

# *Effective dynamics using conditional expectations*

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# Classical $n$ -body system

Atomistic system whose position is given by  $X = (X^1, \dots, X^n) \in \mathbb{R}^n$

**Potential energy:**  $V(X)$ , which contains all the physics of the system

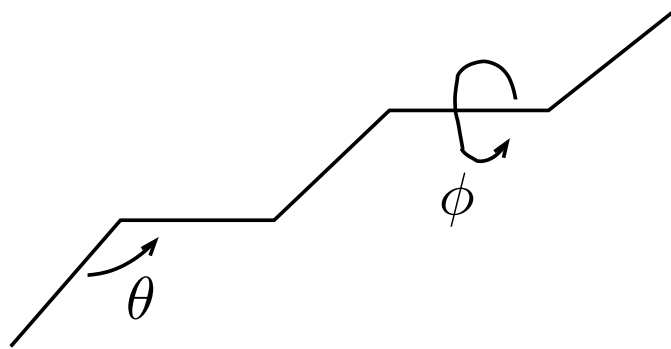
A standard task in molecular dynamics is to compute **thermodynamical averages** wrt Gibbs measure:

$$\langle \phi \rangle = \int_{\mathbb{R}^n} \phi(X) d\mu, \quad d\mu = Z^{-1} \exp(-\beta V(X)) dX$$

with  $\beta = 1/(k_B T)$  and  $Z = \int_{\mathbb{R}^n} \exp(-\beta V(X)) dX$ .

## A simple example

- Chain-like molecules (biological molecules, polymers, ...):



$$V(X) = \sum_i V_2 (X^{i+1} - X^i) + \sum_i V_3(\theta_i) + \sum_i V_4(\cos \phi_i)$$

with

$$V_2 (X^{i+1} - X^i) = \frac{k_2}{2} (|X^{i+1} - X^i| - d_{eq})^2, \quad V_3(\theta_i) = \frac{k_3}{2} (\theta_i - \theta_{eq})^2$$

- end-to-end distance  $\langle |X^n - X^1| \rangle$

Other examples: pressure in a liquid, ...

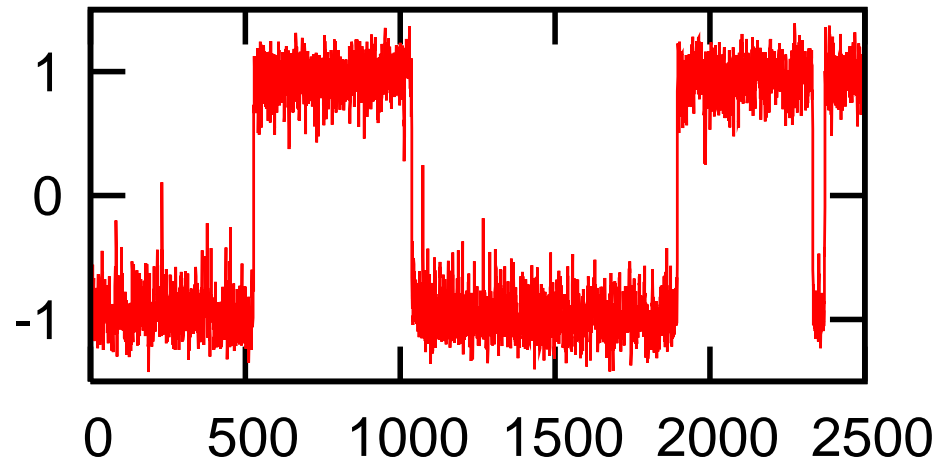
Consider the **dynamics** (overdamped Langevin equation)

$$dX_t = -\nabla V(X_t) dt + \sqrt{2\beta^{-1}} dW_t \quad \text{in } \mathbb{R}^n,$$

which is ergodic for  $d\mu$ : for a.a. initial conditions  $X_0$ , and a.s.,

$$\forall \text{ smooth } \phi, \quad \frac{1}{T} \int_0^T \phi(X_t) dt = \int \phi(X) d\mu + O\left(\frac{1}{\sqrt{T}}\right).$$

In practice, the constant in the  $O(\cdot)$  may be big (**metastability**). Reduced description of the system, that still includes some dynamical information?



$$dX_t = -\nabla V(X_t) dt + \sqrt{2\beta^{-1}} dW_t \quad \text{in } \mathbb{R}^n,$$

Given a reaction coordinate  $\xi : \mathbb{R}^n \mapsto \mathbb{R}$ , propose a dynamics  $z_t$  that approximates  $\xi(X_t)$ .

- preservation of equilibrium properties:

when  $X$  is distributed according to  $d\mu$ , then  $\xi(X)$  distributed according to  $\exp(-\beta A(z)) dz$ :

$$\forall \text{ smooth } \phi_r, \quad \int_{\mathbb{R}^n} \phi_r(\xi(X)) d\mu = \int_{\mathbb{R}} \phi_r(z) \exp(-\beta A(z)) dz$$

The dynamics  $z_t$  should be ergodic wrt  $\exp(-\beta A(z)) dz$ .

- estimation of the distance between  $z_t$  and  $\xi(X_t)$ .

## Some related works

- Mori-Zwanzig projection formalism
- Normal modes decomposition
- system reduction in the framework on SDEs (E, Vanden-Eijnden, ...):

$$\begin{aligned} dx_t &= f(x_t, y_t) dt + \sqrt{2\beta^{-1}} dB_t^1 \\ dy_t &= \frac{1}{\varepsilon} g(x_t, y_t) dt + \sqrt{\frac{2\beta^{-1}}{\varepsilon}} dB_t^2 \end{aligned}$$

If  $x_t$  is held fixed, then  $y_t$  is ergodic w.r.t. a measure independent of  $\varepsilon$ .

Our framework:

- cartesian coordinates not adapted to slow/fast splitting.
- even if  $\xi(x, y) = x$ , the fast dynamics is different:

$$\begin{aligned} dx_t &= f(x_t, y_t) dt + \sqrt{2\beta^{-1}} dB_t^1 \\ dy_t &= \frac{1}{\varepsilon} g(x_t, y_t) dt + \sqrt{2\beta^{-1}} dB_t^2 \end{aligned}$$

## Convergence to equilibrium: Fokker-Planck equation

$$dX_t = -\nabla V(X_t) dt + \sqrt{2\beta^{-1}} dW_t$$

The pdf  $\psi(t, X)$  of  $X_t$  satisfies the **Fokker-Planck equation**:

$$\partial_t \psi = \operatorname{div} (\psi \nabla V) + \beta^{-1} \Delta \psi$$

Introduce the **equilibrium density**

$$\psi_\infty(X) = Z^{-1} \exp(-\beta V(X)), \quad d\mu = \psi_\infty(X) dX$$

We have  $\nabla \psi_\infty = -\beta \psi_\infty \nabla V$ . Hence the Fokker-Planck equation reads

$$\partial_t \psi = \beta^{-1} \operatorname{div} \left[ \psi_\infty \nabla \left( \frac{\psi}{\psi_\infty} \right) \right]$$

We wish to prove that  $\psi(t, \cdot) \rightarrow \psi_\infty(\cdot)$  as  $t \rightarrow \infty$ . Relative entropy argument.

## Convergence to equilibrium: relative entropy

Introduce  $e(t) = H(\psi|\psi_\infty) := \int_{\mathbb{R}^n} \ln \left( \frac{\psi(t, x)}{\psi_\infty(x)} \right) \psi(t, x) dx$ . Then

$$\begin{aligned} \frac{de}{dt} &= \int \ln \left( \frac{\psi}{\psi_\infty} \right) \partial_t \psi + \frac{\psi_\infty}{\psi} \frac{\partial_t \psi}{\psi_\infty} \psi \\ &= \int \ln \left( \frac{\psi}{\psi_\infty} \right) \beta^{-1} \mathbf{div} \left[ \psi_\infty \nabla \left( \frac{\psi}{\psi_\infty} \right) \right] \\ &= -\beta^{-1} \int \nabla \left[ \ln \left( \frac{\psi}{\psi_\infty} \right) \right] \psi_\infty \nabla \left( \frac{\psi}{\psi_\infty} \right) \\ &= -\beta^{-1} \int \left| \nabla \left[ \ln \left( \frac{\psi}{\psi_\infty} \right) \right] \right|^2 \psi =: -\beta^{-1} I(\psi|\psi_\infty) \end{aligned}$$

Assume that  $\psi_\infty$  satisfies a **Log-Sobolev Inequality (LSI)**: there exists  $\rho > 0$  such that, for any probability measure of density  $\nu$ ,

$$H(\nu|\psi_\infty) \leq (2\rho)^{-1} I(\nu|\psi_\infty)$$

Then  $\frac{de}{dt} \leq -2\rho\beta^{-1}e(t)$ , hence  $e(t) \rightarrow 0$ .



- Construction of an effective dynamics
- Assessment of its quality in terms of marginals
- Pathwise convergence in a specific case
- Numerical examples

From the dynamics on  $X_t$ , we obtain

$$\begin{aligned}d\xi(X_t) &= \nabla\xi(X_t) \cdot dX_t + \beta^{-1} \Delta\xi(X_t) dt \\ &= (-\nabla V \cdot \nabla\xi + \beta^{-1} \Delta\xi)(X_t) dt + \sqrt{2\beta^{-1}} \nabla\xi(X_t) \cdot dW_t \\ &= (-\nabla V \cdot \nabla\xi + \beta^{-1} \Delta\xi)(X_t) dt + \sqrt{2\beta^{-1}} |\nabla\xi|(X_t) dB_t\end{aligned}$$

where  $dB_t = \frac{\nabla\xi(X_t) \cdot dW_t}{|\nabla\xi|(X_t)}$ :  $B_t$  is a 1D standard noise.

$$\begin{aligned}\tilde{b}(t, z) &:= \mathbb{E} [(-\nabla V \cdot \nabla\xi + \beta^{-1} \Delta\xi)(X_t) \mid \xi(X_t) = z] \\ \tilde{\sigma}^2(t, z) &:= \mathbb{E} [|\nabla\xi|^2(X_t) \mid \xi(X_t) = z]\end{aligned}$$

and consider

$$d\tilde{z}_t = \tilde{b}(t, \tilde{z}_t) dt + \sqrt{2\beta^{-1}} \tilde{\sigma}(t, \tilde{z}_t) dB_t, \quad \tilde{z}_0 = \xi(X_0)$$

Then, for any  $t$ , the law of  $\tilde{z}_t$  is equal to the law of  $\xi(X_t)$  (Gyongy 1986)

**But**  $\tilde{b}(t, z)$  and  $\tilde{\sigma}(t, z)$  are **extremely difficult** to compute ...

Need for **approximation**:

$$\begin{aligned}\tilde{b}(t, z) &= \mathbb{E} [(-\nabla V \cdot \nabla \xi + \beta^{-1} \Delta \xi) (X_t) \mid \xi(X_t) = z] \\ b(z) &:= \mathbb{E}_\mu [(-\nabla V \cdot \nabla \xi + \beta^{-1} \Delta \xi) (X) \mid \xi(X) = z]\end{aligned}$$

Idea:  $\tilde{b}(t, z) \approx b(z)$  if, in the manifold

$$\Sigma_z = \{X \in \mathbb{R}^n, \quad \xi(X) = z\},$$

$X_t$  **quickly** reaches equilibrium.  $\xi(X_t)$  **much slower** than evolution of  $X_t$  in  $\Sigma_z$ .

Consider the **effective dynamics**

$$dz_t = b(z_t) dt + \sqrt{2\beta^{-1}} \sigma(z_t) dB_t, \quad z_0 = \xi(X_0)$$

with  $\sigma^2(z) := \mathbb{E}_\mu [|\nabla \xi|^2(X) \mid \xi(X) = z]$ .

OK from the **statistical viewpoint**: this dynamics is ergodic wrt  $\exp(-\beta A(z)) dz$ .

- In the general case, obtain estimates between the law of  $z_t$  and the law of  $\xi(X_t)$ , for any  $t$ , in terms of relative entropy.
- the potential  $V$  is frequently of the form

$$V_\varepsilon(X) = V_0(X) + \frac{1}{\varepsilon} q^2(X)$$

- rough dependence of the above estimate in  $\varepsilon$ ?
- exploit this structure to properly define  $\xi(X)$ ?
- consider a more restrictive case to obtain stronger results (pathwise convergence):

$$\mathbb{E} \left[ \sup_{t \in (0, T)} |\xi(X_t) - z_t|^2 \right].$$

## An entropy convergence result

$$dX_t = -\nabla V(X_t) dt + \sqrt{2\beta^{-1}} dW_t, \quad \text{pdf } \psi(t, X)$$

The pdf of  $\xi(X_t)$  is

$$\bar{\psi}(t, z) = \int_{\mathbb{R}^n} \psi(t, X) \delta(\xi(X) - z)$$

On the other hand, we have introduced the effective dynamics

$$dz_t = b(z_t) dt + \sqrt{2\beta^{-1}} \sigma(z_t) dB_t, \quad z_0 = \xi(X_0)$$

Let  $\phi(t, z)$  be the pdf of  $z_t$ . Introduce the relative entropy:

$$E(t) := H(\bar{\psi}|\phi) = \int_{\mathbb{R}} \ln \left( \frac{\bar{\psi}(t, z)}{\phi(t, z)} \right) \bar{\psi}(t, z) dz.$$

We would like  $\bar{\psi} \approx \phi$ , e.g.  $E$  small . . .

## Decoupling assumptions

$$\Sigma_z = \{X \in \mathbb{R}^n, \xi(X) = z\}, \quad d\mu_z \propto \exp(-\beta V(X)) \delta(\xi(X) - z)$$

- assume that the conditioned probability measures  $d\mu_z$  satisfy a log-Sobolev inequality with a large constant  $\rho$  (uniform in  $z$ ). They can be **quickly sampled**.
- assume that the coupling between the dynamics of  $\xi(X_t)$  and the dynamics in  $\Sigma_z$  is **weak**:

$$\text{At equilibrium, } \xi \sim \exp(-\beta A(z)) dz \text{ with } A'(z) = \int_{\Sigma_z} F(X) d\mu_z$$

We assume that  $\|\nabla_{\Sigma_z} F\|_{L^\infty} \leq \kappa$ .

When  $\xi(x, y) = x$ ,  $\nabla_{\Sigma_z} F = \partial_{xy} V$ : weak-coupling between  $x$  and  $y$ .

- assume that  $|\nabla \xi|$  is **close to a constant** on each  $\Sigma_z$ , e.g.

$$\lambda = \left\| \frac{|\nabla \xi|^2(X) - \sigma^2(\xi(X))}{\sigma^2(\xi(X))} \right\|_{L^\infty} < +\infty$$

## A bounded entropy result

$$E(t) := H(\bar{\psi}|\phi) = \int_{\mathbb{R}} \ln \left( \frac{\bar{\psi}(t, z)}{\phi(t, z)} \right) \bar{\psi}(t, z) dz.$$

Under the above assumptions, for all  $t \geq 0$ ,

$$E(t) \leq \frac{\|\nabla \xi\|_{L^\infty}^2}{4} \left( \lambda + \frac{\beta^2 \kappa^2}{\rho^2} \right) H(\psi(0, \cdot) | \psi_\infty)$$

Hence, if  $\xi$  is such that

- $\rho$  is large (**strong mixing** in  $\Sigma_z$ ),
- $\kappa$  is small (**small coupling** between dynamics in  $\Sigma_z$  and on  $z_t$ ),
- $\lambda$  is small ( $|\nabla \xi|$  is **close to a constant** on each  $\Sigma_z$ ),

then the effective dynamics  $z_t$  obtained by conditional expectations is close to the real dynamics  $\xi(X_t)$  in the sense of relative entropy.

Standard expression in MD:  $V_\varepsilon(X) = V_0(X) + \frac{1}{\varepsilon}q^2(X)$

- LSI estimate for the measure  $\mu$  restricted on  $\{X; \xi(X) = z\}$ :  
This measure essentially looks like a gaussian measure of variance  $O(\varepsilon)$ .  
Such gaussian measures satisfy a LSI with  $\rho_\varepsilon = O(1/\varepsilon)$ .

- Mean-force  $F(X)$  estimation:  $A'(z) = \int_{\Sigma_z} F(X) d\mu_z$ ,  $\kappa_\varepsilon = \|\nabla_{\Sigma_z} F\|_{L^\infty}$ ?

$$\begin{aligned} F(X) &= \frac{\nabla V \cdot \nabla \xi}{|\nabla \xi|^2} - \beta^{-1} \operatorname{div} \left( \frac{\nabla \xi}{|\nabla \xi|^2} \right) \\ &= \frac{2q(X)}{\varepsilon} \frac{\nabla q \cdot \nabla \xi}{|\nabla \xi|^2} + O(1), \quad \text{assuming that } \nabla \xi = O(1). \end{aligned}$$

If  $\nabla \xi \cdot \nabla q = 0$ , then  $\kappa_\varepsilon = O(1)$ , otherwise  $\kappa_\varepsilon = O(1/\varepsilon)$ .



$$E(t) \leq \frac{\|\nabla\xi\|_{L^\infty}^2}{4} \left( \lambda + \frac{\beta^2 \kappa^2}{\rho^2} \right) H(\psi(0, \cdot) | \psi_\infty)$$

where  $\lambda$  measures how much  $|\nabla\xi|$  deviates from a constant on  $\Sigma_z$ .

$$\rho_\varepsilon = O(1/\varepsilon).$$

- If  $\nabla\xi \cdot \nabla q = 0$ , then  $\kappa_\varepsilon = O(1)$ , hence  $\kappa_\varepsilon/\rho_\varepsilon = O(\varepsilon)$ , and “good” estimate. In the absence of  $\lambda$ ,

$$E(t) = O(\varepsilon^2).$$

- If  $\nabla\xi \cdot \nabla q \neq 0$ , then  $\kappa_\varepsilon = O(1/\varepsilon)$ , hence  $\kappa_\varepsilon/\rho_\varepsilon = O(1)$ , and

$$E(t) = O(1),$$

hence the laws of  $\xi(X_t)$  and of  $z_t$  are not close one to each other.

Can we get stronger results?



Sufficient conditions for the two dynamics to be identical:

$$\nabla\xi(X) \cdot \nabla q(X) = 0 \text{ for all } X \in \mathbb{R}^2$$

Recall  $V_\varepsilon(X) = V_0(X) + \frac{1}{\varepsilon}q^2(X)$ , hence the limit dynamics lives on

$$\mathcal{M} = \{X \in \mathbb{R}^2; q(X) = 0\}.$$

We ask that the parametrization  $\xi$  of  $\mathcal{M}$  is orthogonal to the direction  $\nabla q$  where sampling is fast.

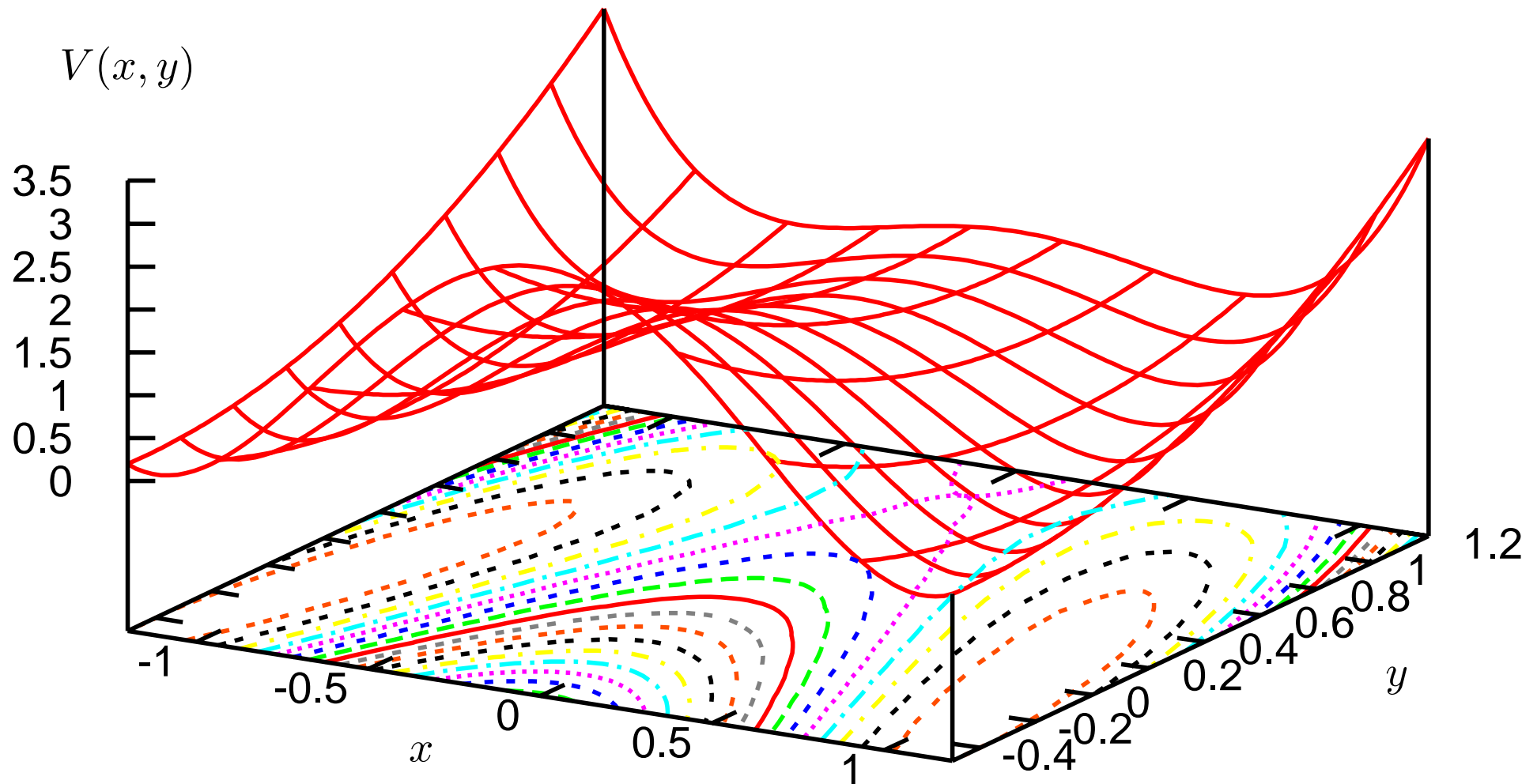
Hence, if the orthogonality condition holds:

- our effective dynamics is a good approximation of the true dynamics  $\xi(X_t)$ , on finite time intervals ( $X \in \mathbb{R}^2$ ).
- good entropy estimate ( $X \in \mathbb{R}^n$ ).

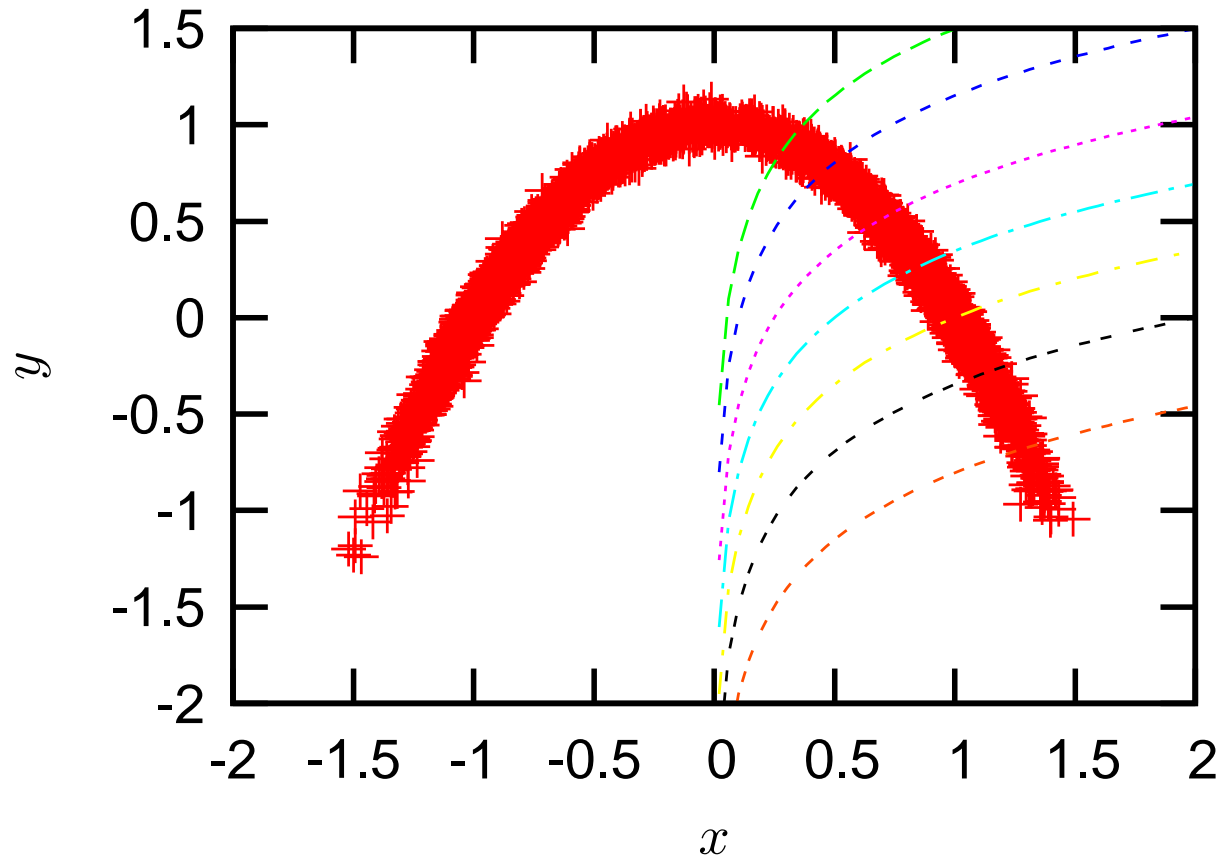
## A toy example

$$V(x, y) = (x^2 - 1)^2 + (y + x^2 - 1)^2$$

Two wells  $V(\pm 1, 0) = 0$ , one saddle point  $V(0, 1) = 1$ .



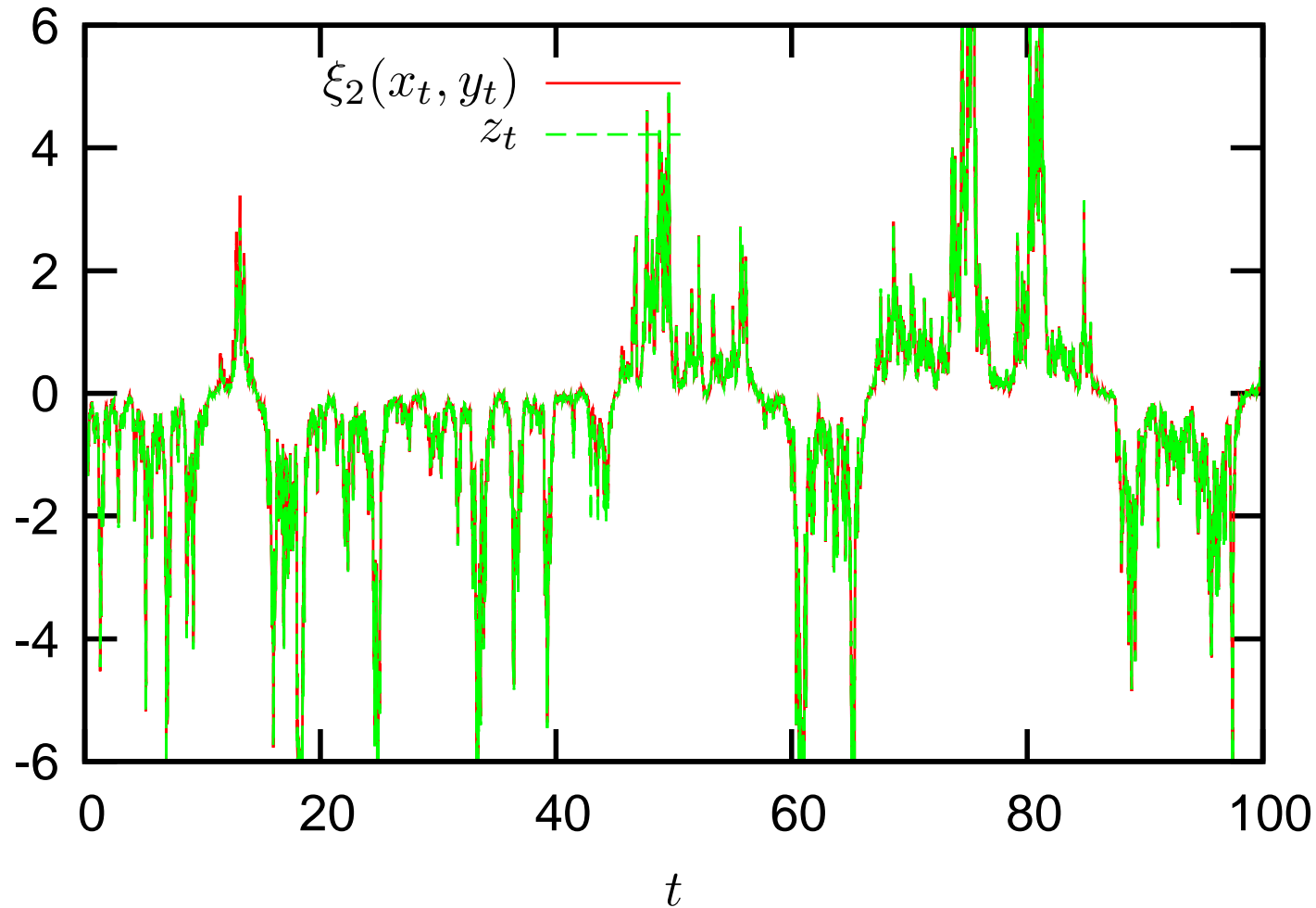
A numerical 2D example:  $\varepsilon = 0.01$ ,  $\beta = 3$



$$V_\varepsilon(x, y) = (x^2 - 1)^2 + \frac{1}{\varepsilon}(x^2 + y - 1)^2$$

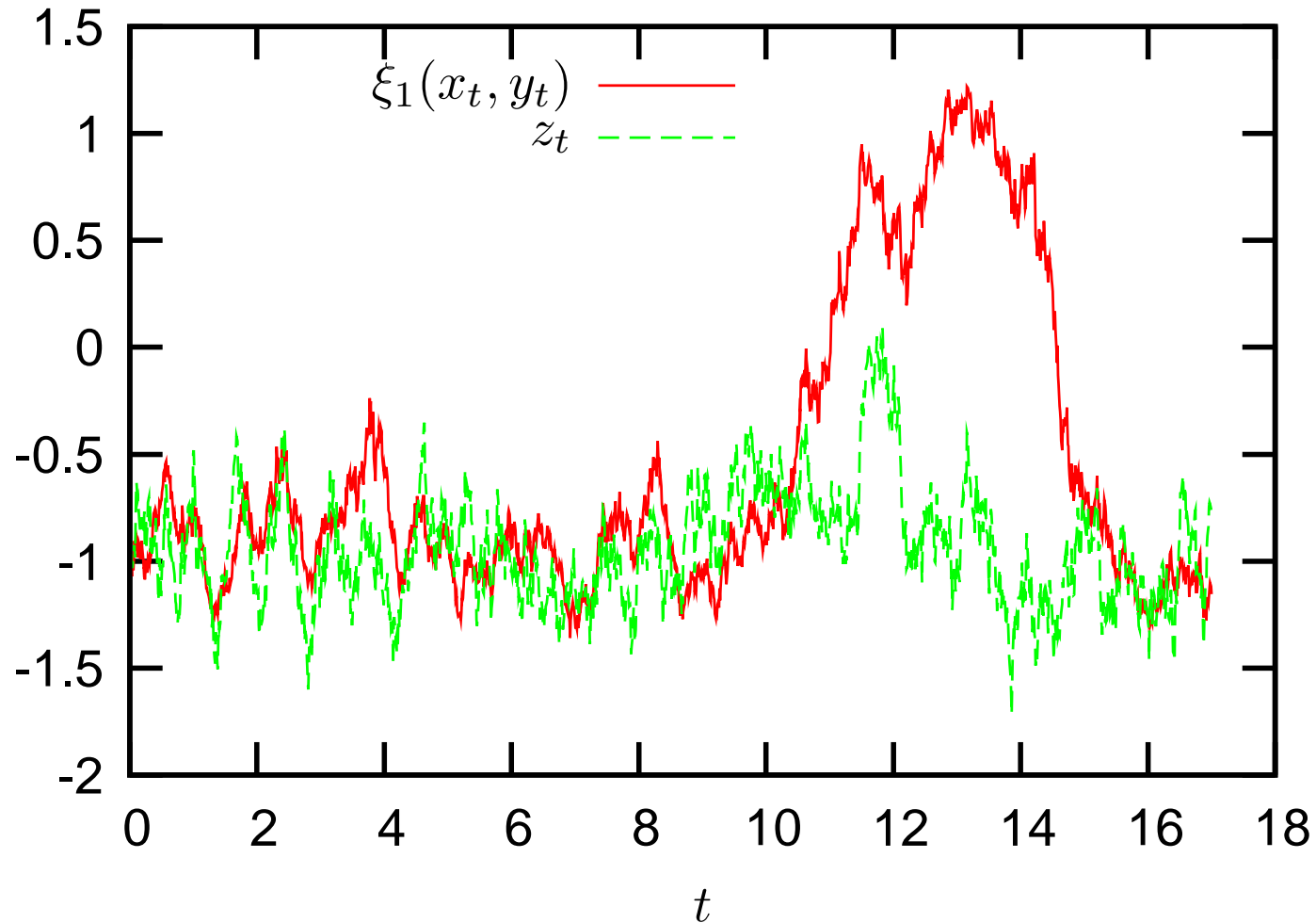
Two possible reaction coordinates:  $\xi_1(x, y) = x$ ,  $\xi_2(x, y) = x \exp(-2y)$

# Effective dynamics with the second reaction coordinate



$$q(x, y) = x^2 + y - 1, \quad \xi_2(x, y) = x \exp(-2y) : \quad \nabla \xi_2 \cdot \nabla q = 0$$

# Effective dynamics with the first reaction coordinate



$$q(x, y) = x^2 + y - 1, \quad \xi_1(x, y) = x : \quad \nabla \xi_1 \cdot \nabla q \neq 0$$

Free energy barriers are often associated with residence times in the wells.

Is our effective dynamics the same as

$$du_t = -A'(u_t) dt + \sqrt{2\beta^{-1}} dB_t,$$

e.g. the dynamics driven by the free energy  $A$  associated with the reaction coordinate  $\xi(X)$ ?

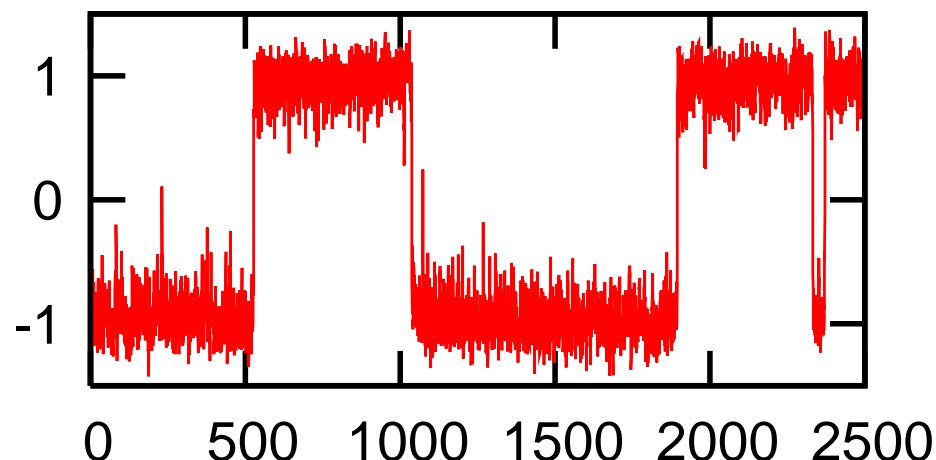
$$\text{Recall } \sigma^2(z) = \mathbb{E}_\mu [|\nabla\xi|^2(X) \mid \xi(X) = z]$$

$$\text{and } dz_t = b(z_t) dt + \sqrt{2\beta^{-1}} \sigma(z_t) dB_t$$

- If  $|\nabla\xi(X)| = 1$ , then  $\sigma(z) = 1$ , and  $b(z) = -A'(z)$ . We recover the free energy. An example:  $\xi(X) = X^j$ .
- In general,  $\sigma(z)$  is not a constant, and  $b(z) \neq -A'(z)$ , even up to a multiplicative constant.



## Approximation of residence times ( $\varepsilon = 0.01$ , $\beta = 3$ )



Reac. Coord.	Full dyn.	Reduced dynamics
$\xi_2(x, y) = x \exp(-2y)$	$32.5 \pm 0.5$	$32.7 \pm 0.5$ (Eff. dyn.)
$\xi_2(x, y) = x \exp(-2y)$	$32.5 \pm 0.5$	$6.4 \pm 0.3$ (based on $A(\xi_2)$ )
$\xi_1(x, y) = x$	$31.6 \pm 0.5$	$24.4 \pm 0.4$

If  $\xi_1(x, y) = x$ , effective dynamics  $\equiv$  dynamics driven by free energy.

It seems that residence times are well approximated by the reduced dynamics when the “good” reaction coordinate and the “good” dynamics are used.

- We have proposed a “natural” way to obtain a closed equation on  $\xi(X_t)$ , by **conditional expectations** (natural extension of the Gyongy result).
- In some cases, we are able to estimate the **accuracy** of this effective equation.
- Encouraging **numerical results** on some simple cases.

F. Legoll, T. Lelièvre, *Effective dynamics using conditional expectations*,  
Arxiv preprint 0906.4865