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Dispersion in the large-deviation regime

Jacques Vanneste

School of Mathematics and Maxwell Institute University of Edinburgh, UK www.maths.ed.ac.uk/~vanneste

with Alexandra Tzella (Birmingham) and Peter Haynes (Cambridge)

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Advection-diffusion

Passive scalar released in a flow: a classical problem.





Concentration C(x, t) obeys the advection–diffusion equation:

$$\partial_t C + \boldsymbol{u} \cdot \nabla C = \kappa \nabla^2 C ,$$

with a flow u(x, t) that is given and satisfies $\nabla \cdot u = 0$. Pdf of particles positions:

$$\dot{X} = u(X,t) + \sqrt{2\kappa} \dot{W}$$

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Advection-diffusion

For $t \gg 1$, the combined effect of advection and diffusion can often be modelled by an effective diffusivity κ_{eff} :

- $\mathbb{E} X \otimes X \sim 2\kappa_{\text{eff}} t$,
- $C \simeq \exp(-x \cdot \kappa_{\text{eff}}^{-1} \cdot x/(4t))$: Gaussian distribution,
- effective equation

$$\partial_t C = \nabla \cdot (\kappa_{\text{eff}} \cdot \nabla C).$$

In simple flows: $\kappa_{\rm eff}$ can be computed explicitly.

- shear flows (Taylor dispersion),
- periodic flows.

e.g. Majda & Kramer 1999



Cellular flow: $\psi = \sin x \sin y$

$$\kappa_{\rm eff} = 2\nu\kappa^{1/2}$$
 for $\kappa \ll 1$,



with $u \sim 0.5327407 \cdots$ Shraiman, Rosenbluth et al, Childress, Soward...

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Limitations of effective diffusivity

Diffusive approximation assumes $x/t^{1/2} = O(1)$ as $t \to \infty$. It cannot describes the tails of C(x, t) which are non-Gaussian.

Large deviations:

- obtain C(x, t) for x/t = O(1),
- recover homogenisation as a limiting case.

Interest:

- Low concentrations can be important:
 - anecdotally: highly toxic chemicals,
 - exactly: FKPP fronts.
- Unifies 'improvements' to homogenisation.

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• Example of extreme-event statistics.

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Large deviations

For $t \gg 1$, the concentration takes the large-deviation form

$$C(x,t) \simeq \exp(-tg(\boldsymbol{\xi}))$$
 for $\boldsymbol{\xi} = x/t = O(1),$

with *g* the rate function, convex with g(0) = g'(0) = 0.

Computing g: define f(q) by

$$\mathrm{e}^{\mathrm{t}f(q)} \asymp \mathbb{E} \,\mathrm{e}^{q \cdot X}$$

f and g are a Legendre transform pair.



f can be estimated

- by Monte Carlo (incl importance sampling),
- ► by solving eigenvalue problems (for $\partial_t u = 0$). Effective equation: $\partial_t C = f(-\nabla)C$.

Haynes & Vanneste 2014a



Large deviations: cellular flow

For Pe \gg 1, particles are trapped inside cells, with rare exits across separatrices.



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Three regimes: (I) $|x|/t = O(Pe^{-3/4})$; (II) $|x|/t = O(\log Pe)$ and (III) |x|/t = O(Pe). Haynes & Vanneste 2014b

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FKPP fronts

Advection-diffusion-reaction equation:

$$\partial_t C + \boldsymbol{u} \cdot \nabla C = \mathrm{Pe}^{-1} \nabla^2 C + \mathrm{Da} C(C-1) ,$$

logistic reaction, with $Da = L/(U\tau)$, Damköhler number.





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Rectangular network



Non-Gaussian behaviour induced by geometry. Applications: urban pollution, porous media...

Rate function *g*:

- for U = V = 0: from g ~ |ξ|²/2 to g ~ (|ξ₁| + |ξ₂|)²/4 (diffusion with κ/2 in L₂-norm vs. κ in L₁-norm),
- ▶ for U, $V \gg 1$: *g* independent of κ , topological dispersion.



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Rectangular network 'Real Manhattan'



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Conclusions

- Large-deviation theory to obtain
 - ► scalar concentrations $C \simeq \exp(-tg(x/t))$ for x/t = O(1),
 - speed of FKPP fronts: $c = g^{-1}(Da)$,
- Assumes $t \gg 1$ but works well for t = O(1).
- ▶ Rate function *g* is calculated by solving an e'value problem.
- Extensions: towards turbulent flows,
 - time-periodic flows,
 - random flows (with A. Renaud),
 - simulation data.
- Complex geometries.

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