Kramers turnover: from energy diffusion to spatial diffusion using metadynamics

Pratyush Tiwary^{1, a)} and B. J. Berne^{1, a)}

Department of Chemistry, Columbia University, New York 10027, USA.

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We consider the rate of transition for a particle between two metastable states coupled to a thermal environment for various magnitudes of the coupling strength, using the recently proposed infrequent metadynamics approach (Tiwary and Parrinello, Phys. Rev. Lett. **111**, 230602 (2013)). We are interested in understanding how this approach for obtaining rate constants performs as the dynamics regime changes from energy diffusion to spatial diffusion. Reassuringly, we find that the approach works remarkably well for various coupling strengths in the strong coupling regime, and to some extent even in the weak coupling regime.

I. INTRODUCTION

It is well known that the dynamics of activated barrier crossing shows strong sensitivity to the strength of coupling to the environment¹. Specifically, for a prototypical setup with two stable states separated by a barrier, the rate constant k for barrier crossing first increases, and then decreases, as the coupling to the environment increases. This non-monotonic behavior of the rate constant, known as Kramers' turnover, is of fundamental interest in chemical dynamics, and manifests itself in a range of practical scenarios, including but not limited to isomerization reactions, protein folding and even excitation energy transfer in light-harvesting systems $^{2-4}$. It has been thoroughly investigated through numerical, analytical and experimental studies over the past few decades $^{1,5-8}$. It is well accepted that in the low coupling regime the rate constant is small due to poor exchange of energy with the environment. Either the system rarely gains enough energy to cross the barrier, or when it does so, it is unable to dissipate this energy and settle in the product state. On the other end, in the high coupling regime, spatial diffusion across the barrier top becomes the rate limiting factor, and increasing the environmental coupling leads to decrease in the rate constant.

In principle one can use molecular dynamics (MD) simulations to directly measure the rate constants for activated barrier crossing without making any assumptions on the nature of the dynamics regime. Here this coupling is generally implemented through a friction coefficient that quantifies the rate of collisions between the system of interest and a thermal bath. This however becomes a challenging task when the barrier height is much larger than the thermal energy k_BT , and it becomes difficult to observe sufficiently many or any barrier crossing events in MD given computer time limitations. To deal with this debilitating problem, over the years several enhanced sampling schemes have been proposed that accelerate barrier crossing events in a controllable manner. While many of these enhanced sampling methods concern recovering the underlying free energy landscape, some are designed to calculate the actual rate of barrier $crossing^{9-14}$.

In this short communication, we consider one such recently proposed method, the so-called 'infrequent metadynamics' approach^{14,15}, which has recently been applied successfully to obtain rate constants in a variety of complex molecular systems 16,17 , and is briefly summarized in Sec. II. Given the potential benefit of this approach and its increasing popularity, in this work we explore its robustness in obtaining rate constants for a model 2-state system¹⁸ with respect to varying coupling strength to a thermal bath implemented via Langevin dynamics¹⁹. Through a large number of independent simulations, we identify the environmental coupling regime in which the dynamics from infrequent metadynamics is correct, by comparing against much longer unbiased MD runs. We find that infrequent metadynamics gives correct rates across several orders of magnitude variation in the environmental coupling, as long as one stays in the high coupling regime. It also reproduces the pronounced change in rate associated with Kramers' turnover. It tends to become less reliable as the coupling constant is made very low and the system enters the deeply underdamped regime. Thankfully, most biomolecular systems, which are the target systems for the method 16,17 , involve large numbers of solvent atoms executing rapid thermal motions where one expects the moderate to high friction regime to be applicable. As we find in this work, the infrequent metadynamics approach is indeed reliable in this regime.

II. DYNAMICS FROM INFREQUENT METADYNAMICS

Metadynamics is an enhanced sampling approach that begins by identifying a small number of slowly changing order parameters, called collective variables $(CVs)^{20,21}$. By periodically adding repulsive bias in the regions of CV space as they are visited, the system is encouraged to escape stable free energy minima where it would normally be trapped. The traditional objective of a metadynamics run is to recover the underlying true free energy surface as a function of the deposited bias²⁰. Recently, Tiwary

^{a)}Electronic mail: pt2399@columbia.edu

and co-workers extended the scope of metadynamics by showing how to extract unbiased rates from biased ones with minimal extra computational burden^{14,15}, inspired by works such as Ref. 10 and 11. The central idea was to deposit bias infrequently enough compared to the time spent in the transition state regions. Through this 'infrequent metadynamics' approach one increases the likelihood of not corrupting the transition states through the course of metadynamics, thus preserving the sequence of transitions between stable states. One can then access the acceleration of transition rates achieved through biasing by appealing to generalized transition state theory¹ and calculating the following simple running average¹⁴:

$$\alpha = \langle e^{\beta V(s,t)} \rangle_t \tag{1}$$

where s is the collective variable being biased, $\beta = \frac{1}{k_B T}$ is the inverse temperature, V(s,t) is the bias experienced at time t and the subscript t indicates averaging under the time-dependent potential.

This approach assumes that one has the correct slow order parameters or CVs that can demarcate all relevant stable states of interest. Whether this is the case or not can be verified *a posteriori* by checking if the transition time statistics conforms to a Poisson distribution¹⁵. The current work assumes one has the right order parameters or collective variables for the problem at hand. For the model potential considered in this work (Sec. III), this is not a problem. For more complicated systems one can identify such CVs in principle through a recently proposed method²².

III. SIMULATION DETAILS

We consider a model 2-state potential (Fig. 1(a)) introduced by De Leon and Berne¹⁸, which has been the subject of numerous similar studies¹ in the past. It is given by

$$V(x,y) = 4y^2(y^2 - 1)e^{-4.485x} + 10(1 - e^{-1.95x})^2 + 1$$
(2)

All units of energy and mass were defined in terms of Eq. 2. Let q = (x, y) denote the configurational coordinates of the system, and m be the mass. All our simulations are performed with m = 8 units and temperature $k_B T = 0.1$ units, with Newton's laws of motion integrated per Langevin dynamics with a timestep 0.05 units:

$$m\ddot{q} + \frac{dV(q)}{dq} + m\gamma\dot{q} = F(t) \tag{3}$$

where γ is the friction coefficient, F(t) is a Gaussian random noise with mean zero and correlation function $\langle F(t)F(t')\rangle = \frac{2m\gamma}{\beta}\delta(t-t')$. There are many available schemes for implementing Langevin dynamics - we use the one from Ref. 19. For unbiased MD as well as metadynamics, the simulations were performed separately for 11 different values of the friction coefficient γ between 10^{-4} and 10. The bias was constructed as a function of the spatial coordinate y. The well-tempered flavor of metadynamics²¹ was used, with the so-called biasing factor that controls the gradual decay of Gaussian amplitude set to 6. An initial Gaussian height of 0.1 k_BT and width of 0.1 units were used. Two different biasing frequencies were used to ascertain sensitivity to the biasing stride - once every 20,000 integration steps and once every 100,000 steps. These will be denoted as the fast and slow deposition schemes respectively.



FIG. 1: (a) The 2-state model potential considered in this work, as described in Sec. III and in Ref. 18. All energies are in k_BT units, with contours spaced every 2 k_BT . The barrier here is approximately 10 k_BT . (b) The rate ν of transitions between the two stable states in (a), as a function of the friction coefficient γ in Eq. 3. The red circles correspond to unbiased MD estimates. The blue diamonds and magenta stars are for the slow deposition and fast deposition schemes respectively. All units are in accordance with Eq. 2.

IV. RESULTS AND DISCUSSION

In Fig. 1(b) we provide the transition rate ν as a function of the friction coefficient γ in Eq. 3, which is kept same for both x and y coordinates. The rate was calculated as $\nu = 1/\langle t \rangle$ where $\langle t \rangle$ is the average residence time in either of the basins. To filter out spurious recrossing events, we used a minimum residence time criterion of 10^4 time units, or 2×10^5 integration steps, to count a transition event as successful. This criterion, which is similar to checking for a plateau region in the reactive flux formalism⁹, was enforced uniformly across unbiased MD and both the metadynamics schemes. While the absolute magnitude of the rate constant itself shows some sensitivity to the employed minimum residence time cutoff, the conclusions of this work as described below were found to be very robust to this choice. The metadynamics runs with the slow and fast deposition schemes corresponded to acceleration as per Eq. 1 relative to unbiased MD, of approximately 35 and 45 respectively. Naturally, the simulation lengths for the unbiased MD runs were thus correspondingly much longer. The following salient features can be seen from Fig. 1 (b):

- 1. Both the deposition schemes for metadynamics agree with unbiased MD in the magnitude of the friction γ at which turnover in rate occurs, and the dynamics switches from energy diffusion to spatial diffusion.
- 2. All three schemes viz. unbiased MD, metadynamics with slow bias deposition and with fast bias deposition agree very well across the spatial diffusion regime, which is not entirely surprising given that the bias was constructed explicitly as a function of the spatial coordinate y.
- 3. What is more surprising is that MD and metadynamics continue to agree to some extent even in the energy diffusion regime (i.e. low friction γ). As the friction γ is continued to reduce, at some point the rate curves from both the fast and slow deposition schemes cleave off the curve from unbiased MD. The cleavage occurs earlier (that is at higher friction) for the fast deposition scheme than for the slow deposition scheme. This is because the latter is less likely to corrupt dynamical trajectories that take a long time to commit to either of the two stable states even when they have crossed the barrier.

Thus, to summarize, in this short communication we demonstrate through a numerical example that infrequent metadynamics performs remarkably well in obtaining the rate constant for various environmental coupling strengths, especially in the moderate to high friction regime, which is the regime of interest for most solvated biomolecular systems. At very low frictions, the system essentially undergoes a change in reaction coordinate from spatial to one that quantifies energy transfer between the reaction coordinate and the other intramolecular degrees of freedom (i.e. IVR or intramolecular vibrational relaxation). Yet, to some extent even in this regime we find that metadynamics can give accurate rate constants. This work thus supports the use of the approach developed in Ref. 14 for obtaining kinetic rate constants.

Finally, we would like to suggest that such a test that ascertains the accuracy of rates across dynamics regime change be performed on any method before it is judged to be trustworthy as a tool for enhancing molecular dynamics and obtaining rate constants.

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