

Rare events in particle systems and stochastic PDE

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LSU

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I. The simple random walk

Independent coin flips: independent random variables with

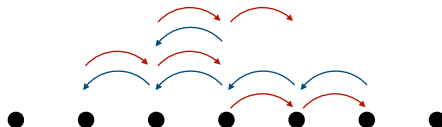
$$\mathbb{P}[X_i = 1] = \mathbb{P}[X_i = -1] = 1/2.$$

The simple random walk: the process $(S_n)_{n \in \mathbb{N}_0}$ with $S_0 = 0$ and

$$S_n = X_1 + \dots + X_n.$$

A normal random variable: with mean μ and variance σ ,

$$\mathbb{P}[a \leq Y < b] = \int_a^b \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{|x - \mu|^2}{2\sigma^2}\right).$$



A realization of S_{11} .

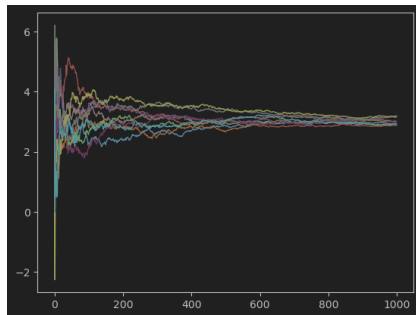
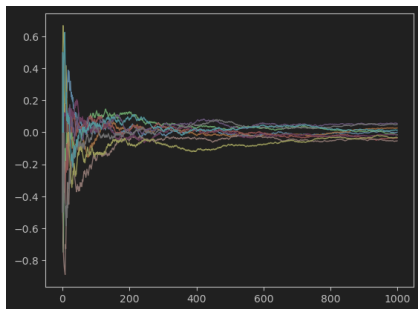
I. The simple random walk

The Law of Large Numbers: for independent coin flips,

$$\lim_{n \rightarrow \infty} \frac{X_1 + \dots + X_n}{n} = \mathbb{E}[X_1] = 0.$$

Normal random variables: independent with mean 3 and variance 25,

$$\lim_{n \rightarrow \infty} \frac{Y_1 + \dots + Y_n}{n} = \mathbb{E}[Y_1] = 3.$$



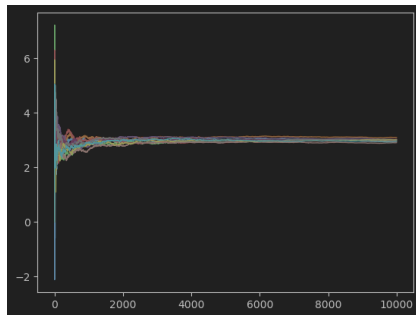
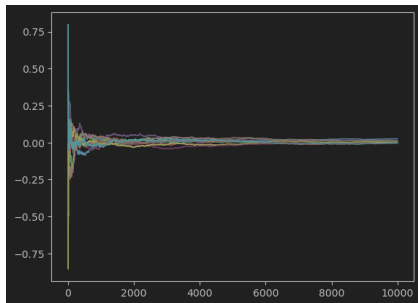
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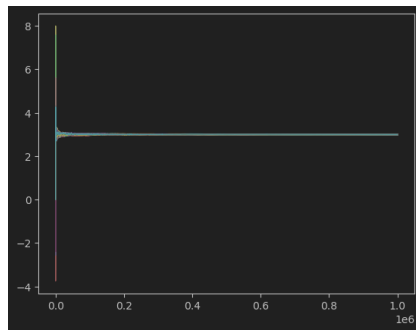
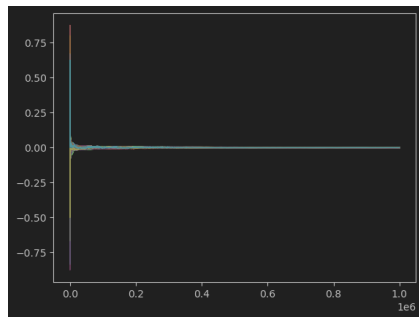
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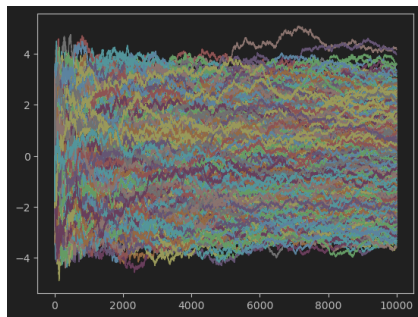
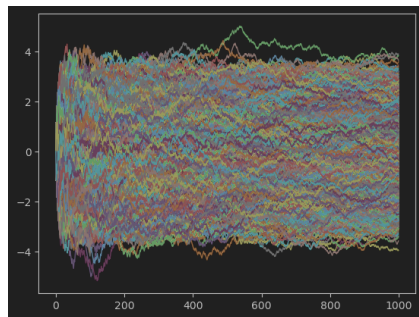
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The Central Limit Theorem: for independent coin flips, for every $a \leq b \in \mathbb{R}$,

$$\lim_{n \rightarrow \infty} \mathbb{P}\left[a \leq \frac{X_1 + \dots + X_n}{\sqrt{n}} < b\right] = \int_a^b (2\pi)^{-\frac{1}{2}} \exp\left(-\frac{x^2}{2}\right) dx.$$

Normal random variables: for independent mean zero variance one variables,

$$\mathbb{P}\left[a \leq \frac{Y_1 + \dots + Y_n}{\sqrt{n}} < b\right] = \int_a^b (2\pi)^{-\frac{1}{2}} \exp\left(-\frac{x^2}{2}\right) dx.$$



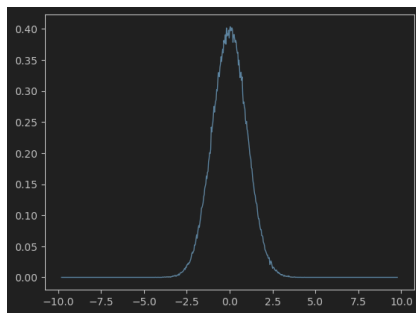
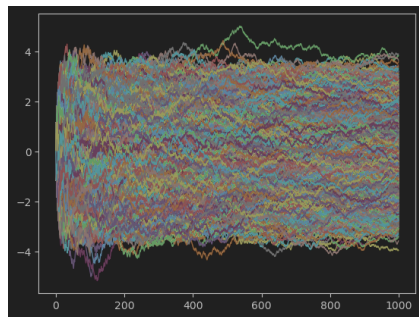
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I. The simple random walk

Linear central limit expansion: combining the LLN and CLT,

$$\frac{X_1 + \dots + X_n}{n} \simeq \mathbb{E}[X_1] + \frac{1}{\sqrt{n}} \mathcal{N}(0, 1).$$

Understanding rare events: for $\delta \in (0, \infty)$,

$$\begin{aligned} \mathbb{P} \left[\frac{X_1 + \dots + X_n}{n} \geq \delta \right] &= \mathbb{P} \left[\frac{X_1 + \dots + X_n}{\sqrt{n}} \geq \sqrt{n}\delta \right] \\ &\simeq \int_{\sqrt{n}\delta}^{\infty} (2\pi)^{-1} \exp \left(-\frac{x^2}{2} \right) dx \\ &\simeq \exp \left(-\frac{n\delta^2}{2} \right). \end{aligned}$$

Normal random variables: if Y_i are independent $\mathcal{N}(0, 1)$,

$$\begin{aligned} \mathbb{P} \left[\frac{Y_1 + \dots + Y_n}{n} \geq \delta \right] &= \int_{\sqrt{n}\delta}^{\infty} (2\pi)^{-1} \exp \left(-\frac{x^2}{2} \right) dx \\ &= \exp \left(-\frac{n\delta^2}{2} \right). \end{aligned}$$

I. The simple random walk

The simple random walk: for independent coin flips X_i ,

$$\overline{X}_n = \frac{X_1 + \dots + X_n}{n}.$$

A large deviations principle: the $\overline{X}_n$ satisfy an LDP with rate function I if

$$-\inf_{x \in A^\circ} I(x) \leq \liminf_{n \rightarrow \infty} \left(\frac{1}{n} \log \left(\mathbb{P}(\overline{X}_n \in A) \right) \right) \text{ for all } A \subseteq \mathbb{R},$$

and

$$\limsup_{n \rightarrow \infty} \left(\frac{1}{n} \log \left(\mathbb{P}(\overline{X}_n \in A) \right) \right) \leq -\inf_{x \in \overline{A}} I(x) \text{ for all } A \subseteq \mathbb{R}.$$

The rate function: if such a lower semicontinuous rate function exists, then

$$I: \mathbb{R} \rightarrow [0, \infty] \text{ is unique.}$$

The informal LDP: we will write

$$\mathbb{P}(\overline{X}_n \simeq x) \simeq \exp(-nI(x)) \text{ or, } \mathbb{P}(\overline{X}_n \in A) \simeq \exp \left(-n \left(\inf_{x \in A} I(x) \right) \right).$$

I. The simple random walk

The simple random walk: for independent coin flips X_i ,

$$\overline{X}_n = \frac{X_1 + \dots + X_n}{n}.$$

Cramér's theorem: the simple random walk satisfies an LDP with rate function

$$\begin{aligned} I(x) &= \tanh^{-1}(x)x - \log\left(\frac{1}{2}(e^{-\tanh^{-1}(x)} + e^{\tanh^{-1}(x)})\right) \\ &= \frac{x^2}{2} + o(x^2), \end{aligned}$$

and $I(x) = \infty$ if $|x| > 1$.

Linear central limit expansion: combining the LLN and CLT,

$$\overline{X}_n = \frac{X_1 + \dots + X_n}{n} \simeq \mathbb{E}[X_1] + \frac{1}{\sqrt{n}}\mathcal{N}(0, 1) = \frac{Y_1 + \dots + Y_n}{n},$$

for independent normal random variables Y_i .

Large deviations of the CLT expansion: for $\overline{Y}_n = n^{-1}(Y_1 + \dots + Y_n)$,

$$\mathbb{P}(\overline{Y}_n \simeq x) \simeq \exp\left(-\frac{nx^2}{2}\right) = \exp(-n\tilde{I}(x)) \quad \text{with rate function} \quad \tilde{I}(x) = \frac{x^2}{2}.$$

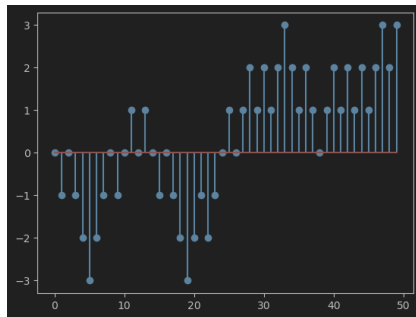
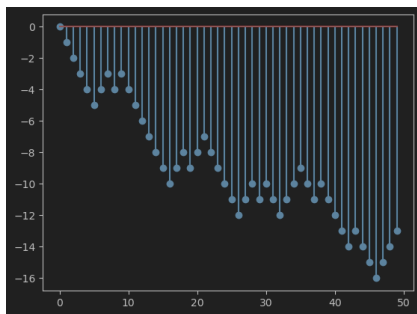
II. Brownian motion

The random walk as a path: for the random walk

$$S_n = X_1 + X_2 + \dots + X_n,$$

we define the path

$$W_t = S_{\lfloor t \rfloor}.$$



II. Brownian motion

The scaling limit: for the path $W_t = S_{\lfloor t \rfloor}$,

$$W_t^n = \frac{1}{\sqrt{n}} S_{\lfloor nt \rfloor}.$$

The increments: if $s < t$,

$$W_t^n - W_s^n = \frac{1}{\sqrt{n}} \left(\sum_{k=\lfloor ns \rfloor}^{\lfloor nt \rfloor} X_i \right) \simeq \frac{n(t-s) \text{ i.i.d. copies of } X_1}{\sqrt{n}}.$$

The central limit theorem: formally, as $n \rightarrow \infty$,

$$W_t^n - W_s^n \simeq \mathcal{N}(0, t-s),$$

a normal mean zero variance $t-s$ random variable.

Independent increments: if $0 \leq s_1 \leq s_2 \leq t_1 \leq t_2 < \infty$,

$$W_{s_2}^n - W_{s_1}^n \text{ and } W_{t_2}^n - W_{t_1}^n \text{ are independent.}$$

Time zero and jumps: we have that

$$W_0^n = 0 \text{ and the jump discontinuities are of order } \frac{1}{\sqrt{n}}.$$

II. Brownian motion

The scaling limit: for the path $W_t = S_{\lfloor t \rfloor}$,

$$W_t^n = \frac{1}{\sqrt{n}} S_{\lfloor nt \rfloor}.$$

Donsker's theorem: on path space, for a random path B_t ,

$$W_t^n \rightharpoonup B_t \text{ in law.}$$

Brownian motion: the path B_t almost surely satisfies that

(i) $B_0 = 0$ and that (ii) the path $t \rightarrow B_t$ is continuous,

that, for every $s \leq t \in [0, \infty)$,

$$\text{(iii)} \quad (B_t - B_s) = \mathcal{N}(0, t - s),$$

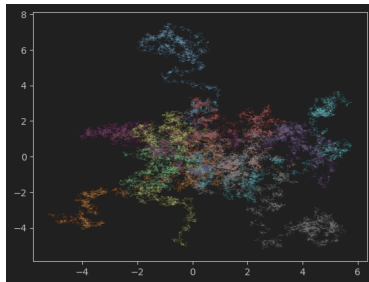
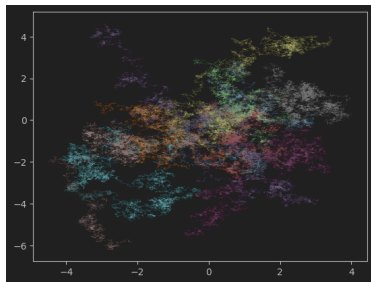
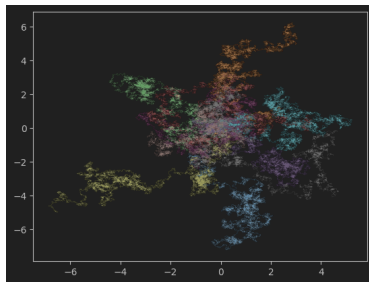
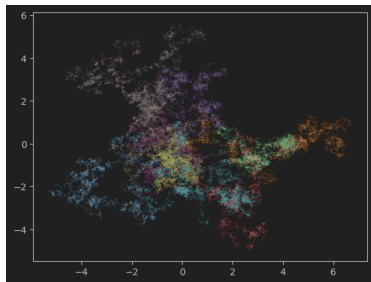
and that, for every $0 \leq s_1 \leq s_2 \leq t_1 \leq t_2$,

(iv) $(B_{s_2} - B_{s_1})$ and $(B_{t_2} - B_{t_1})$ are independent.

Uniqueness: such properties uniquely characterize Brownian motion.

II. Brownian motion

Brownian motion: 10 realizations on $[0, 10]$ in 2-dimensions,



II. Brownian motion

Small noise limit: for a Brownian motion B_t ,

$$W_t^\varepsilon = \sqrt{\varepsilon} B_t.$$

The law of large numbers: almost surely by continuity,

$$\lim_{\varepsilon \rightarrow 0} W_t^\varepsilon = 0.$$

The central limit theorem: we have that

$$\frac{1}{\sqrt{\varepsilon}} (W_t^\varepsilon - 0) = \frac{1}{\sqrt{\varepsilon}} \cdot \sqrt{\varepsilon} B_t = B_t.$$

Schilder's theorem: the paths W^ε satisfy an LDP with rate function

$$I(\pi) = \frac{1}{2} \inf \left\{ \int_0^t |g(s)|^2 ds : \pi \cdot = \int_0^\cdot g(s) ds \right\}.$$

That is, for the minimizer g ,

$$\mathbb{P}(W^\varepsilon \simeq \pi) \simeq \exp(-\varepsilon^{-1} I(\pi)) \quad \text{and} \quad \pi'_t = g_t.$$

III. The Ornstein–Uhlenbeck Process

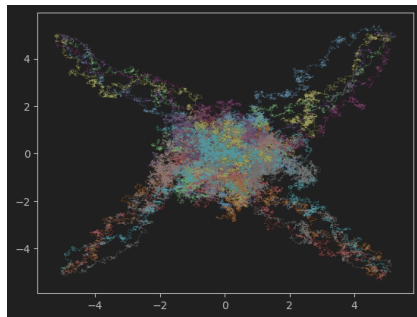
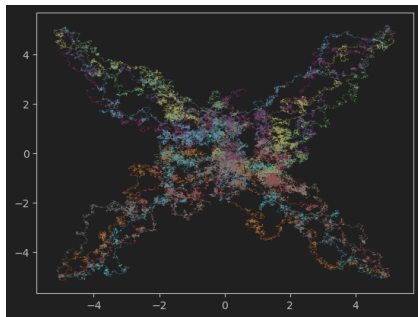
The Ornstein–Uhlenbeck Process: the random path X_t solving

$$dX_t = -X_t dt + dB_t \text{ in } (0, \infty).$$

An explicit representation: in terms of stochastic integration,

$$X_t = e^{-t}X_0 + \int_0^t e^{s-t} dB_s.$$

Example: 5 copies each on $[0, 3]$ and $[0, 10]$ from the four corners,



III. The Ornstein–Uhlenbeck process

The Ornstein–Uhlenbeck Process: the random path X_t solving

$$dX_t = -X_t dt + dB_t \text{ in } (0, \infty).$$

Small noise limit: the process Y_t^ε solves

$$dY_t^\varepsilon = -Y_t^\varepsilon dt - \sqrt{\varepsilon} dB_t.$$

The law of large numbers: almost surely as $\varepsilon \rightarrow 0$,

$$Y_t^\varepsilon \rightarrow Y_t \text{ for } Y_t \text{ solving } Y_t' = -Y_t.$$

The central limit theorem: for every $\varepsilon \in (0, 1)$,

$$Z_t^\varepsilon = \frac{1}{\sqrt{\varepsilon}}(Y_t^\varepsilon - Y_t) \text{ solves } dZ_t^\varepsilon = -Z_t^\varepsilon dt + dB_t,$$

and $Z_t^\varepsilon \simeq X_t$ in law.

Freidlin–Wentzell theorem: the paths Y^ε satisfy an LDP with rate function

$$I(\pi) = \frac{1}{2} \inf \left\{ \int_0^t |g(s)|^2 ds : \pi \cdot = \int_0^\cdot g(s) - \pi(s) \right\}.$$

For the minimizer g ,

$$\pi_t' = -\pi_t + g_t.$$

III. The Ornstein–Uhlenbeck process

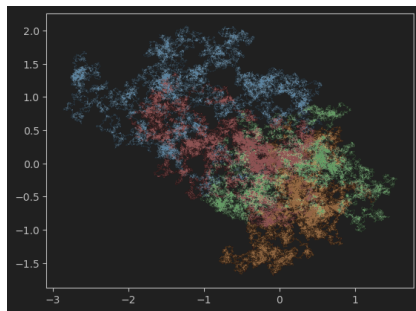
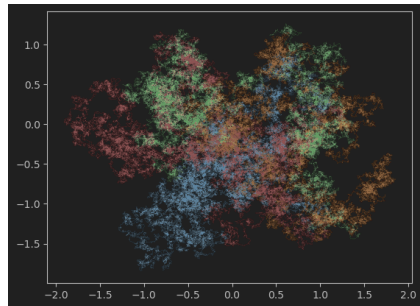
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Example: four Ornstein–Uhlenbeck processes from the origin on $[0, 3]$ and $[0, 10]$,



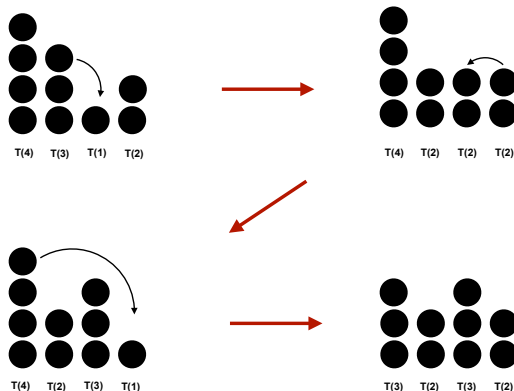
IV. The zero range processs

Random clocks: for $g: \mathbb{N}_0 \rightarrow \mathbb{N}_0$ be nondecreasing with $g(0) = 0$,

$$\mathbb{P}[0 \leq T(k) < a] = \int_0^a g(k) e^{-kt} dt.$$

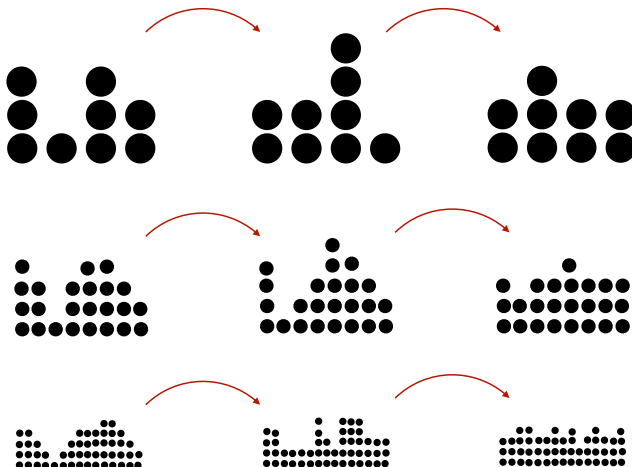
Independent particles: the case $g(k) = k$.

The zero range process: periodic with four sites,



IV. The zero range process

The parabolically rescaled process: the cases $N = 4, 8, 15$,



IV. The zero range process

The parabolically rescaled process: at scale N on $\mathbb{T}^1$,

$$\mu_t^N = \frac{1}{N^d} \sum_{x \in (\mathbb{Z}^d / N\mathbb{Z}^d)} \delta_{\frac{x}{N}} \cdot \eta_{N^2 t}^N(x).$$

The hydrodynamic limit: for $f \in C(\mathbb{T}^d \times [0, T])$ and $\delta \in (0, 1)$,

$$\lim_{N \rightarrow \infty} \mathbb{P} \left[|\langle f, \mu^N \rangle - \langle f, \bar{\rho} \rangle| > \delta \right] = 0,$$

where $\bar{\rho}: \mathbb{T}^d \times [0, T] \rightarrow \mathbb{R}$ is the unique solution of

$$\partial_t \bar{\rho} = \Delta \phi(\bar{\rho}).$$

for the mean local jump rate ϕ .

Independent particles: in this case $\phi(\xi) = \xi$ and

$$\partial_t \bar{\rho} = \Delta \bar{\rho}.$$

Queuing: in the case that $g(k) = \mathbf{1}_{\{k > 0\}}$, we have $\phi(\xi) = \frac{1}{1+\xi}$ and

$$\partial_t \bar{\rho} = \Delta \left(\frac{\bar{\rho}}{1 + \bar{\rho}} \right).$$

IV. The zero range process

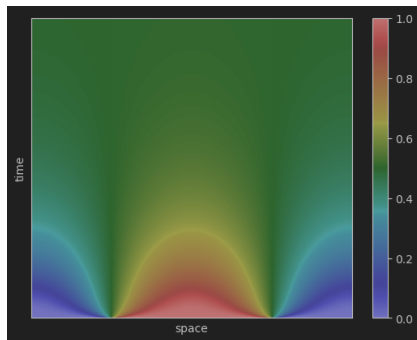
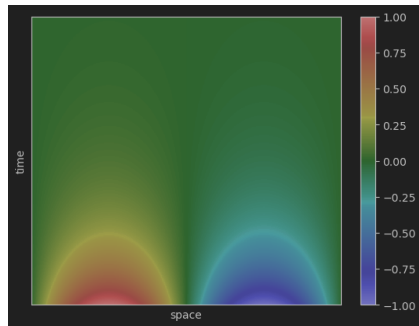
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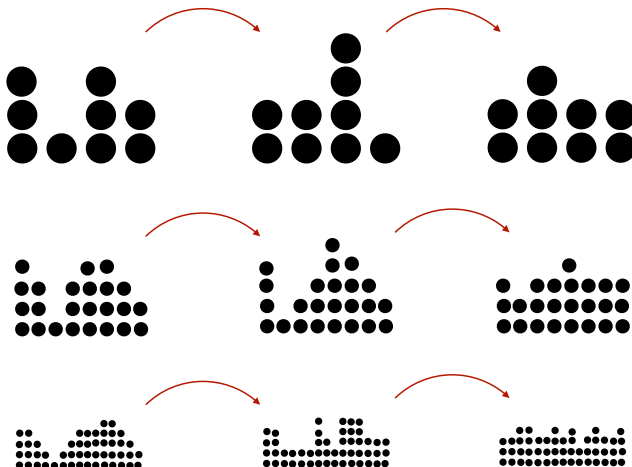
$$\mu_t^N \rightharpoonup \bar{\rho} dx dt.$$

Example: the case of $\phi(\xi) = \xi$ with $\rho_0(x) = \sin(2\pi x)$ and $\rho_0 = \mathbf{1}_{(1/4, 3/4)}$,



IV. The zero range process

The parabolically rescaled process: the cases $N = 4, 8, 15$,



IV. The zero range process

Space-time white noise: a random scalar $d\xi$ on $\mathbb{T}^d \times [0, T]$ satisfying

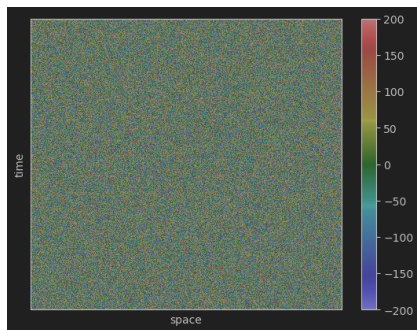
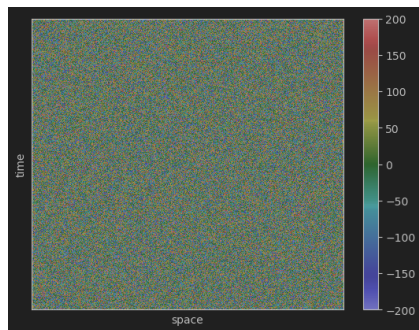
$$\mathbb{E}[d\xi(x, s) d\xi(y, t)] = \delta_0(x - y) \delta_0(s - t),$$

or, as a distribution, for $f, g \in L^2(\mathbb{T}^d \times [0, T])$,

$$\mathbb{E}[\langle f, d\xi \rangle \langle g, d\xi \rangle] = \langle f, g \rangle.$$

Fourier representation: for an orthonormal $L^2(\mathbb{T}^d)$ -basis e_k ,

$$\xi(x, t) = \sum_{k \in \mathbb{Z}^d} e_k(x) B_t^k \quad \text{and} \quad d\xi(x, t) = \sum_{k \in \mathbb{Z}^d} e_k(x) dB_t^k.$$



IV. The zero range process

The parabolically rescaled process: at scale N on $\mathbb{T}^1$,

$$\mu_t^N = \frac{1}{N^d} \sum_{x \in (\mathbb{Z}^d / N\mathbb{Z}^d)} \delta_{\frac{x}{N}} \cdot \eta_{N^2 t}^N(x).$$

The hydrodynamic limit: for $\partial_t \bar{\rho} = \Delta \phi(\bar{\rho})$,

$$\mu_t^N \rightharpoonup \bar{\rho} \, dx \, dt.$$

The generalized Ornstein-Uhlenbeck process: as $N \rightarrow \infty$,

$$N^{\frac{d}{2}} (\mu_t^N - \bar{\rho} \, dx \, dt) \rightharpoonup v,$$

for v satisfying, for a d -dimensional space time white noise $d\xi$,

$$\partial_t v = \nabla \cdot \phi'(\bar{\rho}) \nabla v - \nabla \cdot (\phi^{\frac{1}{2}}(\bar{\rho}) \, d\xi).$$

The case of independent particles: we have that

$$\partial_t v = \Delta v - \nabla \cdot (\sqrt{\bar{\rho}} \, d\xi).$$

IV. The zero range process

The hydrodynamic limit: for $\partial_t \bar{\rho} = \Delta \phi(\bar{\rho})$,

$$\mu_t^N \rightharpoonup \bar{\rho} dx dt.$$

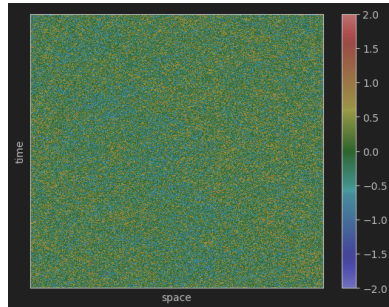
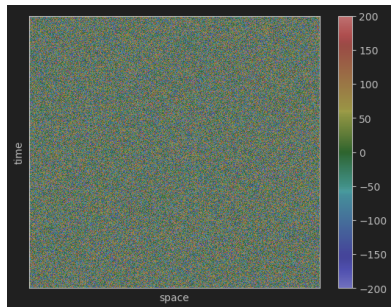
The central limit theorem: as $N \rightarrow \infty$,

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for v satisfying, for a d -dimensional space time white noise $d\xi$,

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Example: the case $\phi(\xi) = \xi$,



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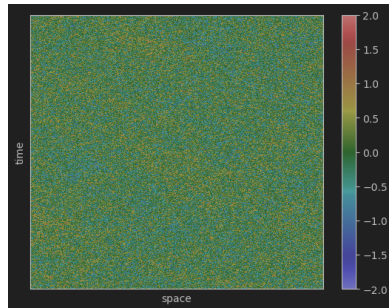
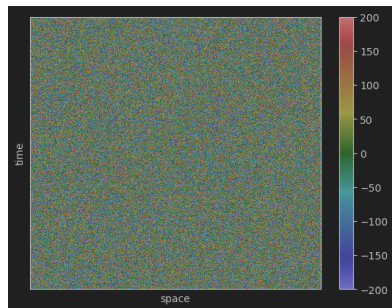
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The linear CLT expansion: does the expansion

$$\mu_t^N \simeq \bar{\rho} + N^{-\frac{d}{2}} v.$$

Space-time white noise: if $\xi = \sum_{k \in \mathbb{Z}^d} e_k B_t^k$, for the rate function

$$I(f) = \frac{1}{2} \inf \{ \|g\|_{L^2}^2 : \partial_t f(x, t) = g(x, t) \},$$

we have that

$$P(\varepsilon^{\frac{d}{2}} \xi \simeq f) \simeq \exp(-\varepsilon^{-d} I(f)).$$

IV. The zero range process

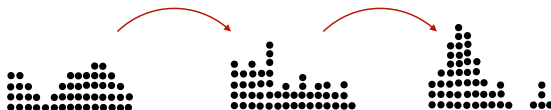
The Ornstein–Uhlenbeck rate function: the linear expansion suggests that, for

$$\tilde{I}(\pi) = \frac{1}{2} \inf \{ \|g\|_{L^2}^2 : \partial_t \pi = \Delta \phi(\pi) - \nabla \cdot (\phi^{\frac{1}{2}}(\bar{\rho})g) \},$$

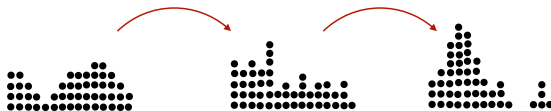
we have

$$\mathbb{P}(\mu_t^N \simeq \pi) \simeq \exp(-N^d \tilde{I}(\pi)).$$

The controlled OU-equation: $\partial_t \bar{\rho} = \Delta \bar{\rho} - \nabla \cdot (\sqrt{\bar{\rho}}g)$ on $\mathbb{T}^1 \times [0, T]$,



The controlled equation: $\partial_t \rho = \Delta \rho - \nabla \cdot (\sqrt{\rho} \cdot g)$ on $\mathbb{T}^1 \times [0, T]$,



IV. The zero range process

The rate function: we have

$$I(\pi) = \frac{1}{2} \inf \left\{ \|g\|_{L^2_{t,x}}^2 : m = \frac{d\pi}{dx} \text{ and } \partial_t m = \Delta \phi(m) - \nabla \cdot (\phi^{\frac{1}{2}}(m)g) \right\}.$$

Large Deviations Principle [Benois, Kipnis, Landim; 1995]

Let μ^N be the parabolically rescaled zero range process on $\mathbb{T}^d \times [0, T]$ (or, $\mathbb{R} \times [0, T]$) started from stationarity. For every closed subset $A \subseteq D([0, T]; \mathcal{M}_+(\mathbb{T}^1))$,

$$\limsup_{N \rightarrow \infty} \left(\frac{1}{N} \log (\mathbb{P}[\mu^N \in U]) \right) \leq - \inf_{\pi \in A} I(\pi),$$

and, for open subset $U \subseteq D([0, T]; \mathcal{M}_+(\mathbb{T}^1))$,

$$\liminf_{N \rightarrow \infty} \left(\frac{1}{N} \log (\mathbb{P}[\mu^N \in A]) \right) \geq - \inf_{\pi \in U} \overline{I}_{\mathcal{S}}^{\text{lsc}}(\pi),$$

for the space $\mathcal{S} = \{ \pi : \partial_t \pi = \Delta \phi(\pi) - \nabla \cdot (\phi(\pi) \nabla H) \text{ for } H \in C_{t,x}^{1,3} \}$.

Counterexample: for LDP of Kac's conservative particle system [Heydecker; 2022].

Fluctuating Boltzmann equation: [Bod., Gal., S.-Ray. Sim.; 2022].

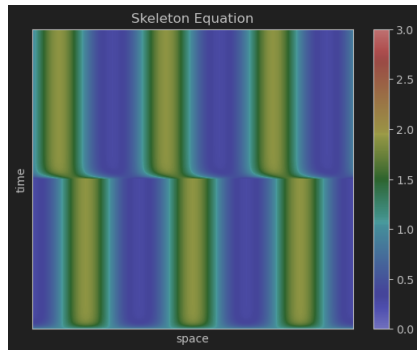
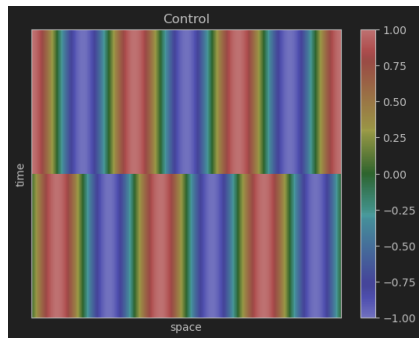
IV. The zero range process

The skeleton equation: on $\mathbb{T}^1 \times [0, T]$,

$$\partial_t \rho = \Delta \rho - \nabla \cdot (\sqrt{\rho} g).$$

Example: for small times,

$$g(x, t) = \sin(6\pi x) \text{ and, for larger times } g(x, t) = \cos(6\pi x).$$



IV. The zero range process

DiPerna–Lions theory [DiPerna, Lions; 1989], [Ambrosio; 2004]

Let $b \in (L_t^1 BV_x)^d$ and $(\nabla \cdot b) \in (L_t^1 L_x^\infty)^d$. Then, for every $\rho_0 \in L^\infty(\mathbb{T}^d)$,

$$\partial_t \rho = \nabla \cdot (\rho b),$$

has a unique weak solution in $(L^1 \cap L^\infty)(\mathbb{T}^d \times [0, T])$.

Counterexample: [Depauw; 2003] shows that this is essentially optimal.

Equivalence of weak and renormalized kinetic solutions [F., Gess; 2023], [F., Gess, Heydecker; 2025]

The skeleton equation

$$\partial_t \rho = \Delta \phi(\rho) - \nabla \cdot (\phi^{\frac{1}{2}}(\rho) g) \quad \text{in } \mathbb{T}^d \times (0, T) \quad \text{with } \rho(\cdot, 0) = \rho_0,$$

is well-posed for L^2 -controls and data with finite entropy.

Full large deviations principle [F., Gess; 2023], [F., Gess, Heydecker; 2025]

The rate functions I and $\overline{I}_{|\mathcal{S}}(\rho)$ are equal, and the zero range process satisfies a LDP with matching upper and lower bounds with rate function I .

V. Stochastic PDE

The Dean–Kawasaki equation: for the empirical densities

$$m_n(x, t) = \frac{1}{n} \sum_{i=1}^n \delta_0(x - B_t^i),$$

we have formally in one-dimension that

$$\partial_t m_n = \frac{1}{2} \Delta m_n - N^{-\frac{d}{2}} \nabla \cdot (\sqrt{m_n} \, d\xi).$$

Supercriticality: no renormalization theory, in general, for

$$\partial_t \rho_\varepsilon = \frac{1}{2} \Delta \phi(\rho_\varepsilon) - \varepsilon^{\frac{d}{2}} \nabla \cdot (\phi^{\frac{1}{2}}(\rho_\varepsilon) \xi).$$

Existence and uniqueness [F., Gess; 2024]

There exists a unique *stochastic kinetic solution* to the equation

$$\partial_t \rho_\varepsilon = \frac{1}{2} \Delta \phi(\rho_\varepsilon) - \varepsilon^{\frac{d}{2}} \nabla \cdot (\phi^{\frac{1}{2}}(\rho_\varepsilon) \circ d\xi^\delta),$$

for nonnegative, integrable initial data. Furthermore, any two solutions $\mathbb{P}$ -a.s. satisfy

$$\max_{t \in [0, T]} \|\rho_1 - \rho_2\|_{L^1} \leq \|\rho_{1,0} - \rho_{2,0}\|_{L^1}.$$

Quantitative LLN and Central Limit Theorem [Clini, F.; 2025]

The scaling limit: let $\delta(\varepsilon)$ be any sequence satisfying, as $\varepsilon \rightarrow 0$,

$$\varepsilon \delta(\varepsilon)^{-(1+\frac{2}{d})} \rightarrow 0 \quad \text{and} \quad \delta(\varepsilon) \rightarrow 0,$$

and for every $\varepsilon \in (0, 1)$ let ρ^ε be the solution

$$\partial_t \rho^\varepsilon = \Delta \phi(\rho^\varepsilon) - \varepsilon^{\frac{d}{2}} \nabla \cdot (\phi^{\frac{1}{2}}(\rho^\varepsilon) \circ \xi^{\delta(\varepsilon)}).$$

Law of Large Numbers: then almost surely with a quantitative rate, as $\varepsilon \rightarrow 0$,

$$\rho^\varepsilon \rightarrow \bar{\rho} \quad \text{for} \quad \partial_t \bar{\rho} = \Delta \phi(\bar{\rho}).$$

Central Limit Theorem: then as distributions, as $\varepsilon \rightarrow 0$,

$$\varepsilon^{-\frac{1}{2}}(\rho^\varepsilon - \bar{\rho}) \rightarrow v \quad \text{for} \quad \partial_t v = \Delta(\phi'(\bar{\rho})v) - \nabla \cdot (\phi^{\frac{1}{2}}(\bar{\rho})\xi).$$

Universality: the central limit theorem is a statement about “universal” behavior:

$$\partial_t \tilde{\rho}^\varepsilon = \Delta \phi(\tilde{\rho}^\varepsilon) - \varepsilon^{\frac{d}{2}} \nabla \cdot (\phi^{\frac{1}{2}}(\bar{\rho}) d\xi^{\delta(\varepsilon)}).$$

Linear diffusions: [Cornalba, Fischer; 2023], [Cornalba, Fischer, Ingmanns, Raithel; 2023], [Djurdjevac, Kremp, Perkowski; 2022]

V. Stochastic PDE

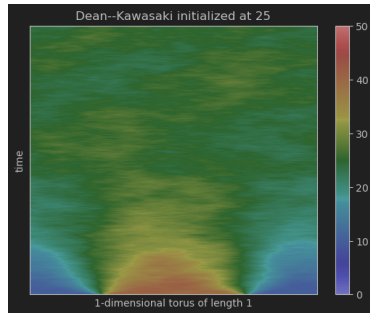
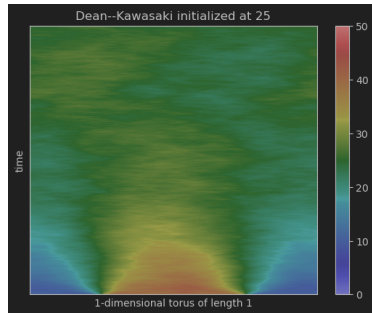
The Large Deviations Principle [F., Gess; 2024]

The LDP: along the scaling limit

$$\varepsilon \delta(\varepsilon)^{-(1+\frac{2}{d})} \rightarrow 0 \quad \text{and} \quad \delta(\varepsilon) \rightarrow 0,$$

the solutions satisfy a LDP in $L^1_{t,x}$ and $C_t \mathcal{M}_+$ with rate function

$$I(\rho) = \frac{1}{2} \inf \{ \|g\|_{L^2}^2 : \partial_t \rho = \Delta \phi(\rho) - \nabla \cdot (\phi^{\frac{1}{2}}(\rho) g) \}.$$



V. Stochastic PDE

The Dean–Kawasaki equation: along the scaling regime $\varepsilon \delta^{-(1+\frac{2}{d})} \rightarrow 0$,

$$\partial_t \rho^\varepsilon = \Delta \rho^\varepsilon - \varepsilon^{\frac{d}{2}} \nabla \cdot (\sqrt{\rho^\varepsilon} \circ \xi^{\delta(\varepsilon)}).$$

