Critical scaling near the yielding transition in granular media

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(Received 28 June 2017; revised manuscript received 24 March 2018; published 6 June 2018)

We show that the yielding transition in granular media displays second-order critical-point scaling behavior. We carry out discrete element simulations in the low-inertial-number limit for frictionless, purely repulsive spherical grains undergoing simple shear at fixed nondimensional shear stress Σ in two and three spatial dimensions. To find a mechanically stable (MS) packing that can support the applied Σ, isotropically prepared states with size L must undergo a total strain γms(Σ, L). The number density of MS packings (νms) vanishes for Σ > Σc, ∝ 0.11 according to a critical scaling form with a length scale ξ ∝ |Σ − Σc|−νγ, where ν ≈ 1.7−1.8. Above the yield stress (Σ > Σc), no MS packings that can support Σ exist in the large-system limit L/ξ ≫ 1. MS packings generated via shear possess anisotropic force and contact networks, suggesting that Σ is associated with an upper limit in the degree to which these networks can be deformed away from those for isotropic packings.

DOI: 10.1103/PhysRevE.97.062901

I. INTRODUCTION

Granular materials consist of macroscopic grains that interact via dissipative contact forces. Their response to external forcing depends on the ratio Σ = τ/p of the applied shear stress τ to the normal stress p, where p is small compared with the stiffness of the grains [1,2]. Granular media, like other amorphous materials [3–6], possess a yield stress Σc. Generally, grains will always rearrange when the applied forces are changed. However, when Σ < Σc, grains move temporarily until finding a solid-like mechanically stable (MS) packing that can support the applied Σ [7–9]. When Σ = Σc, the strain γms required to find MS packings diverges. When Σ > Σc, grains cannot find MS packings, and fluid-like flow persists indefinitely.

In the jamming paradigm [10–16], which is commonly used to understand fluid-solid transitions in granular materials, the packing fraction φ is the controlling variable. Fluid- and solid-like states occur for φ < φJ, and φ > φJ, respectively. A diverging length scale ξJ ∝ |φ − φJ|−ν controlling the mechanical response near φJ [12,14–16]. However, MS packings of frictionless grains at fixed p and varied Σ all have a packing fraction φms(Σ) ∝ φJ [9]. Thus, Σ ∼ Σc may represent a fluid-solid transition distinct from jamming, where the structure of the force and contact networks, not φ, plays a dominant role.

In this paper, we show evidence that the number density of MS packings vanishes at Σ = Σc in the large-system limit, with second-order critical scaling that is not related to φ but instead to the structure of the force and contact networks. We measure γms in systems of frictionless grains subjected to simple shear as a function of Σ and system size L. We postulate a second-order critical-point scaling form for γms with a diverging length scale ξ ∝ |Σ − Σc|−νγ. The data for γms collapse onto two branches: Σ > Σc and Σ < Σc. The functional form of these branches suggests that the number density of MS packings, which is proportional to (γms)−1, vanishes in the large-system limit according to a power law relation in |Σ − Σc|. Specifically, (γms)−1 ∝ |Σ − Σc|νφms for Σ < Σc, and (γms)−1 = 0 for Σ > Σc. For simple shear in two (2D) and three dimensions (3D), we find Σc ≈ 0.11, in agreement with previous studies [1,9,17]. MS packings exist for Σ > Σc in small systems, but the number vanishes as L/ξ increases.

For Σ < Σc, MS packings exist for all L, and large systems (L > ξ) are equivalent to compositions of uncorrelated smaller systems. Our results are insensitive to changes in the boundary conditions and driving method, which we explicitly show by performing additional simulations in a river-bed-like geometry in the viscous or slow-flow limit [18,19].

We find that the packing fraction φms(Σ, L) of MS packings is nearly independent of Σ. However, the anisotropy in both the stress and contact fabric tensors of MS packings increases with Σ, suggesting that Σ is associated with an upper limit to the structural anisotropy that can be realized in the large-system limit [9]. These results may help explain recent studies [20–24] showing that accurately modeling granular flows requires a cooperative length scale that grows as a power law in Σ − Σc. Our results may also be relevant to other amorphous solids that show similar spatial cooperativity near yielding [5,25–27].

The remainder of the paper is organized as follows: In Sec. II, we describe our simulation methods. In Sec. III we present our results, including the critical scaling of γms in Sec. III A and the microstructural properties of MS packings in Sec. III B. Section IV contains a summary and conclusions. We include additional details in Appendix A on the equations of motion and dimensional analysis. Appendix B demonstrates our methods for determining the critical exponents and Σc. Appendix C gives further discussion on the scaling collapse of (γms)−1 versus Σ.

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we vary using a modified velocity Verlet integration scheme. In 3D, Newton’s equations of motion for the wall as well as the plate and grains. We apply downward force per area the details of the top plate, provided no slip occurs between the plate. We have checked that our results are insensitive to and small enough to stop bulk grains from passing through the boundary, which is a plate consisting of rigidly connected boundaries. The system is driven by the upper boundary. We control force and not \( \dot{\gamma} \), so there are fluctuations in \( \Gamma \), but \( \Gamma = 5 \) keeps \( I < 10^{-4} \) even for \( \Sigma > \Sigma_c \). We have explicitly checked that our results are independent of \( \Gamma \) for several values of \( \Gamma \geq 3 \).

For the simple shear simulations, initial states (\( \Sigma = 0 \)) are prepared via uniaxial compression. Specifically, we begin with the top plate at very large \( y \gg L \), and we place the grains sparsely throughout the domain between the top plate and the lower boundary. We then apply finite \( p \) to the top plate and allow it to move freely until an MS packing is found. We then apply finite \( \Sigma \) to the top plate, which can move in all directions. The simulation ends when the upward and horizontal forces from the grains acting on the top plate exactly balance the applied \( \Sigma \). We find similar results when \( \Sigma \) is increased incrementally in small steps, and the total strain is integrated. Average grain displacement profiles are linear for both \( \Sigma < \Sigma_c \) and \( \Sigma > \Sigma_c \) [8,28], as expected for this system.

In addition to simple shear, we study 2D systems of bidisperse frictionless grains in a river-bed-like geometry, depicted in Fig. 1(c), which is similar to the system studied in Ref. [18]. The domain has a no-slip lower boundary at \( y = 0 \), a free upper boundary, and periodic horizontal boundaries in the \( x \) direction. The domain has a no-slip lower boundary at \( y = 0 \), a free upper boundary, and periodic horizontal boundaries in the \( x \) direction.

II. METHODS

As depicted in Fig. 1, we perform discrete element method simulations of simple shear in 3D and 2D, as well as in a 2D river-bed-like geometry subjected to a linear flow profile in the viscous limit. For the simple shear simulations, we study systems of bidisperse frictionless spheres in 3D and disks in 2D. Two-thirds of the grains are small and one-third are large, with diameter ratio 1.2 in 3D and 1.4 in 2D. The lateral directions \( x \) (in 2D and 3D) and \( z \) (in 3D only) are periodic with length \( L \), where \( L \) is the length of the box edge in units of the small grain diameter \( D \). The fixed lower \( y \) boundary consists of a no-slip wall. The system is driven by the upper boundary, which is a plate consisting of rigidly connected small particles, with gaps that are large enough to prevent slip and small enough to stop bulk grains from passing through the plate. We have checked that our results are insensitive to the details of the top plate, provided no slip occurs between the plate and grains. We apply downward force per area \(-p\hat{y}\) and horizontal force per area \( \tau \hat{x} \) to the upper plate and solve Newton’s equations of motion for the wall as well as \( N \) grains using a modified velocity Verlet integration scheme. In 3D, we vary \( N = L^3 \) from \( L = 3, N = 27 \) to \( L = 16, N = 4096 \). In 2D, we vary \( N = L^2 \) from \( L = 7, N = 49 \) to \( L = 40, N = 1600 \). Grains interact via purely repulsive, linear springs with force constant \( K \). For the systems driven by simple shear, we include a viscous damping force \(-B\dot{y}\) in the equations of motion for the top plate and \( N \) grains, where \( \dot{y} \) is the absolute velocity of the wall or a grain and \( B \) is the damping coefficient.

The equations of motion for simple shear, described in detail in Appendix A, are governed by three nondimensional parameters:

\[
\Gamma = \frac{B}{\sqrt{mpD^3}}, \quad \kappa = \frac{K}{pD^{d-1}}, \quad \Sigma = \frac{\tau}{p},
\]

where \( d \) is the spatial dimension. \( \Gamma \) is the dimensionless damping parameter, which we set equal to 5, and \( \kappa \) is a dimensionless grain stiffness. We set \( \kappa = 10^5 \), meaning that \( \phi \approx \phi_J(L) + 0.001 \), where \( \phi_J(L) \) is the jamming packing fraction at a given \( L \). Our results are insensitive to \( \kappa \) in this limit, which we verify for several values of \( \kappa > 200 \). We set \( \Gamma = 5 \), which maintains an inertial number \( I = \frac{\dot{\gamma}D\sqrt{D\rho g}}{p} < 10^{-4} \) (where \( \dot{\gamma} \) is the strain rate and \( \rho \) is the mass density of the grains) in the slow- or creep-flow limit, \( I < 10^{-3} [1,20] \).

We control force and not \( \dot{\gamma} \), so there are fluctuations in \( \Gamma \), but \( \Gamma = 5 \) keeps \( I < 10^{-4} \) even for \( \Sigma > \Sigma_c \). We have explicitly checked that our results are independent of \( \Gamma \) for several values of \( \Gamma \geq 3 \).

The schematic of the simulation procedure for (a) 3D simple shear and (b) 2D simple shear. In panels (a) and (b), MS packings are first created under only a fixed normal force per area \(-p\hat{y}\). We then apply a shear force per area \( \tau \hat{x} \) and search for an MS packing at a given \( \Sigma = \tau/p \) and length \( L \) in units of the small grain diameter \( D \). (c) A schematic of the simulation procedure for the river-bed-like geometry in 2D. MS packings are first created via sedimentation under gravity \(-g\hat{y}\), then driven by a fluid-like drag force in the \( x \) direction.

FIG. 1. Schematics of the simulation procedure for (a) 3D simple shear and (b) 2D simple shear. In panels (a) and (b), MS packings are first created under only a fixed normal force per area \(-p\hat{y}\). We then apply a shear force per area \( \tau \hat{x} \) and search for an MS packing at a given \( \Sigma = \tau/p \) and length \( L \) in units of the small grain diameter \( D \). (c) A schematic of the simulation procedure for the river-bed-like geometry in 2D. MS packings are first created via sedimentation under gravity \(-g\hat{y}\), then driven by a fluid-like drag force in the \( x \) direction.
bed surface, and \( v_i \) is the grain velocity. We find similar results for several different fluid flow profiles. The equations of motion, shown in Appendix A, are again governed by three dimensionless parameters

\[
\Gamma' = \frac{B/m}{\sqrt{g/D}},
\]
\[
\kappa' = \frac{K}{mg},
\]
\[
\Sigma' = \frac{B_0}{mg}.
\]

We again set \( \Gamma' = (B/m)(g/D)^{-1/2} = 5 \) and \( \kappa' = K/mg = 1000 \) and vary the dimensionless shear stress \( \Sigma' = B_0/mg \). Our results are again independent of \( \kappa' \) and \( \Gamma' \) in this regime. We prepare beds via sedimentation with \( L \) and vary the dimensionless shear stress \( \Sigma' = B_0/mg \).

**III. RESULTS**

**A. Critical scaling of \( \gamma_{ms} \)**

We begin with results for simple shear in 3D. We define the shear strain \( \gamma_{ms} \) as the total distance the top plate moves in the \( x \) direction divided by the average of the initial and final \( y \) positions of the top plate. In Fig. 2(a) and 2(b), we show the distribution \( P(\gamma_{ms}) \) for two illustrative values of \( \Sigma \) over a range of system sizes \( L \), obtained using 200 simulations for each \( L \). For small \( L \) above and below \( \Sigma_c \), the distributions are roughly exponential, \( P(\gamma_{ms}) \approx (\gamma_{ms})^{-1} \exp(-\gamma_{ms}/\langle \gamma_{ms} \rangle) \). This form indicates an underlying physical process resembling absorption [29], where objects propagate through space and each stops whenever it encounters an absorber. For absorption processes, the propagation distance distributions are exponential, as in Fig. 2, and the mean “travel distance” is inversely proportional to the density of absorbers. For sheared packings, the mean travel distance is \( \langle \gamma_{ms}(\Sigma, L) \rangle \). Thus, we use \( \langle \gamma_{ms}(\Sigma, L) \rangle^{-1} \) as a measure of the number density of MS packings.

In Fig. 2(c), we plot \( \langle \gamma_{ms} \rangle \) versus \( N = L^3 \) over a range of \( \Sigma \). Figure 2(d) shows that these data can be collapsed by plotting the scaled variables \( L^{-1}/|\Sigma - \Sigma_c|^\beta \) and \( (\gamma_{ms})^{-1}/|\Sigma - \Sigma_c|^\beta \). This collapse implies that finite-size effects for \( \langle \gamma_{ms}(\Sigma, L) \rangle^{-1} \) depend on a diverging correlation length \( \xi \propto |\Sigma - \Sigma_c|^{-\nu} \).

\[
\langle \gamma_{ms}(\Sigma, L) \rangle^{-1} = |\Sigma - \Sigma_c|^{\beta f_\pm} \left( L^{-1}/|\Sigma - \Sigma_c|^\beta \right). \tag{7}
\]

Here, \( f_\pm \) are the critical scaling functions for \( \Sigma > \Sigma_c \) and \( \Sigma < \Sigma_c \), respectively, which capture the finite-size effects. Note that all quantities in Eq. (7) are dimensionless. As shown in Appendix B, we determine the critical values by fitting the data to this functional form, where the critical values are fit parameters. We systematically exclude small system sizes and large deviations \( |\Sigma - \Sigma_c| \). We quantify the quality of the fits by using the reduced chi-squared metric, \( \chi^2 = \sum (\Delta_i)^2/e_i^2 \), where the sum is over all data points \( i \) used in the fit, \( \Delta_i \) is the difference between the data and the fit, and \( e_i \) is the standard deviation of the data at point \( i \). The best fits are obtained for \( \Sigma_c = 0.109 \) and \( \nu = 1.7 \) for all values of \( \beta / \nu \).

**FIG. 2.** Distributions \( P(\gamma_{ms}) \) of the strain \( \gamma_{ms} \) in 3D simple shear between the initial and final MS packings for (a) \( \Sigma < \Sigma_c \), (b) \( \Sigma > \Sigma_c \), and several \( L \). (c) The mean strain \( \langle \gamma_{ms} \rangle \) in 3D simple shear between initial and final MS packings plotted versus system size \( N = L^3 \) for several values of applied stress \( \Sigma \). Solid (dashed) lines correspond to \( |\Sigma - \Sigma_c|/\Sigma_c \) less (greater) than 0.5. Error bars are the standard error of the mean, given by the standard deviation within the sample divided by the square root of the number of trials. (d) The data from panel (c), plus additional data for more \( \Sigma \), collapses when using the scaled variables \( (\gamma_{ms})^{-1}/|\Sigma - \Sigma_c|^\beta \) and \( L^{-1}/|\Sigma - \Sigma_c|^\beta \) for different \( L \). The dashed line gives the large-system limit implied by Eq. (7) and the main plot in panel (d). The data in the inset are at constant \( L \), not constant \( L^{-1}/|\Sigma - \Sigma_c|^\beta \); see text and Appendix C for further discussion.
error in the mean (i.e., the standard deviation within that sample divided by the square root of the number of trials), represented as error bars in Fig. 2(c). We search for fits that satisfy $\chi^2/n \approx 1$ [16], where $n$ is the number of data points minus the number of fit parameters, and the critical values are independent of the range of $|\Sigma - \Sigma_c|$. From this analysis, shown in Appendix B, we estimate $\nu = 1.7 \pm 0.5, \Sigma_c = 0.109 \pm 0.005$, and $\beta/\nu = 0.57 \pm 0.07$. The uncertainty ranges represent the scatter in the fit results plus one standard deviation. Despite the uncertainty, $\nu \approx 1.7$ for yielding appears distinct from $\nu_j \approx 0.6 - 1$ for jamming [10,12,16], suggesting that these are two separate, although possibly related, zero-temperature transitions.

The inset in Fig. 2(d) shows $\gamma_{ms}$ plotted versus $\Sigma$ for different $L$, as well as the large-system limit (dashed, black line) implied by the scaling in the main panel of Fig. 2(d). For $\Sigma < \Sigma_c$, $f_-$ becomes constant at small $L^{-1}/|\Sigma - \Sigma_c|^{\nu}$ (i.e., $L > \xi$). This means that, in the large-system limit, $(\gamma_{ms})^{-1}$ vanishes nonanalytically at $\Sigma = \Sigma_c$ according to $(\gamma_{ms})^{-1} \propto |\Sigma - \Sigma_c|^{\beta}$. Also at small $L^{-1}/|\Sigma - \Sigma_c|^{\nu}$ for $\Sigma < \Sigma_c$, a peak develops in $P(\gamma_{ms})$ at $\gamma_{ms} > 0$, as shown in Fig. 2(a). We interpret this behavior as spatial decorrelation, where large systems behave like compositions of uncorrelated exponentially distributed random variables, yielding a distribution that is peaked at $\gamma_{ms} > 0$ with a mean that is independent of $L/\xi$. For $\Sigma > \Sigma_c$, $f_-$ is finite but tends to zero for small $L^{-1}/|\Sigma - \Sigma_c|^{\nu}$. This means that the number of MS packings vanishes for $\Sigma > \Sigma_c$ as $L/\xi$ increases. If $f_+$ approaches a vertical asymptote, MS packings do not exist for $\Sigma > \Sigma_c$ and finite $L/\xi$. Otherwise, MS packings only vanish for infinite $L/\xi$. Further studies with larger system sizes are required to address this specific point.

Note that the data we present in the inset to Fig. 2(d) approach the $L/\xi \rightarrow \infty$ limiting form (dashed curve) only for $\Sigma < \Sigma_c$ and for $\Sigma > \Sigma_c$, but not near $\Sigma_c$. The data does not collapse near $\Sigma_c$ because the scaled system size $L^{-1}/|\Sigma - \Sigma_c|^{\nu}$ changes significantly as $\Sigma$ is varied at fixed $L$. We show the data in the inset to Fig. 2(d) at constant $L^{-1}/|\Sigma - \Sigma_c|^{\nu}$ in Appendix C.

Figure 3 shows that the results for 2D systems with boundary driven simple shear are similar to those in 3D. Distributions for $P(\gamma_{ms})$ (not shown) are similar to the 3D case, which are shown in Figs. 2(a) and 2(b). In Fig. 3(a), we plot $\langle \gamma_{ms} \rangle$ versus $N = L^2$ for selected values of $\Sigma$. Figure 3(b) shows that these data (plus additional data) collapse by plotting the scaled variables $L^{-1}/|\Sigma - \Sigma_c|^{\nu}$ and $(\gamma_{ms})^{-1}/|\Sigma - \Sigma_c|^{\beta}$. Using a similar fitting analysis to that described above for 3D systems undergoing simple shear, we obtain $\nu = 1.84 \pm 0.3, \Sigma_c = 0.11 \pm 0.01$, and $\beta/\nu \approx 0.57 \pm 0.06$.

In Fig. 4, we display the results for the 2D river-bed-like geometry, which verifies that the scaling behavior is universal with respect to changes in the boundary conditions and driving method. Instead of shear strain, for each simulation we measure the average horizontal distance $\delta_{ms}$ traveled by a grain between initial ($\Sigma' = 0$) and final ($\Sigma' > 0$) MS packings. Figure 4(a) shows the ensemble-averaged values $\langle \delta_{ms} \rangle$ as a function of $\Sigma'$ and $L$. As before, these data collapse when plotted as a function of the scaled variables $L^{-1}/|\Sigma - \Sigma_c|^{\nu}$ and $(\delta_{ms})^{-1}/|\Sigma - \Sigma_c|^{\beta}$. By using a fitting analysis similar to the one discussed above for 3D boundary-driven simple shear, we identify $\Sigma_c = 0.41 \pm 0.015, \beta/\nu = 1.7 \pm 0.2$, and $\nu = 1.75 \pm 0.1$, suggesting that the scaling behavior and the value of $\nu \approx 1.7$–1.8 are generic with respect to changes in the spatial dimension, geometry, boundary conditions, and driving method. We discuss the fitting analysis for this geometry in Appendix B.

B. Microstructure of mechanically stable packings at varying $\Sigma$

To understand why the number density of MS packings vanishes at $\Sigma_c$, we quantify their structure by using the packing fraction $\phi_{ms}$ as well as the stress and contact fabric tensors. Figure 5(a) shows a plot of packing fraction $\phi_{ms}$ of MS
packings generated in 3D via simple shear as a function of $\Sigma$ for varying $L$. Each data point represents the ensemble average of 200 systems. $\phi_{ms}$ shows weak, nonmonotonic dependence on $\Sigma$, consistent with Fig. 10 in Ref. [9]. Specifically, $\phi_{ms}$ rises slightly (by about 0.1%) from $\Sigma = 0$ to $\Sigma = \Sigma_c$ and then decreases slightly for $\Sigma > \Sigma_c$. Figure 5(b) shows the same data plotted as a function of system size $L$. The different symbols represent different values of $\Sigma$, but these curves lie one top of another. As $L$ increases, $\phi_{ms}$ approaches $\phi_J \approx 0.643$, which is indicated by a dashed black line.

The data presented in Fig. 5(b) also allows us to estimate the critical length scale exponent $\nu_J$ for jamming. If we assume that there is a diverging length scale $\xi_j \sim (\phi_J - \phi_{ms})^{-1/\nu}$ related to jamming that controls the system-size dependence in Fig. 5, we expect that $L/\xi_j$ should be a constant and the packing fraction deviation scales as $(\phi_J - \phi_{ms}) \sim L^{-1/\nu}$. The inset to Fig. 5(a) shows that $\nu_J \approx 1/1.2 \approx 0.8$. This result is in agreement with previous studies [10,12,16], which have estimated $\nu_J$ to be between 0.6 and 1. We again note that this value for $\nu_J$ is distinct from $\nu \approx 1.7 - 1.8$ that we estimate for yielding.

The stress and contact fabric tensors [30,31] are given by

$$
\sigma_{\alpha\lambda} = \frac{1}{V} \sum_{i \neq j} r_{ij}^{\alpha} r_{ij}^{\lambda},
$$

(8)

$$
R_{\alpha\lambda} = \frac{1}{N} \sum_{i \neq j} \frac{r_{ij}^{\alpha} r_{ij}^{\lambda}}{|r_{ij}|^2}.
$$

(9)
Here, $\alpha$ and $\lambda$ are Cartesian coordinates, $V$ is the system volume, $r_{ij}^{1/2}$ is the $\alpha$ component of the center-to-center separation vector between grains $i$ and $j$, and $F_{ij}^{\alpha}$ is the $\lambda$ component of the intergrain contact force. The sum over $i$ and $j$ includes all pairs of contacting grains (excluding grain-wall contacts).

Force balance requires $\sigma_{xy} = \sigma_{xy} = -\tau$, $\sigma_{yz} = p$, and $\sigma_{xz} = \sigma_{xz} = 0$. In Fig. 6(b), we show ensemble averages of $\sigma_{xy}/\sigma_{yy}$ as a function of $\Sigma = \tau/p$. The data follows a linear relation with a slope of negative one, confirming that force balance is satisfied. We also find that $\sigma_{yz} = \sigma_{yz} = 0$ and $\sigma_{xz} = \sigma_{xz} = 0$ (not shown). Figure 6(f) shows that the force balance criterion $\sigma_{xy}/\sigma_{yy} = -\Sigma$ requires a proportional change in the corresponding fabric tensor component, $R_{xy}/R_{yy} = -\alpha\Sigma$ with $\alpha \approx 0.4$. Results for 2D simple shear (not shown) are identical: we find $\sigma_{xy}/\sigma_{yy} = -\Sigma$ and $R_{xy}/R_{yy} = -\alpha\Sigma$, but with $\alpha \approx 0.33$. Thus, since MS packings at increasing $\Sigma$ require grain-grain contacts to be increasingly oriented along the compressive direction, the vanishing density of MS packings likely results from an upper limit of the stress and corresponding fabric anisotropies that can be realized in a large system.

Finally, we show in Figs. 6(c), 6(d), 6(g), and 6(h) the excess normal stresses $\sigma_{xx}/\sigma_{yy} \equiv \lambda_x$, and $\sigma_{zz}/\sigma_{yy} \equiv \lambda_z$, as well as the corresponding quantities from the fabric tensor $R_{xx}/R_{yy} \equiv \rho_x$ and $R_{zz}/R_{yy} \equiv \rho_z$. These quantities represent excess compressive stresses and contacts that exist in the periodic $x$ and $z$ directions. For $\Sigma < \Sigma_c$, $\lambda_{x,z}$ and $\rho_{x,z}$ begin at some finite value for small $L$ and tend to zero (or just below, likely due to the uniaxial compression by which we prepare our initial states) at large $L$. For $\Sigma > \Sigma_c$, $\lambda_{x,z}$ and $\rho_{x,z}$ increase with $\Sigma$. We find similar results for 2D simple shear (not shown).

To understand why the normal stress and fabric anisotropies increase with $\Sigma$, we consider the ensemble-averaged stress tensor $\langle \sigma \rangle$ of MS packings in 3D at a given $\Sigma$, which can be written as

$$\langle \sigma \rangle = p \begin{bmatrix} 1 + \lambda_x & -\Sigma & 0 \\ -\Sigma & 1 & 0 \\ 0 & 0 & 1 + \lambda_z \end{bmatrix}. \quad (10)$$

We consider only the stress components in the $x$-$y$ plane, which are decoupled from $z$ in Eq. (10), and its eigenvalue-eigenvector pairs $\{\sigma_1, \lambda_1\}$ and $\{\sigma_2, \lambda_2\}$. The internal stress anisotropy is $\Sigma_i = \tau_i/p_i$, where $p_i = (\sigma_1 + \sigma_2)/2$ and $\tau_i = (\sigma_1 - \sigma_2)/2$ are the internal pressure and shear stress, respectively. From Eq. (10), $\Sigma_i = (2\Sigma^2 + \lambda_1^2)/(2 + \lambda_1)$, and $\sigma_1$ is oriented at an angle that deviates from the compression direction by an angle $\theta^*$, as shown in Fig. 6(a). An expansion of $\Sigma_i$ for small $\lambda_1$ gives $\theta^* = \sqrt{2\lambda_1} + O((\lambda_1)^3)$.

Thus, when $\lambda_1 = 0$, $\Sigma_i = \Sigma$ and $\sigma_1$ and $\sigma_2$ are aligned with the compression and dilation directions, respectively. However, $\Sigma_i$ is minimized by a positive, nonzero value of $\lambda_1 = 2\Sigma^2$ with $\Sigma_{i_{\text{min}}} = (1 + \Sigma^2)^{-1/2}$. This can give $\Sigma_i < \Sigma$, but this rotates the larger eigenvector $\sigma_1$ away from the compression direction.

FIG. 6. Close-up of MS packings in 2D illustrating features of the (a) stress and (e) fabric tensors for the central grain, $\sigma_1$, $\sigma_2$ and $R_{ij}$, $R_i$ denote the eigenvalue-eigenvector pairs from the sum over the contacts (blue and green arrows) for the center grain $i$ [Eqs. (8) and (9)]. The magnitudes of the arrows are proportional to the eigenvalues and the directions are along the eigenvectors. $\theta^*$ is the angle between the larger eigenvector and the compressive direction. (b), (f) Ensemble averages of $\sigma_{xy}/\sigma_{yy}$ and $R_{xy}/R_{yy}$ for MS packings are plotted versus $\Sigma$ for varying $L$, showing $\sigma_{xy}/\sigma_{yy} \approx -\Sigma$ and $R_{xy}/R_{yy} \approx -0.4\Sigma$ for all $L$. Ensemble averages of the normal anisotropies in the $x$ direction in the (c) stress tensor, $\sigma_{xx}/\sigma_{yy} \equiv \lambda_x$, and (g) fabric tensor, $R_{xx}/R_{yy} \equiv \rho_x$, are plotted versus $\Sigma$ for varying $L$. Ensemble averages of the normal anisotropies in the $z$ direction in the (d) stress tensor, $\sigma_{zz}/\sigma_{yy} \equiv \lambda_z$, and (h) fabric tensor, $R_{zz}/R_{yy} \equiv \rho_z$, are plotted versus $\Sigma$ for varying $L$. 

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Near $\Sigma_c$, for finite systems, MS packings are scarce, and $\Sigma_i < \Sigma$ with $\theta^i > 0$ may be preferable, despite the broken symmetry. However, the broken symmetry becomes more difficult to achieve for larger systems. We note that the dependence of $\lambda_z, \kappa_z, \rho_z$, and $\rho_c$ on $L$ in Figs. 6(c), 6(d) and 6(g), 6(h) is suggestive of critical scaling (which we expect if $\xi$ dominates the behavior near $\Sigma_c$) similar to Eq. (7). The scaling results for these quantities are not as conclusive, and we leave a more extensive study of the possible scaling of these quantities for future work.

IV. CONCLUSION

In conclusion, for frictionless spherical grains under shear, we find that the number of MS packings vanishes near $\Sigma = \Sigma_c \approx 0.11$. Finite-size effects depend on a diverging length scale $\xi \propto |\Sigma - \Sigma_c|^{-v}$. We find similar results for the cases of 3D simple shear, shown in Fig. 2, for 2D simple shear, shown in Fig. 3, and in a 2D river-bed-like geometry, shown in Fig. 4. Thus, the critical scaling behavior, including the value of the exponent $v \approx 1.7 \sim 1.8$, is generic with respect to changes in spatial dimension, system geometry, and boundary conditions.

We find that the packing fraction of MS packings at varying $\Sigma$ shows weak, nonmonotonic dependence on $\Sigma$, in agreement with previous work [9]. This suggests that the critical scaling we observe is distinct from that associated with jamming. The force balance criterion, Fig. 6(b), is accompanied by a proportional change in the fabric tensor, Fig. 6(e). Thus, we argue that $\Sigma_c$, corresponds to the maximum anisotropy that can be realized in the large-system limit. This hypothesis is consistent with our finding that finite-sized MS packings with $\Sigma$ near or above $\Sigma_c$ tend to be rotated relative to the axes of the applied deformation, which can reduce the internal force anisotropy of MS packings. However, this effect appears to vanish in the large-system limit, where symmetry dictates that compressive direction be aligned with the largest eigenvalues of the stress and fabric tensors for MS packings.

Finally, we note recent work on jamming by shear [29–32], where MS packings obtained via simple or pure shear at constant volume also display anisotropic stress and contact fabric tensors. These results are distinct from those presented here, since we control normal stress and allow volume to fluctuate. However, we expect future work to unify these two approaches, providing a complete theory of the density of MS packings as a function of volume, stress state, preparation history, and friction.

ACKNOWLEDGMENTS

This research was sponsored by the Army Research Laboratory under Grants No. W911NF-14-1-0005 and No. W911NF-17-1-0164 (A.H.C., N.T.O., and C.S.O.). The views and conclusions contained in this document are those of the authors and should not be interpreted as representing the official policies, either expressed or implied, of the Army Research Laboratory or the U.S. Government. The U.S. Government is authorized to reproduce and distribute reprints for Government purposes notwithstanding any copyright notation herein. M.D.S. also acknowledges support from the National Science Foundation Grant No. CMMI-1463455.

APPENDIX A: EQUATIONS OF MOTION

1. Boundary-driven simple shear in two and three dimensions

For the simple shear simulations in $d = 2$ (2D) and $d = 3$ (3D) spatial dimensions, we solve Newton’s equations of motion for all bulk grains as well as the top plate. The equation of motion for the top plate is

$$M \ddot{a}_{plate} = \sum_{i,j} F^c_{ij} - B \dot{a}_{plate} v_{plate}, \quad (A1)$$

where $M$ is the plate mass, $a_{plate}$ is the plate acceleration, $F^c_{ij}$ is the contact force on plate particle $i$ due to bulk grain $j$, $F_{ext}$ is the external force exerted on the top wall, $B_{plate}$ is a viscous drag coefficient, and $v_{plate}$ is the top plate velocity. Similarly, the equation of motion for each bulk grain $i$ is given by

$$m_i \ddot{a}_i = \sum_j F^c_{ij} - B_i \dot{a}_i v_i, \quad (A2)$$

where $m_i \propto (D_i)^{\nu}$ is the mass of grain $i$, $D_i$ is the diameter of grain $i$, $a_i$ is the acceleration of bulk grain $i$, $F^c_{ij}$ is the contact force on bulk grain $i$ due to bulk grain $j$, $B_i$ is the drag coefficient, and $v_i$ is the velocity of bulk grain $i$. For pairwise contact forces between two bulk grains or between a bulk grain and a plate particle, we use

$$F_{ij} = K \left( \frac{D_{ij}}{r_{ij}} - 1 \right) \Theta \left( 1 - \frac{r_{ij}}{D_{ij}} \right) \hat{r}_{ij}, \quad (A3)$$

where $K$ is a force scale, $r_{ij}$ is the center-to-center distance between grains $i$ and $j$, $D_{ij}$ is the average diameter of grains $i$ and $j$, $\Theta$ is the Heaviside step function, and $\hat{r}_{ij}$ is the unit vector from the center of grain $i$ to the center of grain $j$.

The external force on the wall is given by

$$F_{ext} = (\tau \hat{x} - p \hat{y}) (LD)^{d-1}, \quad (A4)$$

where $\tau$ and $p$ are the shear stress and normal stress, respectively. Each plate particle has the same mass $m$ and drag coefficient $B$ as the small bulk grains, and thus $M = m L^{d-1}$ and $B_{wall} = B L^{d-1}$. Since the number of contacts also scales as $L^{d-1}$, all quantities in Eq. (A1) scale as $L^{d-1}$. Equations (A1) and (A2) are then governed by the nondimensional parameters given in Eq. (3).

2. River-bed model in two dimensions

The river-bed-like geometry that we study consists of a 2D domain of width $L$, with periodic boundary conditions horizontally, containing $N/2$ large and $N/2$ small disk-shaped grains with diameter ratio $r = 1.4$. We use $N = 5L$ grains, so the beds have a height $H \approx 5D$, where $D$ is the diameter of a small grain. There is no upper boundary, and the lower boundary is rigid with a no-slip condition for any grain contacting it. The net force on each grain is given by the sum of contact forces from all other grains, a gravitational force, and a Stokes-drag-like force from a fluid flow that increases linearly with height and moves purely horizontally:

$$m_i \ddot{a}_i = \sum_j F^c_{ij} - m_i g \hat{y} + B_i \left( \frac{v_i}{H} - \hat{y} \right), \quad (A5)$$

Here, $m_i \propto D_i^2$ is the grain mass, $v_i$ and $a_i$ are the velocity and acceleration, respectively, of each grain $i$. $m_i g \hat{y}$ is the gravitational force on grain $i$.
The scaling function in Eq. (B1) for (a) 3D simple shear and (b) 2D river-bed geometries.

APPENDIX B: DETERMINING CRITICAL VALUES

As discussed above, we determine the critical exponents $\nu$ and $\beta$ as well as the value of the yield stress $\Sigma_c$ by fitting scaled data to a scaling function with the values of $\nu$, $\beta$, and $\Sigma_c$ treated as fit parameters. We first estimate $\nu$, $\beta$, and $\Sigma_c$ by collapsing the data according to

$$\langle \gamma_{nm}^{-1} \rangle^1 = L^{-\beta/\nu} g((\Sigma - \Sigma_c)L^{-1/\nu}).$$  \hspace{1cm} (B1)

as shown in Fig. 7. This form is equivalent to Eq. (1) in the main text but is more convenient to use since the scaling function $g$ has only one branch. We fit $\log(\langle \gamma_{nm}^{-1} \rangle^1 L^{\beta/\nu})$ and $(\Sigma - \Sigma_c)L^{-1/\nu}$ to a third-order polynomial. The polynomial coefficients returned from this fit are then used as the initial values in a Levenberg–Marquardt fit to the scaling form in Eq. (1) in the main text, where the critical values $\Sigma_c$, $\nu$, and $\beta$ are then used as fit parameters.

From Fig. 7, it is obvious that the data for large deviations $|\Sigma - \Sigma_c|$ do not collapse as well as the data for small deviations. In addition, we expect that data for small system sizes does not obey the scaling collapse. Thus, we systematically vary the range $X \equiv |\Sigma - \Sigma_c|/\Sigma_c < X_{\text{max}}$ and the minimum system size $L_{\text{min}}$ that we include in our fits, although we are somewhat limited in the maximum $L_{\text{min}}$ we can use before we no longer have enough data for a meaningful fit. We quantify the fits by using the reduced chi-squared metric, $\chi^2/n = \sum_i (\Delta_i)^2/\sigma_i^2$, where the sum is over all data points $i$ used in the fit (a subset of those shown in Fig. 7), $\Delta_i$ is the difference between the data and the fit, and $\sigma_i$ is the standard error of the mean, which we estimate by the standard deviation within that sample divided by the square root of the number of trials. We then measure $\chi^2/n$, where $n$ is the number of data points minus the number of fit parameters in the model. A good fit is characterized by a value of $\chi^2/n \approx 1$.

Figure 8 shows the critical values that yield the best fits for 3D simple shear as a function of $L_{\text{min}}$ for several $\Sigma_c$. Error bars represent one standard deviation. The $\chi^2/n$ values for each fit are plotted in panel (d).

For $X_{\text{max}} = 0.3$, we find $\chi^2/n > 2.5$, signifying a poor fit. For $0.1 \leq X_{\text{max}} \leq 0.2$, we find $\chi^2/n \approx 1.5$, nearly independent of $L_{\text{min}}$. We estimate $\nu = 1.7 \pm 0.5$, $\Sigma_c = 0.109 \pm 0.005$, and $\beta/\nu = 0.57 \pm 0.07$ by the scatter in results for $X_{\text{min}}$. From $X_{\text{max}} = 0.3$, we find $\chi^2/n > 2.5$, signifying a poor fit.
0.1 \leq X_{\text{max}} \leq 0.2$, plus the typical width of the error bars, which represent one standard deviation in the Levenberg–Marquardt fit. We note significant uncertainty in the value of $v$, which agrees with our observation that good scaling collapses are possible with $v$ from 1.2 to 2.2 for the 3D simple shear data.

Figure 9 shows the critical values that yield the best fits for the river-bed-like geometry as a function of $X_{\text{max}}$. As we reduce $X_{\text{max}}$, $\chi^2/n$ steadily decreases to $\approx 1$. However, the critical values are almost independent of $X_{\text{max}}$. We estimate $\Sigma_c' = 0.41 \pm 0.015$, $\beta'/v = 1.7 \pm 0.2$, and $v = 1.75 \pm 0.1$. A similar analysis for 2D simple shear (not shown) yields $v = 1.84 \pm 0.3$, $\Sigma_c = 0.11 \pm 0.01$, and $\beta/v \approx 0.57 \pm 0.06$. The method we describe for obtaining the critical values gives similar results to a brute force search through the parameter space, where we seek a global minimum in $\chi^2/n$.

APPENDIX C: SCALING COLLAPSE OF $(\gamma_{\text{ms}})^{-1}$ VERSUS $\Sigma$

In Fig. 2(d), we showed that the data for $(\gamma_{\text{ms}}(\Sigma, L))^{-1}$ collapsed onto two branches when plotted as a function of the scaled system size $\tilde{L}^{-1} \equiv L^{-1}/|\Sigma - \Sigma_c|^\beta$. The inset to Fig. 2(d) showed a plot of $(\gamma_{\text{ms}}(\Sigma, L))^{-1}$ versus $\Sigma$ for various values of $L$. We also included a curve showing the infinite-system limit for $(\gamma_{\text{ms}}(\Sigma, L))^{-1}$ versus $\Sigma$, which is implied by the scaling in Eq. (7).

In Fig. 10, we show a similar plot, but with data plotted at constant $\tilde{L}^{-1} \equiv L^{-1}/|\Sigma - \Sigma_c|^\beta$. If $\tilde{L}^{-1}$ is held fixed, Eq. (7) reduces to $(\gamma_{\text{ms}})^{-1} = A|\Sigma - \Sigma_c|^\beta$, where $A$ is a constant, $A = f_0(\tilde{L}^{-1})$. The solid curves shown in Fig. 10 are $(\gamma_{\text{ms}}(\Sigma, L))^{-1} = A|\Sigma - \Sigma_c|^\beta$, where the particular value of $A$ for each value of $\tilde{L}^{-1}$ at $\Sigma > \Sigma_c$ and $\Sigma < \Sigma_c$ is determined from the scaling plot in Fig. 2(d). The solid curves pass through the data, reaffirming the scaling behavior in Eq. (7). As $\tilde{L}^{-1}$ approaches zero, these curves approach the infinite-system limit shown by the dashed line.


FIG. 10. A plot of $(\gamma_{\text{ms}}(\Sigma, L))^{-1}$ versus $\Sigma$ for different values of $\tilde{L}^{-1} = L^{-1}/|\Sigma - \Sigma_c|^\beta$. Data shown are within 10% of the relevant value of $\tilde{L}^{-1}$ (e.g., data for $\tilde{L}^{-1} = 100$ include $90 < \tilde{L}^{-1} < 110$). The solid curves are $(\gamma_{\text{ms}}(\Sigma, L))^{-1} = A|\Sigma - \Sigma_c|^\beta$, where $A$ is equal to the value of $f_0$ (for $\Sigma > \Sigma_c$) or $f_1$ (for $\Sigma < \Sigma_c$) at that particular value of $\tilde{L}^{-1}$. The values of $A$ can be determined from the scaling collapse in Fig. 2(d).