On the biophysics and kinetics of toehold-mediated DNA strand displacement

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ABSTRACT
Dynamic DNA nanotechnology often uses toehold-mediated strand displacement for controlling reaction kinetics. Although the dependence of strand displacement kinetics on toehold length has been experimentally characterized and phenomenologically modeled, detailed biophysical understanding has remained elusive. Here, we study strand displacement at multiple levels of detail, using an intuitive model of a random walk on a 1D energy landscape, a secondary structure kinetics model with single base-pair steps and a coarse-grained molecular model that incorporates 3D geometric and steric effects. Further, we experimentally investigate the thermodynamics of three-way branch migration. Two factors explain the dependence of strand displacement kinetics on toehold length: (i) the physical process by which a single step of branch migration occurs is significantly slower than the fraying of a single base pair and (ii) initiating branch migration incurs a thermodynamic penalty, not captured by state-of-the-art nearest neighbor models of DNA, due to the additional overhang it engenders at the junction. Our findings are consistent with previously measured or inferred rates for hybridization, fraying and branch migration, and they provide a biophysical explanation of strand displacement kinetics. Our work paves the way for accurate modeling of strand displacement cascades, which would facilitate the simulation and construction of more complex molecular systems.

INTRODUCTION
Recent advances in DNA nanotechnology have enabled the construction of 2D and 3D nanoscale structures (1–7). Nucleic acids have predictable double-helical structure and generally well-understood thermodynamic (8–11) and mechanical (12) properties, which makes them excellent engineering materials. In addition to static structures, dynamic nanoscale devices such as circuits (13–16), catalysts (17,18), autonomous molecular motors (19–22) and reconfigurable nanostructures (18,23–25) have been engineered using DNA. Inspired by experimental advances, theoretical schemes have been proposed (26,27) to engineer arbitrarily complex chemical dynamics using DNA. If successful, such efforts could enable dynamic DNA circuits to actively control nanoscale devices.

Unfortunately, the biophysical understanding of key kinetic phenomena remains underdeveloped relative to our knowledge of static properties, limiting the development of dynamic DNA nanotechnology. Here, we study the biophysical basis of a molecular mechanism called toehold-mediated strand displacement, which is central to many dynamic DNA devices built to date. Toehold-mediated strand displacement enables control over the kinetics of molecular rearrangement, allowing the engineer to program when and where specific steps take place in a molecular machine.

Figure 1 illustrates strand displacement using domain notation. A domain is a set of contiguous nucleotides designed to be either fully bound or fully unbound in...
stable configurations. Our system initially comprises a two-stranded complex ($S$) and a single-stranded ‘invader’ ($X$). $S$ consists of an ‘incumbent’ strand ($Y$) bound to a ‘substrate’ strand, which has a single-stranded overhang called a ‘toehold’. The invader is fully Watson–Crick complementary to the substrate and may bind reversibly to it using the toehold domain ($h$). This binding is reversible because the toehold may ‘fray’ and eventually dissociate. (We use the term ‘fraying’ to describe the disruption of base pairs at the end of a duplex; if all base pairs fray, the duplex melts or dissociates. Conversely, ‘zipping’ refers to when a new base pair forms at the end of an existing duplex.) Once the toehold is bound, the ‘overhanging’ branch migration domain ($h$) of the invader may compete with the incumbent for binding with the substrate. As the incumbent and invader exchange base pairs with the substrate, the branch point of the three-stranded complex moves back and forth. This ‘three-way branch migration’ (henceforth, ‘branch migration’) process has previously been modeled as an unbiased random walk (28–30), as each step causes no net change in base pairing. Eventually, the incumbent may dissociate, completing strand displacement. Overall, displacement is thermodynamically driven forward by the net gain in base pairs due to the toehold.

Although it is known that bimolecular rate constants for strand displacement can vary a million-fold depending on the length of the toehold, little is known about the underlying biophysics or how to generalize the results to other reaction conditions or molecular modifications. For example, what are the mechanisms and intermediate states involved in a single step of branch migration? How do the kinetics of strand displacement depend on the length of the branch migration domain or on the temperature and buffer conditions? How different is strand displacement in RNA and DNA? What is the effect of sequence mismatches at different positions along the branch migration domain? To begin answering these questions, a fundamental biophysical understanding is necessary.

However, the commonly accepted view of strand displacement biophysics appears at odds with kinetic measurements (30,31). Consider strand displacement with a 1-base toehold and a 20-base branch migration domain. Let us assume that the rate constant for the formation of the toehold base pair is on the order of $10^6$/M/s. [This value for 1-nt association is within a factor of two of experimental values for 6-, 10-, 15-, and 20-mer hybridization rates (30,32), after accounting for the linear length-dependence for short oligonucleotides (33).] Once the toehold has bound, there are two possibilities: (i) the toehold base pair could dissociate, leading to the dissociation of the invader or (ii) the nearest base pair of the substrate-incumbent complex could fray, allowing the invader to compete to replace that base pair and complete the first step of branch migration. Assuming, as may seem reasonable, that the rate at which either base pair frays is similar, process (ii) should be approximately half as fast as process (i). This is because, once the substrate-incumbent base pair frays, there is a 50% chance of the invader replacing the frayed base pair, and a 50% chance of returning to the initial step. Once the first step of branch migration is complete, subsequent forward and backward steps are assumed to occur at the same rate. Therefore, the probability of successfully completing the remaining steps of branch migration before going back to the toehold-only-bound state is 1/20, from the gambler’s ruin analysis [see Section 14.2 of Feller (34)]. The lifetime of the three-stranded intermediate should be independent of concentration. Therefore, at low enough concentrations, the overall reaction can be modeled as an instantaneous second-order process, as reported by Zhang and Winfree (30). Under these conditions, the overall effective rate constant ($k_{eff}$) is given by a hybridization rate constant for the toehold, multiplied by a success probability of displacement once the toehold is bound. Even though the time spent in the three-stranded intermediate is small, $k_{eff}$ depends strongly on the probability of displacement once bound: $k_{eff} \approx 10^6 \times (1/3) \times (1/20) = 1.6 \times 10^4$/M/s for a one-base toehold. This is over three orders of magnitude larger than the $8$/M/s value measured by Zhang and Winfree (30). This is a large unexplained discrepancy, despite the approximate nature of our calculation.

We now summarize the experimental evidence for the exponential acceleration in $k_{eff}$ with toehold length, which was first reported by Yurke and Mills (31). Zhang and Winfree (30) further characterized this exponential acceleration and confirmed that it saturates in the long-toehold limit. Relevant data from both studies (Figure 2A) suggest that the exponential acceleration in itself is not an artifact of particular sequences, although details may well depend on the experimental system (sequences, modifications for read-out, etc.) or conditions (salt, temperature, etc.). The kinetics of zero-toehold (‘blunt end’) strand displacement was investigated by Reynaldo et al. (35), whose measurement of 3.6/M/s at 30°C is similar to the 1.4/M/s at 25°C reported by Zhang and Winfree (30).

Zhang and Winfree (30) built a phenomenological model for predicting $k_{eff}$ from toehold sequence (Figure 2B). They modeled branch migration using two macro-states $I$ and $J$, containing the first and second

![Figure 1. (A) Domain notation. Arrows indicate 3'-ends; Asterisk indicates Watson–Crick complementarity. (B) Toehold h mediates the displacement of the incumbent (Y) by the invader (X). Dots indicate branch migration intermediates, which are not shown.](http://nar.oxfordjournals.org/)

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half of the isoenenergetic branch migration intermediates, respectively, which are connected by a simple transition with rate constant \( k_b \) (see Supplementary Figure S1 and Supplementary Section S1 for more details). Their model fits the data, but it is hard to physically interpret the transition between \( I \) and \( J \). Branch migration is generally thought of as a random walk through many isoenenergetic states, rather than a single reversible transition characterized by a first-order rate constant. Without a more nuanced understanding of the process, it is unclear whether the fitted value of \( k_b = 1.0/s \) can be justified on more fundamental biophysical grounds.

In this work, we model branch migration at a more detailed level that explicitly includes intermediates, thereby highlighting important thermodynamic and kinetic features of the process that are not evident from the phenomenological approach.

First, we analyze a 1D (single-pathway) model of toehold-mediated strand displacement called the intuitive energy landscape (IEL) model. Systematically exploring the parameter space of this simple model suggests that some combination of two factors could explain the dependence of strand displacement rate on toehold length: (i) the branch migration process is not isoenenergetic and contains a free energy penalty for intermediate states and (ii) branch migration is slow relative to fraying of the toehold.

To verify that these factors are not already implicit features of more detailed models of DNA secondary structure thermodynamics, we simulated the experiments of Zhang and Winfree (30) using a secondary structure kinetics (SSK) simulator called Multistrand (36). Multistrand incorporates extensive thermodynamic information—from state-of-the-art nearest-neighbor (NN) thermodynamic models of DNA secondary structure (8–11)—combined with a minimally parameterized rate model. Multistrand predicts that \( k_{e,ff} \) would increase by a factor of \( 10^{1.8} \) as toehold length increases from 0 to 15, in contrast to the experimentally observed factor of \( 10^{0.5} \). This quantitative discrepancy confirms that factors (i) and (ii) suggested by the IEL analysis are not already implicitly incorporated in more detailed models of DNA secondary structure thermodynamics.

We then experimentally investigate possibility (i) suggested by the IEL, by measuring the relative stability of complexes that mimic the geometric structure of branch migration intermediates. Our experiments provide evidence in support of a free energy penalty for branch migration intermediates that is not predicted by the NN models of DNA.

Finally, we use a recently proposed coarse-grained molecular model of DNA (oxDNA) (37–39), which incorporates more physical detail, including geometric and steric effects. In addition to correctly predicting the length-dependence of toehold-mediated strand displacement rates, oxDNA independently predicts the thermodynamic penalty for branch migration intermediates, suggesting it arises from local steric effects between single-stranded overhangs around the branch point. Further, oxDNA predicts that branch migration is slower than fraying of the toehold, as it is geometrically more complex and necessarily involves more thermodynamically unfavorable steps.

**MATERIALS, METHODS AND RESULTS**

**Intuitive Energy Landscape model**

Through a simple approximate calculation, we argued that the current view of strand displacement biophysics is at odds with kinetic measurements (30,31). We now perform a more rigorous intuitive analysis by building a simple single-pathway model called the IEL model. The IEL is simple enough that its kinetic predictions may be analytically or numerically calculated, for a given parameterization. We systematically explore the IEL’s parameter space to obtain biophysical intuition and identify key thermodynamic and kinetic features essential for predictions to match experimental data.

**State space**

The IEL considers an invading strand and a substrate-incumbent complex contained in a virtual box of volume \( V \) in solution, which defines a concentration \( u \) of one molecule per volume \( V \). The state space of the IEL is illustrated in Figure 3. State A corresponds to the invader being unattached to the substrate-incumbent duplex. State B represents the formation of a first base pair within the toehold. For each additional toehold base pair that zips up, we define a new state, with state C indicating the fully formed toehold. From here, each step of branch migration involves replacing an incumbent-substrate base pair with an invader-substrate base pair. We describe the stepping between these intermediates using a simple model in which the system must pass through a single effective transition state of raised free energy, so each complete branch migration step is shown as a single tooth of the ‘sawtooth’ pattern between states C and D. This unknown effective transition state could be potentially as simple as a frayed substrate-incumbent base pair, but could also be more complex. The final stage of successful displacement involves the dissociation of the incumbent (state E) followed by the formation of the final base pair between invader and substrate (state F). Subtleties relating to the zero-toehold case are discussed in Supplementary Section S2 (see Supplementary Figure S2).
Figure 3. Free energy landscape of the IEL at 25°C for a six base toehold. States A–F and the sawtooth amplitude (ΔG_sawtooth) and plateau height (ΔG_p) parameters are described in the text. ΔG_s = 2.6 kcal/mol and ΔG_p = 1.2 kcal/mol are used for illustration.

Energy model

The IEL models the free energy of the virtual box (∆G_box) relative to state A. ∆G_box is defined as in Figure 3. Initial binding (state A to B) incurs a free energy penalty of ∆G_{init} = ∆G_{volume} + ∆G_{assoc} ≈ 9.95 + 1.90 kcal/mol (11) (at concentration u = 50 nM) due to the reduction in the entropy of the box caused by lost translational and orientational degrees of freedom. ∆G_{assoc} is the free energy cost of association at a standard concentration of u_0 = 1 M, and ∆G_{volume} = RT\ln(u_0/u) is a correction for the actual concentration, where R is the universal gas constant and T is the temperature in Kelvin. Figure 3 shows the first base pair of the toehold adjacent to the helix, where it interacts favorably with the adjacent duplex end. For simplicity, we neglect sequence-dependent interaction strengths and use the average value ∆G_{bp} ≈ -1.7 kcal/mol for this and all other base pairs, as taken from the NN model (8–10) of DNA secondary structure thermodynamics. Formation of each successive base pair in the toehold therefore contributes ∆G_{bp}. Our effective transition states, which are local free energy maxima along the sawtooth, are raised above the branch migration intermediates with fully paired substrate. As the physical details of branch migration steps are not well-understood, the effective ‘sawtooth amplitude’ (ΔG_s) and plateau height (ΔG_p) (both in kcal/mol).

In an intuitive model like the IEL, one could consider a range of alternative perturbations to the current biophysical understanding of strand displacement. ΔG_s and ΔG_p, however, parameterize in a simple way aspects of the branch migration process that are known to be poorly characterized. Furthermore, as we show later, experiments and more detailed modeling provide physical justification for the effects that these parameters represent.

Rate model

The system may undergo a transition to either of its neighboring states on the 1D landscape. To ensure that stochastic simulations will eventually converge to the thermodynamic (Boltzmann) equilibrium over the states, transition rates must satisfy detailed balance, i.e.

\[
\frac{k_{ij}}{k_{ji}} = e^{-\frac{\Delta G_{box}(i) - \Delta G_{box}(j)}{RT}}.
\]

Here \( k_{ij} \) is the transition rate from \( i \) to \( j \). Equation (1) only fixes the relative rates; in principle, for each pair of adjacent states \( i \) and \( j \), an independent scaling factor could be chosen. Consequently, the number of candidate rate models is enormous. In the spirit of minimal parameterization to avoid over-fitting the data, we use only two independent scaling factors: \( k_{bi} \) for all bimolecular and \( k_{uni} \) for all unimolecular transitions.

For unimolecular transitions (all except A ≈ B and D ≈ E in Figure 3), we use a model in which all energetically downhill steps have a constant rate \( k_{uni} \). This is a sensible first approximation, given that these steps involve forming a base pair at the end of a duplex or stepping down from the transition states of branch migration, which may resemble forming a base pair at the junction. In other words, for \( \Delta G_{box}(i) > \Delta G_{box}(j) \),

\[
k_{ij} = k_{uni}
\]

and therefore,

\[
k_{ji} = k_{uni} e^{-\frac{\Delta G_{box}(j) - \Delta G_{box}(i)}{RT}}.
\]

This model is a continuous-time variant of the Metropolis (40) scheme for calculating thermodynamic averages.

For bimolecular transitions (A ≈ B or D ≈ E), the IEL assumes that complexes join at the constant rate \( k_{bi} \times u \), and calculates the reverse rate by detailed balance. If \( i \) to \( j \) is a join step,

\[
k_{ij} = k_{bi} u = k_{bi} e^{-\frac{\Delta G_{volume}}{RT}} u_0,
\]

and

\[
k_{ji} = k_{bi} e^{-\frac{\Delta G_{volume}(i) + \Delta G_{volume}(j)}{RT}} u_0.
\]

We choose \( k_{bi} \) to be \( 3 \times 10^9/\text{M/s} \), based on the hybridization rate constant fitted by Zhang and Winfree (30). For simplicity, we choose \( k_{uni} \) such that the dissociation
rate of the last base pair of the toehold (state B to A) is approximately equal to the fraying rate for every other base pair of the toehold (steps from state C toward B). This yields \( k_{uni} = 7.5 \times 10^{7} / \text{s} \) (see Supplementary Section S2 for details). This choice of \( k_{uni} \), at the very least, ensures that two somewhat similar processes have similar rates.

**Analytic and numerical calculations**

We use an analytic formula for calculating absorption probabilities for a 1D random walk with absorbing boundaries (41) to calculate \( k_{eff} \) as a function of toehold length \( h \) for various values \( \Delta G_s \) and \( \Delta G_p \) (Figure 4A).

(For details, see Supplementary Section S2.) To quantify the extent of control provided by toeholds, we define

\[
A_{h_2, h_1} = \log_{10}(k_{eff}(h_1)) - \log_{10}(k_{eff}(h_2))
\]

(6)

to be the orders of magnitude acceleration in \( k_{eff} \) as toehold length increases from \( h_1 \) to \( h_2 \). We will be most interested in \( A_{15,0} \), which corresponds to the difference between the leak rate and the maximal rate measured in Zhang and Winfree (30). [We choose 15 rather than 10 because the length-dependence of hybridization rates (33) suggests that experimental values for \( A_{h,0} \) may not be bounded, but for our purposes there is insignificant change past length 15.] The experimental results of Zhang and Winfree (30) can be matched by the IEL, but only with surprisingly large values of \( \Delta G_s \) and \( \Delta G_p \). IEL(2.6, 0), with a plausible value of \( \Delta G_s = 2.6 \text{ kcal/mol} \), \(~1.5 \text{ base pair stacks} \) and an a priori choice of \( \Delta G_p = 0 \ \text{ kcal/mol} \), predicts \( A_{15,0} = 3.1 \). This is 3.4 orders of magnitude smaller than the experimentally observed value of 6.5. Increasing either \( \Delta G_s \) or \( \Delta G_p \) serves to increase the predicted \( A_{15,0} \), as shown by the contour plot in Figure 4B. The slope of contour lines suggests that \( \Delta G_{s,p} \equiv \Delta G_s + \Delta G_p \) is the key quantity. Indeed, a scatter plot of the dependence on \( \Delta G_{s,p} \) (Figure 4C) using all the data in Figure 4B produces almost no vertical spread. \( \Delta G_{s,p} \) is 7.3 kcal/mol matches the experimentally observed value of \( A_{15,0} = 6.5 \).

Other perturbations of the IEL model were unable to match the experiments. For example, we considered an alternative to the Metropolis method for setting unimolecular rates, the Kawasaki (42) method, which scales both uphill and downhill transition rates based on the corresponding change in \( \Delta G_{box} \). After rescaling \( k_{uni} \) so that dissociation of the last toehold base pair still occurs at the same rate as fraying, the Kawasaki method predicts a value for \( A_{15,0} \) within 10% of the Metropolis method (see Supplementary Figure S3). Going further, and considering that \( k_{bi} \) and \( k_{uni} \) may not be chosen ideally, we first note that logically, a uniform change to both rates will cancel and thus have no effect on \( A_{15,0} \). In contrast, IEL predictions do depend on the ratio \( k_{uni}/k_{bi} \), which substantially affects the probabilities that initial contacts lead to successful zipperung of the toehold and to successful branch migration before dissociation (see Supplementary Figure S4 and Supplementary Section S2). However, even implausibly low values of \( k_{uni}/k_{bi} \) cannot account for the data without a large value of \( \Delta G_{s,p} \).

These features of the IEL can be understood through simple analytical approximations. Conceptually, we can split the strand displacement process into an attachment step (A → B), followed by success or failure of zipperung the toehold, followed by either dissociation of the toehold or successful displacement. We first treat the case of long toeholds, for which in the IEL model \( k_{eff} \) saturates at

\[
k_{eff}(\infty) \approx k_{bi} \cdot p_{zip}
\]

(7)

with \( p_{zip} = k_{uni}/(k_{uni}+k_{bi} \lambda) \) and \( \lambda = e^{-\left(\Delta G_{s} - \Delta G_{ass} \right)/R T} u_0 \) giving the approximate probability that after making the first base pair, the invader does not dissociate and the remaining toehold bases zip up. (For long toeholds successful displacement is guaranteed once the toehold is formed.) For shorter toeholds that do not saturate \( k_{eff} \), the probability of toehold dissociation before branch migration dominates, and we can derive

\[
k_{eff}(h) \approx \frac{k_{uni}}{2b} \cdot e^{-\left(\Delta G_{s} - \Delta G_{ass} \right)/R T} u_0
\]

(8)

### Figure 4.

(A) Predictions of IEL (\( \Delta G_s, \Delta G_p \)) for different values of the sawtooth amplitude \( \Delta G_s \) and plateau height \( \Delta G_p \). \( \Delta G_{s,p} \) needs to be as high as 7.3 kcal/mol (>>4 base-pair stacks) to match experiment (30). (B) Contour plot of orders of magnitude acceleration (\( A_{15,0} \)) as a function of \( (\Delta G_s, \Delta G_p) \) for \( \Delta G_s \in [0,7.3] \ \text{kcal/mol} \) and \( \Delta G_p \in [0,7.3] \ \text{kcal/mol} \). (C) Scatter plot of \( A_{15,0} \) versus \( \Delta G_{s,p} \), using all the data in (B). The points plotted show almost no vertical spread, meaning that all points lie on a line. This indicates that \( \Delta G_{s,p} \) is the predictive quantity. Inset: IEL (1.3, 0) and IEL (5.0, 5.0) for a 1-base toehold. The bimolecular initial binding step, which is identical in both landscapes, is marked in orange to highlight the contrast between the landscapes.
for $h \neq 0$ and, there being two ends where branch migration could start, twice the given value for $h = 0$.

We clearly see that the slope in Figure 4A is governed by $\Delta G_{pp}$, while the role of $h$ (Supplementary Figure S5) and $\Delta G_{rp}$ in determining

$$A_{15,0} \approx \log_{10} \frac{k_{\text{eff}}(\infty)}{k_{\text{eff}}(0)} \approx 0.9 + \log_{10}\left(\frac{2h e^{\Delta G_{rp}/RT\lambda}}{k_{\text{uni}}/k_{\text{bi}} + \lambda}\right)$$

(9)

is immediate from their influence on $k_{\text{eff}}(0)$. The more subtle effect of changing $k_{\text{uni}}/k_{\text{bi}}$ results partly from effects on $p_{\text{app}}$, a slightly more accurate approximation can explain the behavior seen in Supplementary Figure S4. Details on this and other derivations are given in Supplementary Section S2.

**Matching both strand displacement and branch migration rates**

The IEL analysis suggests that we can reconcile kinetic models of displacement at the base pair level with the experimental measurements of Zhang and Winfree (30) by incorporating a thermodynamic penalty for initiating branch migration ($\Delta G_s$) and/or slowing down the rate of branch migration through a relatively large sawtooth amplitude ($\Delta G_p$) associated with each step of branch migration. Although $\Delta G_{rp}$ is constrained to be at least 7.3 kcal/mol to match the data, the individual contributions of $\Delta G_s$ and $\Delta G_p$ are not determined by the analysis.

To compare the inferences of the IEL analysis with experimental measurements, we need to account for the increase in initial binding rate ($A$ to $B$) with toehold length, as observed for short oligonucleotide hybridization (33). (For simplicity, the IEL assumes that the initial binding rate is independent of toehold length.) To this end, we also explored a variant of the IEL, called the Augmented Energy Landscape (AEL) model, that includes the linear increase in the formation rate of the first base pair (see Supplementary Section S3). For the AEL, we find that a smaller value of $\Delta G_{rp} = 5.6$ kcal/mol is required owing to the contribution of the binding rate to the overall acceleration. A good fit to the experimental data shown in Figure 4 is obtained for $k_{\text{bi}} = 3.3 \times 10^7$/M/s and $k_{\text{uni}} = 8.2 \times 10^6$/s (Supplementary Table S1).

We have not yet presented evidence to suggest that the plateau height $\Delta G_p$ is non-zero. However, if $\Delta G_p = 0$, a sawtooth amplitude of 7.3 kcal/mol (5.6 for the AEL) would be required to account for the data, which implies an average branch migration step time of $\sim 3.0$ ms ($\sim 1.6$ ms for the AEL). This is much slower than experimentally inferred step times on the order of 12–20 µs (28,29). Therefore, simultaneously matching both measured branch migration and strand displacement rates requires a significant thermodynamic penalty to initiating branch migration.

The IEL analysis raises two important questions. First, could the necessary values of $\Delta G_s$ and $\Delta G_p$ represent features missing in the IEL’s simplified thermodynamic landscape that are implicitly present in more detailed models? If this is not the case, are $\Delta G_s$ and $\Delta G_p$ purely phenomenological parameters used to fit the data or do they represent real physical effects that arise from the molecular properties of DNA?

**Secondary Structure Kinetics model**

As a first step toward answering these questions, we use a SSK simulator called Multistrand (36) to study strand displacement as a random walk on a more complex energy landscape, the NN secondary structure model (8–11), that incorporates a wealth of existing thermodynamic knowledge. Multistrand extends the Kinfold simulator (43) from single-strand landscapes to landscapes for multiple interacting nucleic acid molecules. Code implementing the Multistrand model is available for public download (see Supplementary Section S4 for details).

**State space**

Multistrand considers a set of strands in a virtual box of volume $V$ in solution, which defines a concentration $u$ of one molecule per volume $V$. The state space consists of all possible sets of Watson–Crick base pairs (such as states in Figure 5A), with two restrictions: (i) no base can have more than one pairing interaction and (ii) secondary structures containing pseudo knots are not allowed. Structures without pseudo knots are tree-like and have nested base pairing (11). The size of Multistrand’s state space grows exponentially in the number of bases (11).

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**Figure 5.** (A) Example states and elementary steps in Multistrand (36), a SSK simulator. States illustrated are each adjacent to state $i$, as they differ from $i$ by only one base pair. Transition rates are chosen to obey detailed balance. Dots and arrows at the top indicate other possible elementary steps from state $i$ (not shown). (B) Multistrand predictions of experimentally measured (30) strand displacement rates as a function of toehold length. Experimental data points and error bars are from Zhang and Winfree (30); the fitted line is their phenomenological model. Standard errors for Multistrand simulations are under 1% (not shown).
Energy model

Let $\Delta G_{\text{box}}(i)$ be the free energy of the virtual box in state $i$, relative to a completely unstructured state with no base pairs. $\Delta G_{\text{box}}(i)$ is the sum of free energies of each isolated complex $c$, $\Delta G(c)$, in state $i$. The free energy of each complex is estimated using the NN model (8–10), which has been extended to multiple interacting nucleic acid strands (11). NN parameters were measured (9,10) in 1 M Na$^+$, which is roughly thermodynamically equivalent to the 12.5 mM Mg$^{2+}$ used by Zhang and Winfree (30), according to salt corrections (44,45) to the NN model. The NN model assumes that the free energy contribution of each base pair is dependent only on the identity and orientation ($\Sigma' \text{' or 3}') of its nearest neighbors on either side and calculates $\Delta G(c)$ by summing contributions from each sub-structure (‘loop’) closed by a base-paired section; where

$$\Delta G(c) = (L - 1)\Delta G_{\text{init}} + \sum_{\text{loops} c} \Delta G(\text{loop})$$

$L$ is the number of strands in complex $c$ and $\Delta G_{\text{init}} = \Delta G_{\text{assoc}} + \Delta G_{\text{volume}}$ is, like in the IEL, the free energy cost of bringing two separate strands together. Stabilizing contributions to $\Delta G(\text{loop})$ mainly arise from base-pair stacks, and destabilizing contributions arise from the entropic cost of closing loops. At duplex ends in either interior or exterior loops, the first overhanging nucleotide contributes a ‘dangle’ energy term (46). When two duplex ends abut, as at a nick, the duplexes are known to ‘coaxially stack’ onto each other (47–50), thereby stabilizing the structure; this interaction is not explicitly incorporated into Multistrand, although in some cases dangle energy terms partially account for it. Multistrand’s energy model is identical to that used in NUPACK (51) and is thus similar to those in Vienna RNA (52) and Mfold (53).

Rate model

Multistrand allows transitions between states $i$ and $j$ if they differ by a single base pair. The rate models we have explored for Multistrand are identical to the IEL, except for scaling factors. Like the IEL, Multistrand’s predictions are not particularly sensitive to the choice between standard unimolecular rate models: The predicted orders of magnitude acceleration in $k_{\text{eff}}$ between toeholds 0 and 15 differ by less than 3% between Metropolis and Kawasaki (Supplementary Figure S6B). For Metropolis, $k_{\text{bi}} = 1.26 \times 10^6 / \text{M/s}$ and $k_{\text{uni}} = 4.4 \times 10^8 / \text{s}$ were calibrated (36) by fitting Multistrand simulations to experimentally measured DNA–DNA hybridization (32) and zippering (33) rates, respectively. Given the rate model, energy model and current state, the choice and timing of the next transition is determined using a Gillespie algorithm (54). Multistrand allows any initial first pair of nucleotides to interact, each at the standard bimolecular rate, resulting in an increase in hybridization rates with increasing toehold lengths. In principle, each $i$ to $j$ transition could have both forward and reverse rates scaled arbitrarily but equally, thus obtaining a distinct kinetic model that still satisfies detailed balance. However, doing so would amount to treating the microscopic barriers that impede transitions between the states of the NN model as different in every case. In the absence of additional information justifying such differences, the Metropolis approach constitutes a minimally parameterized rate model in which microscopic barriers associated with downhill processes (which typically involve base-pair formation) are assumed to be uniform, and their effects are incorporated in the value of $k_{\text{uni}}$. As such, the Metropolis approach allows us to ask whether the free-energy landscape given by the NN model is sufficiently accurate and precise that it captures the factors that determine relative kinetics, and that finer details are only important in setting an overall rate constant $k_{\text{uni}}$.

Comparing Multistrand predictions with data

We simulated the ‘average strength toehold’ experiments of Zhang and Winfree (30) and calculated $k_{\text{eff}}$ as a function of toehold length. For technical details regarding the simulations, see Supplementary Section S4 and Supplementary Table S2. Multistrand captures the qualitative dependence of $k_{\text{eff}}$ on toehold length, despite its minimally parameterized rate model. However, Multistrand predictions quantitatively diverge from experiment in two ways (Figure 5B): (i) the saturation value of $k_{\text{eff}}$ for toeholds longer than five bases is roughly 20 times larger and (ii) increasing toehold length from 0 to 15 increases $k_{\text{eff}}$ by a factor of $10^{3.8}$, in contrast to $10^{6.5}$ observed in experiment.

The first issue could be addressed by uniformly decreasing both $k_{\text{uni}}$ and $k_{\text{bi}}$, which would shift the predicted curve down to match the data for toeholders longer than five bases. Indeed, this re-scaling is also physically reasonable, as there is considerable variation among hybridization and fraying rates in the literature for nucleic acids (32, 33,55–60).

Scaling $k_{\text{uni}}$ and $k_{\text{bi}}$ uniformly would not resolve the second challenge. However, like the IEL, Multistrand is particularly sensitive to the ratio $k_{\text{uni}}/k_{\text{bi}}$. Decreasing $k_{\text{uni}}$ slows down all unimolecular transitions—which brings the branch migration rate closer to experimentally inferred values but makes fraying too slow. Two parameters, $k_{\text{uni}}$ and $k_{\text{bi}}$, are simply not enough (36) to simultaneously match the four distinct time scales involved: rates of hybridization, fraying, branch migration and branch migration initiation. Therefore, even unrealistically low choices of $k_{\text{uni}}/k_{\text{bi}}$ are unable to match observed acceleration in strand displacement rates due to toehold length (see Supplementary Figure S7).

The SSK analysis confirms that understanding what the IEL’s $\Delta G_s$ and $\Delta G_p$ represent requires examining features not present in the NN model. Multistrand models branch migration as a fray–and–replace process and interprets the IEL’s sawtooth transition state as one in which the substrate-incumbent base pair at the junction is frayed. This choice, when coupled with a $k_{\text{uni}}$ calibrated to match fraying rates (33), results in a branch migration rate that is much faster than experimentally inferred step times (28,29).

Indeed, the thermodynamics of the branch migration junction, e.g. states $i$ and $j$ in Figure 5A, is not well
characterized in the standard NN secondary structure model, as it involves overhangs, dangles and coaxial stacking. Reflecting the lack of consensus, tools like NUPACK (51), Vienna RNA (52) and MFold (53) offer several ways of treating dangle contributions; however, none of the three ‘dangle options’ in the NUPACK energy model (11) improved Multistrand predictions (Supplementary Figure S6A).

Measuring relative stability of branch migration intermediates

For the IEL to match measured hybridization, fraying and branch migration rates, and hence strand displacement rates, a non-zero plateau height (ΔGp) was necessary. We hypothesized that the initiation of branch migration incurs a thermodynamic cost due to the second overhang it engenders at the junction, even though the nearest neighbor model for DNA secondary structure thermodynamics predicts no such effect. We present experimental evidence in support of this hypothesis by investigating the free energy landscape of branch migration. The biggest experimental challenge in measuring the relative stability of branch migration intermediates is that they cannot be easily isolated. Indeed, they are interchangeable as branch migration proceeds back and forth, with individual step times just tens of microseconds (28,29).

System description

To overcome this issue, we designed immobile complexes Xi:Yj comprising hairpin Xi and strand Yj (Figure 6). Xi and Yj have poly-T overhangs of length i and j, respectively. Varying (i, j) from (20,0) to (0,20) with i+j = 20 yields complexes that are ‘frozen snapshots’ of branch migration, with no expected branch migration possible. X20:Y00 mimics the binding of the invader by the toehold, and X19:Y01 represents the displacement of 1 base, and so on. X10:Y10 represents the ‘half way stage’ of branch migration and X00:Y20 captures nearly successful displacement. Measuring the relative stability of these frozen snapshots is expected to be indicative of the relative free energies of branch migration intermediates. Experiments involving these complexes will henceforth be referred to as the ‘strand displacement snapshot’ study. To investigate the consequences of short overhangs at the junction, we designed complexes Xi:Yi (for i = 0, 1, 2, 5 and 10). These experiments will be referred to as the ‘local overhang’ study. All our complexes have the same base pairs at the junction and poly-T overhangs, whereas branch migration typically involves different bases at each step. We can therefore study the thermodynamic consequences of junction geometry, without the complication of sequence dependence.

Temperature dependent absorbance experiments

We measure the UV absorbance (at 260 nm) of each complex between 20 and 90 °C, at four different concentrations. As the absorbance of single-stranded DNA (ssDNA) is higher than that of double-stranded DNA, and the fraction of ssDNA is dependent on the temperature, a temperature-dependent absorbance curve is obtained at each concentration (Supplementary Figure S8). Each complex Xi:Yj exhibits two transitions: the bimolecular, lower temperature transition and the unimolecular, higher temperature transition due to the hairpin in Xi closing or opening. The unimolecular transition was identified both by it being independent of concentration and by control melts involving the hairpins only (data not shown). At the concentrations chosen, the bimolecular and unimolecular transitions are distinct.

For each complex, we infer the enthalpy (ΔH) and entropy (ΔS) of formation by fitting the smoothed and

Table 1. Sequences for domains from Figure 6, listed 5’–3’

| Domain | Sequence     | Length |
|--------|--------------|--------|
| d      | CTTCATCATACTACG | 15     |
| e      | CTCCATGTCACTTC | 14     |

Figure 6. (A) Complex Xi:Yj comprises hairpin Xi and strand Yj, with poly-T overhangs of length i and j respectively. Domains d and e are designed to be orthogonal to each other and the overhangs (sequences in Table 1). (B) Varying (i, j) from (20,00) to (00,20) with i+j = 20 mimics the geometry branch migration intermediates [X20:Y00 (start), X10:Y10 (middle) and X00:Y20 (end), respectively]. No branch migration is intended in these complexes.

Figure 7. Smoothed and normalized UV absorbance data while annealing (at 200 nM). The lower temperature transition is the (bimolecular) formation of the complex, whereas the higher temperature transition is the (unimolecular) formation of the hairpin. The mean absorbance between 20 and 35 °C is normalized to 0 and that between 64 and 66 °C (indicated by vertical lines) to 1. The temperature range whose mean absorbance is normalized to 1 is concentration-dependent (Supplementary Table S3). Data acquired by annealing and melting are essentially superimposable. The dashed line indicates the halfway point of the bimolecular transition.
normalized temperature-dependent absorbance curves (Figure 7) to a two-state model (61):}

\[
\frac{[X_i : Y_j]}{[X_i][Y_j]} = e^{-(\Delta H^\circ - T\Delta S^\circ)/RT}
\]  

(11)

where \(\Delta H^\circ\) and \(\Delta S^\circ\) are assumed to be temperature independent. We perform this fitting using a Bayesian analysis and confirm our findings using a simpler descriptive ‘leave-one-concentration-out’ approach. Details are provided in Supplementary Section S5; see Supplementary Figure S9 and Supplementary Tables S3–S5.

From \(\Delta H^\circ\) and \(\Delta S^\circ\), we can calculate the free energy of formation, \(\Delta G^\circ(T) = \Delta H^\circ - T\Delta S^\circ\). By comparing the free energies of different complexes, we infer the contribution of the poly-T overhangs. Our two-state assumption means that \(\Delta G^\circ(T)\) is assumed to be linear in \(T\)—deviations from linearity limit the accuracy of extrapolation from values around the melting temperature of the complexes. Values of \(\Delta G^\circ\) at 55°C, which is closer to the melting temperature, are plotted in Supplementary Figure S10.

**Second overhang causes de-stabilization due to a local effect**

Unlike the predictions of NN thermodynamic models (9,11) of DNA, the immobile complexes we designed are not all of equal free energy; a second overhang at the junction causes a thermodynamic penalty (Figure 8A). At 25°C, we infer a free energy penalty of \(\sim 2.0\) kcal/mol (3.4 RT) as branch migration proceeds from 0 to 10 steps (X20:Y00 versus X10:Y10), with the majority (\(\sim 1.5\) kcal/mol) arising from the first step (X20:Y00 versus X19:Y01). An approximately symmetric decrease is inferred for steps 11 to 20 (X10:Y10 versus X00:Y20).

As the de-stabilization due to an additional overhang plateau so quickly, we suspect that the penalty is due to local effects at the junction, which is supported by the ‘local overhang’ study (Figure 8B). Two one-base overhangs on either side of the junction (X01:Y01) result in a penalty of \(\sim 1.4\) kcal/mol relative to no-overhangs (X00:Y00). Lengthening the overhangs increases this penalty, but each additional base contributes progressively less, with an overall penalty of 3.0 kcal/mol and 3.2 kcal/mol, respectively, for 5-base (X05:Y05) and 10-base (X10:Y10) overhangs.

Our experiments suggest that current NN models of DNA do not capture the free energy landscape of strand displacement accurately enough to capture the kinetics of branch migration. This explains in part the inability of SSK models like Multistrand to match experimentally observed toehold-mediated acceleration.

**Coarse-grained molecular modeling**

A 3D model of DNA at the nucleotide level

Although it is possible to tune the IEL to agree with experimental data, and, moreover, the observed destabilization of duplexes by ssDNA overhangs appears to support a plateau during displacement, a physical explanation of the parameters required is important. Atomically detailed or coarse-grained molecular models have the potential to provide this kind of insight (62). Here, we consider a recently proposed coarse-grained molecular model of DNA (oxDNA) (37–39), specifically the parameterization of Ouldridge (39). Code implementing the model is available for public download (see Supplementary Section S6).

In this model, illustrated in Figure 9, each nucleotide is a 3D rigid body so that the state space of \(N\) model nucleotides has \(6N\) spatial dimensions and \(6N\) momenta. Pairs of nucleotides interact through a number of pairwise effective interactions (shown in Supplementary Figure S11), representing chain connectivity, excluded volume, hydrogen-bonding and stacking interactions between bases. The combination of nearest-neighbor stacking within a strand and hydrogen-bonding between complementary bases drives the formation of helical duplexes. We note here that oxDNA explicitly considers stacking interaction.
DNA tweezers (39), a DNA walker (65) and the ability behavior for a range of systems. Such systems include values used have been seen to produce physically reasonable results for the purposes of this particular study. The parameter values used have been adjusted from those presented previously (39) for a buffer than the other.

Presentation of transition intermediates are more suited to there is no reason to assume that its kinetics and the representation might be significantly different in these two sodium buffers (10). It is plausible that kinetics of strand displacement are computationally demanding to simulate, and thus they require enhanced sampling techniques. We use umbrella sampling (71) to improve equilibration of thermodynamic averages in VMMC simulations. This method involves applying an artificial bias to lower free-energy barriers, thereby accelerating the transitions between (meta)stable states. For LD simulations of kinetics, we use forward flux sampling (72,73) to obtain accurate estimates of the relative rates of strand displacement for different lengths of toehold, as shown in Supplementary Figure S13.

Interaction of kissing hairpins (66) and overstretching of DNA under tension (67). Two of these systems, the tweezers and the walker, depend on strand displacement, and so oxDNA has previously been shown to reproduce this phenomenon. Furthermore, a barrier to initiating branch migration was predicted when studying the tweezers (39), although it was not studied in great detail.

Simulation techniques
We provide a concise summary of our simulation techniques here; for details, see Supplementary Section S6. We use two algorithms to simulate oxDNA: the ‘Virtual Move Monte Carlo’ (VMMC) algorithm of Whitelam et al. (68,69) and the rigid-body Langevin Dynamics (LD) algorithm of Davidchack et al. (70). The first approach randomly attempts and accepts moves of clusters of nucleotides (illustrated in Supplementary Figure S12) in a manner that ensures the system samples from a Boltzmann distribution. The second incorporates noise and damping terms into Newton’s equations in a self-consistent manner, thereby generating states drawn from the canonical ensemble. The sampling of states is more efficient with VMMC, and so it is used for all thermodynamic averages. Langevin algorithms, however, are explicitly dynamical, so they naturally give kinetic information that is harder to infer from VMMC. Consequently, all kinetic results quoted in this work use the LD algorithm.

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Kinetic simulations of displacement
We consider a slightly truncated system (Supplementary Table S6) based on that used by Zhang and Winfree (30). For computational simplicity, we removed the majority of the tail of the incumbent strand, which was used to interact with a reporter. Further, to simplify the order parameter for displacement, we use an approach in which only the expected (native) base pairs between the incumbent and the substrate or the invading strand and the substrate are given a non-zero binding strength. This simplification is reasonable because the sequences were designed to exhibit minimal secondary structure and cross-interactions when single-stranded. Additional simulations of toehold association were also performed to explore the
consequences of this simplification: for further details, see Supplementary Section S6. Simulations of the three strands were performed in a periodic cell of volume $1.67 \times 10^{-20}$ L for toehold lengths between 0 and 7 bases. We initialized the system at $25^\circ C$ with the incumbent fully bound to the substrate and the invading strand separate and measured the overall rate of displacement using forward flux sampling. Order parameters and detailed results are given in Supplementary Tables S7–S10.

Figure 10 reports the measured relative rates of displacement for different toehold lengths in simulation (30). oxDNA agrees remarkably well with the experimental data: in particular, it shows the same exponential dependence on toehold length for short toeholds followed by a plateau for longer ones. The overall acceleration from 0- to 7-base toehold is $10^6$,$^{56}$ close to the experimental value of $10^6$,$^{28}$ It would be unwise to put excessive emphasis on this agreement, but the base-pairing energies of oxDNA are fitted to the same secondary-structure free energies underlying the thermodynamics of Multistrand and the IEL, and therefore the predictions of these discrete models should be equally applicable to oxDNA as to real DNA. By analyzing oxDNA’s representation of displacement, we now aim to physically justify the parameters used in the IEL.

**Free-energy profile of displacement**

We measure the free-energy profile of displacement to see whether oxDNA reproduces the experimental tendency of two ssDNA overhangs to destabilize a branch point and provides a physical explanation for it. We show the free energy of the three-stranded displacement complex as a function of the progress of branch migration, as measured by the identity of the base pair between invading and substrate strands closest to the 3'-end of the substrate, in Figure 11. We observe an increase in free energy of $\sim1.3$ kcal/mol as branch migration is initiated, similar to the plateau height introduced to improve the IEL. Furthermore, as suggested by our strand displacement snapshot experiment, this barrier appears to saturate quickly: once a second ssDNA overhang of three or four bases has been created, there is negligible further increase in the destabilization. When the invading strand’s single-stranded overhang is reduced to a few bases toward the end of branch migration, the free-energy penalty decreases (as in experiment).

The fact that the penalty saturates after around four bases suggests that the cause is local to the branch point. From looking at the branch migration intermediate in Figure 11 in which the helices are coaxially stacked at the junction, one can see that the branch point is densely packed with nucleotides. To maintain coaxial stacking of helices at the junction, the two single-stranded overhangs must both take evasive action, unstacking and bending away from each other. The system can also reduce the overcrowding by breaking the coaxial stacking at the branch point, as in the coaxially unstacked image in Figure 11, but this carries a penalty itself. This overcrowding is worse with two ssDNA overhangs than one, resulting in a free-energy penalty for initiating branch migration and creating a second overhang.

Figure 11 also shows separate free-energy profiles for systems restricted to coaxially stacked or unstacked states (definitions of these states are provided in Supplementary Section S6). The existence of the overcrowding penalty makes the coaxially unstacked state (which is initially unfavorable, but involves minimal overcrowding) relatively more probable at later stages of branch migration. In the absence of coaxial stacking, the plateau height is minimal, consistent with our diagnosis of overcrowding.

To complete the comparison between simulation and experiment, we have estimated the destabilizing effect of single-stranded overhangs on duplex formation within oxDNA. The systems are analogous to those used in our experimental studies and illustrated in Figure 6, but the hairpins have shorter stems of length 12 base pairs. Furthermore, to make comparisons at 25 and $55^\circ C$ for the chosen simulated concentration, we use...
complementary lengths for interstrand binding of six and eight bases, respectively, ($\Delta G^\circ$ can be inferred most accurately from experiments at $\sim$55°C, where the complexes melt, but we are most interested in the value at $\sim$25°C).

The contribution of the single-stranded overhangs to the free energy of association $\Delta G^\circ$ is expected to be independent of the length of the duplexes, however, allowing a direct comparison of this property to be made. The sequences used are given in Supplementary Table S11, and further details are provided in Supplementary Section S6.

The results of the hybridization simulations are reported in Table 2. The presence of two long ssDNA overhangs is destabilizing by $\sim$1.3 and 0.75 kcal/mol at 25 and 55°C, respectively, relative to the case with one single-stranded overhang. We conclude that oxDNA has a plateau height comparable with (but slightly smaller than) that found in experiment, with the caveat that we have not measured sequence-dependent effects at the branch point.

Kinetics of branch migration and fraying

Here, we examine oxDNA’s representation of the mechanism of branch migration in detail. Consistent with the literature (28,29,33), oxDNA predicts that branch migration is slow compared to the rate at which base pairs in the toehold fray and provides a mechanistic explanation for this difference.

### Table 2. Destabilizing $\delta G_{m,n}$ due to dangling ssDNA sections, as found in simulations

| Duplex length | $T$/°C | $m$ | $n$ | $\delta G_{m,n}$(kcal/mol) |
|---------------|--------|-----|-----|----------------------------|
| 6             | 25     | 10  | 0   | 1.46 [1.2]                 |
|               | 0      | 10  | 0   | 1.38 [1.0]                 |
| 8             | 55     | 10  | 0   | 0.87 [0.60]                |
|               | 0      | 10  | 0   | 0.94 [0.78]                |
|               | 10     | 10  | 0   | 1.65 [1.79]                |

$m$ is the length of the excess ssDNA attached to the hairpin, and $n$ the length attached to the shorter strand. Values for $\delta G_{m,n}$, the free energy of formation of the duplex, are extracted from simulations. The reported value, $\delta G_{m,n}$, is given by $\delta G_{m,n} = \Delta G_{m,n} - \Delta G_{0,0}$. $\Delta G_{m,n}$ is then the destabilization of the duplex due to the ssDNA overhangs. The values in square brackets correspond to values inferred from our experimental studies for the closest equivalent lengths of dangling ssDNA. Note that $\delta G_{10,0}$ and $\delta G_{0,10}$ found here are compared to $\delta G_{20,0}$ and $\delta G_{0,20}$ from experiment – the local nature of the destabilisation makes this reasonable.

![Figure 12](http://nar.oxfordjournals.org/) Typical branch migration steps as represented by oxDNA, for a toehold length of three bases. In each case, the nucleotide colored black displaces the nucleotide colored gold. (A) Displacement via invasion: a base from the invading strand enters the incumbent/substrate duplex and competes with a base from the incumbent for base pairing. (i) Initial state, with the gold-colored base on the incumbent bound to the substrate and the black-colored base on the invading strand unbound. (ii) Metastable intermediate with both the gold and black bases competing for base pairing with the substrate. (iii) Final state, with the black base bound to the substrate and the gold base unbound. (B) Displacement via sequential disruption and formation of bonds: a base pair in the incumbent/substrate duplex breaks and the substrate base unstacks and moves across to the invading/substrate duplex. (i) Coaxially unstacked initial state, with the gold-colored base on the incumbent bound to the substrate and the black-colored base on the invading strand unbound. (ii) Metastable intermediate state, in which the base pair involving the gold base has broken, and the base from the substrate strand has unstacked. (iii) Final state, with the substrate base now bound to the black base in the invading strand. Labeled distances between hydrogen-bonding sites are used as coordinates to monitor branch migration steps in detail later in Figure 13.
Typical branch migration steps observed in oxDNA are shown in Figure 12. Two types of process occur most commonly:

1. Invading and incumbent duplexes remain coaxially stacked at the branch point, and a single-stranded base from the invading strand enters the duplex region, competes for base-pairing to the substrate with a base from the incumbent, and eventually displaces it. This process is illustrated in Figure 12A. We refer to this process as ‘branch migration via invasion’.

2. Invading and incumbent duplexes coaxially unstack at the branch point, and a base in the substrate is transferred from one duplex to the other. In this process, one base pair must fully break before the other forms some time later. An example of such a step is shown in Figure 12B. We refer to this process as ‘branch migration via sequential disruption and formation of bonds’.

Invasion is dominant at the start of branch migration, when the majority of systems are coaxially stacked. Branch migration via sequential disruption and formation of bonds becomes more relevant at intermediate stages, when coaxial stacking is less prevalent. The two mechanisms differ in whether the duplexes are coaxially stacked during branch migration, and whether the displacing base invades the duplex before disruption of the initial base pairing. Invasion is geometrically infeasible from an unstacked state. However, the substrate strand could remain stacked at the branch point whilst ‘flipping out’ a base from the incumbent and only then replacing it with a base from the invader. This third type of process is rarer than the alternatives in oxDNA, but cannot be ruled out for real DNA. Importantly, all three processes require the disruption of favorable stacking interactions between neighboring bases and considerable structural rearrangement, for each step of branch migration.

We also note that the branch migration pathways have relatively long-lived metastable intermediates (states represented by Figure 12A.ii and B.ii. These intermediates can be clearly identified on free-energy landscapes for individual steps of the branch migration process that are plotted in Figure 13A and B. Typical trajectories of branch migration are shown in Figure 13C and D. Figure 13C, taken from the start of branch migration, clearly shows the system moving from one base-pairing configuration to another via a metastable intermediate in which both invading and incumbent bases are close to the substrate base (branch migration via invasion). Figure 13D, taken from a later stage of branch migration, shows a change in base pairing occurring via the diffuse metastable intermediate in which both invading and incumbent bases are distant from the substrate base (branch migration via sequential disruption and formation of bonds). We note that trajectories of both types are possible at each stage, but that invasion is particularly dominant for the first step of branch migration. The fact that the intermediates are metastable, despite their high free energy, indicates that they are not easily accessible from the typical configurations of the system.

By contrast, the fraying of a few base pairs in the toehold and the subsequent detachment of the invading strand can occur relatively easily. As illustrated in Figure 14A–C, it is not necessary to disrupt at least one stacking interaction for every base pair that is broken (as it is in branch migration). Further, transition states in fraying are much closer in configuration space to the typical states of the system than they are for branch migration, making them easier to access. This finding is corroborated by Figure 14D, which shows that the coordinates of the system do not have to deviate far from their typical values before detachment occurs. For comparison with Figure 13, Figure 15A shows a free-energy landscape as a function of the separation of the base pairs in a two base-pair toehold. The minimum in the bottom left corresponds to the fully bound state, and the trajectory shown in Figure 15B is the same as that in Figure 14, illustrating both base pairs breaking in quick succession and the system leaving the bound state.
Although the fine details of branch migration processes will be sensitive to detailed chemistry neglected by oxDNA, the conclusion that each step of branch migration necessarily involves the breaking of more stacking interactions and a greater structural rearrangement than fraying of base pairs in the toehold is likely to be a robust one. As a result, if toehold melting and branch migration are to be simultaneously characterized by the IEL, the sawtooth amplitude of branch migration should be larger than the cost of fraying a base pair. We have not directly attempted to infer rates or transition free energies for processes that would correspond to elementary steps in the IEL with oxDNA. As we have discussed, such elementary processes can be relatively complex in oxDNA, involving effects that cannot be captured at the secondary structure level, making the precise definition of rate constants difficult. Nonetheless, understanding the process at an effective secondary structure level is helpful: oxDNA then justifies tuning the IEL to use an effective sawtooth amplitude significantly larger than the free energy of a single base-pair stack to slow the rate of branch migration.

**DISCUSSION**

We have argued that the kinetics of strand displacement can be explained by the destabilizing effect of single-stranded overhangs at the branch migration junction, and the relative slowness of branch migration compared to the fraying of a base pair in the toehold. How consistent are our findings and interpretations with the literature?

Our experiments infer two de-stabilizations: one due to the first ssDNA overhang at the junction (X20:Y00 versus X00:Y00; $\Delta G = 1.2$ kcal/mol at 25°C) and another due to the addition of the second overhang (X10:Y10 versus X20:Y00; $\Delta G = 2.0$ kcal/mol). We have not investigated the sequence dependence of these effects. Vasiliskov et al. (50) have measured the destabilization due to a single base overhang (analogous to X01:Y00 versus X00:Y00, which we did not measure) by immobilizing DNA oligonucleotides in 3D polyacrylamide gel microchips. Most of their sequence-dependent values lie between 0.5 and 1.0 kcal/mol at 37°C, which is similar to our temperature-extrapolated destabilization of 0.9 kcal/mol for X20:Y00 relative to X00:Y00. We would expect the destabilization to be slightly less for X01:Y00 than X20:Y00 from the intuition derived from our experimental studies of two overhangs, which indicate that the destabilization is not fully saturated by a single-base overhang.

oxDNA suggests that the destabilization due to overhangs arises from steric interference, which can be relieved by breaking coaxial stacking at the junction. This suggests that $\Delta G_o$ (and the free energy cost of introducing two overhangs to an overhang-free system) should not be significantly larger than the $\Delta G$ associated with the coaxial...
stacking transition in a nicked duplex. Several attempts to measure this quantity (48,50,74) have inferred values of $|\Delta G| \sim 0.2 - 2.3$ kcal/mol at 37°C and [Na$^+$] = 1 to 15 mM. Increasing [Na$^+$] to 100 mM was observed to increase stability of coaxial stacks by $\sim 0.4$ kcal/mol. We therefore conclude that our temperature-extrapolated destabilization of 2.8 kcal/mol between X01:Y10 and X00:Y00 at 37°C (and in high salt conditions) is not inconsistent with these data.

What do our findings imply for RNA? Walter et al. (75) have measured de-stabilization due to a single base overhang at a junction of duplexes as well as two single base overhangs on either side of a junction for RNA. They report a 2.8 kcal/mol de-stabilization for their analog of X01:Y01 versus X00:Y00 at 37°C, which is much higher than our extrapolation of 1.4 kcal/mol for DNA. As the geometry (A-form helix) and thermodynamics of RNA are significantly different from that of DNA, we would not expect analogous measurements for RNA to be quantitatively similar. For DNA, we find that X20:Y01 is significantly less stable than X01:Y01, and if this holds true for RNA, we would expect $\Delta G_p$ for RNA to be significantly higher than the 2.0 kcal/mol at 25°C we infer for DNA. However, efn2, a NN thermodynamics model for RNA that has been expanded to include experimentally measured terms for coaxial stacking and 1– and 2-nt overhangs at junctions (76,77), does not predict a free energy penalty similar to that for which we infer for DNA strand displacement (see Supplementary Section S7 and Supplementary Figure S14), possibly due to incorporating these terms only in ‘interior loops’.

Is the IEL’s inferred value of $\Delta G_s$ consistent with measured branch migration rates? Using the experimentally measured plateau height $\Delta G_p = 2.0$ kcal/mol and the corresponding inferred values of $\Delta G_s = 5.3$ kcal/mol for the IEL and $\Delta G_s = 3.6$ kcal/mol for the AEL, these models predict branch migration step times of $\sim 103$ and $\sim 53$ µs, respectively. These values are now more in line with previous experimental measurements: Radding et al. (28) report an average branch migration step time of 12 µs at 37°C in 10 mM Na$^+$ and Green and Tibbetts (29) estimate an upper limit of 20 µs at 65°C in 0.3 M Na$^+$. Our estimates are therefore somewhat larger, although it would be expected that the low temperature high salt conditions of Zhang and Winfree (30), which stabilize helices, would reduce branch migration rates. Indeed, such an effect is reported by Radding et al. (28). Finally, we note that Zhang and Winfree (30) infer a branch migration step time of $\sim 2.5$ ms from their phenomenological model, $\sim 12$–24 times larger still than our estimates. We attribute this difference to the absence in their model of a free energy penalty for initiating branch migration, i.e. what we here call $\Delta G_p$.

Are the IEL’s values for $k_{uni}$ and $\Delta G_{bp}$ compatible with measured fraying rates? Wetmur and Davidson (33) have inferred fraying rates from temperature-jump experiments. They predict 0.04 µs for fraying of a base pair at 25°C, which is faster than our values (0.2 µs for the IEL, 2 µs for AEL) and consistent with fraying being at least an order of magnitude faster than branch migration steps. We note, however, that these reported values were not directly measured in experiments and required modeling to extract. Furthermore, related measurements of the kinetics of RNA base pair fluctuations found significantly slower rates, with fraying times on the order of microseconds at 25°C (55–57). We therefore conclude that our parameterization is not inconsistent with well-established kinetic properties of nucleic acids.

We argue that the slowness of branch migration initiation relative to fraying is a key aspect in understanding strand displacement. In contrast, Reynaldo et al. (35) explained the low rate constant of zero-toehold displacement by positing that a certain number of base pairs, $n = 3$ or 4, are necessary to form a stable duplex, and thus shorter duplexes can be ignored or assumed to detach instantly. Our models differ by assigning favorable and increasingly strong thermodynamic energies for toeholds attaching by 1, 2 or more base pairs; there is no penalty for short duplexes. Instead, we can reinterpret their $n$ as the value for which the rate of fraying $n$ base pairs ($k_{uni}e^{-n|\Delta G_{bp}|/RT}$) equals that of initiating branch migration ($k_{uni}e^{-\Delta G_p/RT}$), which for the IEL gives $n = 4.3$. A similar calculation for the AEL gives $n = 3.8$.

CONCLUSIONS

We explain the dependence of strand displacement kinetics on toehold length using two factors: (i) the physical process by which a single step of branch migration occurs is significantly slower than the fraying of a single base pair and (ii) initiating branch migration incurs a thermodynamic penalty, not captured by state-of-the-art nearest neighbor models of DNA, due to the additional overhang it engenders at the junction. The slowness of branch migration relative to fraying is captured in the IEL by a sawtooth amplitude $\Delta G_s = 5.3$ kcal/mol, which is significantly larger than a single base-pair stack ($|\Delta G_{bp}| = 1.7$ kcal/mol). oxDNA provides physical justification for this by suggesting that branch migration steps are slower than fraying as each stage necessarily involves greater structural rearrangement and disruption of favorable stacking interactions. Initiating branch migration is slower than the average branch migration step because of the free energy penalty ($\Delta G_p$) incurred; from our experiments we infer $\Delta G_p \sim 2.0$ kcal/mol. oxDNA reproduces this penalty and suggests that it arises from steric interference of the single strands and duplexes at the branch migration junction. Specifically, the presence of an extra single-stranded overhang after the first step of branch migration causes additional disruption of (coaxial) stacking and reduction of conformational freedom as the strands are forced to bend away from each other.

Our work shows that toehold-mediated strand displacement involves four distinct time scales—rates of hybridization, fraying, branch migration and branch migration initiation—coupled with the energetics of base pairing in the toehold. The IEL model captures these rates via the parameters $k_{\text{bp}}$, $k_{uni}$, $\Delta G_s$, $\Delta G_p$ and $\Delta G_{bp}$. Presuming that these rates and energies are the essential determinants of strand displacement kinetics, the IEL analysis provides a framework for making kinetic predictions about different...
stranded displacement-based systems. For example, the IEL predicts that for short toeholds, \( k_{\text{eff}} \) scales as \( \sim 1/b \), where \( b \) is the branch migration domain length (see Supplementary Section S2 and Supplementary Figure S15). However, the dependence of \( k_{\text{bp}}, k_{\text{uni}}, \Delta G_s, \Delta G_p \) and \( \Delta G_{\text{bp}} \) on reaction conditions (e.g. salt concentrations and temperature) are only partially understood, impeding quantitative predictions for differing reaction conditions. Nonetheless, our current understanding does suggest that determining just those five parameters should be sufficient to predict toehold-mediated strand displacement under new conditions.

We expect that the IEL framework is also suitable for describing other nucleic acids. Toehold-mediated strand displacement has been demonstrated with RNA (78,79), but the dependence of kinetics on toehold length has not been characterized. Although several experimental studies have reported hybridization and fraying rates for RNA that are within the range observed for DNA (55,57, 80–82), branch migration and branch migration initiation rates are not well understood. There is, however, some evidence that with 1-nt toeholds, strand displacement rates for RNA are considerably slower than those for DNA (83), whereas the average base pairing energy is stronger (76); these observations are enough to predict that a plot of \( \log_{10} k_{\text{eff}}(h) \) for RNA would have a lower \( y \)-intercept but higher initial slope than that for DNA. Broadly, then, we would expect the qualitative features of RNA strand displacement kinetics to be similar to that of DNA; quantitative understanding will require further experiments.

Similarly, the effect of structural variations of toehold-mediated strand displacement, such as associative and remote toeholds (84,85), or initial mismatches in the branch migration domain (86) could be understood in terms of their effects on the fundamental rates and energies. These particular variants are likely to slow down branch migration initiation and hence increase \( \Delta A_{15,0} \), which could in turn permit greater design flexibility while engineering nucleic acid devices and systems.

Although the IEL provides a simple framework for understanding strand displacement, it does not predict or explain the underlying biophysical mechanisms, nor can it easily be adapted to handle more complex structural variants or explain sequence-dependent behavior in larger strand displacement cascades and dynamic DNA nanotechnology (24), where the kinetics of both on- and off-pathway reactions (such as ‘leak’ reactions analogous to zero-toehold strand displacement) are of great current interest. oxDNA is well suited to studying the underlying biophysical mechanisms and behaviors of structural variants, but it is computationally expensive for larger systems.

Efficient quantitative simulation of sequence-dependent kinetics for both desired and spurious strand displacement pathways would facilitate design and optimization of large systems in silico. SSK models based on NN thermodynamics, such as Multistrand, would be well-suited for such purposes, but as found here, current implementations require adjustment to match critical thermodynamic and kinetic features. Thermodynamically, accurate coaxial stacking terms and penalties for overhanging single-stranded sections of DNA at junctions should be incorporated into the model. Incorporating the relative rates of fraying and branch migration within the Multistrand model faces several challenges. Because every secondary structure state has a well-defined free energy, branch migration intermediates at the ‘top of the sawtooth’ (frayed junctions with two tails) cannot simply be pushed to higher energies, as was done for the IEL using \( \Delta G_s \). Moreover, Multistrand currently incorporates only one of the two possible branch migration mechanisms suggested by oxDNA (sequential disruption and formation of bonds). Three independent approaches could conceivably be pursued to better capture branch migration kinetics. First, an additional unimolecular rate constant parameter could be used for transitions local to the junction. Second, new transitions, such as ‘shift’ moves in Kinfold (43,87), could be incorporated to model alternative branch migration pathways. Third, Multistrand’s state space could be augmented, possibly by including features such as non-base-pair stacking interactions.

**SUPPLEMENTARY DATA**

Supplementary Data are available at NAR Online, including [88–91].

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