0.1 Continuity of Brownian Path

Consider a particle starting in x = 0 at t = 0, and traversing N points $\{x_i\}_{i=1,\dots,N}$ such that all increments $\Delta x_i = x_i - x_{i-1}$ are independent and described by a gaussian pdf. The density function for such a trajectory $\{x_i\}$ is the usual product of transition probabilities:

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$$d\mathbb{P}_{t_1,\dots,t_N}\left(x_1,\dots,x_N\right) = \left(\prod_{i=1}^N \frac{\mathrm{d}x_i}{\sqrt{4\pi\Delta t_i D}}\right) \exp\left(-\sum_{i=1}^N \frac{\left(\Delta x_i\right)^2}{4D\Delta t_i}\right) \quad {}^{\Delta t_i = t_i - t_{i-1}}_{\Delta x_i = x_i - x_{i-1}} \quad (1)$$

We now show that, taking the continuum limit $\max_i \Delta t_i \to 0$ leads to paths $\{x(\tau)\}$ that are almost surely continuous. In other words, for any interval $T \subseteq \mathbb{R}$, the subset $N \subset \mathbb{R}^T$ of functions that are discontinuous has 0 Wiener measure. Mathematically, we want to show that, as $\Delta t_i \to 0$, the probability that Δx_i is close to 0 approaches certainty:

$$\lim_{\Delta t_i \to 0} \mathbb{P}(|\Delta x_i| < \epsilon) = 1 \quad \forall \epsilon > 0$$

This is just the probability that, during time Δt_i , the particle makes a jump of size lower than ϵ :

$$\mathbb{P}(|\Delta x_i| < \epsilon) = \mathbb{P}(x_{i-1} - \epsilon < x_i < x_{i-1} + \epsilon | x(t_{i-1}) = x_i) =$$

$$= \int_{x_{i-1} - \epsilon}^{x_{i-1} + \epsilon} \frac{\mathrm{d}x_i}{\sqrt{4\pi D\Delta t_i}} \exp\left(-\frac{(x_i - x_{i-1})^2}{4D\Delta t_i}\right) =$$

$$\stackrel{=}{=} \int_{-\epsilon}^{+\epsilon} \frac{\mathrm{d}\Delta x_i}{\sqrt{4\pi D\Delta t_i}} \exp\left(-\frac{(\Delta x_i)^2}{4D\Delta t_i}\right)$$

where in (a) we translated the variable of integration $\Delta x_i = x_i - x_{i-1}$. With another change of variables:

$$\frac{(\Delta x_i)^2}{\Delta t_i} = z^2 \Rightarrow z = \frac{\Delta x_i}{\sqrt{\Delta t_i}} \Rightarrow d\Delta x_i = dz \sqrt{\Delta t_i}$$

we get:

$$\mathbb{P}(|\Delta x_i| < \epsilon) = \int_{|z| < \epsilon/\sqrt{\Delta t_i}} \frac{\mathrm{d}z \sqrt{\Delta t_i}}{\sqrt{4\pi D \Delta t_i}} \exp\left(-\frac{z^2}{4D}\right)$$

And taking the continuum limit leads to:

$$\lim_{\Delta t_i \to 0} \mathbb{P}(|\Delta x_i| < \epsilon) = \int_{-\infty}^{+\infty} \frac{\mathrm{d}z}{\sqrt{4\pi D}} \exp\left(-\frac{z^2}{4D}\right) = 1$$

0.2 Differentiability of Brownian Path

With a very similar calculation (here omitted) we can also show that:

$$\lim_{\Delta t_i \downarrow 0} \left(\left| \frac{\Delta x_i}{\Delta t_i} \right| > k \right) = 1 \quad \forall k > 0$$

meaning that Brownian paths are almost surely everywhere non-differentiable.

Nonetheless, it is sometimes useful to consider "formal derivatives" of a Brownian path, that acquire a definite meaning only when considering a *finite discretization*. For example, we can start from (1) and rewrite it as:

$$d\mathbb{P}_{t_1,\dots,t_N}(x_1,\dots,x_N) = \left(\prod_{i=1}^N \frac{\mathrm{d}x_i}{\sqrt{4\pi\Delta t_i}}\right) \exp\left(-\frac{1}{4D}\sum_{i=1}^N \frac{\Delta t_i}{\Delta t_i} \left(\frac{\Delta x_i}{\Delta t_i}\right)^2\right)$$

Then, in the continuum limit $\Delta t_i \to 0$, the sum in the exponential argument becomes a Riemann integral:

$$\sum_{i=1}^{N} \Delta t_i \left(\frac{\Delta x_i}{\Delta t_i}\right)^2 \xrightarrow{\Delta t \to 0} \int_0^t d\tau \underbrace{\left(\frac{dx_i}{d\tau}\right)^2}_{\dot{x}^2(\tau)} \qquad t = t_N$$

where $t = t_N$. Substituting it back leads to:

$$dx_w(\tau) = \prod_{\tau=0^+}^t \frac{dx(\tau)}{\sqrt{4\pi D d\tau}} \exp\left(-\frac{1}{4D} \int_0^t \dot{x}^2(\tau) d\tau\right)$$

This expression has no rigorous meaning in this form $(\dot{x}(\tau) \text{ does not exists!})$ but can be formally manipulated into other expressions that have a definite meaning, thus proving useful for the discussion.

0.3 Forces on the particle

We want now to generalize the framework we obtained to the case of a diffusing particle subject to *external forces*, e.g. a drop of ink diffusing through a water medium in the presence of gravity.

To do this, we first return to the beginning, deduce a Master Equation for a more general *evolution*, and then choose the right probability distribution reproducing the behaviour in presence of forces.

So, let's start by considering a particle moving on a uniform one-dimensional lattice $(x_i = i \cdot l, t_n = n \cdot \epsilon)$, and satisfying the Markovian property, meaning that the probability $W_i(t_{n+1})$ of being at the position labelled by i at the next time-step t_{n+1} depends only on the current state t_n , that is on the current probabilities $W_i(t_n) \forall j$ and on the current transition probabilities $W_{ij}(t_n)$ from j to i:

$$W_{i}(t_{n+1}) = \sum_{j=-\infty}^{+\infty} W_{ij}(t_{n})W_{j}(t_{n})$$
 (2)

Previously, we assumed that:

$$W_{ij}(t_n) = \delta_{j,i-1}P_{+} + \delta_{j,i+1}P_{-}$$

Which means that the particle only jumps from adjacent positions, one step at a time, and cannot remain at the same place. This Master Equation leads, in d=3 and in the continuum limit, to the usual Diffusion Equation:

$$\frac{\partial}{\partial t}W(\boldsymbol{x},t|\boldsymbol{x_0},t_0) = \nabla^2 W(\boldsymbol{x},t|\boldsymbol{x_0},t_0)$$

We now consider a more general case, where we drop the discretization of the space domain, allowing *jumps* of any size in \mathbb{R} . Then (2) becomes:

$$W(x, t_{n+1}) dx = \int_{-\infty}^{+\infty} dz W(x, t_{n+1}|x - z, t_n) W(x - z, t_n)$$
 (3)

The integrand is the probability of the particle being in [x-z, x-z+dx] at time t_n and making a jump of size z to reach [x, x+dx] at time t_{n+1} . By summing over all possible jump sizes we compute the total probability of the particle being near the arrival position.

If we require *jumps* to be **independent** of each other¹, as it is physically evident by the problem's symmetry, then the *jump* probabilities $W(x, t_{n+1}|x-z, t_n)$ depend only on the *jump size* z.

Assuming a *isolate system*, as the particle cannot *escape*, **probability is conserved**:

$$\begin{split} \int_{\mathbb{R}} \mathrm{d}x \, W(x,t_{n+1}) &\stackrel{!}{=} \int_{\mathbb{R}} \mathrm{d}y \, W(y,t_n) \\ &\stackrel{=}{=} \int_{\mathbb{R}} \mathrm{d}z \int_{\mathbb{R}} \mathrm{d}x W(x,t_{n+1}|x-z,t_n) \, W(x-z,t_n) = \\ &\stackrel{=}{=} \int_{\mathbb{R}} \mathrm{d}z \int_{\mathbb{R}} \mathrm{d}y \, W(y+z,t_{n+1}|y,t_n) W(y,t_n) = \\ &\stackrel{=}{=} \left(\int_{\mathbb{R}} \mathrm{d}z \, W(\bar{y}+z,t_{n+1}|\bar{y},t_n) \right) \left(\int_{\mathbb{R}} \mathrm{d}y \, W(y,t_n) \right) \qquad \forall \bar{y} \in \mathbb{R} \end{split}$$

where in (a) we changed variables $x \mapsto y = x - z$, with dy = dx, and in (b) we used the *independent increments* property (\bar{y} is a arbitrary constant). Comparing the first and last lines leads to:

$$\int_{\mathbb{R}} \mathrm{d}z \, W(y+z, t_{n+1}|y, t_n) = 1$$

Intuitively, if the particle cannot disappear, it must make a jump. Here on, for notation simplicity, we denote:

$$W(y+z,t_{n+1}|y,t_n)\equiv W(+z|y,t_n)$$

Starting from (3) and taking the continuum limit in time we can write a more

¹∧This is a stronger requirement than the Markovian property. In fact, *independent increments* imply a *Markov process*, but the converse is not true. See http://statweb.stanford.edu/~adembo/math-136/Markov_note.pdf

general diffusion equation. We start by constructing the difference quotient:

$$W(x,t_{n+1}) - W(x,t_n) = \int_{\mathbb{R}} dz \, W(+z|x-z,t_n) W(x-z,t_n) - W(x,t_n) =$$

$$= \int_{\mathbb{R}} dz \, W(+z|x-z,t_n) W(x-z,t_n) - \underbrace{\int_{\mathbb{R}} dz \, W(+z|x,t_n) \, W(x,t_n)}_{=1} =$$

$$= \int_{\mathbb{R}} dz \, \left[\underbrace{W(+z|x-z,t_n) W(x-z,t_n)}_{F_z(x-z)} - \underbrace{W(+z|x,t_n) W(x,t_n)}_{F_z(x)} \right] =$$

$$= \int_{\mathbb{R}} dz \, \left[\underbrace{F_z(x-z) - F_z(x)}_{=2} \right] =$$

$$= \int_{\mathbb{R}} dz \, \left[\underbrace{F_z(x) - z \frac{\partial}{\partial x} F_z(x) + \frac{z^2}{2} \frac{\partial^2}{\partial x^2} [F_z(x)] + \cdots - F_z(x)}_{=2} \right] =$$

$$= -\int_{\mathbb{R}} dz \, z \, \underbrace{\frac{\partial}{\partial x} [F_z(x)] + \frac{1}{2} \int_{\mathbb{R}} dz \, z^2 \frac{\partial^2}{\partial x^2} [F_z(x)] + \cdots}_{\mu_1(x,t_n)} =$$

$$= \frac{\partial}{\partial x} \left[\underbrace{\left(\int_{\mathbb{R}} dz \, z \, W(+z|x,t_n) \right)}_{\mu_1(x,t_n)} W(x,t_n) \right] +$$

$$+ \frac{1}{2} \underbrace{\frac{\partial^2}{\partial x^2}}_{=2} \left[\underbrace{\left(\int_{\mathbb{R}} dz \, z \, W(+z|x,t_n) \right)}_{\mu_2(x,t_n)} W(x,t_n) \right] + \cdots$$

where $F_z(x)$ is the probability of a *jump* of size z from the position x. In (a) we expanded F_z about x, and in (b) we exchanged the order of integrals and derivatives. Then we define the k-th moment of the *jump* pdf as follows:

$$\mu_k(x,t) = \int_{\mathbb{R}} dz \, z^k \, W(+z|x,t)$$

This allows us to rewrite the above difference in a more compact form:

$$W(x, t_{n+1}) - W(x, t_n) = \sum_{k=1}^{+\infty} \frac{(-1)^k}{k!} \frac{\partial^k}{\partial x^k} (\mu_k(x, t_n) W(x, t_n))$$

Physically, as probability is conserved, by the continuity equation, the *change* in probability density equals the divergence of a flux, which is just the x derivative in this one-dimensional case. So, if we extract a derivative, we can write the flux explicitly:

$$= \frac{\partial}{\partial x} \left(\sum_{k=1}^{+\infty} \frac{(-1)^k}{k!} \frac{\partial^{k-1}}{\partial x^{k-1}} (\mu_k(x, t_n) W(x, t_n)) \right)$$

$$\equiv -\frac{\partial}{\partial x} J(x, t_n)$$

where $J(x, t_n)$ is the outward flux at x, meaning that if J > 0, then $W(x, t_{n+1}) < W(x, t_n)$ (the particle escapes from x to another place), and otherwise if J < 0 we have $W(x, t_{n+1}) > W(x, t_n)$ (the particle is sucked in x).

If we integrate both sides over x and apply the probability conservation we get the boundary conditions for the flux:

$$\int_{\mathbb{R}} (W(x, t_{n+1}) - W(x, t_n)) dx = \int_{\mathbb{R}} dx \left(-\frac{\partial}{\partial x} J(x, t_n) \right)$$
$$1 - 1 = -J(x, t_n) \Big|_{-\infty}^{+\infty} = J(-\infty, t_n) - J(+\infty, t_n)$$

This means that, in a *isolate system*, the *flux* at $\pm \infty$ must be the same. Finally, normalizing by the time interval we get the complete difference quotient, which will become a time derivative in the continuum limit.

$$\frac{W(x,t_{n+1}) - W(x,t_n)}{t_{n+1} - t_n} = \frac{\partial}{\partial x} \left\{ \sum_{k=1}^{\infty} \frac{(-1)^k}{k!} \frac{\partial^{k-1}}{\partial x^{k-1}} \frac{\mu_k(x,t_n)W(x,t_n)}{t_{n+1} - t_n} \right\}$$
(4)

Letting $t_{n+1} - t_n = \epsilon$, in the limit $\epsilon \to 0$ the left side will be $\dot{W}(x,t)$.

All that's left is to find an explicit definition for the jump pdf W(+z|x,t). Previously, we assumed a **gaussian** pdf for the displacements:

$$z \sim \frac{1}{\sqrt{4\pi D\epsilon}} \exp\left(-\frac{(\Delta x)^2}{4D\epsilon}\right)$$

With this choice, the first two moments become:

$$\mu_1 = 0$$
 $\mu_2 = 2D\epsilon$

And the variance:

$$Var(z) = \mu_2 - \mu_1^2 = 2D\epsilon \propto \epsilon$$

However, for a particle subject to a force we would expect to have a *preferred jump direction*, leading to a *constant velocity motion* in the direction of the force. So we require a different μ_1 :

$$\langle z \rangle = \mu_1 = \int_{\mathbb{D}} zW(+z|x,t) \propto \epsilon f(x)$$

We still want to fix the variance to be proportional to ϵ , as it is expected in a diffusion process.

An appropriate choice for such a distribution is given by:

$$W(+z|x,t) = F\left(\frac{z - \epsilon f(x,t)}{\sqrt{\epsilon \hat{D}(x,t)}} \frac{1}{\sqrt{\epsilon \hat{D}(x,t)}}\right)$$
 (5)

with $F, \hat{D} \colon \mathbb{R} \to \mathbb{R}$ functions, satisfying certain conditions, and with a physical meaning that we will now see.

First of all, we check the normalization:

$$1 \stackrel{!}{=} \int_{\mathbb{R}} dz \, W(+z|x,t) = \frac{1}{\sqrt{\epsilon \hat{D}(x,t)}} \int_{\mathbb{R}} dz \, F\left(\frac{z - \epsilon f(x,t)}{\sqrt{\epsilon \hat{D}(x,t)}}\right) \stackrel{=}{=} \int_{\mathbb{R}} dy \, F(y)$$

where in (a) we changed variables:

$$y = \frac{z - \epsilon f(x, t)}{\sqrt{\epsilon \hat{D}(x, t)}} \qquad dz = \sqrt{\epsilon \hat{D}(x, t)} \, dy$$
 (6)

Then we compute the first moment:

$$\langle z \rangle = \mu_1(x,t) = \int_{\mathbb{R}} dz \, z \, F\left(\frac{z - \epsilon f(x,t)}{\sqrt{\epsilon \hat{D}(x,t)}}\right) \frac{1}{\sqrt{\epsilon \hat{D}(x,t)}} =$$

$$= \int_{\mathbb{R}} dy \left(\epsilon f(x,t) + y\sqrt{\epsilon \hat{D}(x,t)}\right) F(y) =$$

$$= \epsilon f(x,t) \underbrace{\int_{\mathbb{R}} F(y) \, dy}_{-1} + \sqrt{\epsilon \hat{D}(x,t)} \int_{\mathbb{R}} y F(y) \stackrel{!}{=} \epsilon f(x,t)$$

So, in order to have the right normalization and the desired $\langle z \rangle$ we need:

$$\begin{cases} \int_{\mathbb{R}} dy \, F(y) = 1 \\ \int_{\mathbb{R}} dy \, y F(y) = 0 \end{cases}$$

Both conditions are satisfied, for example, by all even normalized functions. For the second moment:

$$\mu_{2}(x,t) = \frac{1}{\sqrt{\epsilon \hat{D}(x,t)}} \int_{\mathbb{R}} dz \, z^{2} F\left(\frac{z - \epsilon f(x,t)}{\sqrt{\epsilon \hat{D}(x,t)}}\right) =$$

$$\stackrel{=}{=} \int_{\mathbb{R}} dy \, (\epsilon f(x,t) + y\sqrt{\epsilon \hat{D}(x,t)})^{2} F(y) =$$

$$= \int_{\mathbb{R}} dy \, F(y) [\epsilon^{2} f^{2} + y^{2} \hat{D} \epsilon + 2\epsilon \sqrt{\epsilon \hat{D} f y}] =$$

$$= \epsilon^{2} f^{2} + \hat{D} \epsilon \int_{\mathbb{R}} dy \, y^{2} F(y) = \epsilon^{2} f^{2} + \hat{D} \epsilon \langle y^{2} \rangle_{F(y)}$$

And so the variance becomes:

$$\operatorname{Var}(z) = \mu_2 - \mu_1^2 = \epsilon \hat{D} \langle y^2 \rangle_{F(y)} \propto \epsilon$$

which is proportional to ϵ as desired. For notational simplicity, we introduce a new function $D: \mathbb{R} \to \mathbb{R}$ such that:

$$\operatorname{Var}(z) = \epsilon \hat{D} \langle y^2 \rangle_{F(y)} \equiv 2D(x, t) \epsilon \Rightarrow \mu_2(x, t) = \epsilon^2 f^2 + 2D(x, t)$$

We note that higher order moments are all of order $O(\epsilon^{3/2})$. For example, the

third moment is:

$$\mu_{3}(x,t) = \frac{1}{\sqrt{\epsilon \hat{D}(x,t)} \int_{\mathbb{R}}} dz \, z^{2} F\left(\frac{z - \epsilon f(x,t)}{\sqrt{\epsilon \hat{D}(x,t)}}\right) =$$

$$= \int_{\mathbb{R}} dy \, (\epsilon f(x,t) + y\sqrt{\epsilon \hat{D}(x,t)})^{3} F(y) =$$

$$= \int_{\mathbb{R}} dy \, \left(\epsilon^{3} f^{3} + y^{3} (\epsilon \hat{D})^{3/2} + 3\epsilon^{2} f^{2} y \sqrt{\epsilon \hat{D}} + 3\epsilon^{2} f \hat{D} y^{2}\right) F(y) =$$

$$= \epsilon^{3} f^{3} + (\epsilon \hat{D})^{3/2} + 3\epsilon^{2} f \hat{D} \langle y^{2} \rangle_{F(y)} = O(\epsilon^{3/2})$$

Substituting back (5) in (4) we arrive to:

$$\frac{W(x,t_{n+1}) - W(x,t_n)}{\epsilon} = -\frac{\partial}{\partial x} \left[W(x,t_n) \underbrace{\frac{\mu_1(x,t_n)}{\epsilon}}_{f(x,t)} \right] + \frac{1}{2} \frac{\partial^2}{\partial x^2} \left[\underbrace{\frac{\mu_2(x,t_n)}{\epsilon}}_{\epsilon f^2 + 2D(x,t)} W(x,t_n) \right] + \underbrace{\frac{1}{3!} \frac{\partial^3}{\partial x^3} \left[W(x,t_n) \frac{\mu_3(x,t_n)}{\epsilon} \right] + \dots}_{O(\epsilon^{1/2})}$$

Taking the limit $\epsilon \to 0$, we are left with:

$$\frac{\partial W(x,t)}{\partial t} = -\frac{\partial}{\partial x} [f(x,t)W(x,t)] + \frac{1}{2} \frac{\partial^2}{\partial x^2} [2D(x,t)W(x,t)] =$$

$$= -\frac{\partial}{\partial x} \Big[f(x,t)W(x,t) - \frac{\partial}{\partial x} \Big(D(x,t)W(x,t) \Big) \Big]$$

This is the **Fokker-Planck equation**, describing the diffusion process in the presence of a *force* f(x,t), and a diffusion parameter D(x,t).

Note that, in absence of forces $f(x,t) \equiv 0$ and with a constant diffusion $D(x,t) \equiv D$ we retrieve the usual diffusion equation:

$$\frac{\partial}{\partial t}W(x,t) = D\frac{\partial^2}{\partial x^2}W(x,t)$$

0.4 Langevin equation

The Fokker-Planck equation involves *probability distributions*, meaning that it describes the behaviour of *ensembles of trajectories* at once. However, we can find an equivalent description by focusing on a *single path*.

We start with a Wiener process, that is a stochastic process with *independent* and *gaussian* increments and *continuous* paths. Considering a time discretization $\{t_i\}$, the evolution of a single trajectory is described by:

$$x(t_{i+1}) = x(t_i) + \Delta x(t_i) \tag{7}$$

where each increment $\Delta x(t_i)$ is sampled from a gaussian pdf:

$$\Delta x_i(t_i) \sim \frac{1}{\sqrt{4\pi D\Delta t_i}} \exp\left(-\frac{(\Delta x)^2}{4D\Delta t_i}\right)$$

To simplify notation, we change variables, so that:

$$\frac{\Delta B^2}{2} = \frac{\Delta x^2}{4D} \Rightarrow \Delta B = \frac{\Delta x}{\sqrt{2D}}$$

If $x \sim p(x)$, and $y = y(x) \sim g(y)$, then by the rule for a change of random variables we have:

$$g(y) = p(x(y)) \frac{\mathrm{d}x(y)}{\mathrm{d}y}$$

In this case:

$$\Delta B \sim \frac{1}{\sqrt{4\pi D\Delta t_i}} \exp\left(-\frac{(\Delta B)^2}{2\Delta t_i}\right) \underbrace{\frac{\mathrm{d}\Delta x}{\mathrm{d}\Delta B}}_{\sqrt{2D}} = \frac{1}{\sqrt{2\pi\Delta t_i}} \exp\left(-\frac{(\Delta B)^2}{2\Delta t_i}\right)$$

Note that now $\langle \Delta B^2(t_i) \rangle = \Delta t_i$, leaving out the D. Substituting in (7) and rearranging we get:

$$x(t_{i+1}) - x(t_i) = \sqrt{2D\Delta B(t_i)} \tag{8}$$

We want now to form a time derivative in the left side, in order to arrive a (stochastic) differential equation for paths. To do this, we first extract a Δt_i factor from $\Delta B(t_i)$ by performing another change of variables:

$$\Delta B(t_i) \equiv \Delta t_i \xi(t_i) \tag{9}$$

so that $\Delta x_i = \sqrt{2D}\Delta t_i \xi_i$, and all the randomness is now contained in the random variable ξ , which is distributed according to:

$$\xi(t_i) \sim \frac{1}{\sqrt{2\pi\Delta t_i}} \exp\left(-\frac{\Delta t_i^2 \xi_i^2}{2\Delta t_i}\right) \underbrace{\frac{\mathrm{d}\Delta B_i}{\mathrm{d}\xi(t_i)}}_{\Delta t_i} = \sqrt{\frac{\Delta t_i}{2\pi}} \exp\left(-\frac{\Delta t_i}{2} \xi_i^2\right) \qquad \xi_i \equiv \xi(t_i)$$

Substituting back in (8) and dividing by Δt_i leads to:

$$\frac{x(t_{i+1}) - x(t_i)}{\Delta t_i} = \sqrt{2D}\xi(t_i)$$

And by taking the continuum limit $\Delta t_i \to 0$ we get the **Langevin equation** for a Brownian particle:

$$\dot{x}(t) = \sqrt{2D}\xi(t) \tag{10}$$

We can see $\xi(t)$ as a highly irregular, quickly varying function, which, in a certain sense, expresses the result of Brownian collisions at a certain instant. In particular, the following holds:

$$\langle \xi(t) \rangle = 0$$
 $\langle \xi(t)\xi(t') \rangle = \delta(t - t')$

meaning that the values of $\xi(t)$ at different instants are completely independent.

Note that, as we saw previously, Brownian paths are not differentiable - and so $\dot{x}(t)$ does not exist, and this is just a *formal* equation, with a definite meaning only in a given *discretization*. Also, note that $\xi(t)$ is a random variable, and so this is an example of a **stochastic differential equation**. It is not clear how to find a solution to such an equation, or even how to *define* what a solution should be - and this will be the main topic of the next section.

We can rewrite (10) in a more *rigorous* form by "multiplying by dt", i.e. performing the change of variables (9), which - in the continuum limit - is $dB = \xi dt$, leading to:

$$dx(t) = \sqrt{2D} dB$$
 $dB \sim \frac{1}{\sqrt{2\pi} dt} \exp\left(-\frac{dB^2}{2 dt}\right)$

Before moving on, we want to generalize this equation to the presence of *external* forces. As we saw previously, this just results in adding a *constant velocity motion* to the particle, leading to the full **Langevin equation**:

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

$$dx(t) = f(x,t) dt + \sqrt{2D(x,t)} dB \qquad dB \sim \frac{1}{\sqrt{2\pi} dt} \exp\left(-\frac{dB^2}{2 dt}\right)$$
(11)

The *physical* meaning of f(x,t) and D(x,t) can be more clearly seen by comparing (11) to the equation of motion of the Brownian particle.

Consider a particle of mass m immersed in a fluid, with a radius a that is much larger than the surrounding molecules (typically $\sim 10^{-9}$ to 10^{-7} m). The forces acting on it will be that of viscous friction $-\gamma \dot{\boldsymbol{r}}$, eventual external forces $\boldsymbol{F}_{\rm ext}$ (e.g. gravity), and a rapidly varying and random term $\boldsymbol{F}_{\rm noise}$, encompassing the effect of the large number of collisions ($\sim 10^{12}/{\rm s}$) with the smaller fluid particles:

$$m\ddot{\boldsymbol{r}}(t) = -\gamma\dot{\boldsymbol{r}} + \boldsymbol{F}_{\mathrm{ext}} + \boldsymbol{F}_{\mathrm{noise}}(t)$$

Dividing both sides by γ :

$$\frac{m}{\gamma}\ddot{\boldsymbol{r}}(t) = -\dot{\boldsymbol{r}} + \frac{\boldsymbol{F}_{\text{ext}}(\boldsymbol{r},t)}{\gamma} + \frac{\boldsymbol{F}_{\text{noise}}(t)}{\gamma}$$
(12)

Assuming a spherical particle, γ is given by Stokes law to be $6\pi a\eta$, where η is the viscosity of the surrounding fluid.

Note that, if we ignore the external force and the random term, the equation becomes:

$$\frac{\mathrm{d}\dot{\boldsymbol{r}}(t)}{\mathrm{d}t} = -\frac{\gamma}{m}v(t)$$

which has solution:

$$\dot{\boldsymbol{r}}(t) = \exp\left(-\frac{t}{\tau_B}\right)\dot{\boldsymbol{r}}(0) \qquad \tau_B = \frac{m}{\gamma}$$

 τ_B is in the scale of 10^{-3} s, and represents the timescale of reaching equilibrium, i.e. 0 velocity. So, for Brownian motion to happen, $\mathbf{F}_{\text{noise}}$ is necessary. Also, if we are interested in the motion on the scale of seconds, we can neglect the acceleration term. This is the **overdamped limit** (in analogy to a damped oscillator with high loss of energy due to attrition, so that it quickly reaches equilibrium without ever "overshooting"). Given that assumption, (12) becomes:

$$\dot{m{r}} = rac{m{F}_{ ext{ext}}}{\gamma} + rac{m{F}_{ ext{noise}}}{\gamma}$$

Which, for a particle moving in one dimension, reduces to:

$$x(t) = \underbrace{\frac{F_{\text{ext}}}{\gamma}}_{f(x,t)} + \underbrace{\frac{F_{\text{noise}}}{\gamma}}_{\sqrt{2D(x,t)}\xi(t)}$$

Comparing with (11) gives the physical meaning of f(x,t) and D(x,t).