



CHORUS

This is the accepted manuscript made available via CHORUS. The article has been published as:

Controlling Strain Bursts and Avalanches at the Nano- to Micrometer Scale

Yinan Cui, Giacomo Po, and Nasr Ghoniem

Phys. Rev. Lett. **117**, 155502 — Published 7 October 2016

DOI: [10.1103/PhysRevLett.117.155502](https://doi.org/10.1103/PhysRevLett.117.155502)

Controlling Strain Bursts and Avalanches at the Nano-to-Micro Scale

Yinan Cui,* Giacomo Po, and Nasr Ghoniem

*Mechanical and Aerospace Engineering Department, University of California,
Los Angeles, 420 Westwood Plaza, Los Angeles CA, 90095*

We demonstrate, through 3-dimensional discrete dislocation dynamics simulations, that the complex dynamical response of nano and micro crystals to external constraints can be tuned. Under load rate control, strain bursts are shown to exhibit scale-free avalanche statistics, similar to critical phenomena in many physical systems. For the other extreme of displacement rate control, strain burst response transitions to quasi-periodic oscillations, similar to stick-slip earthquakes. External load mode control is shown to enable a qualitative transition in the complex collective dynamics of dislocations from self-organized criticality to quasi-periodic oscillations.

Power-law scaling of avalanche phenomena is widely observed in many nonequilibrium natural systems. Examples are found in geologic earthquakes, snow avalanches, sand pile slides, and strain bursts during plastic flow [1, 2]. The realization that such vastly diverse physical systems display common features, implies scale invariance and compels a search into universal fundamental laws. The common scaling raises the possibility that the intricate system behavior can be described by simple local rules, despite the complexity of the underlying internal dynamics. One concept that is widely used to interpret this universality is self-organized criticality (SOC) [3]. In a SOC system, the dynamics has an attractor characterized by infinite correlation time and length, hence displaying scale-free scaling. A key hypothesis behind this abstraction is that the driving force varying rate is much slower than the internal relaxation rate [3, 4] of a system undergoing SOC. Nevertheless, since this condition may not always hold, one wonders if the qualitative aspects of a system’s dynamical behavior change when the driving force changing rate is comparable to its internal relaxation rate? Our objective here is to investigate the relationship between the external driving force and relaxation dynamics associated with strain bursts during nano- and micro-scale plastic deformation of crystals.

At the smallest of physical scales (e.g. nano-to-micro scale), the release of plastic strain by intermittent “bursts” has been found to belong to this power-law scaling behavior [2, 5–8]. One additionally unique aspect of plasticity is that the driving force varying rate can be experimentally tailored. Considering a simple but illustrative case, a pillar is subjected to uniaxial compression in Fig. 1. The force actuator, typically a voice coil, can exert an open-loop stress rate $\dot{\sigma}_0$ and/or be controlled to impose a strain rate $\dot{\epsilon}_0$. For a proportional controller with stiffness K_p , the internal stress rate in the pillar is [9],

$$\dot{\sigma} = \frac{\alpha E}{1 + \alpha}(\dot{\epsilon}_0 - \dot{\epsilon}^p) + \frac{\dot{\sigma}_0}{1 + \alpha} \quad (1)$$

where $\alpha = K_p/K$ is the relative stiffness ratio, $K =$

EA/H is the pillar stiffness, E , A and H are the Young module, cross section area and height of the pillar, respectively. $\dot{\epsilon}^p$ is the plastic strain rate due to all internal dislocation dynamical activities. Once the stiffness ratio α is infinitely large, or $\dot{\sigma}_0$ and $\dot{\epsilon}_0$ are very low, $\dot{\sigma}$ becomes very sensitive to $\dot{\epsilon}^p$, implying that the driving force changing rate ($\dot{\sigma}$) is dominated by and comparable to its internal relaxation rate ($\dot{\epsilon}^p$). This indicates that the corresponding slip statistics are expected to violate SOC.

However, it is generally believed that the machine stiffness K_p only contributes to the cutoff of the power law scaling [6, 8, 10]. The present investigation demonstrates that, if the machine stiffness is extremely high, dislocation avalanche dynamics (and hence strain bursts) undergo a transition from scale-free critical behavior to quasi-periodic oscillations. Interestingly, this is consistent with recent findings on the role of very slow loading rates (low $\dot{\sigma}_0$ and $\dot{\epsilon}_0$) [11, 12], as suggested by Eq. 1. The underlying microstructure mechanism for this dynamical regime transition are disclosed. Considering that the dynamical behaviors under soft or hard machine stiffness conditions are vastly different, the corresponding intermittent plasticity will henceforth be described as either *avalanche* or *burst*, respectively. Moreover, a dislocation based branching model is proposed, giving a clear and precise physical picture of the avalanche dynamical behavior.

The vast majority of existing submicron mechanical testing experiments can only cover a narrow range of machine stiffness. In addition, the time necessary for dislocations to travel through 1 μm sample is estimated at about 1 ns [13]. In state-of-the-art experiments, the feedback loop frequency is ≈ 78 kHz (time constant $\approx 13 \mu\text{s}$) [8], which means that current experimental controller response rate is much slower than sample plastic relaxation rate by 4 orders of magnitude. Namely, the driving force changing rate is much slower than internal relaxation rate. Therefore, most previous experimental conditions correspond to the regime where SOC is observed. Discrete dislocation dynamics (DDD) studies, as a computer simulation tool, make it possible to supple-

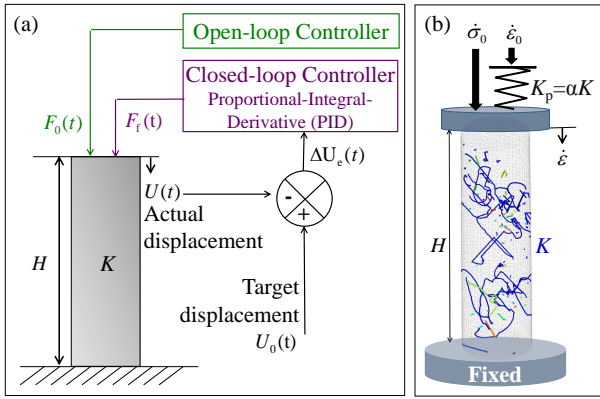


FIG. 1. Simplified sketch of pillar compression. (a) Experimental setup with an open-loop (directly applying a force F_0) and closed-loop control (to realize displacement control); (b) Simulation setup, a proportional dominated closed-loop control is considered here with $F_f = K_p(U_0 - U)$, which is simplified as a spring with a finite machine stiffness K_p . The external stress rate $\dot{\sigma}_0 = \dot{F}_0/A$, target strain rate $\dot{\epsilon}_0 = \dot{U}_0/H$, actual strain rate $\dot{\epsilon} = \dot{U}/H$, where A and H are the cross section area and height of the pillar, respectively. One typical dislocation configuration in a pillar with $d=3000$ b is shown as an example

power law exponent for the corresponding probability density is found to be 1.5, agreeing well with the generally accepted range of 1.35 \sim 1.67 [5, 6, 19–21]. In addition, the power law distribution is consistent across system size, implying the existence of scale-free universality. In contrast, the CCDF of ΔU under pure strain control seems not to exhibit power-law scaling behavior for both small and large system sizes. Meanwhile, most of the data concentrate within one order of magnitude. An analogous breakdown of the power law scaling under pure strain control is also observed for the statistics of burst duration [9].

Then, how to describe the strain burst statistics under pure strain control? When discussing the temporal statistics of earthquakes, distinct dynamical behaviors are distinguished by the coefficient of variation $C = s_x/\bar{x}$ [22], where s_x and \bar{x} are the standard deviation and mean value, respectively. For the cases of $C > 1$ and $C < 1$, the distribution is referred to as “clustered” and “quasi-periodic”, respectively; otherwise, if $C = 1$, it is a random Poisson distribution [22]. Taking the results of ΔU here, C is calculated as 1.9 and 0.9 under pure stress and pure strain control, respectively. This suggests that the dynamical behaviors under pure strain control becomes quasi-periodic. Similar to previous studies [11, 22], quasi-periodicity here is found to be stochastic, due to the intrinsic scatter induced by random cross slip or different dislocation configurations. Quasi-periodic strain bursts under pure strain control are manifested through the smoothed plastic strain rate, as clearly shown in Fig. 2b. Here, the time series of $\dot{\epsilon}^p$ is smoothed over a fixed time window of 0.24 μs . For comparison, the smoothed plastic strain rate under pure stress control, also shown in Fig. 2c, corresponds to a depinning phase transition.

Close examination of dislocation configuration evolution reveals that the mechanisms that control avalanche versus quasi-periodic burst behavior are significantly different, and are highly dependent on the external constraint. First, let’s consider pure strain control. In the submicron regime (e.g. $d=1000$ b), each strain burst is found to be dominated by sequential activation and deactivation of single arm dislocation sources. Once a source is activated, the accompanying plastic strain leads to a decrease in the stress level (see Eq. 1, $\alpha = +\infty$). Even if a weaker source is formed during one burst event, sometimes it also cannot operate due to the lower prevailing stress after relaxation. This makes it difficult to trigger simultaneous operation of multiple dislocation sources (see Fig. 3b), especially for small samples with limited volume. We have recently shown that dislocation sources themselves are transient, because they generally result from the formation of dipolar loops by cross-slip [7]. This rapid stress drop prevents the strain burst from continuously growing into a full-fledged avalanche. Consequently, large-scale cooperative interactions between

93 ment experimental testing and explore regimes that are
 94 currently difficult to access experimentally [6, 14]. The
 95 current research presents the first systematic 3D-DDD
 96 investigation on the slip statistics at submicron scale,
 97 accounting for the effects of the interaction of an external
 98 loading mode [15–17]. Compared with most of existing
 99 two dimensional (2D) DDD studies [2, 18], the key
 100 approximations inherent in 2D techniques are resolved.
 101 Specifically, dislocation junction formation and destruction,
 102 and the occurrence of cross slip are all accounted for
 103 with minimal ad hoc assumptions.

104 The simulation setup is schematically shown in Fig.
 105 1b. We conducted simulations of compression tests on
 106 Cu pillars of different diameters, ranging from 1000-3000
 107 b (≈ 300 nm- 1 μm), where b is the burgers vector mag-
 108 nitude. The aspect ratio H/d is 3. Two extreme machine
 109 stiffness cases are first considered, corresponding
 110 to *pure strain control* ($\alpha = +\infty$) and *pure stress con-*
 111 *trol* ($\alpha = 0$). Here, under pure strain control, the applied
 112 strain rate $\dot{\epsilon}_0 = 960\text{s}^{-1}$. Correspondingly, under
 113 pure stress control, the actual loading rate $\dot{\sigma}_0$ is $E\dot{\epsilon}_0$.
 114 Fifty and twenty separate simulations with different initial
 115 dislocation configurations are carried out under each
 116 loading mode for $d=1000$ b and $d=3000$ b, respectively.

117 Figure 2a presents the results of statistical analysis
 118 of the burst displacement magnitude ΔU . To obtain
 119 maximum resolution of the limited simulation data set,
 120 the complementary cumulative distribution function
 121 (CCDF) is used. Fig. 2a clearly illustrates that ΔU ,
 122 under pure stress control, exhibits a well-defined power law
 123 distribution spanning several orders of magnitude. The

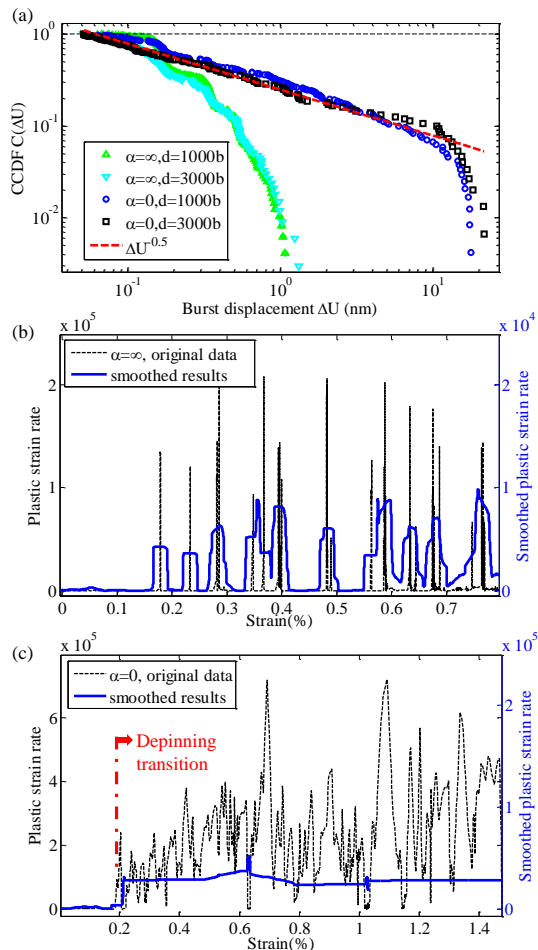


FIG. 2. (a) Statistical properties of burst displacement under pure strain and stress control modes for pillar with diameters $d = 1000$ b and 3000 b. (b-c) Typical evolution of plastic strain rate and its averaged value in $0.24 \mu\text{s}$ windows, showing (b) quasi-periodic strain bursts under pure strain control, and (c) depinning transition dislocation avalanche under pure stress control

180 dislocations that can lead to SOC cannot be realized
 181 under pure strain control. Note that this discussion applies to a sample size ranging from several nanometers
 182 to about 1 micrometer. For smaller pillars, surface nucleation of dislocations becomes dominant [23], and the
 183 rapid stress drop may inhibit correlated surface nucleation, while for larger pillar size, Taylor-type interaction
 184 mechanisms prevail [24, 25], and the rapid stress drop may suppress cooperative dislocation interactions.
 185

186 By contrast, dislocation avalanche under pure stress control is clearly associated with correlated dislocation
 187 motion. According to Eq. 1, when $\alpha = 0$, the stress rate cannot sense the internal dislocation activity. Thus, the
 188 stress level keeps almost constant during each avalanche event (see Fig. 3a). If one activated source leads to the
 189 formation of a weaker one, it can be immediately activated. Thus, distinctly different from the strain control
 190
 191
 192
 193
 194
 195
 196

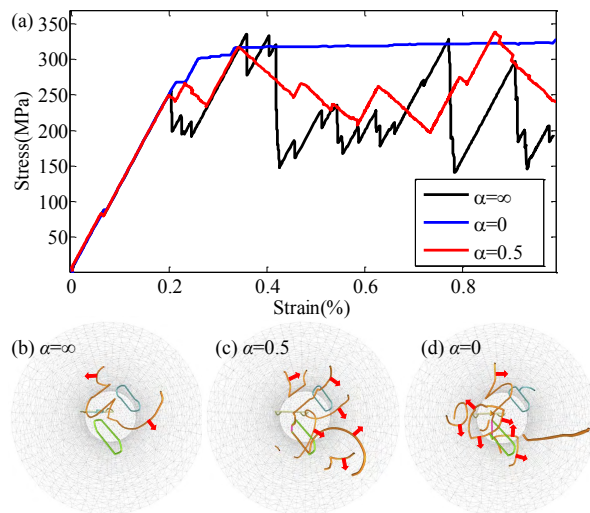


FIG. 3. Typical simulation results under different loading modes for pillar with $d = 1000$ b. (a) Stress-strain curves; (b-d) Snapshots of dislocation configurations (from top view) at a strain value of 0.4%, arrows indicate the bowing out directions of activated sources

197 case discussed above, multiple sources can operate in a
 198 correlated fashion (see Fig. 3d). All correlated sources
 199 contribute then to an increasing magnitude of the strain
 200 burst, turning it into an “avalanche”. Such highly correlated
 201 dynamical behavior suggests a close-to-criticality
 202 nonequilibrium state [3].

203 Since it is difficult to experimentally achieve such extreme
 204 machine stiffness, it is then interesting to examine dislocation
 205 dynamics with finite machine stiffness. All the results in Fig. 3a
 206 correspond to the same size and initial dislocation configuration. The
 207 calculated stress-strain curve with finite machine stiffness
 208 ($\alpha = 0.5, \dot{\sigma}_0 = 0$) in Fig. 3a displays a very similar behavior
 209 to experimental results [8, 21], and exhibits a serrated yield
 210 character with longer decaying stages as compared to pure strain
 211 control. The observation of simultaneous operation of multiple
 212 sources in Fig. 3c suggests that a finite machine stiffness actually
 213 promotes correlated dislocation motion, compared with pure strain
 214 control.

215 To further elucidate the statistical difference between avalanche
 216 versus quasi-periodic dynamics, a simple dislocation based
 217 branching model is proposed. It is inspired by the present 3D-DDD
 218 simulations, and motivated by Zapperi’s sand-pile branching model
 219 [26], in which we translate the branching idea into dislocation
 220 language. The discrete plastic deformation is assumed to mainly
 221 proceed through the intermittent activation of dislocation sources
 222 [27, 28]. One activated source may lead to the stochastic
 223 generation/activation of other sources, similar to a branching
 224 process shown in Fig. 4a.

225 The detailed algorithm proceeds as follows. Assuming

229 a pillar initially with n_s dislocation sources, we can randomly
 230 give each source a specific length λ according to
 231 a given source length probability distribution. The fate
 232 of each source (active or not) is determined by checking
 233 whether the instantaneous applied stress σ_k can reach
 234 the source operation stress,

$$\sigma_k \cdot M \geq \tau_0 + \alpha_1 \mu b \sqrt{\rho} + \alpha_2 \mu b / \lambda \quad (2)$$

235 where M is Schmid factor, the three terms on the
 236 right hand are lattice friction stress, the elastic interaction
 237 stress described by Taylor relation, and the source
 238 strength, respectively. α_1 and α_2 are dimensionless constant,
 239 set to 0.5 and 1 [28], respectively. ρ is the instantaneous
 240 dislocation density, estimated by dividing the total
 241 source length by the pillar volume.

242 Once the weakest source is activated during deformation,
 243 a strain burst begins [28, 29]. After each source is
 244 activated, the burst strain S_k increases by a specific
 245 value $d\varepsilon^p$. Considering that ε^p is much higher than the
 246 applied strain rate $\dot{\varepsilon}_0$ during a strain burst, according
 247 to Eq. 1, σ_k drops by $Ed\varepsilon^p\alpha/(\alpha+1)$, and the total
 248 strain increases by $d\varepsilon^p/(\alpha+1)$. It is assumed that the
 249 activated source is broken (ceases to operate) after it
 250 sweeps the entire slip plane once. However, it can randomly
 251 trigger the generation of additional n_a sources. If
 252 the newly generated source can be activated according
 253 to Eq. 2, it triggers subsequent generation of n_a sources.
 254 Otherwise, the new source is stored for possible dislocation
 255 generation, which may activate during subsequent
 256 deformation stages. This branching source generation
 257 process repeats itself until all dislocation sources cannot
 258 be activated under the combined effect of the instantaneous
 259 applied stress and the resistance stress, given by
 260 the right side of Eq. 2 (see Fig. 4a). At this instance,
 261 this strain burst event stops and the stress continues to
 262 increase till it triggers another strain burst event.

263 In the following, we investigate the slip statistics using
 264 this abstract branching model, and compare to the more
 265 fundamental DDD simulations discussed above. Compression
 266 tests are also modeled for Cu pillars with diameter
 267 $d=1000$ b and 3000 b. Similar to DDD simulation,
 268 surface nucleation is not considered. If the stress is
 269 higher than the surface nucleation stress (about 1.2 GPa
 270 for Cu [30]) or if the strain is higher than 0.5, events
 271 are not recorded. If there is only one activated source,
 272 each burst strain corresponds to the generated plastic
 273 strain when the dislocation sweeps the entire slip plane
 274 once. Therefore, $d\varepsilon^p$ is set to $bM/H/\cos\beta$ [28], where
 275 β is the angle between the normal direction of the slip
 276 plane and the loading orientation. Through examination
 277 of the dislocation configuration evolution, n_a is taken as
 278 the nearest integer of $2 \cdot rand$, where $rand$ represents a
 279 random value from 0 to 1. Accordingly, the probabilities
 280 of n_a being 0, 1 and 2 are 25%, 50% and 25%, respectively.
 281 This is different from previous sand-pile branch-

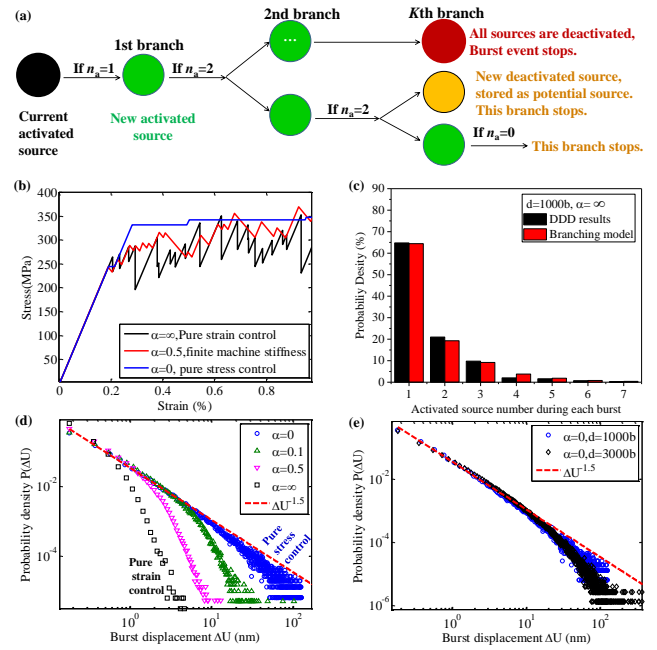


FIG. 4. (a) Schematic showing the random branching dislocation source generation and activation process, n_a is the number of newly generated dislocation sources, green filled circles represent that new source is activated, only activated source may trigger further branching process; (b-d) Typical predicted results for pillar with $d=1000$ b, (b) Stress-strain curve, (c) Comparison of activated source number during each burst under pure strain control, (d) Probability density function of burst displacement for different machine stiffness; (e) Probability density function of burst displacement for different sample sizes

282 ing model [26], where the new activated site number was
 283 taken a constant value of 2. $n_a = 0$ means that the
 284 source is destroyed after operation once, $n_a = 1, 2$ indicate
 285 that other sources are generated due to interactions
 286 with other dislocations, cross slip, forming superjogs, or
 287 forming dipolar loops [7]. Note that, more deactivated
 288 sources may be left in the sample if $n_a=2$, leading to
 289 a slight increase in the dislocation density ρ after each
 290 branching process. This results in an increase in the
 291 elastic interaction resistance stress. Similar to 2D-DDD
 292 simulations [31], the source length is assumed to follow
 293 a Gaussian distribution, with a mean value $\bar{\lambda} = d/2$,
 294 determined according to the yield stress of our DDD
 295 results. Its standard deviation is set to $20\%\bar{\lambda}$, so that
 296 the predicted activated source number for each strain
 297 burst event is statistically equivalent to those obtained
 298 by our DDD results under pure strain control (see Fig. 4c).

299 Fig. 4b presents predicted typical stress-strain curves
 300 under different loading modes, which agree well with our
 301 simulation results in Fig. 3a, including the stress level
 302 and the stepped or serrated burst features. In addition,
 303 the power law scaling of burst displacement ΔU is also
 304 well reproduced under pure stress control for different

pillar sizes in Fig. 4e. The power law exponent of the probability distribution of ΔU agrees with that obtained by the present 3D-DDD. Fig. 4d clearly indicates that as the machine stiffness increases, the power law tails gradually become too wide to recognize proper scale-free power law statistics.

The excellent agreement between the abstract branching model prediction and the fundamental 3D-DDD simulations further verify that hard machine stiffness leads to deviation from scale-free SOC, because the rapid stress relaxation disturbs correlated dislocation motion. The current finding offers a new pathway towards controlling the correlated extent of dislocation dynamics and the intermittent statistics by tuning the machine stiffness. It opens up new possibilities for novel experiments with faster response rate that can reveal the quasi-periodic oscillation dynamics of dislocation systems. The importance of often-neglected interaction with the external loading system on intermittent plastic flow has been demonstrated. The complex dynamics of collective dislocations producing strain bursts is shown to be controlled through simple tuning of the relative value of driving force rate to internal relaxation rate.

This material is based upon work supported by the U.S. Department of Energy, Office of Science, Office of Fusion Energy Sciences, under Award Number DE-FG02-03ER54708, and the US Air Force Office of Scientific Research (AFOSR), under award number FA9550-11-1-0282. We would like to thank Professor Michael Zaiser (Friedrich-Alexander University Erlangen-Nuremberg) and Professor Stephanos Papanikolaou (Johns Hopkins University) for inspiring comments and discussions.

* cuiyinan@ucla.edu

[1] J. T. Uhl *et al.*, *Sci. Rep.* **5**, 16493 (2015).
 [2] M. C. Miguel *et al.*, *Nature* **410**, 667 (2001).
 [3] P. Bak, *How nature works: the science of self-organized criticality* (Copernicus, New York, ADDRESS, 1996).
 [4] P. Sammonds, *Nature Mater.* **4**, 425 (2005).
 [5] D. M. Dimiduk, C. Woodward, R. LeSar, and M. D. Uchic, *Science* **312**, 1188 (2006).
 [6] F. F. Csikor *et al.*, *Science* **318**, 251 (2007).
 [7] T. Crosby, G. Po, C. Erel, and N. Ghoniem, *Acta Mater.* **89**, 123 (2015).
 [8] R. Maass, M. Wraith, J.T. Uhl, J.R. Greer, K.A. Dahmen, *Phys. Rev. E* **91**, 042403 (2015).
 [9] See Supplemental Material, which includes Refs. [32-36].
 [10] M. Zaiser and N. Nikitas, *J. Stat. Mech.: Theory Exp.* **2007**, P04013 (2007).
 [11] S. Papanikolaou *et al.*, *Nature* **490**, 517 (2012).
 [12] K. A. Dahmen, Y. Ben-Zion, and J. T. Uhl, *Nature Phys.* **7**, 554 (2011).
 [13] D. Dimiduk *et al.*, *Philos. Mag.* **90**, 3621 (2010).
 [14] J. A. El-Awady, *Nature Comm.* **6**, (2015).
 [15] N.M. Ghoniem, S.H. Tong, and L.Z. Sun, *Phys. Rev. B*

61, 913 (2000).
 [16] G. Po and N. Ghoniem, (<https://bitbucket.org/model/model/wiki/home>) (2015).
 [17] G. Po, M. Lazar, D. Seif, and N. Ghoniem, *J. Mech. Phys. Solids* **68**, 161 (2014).
 [18] M. Zaiser, *Adv. Phys.* **55**, 185 (2006).
 [19] K. Ng and A. Ngan, *Acta Mater.* **56**, 1712 (2008).
 [20] X. Zhang, B. Pan, and F. Shang, *Europhys. Lett.* **100**, 16005 (2012).
 [21] S. Brinckmann, J.-Y. Kim, and J. R. Greer, *Phys. Rev. Lett.* **100**, 155502 (2008).
 [22] Y. Ben-Zion, *Rev. Geophys.* **46**, (2008).
 [23] J. R. Greer and W. D. Nix, *Phys. Rev. B* **73**, 245410 (2006).
 [24] R. Gu and A. Ngan, *Acta Mater.* **60**, 6102 (2012).
 [25] R. Gu and A. Ngan, *J. Mech. Phys. Solids* **61**, 1531 (2013).
 [26] S. Zapperi, K. B. Lauritsen, and H. E. Stanley, *Phys. Rev. Lett.* **75**, 4071 (1995).
 [27] D. Kiener and A. Minor, *Acta Mater.* **59**, 1328 (2011).
 [28] Y. Cui, P. Lin, Z. Liu, and Z. Zhuang, *Int. J. Plast.* **55**, 279 (2014).
 [29] J. A. El-Awady, M. Wen, and N. M. Ghoniem, *J. Mech. Phys. Solids.* **57**, 32 (2009).
 [30] S.-W. Lee, A. T. Jennings, and J. R. Greer, *Acta Mater.* **61**, 1872 (2013).
 [31] S. Papanikolaou, H. Song, and E. Van der Giessen, *J. Mech. Phys. Solids.* (Submitted).
 [32] M. D. Uchic and D. M. Dimiduk, *Mater. Sci. Eng. A* **400**, 268 (2005).
 [33] J. Weiss, T. Richeton, F. Louchet, F. Chmelik, P. Dobron, D. Entemeyer, M. Lebyodkin, T. Lebedkina, C. Fressengeas, and R. J. McDonald, *Phys. Rev. B* **76**, 224110 (2007).
 [34] C. Fressengeas, A. J. Beaudoin, D. Entemeyer, T. Lebedkina, M. Lebyodkin, V. Taupin, *Phys. Rev. B* **79**, 014108 (2009).
 [35] D. Dimiduk *et al.*, *Modell. Simul. Mater. Sci. Eng.* **15**, 135 (2007).
 [36] N. Friedman, A. T. Jennings, G. Tsekenis, J.-Y. Kim, M. Tao, J. T. Uhl, J. R. Greer, and K. A. Dahmen, *Phys. Rev. Lett.* **109**, 095507 (2012).