Convex hull of a Brownian motion in confinement

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We study the effect of confinement on the mean perimeter of the convex hull of a planar Brownian motion, defined as the minimum convex polygon enclosing the trajectory. We use a minimal model where an infinite reflecting wall confines the walk to its one side. We show that the mean perimeter displays a surprising minimum with respect to the starting distance to the wall and exhibits a non-analyticity for small distances. In addition, the mean span of the trajectory in a fixed direction \( \theta \in [0, \pi/2] \), which can be shown to yield the mean perimeter by integration over \( \theta \), presents these same two characteristics. This is in striking contrast with the one dimensional case, where the mean perimeter is an increasing analytical function. The non-monotonicity in the 2D case originates from the competition between two antagonistic effects due to the presence of the wall: reduction of the space accessible to the Brownian motion and effective repulsion.

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How does one characterize the territory covered by a Brownian motion in two dimensions? This question naturally arises in ecology where the trajectory of an animal during the foraging period is well approximated by a Brownian motion \([1, 2]\) and one needs to estimate the home range of the animal, i.e., the two dimensional (2D) space over which the animal moves around over a fixed period of time \([3]\). The most versatile and popular natural arises in ecology where the trajectory of an animal during the foraging period is well approximated and is, in fact, absent in 1D (as shown in Fig.1(c)).

In this Letter, we address this important issue in a simple minimal model that allows an exact solution. We consider a single planar Brownian motion in the presence of a reflecting infinite wall that confines the Brownian motion in the positive half-space (see Fig.1(a)). This can simply model the habitat of an animal bordering, on one side, a highway or a river that the animal can not cross. We show that the presence of the wall breaks the isotropy of the 2D space in a way that, even in this simple setting, induces a nontrivial effect on the convex hull of the Brownian motion. We compute analytically the mean perimeter \( \langle L^{(d)}_{2D}(t) \rangle \) of the convex hull of the Brownian motion of duration \( t \), starting at an initial distance \( d \) from the wall. We show that it exhibits a scaling form, at all times \( t \),

\[
\langle L^{(d)}_{2D}(t) \rangle = \sqrt{\pi Dt} \hat{L}_{2D} \left( \frac{d}{\sqrt{Dt}} \right)
\]

(1)

where \( \hat{L}_{2D}(x) \) is a nontrivial scaling function that we compute explicitly and is plotted in Fig.1(b). We find a surprising, and rather striking, non-monotonic behavior of the scaling function \( L_{2D}(x) \), revealing a minimum at a certain characteristic scaled distance (see Fig.1(b)). Moreover, the scaling function also exhibits an unexpected singularity as \( x \to 0 \), \( \hat{L}_{2D}(x) \sim -a [x/\ln(1/x)]^{3/2} \) with the prefactor \( a \approx 8\sqrt{2/\pi^3}/[3\Gamma(3/4)] \). We demonstrate that both the non-monotonicity and the small \( x \) singularity of the scaling function are purely 2D effects and are, in fact, absent in 1D (as shown in Fig.1(c)).

One dimension. To get an intuition on the scaling...
Figure 1: What is the influence of a confinement on the convex hull of a 2D Brownian motion? Considering the minimal model of a reflecting confining infinite wall (see (a)), we show that the mean perimeter of the convex hull turns out to be a non-monotonic function of the initial distance \( d \) to the wall (see (b)). This is in striking contrast with the monotonic behavior of the mean span, which is the 1D analogue of the mean perimeter of the convex hull (see (c)). (a) Convex hull of a Brownian trajectory of starting point \( O \) and diffusion coefficient \( D \). (b) Mean rescaled perimeter \( \langle L_{1D}^{(d)}(t)/\sqrt{Dt} \rangle \) of the convex hull as a function of the rescaled distance to the wall \( x = d/\sqrt{Dt} \). Analytical (solid line) and simulation (dots) results (see [24] for details). The inset presents the different geometrical parameters involved in the calculation. (c) Mean rescaled span \( \langle L_{1D}^{(d)}(t)/\sqrt{Dt} \rangle \) of a (semi) confined 1D Brownian motion as a function of the rescaled distance \( x = d/\sqrt{Dt} \) to the reflecting point.

function for the mean perimeter and to appreciate the fact that its non-monotonicity and singularity at \( x = 0 \) is indeed induced by the 2D geometry, it is useful to first compute the analogous quantity in the interesting in itself one dimensional (1D) setting. In 1D, the corresponding quantity is the mean span \( \langle L_{1D}^{(d)}(t) \rangle \) of a Brownian motion on a semi-infinite line, starting at a distance \( d \) from a reflecting origin (see inset of Fig.1(c)). It can be expressed as

$$\langle L_{1D}^{(d)}(t) \rangle = \langle \mathcal{M}_{\text{right}}^{(d)}(t) \rangle + \langle \mathcal{M}_{\text{left}}^{(d)}(t) \rangle,$$

where \( \langle \mathcal{M}_{\text{right}}^{(d)}(t) \rangle \) and \( \langle \mathcal{M}_{\text{left}}^{(d)}(t) \rangle \) denote the mean maximal extensions of the walk up to time \( t \) to the right and to the left respectively.

For convenience, we shift the reflecting wall to \(-d\) and consider the walk starting at the origin 0. We compute the right and the left parts separately, starting with the right. After integration by parts, it is obtained that \( \langle \mathcal{M}_{\text{right}}^{(d)}(t) \rangle = \int_0^\infty dy \left( 1 - S_{\text{right}}^{(d)}(t|y) \right) \), where \( S_{\text{right}}^{(d)}(t|y) = \text{Prob} \left[ \mathcal{M}_{\text{right}}^{(d)}(t) \leq y \right] \) is the cumulative distribution of \( \mathcal{M}_{\text{right}}^{(d)}(t) \), given that the particle starts at the origin. In turn, the cumulative distribution \( S_{\text{left}}^{(d)}(t|y) \) is just the probability that the walker starting at the origin stays within the box \([-d,y]\) up to time \( t \), i.e., the survival probability of the walker with a reflecting wall at \(-d\) and an absorbing wall at \( y > 0 \). In contrast, for the left side, the maximal displacement can at most be \( d \), due to the presence of the reflecting wall and one finds \( \langle \mathcal{M}_{\text{left}}^{(d)}(t) \rangle = \int_{-d}^0 dy \left( 1 - S_{\text{left}}^{(d)}(t|y) \right) \), where \( S_{\text{left}}^{(d)}(t|y) \) is the survival probability of a walker in the semi-infinite region \([y, \infty)\) with an absorbing wall at \(-d < y \leq 0 \). These survival probabilities can be computed using standard techniques [20, 21]. They are best expressed in terms of Laplace transforms with respect to \( t \). Denoting by \( \hat{f}(p) = \int_0^\infty f(t) e^{-pt} dt \) the Laplace transform of a function \( f(t) \), we find

$$\langle \hat{\mathcal{M}}_{\text{right}}^{(d)}(p) \rangle = \int_0^\infty dy \left( \frac{1}{p} - \hat{S}_{\text{right}}^{(d)}(p|y) \right)$$

$$\langle \hat{\mathcal{M}}_{\text{left}}^{(d)}(p) \rangle = \int_{-d}^0 dy \left( \frac{1}{p} - \hat{S}_{\text{left}}^{(d)}(p|y) \right) \tag{3}$$

where

$$\hat{S}_{\text{right}}^{(d)}(p|y) = \frac{1}{p} \left( 1 - \frac{\cosh (\sqrt{\frac{D}{T}} d)}{\cosh (\sqrt{\frac{D}{T}} (y + d))} \right)$$

$$\hat{S}_{\text{left}}(p|y) = \frac{1}{p} \left( 1 - e^{\sqrt{\frac{D}{T}} y} \right). \tag{4}$$

Inverting the Laplace transforms, we get

$$\hat{\mathcal{M}}_{\text{right}}(x) = \frac{\mathcal{M}_{\text{right}}^{(d)}(t)}{\sqrt{D} t}$$

$$= \frac{2}{\sqrt{\pi}} - \sum_{n=1}^{+\infty} (-1)^n \frac{(-1)^n}{4n^2 - 1} \left( \frac{e^{-n^2 x^2}}{\sqrt{n^2 - 1}} - nx \text{erfc} (nx) \right) \tag{5}$$

$$\hat{\mathcal{M}}_{\text{left}}(x) = \frac{\mathcal{M}_{\text{left}}^{(d)}(t)}{\sqrt{D} t} = \frac{2}{\sqrt{\pi}} \left( 1 - e^{-\frac{x^2}{2}} \right) + x \text{erfc} \left( \frac{x}{2} \right) \tag{6}$$

with \( x = d/\sqrt{Dt} \) denoting the scaled distance. This gives, from Eq. (2), \( \langle L_{1D}^{(d)}(t) \rangle = \sqrt{D} t \hat{L}_{1D} \left( \frac{d}{\sqrt{D} t} \right) \) where the 1D scaling function, \( \hat{L}_{1D}(x) = \hat{\mathcal{M}}_{\text{right}}(x) + \hat{\mathcal{M}}_{\text{left}}(x) \), is plotted in Fig. 1(c). The scaling function increases monotonically with \( x \) from \( \hat{L}_{1D}(x \to 0) = 4/\sqrt{\pi} \) (when the walker starts at the wall) to \( \hat{L}_{1D}(x \to \infty) = 4/\sqrt{\pi} \)
(where the walker does not feel the wall and one recovers the mean span of the walker in the absence of the wall).

Several conclusions can be drawn from these expressions. (See Fig. 1(c)): (i) The presence of the reflecting wall preserves the diffusive scaling of the mean perimeter \( \sim \sqrt{Dt} \). (ii) However, for fixed \( d \) and late times, \( \langle L_{1D}(t) \rangle \rightarrow \tilde{L}_{1D}(x \to 0) = \sqrt{\pi} \), which is lower than \( 4/\sqrt{\pi} \) obtained in the absence of confinement \((d \to \infty)\). (iii) The mean perimeter is minimized when the walker starts from the reflecting wall and (iv) is an increasing analytic function of the distance from the reflecting wall.

We show below that in 2D, while points (i) and (ii) continue to hold, (iii) and (iv) are no longer valid. Our main results are indeed the non-monotonicity of the mean perimeter of the convex hull in 2D and its non-analyticity for small starting distances from the reflecting wall. These two features are in striking contrast with the one-dimensional case derived above (see Fig. 1).

**Two dimensions.** We now turn to the 2D case, where we consider a Brownian motion in a semi-infinite medium (delimited by a reflecting wall) starting at a distance \( d \) from this wall (see Fig. 1(a)). To compute the mean perimeter of the convex hull of the walk of duration \( t \), we follow the general set-up developed in Refs. [3] [10]. By adapting Cauchy’s formula [22] for the perimeter of closed convex curves to random curves, it was shown that the mean perimeter of the convex hull of any two-dimensional stochastic process, including that of a Brownian motion, can be expressed as [3] [10]

\[
\langle L_{2D}(t) \rangle = \int_0^{2\pi} d\theta \langle M(\theta, t) \rangle
\]

where \( M(\theta, t) \) is the maximal projection of the trajectory of the walker up to time \( t \) in the direction \( \theta \). As in 1D, it is useful to express the mean in terms of the cumulative distribution of \( M(\theta, t) \)

\[
\langle M(\theta, t) \rangle = \int_0^{+\infty} dM \left( 1 - S(t|M, \theta) \right)
\]

where the cumulative distribution \( S(t|M, \theta) \) can be identified to the survival probability up to time \( t \) of the walker in the semi-infinite plane, but bounded additionally by an absorbing wall at distance \( M \) from the starting point, perpendicular to the direction \( \theta \) (see inset of Fig. 1(b)). It then defines an infinite wedge of top angle \( \alpha/2 = \pi/2 - \theta \) with one absorbing edge and one reflecting edge. However, by adding a twin wedge symmetrically around the reflecting edge, the survival probability of the walker in the original wedge is the same as in the ‘doubled’ wedge with twice the top angle \( \alpha = \pi - 2\theta \), but with two absorbing edges. The survival probability of a Brownian walker, starting initially at the polar coordinates \((r_0, \varphi_0)\), inside a wedge of angle \( \alpha \) with two absorbing edges is [20]

\[
S(t|r_0, \varphi_0) = \frac{r_0}{\sqrt{\pi Dt}} e^{-\frac{r_0^2}{8Dt}} \sum_{m=0}^{+\infty} \frac{\sin \left( \frac{(2m+1)\pi \varphi_0}{\alpha} \right)}{2m+1} \times \left[ I_{\nu+1} \left( \frac{r_0^2}{8Dt} \right) + \frac{I_{\nu+1}}{2} \left( \frac{r_0^2}{8Dt} \right) \right]
\]

where \( I_{\nu}(z) \) is the standard modified Bessel function.

The initial position \((r_0, \varphi_0)\) can be expressed in terms of the original variables \( M, d \) and \( \theta \) (see inset of Fig. 1(b)). For convenience, we introduce two new dimensionless variables as \( u = M/\sqrt{Dt} = r_0 \sin \varphi_0/\sqrt{Dt} \) and \( \varphi = \arccos\left( u \sin \theta / \sqrt{u^2 + 2u \sin \theta + 1} \right) \).

Plugging the result for the survival probability in Eq. 9 into Eqs. 10 and 11 we get

\[
\tilde{L}_{2D}(x) = \int_0^{+\infty} du \left\{ 1 - \frac{u^2 + 2u \sin \theta + 1}{\sqrt{\pi} \cos \theta} e^{-\frac{x^2 + 2xu \sin \theta + u^2}{8 \cos^2 \theta}} \times \sum_{\nu=0}^{+\infty} \sin \left( \frac{(2m+1)\pi}{\alpha} \right) \arccos\left( \sqrt{\frac{x^2 + 2xu \sin \theta + u^2}{8 \cos^2 \theta}} \right) \right\}
\]

with \( \nu = (2m+1)\pi/(2\alpha) - 1/2 \). Integrating over \( \theta \) in Eq. 7 then provides our final result for the mean rescaled perimeter (MRP)

\[
\tilde{L}_{2D}(x) \equiv \frac{L_{2D}(t)}{\sqrt{Dt}} = \int_0^{2\pi} d\theta \tilde{M}(\theta, x)
\]

with \( \tilde{M}(\theta, x) \) given explicitly in Eq. 12. As expected, the MRP is a function of only one parameter, the rescaled distance to the wall \( x = d/\sqrt{Dt} \).

Interestingly, we show in [21] that the MRP, given in Eqs. 13 and 12 for arbitrary \( x = d/\sqrt{Dt} \), simplifies a great deal in the important case \( x = 0 \) of a Brownian motion starting from the reflecting wall (or equivalently starting at any fixed distance but for large times) into

\[
\tilde{M}(\theta, 0) = 2\sqrt{\pi} \cos \theta / \left( \pi - 2\theta \right)
\]

As a simple check, we recover, for \( \theta = 0 \) (i.e., in the direction parallel to the reflecting wall), the result for
the non-confined case $\tilde{M}(0,0) = 2/\sqrt{\pi}$. Indeed, the potential reflections on the wall do not affect the walk in the parallel direction. For $\theta = \pi/2$, outwards orthogonally to the wall, we recover the 1D result obtained above $M(\pi/2,0) = \sqrt{\pi}$, which is higher than in the non-confined case. Indeed, the wall pushes the trajectories further in this direction. Finally, it is straightforward to obtain the MRP of the convex hull by integrating over the angle $\theta$

$$\tilde{L}_{2D}(0) = 2\sqrt{\pi} \text{Si}(\pi) = 6.56495 \ldots$$

where $\text{Si}(z) = \int_0^z \frac{\sin x}{x} dx$. Note that the MRP for a walk starting from the reflecting wall still grows as $\sqrt{Dt}$, but the prefactor $6.56495 \ldots$ is lower than the non-confined value $\sqrt{16\pi} = 7.08982 \ldots$.

As $x$ increases from 0, the scaling function $\tilde{L}_{2D}(x)$ in (13), supplemented by (12), first decreases, achieves a minimum and then increases again before eventually saturating to the constant $\sqrt{16\pi}$ corresponding to the non-confined case (see Fig. 1(b) for a plot). This non-monotonic behavior can already be understood by analyzing the $x \to 0^+$ limit. From Eq. (12), one can show that for small $x$ (see [24] for details)

$$\tilde{M}(\theta, x) = 2\sqrt{\pi} \frac{\cos \theta}{\pi - 2\theta} \sin \theta x$$

$$= \begin{cases} 
\sqrt{\pi} \frac{\cos \theta}{\pi} x^2 + O(x^3) & \text{for } \theta > 0 \\
C(\theta) x^{2\nu_0} + O(x^2) & \text{for } \theta < 0
\end{cases} (16)$$

where $\nu_0 = \theta/(\pi - 2\theta)$ and the amplitude $C(\theta)$ has a complicated expression (see [24]), which is negative for all $\theta < 0$. Interestingly, the leading order correction term in (18) is non-analytic only for $\theta < 0$. Using Laplace’s method, the integration over $\theta$ finally leads to

$$\tilde{L}(x) = 2\sqrt{\pi} \text{Si}(\pi) \sim x \leq 1 - \frac{8\sqrt{2\pi^3}}{3\Gamma\left(\frac{3}{4}\right)} \left(\frac{x}{\ln(1/x)}\right)^{3/2} (17)$$

Let us make a few remarks: (i) While, for any fixed $\theta$, $\tilde{M}(\theta, x)$ has a nonzero linear term in $x$ for small $x$, the linear term disappears when integrated over $\theta$. (ii) Strikingly, the MRP is nonanalytic as $x \to 0$ (contrary to the 1D case) and (iii) it starts decreasing from a value at $x = 0$ which is lower than its $x \to \infty$ limit, so it must display a minimum, as confirmed in Fig. 1(b).

The existence of this surprising minimum can be qualitatively discussed by identifying the temporal regimes of a Brownian motion starting at a distance $d$ from the reflecting wall: (i) At short times (i.e., $x \gg 1$), the walker does not see the reflecting wall and the unconfined value of the MRP $4\sqrt{\pi}$ is recovered. (ii) After a time of order $d^2/D$ (i.e., $x \approx 1$), the reflecting wall starts impacting the trajectory by progressively reducing the space effectively accessible to the Brownian motion and thus decreasing the MRP. (iii) Next, these reflected trajectories start contributing to the outwards part of the convex hull (with respect to the plane). This effective repulsion is an antagonistic effect of the wall that turns out to increase the MRP. Finally, contrary to the 1D case, the MRP displays a complex behavior with $x$, involving a minimum. In addition, this minimum is global, implying in particular that the MRP is not minimized when the Brownian motion starts from the wall anymore, but for a nontrivial value of $x \approx 0.3$. Insights on this minimum may further be gained by considering the mean span in direction $\theta$ given by $\langle \tilde{M}(\theta, t) \rangle$ and $\langle \tilde{M}(\theta, t) \rangle$ involved in (7), whose small $x$ development is obtained by combining the two forms of (10). Knowing that the mean span in direction $\theta$ at $x = 0$ is lower than its large $x$ limit and that it starts decreasing ($C(\theta) < 0$), it displays a minimum for $\theta \neq 0$ and $\theta \neq \pi/2$ (see Fig. 2). After integration over $\theta$, the non-monotonicity remains.

In conclusion, we have studied analytically the mean perimeter of the convex hull of a Brownian motion in 1D and 2D in presence of a reflecting wall at the origin. In 2D, this confinement leads to a striking non-monotonic and non-analytic (for small distances) behavior of the mean perimeter as a function of the scaled starting distance from the wall. The non-monotonicity 2D originates from the competition between two antagonistic effects due to the presence of the wall: reduction of the space accessible to the Brownian motion and effective repulsion. While these two effects are also at work in 1D, they do not lead to a non-monotonicity. Our work opens up several interesting questions for future studies. It would be interesting to know whether this non-monotonicity persists in $d > 2$ dimensions. Computing the mean area of the convex hull in two dimensions in presence of a reflecting wall remains challenging. Finally, it would be interesting to study other forms of confinements, for instance, a Brownian motion in an enclosed space or in the presence of an external confining potential.

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Supplementary Material for “Convex hull of Brownian motion in confinement”

I. DERIVATION OF $\mathcal{M}(\theta, 0)$

We start from Eq. (12) in the main text and first rewrite the 1 on the right hand side (inside the integral) as

$$\frac{u}{\sqrt{\pi \cos \theta}} e^{-u^2/2 \cos^2 \theta} \sum_{m=0}^{+\infty} \frac{(-1)^m}{2m+1} \frac{u^2}{\sqrt{2\pi u^2 \cos^2 \theta}} = 1. \quad (18)$$

A change of variables and the introduction of a regularization parameter $\beta$ yield

$$\dot{\mathcal{M}}(\theta, 0) = \frac{4}{\sqrt{\pi}} \cos \theta \sum_{m=0}^{+\infty} \frac{(-1)^m}{2m+1}$$

$$\times \int_0^{+\infty} dv \ e^{-\beta v} \left( \frac{2}{\pi} \frac{e^v}{\sqrt{v}} - I_v(v) - I_{v+1}(v) \right)_{\beta=1} \quad (19)$$

where $A(\beta, m)$ can be shown to be given by

$$A(\beta, m) = \frac{2}{\beta - 1} \left( \frac{\beta + \sqrt{\beta^2 - 1}}{\sqrt{\beta^2 - 1}} - \frac{\beta + \sqrt{\beta^2 - 1}}{\sqrt{\beta^2 - 1}} \right) \quad (20)$$

Using next that

$$\sum_{m=0}^{+\infty} \frac{(-1)^m a^{(2m+1)/2m+1}}{2m+1} = \sqrt{\arctan \left( a^{-\pi} \right)} \quad (21)$$

and taking the limit $\beta \to 1$ gives the mean extension in the direction $\theta$

$$\dot{\mathcal{M}}(\theta, 0) = 2\sqrt{\pi} \frac{\cos \theta}{\pi - 2\theta}. \quad (22)$$

II. ASYMPTOTIC EXPANSION OF $\dot{\mathcal{M}}(\theta, x)$ FOR SMALL DISTANCES

We give here the main steps for the derivation of the small $x$ development of $\dot{\mathcal{M}}(\theta, x)$ stated in Eq. (16) in the main text. The linear term in $x$ (for $-\pi/2 < \theta < \pi/2$) and the quadratic term (for $0 < \theta < \pi/2$) are obtained by integration over $u$ of the small $x$ development of the survival probability. In turn, these integrals can be carried out by introducing, as performed in the calculation of the order 0 of $\mathcal{M}(\theta, x)$, a regularization factor $\beta$ (see Eq. (19)). The integrals involved can be calculated straightforwardly. However, for $-\pi/2 < \theta < 0$, this method produces a diverging coefficient for the quadratic term, which reveals a non-analytic behavior whose calculation is presented below.
with

\[
C(x, \theta) = \frac{x^2 \cos \theta}{2\sqrt{\pi}} \left\{ \int_0^1 dz \ e^{-\frac{x^2}{8} z} I_{v_0} \left( \frac{x^2 z}{8} \right) - \int_1^{1/\cos^2 \theta} \frac{z}{z-1} \cos \left( \frac{\pi}{\alpha} \arccos \left( \frac{\sqrt{\frac{z-1}{z}}}{\sqrt{\frac{z-1}{z}}} \right) \right) \right. \\
\left. \times e^{-\frac{x^2}{8} z} I_{v_0} \left( \frac{x^2 z}{8} \right) \right\}.
\]

(30)

The constant and linear terms of Eq. (29) contribute to give the constant and linear terms of Eq. (16) of the main text. As for the last term, we use the development of the integrand of Eq. (30) for small \(x\)

\[
e^{-\frac{x^2}{8} I_{v_0} \left( \frac{x^2 z}{8} \right)} \sim x^2 + 2 \pi \frac{\theta}{\alpha} + O(x^2)
\]

(31)

to obtain

\[
C(x, \theta) \sim C(\theta) x^{2+2\pi \theta} + O(x^2)
\]

(32)

where

\[
C(\theta) = \frac{\cos \theta}{2\sqrt{\pi} 16^v \Gamma(1 + v_0)} \left\{ \frac{1}{1 + v_0} - \int_1^{1/\cos^2 \theta} \frac{z}{z-1} \cos \left( \frac{\pi}{\alpha} \arccos \left( \frac{\sqrt{\frac{z-1}{z}}}{\sqrt{\frac{z-1}{z}}} \right) \right) z^{v_0} \right. \\
\left. - \int_1^{+\infty} dz \left[ \frac{z}{z-1} \cos \left( \frac{\pi}{\alpha} \arccos \left( \frac{\sqrt{\frac{z-1}{z}}}{\sqrt{\frac{z-1}{z}}} \right) \right) - 1 \right] z^{v_0} \right\}.
\]

(33)

Replacing \(v_0\) and \(\alpha\) with their values \(\theta/\alpha\) and \(\pi - 2\theta\) finally leads to the small \(x\) development of \(\tilde{M}(\theta, x)\) for \(-\pi/2 < \theta < 0\) (see Eq. (16) of the main text).

\[
\tilde{M}(x) = \frac{2\sqrt{\pi} \cos \theta}{\pi - 2\theta} - \sin \theta x + C(\theta) x^{2+\frac{2\theta}{\pi - 2\theta}} + O(x^2)
\]

(34)

III. NUMERICAL SIMULATIONS

We computed the mean perimeter of the convex hull of a Brownian motion via numerical simulations. We constructed Gaussian random walks of \(10^5\) steps with a time step \(\Delta \tau = 10^{-3}\) when the walker is further than a distance \(d \approx 0.2\) from the wall. When the walker approaches the wall, we take an adapted time step quadratic in the distance \(d\) to the wall \(\Delta \tau = (0.1 d + \lambda)^2\) with \(\lambda = 0.01\). The parameter \(\lambda\) should not be taken too small to prevent the computation time from diverging. The convex hull is then constructed using the Graham scan algorithm (see Ref. 23 of the main text) for each Brownian walk, its perimeter calculated and averaged over \(10^5\) realisations. Agreement is found with our analytical prediction (see Fig.1(b) of the main text).