

# Stochastic Equations and Processes in physics and biology

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## Stochastic differential equations

# The Wiener-Khinchin theorem

**ACF of a real-valued signal  $x(t)$**

$$G(\tau) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T dt x(t)x(t + \tau),$$

**Forward Fourier transform of  $x(t)$**

$$\hat{x}(\omega) = \int_0^T dt e^{-i\omega t} x(t).$$

**Symmetry of  $\hat{x}(\omega)$**

$$\hat{x}(-\omega) = \hat{x}(\omega)^*$$

**Proof of symmetry**

$$\hat{x}(-\omega) = \int_0^T dt e^{i\omega t} x(t) = \left( \int_0^T dt e^{-i\omega t} x(t) \right)^* = \hat{x}(\omega)^*$$

# The Wiener-Khinchin theorem

**Power spectral density  $S(\omega)$  (psd)**

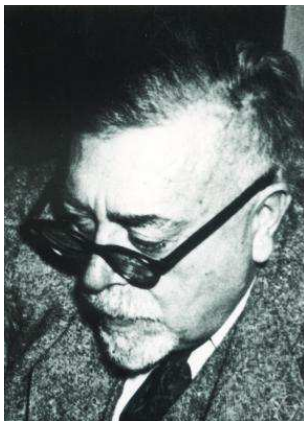
$$S(\omega) = \lim_{T \rightarrow \infty} \frac{1}{2\pi T} |\hat{x}(\omega)|^2,$$

**Wiener-Khinchin theorem**

$$S(\omega) = \frac{1}{2\pi} \lim_{T \rightarrow \infty} \int_{-T}^T d\tau e^{-i\omega\tau} G(\tau)$$

# Norbert Wiener

American mathematician (1894-1964) proved the theorem in 1930 for a deterministic process  $x(t)$



# Aleksandr Yakovlevich Khinchin

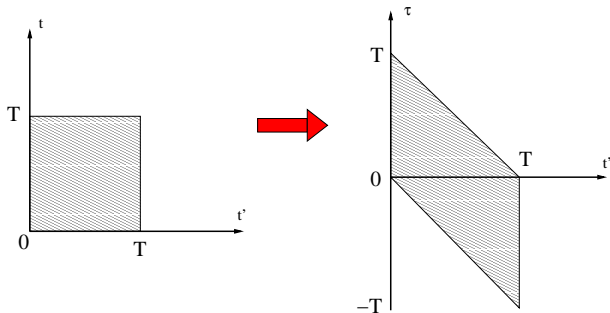
Russian mathematician (1894-1959) proved the theorem in 1934 for a stochastic process  $x(t)$



# Wiener-Khinchin theorem

## Proof

$$\begin{aligned} S(\omega) &= \lim_{T \rightarrow \infty} \frac{1}{2\pi T} \int_0^T dt e^{-i\omega t} x(t) \int_0^T dt' e^{i\omega t'} x^*(t') \\ &= \{(t, t') \rightarrow (t', \tau = t - t')\} \end{aligned}$$



## Proof continued

$$\begin{aligned} S(\omega) &= \lim_{T \rightarrow \infty} \frac{1}{2\pi T} \left[ \int_0^T d\tau e^{-i\omega\tau} \int_0^{T-\tau} dt' x(t') x^*(t' + \tau) \right. \\ &\quad \left. + \int_{-T}^0 d\tau e^{-i\omega\tau} \int_{-\tau}^T dt' x(t') x^*(t' + \tau) \right] \\ &= \frac{1}{2\pi T} \left[ \int_{-T}^0 d\tau e^{-i\omega\tau} (TG(\tau) + O(1)) \right. \\ &\quad \left. + \int_0^T d\tau e^{-i\omega\tau} (TG(\tau) + O(1)) \right] \\ &= \frac{1}{2\pi} \lim_{T \rightarrow \infty} \int_{-T}^T d\tau e^{-i\omega\tau} G(\tau) \end{aligned}$$



# Wiener-Khinchin theorem

Using  $G(-\tau) = G(\tau)$

$$\begin{aligned} S(\omega) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} G(\tau) e^{\pm i\omega\tau} d\tau \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} G(\tau) \cos \omega\tau d\tau \\ &= \frac{1}{\pi} \int_0^{\infty} G(\tau) \cos \omega\tau d\tau \end{aligned}$$

**psd is an even function**

$$S(-\omega) = S(\omega)$$

# Wiener-Khinchin theorem

## Delta function representation

$$\delta(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{\pm i\omega x} d\omega,$$

## ACF and psd

ACF is obtained as inverse Fourier transform of the psd.

$$G(\tau) = \int_{-\infty}^{\infty} e^{\pm i\omega\tau} S(\omega) d\omega.$$

psd is obtained as forward Fourier transform of the ACF.

$$S(\omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{\pm i\omega\tau} G(\tau) d\tau.$$

# Stochastic vs deterministic processes

## Deterministic signal $x(t)$

### Fourier decomposition

Any deterministic signal  $x(t)$  can always be represented as a superposition of periodic functions

$$x(t) = A \cos \Omega t$$

### Fourier transform $\hat{x}(\omega)$

$$\begin{aligned}\hat{x}(\omega) &= \int_0^T dt e^{-i\omega t} A \cos \Omega t = \frac{A}{2} \int_0^T dt e^{-i\omega t} (e^{i\Omega t} + e^{-i\Omega t}) \\ &= \frac{A}{2} \begin{cases} \left( \frac{e^{it(\Omega-\omega)}}{i(\Omega-\omega)} - \frac{e^{-it(\Omega+\omega)}}{i(\Omega+\omega)} \right) \Big|_0^T, & \omega \neq \pm\Omega \\ T, & \omega = \pm\Omega \end{cases} \\ &= T \frac{A}{2} (\delta_{\omega, \Omega} + \delta_{\omega, -\Omega}) + O(T^0)\end{aligned}$$

**psd**  $S(\omega)$

$$\begin{aligned} S(\omega) &= \lim_{T \rightarrow \infty} \frac{\hat{x}(\omega)\hat{x}^*(\omega)}{2\pi T} = \lim_{T \rightarrow \infty} \frac{1}{2\pi T} \left( T \frac{A}{2} (\delta_{\Omega, \omega} + \delta_{\omega, -\Omega}) + O(T^0) \right)^2 \\ &= \lim_{T \rightarrow \infty} \frac{A^2 T}{8\pi} (\delta_{\Omega, \omega} + \delta_{\omega, -\Omega})^2 + O(T^0) + O(T^{-1}) \\ &= \lim_{T \rightarrow \infty} \frac{A^2 T}{8\pi} (\delta_{\Omega, \omega}^2 + \delta_{\omega, -\Omega}^2) + O(T^0) + O(T^{-1}) \end{aligned}$$

**Infinite power of a deterministic signal**

$$\lim_{T \rightarrow \infty} S(\omega = \pm\Omega) \rightarrow \infty$$

## ACF $G(\tau)$

$$\begin{aligned}G(\tau) &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T dt x(t)x(t+\tau) = \lim_{T \rightarrow \infty} \frac{A^2}{T} \int_0^T dt \cos \Omega t \cos \Omega(t+\tau) \\&= \lim_{T \rightarrow \infty} \frac{A^2}{T} \int_0^T dt (\cos^2 \Omega t \cos \Omega \tau - \sin \Omega t \cos \Omega t \sin \Omega \tau) \\&= \lim_{T \rightarrow \infty} \frac{A^2}{T} \int_0^T dt \left( \frac{1}{2} [1 + \cos 2\Omega t] \cos \Omega \tau - \frac{1}{2} \sin 2\Omega t \sin \Omega \tau \right) \\&= \frac{A^2 \cos \Omega \tau}{2} + O(1/T)\end{aligned}$$

### Infinite correlation time

ACF  $G(\tau)$  of a deterministic signal does not vanish as  $\tau \rightarrow \infty$ .

## Limit of large time window $T$

As  $T \rightarrow \infty$ , the Wiener-Khinchin theorem implies

$$\begin{aligned} S(\omega) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-i\omega\tau} G(\tau) d\tau = \frac{A^2}{4\pi} \int_{-\infty}^{\infty} e^{-i\omega\tau} \cos \Omega\tau d\tau \\ &= \frac{A^2}{8\pi} \int_{-\infty}^{\infty} e^{-i\omega\tau} (e^{i\Omega\tau} + e^{-i\Omega\tau}) d\tau \\ &= \frac{A^2}{4} (\delta(\omega - \Omega) + \delta(\omega + \Omega)) \end{aligned}$$

Only works for square-integrable functions

# Stochastic process $x(t)$

## Special case of a random telegraph process

$$x(t) = \pm 1, \quad \lambda = \mu$$

## Recall stationary ACF $G(\tau)$

$$G(\tau) = \frac{(1 - (-1))\mu^2}{(\mu + \mu)^2} \exp[-2\mu|\tau|] = e^{-2\mu|\tau|}$$

$G(\tau)$  is square-integrable on  $(-\infty, \infty)$

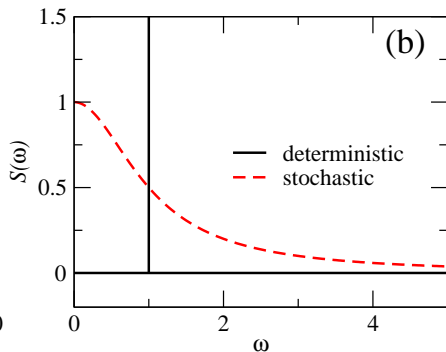
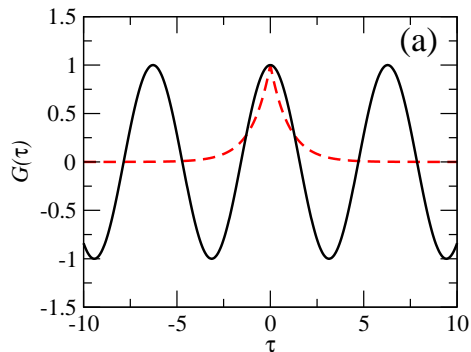
Applying the Wiener-Khinchin theorem

## psd from the Wiener-Khinchin theorem

$$\begin{aligned} S(\omega) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-2\mu|\tau|} e^{-i\omega\tau} d\tau \\ &= \operatorname{Re} \left( 2 \frac{1}{2\pi} \int_0^{\infty} e^{-2\mu\tau} e^{-i\omega\tau} d\tau \right) \\ &= \operatorname{Re} \left( \frac{1}{\pi} \frac{e^{-\tau(2\mu+i\omega)}}{-(2\mu+i\omega)} \Big|_0^{\infty} \right) \\ &= \operatorname{Re} \left( \frac{1}{\pi} \frac{1}{(2\mu+i\omega)} \right) \\ &= \frac{2\mu}{\pi(4\mu^2 + \omega^2)} \end{aligned}$$



# Stochastic vs deterministic processes



**Stochastic process  $\chi(t)$  with zero mean and zero correlation time**

$$\langle \chi(t) \rangle = 0, \quad \langle \chi(t)\chi(t') \rangle = \delta(t - t')$$

**psd**

$$S(\omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \delta(\tau) e^{i\omega\tau} d\tau = \frac{1}{2\pi}$$

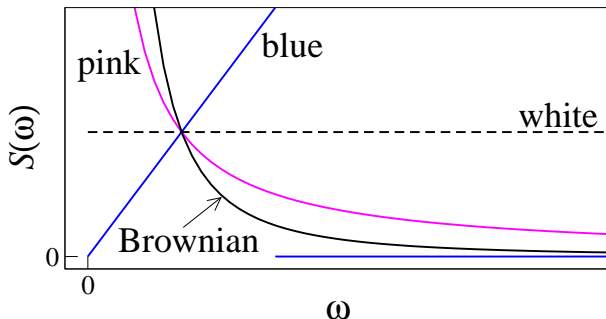
**white noise**

Equal power density for any frequency

# Colored noise

Stochastic process  $\eta(t)$  with any non-constant psd  $S(\omega) \neq \text{constant}$

- Pink noise:  $S(\omega) = \omega^{-1}$
- Brownian noise (red noise):  $S(\omega) = \omega^{-2}$
- Blue noise:  $S(\omega) = \begin{cases} \omega, & \omega \in [0, \omega_0] \\ 0, & \text{otherwise} \end{cases}$



## Flavours of the noise

For each noise color one can define infinitely many different flavours, depending on the distribution of  $x$  at any given time.

## Discretized signal

In practice, we discretize  $x(t)$  by choosing a time step  $\Delta t$ :

$$x(t_i), \quad t_i = t_0 + i\Delta t, \quad (i = 0, 1, 2, 3, \dots)$$

# White Gaussian noise

Generated by

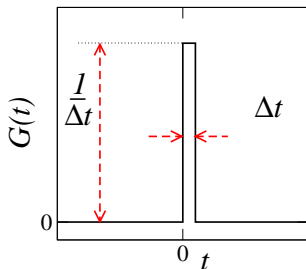
$$x(t_i) \sim \frac{1}{\sqrt{\Delta t}} N(0, 1) = N\left(0, \frac{1}{\Delta t}\right)$$

Indeed

$$\langle x(t_i) \rangle = 0, \quad G(t_k - t_i) = \langle x(t_i)x(t_k) \rangle = \begin{cases} 0, & i \neq k \\ \frac{1}{\Delta t}, & i = k \end{cases}$$

Representation of the Dirac delta-function  $\delta(t)$

$$\delta(t) = \begin{cases} 0, & t \neq 0 \\ \infty, & t = 0 \end{cases}, \quad \int \delta(t) dt = 1$$



**Generated by**

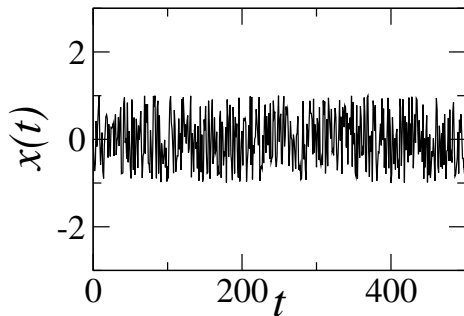
$$x(t_i) \sim \sqrt{\frac{3}{\Delta t}} \text{Uniform}([-1, 1])$$

**Indeed**

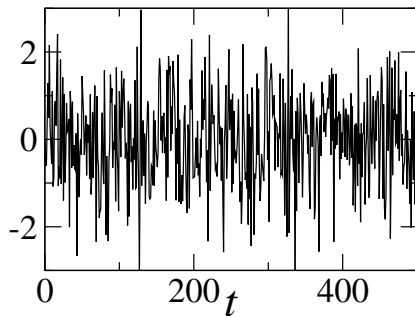
$$\begin{aligned} \langle x(t_i) \rangle &= 0 \\ G(t_k - t_i) &= \langle x(t_i)x(t_k) \rangle = \begin{cases} 0, & i \neq k \\ \frac{1}{\Delta t}, & i = k \end{cases} \end{aligned}$$

# Gaussian and uniform white noise

## White uniform



## White Gaussian



**Stochastic process  $w(t)$  with conditional probability**

$$\partial_t p(w, t | w_0, t_0) = \frac{1}{2} \partial_w^2 p(w, t | w_0, t_0).$$

**Solution of the diffusion equation with  $D = 1/2$**

$$p(w, t | w_0, t_0) = \frac{1}{\sqrt{2\pi(t - t_0)}} \exp\left(-\frac{(w - w_0)^2}{2(t - t_0)}\right).$$



# Wiener process: independents of increments $\Delta w_i$

## Markovian property

For any  $t_i > t_{i-1}$ , the increments  $\Delta w_i = w(t_i) - w(t_{i-1})$  are independent with the distribution

$$p(\Delta w_i) = p(w_i, t_i | w_{i-1}, t_{i-1}) = \frac{1}{\sqrt{2\pi(t_i - t_{i-1})}} \exp\left(-\frac{(w_i - w_{i-1})^2}{2(t_i - t_{i-1})}\right)$$

$$\begin{aligned}\Delta w_i &\sim N(0, (t_i - t_{i-1})) = N(0, \Delta t) \\ \langle (\Delta w_i)^2 \rangle &= \Delta t = t_i - t_{i-1}\end{aligned}$$

# Wiener process as a solution of stochastic differential equation

## Derivative of the Wiener process

**White Gaussian noise**  $\xi(t)$

$$\xi(t) = \frac{\Delta w}{\Delta t} \sim N\left(0, \frac{1}{\Delta t}\right) = \frac{1}{\sqrt{\Delta t}} N(0, 1)$$

**Stochastic differential equation for**  $w(t)$

$$\frac{dw(t)}{dt} = \dot{w}(t) = \lim_{\Delta t \rightarrow 0} \frac{\Delta w}{\Delta t} = \xi(t)$$

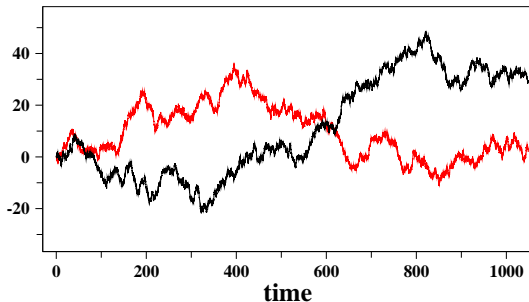
# Wiener process and Brownian motion

## Brownian motion

The Wiener process  $w(t)$  is generated by

$$w(t_{i+1}) = w(t_i) + N(0, 1)\sqrt{\Delta t}, \quad (i = 0, 1, 2, 3, \dots).$$

Describes the coordinate of a Brownian particle with the diffusion coefficient  $D = 1/2$ .



# Computation of Power spectral density

Ensemble average of the square of Fourier transformed  $x(t)$

$$\langle \hat{x}(\omega)\hat{x}^*(\omega') \rangle = \int dt dt' e^{-i\omega t} e^{i\omega' t'} \langle x(t)x(t') \rangle,$$

For a stationary process  $x(t)$

$$\langle x(t)x(t') \rangle = G(t - t'),$$

Connection with  $S(\omega)$

$$\begin{aligned} \langle \hat{x}(\omega)\hat{x}^*(\omega') \rangle &= \int dt dt' e^{-i\omega t} e^{i\omega' t'} G(t - t') = \{(t, t') \rightarrow (t', \tau = t - t')\} \\ &= \int dt' e^{i(\omega' - \omega)t'} \int d\tau e^{-i\omega' \tau} G(\tau) \\ &= 2\pi \int dt' e^{i(\omega' - \omega)t'} S(\omega') = (2\pi)^2 S(\omega') \delta(\omega - \omega'). \end{aligned}$$

# psd of the Wiener process $dW/dt = \xi(t)$

In the Fourier space

$$i\omega \hat{x}(\omega) = \hat{\xi}(\omega),$$

Consequently

$$\langle \hat{x}(\omega) \hat{x}^*(\omega') \rangle = \frac{\langle \hat{\xi}(\omega) \hat{\xi}^*(\omega') \rangle}{(i\omega)(-i\omega')}$$

By the definition of the white noise

$$\langle \hat{\xi}(\omega) \hat{\xi}^*(\omega') \rangle = (2\pi)^2 \left( \frac{1}{2\pi} \int G_{\xi}(\tau) d\tau e^{-i\omega\tau} \right) \delta(\omega - \omega') = (2\pi) \delta(\omega - \omega'),$$

$$\langle \hat{x}(\omega) \hat{x}^*(\omega') \rangle = \frac{2\pi \delta(\omega - \omega')}{\omega^2} = (2\pi)^2 S(\omega) \delta(\omega - \omega')$$

$$S(\omega) = \left( \frac{1}{2\pi} \right) \frac{1}{\omega^2}.$$

# ACF of the Wiener process

## Using the Wiener-Khinchin theorem

$$G(\tau) = \int_{-\infty}^{\infty} \frac{1}{2\pi\omega^2} e^{i\omega\tau} d\omega \sim \tau.$$

## From ACF to psd

The backward Fourier transformation

$$S(\omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \tau e^{-i\omega\tau} d\tau$$

does not exist, as the function  $G(\tau) \sim \tau$  is not a square-integrable on the interval  $(-\infty, \infty)$ . Physically, it means that the Wiener process is not a stationary process.

# Ornstein-Uhlenbeck process

Stochastic ODE for  $x(t)$

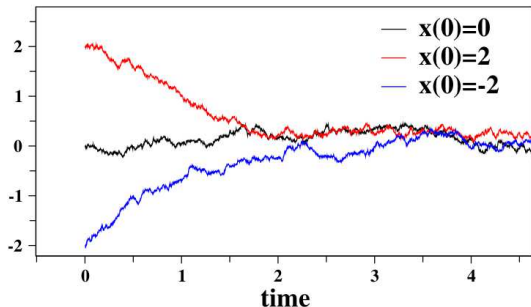
$$\dot{x}(t) = -\frac{x}{\tau} + \frac{1}{\tau}\sqrt{2D}\xi(t),$$

with  $\xi(t)$ ... white noise

$D$ ... strength parameter

$\tau$ ... correlation time

$$D = 0.045, \quad \tau = 1$$

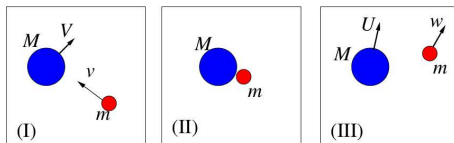


# Ornstein-Uhlenbeck process

Massive particle of mass  $M$  in a gas of molecules  $m$

**Momentum and energy conservation in a single elastic collision**

$$M \frac{V^2}{2} + m \frac{v^2}{2} = M \frac{U^2}{2} + m \frac{w^2}{2}$$
$$M \vec{V} + m \vec{v} = M \vec{U} + m \vec{w}$$





## Change of the momentum of the massive particle

$$\Delta \vec{P} = M\vec{U} - M\vec{V} = -\frac{2m}{M+m}(M\vec{V}) + \frac{2M}{M+m}(m\vec{v})$$

## Number of collisions within time interval $\delta t$

$$\delta t \ll 1, \quad \left\| \frac{\delta P(t)}{P(t)} \right\| \ll 1, \quad I_i(t, \delta t) = (1, 0)$$

## Total momentum change

$$\Delta \vec{P} = - \left( \sum_{i=1}^N \frac{2m}{M+m} I_i(t, \delta t) \right) \vec{P} + \left( \sum_{i=1}^N \frac{2M}{M+m} I_i(t, \delta t) \vec{p}_i \right)$$

# Langevin equation

Large number of uncorrelated collision events within  $dt$

$$\frac{d\vec{P}}{dt} = -\alpha(P)\vec{P} + \vec{\xi}(t)$$

**Paul Langevin (1872-1946): French physicist**



In the Fourier space

$$i\omega\hat{x}(\omega) = -\frac{\hat{x}(\omega)}{\tau} + \frac{\sqrt{2D}}{\tau}\hat{\xi}(\omega)$$

Solving for  $\hat{x}(\omega)$

$$\hat{x}(\omega) = \frac{\sqrt{2D}}{\tau} \frac{\hat{\xi}(\omega)}{i\omega + 1/\tau}$$

Taking ensemble average

$$\langle \hat{x}(\omega)\hat{x}^*(\omega') \rangle = 4\pi D \frac{\delta(\omega - \omega')}{1 + \tau^2\omega^2}.$$

Taking ensemble average

$$S(\omega) = \frac{D}{\pi} \frac{1}{1 + \tau^2\omega^2}.$$

# ACF of the Ornstein-Uhlenbeck process

## Relation to random telegraph process

Note that the functional form of  $S(\omega)$  is identical with the psd of a random telegraph process. Compare

$$S(\omega) = \frac{2\mu}{\pi(4\mu^2 + \omega^2)} \quad \text{vs} \quad S(\omega) = \frac{D}{\pi} \frac{1}{1 + \tau^2\omega^2}$$

This implies that the ACF of the Ornstein-Uhlenbeck process has the form

$$G(t) = \frac{D}{\tau} e^{-|t|/\tau}$$

Check using the Wiener-Khinchin theorem

$$\begin{aligned} G(t) &= \int_{-\infty}^{\infty} \frac{D}{\pi} \frac{e^{i\omega t}}{1 + \tau^2 \omega^2} d\omega \\ &= \text{residual} \left( \frac{De^{i\omega t}}{\pi\tau^2(\omega - i/\tau)(\omega + i/\tau)} \right) \Big|_{\omega=i\tau^{-1}} \\ &= \frac{D}{\tau} e^{-t/\tau}, \quad (t \geq 0) \end{aligned}$$

Correlation time

ACF decay with the characteristic rate of  $\tau$ .

# 2D plasma in magnetic field

## Example

A charged particle, confined to move on a plane in a constant magnetic field  $B$ , which is perpendicular to the plane of motion.

$$\dot{\mathbf{r}} = \mathbf{v}, \quad \dot{\mathbf{v}} = \frac{q}{m} \mathbf{v} \times \mathbf{B} - \gamma \mathbf{v} + \sqrt{2\gamma \frac{kT}{m}} \boldsymbol{\xi}(t),$$

with  $\gamma \dots$  the damping coefficient

$T \dots$  the absolute temperature

$m \dots$  mass of the particle

$q \dots$  charge of the particle

$k \dots$  Boltzmann constant

**Question: Determine the power spectrum of the velocity components  $v_x$  and  $v_y$ .**

# 2D plasma in magnetic field

