Caustics in turbulent aerosols form along the Vieillefosse line at weak particle inertia

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(Received 28 July 2022; accepted 26 January 2023; published 13 February 2023)

Caustic singularities of the spatial distribution of particles in turbulent aerosols enhance collision rates and accelerate coagulation. Here we investigate how and where caustics form at weak particle inertia, by analyzing a three-dimensional Gaussian statistical model for turbulent aerosols in the persistent limit, where the flow varies slowly compared with the particle relaxation time. In this case, correlations between particle- and fluid-velocity gradients are strong and caustics are induced by large strain-dominated excursions of the fluid-velocity gradients. These excursions must cross a characteristic threshold in the plane spanned by the invariants Q and R of the fluid-velocity gradients. Our method predicts that the most likely way to reach this threshold is by a unique optimal fluctuation that propagates along the Vieillefosse line $27R^2/4 + Q^3 = 0$. We determine the shape of the optimal fluctuation as a function of time and show that it is dominant in numerical statistical-model simulations even for moderate particle inertia.

DOI: 10.1103/PhysRevFluids.8.024305

I. INTRODUCTION

In turbulent aerosols, particle inertia allows heavy particles to detach from the flow and generate folds over configuration space, so-called caustics [1–4]. Near such caustic folds, phase-space neighborhoods partially focus onto configuration space, leading to transient divergencies of the spatial particle-number density, analogous to caustics in geometrical optics [5]. At the same time, caustics delineate regions in configuration space where the particle velocities are multivalued, a phenomenon known as the sling effect [2,6]. It leads to anomalously large relative particle velocities [2,7–9], increased collision rates [2,10–13], and large collision velocities [7–9,14]. The latter can significantly affect collision outcomes [15,16].

The process of caustic formation depends sensitively on the characteristic dynamical timescales of the fluid and of the particles [17]. For a dilute monodisperse suspension of dense identical particles, there are three timescales: the Eulerian correlation time τ_c of the flow, its Lagrangian correlation time τ_K , measured along the paths of fluid elements, and the particle relaxation time τ_p , the time it takes the inertial-particle velocity to relax back to the fluid velocity. The particle dynamics is characterized by two dimensionless numbers: The Stokes number $St \equiv \tau_p/\tau_K$ is a dimensionless measure of particle inertia, while the Kubo number $Ku \equiv \tau_c/\tau_K$ determines the degree of persistence of the flow.

The rate of caustic formation has been measured using direct numerical simulation of turbulence [6,18] and kinematic turbulence models [19,20]. Yet the mechanisms of caustic formation are understood only in certain idealized models and limiting cases. In the so-called white-noise limit [17], the fluid velocities seen by the particles are assumed to fluctuate rapidly (Ku \ll 1) and particle inertia is assumed to be large (St \gg 1). For Gaussian random velocity fields in this limit,

caustic formation can be described within a diffusion approximation which predicts that the particle dynamics depends sensitively on the (small) value of the Kubo number [17].

Aerosol models with Gaussian random velocity fields that realistically mimic homogeneous and isotropic turbulence, by contrast, require Kubo numbers of order 10 and are thus in a parameter regime where the particle dynamics depends only weakly upon Ku [17]. This is a motivation for analyzing the opposite so-called persistent limit of large Kubo numbers and small Stokes numbers. It corresponds to slowly varying fluid velocities and short particle relaxation times. Model calculations in one [21–23] and two [24] spatial dimensions show that caustics form in the persistent limit when the matrix of fluid-velocity gradients reaches a large threshold. Consequently, the most likely way for caustics to form in this limit is by rare large fluctuations of the fluid-velocity gradients that exceed this threshold. For small St, these fluctuations are dominated by a single optimal fluctuation that can be calculated using methods from large-deviation theory [25–27], recently reviewed in, e.g., [28,29].

In this paper we compute the optimal fluctuation of the fluid-velocity gradient matrix \mathbb{A} and the corresponding most likely (optimal) path to caustic formation for an incompressible turbulent flow, by analyzing a three-dimensional statistical model in the persistent limit. We first characterize \mathbb{A} in terms of its invariants $Q = -\text{Tr}(\mathbb{A}^2)/2$ and $R = -\text{Tr}(\mathbb{A}^3)/3$ [30,31] and show that it must cross a characteristic threshold line in the Q-R plane for a caustic to form. We then employ large-deviation techniques [28,29] to determine how this threshold is reached. Our method predicts that the optimal fluctuation, given by the most likely way to reach the threshold, moves along the positive branch (R > 0) of the so-called Vieillefosse line [32–35],

$$\frac{27}{4}R^2 + Q^3 = 0, (1)$$

which forms the boundary between extensional and rotational flow configurations. The optimal fluctuation is vorticity-free and unique up to similarity transformations, but exact only for St $\ll 1$. Nevertheless, numerical simulations of a Gaussian statistical model show that it remains dominant even at St numbers as large as St ≈ 0.3 , suggesting a distinct way of caustic formation in three spatial dimensions that can be tested in experiments [36]. In high-Reynolds-number turbulence, the probability distribution of Lagrangian fluid-velocity gradients is strongly skewed along the positive branch of the Vieillefosse line [37–39]. This means that caustic formation is expected to be significantly higher in actual turbulent flows compared with models that neglect this skewness.

II. PROBLEM FORMULATION

The dynamics of a spherical particle in a fluid-velocity field u(x, t) is well approximated by Stokes's law [40]

$$\frac{d}{dt}\mathbf{x}(t) = \mathbf{v}(t), \quad \frac{d}{dt}\mathbf{v}(t) = \tau_{\rm p}^{-1}\{\mathbf{u}[\mathbf{x}(t), t] - \mathbf{v}(t)\},\tag{2}$$

given that the particle is small enough and much denser than the fluid. Here x and v denote particle position and velocity. The particle relaxation time $\tau_p = 2a^2\rho_p/9\rho_f v$ is a function of the particle size a, the kinematic viscosity v of the fluid, and the particle and fluid densities ρ_p and ρ_f , respectively.

In *d* spatial dimensions, a caustic occurs when the separation vectors between d + 1 nearby particles partially align so that the spatial volume $\hat{\mathcal{V}}(t) = |\det \mathbb{J}(t)|$, occupied by the particles, momentarily collapses to zero. Here \mathbb{J} denotes the spatial deformation tensor $J_{ij}[\mathbf{x}(t_0), t] = \frac{\partial x_i(t)}{\partial x_j(t_0)}$. Caustic formation is closely related to the dynamics of the particle-velocity gradients [2,17], written in dimensionless form as

$$\operatorname{St}\frac{d}{dt}\mathbb{Z}(t) = -\mathbb{Z}(t) - \mathbb{Z}(t)^{2} + \mathbb{A}(t), \qquad (3)$$

with $\mathbb{Z}(t_0) = \mathbb{A}(t_0)$ initially. Here $Z_{ij}(t) = \tau_p \partial v_i(t) / \partial x_j(t)$ and $A_{ij}(t) = \tau_p \partial u_i(t) / \partial x_j(t)$ are dimensionless matrices of Lagrangian particle-velocity gradients and fluid-velocity gradients, respectively,

evaluated at the particle position, i.e., $\mathbb{Z}(t) = \mathbb{Z}[\mathbf{x}(t), t]$ and $\mathbb{A}(t) = \mathbb{A}[\mathbf{x}(t), t]$. Time *t* in Eq. (3) is measured in units of the Kolmogorov time $\tau_{\rm K} = \tau_{\rm p} \langle {\rm Tr} \mathbb{A} \mathbb{A}^{\mathsf{T}} \rangle_f^{-1/2}$, where $\langle \cdots \rangle_f$ denotes a Lagrangian average along the paths of fluid elements. The particle-velocity gradients \mathbb{Z} are related to the spatial volume $\hat{\mathscr{V}}(t)$ through

$$\hat{\mathscr{V}}(t) = \hat{\mathscr{V}}(t_0) \exp \int_{t_0}^t ds \operatorname{Tr} \mathbb{Z}(s).$$
(4)

Consequently, a necessary condition for caustic formation $\hat{\mathcal{V}}(t) \to 0$ is that $\text{Tr}\mathbb{Z}(t)$ escapes to negative infinity.

In the small-St limit, the particle dynamics (2) and (3) occurs on much shorter timescales than the dynamics of \mathbb{A} , so the problem can be treated using the persistent limit [21,23,24]. In this limit, changes of \mathbb{A} in Eq. (2) are adiabatic, i.e., \mathbb{A} remains effectively constant [23,24] as \mathbb{Z} varies. Whenever Eq. (3) has a globally attracting stable fixed point \mathbb{Z}_* , where

$$\mathbb{F}(\mathbb{Z}_*, \mathbb{A}) \equiv -\mathbb{Z}_* - \mathbb{Z}_*^2 + \mathbb{A} = 0, \tag{5}$$

 \mathbb{Z} rapidly relaxes to \mathbb{Z}_* within a short transient time of order St. In this case, $\operatorname{Tr}\mathbb{Z}$ remains finite and no caustics form. In the absence of stable fixed points, by contrast, the dynamics drives the particle gradient matrix \mathbb{Z} to infinity, so $\operatorname{Tr}\mathbb{Z} \to -\infty$, implying the formation of a caustic.

In summary, at small St caustics form through bifurcations of the fixed-point equation (5), induced by large but adiabatically slow excursions of \mathbb{A} . The probability and shape of these excursions are determined by optimal fluctuation theory [28,29], which predicts the most likely excursion that leads to a bifurcation and thus to a caustic.

III. FIXED-POINT ANALYSIS

To find the bifurcations of the fixed points $\mathbb{Z}_*(\mathbb{A})$ as \mathbb{A} changes adiabatically, we note that Eq. (5) satisfies

$$\mathbb{F}(\mathbb{P}^{-1}\mathbb{Z}\mathbb{P}, \mathbb{P}^{-1}\mathbb{A}\mathbb{P}) = \mathbb{P}^{-1}\mathbb{F}(\mathbb{Z}, \mathbb{A})\mathbb{P},$$
(6)

where \mathbb{P} is an arbitrary invertible matrix. As a consequence, if $\mathbb{Z}_*(\mathbb{A})$ is a fixed point, then so is $\mathbb{P}^{-1}\mathbb{Z}_*(\mathbb{P}^{-1}\mathbb{A}\mathbb{P})\mathbb{P}$. Due to the symmetry (6), the set of fixed points and their stability are invariant under similarity transformations [41] and thus only depend on the invariants of \mathbb{A} . For incompressible flow, \mathbb{A} is traceless and obeys the characteristic equation [42,43]

$$\mathbb{A}^{3} - \frac{1}{2} \operatorname{Tr}(\mathbb{A}^{2}) \mathbb{A} - \det(\mathbb{A}) \mathbb{1} = 0$$
⁽⁷⁾

with invariants [30]

$$Q(\mathbb{A}) = -\frac{1}{2} \operatorname{Tr}(\mathbb{A}^2), \quad R(\mathbb{A}) = -\det(\mathbb{A}) = -\frac{1}{3} \operatorname{Tr}(\mathbb{A}^3).$$
(8)

An analogous characteristic equation holds for \mathbb{Z}_* , but with a nonvanishing trace term. We characterize the fixed points \mathbb{Z}_* by evaluating $\mathcal{Z}_* \equiv \text{Tr}\mathbb{Z}_*$. To this end, we raise the fixed-point equation $\mathbb{A} = \mathbb{Z}_* + \mathbb{Z}_*^2$ to the second and third powers and take the trace. This gives Q and R in terms of $\text{Tr}(\mathbb{Z}_*^n)$, with $n \leq 6$. From the characteristic equation of \mathbb{Z}_* we obtain, by multiplication with \mathbb{Z}_* and taking the trace, expressions for $\text{Tr}(\mathbb{Z}_*^n)$ that we substitute into the equations for Q and R. In this way, we end up with one single equation for \mathcal{Z}_* :

$$-16[2\mathcal{Z}_*+3]^2R = \{\mathcal{Z}_*[\mathcal{Z}_*+1]^2[\mathcal{Z}_*+2] - 4Q\}\{[\mathcal{Z}_*+1][\mathcal{Z}_*+2]^2[\mathcal{Z}_*+3] - 4Q\}.$$
 (9)

From Eq. (9) we conclude that possible fixed-point values \mathbb{Z}_* depend only on the invariants Q, and R of \mathbb{A} . Equation (3) is unstable in regions in the Q-R plane where Eq. (9) has no solution, allowing $\mathrm{Tr}\mathbb{Z}$ to escape to $-\infty$, so that a caustic forms.

In general, regions with different numbers of solutions of Eq. (9) (ranging from zero to eight, see discussion below) are separated by saddle-node bifurcations of $\mathbb{F}(\mathbb{Z}_*, \mathbb{A}) = 0$, where fixed points are



FIG. 1. Phase diagram with regions of different numbers of solutions of Eq. (9) as a function of Q and R. The lines show the bifurcations described in the text. The green square (R, Q) = (0, 0) shows the most likely gradient configuration at steady state, while for infinitesimal particle inertia the threshold (10) is reached at the red square $(R, Q) = (\frac{1}{256}, -\frac{3}{64})$ (see Sec. IV).

created or destroyed [44]. Exceptions are bifurcations at isolated points or on symmetry lines where other, e.g., higher-order, bifurcations may occur [44]. To find the locations of the saddle-node bifurcations in the Q-R plane, we introduce the coordinate $z_* = (Z_* + 3/2)^2$, because it turns Eq. (9) into a quartic polynomial equation in z_* . Saddle-node bifurcations occur at the simultaneous positive roots of this polynomial and of its derivative with respect to z_* . Solving for these roots, we find two bifurcation lines, shown as solid lines in Fig. 1. The first one (orange line in Fig. 1) is the Vieillefosse line (1) for $R \leq \frac{1}{256}$, which distinguishes rotational and extensional regions in the flow. We note that its positive branch (R > 0) plays an important role for the Lagrangian dynamics of fluid-velocity gradients. Vieillefosse [32,33] showed that Lagrangian fluid-velocity gradients described by the inviscid Navier-Stokes equations self-amplify along the positive branch of (1). When the anisotropic portion of the pressure Hessian is neglected, this self-amplification leads to a finite-time divergence of Lagrangian fluid-velocity gradients along this branch [32–34]. Although the divergence does not occur in more realistic approximations [38,39,45], the self-amplification effect explains why the distribution of Lagrangian fluid-velocity gradients in Navier-Stokes turbulence is strongly skewed towards the positive branch of Eq. (1) [35].

The second bifurcation curve (red line in Fig. 1) is given by

$$Q = 4R - \frac{1}{16} \quad \text{for } R \ge -\frac{1}{32}.$$
 (10)

The two bifurcation lines meet at $R = -\frac{1}{32}$ (not shown) and at $R = \frac{1}{256}$ (red square in Fig. 1). They divide the Q-R plane into regions where Eq. (9) has different numbers of solutions. In addition, transcritical bifurcations, which conserve the number of fixed points [44], occur on the symmetry line (broken lines in Fig. 1)

$$Q = -\frac{3}{64},$$
 (11)

where the z_* derivative of Eq. (9) vanishes. Importantly, the bifurcation line (10) separates the region without any fixed points (yellow region) from the rest of the Q-R plane (blue and crosshatched regions). The latter regions contain a finite number of fixed points and we find numerically that one of them is always stable and smoothly connected to $\mathbb{Z}_* \approx \mathbb{A}$ close to the origin.

In summary, at small St we can invoke timescale separation. While \mathbb{Z} relaxes rapidly, \mathbb{A} changes slowly. As soon as \mathbb{A} crosses the threshold line (10) and enters the yellow region in Fig. 1, the \mathbb{Z} dynamics becomes unstable. As a result, a caustic forms since $\operatorname{Tr}\mathbb{Z}(t) \to -\infty$ in a short time.

IV. OPTIMAL FLUCTUATION

At small particle inertia caustics are rare, because typical fluctuations of \mathbb{A} along particle paths are of order St $\ll 1$, much smaller than the threshold (10), which is of order unity. While there are typically many ways of similar probability to realize a typical fluctuation (here of order St $\ll 1$), this is not the case for rare large fluctuations. Instead, it is a general theme in large-deviation theory [25–27] that rare events (here fluctuations of order unity) are often dominated by unique optimal fluctuations that correspond to the most likely realizations of these events. Here we use large-deviation techniques developed to characterize such rare events [28,29] to determine the optimal fluctuation for \mathbb{A} to reach the threshold, starting from the most likely initial value Q = R = 0.

A. Matrix basis

In homogeneous and isotropic turbulence, the elements A_{ij} of \mathbb{A} are correlated [46], which makes calculations cumbersome. It is therefore convenient to represent \mathbb{A} in a basis of matrices e_1, \ldots, e_8 , $\mathbb{A}(t) = \sum_{i=1}^8 A_i(t)e_i$, chosen in such a way that the random processes $A_i(t)$ are uncorrelated. The basis we use in the following is orthonormal with respect to the inner product $\langle \mathbb{M}, \mathbb{N} \rangle \equiv \text{Tr}(\mathbb{MN})/2$ such that $\langle e_i, e_j \rangle = g_{ij}$, where

The basis elements e_1 , e_2 , and e_3 span the space of traceless antisymmetric matrices and can be written

$$e_1 = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{pmatrix}, \quad e_2 = \begin{pmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \quad e_3 = \begin{pmatrix} 0 & 0 & -1 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix}.$$
 (13)

The remaining e_i , i = 4, ..., 8, spanning the space of traceless symmetric matrices, read

$$\begin{aligned}
\mathbf{e}_{4} &= \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix}, \quad \mathbf{e}_{5} &= \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}, \quad \mathbf{e}_{6} &= \begin{pmatrix} 1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \\
\mathbf{e}_{7} &= \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \quad \mathbf{e}_{8} &= \frac{1}{\sqrt{3}} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -2 \end{pmatrix}.
\end{aligned}$$
(14)

Consequently, the antisymmetric (vorticity) and symmetric (strain) parts of A are expanded as

$$\mathbb{O}(t) = \sum_{i=1}^{3} O_i(t) \mathbb{e}_i, \quad \mathbb{S}(t) = \sum_{i=1}^{5} S_i(t) \mathbb{e}_{i+3}, \tag{15}$$

respectively, where $O_i = A_i$ and $S_i = A_{i+3}$. Note that since $\mathbb{e}_i^{\mathsf{T}} = \sum_{j=1}^8 g_{ij} \mathbb{e}_j$, one has $\langle \mathbb{e}_i, \mathbb{e}_i^{\mathsf{T}} \rangle = \delta_{ij}$. In addition to $\mathbb{e}_1, \ldots, \mathbb{e}_8$, a complete matrix basis also includes the unit matrix $\mathbb{e}_0 = \mathbb{1}$, which is however not needed in the following, because \mathbb{A} is traceless.

B. Correlation functions of fluid-velocity gradients

For homogeneous and isotropic turbulence, the entries of \mathbb{O} and \mathbb{S} have zero mean and correlation functions [47,48]

$$\langle \mathbb{O}_{ik}(t)\mathbb{O}_{jl}(t')\rangle_p = \sigma_0^2 C_{ijkl}^O f_O(t-t'), \quad \langle \mathbb{S}_{ik}(t)\mathbb{S}_{jl}(t')\rangle_p = \sigma_s^2 C_{ijkl}^S f_S(t-t'), \quad \langle \mathbb{O}_{ik}(t)\mathbb{S}_{jl}(t')\rangle_p = 0,$$
(16)

where the angular brackets $\langle \cdot \rangle_p$ denote a steady-state average along inertial-particle trajectories. The correlation functions f_O and f_S are normalized such that $f_O(0) = f_S(0) = 1$ and are well approximated by exponential functions with different correlation times of order τ_K [45,49]. The structure of the rotationally covariant tensors C^S and C^O in Eq. (16) follows from isotropy and incompressibility [45,47]:

$$C_{ijkl}^{O} = \delta_{ij}\delta_{kl} - \delta_{il}\delta_{jk}, \quad C_{ijkl}^{S} = \delta_{ij}\delta_{kl} + \delta_{il}\delta_{jk} - \frac{2}{3}\delta_{ik}\delta_{jl}.$$
 (17)

Along the paths of fluid elements, the Lagrangian flow is homogeneous, which implies that [45,47]

$$\langle \operatorname{Tr}[\mathbb{O}^{\mathsf{T}}(t)\mathbb{O}(t)]\rangle_{f} = \langle \operatorname{Tr}[\mathbb{S}^{\mathsf{T}}(t)\mathbb{S}(t)]\rangle_{f} = \frac{1}{2}\langle \operatorname{Tr}[\mathbb{A}^{\mathsf{T}}(t)\mathbb{A}(t)]\rangle_{f} = \frac{\operatorname{St}^{2}}{2}.$$
(18)

Along inertial particle trajectories, however, preferential concentration [50,51] breaks homogeneity [52], so $\langle \operatorname{Tr}[\mathbb{O}^{\mathsf{T}}(t)\mathbb{O}(t)]\rangle_p < \langle \operatorname{Tr}[\mathbb{S}^{\mathsf{T}}(t)\mathbb{S}(t)]\rangle_p$ [53]. We therefore write the variances σ_0 and σ_s in Eq. (15) as

$$\sigma_O^2 = \frac{1}{6} \mathrm{St}^2 C_O(\mathrm{St}), \quad \sigma_S^2 = \frac{1}{10} \mathrm{St}^2 C_S(\mathrm{St}), \tag{19}$$

where $C_O(\text{St}) \leq \frac{1}{2}$ and $C_S(\text{St}) \geq \frac{1}{2}$ are St-dependent functions with $\lim_{\text{St}\to 0} C_O(\text{St}) = \lim_{\text{St}\to 0} C_S(\text{St}) = \frac{1}{2}$ in the inertialess limit. The prefactors in Eq. (19) account for the three and five independent components of \mathbb{O} and \mathbb{S} , respectively. In terms of the orthonormal basis introduced in Sec. IV A, the correlation functions (16) are conveniently expressed as

$$\langle O_i(t)O_j(t')\rangle_p = \delta_{ij}\sigma_O^2 f_O(t-t'), \quad \langle S_i(t)S_j(t')\rangle_p = \delta_{ij}\sigma_S^2 f_S(t-t'), \quad \langle O_i(t)S_j(t')\rangle_p = 0.$$
(20)

Equation (20) makes explicit that the elements $A_i(t)$ are mutually uncorrelated processes with variances σ_0 and σ_s .

C. Gaussian approximation

The complete statistical information about $\mathbb{A}(t)$ is contained in the cumulant-generating functional [54] of the process

$$\Lambda(\mathbb{G},t) = \ln \left\langle \exp\left(\int_{t_0}^t ds \langle \mathbb{A}(s), \mathbb{G}(s) \rangle\right) \right\rangle_p.$$
(21)

Here \mathbb{G} is a traceless matrix-valued test function. An expansion in powers of $\mathbb{G}(s) = \sum_{i=1}^{8} G_i(s) \mathbb{e}_i$ generates the cumulants of $\mathbb{A}(t)$, which we write in terms of $A_i(t)$. Since the correlation functions for the processes $A_i(t)$ decouple in the orthonormal matrix basis \mathbb{e}_i [see Eq. (20)] we find to second order in \mathbb{G} ,

$$\Lambda(\mathbb{G},t) = \sum_{i=1}^{8} \int_{t_0}^{t} \int_{t_0}^{t} ds \, ds' \sigma_i^2 f_i(s-s') G_i(s) G_i(s') + O(\mathbb{G}^3), \tag{22}$$

with $\sigma_i^2 f_i(t-t') = \sigma_O^2 f_O(t-t')$ for i = 1, ..., 3 and $\sigma_i^2 f_i(t-t') = \sigma_S^2 f_S(t-t')$ for i = 4, ..., 8. The orders $O(\mathbb{G}^3)$ in Eq. (22) contain terms of the form $\langle A_i(s)A_j(t)A_k(u)\rangle_p$, $\langle A_i(s)A_j(t)A_k(u)A_l(v)\rangle_p$, and so on. In a mean-field (or Gaussian) approximation, we decompose the higher-order correlation functions in terms of the second- and first-order ones. This gives $\langle A_i(s)A_j(t)A_k(u)\rangle_p \approx$ $\langle A_i(s)A_j(t)\rangle_p \langle A_k(u)\rangle_p + \cdots = 0$ and

$$\langle A_i(s)A_j(t)A_k(u)A_l(v)\rangle_p \approx \langle A_i(s)A_j(t)\rangle_p \langle A_k(u)A_l(v)\rangle_p + \cdots,$$
(23)

where the ellipsis denotes all permutations of the arguments. Applied to all orders, this approximation makes the terms $O(\mathbb{G}^3)$ in Eq. (22) vanish so that the individual processes $A_i(t)$ become Gaussian. Although the fluid-velocity gradients in the dissipation range of isotropic and homogeneous turbulence are well known not to be Gaussian [55–57], Gaussian models for turbulence have proven to provide valuable information about the behavior of turbulent aerosols [13,17,52]. Uncorrelated Gaussian processes are independent, as can be seen from the fact that $\Lambda(\mathbb{G}, t)$ decomposes completely when the $O(\mathbb{G}^3)$ term in Eq. (22) is absent. The mutual independence of the processes $A_i(t)$ is a crucial ingredient for the calculations that follow.

D. Threshold probability

For the stationary Gaussian process $A_i(t)$, the probability for reaching a given value A_i at a given time t reads

$$P[A_i(t) = A_i] \propto e^{-S(A_i)/\sigma_i^2},$$
(24)

where we omitted the normalization prefactor. Here $S(A_i) = A_i^2/2$ denotes the quadratic action associated with reaching A_i . Due to the mutual independence of the $A_i(t)$, we obtain the probability for \mathbb{A} to reach $\mathbb{A} \equiv \sum_{i=1}^{8} A_i e_i$ as the product of Eq. (24) over *i*,

$$P[\mathbb{A}(t) = \mathbb{A}] = \prod_{i=1}^{8} P[A_i(t) = A_i] \propto e^{-S(\mathbb{A})/\sigma_S^2},$$
(25)

characterized by the weighed sum $S(\mathbb{A})$ of actions $S(A_i)$ over *i*:

$$S(\mathbb{A}) = \frac{1}{2} \left(\frac{\sigma_S}{\sigma_O}\right)^2 \sum_{i=1}^3 A_i^2 + \frac{1}{2} \sum_{i=4}^8 A_i^2 = -\frac{1}{2} \left(\frac{\sigma_S}{\sigma_O}\right)^2 \langle \mathbb{O}, \mathbb{O} \rangle + \frac{1}{2} \langle \mathbb{S}, \mathbb{S} \rangle.$$
(26)

Since σ_S , $\sigma_O \sim$ St, Eqs. (25) and (26) imply that the exponent in Eq. (25) is proportional to St⁻² for small values of St. Also note that $S(\mathbb{A})$ is quadratic in A_i . Both observations are a consequence of approximating $A_i(t)$ as Gaussian processes.

E. Optimal threshold configuration

To obtain the most likely way for the fluid-velocity gradients to reach the threshold (10) and induce a caustic, we must minimize the action (26) under the constraint (10). To this end, we multiply the constraint with a Lagrange multiplier λ and add the product to $S(\mathbb{A})$ to form the Lagrange function

$$\mathscr{L}(\mathbb{A}) = S(\mathbb{A}) - \lambda \left[Q(\mathbb{A}) - 4R(\mathbb{A}) + \frac{1}{16} \right].$$
⁽²⁷⁾

Minimizing \mathscr{L} over \mathbb{A} gives the minimizer \mathbb{A}^* , the most likely configuration of fluid-velocity gradients at the threshold (10). The symmetry of \mathbb{F} under similarity transformations [see Eq. (6)] constrains this minimizer. In order to deduce how, we consider infinitesimal similarity transformations $\mathbb{A} \to \mathbb{A}' = \mathbb{P}^{-1}\mathbb{A}\mathbb{P}$ with det $\mathbb{P} = 1$, generated by traceless matrices

$$\mathbb{A}' \approx \mathbb{A} + \delta \mathbb{A}, \quad \delta \mathbb{A} = \sum_{i=1}^{8} [\mathbb{A}, \mathbb{e}_i] \theta_i, \tag{28}$$

where $[\mathbb{M}, \mathbb{N}] = \mathbb{M}\mathbb{N} - \mathbb{N}\mathbb{M}$ denotes the commutator between two matrices, and the factors $|\theta_i| \ll 1$ parametrize the infinitesimal transformation. For the symmetric and antisymmetric parts of \mathbb{A}' , the

transformation (28) implies $S' = S + \delta S$ and $O' = O + \delta O$, with

$$\delta \mathbb{S} = \sum_{i=1}^{3} [\mathbb{S}, \mathbb{e}_i] \theta_i + \sum_{i=4}^{8} [\mathbb{O}, \mathbb{e}_i] \theta_i, \quad \delta \mathbb{O} = \sum_{i=1}^{3} [\mathbb{O}, \mathbb{e}_i] \theta_i + \sum_{i=4}^{8} [\mathbb{S}, \mathbb{e}_i] \theta_i.$$
(29)

The Lagrange function $\mathscr{L}(\mathbb{A})$ in Eq. (27) must be invariant under similarity transformations of the minimizer \mathbb{A}^* , i.e., $\delta \mathscr{L}(\mathbb{A}) = \mathscr{L}(\mathbb{A}') - \mathscr{L}(\mathbb{A}) = 0$ at $\mathbb{A} = \mathbb{A}^*$, which requires that $\delta S(\mathbb{A}) = S(\mathbb{A}') - S(\mathbb{A}) = 0$ at the minimizer. Applying the infinitesimal transformation (28) to $S(\mathbb{A})$, however, we find

$$\delta S(\mathbb{A}) = \left[\left(\frac{\sigma_S}{\sigma_O} \right)^2 + 1 \right] \sum_{i=4}^8 \langle [\mathbb{S}, \mathbb{O}], e_i \rangle \theta_i.$$
(30)

This expression vanishes, i.e., $S(\mathbb{A})$ is invariant, only in two cases. First, $\theta_i = 0$ for i = 4, ..., 8, which corresponds to the case when \mathbb{P} is a rotation $\mathbb{P}^{\mathsf{T}} = \mathbb{P}^{-1}$. The second case is $[\mathbb{S}, \mathbb{O}] = 0$ which occurs when either $\mathbb{O} = 0$ or $\mathbb{S} = 0$. It can be checked that the constraint (10) is invariant under all similarity transformations, not only rotations, because $\delta Q(\mathbb{A}) = \delta R(\mathbb{A}) = 0$ for all θ_i . Hence, the constraint requires that $\delta S(\mathbb{A}) = 0$ also for finite θ_i , i = 4, ..., 8, meaning that either $\mathbb{O} = 0$ or $\mathbb{S} = 0$. However, $\mathbb{S} = 0$, and thus $\mathbb{A} = \mathbb{O}$, leads to $R(\mathbb{O}) = 0$ and $Q(\mathbb{O}) \ge 0$, so the constraint (10) cannot be satisfied for any real \mathbb{O} . This in turn shows that the vorticity contribution \mathbb{O}^* to the optimal threshold configuration \mathbb{A}^* must vanish, $\mathbb{O}^* = 0$, and that \mathbb{A}^* must be a pure strain $\mathbb{A}^* = \mathbb{S}^*$.

In order to determine the components of S^* , we express A in Eq. (27) in terms of A_i and minimize (27) over A_4, \ldots, A_8 , while $A_1^* = A_2^* = A_3^* = 0$. This leads to two solutions with equal actions $S(A^*) = \frac{3}{64}$, which fix the minimizing configuration A_i^* up to equivalence under similarity transformations. As a consequence of this symmetry, the minimizers are not isolated points but higher-dimensional manifolds. All minimizers found in this way are related by similarity transformations so that we may choose particular representatives to gain insight into the most likely threshold configuration A^* . The simplest representatives correspond to diagonal A^* and they show that A^* takes two positive eigenvalues equal to $\frac{1}{8}$ and one negative eigenvalue equal to $-\frac{1}{4}$ of twice the magnitude. Similarity transformations only affect the order of eigenvalues, so we may put all minimizers into the ordered diagonal form

$$\mathbb{A}^* = \frac{1}{4} \begin{pmatrix} \frac{1}{2} & & \\ & \frac{1}{2} & \\ & & -1 \end{pmatrix}$$
(31)

by a suitable transformation. Expressed in the matrix basis, this corresponds to the simple configuration $A_8^* = (\frac{1}{4})\sqrt{3}/2$ and $A_1^* = A_2^* = \cdots = A_7^* = 0$.

Up to this point, our analysis provides the most likely threshold configuration that induces a caustic in the small-St limit. The strict limit, however, is difficult to approach in numerical simulations, since the threshold probability (25) is exponentially suppressed in St.

F. Small-St correction

We now discuss how to adjust the theory to incorporate the main next-to-leading-order effects in St \ll 1 in an approach similar to that used in two spatial dimensions in Ref. [24]. Our starting point is Eq. (3). Since caustics form on timescales of order St, we require the threshold (10) to be exceeded for a finite time such as to leave the dynamics (3) sufficient time to form a caustic [24]. To account for this, we introduce a St-dependent threshold $A_{th}(St) = \frac{1}{4} + o(St^0)$ [24] that consists of the threshold determined above plus a small positive correction $o(St^0)$ that vanishes as St \rightarrow 0. Although this $o(St^0)$ correction slightly enhances the threshold $A_{th}(St)$, the rate at which trajectories exceed $A_{th}(St)$ nevertheless increases as St increases, because $\sigma_S^2 \sim St^2$ in Eq. (25). Recently, Bätge *et al.* [58] described a one-dimensional model for caustic formation where the time needed for caustics to form is accounted for by a St-dependent preexponential factor instead of a St correction to the action $S(\mathbb{A}^*)$ as introduced here.

We make the ansatz

$$\mathbb{A}^* = A_{\text{th}}(\text{St}) \begin{pmatrix} \frac{1}{2} & & \\ & \frac{1}{2} & \\ & & -1 \end{pmatrix}$$
(32)

for the gradient matrix, i.e., $A_8^* = \sqrt{3}A_{\text{th}}(\text{St})/2$, $A_1^* = \ldots = A_7^* = 0$, leading to the St-dependent threshold line

$$Q = 4R - A_{\rm th}^2({\rm St}) [A_{\rm th}({\rm St}) + \frac{3}{4}],$$
(33)

which reduces to Eq. (10) as $St \rightarrow 0$. The magnitude of the difference $A_{th}(St) - \frac{1}{4}$ determines by how much the gradient threshold $\frac{1}{4}$ is exceeded at the time $t = t_{th}$ when the gradient excursion peaks, and through the dynamics (3) it also fixes the time $t_c > t_{th}$ at which the caustic is generated. Using this information and input from numerical simulations, we model the optimal fluctuation to next-to-leading order in St.

G. Optimal path to caustic formation

In order to find the optimal path to caustic formation, we must understand how the optimal fluctuation connects typical values of $\mathbb{A} \sim St$ to the optimal threshold configuration $\mathbb{A}^* \sim 1$ as a function of time. To make this connection, we use that the most likely way $A_i^*(t)$ for a Gaussian process $A_i(t)$ to reach a large value $A_i \gg \sigma_i$ at a given time *t* is equal to the correlation function (20) of the process, normalized to the threshold value [24], i.e.,

$$A_{i}^{*}(t) = A_{i}f_{i}(t - t_{\text{th}}).$$
(34)

We give a derivation of this formula in the Appendix. Since $A_8^* \gg \sigma_s$ for the optimal threshold configuration (32), we may use Eq. (34) to determine the optimal fluctuation $A_i^*(t)$ of the fluid velocity gradients to reach the threshold. This leads to $A_8^*(t) = \sqrt{3}A_{\text{th}}(\text{St})f_s(t - t_{\text{th}})/2$ and $A_1^*(t) = \cdots = A_7^*(t) = 0$. Denoting by $\lambda_1(t) > \lambda_2(t) > \lambda_3(t)$ the ordered eigenvalues of $\mathbb{A}^*(t)$ along the optimal fluctuation, i.e.,

$$\mathbb{A}^*(t) = \begin{pmatrix} \lambda_1(t) & & \\ & \lambda_2(t) & \\ & & \lambda_3(t) \end{pmatrix}, \tag{35}$$

we obtain

$$\lambda_1(t) = \lambda_2(t) = \frac{1}{2} A_{\text{th}}(\text{St}) f_S(t - t_{\text{th}}), \quad \lambda_3(t) = -A_{\text{th}}(\text{St}) f_S(t - t_{\text{th}})$$
(36)

so that $\mathbb{A}^*(t_{th}) = \mathbb{A}^*$ at the time the threshold is reached [cf. Eq. (32)]. The optimal path to caustic formation $\mathbb{Z}^*(t)$ for the particle-velocity gradients is now obtained by substituting $\mathbb{A}^*(t)$ into Eq. (3). Since $\mathbb{A}^*(t)$ is vorticity-free and diagonal in the chosen coordinates, $\mathbb{Z}^*(t)$ is initially diagonal, $\mathbb{Z}^*(t_0) = \mathbb{A}^*(t_0)$. The initial time t_0 is chosen such that $|t_c - t_0|$ is of the order of the expected time for caustic formation, which implies $|t_c - t_0| \sim \exp[S(\mathbb{A}^*)/\mathrm{St}^2] \gg 1$ for the model considered here. Since Eq. (3) decouples for diagonal matrices,

$$\mathbb{Z}^*(t) = \begin{pmatrix} \zeta_1(t) & & \\ & \zeta_2(t) & \\ & & \zeta_3(t) \end{pmatrix}$$
(37)

remains diagonal along the optimal path to caustic formation. The dynamics of the eigenvalues $\zeta_i(t)$ is then governed by the equations

$$\operatorname{St}\dot{\zeta}_{i}(t) = -\zeta_{i}(t) - \zeta_{i}^{2}(t) + \lambda_{i}(t), \qquad (38)$$

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with i = 1, 2, 3. In words, along the optimal fluctuation, the nine-dimensional matrix equation (3) reduces to three uncoupled equations for the eigenvalues ζ_i , akin to the one-dimensional equation studied in Refs. [21,59]. The theory summarized here demonstrates that the process of caustic formation in the persistent limit is essentially one dimensional, as shown for two dimensions in Ref. [24] and as argued in Ref. [58]. We also see that the leading negative eigenvalue of \mathbb{A}^* must be smaller than $-\frac{1}{4}$ (in our dimensionless formulation) when St $\rightarrow 0$.

We solve (38) numerically for a given normalized strain correlation function $f_S(t)$ [see Eq. (36)] that we extract from numerical simulations (Sec. V). The threshold value $A_{th}(St)$ is then obtained from Eq. (38) by tuning $A_{th}(St)$ to match the time difference $t_c - t_{th}(St)$ for the optimal fluctuation with the numerical value. The optimal fluctuation in the Q-R plane reads

$$Q(t) = -\frac{3}{4}A_{\rm th}({\rm St})^2 f_S^2[t - t_{\rm th}({\rm St})], \quad R(t) = \frac{1}{4}A_{\rm th}({\rm St})^3 f_S^3[t - t_{\rm th}({\rm St})].$$
(39)

For St \rightarrow 0, we have $|t_c - t_0| \rightarrow \infty$ and $A_{th}(St) \rightarrow \frac{1}{4}$ so that Eqs. (39) in this case represent an infinite-time trajectory that connects the green and red squares in Fig. 1, along the right branch of the Vieillefosse line. For finite St, the optimal fluctuation remains on this line, but it now penetrates the yellow region in Fig. 1 for a short time, because $A_{th}(St) > \frac{1}{4}$, and relaxes back to the origin for $t > t_{th}$.

Note that the Vieillefosse line separates regions in the Q-R plane where A has real eigenvalues from regions where two eigenvalues form a complex conjugate pair [30]. Thus, on the Vieillefosse line, all eigenvalues are real and two eigenvalues must be degenerate. Our theory predicts that caustics form when a single eigenvalue of A becomes large and negative, while the other two remain degenerate and positive, with half the magnitude of the negative eigenvalue. For this reason and because the large negative eigenvalue implies R > 0, this entails that the optimal fluctuation must propagate along the positive branch of the Vieillefosse line for small St.

V. NUMERICAL SIMULATIONS

We compare our theory with results of numerical simulations of a Gaussian statistical model for turbulent aerosols at small but finite St. The model captures particle clustering and caustic formation of aerosols in the dissipative range of turbulence qualitatively and, in some cases, even quantitatively [17]. For the simulations, we evolve a large number of particles with Eq. (2). The particles move in a turbulent fluid-velocity field u(x, t), modeled by a fluctuating, statistically homogeneous and isotropic Gaussian vector field with a prescribed correlation function [17]. In addition, we evolve Eq. (3) for $\mathbb{Z}(t)$ along the particle trajectories and track the histories of $\mathbb{A}(t)$. Whenever $\text{Tr}\mathbb{Z}(t)$ reaches a large negative threshold, i.e., a caustic is imminent, we record a caustic and store the corresponding time series of \mathbb{A} . This provides us with an ensemble of fluid-velocity gradients, conditioned on observing a caustic at each end point. From this ensemble, we evaluate the invariants Q(t) and R(t) and compute numerical approximations of their trajectory densities, allowing us to quantify the most likely fluctuations that result in a caustic at time $t = t_c$.

Figure 2 shows snapshots of the trajectory density in the Q-R plane, at different times. Regions of high density are shown in yellow with the maximum marked by a yellow star. The theoretically obtained optimal fluctuation is shown by the green circle. As predicted by the theory, the yellow star performs a large excursion along the positive branch of the Vieillefosse line towards the Stdependent threshold [Figs. 2(a) and 2(b)], reaches it in Fig. 2(c), and returns to the origin [Fig. 2(d)]. At the time of caustic formation in Fig. 2(e), the gradients have almost completely relaxed. Note that typical fluctuations of the fluid-velocity gradients in the Q-R plane [Fig. 2(a)] are tiny compared to the magnitude of the gradient excursion in Fig. 2(c). Although the theory explains the excursion qualitatively, we observe slight deviations in the time dependences of the green circle and the yellow star.

The time evolutions of the optimal fluctuation and of its vorticity contributions are analyzed in more detail in Fig. 3. Vorticity \mathbb{O} contributes to Q and R through the expressions $Q_O = \text{Tr}(\mathbb{O}^T \mathbb{O})/2$ and $R_O = \text{Tr}(\mathbb{O}^T \mathbb{O} \mathbb{S})$, respectively. Figures 3(a) and 3(c) show the vorticity contributions Q_O and



FIG. 2. Snapshots of the trajectory density (heat map) in the Q-R plane at different times (a) $t - t_c = -6$, (b) $t - t_c = -4$, (c) $t - t_c = -2.5$, (d) $t - t_c = -1.25$, and (e) $t - t_c = 0$ for St = 0.3 and Ku ≈ 22 . The yellow star shows the location of maximum density. The Vieillefosse line is shown in orange. The red dashed line shows the St-dependent threshold, Eq. (33) with $A_{th}(St) \approx 0.336$, featuring the optimal threshold configuration (red diamond). The green circle shows the location of the optimal fluctuation from theory with arrows indicating the change with respect to the next snapshot.

 R_O as functions of time, together with the theoretical curves for their optimal fluctuations (dotted lines), which are identically zero. The yellow streaks of high trajectory density for Q_O and R_O remain close to zero at all times, which confirms that in both theory and numerics the optimal fluctuation is dominated by the strain S while vorticity \mathbb{O} remains small.

Figures 3(b) and 3(d) show the trajectory densities of Q and R (including both strain and vorticity) as functions of time, together with the theoretical curves for the optimal fluctuations (dotted lines). The yellow streaks for Q and R perform large excursions, centered at $t - t_c \approx -2.5$, which are in good agreement with our theory. However, the optimal fluctuations in the numerics (yellow streaks) grow faster and are slightly more persistent than predicted. These deviations are likely due to too large St, which leads to inaccuracies in both the optimal fluctuation approach and our approximations.

Outside the regime St $\ll 1$, many fluctuations, not only the optimal one, contribute to caustic formation. Furthermore, the fluid-velocity gradients measured along particle trajectories acquire non-Gaussian statistics and more importantly the timescale separation in the dynamics of \mathbb{Z} and \mathbb{A} becomes less sharp. However, the comparison between numerics and theory in Figs. 2 and 3 shows that optimal fluctuation theory does explain caustic formation qualitatively, even at moderate St. The theory not only reproduces the large excursion of the fluid-velocity gradient along the Vieillefosse line, but also correctly predicts vanishing vorticity, as well as the main features in the time dependence of the optimal fluctuation.



FIG. 3. Trajectory densities (heat maps) of (b) Q and (d) R and of their respective vorticity contributions (a) Q_0 and (c) R_0 , normalized to unity at each time slice, as functions of t prior to caustic formation for St = 0.3 and Ku \approx 22. Yellow regions correspond to high trajectory density. The black dotted line shows the optimal fluctuation obtained from theory [Eq. (39)].

VI. CONCLUSION

We explained caustic formation in turbulent aerosols for small particle inertia (St \ll 1) in three spatial dimensions by means of optimal fluctuation theory. We found that caustics are formed by an instability of spatial particle neighborhoods that occurs when the fluid-velocity gradients exceed a large threshold in the Q-R plane. The most likely way to reach this threshold is by an optimal fluctuation of the fluid-velocity gradients consisting of a large excursion of the strain S, while vorticity \mathbb{O} remains small. We determined the shape of the excursion explicitly within a mean-field (Gaussian) approximation. The theory predicts that the optimal fluctuation propagates along the positive branch of the Vieillefosse line [32–34] to reach the threshold, before it relaxes back to the origin. Using numerical simulations of a Gaussian statistical model, we showed that the optimal fluctuation is dominant not only at infinitesimal St, but also at values of order St $\sim 10^{-1}$, where we observed qualitative agreement with the theory.

The predominance of the strain part in the optimal fluctuation qualitatively explains why collisions are preceded by strong strains and low vorticity in simulations [60]. More generally, our fixed-point analysis shows that at small St, the particle dynamics enters caustic formation only by posing a threshold for the fluid-velocity gradients. This implies that the rate of caustic formation may be estimated solely from the threshold probability (25), without explicitly referring to the particle dynamics, rather than following local particle neighborhoods over long times, as in Refs. [6,18,36].

Finally, our analysis indicates that the tear-shaped elongation of the probability distribution of turbulent fluid-velocity gradients along the positive branch of the Vieillefosse line [35,37–39] may facilitate caustic formation in homogeneous and isotropic turbulence, as it leads to an increased probability of reaching the threshold, perhaps even more so if one takes into account preferential concentration [61].

ACKNOWLEDGMENTS

K.G. thanks J. Vollmer and G. Bewley for discussions regarding the role of the invariants Q and R for caustic formation. We thank D. Mitra for discussions regarding caustic formation in turbulence and A. Bhatnagar for sharing his unpublished data [62], which indicates that also in homogeneous and isotropic turbulence caustics form near the Vieillefosse line at small St, as predicted by our theory. J.M. is funded by a Feodor-Lynen Fellowship of the Alexander von Humboldt-Foundation. B.M. was supported in part by Vetenskapsrådet (VR), Grant No. 2021-4452. The statistical-model simulations were conducted using the resources of HPC2N provided by the Swedish National Infrastructure for Computing (SNIC), partially funded by VR through Grant No. 2018-05973.

APPENDIX: DERIVATION OF EQ. (34)

Equation (34) was derived previously in Ref. [24]; here we give a simplified derivation. We start with a path-integral expression for Eq. (24),

$$P[A_i(t_{\rm th}) = A_i] = \langle \delta[A_i(t_{\rm th}) - A_i] \rangle_p = \int \mathscr{D}A_i(t) \exp\left(-\frac{\mathscr{S}[A_i(t)]}{\sigma_i^2}\right) \delta(A_i(t_{\rm th}) - A_i), \qquad (A1)$$

where $\delta(x)$ denotes the Dirac delta function. For Gaussian processes, the action $\mathscr{S}[A_i(t)]$ is explicitly known [63]:

$$\mathscr{S}[A_i(t)] = \frac{1}{2} \int_{-\infty}^{\infty} ds \int_{-\infty}^{\infty} dt A_i(s) f_i^{-1}(s-t) A_i(t).$$
(A2)

Here $f^{-1}(x)$ is the inverse of the normalized correlation function $f_i(x)$ of $A_i(t)$, defined as

$$\int_{-\infty}^{\infty} ds' f_i^{-1}(s-s') f_i(s'-t) = \delta(s-t).$$
 (A3)

For $A_i \gg \sigma_i$, the path integral in (A1) can be approximated by the saddle point of the action $\mathscr{S}[A_i(t)]$ under the constraint that $A_i(t_{\text{th}}) = A_i$ [enforced by the δ function in (A1)]. We perform the constrained minimisation by adding a Lagrange multiplier λ to $\mathscr{S}[A_i(t)]$:

$$\mathscr{S}_{\lambda}[A_i(t)] = \frac{1}{2} \int_{-\infty}^{\infty} ds \int_{-\infty}^{\infty} dt A(s) f_i^{-1}(s-t)A(t) - \lambda \left(\int_{-\infty}^{\infty} ds \,\delta(t_{\rm th} - s)A(s) - A_i \right). \tag{A4}$$

For computational convenience, we have expressed the constraint $A_i(t_{\text{th}}) = A_i$ as an integral involving a Dirac δ function, as was done in a different context in Ref. [64]. We now compute the optimal fluctuation $A_i^*(t)$ as the constrained saddle point $\delta \mathscr{S}_{\lambda}[A_i^*(t)] = 0$ of the path integral in Eq. (A1). The variation of Eq. (A4) over $A_i(t)$ gives

$$\delta \mathscr{S}_{\lambda}[A_i^*(t)] = \int_{-\infty}^{\infty} ds \,\delta A_i(s) \bigg(\int_{-\infty}^{\infty} dt \, f_i^{-1}(s-t) A_i^*(t) - \lambda \delta(t_{\rm th} - s) \bigg), \tag{A5}$$

which leads to the saddle-point equation

$$\int_{-\infty}^{\infty} dt f_i^{-1}(s-t)A_i^*(t) = \lambda \delta(t_{\rm th} - s).$$
(A6)

Comparison with Eq. (A3) immediately gives

$$A_i^*(t) = \lambda f_i(t - t_{\rm th}). \tag{A7}$$

The Lagrange parameter λ is now simply obtained by using the constraint $A_i^*(t_{\text{th}}) = A_i$ and the normalization condition $f_i(0) = 1$. This gives $\lambda = A_i$ and thus the optimal fluctuation (34). We evaluate the action $\mathscr{S}[A_i(t)]$ at the saddle point by using the optimal fluctuation (34) and obtain

$$\mathscr{S}[A_i^*(t)] = \frac{A_i^2}{2}.$$
(A8)

Hence, we recover the quadratic action in Eq. (24) from Eq. (A1), evaluated at the optimal fluctuation (34).

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