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Stochastic Processes: Data Analysis and Computer Simulation

The course deals with how to simulate and analyze stochastic processes, in particular the dynamics of small particles diffusing in a fluid.



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What you'll learn

- Basic Python programming
- Basic theories of stochastic processes
- Done
 Jimulation methods for a provincin particle
- Application: analysis of financial data

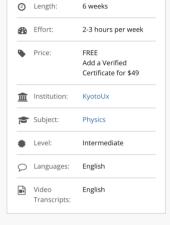
Meet the instructors



Ryoichi Yamamoto Professor, Chemical Engineering Kyoto University



John J. Molina Assistant Professor Kyoto University



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Prerequisites

Basic mathematics expected of a 2nd year undergraduate student (differential and integral calculus and linear algebra).

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Brownian motion 1: basic theories Basic knowledge of stochastic process

Stochastic Processes: Data Analysis and Computer Simulation

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Stochastic process

A deterministic process:

$$X(t) = Func(t)$$

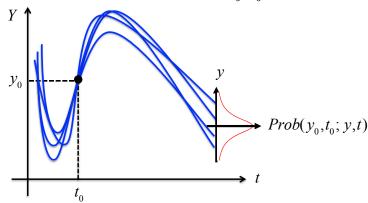
$$X \downarrow x_0 \downarrow x$$

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A stochastic process:

 $Y(t) \neq Func(t) \rightarrow Prob(y_0, t_0; y, t)$

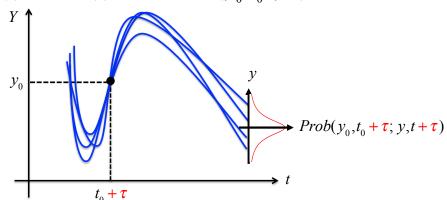


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Stochastic process

A stochastic process:

$$Y(t) \neq Func(t) \rightarrow Prob(y_0, t_0; y, t)$$



A steady stochastic process:

$$Prob(y_0, t_0 + \tau; y, t + \tau) = Prob(y_0, t_0; y, t)$$

Stochastic Processes: Data Analysis and Computer Simulation

Consider a steady stochastic process Y(t) with its mean $\langle Y(t) \rangle = 0$ and define Fourier transformation

$$\tilde{Y}_{T}(\omega) = \int_{-\infty}^{\infty} dt \, e^{i\omega t} Y_{T}(t) \tag{1}$$

and inverse Fourier transformation

$$Y_{T}(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega \, e^{-i\omega t} \tilde{Y}_{T}(\omega) \tag{2}$$

using

$$Y_{T}(t) = Y(t) \quad (|t| \le T/2)$$

$$Y_{T}(t) = 0 \quad (|t| > T/2)$$
(3)

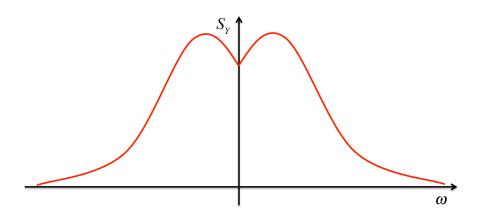
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Stochastic process

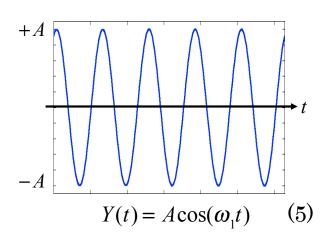
Spectral density / Power spectrum

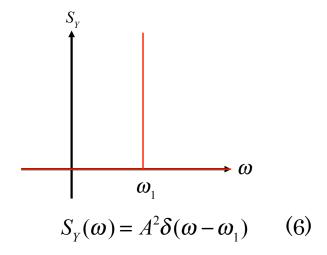
$$S_{Y}(\omega) = \lim_{T \to \infty} \frac{1}{T} \left| \tilde{Y}_{T}(\omega) \right|^{2} \tag{4}$$



Stochastic Processes: Data Analysis and Computer Simulation

Case 1: Single cosine wave



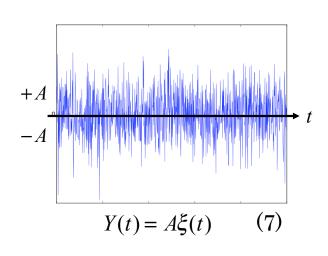


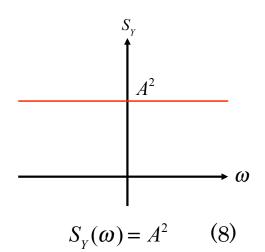
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Stochastic process

Case 2: White noise





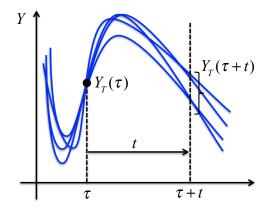
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Auto-correlation function

$$\varphi_{Y}(t) \equiv \lim_{T \to \infty} \frac{1}{T} \int_{-\infty}^{\infty} d\tau \, Y_{T}(\tau) Y_{T}(\tau + t) \equiv \left\langle Y(\tau) Y(\tau + t) \right\rangle_{\tau} \tag{9}$$

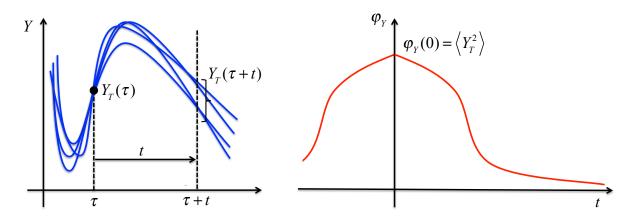


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Stochastic process

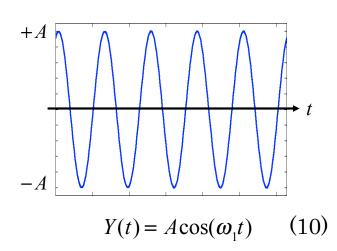
Auto-correlation function

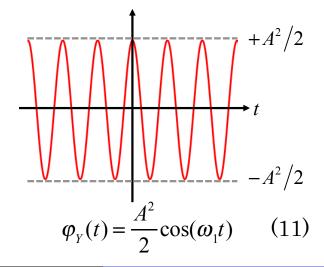
$$\varphi_{Y}(t) \equiv \lim_{T \to \infty} \frac{1}{T} \int_{-\infty}^{\infty} d\tau \, Y_{T}(\tau) Y_{T}(\tau + t) \equiv \left\langle Y(\tau) Y(\tau + t) \right\rangle_{\tau} \tag{9}$$



Stochastic Processes: Data Analysis and Computer Simulation

Case 1: Single cosine wave





Stochastic Processes: Data Analysis and Computer Simulation

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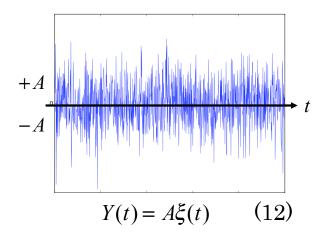
11

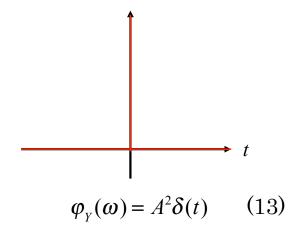
Stochastic process

$$\begin{split} \varphi_{Y}(t) &= \lim_{T \to \infty} \frac{A^{2}}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} d\tau \cos(\omega_{1}\tau) \cos(\omega_{1}(\tau+t)) \\ &= \lim_{T \to \infty} \frac{A^{2}}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} d\tau \sin\left(\omega_{1}\tau + \frac{\pi}{2}\right) \sin\left(\omega_{1}(\tau+t) + \frac{\pi}{2}\right) \\ &= \lim_{T \to \infty} \frac{A^{2}}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} d\tau \sin\left(\omega_{1}\tau + \frac{\pi}{2}\right) \left[\sin\left(\omega_{1}\tau + \frac{\pi}{2}\right) \cos(\omega_{1}t) + \cos\left(\omega_{1}\tau + \frac{\pi}{2}\right) \sin(\omega_{1}t)\right] \\ &= \lim_{T \to \infty} \frac{A^{2}}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} d\tau \left[\sin^{2}\left(\omega_{1}\tau + \frac{\pi}{2}\right) \cos(\omega_{1}t) + \sin\left(\omega_{1}\tau + \frac{\pi}{2}\right) \cos\left(\omega_{1}\tau + \frac{\pi}{2}\right) \sin(\omega_{1}t)\right] \\ &= \cos\left(\omega_{1}t\right) \lim_{T \to \infty} \frac{A^{2}}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} d\tau \sin^{2}\left(\omega_{1}\tau + \frac{\pi}{2}\right) + \sin\left(\omega_{1}t\right) \lim_{T \to \infty} \frac{A^{2}}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} d\tau \sin\left(\omega_{1}\tau + \frac{\pi}{2}\right) \cos\left(\omega_{1}\tau + \frac{\pi}{2}\right) \\ &= \frac{A^{2}}{2} \cos\left(\omega_{1}t\right) + 0 \quad \cdots \quad \text{Eq.} (11) \end{split}$$

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Case 2: White noise





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Stochastic process

From Eq.(9),

$$\varphi_{Y}(t) = \lim_{T \to \infty} \frac{1}{T} \int_{-\infty}^{\infty} d\tau \left[Y_{T}(\tau) \left[\frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega \, e^{-i\omega(\tau + t)} \tilde{Y}_{T}(\omega) \right] \right] \\
= \lim_{T \to \infty} \frac{1}{2\pi T} \int_{-\infty}^{\infty} d\omega \left[e^{-i\omega t} \tilde{Y}_{T}(\omega) \int_{-\infty}^{\infty} d\tau \left[e^{-i\omega \tau} Y_{T}(\tau) \right] \right] \\
= \lim_{T \to \infty} \frac{1}{2\pi T} \int_{-\infty}^{\infty} d\omega \left[e^{-i\omega t} \tilde{Y}_{T}(\omega) \tilde{Y}_{T}^{*}(\omega) \right] \\
= \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega \, e^{-i\omega t} \left[\lim_{T \to \infty} \frac{1}{T} \left| \tilde{Y}_{T}(\omega) \right|^{2} \right] = \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega \, e^{-i\omega t} S_{Y}(\omega) \tag{14}$$

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And also,

$$S_{Y}(\omega) = \int_{-\infty}^{\infty} dt \, e^{i\omega t} \varphi_{Y}(t) \tag{15}$$

Wiener-Khintchine theorem:

$$\varphi_{Y}(t) = \frac{\text{inverse Foulier Eq.(14)}}{\text{Foulier Eq.(15)}} S_{Y}(\omega)$$

Sum rules:

$$\varphi_{Y}(0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega S_{Y}(\omega)$$
 (16)

$$S_{Y}(0) = \int_{-\infty}^{\infty} dt \, \varphi_{Y}(t) \tag{17}$$

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Brownian motion 1: basic theories Brownian motion and the Langevin equation

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Brownian motion and the Langevin equation

Equation of motion of a Brownian particle:

Particle radius: a
Particle mass: m

Solvent viscosity: η

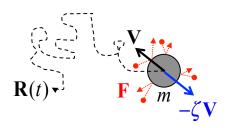
Friction constant: $\zeta = 6\pi\eta a$

Particle position: $\mathbf{R}(t)$

Particle velocity: $V(t) = d\mathbf{R}/dt$

Friction force: $-\zeta V(t)$

Random force: $\mathbf{F}(t)$



Langevin Equation:

$$m\frac{d\mathbf{V}}{dt} = -\zeta\mathbf{V} + \mathbf{F} \tag{21}$$

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Random force:

$$\mathbf{F}(t) = \left(F_x(t), F_y(t), F_z(t)\right)$$

White noise:

$$\langle F_{\alpha}(t) \rangle = 0$$

$$\varphi_{F}(t) \equiv \langle F_{\alpha}(\tau) F_{\beta}(\tau + t) \rangle = 2\tilde{D}\delta_{\alpha\beta}\delta(t)$$

$$\text{where } \alpha, \beta \in x, y, z, \text{ and }$$

$$\delta_{\alpha\beta} = 1 \quad (\alpha = \beta), \quad \delta_{\alpha\beta} = 0 \quad (\alpha \neq \beta)$$

$$(22)$$

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 $\delta(t) = \infty \quad (t = 0), \quad \delta(t) = 0 \quad (t \neq 0)$

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Brownian motion and the Langevin equation

Power spectrum of random force $\mathbf{F}(t)$:

$$S_{F}(\omega) = \lim_{T \to \infty} \frac{1}{T} |\tilde{\mathbf{F}}_{T}(\omega)|^{2}$$

$$= \int_{-\infty}^{\infty} dt \phi_{F}(t) e^{i\omega t}$$

$$= \int_{-\infty}^{\infty} dt \langle \mathbf{F}(\tau) \cdot \mathbf{F}(\tau + t) \rangle e^{i\omega t}$$

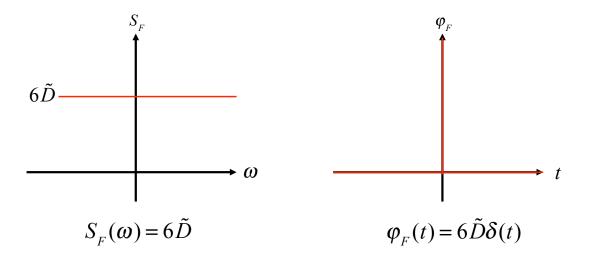
$$= \int_{-\infty}^{\infty} dt 6\tilde{D}\delta(t) e^{i\omega t}$$

$$= 6\tilde{D}$$
(24)

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Property of random force $\mathbf{F}(t)$:



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Brownian motion and the Langevin equation

Fourier transform Eq.(1)

$$-i\omega m \tilde{\mathbf{V}}_{T}(\omega) = -\zeta \tilde{\mathbf{V}}_{T}(\omega) + \tilde{\mathbf{F}}_{T}(\omega)$$
$$\tilde{\mathbf{V}}_{T}(\omega) = \frac{\tilde{\mathbf{F}}_{T}(\omega)}{-i\omega m + \zeta}$$

Power Spectrum of particle velocity V(t):

$$S_{V}(\boldsymbol{\omega}) = \lim_{T \to \infty} \frac{1}{T} |\tilde{\mathbf{V}}_{T}(\boldsymbol{\omega})|^{2}$$

$$= \lim_{T \to \infty} \frac{1}{T} |\tilde{\mathbf{F}}_{T}(\boldsymbol{\omega})|^{2} \frac{1}{m^{2} \boldsymbol{\omega}^{2} + \zeta^{2}} = \frac{6\tilde{D}}{m^{2} \boldsymbol{\omega}^{2} + \zeta^{2}}$$
(25)

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Auto-correlation function of particle velocity V(t):

Using Winner-Khintchine theorem and Eq.(6)

$$\varphi_{V}(t) = \langle \mathbf{V}(\tau) \cdot \mathbf{V}(\tau + t) \rangle = \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega S_{V}(\omega) e^{-i\omega t}$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega \frac{6\tilde{D}}{m^{2}\omega^{2} + \zeta^{2}} e^{-i\omega t}$$

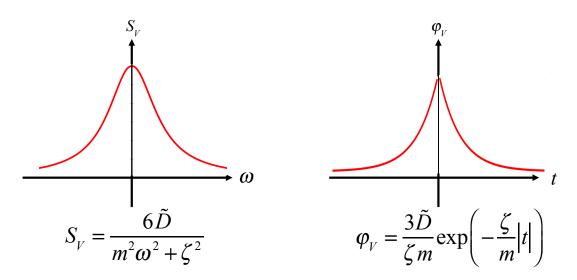
$$= \frac{3\tilde{D}}{\zeta m} \exp\left(-\frac{\zeta}{m}|t|\right)$$
(26)

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Brownian motion and the Langevin equation

Property of particle velocity V(t):



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From Eq.(26)
$$\varphi_V(t=0) = \langle \mathbf{V}^2 \rangle = \frac{3\tilde{D}}{\zeta m}$$

Equipartition of energy $\langle \mathbf{V}^2 \rangle = \frac{3k_B T}{m}$

$$\therefore \frac{3\tilde{D}}{\zeta_m} = \frac{3k_B T}{m} \quad \to \quad \boxed{\tilde{D} = k_B T \zeta} \tag{29}$$

Fluctuation-dissipation theorem

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Brownian motion and the Langevin equation

Displacement:
$$\Delta \mathbf{R}(t) \equiv \mathbf{R}(t) - \mathbf{R}(0) = \int_0^t \mathbf{V}(t_1) dt_1$$

Mean square displacement:

$$\left\langle \left| \Delta \mathbf{R}(t) \right|^{2} \right\rangle = \int_{0}^{t} dt_{1} \int_{0}^{t} dt_{2} \left\langle \mathbf{V}(t_{1}) \cdot \mathbf{V}(t_{2}) \right\rangle$$

$$= \int_{0}^{t} dt_{1} \int_{0}^{t} dt_{2} \frac{3\tilde{D}}{\zeta} \exp\left(-\frac{\zeta}{m} |t_{2} - t_{1}|\right)$$

$$= 2 \int_{0}^{t} dt_{1} \int_{t_{1}}^{t} dt_{2} \frac{3\tilde{D}}{\zeta} \exp\left(-\frac{\zeta}{m} (t_{2} - t_{1})\right) = \frac{6\tilde{D}}{\zeta^{2}} t + Cons.$$

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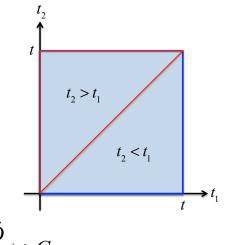
Displacement:
$$\Delta \mathbf{R}(t) \equiv \mathbf{R}(t) - \mathbf{R}(0) = \int_0^t \mathbf{V}(t_1) dt_1$$

Mean square displacement:

$$\left\langle \left| \Delta \mathbf{R}(t) \right|^{2} \right\rangle = \int_{0}^{t} dt_{1} \int_{0}^{t} dt_{2} \left\langle \mathbf{V}(t_{1}) \cdot \mathbf{V}(t_{2}) \right\rangle$$

$$= \int_{0}^{t} dt_{1} \int_{0}^{t} dt_{2} \frac{3\tilde{D}}{\zeta} \exp\left(-\frac{\zeta}{m} |t_{2} - t_{1}|\right)$$

$$= 2 \int_{0}^{t} dt_{1} \int_{t_{1}}^{t} dt_{2} \frac{3\tilde{D}}{\zeta} \exp\left(-\frac{\zeta}{m} (t_{2} - t_{1})\right) = \frac{6\tilde{D}}{\zeta^{2}} t + Cons.$$



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Brownian motion and the Langevin equation

Self diffusion constant:

$$D \equiv \lim_{t \to \infty} \frac{\left\langle \left| \Delta \mathbf{R}(t) \right|^2 \right\rangle}{6t} = \frac{\tilde{D}}{\zeta^2}$$
 (30)

Einstein relation: (from Eq.(29) and (30))

$$D = \frac{k_B T}{\zeta} \tag{31}$$

Stokes-Einstein relation: (from Eq.(31) and Stokes law $\zeta = 6\pi a\eta$)

$$D = \frac{k_B T}{6\pi a\eta} \tag{32}$$

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Brownian motion 1: basic theories

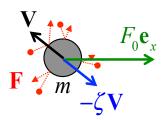
Linear response theory and the Green-Kubo formula

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Linear response theory and the G-K formula

A Brownian particle under the external force $\mathbf{F}_{ext} = F_0 \mathbf{e}_x$



Langevin Equation:

$$m\frac{d\mathbf{V}}{dt} = -\zeta \mathbf{V} + \mathbf{F} + F_0 \mathbf{e}_x \tag{41}$$

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Steady state average under external force, $\lim_{t\to\infty}\langle\cdots\rangle_{ext}$

$$\lim_{t \to \infty} \left\langle \frac{d\mathbf{V}}{dt} \right\rangle_{ext} = (0,0,0)$$

$$\lim_{t \to \infty} \left\langle \mathbf{V} \right\rangle_{ext} = \left(\lim_{t \to \infty} \left\langle V_x \right\rangle_{ext}, 0,0\right)$$

$$\lim_{t \to \infty} \left\langle \mathbf{F} \right\rangle_{ext} = (0,0,0)$$

$$\lim_{t \to \infty} \left\langle F_0 \mathbf{e}_x \right\rangle_{ext} = (F_0,0,0)$$

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Linear response theory and the G-K formula

Thus, the steady drift velocity:

$$\lim_{t \to \infty} \left\langle V_x \right\rangle_{ext} = \frac{F_0}{\zeta} = \frac{DF_0}{k_B T} \tag{42}$$

Here we used the Einstein relation Eq. (31) and finally:

$$D = \lim_{t \to \infty} \left\langle V_x \right\rangle_{ext} \frac{k_B T}{F_0} \tag{43}$$

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The linear response theory (LRT):

References:

- Barrat and Hansen "Basic concepts for simple and complex liquids" (Cambridge, 2003)
- Zwanzig "Non-equilibrium statistical mechanics" (Oxford, 2001)

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Linear response theory and the G-K formula

The linear response theory (LRT):

References:

- Barrat and Hansen "Basic concepts for simple and complex liquids" (Cambridge, 2003)
- Zwanzig "Non-equilibrium statistical mechanics" (Oxford, 2001)

 H_0 : Equilibrium Hamiltonian

 $H_0 + H'(t)$: Hamiltonian under external force F(t)

conjugate with A, $H'(t) \equiv -AF(t)$

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The linear response theory (LRT):

References:

- Barrat and Hansen "Basic concepts for simple and complex liquids" (Cambridge, 2003)
- Zwanzig "Non-equilibrium statistical mechanics" (Oxford, 2001)

: Equilibrium Hamiltonian H_{0}

 $H_0 + H'(t)$: Hamiltonian under external force F(t)

conjugate with A, $H'(t) \equiv -AF(t)$

 $\langle B(t) \rangle_{H_0} \equiv B_0$: Average value of *B* at equilib. under H_0

 $\langle B(t) \rangle_{H_0 + H'(t)} \equiv B_0 + \langle \Delta B(t) \rangle_{H_0 + H'(t)}$

: Average value of B at t under $H_0 + H'(t)$

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Linear response theory and the G-K formula

For a small external force with H'(t) = -AF(t), the time evolution of *B* is determined within LRT as:

$$\left\langle \Delta B(t) \right\rangle_{H_0 + H'} = \int_{-\infty}^t ds \, \Phi_{BA}(t - s) F(s)$$
 (44)

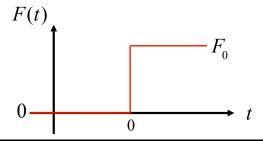
Here $\Phi_{RA}(t)$ is the response function, which is defined as the cross correlation function of $\dot{A} \equiv \frac{dA}{dt}$ and B at equilibrium:

$$\Phi_{BA}(t) = \frac{1}{k_{R}T} \left\langle B(\tau + t)\dot{A}(\tau) \right\rangle_{H_{0}}$$
(45)

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Apply LRT to define self-diffusion constant *D* using equilibrium correlation function. We assume:

$$\begin{split} A(t) &\equiv R_x(t) \,, \quad B(t) \equiv V_x(t) \\ F(t) &= \Theta(t) \,, \quad H'(t) = -AF(t) = -R_x F_0 \Theta(t) \end{split}$$



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Linear response theory and the G-K formula

From LRT Eqs. (44) and (45):

$$\left\langle \Delta B(t) \right\rangle_{H_{0}+H'} = \left\langle V_{x}(t) \right\rangle_{H_{0}+H'} = \frac{F_{0}}{k_{B}T} \int_{0}^{t} ds \left\langle V_{x}(\tau+t-s)V_{x}(\tau) \right\rangle_{H_{0}}$$

$$= \frac{F_{0}}{k_{B}T} \int_{t}^{0} dt' \frac{ds}{dt'} \left\langle V_{x}(\tau+t')V_{x}(\tau) \right\rangle_{H_{0}} = \frac{F_{0}}{k_{B}T} \int_{0}^{t} dt' \left\langle V_{x}(\tau+t')V_{x}(\tau) \right\rangle_{H_{0}}$$

$$\left\langle V_{x} \right\rangle_{H_{0}+H'} = \frac{F_{0}}{3k_{B}T} \int_{0}^{t} dt' \left\langle \mathbf{V}(\tau+t') \cdot \mathbf{V}(\tau) \right\rangle_{H_{0}}$$

$$= \frac{F_{0}}{3k_{B}T} \int_{0}^{t} dt' \left\langle \mathbf{V}(\tau+t') \cdot \mathbf{V}(\tau) \right\rangle_{H_{0}}$$

$$(t' \equiv t-s)$$

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From Eqs. (43) and (46):

$$D = \lim_{t \to \infty} \left\langle V_x(t) \right\rangle_{H_0 + H'(t)} \frac{k_B T}{F_0}$$
$$= \frac{1}{3} \int_0^\infty dt' \left\langle \mathbf{V}(\tau + t') \cdot \mathbf{V}(\tau) \right\rangle_{H_0}$$

$$D = \frac{1}{3} \int_0^\infty dt \varphi_V(t) \tag{47}$$

(Green-Kubo formula for *D*)

Stochastic Processes: Data Analysis and Computer Simulation

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Help

G Outline > Week 3 > Homework 3 > Homework 3

Homework 3

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Homework 3-1

1.0 point possible (graded)

Calculate the auto-correlation function $\varphi_Y(t)$ for a dynamic process $Y(t) = A\sin(\omega_1 t + \pi)$. Choose the correct answer from the following choices.

$$\bigcirc \ arphi_Y(t) = rac{A^2}{2} \mathrm{cos}(\omega_1 t)$$

$$\bigcirc \ arphi_Y(t) = rac{A^2}{2} \mathrm{sin}(\omega_1 t)$$

$$\bigcirc \ arphi_Y(t) = rac{A^2}{2} \mathrm{cos}(\omega_1 t + \pi)$$

$$\bigcirc \ arphi_Y(t) = rac{A^2}{2} \mathrm{sin}(\omega_1 t + \pi)$$

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You have used 0 of 2 attempts

Homework 3-2

1.0 point possible (graded)

Calculate the auto-correlation function $\varphi_Y(t)$ for a dynamic process $Y(t) = A\cos(\omega_1 t) + B\xi(t)$, where $\xi(t)$ is the White noise introduced in Part 1. Choose the correct answer from the following choices.

$$\bigcirc \ arphi_Y(t) = rac{A^2}{2} \mathrm{cos}(\omega_1 t) + AB\sqrt{\mathrm{cos}(\omega_1 t)\delta(t)} + B^2\delta(t)$$

$$\bigcirc \ arphi_Y(t) = rac{A^2}{2} \mathrm{cos}(\omega_1 t) + AB\sqrt{2 \, \mathrm{cos}(\omega_1 t) \delta(t)} + B^2 \delta(t)$$

$$\bigcirc \ arphi_Y(t) = rac{A^2B^2}{2} {
m cos}(\omega_1 t) \delta(t)$$

$$\bigcirc \ arphi_Y(t) = rac{A^2}{2} \mathrm{cos}(\omega_1 t) + B^2 \delta(t)$$

$$\bigcirc \ arphi_Y(t) = rac{A^2}{2} \mathrm{cos}(\omega_1 t) - B^2 \delta(t)$$

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You have used 0 of 2 attempts

Homework 3-3

2.0 points possible (graded)

Estimate the diffusion constant D of spherical particles with radius $a=1\mu$ m immersed in water at T=300K using the Stokes-Einstein relation (Eq.(32))

$$D=rac{k_BT}{6\pi a\eta}$$

and the following parameters

- Viscosity of water at room temperature: $\eta = 0.85 imes 10^{-3}$ Pa•s
- ullet The Boltzmann constant: $k_B=1.38064852 imes 10^{-23}$ J·K $^{-1}$

Choose the value closest to your answer from the following choices.

- $\bigcirc 2.6 \times 10^{-7} \text{m}^2 \cdot \text{s}^{-1}$
- $0.6 \times 10^{-13} \,\mathrm{m}^2 \cdot \mathrm{s}^{-1}$
- $0.6 \times 10^{-19} \,\mathrm{m}^2 \cdot \mathrm{s}^{-1}$
- $\bigcirc 2.6 \times 10^{-7} \text{m}^{-2} \cdot \text{s}$
- $\bigcirc \qquad \qquad \textbf{2.6} \times \textbf{10}^{-13} \text{m}^{-2} \cdot \text{s}$
- $0.6 \times 10^{-19} \text{m}^{-2} \cdot \text{s}$

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You have used 0 of 2 attempts

Homework 3-4

1.0 point possible (graded)

Calculate the right-hand-side of Eq.(47) using the correlation function given in Eq.(26).

$$arphi_V(t) = rac{3 ilde{D}}{\zeta m} {
m exp}igg(-rac{\zeta}{m}|t|igg)$$

$$D=rac{1}{3}\int_0^\infty \mathrm{d}t \ arphi_V(t)$$

Choose the correct result for $oldsymbol{D}$ from the following choices.

 \bigcirc $ilde{m{D}}$

 \bigcirc 6 $ilde{D}t$

 $\int \frac{\tilde{D}}{k_B T C}$

 $\frac{m\hat{L}}{\zeta}$

 $\bigcirc \frac{\tilde{D}}{\zeta^2}$

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You have used 0 of 2 attempts

Homework 3-5

2.0 points possible (graded)

Replace $F_0 o 2F_0$ in Eq.(41), then redo all the calculations to derive the equation corresponding to equation Eq.(47). Choose the correct equation, relating the diffusion constant D to the velocity auto-correlation funciton $\varphi_V(t)$, from the following choices.

- $\bigcirc~D=rac{2}{3}\int_0^\infty \mathrm{d}t~arphi_V(t)$
- $\bigcirc \ D = rac{\sqrt{2}}{3} \int_0^\infty \mathrm{d}t \ arphi_V(t)$
- $\bigcirc \ D = rac{1}{3} \int_0^\infty \mathrm{d}t \ arphi_V(t)$
- $\bigcirc~D=rac{1}{3\sqrt{2}}\int_0^\infty \mathrm{d}t~arphi_V(t)$
- $\bigcirc \ D = \frac{1}{6} \int_0^\infty \mathrm{d}t \ \varphi_V(t)$

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You have used 0 of 2 attempts

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