

Large deviations of linear diffusions

Hugo Touchette

Department of Mathematical Sciences
Stellenbosch University, South Africa

Frontiers in Statistical Physics
Raman Research Institute, Bangalore, India
December 2023

- Work with
Johan du Buisson and Thamu Mnyulwa
PRE **107** 2023; PRE **108** 2023



Dynamical large deviations

- Markov process: $X_t, t \in [0, T]$
- Observable: $A_T[x]$

Probability distribution

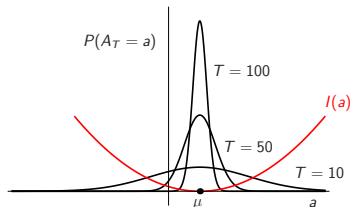
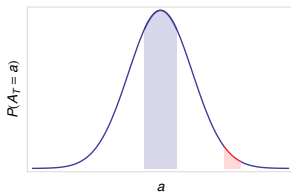
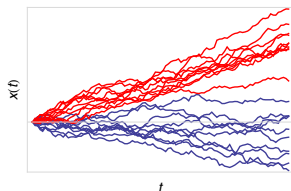
$$P(A_T = a) \approx e^{-T I(a)}$$

Generating function

$$E[e^{TkA_T}] \approx e^{T\lambda(k)}$$

Effective process

- How are fluctuations created?
- Conditioned process: $X_t | A_T = a$
- Markov when $T \rightarrow \infty$



Dynamical large deviations

- Markov process: $X_t, t \in [0, T]$
- Observable: $A_T[x]$

Probability distribution

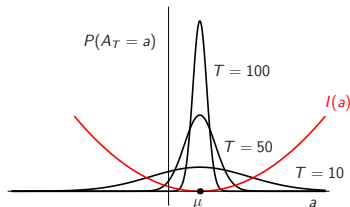
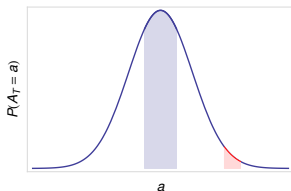
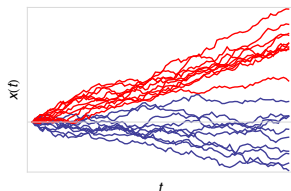
$$P(A_T = a) \approx e^{-T I(a)}$$

Generating function

$$E[e^{TkA_T}] \approx e^{T\lambda(k)}$$

Effective process

- How are fluctuations created?
- Conditioned process: $X_t | A_T = a$
- Markov when $T \rightarrow \infty$



Dynamical large deviations

- Markov process: $X_t, t \in [0, T]$
- Observable: $A_T[x]$

Probability distribution

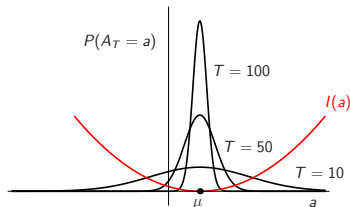
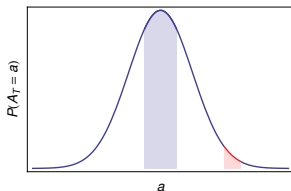
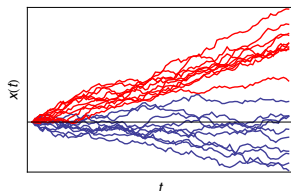
$$P(A_T = a) \approx e^{-T I(a)}$$

Generating function

$$E[e^{TkA_T}] \approx e^{T\lambda(k)}$$

Effective process

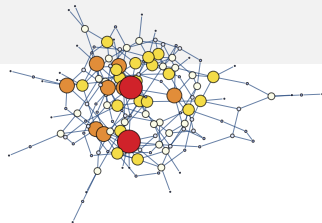
- How are fluctuations created?
- Conditioned process: $X_t | A_T = a$
- Markov when $T \rightarrow \infty$



Applications

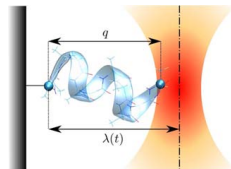
Markov chains

- Random walks
- Jump processes
- Occupation, displacement, currents



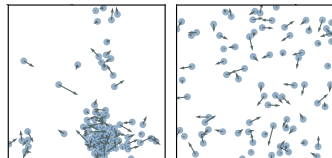
Diffusions

- Langevin equations, SDEs
- Work, heat, entropy production
- Active OUP

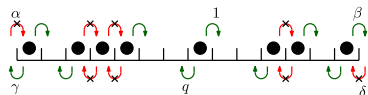


Many-particle dynamics

- Zero-range process
- Exclusion process
- Harmonic chains
- Density and current



[Reviews: HT 2009, 18; Derrida 07; Bertini et al 07]



Linear diffusions

- Quadratic observables: Bercu et al 1997
- Entropy production: Visco 2006; Chernyak et al 2006; Jaksic et al 2016
- Nonequilibrium work: Kwon, Noh & Park 2011, Noh 2014
- Coupled oscillators: Kundu et al 2011, Sabhapandit, 2012; Pal & Sabhapandit 2013
- More general setting: Mazzolo & Monthus 2023

This work

- General linear SDEs in \mathbb{R}^n
- More general class of observables
- Not based on path integrals
- Derive effective process (remains linear)
- Link with control theory (LQG)

Linear diffusions

- Quadratic observables: Bercu et al 1997
- Entropy production: Visco 2006; Chernyak et al 2006; Jaksic et al 2016
- Nonequilibrium work: Kwon, Noh & Park 2011, Noh 2014
- Coupled oscillators: Kundu et al 2011, Sabhapandit, 2012; Pal & Sabhapandit 2013
- More general setting: Mazzolo & Monthus 2023

This work

- General linear SDEs in \mathbb{R}^n
- More general class of observables
- Not based on path integrals
- Derive effective process (remains linear)
- Link with control theory (LQG)

Linear diffusions

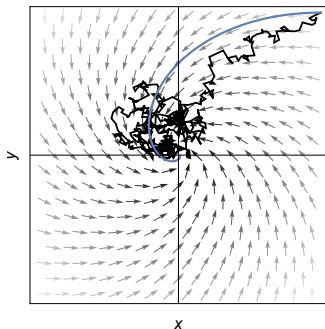
- Dynamics:

$$d\mathbf{X}(t) = -M\mathbf{X}(t)dt + \sigma d\mathbf{W}(t)$$

- $\mathbf{X}(t) \in \mathbb{R}^n$, $\mathbf{W}(t) \in \mathbb{R}^m$
- M positive definite, $\sigma > 0$

- Applications:

- Laser tweezers (thermal noise)
- Electric circuits (Nyquist noise)
- Noisy controlled systems



Stationary density

$$p^*(\mathbf{x}) \propto e^{-\frac{1}{2}\mathbf{x} \cdot \mathbf{C}^{-1}\mathbf{x}}$$

$$D = M\mathbf{C} + \mathbf{C}M^T$$

Stationary current

$$\mathbf{J}^*(\mathbf{x}) = H\mathbf{x}p^*(\mathbf{x})$$

$$H = \frac{D}{2}\mathbf{C}^{-1} - M$$

Linear diffusions

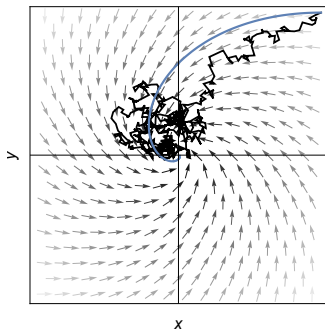
- Dynamics:

$$d\mathbf{X}(t) = -M\mathbf{X}(t)dt + \sigma d\mathbf{W}(t)$$

- $\mathbf{X}(t) \in \mathbb{R}^n$, $\mathbf{W}(t) \in \mathbb{R}^m$
- M positive definite, $\sigma > 0$

- Applications:

- Laser tweezers (thermal noise)
- Electric circuits (Nyquist noise)
- Noisy controlled systems



Stationary density

$$p^*(\mathbf{x}) \propto e^{-\frac{1}{2}\mathbf{x} \cdot \mathbf{C}^{-1}\mathbf{x}}$$

$$D = M\mathbf{C} + \mathbf{C}M^T$$

Stationary current

$$\mathbf{J}^*(\mathbf{x}) = H\mathbf{x}p^*(\mathbf{x})$$

$$H = \frac{D}{2}\mathbf{C}^{-1} - M$$

Example 1: Gradient diffusions

- Conservative drift:

$$\mathbf{F}(\mathbf{x}) = -\nabla U(\mathbf{x})$$

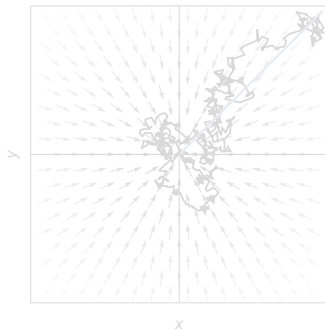
- Noise matrix: $\sigma = \epsilon I$
- 2D example:

$$M = \begin{pmatrix} \gamma & 0 \\ 0 & \gamma \end{pmatrix}$$

- Density:

$$p^*(\mathbf{x}) = \frac{\gamma}{\pi \epsilon^2} \exp\left(-\frac{\gamma}{\epsilon^2} \|\mathbf{x}\|^2\right)$$

- Current: $\mathbf{J}^* = 0$



Equilibrium or reversible system

Example 1: Gradient diffusions

- Conservative drift:

$$\mathbf{F}(\mathbf{x}) = -\nabla U(\mathbf{x})$$

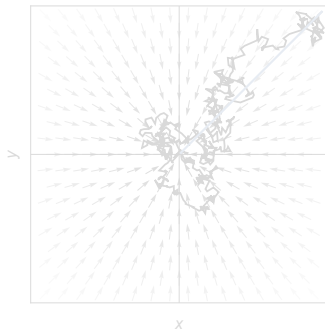
- Noise matrix: $\sigma = \epsilon I$
- 2D example:

$$M = \begin{pmatrix} \gamma & 0 \\ 0 & \gamma \end{pmatrix}$$

- Density:

$$p^*(\mathbf{x}) = \frac{\gamma}{\pi \epsilon^2} \exp\left(-\frac{\gamma}{\epsilon^2} \|\mathbf{x}\|^2\right)$$

- Current: $\mathbf{J}^* = 0$



Equilibrium or reversible system

Example 1: Gradient diffusions

- Conservative drift:

$$\mathbf{F}(\mathbf{x}) = -\nabla U(\mathbf{x})$$

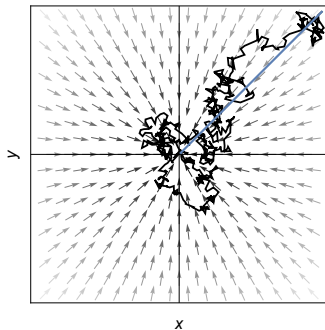
- Noise matrix: $\sigma = \epsilon I$
- 2D example:

$$M = \begin{pmatrix} \gamma & 0 \\ 0 & \gamma \end{pmatrix}$$

- Density:

$$p^*(\mathbf{x}) = \frac{\gamma}{\pi\epsilon^2} \exp\left(-\frac{\gamma}{\epsilon^2} \|\mathbf{x}\|^2\right)$$

- Current: $\mathbf{J}^* = 0$



Equilibrium or reversible system

Example 1: Gradient diffusions

- Conservative drift:

$$\mathbf{F}(\mathbf{x}) = -\nabla U(\mathbf{x})$$

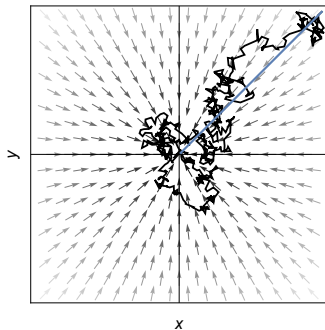
- Noise matrix: $\sigma = \epsilon I$
- 2D example:

$$M = \begin{pmatrix} \gamma & 0 \\ 0 & \gamma \end{pmatrix}$$

- Density:

$$p^*(\mathbf{x}) = \frac{\gamma}{\pi\epsilon^2} \exp\left(-\frac{\gamma}{\epsilon^2} \|\mathbf{x}\|^2\right)$$

- Current: $\mathbf{J}^* = 0$



Equilibrium or reversible system

Example 2: Transverse diffusions

- Drift:

$$\mathbf{F} = -\nabla U + \mathbf{A}, \quad \mathbf{A} \cdot \nabla U = 0$$

- 2D example:

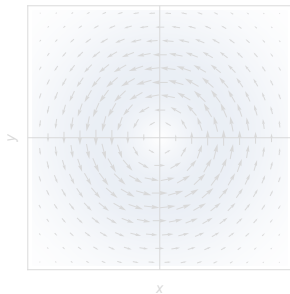
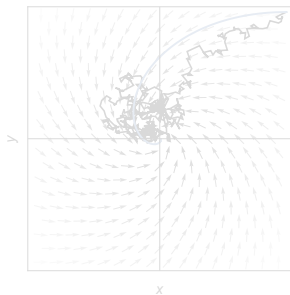
$$M = \begin{pmatrix} \gamma & \xi \\ -\xi & \gamma \end{pmatrix}$$

- Density:

$$p^*(\mathbf{x}) = \frac{\gamma}{\pi \epsilon^2} \exp\left(-\frac{\gamma}{\epsilon^2} \|\mathbf{x}\|^2\right)$$

- Current:

$$\mathbf{J}^*(\mathbf{x}) = \xi \begin{pmatrix} -y \\ x \end{pmatrix} p^*(\mathbf{x})$$



Nonequilibrium or nonreversible system

Example 2: Transverse diffusions

- Drift:

$$\mathbf{F} = -\nabla U + \mathbf{A}, \quad \mathbf{A} \cdot \nabla U = 0$$

- 2D example:

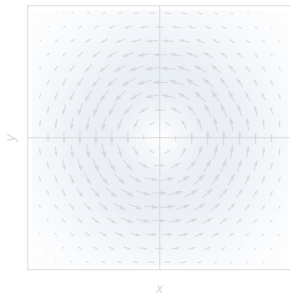
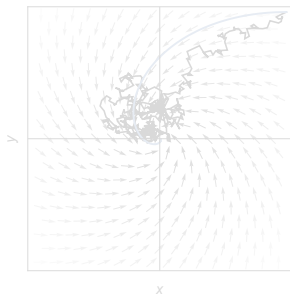
$$M = \begin{pmatrix} \gamma & \xi \\ -\xi & \gamma \end{pmatrix}$$

- Density:

$$p^*(\mathbf{x}) = \frac{\gamma}{\pi \epsilon^2} \exp\left(-\frac{\gamma}{\epsilon^2} \|\mathbf{x}\|^2\right)$$

- Current:

$$\mathbf{J}^*(\mathbf{x}) = \xi \begin{pmatrix} -y \\ x \end{pmatrix} p^*(\mathbf{x})$$



Nonequilibrium or nonreversible system

Example 2: Transverse diffusions

- Drift:

$$\mathbf{F} = -\nabla U + \mathbf{A}, \quad \mathbf{A} \cdot \nabla U = 0$$

- 2D example:

$$M = \begin{pmatrix} \gamma & \xi \\ -\xi & \gamma \end{pmatrix}$$

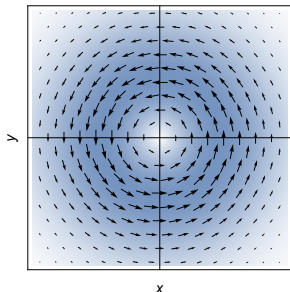
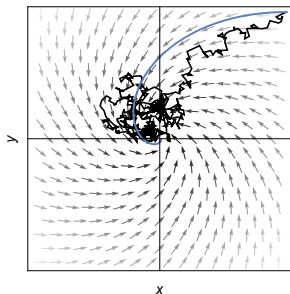
- Density:

$$p^*(\mathbf{x}) = \frac{\gamma}{\pi \epsilon^2} \exp\left(-\frac{\gamma}{\epsilon^2} \|\mathbf{x}\|^2\right)$$

- Current:

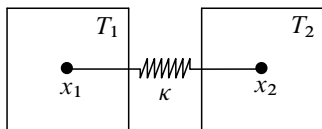
$$\mathbf{J}^*(\mathbf{x}) = \xi \begin{pmatrix} -y \\ x \end{pmatrix} p^*(\mathbf{x})$$

Nonequilibrium or nonreversible system



Example 3: Brownian gyrotator

[Exartier & Peliti 1999; Filiger & Reimann 2007]



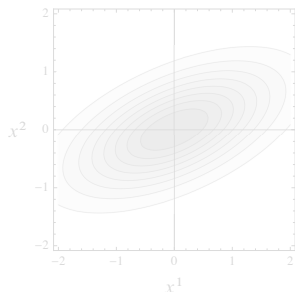
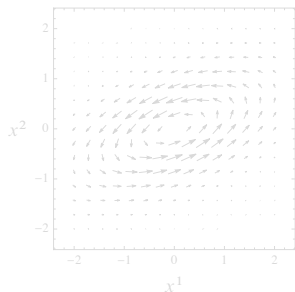
- Drift:

$$M = \begin{pmatrix} \gamma + \kappa & -\kappa \\ -\kappa & \gamma + \kappa \end{pmatrix}$$

- Noise matrix:

$$\sigma = \begin{pmatrix} \sqrt{2T_1} & 0 \\ 0 & \sqrt{2T_2} \end{pmatrix}$$

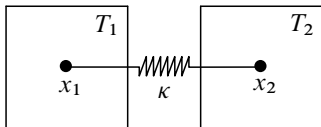
- $J^* \neq 0$ iff $\kappa > 0$ and $T_1 \neq T_2$



Nonequilibrium or nonreversible system

Example 3: Brownian gyrotor

[Exartier & Peliti 1999; Filiger & Reimann 2007]



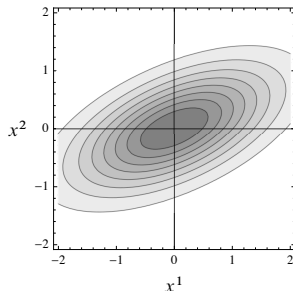
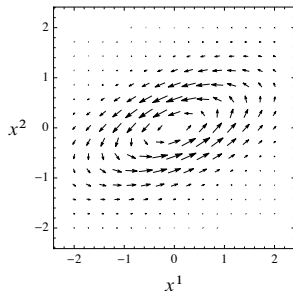
- Drift:

$$M = \begin{pmatrix} \gamma + \kappa & -\kappa \\ -\kappa & \gamma + \kappa \end{pmatrix}$$

- Noise matrix:

$$\sigma = \begin{pmatrix} \sqrt{2T_1} & 0 \\ 0 & \sqrt{2T_2} \end{pmatrix}$$

- $\mathbf{J}^* \neq 0$ iff $\kappa > 0$ and $T_1 \neq T_2$



Nonequilibrium or nonreversible system

Observables

- Linear forms:

$$A_T = \frac{1}{T} \int_0^T \boldsymbol{\eta} \cdot \mathbf{X}(t) dt$$

- Quadratic forms:

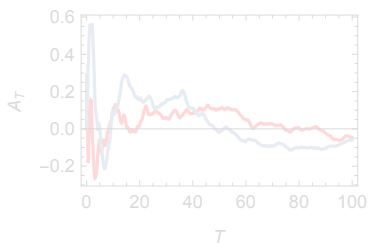
$$A_T = \frac{1}{T} \int_0^T \mathbf{X}(t) \cdot \mathbf{Q} \mathbf{X}(t) dt$$

- Current-type forms:

$$A_T = \frac{1}{T} \int_0^T \boldsymbol{\Gamma} \mathbf{X}(t) \circ d\mathbf{X}(t)$$

Large deviation approximation

$$P(A_T = a) \approx e^{-T I(a)}$$



Examples

- Mechanical work W_T
- Heat exchanged Q_T
- Entropy prod Σ_T
- Integrated currents J_T
- Residence times
- Control costs, reward
- Statistical estimators

Observables

- Linear forms:

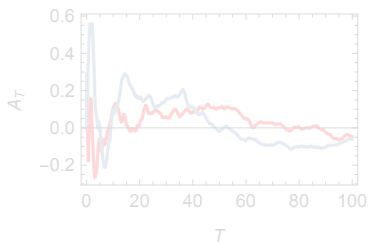
$$A_T = \frac{1}{T} \int_0^T \boldsymbol{\eta} \cdot \mathbf{X}(t) dt$$

- Quadratic forms:

$$A_T = \frac{1}{T} \int_0^T \mathbf{X}(t) \cdot \mathbf{Q} \mathbf{X}(t) dt$$

- Current-type forms:

$$A_T = \frac{1}{T} \int_0^T \boldsymbol{\Gamma} \mathbf{X}(t) \circ d\mathbf{X}(t)$$



Examples

- Mechanical work W_T
- Heat exchanged Q_T
- Entropy prod Σ_T
- Integrated currents J_T
- Residence times
- Control costs, reward
- Statistical estimators

Large deviation approximation

$$P(A_T = a) \approx e^{-T I(a)}$$

Observables

- Linear forms:

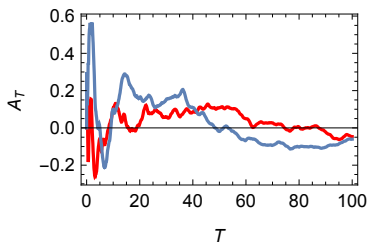
$$A_T = \frac{1}{T} \int_0^T \boldsymbol{\eta} \cdot \mathbf{X}(t) dt$$

- Quadratic forms:

$$A_T = \frac{1}{T} \int_0^T \mathbf{X}(t) \cdot \mathbf{Q} \mathbf{X}(t) dt$$

- Current-type forms:

$$A_T = \frac{1}{T} \int_0^T \boldsymbol{\Gamma} \mathbf{X}(t) \circ d\mathbf{X}(t)$$



Examples

- Mechanical work W_T
- Heat exchanged Q_T
- Entropy prod Σ_T
- Integrated currents J_T
- Residence times
- Control costs, reward
- Statistical estimators

Large deviation approximation

$$P(A_T = a) \approx e^{-T I(a)}$$

Observables

- Linear forms:

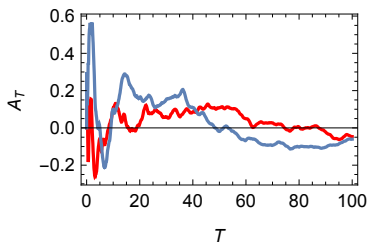
$$A_T = \frac{1}{T} \int_0^T \boldsymbol{\eta} \cdot \mathbf{X}(t) dt$$

- Quadratic forms:

$$A_T = \frac{1}{T} \int_0^T \mathbf{X}(t) \cdot \mathbf{Q} \mathbf{X}(t) dt$$

- Current-type forms:

$$A_T = \frac{1}{T} \int_0^T \boldsymbol{\Gamma} \mathbf{X}(t) \circ d\mathbf{X}(t)$$



Examples

- Mechanical work W_T
- Heat exchanged Q_T
- Entropy prod Σ_T
- Integrated currents J_T
- Residence times
- Control costs, reward
- Statistical estimators

Large deviation approximation

$$P(A_T = a) \approx e^{-T I(a)}$$

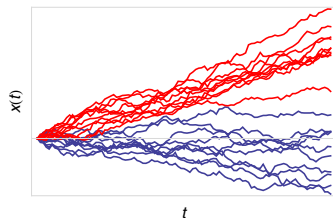
LD functions

- Rate function:

$$I(a) = \max_{k \in \mathbb{R}} \{ka - \lambda(k)\}$$

- SCGF:

$$\lambda(k) = \text{dom eigval}(\mathcal{L}_k)$$

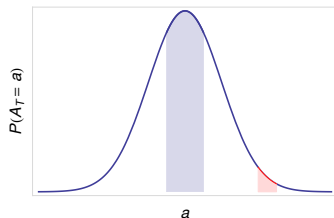


Fluctuation process

$$d\tilde{\mathbf{X}}_t = \tilde{\mathbf{F}}_k(\tilde{\mathbf{X}}_t)dt + \sigma dW_t$$

- Modified drift:

$$\tilde{\mathbf{F}}_k(x) = \mathbf{F}(x) + D\nabla \ln r_k(x), \quad I'(a) = k$$



- Effective process creating fluctuation
- Effective density and current: ρ_k^* , \mathbf{J}_k^*

[Chetrite & HT 2013, 2015]

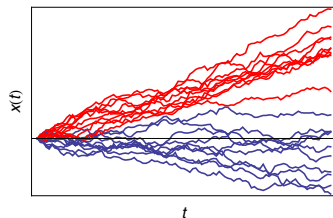
LD functions

- Rate function:

$$I(a) = \max_{k \in \mathbb{R}} \{ka - \lambda(k)\}$$

- SCGF:

$$\lambda(k) = \text{dom eigval}(\mathcal{L}_k)$$

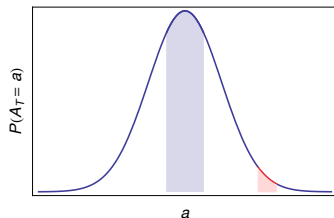


Fluctuation process

$$d\tilde{\mathbf{X}}_t = \tilde{\mathbf{F}}_k(\tilde{\mathbf{X}}_t)dt + \sigma d\mathbf{W}_t$$

- Modified drift:

$$\tilde{\mathbf{F}}_k(\mathbf{x}) = \mathbf{F}(\mathbf{x}) + D\nabla \ln r_k(\mathbf{x}), \quad I'(a) = k$$



- Effective process creating fluctuation
- Effective density and current: p_k^* , \mathbf{J}_k^*

[Chetrite & HT 2013, 2015]

Main result

[du Buisson PhD 2023; du Buisson & HT 2023]

- Generating function:

$$G_k(\mathbf{x}, t) = E[e^{k^t A_t} | \mathbf{X}(0) = \mathbf{x}]$$

- Feynman-Kac equation:

$$\partial_t G_k(\mathbf{x}, t) = \mathcal{L}_k G_k(\mathbf{x}, t), \quad G(\mathbf{x}, 0) = 1$$

- Solution:

$$G_k(\mathbf{x}, t) = e^{\mathbf{x} \cdot B_k(t) \mathbf{x}} e^{\int_0^t \text{Tr}[D B_k(s)] ds} \underset{t \rightarrow \infty}{\sim} e^{\mathbf{x} \cdot B_k^* \mathbf{x}} e^{t \text{Tr}(D B_k^*)}$$

LD solution

- SCGF: $\lambda(k) = \text{Tr}(D B_k^*)$
- Riccati matrix: B_k^*
- Eigenfunction: $r_k(\mathbf{x}) = e^{\mathbf{x} \cdot B_k^* \mathbf{x}}$
- Effective drift: $\mathbf{F}_k(\mathbf{x}) = -M_k \mathbf{x}$

Main result

[du Buisson PhD 2023; du Buisson & HT 2023]

- Generating function:

$$G_k(\mathbf{x}, t) = E[e^{k^t A_t} | \mathbf{X}(0) = \mathbf{x}]$$

- Feynman-Kac equation:

$$\partial_t G_k(\mathbf{x}, t) = \mathcal{L}_k G_k(\mathbf{x}, t), \quad G(\mathbf{x}, 0) = 1$$

- Solution:

$$G_k(\mathbf{x}, t) = e^{\mathbf{x} \cdot B_k(t) \mathbf{x}} e^{\int_0^t \text{Tr}[D B_k(s)] ds} \underset{t \rightarrow \infty}{\sim} e^{\mathbf{x} \cdot B_k^* \mathbf{x}} e^{t \text{Tr}(D B_k^*)}$$

LD solution

- SCGF: $\lambda(k) = \text{Tr}(D B_k^*)$
- Riccati matrix: B_k^*
- Eigenfunction: $r_k(\mathbf{x}) = e^{\mathbf{x} \cdot B_k^* \mathbf{x}}$
- Effective drift: $F_k(\mathbf{x}) = -M_k \mathbf{x}$

Main result

[du Buisson PhD 2023; du Buisson & HT 2023]

- Generating function:

$$G_k(\mathbf{x}, t) = E[e^{k^t A_t} | \mathbf{X}(0) = \mathbf{x}]$$

- Feynman-Kac equation:

$$\partial_t G_k(\mathbf{x}, t) = \mathcal{L}_k G_k(\mathbf{x}, t), \quad G(\mathbf{x}, 0) = 1$$

- Solution:

$$G_k(\mathbf{x}, t) = e^{\mathbf{x} \cdot B_k(t) \mathbf{x}} e^{\int_0^t \text{Tr}[D B_k(s)] ds} \underset{t \rightarrow \infty}{\sim} e^{\mathbf{x} \cdot B_k^* \mathbf{x}} e^{t \text{Tr}(D B_k^*)}$$

LD solution

- SCGF: $\lambda(k) = \text{Tr}(D B_k^*)$
- Riccati matrix: B_k^*
- Eigenfunction: $r_k(\mathbf{x}) = e^{\mathbf{x} \cdot B_k^* \mathbf{x}}$
- Effective drift: $\mathbf{F}_k(\mathbf{x}) = -M_k \mathbf{x}$

Current observables

$$A_T = \frac{1}{T} \int_0^T \Gamma \mathbf{X}(t) \circ d\mathbf{X}(t)$$

- Tilted generator:

$$\mathcal{L}_k = -kM\mathbf{x} \cdot \Gamma\mathbf{x} + (-M + kD\Gamma)\mathbf{x} \cdot \nabla + \frac{1}{2}\nabla \cdot D\nabla + \frac{k^2}{2}\Gamma\mathbf{x} \cdot D\Gamma\mathbf{x}$$

- Algebraic Riccati equation:

$$\frac{k^2}{2}\Gamma^T D\Gamma - \frac{k}{2}(M^T\Gamma - \Gamma M) - (M - kD\Gamma)^T B_k^* - B_k^*(M - kD\Gamma) + 2B_k^* D B_k^* = 0$$

$$C + AB + BA + BQB = 0$$

Effective process

$$M_k = M - 2DB_k^* - kD\Gamma$$

$$p_k^*(\mathbf{x}) \propto e^{-\frac{1}{2}\mathbf{x} \cdot C_k^{-1}\mathbf{x}}, \quad D = M_k C_k + C_k M_k^T$$

$$J_k^*(\mathbf{x}) = H_k \mathbf{x} p_k^*(\mathbf{x}), \quad H_k = \frac{D}{2} C_k^{-1} - M_k$$

Current observables

$$A_T = \frac{1}{T} \int_0^T \Gamma \mathbf{X}(t) \circ d\mathbf{X}(t)$$

- Tilted generator:

$$\mathcal{L}_k = -kM\mathbf{x} \cdot \Gamma\mathbf{x} + (-M + kD\Gamma)\mathbf{x} \cdot \nabla + \frac{1}{2}\nabla \cdot D\nabla + \frac{k^2}{2}\Gamma\mathbf{x} \cdot D\Gamma\mathbf{x}$$

- Algebraic Riccati equation:

$$\frac{k^2}{2}\Gamma^T D\Gamma - \frac{k}{2}(M^T\Gamma - \Gamma M) - (M - kD\Gamma)^T B_k^* - B_k^*(M - kD\Gamma) + 2B_k^* D B_k^* = 0$$

$$C + AB + BA + BQB = 0$$

Effective process

$$M_k = M - 2DB_k^* - kD\Gamma$$

$$p_k^*(\mathbf{x}) \propto e^{-\frac{1}{2}\mathbf{x} \cdot C_k^{-1}\mathbf{x}}, \quad D = M_k C_k + C_k M_k^T$$

$$J_k^*(\mathbf{x}) = H_k \mathbf{x} p_k^*(\mathbf{x}), \quad H_k = \frac{D}{2} C_k^{-1} - M_k$$

Current observables

$$A_T = \frac{1}{T} \int_0^T \Gamma \mathbf{X}(t) \circ d\mathbf{X}(t)$$

- Tilted generator:

$$\mathcal{L}_k = -kM\mathbf{x} \cdot \Gamma\mathbf{x} + (-M + kD\Gamma)\mathbf{x} \cdot \nabla + \frac{1}{2}\nabla \cdot D\nabla + \frac{k^2}{2}\Gamma\mathbf{x} \cdot D\Gamma\mathbf{x}$$

- Algebraic Riccati equation:

$$\frac{k^2}{2}\Gamma^T D\Gamma - \frac{k}{2}(M^T \Gamma - \Gamma M) - (M - kD\Gamma)^T B_k^* - B_k^*(M - kD\Gamma) + 2B_k^* D B_k^* = 0$$

$$C + AB + BA + BQB = 0$$

Effective process

$$M_k = M - 2DB_k^* - kD\Gamma$$

$$p_k^*(\mathbf{x}) \propto e^{-\frac{1}{2}\mathbf{x} \cdot C_k^{-1} \mathbf{x}}, \quad D = M_k C_k + C_k M_k^T$$

$$J_k^*(\mathbf{x}) = H_k \mathbf{x} p_k^*(\mathbf{x}), \quad H_k = \frac{D}{2} C_k^{-1} - M_k$$

Current observables

$$A_T = \frac{1}{T} \int_0^T \Gamma \mathbf{X}(t) \circ d\mathbf{X}(t)$$

- Tilted generator:

$$\mathcal{L}_k = -kM\mathbf{x} \cdot \Gamma\mathbf{x} + (-M + kD\Gamma)\mathbf{x} \cdot \nabla + \frac{1}{2}\nabla \cdot D\nabla + \frac{k^2}{2}\Gamma\mathbf{x} \cdot D\Gamma\mathbf{x}$$

- Algebraic Riccati equation:

$$\frac{k^2}{2}\Gamma^T D\Gamma - \frac{k}{2}(M^T \Gamma - \Gamma M) - (M - kD\Gamma)^T B_k^* - B_k^*(M - kD\Gamma) + 2B_k^* D B_k^* = 0$$

$$C + AB + BA + BQB = 0$$

Effective process

$$M_k = M - 2DB_k^* - kD\Gamma$$

$$p_k^*(\mathbf{x}) \propto e^{-\frac{1}{2}\mathbf{x} \cdot \mathbf{C}_k^{-1} \mathbf{x}}, \quad D = M_k \mathbf{C}_k + \mathbf{C}_k M_k^T$$

$$\mathbf{J}_k^*(\mathbf{x}) = H_k \mathbf{x} p_k^*(\mathbf{x}), \quad H_k = \frac{D}{2} \mathbf{C}_k^{-1} - M_k$$

Example: Stochastic area

[Lévy 1940, 1950, 1951; Teitsworth et al 2017, 2019]

- Process in \mathbb{R}^2 : $((X_t, Y_t))_{t=0}^T$
- Green's theorem:

$$\mathcal{A}_T = \frac{1}{2} \int_0^T X_t dY_t - Y_t dX_t$$

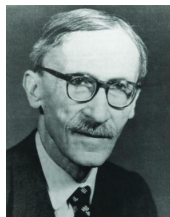
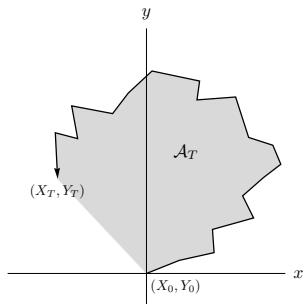
- Brownian motion:

$$p(\mathcal{A}_T = a) = \frac{1}{T} \operatorname{sech} \left(\frac{\pi a}{T} \right), \quad \operatorname{var}(\mathcal{A}_T) = \frac{T^2}{4}$$

- Linear diffusions:

$$a^* = \lim_{T \rightarrow \infty} \left\langle \frac{\mathcal{A}_T}{T} \right\rangle = \left(MC - \frac{D}{2} \right)_{1,2}$$

$\neq 0$ nonreversible systems



Paul Lévy (1886–1971)

Example: Stochastic area

[Lévy 1940, 1950, 1951; Teitsworth et al 2017, 2019]

- Process in \mathbb{R}^2 : $((X_t, Y_t))_{t=0}^T$
- Green's theorem:

$$\mathcal{A}_T = \frac{1}{2} \int_0^T X_t dY_t - Y_t dX_t$$

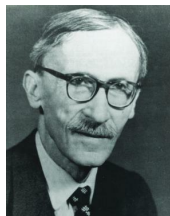
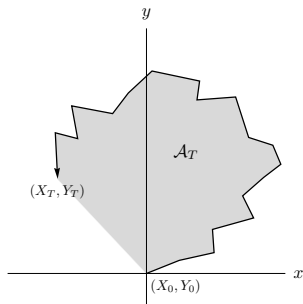
- Brownian motion:

$$p(\mathcal{A}_T = a) = \frac{1}{T} \operatorname{sech} \left(\frac{\pi a}{T} \right), \quad \operatorname{var}(\mathcal{A}_T) = \frac{T^2}{4}$$

- Linear diffusions:

$$a^* = \lim_{T \rightarrow \infty} \left\langle \frac{\mathcal{A}_T}{T} \right\rangle = \left(MC - \frac{D}{2} \right)_{1,2}$$

$\neq 0$ nonreversible systems



Paul Lévy (1886-1971)

Example: Stochastic area

[Lévy 1940, 1950, 1951; Teitsworth et al 2017, 2019]

- Process in \mathbb{R}^2 : $((X_t, Y_t))_{t=0}^T$
- Green's theorem:

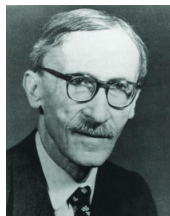
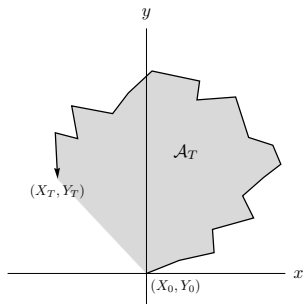
$$\mathcal{A}_T = \frac{1}{2} \int_0^T X_t dY_t - Y_t dX_t$$

- Brownian motion:

$$p(\mathcal{A}_T = a) = \frac{1}{T} \operatorname{sech} \left(\frac{\pi a}{T} \right), \quad \operatorname{var}(\mathcal{A}_T) = \frac{T^2}{4}$$

- Linear diffusions:

$$a^* = \lim_{T \rightarrow \infty} \left\langle \frac{\mathcal{A}_T}{T} \right\rangle = \left(MC - \frac{D}{2} \right)_{1,2} \\ \neq 0 \text{ nonreversible systems}$$



Paul Lévy (1886–1971)

$$M = \begin{pmatrix} \gamma & \xi \\ -\xi & \gamma \end{pmatrix} \quad A_T = \frac{1}{T} \int_0^T \Gamma \mathbf{X}(t) \circ d\mathbf{X}(t) \quad \Gamma = \frac{1}{2} \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

- Typical area: $a^* = \frac{\epsilon^2 \xi}{2\gamma}$
- Riccati matrix: $B_k^* = b_k^* I$
- SCGF:

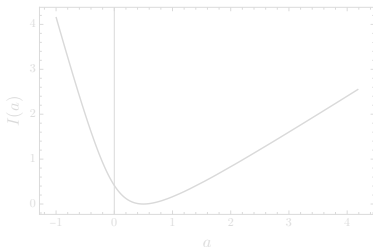
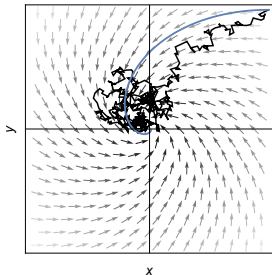
$$\lambda(k) = \gamma - \sqrt{\gamma^2 - \frac{k\epsilon^2(k\epsilon^2 + 4\xi)}{4}}$$

- Rate function:

$$I(a) = \sqrt{\frac{\epsilon^4(\gamma^2 + \xi^2)}{4a^2 + \epsilon^4}} - \frac{2a\xi}{\epsilon^2} - \gamma + \frac{4|a|^3(\gamma^2 + \xi^2)}{\sqrt{a^2\epsilon^4(4a^2 + \epsilon^4)(\gamma^2 + \xi^2)}}$$

- Fluctuation relation:

$$I(-a) = I(a) + \frac{4\xi a}{\epsilon^2}$$



$$M = \begin{pmatrix} \gamma & \xi \\ -\xi & \gamma \end{pmatrix} \quad A_T = \frac{1}{T} \int_0^T \Gamma \mathbf{X}(t) \circ d\mathbf{X}(t) \quad \Gamma = \frac{1}{2} \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

- Typical area: $a^* = \frac{\epsilon^2 \xi}{2\gamma}$
- Riccati matrix: $B_k^* = b_k^* I$
- SCGF:

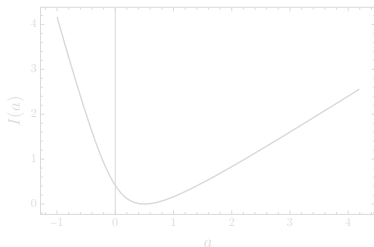
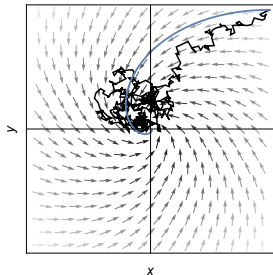
$$\lambda(k) = \gamma - \sqrt{\gamma^2 - \frac{k\epsilon^2(k\epsilon^2 + 4\xi)}{4}}$$

- Rate function:

$$I(a) = \sqrt{\frac{\epsilon^4(\gamma^2 + \xi^2)}{4a^2 + \epsilon^4}} - \frac{2a\xi}{\epsilon^2} - \gamma + \frac{4|a|^3(\gamma^2 + \xi^2)}{\sqrt{a^2\epsilon^4(4a^2 + \epsilon^4)(\gamma^2 + \xi^2)}}$$

- Fluctuation relation:

$$I(-a) = I(a) + \frac{4\xi a}{\epsilon^2}$$



$$M = \begin{pmatrix} \gamma & \xi \\ -\xi & \gamma \end{pmatrix} \quad A_T = \frac{1}{T} \int_0^T \Gamma \mathbf{X}(t) \circ d\mathbf{X}(t) \quad \Gamma = \frac{1}{2} \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

- Typical area: $a^* = \frac{\epsilon^2 \xi}{2\gamma}$
- Riccati matrix: $B_k^* = b_k^* I$
- SCGF:

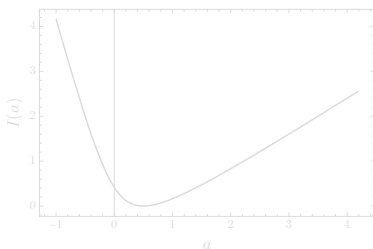
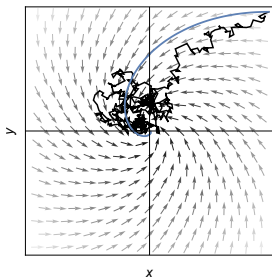
$$\lambda(k) = \gamma - \sqrt{\gamma^2 - \frac{k\epsilon^2(k\epsilon^2 + 4\xi)}{4}}$$

- Rate function:

$$I(a) = \sqrt{\frac{\epsilon^4(\gamma^2 + \xi^2)}{4a^2 + \epsilon^4}} - \frac{2a\xi}{\epsilon^2} - \gamma + \frac{4|a|^3(\gamma^2 + \xi^2)}{\sqrt{a^2\epsilon^4(4a^2 + \epsilon^4)(\gamma^2 + \xi^2)}}$$

- Fluctuation relation:

$$I(-a) = I(a) + \frac{4\xi a}{\epsilon^2}$$



$$M = \begin{pmatrix} \gamma & \xi \\ -\xi & \gamma \end{pmatrix} \quad A_T = \frac{1}{T} \int_0^T \Gamma \mathbf{X}(t) \circ d\mathbf{X}(t) \quad \Gamma = \frac{1}{2} \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

- Typical area: $a^* = \frac{\epsilon^2 \xi}{2\gamma}$
- Riccati matrix: $B_k^* = b_k^* I$
- SCGF:

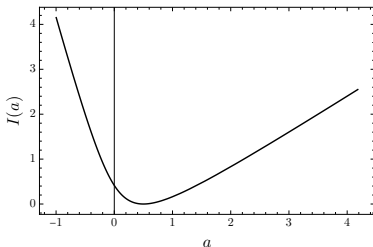
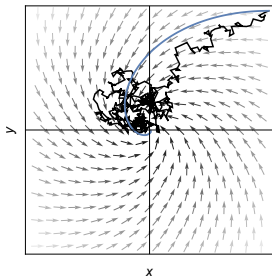
$$\lambda(k) = \gamma - \sqrt{\gamma^2 - \frac{k\epsilon^2(k\epsilon^2 + 4\xi)}{4}}$$

- Rate function:

$$I(a) = \sqrt{\frac{\epsilon^4(\gamma^2 + \xi^2)}{4a^2 + \epsilon^4}} - \frac{2a\xi}{\epsilon^2} - \gamma + \frac{4|a|^3(\gamma^2 + \xi^2)}{\sqrt{a^2\epsilon^4(4a^2 + \epsilon^4)(\gamma^2 + \xi^2)}}$$

- Fluctuation relation:

$$I(-a) = I(a) + \frac{4\xi a}{\epsilon^2}$$

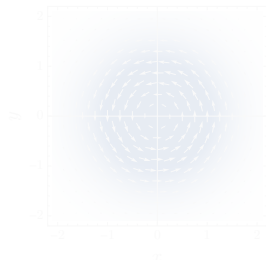
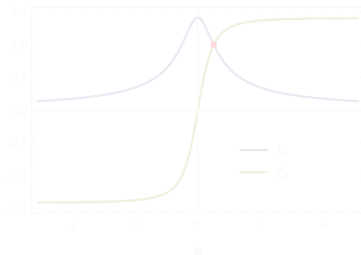


Example: Transverse diffusions (cont'd)

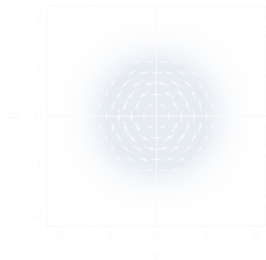
- Effective process:

$$M_k = \begin{pmatrix} \gamma_k & \xi_k \\ -\xi_k & \gamma_k \end{pmatrix} \quad \xi_k = \xi + \frac{k\epsilon^2}{2}$$

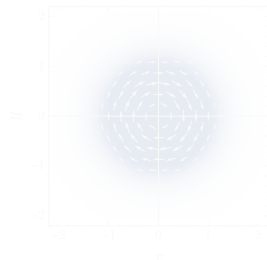
$$\gamma_k = \sqrt{\gamma^2 - \frac{k\epsilon^2(k\epsilon^2 + 4\xi)}{4}}, \quad I'(a) = k$$



$a > a^*$
Deconfinement



$0 < a < a^*$
Confinement



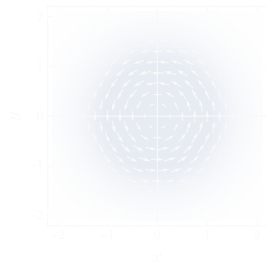
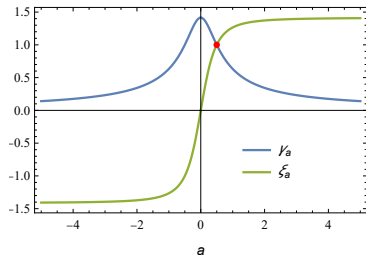
$a < 0$
Current reversal

Example: Transverse diffusions (cont'd)

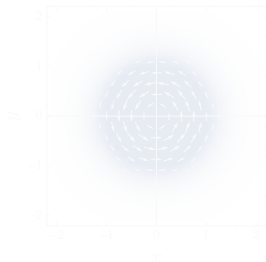
- Effective process:

$$M_k = \begin{pmatrix} \gamma_k & \xi_k \\ -\xi_k & \gamma_k \end{pmatrix} \quad \xi_k = \xi + \frac{k\epsilon^2}{2}$$

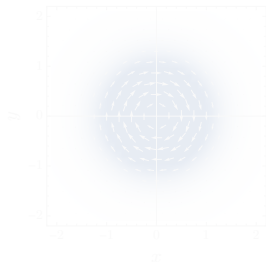
$$\gamma_k = \sqrt{\gamma^2 - \frac{k\epsilon^2(k\epsilon^2 + 4\xi)}{4}}, \quad I'(a) = k$$



$a > a^*$
Deconfinement



$0 < a < a^*$
Confinement



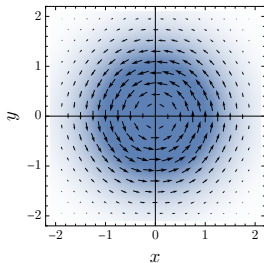
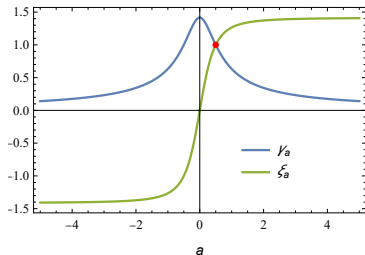
$a < 0$
Current reversal

Example: Transverse diffusions (cont'd)

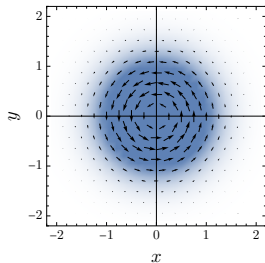
- Effective process:

$$M_k = \begin{pmatrix} \gamma_k & \xi_k \\ -\xi_k & \gamma_k \end{pmatrix} \quad \xi_k = \xi + \frac{k\epsilon^2}{2}$$

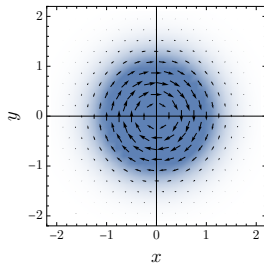
$$\gamma_k = \sqrt{\gamma^2 - \frac{k\epsilon^2(k\epsilon^2 + 4\xi)}{4}}, \quad I'(a) = k$$



$a > a^*$
Deconfinement



$0 < a < a^*$
Confinement



$a < 0$
Current reversal

Conclusions

- Similar results for other observables/processes
- Fluctuations created by effective (modified) linear process
- Underlying Gaussian density and current fluctuations
- Effective process = LQG optimal control

Current work

- Linear approximations for nonlinear SDEs/observables
- Numerical methods (Riccati or control)

- [1] Johan du Buisson
Dynamical Large Deviations of Diffusions
PhD thesis, Stellenbosch University, 2022
- [2] Johan du Buisson, HT
Dynamical large deviations of linear diffusions
Phys. Rev. E **107**, 054111, 2023
- [3] Johan du Buisson, Thamu Mnyulwa, HT
Large deviations of the stochastic area for linear diffusions
Phys. Rev. E **108**, 044136, 2023

Conclusions

- Similar results for other observables/processes
- Fluctuations created by effective (modified) linear process
- Underlying Gaussian density and current fluctuations
- Effective process = LQG optimal control

Current work

- Linear approximations for nonlinear SDEs/observables
- Numerical methods (Riccati or control)

- [1] Johan du Buisson
Dynamical Large Deviations of Diffusions
PhD thesis, Stellenbosch University, 2022
- [2] Johan du Buisson, HT
Dynamical large deviations of linear diffusions
Phys. Rev. E **107**, 054111, 2023
- [3] Johan du Buisson, Thamu Mnyulwa, HT
Large deviations of the stochastic area for linear diffusions
Phys. Rev. E **108**, 044136, 2023

Conclusions

- Similar results for other observables/processes
- Fluctuations created by effective (modified) linear process
- Underlying Gaussian density and current fluctuations
- Effective process = LQG optimal control

Current work

- Linear approximations for nonlinear SDEs/observables
- Numerical methods (Riccati or control)

- [1] Johan du Buisson
Dynamical Large Deviations of Diffusions
PhD thesis, Stellenbosch University, 2022
- [2] Johan du Buisson, HT
Dynamical large deviations of linear diffusions
Phys. Rev. E **107**, 054111, 2023
- [3] Johan du Buisson, Thamu Mnyulwa, HT
Large deviations of the stochastic area for linear diffusions
Phys. Rev. E **108**, 044136, 2023

Conclusions

- Similar results for other observables/processes
- Fluctuations created by effective (modified) linear process
- Underlying Gaussian density and current fluctuations
- Effective process = LQG optimal control

Current work

- Linear approximations for nonlinear SDEs/observables
- Numerical methods (Riccati or control)

- [1] Johan du Buisson
Dynamical Large Deviations of Diffusions
PhD thesis, Stellenbosch University, 2022
- [2] Johan du Buisson, HT
Dynamical large deviations of linear diffusions
Phys. Rev. E **107**, 054111, 2023
- [3] Johan du Buisson, Thamu Mnyulwa, HT
Large deviations of the stochastic area for linear diffusions
Phys. Rev. E **108**, 044136, 2023

Conclusions

- Similar results for other observables/processes
- Fluctuations created by effective (modified) linear process
- Underlying Gaussian density and current fluctuations
- Effective process = LQG optimal control

Current work

- Linear approximations for nonlinear SDEs/observables
- Numerical methods (Riccati or control)

- [1] Johan du Buisson
Dynamical Large Deviations of Diffusions
PhD thesis, Stellenbosch University, 2022
- [2] Johan du Buisson, HT
Dynamical large deviations of linear diffusions
Phys. Rev. E **107**, 054111, 2023
- [3] Johan du Buisson, Thamu Mnyulwa, HT
Large deviations of the stochastic area for linear diffusions
Phys. Rev. E **108**, 044136, 2023