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Stochastic modeling of diffusion in dynamical systems: three examples

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Outline				

Motivation:

dynamical systems, diffusion and stochastic modeling

Diffusion in three random walk-like examples:

- non-chaotic 'slicer' map
- Ø dissipative randomly perturbed standard map
- a simple random dynamical system

Onclusion:

pitfalls when relating the above three layers to each other

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 Microscopic chaos in a glass of water?



water molecules droplet of ink

• dispersion of a droplet of ink by diffusion

- chaotic collisions between billiard balls
- chaotic hypothesis:

microscopic chaos ↓ macroscopic diffusion

Gallavotti, Cohen (1995)

P.Gaspard et al. (1998): experiment on small colloidal particle in water; diffusion due to microscopic chaos based on positive pattern entropy per unit time $h(\epsilon, \tau) \leq h_{KS} = \sum_{\lambda_i > 0} \lambda_i$

The random wind tree model

counterexample:



Ehrenfest, Ehrenfest (1959)

no positive Lyapunov exponent, hence non-chaotic dynamics

Dettmann et al. (1999): generates trajectories and $h(\epsilon, \tau)$ indistinguishable from the colloidal particle dynamics



Microscopic models, diffusion and stochastic modeling

conclusion:

- theory: (chaotic) model \rightarrow diffusion
- experiment: diffusion → (chaotic) model?
- \Rightarrow non-trivial interplay microscopic model \leftrightarrow diffusion

theme of this talk:

add yet a third layer of stochastic modeling



two questions:

- what type of diffusion is generated by a dynamical system?
- 2 can it be reproduced by some stochastic model?

- in the following only diffusion in one dimension
- key quantity: mean square displacement

$$< x^2 >:= \int dx \ x^2
ho(x,t) \sim t^{\gamma}$$

- note: three basic types of diffusion
 - there is not only 'Brownian' (normal) diffusion with $\gamma = 1$ but also anomalous diffusion:
 - 2 subdiffusion with $\gamma < 1$

and

3 superdiffusion with $\gamma > 1$

(plus more exotic types)

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I. The slicer map

Pictorial construction

a one-dimensional 'random walk-like' but fully deterministic system; diffusion of a density of points from uniform initial density in **space (m) - discrete time (n) diagram**:



'slicers' at points (of Lebesgue measure zero) split the density no stretching, hence zero Lyapunov exponent: **no chaos!**

Stochastic modeling of diffusion in dynamical systems

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• consider a chain of intervals $\widehat{M} := M \times \mathbb{Z}$, M := [0, 1]with point $\widehat{X} = (x, m)$ in \widehat{M} , where $\widehat{M}_m := M \times \{m\}$ is the *m*-th cell of \widehat{M}

• subdivide each \widehat{M}_m in subintervals, separated by points called slicers: $\{1/2\} \times \{m\}$, $\{\ell_m\} \times \{m\}$, $\{1 - \ell_m\} \times \{m\}$, where $0 < \ell_m < 1/2$ for every $m \in \mathbb{Z}$ with

$$\ell_m(\alpha) = \frac{1}{\left(|m|+2^{1/\alpha}\right)^{\alpha}}, \, \alpha > 0$$

• slicer map: $S: \widehat{M} \to \widehat{M}$, $\widehat{X}_{n+1} = S(\widehat{X}_n)$, $n \in \mathbb{N}$ with $S(x,m) = \begin{cases} (x,m-1) & \text{if } 0 \le x < \ell_m \text{ or } \frac{1}{2} < x \le 1 - \ell_m, \\ (x,m+1) & \text{if } \ell_m \le x \le \frac{1}{2} \text{ or } 1 - \ell_m < x \le 1. \end{cases}$
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 Main result: diffusive properties

Proposition (Salari et al., 2015)

Given $\alpha \geq 0$ and a uniform initial distribution in \widehat{M}_0 , we have

- $\alpha = 0$: ballistic motion with MSD $\langle \hat{X}_n^2 \rangle \sim n^2$
- 2 $0 < \alpha < 1$: superdiffusion with MSD $\langle \hat{X}_n^2 \rangle \sim n^{2-\alpha}$
- 3 $\alpha = 1$: normal diffusion with linear MSD $\langle \hat{X}_n^2 \rangle \sim n$ non-chaotic normal diffusion with non-Gaussian density

(a) $1 < \alpha < 2$: subdiffusion with MSD $\langle \hat{X}_n^2 \rangle \sim n^{2-\alpha}$ subdiffusion with ballistic peaks

5 $\alpha = 2$: logarithmic subdiffusion with MSD $\langle \hat{X}_n^2 \rangle \sim \log n$ a bit exotic

6 $\alpha > 2$: localisation in the MSD with $\langle \hat{X}_n^2 \rangle \sim \text{const.}$ non-trivial phenomenon Motivation

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Higher order moments

Theorem (Salari et al., 2015)

For $\alpha \in (0, 2]$ the moments $\langle \widehat{X}_n^p \rangle$ with p > 2 even and uniform initial distribution in \widehat{M}_0 have the asymptotic behavior

 $\langle \widehat{X}^{p}_{n}
angle \sim n^{p-lpha}$

while the odd moments (p = 1, 3, ...) vanish.



Matching to stochastic dynamics?

• one-dimensional stochastic Lévy Lorentz gas:

point particle moves ballistically between static point scatterers on a line from which it is transmitted / reflected with probability 1/2

distance r between two scatterers is a random variable iid from the Lévy distribution

$$\lambda(\mathbf{r}) := \beta \mathbf{r}_0^{\beta} \frac{1}{\mathbf{r}^{\beta+1}}, \ \mathbf{r} \in [\mathbf{r}_0, \infty) \ , \ \beta > \mathbf{0}$$

with cutoff ro

 \rightarrow model exhibits only superdiffusion

 \rightarrow all moments scale with the slicer moments for $\alpha \in (0, 1]$ (piecewise linearly depending on parameters)



Matching to stochastic dynamics?

• Lévy walk modeled by CTRW theory:

 \rightarrow moments calculated to $\sim t^{p+1-\beta}$ for $p > \beta$, $1 < \beta < 2$: match to slicer superdiffusion with $\beta = 1 + \alpha$

- \rightarrow but conceptually a totally different process
- correlated Gaussian stochastic processes:

modeled by a generalized Langevin equation with a power law memory kernel

- \rightarrow formal analogy in the subdiffusive regime
- \rightarrow but Gaussian distribution and a conceptual mismatch

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II. The dissipative randomly perturbed standard map

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The standard map and diffusion

• paradigmatic Hamiltonian dynamical system:

standard map

 $x_{n+1} = x_n + y_n \mod 2\pi$

 $y_{n+1} = y_n + K \sin x_{n+1}$

derived from kicked rot(at)or where $x_n \in \mathbb{R}$ is an angle, $y_n \in \mathbb{R}$ the angular velocity with $n \in \mathbb{N}$ and K > 0 the kick strength

• define diffusion coefficient as

$$D(K) = \lim_{n\to\infty} \frac{1}{n} < (y_n - y_0)^2 >$$

with ensemble average over the initial density $< \ldots >= \int dx \, dy \, \varrho(x, y) \ldots , \, x \in [0, 2\pi) , \, y = y_0 \in [0, 2\pi)$

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Diffusion in the standard map

analytical (Rechester, White, 1980) and numerical studies of parameter-dependent diffusion $D_{eff}(K)$:



Manos, Robnik, PRE (2014)

- D(K) is highly irregular
- for some *K* there is superdiffusion with mean square displacement $\langle y_n^2 \rangle \sim n^{\gamma}$, $\gamma > 1$ due to accelerator modes

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The dissipative standard map

model damping in the standard map by $x_{n+1} = x_n + y_n \mod 2\pi$ $y_{n+1} = (1 - \nu)y_n + f_0 \sin x_{n+1}$ with $\nu \in [0, 1]$:



Feudel, Grebogi, Hunt, Yorke, PRE (1996) • islands in phase space for $\nu = 0$ (left) become coexisting periodic attractors (right): 150 found for $\nu = 0.02$, $f_0 = 4$ • simple argument yields $|y_n| < y_{max}$: quick trapping



Dissipative dynamics and random perturbations

Question: What happens to dissipative deterministic dynamics $\mathbf{x}_{n+1} = \mathbf{f}(\mathbf{x}_n)$ under random perturbations?

Consider the dissipative standard map with additive noise:

 $x_{n+1} = x_n + y_n + \epsilon_{x,n} \mod 2\pi$ $y_{n+1} = (1-\nu)y_n + f_0 \sin x_{n+1} + \epsilon_{y,n}$

with iid random variables $\epsilon_n = (\epsilon_{x,n}, \epsilon_{y,n})$ drawn from uniform distribution bounded by $||\epsilon_n|| < \xi$ of noise amplitude ξ

perturbed dynamics $\mathbf{F}(\mathbf{x}_i) = \mathbf{f}(\mathbf{x}_i) + \epsilon_i$:





From attractors to hopping on pseudo attractors

Consequences of the random perturbations:

• beyond a noise threshold $\xi \ge \xi_0$ the attracting sets $W^S(\Lambda_i)$ lose their stability due to holes



- the (invariant) attractors become (quasi-invariant) pseudo attractors from which there is noise-induced escape
- the noise induces a hopping process between all coexisting pseudo attractors

Stochastic modeling of diffusion in dynamical systems

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Intermittency and stickiness

the resulting perturbed dissipative dynamics is intermittent:



 $f_0 = 4 \;,\; \xi = 0.06 \;,\; \nu = 0.002$

• stickiness to pseudo attractors measured by criterion that maximal eigenvalue of the Jacobian matrix along orbit < 1

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Continuous time random walk theory

match simulation results to **CTRW theory** (Montroll, Weiss, Scher, 1973): define stochastic process by master equation with *waiting time distribution* w(t) and *jump distribution* $\lambda(x)$

$$\varrho(\mathbf{x},t) = \int_{-\infty}^{\infty} d\mathbf{x}' \lambda(\mathbf{x}-\mathbf{x}') \int_{0}^{t} dt' \ \mathbf{w}(t-t') \ \varrho(\mathbf{x}',t') + (1 - \int_{0}^{t} dt' \ \mathbf{w}(t')) \delta(\mathbf{x})$$

structure: jump + no jump for points starting at (x, t) = (0, 0)Fourier-Laplace transform yields Montroll-Weiss eqn (1965)

$$\hat{\hat{arrho}}(k,s) = rac{1- ilde{w}(s)}{s}rac{1}{1-\hat{\lambda}(k) ilde{w}(s)}$$

with mean square displacement $\langle x^2 \tilde{(s)} \rangle = -\frac{\partial^2 \hat{\varrho}(k,s)}{\partial k^2}$

according to CTRW theory solving the MW eqn. for

- a power law waiting time distribution $w(t) \sim t^{-(\gamma+1)}$ with jump distribution $\lambda(x) = \delta(|x| - const.)$
- yields a mean square discplacement of < x²(t) >~ t^γ and
- a stretched exponential position pdf, approximately given by $P_n(y) \sim \exp(-cx^{2/(2-\gamma)})$

crucial fit parameter: γ ; check these three predictions in numerical experiments



CTRW theory and mean square displacement

 $< y^2(n) >$ for different noise amplitudes ξ at $\nu = 0.002$:



- transient subdiffusion $\langle y^2(n) \rangle \sim n^{\gamma}$ up to n < 1000
- only small variation of $0.85 < \gamma < 0.95$ for different ξ ; for
- $\xi = 0.06$ we have $\gamma \simeq 0.95$



probability distributions P(t) of escape times *t* from pseudo attractors; dissipation $\nu = 0.002$ with different noise strength ξ :



- transition from power law (stickiness) to exponential
- transition takes longer when $\xi \rightarrow 0$
- the dashed red line represents the CTRW theory prediction of $P(t) \sim t^{-1.95}$ corresponding to $< y^2(n) > \sim n^{0.95}$

CTRW theory and position pdf

$P_n(y)$ for position y at different time steps n:



- $\xi = 0.06 \; , \; \nu = 0.002$ 'Gaussian-like' diffusive spreading up to n < 1000
- localization trivially due to boundedness of pseudo attractors
- \bullet CTRW theory pdf (green lines) for $\gamma = 0.95$ corrects mismatch in tails

Stochastic modeling of diffusion in dynamical systems

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III. A random dynamical system

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Main re	sults			

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• **central theme:** interplay between *dynamical systems*, *diffusion and stochastic modeling*

• main results:

- dynamical systems can feature novel types of (anomalous) diffusion
- naive matching to stochastic models can be misleading and difficult
- **outlook:** perhaps dynamical systems theory can inspire stochastic theory to invent new stochastic processes? *and* take your data *seriously*!!!

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Acknowledgement and references

work performed with all authors on the following references:

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