

The Statistical Physics of Athermal Materials

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Abstract

At the core of equilibrium statistical mechanics lies the notion of statistical ensembles: a collection of microstates, each occurring with a given a priori probability that depends on only a few macroscopic parameters, such as temperature, pressure, volume, and energy. In this review, we discuss recent advances in establishing statistical ensembles for athermal materials. The broad class of granular and particulate materials is immune to the effects of thermal fluctuations because the constituents are macroscopic. In addition, interactions between grains are frictional and dissipative, which invalidates the fundamental postulates of equilibrium statistical mechanics. However, granular materials exhibit distributions of microscopic quantities that are reproducible and often depend on only a few macroscopic parameters. We explore the history of statistical ensemble ideas in the context of granular materials, clarify the nature of such ensembles and their foundational principles, highlight advances in testing key ideas, and discuss applications of ensembles to analyze the collective behavior of granular materials.

1. INTRODUCTION

Athermal, cohesionless particulate materials, such as dry grains and dense, non-Brownian suspensions, exhibit features that are generic to many nonergodic, disordered (glassy) systems (1, 2). However, these granular materials are nonergodic in the extreme sense, as they stay in a single configuration unless driven externally. Exploration of configuration space is completely controlled by the driving protocol, and because cohesive forces are absent, fluid or solid-like behavior emerges purely as a response to external driving (3, 4). Thus, these systems are the epitome of athermal condensed matter. The objective of this article is to describe recent developments in constructing statistical ensembles for such athermal systems. We examine the history of statistical ensembles for granular materials, which began with the ideas of Sam Edwards (5), describe recent developments in theory and experiments, and explore future applications.

Although thermal fluctuations are absent, granular systems nonetheless exhibit large fluctuations (6). This is exemplified by the exponential distribution of contact forces in jammed (solid-like) granular assemblies (7, 8) and by large, intermittent stress fluctuations in quasistatic flows (9). The ubiquitous presence of fluctuations with well-defined distributions that seem to depend on a handful of macroscopic parameters (6) suggests that statistical ensembles could prove to be an important tool for predicting the emergent properties of granular materials. There are two aspects that need to be elucidated: (a) the establishment of the ensemble and (b) how to use the ensemble to calculate and predict the emergent behavior of granular materials. After introducing the history of athermal ensembles in Section 1, we discuss the special role of constraints in the enforcement of mechanical equilibrium and how this differs from thermal ensembles. In Section 2, we review the existing experimental and numerical tests of the ensembles, and in Section 3, we describe how to use the distributions of microstates in an ensemble to calculate collective properties. These same techniques can be extended to systems with slow dynamics, and examples of this are provided in Section 4. Finally, we close with a summary of open questions in the field (Section 5).

1.1. History

Twenty-five years ago, Sam Edwards and coworkers (5) considered the problem of what sets the packing density of granular materials under simple operations such as shaking, stirring, or compression. They proposed a statistical ensemble approach, which was formulated along lines paralleling that of equilibrium statistical mechanics and thermodynamics. In equilibrium statistical mechanics, conservation of energy (First Law) implies that the total energy E of an isolated system cannot fluctuate. The set of microstates (we designate a microstate as ν) with a given total energy E , calculated from the positions and momenta of all particles, form the microcanonical ensemble. The macroscopic, extensive variable E characterizes the ensemble, and all microstates with this energy are assumed to occur with equal probability. Different physical systems are distinguished by the density of states $\Omega_B(E) = \sum_{\nu} \delta(E - E_{\nu})$. Isolated systems are important only for establishing the foundations of thermodynamics, and the canonical ensemble describing a system in contact with a heat bath is characterized by an intensive macroscopic variable: the temperature $\frac{1}{T} = \frac{\partial S(E)}{\partial E}$, where $S(E)$ (Table 1) is the entropy of the isolated system. It is useful to summarize the postulates used to arrive at the framework described above. In addition to energy conservation, which is dynamical in origin, the fundamental assumption underlying statistical mechanics is the equiprobability of microstates with a given energy, which also follows from the maximization of entropy (10).

Edwards and collaborators set about formulating an analogous theory of granular materials. Because of frictional forces and dissipation, energy is not conserved in granular systems; it is therefore not an appropriate state variable. There are two aspects to the Edwards formulation.

Table 1 Similar quantities from ordinary statistical mechanics and the original and generalized Edwards ensemble. In practice, λ is often taken to have a value of unity with units of 1/volume (1/stress for the generalized ensemble), for simplicity. Note that the angoricity, unlike the other temperatures, is defined without a reciprocal.

	Boltzmann	Edwards	Generalized Edwards
Conserved quantity	Energy, E	Volume, V	Force-moment tensor, $\hat{\Sigma}$
Number of valid configurations	$\Omega_B(E)$	$\Omega(V)$	$\Omega(\hat{\Sigma})$
Entropy	$S = k_B \ln \Omega_B$	$S = \lambda \ln \Omega$	$S = \lambda \ln \Omega$
Equilibrating quantity	Temperature	Compactivity	Angoricity tensor
	$\frac{1}{T} = \frac{\partial S(E)}{\partial E}$	$\frac{1}{X} = \frac{\partial S(V)}{\partial V}$	$\alpha_{\mu\lambda} = \frac{\partial S(\Sigma_{\mu\lambda})}{\partial \Sigma_{\mu\lambda}}$
Distribution	$e^{-\frac{E}{k_B T}}$	$e^{-\frac{V}{X}}$	$e^{-\hat{\alpha}:\hat{\Sigma}}$

One that is often forgotten is the assertion that the dynamics of slowly driven granular systems is controlled by the statistical properties of blocked states (11). For infinitely rigid grains, these blocked microstates are composed of grains in mechanical equilibrium such that grains cannot be moved without causing a finite overlap. In the modern parlance, these would be called jammed states (12, 13). Motivated by the central role that the packing density plays in determining the nature of jammed states, Edwards proposed to replace the Hamiltonian or energy function by a volume function and defined the analogous density of states: $\Omega(V) = \sum_{\nu} \delta(V - V_{\nu})$, where ν is now restricted to jammed states and is defined by just the positions of grains. Following the steps of the development of thermodynamics, the Edwards entropy is then defined as $S \equiv \lambda \ln \Omega(V)$, which yields a canonical ensemble characterized by a corresponding temperature-like variable X :

$$\frac{1}{X} = \frac{\partial S(V)}{\partial V}, \tag{1}$$

which Edwards named compactivity. Just as the Boltzmann constant k_B gives the ordinary entropy the correct units of J/K, the constant λ sets the units of compactivity relative to volume. Similarly, one can write quantities analogous to the specific heat, the Boltzmann distribution, etc. A summary of these ordinary statistical mechanical quantities, and their translations into the Edwards ensemble, is given in **Table 1**. Although the Edwards ensemble looks exactly like the ensembles of equilibrium statistical mechanics, it is different in one very important way: It is constrained to consider only those states that are blocked or jammed. Any calculation involving sums over states such as a partition function has to explicitly impose this constraint.

The name compactivity was chosen because it describes how far a system is from its densest possible state, beyond which no further compaction is possible. The extremal values of X are chosen to give a sensible physical interpretation of the limits of validity of the ensemble. The lowest possible compactivity ($X = 0$) has come to be identified with random close packing, the densest packing for which no crystalline order is present (14–16). The upper limit, at $X = \infty$, is typically identified with random loose packing (17, 18), the loosest packing for which mechanical stability is still present. Beyond these values, the ensemble is no longer defined.

Soon after the introduction of the Edwards ensemble, it was realized that contact forces needed to be incorporated into the definition of jammed microstates. Each grain must locally satisfy force and torque balance, and in the presence of friction these forces cannot be deduced just from positions of grains. There was also a heightened awareness (7, 9) of the significance of earlier

observations (19, 20) of what have come to be known as force chains. These are roughly colinear chains of particles through which larger-than-average stresses are transmitted. These considerations led to the proposal of a generalized ensemble (21–23), with stress as an additional governing state variable; the corresponding temperature-like variable was named angoricity by Edwards, from the modern Greek word $\acute{\alpha}\gamma\chi\omicron\varsigma$, meaning stress. There is a growing consensus that the force-moment tensor $\hat{\Sigma}$ is the extensive variable most closely analogous to energy in equilibrium statistical mechanics:

$$\hat{\Sigma} = \sum_{m,n} \mathbf{d}_{mn} \mathbf{f}_{mn}, \quad 2.$$

where \mathbf{d}_{mn} is the vector pointing from the center of particle m to the interparticle contact with its neighbor particle n , \mathbf{f}_{mn} is the interparticle contact force, and the sum is over all pairs of grains. These quantities are illustrated in **Figure 1**, along with sample images illustrating how they are measured in experiments. Unlike the volume V , $\hat{\Sigma}$ depends on both the positions of grains and their interparticle contact forces. Because the force moment is a tensorial quantity, the angoricity is a tensor as well:

$$\alpha_{\mu\lambda} \equiv \frac{\partial S(\hat{\Sigma})}{\partial \Sigma_{\mu\lambda}}. \quad 3.$$

As the μ, λ component of the force-moment tensor goes to zero (corresponding to a loss of rigidity), we expect the corresponding angoricity to tend to infinity. Since the original proposal, a number of different formulations for the stress ensemble have been published (22–27).

1.2. Athermal Versus Thermal Ensembles

The volume and stress ensembles described in the previous section have come to be collectively known as generalized Edwards ensembles (28) and are distinct from ordinary statistical mechanical ensembles in important ways. First, they are completely nonergodic. Because there is no temporal evolution, there is no equivalence of time averages and ensemble averages. This peculiarity arises because granular materials are athermal: At room temperature, the thermal energy

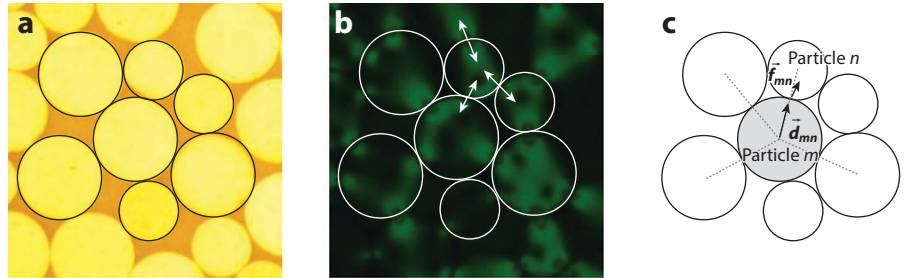


Figure 1

Image from experiments on 2D photoelastic disks. (a) Particles in unpolarized and (b) polarized light. From panel b, it is possible to extract vector contact force $\mathbf{f}_{mn} = -\mathbf{f}_{nm}$ at each interparticle contact, shown schematically as white arrows. (c) Schematic of calculation of force-moment tensor $\hat{\Sigma}$ according to Equation 2, using \mathbf{f}_{mn} and the vector \mathbf{d}_{mn} pointing from the center of particle m to the interparticle contact with particle n . Dashed lines connect the center of particle m with the center of each particle it is in contact with. Adapted from Reference 69.

is many orders of magnitude lower than required to move one grain over another against the force of gravity. For example, a millimeter-sized grain of sand would require $mgd \approx 10^{-5}$ J to move upward by a distance equivalent to its own diameter against the force of gravity. At room temperature (300 K), this corresponds to $\approx 10^{15} k_B T$ and therefore thermal motion is statistically negligible. As such, all granular systems require external driving to explore new configurations, and the Edwards ensemble (or its variants) can be interpreted only as an ensemble of microstates created by independent realizations with certain macroscopic parameters held fixed. Another type of granular ensemble is that of subsystems within a single thermodynamically large system. In this context, thermalization refers to the establishment of equal angoricity and compactivity in these different subsystems. This granular thermalization is distinct from the establishment of effective temperatures in glassy systems (29, 30), where the effective temperatures encode the fluctuations of the slowly relaxing modes and are dynamical in nature.

Although the assumption of equiprobability of all valid microstates is well established for equilibrium statistical mechanics, the situation is less clear for the Edwards ensembles. In fact, tests of the hypothesis for jammed states in experiments, simulations, and exactly solvable models reveal that equiprobability is not universally valid but does hold under some conditions (31–36). In particular, Gao et al. (37) used closely matched simulations and experiments to generate millions of statistically independent packings of seven disk-shaped particles. By enumerating all unique configurations, they observed that the same finite number of microstates appeared in both simulations and experiments, and that the relative frequencies of these microstates were highly nonuniform (they differ by factors of up to a million). However, equiprobability is not essential for the definition of either compactivity (38) or angoricity (25). Furthermore, it has recently been shown (39) that the Edwards entropy can satisfy extensivity even without equiprobability and that a factor of $1/N!$ is necessary to avoid an athermal version of the Gibbs paradox.

If configurations are not equiprobable, one can generalize the definition of the density of states in Table 1, for example $\Omega(V)$, to include the weights of microstates: $\Omega(V) = \sum_v \omega_v \delta(V - V_v)$. The much weaker condition of factorability, $\Omega(V_1 + V_2) = \Omega(V_1)\Omega(V_2)$, is necessary and sufficient for defining angoricity and compactivity (40–42). An important point to note is that, unlike $\Omega_B(E)$, $\Omega(V)$ is not determined by any Hamiltonian. A pragmatic perspective is to assert that different protocols lead to different microstate weights and therefore to different $\Omega(V)$. The ensemble approach is applicable, but each protocol would be like choosing a new Hamiltonian that defines the density of states.

1.3. Conservation Principles

The objective of a statistical ensemble framework is to predict the probability of occurrence of a microscopic configuration, given a set of macroscopic constraints. In equilibrium statistical mechanics, there are natural macroscopic constraints that emerge from conservation laws: energy, volume, and number of particles. In jammed, athermal systems, there are similarly a natural set of macroscopic variables and their associated conservation principles.

Any microscopic jammed state of N grains is completely specified by the positions of the grains and the forces (including frictional forces) at intergrain contacts: $\nu = \{\mathbf{r}_m, \mathbf{f}_{mn}\}$, where \mathbf{r}_m is the position of the m th grain and \mathbf{f}_{mn} is the contact force between grains m and n . As in thermal systems, and in other jammed or glassy systems, a granular packing can exist in many different microscopic states for a fixed set of macroscopic variables such as volume or applied stress. In addition, there is a microscopic indeterminacy (1) arising from the Coulomb condition for static friction, which imposes an inequality constraint: $|\mathbf{f}_T| \leq \mu |\mathbf{f}_N|$ such that the magnitude of the tangential force f_T cannot exceed the static friction coefficient μ times the magnitude of the

normal force f_N . Thus, there is a range of tangential forces that are allowed for any microscopic grain configuration. What, then, is a natural microcanonical ensemble for jammed states?

To address this question, we start by examining the constraints that must be satisfied by jammed (blocked) microstates. For infinitely rigid grains, there is a strict constraint on the $\{\mathbf{r}_m\}$ because grains cannot overlap. The jammed states in the original Edwards ensemble are then all of the just-touching configurations, and the volume occupied by N grains is a macroscopic invariant: It is globally conserved. For grains that are not infinitely rigid, the nonoverlapping constraint vanishes, and it is no longer clear what volume is conserved. Of course, one can fix the total volume V occupied by N grains through the nominal packing fraction, but this is not a consequence of any constraints on jammed states. The constraints that remain are those of mechanical equilibrium, which involve the contact forces $\{f_{mm}\}$. These constraints lead to a robust conservation law involving the force-moment tensor (24, 43, 44) of the granular packing, $\hat{\Sigma}$, defined in Equation 2, and related to the Cauchy stress tensor $\hat{\sigma}$ (45):

$$\hat{\Sigma} = V\hat{\sigma}. \quad 4.$$

The conservation of the force-moment tensor arises purely from local constraints of force and torque balance. It holds for any material that is in mechanical equilibrium, regardless of dimensionality and microscopic details. Here, we demonstrate this conservation principle in a generic two-dimensional (2D) $T=0$ solid using a continuum stress field, but a formulation that explicitly incorporates the discreteness of the granular assembly can also be constructed (25, 46).

We consider a 2D $T=0$ solid under periodic boundary conditions (PBCs). Mechanical equilibrium requires that the continuum stress field $\hat{\sigma}(\mathbf{r})$ obeys force and torque balance everywhere. In the absence of body forces, the continuum stress field satisfies

$$\nabla \cdot \hat{\sigma}(\mathbf{r}) = 0 \quad 5.$$

due to force balance. Local torque balance also requires $\sigma_{\mu\lambda} = \sigma_{\lambda\mu}$. There is an additional constraint of $\sigma_{\mu\mu} > 0$ if the system of interest contains noncohesive interactions (e.g., dry grains), for which tensile forces are not possible.

For a 2D system under PBCs (equivalent to the surface of a torus), there are three classes of topologically distinct loops, labeled A , B , and C in **Figure 2**. For any region enclosed by a topologically trivial (contractible) loop (type C), the total force acting on it can be calculated by applying the divergence theorem to the stress field

$$\oint_C dS \hat{\sigma} \cdot \hat{n} = \int dV \nabla \cdot \hat{\sigma} = 0, \quad 6.$$

yielding the same result regardless of loop shape, which simply reflects force balance. There are also two classes of noncontractible loops: loops labeled A that wrap around the torus in the x direction and those labeled B that wrap around in the y direction. Applying the divergence theorem leads to two vector quantities,

$$\begin{aligned} F_x &= \oint_A dS \hat{\sigma} \cdot \hat{n} \\ \text{and} \\ F_y &= \oint_B dS \hat{\sigma} \cdot \hat{n}, \end{aligned} \quad 7.$$

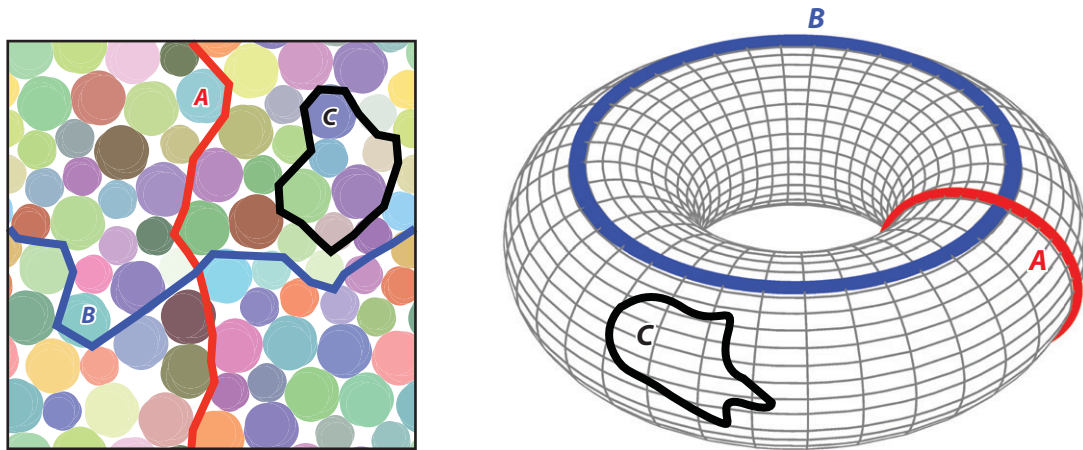


Figure 2

Illustration of invariants. (*Left panel*) The physical system, a two-dimensional granular solid of size $L \times L$, is outlined by the black box. The lines A and B represent the two distinct classes of noncontractible loops in the system, and C represents a trivial loop. (*Right panel*) Representation of the system on the surface of a torus. Loops A and B are noncontractible and correspond to the same labeling as in the left panel. To change F_x (or F_y), a change has to be made on the noncontractible loop B (or A).

that do not depend on the particular paths loops A and B take. These vectors are topological invariants in the sense that all loops belonging to the same topological class, A or B, have the same value of F_x or F_y , respectively. For a granular system, which is inherently discrete, the loops are not, in the strict sense, continuously deformable, given that they have to pass through force-bearing contacts. Although the deformation of loops must obey the constraints of the underlying contact network, this does not change the topological invariance of (F_x, F_y) .

For each dimension (here, there are two: x, y), there is an invariant vector F_i , which represents how much force is propagated through the material. These F vectors are equivalent to the sum of Cauchy traction vectors on a surface in continuum mechanics (47) and are also a formal generalization of the concept of Cartesian load proposed in Reference 26. The F vectors are related to the force-moment tensor, which is defined for discrete systems in Equation 2 and in continuum

as $\hat{\Sigma} = \int dV \hat{\sigma}(\mathbf{r})$ via

$$\hat{\Sigma} = \begin{pmatrix} L_x & 0 \\ 0 & L_y \end{pmatrix} \begin{pmatrix} F_x \cdot \hat{x} & F_x \cdot \hat{y} \\ F_y \cdot \hat{x} & F_y \cdot \hat{y} \end{pmatrix}. \quad 8.$$

The extensive quantity $\hat{\Sigma}$ has been used to construct the stress ensemble framework. However, from the preceding discussion, the natural conserved variables appear to be (F_x, F_y) (28), which scale linearly with system size. This feature is reminiscent of classical lattice models, which have ground-state degeneracy that leads to finite entropy. In these models, the subextensivity of the conserved quantity has nontrivial implications for phase transitions (48). It would be interesting to explore a theory of grains based on the conservation of (F_x, F_y) .

A natural microcanonical ensemble for jammed states of grains is one with a fixed value of the force-moment tensor, which plays the role of energy in equilibrium statistical mechanics. The force-moment tensor $\hat{\Sigma}$ is an extensive quantity (i.e., one that scales with system size), and one can define the generalized density of states:

$$\Omega(\hat{\Sigma}) = \sum_{\nu} \omega_{\nu} \delta(\hat{\Sigma}_{\nu} - \hat{\Sigma}).$$

The constraints on the microstates ν are that (a) forces and torques are balanced on every grain, (b) the Coulomb condition for static friction is satisfied, and (c) all forces are positive. If one now assumes factorization, $\Omega(\hat{\Sigma}_1 + \hat{\Sigma}_2) = \Omega(\hat{\Sigma}_1)\Omega(\hat{\Sigma}_2)$, there is an exact analog of temperature, which is the tensorial angoricity $\hat{\alpha}$:

$$\hat{\alpha}(\hat{\Sigma}) = \frac{\partial \ln \Omega(\hat{\Sigma})}{\partial \hat{\Sigma}}, \quad 9.$$

as was specified by Equation 3.

The preceding derivation is easily generalized to three dimensions. Contractible loops become closed volumes, and the three types of noncontractible surfaces correspond to the three invariants (F_x, F_y, F_z) ; see also Section 3.2.

1.4. Microcanonical and Canonical Formulations

The original Edwards ensemble can be generalized to one in which V and $\hat{\Sigma}$ define a microcanonical ensemble, but with no assumption of equiprobability. This generalized Edwards ensemble has much in common with statistical frameworks developed for glassy systems (1). Collective properties in both are determined by the complexity of the landscape of metastable states. For example, complexity in spin glasses is measured by the logarithm of the number of local free-energy minima that have a given free-energy density (1, 49). An analogous definition applies to the potential energy landscape of supercooled liquids (50, 51). For granular systems, complexity is measured by the generalized entropy $\Omega(V, \hat{\Sigma})$, with the caveat that the microstates in the Edwards ensemble must satisfy the constraints of mechanical equilibrium at zero temperature.

The microcanonical formulation of the generalized Edwards ensemble, which does not assume equiprobability, still assumes that the entropy is extensive (factorization of Ω). Recent numerical simulations establish extensivity in the absence of equiprobability (39). What the microcanonical formulation implies is the following: In a large system of grains that are jammed in a global volume V_g with a global force-moment tensor $\hat{\Sigma}_g$, the probability of finding a microstate within a subsystem is given by the Boltzmann-like distribution

$$P(\nu = \{\mathbf{r}_m, \mathbf{f}_{mm}\}) = \frac{1}{Z} \exp\left(-\frac{V(\{\mathbf{r}_m\})}{X}\right) \exp[-\hat{\alpha} : \hat{\Sigma}(\{\mathbf{r}_m, \mathbf{f}_{mm}\})], \quad 10.$$

where X and $\hat{\alpha}$ depend only on V_g and $\hat{\Sigma}_g$. This is a remarkably strong constraint on the probability of microstates in a nonequilibrium system. In general, this probability can depend on the complete history of preparation (the protocol) and any number of mesoscopic and macroscopic variables. Rigorous tests of Equation 10 are needed to establish the framework for both the microcanonical and canonical ensembles. In the next section, we discuss such tests for both experimental and numerical systems.

Defining the ensemble for static grains is the conceptually difficult part. Calculating the partition function is technically difficult because of the constraints that one still needs to take into account. An important point to make here is that establishment of a Boltzmann-like distribution does not imply that the distribution of individual grain volumes or individual contact forces is exponential (7). This would be true only if granular materials were like ideal gases with no

correlations. As we discuss in Section 3, granular materials are actually strongly interacting systems in which the interactions arise from constraints rather than interaction potentials.

2. TESTING THE ENSEMBLES

In spite of thought experiments proposed by theorists (1, 11), experimentalists were slow to warm to performing laboratory tests of the applicability and validity of the ideas put forth by Edwards. The earliest approach that gained traction was to consider analogs of the specific heat (52). A well-known result of ordinary statistical mechanics is that the energy fluctuations of a system around its mean value \bar{E} are related in a simple way to its specific heat at constant volume, via the relation $c_V = \left(\frac{\partial \bar{E}}{\partial T} \right)_V = \frac{\langle (E - \bar{E})^2 \rangle}{k_B T^2}$. Similarly, there is an analogous quantity $c = \frac{\partial \bar{V}}{\partial X} = \frac{\langle (V - \bar{V})^2 \rangle}{X^2}$ for the Edwards ensemble that allows measurement of relative changes in compactivity via observations of the volume fluctuations:

$$\int_{V_1}^{V_2} \frac{d\bar{V}}{\langle (V - \bar{V})^2 \rangle} = \int_{X_1}^{X_2} \frac{dX}{X^2} = \frac{1}{X_1} - \frac{1}{X_2}. \quad 11.$$

Here, the Boltzmann-like λ factor has been dropped for simplicity. This relation was first used to measure the compactivity of a granular material subject to tapping (52) and later on static packings prepared through varying degrees of fluidization (53). A disadvantage of this volume fluctuation method is that it integrates from a state with known $1/X_1$; typically, this is taken to be random loose packing, where $X_{\text{RLP}} = \infty$. However, it does provide a way to measure the compactivity using only bulk measurements.

As a by-product, the measurements of the volume fluctuations themselves provide a probe of the Edwards ensemble in the sense that a larger number of configurations corresponds to a larger $\langle (V - \bar{V})^2 \rangle$. This susceptibility can be informative in its own right, as a hallmark of the approach to a critical point. An observation of a minimum in the fluctuations as a function of packing density (53) has since been associated with the dilatancy transition (54, 55). Changes in such fluctuations have been observed in a variety of both experiments and simulations of static packings (56–60), but also in some systems with dynamics (61, 62). This last case is a test of Edwards's original idea that ensembles could describe driven granular systems and therefore hints at the validity of this aspect of the Edwards ensemble beyond strictly jammed states.

A second technique for measuring compactivity was proposed by Dean & Lefèvre (63). They observed that although the density of states $\Omega(V)$ and the partition function $Z(X)$ may be unknown for a given granular ensemble, $\Omega(V)$ is identical for the same system under two different conditions. For any particular macroscopic volume V , the probability P of observing that volume is given by

$$P(V) = \frac{\Omega(V)}{Z(X)} e^{-\frac{V}{X}}. \quad 12.$$

If one has measurements of $P(V)$ for the same system under two different preparations (enumerated 1, 2), the ratio of those two histograms is

$$\frac{P_1(V)}{P_2(V)} = \frac{Z(X_2)}{Z(X_1)} e^{\left(\frac{1}{X_2} - \frac{1}{X_1}\right)V}. \quad 13.$$

Therefore, by taking the logarithm of the ratio and fitting the data to determine the slope $\frac{1}{X_2} - \frac{1}{X_1}$, it is possible to measure changes in the compactivity. If one has a reference point (again, commonly $X_{\text{RLP}} = \infty$), then this can also be used as an absolute measurement. This technique was first realized by McNamara et al. (38), who measured these quantities in both simulations and experiments using local Voronoi volumes calculated from tomographs of an assembly of millimetric glass spheres. Because the histograms were nearly Gaussian, it was difficult to make quantitative conclusions beyond a verification that the technique worked: Values of X measured using either Equation 11 or 13 were in approximate agreement. This same overlapping histogram method was used on disk-shaped particles in a quasi-2D packing (64), where it was observed to be properly intensive for assemblies of at least 200 particles (68). Note that this technique for measuring compactivity does not rely on equiprobability and is insensitive to it because the factor $\Omega(V)$ in Equation 13 cancels out whatever its form.

Three other methods of calculating X are available for evaluation. First, using the histograms of local (Voronoi) volumes, it is possible to measure X directly by fitting a gamma distribution and considering the ratio of the mean free volume to the shape factor of the gamma distribution (65). Second, an equation of state relating the structural degrees of freedom and the number of independent boundary forces provides a fourth means of measuring X for isostatic packings (66, 67). A recent comparison of these two methods shows that they are not necessarily in quantitative agreement with the techniques described by Equation 11 and 13 (68). Finally, it is possible, through simulations, to both count particle configurations and to take the limit $T \rightarrow 0$ in a thermal system. Using such methods on a simulation of slowly sheared particles, the thermal T_{eff} (measured via the diffusivity and the fluctuation-dissipation theorem) was found to compare favorably with the compactivity (36).

Although these experiments have demonstrated the utility of the Edwards formulation as a description of a system, this was merely a confirmation that compactivity was measurable, not that it represented a truly temperature-like quantity. A more stringent test, as for ordinary thermodynamics, lies with the Zeroth Law of thermodynamics. A minimal expectation of a temperature-like quantity is that it have the ability to equilibrate between a subsystem and a bath. In experiments on a low-friction subsystem within a high-friction bath, the desired result would be for the compactivity to take on the same value in both. However, for a system of 2D disks floated on a frictionless surface and biaxially compacted, it was observed that although X measured using either Equation 11 or 13 matched for either the subsystem or for the bath, the two values systematically failed to agree between the subsystem and the bath (69). This represents a significant failure of the volume ensemble and compactivity.

Tests of the stress ensemble have a shorter history, but the angoricity appears to be more promising as a valid temperature-like quantity. Because it is a tensor (see Equation 3), it is helpful to calculate it by considering shear (Σ_τ) and compressional (Σ_p) components separately. Each of these components gives rise to its own angoricity, α_τ and α_p , respectively. For example, the shear component (in a 2D system) is $\Sigma_\tau = (\Sigma_1 - \Sigma_2)/2$ and the compressional part is $\Sigma_p = (\Sigma_1 + \Sigma_2)/2$, whereof Σ_1 and Σ_2 are the eigenvalues of the force-moment tensor $\hat{\Sigma}$. The trace of the force-moment tensor has been commonly referred to as $\Gamma \equiv 2\Sigma_p = \text{Tr } \hat{\Sigma}$.

The stress ensemble was first tested using numerically generated packings of frictionless, deformable particles (24). As shown in **Figure 3a**, histograms of Γ obtained from clusters of grains within a packing depend strongly on the density of the packing ϕ (proportional to $1/V$ for a fixed

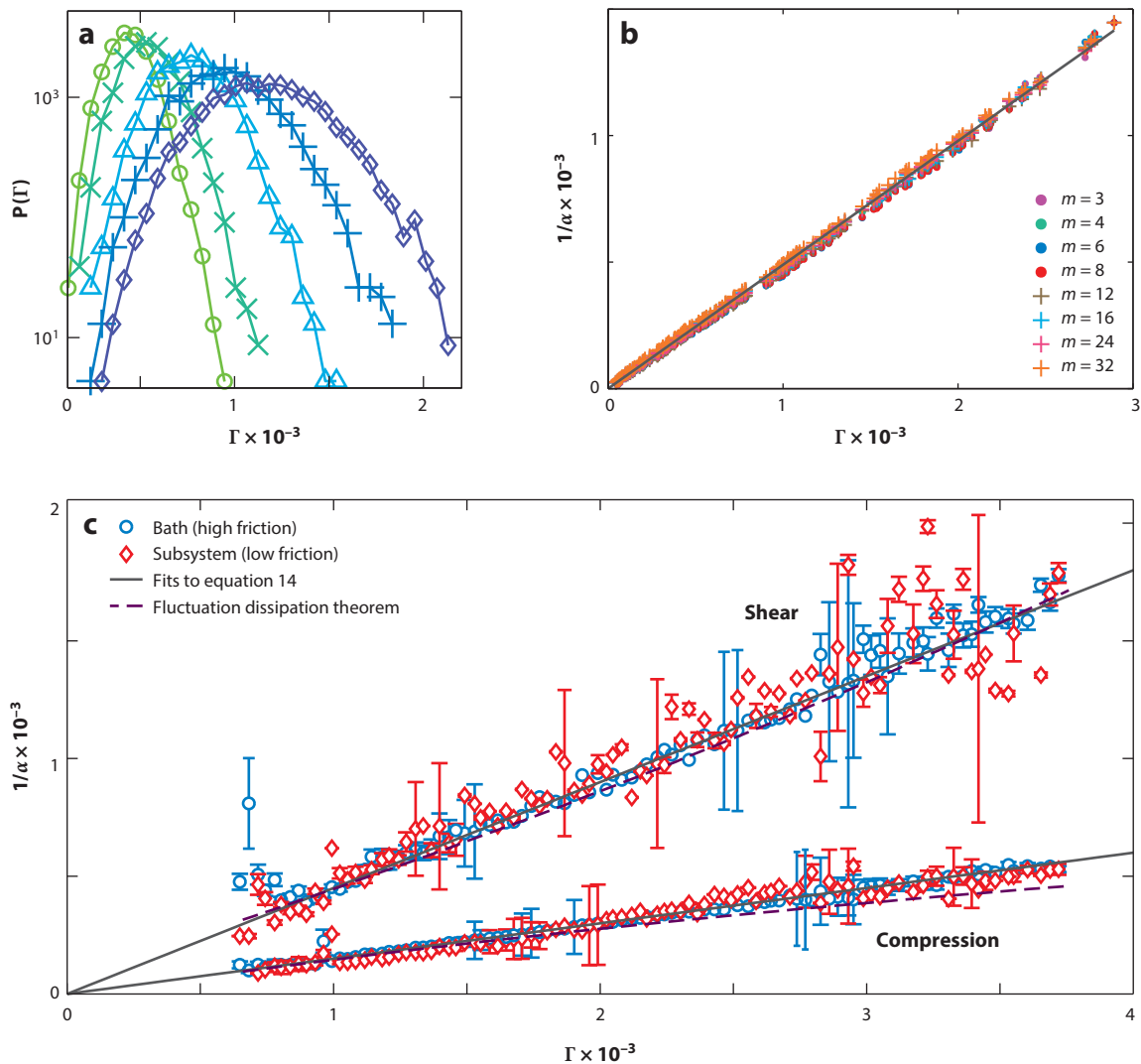


Figure 3

(a) Histograms of the probability of measuring pressure Γ for subsystems of size $m = 8$ within simulations of a frictionless, deformable granular material containing 4,096 particles (24). Packing fraction ϕ increases from left to right. The ratios of these histograms provide a measurement of the angoricity α , using the technique described in Equation 13. (b) Measured values of the reciprocal angoricity $1/\alpha$ as a function of Γ in the same simulations, showing a slope of 0.49 (solid line), which is consistent with theoretical predictions (25). Results are independent of subsystem size m (colored symbols). (c) Equilibration of angoricity in a subsystem of low-friction particles (diamonds) within a high-friction bath (circles), in repeated realizations of a quasi-2D packing (69). The upper data correspond to α_τ (shear), and the lower data correspond to α_p (compression). The solid lines are fits to Equation 14, with slopes of 0.45 and 0.15, respectively. Dashed lines are for values calculated using the fluctuation dissipation theorem (analogous to Equation 11) and are found to be in agreement. Adapted from References 24 and 69.

number of particles). Using the method outlined in Equation 12 and 13, it is possible to (a) test whether the distribution of Γ within a packing has a Boltzmann-like form and (b) extract the differences in angoricity between different packings. This led to the identification of an equation of state:

$$\frac{1}{\alpha_p} \propto \langle \Gamma \rangle. \quad 14.$$

This equation of state was found to be independent of cluster size (**Figure 3b**) for clusters more than an order of magnitude smaller than was necessary for compactivity measurements. The slope of the observed line is 1/2, which is in agreement with the value predicted from a configurational entropy estimate of $2/z_{\text{iso}}$ (25).

A similar technique was applied to packings in experiments (70), where the presence of friction requires considering the full tensorial angoricity. A linear equation of state was observed there as well. These linear equations of state indicate that the configurational entropy of the packings arises via an ideal gas of noninteracting local stresses. A field theoretical model (25) of the spatial stress fluctuations was developed based on numerical data for frictionless disks, and this has been used to predict the stress fluctuations of numerical and experimental packings under pure shear (71).

The generalized Edwards ensemble, combining volume and stress as independent factors, was tested in simulations of three-dimensional (3D) frictionless grains (72). Thermalization was achieved using independently generated configurations, and results from the ensemble were found to be in agreement with predictions from jamming. Experiments in 3D remain hampered by a lack of techniques for measuring vector contact forces in the interior of granular materials. No experiments or simulations have yet addressed the Edwards ensemble as a joint distribution of volumes and forces, as has been proposed by Blumenfeld & Jordan (67). In addition, little is known about how preparation protocol, particle shape, or particle polydispersity affect the ensembles.

Unlike for compactivity, there has been success in applying angoricity in canonical situations as well as microcanonical ones. **Figure 3c** shows results from the same experiment (69) in which compactivity failed to equilibrate. A subsystem of low-friction ($\mu < 0.1$) particles was placed within a bath of higher friction ($\mu = 0.8$) particles. The overlapping histogram method provides angoricities α_p and α_τ for a set of packings prepared at a given volume (and measured pressure Γ). Each of these two temperature-like quantities equilibrated (independently) between the subsystem and the bath: There is a 0th Law. In addition, each angoricity again had a linear equation of state with the same form as Equation 14. The coefficients of these equations of state were of similar magnitude to those observed in simulations (24). However, a configurational entropy calculation (25) would predict a slope of $2/z_{\text{iso}} = 2/3$, which is larger than was observed.

3. STATISTICAL MECHANICS OF JAMMED STATES

As in equilibrium statistical mechanics, establishing the generalized Edwards ensemble provides us with a canonical distribution of microstates: $P_\nu(X, \hat{\alpha})$. Therefore, the machinery of statistical mechanics can be applied to calculate ensemble averaged quantities. In particular, one can define the generating function (canonical partition function):

$$Z(X, \hat{\alpha}) = \sum'_{\{\mathbf{r}_m, \mathbf{f}_{mn}\}} \exp\left(-\frac{V(\{\mathbf{r}_m\})}{X}\right) \exp[-\hat{\alpha} : \hat{\Sigma}(\{\mathbf{r}_m, \mathbf{f}_{mn}\})], \quad 15.$$

where the prime on the summation indicates that the sum over $\{\mathbf{r}_m, \mathbf{f}_{mn}\}$ is to be restricted to only those states that satisfy the constraints of static mechanical equilibrium for every grain. Imposing these constraints poses difficult challenges (26, 73–75). Whereas the force and torque balance are equality constraints and are therefore straightforward to implement, the positivity of forces

and the static friction law are both inequality constraints that are notoriously difficult to implement.

In this section, we discuss the different schemes developed to calculate ensemble averaged properties of jammed solids, emphasizing the approximations involved in implementing the constraints of mechanical equilibrium. In both 2D and 3D, the constraints of force balance can be enforced through the introduction of gauge fields (25, 46, 73). In 2D, these gauge fields have an elegant geometric representation that has been used in recent years to calculate statistical properties of jammed states at various levels of approximation. We first focus on the intertwining of volume and stress in the partition function (Equation 15). Then, we discuss methods applicable in 2D before providing a brief description of a similar mathematical formulation in 3D, along with the difficulties that arise when moving from 2D to 3D.

Equation 15 is a function of both X and $\hat{\alpha}$, and this cannot, in general, be written as a product: $Z(X, \hat{\alpha}) \neq Z(X)Z(\hat{\alpha})$. Recent work (67) draws attention to this fact and obtains an expression for the partition function by combining the loop force formalism (43) (to enforce force and torque balance in 2D) with the quadron method for taking into account the volume degrees of freedom. For the special case of an isostatic assembly of grains, where each geometry leads to a unique configuration of contact forces, the calculations (67) give rise to an equipartition-like equation of state linking mean volume $\langle V \rangle$ and compactivity $\langle V \rangle = (zN/2 + M)X$, where the $zN/2$ structural degrees of freedom and the M independent boundary force degrees of freedom contribute equally.

Most statistical treatments of jammed states have relied on some form of decoupling of the volume and stress terms in the partition function. The original Edwards ensemble is an extreme example: The stress fluctuations were not considered at all. Conversely, the force network ensemble (FNE) (76) is based on a complete decoupling of the geometric and force degrees of freedom. It is a microcanonical ensemble that takes a flat measure on all allowed force states for a given, fixed geometric network. The FNE decoupling is based on a separation of scales: For infinitely rigid grains or packings of soft grains very close to jamming, small geometric displacements lead to large changes in the interparticle contact forces. The FNE is particularly suited to computing probability distributions $P(f)$ of single contact forces f in both lattice geometries and some off-lattice geometries (77–79). It has also proved useful in understanding force fluctuations in frictional packings (80). There are many excellent reviews of the application of the FNE, and we refer the reader to these for a comprehensive discussion (81, 82).

Another variation on the decoupling theme is to consider a canonical stress ensemble that neglects all volume fluctuations (24, 28, 70). This is the analog of the (T, V, N) ensemble of equilibrium statistical mechanics. This is distinct from the FNE because both positions and forces are retained as degrees of freedom through the force-moment tensor. The stress ensemble is most appropriate for describing grains subject to shear stresses at constant volume (70). It is a reasonable approximation in situations where volume fluctuations are negligible compared to stress fluctuations. The experimental tests involving both X and $\hat{\alpha}$, described in the previous section, suggest that $\hat{\alpha}$ can be defined even in cases where the definition of X is murky. In the remainder of this section, therefore, we focus on the pure stress ensemble (which includes the FNE as a special case). The interdependence of volume and stress fluctuations in the generalized ensemble, which is the analog of the (T, P, N) ensemble, has received less attention (67). One of the important questions in this regard is the range of volumes over which jammed packings can be created (15).

3.1. Two-Dimensional Granular Solids

In 2D, the constraints of local force balance (Equation 5) allow for the definition of a vector gauge field, $\mathbf{h}(x, y)$ (23, 25, 46, 73), such that¹

$$\hat{\sigma} = \nabla \times \mathbf{h}. \quad 16.$$

In a discrete formulation at the grain level, this single-valued vector height field is defined on the dual space of voids in a granular packing (43), as illustrated in **Figure 4**. Going around a grain in a counterclockwise direction, the height field gets incremented by the contact force separating two voids. The mapping to this dual space of loops and heights is rigorous and retains all of the information about the original packing (25, 43, 46, 73).

A simpler geometric representation that does not carry information about the real-space geometry (but retains the topology of the contact network and accurately represents the structure in height space) is the tiling picture (27) shown in the left panel of **Figure 4b**. Because the forces on every grain must add up to zero, each grain can be represented by a polygonal tile constructed by connecting the force vectors head-to-tail. Newton's Third Law requires that the edges of tiles exactly coincide because each edge (contact force) is shared by two grains. The vertices of this tiling are the heights, and the faces of the tiles represent grains. Because the force associated with touching grains is equal in magnitude and opposite in direction, force tiles tessellate height space and the vertices of all force tiles fill a parallelogram (see **Figure 4**) in height space spanned by two vectors, F_x and F_y , defined in Equation 8.

The force tilings can form the basis for calculating partition functions and ensemble averages. In the tiling representation, the microcanonical stress ensemble comprises all tilings by N_f tiles of a parallelogram bounded by F_x and F_y , with each tile corresponding to a force-bearing grain with more than two contacts. Grains with two contact forces or fewer are not represented in a tiling. We discuss this aspect in the context of tiling corresponding to frictional grain packings. A canonical stress ensemble would be defined by $\hat{\alpha}$, and in this ensemble the boundary vectors defining the parallelogram would be allowed to fluctuate. In order to perform an actual calculation, the microscopic probabilities, ω_ν , of each tiling would have to be defined and the additional constraints of torque balance and static friction would have to be imposed on the tilings. The latter two are irrelevant for frictionless grains, and all calculations (to date) of statistical properties of frictionless grains have assumed equiprobability $\omega_\nu = 1$. For frictionless grains, the FNE has been used successfully to calculate statistical properties of grains in lattice geometries (83). Interestingly, tilings of frictionless grains exhibit an additional extensive quantity that modifies the stress ensemble. In the rest of this section, we summarize the application of stress ensembles to assemblies of frictionless and frictional grains.

3.1.1. Frictionless grains. For dry granular solids, the forces between grains are purely repulsive. Without friction and for disk-shaped particles, these contact forces are all central forces. It can be easily proven that if forces are frictionless and repulsive, all force tiles must strictly be convex in shape. When all force tiles are convex in shape, they form a planar graph in height space. As illustrated in **Figure 4**, the result is equivalent to the Maxwell-Cremona reciprocal tiling or force tiling (27), where each grain is represented by a polygonal tile. Hence, the area of the parallelogram spanned by F_x and F_y is simply the total area taken up by all tiles in height space

¹The generalized definition of the curl used implies that $\sigma_{\mu\nu} = \sum_{\mu'\nu'} \epsilon_{\mu\mu'\nu'} \partial_{\mu'} W_{\nu'\nu}$ in 3D and $\sigma_{\mu\nu} = \sum_{\mu'} \epsilon_{\mu\mu'\nu} \partial_{\mu'} h_\nu$ in 2D, where ϵ is the Levi-Civita symbol.

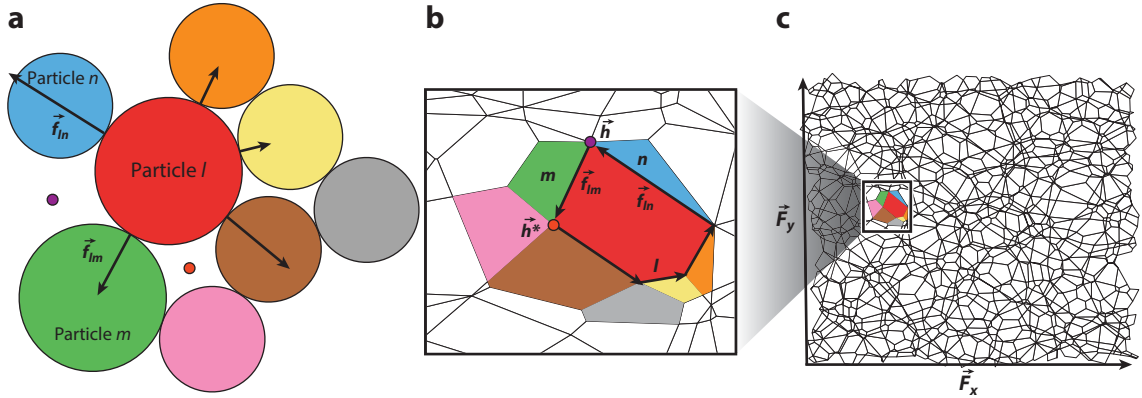


Figure 4

Force tilings for a frictionless grain packing from simulations (courtesy of Corey S. O'Hern). (a) Particle l is supported by balanced contact forces that are exchanged with its neighbors (m , n , etc.). The real space contact network defines voids such as the one bordered by m , l , and n (purple dot), on which the vector loop forces \vec{h} , \vec{h}^* live. Going around a grain in a counterclockwise direction, the height is incremented by the contact force separating two voids. (b) The force tile corresponding to grain l and its neighbors. The force vectors define the edges of the force tiles, the faces are labeled by grain indices, and the heights \vec{h} and \vec{h}^* become the vertices of the tiling. Because of action-reaction and force balance, force tiles tessellate height space, as shown in panel c. The boundaries of the tessellation are defined by the vectors \vec{F}_x and \vec{F}_y , which define the force-moment tensor $\hat{\Sigma}$ (Equation 8).

$$A_{\text{tot}} = \sum_m a_m = |\vec{F}_x \times \vec{F}_y|, \quad 17.$$

where the last equality holds only in the thermodynamic limit or for PBCs (83). Equation 17 demonstrates that there is an additional extensive quantity for 2D frictionless assemblies of grains that is distinct from $\hat{\Sigma} = \sum_m \hat{\sigma}_m$. Although the conservation principle discussed in Section 1.3 is a prerequisite for the additivity and conservation of A_{tot} , the latter holds only in special cases, such as for frictionless grains for which all force tiles are convex. If A_{tot} is conserved, then following the steps in the derivation of Equation 15, an additional Boltzmann term has to be introduced to impose the area constraint on the canonical, microstate probability for frictionless grains:

$$P_\nu = \frac{1}{Z} \exp(-\hat{\alpha} : \hat{\Sigma}_\nu - b \mathcal{A}_\nu), \quad 18.$$

where \mathcal{A}_ν is the area of a microscopic state and

$$b = \frac{\partial S}{\partial A_{\text{tot}}}. \quad 19.$$

For isotropic states, $\hat{\alpha} : \hat{\Sigma}_\nu$ reduces to $\alpha_p \Gamma_\nu$, where Γ_ν is the trace of the force-moment tensor. Γ_ν is proportional to the perimeter of a tile, \mathcal{P} . Furthermore, for a polygon that is convex and isotropic in shape, area is related to perimeter via $\mathcal{A} \propto \mathcal{P}^2$. (Conversely, anisotropic polygons or complex polygons may have a power < 2 .) In this case, Equation 18 becomes

$$P(\mathcal{P}) \propto \frac{1}{Z} \exp(-\alpha_p \mathcal{P} - \beta \mathcal{P}^2), \quad 20.$$

which coincides with the result obtained in a test using FNE (27) and entropy maximization. Within the tiling framework, force and pressure distributions can also be calculated using the FNE

on a lattice and off-lattice with excellent agreement between theory and Monte Carlo simulations (27, 83, 84), although the approach remains to be tested on naturally generated packings with coupled forces and geometry. One particularly interesting result is that the distribution of the magnitude of individual contact forces, $P(f)$, is Gaussian for packings of frictionless grains (27, 83, 84), and this is a direct consequence of the additional area conservation. The absence of this additional conservation principle in frictional grains, to be discussed below, raises the intriguing possibility that the exponential form of $P(f)$ observed in experiments is connected to force chains and friction.

3.1.2. Frictional grains. The tiling framework for frictional grains admits nonconvex and even complex (self-intersecting) polygons. **Figure 5** shows a tiling constructed from experimental data on jammed states created through shear (85). Grains located along force chains, where there is a well-defined direction of large forces, tend to give rise to self-intersecting polygons, as illustrated in **Figure 5**. The additional area conservation principle is, therefore, not applicable to tilings of frictional grains, which are nonplanar because of the presence of self-intersecting polygons. For frictional grains, therefore, the generalized Edwards ensemble encodes all the conservation laws, and there are no additional conserved quantities. The number N_f of tiles corresponds to grains with more than two contacts. In frictionless grains, this number is not far from the actual number of grains. However, in frictional systems at low densities close to random loose packing, force-bearing grains organize into force chains that meander through a host of spectator particles that bear essentially no force (3, 86). For these systems, N_f can be very different from the actual number of grains. The tiling framework shows that for mechanical stability, what matters is N_f . The volume in the original Edwards ensemble refers to the volume occupied by all grains.

In order to perform calculations for frictional grains, assuming equiprobability of microstates, the constraints of torque balance and static friction would have to be imposed explicitly on the sampling of tiles. To our knowledge, such a calculation has not been attempted yet. A mean-field calculation has been performed (73) and leads to an approximation of the (force space)

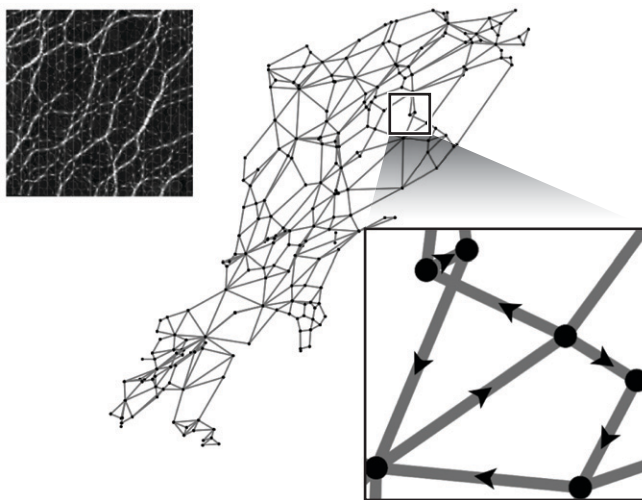


Figure 5

Force tilings constructed from experimental shear-jammed states. Clockwise from left: A photoelastic image (87) showing force chains, force tiling corresponding to this image, and a part of the force tiling showing a self-intersecting polygon. The arrows mark the direction of contact forces.

configurational entropy and hence a prediction for the equation of state for grains with friction. In another development, force tilings constructed from experimental data have been analyzed to demonstrate that the emergence of rigidity in shear-jammed solids (86) can be related to a persistent pattern in the density of vertices in force tiles (85).

There is a clear need for the development of a Metropolis Monte Carlo algorithm that samples tiling according to the Boltzmann-like distribution of the stress ensemble but that also imposes the constraints of torque balance and the Coulomb friction law $|f_t| \leq \mu|f_N|$ on the sampling. The advantage that the tiling framework offers is the ability to encode these constraints in the geometrical language of polygons. It should be emphasized that sampling tilings is not restricted to a fixed geometry of grains in real space. What is fixed is the number of force-bearing grains; for example, their positions and contact forces are dynamical variables in a Monte Carlo scheme. The positions do not enter the tiling picture explicitly, and an assumption being made in sampling tilings without connecting to the real-space network is that there is a real-space network of grains that corresponds to every tiling. This assumption is reasonable if there is a decoupling of forces and positions, and the indeterminacy arising from tangential forces helps in this regard. Even if numerical calculations are difficult, the tiling framework allows us to construct correlations in height space, which are often more illuminating than real-space correlations (85). This is because the primary source of correlations in dry granular ensembles is the constraints of mechanical equilibrium, and these are explicitly taken into account in all tiling configurations.

3.2. Three Dimensions

In the 3D case, the definition of an equivalent microscopic framework to the 2D tiling becomes much more complex (46, 73). At the continuum level, the equivalent to the vector gauge field \mathbf{h} is a tensor field \hat{W} such that Equation 16 becomes¹ $\hat{\sigma} = \nabla \times \hat{W}$. Equivalently to the 2D case, we can use Stokes's theorem to show that the stress in a volume is a boundary term: the surface integral of \hat{W} dotted with the unit normal. Because \hat{W} is a tensor rather than a vector, it is unclear what the analog of the dual tiling space would be; no 3D equivalent of the tiling has been proposed so far. Recently, DeGiuli & McElwaine (73) derived a microscopic interpretation of \hat{W} , analogous to the 2D height field. However, although the microscopic variables are vectorial, they map only to \hat{W} by incorporating significant parts of the local geometry. Finally, it should be noted that for realistic 3D packings, gravitation needs to be incorporated, and the framework based on conservation of $\hat{\Sigma}$ needs to be extended to include gravity as an external field.

4. APPLICATION TO DYNAMICS

As mentioned in the introduction, one of Edwards's original ideas was that the dynamics of slowly driven granular materials can be understood based on the microstate probabilities of jammed states. This seems plausible if the driving is slow enough that there are rare transitions from one jammed state to another, and the transition probabilities are controlled by properties of the jammed states. This picture is reminiscent of frameworks constructed for describing the glassy dynamics of thermal systems, such as the trap model (88), the soft-glassy-rheology (SGR) models (89), and the shear-transformation-zone (STZ) theory (90). In all of these models, there are transitions into and out of trap-like entities, such as free-energy minima. For example, in the SGR model it is assumed that the material can be partitioned into mesoscopic regions. Each region is able to accommodate some elastic energy and strain. Plastic deformations of the mesoscopic region occur because of either (a) global driving or (b) fluctuations due to the elastic energies released by other mesoscopic regions undergoing plastic deformation (89). In this mean-field model, the latter is

represented as an activated process with an effective temperature that is different from the bath temperature (89).

A natural avenue of exploration of granular dynamics based on the ensemble ideas is to adopt the same philosophy as SGR and assume that angoricity plays the role of the effective temperature. The idea that self-activated processes induced by stress fluctuations, or mechanical noise, influence the rheology of slow granular flows has been explored in the literature (91, 92), and it seems reasonable to assume that angoricity determines the strength of this mechanical noise. An angoricity-based, SGR-like framework has been used to analyze stress fluctuations in laboratory granular Couette flows (93, 94). This formalism provides an explanation for the observed logarithmic strengthening (95) of granular materials. A key point is that one needs to use a stress ensemble rather than an energy ensemble (as would be the case for SGR) to obtain the correct scaling with shear rate. Such applications of the stress ensemble to analyze granular rheology are in their infancy, and much remains to be explored.

5. SUMMARY AND OPEN QUESTIONS

Our overview of the application of statistical ensemble ideas to athermal, granular materials highlights recent advances in this area. One of our primary objectives has been to emphasize the similarities and differences between athermal and thermal ensembles. The experimental and numerical tests discussed in this review illustrate the validity of the ensemble framework in several different contexts. The equilibration of angoricity, which is the analog of thermalization in the stress ensemble, has been shown to work both in experiments and simulations. Measurements of angoricity and compactivity based on the method of overlapping histograms (Equation 13) and on fluctuation-response relations (Equation 11) seem to agree with each other, which suggests that even in these athermal ensembles, fluctuations within a system determine its linear response to driving.

The majority of tests have been performed in 2D, and the applicability of ensembles to 3D packings is much less clear. Equations of state, such as those relating the angoricity to the external stress and compactivity to volume, have been measured in only a few systems. In particular, equations of state involving the full tensorial angoricity have remained largely unexplored, as have equations of state relating compactivity and angoricity. Because equiprobability does not seem to be universally valid for jammed systems, equations of state depend upon the probabilities ω_ν of the microstates, and these are expected to be sensitive to preparation protocols, friction coefficients, and particle shapes. The variation in equations of states is, therefore, expected to be greater than that found in thermal ensembles. Validity of the ensemble framework, however, narrows down the choice of macroscopic variables that should enter an equation of state.

From the perspective of calculations based on ensembles, the difficulty of imposing the constraints of mechanical equilibrium has hindered progress. In particular, calculations of partition functions have been hampered by two inequality constraints: positivity of forces and the Coulomb condition for static friction. The situation is hopeful in 2D, where the constraints have elegant geometrical representations, but little is known about how to proceed in 3D.

Within the context of dynamics, the ensemble framework provides a natural basis for stress-induced fluctuations being represented by a type of thermal noise. The rheology of slowly driven systems could potentially be completely linked to angoricity and compactivity by generalizing the idea of thermally activated processes to stress-activated processes that lead to plastic failure. If dynamics can be related to granular equilibrium concepts such as angoricity and compactivity, that would be a big step because granular rheology could then be related to stresses imposed at the boundaries.

To summarize, the past decade has seen much progress in testing and establishing the generalized Edwards ensemble, which establishes the laws of granular equilibrium. We should

emphasize that the ensemble framework is relevant only to static and slowly driven granular materials where the kinetic energy of the grains does not play a significant role. For systems that are governed by the laws of granular equilibrium, stress has emerged as the physical property that dominates collective organization.

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