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# Conformational Nonequilibrium Enzyme Kinetics: Generalized Michaelis–Menten Equation

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#### Abstract

In a conformational nonequilibrium steady state (cNESS), enzyme turnover is modulated by the underlying conformational dynamics. Based on a discrete kinetic network model, we use the integrated probability flux balance method to derive the cNESS turnover rate for a conformation-modulated enzymatic reaction. The traditional Michaelis–Menten (MM) rate equation is extended to a generalized form, which includes non-MM corrections induced by conformational population currents within combined cyclic kinetic loops. When conformational detailed balance is satisfied, the turnover rate reduces to the MM functional form, explaining its validity for many enzymatic systems. For the first time, a one-to-one correspondence is established between non-MM terms and combined cyclic loops with unbalanced conformational currents. Cooperativity resulting from nonequilibrium conformational dynamics has been observed in enzymatic reactions, and we provide a novel, rigorous means of predicting and characterizing such behavior. Our generalized MM equation affords a systematic approach for exploring cNESS enzyme kinetics.

Conformational dynamics is essential for understanding the biological functions of enzymes. For decades, the framework of enzymatic reactions has been the traditional Michaelis–Menten (MM) mechanism,<sup>1</sup> where enzyme-substrate binding initializes an irreversible catalytic reaction to form a product. The average turnover rate v in a steady state (SS) follows a hyperbolic dependence on the substrate concentration [S],  $v = k_2[S]/(K_M + [S])$ , where the catalytic rate  $k_2$  and the Michaelis constant  $K_{\rm M}$  characterize this enzymatic chain reaction. In contrast to the single-conformation assumption for the traditional MM mechanism, recent singlemolecule experiments $^{2-4}$  have revealed the existence of multiple enzymatic conformations, spanning a broad range of lifetime scales from milliseconds to hours. Conformational dynamics, including hopping between different conformations and thermal fluctuations around a single-conformation potential well, must be incorporated into enzymatic reaction models for a quantitative study.<sup>3,5–18</sup> Slow conformational dynamics modulate the enzymatic reaction and allow the enzyme to exist in a conformational nonequilibrium steady state (cNESS), permitting complex deviations from MM kinetics (the hyperbolic [S] dependence for v). However, experimental and theoretical studies have shown MM kinetics to be valid in the presence of slow conformational dynamics under certain conditions, although  $k_2$  and  $K_M$  become averaged over conformations.<sup>3,10,15</sup> Is there a unifying theme governing this surprising behavior?

Non-MM enzyme kinetics have been characterized by cooperativity for many years.<sup>5,6,19</sup> For allosteric enzymes with multiple binding sites, the binding event at one site can alter the reaction activity at another site, accelerating (decelerating) the turnover rate and resulting in positive (negative) cooperativity.<sup>19</sup> Another common deviation from MM kinetics is substrate inhibition, where the turnover rate reaches its maximum value at a finite substrate concentration and then decreases at high substrate concentrations.<sup>19</sup> For a monomeric enzyme, the above non-MM kinetic behavior, referred to – in this case – as 'kinetic cooperativity.

ity,'<sup>19</sup> can be achieved by a completely different mechanism: nonequilibrium conformational dynamics.<sup>5,6,11–15</sup> Can we characterize and predict this interesting behavior in a cNESS?

Recently, theoretical efforts have been applied to study conformation-modulated enzyme kinetics by including dynamics along a conformational coordinate. On the basis of the usual rate approach, some previous work has demonstrated certain non-MM kinetics under specific conditions.<sup>10,12–14,16</sup> Based on an alternative integrated probability flux balance method, non-MM kinetics were linked to a nonzero conformational population current, i.e., broken conformational detailed balance, in a two-conformation model, and a general MM expression was speculated.<sup>15</sup> However, a generalized theory to systematically analyze cNESS enzyme kinetics is still needed. In this Letter, we focus on a monomeric enzyme and apply the integrated flux balance method to derive a generalized form for the turnover rate, which includes non-MM corrections. We show that when conformational detailed balance is satisfied, MM kinetics hold, explaining their general validity. In addition, the deviations from MM kinetics are analyzed with reduced parameters from the generalized form of v. For an extended version of our derivation, we refer readers to ref 20.

To describe the generalized conformation-modulated reaction catalyzed by a monomeric enzyme, we introduce a discrete kinetic network model, which is illustrated in Figure 1a. This  $N \times M$  network consists of a vertical conformation coordinate  $(1 \leq i \leq N)$  and a horizontal reaction coordinate  $(1 \leq j \leq M)$ . For the reaction state index, j = 1 denotes the initial substrate-unbound enzymatic state (E), whereas  $j \geq 2$  denotes intermediate substratebound enzymatic states (ES).<sup>3,14,15,21,22</sup> Without product states, our network corresponds to a dissipative system. For an arbitrary site  $R_{i,j}$ , the reaction rates for the forward  $(R_{i,j} \rightarrow R_{i,j+1})$  and backward  $(R_{i,j} \rightarrow R_{i,j-1})$  directions are given by  $k_{i,j}$  and  $k_{i,-(j-1)}$ , respectively. The rate for enzyme-substrate binding, the only step in our model dependent upon substrate concentration [S], depends linearly on [S] as  $k_{i,1} = k_{i,1}^0$ [S] for binding rate constant  $k_{i,1}^0$ , with [S] maintained constant in most enzymatic experiments. The conformational dynamics are treated via a kinetic rate approach, with the interconversion (hopping or diffusion) rates for



b) flux network



Figure 1: (a) Generalized kinetic network scheme for a conformation-modulated enzymatic reaction. (b) Flux network corresponding to (a) (see text for details).

 $R_{i,j} \to R_{i+1,j}$  and  $R_{i,j} \to R_{i-1,j}$  given by  $\gamma_{i,j}$  and  $\gamma_{-(i-1),j}$ , respectively. We note that local detailed balance results in the constraint  $k_{i,j}\gamma_{-i,j}/(k_{i,-j}\gamma_{i,j}) = k_{i+1,j}\gamma_{-i,j+1}/(k_{i+1,-j}\gamma_{i,j+1})$  for  $j \leq M - 1$ . However, for the purposes of our kinetic analysis, it is unnecessary to impose this relation, as our principal results hold, irrespective of whether it is satisfied. The rate equation for site  $R_{i,j}$  is written as

$$\frac{d}{dt}P_{i,j}(t) = \sum_{i'=1}^{N} \gamma_{i,i';j} P_{i',j}(t) + \sum_{j'=1}^{M} k_{j,j';i} P_{i,j'}(t)$$
(1)

where  $P_{i,j}(t)$  is the probability of an enzyme in site  $R_{i,j}$  at time t, i.e., the survival probability for the site. Here,  $\gamma_{i,i';j} = \gamma_{i-1,j}\delta_{i',i-1} + \gamma_{-i,j}\delta_{i',i+1} - [\gamma_{i,j} + \gamma_{-(i-1),j}]\delta_{i',i}$  denotes the interconversion rates in the *j*-th reaction state and  $k_{j,j';i} = k_{i,j-1}\delta_{j',j-1} + \gamma_{i,-j}\delta_{j',j+1} - [k_{i,j} + k_{i,-(j-1)}]\delta_{j',j}$ denotes the reaction rates for the *i*-th conformation.

Within the framework of a dissipative enzymatic network, the average turnover rate v is equivalent to the inverse of the mean first passage time (MFPT)  $\langle t \rangle$ . Using the residence time  $\tau_{i,j} = \int_0^\infty P_{i,j}(t)dt$  at each site  $R_{i,j}$ , we can express the MFPT in the  $N \times M$  network as a summation of  $\tau_{i,j}$ , i.e,  $\langle t \rangle = \sum_{i,j} \tau_{i,j}$ . Instead of inverting the transition matrix,<sup>3</sup> we evaluate  $\tau_{i,j}$  by inspecting integrated probability fluxes,<sup>15</sup> which correspond to stationary population fluxes normalized by v, and these will be shown to directly reflect conformational nonequilibrium. Along the horizontal reaction coordinate, the integrated flux for  $R_{i,j} \rightarrow$  $R_{i,j+1}$  is given by  $F_{i,j} = k_{i,j}\tau_{i,j} - k_{i,-j}\tau_{i,j+1}$ . Along the vertical conformation coordinate, the integrated flux for  $R_{i,j} \rightarrow R_{i+1,j}$  is given by  $J_{i,j} = \gamma_{i,j}\tau_{i,j} - \gamma_{-i,j}\tau_{i+1,j}$ . In addition, we need to specify the initial condition  $P_{i,j}(t=0)$  for calculating  $\langle t \rangle$ . For a monomeric enzyme, each turnover event begins with the substrate-unbound state, and  $P_{i,1}(t=0)$  defines the initial flux  $F_{i,0}$ . With the definition of  $\{F_{i,j}, J_{i,j}\}$ , we map the original kinetic network to a flux network as shown in Figure 1b. For each site  $R_{i,j}$ , the rate equation in eq 1 is replaced by a flux balance relation,

$$F_{i,j-1} + J_{i-1,j} = F_{i,j} + J_{i,j} \tag{2}$$

which is generalized to the probability conservation law: the total input integrated probability flux must equal the total output integrated probability flux. This conservation law can be extended to complex first-order kinetic structures including the  $N \times M$  network. The flux balance method thus provides a simple means of calculating the MFPT.

To evaluate the MFPT, we begin with the final reaction state (j = M) and propagate all the fluxes back to the initial reaction state (j = 1) based on eq 2. For each site  $R_{i,j}$ , the physical nature of the first-order kinetics determines that all three variables,  $\tau_{i,j}$ ,  $J_{i,j}$ and  $F_{i,j}$ , are linear combinations of terminal fluxes  $F_{i,j=M}$ . The first two variables are formally written as  $\tau_{i,j} = \sum_{i'=1}^{N} a_{i,j,i'}F_{i',M}$  and  $J_{i,j} = \sum_{i'} c_{i,j,i'}F_{i',M}$ , where  $a_{i,j,i'}$  and  $c_{i,j,i'}$ are coefficients depending on rate constants  $\{k, \gamma\}$ . For example, the coefficients for the final reaction state are  $a_{i,M,i'} = 1/k_{i,M}\delta_{i',i}$  and  $c_{i,M,i'} = \gamma_{i,M}/k_{i,M}\delta_{i',i} - \gamma_{-i,M}/k_{i+1,M}\delta_{i',i+1}$ . Because of the direction of our reversed flux propagation, only the coefficients for the initial reaction state are [S] dependent, and they can be explicitly written as  $a_{i,1,i'} = b_{i,i'}/[S]$  and  $c_{i,1,i'} = d_{i,i'}/[S]$ . The substrate-unbound ( $\mathbf{E}_i = R_{i,1}$ ) and substrate-bound ( $\mathbf{ES}_i = \sum_{j=2}^{M} R_{i,j}$ ) states are distinguished by the different [S] dependence of the coefficients. The MFPT is thus given by

$$\langle t \rangle = \sum_{i'=1}^{N} \left[ \frac{\sum_{i=1}^{N} b_{i,i'}}{[S]} + \sum_{i=1}^{N} \sum_{j=2}^{M} a_{i,j,i'} \right] F_{i',M}$$
(3)

The essential part of our derivation is then to solve for the terminal fluxes  $F_{i,M}$ . The SS condition can be interpreted as follows: after each product release, the enzyme returns to the same conformation for the next turnover reaction, i.e.,  $F_{i,M} = F_{i,0}$ .<sup>7</sup> Applying the probability conservation law to each horizontal chain reaction with a single conformation and considering the boundary condition at conformations i = 1 and N, we express the SS condition as a flux constraint,  $J_{i,E} + J_{i,ES} = 0$  for  $i = 1, 2, \dots, N-1$ , where  $J_{i,E} = J_{i,1}$ and  $J_{i,ES} = \sum_{j=2}^{M} J_{i,j}$ . For each combined cyclic loop  $E_i \rightarrow E_{i+1} \rightarrow ES_{i+1} \rightarrow ES_i \rightarrow E_i$ , there may exist a stabilized nonequilibrium conformational population current (see Figure 1b), with  $J_{i,E}$  representing this stationary current normalized by v. However, under certain circumstances,  $J_{i,E}$  can vanish, and the SS condition is further simplified to  $J_{i,ES} = 0$ . We note that satisfaction of the aforementioned constraint resulting from local detailed balance still permits nonzero  $J_{i,E}$ . In general, we assume that there exist  $N_c (\leq N - 1)$  nonzero conformational currents and  $(N - 1 - N_c)$  zero ones. In addition to these (N - 1) current conditions, the normalization condition  $\sum_{i=1}^{N} F_{i,0} = 1$  is needed for fully determining the initial fluxes (due to  $F_{i,0} = F_{i,M}$ ). As a result, we derive an N-equation array for  $F_{i,0}$ ,

$$\mathbf{U} \cdot \mathbf{F} = \begin{bmatrix} 1 & 1 & \cdots \\ C_{1,1} + \left(\frac{d_{1,1}}{[S]}\right) & C_{1,2} + \left(\frac{d_{1,2}}{[S]}\right) & \cdots \\ C_{2,1} + \left(\frac{d_{2,1}}{[S]}\right) & C_{2,2} + \left(\frac{d_{2,2}}{[S]}\right) & \cdots \\ \vdots & \vdots & \vdots \end{bmatrix} \cdot \begin{bmatrix} F_{1,0} \\ F_{2,0} \\ \vdots \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ \vdots \end{bmatrix} \tag{4}$$

with  $C_{i,i'} = \sum_{j=2}^{M} c_{i,j,i'}$ . Notice that for the (i + 1)-th row of matrix **U** in eq 4,  $d_{i,i'}/[S]$  only exists when  $J_{i,E} \neq 0$ , and  $N_c$  rows are [S] dependent for this matrix. We solve for the initial fluxes by the matrix inversion  $F_{i,0} = [\mathbf{U}^{-1}]_{i,1}$ . After a tedious but straightforward derivation,  $F_{i,0}$  is written as

$$F_{i,0}([S]) = f_{i,0} + \sum_{n=1}^{N_c} f_{i,n} / ([S] + s_n)$$
(5)

where each  $s_n$  is assumed to be distinct, and constraints hold for  $\sum_i f_{i,0} = 1$  and  $\sum_i f_{i,n} = 0$ for  $n \ge 1$ .

Substituting eq 5 into eq 3, we obtain the key result of this Letter: the cNESS turnover rate for the  $N \times M$  network with  $N_c$  unbalanced conformational currents is given by a generalized Michaelis–Menten equation,

$$v = \left[A_0 + \frac{B_0}{[S]} + \sum_{n=1}^{N_c} \frac{B_n}{[S] + s_n}\right]^{-1}$$
(6)

where the reduced parameters are  $A_0 = \langle 1/k_2^{\text{eff}} \rangle_{[S] \to \infty}$ ,  $B_0 = \langle K_M^{\text{eff}}/k_2^{\text{eff}} \rangle_{[S]=0}$ , and  $B_n = \sum_i [1/k_{i,2}^{\text{eff}} - K_{i,M}^{\text{eff}}/(k_{i,2}^{\text{eff}}s_n)] f_{i,n}$ . For each conformational channel, we introduce an effective cat-

alytic rate  $k_{i,2}^{\text{eff}} = (\sum_{i'=1}^{N} \sum_{j=2}^{M} a_{i',j,i})^{-1}$  and an effective Michaelis constant  $K_{i,M}^{\text{eff}} = k_{i,2}^{\text{eff}} \sum_{i'} b_{i',i}$ , which describe the kinetics within that channel in the decomposed representation of the scheme, wherein the N two-state chain reactions are effectively independent, each with probability  $F_{i,0}$ . The conformational average is defined as  $\langle x \rangle_{[S]} = \sum_{i} x_i F_{i,0}([S])$  for a conformation-dependent variable  $x_i$ . In the right hand side of eq 6, the first two terms retain the traditional MM form, whereas the remaining  $N_c$  terms introduce non-MM rate behavior, with a 1:1 correspondence between non-MM terms and combined cyclic loops with nonzero conformational currents. Our derivations clearly show that these non-MM terms are induced by the [S]-dependent conformational distribution **F** resulting from nonequilibrium conformational currents. Therefore, MM kinetics are valid when conformational detailed balance is satisfied, where all  $B_n$  vanish due to  $F_{i,0} = f_{i,0}$ .



Figure 2: (a) Three non-MM turnover rates for the single-loop model with  $A_0 = B_0 = s_1 = 1$ . The circles  $(B_1 = -1)$  and the up-triangles  $(B_1 = 2)$  exhibit positive and negative cooperativity, respectively. The two solid lines are the fit using the Hill equation. The dashed line  $(B_1 = -2)$  shows substrate inhibition behavior. (b) Phase diagram of enzyme kinetics for the single-loop model. Two lines,  $B_1 = 0$  and  $B_1 = -B_0$ , separate three regimes of kinetics.

With nonzero conformational currents, the enzyme kinetics are expected to exhibit cooperative non-MM behavior. As a demonstration, the single-loop model with only one current  $J_{1,E}$  and one non-MM term  $B_1/([S] + s_1)$  is first considered. With other parameters fixed, we calculate turnover rates v for the three values of  $B_1$  in Figure 2a. For the two turnover rates monotonically increasing with [S]  $(B_1 = -1 \text{ and } 2)$ , we fit them with the Hill equation,  $v/v_{\text{max}} = [S]^{n_{\text{H}}}/(\kappa + [S]^{n_{\text{H}}})$ , where the Hill constant  $n_{\text{H}} > 1$   $(n_{\text{H}} < 1)$  indicates positive (negative) cooperativity. The fitting results show that cooperativity is completely determined by the sign of  $B_1$ : positive for  $B_1 < 0$  and negative for  $B_1 > 0$ . This result is also reflected in eq 6, where negative (positive)  $B_1$  increases (decreases) the MM turnover rate  $(A_0 + B_0/[S])^{-1}$ . The dashed line in Figure 2a shows that a largely negative  $B_1 + B_0$  leads to substrate inhibition behavior. The cNESS substrate inhibition shows positive cooperativity at low substrate concentrations, and then the turnover rate decreases to a nonzero value  $A_0^{-1}$ in the substrate-saturation limit. Next, we plot the phase diagram of enzyme kinetics for the single-loop model in Figure 2b, which only depends on  $B_0$  and  $B_1$ . From this phase diagram,  $\alpha = B_1/B_0$  is defined as a unique non-MM indicator for single-loop systems, with negative cooperativity for  $\alpha > 0$ , positive cooperativity for  $-1 \le \alpha < 0$ , and substrate inhibition for  $\alpha < -1$ .



Figure 3: (a)–(c) Three cases in which a current  $J_{1,E}$  circulating counterclockwise within a two-conformation loop can be modulated by  $\Delta \overline{\Delta \tau^{\text{eff}}}$  (see text for details); such modulation underlies the emergence of kinetic cooperativity. In each conformational channel, a horizontal arrow proceeds from the state with the faster effective characteristic residence time (see text for details) to the state with the slower one, with  $J_{1,E}$  superimposed onto this view. Note that there are also analogous cases for  $J_{1,E}$  proceeding in the clockwise direction.

The direction of a conformational current alone does not predict its influence on the cooperativity, which raises the question of how currents are modulated to govern cooperative behavior. For the two-conformation network, the simplest single-loop model, we can rewrite the non-MM term as

$$\frac{B_1}{[S] + s_1} \propto \Delta \overline{\Delta \tau^{\text{eff}}} \times J_{1,\text{E}}([S])$$
(7)

where  $\Delta \overline{\Delta \tau^{\text{eff}}} = \overline{\Delta \tau^{\text{eff}}}_1 - \overline{\Delta \tau^{\text{eff}}}_2$ , with  $\overline{\Delta \tau^{\text{eff}}}_i = \overline{\tau^{\text{eff}}_{\text{E}_i}} - \tau^{\text{eff}}_{\text{ES}_i}$ . Here, the  $E_i$  and  $\text{ES}_i$  residence times in the decomposed representation, each independent of the non-MM term [and thus

of  $J_{1,\mathrm{E}}([\mathrm{S}])]$ , are given by  $\tau_{\mathrm{E}_i}^{\mathrm{eff}}([\mathrm{S}]) = K_{i,\mathrm{M}}^{\mathrm{eff}}/(k_{i,2}^{\mathrm{eff}}[\mathrm{S}])$  and  $\tau_{\mathrm{ES}_i}^{\mathrm{eff}} = 1/k_{i,2}^{\mathrm{eff}}$ , respectively. Also,  $\overline{\tau_{\mathbf{E}_i}^{\text{eff}}} = \tau_{\mathbf{E}_i}^{\text{eff}}([\mathbf{S}] = s_1)$ , where  $s_1$  is the value of  $[\mathbf{S}]$  at which  $|J_{1,\mathbf{E}}([\mathbf{S}])|$  is at half its maximum and thus represents a characteristic non-MM substrate concentration. Therefore,  $\overline{\tau_{E_i}^{eff}}$ represents a characteristic value of  $\tau_{\mathbf{E}_i}^{\mathrm{eff}}([\mathbf{S}])$ , with corresponding characteristic residence time gradient  $\overline{\Delta \tau^{\text{eff}}}_i$ . Thus,  $\Delta \overline{\Delta \tau^{\text{eff}}}$  represents the difference in characteristic residence time gradient between the two decomposed conformational channels. Cooperativity depends upon  $J_{1,E}$ modulated by  $\Delta \overline{\Delta \tau^{\text{eff}}}$ , i.e., it is governed by the relative modulation of the current between the two decomposed chain reactions. In the two-conformation model,  $J_{1,\mathrm{E}}$  proceeds from  $E_i$  to  $ES_i$  in one conformational channel and from  $ES_i$  to  $E_i$  in the other, as illustrated in Figure 3a–c for a counterclockwise current, which corresponds to  $J_{1,E} > 0$  based upon our original definition of  $J_{i,j}$ . In each two-state chain reaction, enzyme turnover is accelerated (decelerated) when  $J_{1,\mathrm{E}}$  proceeds from the state with the slower (faster) effective characteristic residence time to the state with the faster (slower) one. In Figure 3a (b), turnover is accelerated (decelerated) in both chain reactions, resulting in overall turnover acceleration (deceleration), i.e., positive cooperativity or substrate inhibition (negative cooperativity). In Figure 3c, turnover is accelerated in conformation 1 and decelerated in conformation 2 (the opposite [not shown] is possible as well), with the cooperativity depending upon the relative modulation of the current between the two decomposed chains. Kinetic cooperativity is thus explained as follows: when  $J_{1,E}$  proceeds in the direction that, on average, corresponds to decreasing (increasing) effective characteristic residence time, positive cooperativity or substrate inhibition (negative cooperativity) occurs.

Interestingly, when the effective characteristic residence time gradient is conformation invariant, modulation of the conformational current is balanced, resulting in MM kinetics, even in the presence of circulating current (i.e., when  $\Delta \overline{\Delta \tau^{\text{eff}}} = 0$  and  $J_{1,\text{E}} \neq 0$  for the two-conformation network). This scenario represents a unique type of nonequilibrium symmetry in multidimensional kinetic networks and is not precluded by the satisfaction of the aforementioned constraint resulting from local detailed balance. Additionally, we note that for the  $2 \times 2$  model,  $J_{1,E}$  vanishes under a simple conformational detailed balance condition,

$$\frac{\gamma_{1,1}}{\gamma_{1,2}}K_{1,\mathrm{M}} = \frac{\gamma_{-1,1}}{\gamma_{-1,2}}K_{2,\mathrm{M}}$$
(8)

where  $K_{i,M} = (k_{i,-1} + k_{i,2})/k_{i,1}^0$ . Explicit calculations for this model are provided in the Supporting Information.



Figure 4: Enzyme kinetics for the two-loop model with  $A_0 = B_0 = s_1 = 1$  and  $s_2 = 4$ . (a) Three turnover rates v that are non-monotonic functions of [S]. Each line shows a typical type of non-MM kinetic behavior from the regime labeled by the same number in (b). (b) Phase diagram determined by two non-MM parameters  $B_1$  and  $B_2$ . There are five regimes of non-MM behavior (see text for details).

For the two-loop models with two non-MM terms, the cNESS enzyme kinetics become more complicated, as illustrated in a typical phase diagram in Figure 4b. Except for an unphysical regime where v shows divergence and negativity, five regimes of enzyme kinetics can be characterized in the phase space composed of  $B_1$  and  $B_2$ . Similar to the singleloop model, when monotonically increasing to the maximum value  $v_{\text{max}}$  in the substratesaturation limit ([S]  $\rightarrow \infty$ ), v can exhibit negative (Regime 1) and positive (Regime 2) cooperativity. The separation line between these two kinetic regimes, however, is hard to rigorously define. The dashed separation line in Figure 4b corresponds to  $n_{\text{H}} = 1$ , where the Hill constant is empirically calculated using  $n_{\text{H}} = \log 81/\log([S]_{0.9v_{\text{max}}}/[S]_{0.1v_{\text{max}}})$ ,<sup>19</sup> and  $[S]_v$ is the substrate concentration for v. In Regimes 3-5, the turnover rate v is a non-monotonic function of [S] (examples shown in Figure 4a). In Regime 3, with  $v_{\text{max}}$  occurring at a finite  $[S]_{v_{\text{max}}}$ , the turnover rate exhibits the same substrate inhibition behavior as the single-loop model. Alternatively, an additional local minimum of v can appear at  $[S]_{v_{\min}}(> [S]_{v_{\max}})$ , and v increases at high substrate concentrations instead. Two examples are shown by the dashed and dotted lines in Figure 4a. Based on a criterion whether the global  $v_{\max}$  appears as  $[S] \rightarrow \infty$  or at the finite  $[S]_{v_{\max}}$ , this non-MM kinetic behavior is further divided into Regimes 4 and 5, respectively.

For the generalized  $N_c$ -loop model, cNESS enzyme kinetics can be similarly analyzed using reduced parameters from eq 6. In the case that all the non-MM parameters  $B_n$  are positive (negative), the turnover rate exhibits negative cooperativity (positive cooperativity or substrate inhibition). With the coexistence of positive and negative non-MM parameters, cooperativity can be qualitatively determined by the small-[S] expansion of the turnover rate in eq 6,  $v \sim B_0^{-1}[S] - (A_0B_0^{-2} + \sum_{n=1}^{N_c} B_n s_n^{-1} B_0^{-1})[S]^2 + O([S]^3)$ . For a largely negative  $\sum_n B_n/s_n$ , the positive quadratic [S] term dominates in v, resulting in positive cooperativity. When this summation becomes largely positive, the cancellation between linear and nonlinear terms can slow down the increase of v with [S], inducing negative cooperativity. The sign of  $\sum_n B_n/s_n$  is thus a qualitative indicator of cooperativity. To investigate the substrate inhibition behavior, we expand v in the substrate-saturation limit as  $v \sim A_0^{-1} - (B_0 + \sum_n B_n)A_0^{-2}[S]^{-1} + O([S]^{-2})$ . For  $\sum_n B_n < -B_0$ , v is a decreasing function of [S], and the maximum turnover rate  $v_{\text{max}}$  must appear at a finite [S]. The investigation of other types of non-monotonic behavior for v needs the explicit rate form in eq 6.

In summary, we study cNESS enzyme kinetics induced by population currents from conformational dynamics. Applying the flux balance method to a discrete  $N \times M$  kinetic model, we derive a generalized Michaelis–Menten equation to predict the [S] dependence of the turnover rate. Using reduced non-MM parameters,  $B_n$  in eq 6, our generalized MM equation provides a systematic approach to explore cNESS enzyme kinetics. Compared to the typical rate matrix approach, our flux method characterizes non-MM enzyme kinetics in a much simpler way. For example, a unique kinetic indicator  $\alpha = B_1/B_0$  is defined for the single-loop model, and phase diagrams are plotted for the single- and two-loop models. Our study can be extended to other important biophysical processes following the MM mechanism, e.g., the movement of molecular motors induced by ATP binding.

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### Supporting Information Available

Four-site, single-loop model calculations

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