

## Supplemental Material: Slowest kinetic modes revealed by metabasin renormalization

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In this Supplemental Material, we will show that the renormalization procedure developed in the paper can be applicable to the discrete-time kinetic equations, also known as (discrete-time) Markov state models [1], with small modifications.

Suppose that the kinetic state is described by the distribution of probability,  $p_i$ , for  $i = 1, 2, \dots, n$ , where  $n$  denotes the number of states, the kinetic equations are given by

$$p_i(t+1) = \sum_{j=1}^n t_{ij} p_j(t) + p_i(t) \left(1 - \sum_{j=1}^n t_{ji}\right) \quad \text{for } i = 1, 2, \dots, n, \quad (1)$$

where  $p_i(t)$  is the probability distribution of the system state  $i$  at discrete times  $t = 0, 1, 2, \dots$ , and  $t_{ij}$  is the transition probability from state  $j$  to state  $i$  for  $j \neq i$ , otherwise  $t_{ii} = 0$ . With the transition *probability* matrix  $T$  defined by  $(T)_{ij} = t_{ij} + \delta_{ij}(1 - \sum_{j'=1}^n t_{j'i})$ , the equations can be expressed in a matrix form:  $\mathbf{p}(t+1) = T\mathbf{p}(t)$ . We assume the equilibrium,  $\lim_{t \rightarrow \infty} \mathbf{p}(t)$ , to be a unique static state. Accordingly, the eigenvalues of  $T$  satisfy  $1 = \lambda_0 > \lambda_1 \geq \dots \geq \lambda_{n-1} > 0$  [1]. The equilibrium  $\mathbf{p}(\infty)$  coincides with the zeroth eigenvector of  $T$ , and the first, second,  $\dots$  eigenvectors of  $T$  represent the slowest relaxation modes.

Metabasins (MBs) in transition probability matrices, are determined similarly with the use of monotonic sequences [2, 3]. A sequence of states  $i_1 \rightarrow i_2 \rightarrow \dots$  is called monotonic if it consists only of most probable transitions. The monotonic sequences with the same terminal state belong to the same MB.

Similarly to the transition rate matrix  $K$ , the columns and rows of  $T$  are rearranged in the MB ordering,  $\sigma(1, 1), \sigma(1, 2), \dots, \sigma(2, 1), \sigma(2, 2), \dots$ , and the resultant matrix is denoted by  $T_\sigma$ , where  $\sigma(\ell, i)$  returns the number  $j$  of state  $j$  that is the  $i$ th energy state in  $\text{MB}_\ell$ . Then, we consider the block diagonal matrix  $\text{diag}(T_1, \dots, T_\ell, \dots, T_m)$  to be the unperturbed matrix, where  $T_\ell$  is given by

$$(T_\ell)_{ij} = t_{\sigma(\ell, i), \sigma(\ell, j)} + \delta_{ij} \sum_{j'=1}^{n_\ell} (1 - t_{\sigma(\ell, j'), \sigma(\ell, i)}),$$

where  $n_\ell$  is the size of  $\text{MB}_\ell$ . Note that  $j$ th eigenvalues,  $\lambda_{\ell, j}$ , of  $T_\ell$  satisfy  $1 = \lambda_{\ell, 0} > \lambda_{\ell, 1} \geq \dots \geq \lambda_{\ell, n_\ell-1} > 0$ . To consider the intra-MB relaxation modes, we form  $\Lambda_\ell = D_\ell^{-1} T_\ell D_\ell$ , with the use of the diagonal matrix  $D_\ell = \text{diag}(\sqrt{\mathbf{p}_{\ell, 0}})$ , where  $\mathbf{p}_{\ell, 0}$  is the local equilibrium in  $\text{MB}_\ell$ . It is noteworthy that  $\Lambda_\ell$  is the symmetric matrix and can be diagonalized with an orthogonal matrix  $S_\ell = [\sqrt{\mathbf{p}_{\ell, 0}}, \mathbf{v}_{\ell, 1}, \dots, \mathbf{v}_{\ell, n_\ell-1}]$ , where  $\mathbf{v}_{\ell, j}$  is the  $j$ th eigenvectors of  $\Lambda_\ell$ .

To consider the inter-MB transitions, we form the symmetric matrix  $\Lambda' = S^T D^{-1} T_\sigma D S$ , with  $S = \text{diag}(S_1, S_2, \dots)$  and  $D = \text{diag}(\sqrt{\mathbf{p}_{\text{eq}}})$ , where  $\mathbf{p}_{\text{eq}}$  is the equilibrium of  $T_\sigma$ . Moreover, to introduce the division of intra-MB relaxation modes into slow and fast modes, we set a certain threshold  $\lambda_{\text{cut}}$  satisfying  $1 \geq \lambda_{\text{cut}} > 0$ : the slow relaxation modes are  $1 \geq \lambda_{\ell, j} \geq \lambda_{\text{cut}}$  and the fast relaxation modes are  $\lambda_{\text{cut}} > \lambda_{\ell, j} \geq 0$ . We, then, reorder the columns and rows of  $\Lambda'$  in the slow-to-fast relaxation block order, and the resultant matrix is denoted by  $\Lambda_{\text{slow-fast}}$ . Note that  $\Lambda_{\text{slow}}$ , defined by the first  $n_{\text{slow}} \times n_{\text{slow}}$  submatrix with  $n_{\text{slow}}$  denoting the number of unperturbed slow relaxation modes, generally describes an approximate transitions between the slow modes. To obtain the accurate results, we need a renormalized transition matrix  $\Lambda_{\text{slow}}^{\text{RG}}$ . To this end, we use a Jacobi rotation  $\Lambda_{\text{slow-fast}} \mapsto \Lambda_{\text{slow-fast}}^{\text{RG}} = G^T \Lambda_{\text{slow-fast}} G$  such that the resultant couplings between slow and fast modes,  $(\Lambda_{\text{slow-fast}}^{\text{RG}})_{ij}$  with  $i \leq n_{\text{slow}} < j$ , are vanishing. Now, the renormalized transition matrix  $\Lambda_{\text{slow}}^{\text{RG}}$  is defined by the first  $n_{\text{slow}}$ -by- $n_{\text{slow}}$  submatrix of  $\Lambda_{\text{slow-fast}}^{\text{RG}}$ , which reproduces the exact slowest relaxations of the kinetic equation (1).

Table I summarizes the differences between the renormalization procedures for discrete-time and for continuous-time kinetic equations, from which we clearly see that the renormalization procedure for discrete-time kinetic equations is constructed almost in the same way as that for continuous-time kinetic equations developed in the paper.

TABLE I. Differences between discrete-time and continuous-time renormalization procedures: The listed differences are absorbed by the matrix  $\Lambda'$  and the definitions of slow modes. In other words, all the procedures after specifying these are exactly the same way between for the discrete-time equation and for the continuous-time equation.

Time	Matrix to be normalized	$\Lambda'$	Slow modes
Discrete time	Transition probability matrix $T$	$S^T D^{-1} T_\sigma D S$	$1 \geq \lambda_{\ell, i} \geq \lambda_{\text{cut}}$
Continuous time	Transition rate matrix $K$	$S^T D^{-1} K_\sigma D S$	$0 \geq \lambda_{\ell, i} \geq \lambda_{\text{cut}}$

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- [1] *An Introduction to Markov State Models and Their Application to Long Time scale Molecular Simulation*, edited by G. R. Bowman, V. S. Pande, and F. Noé (Springer, New York, 2013).
  - [2] For another formularization for MB decompositions of discrete-time Markov state models, see K. Klemm, C. Flamm, and P. F. Stadler, *Eur. Phys. J. B* **63**, 387 (2008).
  - [3] T. Okushima, T. Niiyama, K. S. Ikeda, and Y. Shimizu, *Phys. Rev. E* **80**, 036112 (2009).