

# Eyring-Kramers law and exit rates for the overdamped Langevin process

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## Plan

- I. The kinetic Monte Carlo method and the Eyring-Kramers law
- II. Justification of the use of the kinetic Monte Carlo method to sample the exit from a metastable domain :
  - The quasi-stationary distribution approach
  - Eyring-Kramers law for the exit rates

# Part I

The kinetic Monte Carlo method and the Eyring-Kramers law

## Overdamped Langevin process

**Molecular dynamics** : simulate the evolution of atomistic systems (e.g. macromolecules or clusters of atoms)

→ compute macroscopic quantities or understand macroscopic mechanisms

Evolution of the positions of the atoms : **overdamped Langevin** process

$$dX_t = -\nabla f(X_t)dt + \sqrt{h} dB_t,$$

where

- $X_t \in \mathbb{R}^d$  ( $d = 3N$ )
- $f : \mathbb{R}^d \rightarrow \mathbb{R}$  is the potential energy of the system (interactions between the atoms)
- $h = 2k_B T_e$

## Metastability of the Overdamped Langevin process

When

$$h \ll \Delta f \text{ ("}h \rightarrow 0\text{")}$$

the drift part  $-\nabla f(X_t)dt$  is dominant in

$$dX_t = -\nabla f(X_t)dt + \sqrt{h} dB_t$$

$\rightarrow (X_t)_{t \geq 0}$  is trapped over a long period of time in a neighborhood  $\Omega$  of a local minimum  $x_0$  of  $f$  (metastable region) before going into another metastable region.

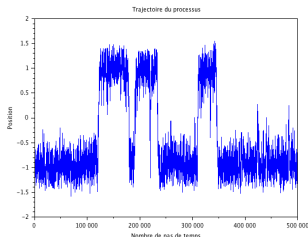
$\rightarrow$  [Metastability](#)

## Illustration of the metastability in a very simple example

$$f(x) = (x - 1)^2(x + 1)^2$$



- 2 metastable regions  $\mathbb{R}_-^*$  et  $\mathbb{R}_+^*$  (basin of attractions)
- time evolution of  $(X_t)_{t \geq 0}$  :



- Transitions between the two metastable regions = **rare events**  
How to efficiently sample the exit event from a metastable region ?

## Macroscopic evolution

- Macroscopic state = metastable region  $\Omega \subset \mathbb{R}^d$

### Macroscopic evolution

How long does  $(X_t)_{t \geq 0}$  stay in  $\Omega$  and where does it exit  $\Omega$  (next visited state)?

→ Applications in biology, medicine, material science, nuclear physics

- Macroscopic transition = the move from one metastable region to another

Not accessible by integrating the trajectories of  $(X_t)_{t \geq 0}$

→ Major issue in computational statistical physics

How to efficiently sample the exit event  $(\tau_\Omega, X_{\tau_\Omega})$  from a metastable state  $\Omega$ ?

Where

$$\tau_\Omega := \inf\{t \geq 0, X_t \notin \Omega\}$$

and  $X_{\tau_\Omega} \in \partial\Omega$  = the exit point location from  $\Omega$

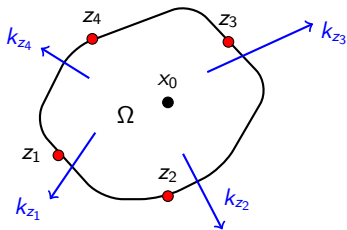
## Sample the macroscopic evolution

**Kinetic Monte Carlo method (kMC)** : generates a [jump Markov process](#)

**Ingredients** to sample the exit from the region  $\Omega$  with kMC : for all saddle points  $z_j \in \partial\Omega$  ( $j \in \{1, \dots, n\}$ ), the **exit rate**

$$k_{z_j}$$

associated with an exit from  $\Omega$  around  $z_j$ .



Domain  $\Omega$  ( $n = 4$ )



## How sample the exit event from $\Omega$ with kMC ?

First denote by :

- $T \geq 0$  the residence time in  $\Omega$
- $Y \in \{z_1, \dots, z_n\}$  the random variable such that  $Y = z_j$  if the exit from  $\Omega$  occurs around  $z_j$

With a kMC method, the exit event  $(T, Y)$  from  $\Omega$ , is sampled as follows :

1.  $T \sim \mathcal{E}\left(\sum_{\ell=1}^n k_{z_\ell}\right)$
2.  $Y$  is independent of  $T$
3.  $Y$  has the following law :

$$\mathbb{P}[Y = z_j] = \frac{k_{z_j}}{\sum_{\ell=1}^n k_{z_\ell}}$$

Remark :

$(T, Y)$  = exit event generated by the kMC method

$(\tau_\Omega, X_{\tau_\Omega})$  = exit event for  $(X_t)_{t \geq 0}$

## It remains to compute $k_{z_j}$ ...

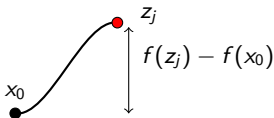
In practice,  $k_{z_j}$  is computed with the Eyring-Kramers law :

$$\boxed{k_{z_j} = A_j e^{-\frac{2}{h}(f(z_j) - f(x_0))}} \quad \text{where :}$$

- $A_j$  : prefactor (depends on the process  $(X_t)_{t \geq 0}$  and the nature of the saddle point  $z_j$ ) :

$$A_j = \frac{|\lambda(z_j)|}{\pi} \frac{\sqrt{|\det \text{Hess } f(x_0)|}}{\sqrt{|\det \text{Hess } f(z_j)|}} \quad \text{if } \nabla f(z_j) = 0$$

- $f(z_j) - f(x_0)$  : activation energy.



**Question :**  $(T, Y)$  (+  $k_{z_j}$  computed with Eyring-Kramers)  $\simeq (\tau_\Omega, X_{\tau_\Omega})$  ?

## **kMC et accelerated dynamics algorithms**

The kMC model is also used to build accelerated dynamics algorithms such as :

- *Hyperdynamics* [A.F. Voter, Phys. Rev. Lett. (1997)].
- *Temperature Accelerated Dynamics* (TAD) [M.R. Sorensen et A.F. Voter, J. Chem. Phys. (2000)].

More precisely, these 2 algorithms are based on :

$$(T, Y) (+ k_{z_j} \text{ computed with Eyring-Kramers}) \simeq (\tau_\Omega, X_{\tau_\Omega})$$

Recall, the question was : **is this approximation valid ?**

**One advantage of TAD** : does not require to know all the saddle points  $z_j$ .

[D. Perez, B. P. Uberuaga, Y. Shim, J. G. Amar, and A.F. Voter, *Accelerated Molecular Dynamics Methods : Introduction and Recent Developments*. Annual Reports in computational chemistry (2009)].

## Mathematical contributions

The raised question is linked with the following two areas in the mathematical literature :

### 1. Exit problem.

**Large deviations** : Freidlin-Wentzell and also Berglund, Day, Galves *and al.*, Gentz

Combined with PDE techniques : Devinatz, Friedman, Kamin, Perthame

### 2. Investigate the (low-lying) spectrum as $h \rightarrow 0$ of the infinitesimal generator $\frac{h}{2}\Delta - \nabla f \cdot \nabla$ of $(X_t)_{t \geq 0}$ in $L^2(\mathbb{R}^d, e^{-\frac{2}{h}f})$

- **Potential theory** :

Bovier *and al.*, Landim, Seo *and al.*

- **Spectral methods** (semi-classical analysis) :

Helfer, Hérau, Holley, Michel, Miclo, Nier, Sjöstrand

## Part II

Justification of the use of the kinetic Monte Carlo method to sample the exit from a metastable domain

## Quasi-stationary distribution (QSD)

Let  $\Omega$  be a  $C^\infty$  bounded domain of  $\mathbb{R}^d$  and  $f : \bar{\Omega} \rightarrow \mathbb{R}$  be  $C^\infty$ .

**Definition (QSD).**  $\nu_h$  is a QSD if

$$X_0 \sim \nu_h \Rightarrow \forall t \geq 0, \mathcal{L}(X_t | t < \tau_\Omega) = \nu_h$$

**Proposition.** There exists a unique QSD  $\nu_h$  and

$$\forall X_0 \in \Omega, \quad \lim_{t \rightarrow +\infty} \mathcal{L}(X_t | t < \tau_\Omega) = \nu_h$$

In other words

$$\nu_h = \text{local equilibrium of the process } (X_t)_{t \geq 0} \text{ in } \Omega$$

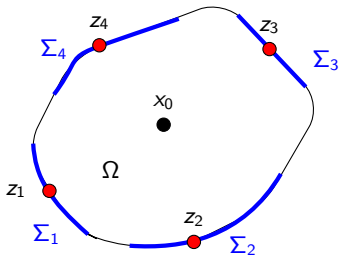
If  $\Omega$  metastable (reaching  $\nu_h \ll$  exiting  $\Omega$ ), it is relevant to study the exit event when

$$X_0 \sim \nu_h$$

## Definition of the exit rates in a metastable region $\Omega$

Exit rate associated with an exit through a neighborhood  $\Sigma_j \subset \partial\Omega$  of  $z_j$  :

$$k_{z_j}^L = \frac{\mathbb{P}_{\nu_h}[X_{\tau_\Omega} \in \Sigma_j]}{\mathbb{E}_{\nu_h}[\tau_\Omega]}$$



**Proposition.** When  $X_0 \sim \nu_h$ ,

$$\tau_\Omega \sim \mathcal{E}\left(\sum_{\ell=1}^n k_{z_\ell}^L\right), \quad \tau_\Omega \perp\!\!\!\perp X_{\tau_\Omega}, \quad \text{and} \quad \mathbb{P}_{\nu_h}[X_{\tau_\Omega} \in \Sigma_j] = k_{z_j}^L / \sum_{\ell=1}^n k_{z_\ell}^L$$

→ We can simulate  $(\tau_\Omega, X_{\tau_\Omega})$  with a **jump Markov process** (kMC)

**Eyring-Kramers law** for the exit rates  $k_{z_j}^L$  ?

## Eyring-Kramers law : case when $\partial_n f > 0$

Assumptions :

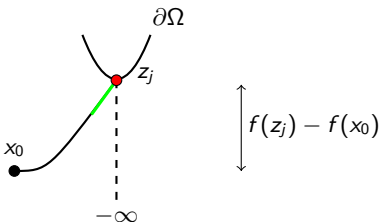
- $\partial_n f > 0$  on  $\partial\Omega$
- $f$  has unique critical point  $x_0$  in  $\Omega$ , and

$$f(x_0) = \min_{\bar{\Omega}} f < \min_{\partial\Omega} f$$

- $f : \bar{\Omega} \rightarrow \mathbb{R}$  and  $f|_{\partial\Omega}$  are Morse functions

Denote by  $z_1, \dots, z_n$  the **local minima** of  $f|_{\partial\Omega}$

Remark :  $\nabla f(z_j) \neq 0$  and  $z_1, \dots, z_n$  are the geometric saddle points of  $f$  on  $\partial\Omega$ , if we extend  $f$  by  $-\infty$  outside  $\bar{\Omega}$  (absorbing boundary).



Remark : such edge shaped barriers were considered by Kramers (1940), Matkowsky and Schuss (1982)



## Eyring-Kramers law when $\partial_n f > 0$

**Theorem 1.** *If each  $z_j$  is sufficiently far from the other  $z_k$ 's (for an Agmon distance), then, for all  $j \in \{1, \dots, n\}$ , when  $h \rightarrow 0$  :*

$$k_{z_j}^L = \frac{\partial_n f(z_j)}{\sqrt{h\pi}} \frac{\sqrt{\det \text{Hess } f(x_0)}}{\sqrt{\det \text{Hess } f|_{\partial\Omega}(z_j)}} e^{-\frac{2}{h}(f(z_j)-f(x_0))} (1 + O(h))$$

→ This is the expected Eyring-Kramers law for such a  $z_j$  ( $\nabla f(z_j) \neq 0$ )

[*Sharp asymptotics of the first exit point density.* Annals of PDE (2019). G. Di Gesù, T. Lelièvre, D. Le Peutrec, and B.N.]

[*Jump Markov models and transition state theory : the quasi-stationary distribution approach.* Faraday Discussions (2016). G. Di Gesù, T. Lelièvre, D. Le Peutrec, and B.N.]

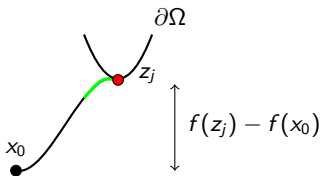
**And if  $f$  has saddle points  $z_j \in \partial\Omega$  with  $\nabla f(z_j) = 0$ ?**

**Generalization :  $f$  has saddle points  $z_j \in \partial\Omega$  with  $\nabla f(z_j) = 0$**

In many applications of the kMC methods and the accelerated dynamics algorithms :  $\Omega =$  the basin of attraction of a local minimum  $x_0$  of  $f$  in  $\mathbb{R}^d$  for the dynamics

$$\dot{x}(t) = -\nabla f(x(t)), t \geq 0$$

→  $f$  has critical points on  $\partial\Omega$  :



→ Practical interest in considering such a geometrical setting

**Eyring-Kramers law : case when  $f$  has saddle points  $z_j \in \partial\Omega$  with**  
$$\nabla f(z_j) = 0$$

Assume that :

- $\Omega$  is the basin of attraction of a local minimum  $x_0$  of  $f$  in  $\mathbb{R}^d$  for the dynamics

$$\dot{x}(t) = -\nabla f(x(t)), t \geq 0$$

Denote by

$$z_1, \dots, z_n$$

the saddle points of  $f$  on  $\partial\Omega$  ( $\nabla f(z_j) = 0$ ).

- $f : \bar{\Omega} \rightarrow \mathbb{R}$  and  $f|_{\partial\Omega}$  are Morse functions

Remark : we smooth  $\partial\Omega$  (note that  $\partial\Omega$  is already smooth around the  $z_j$ 's)

**Eyring-Kramers law when  $f$  has saddle points  $z_j \in \partial\Omega$  with  $\nabla f(z_j) = 0$**

**Theorem 2.** *If each  $z_j$  is sufficiently far from the other  $z_k$ 's (for another Agmon distance), then, for all  $j \in \{1, \dots, n\}$ , when  $h \rightarrow 0$  :*

$$k_{z_j}^L = \frac{|\lambda(z_j)|}{\pi} \frac{\sqrt{|\det \text{Hess } f(x_0)|}}{\sqrt{|\det \text{Hess } f(z_j)|}} e^{-\frac{2}{h}(f(z_j) - f(x_0))} (1 + O(h))$$

→ This is the expected Eyring-Kramers law for such a saddle point  $z_j$  ( $\nabla f(z_j) = 0$ )

[Eyring-Kramers exit rates for the overdamped Langevin dynamics : the case with saddle points on the boundary. Submitted. 2022. T. Lelièvre, D. Le Peutrec, and B.N.]

## Conclusion

When :

*a.*  $\partial_{\mathbf{n}} f > 0$

or,

*b.*  $f$  has saddle points  $z_j \in \partial\Omega$  with  $\nabla f(z_j) = 0$

we have justified that

$$(\tau_{\Omega}, X_{\tau_{\Omega}})$$

can be sampled with the **kMC method** parametrized by exit rates computed with the **Eyring-Kramers law**.

## Extension

The kinetic Langevin process :

$$\begin{cases} dx_t = v_t dt, \\ dv_t = -\nabla f(x_t) dt - \gamma(x_t, v_t) v_t dt + \Sigma(x_t, v_t) dB_t, \end{cases} \quad (1)$$

where

$$\Omega = \mathcal{O} \times \mathbb{R}^d \text{ and } \mathcal{O} \subset \mathbb{R}^d$$

Existence and uniqueness of the QSD for such hypoelliptic processes :

- [A. Guillin, B.N., and L. Wu, 2020, submitted]
- [T. Lelièvre, J. Reygner, and M. Ramil, 2021, Stoch. Process. their Appl.]
- [A. Guillin, B.N., and L. Wu, 2022, Probab. Theory Relat. Fields]

Thank you