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Alessandro Sozza,^{1,a)}  Gábor Drótos,^{2,3}  Emilio Hernández-García,²  and Cristóbal López² 

AFFILIATIONS

¹Istituto dei Sistemi Complessi, CNR, via dei Taurini 19, 00185 Rome, Italy and INFN sez. Roma 2 “Tor Vergata,” Rome, Italy

²IFISC (CSIC-UIB), Institute for Cross-Disciplinary Physics and Complex Systems, Campus Universitat de les Illes Balears, 07122 Palma de Mallorca, Spain

³MTA-ELTE Theoretical Physics Research Group, Pazmany Peter setany 1/A, H-1117 Budapest, Hungary

^{a)} Author to whom correspondence should be addressed: asozza.ph@gmail.com

ABSTRACT

We study the effect of turbulence on a sedimenting layer of particles by means of direct numerical simulations. A Lagrangian model in which particles are considered as tracers with an additional downward settling velocity is integrated together with an isotropic homogeneous turbulent flow. We study the spatial distribution of particles when they are collected on a plane at non-asymptotic times. We relate the resulting coarse-grained particle density to the history of the stretching rate along the particle trajectory and the projection of the density onto the accumulation plane and analyze the deviation from homogeneity in terms of the Reynolds number and the settling velocity. We identify two regimes that arise during the early and well-mixed stages of advection. In the former regime, more inhomogeneity in the particle distribution is introduced for decreasing settling velocity or increasing Reynolds number, while the tendencies are opposite in the latter regime. A resonant-like crossover is found between these two regimes where inhomogeneity is maximal.

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I. INTRODUCTION

Sedimentation of particles in a turbulent flow is a crucial problem both for theory and applications. For example, it plays a key role in the process of rain formation in clouds.^{1,2} In the marine environment, sinking of particles is an important mechanism for many physical processes: in the sequestration of carbon dioxide^{3,4} and in the downward transport of organic and inorganic aggregates, such as marine snow,^{5,6} larval eggs, and microplastics.^{7,8} Experimentally, a way to estimate the downward fluxes of particles in the ocean interior is performed by placing sediment traps.^{9–11} An open question concerns the identification of the mechanisms that lead to the observed size and spatial distributions of particles, which are collected at a given depth by the traps.

The interaction between particles and flow is determinant to establish the spatial distribution of particles.¹² Advection of a homogeneous distribution of passive particles in an incompressible flow generally results in a homogeneous concentration of particles. Deviations from homogeneity may arise from some type of compressibility, either in the flow itself or in the motion of particles. In this case, particle dynamics is restricted to a

lower-dimensional or even fractal subspace. Some exemplary cases of this phenomenon are found in the motion of particles under significant inertial effects,^{13–16} in gyrotactic algae,¹⁷ in the action of buoyancy that forces particles to relax to a specific isopycnal depth,¹⁸ or even confined to move on a horizontal sheet¹⁹ or on a free surface.²⁰ Another situation arises when considering initially inhomogeneous distributions. In this case, even with passive tracers in incompressible flow, it is possible to observe inhomogeneities at non-asymptotic time scales. Cuts or projections to a lower-dimensional manifold can give rise to additional inhomogeneity in this case. Under complex flow acting for sufficiently long times, the particle distribution will generally recover homogeneity, but for the finite times characteristic of realistic situations (for example, sedimentation in the ocean), distributions are far from this asymptotic limit.

In this paper, we investigate the dynamics of a sedimenting layer of particles under three-dimensional turbulence and discuss the role of the flow to create inhomogeneities. Particles initially distributed homogeneously on an upper plane are let to fall down in a turbulent flow and are collected at a lower accumulation plane. We will discuss how the final coarse-grained density of particles is

related to the properties of the flow. Two contributions were identified from previous works: stretching of the particle layer and projection on the collecting surface. Different from the previous works in which large-scale oceanic simulations^{21,22} or chaotic dynamical systems²³ were used, our attention is focused on the small-scale inhomogeneities due to an isotropic homogeneous turbulent flow. In Sec. II, we formulate the numerical setup and introduce the main theoretical tools. Section IV describes our results and discusses them, and our conclusions are summarized in Sec. V.

II. FORMULATION OF THE PROBLEM

We begin by considering a homogeneous and isotropic turbulent flow described by an incompressible velocity field $\mathbf{u}(\mathbf{x}, t)$ (e.g., $\nabla \cdot \mathbf{u} = 0$) ruled by Navier–Stokes equations,

$$\partial_t \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u} = -\nabla p + \nu \Delta \mathbf{u} + \mathbf{f}, \quad (1)$$

where p is the pressure, ν is the kinematic viscosity, and \mathbf{f} is the mechanical random forcing with imposed energy input ε . In the absence of forcing and viscosity, the system conserves energy $E = \frac{1}{2}(\mathbf{u}^2)$. When forcing and viscosity are at work, a turbulent steady state can be reached, where energy is conserved only in a statistical sense and transferred from large scales to small scales with a constant flux.²⁴ The energy input ε , together with the kinematic viscosity ν , defines the Kolmogorov microscales for the length $\eta = (\nu^3/\varepsilon)^{1/4}$, time $\tau_\eta = (\nu/\varepsilon)^{1/2}$, velocity $u_\eta = (\nu\varepsilon)^{1/4}$, and acceleration $a_\eta = (\varepsilon^3/\nu)^{1/4}$. These scales will be used to define dimensionless parameters.

We now discuss the equations of motion for the particles. We consider small spherical particles of size a and density ρ_p transported by the incompressible velocity field $\mathbf{u}(\mathbf{x}, t)$. A standard modeling setup is a simplified form of the Maxey–Riley equations²⁵ for the velocity of the particles \mathbf{v} ,

$$\frac{d\mathbf{v}}{dt} = \beta \frac{d\mathbf{u}}{dt} - \frac{\mathbf{v} - \mathbf{u}}{\tau_p} + (1 - \beta)\mathbf{g}, \quad (2)$$

where $\beta = 3\rho_f/(2\rho_p + \rho_f)$ is the density contrast (ρ_f is the density of the fluid), $\tau_p = a^2/(3\beta\nu)$ is the Stokes relaxation time, and \mathbf{g} is the gravitational acceleration. If the flow is turbulent, we can define two dimensionless parameters, the Stokes number $St = \tau_p/\tau_\eta$ and the Froude number $Fr = a_\eta/g$. In the limits $St \rightarrow 0$ and $Fr \rightarrow 0$ but such that St/Fr remains constant, we can neglect the inertial effects without omitting the gravity term (since the settling velocity is^{23,26–28} $v_s \propto St/Fr$), leading to the reduced first-order differential equation,

$$\mathbf{v}(t) \equiv \frac{d\mathbf{X}(t)}{dt} = \mathbf{u}(\mathbf{X}(t), t) - v_s \hat{\mathbf{z}}. \quad (3)$$

In this expression, we neglect terms of first order in St . Thus, particles are “tracers” transported by the incompressible velocity field $\mathbf{u}(\mathbf{x}, t)$ that additionally sink with a constant settling velocity $v_s = (1 - \beta)g\tau_p$ along the vertical direction z (characterized by the $\hat{\mathbf{z}}$ unit vector). The model defined by Eq. (3) has been largely studied in the literature and in previous studies on this specific subject.^{11,21,23,26,29} We remark that the model is derived within the assumption that particles are small, with particle Reynolds number $Re_p = v_s a/\nu \ll 1$, and not interacting so that each particle evolves independently from the others. Furthermore, imposing

$St < 1$ restricts the validity of the reduced model to settling velocities $v_s < (1 - \beta)g\tau_\eta$.

We introduce a dimensionless settling parameter $\Phi = v_s/U$, with U being the root mean square velocity $U = (2E/3)^{1/2}$. Notice that for $\Phi \gg 1$ (i.e., $v_s \gg U$), the motion of the particles is ballistic and turbulence is reduced to a small perturbation. On the contrary, when $\Phi \approx 1$ or $\Phi \ll 1$, trajectories are strongly controlled by turbulence and a random-like motion arises. The constraint $v_s < (1 - \beta)g\tau_\eta$, ensuring $St < 1$, reads as $\Phi < 15^{1/4}(1 - \beta)Fr^{-1}Re_\lambda^{-1/2}$ in dimensionless quantities (see the definition of Re_λ in Sec. III).

At the initial time $t = 0$, particles are homogeneously released at random positions on a horizontal plane $z = L$, after which they move following Eq. (3). In order to investigate the evolution and deformation of the layer of particles, we need to calculate, among other quantities, the local stretching rates along each particle trajectory. We introduce the Jacobian matrix $\mathbb{J}(t)$ describing separation in time $\delta\mathbf{X}(t)$ of particle trajectories initialized at an infinitesimal distance $\delta\mathbf{X}(0)$, i.e.,

$$\delta X_\alpha(t) = \sum_{\beta=1,2,3} \mathbb{J}_{\alpha\beta}(t) \delta X_\beta(0), \quad (4)$$

with

$$\mathbb{J}_{\alpha\beta}(t) = \frac{\partial X_\alpha(t)}{\partial X_\beta(0)}. \quad (5)$$

Using the chain rule, the evolution of $\mathbb{J}_{\alpha\beta}$ is given by

$$\frac{d}{dt} \mathbb{J}_{\alpha\beta}(t) = \sum_{\gamma=1,2,3} \partial_\gamma u_\alpha(\mathbf{X}(t), t) \mathbb{J}_{\gamma\beta}(t), \quad (6)$$

where $\partial_\gamma u_\alpha(\mathbf{X}(t), t)$ is the fluid velocity gradient measured at the position of the particle that started at $\mathbf{X}(0)$. The initial condition is $\mathbb{J}_{\alpha\beta}(0) = \delta_{\alpha\beta}$. Since initially the particle surface is horizontal, the first and second columns of the matrix $\mathbb{J}_{\alpha\beta}(t)$ give at each time two vectors, $\mathbf{t}_1(t)$ and $\mathbf{t}_2(t)$, tangent to that falling surface.

We are interested in quantifying the final distribution of particles deposited on a horizontal plane at a fixed depth, say $z = 0$. At that plane, we can define a particle surface density $\rho(\mathbf{x}_h)$, with $\mathbf{x}_h = (x, y)$ denoting the horizontal components. The relationship between the homogeneous density ρ_0 at the upper release plane and the density $\rho(\mathbf{x}_h)$ at the lower collecting plane is given by a *total density factor* $F(\mathbf{x}_h)$ defined by $\rho(\mathbf{x}_h)/\rho_0 \equiv F(\mathbf{x}_h)$. As demonstrated in previous work,^{23,30} this total factor is the product of two contributions: $F(\mathbf{x}_h) = S(\mathbf{x}_h)P(\mathbf{x}_h)$. S , the *stretching factor*, characterizes the stretching accumulated by the falling surface around the trajectory that reaches the lower plane at \mathbf{x}_h , whereas P , the *projection factor*, takes into account the orientation-dependent footprint of the falling surface on the horizontal collecting plane in the neighborhood of \mathbf{x}_h . These two factors can be calculated^{23,30} (cf.^{31,32} as well) from the tangent vectors $\mathbf{t}_1(t)$ and $\mathbf{t}_2(t)$ [thus, from Eq. (6)] as

$$\begin{cases} S = |\mathbf{t}_1 \times \mathbf{t}_2|^{-1}, \\ P = \frac{|v_z|}{|\hat{\mathbf{n}} \cdot \mathbf{v}|}. \end{cases} \quad (7)$$

All quantities are computed at the final time t_h at which the particle trajectory reaches position \mathbf{x}_h on the collecting plane. v_z is the vertical component of the particle velocity \mathbf{v} at that time, and $\hat{\mathbf{n}}$ is the unit vector normal to the surface that can be computed by normalizing \mathbf{n} , the vector normal to the surface given by the cross product

$\mathbf{n}(t_h) = \mathbf{t}_1(t_h) \times \mathbf{t}_2(t_h)$. If the falling surface reaches the accumulation plane horizontally around \mathbf{x}_h , $\hat{\mathbf{n}}$ at that location points along the z axis and $P = 1$, meaning that there is no projection effect. Note that P diverges where $\hat{\mathbf{n}} \cdot \mathbf{v} = 0$, i.e., where the particle velocity arrives at the collecting plane tangent to the falling surface. These locations define *caustics* that form lines and typically occur when the falling surface develops folds. On the other hand, the area of an infinitesimal surface element at time t is $|\mathbf{t}_1(t) \times \mathbf{t}_2(t)| dA_0$, where dA_0 is the initial area. Thus, $S = 1$ if the surface reaches the accumulation plane unstretched.

III. NUMERICAL SIMULATIONS

We solve Eq. (1) with a pseudo-spectral method on a triply periodic cubic domain of size $L = 2\pi$ containing $M^3 = 32^3 - 256^3$ grid points to obtain statistically steady flows with Taylor-microscale Reynolds number $Re_\lambda = U\lambda/\nu \approx 19-93$, where $\lambda = U\sqrt{15\nu/\varepsilon}$ is the Taylor microscale and U is the root-mean-square velocity fluctuation. Time marching is performed using a second-order Runge-Kutta scheme. The forcing acts only at large scales in a shell of wavenumbers $k \leq k_f$ and maintains a constant energy input $\langle \mathbf{f} \cdot \mathbf{u} \rangle = \varepsilon$, which equates, on average, the energy dissipation rate. This is obtained by taking $\mathbf{f}(\mathbf{x}, t) = \varepsilon \Theta(k_f - k) \mathbf{u}(\mathbf{x}, t) / 2E_{k \leq k_f}$, where Θ is the Heaviside step function and $E_{k \leq k_f}$ is the kinetic energy restricted to the wavenumbers smaller than k_f .³³⁻³⁵ We ensure that small-scale fluid motion is well resolved by imposing the Kolmogorov length scale $\eta = (\nu^3/\varepsilon)^{1/4}$ of the resulting flow to be of the same order as our grid spacing, $k_{max}\eta > 1.8$, where $k_{max} = M/3$. Table I reports the most important Eulerian parameters used in the simulations. Additional numerical details are given in Ref. 36.

After the flow has reached statistical steady state, $N = 1.2 \times 10^6$ particles are initialized with homogeneously random positions on a plane at a fixed horizontal position $z_0 = L$. The trajectory of each of them is evolved with Eq. (3). The associated Jacobian matrix $\mathbb{J}_{\alpha\beta}(t)$ giving deformations close to that trajectory is simultaneously evolved with Eq. (6) and initial condition $\mathbb{J}_{\alpha\beta}(0) = \delta_{\alpha\beta}$. Fluid velocity and its gradients are calculated by third-order spatial interpolation on the particles' positions. The integration time step dt is chosen to be smaller than the time needed to cross a grid cell, which is equal to satisfying the condition $v_s dt/dx < 1$, where $dx = L/M$. Deformation of the evolving surface is characterized by its tangent vectors $\mathbf{t}_1(t)$ and $\mathbf{t}_2(t)$, given by the first two columns of

TABLE I. Parameters of the four turbulent flows used here: resolution M , kinematic viscosity ν , kinetic energy $E = \frac{1}{2} \langle u^2 \rangle$, root mean square velocity $U = (2E/3)^{1/2}$, Kolmogorov velocity $u_\eta = (\nu\varepsilon)^{1/4}$, eddy turnover time $\mathcal{T} = E/\varepsilon$, Kolmogorov time scale $\tau_\eta = (\nu/\varepsilon)^{1/2}$, integral length scale $\mathcal{L} = E^{3/2}/\varepsilon$, Taylor microscale $\lambda = U(15\nu/\varepsilon)^{1/2}$, Kolmogorov length scale $\eta = (\nu^3/\varepsilon)^{1/4}$, and Taylor-microscale Reynolds number $Re_\lambda = U\lambda/\nu$. All the simulations are performed with energy dissipation rate $\varepsilon = 0.1$ and domain size $L = 2\pi$.

M	ν	E	U	u_η	\mathcal{T}	τ_η	\mathcal{L}	λ	η	Re_λ
32	4×10^{-2}	0.47	0.56	0.25	4.65	0.63	3.17	1.36	0.16	19
64	2×10^{-2}	0.52	0.59	0.21	5.25	0.45	3.80	1.02	0.09	30
128	7×10^{-3}	0.61	0.64	0.16	6.11	0.26	4.77	0.65	0.04	60
256	3×10^{-3}	0.62	0.65	0.13	6.24	0.17	4.93	0.33	0.02	93

$\mathbb{J}_{\alpha\beta}(t)$, and by the normal vector $\mathbf{n}(t) = \mathbf{t}_1(t) \times \mathbf{t}_2(t)$. To limit numerical errors arising from exponentially different values of the components of $\mathbb{J}_{\alpha\beta}(t)$, a Gram-Schmidt orthonormalization is applied periodically to the vectors $\mathbf{t}_1(t)$, $\mathbf{t}_2(t)$, and $\mathbf{n}(t)$, and a new initial condition for $\mathbb{J}_{\alpha\beta}$ is built by using the resulting vectors as columns. The stretching factor S in Eq. (7) is computed as a product of the partial stretching factors obtained before each reinitialization. We consider 17 different values of the settling velocity v_s . The largest values do not satisfy the constraint imposed by $St < 1$ (a validity condition for the model, see Sec. II), but $St = 1$ is clearly marked in every figure.

As we let particles fall and be transported by the flow, we observe the deformation of the initially flat and homogeneous distribution of particles into a crumpled surface [see Fig. 1]. Since $\nabla \cdot \mathbf{u} = 0$, the dynamics defined by Eq. (3) is also incompressible ($\nabla \cdot \mathbf{v} = 0$), and we expect that a homogeneous distribution (in the three-dimensional space) is recovered after a sufficient number of eddy turnover times. Such a return to homogeneity can be obtained either at large times or, equivalently, at large depths. At finite times or depths, we suggest that the settling parameter $\Phi = v_s/U$ determines the morphology of the surface.

Integration of particle trajectories is performed until the particles reach the bottom plane at $z = 0$. In principle, there may be particles that are trapped forever in the flow above the bottom plane, but for the parameters used here, all particles arrive at the bottom plane within a finite time. When a particle reaches the bottom plane at $z = 0$, we register its position $\mathbf{X}(t_h) = (\mathbf{x}_h, 0)$, its velocity \mathbf{v} , and its arrival time t_h . With this information and the values of the stretching computed along the trajectory, we are able to compute the total stretching S , the projection P , and the total factor F for each particle.

We recall that the simulations of the fluid dynamics are implemented with periodic boundaries, which means that the accumulation plane is neither a physical barrier nor a wall. For the particles, however, the domain is periodic only in the horizontal directions. In the vertical direction, it is semi-finite with an absorbing boundary

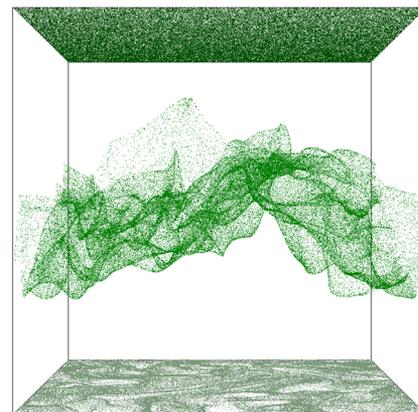


FIG. 1. Particle distribution in the initial homogeneous configuration at $z = L$ (upper plane) at an intermediate time (crumpled surface) and particles finally deposited at the lower plane $z = 0$.

condition at the bottom, on the accumulation plane, where particle trajectory integration is stopped. We also remark that caution should be taken when considering fast settling particles in a periodic flow,^{2,37} since they can perceive spurious correlations of the turbulent flow, when the time it takes a particle to fall through the domain is smaller than the correlation time of the underlying flow. In such standard setups,³⁷ particles were recirculating along the periodic domain, and then, if falling sufficiently fast, they could artificially encounter the same eddy several times. In our setup, a fast particle can sample parts of the same eddy twice at most. Also, the density factor accumulates stretching contributions from the whole particle trajectory, of which the boundary region is just a tiny part. Thus, we expect the results described below to be independent of the use of periodic boundary conditions. In fact, we have computed the average time-dependent stretching on particles with trajectories stopped at the same accumulation layer, but with a domain size L for the flow simulation twice as large, and found no difference with the result under the setup described here.

IV. RESULTS AND DISCUSSION

A. Direct inspection of spatial variations

Particles reach the bottom with different times of arrival. Hence, neighboring particles on the accumulation plane may have visited different regions of the domain, experienced very different histories of stretching and folding, and finally be collected at different moments. Similarly, particles that are initially close may have diverged and concluded their trajectories in very distant regions and at very diverse times as well.

First, we aim to obtain a direct quantitative insight into the inhomogeneities in the distribution of particles collected on the accumulation plane. A suitable way to characterize this concentration field is to compute a coarse-grained surface density ρ_{ij} , where the indices (i, j) label a set of boxes on the collecting plane: particle positions on that collecting plane are located within a two-dimensional grid with resolution M_B and counted in each cell of size $L_B = L/M_B$ so that $\rho_{ij} = n_{ij}/L_B^2$, where n_{ij} is the number of particles in the cell (i, j) . Summing over all cells, one obtains the total number of particles as $\sum_{i,j=1}^{M_B} n_{ij} = N$. The initial density, namely, ρ_0 , is equal to N/L^2 so that $\rho_{ij}/\rho_0 = (n_{ij}L^2)/(NL_B^2) = (n_{ij}M_B^2)/N$. In the homogeneous case, when $n_{ij} = N(L_B/L)^2$, we obtain $\rho_{ij}/\rho_0 = 1$. If the particle distribution becomes inhomogeneous, the presence of voids and clusters will be registered where $\rho_{ij}/\rho_0 < 1$ and $\rho_{ij}/\rho_0 > 1$, respectively.

In the absence of folds, ρ_{ij}/ρ_0 is a coarse-grained version of $F(\mathbf{x}_h) = \rho(\mathbf{x}_h)/\rho_0$. If more than one branch of the surface appears at a particular position due to some folding of the surface, ρ/ρ_0 will correspond to a sum of the coarse-grained values of F characterizing the different branches.

In Fig. 2, examples for the spatial distribution of the coarse-grained particle density, the total density factor, and the separated contributions of stretching and projection are shown on the accumulation plane for a given settling parameter (chosen near the maximal observed inhomogeneity, as characterized by the Poisson dispersion index χ defined below). F , S , and P have also been coarse-grained by taking the arithmetic average in the same cells as those that define ρ_{ij} . (Note that summation over different branches is not

actually performed for this qualitative inspection.) We observe the emergence of clustering of particles in the coarse-grained density and in the total density factor, which are in reasonable agreement with each other, even if a perfect agreement is not expected, since the presence of folds is obvious. At most points, we observe that $S < 1$, meaning that the infinitesimal area $|\mathbf{t}_1 \times \mathbf{t}_2| dA_0$ has grown larger than the original dA_0 . Also, the most noticeable features in P are large values that arise from the lines at which $\hat{\mathbf{n}} \cdot \mathbf{v} \rightarrow 0$, i.e., from the caustic lines at which P diverges. In fact, a comparison with the maps of stretching and projection suggests that the largest inhomogeneities are due to the formation of caustics, the abundance of which increases with the Reynolds number Re_λ , leading to the formation of a complex web of filaments. The dominance of caustics is similar to the case of advection of inertial heavy particles, but in that case, they arise from the compressibility of the particle flow,^{38,39} whereas here, the particle flow is incompressible ($\nabla \cdot \mathbf{v} = 0$) and develops caustics because of the two-dimensional character of the initial distribution, together with the bending action of the flow and the projection effect on the bottom surface. These three effects concur in the formation of the final distribution of particles.

B. Statistical characterization of inhomogeneities in the collecting plane

Next, we quantitatively investigate the degree of inhomogeneity and its dependence on Φ and Re_λ by evaluating the so-called Poisson dispersion index χ of the particle number distribution n_{ij} defined over the coarse-graining boxes of the accumulation plane. We also discuss implications of the choice of the box size L_B for coarse-graining.

As a first step, the average and the standard deviation of the set of values $\{n_{ij}\}$ on the accumulation plane are considered (similarly as in Refs. 23 and 30). Since the number of particles is conserved and periodic boundary conditions are prescribed in the horizontal direction, the spatial average of n_{ij} is the same as the initial number: $\bar{n}_{ij} = n_0 = N/M_B^2$ (where the bar represents the average with respect to boxes). Simple quantifiers of inhomogeneity are the standard deviation σ_n and its square, the variance. The latter is conveniently normalized by n_0 to quantify deviations from a homogeneous Poisson distribution by the Poisson dispersion index as $\chi = \sigma_n^2/\bar{n} = \sigma_n^2/n_0$. Note that $\chi = 1$ corresponds to a homogeneous but random distribution, describing particles arriving at uniformly random positions on the accumulation plane. In such a case, a nonzero standard deviation σ_n results from the finite number of particles, which, after coarse-graining, leads to a Poisson distribution of n over the boxes. True inhomogeneity, with clusters and voids, is indicated by $\chi \neq 1$.

How to choose L_B for coarse-graining is not obvious. On the one hand, it is not meaningful to take L_B below some mean distance between the particles ($\rho^{-1/2}$). On the other hand, L_B may be chosen below the spatial resolution L/M of the fluid flow in order to resolve small-scale folds of the particle sheet, which may have an important effect on the observed inhomogeneity. In Fig. 3(a), we present the dispersion index as a function of the settling parameter Φ and where the size of the coarse-graining boxes is chosen to depend on the resolution M of the fluid model as $L_B = 2L/M$ and thus also on the Reynolds number [cf. Table I]. This box size is

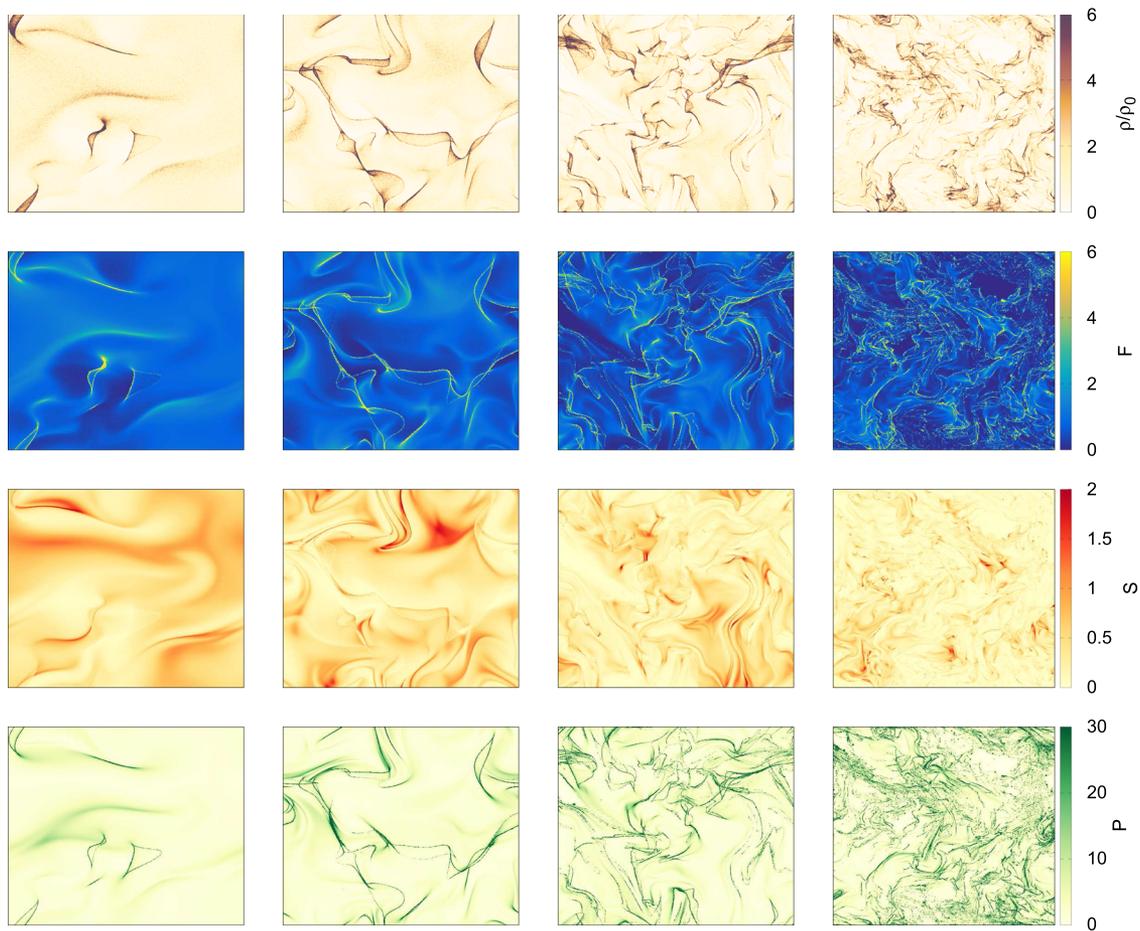


FIG. 2. Color map of the final distribution at the accumulation plane of the coarse-grained particle density ρ/ρ_0 (first row; box resolution of the coarse-graining is $M_B = 512$), the total density factor $F(\mathbf{x}_h)$ (second row), and the separated contributions due to the stretching $S(\mathbf{x}_h)$ and the projection $P(\mathbf{x}_h)$ factors (third and fourth row). $Re_\lambda = 19, 30, 60, 93$ in the four columns from left to right. Computations are for $\Phi = 3$, a value for which a large Poisson dispersion index χ is attained.

near the smallest characteristic scale (the Kolmogorov length scale) of the fluid motion but varies between relatively coarse ($L/16$) and much finer ($L/128$) values compared to the domain size. Irrespective of Re_λ , particles are found uniformly distributed on the accumulation plane for large Φ ($\chi \approx 1$), which is a result of the lack of time for the surface to deform (remember that the surface is represented by randomly initialized particles). At intermediate Φ , we start to observe considerable inhomogeneities characterized by $\chi > 1$. A maximum of clustering is found between $\Phi = 1$ and 4 when the particle settling velocity v_s is of the same order as the root-mean-square fluid velocity U . Note also that the accumulation plane would be reached during one unit of the integral time scale T by a particle uniformly settling with Φ between 1.5 and 2.5 in all simulations [see Table I]. Decreasing Φ further results in a slight decrease in χ .

Figure 3(a) also shows that the limiting value of χ for $\Phi \rightarrow 0$ strongly depends on the Reynolds number. For any Φ , in fact, a higher Re_λ implies a smaller χ . This result means that

inhomogeneities at the Kolmogorov length scale are actually attenuated as the velocity field becomes increasingly complicated, which can be attributed to an increased mixing.

One may, of course, also compare inhomogeneities observed at the same spatial resolution L_B in flows with different Kolmogorov scale and Re_λ . Results are shown for a large and a small L_B in Figs. 3(b) and 3(c), respectively. While the characteristics of the individual lines are the same as in Fig. 3(a), curves for different Re_λ cross at a value of Φ a bit above $\Phi = 1$. That is, it depends on the settling velocity whether increasing turbulence strength attenuates or enhances inhomogeneity observed at a given spatial resolution. The settling parameter of Fig. 2 is just large enough to fall into the latter category.

It is worth noting that inhomogeneities observed at a small resolution L_B are typically weaker than those at a larger resolution for any given Reynolds number: compare the range of χ between Figs. 3(b) and 3(c) and see Figs. 3(d) and 3(e) for a direct representation for given (large) values of Φ . On the finest spatial scales, where

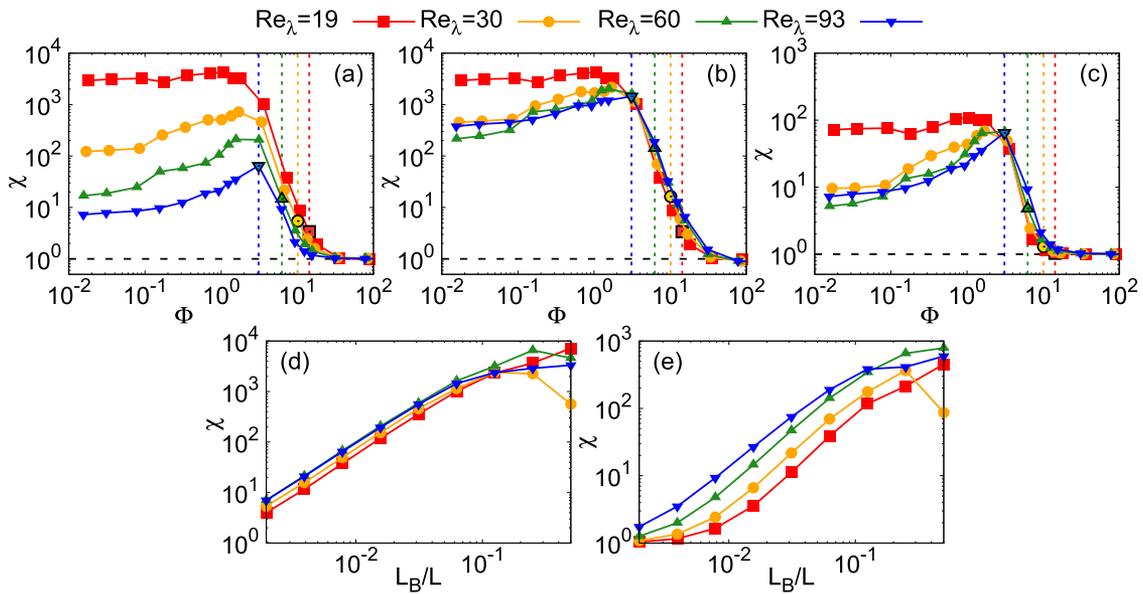


FIG. 3. The Poisson dispersion index χ computed for the two-dimensional horizontal distribution of particles at the accumulation level as a function of the settling parameter Φ [(a)–(c)] and the coarse-graining box size L_B [(d) and (e)] for different values of Re_λ . $L_B = 2L/M$ (a), $L_B = L/16$ (b), $L_B = L/128$ (c), $\Phi = 3.35$ (d), and $\Phi = 6.7$ (e). The settling parameter Φ corresponding to $St = 1$, an upper bound of the validity range of 3, is marked by a black contoured symbol and a dashed line for each Re_λ .

for fast settling, initial randomness dominates over later mixing, χ appears to converge to 1.

We now see that the degree of observed inhomogeneity strongly depends on the spatial resolution, but its dependence on the settling velocity and on the turbulence strength (Re_λ) is robust for any given resolution. We can conclude about the existence of two regimes from the point of view of parameter dependence, one for large Φ and the other for small Φ , where the effect of increasing mixing by the flow is opposite. We will elaborate on this point and on the crossover between the two regimes in Subsection IV C, where we analyze the mechanisms underlying our observations.

When using the correlation dimension^{1,13,18} for estimating inhomogeneities as a function of Φ (not shown), the same qualitative behavior is observed as with the Poisson dispersion index. This suggests that our conclusions are robust, and they do not depend on the choice of the particular statistical quantifier.

C. Statistics of stretching and projection over trajectories

We attempt to explore the mechanisms leading to the dependence of χ on Φ and Re_λ presented in Fig. 3 by investigating corresponding properties of the two mechanisms contributing to inhomogeneities, namely, the stretching and the projection effects. For the statistical quantification of their local characteristics, we treat different branches of the sedimenting surface separately, without any summation. Furthermore, at difference with Sec. IV B,^{23,30} we explore in this section the statistics with respect to the uniform distribution of particles in the initial layer, or equivalently, we weight each particle trajectory equally. This provides a point of view complementary to the statistics over boxes in the collecting layer

explored in Sec. IV B to compute σ_n and χ . In particular, we compute here arithmetic averages $\langle A \rangle$, standard deviations σ_A , and correlation coefficients of $A = S$, P , and also F over the individual values obtained for the individual particles, e.g.,

$$\langle A \rangle = \frac{1}{N} \sum_{k=1}^N A_k, \quad (8)$$

where k runs over different particles. In the limit of infinitely many particles,

$$\begin{aligned} \langle A \rangle &= \frac{\int A(\mathbf{x}_0) d^2 \mathbf{x}_0}{\int 1 d^2 \mathbf{x}_0} \\ &= \frac{\int A(\mathbf{x}_h) F(\mathbf{x}_h) d^2 \mathbf{x}_h}{\int F(\mathbf{x}_h) d^2 \mathbf{x}_h}, \end{aligned} \quad (9)$$

where the $d\mathbf{x}_0$ integrals are taken over the complete initial release plane, and the integral over $d\mathbf{x}_h$ over each branch of the surface sedimented on the collecting plane with a subsequent summation of the values obtained for the different branches. We have used that the number of particles is conserved, $\rho_0 d^2 \mathbf{x}_0 = \rho(\mathbf{x}_h) d^2 \mathbf{x}_h$. The second expression in (9) illustrates why such a uniform weighting according to the initial (uniform) distribution of the particles is equivalent to weighting the points in the collecting plane with the total density factor F (or the final density at those points if the sedimenting surface reaches the collecting plane in a single branch). Note that this kind of evaluation for a finite number N of particles corresponds to an “implicit” coarse-graining on the collecting plane, with a grid provided by the particles’ positions.

To better understand the contribution of stretching and projection to the inhomogeneities, we first report in Fig. 4 the probability

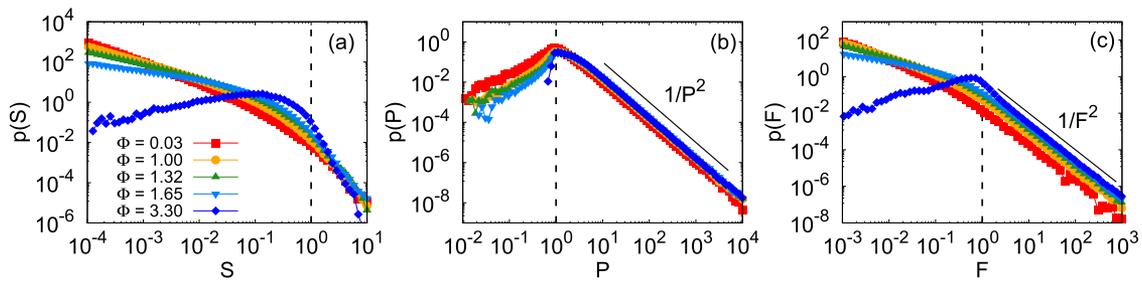


FIG. 4. Probability density functions (statistics over released particles) of stretching $p(S)$ (a), projection $p(P)$ (b), and the total density factor $p(F)$ (c) at $Re_\lambda = 60$ for the indicated values of the settling parameter Φ .

density functions of S , P , and F over the individual trajectories. The distribution of F combines the behavior of S and P . The low values of the total density factor $F \ll 1$ are controlled by low values of stretching, whereas large values $F \gg 1$ are produced by the large values in P , associated with caustics. The distributions of stretching appear to behave as power laws for small values of the settling parameter Φ . When Φ is below roughly 1 [the value giving the maximum of clustering; see Figs. 3(a)–3(c)], the weight given to very small values of S increases as Φ decreases, since the areas of the surface elements arriving on the collecting plane can grow without limits. On the contrary, the distribution of P remains mostly unchanged for varying values of Φ and does not depend on Re_λ (not shown), revealing a universal geometric feature of the projection near caustics. Indeed, in Fig. 4, we observe $p(P) \sim P^{-2}$ and $p(F) \sim F^{-2}$ for $P \gg 1$ and $F \gg 1$, respectively, which can be explained by the formation of caustics. It is a well-known result that, generically, the density profile at

a line caustic diverges as $F \propto x^{-1/2}$, where x is the transverse spatial distance to the caustic.³⁸ Considering the transformation between variables x and F (assuming homogeneity in the direction parallel to the caustic), $p(F)dF = p(x)dx$, and that, as seen before, the density factor gives the proper weight to the horizontal locations $p(x) \sim F$, one obtains $p(F) = |dx/dF|p(x) \propto F^{-2}$.

To place the corresponding properties of S and P into a narrower context, we now investigate the average and standard deviation (statistics over released particles) of the total density factor F . $\langle F \rangle$ and σ_F are plotted in Fig. 5. The former characterizes the average dilution ($\langle F \rangle < 1$) or concentration ($\langle F \rangle > 1$) of particles on the collecting plane with respect to the initial release density ρ_0 (remember that different branches generated by folding of the falling surface are treated separately). Meanwhile, σ_F describes the degree of inhomogeneity among the different particles.

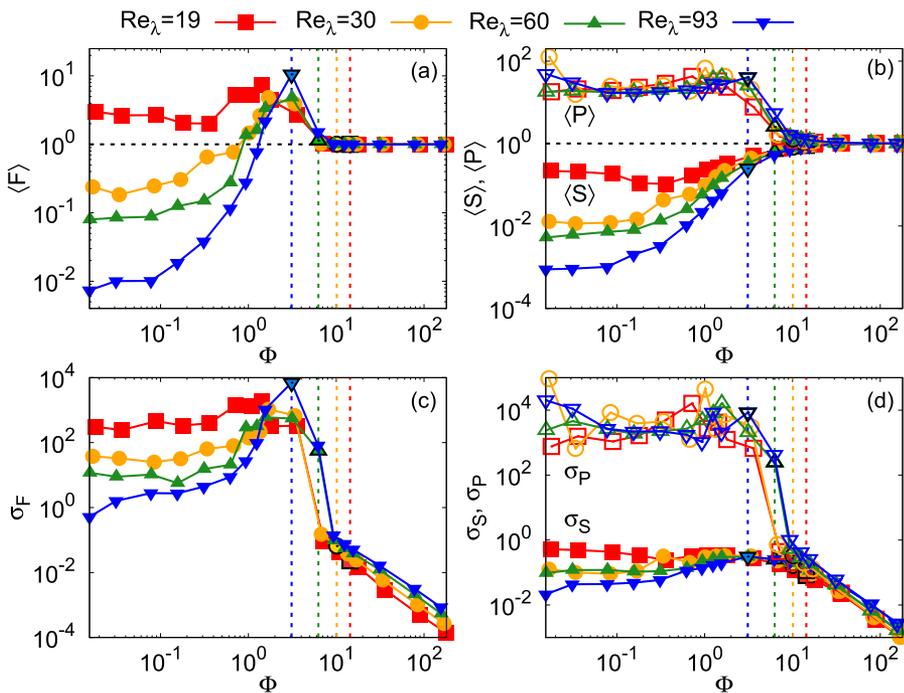


FIG. 5. Average and standard deviation of F , S , and P for different values of Re_λ as a function of the settling parameter Φ . (a) Average of F , (b) average of S and P , (c) standard deviation of F , and (d) standard deviation of S and P . The settling parameter Φ corresponding to $St = 1$, an upper bound of the validity range of 3, is marked by a black contoured symbol and a dashed line for each Re_λ .

The shape of σ_F as a function of Φ in Fig. 5 is very similar to that of χ in Fig. 3 (except for the large- Φ asymptotics, of course), which suggests that the accumulated inhomogeneities (represented by χ) are closely related to the trajectory-wise processes of stretching and projection, as opposed to summation of the density over different branches of the sedimented surface, which could also have a dominating effect. Note, however, that a quantitative comparison would be difficult so that summation may well be important too.

For increasing Φ , $\langle F \rangle$ converges to 1, as expected in the lack of time for deformation and bending, while it generally exhibits a shift toward net dilution, or area expansion, for decreasing Φ below $\Phi \approx 1$, which will be understood by analyzing S and P . Between $\Phi = 1$ and 10, $\langle F \rangle$ exhibits a prominent maximum, just like χ and σ_F . This maximum suggests again that either $v_s \approx U$ or a settling time near the integral time scale (or both circumstances) results in a kind of resonance where maximal net deformation and maximal inhomogeneity in the deformation take place. This resonance represents, furthermore, a crossover between the regimes of large and small Φ with different tendencies.

Very close to $\Phi = 1$, just as for χ , we find a crossing in the Re_λ -dependences too: for $\Phi < 1$, an increasing Reynolds number results in a shift toward net dilution or expansion (decreasing $\langle F \rangle$) and a decrease in inhomogeneity (σ_F), but these tendencies revert near $\Phi = 1$. We conclude that the effects of increasing the strength of turbulence are far from trivial but are certainly different in the regimes of small and large settling parameter.

Much insight becomes accessible about explanations for the above tendencies by analyzing the statistics of S and P . The decay of both σ_S and σ_P in Fig. 5 follows the same power law for large Φ as σ_F . σ_S is not affected much by the resonance between $\Phi = 1$ and 10 but typically becomes slightly decreasing for $\Phi < 1$ while exhibiting only a minor degree of inhomogeneity there. Based on this observation and the similar magnitudes of σ_F and σ_P in Fig. 5, most of the inhomogeneity in F near and below the resonance might appear to originate from the inhomogeneity of P . Note, however, the rather erratic behavior, lacking a clear tendency, of σ_P for decreasing Φ , contrasting the behavior of σ_F . While this relationship will be further commented on later and a comprehensive understanding of all aspects is beyond the scope of the current analysis, a universal conclusion about the small- Φ behavior of the degree of inhomogeneity in any quantity seems to be a convergence to some constant value, despite the arbitrarily long time available for deformation for $\Phi \rightarrow 0$.

This behavior of the standard deviation appears to apply to mean values as well, as Fig. 5 illustrates for $\langle S \rangle$ and $\langle P \rangle$. With $\langle P \rangle$ being mostly constant for $\Phi < 1$, the decrease in $\langle F \rangle$ can naturally be linked to the decrease in $\langle S \rangle$ for decreasing Φ observed in this regime in Fig. 5. The decrease in $\langle S \rangle$ below 1 actually describes a stretching (expansion, corresponding to a dilution of the density) of increasing strength, which is presumably related to the longer time available for the development of deformation. The same effect may underlie the even sharper response of $\langle P \rangle$ for decreasing Φ between 10 and 1, before the increase in $\langle P \rangle$ saturates. The difference in the sharpness and the saturation of $\langle P \rangle$ is what gives rise to the resonance-like behavior in $\langle F \rangle$, even though $F = SP$ only pointwise, and $\langle F \rangle \neq \langle S \rangle \langle P \rangle$ in general due to spatial correlations.

Figure 5 also provides with the opportunity to study the effects of varying the Reynolds number. Both $\langle P \rangle$ and σ_P depend weakly

and irregularly on Re_λ for $\Phi < 1$. This might be regarded as an indication of a saturation in all effects of projection, which cannot be enhanced further by modifying the circumstances (Reynolds number and settling parameter). The explanation of such a saturation might be the reaching of a “maximal randomness” in the orientation of the normal vector of an arbitrarily chosen point of the sedimenting surface.²³

The fact that $\langle S \rangle$ does not saturate but decreases with increasing Re_λ in the same range of Φ (Fig. 5; similar to $\langle F \rangle$) suggests that a similar saturation is not reached in the stretching, the net effect of which may grow without any limit, as shown by the distribution of S in Fig. 4(a). The dependence of $\langle S \rangle$ (and $\langle F \rangle$) on Re_λ might simply be understood as stronger deformation resulting from stronger turbulence. Especially, in view of this, explaining why inhomogeneity is attenuated with the increase in Re_λ as indicated by σ_S might be linked to the long-term homogenization in an increasingly complicated flow with the increase in mixing capability. The attenuation of inhomogeneity with the decrease in Φ might be explained in a similar way but relying on the longer time available for mixing instead of the increasing mixing capability of the flow. How σ_S depends on Re_λ and Φ for $\Phi < 1$ appears to be transferred to σ_F (Fig. 5), which suggests that inhomogeneities in stretching do have an important effect on the final inhomogeneities despite their much smaller magnitude.

By now, mixing is understood to be a central process in shaping the inhomogeneities for $\Phi < 1$. We have seen that more mixing (smaller Φ or larger Re_λ) attenuates inhomogeneities on the long term (at least when investigated at a predefined spatial resolution, which is determined here by the finite number of particles, cf. Sec. IV B). Without mixing, however, there would be no inhomogeneities at all.

We resolve this apparent contradiction by considering the short-term effects of mixing. In particular, when compared to the small- Φ regime, mixing works in the opposite way in the large- Φ regime. That is, σ_P and σ_S (also $\langle P \rangle$ and $\langle S \rangle$) increase with the increase in Re_λ (see Fig. 5). The presumable explanation precisely lies in the time available for mixing, which is around or less than the integral time scale. It seems plausible that saturation is not reached in the effects of the projection, nor homogenization is performed, which is confirmed by σ_P and σ_S growing from 0 with the decrease in Φ and increase in Re_λ in Fig. 5. As long as the sheet is not deformed very much, stronger turbulence or longer time naturally results in the intensification of both the net effects of deformation and their inhomogeneity. For the net effects $\langle P \rangle$ and $\langle S \rangle$, this is similar to the $\Phi < 1$ regime except that $\langle P \rangle$ saturates there.

Comparing the Re_λ -dependence of $\langle S \rangle$ and $\langle P \rangle$ for $\Phi \gtrsim 1$, the former becomes weaker than the latter, and this is what we suppose to yield a change in the dependence of $\langle F \rangle$ on Re_λ between the two regimes. At the same time, the similar change for σ_F is more straightforwardly explained by the same change for σ_S and σ_P , corresponding to an inherent difference between the short-term and long-term behaviors. It is interesting to observe that introducing stronger turbulence enhances and attenuates inhomogeneities before and after the crossover.

So far, we have learnt that observable inhomogeneities (as in Fig. 3) are strongly determined by trajectory-wise processes (investigated in Figs. 4 and 5). Increasing mixing by the flow has been identified to introduce and enhance inhomogeneities on the short

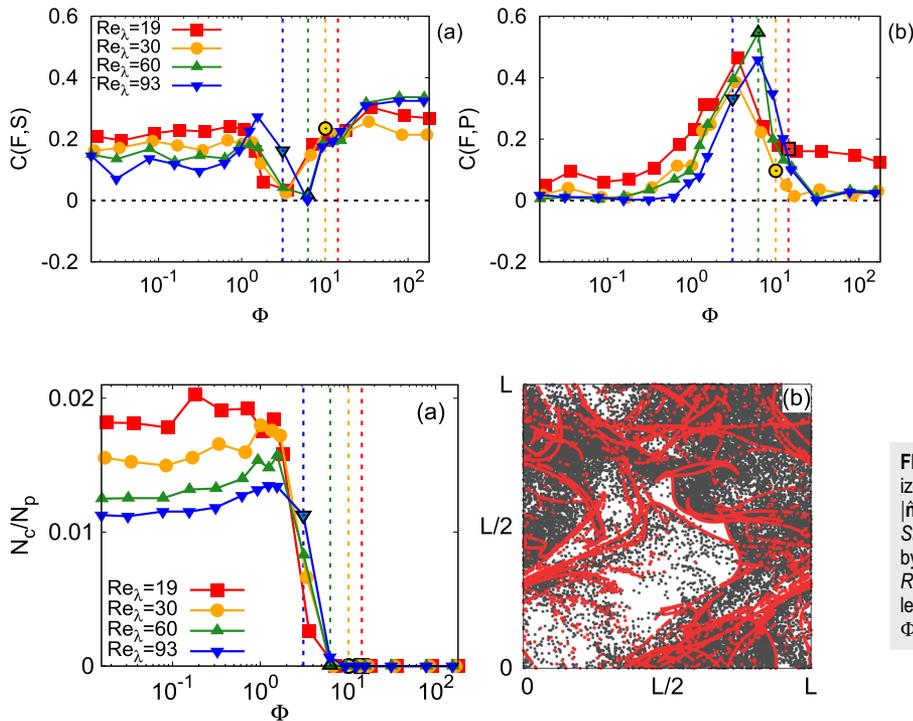


FIG. 6. Pearson correlation coefficients: (a) $C(F, S)$ and (b) $C(F, P)$ for different values of Re_λ as a function of the settling parameter Φ . The settling parameter Φ corresponding to $St = 1$, an upper bound of the validity range of 3, is marked by a black contoured symbol and a dashed line for each Re_λ .

FIG. 7. (a) Fraction of particles N_c/N_p in caustics characterized by having $\hat{\mathbf{n}} \cdot \mathbf{v} = 0$ (numerically, the requirement is $|\hat{\mathbf{n}} \cdot \mathbf{v}| < 0.01$). The settling parameter Φ corresponding to $St = 1$, an upper bound of the validity range of 3, is marked by a black contoured symbol and a dashed line for each Re_λ . (b) Positions of the particles sedimented on the collecting plane. In red, particles in caustics as defined above. $\Phi = 1$ and $Re_\lambda = 19$.

term and to attenuate them on the long term (at the spatial resolution corresponding to the finite number of particles). We point out, however, that summation over the increasingly numerous branches of the falling layer may contribute to the attenuation of inhomogeneities. Irrespective of that, the strongest inhomogeneities have been linked to the projection of the falling layer onto the accumulation plane (e.g., caustics). However, the impacts of projection have been found to saturate after entering the well-mixed regime, where the parameter dependence of stretching effects appears to be dominant, conforming with the above-mentioned attenuation of observable inhomogeneities.

The latter claims, indirectly derived from Fig. 5, are supported by an analysis of spatial correlations. We compute the Pearson correlation coefficient of F with S and P [$C(F, S)$ and $C(F, P)$] using statistics over particles. The value of the correlation coefficient is influenced by both net effects and inhomogeneities. If stretching and projection were uncorrelated (which is not the case, see Fig. 2), we would have $C(F, S) = \frac{\sigma_S}{\sigma_F} \langle P \rangle$ and a corresponding formula for $C(F, P)$, which suggests that both averages and standard deviations are relevant.³⁰

Results for the correlation coefficients are plotted in Fig. 6. For increasing Φ beyond the crossover, stretching seems to be dominant in forming the spatial structures of the final density (although this is not observed for all Reynolds numbers). This suggests that undulations of the surface become negligible compared to the effect of stretching for increasing Φ , in accordance with the same conclusion of Ref. 30. In the vicinity of the resonance, projection takes over, and the correlation with stretching falls to zero. This is presumably due to the increase in the inhomogeneity of projection, without a similar increase in stretching [see Fig. 5(d)]. Such a result is in

agreement with the qualitative observation of the spatial distributions in the proximity of the resonance, displayed in Fig. 2, where the filamentary structures of P and F appear to be well correlated. The dominance reverts again for $\Phi < 1$, which is in accordance with the observation that both $\langle F \rangle$ and σ_F follow the corresponding features of S (both as a function of Φ and Re_λ). In relation with Fig. 5, we explained this via the saturation of P , corresponding to the unit vector $\hat{\mathbf{n}}$ normal to the (wrapped and contorted) surface taking already a random orientation, which cannot become more disordered by decreasing Φ .

In Fig. 7, we present further evidence supporting that the reason for the huge increment in P in vicinity of the resonance (coming from large Φ where the surface is flat, see Fig. 5) is that caustics appear where the density formally diverges. Thus, in Fig. 7(a), we plot the fraction of particles in caustics (numerically requiring $|\hat{\mathbf{n}} \cdot \mathbf{v}| < 0.01$) as a function of Φ , observing the two regimes: very small for large values of Φ and non-negligible values for $\Phi < 1$ and any values of Re_λ . The example of Fig. 7(b) illustrates by a direct plotting of the positions of the particles on the accumulation plane that caustics are closely related to inhomogeneities in the sedimented particles' distribution. The filamentary pattern of caustic lines in Fig. 7(b) is recognized to be the same as that of the maxima of P in Fig. 2 and spreads through the collecting plane.

V. CONCLUSIONS

We performed direct numerical simulations of sinking non-inertial particles in a turbulent flow, exploring a range of settling velocities and Reynolds numbers. We focused our attention to the inhomogeneities of the particle distribution that take place when

particles released on a plane at a fixed height are collected on a certain accumulation depth.

Although the Lagrangian dynamics is incompressible, advection of the two-dimensional surface by the flow and accumulation on a plane can lead to the emergence of inhomogeneities by a combination of stretching and projection effects.^{23,30} Our results indicate the existence of two different regimes from this point of view: the inhomogeneities grow during the initial stages of the dispersion, while they undergo attenuation when approaching the long-term asymptotics of a well-mixed state. With a fixed domain size, the settling time and thus the degree of inhomogeneity in the accumulated density are controlled by the settling velocity: the initial and the long-term regimes are realized for large and small settling velocity, respectively.

Between the two regimes, we have found a “resonant” range of settling velocity where inhomogeneity can become maximal. The maximum might approximately be determined by the coincidence of the settling velocity with the root-mean-square velocity of the flow, by the coincidence of the typical settling time with the integral time scale of the flow, or by an interplay of the two circumstances.

The range of settling velocity hosting this resonance-like behavior not only marks a change in behavior of the degree of inhomogeneity as a function of the settling velocity itself but also as a function of the Reynolds number. During the initial transients, a more complicated flow (higher Reynolds number) enhances inhomogeneity, while it facilitates approaching homogeneous mixing in the regime leading to the long-term asymptotics.

We have also investigated the contributions of the two basic inhomogenizing mechanisms in the two regimes. For large settling velocities, when the surface is bended very little without developing overhangs, stretching is predominant. When getting close to resonant-like settling velocities, folds appear, yielding projection caustics in the sedimented density. For this reason, effects of projection become dominant, and this is responsible for the crossover in some properties at the resonance-like region. With further decrease in the settling velocity, the magnitude of the inhomogeneities remains determined by projection, but the increasing effects of projection saturate soon as mixing becomes strong. The parameter dependence of observable inhomogeneities then conforms with the increasing homogeneity of stretching as mixing becomes stronger, although summation over a large number of different branches of the falling particle layer may also be important.

The above results give an opportunity to comment on some previous work. Although our setup shows a few important differences with the problem of sedimentation in mesoscale oceanic flows addressed in Ref. 30, common points are prominent enough to locate the mesoscale oceanic setup on the axis of the settling velocity. In particular, although anisotropy in the velocity field of the ocean is pronounced (with large differences between horizontal and vertical velocities), one can safely state that the settling velocity of typical biogenic particles²¹ is (several times) larger than vertical velocities of flow. As for the typical sedimentation time, it is the same order of magnitude as the characteristic time scale of the mesoscale oceanic flow. These circumstances may mean that the parameters are not far from the resonance-like maximum of inhomogeneity and fall into the regime of initial transients identified for $\Phi > 1$ in the present paper. Considerable inhomogeneities appear in the

corresponding oceanic simulations and are enhanced for decreasing settling velocity and increasing mesoscale turbulence strength.³⁰

Finally, we indicate the relevance of these studies for sinking biogenic particles in an eddy-resolving oceanic velocity field. A careful study has been performed in Ref. 30 for a mesoscale oceanic flow, based on the analysis of Ref. 21 of sizes and densities of particles for which our modeling approach is valid. These biogenic particles (examples of which are dead plankton bodies, zooplankton fecal pellets, or small aggregates and marine snow) have typical sizes a ranging between 10^{-6} m and 10^{-3} m and typical densities between $\rho_p = 1050$ kg m⁻³ and 2700 kg m⁻³ so that β is bounded between 0.5 and 1.0. Oceanic turbulence is characterized by $\varepsilon = 10^{-4}$ m²/s³ – 10^{-8} m²/s³, $\nu = 10^{-6}$ m²/s,⁴⁰ for which we obtain a Kolmogorov length scale $\eta = (\nu^3/\varepsilon)^{1/4} = 0.3$ mm – 3 mm, Kolmogorov time scale $\tau_\eta = (\nu/\varepsilon)^{1/2} = 0.1$ s – 10 s, Kolmogorov velocity $u_\eta = (\nu\varepsilon)^{1/4} = 0.3$ mm/s – 3 mm/s, and acceleration $a_\eta = (\varepsilon^3/\nu)^{1/4} = 30$ mm/s² – 0.03 mm/s², leading to a Froude number $Fr = \times 10^{-6}$ – $\times 10^{-3}$ and a Stokes number $St = 10^{-7}$ – 0.5. The range of Reynolds numbers used in our numerical simulations indicates that we are dealing with spatial scales of the flow between 6 cm and 1 m. Note that our configuration represents two relevant situations in this context: one is sedimentation on the seafloor and the other is the collection of particles in sediment traps located at a given depth. For this last situation, the impact of boundary conditions at the bottom is irrelevant. However, for sedimentation on the seafloor, in the case of a no-slip boundary condition, a boundary layer close to the bottom is formed and turbulence is drastically reduced and modified there, which does not affect the processes in the bulk.³⁰

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DATA AVAILABILITY

The data that support the findings of this study are available from the corresponding author upon reasonable request.

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