

Biophysics Homework II (2021-2022)

ICTP Quantitative Life Sciences Diploma

Additional theory background: numerical integration of SDEs. There exist plenty numerical integration techniques of the one-dimensional stochastic differential equation $dX = f(X, t)dt + g(X, t)dB$, whose solution is formally written as

$$X(t) = \int f(X(s), s)ds + \int g(X(s), s)dB(s). \quad (1)$$

In the computer, it is only possible to resolve SDEs in discrete time $0, \Delta t, 2\Delta t, \dots$, with Δt a finite simulation time step. Two commonly used approaches are the following [see e.g. W. Rümelin, SIAM J. Numer. Anal. 19, 604 (1982)].

- Euler scheme: Given X_t , the value of the process at $t + \Delta t$ is obtained as follows

$$X_{t+\Delta t} = X_t + f(X_t, t)\Delta t + g(X_t, t)\Delta B_t, \quad (2)$$

where $\Delta B_t \sim \mathcal{N}(0, \sqrt{\Delta t})$ is a Gaussian random number with zero mean and standard deviation equal to $\sqrt{\Delta t}$. Note that, to obtain $X_{t+\Delta t}$ in (2) one needs to know: the value of the process in the previous time step X_t , the value of the functions f and g evaluated at the previous instances t and X_t , and a random number ΔB_t that is generated anew at each time t . This is why this method converges to Eq. (1) when interpreted in the Ito sense.

- Heun scheme: Given X_t , the value of the process at $t + \Delta t$ is obtained as follows. First, one computes a prediction of the value in the next time step $\tilde{X}_{t+\Delta t}$

$$\tilde{X}_{t+\Delta t} = X_t + f(X_t, t)\Delta t + g(X_t, t)\Delta \tilde{B}_t, \quad (3)$$

with $\Delta \tilde{B}_t \sim \mathcal{N}(0, \sqrt{\Delta t})$ a Gaussian random number with standard deviation equal to $\sqrt{\Delta t}$. and then one uses both the real previous value X_t and the prediction $\tilde{X}_{t+\Delta t}$ to get the value $X_{t+\Delta t}$, as follows:

$$X_{t+\Delta t} = X_t + \frac{\Delta t}{2}[f(X_t, t) + f(\tilde{X}_{t+\Delta t}, t + \Delta t)] + \frac{1}{2}[g(X_t, t) + g(\tilde{X}_{t+\Delta t}, t + \Delta t)]\Delta B_t. \quad (4)$$

Again here $\Delta B_t \sim \mathcal{N}(0, \sqrt{\Delta t})$ which in each run may be different to $\Delta \tilde{B}_t$, i.e. one needs to produce two Gaussian random numbers to generate $X_{t+\Delta t}$ from X_t . The Heun scheme converges to the Stratonovich solution of Eq. (1) in the limit of Δt small.

1. **Active Brownian particles.** The following model has been extensively used to describe the motion of active self-propelled microswimmers (e.g. bacteria, Janus particles, etc.) in two dimensions [C Bechinger et. al, Rev. Mod. Phys. **88** (4) 045006 (2016)]

$$\dot{\theta} = \omega + \sqrt{2D_r}\xi_\theta \quad (5)$$

$$\dot{x} = v \cos(\theta) + \sqrt{2D_t}\xi_x \quad (6)$$

$$\dot{y} = v \sin(\theta) + \sqrt{2D_t}\xi_y. \quad (7)$$

Here, ω is the average angular velocity of the swimmer, v the self-propulsion speed, D_r the rotational diffusion coefficient, and D_t the translational diffusion coefficient. The three noises are zero-mean Gaussian, white, and independent to each other, each with autocorrelation function $\langle \xi_\theta(t)\xi_\theta(t') \rangle = \langle \xi_x(t)\xi_x(t') \rangle = \langle \xi_y(t)\xi_y(t') \rangle = \delta(t - t')$. For simplicity we may assume $\theta(0) = x(0) = y(0) = 0$.

(a) Derive an analytical formula for the probability density $P(\theta, t)$.

Does θ reach a stationary state?

(b) Compare the theoretical result obtained in (a) with the probability density of θ obtained from 10^3 simulations using the parameter values in (d) for observation times $t = 1, 5$, and 10 s.

(c) Using stochastic calculus derive a stochastic differential equation for the distance $r(t) = \sqrt{x^2(t) + y^2(t)}$ of the swimmer from the center (both in Ito and Stratonovich). What is the value of $\langle dr(t) \rangle$ if $D_t = 0$?

(d) Consider the parameter values $D_t = 0.2\mu\text{m}^2/\text{s}$, $D_r = 0.17\text{rad}^2/\text{s}$, $v = 30\mu\text{m}/\text{s}$, and $\omega = 10\text{rad}/\text{s}$. Using Euler's numerical simulation scheme, plot five stochastic trajectories of duration 10 s, using a simulation time step $\Delta t = 0.01$ s. Repeat the same procedure setting $\omega = 0$ and discuss the obtained result.

2. **Itô's lemma.** Use Itô calculus rules to write the following stochastic processes on the standard form

$$dX = f(X, t)dt + g(X, t) \cdot dB, \quad (8)$$

i.e. find the functions $f(X, t)$ and $g(X, t)$, for:

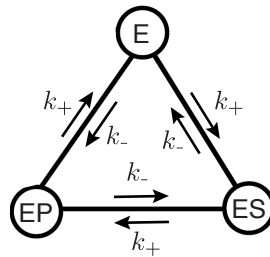
(a) $X = B^2$.

(b) $X = A \cos(\omega t + B)$.

(c) $X = B/(1 + t/\tau)$.

(d) $X = e^{-B^2/2t}$.

3. **Statistics of enzymatic reactions.** Consider a three-state continuous-time Markov jump model of the cyclic enzymatic reaction illustrated in the figure



Here the circles denote the three different states of the enzyme (E: free enzyme; ES: enzyme bound to substrate; EP: enzyme bound to product), and k_+ and k_- are respectively the clockwise and counterclockwise transition rate in the state space.

- (a) Write down the Master equation associated with the dynamics of the enzyme, both as a system of equations and also in its matrix form $d\vec{P}/dt = \mathbf{W}\vec{P}$, with \vec{P} a probability column vector.
- (b) Compute the eigenvalues and eigenvectors of the transition matrix \mathbf{W} .
Hint: find and use the known expressions for the eigenvalues/vectors of a circulant matrix.
- (c) Calculate the stationary distribution and stationary current of the system, and the characteristic relaxation times of a system that is initially out of steady state.
Hint: the relaxation times can be found from the eigenvalue spectrum of \mathbf{W} .
- (d) Evaluate, with the help of a computer, the value of the three components of the probability vector $\vec{P}(t)$ describing the probability of an ensemble of enzymes to be in any of the three possible states. Assume that the enzymes are initially free, i.e. at time $t = 0$ the system is in state E with probability one. Compare the results of two different methods: (i) numerical integration of the system of ordinary differential equations given by the Master equation; (ii) performing the average over many (at least 10^3) stochastic trajectories obtained using the Gillespie algorithm. Plot few trajectories obtained from the Gillespie algorithm method.
- Values of the parameters $k_+ = 10\text{Hz}$, $k_- = 1\text{Hz}$, total integration time $t_{\text{max}} = 3\text{s}$.

4. **Four-state model of molecular motor.** Experimentalists sketched a Markov jump model of molecular motor stepping:

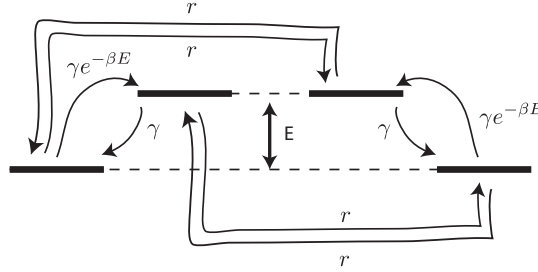


Figure 1: Sketch of the model.

Here, r and γ are two rate parameters, $\beta = 1/k_B T$ and E is the energy difference between two adjacent energy levels, and T is the temperature of the environment.

- (a) Write the Master equation of the model in matrix form.
- (b) Calculate the probability to be in any of the state in the long time limit, i.e. its stationary value. Compare this result with the associated Boltzmann distribution.
- (c) What is the stationary current between each of the states? Sketch the direction of the current between all the states. For which value of E is the current equal to zero?
5. The dynamics of a photoreceptor neuron subject to a light source of frequency ω can be modelled as a two-state continuous-time Markov process with *time-dependent* transition rates

$$\omega_{21}(t) = \mu \cos^2(\omega t), \quad (9)$$

$$\omega_{12}(t) = \mu \sin^2(\omega t), \quad (10)$$

where $\mu > 0$ is a characteristic rate parameter.

- (a) Plot the transition rates as a function of time, for a given value of ω . What happens if the frequency of the incident light is doubled?
- (b) Write the Master equation describing the dynamics of the probabilities $P_1(t)$ and $P_2(t)$ for the neuron to be at state 1 and 2 at time t , respectively.
- (c) Derive analytical expressions for $P_1(t)$ and $P_2(t)$.
- (d) Evaluate $P_1(t)$ and $P_2(t)$ in the limit of t large, i.e. when $t \gg \mu^{-1}$.

Deadline: 7th January, 2021.