforward task to surface so many diverse open problems to fill an entire book. As a mathematician by training, I knew about stochastic calculus, but I did not anticipate the troubles of deriving diffusion approximations and inferring their parameters from real data.

When delving into the diffusion approximation literature, I noticed that there were several, at first sight, contradicting approaches, some of them formulated in generality, others being carried out for particular problems. Their appropriateness, differences and conformities, however, were unclear as well as their extension to more complex, e.g. multidimensional, processes. Furthermore, parameter estimation for diffusions is a challenging problem, in particular if the application of interest involves multi-dimensional processes, few observation times, latent variables and considerable measurement error. Under these circumstances, probably the only applicable technique is a popular Bayesian approach which is used in a number of scientific papers. I was astonished that I could not find any textbook which comprehensively explained it. Moreover, the method has a well-known but hard-to-grasp convergence problem, which has not been detailed in any book or review so far. Since I am convinced that these are subjects of wide-spread interest and importance, I dedicated to them the major chapter in each of the first two parts of this book. The third part finally addresses the initial project which triggered the theoretical questions: to estimate a diffusion model for the spread of diseases.

In contrast to existing literature, this book treats modelling and inference for diffusions under one umbrella. It thus covers both steps that necessarily arise in a real application. Importance is attached to presenting the methods both comprehensibly and mathematically well-founded. As such, the book addresses both theoreticians, like mathematicians and statisticians, as well as practitioners, like bioinformaticians and biologists. The reader is required to have basic knowledge about deterministic differential equations, probability theory and statistics. An introduction to stochastic calculus, in particular to diffusions, is provided in this book.

Everybody who supported me during the writing of my thesis “Bayesian Inference for Diffusion Processes with Applications in Life Sciences”, submitted in 2010 at Ludwig-Maximilians-Universität Munich under my maiden name Christiane Dargatz, also supported the making of this book. My sincere gratitude is due to my supervisors Ludwig Fahrmeir and Gareth Roberts. They enriched my work through their advice, ideas and encouragement. I deeply appreciate the careful proof-reading and helpful comments by Michael Höhle. Katrin Schneider and Lothar Schermelleh deserve my thanks for having initiated the collaboration on the FRAP project and having collected all the data. I thank my former and present colleagues at Ludwig-Maximilians-Universität Munich and Helmholtz Centre Munich for their interest and helpful discussions. Furthermore, I thank Niels Thomas and Alice Blanck from Springer for the friendly and constructive cooperation. My family has been a constant source of support, which I greatly acknowledge. My heartfelt gratitude is due to my husband Florian, who caringly accompanied me and my work all chapters long.

Munich, Germany

Christiane Fuchs

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Notation

Symbols

$\mathbb{N} = \{1, 2, 3, \dots\}$	the natural numbers
$\mathbb{N}_0 = \mathbb{N} \cup \{0\}$	the non-negative whole numbers
\mathbb{Z}	the whole numbers
\mathbb{Q}	the rational numbers
\mathbb{R}	the real numbers
\mathbb{R}_+	the strictly positive real numbers
$\mathbb{R}_0 = \mathbb{R}_+ \cup \{0\}$	the non-negative real numbers
x	a column vector
x'	the transpose of x
$\mathbf{0} = (0, \dots, 0)'$	the null vector
$e_i = (0, \dots, 1, \dots, 0)'$	the i th unit vector
I	the identity matrix
$\text{diag}(\mathbf{A})$	the main diagonal of the quadratic matrix \mathbf{A}
$\text{diag}(\mathbf{a})$	quadratic matrix with main diagonal \mathbf{a} and zero entries otherwise
$ a $	the absolute value of $a \in \mathbb{R}$
$\ \mathbf{A}\ $	Euclidean distance, i. e. $\ \mathbf{A}\ ^2 = \text{tr}(\mathbf{A}'\mathbf{A})$ for a vector or matrix \mathbf{A}
$ \mathbf{k} = \sum_{i=1}^n k_i$, where $\mathbf{k} = (k_1, \dots, k_n)'$	(in the context of Sect. B.1)
$\mathbf{u} \diamond \mathbf{v} = (u_1 v_1, \dots, u_n v_n)'$, where $\mathbf{u} = (u_1, \dots, u_n)'$ and $\mathbf{v} = (v_1, \dots, v_n)'$	
$(\Omega, \mathcal{F}^*, \mathcal{F}, \mathbb{P})$	a filtered probability space with sample space Ω , σ -algebra \mathcal{F}^* , natural filtration $\mathcal{F} = (\mathcal{F}_t)_{t \geq 0}$ and probability measure \mathbb{P}
\mathcal{L}	the σ -algebra of Lebesgue subsets of \mathbb{R}

\mathbb{L}	Lebesgue measure
\mathbb{W}	Wiener measure
$\mathbb{P}_1 \ll \mathbb{P}_2$	the measure \mathbb{P}_1 is absolutely continuous with respect to \mathbb{P}_2
$\mathbb{P}_1 \perp \mathbb{P}_2$	the measures \mathbb{P}_1 and \mathbb{P}_2 are mutually singular, i. e. they have disjoint support
$\mathbb{P}_1 \otimes \mathbb{P}_2$	factorisation of measures
$\mathcal{L}(X)$	the distribution of the random variable X
$\mathbf{1}(A)$	indicator function; equal to one if A is true and zero otherwise
$\delta(x - y)$	Dirac delta function; equal to ∞ if $x = y$ and zero otherwise
$a = o(h)$	$\Leftrightarrow \lim_{h \rightarrow 0} (a/h) = 0$
$\biguplus_{i=1}^n A_i$	disjoint union of sets A_1, \dots, A_n
Γ	Gamma function, defined as $\Gamma(x) = \int_0^\infty t^{x-1} \exp(-t) dt$
d/dt	total derivative, i. e.

$$\frac{d}{dt} f(t, \xi_1, \dots, \xi_m) = \frac{\partial f(t, \xi_1, \dots, \xi_m)}{\partial t} + \sum_{j=1}^n \frac{\partial f(t, \xi_1, \dots, \xi_m)}{\partial \xi_j} \frac{d\xi_j}{dt}$$

$\partial/\partial t$ partial derivative, where all arguments but t remain constant

Abbreviations

SDE	stochastic differential equation
ODE	ordinary differential equation
i.i.d.	independent and identically distributed
a.s.	almost surely

Distributions

Normal distribution

$\mathbf{X} \sim \mathcal{N}(\mu, \Sigma)$ with $\mu \in \mathbb{R}^n$, $\Sigma \in \mathbb{R}^{n \times n}$ symmetric and positive definite

Log-normal distribution

$\mathbf{X} \sim \mathcal{LN}(\mu, \Sigma) \Leftrightarrow \log(\mathbf{X}) \sim \mathcal{N}(\mu, \Sigma)$

Truncated normal distribution

$\mathbf{X} \sim \mathcal{N}_{\text{trunc}}(\mu, \sigma^2)$; generates random numbers from $\mathcal{N}(\mu, \sigma^2)$ restricted to the positive real line, i. e. the density $f(x)$ of this distribution is proportional to the density of $\mathcal{N}(\mu, \sigma^2)$ for $x > 0$ and zero otherwise

Multivariate t distribution

$\mathbf{X} \sim t_\nu(\mu, \Sigma)$ with $\nu \in \mathbb{R}_+$ degrees of freedom, $\mu \in \mathbb{R}^n$, $\Sigma \in \mathbb{R}^{n \times n}$ symmetric and positive definite and density

Chapter I

Introduction

$$f(\mathbf{x}) = \frac{\Gamma\left(\frac{\nu+n}{2}\right)}{\Gamma\left(\frac{\nu}{2}\right)(\nu\pi)^{n/2}} |\boldsymbol{\Sigma}|^{-\frac{1}{2}} \\ \left(1 + \frac{1}{\nu}(\mathbf{x} - \boldsymbol{\mu})' \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu})\right)^{\frac{\nu+n}{2}} \text{ for } \mathbf{x} \in \mathbb{R}^n$$

Uniform distribution	$X \sim U(A)$ for a discrete or continuous set A
Poisson distribution	$X \sim \text{Po}(\lambda)$ with $\lambda \in \mathbb{R}_+$
Exponential distribution	$X \sim \text{Exp}(\lambda)$ with $\lambda \in \mathbb{R}_+$
Gamma distribution	$X \sim \text{Ga}(a, b)$ with $a, b \in \mathbb{R}_+$ and density
	$f(x) = \frac{b^a}{\Gamma(a)} x^{a-1} \exp(-bx) \quad \text{for } x \in \mathbb{R}_+$
Inverse gamma distribution	$X \sim \text{IG}(a, b) \Leftrightarrow 1/X \sim \text{Ga}(a, b)$

Statistical processes at molecular level up to the various types of inference are discussed in [Chapters 2–5](#). Mathematical models are indispensable tool for the understanding of such complex natural phenomena.

It is difficult to describe the time continuous evolution of a given system, stochastic models are often favoured as they allow computationally simple simulation and estimation techniques. Such models, however, do not capture the randomness of the underlying dynamics and therefore turn out to be inadequate in many applications. The estimation of exact individual based stochastic models, for the other hand, typically proves to be impossible in practice when the considered system involves large numbers of objects. A related subproblem concerns the application of stochastic differential equations (SDEs), whose solutions are given by diffusion processes. Hence, diffusions have become an increasingly important tool for the numerical analysis of real world phenomena.

However, approximation of a given dynamic system is often done heterogeneously in the literature, leading to difficulties that do not correctly reflect the true dynamics of the original process. Furthermore, the statistical inference for diffusions typically turns out to be demanding in real data situations as described below. Hence, the statistical estimation of complex diffusion models is applied to real datasets most frequently. These issues are addressed in the present book.

1.1 Aims of This Book

The main objectives of this book are threefold. First of all, given a dynamical system of interest with the aim to describe its possible evolutions by means of a stochastic process, one needs to construct this process such that it appropriately captures the characteristics of the considered real phenomenon. In applications in life sciences, the original process typically concerns e.g. the quantity of objects such as the number of infectious individuals in a population or the number of patients in a

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