

# Kinetics of Filament Bundling with Attractive Interactions

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**ABSTRACT** We study the kinetics of filament bundling by variable time-step Brownian-dynamics simulations employing a simplified attractive potential based on earlier atomic-level calculations for actin filaments. Our results show that collisions often cluster in time, due to memory in the random walk. The clustering increases the bundling opportunities. Small-angle collisions and collisions with short center-to-center distance are more likely to lead to bundling. Increasing the monomer-monomer attraction decreases the bundling time to a diffusional limit, which is determined by the capture cross-section and diffusion coefficients. The simulations clearly show that the bundling process consists of two sequential phases: rotation, by which two filaments align parallel to each other; and sliding, by which they maximize their contact length. Whether two filaments bundle or not is determined by the competition between rotation to a parallel state and escape. Increasing the rotational diffusion coefficient and attraction enhances rotation; decreasing attraction and increasing the translational diffusion coefficients enhance escape. Because of several competing effects, the filament length only affects the bundling time weakly.

## INTRODUCTION

Polyelectrolytes such as DNA and F-actin often aggregate or bundle together *in vivo* or *in vitro* (Kawamura and Maruyama, 1970; Baeza et al., 1987; Bloomfield, 1996, 1997; Tang and Janmey, 1996; Tang et al., 1996). The aggregation of very long DNA molecules allows them to be stored in a very small volume. The bundling of actin filaments can enhance their rigidity, which is crucial for their cytoskeletal role of supporting cell extensions, and may affect the internal mechanical properties of the cell as well. On the other hand, the formation of amyloid fibrils, which are stable, ordered, filamentous protein aggregates consisting of multiple bundled protofilaments (Rochet and Lansbury, 2000), cause amyloidoses including many neurodegenerative diseases. Although there have been many theoretical studies in this field (Oosawa, 1968; Ray and Manning, 1994; Grønbech-Jensen et al., 1997; Ha and Liu, 1997; Kornyshev and Leikin, 1998; Shklovskii, 1999; Gelbart et al., 2000; Stevens, 2001; Diehl et al., 2001; Moreira and Netz, 2001; Deserno and Holm, 2002; Lau and Pincus, 2002; Manning, 2003), most of these have focused on deriving the attractive interaction between like-charged polymers. When the attractive interaction is mediated by counterions or bundling proteins, it is generally found to be short-ranged. Much less attention has been paid to the bundling process itself. Despite some theoretical studies of the thermodynamics of bundling (van der Schoot and Odijk, 1992; Sear, 1997; Khokhlov and Semenov, 1985; Yu and Carlsson, 2003), there has been no comprehensive theoretical analysis of bundling kinetics. There have been several experimental studies of bundling as a function of properties such as counterion concentration and

filament length (Tang and Janmey, 1996; Tang et al., 1996), but we are not aware of systematic experimental studies of the bundling kinetics of filaments with short-ranged attractive interactions. Our purpose in this article is to establish the mechanism of bundling and develop simplified mathematical models of the bundling kinetics.

The realistic study of biopolymer bundling is often hampered by the absence of suitable interaction potentials. In our previous work (Yu and Carlsson, 2003), we derived the potentials between actin filaments in counterion solutions and simplified the attractive potential under a limited counterion concentration range as a sum of short-ranged monomer-monomer interactions. With this simplified potential, we studied the thermodynamics of bundling. The potentials of this form include the main features of the filament interaction: large anisotropy, short range, and steric exclusion. In this article, we use potentials of this form to reveal the bundling mechanism and to study its kinetics by Brownian dynamics simulations. The simulations are carried out under periodic boundary conditions, using variable time steps. Simplified mathematical models are then used to explain and summarize the simulation results.

## METHODS

Our simplified attraction potential between two actin filaments in a two-filament conformation  $C_{2f}$  is of the form (Yu and Carlsson, 2003)

$$E(C_{2f}) = \begin{cases} - \sum_{i \in a, j \in b} E_{mm} \exp[-(R_{ij} - R_c)/R_d] \\ \quad \times H(R_{\max} - R_{ij}), & \forall R_{ij} \geq R_c, \\ \infty, & \exists R_{ij} < R_c \end{cases} \quad (1)$$

where  $a$  and  $b$  are two filaments containing monomers  $i$  and  $j$ , respectively;  $R_{ij}$  is the distance between the centers of  $i$  and  $j$ ;  $R_d$  is a decay length ( $R_d = 7 \text{ \AA}$ );  $R_{\max}$  is the distance cutoff for this short-ranged interaction; and  $H(x)$  is

Submitted June 9, 2004, and accepted for publication September 7, 2004.

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0006-3495/04/12/3679/11 \$2.00

doi: 10.1529/biophysj.104.047373

the Heaviside step function, which equals 0 for  $x < 0$  and 1 for  $x \geq 1$ . The monomers are arranged in a straight line. Eq. 1 implies that the bundling energy for parallel filaments varies fairly linearly with filament length. We let the monomer-monomer interaction vanish for  $R_{ij} > R_{\max}$  ( $100 \text{ \AA}$ ), and the closest approach distance allowed is  $R_c = 75 \text{ \AA}$ , the filament diameter. Any step leading to steric overlap (i.e.,  $R_{ij} < R_c$ ) is rejected in our Brownian dynamics simulations.  $E_{\text{mm}}$  is the maximal attractive energy of two monomers when they are in closest contact. The  $E_{\text{mm}}$  value found for attraction mediated by divalent metal ions at 32 mM is  $E_{\text{mm}}^{\text{ion}} = 0.0051 \text{ eV}$  (Yu and Carlsson, 2003), which may increase when mediated by bundling proteins such as fascin. Other aspects of the interaction may also be changed by bundling proteins. Note that the lowest energy for two filaments of equal length  $L$  (measured in monomers) is  $\sim -2LE_{\text{mm}}$  instead of  $-LE_{\text{mm}}$ , because a monomer can interact with more than one monomer in the other filament. Eq. 1 neglects the long-ranged interactions. Even for a simple system consisting of only two similarly charged plates and their counterions (no salt), the long-ranged monopolar repulsion is generally neutralized by the fluctuation-driven attraction (Lau and Pincus, 2002). In not-too-dilute salt solutions, the interaction will be short-ranged due to Debye-Hückel screening. As shown below, the short-ranged interaction simplifies the simulations greatly, because we can use large time steps when two filaments are far away from each other.

We use Brownian dynamics to simulate the filament motion, with diffusion coefficients calculated according to Doi and Edwards (1986). The three diffusion coefficients are

$$D_h = \frac{T \ln(L_f/2R_c)}{2\pi\eta_w L_f}, \quad (2)$$

$$D_v = \frac{T \ln(L_f/2R_c)}{4\pi\eta_w L_f}, \quad (3)$$

and

$$D_r = \frac{3T \ln(L_f/2R_c)}{\pi\eta_w L_f^3}, \quad (4)$$

where  $D_h$  and  $D_v$  characterize the diffusion parallel and perpendicular to the filament, and  $D_r$  is the rotational coefficient.  $L_f$  is the filament length ( $L_f = L \times 27.3 \text{ \AA}$ , where  $27.3 \text{ \AA}$  is the height of a monomer),  $\eta_w$  is the dynamic viscosity of water,  $T$  is the temperature, and our temperature units are such that Boltzmann's constant is unity.

We carry out simulations in a periodic-boundary cubic cell (Fig. 1). One filament is chosen as the target, and this filament moves through other filaments, called *environmental* filaments. These constitute a solution of the desired filament concentration. Thus, we divide the whole space into periodic-boundary cubic cells, whose size is determined by the filament concentration, and fix an environmental filament at the center of each cell. The orientations of the environmental filaments could be assigned randomly for each filament, but we fix them in the  $z$  direction. This does not affect the results noticeably if the solution is dilute, and simplifies the treatment of steric exclusion. The target filament is initially put at a random position with a random orientation.

Because the environmental filaments are frozen, the motion of the target filament should be taken as its motion relative to the environmental filament at the center of the same cell, i.e., its own motion with that of the central filament subtracted. Its own motion includes two parts: translation and rotation. To explain how we calculate the filament motion, we consider a horizontal filament for simplicity. Then translation has two directions: horizontal  $\hat{h}$  (along the filament) and vertical  $\hat{v}$  (perpendicular to the filament), where  $\hat{h}$  and  $\hat{v}$  are unit vectors. For each time step  $\Delta t$ ,  $\hat{v}$  is randomly chosen from the directions perpendicular to the filament. The random displacements along  $\hat{h}$  and  $\hat{v}$  can be expressed as

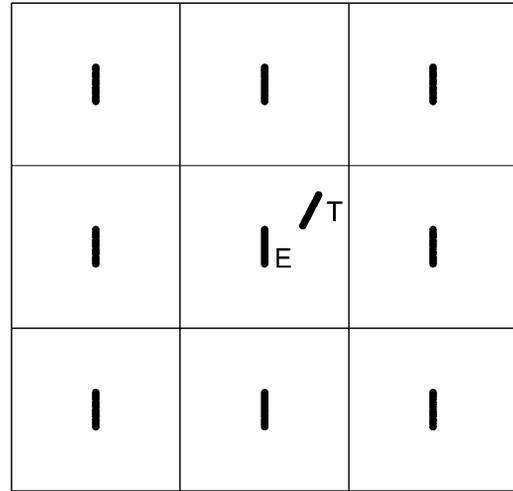


FIGURE 1 The Brownian dynamics simulations are carried out under periodic boundary conditions, with variable time steps. One filament is chosen as the target (marked  $T$ ), which moves through other environmental filaments (marked  $E$ ), centered in cells. Environmental filaments are frozen; target filament moves relative to the environmental filament at the center of its cell.

$$\Delta h = \sqrt{2D_h \Delta t} \xi \gamma, \quad (5)$$

$$\Delta v = \sqrt{4D_v \Delta t} \xi \gamma, \quad (6)$$

where  $\gamma = \pm 1$  defines the direction of the random Brownian motion, and  $\xi$  is a random variable, which satisfies  $\xi^2 = 1$ . We choose  $\xi$  randomly from 0 to 2. The values of  $\gamma$  and  $\xi$  are different in Eqs. 5 and 6, and change in every step. Thus, the translation step is

$$\begin{aligned} \Delta \vec{r} = & \Delta h \hat{h} + \Delta v \hat{v} + (\vec{F} \cdot \hat{h} D_h \Delta t / T) \hat{h} \\ & + [\vec{F} - (\vec{F} \cdot \hat{h}) \hat{h}] D_v \Delta t / T, \end{aligned} \quad (7)$$

where  $F$  is the force, driving the motion described by the last two terms.

To treat rotation, we define  $\theta$  as the angle between the filament axis and the  $z$  axis, and  $\phi$  as the corresponding azimuthal angle. The rotation is also divided into two components,

$$\Delta \theta = \Delta \alpha \sin \psi + \tau_\theta D_r \Delta t / T, \quad (8)$$

$$\Delta \phi = \Delta \alpha \cos \psi / \sin \theta + \tau_\phi D_r \Delta t / T, \quad (9)$$

where  $\Delta \alpha = \sqrt{4D_r \Delta t} \xi$  is the magnitude of a random angle change  $\Delta \vec{\alpha}$ , and  $\psi$  is the angle between  $\Delta \vec{\alpha}$  and  $\hat{e}_\phi$  (the direction of  $\phi$ ), which is randomly chosen from 0 to  $2\pi$  so that the direction of  $\Delta \vec{\alpha}$  is random.  $\tau_\theta$  and  $\tau_\phi$  are torques driving rotation in the  $\hat{e}_\theta$  (direction of  $\theta$ ) and  $\hat{e}_\phi$  directions. When  $\theta = 0$ ,  $\Delta \phi$  is randomly chosen from 0 to  $2\pi$ . Similarly, we obtain the motion of the central environmental filament. By subtracting the motion of the central filament from the target filament's motion, we obtain the relative motion. Defining  $\Delta \vec{r}_c$ ,  $\Delta \theta_c$ , and  $\Delta \phi_c$  to be the translation and rotation of the central filament, we subtract its motion by two steps. First, we rotate the target filament by  $-\Delta \theta_c$  about  $\hat{e}_z \equiv \cos(\Delta \phi_c + \pi/2) \hat{e}_x + \sin(\Delta \phi_c + \pi/2) \hat{e}_y$  from the center of the central filament, where  $\hat{e}_x$  and  $\hat{e}_y$  are the directions of the  $x$  and  $y$  axes. Then we move the target filament by  $-\Delta \vec{r}_c$ . We ignore rotation about the filament axis because in our model the filament

is isotropic. As mentioned above, any step leading to steric overlap is rejected.

The variable time step is chosen to depend as follows on the minimal monomer-monomer distance between two filaments,  $d_{ff} = \min_{i \in a, j \in b} R_{ij}$ . For  $d_{ff} > 120 \text{ \AA}$ , where  $120 \text{ \AA}$  is an outer cutoff distance,  $\Delta t = \max(\Delta t_0, (d_{ff} - 120)^2 / (D_h \times 10^5))$ . This form guarantees that the target filament can move 100 steps without reaching  $d_{ff} = 100 \text{ \AA}$ , the contact distance. The time  $\Delta t_0$  is the basic step used for  $d_{ff} < 120 \text{ \AA}$ . It is chosen so that the deterministic part of the motion is much less than the random part of the motion, in the presence of the interfilament force, and the energy change during a time step is much less than  $kT$ . This results in  $\Delta t_0$  decreasing with increasing  $E_{mm}$ , and in our simulations, the  $\Delta t_0$  values range from 3 to 30 ps. When the filaments contact and align parallel to each other, even very small motions can cause steric overlap so that almost all motion steps are refused. To solve this problem, we separate the motion of the target filament into three independent parts when they contact and align parallel to each other. The three parts are: rotation, sliding along  $\hat{e}_z$  and translation in the  $x, y$  plane. Although most of the rotations cause steric overlap, most sliding steps and a substantial part of the translation steps in the  $x, y$  plane will be accepted. We also decrease the time step by a factor of 15 when the interaction energy reaches 40% of its lowest value.

The bundling criteria are that the interaction energy of two filaments reaches 90% of its lowest value and the  $z$  difference is  $< 5 \text{ \AA}$ . Generally, the sliding time is very short ( $\sim 10^{-5} \text{ s}$ ), and changing the 90% criterion to 80% affects the bundling time only slightly if the filament is not too short.

## RESULTS

We run 150 bundling trajectories (or more) to obtain the average bundling times. Most of the simulations are carried out in the  $4\text{-}\mu\text{m} \times 4\text{-}\mu\text{m} \times 4\text{-}\mu\text{m}$  cell, and the corresponding filament concentration is  $2.596 \times 10^{-11} \text{ M}$ . We also vary the filament concentration to evaluate its effect on the average bundling time. Our calculations show that the bundling rate (defined as the inverse of the average bundling time) is proportional to the filament concentration, as expected.

### Time distribution of collisions

Collisions are prerequisite for bundling. We define a collision as beginning when  $d_{ff} < 100 \text{ \AA}$  (contact distance) and ending when  $d_{ff} > 120 \text{ \AA}$  (escape distance). Decreasing the escape distance increases the number of collisions ( $n_c$ ). When the escape distance equals  $100 \text{ \AA}$ ,  $n_c$  increases abruptly and depends strongly on the time step. For larger escape distances,  $n_c$  does not depend on the time step. We chose  $120 \text{ \AA}$  because it is consistent with our scheme for adjusting the time step. Changing this value does not affect our results significantly.

Fig. 2 shows the number of collisions  $N_c$  before time  $t$  for typical bundling runs using two different interaction strengths. We see that filaments with weak attraction require more time, and also many more collisions, to bundle: 303 for  $E_{mm} = E_{mm}^{\text{ion}}$  and 9 for  $E_{mm} = 8E_{mm}^{\text{ion}}$ . The collisions tend to form clusters, as previously observed by Northup and Erickson (1992). They called the collision clusters *encounters*, and we will also use this terminology.

The clustering of collisions is caused by the random-walk memory, as we discuss in the Appendix. A particle in

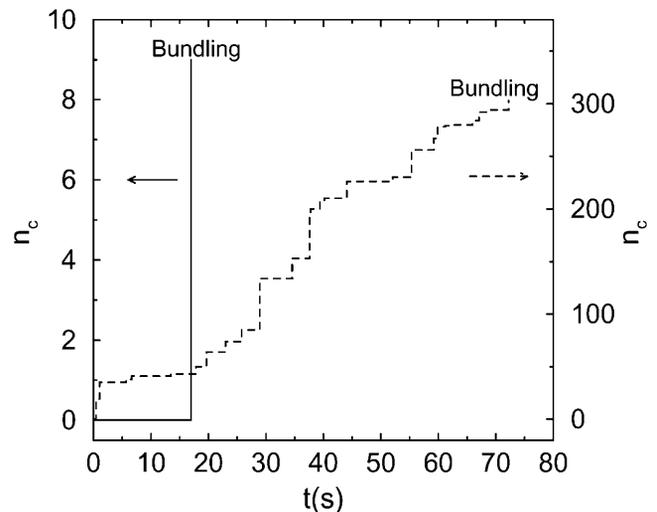


FIGURE 2 Number of collisions  $n_c$  before time  $t$  for two simulation trajectories. (Solid line)  $E_{mm} = 8E_{mm}^{\text{ion}}$ ; (dotted line)  $E_{mm} = E_{mm}^{\text{ion}}$ .  $L = 25$  for both lines. Their lowest interaction energies are  $-0.269 \text{ eV}$  ( $-10.4 \text{ kT}$ ) and  $-2.15 \text{ eV}$  ( $-83.1 \text{ kT}$ ), respectively.

a random walk tends to return to its original position with a probability density proportional to  $\exp[-r^2/(4Dt)]$ , which can be considered as coming from a pseudo-potential  $-Tr^2/(4Dt)$ , where  $r$  is the distance to the original position,  $D$  is its diffusion coefficient, and  $t$  is the time. The resulting clustering increases the contact time so that the filaments have more opportunities to rotate and find the right orientation for bundling. This increases the expected bundling rate by an approximate factor of  $s_c$  (the number of collisions in an encounter) if the orientations between neighboring collisions are not strongly correlated. For example, assuming that only a fraction 0.01 of collisions lead to bundling due to the orientation requirement, it is expected that the bundling rate for  $s_c = 1$  will be 0.01 of the rate obtained without the orientation constraint; but if  $s_c = 100$ , the bundling rate will be almost the same as that without the orientation constraint. In Fig. 2, when  $E_{mm} = 8E_{mm}^{\text{ion}}$ , although the bundling probability per collision is  $1/9$ , the rate is the same as if the bundling probability were 1 because the first and ninth collisions occur almost at the same time. Previous studies (Schlosshauer and Baker, 2002) found that for sticking angular constraints of  $\sim 5\text{--}15^\circ$  of two balls, the reaction rate is  $\sim 2\text{--}3$  orders-of-magnitude higher than expected from a simple geometric model. Earlier calculations also showed that the reduction in reaction rate caused by orientation requirements is significantly less than suggested by the reduction in the probability for a properly oriented collision (Šolc and Stockmayer, 1971, 1973; Schmitz and Schurr, 1972; Shoup et al., 1981; Zhou, 1993).

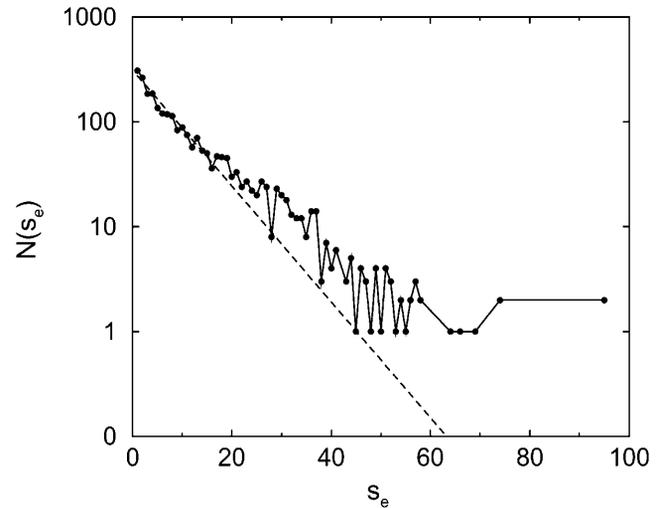
We define an encounter as a cluster of collisions with the time spacing between all sequential collisions as  $< 0.1 \text{ s}$ . Within the same encounter, neighboring collisions have a very short time spacing, which is generally on the order of

microseconds. However, the time spacing between two neighboring encounters can be on the order of seconds or more. In Table 1, we show the calculated averages and variances of collisions and encounters in two time intervals from the 150 simulation trajectories with  $L = 25$  and  $E_{\text{mm}} = E_{\text{mm}}^{\text{ion}}$ . If collisions/encounters occur independently with a uniform probability, they should obey the Poisson distribution, and thus the average should equal the variance. For collisions, the variance is found to be much larger than the average, which shows that some collisions (in the same encounter) are closely correlated. For encounters, the average almost equals the variance. Furthermore the distribution data (not shown) shows that encounters obey the Poisson distribution. Both of the observations suggest that encounters are almost independently distributed.

Fig. 3 shows the distribution of the encounter size  $s_e$ , averaged over 150 trajectories for  $L = 25$  and  $E_{\text{mm}} = E_{\text{mm}}^{\text{ion}}$ . We use 150 trajectories for the other calculated averages as well. From Fig. 3, we can see that the distribution looks reasonably geometric, as expected. We analyze the statistics of an encounter as follows. The first collision plays the role of the seed of the encounter, and it induces the next collision in the same encounter with a probability  $p_{\text{nc}}$ , where *nc* means *next collision*. Similarly, the second induces the third, etc., and at each stage the encounter can end with a probability of  $1 - p_{\text{nc}}$ . Thus the probability that  $s_e = n$  is  $(1 - p_{\text{nc}})p_{\text{nc}}^{n-1}$ , and the average encounter size is  $\bar{s}_e = 1/(1 - p_{\text{nc}})$ . As a first-order approximation, we use one value of  $p_{\text{nc}}$  for all encounters in Fig. 3. We obtain  $p_{\text{nc}} = 0.88$  and  $\bar{s}_e = 8.4$  by fitting the data in Fig. 3, but the value of  $\bar{s}_e$  obtained directly from the simulations is 10.7. We obtain a more refined estimate as follows. We evaluate the dependence of  $p_{\text{nc}}$  on the center-to-center  $z$ -direction displacement of the first collision of an encounter, which we call the *encounter  $z$ -displacement* ( $\Delta z_e$ ), by dividing all encounters (2490) into two groups: *group I* for  $|\Delta z_e| < 150 \text{ \AA}$  (345) and *group II* for others (2145). Fitting gives  $p_{\text{nc}} = 0.93$  and  $\bar{s}_e = 15.0$  for group I, and  $p_{\text{nc}} = 0.871$  and  $\bar{s}_e = 7.7$  for group II. The overall average of  $\bar{s}_e$  is 8.8, closer to the value obtained directly. Thus, collisions with smaller  $\Delta z$  tend to induce larger encounters. Similarly, we study the dependence of  $p_{\text{nc}}$  on the filament-filament angle of the first collision of an encounter ( $\theta_e$ ). For the group with  $\theta_e < 45^\circ$  or  $\theta_e > 135^\circ$  (798 encounters),  $p_{\text{nc}} = 0.87$  and  $\bar{s}_e = 7.5$ ; for the group of the other encounters (1692),  $p_{\text{nc}} = 0.89$  and  $\bar{s}_e = 8.8$ . Thus  $\theta_e$  does not strongly affect  $p_{\text{nc}}$ .

**TABLE 1** Average number of collisions  $n_c$ , encounters  $n_e$ , and the corresponding variances ( $\sigma^2(n_c) \equiv \bar{n}_c^2 - \bar{n}_c^2$  and  $\sigma^2(n_e)$ ) during two different time intervals ( $\delta t$ )

$\delta t$ (s)	$\bar{n}_c$	$\sigma^2(n_c)$	$\bar{n}_e$	$\sigma^2(n_e)$
1.0	3.2	68.5	0.30	0.28
2.0	6.3	133.0	0.60	0.54



**FIGURE 3** Distribution of the encounter size  $s_e$ .  $N(s_e)$  is the number of encounters with  $s_e$  collisions. (Solid line) Results from 150 trajectories for  $L = 25$  and  $E_{\text{mm}} = E_{\text{mm}}^{\text{ion}}$ . (Dotted line) Fit by the geometric distribution  $N(s_e) = 311 \times 0.88^{s_e} = 311 \exp(-s_e/7.86)$ . The geometric distribution is the discrete exponential distribution.

### Flowchart for bundling

A typical simulation trajectory can be visualized according to the model shown in Fig. 4 *a*. It begins with the first free state ( $1F$ ); the target filament collides with an environmental filament. Then it has three options: short free state ( $f$ ), free state ( $F$ ), and bundling ( $B$ ). State  $f$  means the target filament will return to state  $c$  soon, in  $\ll 0.1$  s, whereas state  $F$  means the encounter ends and the target filament will return after at least 0.1 s. The probabilities from state  $c$  to states  $F$ ,  $f$ , and  $B$  are  $p_{cF}$ ,  $p_{cf}$ , and  $p_{cB}$ , respectively. The states labeled with lower-case letters are those with short lifetimes. We define the encounter state ( $E$ ) in Fig. 4 *b* as the combination of states  $c$  and  $f$ . From our results, the time in state  $1F$  is smaller but very close to that in state  $F$ . Thus we combine states  $1F$  and  $F$ , and simplify the flow chart as in Fig. 4 *c*. In Fig. 4 *c*,

$$t_b = t_F + t_E + (1 - p_{EB})t_b. \quad (10)$$

Here the  $t_F$  and  $t_E$  terms account for the time spent in states  $F$  and  $E$ , and the last term accounts for the probability that the filament is recycled to state  $F$ , starting the process over. So

$$t_b = (t_F + t_E)/p_{EB}, \quad (11)$$

where  $t_b$  is the bundling time, and  $t_X$  is the time in state  $X$  ( $X = c, f, F, 1F, E$ ). Our results show that  $t_F$  and  $t_{1F}$  are much larger than  $t_c$ ,  $t_f$ , and  $t_E$ . Therefore

$$t_b \approx t_F/p_{EB}. \quad (12)$$

Similarly, in Fig. 4 *a*,  $t_b = t_{1F} + (t_c + p_{cf}t_f + p_{cF}t_F)/p_{cB}$ , and in Fig. 4 *b*,  $t_b = t_{1F} + (t_E + (1 - p_{EB})t_F)/p_{EB}$ . Thus,

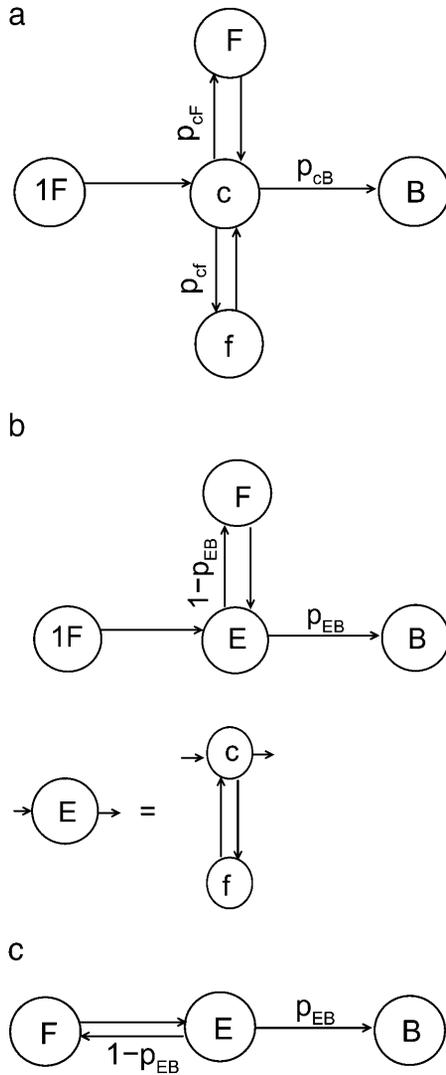


FIGURE 4 Flowchart of filament states.  $1F$ , first free state;  $F$ , free state between two encounters;  $c$ , collision;  $E$ , encounter;  $f$ , short free state between two collisions; and  $B$ , bundling. (a) Flowchart for collisions; (b) flowchart for encounters; and (c) flowchart obtained by combining states  $1F$  and  $F$ . Transition probabilities ( $p_{cf}$ , etc.) are defined in text.

$$t_E = (t_c + p_{cf}t_f)p_{EB}/p_{cB} \quad (13)$$

and

$$(1 - p_{EB})/p_{EB} = p_{cF}/p_{cB}. \quad (14)$$

In Fig. 4 *a*,  $p_{cf}$  equals  $p_{nc}$ , because state  $f$  always induces the next collision in a very short time. In addition,  $p_{cF} + p_{cf} + p_{cB} = 1$ . Therefore,

$$p_{EB} = p_{cB}/(1 - p_{cf}) = p_{cB}/(1 - p_{nc}) = \bar{s}_c p_{cB}, \quad (15)$$

and

$$t_E = \bar{s}_c (t_c + p_{cf}t_f). \quad (16)$$

Eqs. 15 and 16 define the relation between encounters and collisions.

### Bundling probability of collision/encounters

We now consider the dependence of the bundling probability on the angle ( $\theta_c$  or  $\theta_e$ ) and the center-to-center distance ( $d_{cc}$ ) when the collision/encounter happens. Bundling is an orientationally constrained reaction. Therefore, the bundling probability  $p_{cB}$  of a collision, which is the probability that bundling occurs before the collision ends, is sensitive to the collision angle if  $E_{mm}$  is not extremely large. If  $E_{mm}$  is extremely large, every collision leads to bundling and  $p_{cB} = 1$ . We find that the distribution of collision angles ( $\theta_c$ ) is proportional to  $\sin(\theta_c)$ , as expected. The collisions near  $\theta_c = 90^\circ$  are thus the most abundant. The solid line of Fig. 5 *a* shows the dependence of  $p_{cB}$  on  $\theta_c$  for  $L = 25$  and  $E_{mm} = E_{mm}^{ion}$ . Collisions of smaller angle (those near  $180^\circ$  are equivalent to angles near  $0^\circ$ ) have larger bundling probability. The most abundant collisions near  $\theta_c = 90^\circ$  have nearly zero bundling probability. But the situation is quite different for encounters. The encounter angle  $\theta_e$  has no significant effect on  $p_b$ . The encounters near  $\theta_e = 90^\circ$  can lead to bundling, because the first collision of such an encounter can induce a collision of smaller angle and thus a larger bundling probability. As expected, the bundling probability of an encounter is much larger than that of a collision. Increasing the attraction increases the bundling probability and widens the range of collision angles that allow bundling (Fig. 5 *b*). The dot-dashed line for strong attraction  $E_{mm} = 8E_{mm}^{ion}$  is much more bumpy, because a bundling event includes much fewer collisions. Thus the total number of collisions is much less than those of the other two cases, and the sampling is insufficient. Similarly, curves for encounters are more bumpy than the corresponding ones for collisions in Fig. 5, *a* and *c*.

Fig. 5 *c* shows that collisions/encounters with smaller  $d_{cc}$  tend to have larger bundling probabilities. Compared with the encounter angle,  $d_{cc}$  has a much stronger effect on  $p_{EB}$ . This is due to its effect on the encounter size.  $p_{cB}$  is also enhanced by smaller  $d_{cc}$  because the geometry is closer to the final bundled geometry. We extract the encounters with  $d_{cc} < 200 \text{ \AA}$ , and we find that their averaged size is 13.7, which is significantly larger than the global average of 10.7. Then we extract the encounters of  $\theta < 45^\circ$ , and find an average size of 10.5, very close to the global average 10.7. The encounter  $d_{cc}$  thus affects the encounter size and therefore the bundling probability, whereas the encounter angle does not.

### Bundling process

The bundling process is seen most clearly in the case of strong attraction ( $E_{mm} = 8E_{mm}^{ion}$ ), shown in Fig. 6 and the movie as supplemental material. Stronger attraction in-

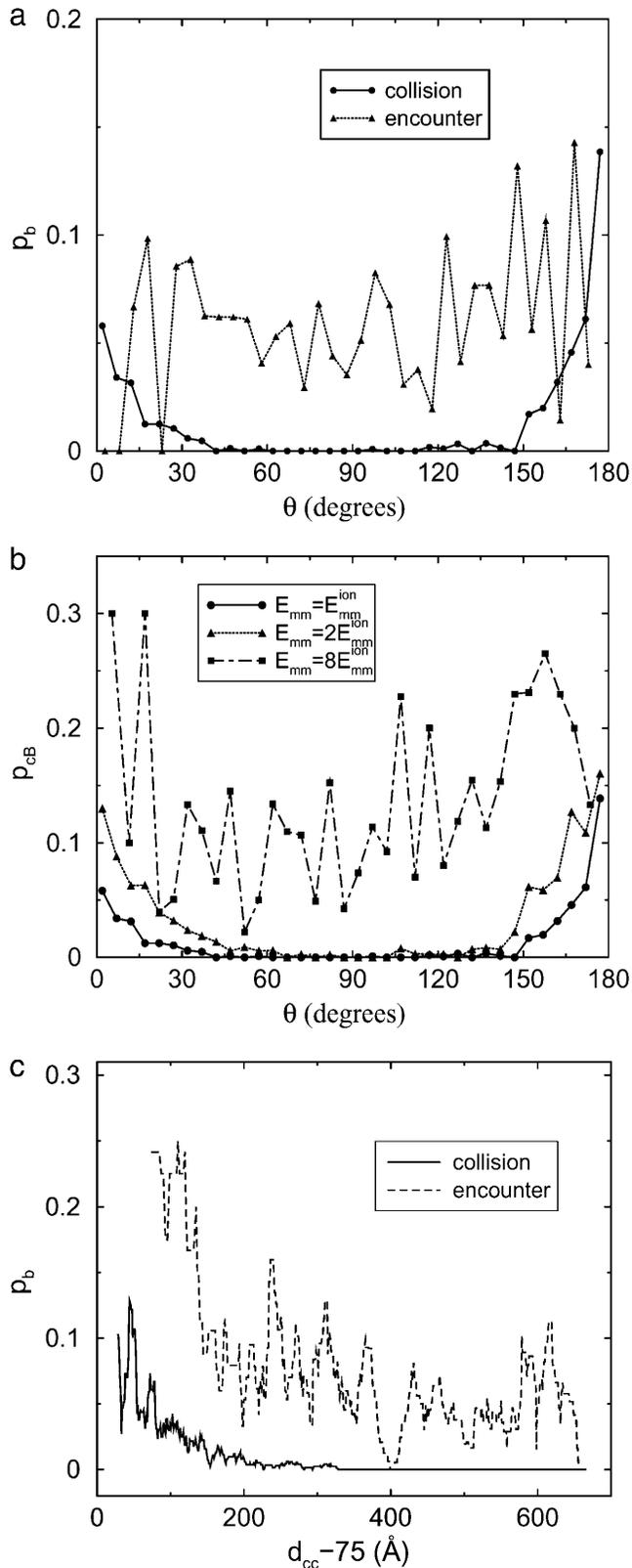


FIGURE 5 Dependence of bundling probability on filament geometry when collision/encounter begins. The value  $p_b$  is the probability that bundling happens in a given collision/encounter ( $p_{cB}$  or  $p_{EB}$ ). The value  $\theta$  is the angle between two filaments when they begin to collide. (a) Dependence

increases the duration of a collision, so it can induce more changes in the filaments' relative position and orientation. In Fig. 6 *a*, the energy fluctuates randomly before  $t = 7.2 \mu\text{s}$ ; then it steeply drops to  $-1.2 \text{ eV}$  and finally it drops more slowly, with fluctuations. In Fig. 6 *b*,  $\theta$  also fluctuates before  $t = 7.2 \mu\text{s}$ ; then it quickly drops to near  $0^\circ$  and stays there. The situation for the  $z$ -displacement  $\Delta z$  is similar to that for  $\theta$ . However, the drop in  $\Delta z$  occurs later than the drop in  $\theta$ . It fluctuates until  $E = -1.2 \text{ eV}$  and then drops to zero (bundled state). These results clearly show that there are two phases: rotation and sliding. The initial rotation appears to be random. When  $\theta$  reaches a certain limit, two filaments quickly align parallel to each other, and the energy drops steeply. Then the sliding begins and leads to bundling.

For weak attraction, a collision lasts a short time, and the driving force for rotation is weak, so only small geometric changes occur during a collision. Therefore, a high bundling probability for a collision requires small geometric differences in the collision angle and  $d_{cc}$ , as shown previously in Fig. 5. In Fig. 7, we show the final encounter of a typical such trajectory. Compared with Fig. 6 *b* which has  $\theta = 40^\circ$  and  $\Delta z = 245 \text{ \AA}$ , the final collision, which leads to bundling, has smaller values:  $\theta = 8^\circ$  and  $\Delta z = 20 \text{ \AA}$ . Many collisions are required to obtain a small-angle collision so that the filaments can bundle.

### Bundling time

The bundling time depends on the attractive interaction, the filament length, and the diffusion coefficients, which are in turn determined by the filament geometry (length and radius) and the solution viscosity. In addition to performing simulations with the correct parameters, we also artificially change a single parameter at a time to evaluate its role in bundling. In our simulations, most of the trajectories, including almost all of those with the correct parameters, end in a bundled state, but some of the artificial simulations, especially those of extremely small  $D_r$ , do not always end with bundling within the time limit of our simulations. We obtain the bundling time  $t_b$  by dividing the total running time of all trajectories by the number of bundling trajectories ( $N_b$ ):  $t_b = \sum_i t_i / N_b$ . This is equivalent to assuming that those trajectories which do not bundle, on average, need an extra time  $t_b$  to bundle, i.e.,  $Nt_b = \sum_i t_i + N_{nb}t_b$ .

The model given in Fig. 4 can help us analyze the bundling time. There are two timescales in a trajectory: the encounter timescale and the collision timescale. The trajectory consists of encounters and an encounter consists of collisions. We then define the trajectory size  $s_t$  as the number

of  $p_b$  on  $\theta$ . (b) Dependence of  $p_{cB}$  on  $\theta$  and the maximal attractive energy  $E_{mm}$  of two monomers. (c) Dependence of  $p_b$  on the center-to-center distance  $d_{cc}$  of two filaments.  $L = 25$  for *a-c* and  $E_{mm} = E_{mm}^{\text{ion}}$  for all unlabeled curves.

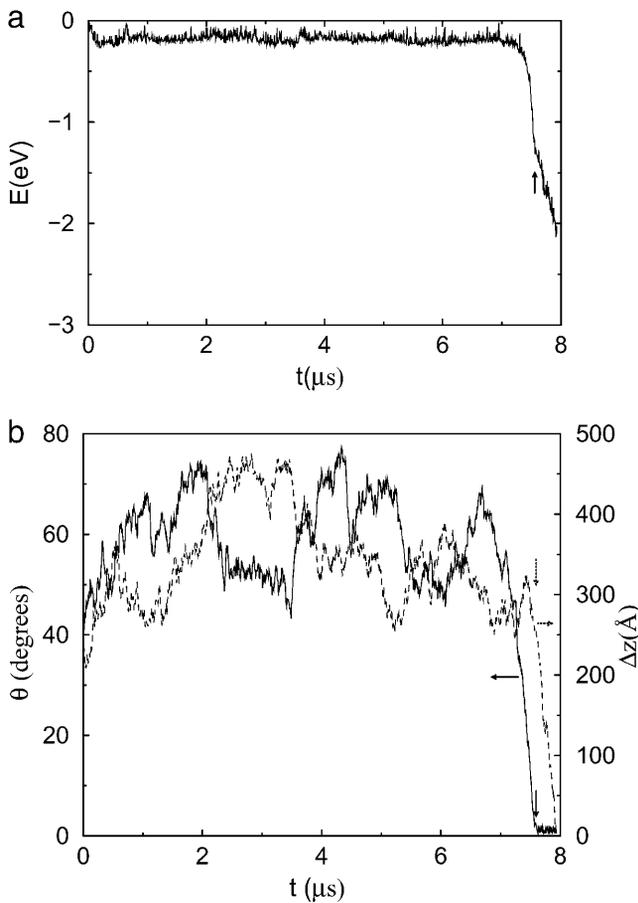


FIGURE 6 Bundling for strong attraction  $E_{mm} = 8E_{mm}^{ion}$  and  $L = 25$ . (a) Energy during last collision, which leads to bundling. The value  $t = 0$  is the beginning time for this collision. (b) Angle  $\theta$  (solid line) and displacement  $\Delta z$  along the  $z$  axis between the centers of two filaments (dotted line) during the last collision. Downward arrows mark separation of the rotation and sliding phases of bundling.

of encounters in a trajectory, and  $s_t$  should obey the geometric distribution according to the previous analysis of  $s_e$ . In addition, the distribution of the time in state  $F$ ,  $t_F$ , obeys a continuous geometric (or exponential) distribution. For the 150 trajectories for  $L = 25$  and  $E_{mm} = E_{mm}^{ion}$ , the average  $t_F$  is 16.2 s and the standard deviation is 15.4 s; for an exactly exponential distribution, their values would be identical. In Fig. 4, the time in state  $E$  can be ignored in comparison with  $t_F$  according to our results, so

$$t_b = \sum_i^{s_t} t_{Fi}, \quad (17)$$

where  $t_{Fi}$  is the  $i^{\text{th}}$   $t_F$  in a given trajectory. Thus the distribution of bundling times of a set of trajectories should be a geometric distribution. The average bundling time of the 150 trajectories for  $L = 25$  and  $E_{mm} = E_{mm}^{ion}$  is 55.9 s and the standard deviation is 55.1 s, close to the average. This property strongly suggests an exponential distribution,

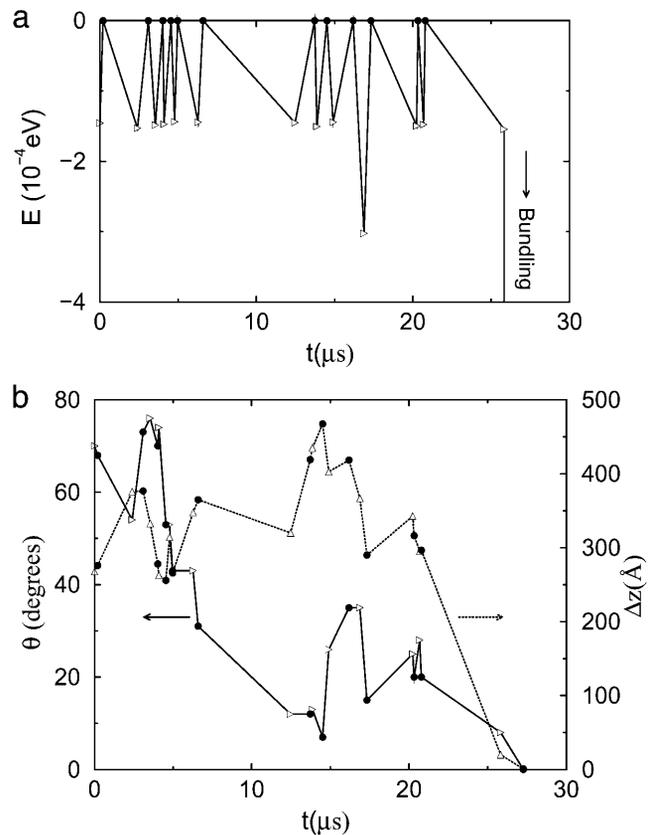


FIGURE 7 Bundling for weak attraction  $E_{mm} = E_{mm}^{ion}$  and  $L = 25$ . Energy (a), angle  $\theta$  (b, solid line), and  $z$ -displacement  $\Delta z$  (b, dotted line) during last encounter. (Open triangles) Collisions begin; (solid circles) collisions end. Last collision leads to bundling.

which is confirmed by fitting the distribution. The error of the bundling time of one trajectory is exactly the real average bundling time  $\hat{t}_b$ . Thus the error of the average bundling time of  $N_{\text{traj}}$  trajectories is  $\hat{t}_b / \sqrt{N_{\text{traj}}}$ , i.e., the relative error is  $1 / \sqrt{N_{\text{traj}}}$ . We keep  $N_{\text{traj}} \geq 150$ , so relative errors are within 8%.

Fig. 8 shows the dependence of bundling time (Fig. 8 a) and its corresponding rate constant (Fig. 8 b) on attraction strength and filament length. In Fig. 8 a, increasing the attraction between the filaments decreases the bundling time. There is a diffusion limit for the bundling time. If the attraction is strong enough,  $p_{EB} = 1$  and  $t_b = t_F$  according to Eq. 12;  $t_F$  is determined entirely by the diffusion coefficients and the filament length. The dependence of  $t_b$  on filament length ( $L$ ) is weak. Although increasing the filament length can expand the capture region and thus decrease  $t_F$  and increase  $s_e$ , it also decreases the diffusion coefficients, which increases  $t_F$ . If the translational diffusion coefficients  $D_h$  and  $D_v$  changed by the same factor as the rotational diffusion coefficient  $D_r$ , adjusting the time step  $\Delta t$  by the inverse of this factor would leave Eqs. 6–10 unchanged and the bundling probability for a collision ( $p_{cB}$ ) would be the same. But increasing  $L$  decreases  $D_r$  more than  $D_h$  and  $D_v$ , by

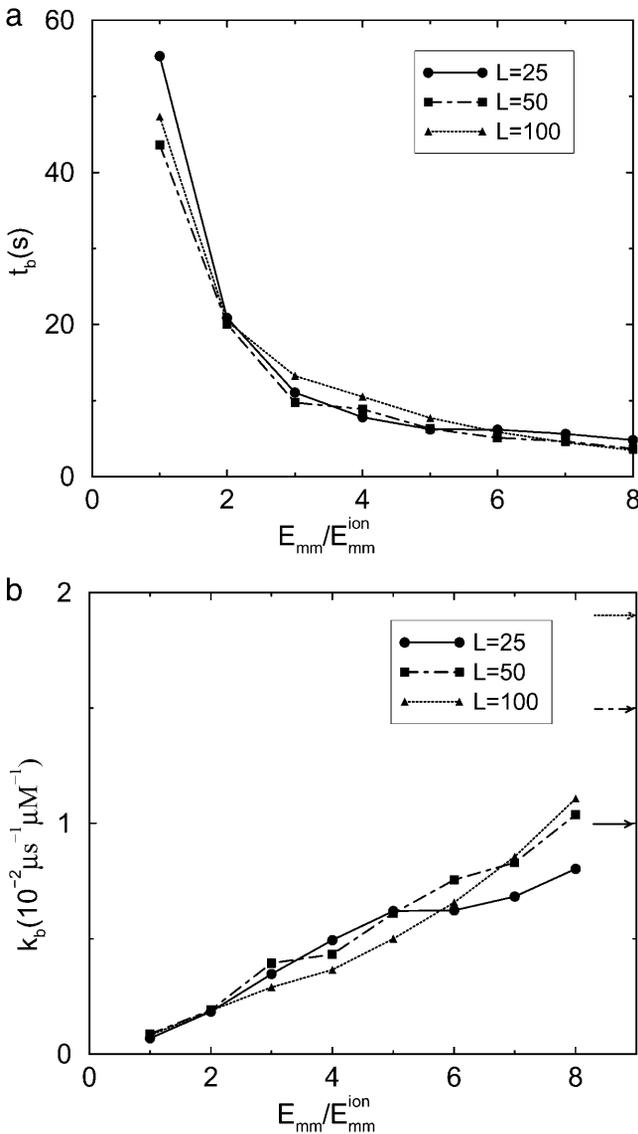


FIGURE 8 Dependence of bundling time  $t_b$  (a) and its corresponding rate constant  $k_b$  (b) on filament length and attraction. Increasing potential decreases the bundling time. Large  $D_r$  and  $E_{mm}$  give upper limits for bundling rate constants, whose values for the three lengths are marked by arrows in b.

a factor of  $L^{-2}$ . This results in a decrease of  $p_{cB}$  because  $p_{cB}$  is determined by the competition between rotation (dominated by  $D_r$ ) and escape (controlled mainly by  $D_v$  and affected less by  $D_r$  and  $D_h$ ). Because  $p_{EB} = \bar{s}_e p_{cB}$  (Eq. 15), the dependence of  $p_{EB}$  on  $L$  is also determined by the same two competing factors as that of  $t_F$  on  $L$ . Therefore,  $t_b \sim t_F/p_{EB}$  is only weakly affected by  $L$ . The dependences of the bundling parameters on  $L$  at  $E_{mm} = E_{mm}^{ion}$  are shown in Table 2. The dependences are weak, as expected. We also see  $t_{1F} \sim t_F$ , supporting our earlier analysis.

Increasing  $D_r$  should increase  $p_{EB}$  by increasing the orientational search speed, and decrease  $t_F$  by increasing the collision rate. Therefore, increasing  $D_r$  should decrease the bundling time  $t_b$ . This expectation is confirmed by the

TABLE 2 Dependence of bundling parameters on  $L$  for

$L$	$s_e$	$1/p_{cB}$	$1/p_{EB}$	$t_F$ (s)	$t_{1F}$ (s)	$t_b$ (s)
25	10.7	178	16.7	3.33	2.91	55.3
50	17.9	285	15.9	2.73	2.80	43.6
100	33.5	702	20.9	2.27	2.19	47.3

$s_e$ , average number of collisions in an encounter;  $1/p_{cB}$ , average number of collisions needed for bundling;  $1/p_{EB}$ , average number of encounters needed for bundling;  $t_F$ , average time in state  $F$ ;  $t_{1F}$ , average time in state  $1F$ ; and  $t_b$ , average bundling time.

numerical results for the dependence of  $t_b$ ,  $t_F$ , and  $p_{EB}$  on  $D_r$ , shown in Table 3. The inverse of  $p_{EB}$  is the number of encounters needed for bundling, which decreases with increasing  $D_r$ . This effect is stronger than the decrease in  $t_F$ .

If  $D_r$  and  $E_{mm}$  are large enough, the two filaments will contact and bundle almost as soon as  $d_{cc} = L_f$ , since the orientational search will be faster than the translational search. We use a simple model to estimate the bundling rate constant in this limit. In this model, an environmental filament is centered in a sphere of radius  $R_s \gg L_f$  and the target filament is randomly located at  $r$  ( $r < R_s$ ). When  $d_{cc} = L_f$ , the target is captured. When the target is captured, bundling is almost instantaneous, so  $t_b$  is the capture time  $w(r)$ . If the target moves out of the sphere, it is assumed to run into another sphere, so  $dw/dr|_{r=R_s} = 0$ . Therefore, according to Gardiner (1985),

$$\nabla^2 w + 1/\bar{D} = 0, \quad (18)$$

with

$$w(L_f) = 0 \quad \text{and} \quad w'(R_s) = 0,$$

where  $\bar{D} \sim 2.0 \times (D_h + 2D_v)/3.0$  is twice the average translational diffusion coefficient (twice because of the relative motion). The solution is

$$w(r) = -\frac{r^2}{6\bar{D}} - \frac{R_s^3}{3\bar{D}r} + \frac{L_f^2}{6\bar{D}} + \frac{R_s^3}{3\bar{D}L_f}. \quad (19)$$

TABLE 3 Dependence of bundling time  $t_b$ , bundling probability  $p_{EB}$  of an encounter and time  $t_F$  in state  $F$  on  $D_r$  for  $L = 25$  and  $E_{mm} = E_{mm}^{ion}$

$D_r$	$t_F$ (s)	$1/p_{EB}$	$t_b$ (s)
0.01	4.95	135	686
0.03	4.61	77.3	443
0.1	4.26	41.5	178
0.3	3.85	25.2	97
1	3.32	16.5	55
3	2.91	12.9	35
10	2.53	11.7	29

We vary  $D_r$  artificially while keeping other parameters fixed. Units of  $D_r$  are chosen such that original value is 1.

The average of  $w(r)$  shows that the last term dominates, i.e.,  $t_b \sim V/4\pi\bar{D}L_f$  where  $V = 4\pi R_s^3/3$  is the volume of the sphere. So the bundling rate constant  $k_b$  for large  $D_r$  and  $E_{mm}$  is  $4\pi\bar{D}L_f$ . The values for  $L = 25, 50,$  and  $100$  are  $1.0 \times 10^{-2}, 1.5 \times 10^{-2},$  and  $1.9 \times 10^{-2} \mu\text{M}^{-1} \mu\text{s}^{-1}$ , which are marked as arrows in Fig. 8 *b*. These limits are consistent with the numerical results. Because the effect of the reduced  $D_r$  coming from increasing  $L$  is ignored in the limits, the dependence of  $k_b$  on  $L$  is stronger than that in the numerical results, but still weak.

### Two-step potential well model

To explain the dependence of  $t_b$  on  $E_{mm}$ , we use the two-step potential well model shown in Fig. 9 *a*. Particles reaching  $x = 0$  are regarded as bundled and the particles are restricted to the region  $x \leq X_c$ . The first well of depth  $-E_1$  represents the rotation phase, and the second well, of depth  $-E_2$  relative to the bottom of the first well, represents the sliding phase. When two filaments collide, the particle enters the first well; when they rotate and align parallel with each other, the particle drops into the second well. The first well thus shuttles the particle to the second well. The capture time for the particle at  $x$ ,  $w(x)$ , satisfies Gardner (1985),

$$w''(x) + \beta F w'(x) + 1/D = 0, \quad (20)$$

with

$$w(0) = 0 \quad \text{and} \quad w'(X_c) = 0,$$

where  $\beta$  is inverse temperature,  $F = -(\partial E)/(\partial x)$  is the force, and  $D$  is the diffusion coefficient. Letting  $u(x) \equiv Dw(x)$ ,

$$u''(x) + \beta F u'(x) + 1 = 0, \quad (21)$$

with

$$u(0) = 0 \quad \text{and} \quad u'(X_c) = 0.$$

The solution for  $x > x_1$  is

$$u(x) = \exp(-\beta E_2) \exp(-\beta E_1) u(x; 0, 0) + \exp(-\beta E_2) (1 - \exp(-\beta E_1)) u(x; +\infty, 0) + (1 - \exp(-\beta E_2)) \times \exp(-\beta E_1) u(x; 0, +\infty) + (1 - \exp(-\beta E_2)) (1 - \exp(-\beta E_1)) u(x; +\infty, +\infty), \quad (22)$$

where  $u(x; a, b)$  is the solution for  $E_1 = a$  and  $E_2 = b$  ( $a$  and  $b$  can be 0 or  $\infty$  here). The  $u$  functions in Eq. 22 are easily obtained, but we do not specify them here because they are not needed for our result.

Usually, the second well is much deeper and narrower than the first, and two filaments in the sliding phase almost always bundle ( $\beta E_2 \gg 1$ ). The first two terms can thus be ignored. So

$$u(x) \approx \exp(-\beta E_1) u(0, +\infty) + (1 - \exp(-\beta E_1)) u(+\infty, +\infty), \quad (23)$$

and the bundling time

$$t_b = t_\infty + (t_0 - t_\infty) \exp(-\beta E_1), \quad (24)$$

where  $t_\infty$  is the bundling time for  $E_1 = \infty$  and  $t_0$  is for  $E_1 = 0$ . We can assume that  $E_1$  is proportional to  $E_{mm}$ , i.e.,  $\beta E_2 = E_{mm}/E_d$ , where  $E_d$  is a constant, and

$$t_b = t_\infty + (t_0 - t_\infty) \exp(-E_{mm}/E_d). \quad (25)$$

Thus dependence on  $E_{mm}$  is exponential. From Fig. 9 *b*, we can see that Eq. 25 fits the simulation results closely.

## CONCLUSIONS

In summary, we have found the following. In a bundling event, collisions are clustered in time. Collisions of smaller angles and smaller  $d_{cc}$  are more likely to lead to bundling. The bundling process consists of two phases: rotating and sliding. The bundling rate depends only weakly on filament length. Increasing the attraction decreases the bundling time to a lower limit, which is determined by diffusion properties. A simple two-step-well model predicts an exponential dependence of the bundling time on the interaction strength, which is confirmed by the simulation results.

## APPENDIX: COLLISION CLUSTERING

In this Appendix, we show, by studying a simple system, that the phenomenon of collision clustering (Northup and Erickson, 1992) is a universal and a direct property of random walks. In this system, two noninteracting particles are restricted to a one-dimensional lattice of unit

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spacing ranging from  $-X_m$  to  $X_m$ , where one is fixed at the center  $x = 0$ , and the other moves randomly. When the free particle meets the fixed one, a collision is counted. The free particle is reflected at the boundaries. The initial position of the free particle is randomly chosen. Fig. 10 shows a typical simulation trajectory for  $X_m = 100$ . Although there is no interaction between

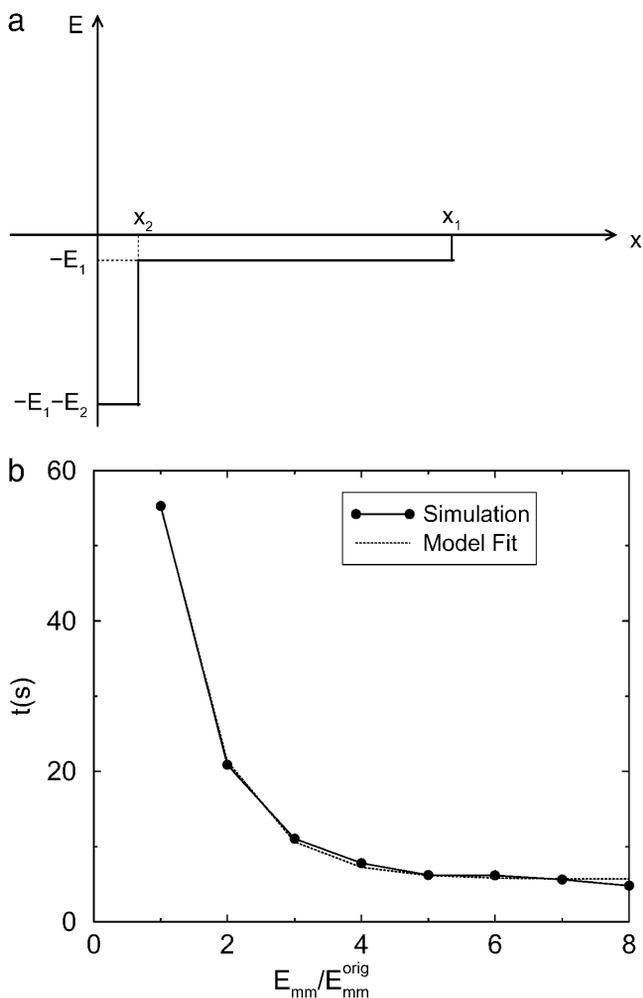


FIGURE 9 Two-step potential well model (a) and fitting results (b) for  $L = 25$  using Eq. 25. The fit parameters are  $t_{\infty} = 5.68$  s,  $t_0 = 162.2$  s, and  $E_d = 0.87 E_{mm}^{\text{orig}}$ .

the two particles, collision clustering is obvious in Fig. 10 a, which shows the times of the first 200 sequential collisions.

From this simulation trajectory, we obtain the distribution of time spacings  $t_{cc}$  between two successive collisions in Fig. 10 b. The  $t_{cc}$  distribution can be also calculated by the following method. First, we find all paths, from  $x = 0$  to  $x = 0$ , of length  $t_{cc}$ ; then we sum up the probabilities of these paths and obtain the probability of  $t_{cc}$ . For a path, the free particle at each point except the ending point and the boundary points can have two choices, so the probability of this path is  $2^{-n}$ , where  $n$  is the number of these points. As an example, for  $t_{cc} = 2$ , there are two paths:  $0 \rightarrow 1 \rightarrow 0$  ( $2^{-2}$ ) and  $0 \rightarrow -1 \rightarrow 0$  ( $2^{-2}$ ), where the numbers in parentheses are the corresponding probabilities. Thus, the probability of  $t_{cc} = 2$  is  $1/2$ . Similarly, for  $t_{cc} = 4, 6, 8$ , and  $10$ , the probabilities are  $1/8, 1/16, 5/128$ , and  $7/256$ , respectively. These values are consistent with the simulation data in frame b. This distribution has two features. First, most of the probability is concentrated at small  $t_{cc}$  values, and the probability decreases with increasing  $t_{cc}$ . Thus, two successive collisions tend to cluster with small  $t_{cc}$  values. Second, the decrease slows down quickly with increasing  $t_{cc}$ , which is shown by the ratios of  $P(t_{cc})$  to  $P(t_{cc}-2)$  in Fig. 10 b. Thus, there are still substantial probabilities for large  $t_{cc}$  values. For instance, there is a  $t_{cc}$  of  $\sim 10,000$  steps in Fig. 10 a. This feature guarantees that two successive collision clusters can be separated by large  $t_{cc}$  values, with a substantial probability. Collision clustering naturally results from these two properties.

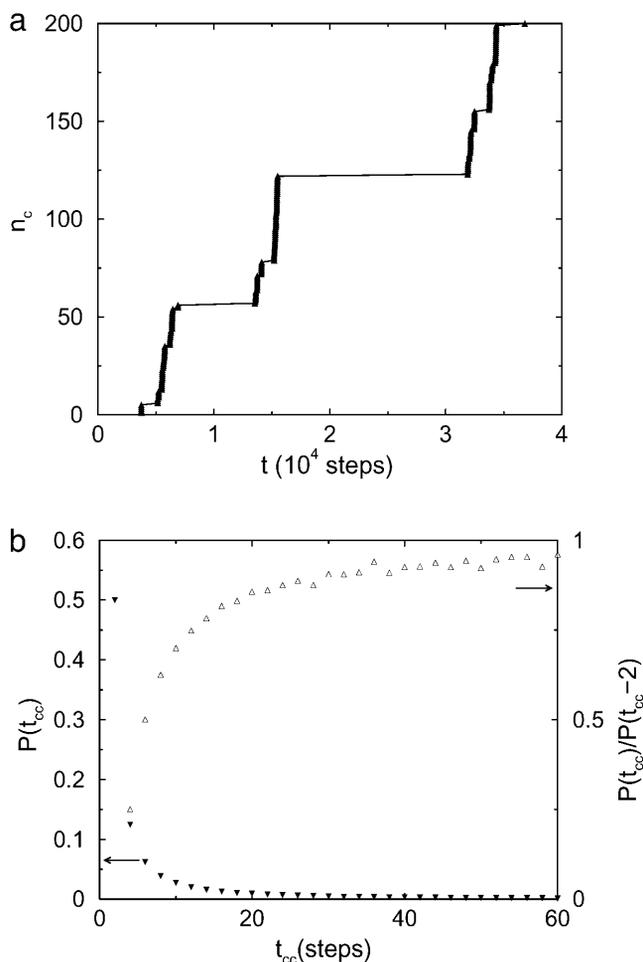


FIGURE 10 (a) Number of collisions  $n_c$  before time  $t$ , in which solid triangles show times of the first 200 sequential collisions in a simulation trajectory with  $X_m = 100$  and  $10^9$  steps; (b) probability distribution  $P(t_{cc})$  of  $t_{cc}$ , the time spacing between two successive collisions, in the same simulation trajectory. Note that probability of odd  $t_{cc}$  is zero. For large, even  $t_{cc}$  values, the probabilities are close to zero, but are not zero. Probabilities of  $t_{cc}$  values  $> 60$  steps are not shown. Open triangles in b are probability ratios.

We thank Le Yang, Jie Zhu, and Professors Rohit Pappu and David Sept, for informative discussions. We also thank the referees for helpful comments.

The work was supported by the National Science Foundation under grant DMS-0240770.

## REFERENCES

- Baeza, I., P. Garislio, L. M. Rangel, P. Chavez, L. Cervants, C. Arguello, C. Wong, and C. Montanez. 1987. Electron microscopy and biochemical properties of polyamines-compacted DNA. *Biochemistry*. 26:6387–6392.
- Bloomfield, V. A. 1996. DNA condensation by multivalent cations. *Curr. Opin. Struct. Biol.* 6:334–341.
- Bloomfield, V. A. 1997. Polyelectrolyte effects in DNA condensation by polyamines. *Biophys. Chem.* 11:339–343.

- Deserno, M., and C. Holm. 2002. Theory and simulations of rigid polyelectrolytes. *Mol. Phys.* 100:2941–2956.
- Diehl, A., H. A. Carmona, and Y. Levin. 2001. Counterion correlations and attraction between like-charged macromolecules. *Phys. Rev. E.* 64: 011804-1–011804-6.
- Doi, M., and S. F. Edwards. 1986. *The Theory of Polymer Dynamics*. Oxford University Press, New York. Chap. 8.
- Gardiner, C. W. 1985. *Handbook of Stochastic Methods for Physics, Chemistry, and the Natural Sciences*. Springer-Verlag, New York. Chap. 5.
- Gelbart, W. M., R. F. Bruinsma, P. A. Pincus, and V. A. Parsegian. 2000. DNA-inspired electrostatics. *Phys. Today.* 53:38–44.
- Grønbech-Jensen, N., R. J. Mashl, R. F. Bruinsma, and W. M. Gelbart. 1997. Counterion-induced attraction between rigid polyelectrolytes. *Phys. Rev. Lett.* 78:2477–2480.
- Ha, B.-Y., and A. J. Liu. 1997. Counterion-mediated attraction between two like-charged rods. *Phys. Rev. Lett.* 79:1289–1292.
- Kawamura, M., and K. Maruyama. 1970. Polymorphism of F-actin. I. Three forms of paracrystals. *J. Biochem. (Tokyo).* 68:885–899.
- Khokhlov, A. R., and A. N. Semenov. 1985. On the theory of liquid-crystalline ordering of polymer chains with limited flexibility. *J. Stat. Phys.* 38:161–182.
- Komyshv, A. A., and S. Leikin. 1998. Electrostatic interaction between helical macromolecules in dense aggregates: an impetus for DNA poly- and meso-morphism. *Proc. Natl. Acad. Sci. USA.* 95:13579–13584.
- Lau, A. W. C., and P. Pincus. 2002. Counterion condensation and fluctuation-induced attraction. *Phys. Rev. E.* 66:041501-1–041501-14.
- Manning, G. S. 2003. Comments on selected aspects of nucleic acid electrostatics. *Biopolymers.* 69:137–143.
- Moreira, A. G., and R. R. Netz. 2001. Binding of similarly charged plates with counterions only. *Phys. Rev. Lett.* 87:078301-1–078301-4.
- Northrup, S. H., and H. P. Erickson. 1992. Kinetics of protein-protein association explained by Brownian dynamics computer simulation. *Proc. Natl. Acad. Sci. USA.* 89:3338–3342.
- Oosawa, F. 1968. Interaction between parallel rodlike macroions. *Biopolymers.* 5:1633–1647.
- Ray, J., and G. S. Manning. 1994. An attractive force between two rodlike polyions mediated by sharing of condensed counterions. *Langmuir.* 10:2450–2461.
- Rochet, J. C., and P. T. Lansbury. 2000. Amyloid fibrillogenesis: themes and variations. *Curr. Opin. Struct. Biol.* 10:60–68.
- Schlosshauer, M., and D. Baker. 2002. A general expression for bimolecular association rates with orientational constraints. *J. Phys. Chem. B.* 106:12079–12083.
- Schmitz, K. S., and J. M. Schurr. 1972. Role of orientation constraints and rotational diffusion in bimolecular solution kinetics. *J. Phys. Chem.* 76:534–545.
- Sear, R. P. 1997. Cohesion and aggregation of flexible hard rods with an attractive interaction. *Phys. Rev. E.* 55:5820–5824.
- Shklovskii, B. I. 1999. Screening of a macroion by multivalent ions: correlation-induced inversion of charge. *Phys. Rev. E.* 60:5802–5811.
- Shoup, D., G. Lipari, and A. Szabo. 1981. Diffusion-controlled bimolecular reaction rates. The effect of rotational diffusion and orientation constraints. *Biophys. J.* 36:697–714.
- Šolc, K., and W. H. Stockmayer. 1971. Kinetics of diffusion-controlled reaction between chemically asymmetric molecules. I. General theory. *J. Chem. Phys.* 54:2981–2988.
- Šolc, K., and W. H. Stockmayer. 1973. Kinetics of diffusion-controlled reaction between chemically asymmetric molecules. II. Approximate steady-state solution. *Int. J. Chem. Kinet.* 5:733–752.
- Stevens, M. J. 2001. Simple simulations of DNA condensation. *Biophys. J.* 80:130–139.
- Tang, J. X., and P. A. Janmey. 1996. The polyelectrolyte nature of F-actin and the mechanism of actin bundle formation. *J. Biol. Chem.* 271:8556–8563.
- Tang, J. X., S. Wong, P. T. Tran, and P. A. Janmey. 1996. Counterion induced bundle formation of rodlike polyelectrolytes. *Ber. Bunsenges. Phys. Chem.* 100:796–806.
- van der Schoot, P., and T. Odijk. 1992. Statistical theory and structure factor of a semidilute solution of rodlike macromolecules interacting by van der Waals forces. *J. Chem. Phys.* 97:515–524.
- Yu, X. P., and A. E. Carlsson. 2003. Multiscale study of counterion-induced attraction and bundle formation of F-actin using an Ising-like mean field model. *Biophys. J.* 85:3532–3543.
- Zhou, H.-X. 1993. Brownian dynamics study of the influences of electrostatic interaction and diffusion on protein-protein association kinetics. *Biophys. J.* 64:1711–1726.