Dynamics Reading Group Optimal paths: Revisited

> Paul Ritchie Supervisor: Jan Sieber

### Overview

Stochastic differential equation:

$$\dot{x} = f(x(t),t) + \sqrt{2D}\eta(t)$$

The optimal path is the most probable path for the transition between a given starting point  $x_0$  at time  $t_0$  to a given end position  $x_T$  at time  $T_{end}$ .



#### Limit: $\delta \ll \Delta t \ll 1$

**Optimisation problem:** Optimal path derived from optimising a functional of the probability for passing through gates along a path.

### Introduction

Probability density function P(x,t) of the random variable x(t) is governed by the Fokker-Planck equation:

$$\frac{\partial P(x,t)}{\partial t} = D \frac{\partial^2 P(x,t)}{\partial x^2} - \frac{\partial}{\partial x} (f(x,t)P(x,t))$$

where a potential U(x,t) satisfies:

$$\frac{\partial U(x,t)}{\partial x} = -f(x,t)$$

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# Introduction



Fokker-Planck run for  $\dot{x} = -1 + \eta$ ,  $x_0 = 0, T = 3$ 

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### Identities

Fourier Transform  $\hat{P}(k,t)$  of P(x,t)

$$\hat{P}(k,t) = \int_{-\infty}^{\infty} P(x,t) e^{-ikx} \mathrm{d}x$$

Dirac delta identity

$$\int_{-\infty}^{\infty} f(x)\delta(x-x_0)\mathrm{d}x = f(x_0)$$

Inverse Fourier Transform

$$P(x,t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{P}(k,t) e^{ikx} dk$$

Gaussian integral

$$\int_{-\infty}^{\infty} e^{-\alpha x^2} \mathrm{d}x = \sqrt{\frac{\pi}{\alpha}}$$

# Notation

- $t_k = t_{k-1} + \Delta t$ , for k = 1, .., N + 1
- $x_k = x(t_k)$ : Realisation of random variable x at time  $t_k$  conditioned on having passed through gates 1, ..., k 1
- $\tilde{x}_k$ : Location of path and represents centre of gate k at time  $t_k$
- $P_k(x_k)$ : Probability density function for being at  $x_k$  assuming passed through gates 1, ..., k-1 at time  $t_k$
- $\mathbb{P}_k$ : Probability of passing through gate k conditioned on having passed through gates 1,...,k-1
- $\tilde{P}_k(x_k)$ : Probability density function for being at  $x_k$  assuming passed through gates 1, ..., k at time  $t_k$
- $\mathbb{P}_k^{(T)}$ : Total probability of passing through first k gates
- $\mathbb{P} = \mathbb{P}_{N+1}^{(T)}$ : Probability of passing through all N+1 gates

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• Case 1: Pure diffusion

$$\mathbb{P}(\tilde{x}, \delta, \Delta t) = \left(\frac{\delta}{\sqrt{4\pi D\Delta t}}\right)^{N+1} \exp\left(-\frac{1}{4D} \int_{t_0}^{T_{\text{end}}} \left(\frac{\mathrm{d}\tilde{x}}{\mathrm{d}\tau}\right)^2 \mathrm{d}\tau\right)$$

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• Case 1: Pure diffusion

$$\mathbb{P}(\tilde{x},\delta,\Delta t) = \left(\frac{\delta}{\sqrt{4\pi D\Delta t}}\right)^{N+1} \exp\left(-\frac{1}{4D}\int_{t_0}^{T_{\text{end}}} \left(\frac{\mathrm{d}\tilde{x}}{\mathrm{d}\tau}\right)^2 \mathrm{d}\tau\right)$$

• Case 2: Absorbing medium

$$\mathbb{P}(\tilde{x}, \delta, \Delta t) = \left(\frac{\delta}{\sqrt{4\pi D\Delta t}}\right)^{N+1} \exp\left(-\int_{t_0}^{T_{\text{end}}} \frac{1}{4D} \left(\frac{\mathrm{d}\tilde{x}}{\mathrm{d}\tau}\right)^2 + A(\tilde{x}(\tau))\mathrm{d}\tau\right)$$

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• Case 3: Fokker-Planck equation

$$\mathbb{P} = \left(\frac{\delta}{\sqrt{4\pi D\Delta t}}\right)^{N+1} \exp\left(\frac{U(x_0) - U(x_T)}{2D} - \int_{t_0}^{T_{\text{end}}} \mathbb{L}(\tilde{x}(\tau)) \mathrm{d}\tau\right)$$

where

$$\mathbb{L}(x) = \frac{1}{4D} \left(\frac{\mathrm{d}x}{\mathrm{d}\tau}\right)^2 + V_s(x)$$

and

$$V_s(x) = \frac{1}{4D} \left(\frac{\mathrm{d}U}{\mathrm{d}x}\right)^2 - \frac{1}{2} \frac{\mathrm{d}^2 U}{\mathrm{d}x^2}$$

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To minimise  $\mathbb{L}$ , solve the Euler-Lagrange equation:

$$\frac{\partial \mathbb{L}}{\partial x} - \frac{\mathrm{d}}{\mathrm{d}\tau} \frac{\partial \mathbb{L}}{\partial \dot{x}} = 0$$

A  $2^{nd}$  order BVP is derived that the most likely trajectory will satisfy:

$$\ddot{x} = 2D \frac{\mathrm{d}V_s}{\mathrm{d}x}, \qquad \begin{cases} x(t_0) &= x_0\\ x(T_{\mathrm{end}}) &= x_T \end{cases}$$

where

$$V_s = \frac{1}{4D} \left(\frac{\mathrm{d}U}{\mathrm{d}x}\right)^2 - \frac{1}{2} \frac{\mathrm{d}^2 U}{\mathrm{d}x^2}$$

Consider the Ornstein-Uhlenbeck process:

$$\dot{x} = -ax(t) + \sqrt{2D}\eta(t)$$

Optimal path satisfies:

$$\ddot{x} = a^2 x, \qquad \begin{cases} x(t_0) &= x_0 \\ x(T_{\text{end}}) &= x_T \end{cases}$$

Solution can be obtained analytically:

$$x(t) = \frac{x_0 \sinh(a(T_{\text{end}} - t)) + x_T \sinh(a(t - t_0))}{\sinh(a(T_{\text{end}} - t_0))}$$

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# Time dependent potentials U(x,t)

New PDE is:

$$\frac{\partial P_s}{\partial t} = D \frac{\partial^2 P_s}{\partial x^2} + \left( \frac{U^{\prime\prime}}{2} - \frac{U^{\prime 2}}{4D} + \frac{\dot{U}}{2D} \right) P_s$$

 $2^{nd}$  order BVP remains the same:

$$\ddot{x} = 2D \frac{\mathrm{d}V_s}{\mathrm{d}x}$$

where

$$V_s = \frac{U'^2}{4D} - \frac{U''}{2} - \frac{\dot{U}}{2D}$$

Consider the Ornstein-Uhlenbeck process:

$$\dot{x} = -a(t)x(t) + \sqrt{2D}\eta(t)$$

where a is not constant, instead

$$a(t) = a_0 - \epsilon t$$

The optimal path satisfies:

$$\ddot{x} = a(t)^2 x + \epsilon x$$

To be solved numerically

$$a_0 = -0.2, \qquad \epsilon = -0.05$$



$$D = 0.05$$

D = 0.1

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