Segregation mechanisms in a numerical model of a binary granular mixture

George C. M. A. Ehrhardt* and Andrew Stephenson

Theoretical Physics Group, Department of Physics and Astronomy, The University of Manchester, M13 9PL, United Kingdom

Pedro M. Reis

Manchester Centre for Nonlinear Dynamics, The University of Manchester, M13 9PL, United Kingdom

(Received 10 March 2004; published 1 April 2005)

A simple phenomenological numerical model of a binary granular mixture is developed and investigated numerically. We attempt to model a recently reported experimental system where a horizontally vibrated binary monolayer was found to exhibit a transition from a mixed to a segregated state as the filling fraction of the mixture was increased. This numerical model is found to reproduce much of the experimentally observed behavior, most importantly the transition from the mixed to the segregated state. We use the numerical model to investigate granular segregation mechanisms and explain the experimentally observed behavior.

DOI: 10.1103/PhysRevE.71.041301 PACS number(s): 45.70.–n, 02.50.Ey, 64.75.+g, 82.70.Dd

I. INTRODUCTION

Granular systems exhibit a wide range of intriguing and often counterintuitive phenomena. Segregation of two or more species of grains due to vibration or shearing is one such example [1,2]. Many mechanisms, including buoyancy, temperature gradients, differing angles of repose, and differing roughness of particles, have been proposed [2–5]. Recently, segregation of a vibrated binary monolayer has been demonstrated in a series of experiments [1,6,7] that imply/suggest the existence of a segregation critical point as the filling fraction is varied, with associated growth of fluctuations and time scales in the vicinity of that point.

In this paper we propose a phenomenological numerical model of the experimental system which captures its essential features. We show that there is qualitative agreement between the numerical model and the experiment and, in particular, that the quantitative measures reported in [6,7] are of the same form in both cases. We then use the numerical model to study segregation mechanisms, giving one definite mechanism and demonstrating that a second may also play a role.

The experimental system used in [6,7] consists of a smooth horizontal tray of dimensions $L_x=180$ mm by $L_y=90$ mm vibrated sinusoidally parallel to its major axis at a frequency of $f=12$ Hz and amplitude of $A=1.74\pm0.01$ mm. The grains are high-precision phosphor-bronze spheres (denoted by $c$ in the following) of radius $R_c=0.75$ mm and mass $m_c=16.8$ mg and poppy seeds (denoted by $p$ in the following) of average radius and mass $R_p=0.54$ mm and $m_p=0.52$ mg respectively. The poppy seeds are rough, non-spherical polycrystalline [8] particles.

Particles placed on the oscillating tray predominantly move periodically with the driving. They also have some quasirandom component to their motion, caused in the case of the $p$’s by their nonsphericity which means that they do not roll smoothly but instead “stick and slip” and scatter. The $c$’s are affected less by the driving since they roll as the tray moves beneath them. However they too, when placed individually on the tray, are seen to have some quasirandom component to their motion. The overall impression is of a granular bed of agitated particles with the predominant motion being parallel to the driving.

The control parameter used in the experiments was the two-dimensional filling fraction, termed the compacity $C=(N_c\pi R_c^2+N_p\pi R_p^2)/(L_xL_y)$, with $N_c$ fixed. The value of $N_p$ is also important but for simplicity the experiments focussed on $N_c=1600$ and varied $N_p$. The system was quasi-two-dimensional in that $0.49 < C < 1.2$, high $C$ being achieved by poppy seeds “riding up” and overlapping each other to some extent, although never so much that they were above and overlapping the $c$’s.

Starting from a uniformly mixed initial distribution, the final state reached by the system varies with $C$. For $C \approx 0.65$ the system remains mixed while above that value small mobile segregated clusters of $c$’s form. As $C$ increases, these clusters grow in size and become anisotropic, forming stripes perpendicular to the direction of driving for $C$ large. For $C \approx 0.93$ the $c$’s within these stripes crystallize to form a dense close-packed hexagonal lattice. These three phases were termed binary gas, segregation liquid, and segregation crystal. The existence of a phase transition from the binary gas to segregation liquid, with an associated critical value of $C$, was reported in [7]. The details of these results are given in [6,7].

The experimental system has several appealing properties from the point of view of studying granular matter: except at high compacities where stripes form, the final state reached is a function only of the compacity, i.e., the initial conditions are not relevant. The particles are always in contact with the tray and hence are always effectively “thermalized.” This is in contrast to the behavior of particles in granular systems such as sandpiles where the grains spend most of their time locked in position, or some vertically vibrated systems where much of the time is spent in free flight [9]. The constant “thermalization” of the particles, their ability to explore many possible states, and the irrelevance of initial conditions

*Presently at The Abdus Salam ICTP, Strada Costiera 11, 34014 Trieste, Italy. Electronic address: gehrhard@ictp.trieste.it
suggests that the system might be a good choice for studying the “statistical mechanics” of granular matter. However detailed balance, equipartition of energy and other rigorous features of equilibrium statistical mechanics are of course not obeyed in this system [10].

II. DESCRIPTION OF THE NUMERICAL MODEL

We here develop a phenomenological numerical model of the experimental system. The aim is to capture the essential features of the experiment in the numerical model, thereby discovering what those features are and in particular what the segregation mechanism(is) are.

The following features are, we believe, necessary: there is a tray of dimensions $L_x \times L_y$ whose base moves with sinusoidal velocity $A \sin(\omega t)$. There are two particle species, named $c$ and $p$. The particles moving relative to the tray surface are subject to a frictional force. The particles feel a randomisation of their velocity caused by their stick-slip motion with the oscillating tray and their nonsphericity (thermalization). The particles collide inelastically.

We therefore model the system as two-dimensional with the particles behaving as hard disks of mass $m_\alpha$ and radius $R_\alpha$, where $\alpha$ denotes the species ($c$ or $p$). Except during collisions, each particle $i$ of type $\alpha$ obeys the Langevin equation

$$m_\alpha \dot{v}_{\alpha i} = - \gamma_\alpha (v_{\alpha i} - v_{\text{tray}}) + \eta_{\alpha i}(t),$$

where $v_{\text{tray}} = \omega A \sin(\omega t)$ and $\gamma$ provides a linear damping. $\eta_{\alpha i}(t)$ is Gaussian white noise of mean zero and standard deviation $\langle \eta_{\alpha i}(t) \cdot \eta_{\alpha i}(t') \rangle = 2\sigma_\alpha^2 \delta(t-t')$ and this provides the “thermalization.” Note that the $v_{\text{tray}}$ term provides a collective forcing while the noise term, $\eta_{\alpha i}$ is unique to each particle. The particles interact through smooth hard-disk inelastic collisions with coefficient of restitution $r_{\alpha,\beta}$. Thus, in the centre of mass frame, a particle $i$ of type $\alpha$, undergoing a collision with a particle $j$ of type $\beta$, has its velocity changed according to: $v_{\beta j} \rightarrow r_{\alpha,\beta} p v_{\beta j}$, $v_{\alpha j} \rightarrow v_{\alpha j}$, where $v_{\beta j}$ and $v_{\alpha j}$ are the velocity components parallel and perpendicular to the line joining the centres of the particles $i$ and $j$. For simplicity, the disks have been taken to be smooth sided so that angular momentum is not considered. A similar model has been used in [11] to describe colloidal particles driven by an external electric field. Also [12] has used a similar model without noise to model granular particles driven by a vertically oscillating air column [13]. There exists many other models for driven granular systems, for example [14] gives a general class of stochastic multiplicative driving terms which includes our (nonmultiplicative) driving term as a special case, the references given here are in no way intended to be a complete list.

These are the essentials. We have also kept the walls of the box stationary for simplicity but modeled the motion of the end walls by considering that in collisions they have a velocity $i \max(\omega A \sin(\omega t),0)$ for the left wall and $i \min(\omega A \sin(\omega t),0)$ for the right wall. We felt that this was necessary in order to model the low-density region near each end wall caused by the vigorous collisions with the end walls. The width of the low-density region is independent of the system size, thus we used $A_1 = A/10$ rather than $A_1 = A$ in the results described here in order to reduce the “finite size effect” of this region as the system size is changed.

There are many approximations of the real system made here, the most significant ones are: the friction term $\gamma v$ is only an approximation, it is chosen as being the simplest possible form. The noise is in fact due to the stick and slip interactions between the particles and the oscillatory surface, and also their non-sphericity when interacting with each other and with the tray. We do not try to directly model this since it would require detailed specification of the shape of each particle and its actual interaction with the tray, which is not known. Even a single high-precision phosphor-bronze sphere conducts a quasirandom walk when placed on the oscillating tray, indicating that the randomness can depend on very small imperfections of the particles and the tray (and possibly also in the driving). Both because of this immense difficulty and in order to have a reasonably simple numerical model whose behavior we can understand, we instead choose to include the noise phenomenologically. We assume that the noise the particles receive is independent of their neighbors and of the phase of the tray cycle, both of which are unlikely to be accurately what happens in the experiments. The assumption that the noise is Gaussian and white is an approximation. For simplicity we are using a two-dimensional numerical model, which ignores the overlapping of $p$’s and the rolling of particles. The particles and walls are assumed to be smooth and so angular momentum is not considered. We have ignored the polydispersity of the $p$’s. Polydispersity was included to check its importance and was found to leave the qualitative behavior unchanged. The final approximation is that the coefficients of restitution are constant which is a commonly made one [15]. Despite these simplifications and approximations, in Sec. III we show that our phenomenological numerical model captures much of the behavior observed experimentally.

A. Parameter values

Static parameter values such as mass and size can be measured reasonably accurately, for the poppy seeds we have used the mean values of a sample of measurements [16]. Dynamic parameters were less accurately known, $\gamma_\alpha$ was estimated from the distances traveled by single particles striking the moving end walls. Using the result for the noiseless case,

$$x(\infty) = x(0) + v_\perp(0)m/\gamma$$

and estimating $v_\perp(0)$ to be equal to the maximum velocity of the end wall gives approximate values for $\gamma_\alpha$. The value for the noise is the hardest to determine since no velocity or accurate $r(t)$ path measurements were available. We merely estimated that the mean square velocity due to the noise, $\langle v^2 \rangle = \sigma_\alpha^2/m_\alpha \gamma_\alpha$ should be equal to $\omega A/F^2$ where the factor $F=3$ for the $p$ case and $F=13$ for the $c$ case.

Clearly these last estimates are rather crude. However, extensive study of a wide range of these parameters has shown that the qualitative features are robust to variation of these estimates.
The coefficients of restitution, $r_{\alpha,\beta}=r_{\beta,\alpha}$, are estimated to be: $r_{c,c}=0.9$, $r_{c,p}=0.2$, $r_{p,p}=0.1$, $r_{c,w}=0.9$, $r_{p,w}=0.2$ where $w$ denotes a side wall. Other values have been studied, but the qualitative behavior is unchanged. The main effect of increasing(decreasing) $r_{\alpha,\beta}$ is to increase(decrease) the granular temperature which in general merely moves the onset of segregation to slightly higher(lower) compacity values.

The parameters then are listed in Table I.

### B. Simulation method

We have simulated the numerical model via an event driven code [15]. The process is as follows:

1. For each particle, predict when it will next collide.
2. Identify the first collision to occur.
3. Move the particle(s) involved so that they touch.
4. Update the velocities of the particle(s) involved (change in velocity due to damping and noise since they were last updated).
5. Collide the particle(s) [17].
6. Repredict the next collision(s) of the particle(s) and their neighbors.
7. Repeat from 2 until time has advanced by $t_{\text{minupdate}}/2$.
8. Update all particles that have not been updated in the last $t_{\text{minupdate}}/2$ s.
9. Repeat from 1 until time has advanced by $t_{\text{takedata}}$.
10. Record data.
11. Repeat from (1).

The prediction of collisions assumes that the particle’s velocities do not change during their motion. The error caused by this is small provided that $t_{\text{minupdate}} \ll \min(t_{\alpha})$, where $t_{\alpha}=m_{\alpha}/\gamma_{\alpha}$ is the time constant of the velocity decay.

We set $t_{\text{minupdate}}=0.01 \times \min(t_{\alpha})$.

The hard sphere model with inelastic collisions can undergo inelastic collapse [18], where the particles undergo an infinite number of collisions in a finite time. Clearly a simulation that implements each collision will “stall” in these circumstances. One way around this unphysical singularity is the tc model [19] which prevents the collapse by setting $r_{\alpha,\beta}=1$ for any particle which collided within the last $t_{\text{colmin}}$ seconds. We found that for $t_{\text{colmin}}<1 \times 10^4$ the results did not change. For the results presented here we chose $t_{\text{colmin}}=5 \times 10^5$. The time between consecutive collisions of a particle follows a broad distribution which would diverge for small times were it not for the “cutoff” due to $t_{\text{colmin}}$. For times large compared to $t_{\text{colmin}}$ the distribution is close to exponential with a characteristic time of order $10^{-2}$ which depends on the compacity. For comparison, $\tau_r=3.9$ and $\tau_p=5.2 \times 10^{-2}$.

The initial conditions were created by running a reduced size system $N_c/25$, $N_p/25$, of twice the aspect ratio with particles initially placed randomly on a square lattice. An external force was applied to compress the particles into an area of size $L_x/5 \times L_y/5$. The particles were then allowed to move in this box without the external force until equilibrated. The initial condition was then created by tessellating the full system size with 25 replicas of the reduced system. To prevent long-range order the replicas were randomly inverted in both the $x$ and $y$ directions (the differing noises received by the particles in conjunction with the chaotic behavior of the particles would rapidly remove any correlations in any case). During this stage, both particle species had the same properties, including $r_{\alpha,\beta}=1$, except for their radii. This method produced initial conditions which appeared to be as homogeneous as those for the experiment.

The simulations were run on standard PCs. For comparison with the experiments our results are for the same system size (18 cm × 9 cm), except where stated otherwise. The aspect ratio is here 2:1 in all cases.

### III. RESULTS

As with the experiments, we used the number of ps, $N_p$, as our control parameter. The main quantities measured were those found experimentally [1,6,7] and related to the $c$’s only: the mean stripe width, $W$, and the local density, $\rho_i$, of the $i$th particle. We also visualized the system and watched its behavior. In addition, we also measured the area available to the $p$’s, and the kinetic energy or “granular temperature” of the particles.

$W$ measures the mean width of the stripes in the $x$ direction by pixellating an image of the system, deciding which pixels $P_{x,y}$ are within a $c$ domain, then running along each pixel row $P_x$ and counting the width and number of domains. All the rows $P_x$ are summed and the average domain width found. Whether $P_{x,y}$ is within a domain is decided by blurring the image with a Gaussian smoothing function and then setting a threshold. This measure was used in [6,7] even when the domains had not formed into stripes since it provided a simple measure of the domain sizes in the longitudinal direction.

The normalized local density $\rho_i$ was found by Voronoi tessellation [20] of the $c$’s, such that each $c_i$ has around it a polygonal area all points of which are closer to $c_i$ than to any other $c$. $\rho_i$ is then the minimum possible area, which is $2 \sqrt{3} R^2_i$, divided by the polygonal area. Polygonal cells on the edge of the system which are not bounded are discarded.

In addition to $W$ as a measure of the amount of coarsening, we also measured the area available to the $p$’s. This “available area” is just the fractional area of the system in which a $p$ could be placed without overlapping a $c$. Each $c$ has a circular “excluded area” around it of radius $R_{c}+R_{p}$, inside which the center of a $p$ cannot be placed. If all $N_c c$’s...
are widely separated, the available area is \(1 - N_c \pi (R_c + R_p)^2 / L_x L_y\), while if all the \(c\)'s are hexagonally close-packed in one domain the available area will be larger \(\approx 1 - N_c 2\sqrt{3} R_c^2 / L_x L_y\) since the excluded areas now overlap. Thus the available area gives a measure of how segregated the \(c\)'s are. In Sec. IV B we will show that the available area is relevant to noise segregation.

The parameters used are given in Table I, the system sizes used are the experimental size, \(L_x = 0.18\) m, \(L_y = 0.9\) m, or 1/4 of the experimental area, \(L_x = 0.9\) m, \(L_y = 0.45\) m.

Figure 1 shows 6 images of the system at differing compacities for late times, the lowest compacity \((C=0.446)\) shows a binary mixture, the second \((C=0.582)\) a segregated “liquid” which has mobile and transient clusters and shows a slight anisotropy, and the third \((C=0.717)\) shows a system that has coarsened into stripes. The first two are in a steady state while the third is still evolving slowly due to the high compacity (for a strictly two-dimensional system) which causes particles to be “blocked” by other particles.

Figure 3 shows \(W_c(t)\) for several compacities. As \(C\) increases, the \(c\)'s, which were initially mixed with the \(p\)'s, coarsen into domains whose size increases with \(C\). The early-time coarsening is rapid, followed by slower coarsening and then saturation at some relatively steady value. For the lowest compacities, there is no coarsening, the system remains in a mixed, disordered state. The highest \(C\) value shown has not reached a final steady state, the stripes are still moving and merging at a very slow rate compared to the initial coarsening. At higher compacities, the system becomes blocked or jammed, the particles being unable to rearrange themselves in two dimensions, the slow time scales of the top curve show the onset of this jamming. This jamming does not occur to the same extent in the experimental system since the particles do not always form a monolayer and \(p\)'s can move out of the way of \(c\)'s by “riding up” on top of each other.

Figure 4 shows a plot of the value of the area available to the \(p\)'s against time, it is similar to that for \(W\). It is a more repeatable measure than \(W\) since it is less affected by stripes.
merging, for example, two runs with the same parameters naturally differ due to the chaotic behavior of the particles. This means that for high compacities merging of stripes in two realisations at the same parameters may occur at different times, causing the $W_C(t)$ curves to differ between runs at late times [since $W_C(t)$ is inversely proportional to the number of domains]. The available area is less affected by this and thus provides a “cleaner” measure of the coarsening. Thus for later results we will use the available area, although $W$ provides similar, if more noisy, results. Experimentally, the available area has not been measured since the experimental system was only imaged in its central region, thus the number of $c$’s changes with time making the measure somewhat arbitrary (the changes due to $c$’s entering or leaving the system would be more significant than the coarsening). We

FIG. 2. Examples of a binary mixture, segregated liquid, and segregated stripes. Compacities are 0.446, 0.582, and 0.717, respectively. $p$’s are colored black. The pictures were taken after 100 s for the gas and 200 s for the liquid and crystal, by which time the first two have reached a steady state while the third is still slowly evolving.

![Image](a)

![Image](b)

![Image](c)

FIG. 3. Plots of $W$ against time for various compacities. From the right, top to bottom, the compacities decrease from $C=0.717$ to $C=0.401$ in uniform increments. Thus it can be seen that for low $C$ the system does not coarsen while when $C$ is increased, the system coarsens to a roughly constant value which increases with $C$. For the largest value of $C$, the system has undergone rapid initial coarsening but then slowed as the large domains move more slowly, especially at this high compacity where the system is becoming somewhat “jammed.” $W$ was measured every 0.0209 s.

![Image](High C)

![Image](Low C)

FIG. 4. Plots of the (fractional) available area for the $p$’s against time for various compacities. From the right, top to bottom, the compacities decrease from $C=0.717$ to $C=0.401$ in uniform increments. The conclusions are similar to those for Fig. 3 except that the curves here always remain in the same order. The available area was measured every 0.838 s.
were found by fitting exponentials to the available-area curve. This was done by fitting exponentials to the time series of the type shown in Figs. 3 and 4 and using the late-time values. We do not claim that the time series are exponential, but the fits provide us with a reasonable measure of the late-time values. The simulation value of sat for the experimental results, as reported in Ref. [7], note that for the highest compacity curve, the area available has increased by \( \approx 0.15 \) while the maximum range of the available area stated before is \( \approx 0.30 \).

In Figs. 5 and 6 we plot the late time values of \( W \) and available area as a function of the compacity. This was done by fitting exponentials to the time series of the type shown in Figs. 3 and 4 and using the late-time values. We do not claim that the time series are exponential, but the fits provide us with a reasonable measure of the late-time values. The \( W \) vs \( C \) curve is the same measure as that used experimentally in [7,21] to claim a mixed state to segregated state phase transition for the experimental system. In [7] a square root curve was fitted to the right-hand side of the data, the data to the left of the transition being taken to be roughly constant, i.e., \( W_{\text{sat}}(C) = B \) for \( C < C_{\text{transition}} \) and \( W_{\text{sat}}(C) = B + D(C - C_{\text{transition}}) \) for \( C > C_{\text{transition}} \), where \( B \) and \( D \) are constants. From Fig. 5 we see that this is not the case for the simulation data, which, as \( C \) is increased, initially rises slowly, then increasing rapidly before slowing again for large \( C \). There is no indication of a discontinuity in the gradient of the order parameter, merely a rapid increase. In [21] the same experimental data is presented with a sigmoid shaped “guide to the eye” curve rather than a square root. In our opinion the simulation data is similar to the experimental data but not to the square root form suggested in [7]. The main conclusion from both the experimental and simulation data is that \( W_{\text{sat}}(C) \) has a roughly sigmoid shape, with \( W_{\text{sat}}(C) \) increasing rapidly with \( C \) in the central region.

The simulation value of \( C \) at which the rapid increase occurs \(( \approx 0.58 \) differs somewhat from the experimental value \(( 0.647 \pm 0.049) \) [7]. In the next section we show how this value changes continuously as we vary the noise strength or other parameters. Thus, by increasing the noise strength we can increase the value of \( C \) at which the rapid increase occurs.

Figure 7 shows histograms of the local Voronoi density for various compacities, the results are qualitatively similar to the experimentally reported ones [6]. At the lowest \( C \) values, corresponding to the mixed state, the distribution is peaked at low densities as one would expect for unclustered \( c \)’s. At high \( C \) the distribution is peaked at large densities as one would expect for clustered (i.e., segregated) \( c \)’s. There is a crossover between these two cases, with Fig. 7(d) showing a broad histogram due to almost the full range of densities being present in almost equal weights. Unlike the experimental results however, the high \( C \) distribution also has a peak at low density caused by a small fraction of isolated \( c \)’s which are not present in the experiment. There is a crossover between these two extremes as \( C \) is increased, the central 4 figures (b–e) show this crossover. Following [6], we plot the location of the peak(s) and their widths as a function of compacity in Fig. 8. In [6] the peak width used was the full width 3/4 maximum since the peak did not extend far enough above the rest of the distribution for a full width 1/2 maximum to be meaningful. Here we did the same although a full width 3/4 maximum also does not exist in one case. Figure 8 is qualitatively different from its experimental equivalent and we conclude that although the histograms are qualitatively similar, the results derived from them are not.

The results presented here show that our numerical model reproduces the segregation and its qualitative behavior with \( C \) of binary gas, segregated liquid, and segregated stripes as seen experimentally. This behavior is not immediately obvious from the model rules—it emerged from the set of rules which we believed contained the important microscopic features of the experiment. We have also shown that \( W \) as a function of time and its saturated value as a function of compacity behave in a qualitatively similar manner to the experiment. These reproductions of experimentally observed behavior lead us to conclude that the “necessary features” listed

![Figure 5](image1.png)

**FIG. 5.** Plot of the saturated (late time) mean stripe width as a function of compacity. The saturated values were found by fitting exponentials to the \( W(t) \) curves. Two runs were done at each compacity, and the results averaged. The error bars are based on the difference between these two runs. Inset: the fitted curve of \( W \) against compacity for the experimental results, as reported in Ref. [7].

![Figure 6](image2.png)

**FIG. 6.** Plot of the saturated (i.e., late time) (fractional) area available to the \( p \)’s as a function of compacity. The saturated values were found by fitting exponentials to the available-area \( (t) \) curves. Two runs were done at each compacity, and the results averaged. The error bars are based on the difference between these two runs. Note that the error bars are smaller than for the \( W \) vs \( C \) plot since, as stated in the text, the available area is a “cleaner” measure.
in Sec. II capture the essential behavior of the system. Nonetheless, our phenomenological numerical model does not quantitatively reproduce the experiment (it was never expected to do so), in particular the Voronoi histograms differ from the experimental ones sufficiently that the data derived from them (Fig. 8) is qualitatively different.

IV. SEGREGATION MECHANISMS

Having shown that our numerical model is relevant to the experiment, we now use it to investigate segregation mechanisms.

A. Segregation due to oscillatory driving

That the domains are anisotropic and indeed form stripes at high $C$ indicates that the anisotropy of the driving is significant. Experimentally the stripes form perpendicular to the driving even for aspect ratios greater than 1, e.g., $L_y/L_x=2$. The side-to-side driving causes the two species to move at different rates due to their differing masses and friction coefficients. A single particle will oscillate with

$$k x = \frac{A}{\gamma} \frac{1 + \omega^2 (m_a/\gamma a)^2 \sin(\omega t + \phi)}$$

where the average is over the noise and $\phi$ is a phase shift relative to the driving. Thus $\langle x_c(t) \rangle = 6.0 \times 10^{-6} \sin(\omega t)$ and $\langle x_p(t) \rangle = 4.3 \times 10^{-5} \sin(\omega t)$. Thus the $p$’s would “like” to move a distance of order their radius during a cycle while the $c$’s hardly move. Consider a state of only $p$’s, if we remove the noise then all the $p$’s move in the same sinusoidal way, if we transfer to the non-inertial reference frame in which they are at rest, we see that this is identical to the state with no sinusoidal driving for which the dissipation causes the particles to be stationary. The same can be said of a state of only $c$’s. In a mixed state, however, the $p$’s will collide with the $c$’s and the system will “scatter” into a different state. Stable states, i.e., those that do not undergo further scatterings, will be those for which the $c$’s are separated from the $p$’s in the $x$ direction by distances of at least the

![FIG. 7. Normalized histograms of the Voronoi densities plotted against normalised density, for $C=0.401(a), 0.559(b), 0.582(c), 0.604(d), 0.627(e)$, and $0.740(f)$. Note the crossover with increasing $C$ from a single peak on the left to two peaks, then a larger peak on the right. For each compacity the data was measured in the steady state from 45 frames of 1600 $c$’s each.](image)

![FIG. 8. Plots of data derived from the Voronoi histograms as shown in Fig. 7. Squares show the position of the peak(s) in the histograms, representing the “most probable” densities, the lower curve is for the leftmost peak and the upper curve is for the rightmost peak. Filled circles show the full width of each peak at 3/4 maximum where the peak does have a distinct 3/4 maximum). The curves differ from their experimental equivalents where there was only one maximum (thus just two single curves) which moved steadily to the right and where the width was peaked at an intermediate compacity. These differences are due mainly to the presence in the simulation density histograms of a second peak at low density which is always present and is caused by a small number of isolated $c$’s, something not observed in the experiment.](image)

041301-7
amplitude of the p’s oscillations. For the packing fractions considered here, this can only happen by the two species segregating into domains. The fact that the area available to the p’s increases with time is consistent with this interpretation. This argument only holds when the driving is not too large compared to the dissipation. For example if the amplitude of oscillation were of order the system size then it would not be valid. An argument similar to this was also given in [12].

We have studied systems with very low noise and the results obtained agree with the heuristic argument given above. The addition of noise, which causes particles to diffuse, will tend to cause mixing. Thus there is a competition between the periodic driving which causes segregation and the noise which prevents it. This is shown in the results presented below.

For the standard parameters but with $\sigma_\alpha \rightarrow 0.1 \sigma_\alpha$, the system segregates for all the compacities studied in Sec. III (see Fig. 9), indicating that, as expected, the noise acts to prevent segregation. To confirm this we then gave the two species identical parameters (the p parameters) and set all $r_{\alpha,\beta}=1$. The only difference was that the p’s experienced the periodic driving term $\gamma_{\alpha,\beta}$, while the c’s did not. The results are shown in Fig. 10 and demonstrate that a difference in the periodic driving alone can cause segregation.

For $C=0.498$ with the same parameters except $\gamma_\alpha \rightarrow 2 \gamma_\alpha$ (to reduce the granular “temperature” which is higher than usual due to the $r=1$) we varied $\sigma$ and, as shown in Fig. 11, the system segregates for low $\sigma$ and remains mixed for high $\sigma$.

These results confirm that the state of the system is a result of a competition between the periodic driving which causes segregation and the noise or “granular temperature” which prevents it. Varying our control parameter, $N_p$, causes us to go from a mixed to a segregated state because increasing $N_p$ both reduces the granular temperature (since it increases the number of collisions which are highly inelastic) and also increases the “pressure” that the c’s feel due to collisions with a greater number of oscillating p’s. Figure 12 shows results for all properties set to p values including all
r=0.1 but with only the p’s feeling the periodic driving, showing that the results remain of the same form for r<1.

Finally, we find that the system with standard parameters does not segregate if the oscillatory motion is turned off. These results show that the differential oscillatory driving can cause segregation and present good evidence that it is the responsible mechanism in the system in conjunction with the noise which acts to prevent segregation. The transition from a mixed to a segregated state as C is varied is due to the compacity changing the relative strengths of these two competing effects.

B. Noise segregation

It was suggested in [6,7] that the segregation mechanism might be similar to the depletion interaction in equilibrium binary systems [22]. As an example consider a colloidal suspension containing non-adsorbing polymers. In the ideal case where there are no forces present, the free energy depends only on the entropy of the system. Treating the polymers as spheres [23], it is clear that each colloidal particle has an “excluded volume” around it of radius $R_{\text{colloid}}+R_{\text{sphere}}$ which the center of the polymer cannot enter. Thus the volume available to the polymers is the volume of the system less the excluded volumes around the colloidal particles and the system edges. However, the excluded volumes overlap when colloidal particles are closer than $2(R_{\text{colloid}}+R_{\text{sphere}})$, thus the volume available to the polymers, and hence their entropy, is larger if the colloidal particles are close to each other. This entropic “effective potential” can be large enough to cause the colloid to coagulate. This mechanism was one reason for measuring the area available to the p’s in our simulations. Segregation has been observed in simulations of hard spheres of two different sizes [24] for large size ratios, e.g., $R_1/R_2=10$, and also experimentally in binary mixtures of hard sphere colloids [25].

FIG. 12. Plot of the area (fractional) available to the p’s as a function of compacity, the parameters are as for Fig. 10 except that here the coefficients of restitution are all 0.1 rather than 1. It can be seen that this does not qualitatively change the results. Note that the area of the system is 1/4 that of the experimental system.

FIG. 13. Segregation of c’s for the standard parameters but with no periodic driving and $\sigma_c=0.05\sigma_x$. Notice the domains of c’s at the edges of the system. Note that the area of the system is 1/4 that of the experimental system. p’s are colored black.

We may equivalently view the entropy argument from the kinetic point of view. Two particles which are close to each other such that no third particle may fit in the space between them will feel a pressure on all sides due to collisions with other particles except on their neighboring sides. Thus the particles feel an effective attractive force. This pressure argument may be extended to systems that are not in equilibrium, for example the granular experiment studied here. The size ratio $R_c/R_p$ is much closer to unity than for simulated equilibrium segregating systems [24], implying that the difference in size alone does not cause segregation.

In our out-of-equilibrium system there are several possible differences between the two species besides a difference in size. For our system, it seemed possible that the lighter, faster moving p’s might, through the differential pressure mechanism, cause the c’s to coagulate even in the absence of periodic driving. As stated before, this was not observed for the standard parameters. We therefore increased the noise of the p’s and/or reduced the noise of the c’s in order to increase the p to c temperature and hence pressure ratio. Noticeable segregation occurred over a wide range of compacity values for $\sigma_p=10\sigma_x$ for $\sigma_p=2\sigma_x$ and $\sigma_c=0.1\sigma_x$, and also for $\sigma_c=0.05\sigma_x$. While the first of these cases is outside the reasonable parameter range, the second and third are at parameters which might be physical. Figure 13 shows an example for the third case. Notice that many of the c’s have coagulated at the walls as one would expect since the pressure argument for two-particle attraction also applies to the particle-wall case. Although this marked congregation at the walls is observed in experiments with colloids, it is not observed in the experiments of [6,7]. Reintroducing the periodic driving of the end walls prevented congregation at the $x=0,L_x$ walls whose large momentum transfer to the particles, as stated earlier, gives rise to low-density regions next to them. However, for all parameter values studied that displayed segregation with no driving, the c’s still congregated at the $y=0,L_y$ walls when the driving was turned on. Since stripes that touch the top and bottom walls are stable in the experiment, it seems unlikely that agitation due to the motion of the top and bottom walls is what prevents the liquid domains from coagulating there. It is possible to remove this experimentally unobserved effect by giving $\sigma_z$ and $\sigma_y$ differing values such that $\sigma_x$ is lower.
We have studied two species with standard addition to the differential driving discussed above. Pressure segregation may play a role in the experiment in task and at present all we can conclude is that differential to-side movement ratios. This is likely to prove a difficult mixtures in order to get readings for different noise to side-

We therefore conclude that this differential pressure segregation mechanism may play a role in the experiment. We had to “tune” the parameters which implies that the mechanism is less robust than the oscillatory driving mechanism. Accurate experimental measurements of the parameters would help to resolve whether this mechanism is indeed present. Directly distinguishing the two mechanisms mentioned would require accurate tracking of all particles and their collisions, coupled with investigations of other binary mixtures in order to get readings for different noise to side-

Directly distinguishing the two mechanisms mentioned would require accurate tracking of all particles and their collisions, coupled with investigations of other binary mixtures in order to get readings for different noise to side-

To demonstrate that different temperatures alone can cause segregation we have studied two species with standard \( p \) parameters but all \( r_{\alpha,\beta}=1 \). The temperature difference is produced by \( \sigma_p \rightarrow 40\sigma \) and also setting all \( \gamma \rightarrow 100\gamma \) so that the time constants \( \tau=m/\gamma \) are sufficiently small that particles remember the temperature of their heat baths rather than only the temperature of their previous collision partners. This imposed temperature difference causes coagulation of the lower temperature particles as shown in Fig. 15. While these parameters are very different from the standard ones, they clearly show another segregation mechanism in a nonequilibrium system and that the heuristic arguments regarding pressure differences out of equilibrium are valid.

C. Further mechanisms

One further possible mechanism is that the noise of the particles is correlated among the particles but differently for the \( c \)'s and the \( p \)'s. This could cause segregation in the same way as the periodic driving since it also would produce different collective motions for the two species. The correlation could be caused by, for example, the \( p \)'s all changing from sticking to the tray base to slipping at the same time in the periodic cycle. While this mechanism is at least plausible, the collective motions due to correlated noise and due to periodic driving (which would have caused the correlated noise in the first place) would not be clearly distinguishable.

At the level of this phenomenological numerical model which breaks the driving into a periodic and a noise component, any such mechanism is, therefore, not meaningful.

It is possible that there are other segregation mechanisms not discussed here, however we believe that we have considered the ones most likely to be relevant in the experiment.

V. CONCLUSION

In this paper we have introduced and numerically studied a (relatively) simple phenomenological numerical model of a recently reported granular segregation experiment. We have measured the same quantities as measured experimentally and shown that our numerical model reproduces most of the features of the experiment, the most important being a transition from a mixed to a segregated state as the compacity is increased. This behavior is not \textit{a priori} built into the numerical model—it emerges from the simple rules governing the motion of the particles. This is significant as it shows that we have a set of basic features necessary for an explanation of the experimentally observed behavior.

We then used our numerical model to investigate and identify segregation mechanisms and elucidate the experimental behavior. We showed that the transition from mixed to segregated state in the numerical model is caused by competition between the different driving felt by the \( c \)'s and \( p \)'s, which acts to cause segregation, and the noise, which acts to prevent segregation. We are led to conclude that this is also the main mechanism present in the experiment. We have also considered and demonstrated segregation due to different pressures and shown that it is possible that this might play a role in the experiment. The differential driving segregation mechanism is applicable to many binary driven systems [11,13].

FIG. 14. Segregation of \( c \)'s for standard parameters but with \( \sigma_{cx} \rightarrow 0.05\sigma_{cx} \) and \( \sigma_{px} \rightarrow 2\sigma_{px} \) and no driving of the tray base. Thus we get segregation but without coagulation at the top and bottom walls. The left and right walls are oscillating and prevent coagulation there. Note that the area of the system is 1/4 that of the experimental system. \( p \)'s are colored black.

FIG. 15. Coagulation of \( c \)'s after 96 s due only to temperature difference \( T_p=1600T_c \). The system is still evolving, the groups of \( c \)'s in the center will eventually attach to the sides of the system. Note that the area of the system is 1/4 that of the experimental system \( p \)'s are colored black.
SEGREGATION MECHANISMS IN A NUMERICAL MODEL...

This work goes some way to explaining the intriguing experimental results of [1,6,7]; Experiments with other particle types in conjunction with more accurate experimental measurements, particularly with regard to particle positions and velocities, should allow more accurate comparisons with our numerical model and also refinements of the model. In particular more accurate values for the parameters used.

Now that we have shown our numerical model to be relevant to the experimental system, it is possible to use it further to investigate the “granular statistical mechanics” of this type of system. In particular, it may be of use in developing and testing theories for agitated granular mixtures before attempting the more difficult task of accurately comparing with experiments.

Note added in proof: Recently, we learned of work by Pooley and Yeomans [26] in which, analytically, they show stripe formation for binary fluids where one fluid is driven by an external periodic force, and suggest as here that the same mechanism causes stripe formation in horizontally driven binary granular mixtures.

ACKNOWLEDGMENTS

G.C.M.A.E. thanks Timo Aspelmeier and Sam Carr for useful discussions. P.M.R. thanks Tom Mullin for many stimulating conversations. P.M.R. was supported by a doctoral scholarship from the Portuguese Foundation of Science and Technology.

[8] The two-dimensional projected areas of the particles follow to a reasonable approximation a Gaussian distribution with a standard deviation of 17% of the mean.
[16] 2500 ps were measured to calculate the mean radius.
[17] In the uncommon event that the particles are no longer moving together after the velocity update, they are not collided.