

# Notes on Position Distributions of a Walker

Theophilus Dwamena Frimpong

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## 1 Simple Diffusion Process

Consider a particle undergoing simple diffusion in one dimension. The probability density function  $P(x, t)$  satisfies the diffusion equation:

$$\frac{\partial P(x, t)}{\partial t} = D \frac{\partial^2 P(x, t)}{\partial x^2}, \quad (1)$$

with the initial condition:

$$P(x, 0) = \delta(x). \quad (2)$$

We apply the Fourier transform:

$$\tilde{P}(k, t) = \int_{-\infty}^{\infty} e^{ikx} P(x, t) dx$$

Taking the transform of both sides of equation (1), we get:

$$\frac{\partial \tilde{P}(k, t)}{\partial t} = -Dk^2 \tilde{P}(k, t)$$

This is an ordinary differential equation in time  $t$ , with solution:

$$\tilde{P}(k, t) = \tilde{P}(k, 0) e^{-Dk^2 t}$$

Since  $P(x, 0) = \delta(x)$ , it follows that  $\tilde{P}(k, 0) = 1$ , so:

$$\tilde{P}(k, t) = e^{-Dk^2 t}$$

Now we apply the inverse Fourier transform to find  $P(x, t)$ :

$$P(x, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-ikx} \tilde{P}(k, t) dk \quad (3)$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-ikx} e^{-Dk^2 t} dk \quad (4)$$

This integral is a standard Gaussian integral of the form:

$$\int_{-\infty}^{\infty} e^{-ak^2+ibk} dk = \sqrt{\frac{\pi}{a}} e^{-b^2/(4a)} \quad \text{for } a > 0$$

We identify  $a = Dt$ ,  $b = x$ . Then:

$$P(x, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-Dk^2t-ikx} dk \tag{5}$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-Dt(k^2 + \frac{ikx}{Dt})} dk \tag{6}$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-Dt(k + \frac{ix}{2Dt})^2} e^{-x^2/(4Dt)} dk \tag{7}$$

Now make the substitution  $u = k + \frac{ix}{2Dt}$ , so  $dk = du$ , and the limits remain  $(-\infty, \infty)$ :

$$P(x, t) = \frac{1}{2\pi} e^{-x^2/(4Dt)} \int_{-\infty}^{\infty} e^{-Dtu^2} du = \frac{1}{2\pi} e^{-x^2/(4Dt)} \sqrt{\frac{\pi}{Dt}}$$

Thus:

$$P(x, t) = \frac{1}{\sqrt{4\pi Dt}} \exp\left(-\frac{x^2}{4Dt}\right)$$

We now plot this analytic against simulations as shown in [Figure 1](#).

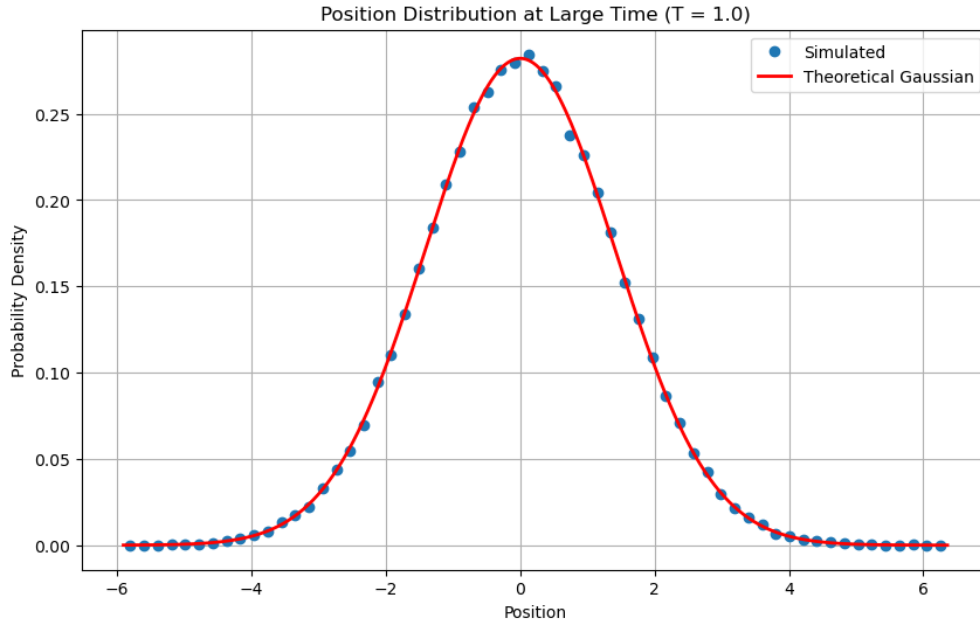


Figure 1: Probability distribution of positions for simple diffusion process with  $D = 1$  and large time  $t = 500$ .

[Python notebook.](#)

## 2 Simple Diffusion with Resetting

We start with the Master Equation (Fokker-Planck equation) for the probability  $p(x, t|x_0)$  [2]:

$$\frac{\partial p(x, t|x_0)}{\partial t} = D \frac{\partial^2 p(x, t|x_0)}{\partial x^2} - rp(x, t|x_0) + r\delta(x - x_0) \quad (8)$$

where  $D$  is the diffusion constant,  $r$  is the resetting rate, and  $\delta(x - x_0)$  is the Dirac delta function representing the resetting to the initial position  $x_0$ .

To find the stationary position distribution, denoted as  $p_{st}(x|x_0)$ , we set the time derivative to zero, as the distribution does not change in the steady state:

$$\frac{\partial p(x, t|x_0)}{\partial t} = 0$$

Substituting this into equation (8), we obtain:

$$0 = D \frac{d^2 p_{st}(x|x_0)}{dx^2} - rp_{st}(x|x_0) + r\delta(x - x_0)$$

Rearranging the terms, we get a second-order ordinary differential equation:

$$D \frac{d^2 p_{st}(x|x_0)}{dx^2} - rp_{st}(x|x_0) = -r\delta(x - x_0) \quad (9)$$

Dividing by  $D$ , we have:

$$\frac{d^2 p_{st}(x|x_0)}{dx^2} - \frac{r}{D} p_{st}(x|x_0) = -\frac{r}{D} \delta(x - x_0)$$

Let's define a characteristic inverse length scale  $\alpha_0$  as:

$$\alpha_0 = \sqrt{\frac{r}{D}} \quad (10)$$

Substituting this into the differential equation, we get:

$$\frac{d^2 p_{st}(x|x_0)}{dx^2} - \alpha_0^2 p_{st}(x|x_0) = -\alpha_0^2 \delta(x - x_0) \quad (11)$$

This is a non-homogeneous linear second-order differential equation. Consider the homogeneous part of the equation:

$$\frac{d^2 p_{st,h}(x)}{dx^2} - \alpha_0^2 p_{st,h}(x) = 0$$

The characteristic equation is  $m^2 - \alpha_0^2 = 0$ , which yields roots  $m = \pm\alpha_0$ . The general solution to the homogeneous equation is:

$$p_{st,h}(x) = Ae^{\alpha_0 x} + Be^{-\alpha_0 x}$$

The physical requirement is that the probability distribution  $p_{st}(x|x_0)$  must decay to zero as  $|x| \rightarrow \infty$ . This implies that for  $x > x_0$ , the  $e^{\alpha_0 x}$  term must vanish ( $A = 0$ ), and for  $x < x_0$ , the  $e^{-\alpha_0 x}$  term must vanish ( $B = 0$ ). Therefore, we propose a solution of the form:

$$p_{st}(x|x_0) = Ce^{-\alpha_0 |x - x_0|}$$

This form ensures that the probability decays exponentially away from the resetting point  $x_0$ . We can write this explicitly for the two regions:

$$p_{st}(x|x_0) = \begin{cases} Ce^{-\alpha_0(x-x_0)} & \text{for } x > x_0 \\ Ce^{\alpha_0(x-x_0)} & \text{for } x < x_0 \end{cases}$$

The solution  $p_{st}(x|x_0)$  must be continuous at  $x = x_0$ . For  $x = x_0$ :  $Ce^{-\alpha_0(x_0-x_0)} = Ce^{\alpha_0(x_0-x_0)} \implies C = C$ . This condition is satisfied.

The presence of the Dirac delta function on the right-hand side of equation (11) implies a discontinuity in the first derivative of  $p_{st}(x|x_0)$  at  $x = x_0$ . To find this discontinuity, we integrate equation (11) from  $x_0 - \epsilon$  to  $x_0 + \epsilon$  and then take the limit as  $\epsilon \rightarrow 0$ :

$$\int_{x_0-\epsilon}^{x_0+\epsilon} \left( \frac{d^2 p_{st}}{dx^2} - \alpha_0^2 p_{st} \right) dx = \int_{x_0-\epsilon}^{x_0+\epsilon} (-\alpha_0^2 \delta(x-x_0)) dx$$

Applying the fundamental theorem of calculus to the first term and noting that  $\int_{x_0-\epsilon}^{x_0+\epsilon} p_{st} dx \rightarrow 0$  as  $\epsilon \rightarrow 0$  (since  $p_{st}$  is continuous), and  $\int_{x_0-\epsilon}^{x_0+\epsilon} \delta(x-x_0) dx = 1$ , we get:

$$\left[ \frac{dp_{st}}{dx} \right]_{x_0-\epsilon}^{x_0+\epsilon} - \alpha_0^2 \lim_{\epsilon \rightarrow 0} \int_{x_0-\epsilon}^{x_0+\epsilon} p_{st} dx = -\alpha_0^2 \cdot 1$$

$$\left. \frac{dp_{st}}{dx} \right|_{x_0^+} - \left. \frac{dp_{st}}{dx} \right|_{x_0^-} = -\alpha_0^2$$

Now we calculate the derivatives of our proposed solution in the two regions: For  $x > x_0$ :

$$\frac{dp_{st}}{dx} = \frac{d}{dx} \left( Ce^{-\alpha_0(x-x_0)} \right) = -C\alpha_0 e^{-\alpha_0(x-x_0)}$$

So, at  $x = x_0^+$ :

$$\left. \frac{dp_{st}}{dx} \right|_{x_0^+} = -C\alpha_0 e^{-\alpha_0(x_0-x_0)} = -C\alpha_0$$

For  $x < x_0$ :

$$\frac{dp_{st}}{dx} = \frac{d}{dx} \left( Ce^{\alpha_0(x-x_0)} \right) = C\alpha_0 e^{\alpha_0(x-x_0)}$$

So, at  $x = x_0^-$ :

$$\left. \frac{dp_{st}}{dx} \right|_{x_0^-} = C\alpha_0 e^{\alpha_0(x_0-x_0)} = C\alpha_0$$

Substitute these into the jump condition:

$$(-C\alpha_0) - (C\alpha_0) = -\alpha_0^2$$

$$-2C\alpha_0 = -\alpha_0^2$$

Solving for  $C$ :

$$C = \frac{-\alpha_0^2}{-2\alpha_0} = \frac{\alpha_0}{2}$$

Substituting the value of  $C$  back into our proposed solution, we obtain the stationary position distribution:

$$p_{st}(x|x_0) = \frac{1}{2} \cdot \sqrt{\frac{r}{D}} \cdot \exp \left( -\sqrt{\frac{r}{D}} \cdot |x - x_0| \right)$$

We now plot this analytic against simulations for  $r = \{0, 0.01, 0.025, 0.1\}$  as shown in Figure 2. It is important to note that for  $r=0$ , we have a simple diffusion with no resetting hence:

$$P(x, T) = \frac{1}{\sqrt{4\pi DT}} \exp \left( -\frac{(x-x_0)^2}{4DT} \right) \quad (12)$$

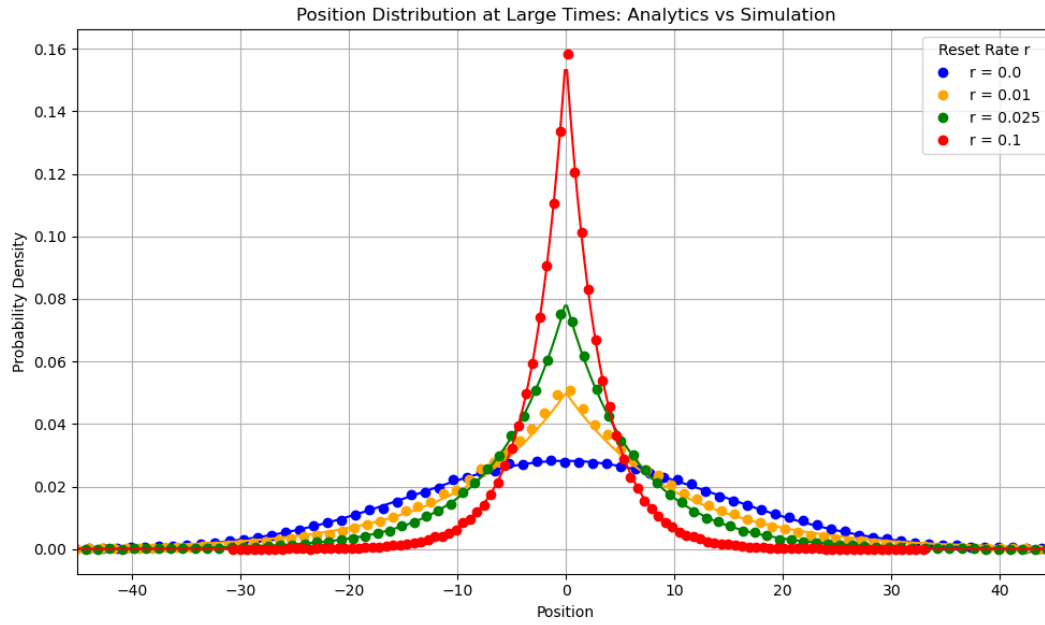


Figure 2: Probability distribution of positions of simple diffusion processes with resetting for different reset probabilities  $r = \{0, 0.01, 0.025, 0.1\}$  based on  $10^6$  simulations.

[Python notebook.](#)

### 3 Sluggish Random Walk

Recall the Fokker–Planck equation with a position-dependent diffusion coefficient [1]:

$$\frac{\partial P(x, t)}{\partial t} = \frac{\partial^2}{\partial x^2} \left[ \frac{1}{|x|^\alpha} P(x, t) \right], \quad (13)$$

with the initial condition:

$$P(x, 0) = \delta(x - x_0).$$

We propose a scaling form:

$$P(x, t) = \frac{1}{t^\mu} H\left(\frac{x}{t^\mu}\right), \quad (14)$$

where  $\mu$  is a scaling exponent to be determined and  $z = \frac{x}{t^\mu}$  is the scaling variable.

We begin with LHS of eqn (13):

$$\begin{aligned} (14) \implies \frac{\partial P(x, t)}{\partial t} &= \frac{d}{dt} \left( \frac{1}{t^\mu} \cdot H\left(\frac{x}{t^\mu}\right) \right) \\ &= -\mu t^{-\mu-1} \cdot H\left(\frac{x}{t^\mu}\right) + t^{-\mu} \cdot H'\left(\frac{x}{t^\mu}\right) \cdot \frac{d}{dt} \left( \frac{x}{t^\mu} \right) && \text{Product and chain rules} \\ &= -\frac{\mu}{t^{\mu+1}} H(z) + t^{-\mu} \cdot H'(z) \cdot (x(-\mu)t^{-\mu-1}) && \text{since } z = \frac{x}{t^\mu} \\ &= -\frac{\mu}{t^{\mu+1}} H(z) - \frac{\mu z}{t^{\mu+1}} H'(z) \\ \frac{\partial P(x, t)}{\partial t} &= \frac{-\mu}{t^{\mu+1}} \left[ H(z) + z H'(z) \right]. \end{aligned} \quad (15)$$

For the right-hand side of equation (13):

$$\begin{aligned} \frac{\partial^2}{\partial x^2} \left[ \frac{1}{|x|^\alpha} P(x, t) \right] &= \frac{\partial^2}{\partial x^2} \left[ \frac{1}{|x|^\alpha} \cdot \frac{1}{t^\mu} \cdot H\left(\frac{x}{t^\mu}\right) \right] \\ \implies \frac{1}{|x|^\alpha} P(x, t) &= \frac{1}{t^{\mu(1+\alpha)}} \cdot \frac{1}{|z|^\alpha} \cdot H(z) \end{aligned}$$

For the second derivative with respect to  $x$ , we use the chain rule:

$$\frac{d}{dx} = \frac{dz}{dx} \cdot \frac{d}{dz} = \frac{1}{t^\mu} \cdot \frac{d}{dz} \implies \frac{d^2}{dx^2} = \frac{1}{t^\mu} \cdot \frac{d}{dz} \left( \frac{1}{t^\mu} \cdot \frac{d}{dz} \right) = \frac{1}{t^{2\mu}} \cdot \frac{d^2}{dz^2}$$

Thus:

$$\frac{\partial^2}{\partial x^2} \left[ \frac{1}{|x|^\alpha} P(x, t) \right] = \frac{1}{t^{\mu(1+\alpha)}} \cdot \frac{1}{t^{2\mu}} \cdot \frac{d^2}{dz^2} \left[ \frac{1}{|z|^\alpha} \cdot H(z) \right] = \frac{1}{t^{\mu(\alpha+3)}} \cdot \frac{d^2}{dz^2} \left[ \frac{1}{|z|^\alpha} \cdot H(z) \right] \quad (16)$$

Equating expression (15) with (16), we obtain:

$$-\frac{\mu}{t^{\mu+1}} \left[ H(z) + z H'(z) \right] = \frac{1}{t^{\mu(\alpha+3)}} \cdot \frac{d^2}{dz^2} \left[ \frac{1}{|z|^\alpha} \cdot H(z) \right] \quad (17)$$

For this to hold, powers of  $t$  must be equal. Thus  $\mu + 1 = \mu(\alpha + 3) \implies 1 = \mu(\alpha + 2)$

$$\implies \boxed{\mu = \frac{1}{\alpha + 2}} \quad (18)$$

We now derive the ODE for  $H(z)$ . Recall:

$$P(x, t) = \frac{1}{t^\mu} H(z), \quad \mu = \frac{1}{\alpha + 2}, \quad z = \frac{x}{t^\mu}$$

From equation (17):

$$-\mu \left[ H(z) + zH'(z) \right] = \frac{d^2}{dz^2} \left[ \frac{1}{|z|^\alpha} H(z) \right] \quad (19)$$

Let:

$$Q(z) = \frac{1}{|z|^\alpha} H(z),$$

then:

$$\begin{aligned} \frac{dQ}{dz} &= -\alpha \frac{\text{sgn}(z)}{|z|^{\alpha+1}} H(z) + \frac{1}{|z|^\alpha} H'(z) \\ \frac{d^2Q}{dz^2} &= \alpha(\alpha-1) \cdot \frac{1}{|z|^{\alpha+2}} \cdot H(z) - 2\alpha \cdot \frac{1}{|z|^{\alpha+1}} \cdot H'(z) + \frac{1}{|z|^\alpha} \cdot H''(z) \end{aligned}$$

$\implies$  (19) becomes:

$$\begin{aligned} -\mu \left[ H(z) + zH'(z) \right] &= \left[ \alpha(\alpha-1) \frac{1}{|z|^{\alpha+2}} \cdot H(z) - 2\alpha \cdot \frac{1}{|z|^{\alpha+1}} \cdot H'(z) + \frac{1}{|z|^\alpha} H''(z) + H''(z) \right] \\ -\mu \left[ H(z) + zH'(z) \right] &= \frac{1}{|z|^\alpha} \cdot \frac{d^2}{dz^2} \left[ \alpha(\alpha-1) \cdot \frac{1}{|z|^2} \cdot H(z) - 2\alpha \frac{1}{|z|} H'(z) + H''(z) \right] \\ \implies -\mu z^\alpha \cdot \mathcal{H}(z) - \mu z^{\alpha+1} \cdot H'(z) &= \alpha(\alpha+1) z^{-2} \cdot H(z) - 2\alpha z^{-1} \cdot H'(z) + H''(z) \\ H''(z) + \left( \mu z^{\alpha+1} - \frac{2\alpha}{z} \right) H'(z) + \left( \mu z^\alpha + \frac{\alpha(\alpha+1)}{z^2} \right) H(z) &= 0. \end{aligned} \quad (20)$$

Hence the scaling function  $H(z)$  must satisfy the nonlinear second-order ODE in equation (20). To solve this equation, we propose an ansatz for  $H(z)$  of the form:

$$H(z) = A|z|^a e^{-B|z|^b}, \quad \text{as } |z| \rightarrow \infty,$$

where  $A$ ,  $a$ ,  $b$ , and  $B$  are constants to be determined.

We compute the first and second derivatives for large  $z$  (assuming  $z > 0$  for simplicity):

$$\begin{aligned} H'(z) &= A \left( a|z|^{a-1} \text{sgn}(z) - Bb|z|^{a+b-1} \text{sgn}(z) \right) e^{-B|z|^b}, \\ H''(z) &= A e^{-B|z|^b} \left[ a(a-1)|z|^{a-2} - 2aBb|z|^{a+b-2} + B^2b^2|z|^{a+2b-2} - Bb(a+b-1)|z|^{a+b-2} \right]. \end{aligned}$$

For large  $z$ , the leading behavior is:

$$\begin{aligned} H(z) &\sim A z^a e^{-Bz^b}, \\ H'(z) &\sim -ABb z^{a+b-1} e^{-Bz^b}, \\ H''(z) &\sim AB^2b^2 z^{a+2b-2} e^{-Bz^b}. \end{aligned}$$

At large  $z$ , the dominant terms in the ODE (20) are:

$$\begin{aligned} H''(z) &\sim AB^2b^2 z^{a+2b-2} e^{-Bz^b}, \\ \mu z^{\alpha+1} H'(z) &\sim -A\mu Bb z^{a+b+\alpha} e^{-Bz^b}. \end{aligned}$$

We match the powers of  $z$  in the dominant balance:

$$a + 2b - 2 = a + b + \alpha \quad \Rightarrow \quad b = \alpha + 2.$$

Using the definition  $\mu = \frac{1}{\alpha+2}$ , we find:

$$b = \frac{1}{\mu}, \quad \text{so} \quad B = \mu^2.$$

To find  $a$ , we use:

$$a = \frac{1-2\mu}{\mu} = \alpha.$$

Putting everything together, the scaling function  $H(z)$  takes the explicit form:

$$H(z) = A|z|^{\frac{1-2\mu}{\mu}} e^{-\mu^2|z|^{1/\mu}}.$$

To determine the normalization constant  $A$ , we impose:

$$\int_{-\infty}^{\infty} H(z) dz = 1.$$

Since  $H(z)$  is an even function, we compute:

$$\int_{-\infty}^{\infty} H(z) dz = 2A \int_0^{\infty} z^{\frac{1-2\mu}{\mu}} e^{-\mu^2 z^{1/\mu}} dz.$$

Let  $u = \mu^2 z^{1/\mu} \Rightarrow z = \left(\frac{u}{\mu^2}\right)^{\mu}$ ,  $dz = \mu \left(\frac{u}{\mu^2}\right)^{\mu-1} \cdot \frac{du}{\mu^2}$ . Then,

$$\int_0^{\infty} z^{\frac{1-2\mu}{\mu}} e^{-\mu^2 z^{1/\mu}} dz = \mu^{1-2\mu} \int_0^{\infty} u^{-\mu} e^{-u} du = \mu^{1-2\mu} \Gamma(1-\mu).$$

Hence, the normalization constant is:

$$A = \frac{\mu^{1-2\mu}}{2\Gamma(1-\mu)}.$$

Hence the final expression for the scaling function is

$$\boxed{H(z) = \frac{\mu^{1-2\mu}}{2\Gamma(1-\mu)} |z|^{\frac{1-2\mu}{\mu}} \exp\left(-\mu^2 |z|^{1/\mu}\right)} \quad (21)$$

Substitute equations (18) and (21) into equation (14) to obtain a complete expression for  $P(x, t)$ . First, we substitute the expression for  $H(z)$  from equation (21) into equation (14). We replace  $z$  with  $\left(\frac{x}{t^\mu}\right)$ :

$$P(x, t) = \frac{1}{t^\mu} \left( \frac{\mu^{1-2\mu}}{2\Gamma(1-\mu)} \left| \frac{x}{t^\mu} \right|^{\frac{1-2\mu}{\mu}} e^{-\mu^2 \left| \frac{x}{t^\mu} \right|^{\frac{1}{\mu}}} \right)$$

Next, we simplify the terms involving the absolute value  $\left| \frac{x}{t^\mu} \right|$ :

$$\begin{aligned} \left| \frac{x}{t^\mu} \right|^{\frac{1-2\mu}{\mu}} &= \frac{|x|^{\frac{1-2\mu}{\mu}}}{(t^\mu)^{\frac{1-2\mu}{\mu}}} = \frac{|x|^{\frac{1-2\mu}{\mu}}}{t^{1-2\mu}} = \frac{|x|^{\frac{1}{\mu}-2}}{t^{1-2\mu}} \\ \left| \frac{x}{t^\mu} \right|^{\frac{1}{\mu}} &= \frac{|x|^{\frac{1}{\mu}}}{(t^\mu)^{\frac{1}{\mu}}} = \frac{|x|^{\frac{1}{\mu}}}{t} \end{aligned}$$

Now, we substitute these simplified terms back into the expression for  $P(x, t)$ :

$$\begin{aligned} P(x, t) &= \frac{1}{t^\mu} \frac{\mu^{1-2\mu}}{2\Gamma(1-\mu)} \frac{|x|^{\frac{1}{\mu}-2}}{t^{1-2\mu}} e^{-\mu^2 \frac{|x|^{\frac{1}{\mu}}}{t}} \\ &= \frac{\mu^{1-2\mu}}{2\Gamma(1-\mu)} \frac{|x|^{\frac{1}{\mu}-2}}{t^{1-\mu}} e^{-\mu^2 \frac{|x|^{\frac{1}{\mu}}}{t}} \\ &= \frac{\left(\frac{1}{\alpha+2}\right)^{\frac{\alpha}{\alpha+2}} |x|^\alpha}{2\Gamma\left(1 - \frac{1}{\alpha+2}\right) t^{\frac{\alpha+1}{\alpha+2}}} e^{-\left(\frac{1}{\alpha+2}\right)^2 \frac{|x|^{\alpha+2}}{t}} \end{aligned} \quad \text{substituting } \mu = \frac{1}{\alpha+2}$$



Hence,

$$P(x, t) = \frac{\left(\frac{1}{\alpha+2}\right)^{\frac{\alpha}{\alpha+2}}}{2\Gamma\left(\frac{\alpha+1}{\alpha+2}\right)} \cdot \frac{|x|^\alpha}{t^{\frac{\alpha+1}{\alpha+2}}} \cdot \exp\left(-\frac{|x|^{\alpha+2}}{(\alpha+2)^2 t}\right)$$

We now plot this analytic against simulations for  $\alpha = \{0, 0.5, 1.5\}$  as shown in [Figure 3](#).

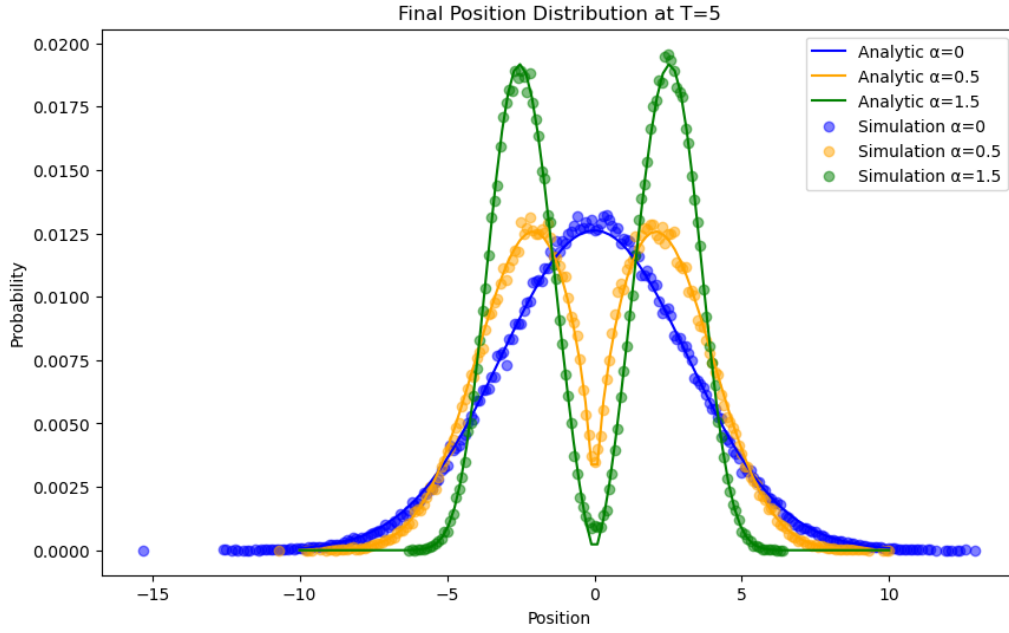


Figure 3: Probability distribution of positions of sluggish random walk for different levels of sluggishness ( $\alpha \in \{0, 0.5, 1.5\}$ ) based on  $10^6$  simulations.

[Python notebook](#)

## 4 Sluggish Random Walk with Resetting (Novel Derivation)

We propose that the probability density  $p(x, t|x_0)$  evolves according to:

$$\frac{\partial p(x, t|x_0)}{\partial t} = \frac{\partial^2}{\partial x^2} [|x|^{-\alpha} p(x, t|x_0)] - r p(x, t|x_0) + r \delta(x - x_0). \quad (22)$$

For the stationary distribution  $p_{\text{st}}(x|x_0)$ , we set  $\partial p/\partial t = 0$ :

$$0 = \frac{d^2}{dx^2} [|x|^{-\alpha} p_{\text{st}}(x|x_0)] - r p_{\text{st}}(x|x_0) + r \delta(x - x_0). \quad (23)$$

Rearranging gives the key equation:

$$\frac{d^2}{dx^2} [|x|^{-\alpha} p_{\text{st}}(x|x_0)] - r p_{\text{st}}(x|x_0) = -r \delta(x - x_0). \quad (24)$$

For resetting to the origin ( $x_0 = 0$ ), the problem has even symmetry. We solve for  $x > 0$  and extend by symmetry. Thus for  $x \neq 0$ , we solve the homogeneous equation:

$$\frac{d^2}{dx^2} [|x|^{-\alpha} p_{\text{st}}(x)] - r p_{\text{st}}(x) = 0. \quad (25)$$

Define  $q(x) \equiv |x|^{-\alpha} p_{\text{st}}(x)$ , yielding:

$$\frac{d^2 q}{dx^2} - r |x|^\alpha q(x) = 0. \quad (26)$$

To reduce this to a known form, we introduce the change of variables:

$$z = \frac{2\sqrt{r}}{\alpha+2} x^{(\alpha+2)/2}, \quad (\text{dimensionless coordinate}) \quad (27)$$

$$q(x) = x^{1/2} f(z), \quad (\text{ansatz to remove singularity}) \quad (28)$$

with derivatives:

$$\frac{dz}{dx} = \sqrt{r} x^{\alpha/2}, \quad (29)$$

$$\frac{d^2 z}{dx^2} = \frac{\alpha}{2} \sqrt{r} x^{\alpha/2-1}. \quad (30)$$

Our goal is to transform equation (25) into a differential equation for  $f(z)$  (Bessel form).

Compute  $\frac{d^2 q}{dx^2}$  using (28), (29), and (30):

$$\frac{dq}{dx} = \frac{d}{dx} (x^{1/2} f(z)) = \frac{1}{2} x^{-1/2} f(z) + x^{1/2} \frac{df}{dz} \frac{dz}{dx}, \quad (31)$$

$$\frac{d^2 q}{dx^2} = -\frac{1}{4} x^{-3/2} f(z) + x^{-1/2} \frac{df}{dz} \frac{dz}{dx} + x^{1/2} \frac{d^2 f}{dz^2} \left( \frac{dz}{dx} \right)^2 + x^{1/2} \frac{df}{dz} \frac{d^2 z}{dx^2}. \quad (32)$$

Substitute into (25) and divide by  $x^{\alpha+1/2}$ :

$$\frac{d^2 f}{dz^2} + \frac{1}{z} \frac{df}{dz} - \left( 1 + \frac{\nu^2}{z^2} \right) f(z) = 0, \quad \nu = \frac{1}{\alpha+2}. \quad (33)$$

This is the modified Bessel equation of order  $\nu$ . Its general solution is:

$$f(z) = C_1 I_\nu(z) + C_2 K_\nu(z),$$

where  $I_\nu(z)$  and  $K_\nu(z)$  are the modified Bessel functions of the first and second kind, respectively. To ensure that the stationary distribution  $p_{\text{st}}(x)$  is normalizable as  $x \rightarrow \infty$ , we must discard the exponentially growing solution  $I_\nu(z)$ . Hence we set  $C_1 = 0$ .

Therefore,

$$q(x) = x^{1/2} K_\nu \left( \frac{2\sqrt{r}}{\alpha+2} x^{(\alpha+2)/2} \right),$$

and using  $p_{\text{st}}(x) = x^\alpha q(x)$ , we obtain:

$$p_{\text{st}}(x) = x^{\alpha+1/2} K_\nu \left( \frac{2\sqrt{r}}{\alpha+2} x^{(\alpha+2)/2} \right), \quad x > 0.$$

Finally, using the even symmetry  $p_{\text{st}}(-x) = p_{\text{st}}(x)$ , the solution over the full real line is:

$$p_{\text{st}}(x) = C |x|^{\alpha+1/2} K_\nu \left( \frac{2\sqrt{r}}{\alpha+2} |x|^{(\alpha+2)/2} \right), \quad (34)$$

where the constant  $C$  is determined from normalization and the jump condition at  $x = 0$ .

We determine  $C$  using the jump condition from integrating the stationary Fokker-Planck equation (24) across an infinitesimal region around the origin. Recall (24), integrate both sides from  $-\epsilon$  to  $+\epsilon$ :

$$\int_{-\epsilon}^{\epsilon} \frac{d^2}{dx^2} [|x|^{-\alpha} p_{\text{st}}(x)] dx - \int_{-\epsilon}^{\epsilon} r p_{\text{st}}(x) dx = -r.$$

As  $\epsilon \rightarrow 0$ , the second term vanishes, and we get:

$$\frac{d}{dx} [|x|^{-\alpha} p_{\text{st}}(x)] \Big|_{x=0^+} - \frac{d}{dx} [|x|^{-\alpha} p_{\text{st}}(x)] \Big|_{x=0^-} = -r. \quad (35)$$

By symmetry, the left and right derivatives are equal in magnitude but opposite in sign, so:

$$2 \frac{d}{dx} [x^{-\alpha} p_{\text{st}}(x)] \Big|_{x=0^+} = -r.$$

For small argument  $z \rightarrow 0$ , the modified Bessel function has the following behavior:

$$K_\nu(z) \sim \frac{\Gamma(\nu)}{2} \left( \frac{z}{2} \right)^{-\nu}, \quad \nu > 0.$$

Near  $x = 0$ , the argument of the Bessel function becomes small:

$$z = \frac{2\sqrt{r}}{\alpha+2} x^{(\alpha+2)/2} \rightarrow 0.$$

Therefore,

$$K_\nu(z) \sim \frac{\Gamma(\nu)}{2} \left( \frac{2\sqrt{r}}{\alpha+2} \cdot \frac{1}{2} \right)^{-\nu} x^{-\frac{(\alpha+2)\nu}{2}} = \frac{\Gamma(\nu)}{2} \left( \frac{2\sqrt{r}}{\alpha+2} \right)^{-\nu} x^{-1/2}.$$

Then for small  $x > 0$ ,

$$p_{\text{st}}(x) \approx C x^{\alpha+1/2} \cdot x^{-1/2} = C x^\alpha \cdot \frac{\Gamma(\nu)}{2} \left( \frac{2\sqrt{r}}{\alpha+2} \right)^{-\nu} x^{-1/2} = C \cdot \frac{\Gamma(\nu)}{2} \left( \frac{2\sqrt{r}}{\alpha+2} \right)^{-\nu} x^{\alpha-1/2}.$$

Thus,

$$x^{-\alpha} p_{\text{st}}(x) \sim C \cdot \frac{\Gamma(\nu)}{2} \left( \frac{2\sqrt{r}}{\alpha+2} \right)^{-\nu} x^{-1/2}.$$

Differentiate:

$$\frac{d}{dx} [x^{-\alpha} p_{\text{st}}(x)] \sim -\frac{1}{2} C \cdot \frac{\Gamma(\nu)}{2} \left( \frac{2\sqrt{r}}{\alpha+2} \right)^{-\nu} x^{-3/2}.$$

Then as  $x \rightarrow 0^+$ , the derivative diverges like  $x^{-3/2}$ , but we can still compute the jump condition in a distributional sense. Using Eq. (35), we match:

$$2 \cdot \left( -\frac{1}{2} C \cdot \frac{\Gamma(\nu)}{2} \left( \frac{2\sqrt{r}}{\alpha+2} \right)^{-\nu} x^{-3/2} \right) = -r \quad \Rightarrow \quad C = \frac{r^{\frac{2\alpha+3}{2(\alpha+2)}}}{\Gamma\left(\frac{\alpha+1}{\alpha+2}\right)} (\alpha+2)^{-\frac{\alpha+1}{\alpha+2}}.$$

Combining everything, the normalized stationary distribution is:

$$p_{\text{st}}(x) = \frac{r^{\frac{2\alpha+3}{2(\alpha+2)}}}{\Gamma\left(\frac{\alpha+1}{\alpha+2}\right) (\alpha+2)^{\frac{\alpha+1}{\alpha+2}}} |x|^{\alpha+1/2} K_{\frac{1}{\alpha+2}} \left( \frac{2\sqrt{r}}{\alpha+2} |x|^{(\alpha+2)/2} \right) \quad (36)$$

We now plot this analytic against simulations for  $\alpha = \{0, 0.5, 1.5\}$  and reset probabilities  $r = \{0, 0.01, 0.025, 0.1\}$  as shown in Figure 4 and Figure 5.

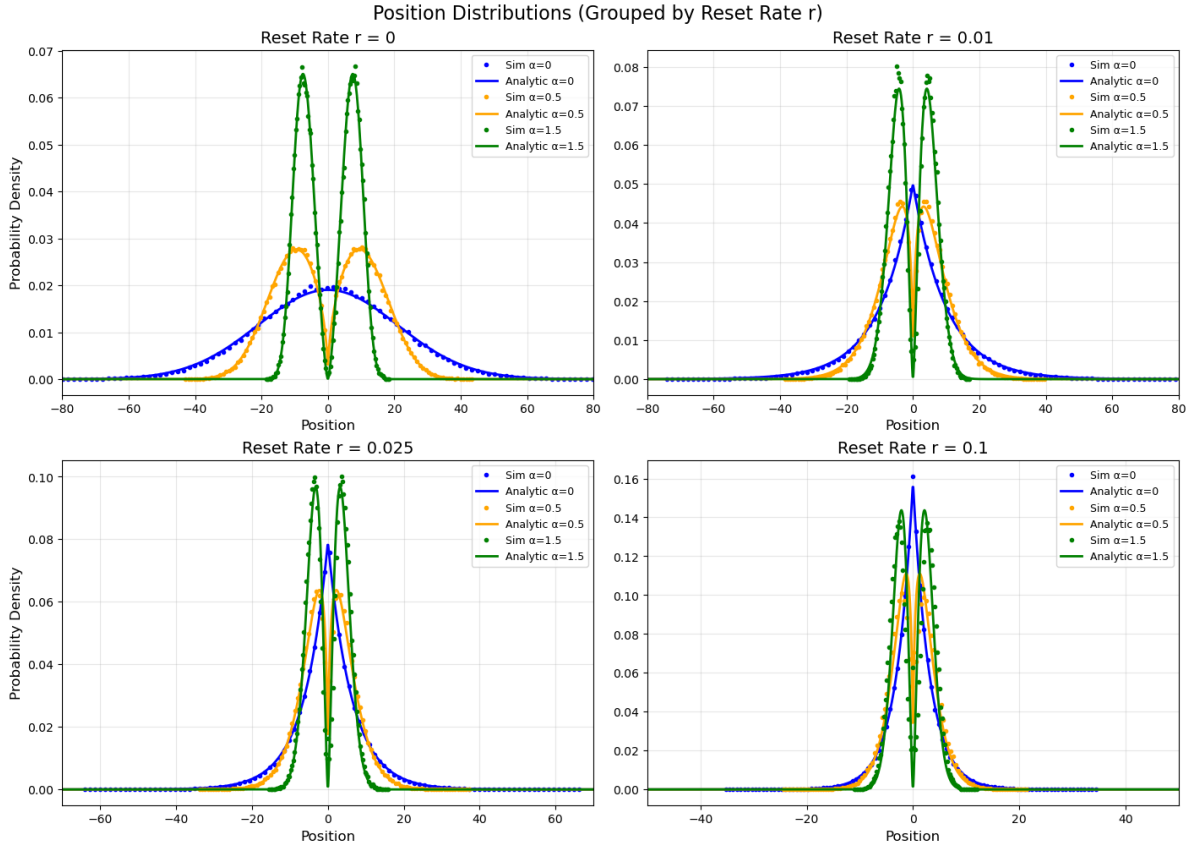


Figure 4: Probability distribution of positions of sluggish random walks with resetting grouped by reset probabilities  $r = \{0, 0.01, 0.025, 0.1\}$  for different levels of sluggishness ( $\alpha \in \{0, 0.5, 1.5\}$ ).

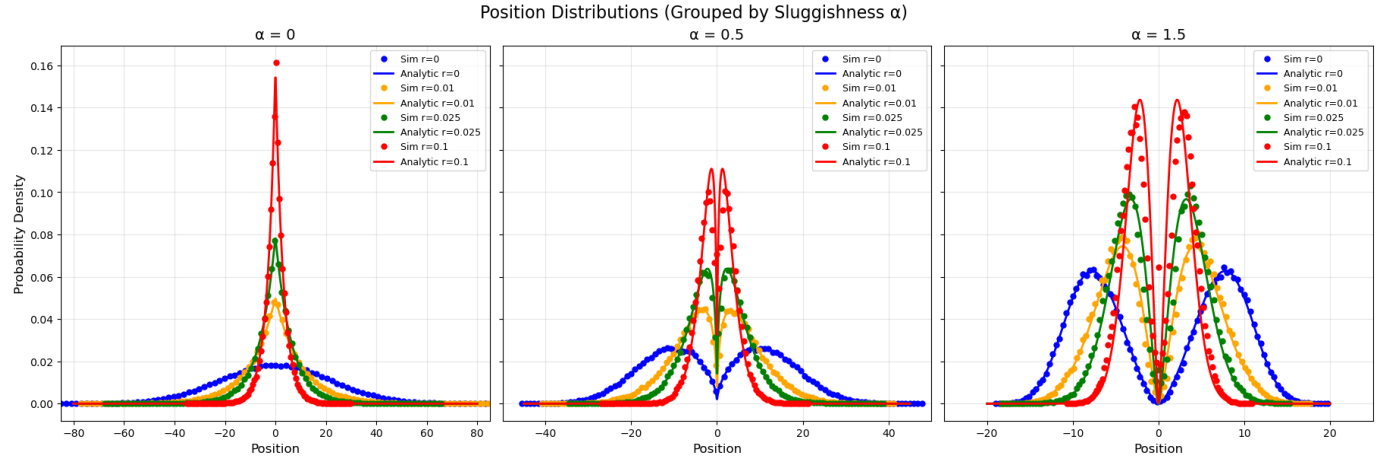


Figure 5: Probability distribution of positions of sluggish random walks with resetting grouped by different levels of sluggishness ( $\alpha \in \{0, 0.5, 1.5\}$ ) for reset probabilities  $r = \{0, 0.01, 0.025, 0.1\}$ .

[Python notebook](#)

## References

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