Dynamics of a passive sliding particle on a randomly fluctuating surface

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We study the motion of a particle sliding under the action of an external field on a stochastically fluctuating one-dimensional Edwards-Wilkinson surface. Numerical simulations using the single-step model shows that the mean-square displacement of the sliding particle shows distinct dynamic scaling behavior, depending on whether the surface fluctuates faster or slower than the motion of the particle. When the surface fluctuations occur on a time scale much smaller than the particle motion, we find that the characteristic length scale shows anomalous diffusion with \( \xi(t) \sim t^{2/3} \), where \( \phi \approx 0.67 \) from numerical data. On the other hand, when the particle moves faster than the surface, its dynamics is controlled by the surface fluctuations and \( \xi(t) \sim t^\phi \). A self-consistent approximation predicts that the anomalous diffusion exponent is \( \phi = \frac{3}{4} \), in good agreement with simulation results. We also discuss the possibility of a slow cross-over towards asymptotic diffusive behavior. The probability distribution of the displacement has a Gaussian form in both the cases.

I. INTRODUCTION

The advection of a passive scalar field (such as temperature) by a turbulent fluid is a well-known problem [1], and is an example of a coupled semi-autonomous system, where one of the fields evolves on its own, but affects the dynamics of the other. Many such systems have been studied in recent times, such as the dynamics of a particle passively sliding on a randomly fluctuating surface [2–4], phase separation on a rough substrate [4] and clustering of particles on a stochastically fluctuating surface under the influence of gravity [5]. These studies have shown that such systems possess a number of non-trivial features. For example, hard-core particles cluster into a fluctuation-dominated phase separated state [5] and non-interacting particles have non-trivial density correlations [2] in the steady state. It is thus highly desirable to aim for an understanding of the relation between the time evolution of the stochastic fluctuations of the underlying field and the dynamics of the passive scalar, and this is our primary motivation.

In this paper, we study the dynamics of a passive sliding particle moving on a randomly fluctuating surface. For concreteness, we choose the surface to be the one-dimensional Edwards-Wilkinson surface with uncorrelated noise [6], whose steady state and dynamical properties are well-known. As we shall see, even this simple example has unexpectedly rich properties.

For clarity of presentation, we shall state the problem here. Let us consider a one-dimensional fluctuating Edwards-Wilkinson surface, whose equation of motion has the form

\[
\frac{\partial h}{\partial t} = \nu \nabla^2 h + \eta \quad ; \langle \eta(x,t)\eta(x',t') \rangle = 2D\delta(x-x')\delta(t-t')
\]

(1)

The Langevin equation for the motion of a particle sliding on the surface has the form

\[
\frac{dX(t)}{dt} = -\Gamma \partial_x h |_{x=X(t)}
\]

(2)

where \( X(t) \) is the position of the particle on the surface and \( \Gamma \) is a friction coefficient. In principle, one could include a random noise term for the motion of the particle also, but since its only effect is to add a diffusive term to the mean square displacement, we shall drop it from further calculations.

The time evolution of the mean-square displacement of the sliding particle is given by the equation,

\[
\frac{d}{dt}\langle X^2(t) \rangle = 2\Gamma^2 \int_0^t dt' \Phi(t, t')
\]

(3)

where

\[
\Phi(t, t') = \langle \partial_x h[X(t), t] \partial_x h[X(t'), t'] \rangle
\]

(4)

is the ‘effective’ noise correlator for the motion of the sliding particle, and is equal to the slope correlator of the fluctuating surface evaluated along the particle trajectories. The angular brackets on both side represents average over different realizations of the noise \( \eta(x,t) \). Clearly, this is a highly non-trivial quantity to compute, since it is not obvious a priori how the slope correlator would behave when averaged over such a set of highly restricted paths.

We outline our main results here. Numerical simulations show that when the particle moves much faster than the surface, its dynamics is controlled by the surface fluctuations, and the mean square displacement behaves as
the sliding particle is moved by a distance $t \sim \nu t$. An argument using the concept of over-turning valleys confirms this result. On the other hand, when the surface fluctuation is fast, the particle moves in a very dynamic landscape, and the mean square displacement shows anomalous diffusion with $(X^2(t)) \sim t^{2\phi}$ over the time scales of simulations. A self-consistent approximation which assumes no correlation between a particle trajectory and the underlying surface height configuration, predicts that the anomalous diffusion exponent is $\phi = \frac{2}{5}$, in reasonably good agreement with numerical results. We also discuss the possibility that the observed anomalous diffusion might be simply transient behavior, and give reasons for the same.

The rest of this paper is arranged in the following way. In the next section, we present the self-consistent approximation to evaluate the correlator in Eq.4, and discuss its predictions and limitations. In Sec. III, we present the results of numerical simulations of the problem using the single-step algorithm. In Sec. IV, we summarize our findings and discuss the outstanding questions.

II. THE SELF-CONSISTENT APPROXIMATION

Let us define $z = X(t) - X(t')$, which, like $X(t)$ itself, is a random variable. Let us now sort the set of all available surface configurations using $z$, i.e., for fixed $t$ and $t'$, let $S_z$ denote the set of surface configurations where a sliding particle is displaced by a distance $z \pm \Delta z$ between times $t$ and $t'$. The fraction of such configurations is simply $P(t,t',z)\Delta z$, where $P(t,t',z)$ is the (unknown) probability distribution function for particle displacements between times $t$ and $t'$. It then follows that

$$\Phi(t,t') = \int_{-\infty}^{\infty} dz P(t,t',z) \langle \partial_x h[X(t),t] \partial_x h[X(t) + z,t'] \rangle_{S_z}$$

(5)

Note that the average in Eq.5 is a restricted average, i.e., it is evaluated over surface configurations in which the sliding particle is moved by a distance $z$ between $t$ and $t'$.

Up to now, our treatment has been exact. In order to make further progress, we make an approximation by assuming that the restricted slope correlator in Eq.5 is simply the standard slope correlator of the surface, evaluated at the set of points $(0,t)$ and $(z,t')$. In other words, we assume that the particle trajectory is not so strongly correlated with the surface slope configuration, such that averages of surface characteristics are not affected. Numerical results show that this approximation is justified except when the particle motion is slower than the surface fluctuations. We shall discuss this case shortly, and assume for the time being that this assumption is valid. This approximation is the most crucial step in our calculation.

For the Edwards-Wilkinson surface, the slope correlator can be computed exactly, and the result is (for $t > t'$),

$$\langle \partial_x h(0,t) \partial_x h(z,t') \rangle = \frac{D}{\nu} \sqrt{\frac{\pi}{\nu(t-t')}} \exp \left(-\frac{z^2}{\nu(t-t')} \right)$$

(6)

Let us now conjecture the following asymptotic scaling form for the probability distribution function of particle displacement:

$$P(t,t',z) = \frac{1}{\xi} f \left( \frac{z}{\xi} \right)$$

(7)

where $\xi(t-t')$ is the characteristic dynamic length scale of particle motion. The scaling function $f(x)$ is required to satisfy the normalization $\int_{-\infty}^{\infty} f(x) dx = 1$. It also follows that the mean-square displacement is

$$\langle X^2(t) \rangle = \mu \xi^2$$

(8)

where $\mu = \int_{-\infty}^{\infty} f(x)x^2 dx$. We now substitute Eq.6 and Eq.7 into Eq.5. The effective noise correlator thus becomes

$$\Phi(t,t') = \frac{G(T)}{\sqrt{\nu T}} ; T = |t - t'|$$

(9)

where

$$G(T) = \frac{D \sqrt{\pi}}{\nu} \int_{0}^{\infty} d\eta \frac{f(\sqrt{\eta})}{\sqrt{\eta}} e^{-\eta^2/\eta} ; \eta = \frac{\xi^2}{\nu T}$$

(10)

We note that $G(T)$ is proportional to the Laplace transform of the function $g(x) = x^{-\frac{1}{2}}f(x)$, i.e., $G(T) = \frac{D}{\nu} \hat{g}(\lambda)$ with $\lambda = \frac{\xi^2}{\nu T}$. We now conjecture a power-law growth of the characteristic length scale $\xi$ at late times:

$$\xi(t) \approx a t^\phi \quad ; t \gg t_0$$

(11)

where $\phi$ is the dynamical exponent that characterize the particle motion and $t_0$ is a microscopic time scale. From the preceding equations, it is clear that consistency requires $\phi > \frac{1}{2}$. For, if $\phi < \frac{1}{2}$, the l.h.s in Eq.3 would vanish at large $t$, whereas the r.h.s would not. The value $\phi = \frac{1}{2}$ is also inconsistent with this equation, since the l.h.s would be a constant whereas the r.h.s would grow as $\sqrt{T}$ at large time $t$. Thus the only possible values are $\phi > \frac{1}{2}$. We now focus on the large-$t$ limit, where the Laplace transform variable $\lambda \to \infty$. The behavior of $G(T)$ in this limit follows from the following theorem [8]:

$$\lim_{x \to 0} x^{-\rho}g(x) = \lim_{\lambda \to \infty} \frac{\lambda^{\rho+1} \hat{g}(\lambda)}{\rho!} ; \rho > -1$$

(12)

It is reasonable to assume that $f(0)$ is a non-vanishing constant, in which case $g(x) \sim x^{-\frac{1}{2}}$ as $x \to 0$. From Eq.12, it then follows that $\rho = -\frac{1}{2}$, and hence $\hat{g}(\lambda) \sim \frac{1}{\sqrt{\lambda}}$ as $\lambda \to \infty$. Thus, the asymptotic form of $G(T)$ is
\[ G(T) \simeq \frac{\pi D f(0) \sqrt{\nu T}}{\nu} \xi(T) \quad ; T \to \infty \] (13)

After substituting Eq.8 and Eq.9 into Eq.3, and using the large T-form in Eq.13, the resulting self-consistent integral equation for \( \xi(t) \) is

\[ \xi \frac{dT}{dt} = \frac{\Gamma^2 \pi D}{2\mu} f(0) \int_0^t \frac{dT}{\xi(T)} \] (14)

We now substitute the power-law scaling form Eq.11 for \( \xi(T) \), which gives

\[ a^3(1 - \phi)t^{2\phi-1} \sim \frac{\pi D \Gamma^2 f(0)}{2\nu} t^{1-\phi} \quad ; t \to \infty \] (15)

After equating the powers of time on both sides, we find \( \phi = \frac{2}{3} \), which is the principal result of this paper, along with the pre-factor \( a \sim \left( \frac{D \Gamma^2}{\nu} \right)^{\frac{1}{3}} \). We have thus arrived at the somewhat surprising result that the sliding particle moves faster than the surface fluctuations, whose correlation length scales with time as \( \sqrt{\nu t} \). This super-diffusive behavior formally has its origin in the long-range nature of the effective noise correlator \( \Phi(t, t') \), which, from Eq.9 and Eq.13, has a power-law tail of the form \( (t - t')^{-\phi} \). Similar anomalous diffusive motion of advected particles with \( \phi = \frac{2}{3} \) has been shown to occur in the one-dimensional (non-linear) Burger’s equation with noise [3,7], using a mean-field approach.

Before proceeding to discuss the numerical results, we would like to point out that the calculation presented above is valid only if \( \lim_{t \to \infty} \frac{\sqrt{T}}{\xi(T)} \) vanishes, else we would not be able to use Eq.12 to arrive at our result. Indeed, it is quite possible that in the asymptotic limit, \( \xi(t) \sim \sqrt{\nu t} \), but that this regime lies outside the validity of this approximation. In such a case, the anomalous diffusion at early times can be still shown to exist, but with a different exponent. To see this, let us consider Eq.10 again at sufficiently early times \( t \) so that \( \xi^2(t) \ll \sqrt{\nu t} \). In this case, \( G(T) \simeq \frac{D}{2} \) is a constant, and after substituting in Eq.9, and subsequently in Eq.3, we find that \( \xi(t) \sim t^{3/4} \) when \( \xi^2 \ll \sqrt{\nu t} \). From numerical simulations (to be presented in the subsequent section), we are unable to rule out this possibility. Further discussions on this point are presented at the end of Sec. III.

III. NUMERICAL RESULTS

In this section, we discuss the results of numerical simulations of the problem. The surface is constructed as a set of height variables \( \{h_i\} \), where \( i = 1, 2, \ldots N \) and \( N \) is the lattice size. Periodic boundary conditions are imposed on the surface. We use the single-step model of the surface [9], where the height difference between adjacent lattice points is restricted, i.e., \( \tau_i^\pm = \pm[h_{i+1} - h_i] = \pm 1 \), where we have defined the clockwise/anti-clockwise slope variables \( \tau_i^\pm \). The surface evolves through the following set of dynamical rules:

- If \( \tau_i^+ = 1 \) and \( \tau_i^- = -1 \), \( h_i \to h_i + 2 \) with probability \( q \) (\( q \leq 1 \)).
- If \( \tau_i^+ = -1 \) and \( \tau_i^- = 1 \), \( h_i \to h_i - 2 \) with probability \( q \).

The large-distance, long-time properties of this model are identical to the continuum model in Eq.1. We start from a ‘ordered height configuration, where \( h_i = 1 \) is \( i \) is odd, and \( h_i = 0 \) when \( i \) is even. A lattice site is selected at random, and its height is updated in accordance with the above rules. For the surface configuration, one Monte Carlo step is counted after \( N \) such attempted updates, where \( N \) is the lattice size. The surface then evolves through \( T \simeq 10N^2 \) Monte Carlo time steps in order to reach the steady state. We used two lattice sizes, \( N = 1024 \) and \( N = 4096 \) in our simulations.

For a certain surface configuration in the steady state, we place 100 particles on the surface at randomly selected locations. At each Monte Carlo step, after a surface update is complete, a particle is selected at random (with initial position at lattice site \( i \), say) and its position is updated according to the following set of dynamical rules:

- If \( \tau_i^+ = 1 \) and \( \tau_i^- = 1 \), the particle moves one lattice step in the clockwise direction with probability \( p \) (\( p \leq 1 \)).
- If \( \tau_i^+ = -1 \) and \( \tau_i^- = -1 \), the particle moves one lattice step in the anti-clockwise direction with probability \( p \).

Essentially, the rules say that the particle only slides down when the slope is favorable, and remains stationary when it is at a local minimum or maximum. The particle displacements are measured such that for every move in clockwise direction, \( X(t) \to X(t) + 1 \), and for every move in anti-clockwise direction, \( X(t) \to X(t) - 1 \). The particles are assumed non-interacting, which is consistent with the single-particle picture. When the probability \( p > q \), the particles move faster relative to the surface motion and vice-versa. Also, without any loss of information, we set \( \max(p, q) = 1 \). The parameter we tune in the simulations is the ratio \( r = \frac{p}{q} \). The simulations were run up to \( 10^5 \) Monte Carlo steps for \( N = 1024 \) and \( 10^6 \) Monte Carlo steps for \( N = 4096 \). We computed the probability distribution of the displacement of the particles, as well as the mean square displacement, as functions of time. The results were averaged over 100 different starting configurations for \( N = 4096 \) and 1000 starting configurations for \( N = 1024 \).

In Fig.1, we have shown the results for the RMS displacement of the sliding particle plotted against the quantity \( qt \), where \( t \) is the number of Monte Carlo steps. The
fraction $q$ is used to scale the time so that all the surface configurations pass through an almost equal number of updates (In the continuum language, this is equivalent to using the combination $\nu t$). We have shown the results for four values of the update frequency ratio $r$: 0.01, 0.1, 1 and 10. The results for $r = 0.01$ and $r = 0.1$ offers evidence for the power-law behavior $\xi \sim t^2$ as predicted by the self-consistent argument. For $r > 1$, on the other hand, the growth of the RMS displacement with time is much slower. In this case, the observed growth exponent is closer to $\frac{1}{2}$ at all times. The case $r = 1$ displays intermediate behavior, with an effective exponent close to 0.56.

![Image](https://example.com/image1.png)

**FIG. 1.** The figure shows the time evolution of the root-mean-square displacement of the sliding particle for several values of the update frequency: $r = 10, 1, 0.1$ and 0.01 (top to bottom). The short straight lines are guides to the eye, and have slopes 0.5, 0.56, 0.68 and 0.67 (top to bottom). When $r < 1$, the dynamic exponent is close to 2/3, whereas when $r > 1$, it is close to 1/2. For $r = 0.01$, there is an early-time regime where the growth exponent is smaller than 2/3, but the asymptotic behavior is identical to that of $r = 0.1$. The lattice size in these simulations is $N = 4096$.

Although it is tempting to say that the cases $r > 1$ and $r < 1$ are characterized by different dynamical exponents, we would like to be cautious here. Our numerical data does not rule out the possibility that the observed anomalous diffusion of the sliding particle when $r < 1$ is simply a transient regime, and the asymptotic behavior might be identical for all values of $r$. In the scenario discussed in the last paragraph of the previous section, the observed behavior might reflect a slow cross-over from $\xi \sim t^{3/4}$ to $\xi \sim t^{1/2}$. The data for $r = 0.1$ shows some bending for $t > 10^5$, but at the moment we cannot conclusively say whether this shows the cross-over towards a diffusive regime, or simply a finite size effect. We have checked that a smaller system size ($N = 1024$) shows bending as early as $t \approx 10^4$ (Fig. 2), which makes it likely that this is a finite size effect.

![Image](https://example.com/image2.png)

**FIG. 2.** The figure illustrates finite size effects in the time evolution of the root-mean-square displacement of the sliding particle for $r = 0.1$ (top) and $r = 0.01$ (bottom), for lattice sizes $N = 1024$ and $N = 4096$. The short straight line at the top is a fit, and has slope 0.67.

*Particle trapping in the valleys:* The origin of the distinct scaling behavior when $r > 1$ may be understood through the following argument. When the particle moves faster than the underlying surface fluctuations, after the initial downward slide towards the local minimum, it gets 'trapped' there. At all future times, the motion of the particle is 'slaved' to the dynamics of this local minimum. The long-time dynamics of the particle may be visualized using the concept of *valleys* in the surface. A valley is defined as a linear stretch of the surface below a certain reference height, which we choose to be the instantaneous height at the starting location of the particle (Fig. 3). It is well-known that any instantaneous height configuration of a EW surface relative to a reference point on the surface is a random walk (RW), which returns to the origin after $N$ steps. Thus, the length distribution of the valleys is simply the distribution of return times of a one-dimensional RW to its starting point, which follows a power-law decay: $P(l) \sim l^{-\nu}$ for $l \ll N$ [10]. Let us now consider a section of the surface with length $\xi$, and let $N_\xi(l)$ be the number of valleys of length $l$ within this section. Since we require $\int dl N_\xi(l) = \xi$, it follows that $N_\xi(l)$ has the scaling form

$$N_\xi(l) \sim \sqrt{\xi} l^{-\frac{2}{\nu}}$$

(16)

The slope correlations of the surface over a length $l$ will decay by a time $\tau(l) \sim l^2/\nu$ from Eq.6. For a valley of length $l$, this time is roughly the time scale for the over-turning of the valley, whereby the valley is transformed into a hill. An event of over-turning leads to a displacement of a particle trapped inside the valley by a distance $\sim l$. The total time required to produce a displacement $\sim \xi$ is then the sum of the trapping times inside all the valleys within $\xi$, which is given by
After using Eq. 16, this equation gives \( t(\xi) \approx \int_0^{\xi} dl \tau(l) N_\xi(l) \) \( (17) \)

Equivalently, for a fixed time \( t \), the typical displacement of the sliding particle is \( \xi(t) \sim \nu t^{1/2} \). The result shows that in the regime \( r > 1 \), the dynamical exponent for the sliding particle is \( \frac{1}{2} \).

![Figure 3](image-url)  
**FIG. 3.** Illustration of a valley in the single step model: The open circle shows the position of the particle at time \( t = 0 \) and the filled circle shows its current position. The horizontal straight line divides the surface into valleys and hills.

![Figure 4](image-url)  
**FIG. 4.** The figure shows the scaled probability distribution \( f(x) = \langle X^2(t) \rangle^{1/2} P(0, t, z) \) plotted against the scaled distance \( x = r/\sqrt{\langle X^2(t) \rangle} \) for three widely separated values of time. The good scaling collapse justifies the scaling form used in Eq. 7. The lattice size is \( N = 1024 \) and the update frequency ratio is \( r = 0.1 \) in this figure. The parameter \( \sigma \approx 1.45 \) in the Gaussian fit, and is found to be the same for all values of \( r \). The data represents an average over 1000 different surface starting configurations.

Finally, we discuss the results for the probability distribution \( P(0, t, z) \) of the displacement of the particle at time \( t \), relative to its location at time \( t = 0 \). In Fig. 4, we have displayed the results for \( r = 0.1 \) at three widely separated instants of time, \( t = 10^2, 10^3 \) and \( 10^4 \). When the scaling function \( f(x) = \langle X^2(t) \rangle^{1/2} P(0, t, z) \) of the probability distribution is plotted against the scaled distance \( x = z/\sqrt{\langle X^2(t) \rangle} \), we find good collapse of the data, which provides a *a posteriori* justification for using this scaling form. It is also remarkable that the scaling function is very well represented by a Gaussian (see the fit in the figure).

**IV. CONCLUSIONS**

To conclude, we have shown by means of a self-consistent argument that the motion of a passive sliding particle on a fluctuating surface may show non-trivial behavior even when the surface is Gaussian, like the Edwards-Wilkinson surface. Specifically, the effective dynamical exponent for the motion of the sliding particle may be the same as, or different from the dynamical exponent of the surface fluctuations, depending on the relative time scale of surface and particle motion. The observed apparent super-diffusive dynamics of particle displacement in the EW surface urges us to be cautious when the measurements of the passive sliding particle is used to measure the dynamical exponent or other dynamical properties of the surface itself, as has been suggested in [3]. It might be interesting to extend this study to look into the effects of changing the relative update frequency of the particle and surface configurations on the steady state characteristics of the problem in a many-particle context [2,4]. Lastly, simulations on a larger time scale using much bigger lattices would be necessary to determine conclusively if the observed anomalous diffusion is indeed a cross-over effect.

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