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Continuous Time Random Walks on Bounded Domains[∗]

Nathanial Burch Department of Mathematics Colorado State University burch@math.colostate.edu

R. B. Lehoucq Applied Math and Applications Sandia National Laboratories rblehou@sandia.gov

A useful perspective to take when studying anomalous diffusion processes is that of a continuous time random walk and its associated generalized master equation. We derive the generalized master equations for continuous time random walks that are restricted to a bounded domain and compare numerical solutions to kernel density estimates of the probability density function computed from simulations. The numerical solution of the generalized master equation represents a powerful tool in the study of continuous time random walks on bounded domains.

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I. INTRODUCTION

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INTRODUCTION

Study of continuous time random walks on bounded domains.

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INTRODUCTION

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LEVIENCOPY Example, contaminant flow in mains. Recent eff Anomalous diffusion processes have been observed in many applications, for example, contaminant flow in groundwater [1], dynamic motions in proteins [2], turbulence in fluids [3], and dynamics of financial markets [4] have all been verified experimentally to exhibit characteristics of anomalous diffusion; see [5] for a review. A diffusion process is termed anomalous when the mean square displacement satisfies

$$
\langle X^2(t) \rangle = \int_{\mathbb{R}} x^2 v(x, t) \, dx \sim t^{\gamma}, \quad \gamma \neq 1, \qquad (1)
$$

unlike normal diffusion, where $\gamma = 1$. In (1), v is the probability density function of the random variable $X(t)$, which is the displacement of a diffusing particle at time t. When $0 < \gamma < 1$ such a process is subdiffusive, while $\gamma > 1$ indicates a superdiffusive process. A thorough survey of theoretical considerations for anomalous diffusion processes can be found in [6].

One common perspective to take when studying anomalous diffusion processes is that of a continuous time random walk (CTRW) and its associated generalized master equation $[6, 7]$. As discussed in $[6, 8, 9]$, this perspective is especially useful when the diffusion process lacks finite characteristic scales, e.g., mean square displacement of a particle or the mean wait-time between collisions. Though the relationship between CTRW in free space and anomalous diffusion processes has been well-studied, the same cannot be said for the subsequent relationship on bounded domains. Of the existing research, much is concerned with graphs and lattices and

there exists comparatively little work into the generalized master equations for CTRW on general bounded domains. Recent efforts, namely [8], however, have made advances to remedy this by investigating certain Markovian CTRW with absorbing and reflecting boundary conditions. The analysis in [8] is limited in relying on special cases so that explicit, closed-form, solutions to the generalized master equations can be found for simple onedimensional domains. This analysis becomes difficult when the Markovian assumption is removed, the domains in two and three dimensions are not simple, and the step density is not suitably chosen, e.g., it is approximated from data.

ing handom was in as a secondar way in the bare of the control in the particular constraints and the spectral of the matrices of the stationary constraints in the stationary in the stationary in the stationary in the stat There is also a well-known relationship between the generalized master equations for CTRW in free space and fractional diffusion equations. For bounded domains, considerably more research exists for fractional diffusion than for integro-differential equations, such as the aforementioned generalized master equations. For instance, the paper [10] gives a probabilistic interpretation of the Lévy-Feller fractional diffusion equation with absorbing boundaries, where the fraction of the Laplacian is restricted to $\alpha \in (1, 2)$, i.e., the cases $\gamma \geq 2$ in equation (1) are not considered. Other work, e.g., [11], considers fractional diffusion equations on bounded domains with reflecting boundaries. However, even for fractional diffusion, there is little notion of general boundary conditions outside of specialized domains, e.g., rectangles and parallelepipeds in two and three dimensions, respectively.

In this paper, we derive the generalized master equations for both Markovian and non-Markovian continuous time bounded random walks (CTBRW) with either absorbing or insulated boundaries. An insulated boundary restricts the random walker from taking a step past the boundary, e.g., a special case of insulated boundaries is the reflective behavior described in [8]. Boundary conditions such as these appear naturally when a diffusion process is restricted to a bounded domain, e.g., contami-

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nant flow in an underground aquifer. The boundary conditions for a random walker induce volume constraints on the solution of the generalized master equation and the resulting equations are then studied via a variational formulation and conforming finite element method described in [12, 13]. This computational approach allows for the study of a wide-class of problems on nontrivial bounded domains in two and three dimensions, a capability currently unavailable.

We demonstrate the numerical solutions to the generalized master equations agree with kernel density estimates of the solution from CTBRW simulations. This renders the aforementioned finite element formulation a powerful tool in studying CTBRW as models of anomalous diffusion because computationally intensive CTBRW simulations may be avoided.

II. CTRW IN BOUNDED DOMAINS

We consider separable CTRW, i.e., wait-times are independent of the choice of step. The wait-time density is denoted with ω and the step density with $J(y, x)$. That is, $J(y, x)$ is the probability density of taking a step from y to x and, consequently, $\int_{\mathbb{R}} J(y, x) dx = 1$. Note, however, that $\int_{\mathbb{R}} J(y, x) dy \neq 1$ in general. It is well-known, see for instance [4, 6, 14], that the probability density function of the CTRW, $u(x, t)$, satisfies the generalized master equation

$$
u_t(x,t) = \int_0^t \Lambda(t - t') L_{\mathbb{R}}^J u(x, t') dt',
$$
 (2)

where the Laplace transform of the memory kernel Λ is

$$
\widehat{\Lambda}(\zeta) = \frac{\zeta \widehat{\omega}(\zeta)}{1 - \widehat{\omega}(\zeta)}
$$

and we have introduced the operator

$$
L_f^fu(x,t):=\int_I\left(u(y,t)f(y,x)-u(x,t)f(x,y)\right){\mathrm d}y.
$$

The analogous operator to $L_I^f u(x,t)$ for a CTRW on a lattice has been studied previously, see, e.g., [15].

For this paper, we consider two choices of Λ in (2):

$$
\Lambda(t - t') = \frac{1}{2\tau}\delta(t - t')
$$
 (3a)

$$
\Lambda(t - t') = \frac{1}{\tau^2} \exp\left(-\frac{t - t'}{\tau/2}\right),\tag{3b}
$$

which are tantamount to specifying that wait-times are distributed as

$$
Exp(2\tau), \text{ i.e., } \omega(t) = \frac{1}{2\tau} \exp\left(-\frac{t}{2\tau}\right) \qquad (4a)
$$

Gamma(2,
$$
\tau
$$
), i.e., $\omega(t) = \frac{t}{\tau^2} \exp\left(-\frac{t}{\tau}\right)$, (4b)

respectively, both of which imply finite mean wait-times. In fact, (4a) and (4b) imply the underlying CTRW are compound Poisson and renewal reward processes, respectively. With (3), (2) reduces to

$$
u_t(x,t) = \frac{1}{2\tau} L_{\mathbb{R}}^J u(x,t)
$$
 (5a)

$$
u_t(x,t) + \frac{\tau}{2} u_{tt}(x,t) = \frac{1}{2\tau} L_{\mathbb{R}}^J u(x,t).
$$
 (5b)

Since the mean wait-time is finite, (5a) and (5b) are models for either normal diffusion or anomalous superdiffusion, depending on whether $\int_{\mathbb{R}} (x - y)^2 J(y, x) dx$ is finite or infinite, respectively. By selecting a heavy-tailed waittime density, we may obtain models for subdiffusion, normal diffusion, or superdiffusion, depending now upon the interplay between the characteristic step-length variance and characteristic mean wait-time. We refer the reader to [6] for further information.

Boundary conditions for CTBRW manifest themselves in the definition of the step density $J(y, x)$ and are now described. We let ϕ be a symmetric probability density that should be interpreted as the step density in the absence of boundary conditions.

We first describe the behavior of fully absorbing boundaries. Once a random walker reaches, or steps beyond, the boundary $\partial\Omega$, he is banned from Ω for all future time. This description gives the step density

$$
J(y,x) = \begin{cases} \phi(x-y), & y \in \Omega, \\ \delta(x-y), & y \notin \Omega, \end{cases}
$$
 (6)

so that a random walker may step from $y \in \Omega$ to $x \in \mathbb{R}$ via the radial density $\phi(x - y)$. It is convenient then to set $u(x, t) = 0$ for $x \notin \Omega$ and inserting (6) into (2) gives

$$
\begin{cases}\n u_t(x,t) = \int_0^t \Lambda(t-t') L_{\mathbb{R}}^{\phi} u(x,t') dt', & x \in \Omega, \\
 u(x,t) = 0, & x \notin \Omega\n\end{cases}
$$

and, thus,

$$
u_t(x,t) = \frac{1}{2\tau} L_{\mathbb{R}}^{\phi} u(x,t), \quad x \in \Omega, \qquad (7a)
$$

$$
u_t(x,t) + \frac{\tau}{2} u_{tt}(x,t) = \frac{1}{2\tau} L_{\mathbb{R}}^{\phi} u(x,t), \quad x \in \Omega.
$$
 (7b)

The equation (7a) was studied in the context of Markovian CTRW in [8], while (7b) belongs to a non-Markovian CTRW.

The case of fully insulated boundaries restricts a random walker from reaching, or stepping beyond, ∂Ω. One interpretation of this description gives rise to

$$
J(y,x) = \chi_{\Omega}(x)\phi(x-y)
$$

+ $\delta(x-y)\int_{\mathbb{R}\setminus\Omega}\phi(z-y)\,\mathrm{d}z$, $y \in \Omega$, (8)

The step density (8) states that a random walker may step from $y \in \Omega$ to $x \in \Omega$ via the radial density $\phi(x-y)$.

Further, there is a nonzero probability, $\int_{\mathbb{R}\setminus\Omega} \phi(z-y) \, \mathrm{d}z$, of the walker at $y \in \Omega$ not taking a step. Together, these guarantee that the random walker remains in Ω for all time and, consequently, defining $J(y, x)$ for $y \notin \Omega$ in (8) is not required. Insertion of (8) into (2) gives

$$
u_t(x,t) = \int_0^t \Lambda(t-t') L_\Omega^\phi u(x,t') dt', \quad x \in \Omega
$$

and, thus,

$$
u_t(x,t) = \frac{1}{2\tau} L^{\phi}_{\Omega} u(x,t), \quad x \in \Omega, \qquad (9a)
$$

$$
u_t(x,t) + \frac{\tau}{2}u_{tt}(x,t) = \frac{1}{2\tau}L_{\Omega}^{\phi}u(x,t), \quad x \in \Omega.
$$
 (9b)

Now, we relate the equations (7) and (9) to nonlocal boundary value problems that have been postulated and studied in various different settings [8, 12, 13, 16, 17]. A nonlocal boundary value problem augments (5) by constraining the solution on a nonzero volume, generalizing the notion of classical boundary conditions to that of a volume constraint. Such volume constraints need not be relegated to the exterior of Ω . We specify an initial density $u_0(x)$ on Ω , satisfying $u_0 \ge 0$ and $\int_{\Omega} u_0(x) dx = 1$.

The nonlocal Dirichlet boundary value problems are

$$
\begin{cases}\n u_t(x,t) = \frac{1}{2\tau} L_{\mathbb{R}}^{\phi} u(x,t), & x \in \Omega, \\
 u(x,t) = 0, & x \notin \Omega, \\
 u(x,0) = u_0(x), & x \in \Omega\n\end{cases}
$$
\n(10a)

and

$$
\begin{cases}\nu_t(x,t) + \frac{\tau}{2} u_{tt}(x,t) = \frac{1}{2\tau} L_{\mathbb{R}}^{\phi} u(x,t), & x \in \Omega, \\
u(x,t) = 0, & x \notin \Omega, \\
u(x,0) = u_0(x), & x \in \Omega, \\
u_t(x,0) = 0, & x \in \Omega.\n\end{cases}
$$
\n(10b)

The nonlocal Dirichlet boundary condition constrains u for $x \notin \Omega$, analogous to the classical Dirichlet boundary condition that does so at the points on the boundary.

The nonlocal Neumann boundary value problems are

$$
\begin{cases} u_t(x,t) = \frac{1}{2\tau} L_{\Omega}^{\phi} u(x,t), & x \in \Omega, \\ u(x,0) = u_0(x), & x \in \Omega \end{cases}
$$
 (11a)

and

$$
\begin{cases}\n u_t(x,t) + \frac{\tau}{2} u_{tt}(x,t) = \frac{1}{2\tau} L_{\Omega}^{\phi} u(x,t), & x \in \Omega, \\
 u(x,0) = u_0(x), & x \in \Omega, \\
 u_t(x,0) = 0, & x \in \Omega.\n\end{cases}
$$
\n(11b)

The integrals in (11), in contrast to those in (10), are over Ω rather than all of R. This implies a constraint on diffusion so that it occurs strictly inside Ω , i.e., density neither enters nor exits Ω , which is analogous to the classical Neumann boundary condition.

In summary, the descriptions of the boundary conditions for the CTBRW determine J in (2) so that (2) reduces to an appropriate nonlocal boundary value problem in (10) or (11). Evidently, these nonlocal boundary value problems describe the time-evolution of the probability density of the state of the corresponding CTBRW. The analysis in [12, 13] allows us to analyze (10) and (11) via a variational formulation and conforming finite element method so extending the class of problems computationally tractable.

We simulate N random walkers and a kernel density estimate of u at various points in time is computed. This kernel density estimate is compared to the finite element solution of the associated nonlocal boundary value problem. We select ϕ to be a Lévy stable density with stability index α , characterized via

$$
\phi(s) = \mathcal{F}^{-1}\left\{\exp\left(-\varepsilon^{\alpha}|\xi|^{\alpha}\right)\right\}(s),\tag{12}
$$

and choose $\alpha = 3/2$ and $\varepsilon = 0.25$. For simulations with absorbing boundaries, we use $u_0(x) = 2x$ and for insulated boundaries, $u_0(x) = \frac{\pi}{2} \sin(\pi x)$. These choices of u_0 , in consideration of the respective boundary conditions, were opportune and have no effect on our conclusions.

A walker begins at a random location $x_0 \in (0,1)$ according to the initial density $u_0(x)$. For each k, a wait-time t_k is generated from ω and the arrival-time $a_k = a_{k-1} + t_k$ is recorded. A step s_k is generated from ϕ , the new location $x_k = x_{k-1} + s_k$ is recorded, and then boundary conditions are imposed. For instance, if $x_k \notin (0, 1)$ for the case of absorbing boundary conditions, the random walk is stopped. In the case of insulated

Absorbing boundaries	Insulated boundaries
$a_0=0$	$a_0=0$
simulate $x_0 \sim u_0(x)$	simulate $x_0 \sim u_0(x)$
for k from 1 to T	for k from 1 to T
simulate $t_k \sim \omega(t)$	simulate $t_k \sim \omega(t)$
$a_k = a_{k-1} + t_k$	$a_k = a_{k-1} + t_k$
simulate $s_k \sim \phi(s)$	simulate $s_k \sim \phi(s)$
$x_k = x_{k-1} + s_k$	$x_k = x_{k-1} + s_k$
if $x_k \notin (0,1)$	if $x_k \notin (0,1)$
break	$x_k = x_{k-1}$
end	end
end	end

TABLE I: Pseudo code for simulating CTBRW.

boundary conditions, if $x_k \notin (0, 1)$, we set $x_k = x_{k-1}$, i.e., the walker waits at the current position. Again, this treatment of an insulated boundary differs from the reflective behavior in [8] and is merely one approach for treating insulated boundaries. Deciding on the appropriate treatment is application specific and depends largely on the mechanism driving the CTBRW. Note that the position of the random walker is known for all time, e.g., the walker is at position x_k for the time interval $[a_k, a_{k+1})$.

Data from the CTBRW simulations are used to estimate the density $u(x, t)$. Let $p_i(t)$ denote the *i*-th random

FIG. 1: (Color Online) Panels (a)–(d) show kernel density estimates of the CTBRW simulations (solid) on $\Omega = (0,1)$ with $N = 8 \cdot 10^4$ and numerical solutions (dashed) of the nonlocal boundary value problems (10a)–(11b), respectively. The horizontal axis is x and the vertical axis is the value of the density. The ten curves represent ten different values of $t \in [0, 0.5]$.

walker's position at time t and partition $\Omega = (0, 1)$ into n subintervals Ω_i . Then, define the kernel density estimate

 $\mu_N(x, t) := \sum_{n=1}^n$ $k=1$ $\chi_{\Omega_k}(x)$ $\begin{pmatrix} 1 \end{pmatrix}$ N_h $\stackrel{N}{\blacklozenge}$ $i=1$ $\chi_{\Omega_k}(p_i(t))$ (13)

Though results exist that give the "optimal" bandwidth, i.e., h, so not to over-smooth or under-smooth the data, it is convenient in this case to pick h to be the mesh size induced by the finite element discretization. We denote the numerical solutions to (10) and (11) with u_h .

We present simulation results for $N = 8 \cdot 10^4$ random walkers with $h = 0.01$ and $t \in [0, 0.5]$. To produce a, visually, more pleasing comparison between u_h and μ_N , the kernel density estimate in (13) is plotted as a continuous piecewise linear function by connecting the heights of μ_N at each of the midpoints of the subintervals Ω_i . FIG. 1

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shows results of the CTBRW simulations on (0, 1).

III. CONCLUSIONS

The results in Section II corroborate that the nonlocal boundary value problems in (10) and (11) are indeed the generalized master equations for CTBRW with appropriate boundary conditions. Consequently, a rapid means of investigating statistics of the CTBRW, e.g., exit-times, exists via finding numerical solutions to generalized master equations. and thus renders the recently developed variational formulation and numerical methods are powerful tools. Without this capability, estimating such statistics requires simulations of the CTBRW, a computationally demanding task.

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