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Optimizing the diffusion for equilibrium and nonequilibrium dynamics

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ANR SINEQ

Diffusion dependent Overdamped Langevin dynamics

- **Aim:** Estimation of $\mathbb{E}_\pi[f] = \int_{\mathcal{X}} f(q)\pi(q)\mathrm{d}q$, $\pi \propto e^{-\beta V}$
with the estimator

$$\hat{I}_N := \frac{1}{N} \sum_{i=1}^N f(q^i), \quad q^i \sim \pi$$

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- **Challenge:** **Find optimal diffusion coefficient \mathcal{D} to accelerate convergence**

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Quantify convergence rate

- Related to the spectral gap of the dynamics' generator $\mathcal{L}_{\mathcal{D}}$:

$$\mathcal{L}_{\mathcal{D}}\varphi = (-\mathcal{D}\nabla V + \beta^{-1} \operatorname{div} \mathcal{D}) \cdot \nabla \varphi + \beta^{-1} \mathcal{D} : \nabla^2 \varphi$$

Then for any initial distribution π_0 , the law π_t of the process q_t satisfies²

$$\left\| \frac{\pi_t}{\pi} - 1 \right\|_{L^2(\pi)} \leq e^{-\Lambda(\mathcal{D})\beta^{-1}t} \left\| \frac{\pi_0}{\pi} - 1 \right\|_{L^2(\pi)}$$

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 - **Examples**: Approach mainly used in Bayesian Inference³: $\mathcal{D} \equiv (\nabla^2 V)^{-1}$
- Other works⁴ suggest $\mathcal{D} \propto e^{\beta V} \mathbf{I}_d$

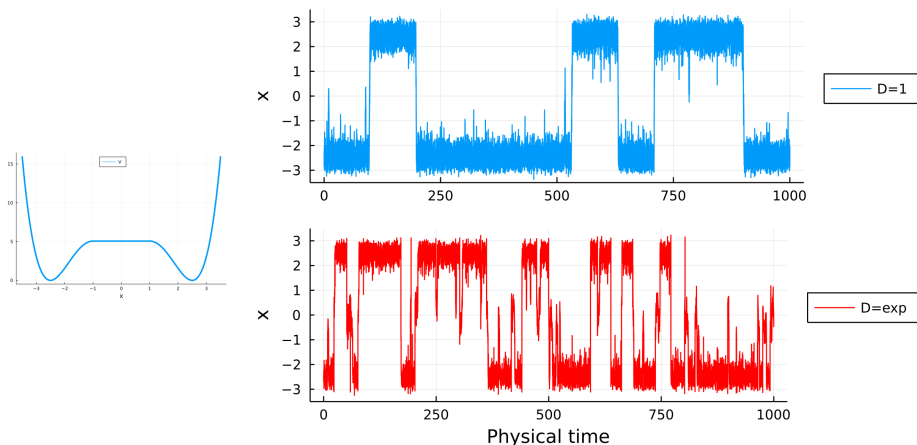
²Lelièvre/Nier/Pavliotis (2013)

³Girolami/Calderhead (2011)

⁴Roberts/Stramer (2002), Lelièvre/Pavliotis/Robin/Santet/Stoltz (In prep.)

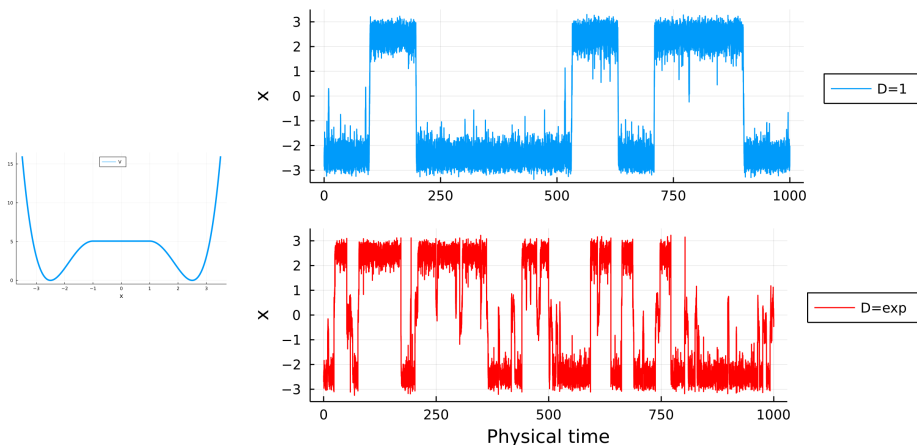
Which diffusion coefficient? Metastability case

- Example with $\mathcal{D}_{\text{One}} = 1$ and $\mathcal{D}_{\text{exp}} = e^{\beta V} \mathbf{1}_{[-10,10]} + \mathbf{1}_{\mathbb{R} \setminus [-10,10]}$ (both normalized in $L^1(\pi)$)



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- \mathcal{D}_{exp} helps to cross energy barriers (if $V \uparrow$, then $D \uparrow$)

Formulation of the optimization problem

- Using $\mathcal{L}_{\mathcal{D}} = -\beta^{-1} \nabla^* \mathcal{D} \nabla$ on $L^2(\pi)$, the **spectral gap** of $-\beta \mathcal{L}_{\mathcal{D}}$ is

$$\Lambda(\mathcal{D}) = \min_{u \in H^1(\mathbb{T}^d) \setminus \{0\}} \left\{ \frac{\int_{\mathbb{T}^d} \nabla u^\top \mathcal{D} \nabla u \, d\pi}{\int_{\mathbb{T}^d} u^2 \, d\pi} \mid \int_{\mathbb{T}^d} u \, d\pi = 0 \right\}$$

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- L^p constraint on \mathcal{D} :** $\mathcal{D} \in L^p_\pi(\mathbb{T}^d, \mathcal{M}_{a,b})$ for $1 \leq p \leq +\infty$, $a, b \geq 0$ if

$$e^{-\beta V(q)} \mathcal{D}(q) \in \mathcal{M}_{a,b} = \left\{ M \in \mathcal{S}_d^+ \mid \forall \xi \in \mathbb{R}^d, a |\xi|^2 \leq \xi^\top M \xi \leq b^{-1} |\xi|^2 \right\} \text{ a.e.}$$

endowed with

$$\|\mathcal{D}\|_{L^p_\pi} = \left(\int_{\mathbb{T}^d} |\mathcal{D}(q)|_F^p e^{-\beta p V(q)} \, dq \right)^{1/p}$$

$$\mathfrak{D}_p^{a,b} = \left\{ \mathcal{D} \in L^\infty_\pi(\mathbb{T}^d, \mathcal{M}_{a,b}) \mid \|\mathcal{D}\|_{L^p_\pi} \leq 1 \right\}$$

Theoretical analysis of the optimization problem

- $V \in \mathcal{C}^\infty(\mathbb{T}^d)$, V and π bounded on \mathbb{T}^d
- π satisfies a Poincaré inequality
- $\mathfrak{D}_p^{a,b}$ weakly closed for L_π^p
- $\mathcal{D} \mapsto \Lambda(\mathcal{D})$ concave

Theorem [Existence of a maximizer]

For any $p \in [1, +\infty)$, there exists

$$\mathcal{D}_p^\star = \arg \max_{\mathcal{D} \in \mathfrak{D}_p^{a,b}} \Lambda(\mathcal{D})$$

The maximizer is such that

- $\|\mathcal{D}\|_{L_\pi^p} = 1$;
- For any open set $\Omega \subset \mathbb{T}^d$, there exists $q \in \Omega$ such that $\mathcal{D}_p^\star(q) \neq 0$

Euler-Lagrange equation

If $\Lambda(\mathcal{D}_p^\star)$ is isolated with $u_{\mathcal{D}_p^\star}$ an associated unit eigenvector,

$$\left. \frac{d}{dt} \Lambda(\mathcal{D}_p^\star + t\delta\mathcal{D}) \right|_{t=0} + \gamma \left. \frac{d}{dt} \|\mathcal{D}_p^\star + t\delta\mathcal{D}\|_{L_\pi^p}^p \right|_{t=0} = 0$$

leads to

$$\mathcal{D}_p^\star(q) = \alpha_p |\mathcal{D}_p^\star(q)|_{\mathbb{F}}^{2-p} e^{\beta(p-1)V(q)} \nabla u_{\mathcal{D}_p^\star}(q) \otimes \nabla u_{\mathcal{D}_p^\star}(q)$$

- The matrix $\mathcal{D}_p^\star(q)$ of **rank 1** a.e.

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$$\boxed{\mathcal{D}_p^\star \propto e^{\beta V} \left| u'_{\mathcal{D}_p^\star} \right|^{2/(p-1)}}$$

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Else, we still expect

$$\boxed{\mathcal{D}_p^\star \propto e^{\beta V} \left(\sum_{i=1}^N \left| u'_{\mathcal{D}_p^\star, i} \right|^2 \right)^{1/(p-1)}}$$

Numerical approximation of the optimization problem

- For simplicity, $\mathcal{D}(q) = \mathcal{D}(q)\mathbf{I}_d$
- Piecewise constant approximation for \mathcal{D} on \mathbb{T}^d
- \mathbb{P}_1 Finite Elements approximation to compute $(\Lambda(\mathcal{D}), u_{\mathcal{D}})$:

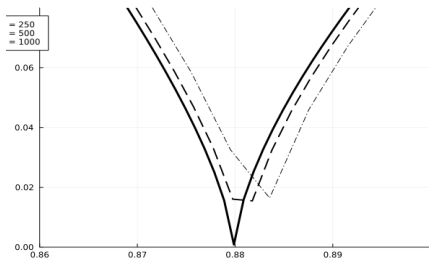
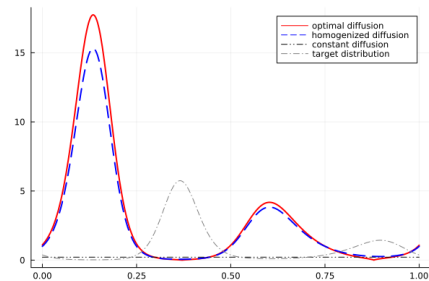
$$\boxed{A(\mathcal{D})u_{\mathcal{D}} = \Lambda(\mathcal{D})Bu_{\mathcal{D}}}$$

with

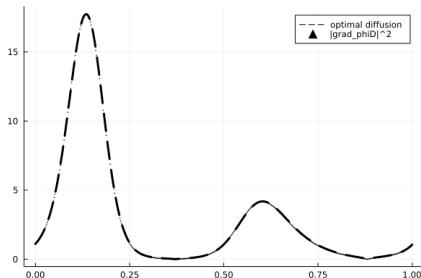
$$A_{i,j}(\mathcal{D}) = \int \nabla \varphi_j^{\top} \mathcal{D} \nabla \varphi_i \, d\pi, \quad B_{i,j} = \int \varphi_j \varphi_i \, d\pi$$

- **Generalized eigenvalue problem:** A sym., B pos. def. sym.

Numerical results - 1

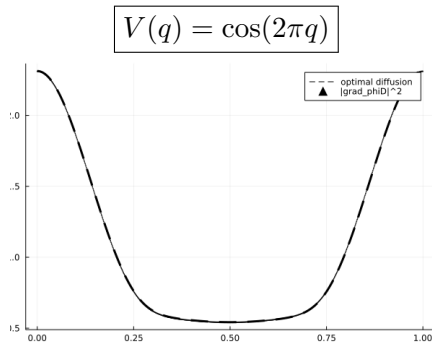
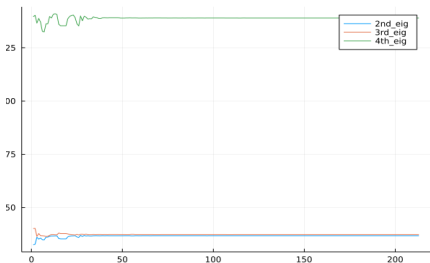
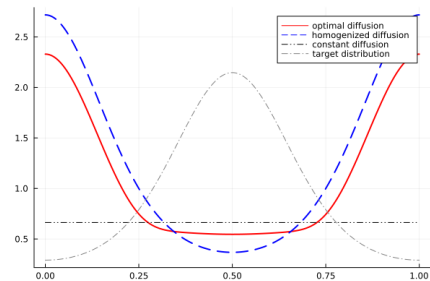


$$V(q) = \sin(4\pi q)(2 + \sin(2\pi q))$$



Isolated eigenvalue

Numerical results - 2



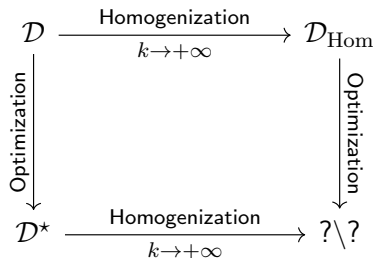
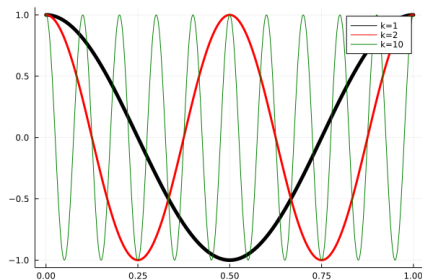
Multiple eigenvalue

Optimal diffusion in the homogenized limit

- Previous procedure only helpful in **low dimensions**
- Need to solve a **high-dimensional generalized eigenvalue problem**

Goal: Obtain a **good approximation** of the optimal diffusion

- Idea 1: study the asymptotic behaviour of the optimal diffusion in the **homogenized limit**
- Idea 2: optimize the periodic homogenization limit



Periodic homogenization procedure

- Decrease the period: $(\mathbb{Z}/k)^d$ -periodic functions $V_{\#,k}(q) = V(kq)$ and $\mathcal{D}_{\#,k}(q) = \mathcal{D}(kq)$
- Write the spectral gap problem:

$$\Lambda_{\#,k}(\mathcal{D}) = \min_{u \in H^1(\mathbb{T}^d) \setminus \{0\}} \left\{ \frac{\int_{\mathbb{T}^d} \nabla u^\top \mathcal{D}_{\#,k} \nabla u e^{-\beta V_{\#,k}}}{\int_{\mathbb{T}^d} u^2 e^{-\beta V_{\#,k}}} \left| \int_{\mathbb{T}^d} u e^{-\beta V_{\#,k}} = 0 \right. \right\}$$

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- Use **H-convergence**:⁵ $\exists \overline{\mathcal{D}} \in \mathfrak{D}_p^{a,b}$, $\Lambda_{\#,k}(\mathcal{D}) \xrightarrow[k \rightarrow +\infty]{} \Lambda_{\text{Hom}}(\mathcal{D})$ with

$$\Lambda_{\text{Hom}}(\mathcal{D}) := \min_{u \in H^1(\mathbb{T}^d) \setminus \{0\}} \left\{ \frac{\int_{\mathbb{T}^d} \nabla u^\top \overline{\mathcal{D}} \nabla u}{\int_{\mathbb{T}^d} u^2} \left| \int_{\mathbb{T}^d} u = 0 \right. \right\}$$

- $\overline{\mathcal{D}}$ can be expressed using \mathcal{D} and corrector functions appearing in the H-convergence procedure

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Optimization of the homogenized limit

Goal: compute

$$\Lambda_{\text{Hom}}^* = \sup_{\mathcal{D} \in \mathfrak{D}_p^{a,b}} \Lambda_{\text{Hom}}(\mathcal{D})$$

$$\begin{array}{ccc} \mathcal{D} & \xrightarrow[k \rightarrow +\infty]{\text{Hom.}} & \mathcal{D}_{\text{Hom}} \\ & & \downarrow \text{Opt.} \\ & & \mathcal{D}_{\text{Hom}}^* \end{array}$$

Theorem [Analytic expression]

- **Linear constraint:** For a fixed $M \in \mathcal{S}_d^{++}$, under the constraint, $\int_{\mathbb{T}^d} \mathcal{D} \, d\pi = M$,

$$\mathcal{D}_{\text{Hom}}^*(q) = M/\pi(q)$$

is a maximizer.

- **L_π^p constraint, $d = 1$:** Under the constraint $\|\mathcal{D}\|_{L_\pi^p} \leq 1$,

$$\mathcal{D}_{\text{Hom}}^*(q) = e^{\beta V(q)}$$

is a maximizer.

Homogenization of the optimal diffusion

Goal: optimize for a given $k \geq 1$, then let $k \rightarrow +\infty$

- Recall the oscillating potential $V_{\#,k}(q) = V(kq)$. Let $\mathfrak{D}_{\#,k,p}^{a,b} \equiv \mathfrak{D}_p^{a,b}$ but defined with $V_{\#,k}$ instead of V .

- Let

$$\Lambda^k(\mathcal{D}) = \min_{u \in H^1(\mathbb{T}^d) \setminus \{0\}} \left\{ \frac{\int_{\mathbb{T}^d} \nabla u^\top \mathcal{D} \nabla u e^{-\beta V_{\#,k}}}{\int_{\mathbb{T}^d} u^2 e^{-\beta V_{\#,k}}} \left| \int_{\mathbb{T}^d} u e^{-\beta V_{\#,k}} = 0 \right. \right\}$$

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Lemma

There exists a maximizer $\mathcal{D}^{k,\star} \in \mathfrak{D}_p^{a,b}$ such that, denoting by $\mathcal{D}_{\#,k}^{k,\star}(q) = \mathcal{D}^{k,\star}(kq)$,

$$\Lambda^k(\mathcal{D}_{\#,k}^{k,\star}) = \Lambda^{k,\star}$$

Commutation between Homogenization and Optimization

$$\begin{array}{ccc} \Lambda(\mathcal{D}) & \xrightarrow[k \rightarrow +\infty]{\text{Hom.}} & \Lambda_{\text{Hom}}(\mathcal{D}) \\ \text{Opt.} \downarrow & & \downarrow \text{Opt.} \\ \Lambda^{k,\star} & \xrightarrow[k \rightarrow +\infty]{\text{Hom.}} & \Lambda_{\text{Hom}}^{\star} \end{array}$$

Theorem

The sequence $(\Lambda^{k,\star})_{k \geq 1}$ converges to $\Lambda_{\text{Hom}}^{\star} := \Lambda(\mathcal{D}_{\text{Hom}}^{\star})$.

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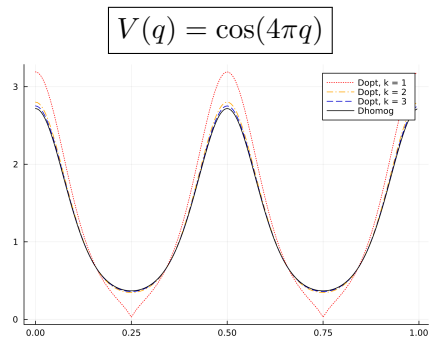
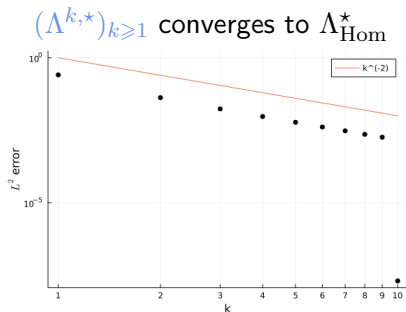
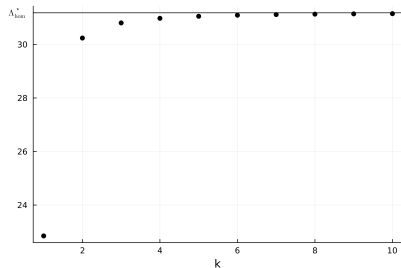
Theorem

The sequence $(\Lambda^{k,\star})_{k \geq 1}$ converges to $\Lambda_{\text{Hom}}^{\star} := \Lambda(\mathcal{D}_{\text{Hom}}^{\star})$.

- This implies that a good proxy ($d = 1$) is $\mathcal{D}_{\text{Hom}}^{\star} = e^{\beta V}$
- In this case, $\overline{\mathcal{D}} = (\int_{\mathbb{T}} e^{-\beta V})^{-1} := Z^{-1}$, and

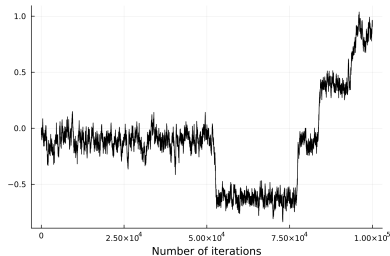
$$\Lambda_{\text{Hom}}^{\star} = 4\pi^2 Z^{-1}$$

Numerical results - 3

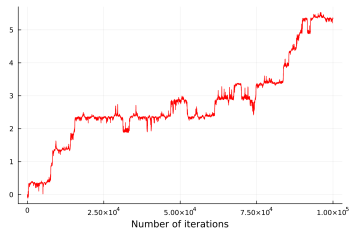


$(\mathcal{D}^{k,\star})_{k \geq 1}$ and $\mathcal{D}_{\text{Hom}}^*$ (rescaled with $1/k$ unit cell)

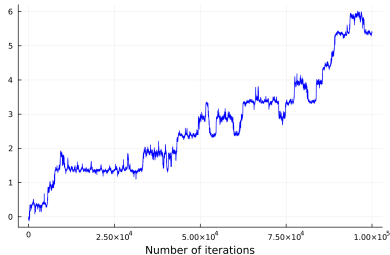
Numerical results - Application to sampling experiments - 1



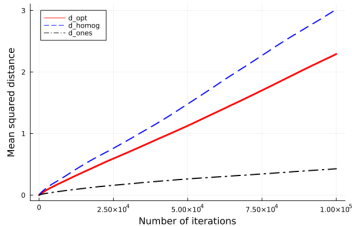
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Constant diffusion coefficient



Optimal diffusion coefficient

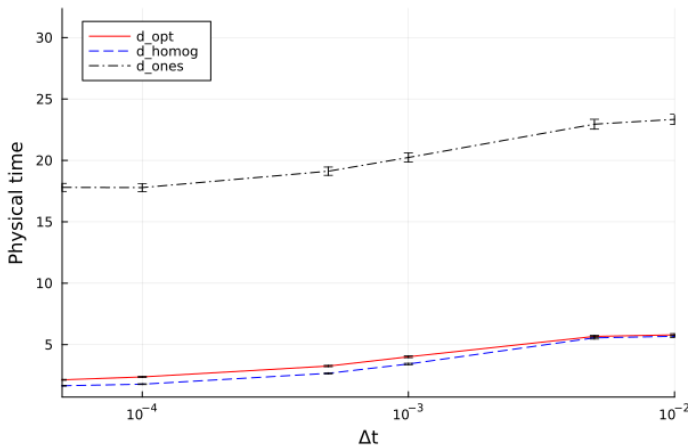


Homogenized diffusion coefficient

Mean square distance (averaged)

Numerical results - Application to sampling experiments - 2

Diffusion coefficient	Constant	Homogenized	Optimal
Spectral gap	2.16	10.57	11.23



Transition times between the two wells, $N_{\text{transitions}} = 10^5$

Conclusion

- Using a position-dependant diffusion coefficient can help **sample rare events, cross energy barriers**, etc.
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- Good approximation with **homogenization procedure**: $\mathcal{D}_{\text{Hom}}^* = e^{\beta V}$
- In high-dimension, use **free energy F** and coordinate reaction ξ :

$$\mathcal{D}(q) \propto e^{\beta F(\xi(q))}$$

Conclusion

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- Optimization problem can be solved numerically in **low dimensions**
- Good approximation with **homogenization procedure**: $\mathcal{D}_{\text{Hom}}^* = e^{\beta V}$
- In high-dimension, use **free energy F** and coordinate reaction ξ :

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- Spectral gap is only **one** criterion: sampling issues when $\mathcal{D}^*(q) \approx 0$
- Normalization constraint on \mathcal{D} : which one to choose ?

- Adapt to nonequilibrium dynamics: non-gradient force F , use biasing scalar function $\mathcal{E} > 0$

$$dq_t^\eta = \mathcal{E}(q_t^\eta) (-\nabla V(q_t^\eta) + \eta F(q_t^\eta)) dt + \sqrt{2\mathcal{E}(q_t^\eta)} dW_t.$$

- $\mathcal{L}_\mathcal{E}^\eta$ is not self-adjoint: optimize real part of the spectral gap
- Antisymmetric part in the gen. eigen. problem, complex eigenvalues

$$(A - \eta M) u_\mathcal{E} = \Lambda(\mathcal{E}) B(\mathcal{E}) u_\mathcal{E}$$

$$A \in \mathcal{S}_d, M \in \mathcal{A}_d, B \in \mathcal{S}_d^{++}$$

- Adapt to QSD: from a metastable state A , if $\nu^{\mathcal{D}}$ is a QSD *i.e.*

$$\begin{cases} \mathcal{L}_{\mathcal{D}}^* \nu_{\mathcal{D}} = \lambda(\mathcal{D}) \nu_{\mathcal{D}} & \text{on } A \\ \nu_{\mathcal{D}} = 0 & \text{on } \partial A \end{cases}$$

- Then to accelerate convergence, we maximize $\lambda_2 - \lambda_1$ which amounts to

$$\max_{\mathcal{D} + \text{proper normalization}} \lambda_2(\mathcal{D})$$

- It is observed numerically that there is indeed an optimum when $\mathcal{D} \propto e^{\alpha V}$, with $\alpha \approx 1.5$ (**normalization constraint effect**)

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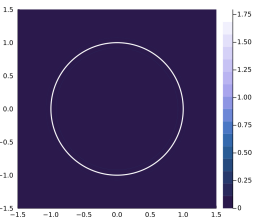
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Thank you !

Which diffusion coefficient? Anisotropic case

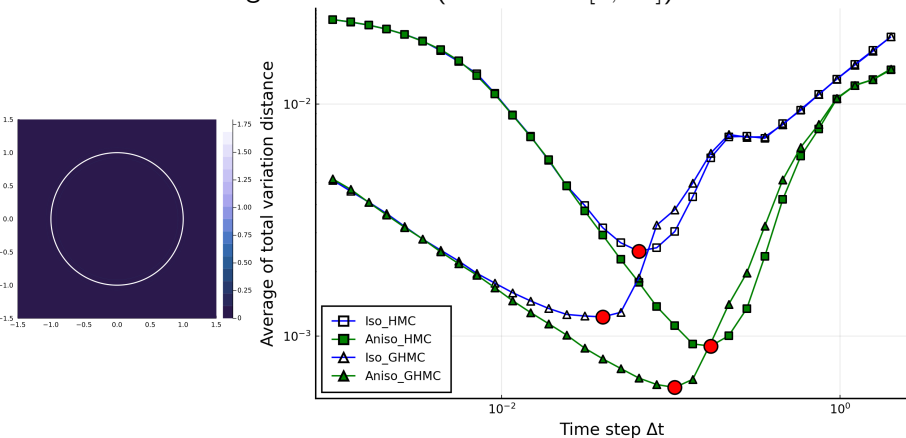
- Anisotropic diffusion coefficient $\mathcal{D}_{\text{Tan}}(q) = \varepsilon \mathbf{I}_2 + \tilde{q} \tilde{q}^T / \|q\|^2$, $\tilde{q} = (-y \ x)^T$
- Isotropic diffusion coefficient $\mathcal{D}_{\text{One}} \equiv (1 + \varepsilon) \mathbf{I}_2$, $\varepsilon = 0.1$



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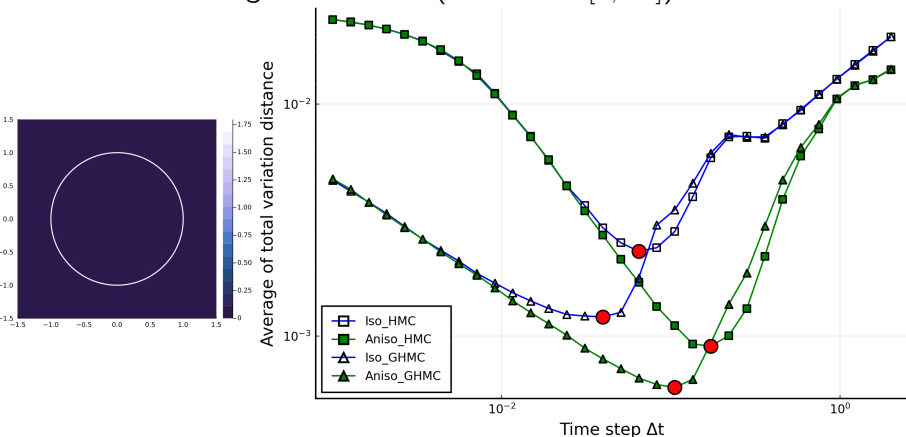
Computing: after fixed number of iterations, distance to the invariant measure of the angle distribution (uniform on $[0, 2\pi]$)



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⇒ Compromise: **small**/**large** time steps (exploration vs rejection)

Definition [H -convergence]

A sequence $(\mathcal{A}^k)_{k \geq 1} \subset L^\infty(\mathbb{T}^d, \mathcal{M}_{a,b})$ H -converges to $\overline{\mathcal{A}} \in L^\infty(\mathbb{T}^d, \mathcal{M}_{a,b})$ if, for any $f \in H^{-1}(\mathbb{T}^d)$ such that $\langle f, \mathbf{1} \rangle_{H^{-1}, H^1} = 0$, the sequence $(u^k)_{k \geq 1} \subset H^1(\mathbb{T}^d)$ of solutions to

$$\begin{cases} -\operatorname{div}(\mathcal{A}^k \nabla u^k) = f & \text{on } \mathbb{T}^d, \\ \int_{\mathbb{T}^d} u^k(q) dq = 0 \end{cases}$$

satisfies in the limit $k \rightarrow +\infty$,

$$\begin{cases} u^k \rightharpoonup u & \text{weakly in } H^1(\mathbb{T}^d), \\ \mathcal{A}^k \nabla u^k \rightharpoonup \overline{\mathcal{A}} \nabla u & \text{weakly in } L^2(\mathbb{T}^d)^d, \end{cases}$$

where $u \in H^1(\mathbb{T}^d)$ is the solution of the homogenized problem

$$\begin{cases} -\operatorname{div}(\overline{\mathcal{A}} \nabla u) = f & \text{on } \mathbb{T}^d, \\ \int_{\mathbb{T}^d} u(q) dq = 0 \end{cases}$$

Definition [Correctors]

If $\mathcal{A} = \mathcal{D} \exp(-\beta V)$, $(w_i)_{1 \leq i \leq d} \subset H^1(\mathbb{T}^d)$ is the family of unique solutions to the problem

$$\begin{cases} -\operatorname{div}(\mathcal{A}(e_i + \nabla w_i)) = 0, \\ \int_{\mathbb{T}^d} w = 0 \end{cases}$$

Then for any $\xi \in \mathbb{R}^d$,

$$\xi^\top \overline{\mathcal{D}} \xi = \xi^\top \left(\int_{\mathbb{T}^d} \mathcal{D}(q) d\pi \right) \xi - \int_{\mathbb{T}^d} \nabla w_\xi^\top \mathcal{D} \nabla w_\xi d\pi.$$