Size effects in dislocation depinning models for plastic yield

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Abstract. Typically, the plastic yield stress of a sample is determined from a stress–strain curve by defining a yield strain and reading off the stress required to attain it. However, it is not a priori clear that yield strengths of microscale samples measured this way should display the correct finite size scaling. Here we study plastic yield as a depinning transition of a $1+1$ dimensional interface, and consider how finite size effects depend on the choice of yield strain, as well as the presence of hardening and the strength of elastic coupling. Our results indicate the existence of a crossover length that depends on the yield strain. It is only above this length scale that standard finite size scaling is expected to hold. These results are also expected to be particularly relevant for simulations of single dislocations, such as those used to study strengthening due to included particles.

Keywords: interfaces in random media (theory), avalanches (theory), defects (theory), plasticity (theory)
1. Introduction

A complete understanding of microscale plasticity is important both for fundamental and technological reasons, and much experimental and theoretical effort has been expended on this problem in recent years. A key result that has emerged—and that is not intuitively obvious from our everyday experience of classical bulk plasticity—is that on the microscale plastic flow is subject to size effects, with recent experiments showing that in submicron samples, the yield stress increases with decreasing sample size [1]. However, the origins of these finite size effects are not clear, and many mechanisms have been proposed (for reviews, see [2]–[4]).

In typical experimental studies, yield stress is determined from measured stress–strain curves by defining a yield strain $\gamma_Y$, typically a few per cent, and reading off the stress required to attain such a strain. This procedure suffers from the fact that measured stress–strain curves are not smooth and the transition to plastic yield is generally preceded by power law distributed avalanches [5]. This raises the question of how the yield strain should be selected to ensure that the measured yield stress is not systematically over- or under-estimated in a way that affects finite size scaling. Furthermore, it is not a priori clear that one can compare experimental results in which different threshold strains have been used, making it difficult to synthesize results from different studies.

In this paper, we approach this problem from a theoretical perspective, by studying numerical simulations of depinning of a 1D interface at zero temperature. This is a minimal model for plastic yield in a crystal: the interface can be thought of as representing a dislocation, driven through a random potential generated by disorder and other dislocations in the sample. Although more realistic models for plastic deformation exist, a key advantage of the depinning model is that it has previously been studied in a variety of contexts, using both numerical and analytical techniques [6]–[18], and it therefore serves as a well-characterized ‘benchmark’ against which we can compare the effect of varying our measurement procedure. Furthermore, as we discuss later, the key ideas illustrated

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**Contents**

1. **Introduction**  
2. Effects of target choice: the quenched Edwards–Wilkinson equation  
3. Other depinning models: KPZ and mean field  
4. Effects of hardening and machine stiffness  
   4.1. Hardening  
   4.2. Strain controlled experiments  
5. **Conclusion**  
   Acknowledgment  
   References

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Figure 1. (a) Typical ‘stress–strain’ (driving force, $f$, versus mean interface displacement, $\bar{u}$) curves from simulations of the quenched Edwards–Wilkinson equation, with (circles) and without (triangles) hardening. The black solid line indicates the $f \propto \bar{u}$ behaviour of the model with hardening as $\bar{u} \to \infty$ (see section 4). Depinning forces are determined by selecting a target value of $\bar{u}$ and reading off the driving force required to attain this target. (b) Data from a hardening simulation recast as a strain controlled experiment with machine stiffness $k$, using the transformations $\sigma_T = f - k\bar{u}/L^2$, $\gamma_T = f/k$, as described in section 4. The yield stress, $\sigma_{T,y}$ is determined by selecting a yield strain $\gamma_{T,y}$ and reading off the stress required to attain it.

in our study of this model are relevant to experiments, even without exact quantitative agreement.

Despite the relative simplicity of the model, it exhibits several key features in common with plastic deformation in real samples. The most important feature is the depinning phase transition, in which the long-time velocity of the interface is zero when the driving force $f$ is below a critical force $f_c$ and positive when $f > f_c$. This transition is analogous to plastic yield, in which stresses above a finite threshold are able to induce large strain increases that ideally are limited only by sample failure.

However, as illustrated in figure 1(a), when the interface is subject to an adiabatically increasing drive, as $f$ approaches $f_c$ the interface undergoes a series of avalanches with a broad size distribution. Accordingly, given a single force–position curve, it is not possible to state with certainty whether the interface has truly depinned. (Indeed, there exist...
alternative methods to exactly determine $f_c$ for a given system [19], but they are not directly relevant for this work.)

This suggests a method for imitating the experimental procedure for measuring yield stress. The motion of an edge dislocation with Burgers vector $b$ over distance $x$ in a cubic system of linear size $L$ results in a shear strain increase of $\gamma = bx/L^2$; the driving force is proportional to the external stress. One can therefore consider a plot of the mean position of the interface versus driving force to represent plastic strain versus stress. In analogy to the approach taken in experimental analysis, we define a target displacement $\Delta = \gamma L^2/b$ to define the onset of depinning, and study the effects of varying this target.

In particular, we determine the finite size scaling of the measured $f_c$ at various yield stresses. Like other phase transitions, the depinning transition is subject to finite size effects, so that ideally the depinning force in a system of linear size $L$ has the asymptotic form

$$f_c(L) = f_\infty + aL^{-1/\nu}$$

(1)

where $f_\infty$ is the depinning force in the limit $L \to \infty$ and $\nu$ is the finite size exponent, determined by the universality class of the model. Because the ‘correct’ behaviour is already known, the effects of changing the yield strain can be clearly distinguished.

In addition, the depinning model can also be related to strain controlled experiments, by driving the interface via a spring of stiffness $k$, and adjusting the externally-applied stress so that the total strain—the elastic strain of the spring, partially relieved by the plastic deformation arising from interface motion—increases smoothly. The set-up and a typical stress–strain curve are illustrated in figure 1(b). Again in this context, we study the effect of the choice of yield strain on measured yield stress, and discuss how these effects are related to strain hardening in a stress controlled experiment.

Beyond the possibility of gaining insight into experiments on finite size effects in plasticity, we observe that studies of strengthening due to included particles are often carried out within the framework of studying the motion of a single dislocation [20]–[22]. In this context, our results are directly relevant, as we study the effect of simulation size and aspect ratio and show that a poor choice can lead to spurious results.

A related issue has previously received some attention, as it has been pointed out [16, 17, 23] that in a simulation of a system of finite size $L \times M$ with periodic boundary conditions in both directions, a ‘natural’ choice of $M$ is $M \sim L^\zeta$, where $\zeta$ is the interface roughness exponent and $L^\zeta$ gives the interface width at criticality. In this way, one ensures that for all $L$, the system contains the same number of critical configurations. On the other hand, our target scales as $L^2$, so although it is natural for the plasticity scenario at hand, the system aspect ratios increase with $L$. We will see that this affects the distribution of measured depinning forces, as anticipated [16, 17].

The outline of the paper is as follows. In section 2, we use the quenched Edwards–Wilkinson model for interface depinning as a case study for the role of target choice on measured properties, considering not only the finite size scaling of the mean depinning force but also how the choice of target affects its whole distribution. The ideas developed in section 2 are independent of universality class, which we demonstrate in section 3 via results from the Kardar–Parisi–Zhang (KPZ) model and a mean field model. Finally, in section 4 we discuss two closely related extensions to the quenched Edwards–Wilkinson model, namely strain hardening and the above-mentioned strain
controlled experiments. We show that finite size effects in these models are more complex than in the other models we study, depending on both the finite size scaling of the depinning force and also the hardening amplitude or stiffness of the coupled spring.

2. Effects of target choice: the quenched Edwards–Wilkinson equation

We start by considering a simple linear model for the interplay between elasticity, random pinning and an external drive, namely, the quenched Edwards–Wilkinson model [6, 7]. This will allow us to discern the effect of the choice of target interface displacement in a straightforward manner, giving results that will aid the interpretation of more complex models that follow.

In general, an overdamped elastic interface characterized by height function \( u(x) \) and subject to a random potential and a uniform external drive \( f \) is described by overdamped equations of motion

\[
\frac{1}{\Gamma} \dot{u}(x) = f_{\text{tot}}(x) = f + f_{\text{el}}(x, u(x)) - \eta(x, u(x)),
\]

where the damping constant \( \Gamma \) is hereafter set to unity. The self-interaction of the interface, \( f_{\text{el}} \), promotes it to be flat, and may be local or long-range, depending on the model at hand. The self-interaction is in competition with the quenched disorder \( \eta \), which causes the interface to prefer to pass through sites with low potential energy. In the quenched Edwards–Wilkinson model that we study in this section, the elastic force at position \( x \) on the line is given by \( \nabla^2 u(x) \).

In this work, we study a cellular automaton version of this model, by discretizing \( x \) and \( u \). The elastic force is given by \( u_{x+1} + u_{x-1} - 2u_x \). We allow only forward motion of the interface: at each simulation step, a site \( x \) may increase \( u_x \) by one unit if the total force acting on it is positive, or remain stationary if the force is zero or negative. The random potential is implemented by drawing a random number for each \( x, u_x \). In general, these are from a uniform distribution over \([0, 1]\), but in section 3 we will also study Gaussian distributed disorder, because in mean field models, the disorder distribution can be important [24]. We use periodic boundary conditions in the \( x \) direction, and open boundary conditions in the \( u \) direction.

In order to generate ‘stress–strain’ curves in a stress controlled simulation, we start with a flat interface with \( u_x = 0 \) everywhere and \( f = 0 \). We increase the drive \( f \) adiabatically, by finding the site \( x^* \) with the smallest required driving force for motion, and increasing \( f \) so that \( x^* \) can move. This may trigger an avalanche via changes in \( f_{\text{el}} \), and we update sites in parallel until no more can move. We record the value of \( f \) and the mean position of the interface attained after it has stopped moving, and then increase the drive again. As mentioned in the Introduction, the record of driving forces and mean interface positions is the ‘stress–strain’ curve, since the motion of a dislocation by distance \( x \) increases the plastic strain by \( x/L^2 \) (we work in units where the Burgers vector \( b \) is unity). We continue to increase the drive until the mean interface position passes the target \( \gamma L^2 \). The driving force required for the target to be attained is denoted \( f_c \), and for simplicity we will call it a ‘depinning’ threshold, even if this is not strictly accurate.
Figure 2. (a) Mean measured depinning force \( \langle f_c \rangle \) plotted against system linear size \( L \), from simulations of the quenched Edwards–Wilkinson equation. The target strains used are \( \gamma = 0.01 \) (circles), 0.07 (triangles), 0.25 (squares) and 2 (inverted triangles). For all targets, \( f_c \) approaches the same value as \( L \to \infty \), but the behaviour for smaller \( L \) depends on \( \gamma \). Error bars (typically smaller than the symbols) represent the standard error. (b) The same data, collapsed using (1) and (3). Here, the absolute value has been plotted: for small \( L\gamma^{1/(2-\zeta)} \), \( \langle f_c \rangle \) is smaller than \( f_\infty \). The black solid line is proportional to \( L^{-0.75} \), which is the expected finite size scaling for the model for large \( L \).

Figure 2(a) shows the mean force required for the interface to attain the target mean position \( \gamma L^2 \), as a function of system linear size \( L \), for a range of \( \gamma \) values between 0.01 and 2. In the large-\( L \) limit, we find that \( f_c(L) \) approaches \( f_\infty = 0.424 \) according to \( L^{-1/\nu} \), with \( 1/\nu = 0.75 \), in agreement with previous studies of finite size scaling [14, 16]. However, for small \( L \), a distinct \( \gamma \) dependence becomes apparent: for small \( \gamma \), \( f_c \) initially grows with \( L \), but for large \( \gamma \), \( f_c(L) \) decreases monotonically.

This result can be understood by considering that there are two length scales in the direction of interface propagation. The first is the interface width at depinning, which is given by \( wL^\zeta \), with the roughness exponent \( \zeta = 1.25 \) for the quenched Edwards–Wilkinson equation [10, 15] and \( w \) a constant. The second length scale is our choice of target, which is \( \gamma L^2 \). The two length scales are equal when

\[
L^{2-\zeta} = \frac{w}{\gamma},
\]

giving a crossover length scale \( L_c = (w/\gamma)^{1/(2-\zeta)} \).
Size effects in dislocation depinning models for plastic yield

Figure 3. Standard deviation in the measured depinning force, $\sigma(f_c)$, plotted against system linear size $L$. For sufficiently large targets, $\sigma(f_c) \sim L^{-0.75}$, as indicated by the solid black line. The target strains used are $\gamma = 0.07$ (circles), 0.15 (triangles), 2 (squares). Because the distributions of $f_c$ are not Gaussian (see figure 4), the error bars have been estimated using the method in [25].

If $L < L_c$, the target is shorter than the interface width at criticality, and the simulation is halted during the early stages of interface motion, before the depinning transition can occur. In this regime, increasing $L$—and thereby the target—should increase the final driving force applied before the simulation is stopped. On the other hand, for $L > L_c$, we can expect the interface to undergo the depinning transition, and the standard finite size scaling of (1) should hold. At $L_c$, there is a transition from the former behaviour to the latter. However, if $\gamma$ is sufficiently large, the target is always larger than the interface width, that is, $L_c \to 0$ as $\gamma \to \infty$.

To verify this interpretation, we collapse the data from simulations with different $\gamma$ values. From (3), we see that we must rescale $L \to L \gamma^{-1/(2-\zeta)}$. Then from the finite size scaling (1) and the fact that $\nu = 1/(2 - \zeta)$ [9], we rescale $(f_c - f_{\infty})$ by $\gamma$. As seen in figure 2(b), the data collapse is very good. Furthermore, that the correct finite size scaling laws are recovered for sufficiently large targets is further evidenced by the measured scaling of the standard deviation of $f_c(L)$, as plotted in figure 3. This has the asymptotic form $L^{-1/\nu}$, as expected from finite size scaling.

We can go beyond the moments of the distribution of $f_c$ to consider the distribution itself. It has previously been shown [16, 17] that the distribution of $f_c$ measured in a finite sample of size $L \times M$ (with open boundary conditions in the $M$ direction) depends on the ratio of $M$ to the interface width $wL^\zeta$. Heuristically, the system can be divided into $n = M/(wL^\zeta)$ subsystems which are approximately independent, each with its own depinning force. The depinning force in the entire $L \times M$ system is the maximum of these, and if $n \to \infty$ the measured depinning force is described by extreme value statistics, in particular, the Gumbel distribution, which has cumulative distribution $P(x) = \exp(-e^{-(x-\mu)/\beta})$, where $\mu$ is the modal value of $x$ and the standard deviation of the distribution is $\beta \pi / \sqrt{6}$. In figure 4, we plot the distribution of $f_c$ (rescaled to have zero mean and unit standard deviation) for $L = 100$, for a range of $\gamma$ values. The targets range from a few interface widths to $\sim 10^4$ interface widths. In agreement with [16, 17], as $\gamma$ increases, the distribution approaches the Gumbel limit, albeit very slowly in the tails of the distributions.
3. Other depinning models: KPZ and mean field

The role of the parameter $\gamma$ relative to the interface width does not depend specifically on the choice of interaction $f_{el}$ used in the quenched Edwards–Wilkinson equation, and we have verified that models from other universality classes are affected by the choice of $\gamma$ in the same way. For example, in the KPZ model [26], where the elastic force $f_{el} = \nabla^2 u(x)$ is supplemented by a nonlinear term proportional to $(\nabla u)^2$, we measure a roughness exponent of $\zeta = 0.635$ (consistent with other measurements [15, 27]) and a finite size exponent of $1/\nu \approx 0.55$. Per (3), $f_c(L)$ curves for different $\gamma$ can be collapsed via the rescaling $L \rightarrow \gamma^{1/(2-\zeta)} L$, $(f_c - f_\infty) \rightarrow \gamma^{-1/(\nu(2-\zeta))} (f_c - f_\infty)$, as shown in figure 5.

The quenched Edwards–Wilkinson and KPZ models both have local interactions, and the interface width scales with length as $L^\zeta$. We now turn to a mean field model [13, 28, 29], in which every site interacts with every other site, and the interface width does not depend on the length scale it is measured over ($\zeta = 0$). We couple the interface to its
Figure 5. Data collapse for the KPZ model, based on the same arguments as the quenched Edwards–Wilkinson model (equations (1) and (3)), using fitted parameters $\zeta = 0.635$, $f_\infty = 0.226$, $1/\nu = 0.55$. The symbols represent $\gamma = 0.1$ (circles), $\gamma = 0.01$ (squares) and $\gamma = 0.005$ (diamonds). The standard error is smaller than the symbol size.

Figure 6. Rescaled depinning force versus target displacement $\gamma L^2$, in the mean field model. The solid black lines are proportional to $L^{-1/2}$. Closed symbols represent data for $J = 1$, open symbols represent data for $J = 0.1$, the $\gamma$ values are 0.005 (squares), 0.01 (triangles), 0.1 (circles) and 1 (diamonds). For $J = 0.1$, $f_\infty \approx 0.64$, for $J = 1$, $f_\infty \leq 10^{-3}$.

centre of mass, $\bar{u}$. The total force on site $x$ is

$$f_x = f + J(\bar{u} - u_x) - \eta_{u,x}. \tag{4}$$

In figure 6 we show the finite size behaviour of the depinning force for this model. The finite size exponent takes the mean field value of $\nu = 2$. For fixed $J$ we can collapse data from simulations with different $\gamma$ values following the same argument as for the quenched Edwards–Wilkinson model, but with the interface width constant. As seen in figure 6, the data collapse is excellent.

The coupling strength $J$ affects the finite size scaling in two ways. First, as $J$ increases, the interface becomes flatter. Accordingly, as seen in figure 6, for larger coupling strengths, the peak in $f_c(L)$ occurs for smaller targets. The second effect of $J$ is to alter the depinning
threshold, so that as $J$ increases, $f_\infty$ decreases. Heuristically, this can be understood by considering that as $J$ increases, so does the forward force on sites lagging behind the mean interface position, so that the external driving force needs only be able to overcome the local pinning force on a few sites to start a large avalanche.

For completeness, we have also made simulations using Gaussian distributed quenched disorder, with $\langle \eta \rangle = 0.5, \sigma(\eta) = 0.2$, and find very similar behaviour to that observed with uniform-distributed disorder.

4. Effects of hardening and machine stiffness

Finally, we discuss two related scenarios: strain hardening and strain controlled experiments. In materials subject to hardening, the external stress required to induce plastic deformation grows with the plastic strain. A simple model for this can be incorporated into the model for interface depinning, by adding a force term that is opposed to the interface motion, and proportional to the strain associated with the interface displacement, $\bar{u}/L^2$, where $\bar{u}$ is the average of $u_x$ over $x$. The total force on site $x$ is then

$$f_x = f_{el}(x, u_x) - \eta(x, u_x) + f - k \frac{\bar{u}}{L^2},$$

where $k$ is a hardening constant. The hardening force, $f_H = -k\bar{u}/L^2$, effectively reduces the external force acting on the dislocation as the dislocation displacement increases, and for large displacements the hardening dominates and the driving force required to move the interface is proportional to the mean displacement. As a result, the interface remains critical and never depins.

Alternatively, in the case of a strain controlled experiment, one can imagine that a subsystem undergoing plastic deformation is coupled to a spring of stiffness $k$, and a total deformation $\gamma_T$ is imposed (as illustrated in the inset of figure 1(b)). In the absence of a plastic subsystem, $\gamma_T$ would be attained using a stress of $\sigma_T = k\gamma_T$, as per Hooke’s law, but when plastic deformation occurs, the spring extension/compression is reduced by $\bar{u}/L^2$ so that $\sigma_T$ is given by

$$\sigma_T = k\left(\gamma_T - \frac{\bar{u}}{L^2}\right).$$

Simulations of an interface subject to strain hardening can be reinterpreted in terms of strain controlled experiments [30, 31], by treating $f$ as $k\gamma_T$ and plotting $\sigma_T$ versus $\gamma_T$, as illustrated in figure 1(b). We work in units where stresses and forces are interchangeable.

We will first discuss how size effects in interface depinning are affected by strain hardening, and then discuss size effects in strain controlled experiments.

4.1. Hardening

We have simulated the quenched Edwards–Wilkinson model with an added hardening term. A typical ‘stress–strain’ curve is shown in figure 1(a). For small driving force and interface displacement, the behaviour is that of the quenched Edwards–Wilkinson model without hardening, but as the driving force and displacement are increased, the ‘stress–strain’ curve departs from its zero-hardening counterpart. When hardening
Size effects in dislocation depinning models for plastic yield

Figure 7. Size dependence of depinning force for the quenched Edwards–Wilkinson equation subject to hardening of strength $k$. For sufficiently large $k \gamma L^2$, data from simulations with different $\gamma$ values can be collapsed by subtracting $k \gamma$ from $\langle f_c \rangle$, to give finite size scaling of the form described by (8), as indicated by the black solid line. This collapse does not work for small $k \gamma L^2$, as seen in the data for $k = 1, \gamma = 0.01$ (open circles), which only fall on to the collapsed curve for sufficiently large $L$. The dashed horizontal line indicates $f_\infty$, the $L \to \infty$ limit of $f_c(L)$ in the absence of hardening. Solid symbols represent $k = 10$ and open symbols represent $k = 1$; the shapes represent $\gamma = 0.01$ (circles), $\gamma = 0.15$ (triangles) and $\gamma = 2$ (inverted triangles). Error bars represent the standard error.

dominates, if a force of $f^*$ has been applied to drive the interface to position $\bar{u}^*$ and it has come to a stop, the force required to start it moving again is

$$f = f^* + k \frac{\bar{u}^*}{L^2}. \quad (7)$$

This is the behaviour seen in the ‘stress–strain’ curve, which approaches a line of slope $k/L^2$ in the limit $\bar{u} \to \infty$.

In figure 7, we plot typical $f_c$ versus $L$ curves. With the introduction of hardening, it is no longer possible to collapse data for the whole space of parameters $\gamma$ and $k$. Instead, the observed behaviour depends on $k \gamma$. For large $k \gamma$, $f_c(L)$ curves from simulations with the same $k$ can be collapsed by subtracting $k \gamma$ from $\langle f_c \rangle$, but data from simulations with different $k$ cannot be collapsed onto a single form. Furthermore, if $k \gamma$ is small, this procedure does not yield data collapse, as exemplified in figure 7 by the $f_c(L) - k \gamma$ curve for $\gamma = 0.01, k = 1$.

We can interpret this by considering the maximum hardening force applied during simulations, which is applied when the interface is at its target: $\bar{u} = \gamma L^2$. For small $k \gamma$, the hardening force is always negligible, and the size effects are the same as those discussed in previous sections. When $k \gamma$ is large, hardening becomes important. A first approximation to understand its role is to imagine that the interface is driven to the target without hardening, and then the hardening force is turned on. Assuming the target is sufficiently large, the force required to attain it (with zero hardening) is given by (1). Then, from (7), the force required to continue to move the interface is

$$f = f_\infty + a L^{-1/\nu} + k \gamma, \quad (8)$$

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11
Figure 8. Size dependence of the measured yield stress $\sigma_{T,y}$ in strain controlled simulations of the quenched Edwards–Wilkinson model, with spring constant $k = 10$ (open symbols), and $k = 1$ (filled symbols). The yield strains used are $\gamma_{T,y} = 1$ (triangles), 0.5 (circles), 0.4 (inverted triangles) and 0.04 (squares). The black lines indicate the expected finite size scaling of (9), with $f_\infty = 0.424$ and $1/\nu = 0.75$. When $k\gamma_{T,y}$ is too small, this scaling breaks down because the interface never reaches criticality, as seen for the lower two curves ($k = 10, \gamma_{T,y} = 0.04$, and $k = 1, \gamma_{T,y} = 0.4$).

where $f_\infty$ is the $L \to \infty$ limit of $f_c(L)$ when $k = 0$. This form agrees well with our simulation data, as seen in figure 7. Because of the extra $\gamma$-dependent term, it is not possible to use linear rescalings to eliminate the $\gamma$ dependence of $(f_c - f_\infty)$, and exact data collapse is not possible.

4.2. Strain controlled experiments

In strain controlled simulations, the choice of yield strain is more crucial than in stress controlled simulations. In figure 8, we plot the size dependence of $\sigma_{T,y}$ for several choices of yield strain $\gamma_{T,y}$, for strain controlled simulations of an interface described by the quenched Edwards–Wilkinson equation. For sufficiently large $\gamma_{T,y}$, the behaviour described in (1) is recovered and $\sigma_{T,y}(L)$ approaches the same $L \to \infty$ limit found in stress controlled simulations, with the same finite size exponent $1/\nu = 0.75$. However, for smaller $\gamma_{T,y}$, this picture breaks down.

We can interpret this as follows. Recall that in stress controlled simulations of the quenched Edwards–Wilkinson model, the target $\gamma L^2$ grows faster than the interface width $wL^\xi$ and therefore as $L \to \infty$ the correct finite size scaling must occur. However, in strain controlled simulations, the target $\gamma_{T,y} = f_y/k$ may correspond to a driving force $f_y$ that is always smaller than the true depinning force (as given in (1)). In that case, the interface never reaches criticality for any $L$, and the standard finite size scaling does not hold. Note that the minimum strain target depends on both the depinning force of the interface and also the spring constant $k$, as shown in figure 8.

If the target is large enough for the interface to reach criticality, then the size effects can be predicted by reinterpreting the strain controlled simulation as a stress controlled simulation with hardening. The finite size behaviour of $f = k\gamma_T$ is given by (1) and (7),

\[
doi:10.1088/1742-5468/2013/04/P04029
\]
so that
\[ \sigma_{T,y} = f - k \frac{\bar{u}_y}{L^2} = f_\infty + aL^{-1/\nu}, \] (9)

where \( \bar{u}_y \) is the mean interface displacement when the total strain is equal to the yield strain. In other words, for sufficiently large yield strains, stress controlled and strain controlled simulations give the same results, and furthermore, in that case, the measured yield stress in a strain controlled experiment does not depend on the yield strain.

5. Conclusion

The question of how to measure yield stress is an important methodological problem in the determination of finite size effects in microscale plasticity, which to our knowledge has received little attention previously. In this paper, we have gained insight into how to avoid spurious size effects in a depinning model for the onset of plastic yield when the yield strain is arbitrarily defined. Reassuringly, in all the models we study, for sufficiently large systems the correct finite size scaling is recovered. However, for smaller systems, the finite size effects are not described by the scaling law (1), and instead depend on various details of the model under study, a result which has implications, especially for simulation studies of depinning and plasticity, where until recently only relatively small sizes were accessible computationally.

We also observe that the idea that the crossover between ‘small’ and ‘sufficiently large’ systems is determined by the roughness of a self-affine interface is one that has broader relevance. For example, experimental studies of polycrystalline copper samples under tension [32] have revealed that profiles of the surface height \( y(x) \) (with \( x \) along the sample axis) are self-affine, that is, \( W(L) \equiv \langle |y(x+L) - y(x)| \rangle \sim L^H \), with \( H \approx 0.75 \) for imposed strains \( \gtrsim 10\% \). This surface profile is related to the variation in the strain, which in a sample of length \( L \) is \( \Delta \gamma = W(L)/L \sim L^{-0.25} \). Imposing the requirement that the total strain should be larger than its variance, a minimum system size \( L_c \) is determined by setting \( \gamma = \Delta \gamma \). In the case of polycrystalline copper, this length scale turns out to be small enough to have no experimental effect: using the data from Zaiser et al’s experiments [32], \( L_c \approx 1.6 \times 10^{-9} \text{ m} \) for \( \gamma_{\text{tot}} = 10\% \). For different materials and loading conditions, however, it may be possible for \( L_c \) to be comparable with experimental sample sizes, leading to intricate size effects.

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13
Monte Carlo dynamics of driven elastic strings in disordered media

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