

How Particle Shape Affect Granular Segregation in Industrial and Geophysical Flows

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How particle shape affects granular segregation in industrial and geophysical flows

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Industrial and environmental granular flows commonly exhibit a phenomenon known as “granular segregation,” in which grains separate according to physical characteristics (size, shape, density), interfering with industrial applications (cement mixing, medicine and food production) and fundamentally altering the behavior of geophysical flows (landslides, debris flows, pyroclastic flows, riverbeds). While size-induced segregation has been well studied, the role of grain shape is not well understood. Here we conduct numerical experiments to investigate how grain shape affects granular segregation due to grain-grain and grain-fluid interactions. To isolate the former, we compare dry, bidisperse mixtures of spheres alone with mixtures of spheres and cubes in a rotating drum. Results show that while segregation generally increases with particle size ratio, the presence of cubes decreases segregation levels compared to cases with only spheres. Further, we find hysteresis in segregation trends with size ratio; segregation is lowest when the small grains are cubic as they approach a jammed state with reduced mobility. We find similar hysteresis in simulations of a shear-driven coupled fluid-granular flow (e.g., a riverbed), demonstrating that this phenomenon is not unique to rotating drums; however, in contrast to the dry system, we find that total segregation increases in the presence of cubic grains, and fluid drag effects can qualitatively change segregation trends. Our findings demonstrate competing shape-induced segregation patterns in wet and dry flows—independently from grain size controls—with implications for many industrial and geophysical processes.

segregation | brazil nut effect | armoring | rivers | shape

Granular materials are commonly found in our daily lives in a multitude of industrial applications (e.g., cement, pharmaceuticals, food grains) (Figure 1a,b) and in nature (e.g., rocks, sand, snow, soil). Because these materials can behave as solids, liquids or gases under the influence of external forces, they have no single constitutive equation and we have yet to gain a complete understanding of their complex behavior (1, 2). Further, mixtures of granular materials commonly exhibit an emergent phenomenon known as “segregation” in which grains of different size, shape, density and roughness self-organize to prevent uniform mixing (3–5). One of the most common examples of granular segregation is the “brazil nut effect,” which occurs when smaller grains fill in voids beneath the large grains when disturbed, causing large grains to migrate toward the surface over time (Figure 1c) (6). You have likely experienced this when eating a jar of nuts or pouring cereal into a bowl. Granular segregation can be a severe nuisance, interfering with a variety of mixing processes in the cement, food and pharmaceutical industries (3).

Granular segregation is also pervasive in nature, where sediment grain size ranges from very fine silt to massive boulders (7). Geophysical flows such as debris flows (Figure 1c)

(8), landslides (9), pyroclastic flows (10), and slow-moving, lobate arctic soil patterns (11) exhibit strong segregation, in which large boulders tend to organize at the front of the flow, increasing runout distance and destructive potential (3, 12). Segregation also occurs for granular beds driven by shear flows, such as wind-blown or subaqueous ripples and dunes (13, 14), beaches (15), and riverbeds where large grains can armor the surface and influence erosion rates and sediment transport dynamics (16, 17). These processes are ubiquitous not only on Earth but on other planetary bodies, including asteroids (18) and any planet or moon with a granular surface (Figure 1d)(19–21).

While granular segregation for the simplified case of spherical grains has been extensively studied (22–24), our ability to predict and control its effect in industrial or natural settings is limited; complex interactions between size, density, frictional, shape effects and disturbance rate can lead to unexpected outcomes (3, 25). One of the least explored aspects of segregation is the role of shape, though the presence of non-spherical grains is ubiquitous in most industrial and natural flows (26). Some previous studies have examined the role of grain shape in controlling rotating drum segregation patterns, showing that the presence of angular shapes can dissipate more rotational energy, affecting how they interact with the wall and with each other (5, 26–28). Grain shape has been shown to alter mobility in a variety of flow regimes, with sharp edges of cubes dissipating energy faster than spheres and decreasing mobility (28–30). However, findings from these studies are often seemingly conflicting and have been difficult to synthesize because

Significance Statement

Granular materials like cereal, pharmaceuticals, sand and concrete commonly organize such that grains segregate according to size rather than uniformly mixing. For example, in a jar of nuts, the largest ones are commonly found at the top. Here, we use computer simulations to explore how grain shape controls this phenomenon in industrial and natural settings. We find that even small differences in shape can substantially change the amount and style of segregation, with different effects depending on whether the system is wet or dry. This study demonstrates the importance of grain shape in different systems ranging from food and medicine production to geophysical hazards and processes such as landslides, river erosion, and debris flows on Earth and other celestial bodies.

F.D.C. performed research; F.D.C. and R.C.G. designed research; F.D.C., D.P. and R.C.G. analyzed data, and wrote the paper.

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Fig. 1. Processes of granular segregation. (a) Brazil nut effect in a jar of nuts. (b) Process of mixing cement. Inset: Granular mixture in a rotary drum composed by marbles with diameters of $d_s = 4$ and $d_b = 8$ mm. (c) Granular segregation in the front of the Illgraben debris flow. Photo by Pierre Zufferey. Image credit: American Geophysical Union. (d) Debris flow deposit-terminations in Kepler moon crater (latitude 8.32°N , longitude 37.69°W)(20). (1) Finer-grained fractions (fines), (2) coarse dark levees, and (3) terminal deposits.

52 it is nontrivial to disentangle the role of shape and size, and
 53 different filling levels and rotational speeds used in different
 54 studies can result in complex, unpredictable radial segregation
 55 patterns that are challenging to compare (4, 29–34). Only
 56 recently has a universal rule been proposed for segregation
 57 with different shapes; (5, 26) found that segregation levels for
 58 bidisperse grains (disks, rods, spheres) in a numerical model
 59 depends largely on the volume ratio between the two species.
 60 According to their results, segregation intensity increases loga-
 61 rithmically with volume ratio, and grains with equal volume
 62 exhibit zero segregation. This promising work demonstrates
 63 that differences in grain volume can account for shape effects
 64 on segregation; however, their results show that segregation
 65 levels can still vary substantially for different shapes even
 66 within the same volume ratio. Many other questions remain,
 67 including the effect of angular shapes, shapes that exhibit
 68 negative curvature, and the presence of a fluid.

69 Here we use numerical models building in complexity to
 70 explore the role of grain shape in controlling granular seg-
 71 regation. First, we examine a partially-filled rotating drum
 72 filled with dry, bidisperse grains (spheres and cubes) at a
 73 low rotational velocity. We choose this setup because it is
 74 relevant not only for industrial mixing applications, but also

for geophysical flows such as debris flows and landslides. We
 choose to compare spheres with cubes because they are not
 too dissimilar in shape; thus our findings may demonstrate
 how even mild shape differences control segregation, leaving
 more extreme shapes (long rods, stars, etc.) to future studies.
 We explicitly control for grain size by comparing results for
 bidisperse spheres alone with results for bidisperse mixtures of
 spheres and cubes and find that the presence of cubic grains
 not only changes segregation levels, but leads to hysteresis in
 segregation as small cubic grains approaching a jammed state.
 Next, we test numerically whether this finding applies in an
 entirely different system in which fluid shear drives motion
 over a granular bed (e.g., a riverbed). While we find similar
 hysteresis, results show that the presence of fluid drag can
 qualitatively alter segregation trends, resulting in 1) larger
 segregation levels in runs with cubic grains for all cases and
 2) inverse segregation in which smaller cubes organize at the
 bed surface. Our work shows that grain shape can exert a
 fundamental control on segregation, both quantitatively and
 qualitatively, in industrial and geophysical flows. These find-
 ings demonstrate the need for more attention on grain shape
 to understand granular dynamics, with implications for efforts
 to control granular segregation in industry, predict the behav-

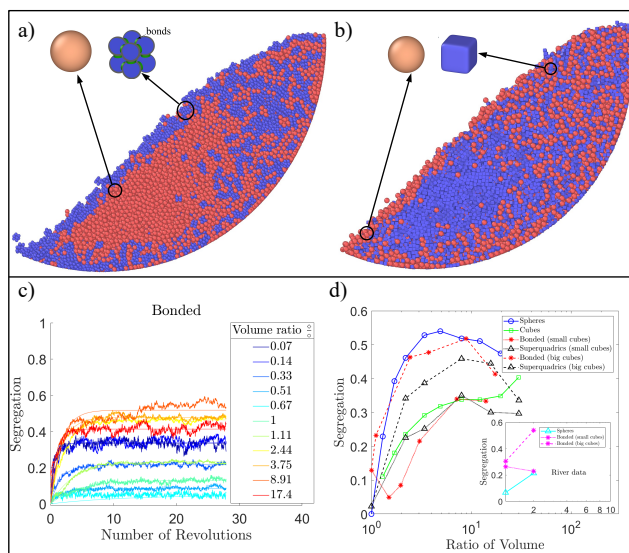
98 ior of destructive geophysical flows, and understand sediment
 99 dynamics in rivers and windblown dunes that are pervasive
 100 on Earth and other planets.

101 Grain shape controls on segregation in a dry rotating 102 drum

103 To isolate the purely granular effects of shape while controlling
 104 for size differences, we run dry, bidisperse models in a rotating
 105 drum with varying volume ratio for cases with 1) only spheres
 106 or only cubes with varying volume ratio ($1.3 \leq V_l/V_s \leq 30$),
 107 where V_l and V_s are the volumes of each particle for the large
 108 and small species, respectively; 2) mixtures of spheres and
 109 cubes varying the volume ratio ($0.03 \leq V_{\square}/V_{\circ} \leq 30$), where
 110 V_{\square} and V_{\circ} are the volumes of each particle for the cubical and
 111 spherical species, respectively. By examining differences in
 112 segregation between these cases for the same volume ratios,
 113 we can truly isolate the effects of shape.

114 We use the open-source code LIGGGHTS, which is based on
 115 the Discrete Element Method (DEM), to compute granular dynam-
 116 ics. While LIGGGHTS was originally designed to simulate
 117 spherical grains, we take advantage of two recently developed
 118 capabilities to simulate cubic grains: bonded spheres (Figure
 119 2a) and superquadrics (Figure 2b). Superquadrics allow
 120 simulations of near-realistic shapes such as rods, ellipsoids and
 121 more angular shapes such as cubes (albeit with slightly rounded
 122 edges). However, state of the art coupled fluid-granular models
 123 are not yet able to simulated superquadrics because fluid drag
 124 formulations only work for groups of spherical grains (35, 36).
 125 Therefore, we also use bonded spheres to create lumpy cubic
 126 grains of various sizes (hereafter referred to as “bonded cubes”)
 127 in order to test whether this approach can be a good approx-
 128 imation for real shapes in fluid simulations. These bonded
 129 cubes also allow us to explore effects of grain shapes with
 130 negative curvature (Figure 2a) (37). We calculate the total
 131 volume of the bonded cubes as the total volume of the bonded
 132 spheres, plus the volume of the void space in the middle of the
 133 grain. We slightly increase the density of each bonded spheres
 134 to account for this void space, allowing for equal effective
 135 density of bonded cubic grains and other grains (see Methods).
 136 Cubes and spheres are initially randomly distributed within
 137 the drum at equal volumes between the two species, with a
 138 packing fraction of around 30%, and the drum is driven at a
 139 low rotational speed representative of a variety of industrial
 140 and natural flows. For all cases, we calculate the final amount
 141 of segregation once the system has reached a quasi-steady
 142 state (Figure 2c) and the time that the mixtures take to reach
 143 it (See Methods). Segregation is calculated such that $S=0$
 144 represented a completely mixed system (equal proportions
 145 of both grains), and $S=1$ represents a completely segregated
 146 system (only a single type of grain present) (see Methods). We
 147 validated segregation calculations by computing the amount
 148 of segregation in a simulation where both species were equal
 149 sized spheres, finding that the amount of segregation through
 150 time was zero (Supplemental Material). We also validated the
 151 model setup by comparing results with a physical experiment
 152 in a rotating drum, using the same rotation rate and marbles
 153 of the same size (Figure 1b; Supplemental Material).

154 Our results illuminate the importance of both grain size
 155 and shape in controlling segregation, clearly demonstrating
 156 that shape alone can substantially affect segregation levels.
 157 We observe similar qualitative behavior for all runs; Figure 2c



158 **Fig. 2.** (a) Snapshot of particle positions of a mixture of spheres and cubic grains
 159 developed by bonded spheres ($V_{\square}/V_{\circ} = 3.75$). (b) Snapshot of particle positions of
 160 a mixture of spheres and cubic grains developed by superquadrics ($V_{\square}/V_{\circ} = 0.45$).
 161 (c) Temporal evolution of the amount of segregation for mixtures of spheres and cubic
 162 grains (bonded spheres). (d) Final amount of segregation as a function of the ratio of
 163 volume for rotary drum cases. Inset: Final amount of segregation for fluid-sheared
 164 granular beds.

165 shows the evolution of the amount of segregation for mixtures
 166 of spheres and bonded particles, where cooler colors represent
 167 to small volume ratios and warmer colors to large volume ratios.
 168 In all cases the amount of segregation starts at zero, where
 169 the particles are randomly distributed and then increases until
 170 it reaches a final steady state (see Supplemental Materials
 171 for other time series). In agreement with previous studies
 172 (5, 26, 38), final segregation for all cases tends to increase
 173 with volume ratio for ratios up to about 12 (Figure 2c,d). Large
 174 grains, regardless of shape, tend to migrate toward the surface
 175 and walls of the drum as the brazil nut effect. However,
 176 once volume ratios are large enough, it is not possible to
 177 define well segregated regions in the mixture. This is due
 178 to the onset of a segregation inversion in which large grains
 179 begin to accumulate at the center of the drum (Supplemental
 180 Materials). This result agrees with previous studies that found
 181 inverse segregation for large size ratios, where depending on
 182 the roughness of the walls and the weight of the grains, large
 183 grains may condensate in the center of the drum (30, 39, 40).

184 The effect of grain shape lies in substantial quantitative
 185 differences in segregation levels in all runs. The presence of
 186 cubic grains decreases segregation in most cases, except for the
 187 case of equal volume in which non-spherical grains produce
 188 slightly higher segregation levels, in contrast to previous find-
 189 ings (5, 26) (Figure 2d). Runs with superquadric cubes alone
 190 exhibit nearly half the segregation levels as spheres alone.
 However, the most surprising result is found for mixtures
 of cubes and spheres. We observe a hysteresis in segrega-
 tion trends, where segregation levels are lowest for cases in
 which cubes are smaller than spheres, and higher for cases
 in which cubes are larger than spheres for the same volume
 ratio; this occurs for both superquadric and bonded cubes
 (Figure 2d). Superquadrics exhibit lower segregation than

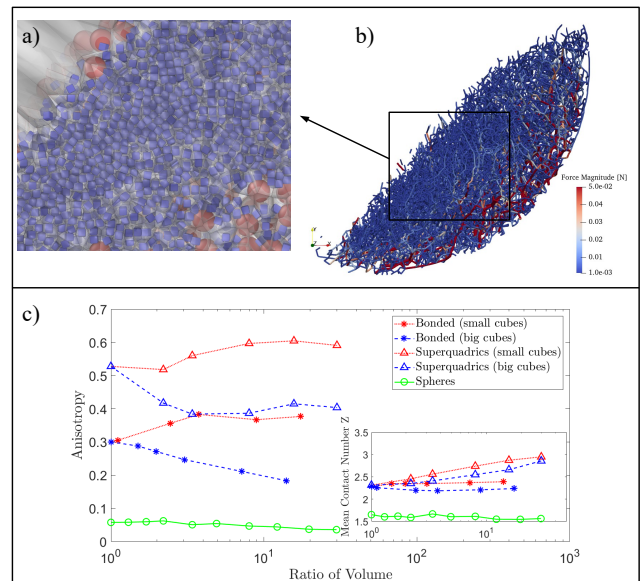
191 comparable bonded cube cases; while runs with large bonded
 192 cubes and small spheres are nearly identical runs with spheres
 193 alone, runs with larger superquadric cubes experience less
 194 segregation. The lowest values of segregation occur for cases
 195 with small superquadric or bonded cubes mixed with spheres.

196 Anisotropy and the approach to shear jamming

197 Our results demonstrate clear shape controls on granular seg-
 198regation and previously unobserved hysteresis in segregation
 199 depending on grain shape. The shape of the smaller grains
 200 controls granular segregation behavior in the drum, with possi-
 201 ble implications for industrial applications where segregation
 202 is unwanted and our ability to understand and predict segre-
 203-gation processes in granular flows. But how can we explain
 204 the observed hysteresis and importance of the shape of small
 205 grains?

206 To better understand our results, we analyze snapshots of
 207 model runs at steady state segregation and dive into the small
 208 scale interactions between individual grains. By examining
 209 grain dynamics in zoomed in videos of model runs (Supple-
 210 mental Materials), we observe that cubic grains tend to align
 211 along their edges and inhibit mobility in the center of the
 212 drum. We hypothesize that this behavior can be associated
 213 with an approach to a shear jamming state, in which grains
 214 experience an increase in the number of contact points with
 215 other grains as well as anisotropy related to the direction of
 216 the applied stress, thus leading to structure between grains
 217 and reduced mobility (41). While jamming behavior is not
 218 often associated with granular segregation dynamics in the
 219 literature, it is intuitive that they should be related; if granu-
 220-lar segregation depends largely on percolation of small grains
 221 through a network of larger grains, then the mobility of the
 222 small grains depending on shape and size must matter. While
 223 most jamming has been observed in monodisperse systems, it
 224 has recently been observed in bidisperse mixtures (42); further,
 225 because segregation tends to separate grains based on physi-
 226-cal characteristics, through time grains tend to interact with
 227 other grains of the same size and shape. Previous studies of
 228 hopper flow have observed unexplained slowing of segregation
 229 when a critical concentration of small grains is reached, likely
 230 due to the onset of jamming (43, 44). Though the continuous
 231 fluid-like motions of grains in the rotating drum prevent a fully
 232 jammed state, we propose that decreases in grain mobility
 233 may be explained by an approach to this state.

234 In order to test this, we examine snapshots of the model at
 235 final segregation levels to see if they exhibit certain charac-
 236-teristics that are hallmarks of the approach to a shear jammed
 237 state: organized force chains, anisotropy, and an increase in
 238 the number of contact points (Figure 3) (41, 45). Figure 3a
 239 shows a voronoi tessellation of a model run with superquadric
 240 cubes ($V_{\square}/V_{\circ} = 0.45$), illustrating that cubes in the center of
 241 the drum tend to be tightly packed with minimal void space
 242 between neighbors. This is due to the preferential alignment
 243 of the cubes with the direction of shear making the orientation
 244 of the cubes uniform and letting them align face to face. A
 245 more random orientation of cubes would increase the effective
 246 space any individual cube takes up, preventing the cubes from
 247 packing together as densely. Figure 3b shows a snapshot of
 248 force chains throughout the drum; long force
 249 chains are strongly aligned in the direction of shear, with
 250 strong connections between cubic grains in the center of the



251 **Fig. 3.** (a) Zoom in of an instantaneous snapshot of particle positions of a mixture
 252 of spheres and cubic grains developed by superquadrics ($V_{\square}/V_{\circ} = 0.45$) and its
 253 voronoi tessellation. (b) Network of force chains for the same case ($V_{\square}/V_{\circ} = 0.45$).
 254 (c) Anisotropy as a function of the ratio of volume. Inset: Mean contact number per
 255 particle Z as a function of the ratio of volume.

256 drum. This differs from cases in which cubes are larger than
 257 spheres, where force chains are shorter and more randomly
 258 oriented (Supplemental Material). The alignment of force
 259 chains suggests anisotropy, another common hallmark of the
 260 approach to jamming, in which grains are preferentially aligned
 261 and therefore exhibit greater structure (and lower mobility)
 262 than randomly oriented grains (41, 46, 47). Fabric anisotropy
 263 refers to properties such as shear strength and dilatancy taking
 264 on different values along different directions due to the state
 265 of the granular material's microstructure, where microstruc-
 266-ture refers to the arrangement of particles, void spaces, and
 267 interparticle contacts (48). Though it could be characterised
 268 with the arrangement of particles and void spaces, this study
 269 characterises the microstructure with a fabric tensor based on
 270 inter-particle contacts due to forces being transmitted along
 271 these contacts, forming force chains (see Methods for the math-
 272-ematical formulation of the fabric tensor). Granular mixtures
 273 with fabric anisotropy may exhibit jamming along directions
 274 with enhanced shear strength and thus diminished mobility in
 275 those directions. We calculate fabric anisotropy for all model
 276 runs (49) (see Methods) and find that runs with small cubic
 277 grains exhibit higher anisotropy than the equivalent runs with
 278 large cubic grains (Figure 3c). All cases with cubes exhibit
 279 higher anisotropy than for spheres alone, and superquadric
 280 cubes exhibit higher anisotropy than bonded cubes. This
 281 can be explained due to the negative curvature of bonded
 282 cubes, such that they can fit together in a variety of ways,
 283 whereas superquadrics preferentially align face to face. The
 284 mean contact number for all cubic runs is also substantially
 higher than that of spherical runs (Figure 3c, inset). The
 mean contact number is higher for all cases involving cubes
 than for spheres; again, cases with small cubes experience a
 higher contact number than for large cubes, and superquadrics
 experience higher contact numbers than bonded cubes. This

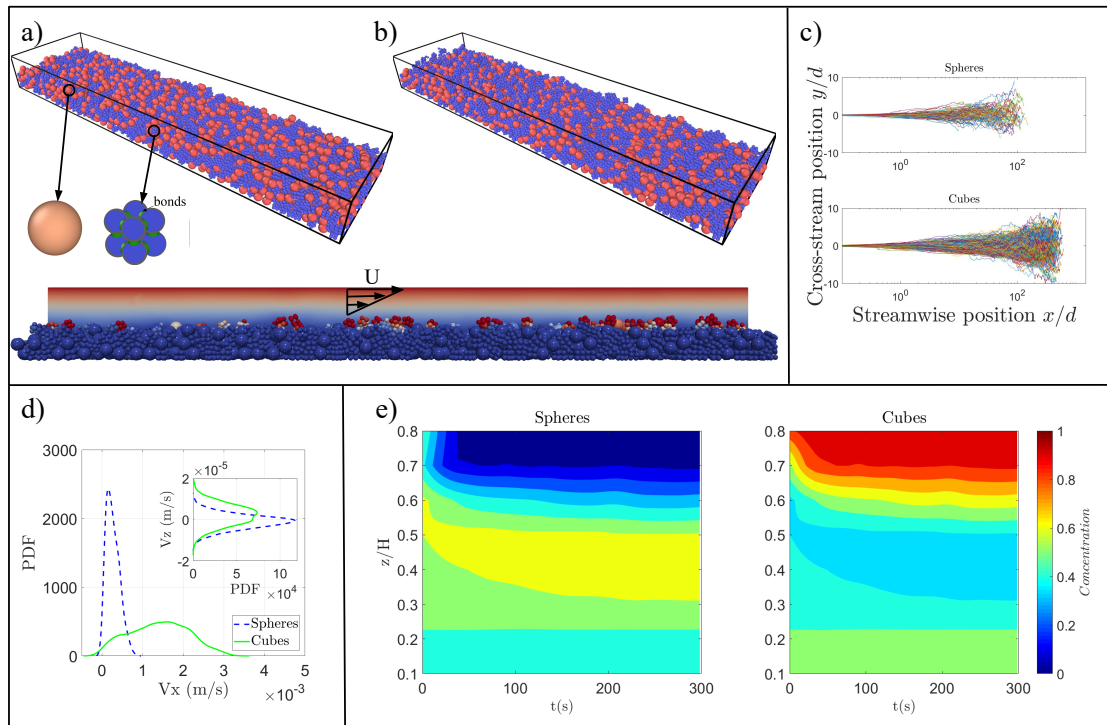


Fig. 4. River data. Snapshots of particle positions of a mixture of spheres and cubic grains developed by bonded spheres ($V_{\square}/V_{\circ} = 0.5$) at: (a) Initial condition ($t = 0$ s), and (b) final state ($t = 300$ s). (c) Particle trajectories. (d) Probability Density Functions (PDF) of downstream velocities and vertical velocities (inset) for each species of particles. (e) Temporal evolution of concentration of spheres and cubic grains as a function of the channel height.

analysis suggests that the presence of small cubes leads to an approach to shear jamming, causing cubes to align and experience reduced mobility in the center of the drum. This leads to a decrease in their ability to percolate, and therefore a decrease in segregation.

These findings demonstrate a new mechanism by which shape can control granular segregation levels, illustrating a previously unexplored link between shear jamming and segregation efficiency. Further, the hysteresis we have identified in segregation levels demonstrates the importance of the small size fraction in controlling segregation. This aligns with the prevailing hypothesis that the brazil nut effect is generally controlled by percolation of small grains; thus we anticipate that the shape of the smallest size fraction may be the dominant factor for shape-controlled segregation.

Our results also show that superquadratic and bonded cubes exhibit similar segregation behaviors. In the next section we use bonded cubes in a coupled fluid-granular simulation to show that shape-induced hysteresis in segregation also occurs in an entirely different system: shear flow over a granular bed (e.g., a riverbed).

Grain shape controls on a fluid-sheared granular bed

Fluid shear flow over granular beds sculpts planetary landscapes, as wind creates ripples and dunes and rivers transport sediment, carving mountain ranges and delivering nutrients to the ocean. Granular segregation is ubiquitous in these types of flows, especially in rivers or on beaches where large grains commonly armor the bed surface. This armoring can change

the morphology and dynamics of the flow, with implications for flooding, erosion and landscape evolution processes (15, 16). It is thought to occur due to a variety of processes, including preferential removal of fine grains due to sediment supply limitations (50) and granular segregation via the brazil nut effect as grains are disturbed by fluid near the surface of the bed (51, 52) and experience creep at slower rates deeper into the bed (17). However, most formulations of bedload transport in rivers assume spherical grains that do not represent natural sediment. Only recently has grain shape been shown to affect fluvial sediment transport via changes in fluid drag (53, 54) and interactions with the granular bed (53, 55). The role of grain shape in controlling granular segregation processes in natural fluid flows has been unexplored.

To begin to explore grain shape effects on segregation in natural systems, we run simulations of a Couette flow over a granular bed with bidisperse spheres and bonded cubes (Figure 4a), tracking segregation of the bed and grain velocities through time. We use the Coupled Fluid Dynamics/Discrete Element Method (CFDEM) modeling software, which couples the LIGGGHTS granular dynamics and OpenFOAM fluid dynamics models (56), to observe laminar flow over a granular bed in a rectangular channel with periodic boundary conditions in the streamwise direction. The flow velocity is set to be just above the threshold of motion ($\theta/\theta_{cr} \approx 1.5$) for the largest grains in the channel (see Methods). We choose to use laminar flow for simplicity, in order to focus on first order interactions between fluid and grains; while future studies may examine the role of turbulence characteristic in many natural flows, studies have shown that sediment transport in laminar flows

343 is fundamentally similar to that of turbulent flows (57).

344 Our results show that granular segregation driven by fluid
345 shear exhibits hysteresis similar to that seen in the dry rotating
346 drum (Figure 2d inset), suggesting that our findings are not
347 unique to that system. Runs with small cubes and large
348 spheres experience only a third of the segregation level seen
349 for runs with large cubes and small spheres. However, we
350 find that the effects of fluid-grain interactions can lead to
351 both quantitative and qualitative differences in segregation
352 trends. In contrast to the rotating drum case, the presence of
353 bonded cubes leads to higher segregation levels than spheres
354 alone (Figure 2d inset); further, bonded cubes always organize
355 at the top of the bed, even when they are smaller than the
356 spheres. This can be seen for the case shown in Figure 4,
357 in which $V_{\square}/V_{\circ} = 0.5$. Beginning from a fully mixed state
358 (Figure 4a), the smaller blue cubes preferentially organize at
359 the bed surface through time (Figure 4b,e). This demonstrates
360 a new shape-induced reverse brazil nut effect in natural flows
361 that may offset the armoring phenomenon.

362 Why do we observe qualitatively different segregation trends
363 in the presence of a fluid? By analyzing individual grain
364 velocities and distances, we find that cubes travel further
365 distances downstream than spheres over the course of the
366 model (Figure 4c). PDFs of grain velocities show that cubes
367 experience faster instantaneous downstream velocities than
368 spheres (albeit with much larger variability)(Figure 4d) and
369 upward directed vertical velocities, while spheres subtly tend
370 toward downward directed vertical velocities (Figure 4d inset).
371 We can better understand this behavior by examining the
372 concentration of each grain shape with respect to the total
373 number of grains in a series of layers at different depths in the
374 bed at the end of the model run (Figure 4e). At $t=0$, grains are
375 randomly mixed throughout the bed. As time progresses, they
376 experience rapid segregation in which cubes accumulate at the
377 bed surface (approx. where $z/H = 0.7$). A zone of low cube
378 concentration grows through time with depth just beneath the
379 surface; in contrast, spheres accumulate just beneath the bed
380 surface in a concentrated layer that grows in depth over time.

381 We interpret these results to illustrate the role of the fluid
382 in driving segregation patterns in a granular bed. Because
383 cubes experience a higher drag force than spheres (53, 58),
384 once they reach the surface they can move faster and are
385 more likely to continue moving. This likely prevents them
386 from settling back into the bed, decreasing their ability to
387 percolate and leading to higher segregation levels, causing
388 them to collect on the bed surface even if they are smaller
389 than the spheres. At depth, however, grain-grain interactions
390 dominate, causing spheres to migrate upwards and collect
391 just beneath the surface above which fluid effects take over
392 (see high concentrations of spheres at $z/H = 0.4-0.5$). These
393 findings point toward the need for further exploration of the
394 role of fluid effects in non-spherical granular flows and may
395 begin to explain enigmatic observations in riverbeds, where in
396 some situations large grains armor the surface, while in other
397 situations finer grains are found at the top (59, 60). Further
398 work is needed to determine whether our findings apply to
399 natural rivers, where dense sediment of many different shapes
400 are found in turbulent flows.

401 Discussion

402 Our findings show that grain shape cannot be ignored in
403 granular segregation processes, even when volume effects are
404 accounted for. Shape-induced segregation trends can vary
405 both quantitatively and qualitatively depending on competi-
406 tion between grain-grain and grain-fluid effects. In dry flows,
407 we observe hysteresis in which small cubic grains can experience
408 high anisotropy and contact numbers, approaching a shear jam-
409 ming state that reduces mobility and segregation, with possible
410 implications for industrial applications where segregation is a
411 nuisance. While we see similar hysteresis in fluid shear-driven
412 flows, cubic grains of any size instead increase segregation
413 levels compared to spheres alone; fluid-grain interactions can
414 even lead to qualitative shifts in behavior, producing a re-
415 verse brazil nut effect in which small cubes accumulate at
416 the surface. These results illuminate competing segregation
417 effects due to grain-grain and grain-fluid interactions, which
418 could lead to different qualitative behavior depending on the
419 volume fraction and inertial regime of different industrial and
420 geophysical flows.

421 Our methods demonstrate a way to isolate the role of
422 grain shape from size disparities by comparing results for the
423 same volume ratio with different shape combinations. Future
424 studies can use this approach to examine different shapes,
425 mixtures with more than two grain classes, and to see whether
426 our results hold for rotating drums with different rotation
427 rates and filling levels. Studies can also explore whether our
428 results can be harnessed in industrial applications to decrease
429 segregation in mixing processes by adding non-spherical grains
430 to mixtures. While our analysis suggests that small grains are
431 inherently important to segregation processes, further studies
432 could explore whether it is the size or abundance of cubic
433 grains that most strongly controls segregation; because we use
434 an equal total volume of each species in our models, small
435 grains are more abundant than large ones. The fact that
436 runs with large superquadric cubes exhibit lower segregation
437 rates than those with spheres alone illustrates that even small
438 numbers of cubes can have an effect on segregation dynamics.
439 It is possible that experiments with abundant large cubic
440 grains could experience shear jamming effects similar to those
441 we see for small cubes.

442 The link between the approach to shear jamming and seg-
443regation has implications not only for industry but also for
444 geophysical flows. A recent study demonstrated that debris
445 flow rheology is controlled by the solid volume fraction, and
446 therefore the distance to the jamming transition (61). An-
447 other recent study found that the temporal evolution of angular
448 grains in a pyroclastic flow determines flow rheology (62). In-
449 deed, changes in packing fraction known to affect rheology
450 have also been shown to result in qualitative shifts in segre-
451-gation trends (63). In light of these studies and our findings,
452 we suggest that grain shape exerts a fundamental control on
453 both the segregation and rheology—and therefore destructive
454 potential—of geophysical flows. While our fluid shear-driven
455 model applies to riverbeds, beaches, and possibly windblown
456 settings—examples of dilute suspensions where the volume of
457 moving sediment is low compared to the volume of the fluid
458 (2)—future work could explore whether similar competition
459 between shape-induced grain-grain and grain-fluid controls
460 on segregation applies in industrial and natural systems that
461 behave as dense suspensions (64), such as cement mixers (65),

debris flows and landslides (2). Further work could explore shape-induced granular segregation processes in non-inertial systems over longer timescales, such as hillslopes that evolve through slow soil creep or crystal segregation in magmas (66).

Materials and Methods

Model Description. In our numerical simulations for the purely granular effects, we used the open source code LIGGGHTS (67, 68) and its modified version that includes bond equations (35) to compute the interactions of each individual particle and the wall by solving the linear and angular momentum equations, given by Eqs. 1 and 2, respectively:

$$m \frac{d\vec{u}}{dt} = \sum_{i \neq j}^{N_c} \vec{F}_{c,ij} + \sum_i^{N_w} \vec{F}_{c,iw} + m\vec{g} \quad [1]$$

$$I \frac{d\vec{\omega}}{dt} = \sum_{i \neq j}^{N_c} \vec{T}_{c,ij} + \sum_i^{N_w} \vec{T}_{c,iw} \quad [2]$$

where \vec{g} is the acceleration of gravity and, for each solid particle, m is the mass, \vec{u} the velocity, I the moment of inertia, $\vec{\omega}$ the angular velocity, \vec{F}_c the resultant of contact forces, and \vec{T} the resultant of contact torques. The indices in F_c and T correspond to the collisions between particles i and j , and between particle i and the wall w .

To compute the contact forces between particles $\vec{F}_{c,ij}$ and between particles and the rotational wall $\vec{F}_{c,iw}$, we use the Hertzian contact theory (69) which consists of a system with two springs to represent the normal and tangential forces acting between two spheres colliding. The DEM parameters used in this work are taken from previous studies (17, 36) and are detailed in Tab. 1.

Table 1. DEM Simulation parameters.

Particle density ρ (kg/m ³)	1190
Young's Modulus E (MPa)	5
Poisson Ratio σ	0.45
Particle-particle friction coefficient μ_p	0.5
Particle-wall friction coefficient μ_w	0.5
Coefficient of restitution ϵ	0.5
Time step ΔT (s)	1×10^{-6}
Angular velocity of the drum Ω (rpm)	12

For the fluid-sheared granular bed, the computations were carried out by using the open-source code CFDEM (56), that couples LIGGGHTS (described previously) and OpenFOAM (which computes the fluid motion in an Eulerian frame). For this case, the LIGGGHTS code solves a modified Eq. 1, where we add the fluid contributions given by $\vec{F}_D + \vec{F}_{stress} + \vec{F}_{am}$ in the right-hand side, where \vec{F}_D is the drag force caused by the fluid on particles, $\vec{F}_{stress} = V_p[-\nabla P + \nabla \cdot \bar{\tau}]$ is the force caused by the fluid stresses, and \vec{F}_{am} is the added mass force which is important for simulations involving liquids (36). P is the fluid pressure and $\bar{\tau}$ is the deviatoric stress tensor of the fluid. On the other hand, OpenFoam computes the conservation of mass and momentum of the fluid by the following equations:

$$\frac{\partial \rho_f \epsilon_f}{dt} + \nabla \cdot (\rho_f \epsilon_f \vec{u}_f) = 0 \quad [3]$$

$$\frac{\partial \rho_f \epsilon_f \vec{u}_f}{dt} + \nabla \cdot (\rho_f \epsilon_f \vec{u}_f \vec{u}_f) = -\epsilon_f \nabla P + \epsilon_f \nabla \cdot \bar{\tau} - \frac{\vec{F}_D}{V_{cell}} \quad [4]$$

where \vec{u}_f is the velocity of the fluid phase, ϵ is the volume fraction of the fluid in a calculation cell, and V_{cell} is the volume of the considered calculation cell. The estimations of the drag force

\vec{F}_D imposed on each particle come from experimental correlations based in the flow regime and the volume fraction (70). The CFD parameters used in this work are detailed in Tab. 2.

Table 2. CFD Simulation parameters.

Fluid density ρ_f (kg/m ³)	1050
Fluid viscosity μ_f (mPa.s)	72.2
Top mean velocity U (m/s)	0.02
Mean fluid height h_f (m)	0.004
Time step ΔT_f (s)	5×10^{-5}
Channel dimensions X, Y, Z (m)	$0.1 \times 0.025 \times 0.01$

With the conditions described above, the Reynolds number $Re_f = \rho_f U h_f / \mu_f$ is around 1.5 that assures the flow is in a laminar regime. The shields number $\theta = \frac{\mu_f U / h_f}{(\rho_p - \rho_f) g d_p}$ has values ranging from 0.13 to 0.18 (depending on the size of particles), and the threshold of motion for this case is $\theta_{cr} \approx 0.1$ (52, 71).

Numerical setup and validation. For the particles, we used: (i) Spheres with sizes varying from 1.5mm to 4.5mm. (ii) Cubical particles formed from bonded spheres, that were implemented numerically by placing into permanent contact 8 spheres, that do not overlap with each other, with bonds half the diameter of spheres and being considered solid, as shown in Fig. 2(a); in order to prevent any gravitational stratification, we match the mass of the 8 bonded spheres to the solid spheres to estimate the density of individual grains that composed a bonded cube. (iii) Cubical particles formed from superquadric shapes (Fig. 2(b)) which are determined by the following equation:

$$\left(\left| \frac{x}{a} \right|^{n_2} + \left| \frac{y}{b} \right|^{n_2} \right)^{\frac{n_1}{n_2}} + \left| \frac{z}{c} \right|^{n_1} - 1 = 0 \quad [5]$$

where a, b, c are the lengths of the particles semi-axis, and n_1 and n_2 determine the particle shape and the surface blockiness (5, 26, 72). To obtain the cubical particle shown in Fig. 2(b), we set n_1 and n_2 equal to 8.

For the case of the purely granular interactions, we consider a rotary drum with a diameter D of 0.3m and a width W of 0.05m driven by a rotational speed of 12RPM for 140s; meanwhile, for the case of the fluid-sheared granular bed, we used a rectangular channel with dimensions of 0.1m in the streamwise direction, 0.025m in the cross-stream direction, and 0.01m in depth; where we imposed a velocity at the top wall of 0.02m/s for 300s. For both cases, two species of particles were randomly placed in equal ratios. In order to run the numerical simulations, first we let the mixture of particles to settle for 1 second and to rest for another 1s.

As part of the validation of our dry model, we also carried out an experiment with a rotary drum filled with glass beads of various sizes (see Supplemental Material (73) for a video comparing the experiments and numerical simulations).

Calculation of the amount of segregation. Figure 2c shows the evolution of the amount of segregation for mixtures of spheres and bonded particles, where cooler colors represent to small volume ratios and warmer colors to large volume ratios. In all cases the amount of segregation starts at zero, where the particles are randomly distributed and then increases until it reaches a final steady state. For each case, we fitted the curves of the temporal evolution of the amount of segregation by using the following expression:

$$S(t) = S_f \left(1 - e^{-t/t_s} \right). \quad [6]$$

where S_f is the amount of segregation at the steady state, t is time, and t_s is the time that a case takes to reach the steady state from its initial condition.

By fitting the curves shown in Fig. 2 (c), the final amount and the time of segregation for each case were determined.

The amount of segregation that a system reaches is an important parameter to estimate the final behavior of a mixture of particles; however, it is an empirical parameter that varies with the local

560 domain, number of species, and the distribution of particle size.
 561 Although there are several studies that focus on determining the
 562 amount of segregation, calculations are inherently biased depending
 563 on the choice of window size. To quantify segregation, we calculated
 564 the fraction of each species with respect to the total number of
 565 particles throughout the entire domain, based on dividing the rotary
 566 drum in sub-domains as shown in Ref. (74). This formulation is
 567 useful because it can be applied to systems with any number of
 568 different species, rather than being limited to bidisperse systems.
 569 Based on an exhaustive analysis of the number of subdomains
 570 needed in the rotary drum, we found that the size of a subdomain
 571 is best determined by the sum of the sizes of each species (see
 572 Supplemental Material (73) for the study of subdomain sizes).

573 The domain of the drum is divided in M number of subdomains
 574 of rectangular shape to estimate the amount of segregation of Q
 575 types of species present in the mixture. For our study, we consider
 576 a distribution of equal volume ratio for the granular bed; then,
 577 the domain does not contain the same number of particles of each
 578 species. Therefore, to determine the fraction of one species with
 579 respect to the highest number of particles relative to the other
 580 species is given by the following equation:

$$581 \quad P_{ki} = \frac{n_{ki} f_k}{\max((n_{1i} f_1), (n_{2i} f_2), \dots, (n_{Qi} f_Q))} \leq 1. \quad [7]$$

582 where n_{ki} is the number of particles of the k^{th} species in the
 583 subdomain i , and f_k is the factor of participation based in the total
 584 number of particles of each species given by:

$$585 \quad f_k = \frac{\max(\sum_{i=1}^M n_{1i}, \sum_{i=1}^M n_{2i}, \dots, \sum_{i=1}^M n_{Qi})}{\sum_{i=1}^M n_{ki}}. \quad [8]$$

587 The instantaneous amount of segregation S is obtained from
 588 the arithmetic mean of the individual fractions of each species of
 589 particles k in all M subdomains, and is calculated by the following
 590 equation:

$$591 \quad S = 1 - \left(\frac{1}{N} \sum_{i=1}^M \left[\frac{1}{Q-1} \left(\sum_{k=1}^Q P_{ki} - 1 \right) \sum_{k=1}^Q (n_{ki}) \right] \right). \quad [9]$$

592 where N is the total number of particles in the mixture.

593 We used the segregation time series to calculate final segregation
 594 levels for each run. We fitted the curves of the temporal evolution
 595 of the amount of segregation by using the following expression:

$$597 \quad S(t) = S_f (1 - e^{-t/t_s}). \quad [10]$$

598 where S_f is the amount of segregation at the steady state, t is time,
 599 and t_s is the time that a case takes to reach the steady state from
 600 its initial condition. By fitting the curves shown in Figure 2c, for
 601 example, the final amount and the time of segregation for each case
 602 were determined.

603 **A. Calculation of Anisotropy.** The amount of anisotropy that a granular
 604 system exhibits is determined by the contact fabric tensor \hat{R} ,
 605 which is calculated by the following equation:

$$606 \quad \hat{R} = \frac{1}{N_c} \sum_{i \neq j} \frac{\vec{r}_{ij}^+}{|\vec{r}_{ij}^+|} \otimes \frac{\vec{r}_{ij}^-}{|\vec{r}_{ij}^-|}. \quad [11]$$

607 where \vec{r}_{ij}^+ is the contact vector from the center of particle i to
 608 the interparticle contact between particles i and j , \otimes is the vector
 609 outer product, and N_c is the total number of particles with at least
 610 two contacts. The dimensionless fabric anisotropy tensor \hat{AF} is
 611 proportional to the deviatoric part of the contact fabric tensor \hat{R}
 612 and can be estimated by the following expression (49):

$$613 \quad \hat{AF} = \frac{5}{2} (3\hat{R} - \hat{I}). \quad [12]$$

614 where \hat{I} is the identity tensor. Finally, the amount of anisotropy
 615 that a system shows AF is given by the norm of the dimensionless
 616 fabric anisotropy tensor and can be computed by:

$$AF = \sqrt{\hat{AF} : \hat{AF}}. \quad [13] \quad 617$$

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PNAS



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2 **Supporting Information for**

3 **How particle shape affects granular segregation in industrial and geophysical flows**

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7 **This PDF file includes:**

8 Supporting text

9 Figs. S1 to S7

10 Legends for Movies S1 to S4

11 **Other supporting materials for this manuscript include the following:**

12 Movies S1 to S4

13 Supporting Information Text

14 In this document we provide additional information for key aspects of the manuscript. In section I, we show the validation of
15 our numerical model with an experimental setup with similar dimensions. In section II, we provide some information about
16 how we determine the window size to compute the segregation level. In section III, we show instantaneous snapshots of the
17 particles' final state for all the cases we used to produce our dry rotating drum results. In section IV, we show instantaneous
18 snapshots of the final state of force networks for all the cases we used to reproduce the results described in section III.

19 **I. Validation of the numerical model.** We carried out an experiment to qualitatively validate our numerical model. The
20 experimental setup (Fig. S1i) consisted of an acrylic drum with a diameter of $D = 30\text{cm}$, a stepper motor driver which rotates
21 the drum and is controlled by an Arduino UNO through the GRBL project (<https://github.com/grbl/grbl>), a set of LED lamps,
22 and a DSLR Nikon D7500 camera. The camera and the motor driver were automatically synchronized to record and save the
23 movies. For this specific case, we placed small marbles ($d_s = 4\text{mm}$, yellow color in Fig. S1ii) at the bottom and then the bigger
24 ones ($d_b = 8\text{mm}$, red color in Fig. S1ii) just above the smaller ones. We used this initial condition in order to simulate the
25 exact initial condition for the experiment and the numerical model; however, note that it differs from the initial condition
26 in our main model results, which begin with a fully mixed system. This initial condition would be very difficult to implement
27 in a physical experiment. Finally, we ran the experiment for 140s at a rotation speed of 12RPM (similar conditions to the
28 simulations carried out for the manuscript). From Figure S1ii and movie S1, we found that the behavior of the experiment and
29 the numerical model is highly similar in terms of segregation levels and time to reach a final segregated state.

30 **II. Determination of the window size for segregation level calculations.** We studied the influence of the size of the subdomains
31 detailed in Figure S2 in the calculation of the amount of segregation of the system. The amount of segregation is strongly
32 dependent on the number and the size of subdomains we have in the system. Even though many studies focused in different
33 techniques to study either level of segregation or mixing, there is no a defined recipe to follow when picking the correct window
34 size. For our study we divided the domain into small squared sections in the x and y directions ($\Delta = dx = dy$) throughout the
35 width of the drum, then we varied the size of Δ from $0.25(d_s + d_b)$ (where the window is around half of the size of one species)
36 to $10(d_s + d_b)$ (where the window is around 20 times the size of one species).

37 Figure S3 shows the trends of the temporal evolution of the amount of segregation for different window sizes. In the case of
38 a very small size ($\Delta = 0.25(d_s + d_b)$), the amount of segregation is very high because we are looking at a region where only
39 one particle can fit inside, having thus a totally segregated system in that subdomain ($S \approx 1$). For the case of a big size
40 ($\Delta = 10(d_s + d_b)$), the region is big enough that we can encounter both species in same proportions, having thus a totally
41 mixed system in that subdomain ($S \approx 0$). However, for the cases where size is close to the sum of both species diameters
42 ($\Delta \approx 1(d_s + d_b)$), the behavior of the evolution of the amount of segregation is similar.

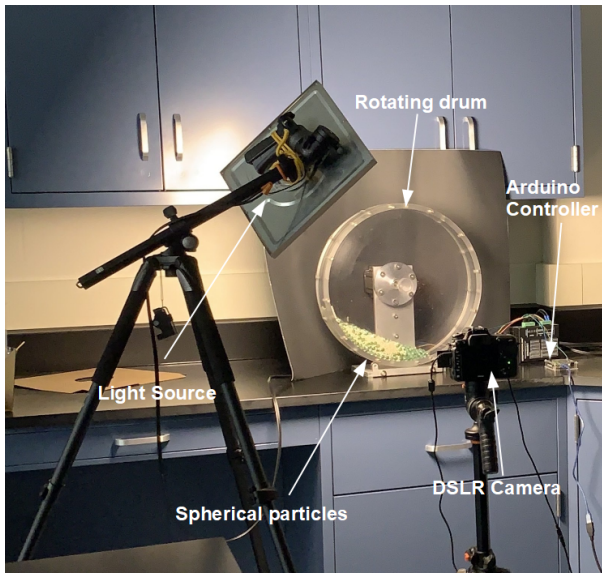
43 To identify which size is better to compute the amount of segregation we have two references, 1) the initial condition for all
44 the cases is that particles are randomly distributed which S should be around zero, and 2) the amount of segregation for spheres
45 of equal size ($d_s = d_b = 1$) is zero as the system does not have either segregation by size or shape. Therefore, the best window
46 size to compute the level of segregation is $\Delta = 1(d_s + d_b)$, because all the cases start in the same point at $t=0\text{s}$ (Fig S3(d)) and
47 the results are consistent with what we see qualitatively from the particle positions snapshots (Fig S5). Finally, to get the data
48 that are in Figure 2 of the manuscript, we shifted the curves to the zero reference that is given by the case of $V_b/V_s=1$.

49 Figure S4 shows the temporal evolution of the amount of segregation for all the tested cases and their exponential fits.

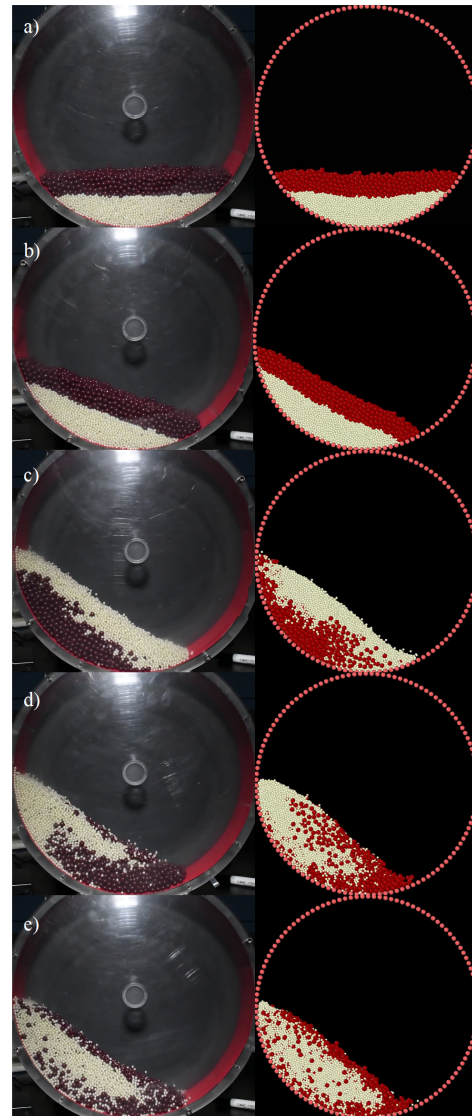
50 **III. Final state of particles position for all the tested cases.** Figure S5 shows instantaneous snapshots of particle positions for
51 size segregation in both cases spheres and cubes.

52 Figure S6 shows instantaneous snapshots of particle positions for size and shape segregation in for cubes made by bonded
53 particles and superquadrics.

54 **IV. Final state of force networks for all the tested cases.** Figure S7 shows the network of force chains for mixtures of spheres
55 and cubical particles at $t=140\text{s}$ of simulation.



(i)



(ii)

Fig. S1. (i) Experimental setup. (ii) Comparison between the experiment (left side of the figure) and the numerical model (right hand of the figure) at: (a) $t=0s$, (b) $t=1s$, (c) $t=2s$, (d) $t=4s$, and (e) $t=8s$

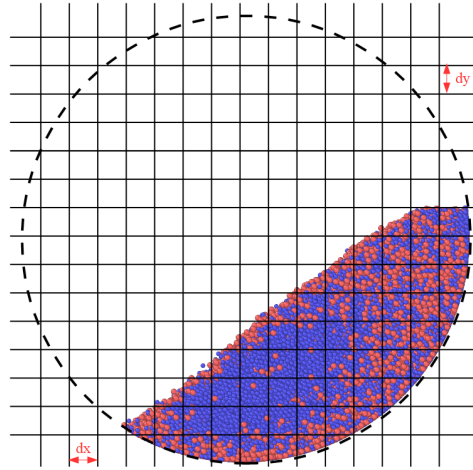


Fig. S2. Scheme of domain and subdomains considered to compute the level of segregation

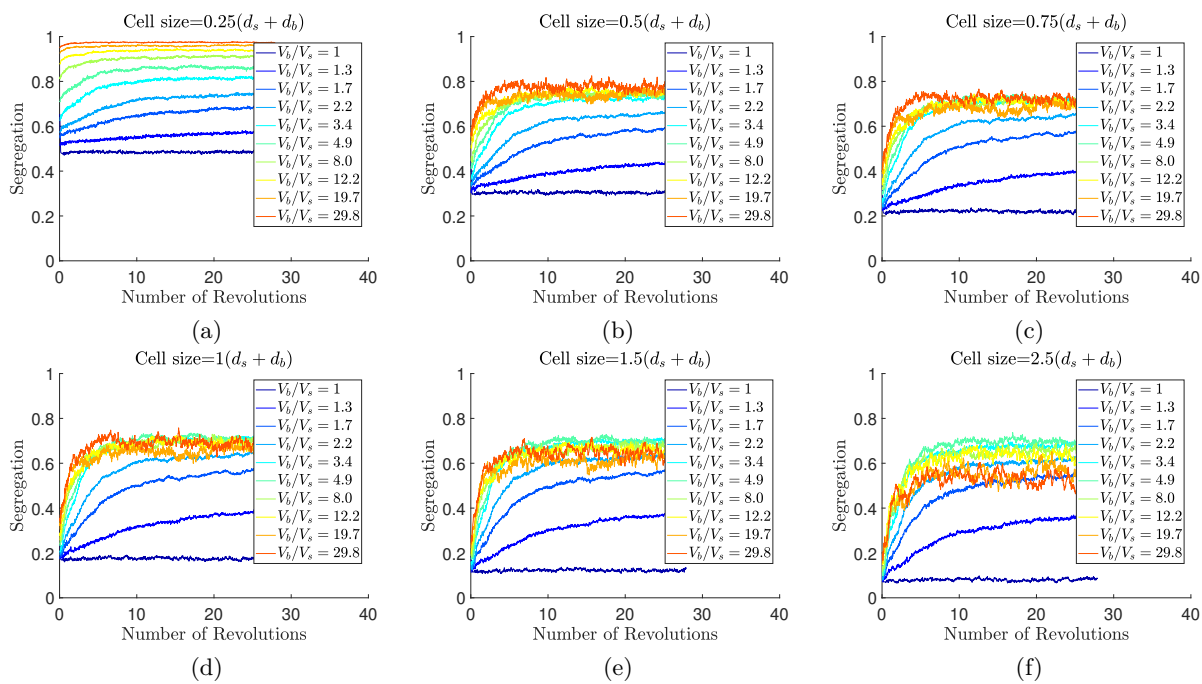


Fig. S3. Temporal evolution of the amount of segregation for mixtures containing spheres of different sizes with different window sizes. (a) $0.25d_s + d_b$, (b) $0.5d_s + d_b$, (c) $0.75d_s + d_b$, (d) $1d_s + d_b$, (e) $1.5d_s + d_b$, and (f) $2.5d_s + d_b$. Cooler and warmer colors show cases with small and high volume ratios, respectively.

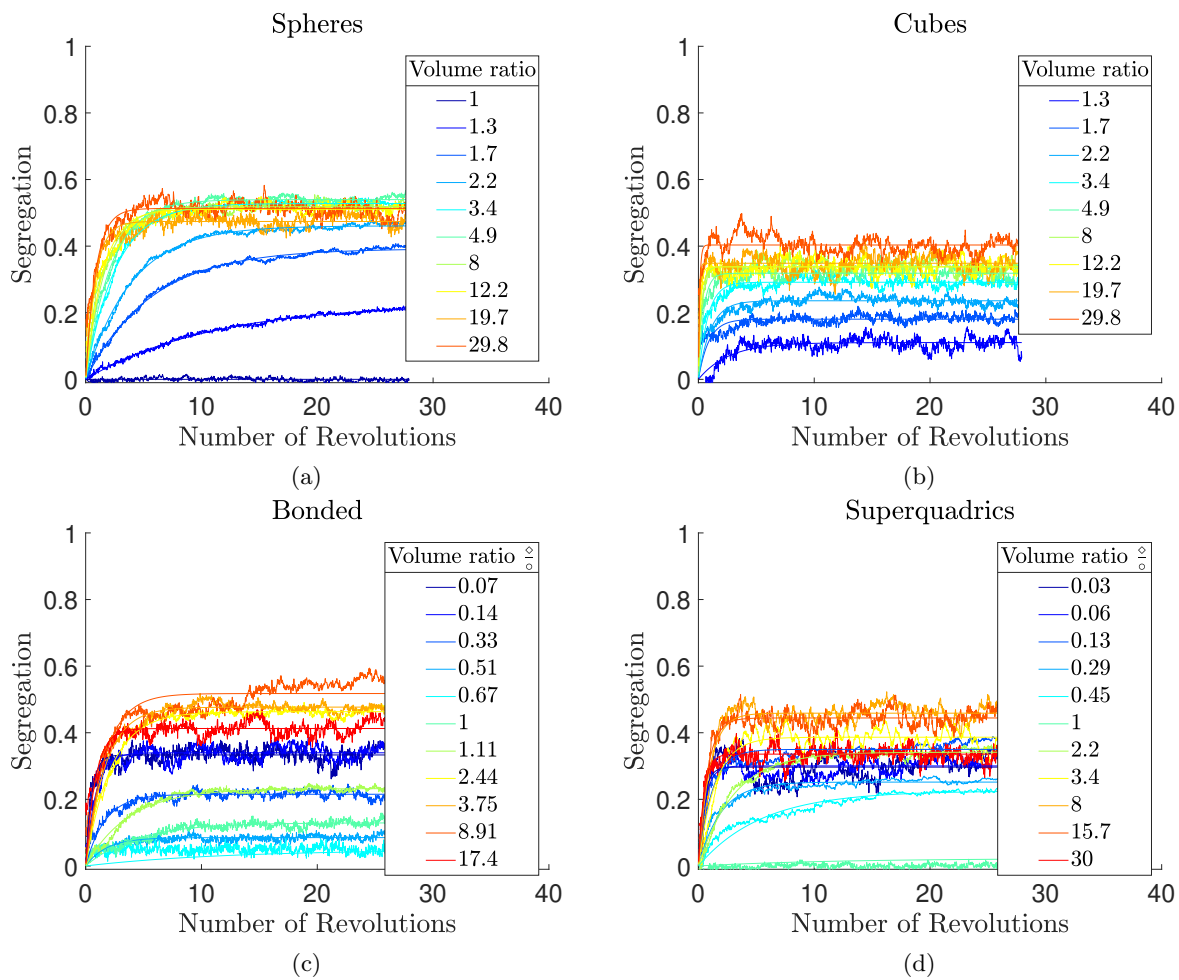


Fig. S4. Temporal evolution of the amount of segregation for mixtures containing. (a) Spheres of different sizes, (b) cubical bonded particles of different sizes, (c) spheres and cubical bonded particles, and (d) spheres and cubical superquadrics.

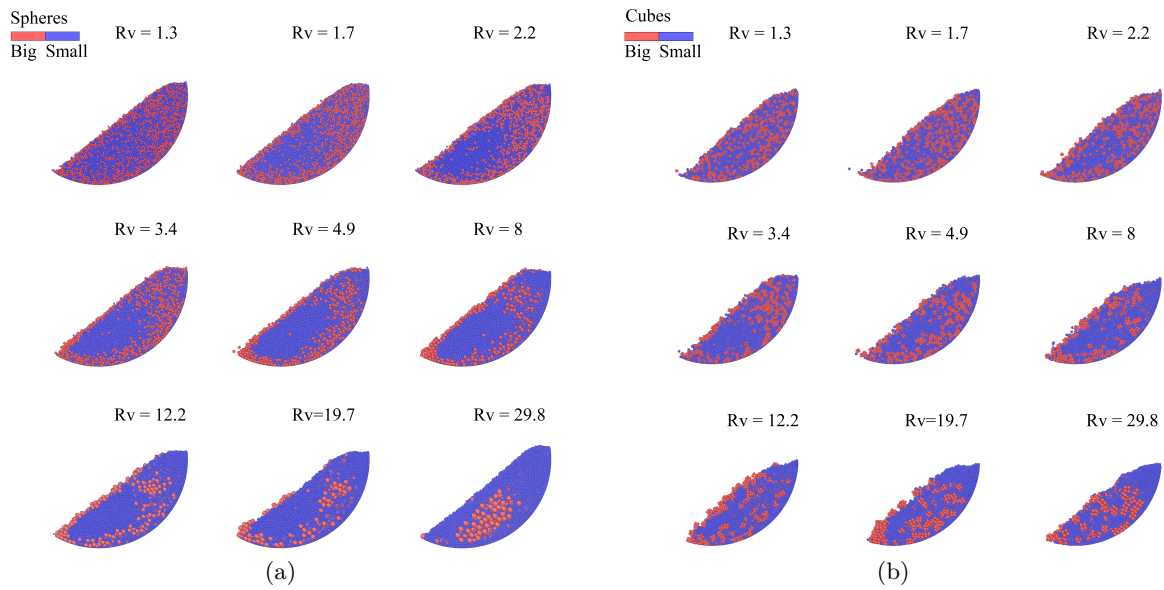


Fig. S5. Snapshots of particle positions for size segregation on: (a) spheres, (b) bonded cubes

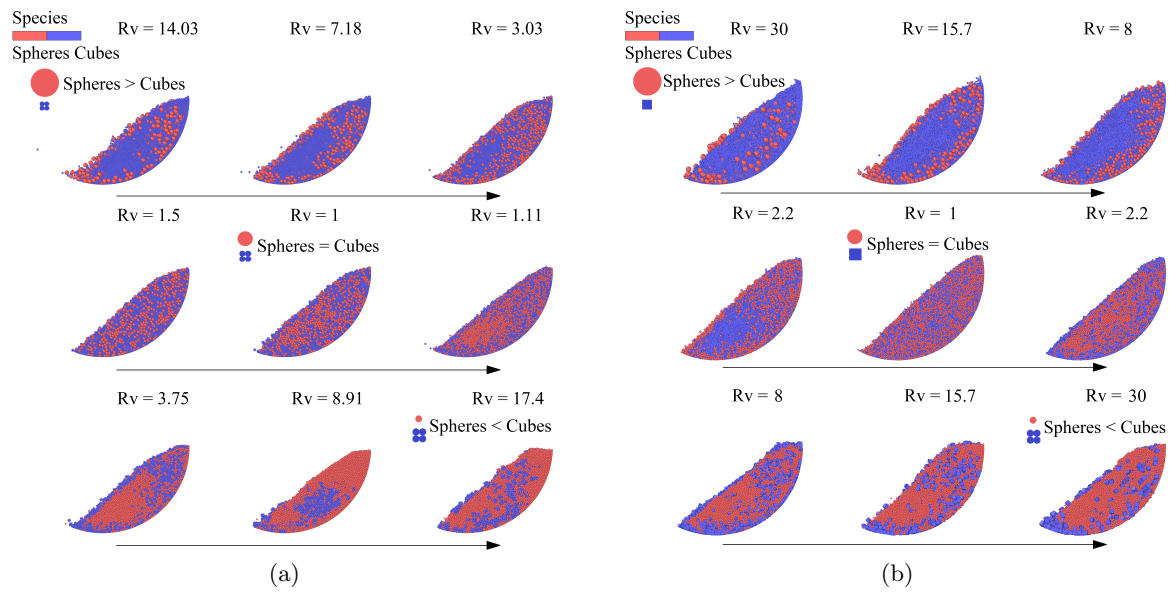


Fig. S6. Snapshots of particle positions for size and shape segregation on spheres and: (a) bonded spheres, and (b) superquadric cubes.

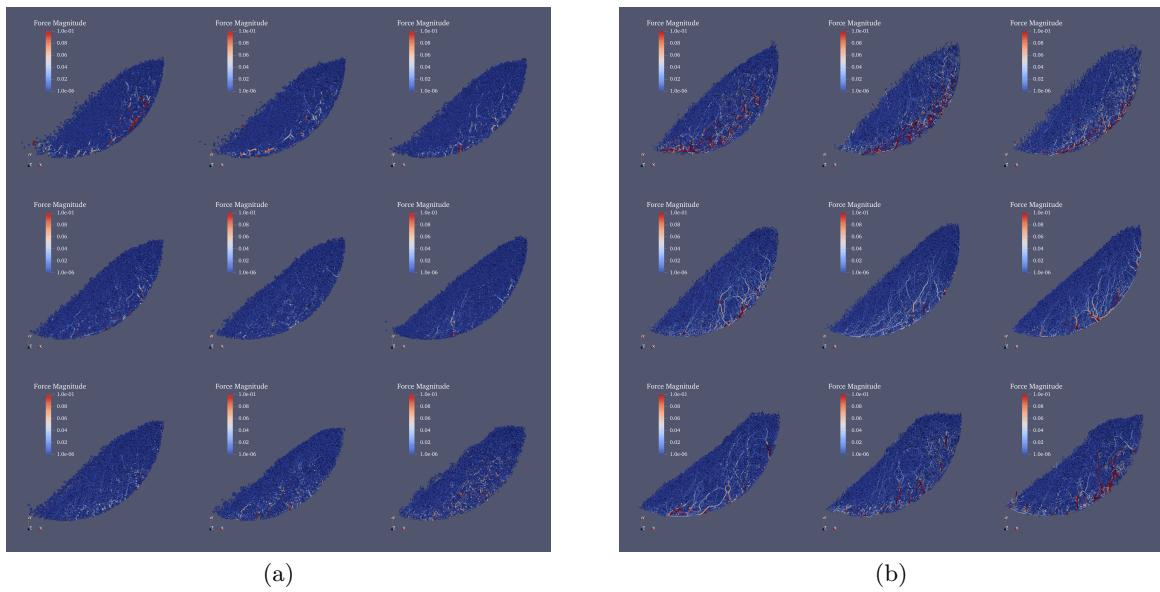


Fig. S7. Snapshots of network of force chains for size and shape segregation on mixtures with spheres and: (a) bonded particles, and (b) superquadric cubes.

- 56 **Movie S1.** Validation of the numerical model with an experiment in real time
- 57 **Movie S2.** Numerical model with a mixture of spheres and bonded spheres for the case $V_{\square}/V_{\circ} = 0.07$
- 58 **Movie S3.** Numerical model with a mixture of spheres and cubical superquadrics for the case $V_{\square}/V_{\circ} = 0.06$
- 59 **Movie S4.** CFD-DEM model of a granular bed composed by spheres and bonded particles and driven by a
60 Couette laminar flow.