

Foundations of Multiscale Modelling: Kinetics
(Seminars/Labs)
Lesson 1

Feb, 2022

Langevin Equation and Brownian Dynamics (5 points)

Stochastic differential equation

$$M \frac{dv}{dt} = -\gamma v + F(t),$$

where $F(t)$ is the random force (Gaussian noise):

$$\langle F(t) \rangle = 0, \quad \langle F(t)F(t') \rangle = \Gamma \delta(t - t'), \quad \Gamma = 2k_B T \gamma.$$

(a) Find an analytic expression for the VACF:

$$K_v(t) = \frac{\langle v(0)v(t) \rangle}{\langle v^2(0) \rangle}.$$

- An ensemble of different realizations E of $v(t, E)$ with $F(t, E)$;
- the average $\langle \dots \rangle$ is carried out over all realizations E .

Box-Muller transform

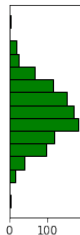
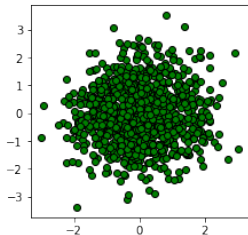
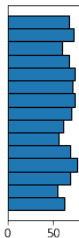
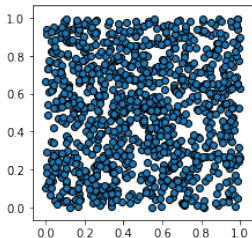
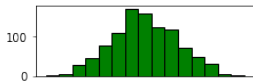
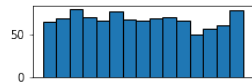
(b)–(c) Produce $\xi \sim \mathcal{N}(0, 1)$ from a uniform distribution.

Lemma

If $U_1, U_2 \sim \mathcal{U}(0, 1)$ are independent, then

$$Y_1 = \sqrt{(-2 \ln U_1)} \cos(2\pi U_2), \quad Y_2 = \sqrt{(-2 \ln U_1)} \sin(2\pi U_2)$$

are independent and $Y_1, Y_2 \sim \mathcal{N}(0, 1)$.



(d) Write a program that simulates the Langevin equation using

$$v_{n+1} - v_n = -\frac{\gamma}{M} v_n \Delta t + \frac{1}{M} (\Gamma \Delta t)^{1/2} \xi_n.$$

(e) Calculate the VACF and compare it with the analytic result (a) by putting them on one plot.

- For simplicity, assume $\gamma = 1$, $M = 1$ and $k_B T = 1$;
- in python, for ξ_n use *random.normal* from *numpy*;
- you should gain a good agreement for the VACF $K_v(t)$ in (e).

Generalized Langevin Equation* (2 points)

Assuming that the friction depends on the history, we have

$$M \frac{dv}{dt} = - \int_0^t G(t - \tau) v(\tau) d\tau + F(t),$$

where the random force $F(t)$ **is not** a Gaussian white noise:

$$\langle F(t) \rangle = 0, \quad \langle F(t) F(t') \rangle = 2k_B T G(t - t').$$

(a) Program the finite difference scheme for the equation and calculate the VACF. Consider two cases,

$$G(t) = e^{-a|t|}, \quad G(t) = e^{-b|t|^2}.$$

- choose $M = 1$, $k_B T = 0.5$, and a and b from your own choice;
- the convolution integral can be represented as a sum over already obtained history of the system.

How to simulate the colored noise $\xi(t) \equiv F(t)$ properly?

An exponentially correlated noise,

$$\langle \xi(t) \xi(t') \rangle = e^{-a|t-t'|},$$

can be obtained from the following equation

$$\frac{d\xi}{dt} = -a\xi + \sqrt{2a} \eta(t).$$

Here, $\eta(t)$ is a Gaussian white noise,

$$\langle \eta(t) \eta(t') \rangle = \delta(t - t').$$

- program the finite-difference scheme for the system of two equations (generate ξ_n from the colored noise equation);
- this equation does not depend on the velocity $v(t)$.

Colored noise: general approach

The colored noise $\xi(t)$,

$$\langle \xi(t) \rangle = 0, \quad \langle \xi(t) \xi(t') \rangle = G(t - t'),$$

can be discretized to $\boldsymbol{\xi} = (\xi_1, \dots, \xi_n)^T$, where

$$\xi_i = \xi(t_i), \quad \langle \xi_i \rangle = 0, \quad \langle \xi_i \xi_j \rangle = G_{ij}.$$

Multivariate normal distribution with the covariance matrix

$$\Sigma = (G_{ij})_{i,j=1}^n \in \mathbb{R}^{n \times n}, \quad \Sigma \equiv \langle \boldsymbol{\xi} \boldsymbol{\xi}^T \rangle, \quad \Sigma^T = \Sigma > 0.$$

To generate $\boldsymbol{\xi}$, apply the following method:

- 1 find the Cholesky decomposition: $C^T C = \Sigma$;
- 2 generate $\mathbf{x} = (x_1, \dots, x_n)$, $x_i \sim \mathcal{N}(0, 1)$ are i.i.d.;
- 3 make linear transformation: $\boldsymbol{\xi} = C^T \mathbf{x}$.

(b) For the case $G(t) = e^{-a|t|}$ find an analytic solution for the VACF:

$$M \frac{d}{dt} K_v(t) = - \int_0^t G(t - \tau) K_v(\tau) d\tau.$$

Compare the VACF with the simulations result (a) by putting them on one plot.

- Use the Laplace transform

$$\mathcal{L}\{f\}(s) = \int_0^\infty f(t) e^{-st} dt$$

to reduce the equation to an algebraic form;

- use the inverse Laplace transform to get the result in real time;

Fluctuation-Dissipation Theorem (2 points)

Assume that the quantities A and B are correlated as

$$\langle A(0)B(t) \rangle = Ce^{-|t|/\tau}.$$

Find the imaginary part of the linear response function $\chi_{AB}(\omega)$.

How would you describe the physical meaning of this function?

- Use lecture notes.



pythonTM



Upload here:

<https://skoltech.instructure.com/login/canvas>

<https://jupyter.org>

Jupyter advantages:

- code can be divided into several problems within *one* file (cells);
- *your* runs results are presented above the code (graphs, etc.);
- scans of analytics can be attached in ipynb as jpg/png files;
- analytic expressions can be written directly using LaTeX.

Telegram chat for the course (to be in touch):

