

Fluctuations in conservative systems and SPDEs

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joint work with Ben Fehrman [<https://arxiv.org/abs/1910.11860>]

and Daniel Heydecker [<https://arxiv.org/abs/2303.11289>]



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Content

General perspective

From interacting particle systems to conservative SPDEs

From large deviations to PDE with irregular drift

Further Applications

Principle viewpoint

- Consider large, interacting, stochastic systems (complex dynamics, many model parameters)
- Effective behavior by universal scaling limits. Hydrodynamic limits leading to partial differential equations (PDE)
- Stochastic fluctuations often are essential, e.g. meta-stability
- **Aim:** Universal scaling limits, capturing both the average behavior and fluctuations
- **Ansatz:** Conservative SPDEs

$$\partial_t u = \operatorname{div} F(u, Du) + \sqrt{\varepsilon} \operatorname{div}(G(u)\xi),$$

with ξ space-time white noise, as universal fluctuating continuum models.

- Many open problems, e.g. well-posedness, regularity, stochastic dynamics [Fehrman, G., 2019, 2020, 2022], [Mariani, 2010], [Lions, Souganidis 1998ff], [Gassiat, G., Lions, Souganidis 2021], ...

Nonequilibrium statistical mechanics - fluctuating gradient flows

- Gradient flow on an (infinite dimensional) “Riemannian manifold”
[Jordan, Kinderlehrer, Otto, 1998]

$$\partial_t \rho = -M(\rho) \frac{\partial E}{\partial \rho}(\rho). \quad (\text{e.g. } \dot{X} = -M(X) \nabla E(X)).$$

- Formal non-equilibrium stationary Gibbs state $\mu = \frac{1}{Z} e^{-N^{-1} E(\rho)} d\rho$.
- Detailed-balance: Fluctuating gradient flow
([Öttinger 2005], fluctuating hydrodynamics [Spohn 1991])

$$\partial_t \rho = -M(\rho) \frac{\partial E}{\partial \rho}(\rho) + N^{-\frac{1}{2}} M^{\frac{1}{2}}(\rho) \xi.$$

- Decorrelation length \approx typical particle distance / grid-size: $\xi \rightarrow \xi^N$
- Nonlinear diffusion

$$\partial_t \rho^N = \partial_{xx} \Phi(\rho^N) + N^{-\frac{1}{2}} \partial_x \left(\Phi^{\frac{1}{2}}(\rho^N) \xi^N \right).$$

Fluctuating hydrodynamics and macroscopic fluctuation theory

[Spohn 1991]

[Bertini, De Sole, Gabrielli, Jona-Lasinio, Landim 2015]

Recall: Fluctuating gradient flow

$$\partial_t \rho^N = -M(\rho^N) \frac{\partial E}{\partial \rho}(\rho^N) + N^{-\frac{1}{2}} M^{\frac{1}{2}}(\rho^N) \xi.$$

Rare events: (Im-)probability of a fluctuation ρ in small noise $N^{-\frac{1}{2}} \rightarrow 0$ limit

$$\mathbb{P}[\rho^N \approx \rho] = e^{-N I(\rho)} \quad N \text{ large.}$$

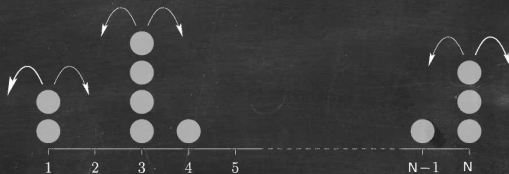
Informally, “by” contraction principle LDP has rate function

$$\begin{aligned} I(\rho) &= \inf \left\{ \int_{t,x} g^2 : \partial_t \rho = -M(\rho) \frac{\partial E}{\partial \rho}(\rho) + M^{\frac{1}{2}}(\rho) g \right\} \\ &= \|M^{-\frac{1}{2}}(\rho) (\partial_t \rho + M(\rho) \frac{\partial E}{\partial \rho}(\rho))\|_{L^2}^2 \end{aligned}$$

Macroscopic fluctuation theory: Postulate $I(\rho)$ as energy for non-equilibrium systems.

From interacting particle systems to conservative SPDEs

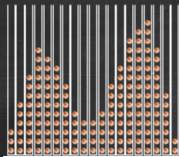
The zero range process (could also consider simple exclusion, independent particles, ..).



- State space $\mathbb{M}_N := \mathbb{N}_0^{\mathbb{T}_N}$, i.e. configurations $\eta : \mathbb{T}_N \rightarrow \mathbb{N}_0$: System in state η if container k contains $\eta(k)$ particles.
- Local jump rate function $g : \mathbb{N}_0 \rightarrow \mathbb{R}_0^+$.
- Translation invariant, asymmetric, zero mean transition probability $p(k, l)$.
- Markov jump process $\eta(t)$ on \mathbb{M}_N .
- $\eta(k, t)$ = number of particles in box k at time t .

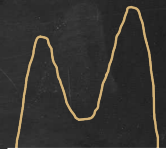
- Hydrodynamic limit? Multi-scale dynamics

Microscopic scale: Particles



Gridsize = $\frac{1}{N}$

Macroscopic scale: PDEs



Mean dynamics

- Empirical density field: $\mu^N(x, t) := \frac{1}{N} \sum_k \delta_{\frac{k}{N}}(x) \eta(k, tN^2)$.
- [Hydrodynamic limit - Ferrari, Presutti, Vares; 1987]

$$\mu^N(t) \rightharpoonup^* \bar{\rho}(t) dx$$

with

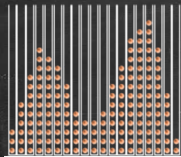
$$\partial_t \bar{\rho} = \partial_{xx} \Phi(\bar{\rho})$$

with Φ the mean local jump rate $\Phi(\rho) = \mathbb{E}_{\nu_\rho} [g(\eta(0))]$.

- Loss of information:
 - Fluctuations, rare events?
 - Error: $\mu^N = \bar{\rho} + O(N^{-\frac{1}{2}})$.

Fluctuating Hydrodynamics?

Microscopic scale: Particles



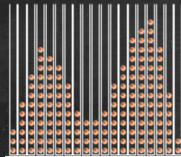
Gridsize = $\frac{1}{N}$

Macroscopic scale: PDEs



Mean dynamics

Microscopic scale: Particles



Gridsize = $\frac{1}{N}$

Mesoscopic scale: Conservative SPDEs



Fluctuation correction

Macroscopic scale: PDEs



Mean dynamics

Ansatz: Conservative SPDEs

$$\partial_t \rho^N = \partial_{xx} \Phi(\rho^N) + N^{-\frac{1}{2}} \partial_x \left(\Phi^{\frac{1}{2}}(\rho^N) \xi^N \right).$$

Informal justification:

1. Physics: Fluctuation-dissipation relation, Fluctuating Wasserstein gradient flow, ∞ -dim Fokker-Planck equations

From large deviations to parabolic-hyperbolic PDE with irregular drift

Rare events: (Im-)probability to observe a fluctuation ρ : $\mathbb{P}[\mu^N \approx \rho] = e^{-N I(\rho)}$.

Zero range process

$$I(\rho) = \inf \left\{ \int_{t,x} |\partial_x H|^2 \Phi(\rho) : \underbrace{\partial_t \rho = \partial_{xx} \Phi(\rho) + \partial_x (\Phi(\rho) \overbrace{\partial_x H}^{\in L^2_{\Phi(\rho)}})}_{\text{"controlled nonlinear Fokker-Planck equation"}} \right\}.$$

Theorem ([Large deviation principle, Kipnis, Olla, Varadhan, Benois, Landim, 80's])

For every open set $O \subseteq D([0, T], \mathcal{M}_+)$ we have

$$\mathbb{P}[\mu^N \in \bar{O}] \lesssim e^{-N \inf_{\mu \in \bar{O}} I(\mu)}$$

$$e^{-N \inf_{\mu \in O} J(\mu)} \lesssim \mathbb{P}[\mu^N \in O] \leq \mathbb{P}[\mu^N \in \bar{O}] \lesssim e^{-N \inf_{\mu \in \bar{O}} I(\mu)}$$

where $J = \overline{I|_A}$ and A is the set of nice fluctuations $\mu = \rho dx$ with ρ a solution to

$$\partial_t \rho = \partial_{xx} \Phi(\rho) + \partial_x (\Phi(\rho) \partial_x H)$$

for some $H \in C_{t,x}^{1,3}$. Problem: $\overline{I|_A} = I$?

This is a frequently observed problem: E.g. Fluctuations around Boltzmann equation

Theorem (The skeleton equation, [Fehrman, G. 2022])

Let $g \in L^2_{t,x}$, ρ_0 non-negative and $\int \rho_0 \log(\rho_0) dx < \infty$. There is a unique weak solution to

$$\partial_t \rho = \Delta \Phi(\rho) + \operatorname{div}(\Phi^{\frac{1}{2}}(\rho)g).$$

The map $g \mapsto \rho$, $L^2_{t,x} \rightarrow L^1_{t,x}$ is weak-strong continuous. E.g. $\Phi(\rho) = \rho^m$, $m \in [1, \infty)$.

Comment on the proof: Extending DiPerna-Lions, Ambrosio, Le Bris-Lions to non-linear PDE & combination with kinetic solution theory.

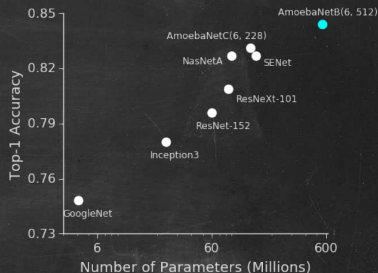
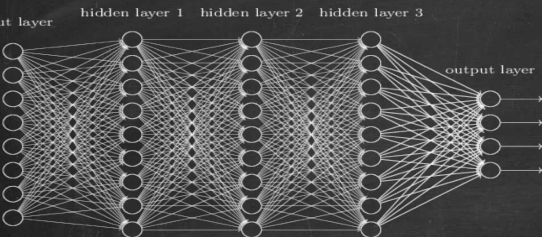
Theorem (LDP for zero range process, [G. Heydecker 2023])

The zero range process satisfies the full large deviations principle with rate function

$$I(\rho) = \|\partial_t \rho - \partial_{xx} \Phi(\rho)\|_{H^{-1}_{\Phi(\rho)}}.$$

Further Examples: Machine learning

Feed-forward neural network



Collecting all parameters $\theta = (\theta_1, \dots, \theta_M) \in \mathbb{R}^M$

Stochastic gradient descent / empirical risk minimization

$$\theta_{n+1} = \theta_n - \eta \nabla_{\theta} l(f_{\theta}(X), Y),$$

Scaling limits: Small learning rate η , overparametrization $M \rightarrow \infty$.

Shallow network: Empirical distribution $\mu_t^M := \frac{1}{M} \sum_i \delta_{\theta_t^i} \rightarrow \mu_t$ solution to

$$\partial_t \mu_t = \operatorname{div}(\nabla V(\mu_t, \cdot) \mu_t) + \sqrt{\sigma} \operatorname{div}(T(\mu_t, \cdot) \mu \circ \xi),$$

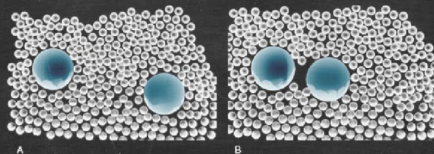
where ξ is space-time white noise and V, A, T_{μ} are non-local operators, see [Chen, Rotskoff, Bruna, Vanden-Eijnden, 2020], [Gess, Gvalani, Konarovskyi, 2022].

Fluctuating mean field systems

- Interacting agents with common noise

$$dX_t^i = \sigma(X_t^i, \frac{1}{L} \sum_{j \neq i} \delta_{X_t^j}) \circ dB_t + \alpha(X_t^i, \frac{1}{L} \sum_{j \neq i} \delta_{X_t^j}) dW_t^i$$

with B_t^i environmental noise, W_t^i individual noise.



- Informally: Conditioned empirical measure

$$\mu^L := \frac{1}{L} \sum_i \delta_{X^i} \rightarrow \mu$$

with $\mu = m dx$ satisfying the nonlocal conservative SPDE

$$\partial_t m = \frac{1}{2} \Delta (\alpha^2(x, m)m) + \nabla \cdot (\sigma(x, m)m \circ dB_t).$$

See, e.g. [Kurtz, Xiong 2001], [Coghi, Gess, 2019].

Stochastic thin films

$$\partial_t h = -\operatorname{div}(h^m \nabla \Delta h) + \operatorname{div}(h^{m/2} \xi)$$

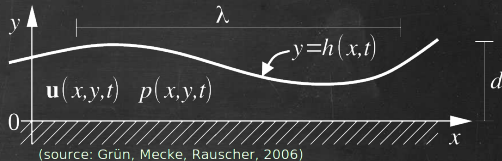
Fluctuating gradient flow structure

$$\partial_t h = -M(\rho) \frac{\partial E}{\partial h}(h) + M^{\frac{1}{2}}(h) \xi,$$

with $E(h) = \frac{1}{2} \int |\nabla h|^2 dx$ and $M(h) = \operatorname{div}(h^m \nabla \cdot)$,

Relevance to include thermal fluctuations:

- Improved prediction of empirical film rupture time scales
[Grün, Mecke, Rauscher 2006]
- Corrected spreading rate of droplets
[Davidovitch, Moro, Stone 2005]



Conclusion:

- Conservative SPDEs as universal fluctuating continuum models.
- Compared to classical hydrodynamic description they include information on fluctuations.
- Central limit fluctuations and large deviations of microscopic systems are correctly captured by the conservative SPDEs.

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