

Exponentially growing/decaying steps

Consider

$$a_m = a^m \quad \text{with } a = \text{const.}$$

For $a > 1$ the steps are growing while for $a < 1$ the steps are decaying.

$$C_{N,l} = c_l \sum_{m=1}^N a_m^l = c_l \sum_{m=1}^N a^{ml} = c_l \left(\frac{1 - a^{l(N+1)}}{1 - a^l} - 1 \right) = c_l a^l \left(\frac{1 - a^{lN}}{1 - a^l} \right).$$

The asymptotic behavior of the cumulants is given by

$$C_{N,l} \sim \begin{cases} c_l \frac{a^l}{1-a^l} & a < 1 \\ c_l \frac{a^{l(N+1)}}{a^l - 1} & a > 1 \end{cases}.$$

The common step $\widetilde{\Delta x}$ may have different PDFs. In what follows we will consider two examples.

Gaussian steps with $std = \sigma$

We already showed that the convolution of Gaussian PDFs yields a Gaussian PDF. Therefore, we know the exact PDF of the position after N steps.

$$\ln(\widetilde{p}_N(k)) = -\frac{k^2}{2} C_{N,2} = -\frac{k^2}{2} \sigma^2 a^2 \left(\frac{1 - a^{2N}}{1 - a^2} \right);$$

Defining a scaled variable

$$\xi = \frac{x_N}{\sigma a \sqrt{\frac{1 - a^{2N}}{1 - a^2}}};$$

we find that its PDF is

$$\begin{aligned} p_\xi(\xi) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{\xi^2}{2}}; \\ p_x(x_N) &= \frac{1}{\sigma a \sqrt{\frac{1 - a^{2N}}{1 - a^2}} \sqrt{2\pi}} e^{-\frac{x_N^2}{2\sigma^2 a^2 \frac{1 - a^{2N}}{1 - a^2}}}; \\ \langle x_N^2 \rangle &= \sigma^2 a^2 \frac{1 - a^{2N}}{1 - a^2} \sim \begin{cases} \sigma^2 \frac{a^{2(N+1)}}{a^2 - 1} & a > 1 \\ \frac{\sigma^2 a^2}{1 - a^2} & a < 1 \end{cases}. \end{aligned}$$

For $a > 1$ we have anomalous diffusion and the MSD grows exponentially with the number of steps while for $a < 1$ the MSD is bounded and does not grow indefinitely with the number of steps. However, for all values of a the PDF is Gaussian.

A Simple Walk with Decaying Steps

In this example we consider a simple RW with decaying steps. Namely,

$$x_N = \sum_{m=1}^N a^m \varepsilon_m;$$

with $a < 1$ and ε_m being a random variable taking the values 1 and -1 with the same probability (1/2). The PDF of the m 'th step is given by

$$p_{x_m}(x_m) = \frac{1}{2} (\delta(x_m - a^m) + \delta(x_m + a^m)).$$

The corresponding characteristic function is:

$$\tilde{p}_{x_m}(k) = \cos(ka^m).$$

From the convolution theorem we find

$$\tilde{p}_N(k, a) = \prod_{m=1}^N \cos(ka^m) = \cos(ka) \cos(ka^2) \dots \cos(ka^N).$$

We can also see that

$$\tilde{p}_N(k, a) = \tilde{p}_{\text{ceiling}(N/2)}(k/a, a^2) \tilde{p}_{\text{floor}(N/2)}(k, a^2) = \prod_{m=1}^{\text{ceiling}(N/2)} \cos\left(\frac{k}{a} a^{2m}\right) \prod_{r=1}^{\text{floor}(N/2)} \cos(ka^{2r}),$$

and, in particular,

$$\begin{aligned} \tilde{p}_\infty(k, a) &= \tilde{p}_\infty(k/a, a^2) \tilde{p}_\infty(k, a^2) = \prod_{m=1}^{\infty} \cos\left(\frac{k}{a} a^{2m}\right) \prod_{r=1}^{\infty} \cos(ka^{2r}). \\ \tilde{p}_\infty(k, a) &= \tilde{p}_\infty(k/a, a^2) \tilde{p}_\infty(k, a^2) = \tilde{p}_\infty(k/a^2, a^3) \tilde{p}_\infty(k/a, a^3) \tilde{p}_\infty(k, a^2) \\ &= \tilde{p}_\infty(k/a^2, a^3) \tilde{p}_\infty(k/a, a^3)^2 \tilde{p}_\infty(k, a^3) \end{aligned}$$

For the case of $a = 1/2$ we can solve exactly.

$$\begin{aligned} \tilde{p}_N(k, 1/2) &= \prod_{m=1}^N \cos\left(k \frac{1}{2^m}\right) = \cos\left(k \frac{1}{2^1}\right) \cos\left(k \frac{1}{2^2}\right) \cos\left(k \frac{1}{2^3}\right) \dots \cos\left(k \frac{1}{2^N}\right) \\ &= \frac{\sin(k)}{2 \sin(k/2)} \frac{\sin(k/2)}{2 \sin(k/4)} \frac{\sin(k/4)}{2 \sin(k/8)} \dots \frac{\sin(k/2^{N-1})}{2 \sin(k/2^N)} = \frac{\sin(k)}{2^N \sin(k/2^N)} \\ &\sim \frac{\sin(k)}{k} \end{aligned}$$

We used

$$\cos(a/2) = \frac{\sin(a)}{2 \sin(a/2)}.$$

The inverse Fourier transform of the sinc function is the uniform distribution in the interval $[-1, 1]$,

$$p_\infty(x, 1/2) = \begin{cases} 1/2 & -1 < x < 1 \\ 0 & \text{else} \end{cases}.$$

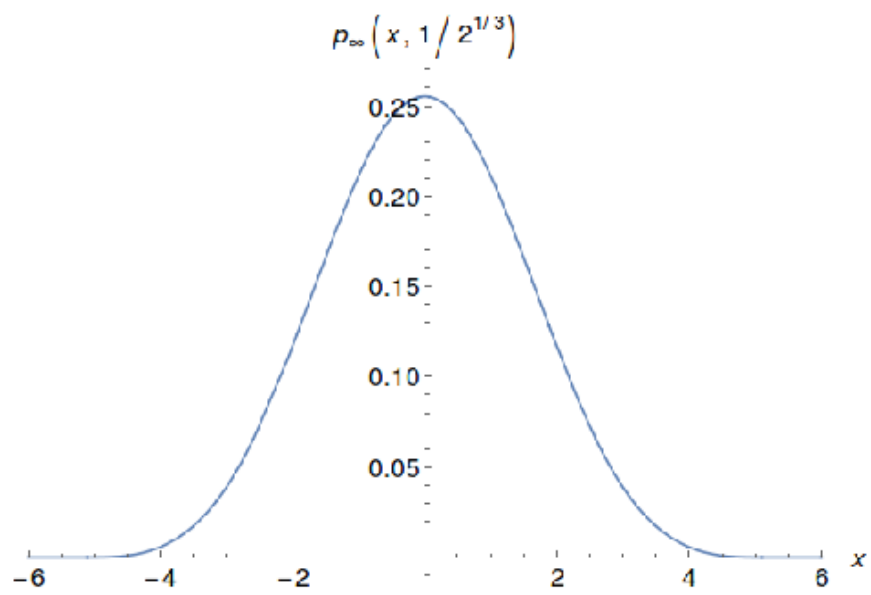
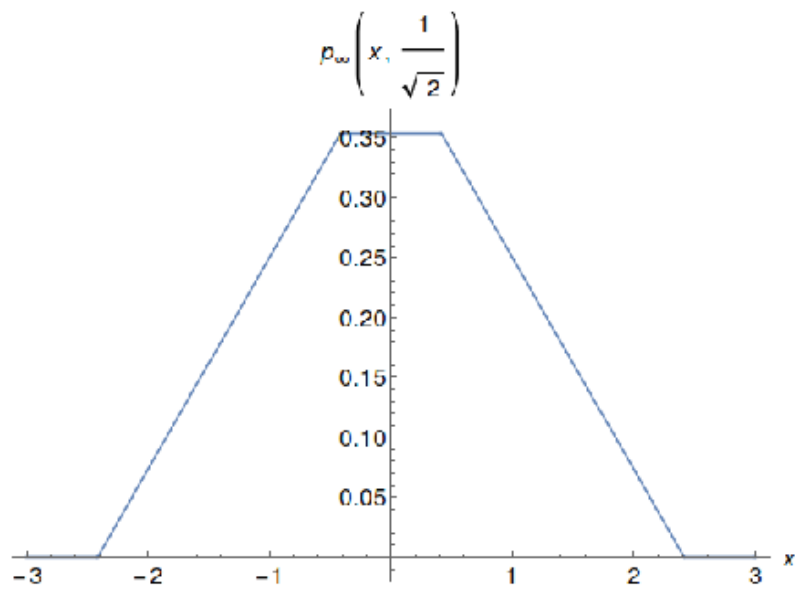
$$\frac{1}{2} \int_{-\infty}^{\infty} e^{-ikx} \Theta(1+x) \Theta(1-x) dx = \frac{1}{2} \int_{-1}^1 e^{-ikx} dx = \frac{e^{-ik} - e^{ik}}{2(-ik)} = \frac{\sin(k)}{k}.$$

Since we know the PDF for $a = 1/2$ we can find the PDF for $a = 1/2^{1/m}$ (with m being an integer). For example for $a = 1/\sqrt{2}$ we can write

$$\begin{aligned} \tilde{p}_\infty(k, 1/\sqrt{2}) &= \tilde{p}_\infty(\sqrt{2}k, 1/2) \tilde{p}_\infty(k, 1/2) \\ &\rightarrow p_\infty(x, 1/\sqrt{2}) = p_\infty\left(\frac{x}{\sqrt{2}}, 1/2\right) * p_\infty(x, 1/2). \end{aligned}$$

(the $*$ denotes the convolution). Therefore, the PDF for $a = 1/\sqrt{2}$ is given by the convolution of two step functions, one in the interval $[-1, 1]$ and one in the interval $[-\sqrt{2}, \sqrt{2}]$.

$$p_\infty(x, 1/\sqrt{2}) = \frac{1}{4\sqrt{2}} \int_{-\infty}^{\infty} \Theta(u-1) \Theta(u+1) \Theta(x-u-\sqrt{2}) \Theta(x-u+\sqrt{2}) du$$



Note that it is also possible to find a recursion relation without using the Fourier transform. Let's define as $p_\infty^s(x, a)$ the PDF for the case when the walker start their steps as a^2 rather than a (thus step sizes go like $a^2, a^3 \dots$). This PDF should look like $p_\infty(x, a)$ except that the distance is scaled by a factor of a . The following relation between the probabilities hold

$$p_\infty\left(\frac{x}{a}, a\right) \frac{dx}{a} = p_\infty^s(x, a) dx.$$

Now we should remember that $p_\infty(x, a)$ is just the average of $p_\infty^s(x - a, a)$ and $p_\infty^s(x + a, a)$ because after the first step the walker is either in $+a$ or in $-a$. Therefore we find

$$p_\infty(x, a) = \frac{1}{2} (p_\infty^s(x - a, a) + p_\infty^s(x + a, a)) = \frac{1}{2a} \left(p_\infty\left(\frac{x - a}{a}, a\right) + p_\infty\left(\frac{x + a}{a}, a\right) \right).$$

Another way to understand the above relation is to look at the Fourier transform

$$\tilde{p}_\infty(k, a) = \prod_{m=1}^{\infty} \cos(ka^m) = \cos(ka) \prod_{m=1}^{\infty} \cos((ka) a^m) = \cos(ka) \tilde{p}_\infty(ka, a) = \frac{1}{2} (e^{-ika} + e^{ika}) \tilde{p}_\infty(ka, a),$$

which implies that

$$p_\infty(x, a) = \frac{1}{2a} \left(p_\infty\left(\frac{x - a}{a}, a\right) + p_\infty\left(\frac{x + a}{a}, a\right) \right).$$

Note that we could also write the more general relation for any N

$$\tilde{p}_{N+1}(k, a) = \prod_{m=1}^{N+1} \cos(ka^m) = \cos(ka) \prod_{m=1}^N \cos((ka) a^m) = \cos(ka) \tilde{p}_N(ka, a) = \frac{1}{2} (e^{-ika} + e^{ika}) \tilde{p}_N(ka, a),$$

and

$$p_{N+1}(x, a) = \frac{1}{2a} \left(p_N\left(\frac{x - a}{a}, a\right) + p_N\left(\frac{x + a}{a}, a\right) \right).$$

Correlated walks

The correlation of two variables is defined as

$$\rho(x, y) = \rho(y, x) = \frac{\langle (x - \langle x \rangle)(y - \langle y \rangle) \rangle}{\sigma_x \sigma_y},$$

$$\sigma_x^2 = \langle (x - \langle x \rangle)^2 \rangle.$$

The correlation coefficient is in the range

$$-1 \leq \rho(x, y) \leq 1.$$

****Proof

$$\begin{aligned} \text{Var}(tx + y) &= \langle (tx + y - \langle tx + y \rangle)^2 \rangle \geq 0 \\ \text{Var}(tx + y) &= \langle (tx + y)^2 - \langle tx + y \rangle^2 \rangle = t^2 \langle x^2 \rangle + \langle y^2 \rangle + 2t \langle xy \rangle - t^2 \langle x \rangle^2 - \langle y \rangle^2 - 2t \langle x \rangle \langle y \rangle \\ &= t^2 \text{Var}(x) + \text{Var}(y) + 2t \text{Cov}(x, y) \end{aligned}$$

Denoting

$$\begin{aligned} A &= \text{Var}(x); \\ B &= 2\text{Cov}(x, y); \\ C &= \text{Var}(y); \end{aligned}$$

we find that the quadratic polynomial of t is always non-negative and that occurs just if the discriminant is non-positive, i.e.,

$$\begin{aligned} B^2 - 4AC &\leq 0; \\ 4Cov(x, y)^2 - 4Var(x)Var(y) &\leq 0; \\ Cov(x, y)^2 &\leq Var(x)Var(y) \\ &\rightarrow -1 \leq \rho(x, y) \leq 1. \end{aligned}$$

When $\rho = 0$ the variables are not correlated because $\langle \Delta x \Delta y \rangle = \langle \Delta x \rangle \langle \Delta y \rangle$. When $\rho = 1$, the variables are perfectly correlated, i.e., $\Delta x / \sigma_x$ is an exact copy of $\Delta y / \sigma_y$. When $\rho = -1$, the variables are perfectly anti-correlated, i.e., $\Delta x / \sigma_x$ is an exact copy of $-\Delta y / \sigma_y$.

Correlation function is defined as

$$C(m, m') = \langle \Delta \vec{x}_m \Delta \vec{x}_{m'} \rangle = Cov(\Delta \vec{x}_m, \Delta \vec{x}_{m'}),$$

where m, m' are the numbers of steps taken along the path of a random walker ($m \rightarrow m+1 \rightarrow m+2 \dots m'$).

Denoting the total displacement as

$$\vec{X}_N = \sum_{m=1}^N \Delta \vec{x}_m,$$

we can write the average squared displacement as

$$\langle \vec{X}_N^2 \rangle = \left\langle \sum_{m=1}^N \Delta \vec{x}_m \sum_{m'=1}^N \Delta \vec{x}_{m'} \right\rangle = \sum_{m=1}^N \sum_{m'=1}^N \langle \Delta \vec{x}_m \Delta \vec{x}_{m'} \rangle = \sum_{m=1}^N \sum_{m'=1}^N C(m, m').$$

If we assume that the correlation function is translation invariant, i.e.,

$$C(m, m') = C(|m - m'|) \quad \text{or} \quad C(m) = \langle \Delta \vec{x}_{m+m_0} \Delta \vec{x}_{m_0} \rangle \quad \text{for any } m_0,$$

we can write the MSD as

$$\begin{aligned} \langle \vec{X}_N^2 \rangle &= \sum_{m=1}^N \sum_{m'=1}^N C(m, m') = NC(0) + 2(N-1)C(1) + 2(N-2)C(2) + \dots + 2(1)C(N-1) \\ &= NC(0) + 2 \sum_{k=1}^{N-1} (N-k)C(k). \end{aligned}$$

In the continuum limit when $t' = k\tau$ and $t = N\tau$,

$$\frac{C(m)}{\tau^2} = \left\langle \frac{\Delta \vec{x}_{m_0}}{\tau} \cdot \frac{\Delta \vec{x}_{m+m_0}}{\tau} \right\rangle = \langle \vec{v}(0) \cdot \vec{v}(t) \rangle,$$

the velocity auto-correlation function. Using this function we can express the MSD as

$$\langle \vec{X}_t^2 \rangle = \langle \vec{v}^2(0) \rangle t\tau + 2 \int_0^t \langle \vec{v}(0) \cdot \vec{v}(t') \rangle (t-t') dt'$$

We define the diffusion coefficient as

$$D = \lim_{t \rightarrow \infty} \frac{1}{2} \frac{d}{dt} \langle \vec{X}_t^2 \rangle = \langle \vec{v}^2(0) \rangle \tau + \int_0^t \langle \vec{v}(0) \cdot \vec{v}(t') \rangle dt'.$$

If the correlation time of the velocity is much larger than the time between consecutive steps we can neglect the first term and get the Kubo formula

$$D = \int_0^t \langle \vec{v}(0) \cdot \vec{v}(t') \rangle dt'.$$

Back to the RW, the asymptotic dependence of D on the number of steps determines the type of diffusion we get.

If $D \sim \text{const}$ the RW corresponds to normal diffusion. $\langle \vec{X}_N^2 \rangle \sim N$ or $\langle \vec{X}_t^2 \rangle \sim (2d)Dt$.

If $D \sim \text{diverges}$, the RW corresponds to super-diffusion, i.e., $\langle \vec{X}_N^2 \rangle \sim N^{1+\mu}$ with $0 < \mu \leq 1$ or $\langle \vec{X}_t^2 \rangle \sim t^{1+\mu}$.

If $D \sim \text{vanishes}$, the RW corresponds to sub-diffusion, i.e., $\langle \vec{X}_N^2 \rangle \sim N^\mu$ with $0 < \mu \leq 1$ or $\langle \vec{X}_t^2 \rangle \sim t^\mu$.