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ABSTRACT

Transient bonds between fast linkers and slower particles are widespread in physical and biological systems. Despite their diverse structure and function, a commonality is that the linkers diffuse on timescales much faster compared to the overall motion of the particles they bind to. This limits numerical and theoretical approaches that need to resolve these diverse timescales with high accuracy. Many models, therefore, resort to effective, yet *ad hoc*, dynamics, where linker motion is only accounted for when bound. This paper provides a mathematical justification for such coarse-grained dynamics that preserves detailed balance at equilibrium. Our derivation is based on multiscale averaging techniques and is broadly applicable. We verify our results with simulations on a minimal model of fast linker binding to a slow particle. We show how our framework can be applied to various systems, including those with multiple linkers, stiffening linkers upon binding, or slip bonds with force-dependent unbinding. Importantly, the preservation of detailed balance only sets the ratio of the binding to the unbinding rates, but it does not constrain the detailed expression of binding kinetics. We conclude by discussing how various choices of binding kinetics may affect macroscopic dynamics.

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I. INTRODUCTION

Transient bonds between fast molecules and other slower molecules are found ubiquitously throughout physical systems. Such bonds enable momentum transfer at microscopic scales and are at the root of diverse phenomena, including linkers that tune the material properties of polymers,^{1–4} shape spatial organization and function of biological systems,^{5–10} and determine the rate of tethered chemical reactions.¹¹ Stochastic modeling of transient binding of linkers via numerical or theoretical approaches is, therefore, of widespread interest.

The diversity of phenomena goes hand in hand with a variety of systems and hence of lexical terms: Examples include cross-links in polymer meshes,^{1,2} metal–ligand bonds in self-assembled porous materials,^{12,13} complementary DNA pairs hybridizing between colloids,^{14–21} and fast myosin motors binding to slender actin fibers.^{5,6,22} Henceforth, the word *linker* will refer to any molecule with (a) a binding end that can form a bond with another molecule

and (b) that relaxes *rapidly* to a finite length. In the previous examples, the linker is the polymer linker, ligand, DNA, or myosin molecule. The *bond* will refer to the chemical bond formed between a linker and a slower molecule (or another linker). In the previous examples, the bond is the weak polymer–polymer adhesion, the metal–ligand chemical bond, the hybridized DNA section, or the high-affinity myosin head after ATP hydrolysis.

Despite enormous variations in mechanochemical properties of linking molecules, one unifying feature is their diffusion on characteristic timescales much faster than the overall motion of the fibers or objects they link.^{10,23–25} This rapid diffusion of (often many) linking molecules creates disparate timescales and length scales that must be resolved, often creating a bottleneck for numerical or theoretical investigations.^{12,13,26} To alleviate this, it is common to rely on coarse-grained descriptions of the linkers, for their dynamics, binding kinetics, or both. In these coarse-grained scenarios, cross-linkers are replaced by effective laws so that their detailed dynamics need not be considered directly.^{22,27–36} For example, if a fast linker binds

transiently to a slow particle, a coarse-grained simulation or model would only specify the dynamics of the slow particle; see illustration in Fig. 1. While the proposed effective dynamics may be relatively intuitive in some scenarios, there does not seem to be a systematic procedure for justifying and comparing this coarse-graining across different systems.^{22,29,31} This paper provides a mathematical justification for such dynamics and, more importantly, a framework to derive effective dynamics in various settings.

In establishing the justification for the effective dynamics with fast linkers, several questions arise. Importantly, when linker dynamics are in thermodynamic equilibrium, with binding and unbinding rates that depend on mutual distance, detailed balance has to be enforced.^{24,25,37} An immediate consequence is that the equilibrium distribution of the system is determined by the binding and unbinding rates. As these detailed rates are often unmeasured, a variety of choices are taken in the literature.^{22,27–34} These choices of specific forms of binding and unbinding spatial dependence introduce ambiguity in interpreting the resulting dynamics. We, therefore, briefly discuss choices of kinetic rates and their consequences in the coarse-graining procedure.

The paper is organized as follows: We first consider a minimal system [Fig. 1(b)] made of a single fast linker binding to a slow particle (Sec. II) and catalog the variety of modeling choices at this microscopic level. Next, we rigorously coarse-grain fast linker dynamics to obtain an effective model for the slow particle (Sec. III) that we validate with numerical simulations (Sec. IV). We then show, with three examples, how we can apply our formalism to more complex yet common setups (Sec. V): We investigate (i) two

fast reactive linkers connecting to each other, (ii) a linker that stiffens upon binding, and (iii) a slip bond with force-dependent unbinding. Finally, we discuss how the choice of microscopic kinetic rates can affect the coarse-grained dynamics both at short and long timescales in nontrivial ways (Sec. VI). We hope our framework will help to investigate systems with fast transient cross-linkers more systematically.

II. GENERAL SYSTEM WITH TRANSIENT CROSS-LINKERS

A. Microscopic dynamics

We consider the motion of a relatively slow particle representing, for example, a slender actin filament, a cell, or a colloid.^{22,24,25} The slow particle diffuses, for simplicity, in one spatial dimension [see Fig. 1(b), blue particle] and we will later discuss how to extend the results to 3D. The position of the slow particle at time t is $x_s(t)$. The diffusion coefficient of the particle is $D_s = k_B T / \gamma_s$, where k_B is Boltzmann's constant, T is the temperature, and γ_s is the friction coefficient of the particle. The particle evolves in an external potential $\mathcal{U}_s(x_s)$, which could represent connections with other particles. The forces in the unbound state on the particle are thus simply $\mathcal{F}_u = -\partial_s \mathcal{U}_s(x_s)$.

We also track the motion of a relatively fast linker, for example, a myosin head²² or the sticky ends of a single-stranded DNA filament.^{24,29} The fast linker's position is x_ℓ , and the linker diffuses with diffusion coefficient $D_\ell = k_B T / \gamma_\ell$, where γ_ℓ is the friction coefficient of the fast particle [see Fig. 1(b), pink linker]. Here, the linker is attached to another slow object or an immobile surface; for now, we will consider it connected to a fixed point. This assumption ensures that the binding is localized in space. The linker usually resists extension, as it is made of a polymer or a protein that resists uncoiling.^{23,24} It is hence reasonable to assume the linker is submitted to a recoil force, $-k_\ell(x_\ell - x_{\ell,0})$, where k is a spring constant,³⁸ and $x_{\ell,0}$ is the rest length of the linker. Note that, as long as the force is conservative, extending our approach to other expressions is straightforward.

The unbound dynamics are

$$\begin{cases} \frac{dx_s}{dt} = \frac{\mathcal{F}_u(x_s, x_\ell)}{\gamma_s} + \sqrt{\frac{2k_B T}{\gamma_s}} \eta_s(t) \\ = -\frac{\partial_s \mathcal{U}_s(x_s)}{\gamma_s} + \sqrt{\frac{2k_B T}{\gamma_s}} \eta_s(t), \\ \frac{dx_\ell}{dt} = -\frac{k_\ell}{\gamma_\ell} (x_\ell - x_{\ell,0}) + \sqrt{\frac{2k_B T}{\gamma_\ell}} \eta_\ell(t), \end{cases} \quad (1)$$

where the $\eta_i(t)$ are uncorrelated Gaussian white noises, where $\langle \eta_i(t) \rangle = 0$ and $\langle \eta_i(t) \eta_j(t') \rangle = \delta_{ij} \delta(t - t')$, where δ_{ij} is the Kronecker symbol and $\langle \cdot \rangle$ is an average over realizations of the noise. Without loss of generality, we will shift the domain such that the rest length of the fast variable is at the center of the domain, i.e., $x_{\ell,0} = 0$.

The slow particle may transiently bind to the linker [see Fig. 1(b), orange bond]. In this entire paper, we will consider that when the bond is formed, it corresponds to a stiff spring with spring constant k_b added between the particle and the linker. Hence, in the

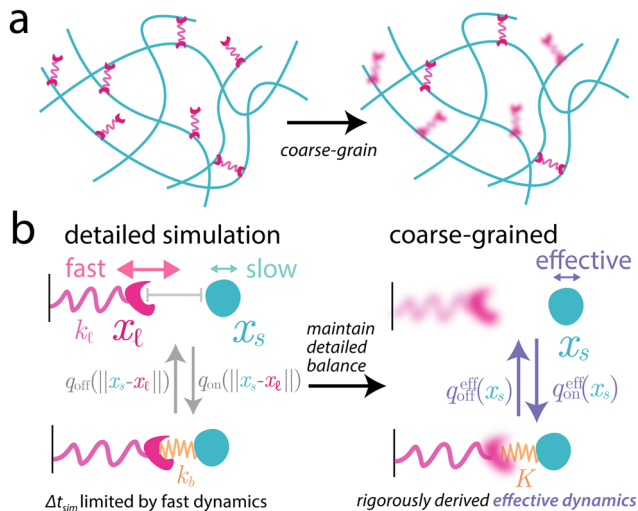


FIG. 1. Coarse-graining principle for a fast linker forming a transient bond. (a) Cartoon of a common coarse-graining procedure where fast linkers (pink) between slow fibers (blue) are modeled only when they are bound. In this work, we study a minimal system where (b) a fast jiggling linker (pink) binds and unbinds rapidly to a binding site on a slow particle with rates q_{on} and q_{off} that depend on the relative distance. The 1D position of the linker is x_ℓ and that of the slow particle x_s . In this paper, we use multiscale averaging techniques to coarse-grain the dynamics of the fast linker while preserving detailed balance. In the coarse-grained model, only the slow particle dynamics are specified, with effective kinetic rates $q_{\text{on}}^{\text{eff}}(x_s)$ and $q_{\text{off}}^{\text{eff}}(x_s)$ and forces that only depend on the position of the slow particle x_s .

bound state, the forces on the slow particle are $\mathcal{F}_b = -k_b(x_s - x_\ell) - \partial_s \mathcal{U}_s(x_s)$. The bound dynamics are thus

$$\begin{cases} \frac{dx_s}{dt} = \frac{\mathcal{F}_b(x_s, x_\ell)}{\gamma_s} + \sqrt{\frac{2k_B T}{\gamma_s}} \eta_s(t) \\ = -\frac{k_b}{\gamma_s}(x_s - x_\ell) - \frac{\partial_s \mathcal{U}_s(x_s)}{\gamma_s} + \sqrt{\frac{2k_B T}{\gamma_s}} \eta_s(t), \\ \frac{dx_\ell}{dt} = \frac{k_b}{\gamma_\ell}(x_s - x_\ell) - \frac{k_\ell}{\gamma_\ell}(x_\ell) + \sqrt{\frac{2k_B T}{\gamma_\ell}} \eta_\ell(t). \end{cases} \quad (2)$$

Our model for the bound dynamics is not unique. For example, one could consider that instead of a stiff spring, the bond formed is a rigid rod constraining the dynamics.^{24,32,33,39} Yet, we choose this spring bond model as it is a simple starting point and because it is physically satisfying. Indeed, with the spring model, the linker and the particle relax toward each other, while in the rigid rod model, they can stay unnaturally far apart. Ultimately, we will consider the dynamics in the limit where the bond is very stiff, corresponding to a so-called *soft constraint*.^{40–42}

The equilibrium distribution corresponding to these choices of dynamics can be decomposed over the two states (bound and unbound) as $\pi = (\pi_u, \pi_b)^T$. The component of the equilibrium distribution corresponding to the unbound state is

$$\pi_u(x_s, x_\ell) = \frac{1}{Z_u} e^{-\mathcal{U}_s(x_s)/k_B T - k_\ell(x_\ell)^2/2k_B T} \quad (3)$$

and the bound one is

$$\pi_b(x_s, x_\ell) = \frac{1}{Z_b} e^{-k_b(x_s - x_\ell)^2/2k_B T - \mathcal{U}_s(x_s)/k_B T - k_\ell(x_\ell)^2/2k_B T}, \quad (4)$$

where Z_u and Z_b are constant prefactors that are set by a normalization condition on the total equilibrium distribution $\int dx_\ell dx_s (\pi_u + \pi_b) = 1$ and by detailed balance, which we turn to now.

B. Possible kinetic rates and detailed balance

We consider that the linker and the particle bind to each other with rate q_{on} and unbind with rate q_{off} . To be physically accurate, it is reasonable to assume that both rates may depend on the spatial variables (x_s, x_ℓ) , *a priori*. While the exact expression of the rates for our coarse-graining approach does not matter, it is crucial to recall how these rates should be specified to satisfy detailed balance.

If the system we consider is at equilibrium, the rates *must satisfy detailed balance*.^{24,37} Here, this means the probability flux at equilibrium of going from one state to the other is equal to the inverse flux, i.e.,

$$\pi_u(x_s, x_\ell) q_{\text{on}}(x_s, x_\ell) = \pi_b(x_s, x_\ell) q_{\text{off}}(x_s, x_\ell). \quad (5)$$

Here, this relation simplifies to

$$\frac{q_{\text{on}}}{q_{\text{off}}} = \frac{Z_u}{Z_b} e^{-k_b(x_s - x_\ell)^2/2k_B T}. \quad (6)$$

To make this expression more explicit, we can redefine the constants $Z_u = Z$ and $Z_b = Z q_{\text{off}}^0/q_{\text{on}}^0$, where Z is a global normalization constant such that $\int dx_\ell dx_s (\pi_u + \pi_b) = 1$. Here, q_{off}^0 and q_{on}^0 set the

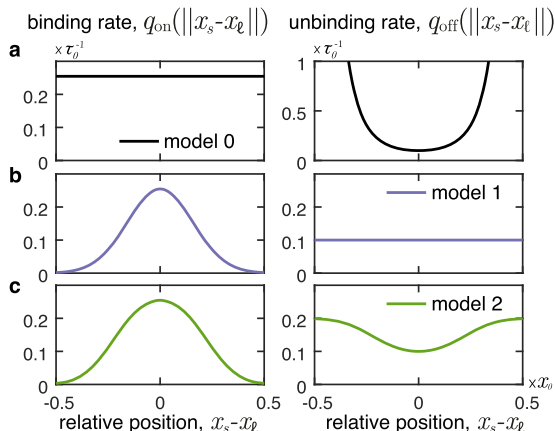


FIG. 2. Possible choices of binding and unbinding kinetics agreeing with detailed balance. (a) Constant binding rate, (b) constant unbinding rate, and (c) both rates varying; see text for detailed expressions. Here, we chose a bond spring constant $k_b L^2/k_B T = 40$ and a fixed macroscopic probability $\Pi_b = 0.5$.

typical range of the kinetic rates and are related via the typical free energy of bond formation E_0 such that $q_{\text{on}}^0/q_{\text{off}}^0 \equiv e^{-E_0/k_B T}$. We obtain

$$\frac{q_{\text{on}}}{q_{\text{off}}} = \frac{q_{\text{on}}^0}{q_{\text{off}}^0} e^{-k_b(x_s - x_\ell)^2/2k_B T}. \quad (7)$$

Since this is the only relation that constrains q_{on} and q_{off} , it is clear that *the choice of q_{on} and q_{off} is not unique* and that at least one of the rates has to *depend on space*.

1. Possible expressions of the rates

Therefore, there are infinite possibilities in how we can specify rates consistent with detailed balance, reflecting the diversity of choices made in the literature.^{22,25,27–34,36,43–45} However, we can catalog a few simple commonly chosen examples, see Fig. 2, and discuss to what extent they are consistent with physical intuition.

- Model 0: Unbinding is faster further away from the target, but the binding rate is constant,

$$\begin{cases} q_{\text{on}} = q_{\text{on}}^0, \\ q_{\text{off}}(x_s, x_\ell) = q_{\text{off}}^0 e^{k_b(x_s - x_\ell + \ell_b)^2/2k_B T}. \end{cases} \quad (8)$$

This binding term provides the convenient feature of avoiding the need to resolve detailed dynamics of the fast linker. Furthermore, the unbinding form is motivated by the intuitive observation that bonds break faster when a larger force is exerted. One such example is molecular motor detachment,^{46–48} although this is often modeled as a slip bond, which we discuss in Sec. V.

However, we note that this choice suffers from the numerically undesirable feature of the off rate increasing exponentially as a function of distance.

- Model 1: Binding is faster closer to the target, but unbinding is constant,

$$\begin{cases} q_{\text{on}}(x_s, x_\ell) = q_{\text{on}}^0 e^{-k_b(x_s - x_\ell)^2 / 2k_B T}, \\ q_{\text{off}} = q_{\text{off}}^0, \end{cases} \quad (9)$$

which is a model used in Refs. 28, 31, and 36. Constant unbinding seems rather unphysical since one would expect the bond to break if the linker and the particle are brought too far apart. We can instead opt for an expression where both rates depend on space, for example,

- model 2, where binding is faster closer to the target, unbinding is faster further away from the target,

$$\begin{cases} q_{\text{on}}(x_s, x_\ell) = \frac{q_{\text{on}}^0}{1 + e^{k_b(x_s - x_\ell)^2 / 2k_B T}}, \\ q_{\text{off}}(x_s, x_\ell) = \frac{q_{\text{off}}^0}{1 + e^{-k_b(x_s - x_\ell)^2 / 2k_B T}}, \end{cases} \quad (10)$$

which solves the issues raised above.

While these are possible choices of the binding rates that agree with detailed balance, these are not the only ones, and one could consider various other kinetics, possibly with a kinetic barrier to overcome to unbind or to bind.

One might then wonder if the choice of the microscopic kinetic rates affects the long-time dynamics of the system. We raise the question here but we will not attempt to answer it thoroughly. Rather, our goal is to show how to systematically coarse-grain the dynamics of the fast linker once the microscopic dynamics are properly chosen.

2. Binding models with uniform kinetic rates

Inspired by a previous model,⁴⁹ we also discuss an alternative binding scheme. Specifically, if one wanted the binding and unbinding rates to be uniform, at least over a certain length scale, the *only* option is that the dynamics are not specified via Eqs. (1) and (2) but have to be changed. We illustrate this point briefly below.

For example, one could define uniform rates, as

$$\begin{cases} q_{\text{on}} = q_{\text{on}}^0, \\ q_{\text{off}} = q_{\text{off}}^0, \end{cases} \quad (11)$$

where the rates q_{on}^0 and q_{off}^0 are nonzero. In that case, the bound π_b and unbound π_u parts of the equilibrium distribution must be the same, which constrains the forces to be the same in each state. Otherwise, detailed balance is broken. For example, the bound and unbound dynamics could both satisfy

$$\begin{cases} \frac{dx_s}{dt} = -\frac{k_b}{\gamma_s}(x_s - x_\ell) - \frac{\partial_x \mathcal{U}_s(x_s)}{\gamma_s} + \sqrt{\frac{2k_B T}{\gamma_s}} \eta_s(t), \\ \frac{dx_\ell}{dt} = \frac{k_b}{\gamma_\ell}(x_s - x_\ell) - \frac{k_\ell}{\gamma_\ell}(x_\ell) + \sqrt{\frac{2k_B T}{\gamma_\ell}} \eta_\ell(t). \end{cases} \quad (12)$$

As a result, the dynamics are not very interesting as they simply amount to a global quadratic confining potential with spring constant k_b .

Tethered dynamics are more in line with physical intuition if the confining potential associated with the bond is restricted to a patch in space. Following Refs. 49 and 50, this can be achieved by defining kinetic rates that are nonzero only when the linker and the particle are close, typically

$$\begin{cases} q_{\text{on}} = q_{\text{on}}^0 H(|x_s - x_\ell| < R), \\ q_{\text{off}} = q_{\text{off}}^0 H(|x_s - x_\ell| < R), \end{cases} \quad (13)$$

where R is a characteristic distance that sets a reasonable maximum binding distance.^{32,33} One can define $R = \sqrt{2k_B T / k_b}$, where k_b is the spring constant of the bond and R corresponds, therefore, to thermal vibrations around that bond. When the linker and the particle are nearby $|x_s - x_\ell| < R$, the dynamics follow Eq. (12) while they are given by Eqs. (1) and (2) otherwise. This ensures that detailed balance is satisfied in both regions, since when $|x_s - x_\ell| \geq R$ the kinetic rates are zero. We refer to models with uniform kinetic rates as in Eq. (13) as Doi models^{37,51,52} in analogy with reaction models without linkers.

However, such uniform kinetic rates are unphysical in two ways: (i) Bond dissociation is not allowed when the bond is stretched beyond R , which seems rather unlikely; and (ii) when they are close, the interaction between the linker and particle is the same regardless of the state of the bond—see Eq. (12), which also seems rather unlikely.

We will briefly show in Sec. VI how also at the coarse-grained level, important physical differences arise between Doi models with uniform kinetic rates as in Eq. (13) and models with spatially dependent kinetic rates as say in Eq. (10). This means, again, that the choice of kinetic rates affects short-time dynamics at least in nontrivial ways; hence, that is an important assumption of any model.

III. COARSE-GRAINED DYNAMICS WITH FAST LINKERS

Our aim is now to rigorously coarse-grain the linker dynamics to obtain the effective long-time dynamics of the particle. This is useful both to gain analytic insight and to propose consistent and fast simulation schemes. We use multiscale averaging⁵³ to coarse-grain the dynamics, a technique that is broadly used in the field to properly average over short length scales and timescales.^{10,24,27,28,39,47} We will show the approach is valid over a broad range of parameters in Sec. IV. We provide Nomenclature as a summary of the main notations used throughout the discussion.

A. Set up of the dynamics

The set of stochastic Eqs. (1) and (2) defines a Markov process that is conveniently studied via the Kolmogorov backward equation^{53,54} on the functions $f(x_s, x_\ell, t) = (f_u(x_s, x_\ell, t), f_b(x_s, x_\ell, t))^T$ such that

$$\partial_t f = \mathcal{L}f, \quad f(x_s, x_\ell, 0) = g(x_s, x_\ell), \quad (14)$$

where \mathcal{L} is the generator of the system and g is any scalar function. Here, the functions $f(x_s, x_\ell, t) = \int p(x'_s, x'_\ell, t | x_s, x_\ell) g(x'_s, x'_\ell) dx'_s dx'_\ell$ give the expectation of any scalar function $g(x_s(t), x_\ell(t))$, given an initial condition $x_s(0) = x_s, x_\ell(0) = x_\ell$, where p is the probability

density that the system evolved from the initial condition to (x'_s, x'_ℓ) at time t . Once we know how such functions f evolve, we may calculate any statistic g of our stochastic process.

The generator \mathcal{L} can be calculated from the forward equation, the Fokker–Planck equation, associated with the dynamics Eqs. (1) and (2), see the supplementary material, Sec. 1, for further details. The full generator can be written as

$$\mathcal{L} = \mathcal{Q} + \mathcal{V}, \quad (15)$$

where

$$\mathcal{Q} = \begin{pmatrix} -q_{\text{on}} & q_{\text{on}} \\ q_{\text{off}} & -q_{\text{off}} \end{pmatrix} \text{ and } \mathcal{V} = \begin{pmatrix} \mathcal{V}_u & 0 \\ 0 & \mathcal{V}_b \end{pmatrix},$$

with

$$\begin{aligned} \mathcal{V}_u &= -\frac{k_\ell}{\gamma_\ell} x_\ell \partial_\ell + \frac{k_B T}{\gamma_\ell} \partial_{\ell\ell} - \frac{\partial_x \mathcal{U}(x_s)}{\gamma_s} x_s \partial_s + \frac{k_B T}{\gamma_s} \partial_{ss}, \\ \mathcal{V}_b &= -\frac{k_\ell}{\gamma_\ell} x_\ell \partial_\ell + \frac{k_B T}{\gamma_\ell} \partial_{\ell\ell} - \frac{\partial_x \mathcal{U}(x_s)}{\gamma_s} x_s \partial_s + \frac{k_B T}{\gamma_s} \partial_{ss} \\ &\quad - \frac{k_b}{\gamma_\ell} (x_\ell - x_s) \partial_\ell - \frac{k_b}{\gamma_s} (x_s - x_\ell) \partial_s. \end{aligned}$$

B. Nondimensionalization

To highlight the ratio between the different temporal and spatial scales at play, we non-dimensionalize our equations. Since the particle's motion is much slower than that of the linker, we can identify a small number $\varepsilon = \gamma_\ell/\gamma_s$. We seek the dynamics of the slow particle over a typical long timescale τ_0 such that its motion is diffusive, extending over the range $x_0 = \sqrt{D_0 \tau_0}$. We now can call $q'_{\text{on}} = q_{\text{on}} \tau_0$ and similarly $q'_{\text{off}} = q_{\text{off}} \tau_0$, remembering that these are both functions of space. We consider for now that we observe the dynamics over timescales τ_0 where transient binding and unbinding are still relevant such that $q'_{\text{on}} = O(1)$. Typical slow dynamics are captured by the timescale τ_0 ; so, again we consider $\kappa = k_\ell \tau_0/\gamma_s = O(1)$. We also write $\lambda = k_b \tau_0/\gamma_s = O(1)$ for now. Since $\varepsilon = \gamma_\ell/\gamma_s$, we have $k_\ell \tau_0/\gamma_\ell = k_\ell \tau_0/\gamma_s \varepsilon$.

In this scaling, we have the nondimensional generator $\mathcal{L}' = \frac{1}{\varepsilon} \mathcal{L}'_0 + \mathcal{L}'_1$ such that

$$\mathcal{L}'_0 = \begin{pmatrix} -\kappa x_\ell \partial_\ell + \partial_{\ell\ell} & 0 \\ 0 & -\kappa x_\ell \partial_\ell - \lambda(x_\ell - x_s) \partial_\ell + \partial_{\ell\ell} \end{pmatrix}$$

and

$$\mathcal{L}'_1 = \begin{pmatrix} -q'_{\text{on}} - \frac{\partial_s \mathcal{U}(x_s)}{k_B T} \partial_s + \partial_{ss} & q'_{\text{on}} \\ q'_{\text{off}} & -q'_{\text{off}} - \lambda(x_\ell - x_s) \partial_s - \frac{\partial_s \mathcal{U}(x_s)}{k_B T} \partial_s + \partial_{ss} \end{pmatrix}.$$

In the following, we will drop the $'$ notations for simplicity.

C. Averaging procedure

We then look for a solution to $\partial_t f = \mathcal{L}f$ of the form $f = f_0 + \varepsilon f_1 + \varepsilon^2 f_2$. To first order, we have

$$\mathcal{L}_0 f_0 = \begin{pmatrix} -\kappa x_\ell \partial_\ell + \partial_{\ell\ell} & 0 \\ 0 & -\kappa x_\ell \partial_\ell - \lambda(x_\ell - x_s) \partial_\ell + \partial_{\ell\ell} \end{pmatrix} f_0 = 0. \quad (16)$$

Notice that $-\kappa x_\ell - \lambda(x_\ell - x_s) = -(\kappa + \lambda)(x_\ell - x_s \frac{\lambda}{\kappa + \lambda})$ such that the general solution to Eq. (16) is

$$\begin{aligned} f_0 &= \begin{pmatrix} g_1(x_s, t) \\ g_2(x_s, t) \end{pmatrix} + \begin{pmatrix} h_1(x_s, t) \\ 0 \end{pmatrix} \int_0^{x_\ell} e^{+\kappa x^2} dx \\ &\quad + \begin{pmatrix} 0 \\ h_2(x_s, t) \end{pmatrix} \int_0^{x_\ell} e^{+(\kappa + \lambda)(x - \frac{\lambda}{\kappa + \lambda} x_s)^2} dx \end{aligned} \quad (17)$$

and for which the associated equilibrium distribution is of the form

$$\pi_0 = \frac{1}{\sqrt{2\pi}} \begin{pmatrix} \alpha e^{-\kappa x_\ell^2/2} \\ \beta e^{-(\kappa + \lambda)(x_\ell - \frac{\lambda}{\kappa + \lambda} x_s)^2/2} \end{pmatrix}. \quad (18)$$

Here, α and β are free parameters and are not constrained by detailed balance; they can be any real number. In fact, there is no exchange between the bound and the unbound times at these very short timescales. Notice that this means that the averaging approach does not know *a priori* that it should preserve detailed balance.

In any case, we require $\langle f_0, \pi_0 \rangle_\ell$ to be bound; $\langle f, p \rangle_\ell$ is the inner product where the integration is only carried over the fast variable x_ℓ . This imposes $h_1 \equiv 0$ and $h_2 \equiv 0$, and $f_0 = (g_1(x_s, t), g_2(x_s, t))^T$ is actually independent of the fast variable.

Seeking the next order $\mathcal{L}_0 f_1 = -\mathcal{L}_1 f_0 + \partial_t f_0$. A solution exists for f_1 if the Fredholm alternative is satisfied,⁵³ i.e., if $\langle (\partial_t f_0 - \mathcal{L}_1 f_0), \pi_0 \rangle_\ell = 0$ is true for any π_0 in the null space of \mathcal{L}'_0 . This corresponds to any real value combination of α and β ; hence, we may pick the convenient choice of $(\alpha, \beta) = (1, 0)$ and $(\alpha, \beta) = (0, 1)$. We obtain

$$\begin{aligned} \partial_t g_1 &= -\frac{\int e^{-\kappa x_\ell^2/2} q_{\text{on}} dx_\ell}{\int e^{-\kappa x_\ell^2/2} dx_\ell} (g_1 - g_2) - \frac{\partial_s \mathcal{U}(x_s)}{k_B T} \partial_s g_1 + \partial_{ss} g_1, \\ \partial_t g_2 &= \frac{\int e^{-(\kappa + \lambda)(x_\ell - \frac{\lambda}{\kappa + \lambda} x_s)^2/2} q_{\text{off}} dx_\ell}{\int e^{-(\kappa + \lambda)(x_\ell - \frac{\lambda}{\kappa + \lambda} x_s)^2/2} dx_\ell} (g_1 - g_2) \\ &\quad - \frac{\partial_s \mathcal{U}(x_s)}{k_B T} \partial_s g_2 + \partial_{ss} g_2 \\ &\quad - \frac{\lambda \left(\int (x_\ell - x_s) e^{-(\kappa + \lambda)(x_\ell - \frac{\lambda}{\kappa + \lambda} x_s)^2/2} dx_\ell \right)}{\int e^{-(\kappa + \lambda)(x_\ell - \frac{\lambda}{\kappa + \lambda} x_s)^2/2} dx_\ell} \partial_s g_2. \end{aligned} \quad (19)$$

D. Effective kinetic rates and dynamics

We have just obtained the coarse-grained backward equations for the dynamics.

We can carry over the last line's integral and return to dimensional units, to directly read off the effective on and off rates as follows:

$$\begin{cases} q_{\text{on}}^{\text{eff}}(x_s) = \frac{\int e^{-k_\ell x_\ell^2/2k_B T} q_{\text{on}}(x_\ell, x_s) dx_\ell}{\int e^{-k_\ell x_\ell^2/2k_B T} dx_\ell}, \\ q_{\text{off}}^{\text{eff}}(x_s) = \frac{\int e^{-(k_\ell+k_b)\left(x_\ell - \frac{k_b}{k_\ell+k_b}x_s\right)^2/2k_B T} q_{\text{off}}(x_\ell, x_s) dx_\ell}{\int e^{-(k_\ell+k_b)\left(x_\ell - \frac{k_b}{k_\ell+k_b}x_s\right)^2/2k_B T} dx_\ell}. \end{cases} \quad (20)$$

We find that, coherently, the effective on and off rates are weighted averages over the distribution of positions of the fast variable in the respective states. A similar expression for the coarse-grained rates was obtained from a first principles derivation at equilibrium.²⁵

Importantly, beyond the coarse-grained rates, from Eq. (19) we can also read off the *coarse-grained dynamics* of the slow particle at $O(1)$ in the small parameter $\varepsilon = \gamma_\ell/\gamma_s$, in the unbound and bound states given by

$$\begin{cases} \frac{dx_s}{dt} = -\frac{\partial_s \mathcal{U}(x_s)}{\gamma_s} + \sqrt{\frac{2k_B T}{\gamma_s}} \eta_u(t) \quad (\text{unbound}), \\ \frac{dx_s}{dt} = -\frac{\partial_s \mathcal{U}(x_s)}{\gamma_s} - \frac{K}{\gamma_s} x_s + \sqrt{\frac{2k_B T}{\gamma_s}} \eta_b(t) \quad (\text{bound}). \end{cases} \quad (21)$$

Here, $K = k_\ell k_b / (k_\ell + k_b)$ is an effective spring constant. In the [Appendix](#), we verify that this coarse-graining maintains detailed balance at the macroscopic level.

Overall, the coarse-grained dynamics we obtain are consistent with physical intuition, at this lowest order in ε . The unbound dynamics of the slow particle are not changed by the coarse-grained approach. However, in the bound state, an extra recoil force is exerted on the particle, corresponding to a force averaged over the fast-moving linker. Interestingly, the spring constant K of the bond formed at the coarse-grained level corresponds to the effective spring constant corresponding to two springs in series, with spring constants k_ℓ and k_b . This is precisely what is expected physically and from the sketch of the setup; see [Fig. 1\(b\)](#). One novelty of this calculation is that it allows one to give meaning to the spring constant K used in effective models such as those used in [Refs. 28, 31, and 36](#).

Both in the bound and unbound states, the coarse-grained dynamics are damped by the friction coefficient γ_s . Although this is rather intuitive, notice again that this is only true when ε is small enough, otherwise the effective friction would be increased by the presence of the linker.^{24,27} All in all, the lowest-order dynamics evolve as if the linker were moving so fast that it loses memory of previous binding and unbinding events.

Beyond these $O(1)$ terms in the separation of scales $\varepsilon = \gamma_\ell/\gamma_s$, we can derive further terms at $O(\varepsilon)$ and beyond, which do account for more and more memory between binding events. This can be done by proceeding with the coarse-graining approach explained above to further terms in the expansion. We report in the supplementary material, Sec. 2, the coarse-grained dynamics at $O(\varepsilon)$. The effect of memory is twofold: It modifies the binding rates now containing $q_{\text{on}} q_{\text{off}}$, q_{on}^2 , and q_{off}^2 couplings; and it also modifies the forces. In particular, at $O(\varepsilon)$ there is now a force in the unbound state, which arises from a remnant memory of the recoil force on the bound fast linker as the linker unbinds. Finally, at $O(\varepsilon)$, diffusion in the bound state is now damped by the presence of the linker. Our coarse-graining approach is thus a robust tool to systematically derive coarse-grained dynamics to any order.

E. General coarse-grained dynamics

The averaging approach can be extended in a straightforward way, following the averaging steps above, to more arbitrary dynamics, and we summarize effective dynamics in full generality in [Eqs. \(22\)–\(26\)](#). The initial forces on the slow particle in the bound $\mathcal{F}_b(x_s, x_\ell)$ and the unbound $\mathcal{F}_u(x_s, x_\ell)$ states can be arbitrary forces. All forces on the particle and linker should be conservative so as to define an equilibrium distribution. [Eqs. \(22\)–\(26\)](#) then provide the general formulas for the effective dynamics, effective binding and unbinding rates as well as the effective force on the particle in the unbound and bound states. The diffusive part of the slow particle motion is not affected by the coarse-graining procedure since we assume diffusion coefficients do not depend on space. Finally, the formulas can be extended in a straightforward way to 3D dynamics and to multiple fast linkers. We will show in [Sec. V](#) how to use these formulas with specific examples:

$$\text{Effective dynamics} \quad \frac{dx_s}{dt} = \frac{1}{\gamma_s} \mathcal{F}_{u/b}^{\text{eff}}(x_s) + \sqrt{\frac{2k_B T}{\gamma_s}} \eta_{u/b}(t), \quad (22)$$

$$\text{Effective binding rate} \quad q_{\text{on}}^{\text{eff}}(x_s) = \frac{\int q_{\text{on}}(x_s, x_\ell) \pi_u(x_s, x_\ell) dx_\ell}{\int \pi_u(x_s, x_\ell) dx_\ell}, \quad (23)$$

$$\text{Effective unbinding rate} \quad q_{\text{off}}^{\text{eff}}(x_s) = \frac{\int q_{\text{off}}(x_s, x_\ell) \pi_b(x_s, x_\ell) dx_\ell}{\int \pi_b(x_s, x_\ell) dx_\ell}, \quad (24)$$

$$\text{Effective force on the slow unbound particle} \quad \mathcal{F}_u^{\text{eff}}(x_s) = \frac{\int \mathcal{F}_u(x_\ell, x_s) \pi_u(x_s, x_\ell) dx_\ell}{\int \pi_u(x_s, x_\ell) dx_\ell}, \quad (25)$$

$$\text{Effective force on the slow bound particle} \quad \mathcal{F}_b^{\text{eff}}(x_s) = \frac{\int \mathcal{F}_b(x_\ell, x_s) \pi_b(x_s, x_\ell) dx_\ell}{\int \pi_b(x_s, x_\ell) dx_\ell}. \quad (26)$$

Notice how each formula from Eqs. (22)–(26) can be interpreted intuitively. The effective force in a given state or the rate of switching from that state is simply the spatial average weighted by the local probability distribution to be in that state. While this seems consistent *a posteriori*, and consistent with detailed balance at the coarse-grained level, the formulas in Eqs. (23) and (24) are not the only possible expressions for the effective rates that obey detailed balance and accurately describe the dynamics at lowest order in ε . The results of Eqs. (22)–(26) are therefore not trivial. Notice that in a related work, a similar expression for the coarse-grained rates $q_{\text{off}}^{\text{eff}}$ and $q_{\text{on}}^{\text{eff}}$ was obtained using averaging techniques [Eqs. (6) or (19) of Ref. 25]. However, a crucial addition here is that we provide also the coarse-grained dynamics of the slow particle, through the effective equations of motion Eq. (22) and effective forces Eqs. (25) and (26).

Interestingly, the expressions in Eqs. (22)–(26) are straightforward to compute. One simply needs to know the equilibrium probability distribution of the free and bound linkers to obtain the effective dynamics. Specifically, one only needs to know the dependence of the equilibrium distribution on the linker coordinate x_ℓ and not the details of the landscape for the slow particle x_s . In practice, one could then simulate (or calculate) a free and bound linker, obtain their local probability distributions, and integrate them to obtain the effective dynamics. We note the caveat that the full probability distribution needs to be evaluated if direct interactions between linkers exist.

F. Limit regimes

We will now comment on the effective dynamics and kinetic rates obtained for the minimal system of Fig. 1(b) [Eqs. (1) and (2)] in a few limiting regimes of the system parameters k_ℓ and k_b . In the following, it will be useful to specify the expression of the rates and we will use for example Eq. (9).

In all limiting regimes, some variables such as $q_{\text{on}}^0/q_{\text{off}}^0$ have to be specified with respect to the limits. In practice, the macroscopic probability of being bound, i.e., Π_b (or unbound, Π_u) can be measured experimentally^{24,55} and is easier to probe than spatially dependent kinetic rates. Hence, we constrain the different functional forms for the kinetic rates to predict the same Π_b . The macroscopic probability of being bound (respectively unbound) is $\Pi_b = \int dx_s dx_\ell \pi_b(x_s, x_\ell)$ [respectively, $\Pi_u = \int dx_s dx_\ell \pi_u(x_s, x_\ell)$]. It is easy to show that here

$$\frac{\Pi_b}{\Pi_u} = e^{-E_0/k_B T} \sqrt{\frac{K}{k_b}} \frac{\int e^{-\mathcal{U}_s(x_s)/k_B T} e^{-k_{\text{eff}} x_s^2/2k_B T} dx_s}{\int e^{-\mathcal{U}_s(x_s)/k_B T} dx_s} \quad (27)$$

such that the macroscopic bound and unbound probabilities measure in particular the energy of the bond, weighted by the geometry of the system. Of course, this does not constrain the local values of the kinetic rates. Rather, it sets the value of some parameters, here of E_0 .

1. Stiff bond

We first investigate the limit regime where the bond is very stiff, i.e., $k_b \gg k_\ell$. This is the so-called *soft* constraint limit.^{40–42} In that case, we expect the fast and slow particles are constrained to move

synchronously in the bound state. For this limit to make sense, we need, according to Eq. (27), to constrain $q_{\text{on}}^0 \sim \sqrt{k_b} q_{\text{off}}^0$.

The effective force in the bound state converges to $\mathcal{F}_u^{\text{eff}} = -k_\ell x_s - \partial_s \mathcal{U}_s$, i.e., simply to a recoil force exerted with a spring constant corresponding to that of the fast linker (and the force deriving from the external potential). This makes sense since in the limit $k_b \gg k_\ell$, since the springs are in series, we expect the weakest spring to take over and $K \sim k_\ell$.

The effective kinetic rates, according to Eq. (20) [and choosing the kinetic rates as in Eq. (9)], are

$$q_{\text{on}}^{\text{eff}} = q_{\text{on}}^0 \sqrt{\frac{K}{k_b}} e^{-K x_s^2/2k_B T}, \quad q_{\text{off}}^{\text{eff}} = q_{\text{off}}^0 \quad (28)$$

In the limit regime where $k_b \gg k_\ell$, we have $q_{\text{on}} \sim q_{\text{off}}^0 e^{-k_\ell x_s^2/2k_B T}$ and $q_{\text{off}} = q_{\text{off}}^0$. Hence the binding rate remains a function of x_s , and therefore spatially dependent. Qualitatively, the particle is more likely to bind when it is closer to the average linker. It unbinds though at the same rate regardless of the linker position. Similar results may be found with other initial choices of kinetic rates such as with Eqs. (8) or (10).

A few models in the literature actually take the spring constant for the binding kinetics to be the linker's spring constant $K = k_\ell$,^{28,31,36} making the implicit, albeit rather physical, assumption that the bond's spring constant is much stiffer than the linker's. This underlines that the meaning of the parameters in the kinetic rates is underappreciated in the field.

Notice that here we explored a limit regime after taking the limit of fast linker dynamics. This is not an issue here since the limits commute. Indeed, one could initially consider an infinitely stiff bond, and then take the limit of fast linker dynamics, and get the same result—see the supplementary material, Sec. 1, for further details.

2. Stiff linker

When the fast linker is stiff, then $k_\ell \gg k_b$. We then simply have that the force in the bound state is determined by the spring constant of the bond $\mathcal{F}_u^{\text{eff}} = -k_b x_s - \partial_s \mathcal{U}_s$, as expected since now $K \sim k_b$ in that limit. With the same choice of kinetic rates as in Eq. (9), we simply have that the effective on rate is faster near the average linker position $q_{\text{on}}^{\text{eff}} = q_{\text{on}}^0 e^{-\frac{k_b x_s^2}{2k_B T}}$ and the effective off rate is constant. All in all, this is a simple consequence of the dynamics of two springs in series, when one of the spring constants is stiff compared to the other.

IV. VALIDATION OF THE COARSE-GRAINING APPROACH

We now use numerical simulations to test the derived effective forces and kinetic rates and determine the range of parameters over which the averaging procedure is valid.

A. Simulation setup

We simulate the dynamics specified through Eqs. (1) and (2) [with no confining potential on the slow particle $\mathcal{U}(x_s) \equiv 0$]. We use the same nondimensional variables τ_0 and $x_0 = \sqrt{D_0} \tau_0 = \sqrt{k_B T \tau_0} / \gamma_s$ such that the problem is fully characterized by five nondimensional numbers $\varepsilon, \kappa, \lambda, q_{\text{off}}^0$ and Π_b / Π_u . Here, $\varepsilon = \gamma_\ell / \gamma_s$ is

not necessarily small and represents the ratio between friction coefficients; $\kappa = k_\ell \tau_0 / \gamma_s$ is the nondimensional spring constant of the linker; $\lambda = k_b \tau_0 / \gamma_s$ is that of the bond, $q_{\text{off}}^0 = q_{\text{off}}^0 \tau_0$ the nondimensional typical off rate. Finally, the on rate is set by the conservation of macroscopic probability through Eq. (27), so by the ratio Π_b / Π_u , once a choice of functional forms for the kinetic rates has been made. Here, we will keep Eq. (10) as an example.

We discretize the dynamics of both slow and fast particles with a standard Euler–Maruyama scheme with time step Δt until a terminal time of $T = 1000 \tau_0$ with $M = 1000$ simulations for each set of parameters. Initial configurations are sampled with the equilibrium distribution of the system. We impose periodic boundary conditions on the slow particle at a nondimensional distance L so that the slow particle does not escape far away from the domain. The domain size $L = 10x_0$ and $\Delta t = 0.01 \tau_0$ are chosen sufficiently large and small, respectively, such that our results do not depend on the specific value.

To estimate the effective rates from simulation, we note that this is a doubly stochastic Poisson process, or a Cox process,⁵⁶ because the stochastic dynamics of the particle positions drive the stochastic Poisson events of binding and unbinding. While sophisticated methods for inference on Cox intensities exist,⁵⁶ a simple binning approach suffices here. We discretize x_s into bins of width Δx and count the number of occurrences of each event in each bin. For bin i with center x_i , the estimated flux to the other state is then $\hat{f}(x_i) = N_i / (TM\Delta x)$, where $\hat{\cdot}$ notation means estimate and N_i is the number of events (either binding or unbinding) occurring in bin i . Notably, this estimates the macroscopic kinetic rates, for instance, for unbinding $\hat{f}_{\text{off}}^{\text{eff}}(x_s) \approx q_{\text{off}}^{\text{eff}}(x_s) \pi_{\text{on}}^{\text{eff}}(x_s)$. The marginal densities of being bound and unbound, $\pi_{\text{on}}^{\text{eff}}(x_s)$ and $\pi_{\text{off}}^{\text{eff}}(x_s)$, are straightforwardly computed by the fraction of time in each state and bin for x_s . Then, the microscopic rates are estimated by $\hat{q}^{\text{eff}}(x_s) = \hat{f}^{\text{eff}}(x_s) / \hat{\pi}^{\text{eff}}(x_s)$. Finally, the forces $\mathcal{F}^{\text{eff}}(x_s)$ are taken as the average $(x_s(t + \Delta t) - x_s(t))$ and accumulated in the corresponding bin for $x_s(t)$.

B. Agreement between simulations and averaging approach

We first compare effective rates and forces at long times. In Figs. 3(a) and 3(b), the kinetic rates and forces obtained numerically and with averaging theory Eqs. (20) and (21) [alternatively with Eqs. (22)–(26)] are in excellent agreement. The effective probability distribution function is also in agreement with the marginal distribution Eq. (A1) [Fig. 3(c)]. Overall, at the coarse-grained level, detailed balance holds numerically as well as analytically [Fig. 3(d)]. While we present in Fig. 3 the results for a small value of $\varepsilon = 0.1$, similarly good agreement can be obtained for large values of $\varepsilon \approx 10$ as long as the simulation times are long enough, which is expected from the coarse-graining procedure.

Overall, the analytical and numerical curves shown in Fig. 3 compare the effective rates and forces averaged over long times, but do not necessarily validate the short and intermediate time dynamics being correct in the coarse-graining procedure. To validate this, we numerically compute the autocorrelation from trajectories of the explicit microscopic model and coarse-grained as a function of ε . In Fig. 4, the dynamics of the coarse-grained model agree with

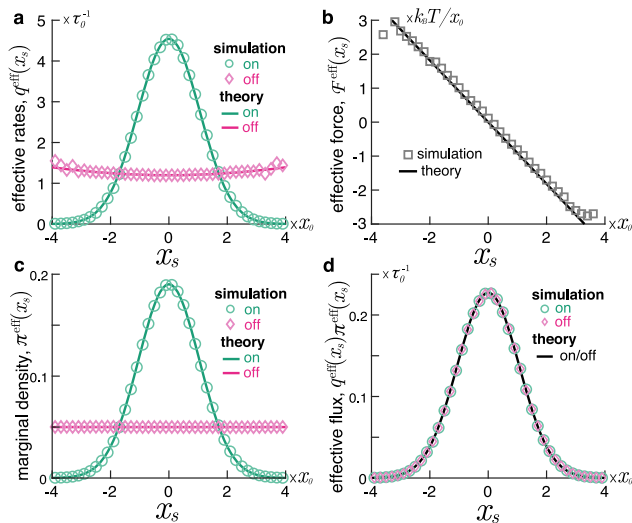


FIG. 3. Numerical validation of the coarse-graining procedure, here in the case of a fast linker transiently binding to a slow particle in 1D. Both binding and unbinding are chosen to be spatially dependent, as in model 2, Eq. (10), with numerical parameters $\varepsilon = 0.1$, $\kappa = 1$, $\lambda = 10$, $q_{\text{off}}^0 = 1/\tau_0$, and q_{on}^0 chosen such that $\Pi_b / \Pi_u = 1$ determined by the relation (27). 1000 Simulations are run until $T = 1000 \tau_0$. (a) Microscopic binding and unbinding effective rates, (b) effective force, (c) effective marginal probabilities of being bound or unbound, (d) macroscopic transition rates. Theory curves are obtained with the expressions summarized in Eqs. (22)–(26).

those of the explicit model only for small ε . At $\varepsilon \gtrsim 1$, the autocorrelation displays significant deviation. Intuitively, the coarse-grained model loses the memory of recent binding and unbinding events and therefore has faster decay in correlation. This highlights that the coarse-grained dynamics described in Eqs. (22)–(26) are only valid when $\varepsilon \lesssim 1$, i.e., when the timescales associated with fast linker or slow particle relaxation are disparate enough. To account for these memory effects, one could add the $O(\varepsilon)$ terms to the coarse-grained

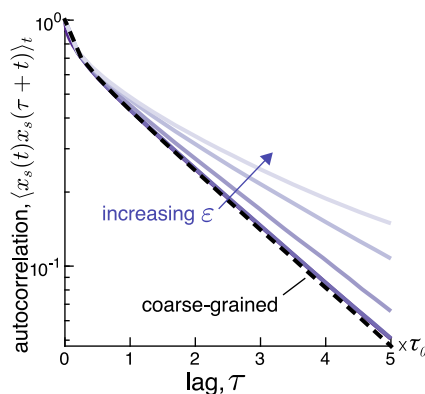


FIG. 4. A comparison between the detailed and coarse-grained model of the autocorrelation of the slow particle position $(x_s(t + \tau)x_s(t))_t$. In the limit of separation of timescales corresponding to small ε , the dynamics approach those of the coarse-grained model. Parameter values are the same as in Fig. 3, with purple curves corresponding to $\varepsilon = \{10^{-2}, 10^{-1}, 10^0, 10^1, 10^2\}$. The curves for the lowest two values of ε are indistinguishable.

dynamics we derive in the supplementary material, Sec. 2. Altogether, our numerical results thus show the averaging approach is robust to infer coarse-grained effective dynamics.

V. APPLICATIONS

Having shown the validity of the averaging approach, we now explore how to use Eqs. (22)–(26) in specific situations. We investigate (i) a pair of connecting linkers, (ii) a linker stiffening upon binding, and (iii) a slip bond with force-dependent unbinding. According to the specific example under scrutiny, we will focus either more on the effective forces or on the effective kinetic rates, and comment on the physical meaning of the results.

A. Pair of connecting linkers

We first consider two fast linkers that can connect to each other (see Fig. 5). This could represent for example two complementary DNA strands transiently hybridizing, which finds some applications for example in the field of DNA-coated colloids.^{20,24,29,39,57,58} We consider that one of the linkers, in position x_2 , is tethered to a fixed plate while the other, in position x_1 , is tethered to a mobile slow particle, itself in position x_s . Both linkers are described by springs with the same spring constant k_ℓ and fluctuate rapidly.

Unbound dynamics: The unbound dynamics are given by

$$\begin{cases} \frac{dx_s}{dt} = -\frac{k_\ell}{\gamma_s}(x_s - x_1) + \sqrt{\frac{2k_B T}{\gamma_s}}\eta_s(t), \\ \frac{dx_1}{dt} = -\frac{k_\ell}{\gamma_\ell}(x_1 - x_s) + \sqrt{\frac{2k_B T}{\gamma_\ell}}\eta_1(t), \\ \frac{dx_2}{dt} = -\frac{k_\ell}{\gamma_\ell}(x_2) + \sqrt{\frac{2k_B T}{\gamma_\ell}}\eta_2(t), \end{cases} \quad (29)$$

where the $\eta_i(t)$ are uncorrelated Gaussian white noises and here we identify the force on the slow particle in the unbound state as

$$\mathcal{F}_u = -k_\ell(x_s - x_1). \quad (30)$$

The unbound term of the equilibrium distribution corresponding to this choice of dynamics is

$$\pi_u(x_s, x_1, x_2) = \frac{1}{Z_u} e^{-\frac{k_\ell x_2^2 + k_\ell(x_s - x_1)^2}{2k_B T}}. \quad (31)$$

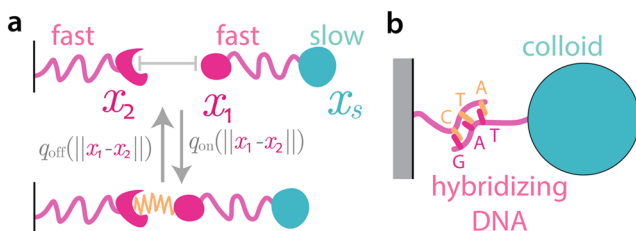


FIG. 5. (a) Cartoon of two fast linkers that can connect to each other, representing for example (b) two complementary DNA strands transiently hybridizing, one of them being attached to a colloid whose motion is of interest.

Bound dynamics: The linkers can transiently form a bond with spring constant k_b ,

$$\begin{cases} \frac{dx_s}{dt} = -\frac{k_\ell}{\gamma_s}(x_s - x_1) + \sqrt{\frac{2k_B T}{\gamma_s}}\eta_s(t), \\ \frac{dx_1}{dt} = -\frac{k_\ell}{\gamma_\ell}(x_1 - x_s) + \sqrt{\frac{2k_B T}{\gamma_\ell}}\eta_1(t) + \frac{k_b}{\gamma_\ell}(x_2 - x_1), \\ \frac{dx_2}{dt} = -\frac{k_\ell}{\gamma_\ell}(x_2) + \sqrt{\frac{2k_B T}{\gamma_\ell}}\eta_2(t) - \frac{k_b}{\gamma_\ell}(x_2 - x_1), \end{cases} \quad (32)$$

and we identify the force on the slow particle in the bound state as identical to that in the unbound state,

$$\mathcal{F}_b = -k_\ell(x_s - x_1). \quad (33)$$

The bound term of the equilibrium distribution corresponding to this choice of dynamics is

$$\pi_{\text{bound}}(x_s, x_1, x_2) = \frac{1}{Z_b} e^{-\frac{k_\ell x_2^2 + k_\ell(x_s - x_1)^2 + k_b(x_2 - x_1)^2}{2k_B T}}. \quad (34)$$

Kinetic rates and detailed balance: Here, we will not specify the kinetic rates in detail. However, since we specify the dynamics, the binding and unbinding rates must satisfy detailed balance

$$\frac{q_{\text{on}}}{q_{\text{off}}} = \frac{q_{\text{on}}^0}{q_{\text{off}}^0} e^{-k_\ell(x_2 - x_1)^2 / 2k_B T}.$$

Effective force in the unbound state: We can now use Eqs. (22)–(26) to obtain the effective force. Here, compared to our foundational example with one fast linker in Sec. II, we have two fast linkers, hence we need to carry the integral over those two fast degrees of freedom. We find with Eq. (25) that

$$\begin{aligned} \mathcal{F}_u^{\text{eff}}(x_s) &= \frac{\int \mathcal{F}_u(x_1, x_2, x_s) \pi_u(x_1, x_2, x_s) dx_1 dx_2}{\int \pi_u(x_1, x_2, x_s) dx_1 dx_2} \\ &= \frac{\int (-k_\ell(x_s - x_1)) e^{-k_\ell x_2^2 / 2k_B T} e^{-k_\ell(x_s - x_1)^2 / 2k_B T} dx_1 dx_2}{\int e^{-k_\ell x_2^2 / 2k_B T} e^{-k_\ell(x_s - x_1)^2 / 2k_B T} dx_1 dx_2} \\ &= 0 \end{aligned} \quad (35)$$

for symmetry reasons. In the unbound state, quite logically, at the coarse-grained level the particle does not feel any effective force from its unbound fast linker.

Effective force in the bound state: In the bound state, using Eq. (26) we find that

$$\begin{aligned} \mathcal{F}_b^{\text{eff}}(x_s) &= \frac{\int \mathcal{F}_b(x_1, x_2, x_s) \pi_{\text{bound}}(x_1, x_2, x_s) dx_1 dx_2}{\int \pi_{\text{bound}}(x_1, x_2, x_s) dx_1 dx_2} \\ &= -\frac{\frac{k_\ell}{2} k_b}{\frac{k_\ell}{2} + k_b} x_s. \end{aligned} \quad (36)$$

In the bound state, we thus obtain that the effective force on the particle is a spring force. The force is centered around 0 as this is the average position of both springs. The spring constant at the coarse-grained level is $\frac{k_\ell k_b}{\frac{k_\ell}{2} + k_b}$, which corresponds, logically, to the effective spring constant of three springs in series, with spring constants k_ℓ, k_b, k_ℓ .

B. Stiffening linker upon binding

We now turn to another example where the linker stiffens when bound—see Fig. 6(a). Such stiffening occurs when the linker undergoes conformational changes upon binding²⁸ or else upon single-stranded DNA hybridizing into double-stranded, resulting in a stiffer connection.^{16,18,21,24,59,60}

Unbound dynamics: The unbound dynamics are still given by Eq. (1).

Bound dynamics: The bound dynamics are now changed compared to Eq. (2), since the linker now has a stiffer spring constant say $k'_\ell > k_\ell$,

$$\begin{cases} \frac{dx_s}{dt} = -\frac{k_b}{\gamma_s}(x_s - x_\ell) + \sqrt{\frac{2k_B T}{\gamma_s}}\eta_s(t), \\ \frac{dx_\ell}{dt} = -\frac{k'_\ell}{\gamma_\ell}x_\ell - \frac{k_b}{\gamma_\ell}(x_\ell - x_s) + \sqrt{\frac{2k_B T}{\gamma_\ell}}\eta_\ell(t). \end{cases} \quad (37)$$

The bound term of the equilibrium distribution corresponding to this choice of bound dynamics is

$$\pi_b(x_s, x_\ell) = \frac{1}{Z_b} e^{-k'_\ell x_\ell^2/2k_B T} e^{-k_b(x_s - x_\ell)^2/2k_B T}. \quad (38)$$

Kinetic rates: Again without specifying the kinetic rates in detail, since we set the dynamics, the binding and unbinding rates must satisfy detailed balance and hence are related via

$$\frac{q_{\text{on}}}{q_{\text{off}}} = \frac{q_{\text{on}}^0}{q_{\text{off}}^0} e^{-k_b(x_s - x_\ell)^2/2k_B T} e^{-(k'_\ell - k_\ell)x_\ell^2/2k_B T}.$$

This relation favors binding when the tether is not too extended. Otherwise, the energetic price to pay to “stiffen” is too costly. Reciprocally, the bond is more likely to break when the tether is quite extended, and the energetic gain to loosen the bond is quite high. One could thus choose, in line with physical intuition,

$$\begin{cases} q_{\text{on}} \sim q_{\text{on}}^0 e^{-k_b(x_s - x_\ell)^2/2k_B T}, \\ q_{\text{off}} \sim q_{\text{off}}^0 e^{(k'_\ell - k_\ell)x_\ell^2/2k_B T}. \end{cases} \quad (39)$$

The detailed prefactors should be specified in agreement with detailed balance.

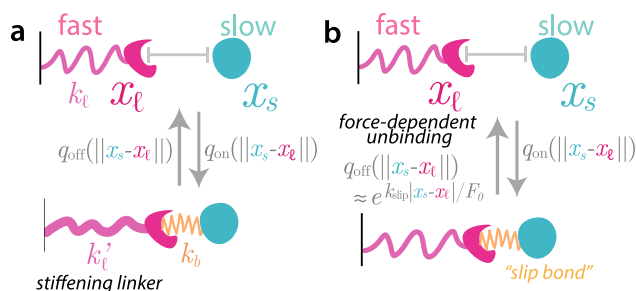


FIG. 6. (a) Cartoon of a fast linker stiffening upon binding and (b) a fast linker connecting to the particle via a slip bond with force-dependent unbinding.

Effective force in the bound state: The effective force in the unbound state naturally vanishes. Hence, we focus on the effective force in the bound state, again obtained via Eq. (26),

$$\mathcal{F}_b^{\text{eff}}(x_s) = -\frac{k'_\ell k_b}{k'_\ell + k_b} x_s. \quad (40)$$

The coarse-grained force is that associated with two springs in series with spring constants k'_ℓ and k_b . However, here it was not obvious *a priori* which of the spring constants for the linker, either k'_ℓ or k_ℓ or a mix of both, should contribute to the force.

Effective rates: We now use the rates defined in Eq. (39) to study coarse-grained kinetic rates as well. Let $K = k_\ell k_b / (k_\ell + k_b)$ and $K' = k'_\ell k_b / (k'_\ell + k_b)$. Then, we obtain

$$\begin{cases} q_{\text{on}}^{\text{eff}}(x_s) \sim e^{-Kx_s^2/2k_B T}, \\ q_{\text{off}}^{\text{eff}}(x_s) \sim e^{(K' - K)x_s^2/2k_B T}. \end{cases} \quad (41)$$

The effective binding rate is determined by the typical, unstiff, radius of the interaction, $\sqrt{k_B T / K}$, similarly as in our foundational example in Sec. II. The unbinding rate is now larger at larger distances, according to how much stiffening occurred. Notice that even if the bond is really stiff, meaning $k_b \gg k_\ell, k'_\ell$, then $K' - K \simeq k'_\ell - k_\ell$ still converges to a finite value.

C. Slip bonds with force-dependent unbinding

Another quite common situation is that of slip bonds with force-dependent unbinding^{28,43}—see Fig. 6(b). In these cases, the unbinding rate scales as⁴³

$$q_{\text{off}}(x_s, x_\ell) = q_{\text{off}}^0 e^{k_{\text{slip}}|x_s - x_\ell|/F_0}, \quad (42)$$

where k_{slip} is a characteristic bond spring constant and F_0 a characteristic force threshold. Above this threshold, the unbinding rate is indeed greatly enhanced.

Kinetic rates and satisfying detailed balance: How shall we proceed to model the dynamics with such force-dependent unbinding rate, when the system is still at equilibrium? Compared to our foundational example in Sec. II, here we might modify the binding rate to satisfy detailed balance. Notice that this is not the only possibility to satisfy detailed balance and one could also consider alternative bound dynamics. For the sake of the example, here we will consider

$$q_{\text{on}}(x_s, x_\ell) = q_{\text{on}}^0 e^{-k_b(x_s - x_\ell)^2/2k_B T} \quad (43)$$

and an additional force on the bound particle that we specify in Eq. (44) below. The expressions for the rates are satisfactory since they are coherent with faster binding and slower unbinding when the particle and linker are closer. Similar choices motivated by their intuitive behavior date back to Bell and Dembo^{43,44} and are derived in Ref. 25 as a coarse-graining of a microscopic model for binding. The detailed prefactors should be specified in agreement with detailed balance.

Unbound dynamics: The unbound dynamics are still given by Eq. (1).

Bound dynamics: The bound dynamics are now changed compared to Eq. (2) since the linker and particle are now subjected to additional forces,

$$\begin{cases} \frac{dx_s}{dt} = -\frac{k_b}{\gamma_s}(x_s - x_\ell) \\ \quad - \frac{k_{\text{slip}} k_B T}{\gamma_s F_0} \text{sgn}(x_s - x_\ell) + \sqrt{\frac{2k_B T}{\gamma_s}} \eta_s(t), \\ \frac{dx_\ell}{dt} = -\frac{k_\ell}{\gamma_\ell} x_\ell - \frac{k_b}{\gamma_\ell}(x_\ell - x_s) \\ \quad + \frac{k_{\text{slip}} k_B T}{\gamma_\ell F_0} \text{sgn}(x_s - x_\ell) + \sqrt{\frac{2k_B T}{\gamma_\ell}} \eta_\ell(t). \end{cases} \quad (44)$$

In higher dimensions, $\text{sgn}(x) = x/\|x\|$ where $x = x_s - x_\ell$. The bound term of the equilibrium distribution corresponding to this choice of bound dynamics is

$$\pi_b(x_s, x_\ell) = \frac{1}{Z_b} e^{-k_\ell x_\ell^2/2k_B T} e^{-k_b(x_s - x_\ell)^2/2k_B T} e^{-k_s|x_s - x_\ell|/F_0}. \quad (45)$$

Effective force in the bound state: The effective force in the unbound state naturally vanishes. Hence, we focus on the effective force in the bound state, which has a lengthy expression that we do not report here. In the case of a small slip force $k_{\text{slip}}k_B T/F_0 \ll k_\ell|x_s|$, we obtain

$$\begin{aligned} \mathcal{F}_b^{\text{eff}}(x_s) \simeq & -\frac{k_\ell k_b}{k_\ell + k_b} x_s \\ & - \frac{k_\ell}{k_\ell + k_b} \frac{k_{\text{slip}} k_B T}{F_0} \text{erf}\left(\sqrt{\frac{k_\ell}{k_\ell + k_b} \frac{k_\ell x_s^2}{2k_B T}}\right). \end{aligned} \quad (46)$$

The coarse-grained force contains now two contributions. The first one is that associated with two springs in series with spring constants k_ℓ and k_b that we have seen before. The second one corresponds to the slip bond, which grows stronger when the particle goes further away from the target. Notice that this latter slip bond friction force is screened by a factor $\frac{k_\ell}{k_\ell + k_b}$ in the coarse-grained state. In the case of a stiff bond, $k_b \gg k_\ell$, the slip force is entirely screened, and the particle “slips”; the linker essentially accommodates changing configurations by rapidly adjusting its length.

Effective rates: We now use the rates defined in Eqs. (42) and (43) to study coarse-grained kinetic rates as well. Let $K = k_\ell k_b / (k_\ell + k_b)$. Then, we obtain, in the case of a small slip force $k_{\text{slip}}k_B T/F_0 \ll k_\ell|x_s|$,

$$\begin{cases} q_{\text{on}}^{\text{eff}}(x_s) \sim e^{-Kx_s^2/2k_B T}, \\ q_{\text{off}}^{\text{eff}}(x_s) \sim e^{\frac{k_\ell}{k_\ell + k_b} \frac{k_{\text{slip}}|x_s|}{F_0}}. \end{cases} \quad (47)$$

The effective binding rate is determined by the typical, unstiff, radius of the interaction, $\sqrt{k_B T/K}$, similarly as in our foundational example in Sec. II. The unbinding rate is now larger at larger distances, with force-dependent unbinding.

VI. MACROSCOPIC CONSEQUENCES FOR THE CHOICE OF BINDING KINETICS

A. Coarse-graining various microscopic binding models: The Doi model

How might the effective dynamics change from our foundational example in Sec. II when we consider the Doi model for binding?

Effective force in the unbound state: According to Eq. (25) and with Eqs. (12) and (13) describing the Doi model in our context, the unbound friction force is simply

$$\mathcal{F}_u^{\text{eff}}(x_s) = -k_b \frac{\int_{x_s-R}^{x_s+R} (x_s - x_\ell) e^{-\frac{k_b(x_s - x_\ell)^2 + k_\ell x_\ell^2}{2k_B T}} dx_\ell}{\int \pi_u(x_s, x_\ell) dx_\ell}, \quad (48)$$

which has a cumbersome, nonvanishing, expression that we do not report here. However, we can make a finite perturbation of the obtained expression in the limit of a stiff bond, when $k_b \gg k_\ell$, and to simplify the expression further, we assume that the radius to bind R is given by the typical spatial scale of the bond $R = \sqrt{2k_B T/k_b}$, such that

$$\mathcal{F}_u^{\text{eff}}(x_s) \simeq -0.43 \sqrt{\frac{k_\ell}{k_b}} k_\ell x_s. \quad (49)$$

Since x_ℓ wiggles around 0 and even in the unbound state, close to the linker, the particle feels a recoil force, then it makes sense that the particle can now feel a recoil force everywhere. The magnitude of this force is slightly decreased because the bond can only form when the particle is close enough to the linker. Notice how the scaling of the effective spring constant is entirely nontrivial as $\sqrt{\frac{k_\ell}{k_b}} k_\ell$.

Effective force in the bound state: According to Eq. (26), the bound friction force is simply

$$\mathcal{F}_b^{\text{eff}}(x_s) = -\frac{k_\ell k_b}{k_\ell + k_b} x_s, \quad (50)$$

which is similar to the results obtained with our foundational example.

Effective kinetic rates: Again, the expression for the effective binding and unbinding rates obtained from Eqs. (23) and (24) is rather cumbersome. When $k_b \gg k_\ell$, and assuming $R = \sqrt{2k_B T/k_b}$, the kinetic rates are smoothly decaying to 0 (instead of a sharp Heaviside function as in the microscopic equations), at a characteristic distance $x_s \simeq \sqrt{\frac{k_B T}{k_\ell} \sqrt{\frac{k_\ell}{k_b}}}$. Again, one sees how this scaling is entirely nontrivial. Systematic coarse-graining is therefore essential for faster numerical simulations and enhanced theoretical investigations.

B. Macroscopic consequences

We finish by briefly exploring how the microscopic choices for $q_{\text{on}}(x_s, x_\ell)$ and $q_{\text{off}}(x_s, x_\ell)$ can affect the macroscopic dynamics—see Sec. II B. To simplify the exploration, here we directly simulate the coarse-grained equations for the slow particle, using the coarse-grained forces and kinetic rates in Eqs. (22)–(26). We test the impact of different initial choices of microscopic binding, specifically with model 1 corresponding to Eq. (9) with

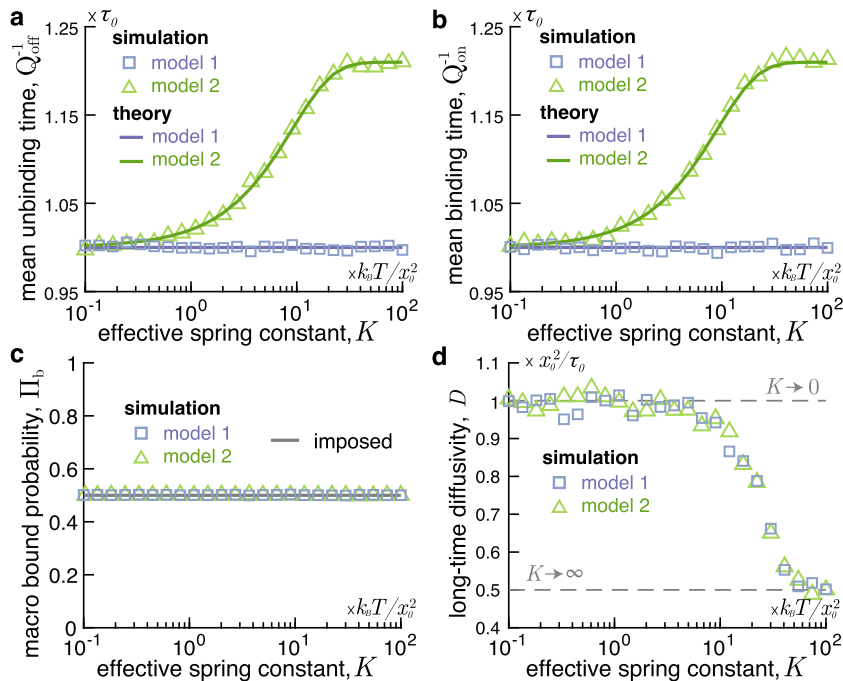


FIG. 7. Different choices of $q_{\text{on}}(x)$ and $q_{\text{off}}(x)$ yield different macroscopic dynamics. Model 1 corresponds to Eq. (9) with spatially dependent binding only and model 2 corresponds to Eq. (10) with both spatially dependent binding and unbinding. Simulation parameters are $q_{\text{off}}^0 = 0.1\tau_0^{-1}$ and q_{on}^0 , and are chosen such that $\Pi_b/\Pi_u = 1$. 1000 simulations are run until $T = 1000\tau_0$. (a) Mean unbinding time as a function of the effective spring constant K , showing divergence between models 1 and 2. Theory curves are obtained with Eq. (51). (b) Mean binding time as a function of the effective spring constant K with similar trend. (c) Macroscopic bound probabilities Π_b are set to be fixed and are the same for both models. (d) Macroscopic diffusion coefficient $D = \lim_{t \rightarrow \infty} \langle x_s(t)^2 \rangle / (2t)$ for both models.

spatially dependent binding only and model 2 from Eq. (10) with both spatially dependent binding and unbinding. To compare different microscopic models in a sensible fashion, we constrain the macroscopic bound probability $\Pi_b = 0.5$ [see Fig. 7(c)] for all microscopic models. We then calculate the macroscopic mean binding and unbinding times for the particle [Figs. 7(a) and 7(b)], as well as the particle’s long-time diffusion coefficient [Fig. 7(d)], determined by its mean-squared displacement. Importantly, we vary the effective bond spring constant $K = k_\ell k_b / (k_b + k_\ell)$ to probe how model parameters affect macroscopic dynamics.

We find that both the choice of microscopic binding model and the parameter K significantly affect the macroscopic binding and unbinding times [Figs. 3(a) and 3(b)]. Since the macroscopic binding probability Π_b is fixed, the mean binding and unbinding times are identical, and hence Figs. 7(a) and 7(b) appear the same. At long times, we expect, as for the coarse-graining over x_ℓ , that the kinetic rates do not depend on the position x_s anymore and would verify

$$Q_{\text{off}} = \frac{\int q_{\text{off}}^{\text{eff}}(x_s) \pi_b^{\text{eff}}(x_s) dx_s}{\int \pi_b^{\text{eff}}(x_s) dx_s}, \quad (51)$$

and similarly for the binding rate Q_{on} , in agreement with other works.²⁵ Equation (51) gives the mean unbinding time as $1/Q_{\text{off}}$ and reproduces perfectly the numerical results [lines in Fig. 3(a)].

For the different microscopic models, the response to various binding constants K is nontrivial. For model 1, the macroscopic binding and unbinding times remain constant. In model 1, $q_{\text{off}}^{\text{eff}}(x_s)$ does not depend on space, and, therefore, its macroscopic counterpart $q_{\text{off}}^{\text{eff}}$, which verifies Eq. (51), seems to have no direct dependence

on K . However, $q_{\text{on}}^{\text{eff}}(x_s)$ is inherently spatially dependent and hence its macroscopic counterpart may depend on K as well, via the associated length scale $\sqrt{k_B T / K}$. In contrast, with model 2 the macroscopic kinetic rates increase with the binding constant K . When K is larger, the bond is stronger, which decreases the binding time but increases the unbinding time. Hence, the macroscopic dependence where both kinetic rates increase with K is interesting. In fact, to properly coarse-grain to this now macroscopic scale, one should also take into account the probability distribution of being in each state with a given extension. Such entangled behavior requires proper integration beyond simple intuition. Eventually, we find that the macroscopic binding times strongly depend on microscopic choices for the kinetic rates, which means one should use caution when designing such models.

In addition, for both microscopic model choices, the long-time self-diffusion coefficient of the particle depends on K [Fig. 3(d)]. In fact, at small K we find $D \rightarrow k_B T / \gamma_s$ since the bond is weak enough that it barely affects particle motion. At larger K values, we find $D \rightarrow k_B T / 2\gamma_s$ since when it is bound, the bond is strong enough that it prevents any motion, and the particle is bound half of the time ($\Pi_b = 0.5$). The dependence on K appears to be similar for both microscopic binding models 1 and 2. In more varied models for binding, there is no reason this should stay true. We leave the general exploration of macroscopic transport properties, such as diffusion coefficients, and their dependence on microscopic binding kinetics, for future work.

VII. CONCLUSION

In this work, we have attempted to unify and justify various coarse-graining approaches for linker dynamics.^{22,25,27–34,36,43–45}

In these earlier approaches, we have identified many different choices for how binding and unbinding depend on distance, yet (1) these are often *heuristically* motivated, or (2) they only provide coarse-grained binding rates and not the full dynamics of the slow particles including forces and friction. Here, we have addressed all of these issues by providing a systematic derivation of effective dynamics, including effective friction, forces, and binding kinetics for linker molecules that obey detailed microscopic descriptions. Our coarse-graining approach is based on averaging techniques and preserves detailed balance, in line with our assumption of equilibrium dynamics. We have verified our approach with numerical simulations and found excellent agreement in both the effective kinetic rates and dynamics after averaging. This averaging analysis hinges on $\varepsilon = \gamma_\ell/\gamma_s$ being small, corresponding to a large separation of timescales, between the fast linker and slow particle relaxation. When $\varepsilon \gtrsim 1$, the coarse-grained dynamics diverge from the detailed model. We showed how our general framework may be applied to diverse microscopic scenarios, including those with two linkers binding to each other, a linker stiffening upon binding, or a slip bond with force-dependent unbinding. Finally, we showed that different microscopic kinetic rates result in fundamentally different dynamics at the macroscopic scale, raising caution in making these choices without care.

Many choices for effective dynamics in the literature clearly violate detailed balance and hence operate out of equilibrium.^{28,61} The coarse-graining in Ref. 25 computes the effective rates in a mechanical model with nonequilibrium fluctuations, but relating nonequilibrium rates and dynamics seems to be missing a unified framework. Our systematic approach to coarse-graining may provide a first step toward addressing this. The authors in Ref. 62 interestingly note that effective equilibria can arise by switching between nonconservative systems, providing hope that this pursuit is both fruitful and interesting. Thus, extending our method to out-of-equilibrium systems would be of broad relevance, in particular, to explore transient bonds with activated cleaving that are common in viral linker-mediated motion^{63,64} and also artificial nano-motors.⁷

Although we have investigated relatively simple setups here, the tools we introduce and the lessons learned are applicable across many more complex systems. The formulas we derived for effective dynamics can be applied with ease via Eqs. (22)–(26) and provide a baseline for more systematic coarse-graining found in simulations and theoretical studies across the literature. One such example is molecular motor binding in intracellular transport. There, coarse-graining ranges from nonspatial effective kinetic rates,^{65,66} motor linkers obeying a worm-like chain model,^{67,68} or Doi-like motors that bind when within a specific radius of interaction.⁴⁸ Coarse-grained cross-linked cytoskeletal networks are studied extensively,^{5,6,69–71} especially in self-organization in the mitotic spindle^{72–75} and actomyosin network mechanics.^{76,77} Transient dynamics of (typically coarse-grained) cross-linkers are also fundamental in controlling viral responses in bio-gels^{78,79} and building chromosomal territories.⁸⁰ Our framework also readily extends to systems that are not “cross”-linked, such as membrane-filament interactions that drive cell protrusion^{81,82} and adhesions.^{9,25,45}

SUPPLEMENTARY MATERIAL

See the supplementary material for further mathematical details of the coarse-graining procedure, including calculations of higher order corrections.

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AUTHOR DECLARATIONS

Conflict of Interest

The authors have no conflicts to disclose.

Author Contributions

S.M. and C.E.M. contributed equally to this work.

Sophie Marbach: Conceptualization (equal); Investigation (equal); Visualization (equal); Writing – original draft (equal); Writing – review & editing (equal). **Christopher E. Miles:** Conceptualization (equal); Investigation (equal); Visualization (equal); Writing – original draft (equal); Writing – review & editing (equal).

DATA AVAILABILITY

The data that support the findings of this study are available from the corresponding author upon reasonable request.

NOMENCLATURE

x_s	coordinate of the slow particle
x_ℓ	coordinate of the fast linker
x_0	characteristic length scale for nondimensionalization
γ_s	friction coefficient on the slow particle
γ_ℓ	friction coefficient on the fast linker
$\varepsilon = \gamma_\ell/\gamma_s$	ratio of friction coefficients
k_ℓ	spring constant describing the linker
k_b	spring constant describing the bond between the particle and the linker
$k_B T$	thermal energy
τ_0	characteristic timescale for nondimensionalization
$\pi_{u/b}(x_s, x_\ell)$	microscopic unbound (respectively, bound) probability distribution
$\pi_{u/b}^{\text{eff}}(x_s)$	coarse-grained unbound (respectively, bound) probability distribution
$\Pi_{u/b}$	macroscopic unbound (respectively, bound) probability
$\mathcal{F}_{u/b}(x_s, x_\ell)$	microscopic force on the slow particle in the unbound (respectively, bound) state

$\mathcal{F}_{u/b}^{\text{eff}}(x_s)$	coarse-grained force on the slow particle in the unbound (respectively, bound) state
$q_{\text{on/off}}(x_s, x_\ell)$	microscopic binding (respectively, unbinding) rate
$q_{\text{on/off}}^{\text{eff}}(x_s)$	coarse-grained binding (respectively, unbinding) rate
$Q_{\text{on/off}}$	macroscopic binding (respectively, unbinding) rate

APPENDIX: DETAILED BALANCE AT THE COARSE-GRAINED LEVEL

Finally, for physical consistency, we need to check that the marginal equilibrium distribution, i.e., the equilibrium distribution integrated over the fast degrees of freedom $\pi^{\text{eff}}(x_s) = \int \pi(x_s, x_\ell) dx_\ell$, is indeed a stationary solution of the effective dynamics obtained. The marginal distribution is

$$\pi^{\text{eff}}(x_s) = \begin{pmatrix} \pi_u^{\text{eff}}(x_s) \\ \pi_b^{\text{eff}}(x_s) \end{pmatrix} = \frac{\sqrt{2\pi} e^{-\mathcal{U}_s(x_s)/k_B T}}{Z} \begin{pmatrix} \frac{1}{\sqrt{k_\ell}} \\ \frac{q_{\text{on}}^0}{q_{\text{off}}^0 \sqrt{k_\ell + k_b}} e^{-K \frac{x_s^2}{2k_B T}} \end{pmatrix}. \quad (\text{A1})$$

It is clear that the marginal distribution of either state is indeed a stationary solution of the dynamics in each state as specified in Eq. (21).

To check that the marginal distribution is a stationary distribution of the dynamics as a whole, we still need to check that detailed balance is satisfied at the coarse-grained level. This is not guaranteed *a priori* since the averaging technique does not use at any point that it should preserve detailed balance. At this point, it is important to notice that in fact the effective rates are related to the equilibrium probability distribution as

$$q_{\text{on}}^{\text{eff}}(x_s) = \frac{\int q_{\text{on}}(x_s, x_\ell) \pi_u(x_s, x_\ell) dx_\ell}{\int \pi_u(x_s, x_\ell) dx_\ell} \quad (\text{A2})$$

and similarly for $q_{\text{off}}^{\text{eff}}$. Hence, we simply have that

$$\begin{aligned} \pi_u^{\text{eff}} q_{\text{on}}^{\text{eff}}(x_s) &= \pi_u^{\text{eff}} \frac{\int q_{\text{on}}(x_s, x_\ell) \pi_u(x_s, x_\ell) dx_\ell}{\int \pi_u(x_s, x_\ell) dx_\ell} \\ &= \int q_{\text{on}}(x_s, x_\ell) \pi_u(x_s, x_\ell) dx_\ell \\ &= \int q_{\text{off}}(x_s, x_\ell) \pi_b(x_s, x_\ell) dx_\ell \\ &= \int q_{\text{off}}(x_s, x_\ell) \pi_b(x_s, x_\ell) dx_\ell \frac{\pi_b^{\text{eff}}}{\int \pi_b(x_s, x_\ell) dx_\ell} \\ &= \pi_b^{\text{eff}} q_{\text{off}}^{\text{eff}}(x_s). \end{aligned}$$

Detailed balance is therefore also true at the coarse-grained level. Since the marginal distribution is consistent with the dynamics in each state and with detailed balance, it is indeed the stationary solution to the effective dynamics.

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