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## Signal-Pair Correlation Analysis of Single-Molecule Trajectories\*\*

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Single-molecule (SM) methods have dramatically expanded the study of dynamic processes in biomolecules. Since the first studies of single ion channels, [1] techniques such as Förster Resonance Energy Transfer (FRET) and force spectroscopy [2] have emerged which can be used to study a broad range of phenomena. By directly observing structural changes in a single molecule over time, the states occupied by the molecule can be identified, the possible transitions between them mapped out, and the transition rates measured. Such information has been used to build uniquely detailed pictures of various macromolecular processes, from ion-channel function [1] to molecular-motor motion, [3] enzymatic activity, [4] RNA-based regulation, [5] and biomolecular folding. [6,7]

A key feature of SM approaches is the ability to observe and characterize sub-populations, especially rare or transient states. Kinetic analysis of these states can, however, be challenging. Often the measured signal trajectories are mapped onto a set of discrete states, [8] for example by using thresholding or step-finding algorithms, [9] or more-sophisticated maximum-likelihood methods, [10,11] especially in combination with hidden Markov modeling (HMM). [12,13] Dwell-time distributions are then used to obtain kinetic information from the state trajectories. [14] A drawback of this approach is that any factors hindering state identification (e.g. noisy or overlapping signals, large differences in lifetimes) can introduce biases for which it is difficult to control.

An alternate strategy is to extract kinetics directly by analyzing correlations in the signal record, which has a number of advantages: it is more robust against noise and filtering artifacts, correlation fitting functions are easily calculated, and the fits can be used to test kinetic models directly. [15] Signal-intensity correlations have been used previously to analyze phenomena such as photo-physical processes [16,17] and two-state folding, [18] but such analysis has proven challenging for multi-state systems or processes having similar timescales, because the rates for all the transitions in the system are folded into a single correlation function from which they are difficult to recover separately. [19]

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Herein we present a new type of correlation analysis, based not on the entire signal but rather on discrete ranges of the signal associated with different states, which allows transition rates and kinetic schemes to be determined even in multi-state systems. Only part of the signal associated with a given state is needed for correlation analysis, hence the ranges can be chosen to minimize overlap between states in noisy data. A related approach was recently applied to protein diffusion measured by SM FRET.<sup>[20]</sup> This signal-pair correlation method works even for trajectories with states that overlap because of noise, states with low occupancy, and rates that are very similar or differ by orders of magnitude, all issues that can hinder other approaches.

The method involves two steps: First, the signal (extension, FRET efficiency, current ...) is divided into discrete ranges and the time correlations between all pairs of ranges ("signal pairs") are calculated. Next, specific kinetic models are tested by assigning a certain signal range to each state and fitting all cross-correlations between them with the functions derived for a given kinetic scheme. By repeating the fits for all possible schemes, the correct scheme can be validated empirically and the associated rates determined. The selection of signal ranges is simplified by using signal-pair histograms. These contain valuable information about the states present and the transitions between them, without the need to identify transitions or states.

To demonstrate this signal-pair correlation analysis, we used force-spectroscopy measurements of folding in three different molecules: 1) a two-state-folding DNA hairpin, for comparing the correlation analysis to the thresholding method;<sup>[21]</sup> 2) a three-state-folding DNA hairpin with a known kinetic scheme and overlapping states;<sup>[7]</sup> and 3) the hamster prion protein (HaPrP), which can fold into nonnative structures.<sup>[22]</sup> Folding trajectories consisting of records of the molecular extension as a function of time were measured at equilibrium for all molecules using dual-beam optical tweezers with a passive force clamp (Figure 1 a).<sup>[23]</sup>

We first analysed the two-state-folding DNA hairpin (Figure 1b, inset). Two states are very clearly observed both in the trajectory itself (Figure 1b) and in the extension histogram (Figure 1c): one at 548 nm representing the folded hairpin (F), the other at 559 nm for the unfolded hairpin (U). We also calculated 2D signal-pair histograms indicating how often an initial extension  $e_1$  at time t led to an extension  $e_2$  after time  $t+\tau$ ; three different delay times  $\tau$  are shown (Figure 1d-f). The number of events N in each 2D bins is given by Equation (1)

$$N(\tau, e_1, e_2) = \{ e(t) \in e_1 \pm \Delta e \mid e(t + \tau) \in e_2 \pm \Delta e \}$$
 (1)

where  $\Delta e$  is the bin half-width (in this case, 0.25 nm). Note that these signal-pair histograms are significantly different

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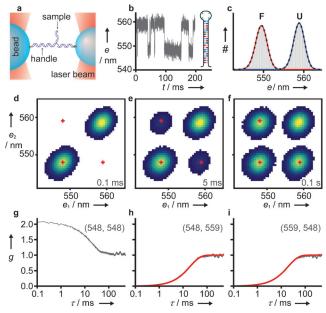


Figure 1. Signal-pair correlation analysis of DNA hairpin folding measured by force spectroscopy. a) Measurement scheme: the hairpin is held under tension between two optical traps, and the extension measured as a function of time. b) Extension trajectory showing two states. Inset: scheme of the hairpin with color-coded sequence (A/T: dark/light blue, G/C: dark/light red). c) The extension histogram shows two distinct peaks, folded (F) and unfolded (U). d)−f) Histograms of signal pairs at different time delays reveal dynamic information. Color scale: dark blue to yellow (2% to ≥50% of largest category ("bin")). Red crosses mark signal pairs ( $e_1$ ,  $e_2$ ) for correlation functions in (g)−(i). g) Autocorrelation of the folded state. h),i) Cross-correlations from F to U (h) and U to F (i) fit well to a two-state model (red).

from the similar-appearing and commonly-used transition maps, [24] since we only plot signal pairs for a given  $\tau$ , without assigning states or transitions. For  $\tau=0.1$  ms, two peaks are seen (Figure 1 d) along the identity line  $e_1=e_2$ , representing the two states. The width of the peaks arises from dynamic processes that are fast on the timescale of  $\tau$  but do not significantly change the molecular extension: in this case, the diffusion of the beads attached to the hairpin. [23] As  $\tau$  increases, two peaks emerge, (Figure 1 e) and grow in magnitude (Figure 1 f), reflecting a slower, dynamic process that changes the extension, namely the folding/unfolding of the hairpin.

To analyze the kinetics quantitatively, we calculated signal-pair correlation functions,  $g(\tau,e_1,e_2)$ , derived from Equation (1), varying  $\tau$  instead of  $e_1$  and  $e_2$  [Eq. (2)].

$$g\left(\tau,e_{1},e_{2}\right)=\frac{p\left(\left\{ e_{1},t\right\} \mid\left\{ e_{2},t+\tau\right\} \right)}{p\left(e_{1}\right)p\left(e_{2}\right)}=\frac{N\left(\tau,e_{1},e_{2}\right)T^{2}}{(T-\tau)N(e_{1})N(e_{2})}\tag{2}$$

Here,  $p(\{e_1,t\}|\{e_2,t+\tau\})$  is the joint probability to measure extension  $e_1$  at time t and extension  $e_2$  at time  $t+\tau$ ,  $p(e_i)$  is the probability to measure extension  $e_i$ ,  $N(e_i)$  is the number of time bins measuring extension  $e_i$ , and T is the total number of time bins. Correlation functions were calculated for all signal pairs (1 nm wide ranges) with sufficient counts; three are shown in Figure 1 g-i with signal pairs near the peak centers in

Figure 1 d–f. As expected, the autocorrelation ( $e_1 = e_2$ : Figure 1 g) shows a positively correlated signal, whereas the cross-correlations ( $e_1 \neq e_2$ : Figure 1 h,i) are negatively correlated and start from 0. All correlations change on the same time scale (about 10 ms).

To determine the microscopic folding rates, we fitted the cross-correlation signals to functions derived by matrix methods as described previously for fluorescence correlation analysis<sup>[25]</sup> (see Supporting Information). We reduced the number of fit parameters to a single parameter by using the fractional occupancies of the states obtained from the 1D histogram (Figure 1c) to relate folding and unfolding rates, according to the principle of detailed balance: if  $f_i$  is the fraction of state i and  $k_{i,j}$  the rate of the transition from state i to j, then  $k_{ij} = k_{ji}$  ( $f_i/f_j$ ). Fitting 25 extension-pair combinations close to the ones shown in Figure 1 h,i, we obtain  $k_{E,U} = k_{U,F} = (25 \pm 2) \, \mathrm{s}^{-1}$ , where the uncertainty is the standard error on the mean. These values agree with the results of the thresholding method ( $k_{E,U} = (27 \pm 2) \, \mathrm{s}^{-1}$  and  $k_{U,F} = (25 \pm 2) \, \mathrm{s}^{-1}$ ).

Having demonstrated that signal-pair correlation analysis works for a simple two-state system, we next applied it to a DNA hairpin whose stem sequence was designed to produce a partially folded, on-pathway intermediate<sup>[7]</sup> (Figure 2a, inset). This intermediate (I) is seen in the trajectory (Figure 2a) and extension histogram (Figure 2b) at 521 nm; F is at 508 nm and U at 528 nm. Note that around 30% of I overlaps U, making these states difficult to resolve well by thresholding analysis (Supporting Information, Figure S1–S3). As with the two-state hairpin, the signal-pair histogram for  $\tau = 0.1 \text{ ms}$ shows peaks only along the identity line  $e_1 = e_2$ : in this case, there are three peaks (Figure 2c). At  $\tau = 1$  ms, small crosspeaks appear between F and I as well as between I and U (Figure 2d). By  $\tau = 10$  ms, significant cross-peaks also arise between F and U (Figure 2e). The three cross-correlations shown in Figure 2 f-h correspond to the signal pairs (ranges 1 nm wide) centered on the red crosses in Figure 2c-e and reflect folding dynamic processes at approximately 1 and 10 ms.

To extract the microscopic rate constants and confirm empirically the expected sequential folding pathway of this hairpin, we chose three signal ranges corresponding to F, I, and U, balancing maximal signal with minimal overlap between states within each range. We then globally fitted the six cross-correlations between the chosen ranges (e.g. as in Figure 2 f-h and Supporting Information, Figure S5) to functions derived for each of the three different kinetic schemes in Figure 2i. Again, we used the fractional state occupancies from the extension histogram (Figure 2b) and detailed balance to reduce the number of fitting parameters for all six correlations to  $k_{\rm I,F}$  and  $k_{\rm U,I}$  (see the Supporting Information). Only the sequential model (Figure 2 f-h, red) fits the data well, confirming that the intermediate is on-pathway. Fits to 125 extension pairs close to those shown in Figure 2c-e yield  $k_{\rm LF} = (0.21 \pm 0.01) \, {\rm ms}^{-1}$ ,  $k_{\rm U,I} = (79 \pm 4) \, {\rm s}^{-1}$ ,  $k_{\rm F,I} = (34 \pm 1) \, {\rm s}^{-1}$ , and  $k_{\rm L,U} = (0.43 \pm 0.03) \, {\rm ms}^{-1}$ . We note that neither these four rates nor the kinetic model could be extracted from a standard correlation analysis (Supporting Information, Figure S4). By repeating the signal-pair correlation analysis for trajectories measured at different forces (Supporting



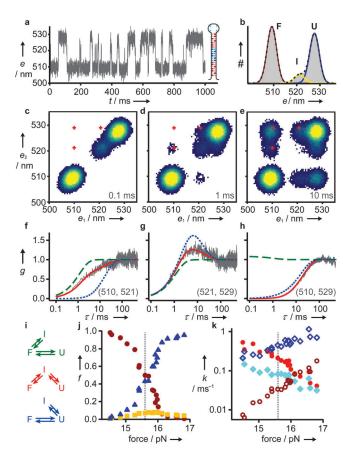


Figure 2. Analysis of a three-state folding DNA hairpin. a) The extension trajectory reveals 3 states. Inset: scheme of the hairpin with colorcoded sequence as in Figure 1. b) The extension histogram shows an intermediate state (I) overlapping the unfolded state (U). c)-e) Signalpair histograms at different time delays. f)-h) The cross-correlations for signal pairs (e1,e2) were fitted globally to three different kinetic schemes i) off-pathway (green from F, blue from U) or on-pathway (red). The fits indicate the intermediate is on-pathway. j) Forcedependent state occupancies from extension histograms: F (red circles), I (orange squares), U (blue triangles). k) Microscopic rates for the on-pathway mechanism:  $k_{\text{F,I}}$  (dark red open circles),  $k_{\text{I,F}}$  (light red filled circles),  $k_{\text{LU}}$  (dark blue open diamonds), and  $k_{\text{LL}}$  (light blue filled diamonds). Dotted gray lines in (j) and (k) mark the force for plots (a)-(h).

Information, Figure S5), where the state occupancies change from mostly folded to mostly unfolded as the force is increased (Figure 2j), the force-dependence of each of the microscopic rates was determined (Figure 2k). The method clearly works well even for low-occupancy states (Supporting Information, Figure S5). Knowing the force-dependent rates allows properties such as the height and location of the folding energy barrier, to be determined for each transition. [26]

As a final example, we applied the method to a folding trajectory of the protease-resistant fragment of the C179A/ C214A mutant of HaPrP (Figure 3a, inset). The trajectory (Figure 3a) and extension histogram (Figure 3b) show F at 579 nm, U at 598 nm, and I at 591 nm. The low occupancy and short lifetime of I, compounded with the strong overlap between I and U (about 75% of I overlaps U), make these data very difficult to analyze with standard methods. The

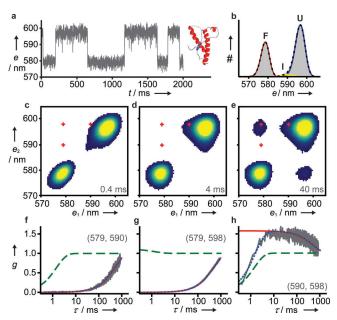


Figure 3. Analysis of HaPrP folding. a) The folding trajectory shows two major states. Inset: protein structure (Protein data bank (PDB) code: 1B10). b) The extension histogram reveals a rarely-occupied intermediate state I (yellow). c)-e) Signal-pair histograms at different time delays. Red crosses: signal pairs used for correlation functions. f)-h) Cross-correlations for the signal pairs centered at  $(e_1,e_2)$ , fit globally to the same models as in Figure 2i, reveal that I is offpathway, entered exclusively from U.

signal-pair histogram at  $\tau = 0.4$  ms (Figure 3c) shows two major peaks for F and U, and a small peak for I. Due to its short lifetime, this peak for I vanishes by  $\tau = 4$  ms (Figure 3 d), but cross-peaks between I and U remain. By  $\tau = 40 \text{ ms}$ (Figure 3e), additional cross-peaks appear between F and U.

Cross-correlations calculated for the signal pairs (ranges 1 nm wide) shown in Figure 3c (red crosses) reveal that F changes into the U and I after hundreds of ms (Figure 3 f,g). The I to U correlation (Figure 3h) also has a rapid component at approximately 1 ms. Fitting all six signal-pair crosscorrelations globally to the kinetic schemes in Figure 2i (Supporting Information Figure S6), we find that sequential folding is inconsistent with the data. Instead, the fits indicate that I is off-pathway, reached only from U (Figure 2i, blue). Fitting 125 signal-pair combinations close to those shown in Figure 3c yielded  $k_{EU} = (1.43 \pm 0.03) \text{ s}^{-1}$ ,  $k_{UF} = (0.83 \pm 0.03) \text{ s}^{-1}$ 0.02)  $s^{-1}$ ,  $k_{LU} = (0.6 \pm 0.1) \text{ ms}^{-1}$ , and  $k_{UJ} = (26 \pm 4) \text{ s}^{-1}$ . This example shows that the method can identify rates and kinetic schemes even with states that are difficult to distinguish and rarely populated, and with processes occurring on timescales differing by several orders of magnitude.

The method we have demonstrated herein is clearly applicable to a broad range of systems, measurement modalities, and simulations, beyond force spectroscopy of folding. It is also readily generalized to combine different types of signal, for example, from multi-parameter fluorescence spectroscopy, and can be extended beyond signal-pairs to signalmultiples to improve the state selection. A key feature of the method is that the assignment of states, while intuitive

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(and simplified using the signal-pair histograms), is done at the end of the analysis-when choosing correlations for fitting—thereby avoiding many of the biases that arise when converting signals into state trajectories, especially with noisy data or short-lived states. Because of this, the method is also useful as a complementary tool for validating the results from other analytical approaches, such as hidden Markov models based on specific assumptions about the system under study. Signal-pair correlation analysis thus provides a powerful approach for extending the reach of single-molecule data.

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# **Supporting Information**

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#### Sample preparation and measurement

The DNA hairpin constructs were produced as described previously<sup>[1]</sup>. Briefly, a 621-basepair (bp) segment of double-stranded (ds) DNA labeled with digoxigenin and a 1,036-bp fragment of dsDNA labeled with biotin were bound to the hairpin sequences (Integrated DNA Technologies) shown below. The DNA handles were attached to 600-nm diameter polystyrene beads (Bangs Laboratories) coated with avidin-DN (Vector Laboratories) and 730-nm diameter beads coated with polyclonal anti-digoxigenin (Roche Diagnostics).

The hairpin sequences were: 5' gegge ggtta atagg gegee tttt ggege cetat taace geege 3' and 5' gagte aacgt etgga teeta tttt tagga teeta acgtt gaete 3' for the two- and three-state-folding hairpins, respectively.

We used truncated wild-type Syrian Hamster Prion protein (90–232), cloned into the pET-15b plasmid, with the cysteines at residues 179 and 214 exchanged for alanines, and two new cysteines introduced for handle attachment, one within the thrombin cleavage site of the HisTag at the N-terminus and the other replacing serine 232 at the C-terminus of the prion protein. The protein was produced and purified similarly to a method described previously<sup>[2]</sup>. Purity and identity were checked by SDS PAGE and Western blotting (anti-prion(109-112) clone 3F4, Millipore); native conformation was confirmed by circular dichroism spectroscopy. The DNA handles were attached as described elsewhere<sup>[3]</sup>: the reduced terminal cysteines were reacted with sulfhydryl-labeled DNA handles of different lengths prepared by PCR (one 798 bp, labelled by biotin, the other 1261 bp, labelled with digoxigenin).

All molecular constructs were diluted to ~ 500 fM in 50 mM MOPS, pH 7.0, with 200 mM KCl and oxygen scavenging system and measured at 23C using a custom dual-beam optical trap, similar to previous reports<sup>[3]</sup>. The extension was measured as a function of time at 10-50 kHz sampling rate with fixed trap separation, at various different forces. Data in figure 1 were measured at non-constant force<sup>[4]</sup>, data in figures 2 and 3 were measured at constant force using a passive force clamp<sup>[5]</sup>.

#### **Data analysis**

The data were median-filtered offline in 0.1 ms and 0.4 ms windows respectively for hairpin and prion data. Measurement times were 40 s, 15 s, and 13.6 min for the data in figures 1, 2, and 3, respectively. 1D extension histograms were binned in 0.2 nm increments and fitted with the point-spread function of the instrument. The fractions were calculated as the area of one state divided by the total area. Correlation functions were calculated for 1 nm signal ranges using Python (www.python.org; code is available upon request). Further data processing, including global fitting, was done with Mathematica (Wolfram Research).

For each state pair, correlation fit functions,  $g_{x,y}(t)$ , were derived from the following equation using a matrix-based formalism<sup>[6]</sup>:

$$g_{x,y}(t) = \frac{\mathbf{1}^{\mathrm{T}} \mathbf{V}_{x,y} e^{\mathbf{K}t} \mathbf{V}_{x,y} \mathbf{p}_{ss}}{\left(\mathbf{1}^{\mathrm{T}} \mathbf{V}_{x,y} \mathbf{p}_{ss}\right)^{2}}$$
(S1)

where x is the initial and y the final state,  $\mathbf{1}$  is the unit vector, and  $\mathbf{T}$  denotes transpose.  $\mathbf{K}$  is the rate matrix whose elements  $K_{i,j}$  denote the rates from j to i, with  $K_{j,j} = -\sum_{i \neq j} K_{i,j}$ . By setting certain microscopic rates to zero, fit functions for defined schematic schemes can be derived.  $\mathbf{V}_{x,y}$  is the matrix of transitions being monitored and has only one non-zero element:  $V_{i=x,j=y} = 1$ . Finally,  $\mathbf{p}_{ss}$  is the steady-state probability distribution and can be derived by solving  $\mathbf{K}\mathbf{p}_{ss} = 0$ .

The requirement for detailed balance reduces the number of fitting parameters needed for the three-state models shown in Fig. 2 i from 4 to 2, as listed in table S1.

scheme	rate-fraction relations	
I-F-U	$k_{F,I} = k_{I,F}(f_F/f_I)$	$k_{F,U} = k_{U,F}(f_{F}/f_{U})$
F-I-U	$k_{F,I} = k_{I,F}(f_F/f_I)$	$k_{I,U} = k_{U,I}(f_I/f_U)$
F-U-I	$k_{F,U} = k_{U,F}(f_F/f_U)$	$k_{I,U} = k_{U,I}(f_I/f_U)$

Table S1: Relations between fractions and rates for specific three-state kinetic schemes

The selection of the extension ranges for each state is a compromise between ensuring a small contribution from unwanted states and a high occupancy of the desired state to obtain good statistics. Both criteria were judged based on the 1D and 2D histograms of the measurements. Standard errors were obtained by fitting a set of 5 different extension ranges chosen from within the same cross-peaks in the 2D histogram. For the two-state system this resulted in calculating 25 correlations; for the three-state systems, 125.

### Comparison to thresholding analysis and correlation analysis of the entire signal

In order to illustrate concretely some of the advantages of the signal-pair correlation analysis method, we compared it to the results of a dwell-time analysis of the states identified by thresholding and a correlation analysis of the entire signal.

The comparison with the thresholding-based approach was performed with the three-state-folding hairpin at 15.7 pN (the example shown in figure 2 a-h), where there is a significant overlap between the peaks for states I and U. For the dwell-time analysis, the data were first median-filtered, using a range of filtering window sizes (since filtering is a typical way to reduce noise and therefore the overlap of states). Thresholds were then chosen to separate the three states, based on the 1D extension histogram of the filtered trajectory (figure S1 a). Whereas the threshold between F and I was straightforward to determine, the position of the threshold between I and U state was less clear due to the significant overlap between the states. Therefore three different thresholds were chosen. From the resulting state trajetories, the dwell-times in each state were collected and the dwell-time survival distributions (figure S1 b-d) were fitted with a single exponential function. The resulting lifetimes are shown in figure S2 for the three different choices of I-U threshold and 7 different choices of filtering window, and compared to the results of the signal-pair analysis performed at 5 comparable filtering windows.

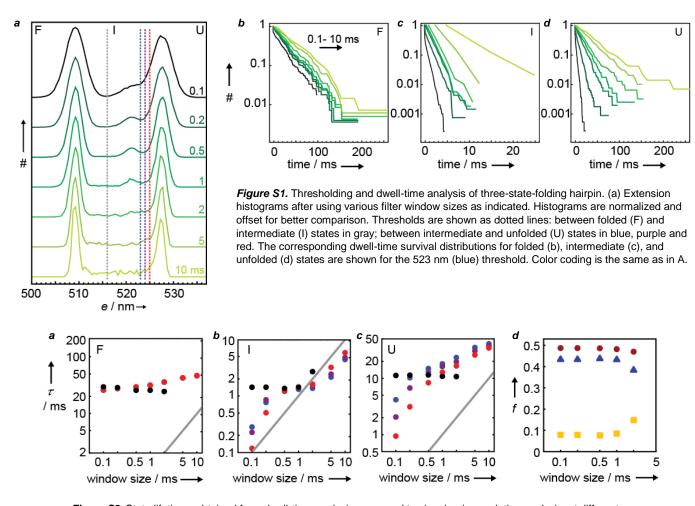
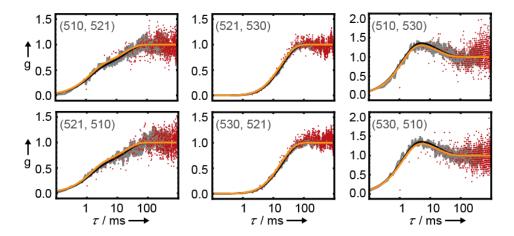


Figure S2. State lifetimes obtained from dwell-time analysis compared to signal-pair correlation analysis, at different filtering window sizes and threshold locations. (a)-(c) Lifetimes for folded (F), intermediate (I), and unfolded (U) states were obtained using different thresholds between I and U: 523 nm (blue), 524 nm (purple), and 525 nm (red). Lifetimes from signal-pair correlation analysis are shown in black. The gray line shows where the lifetime equals window size. (d) The fractional occupancy of states F (red circles), I (yellow squares) and U (blue triangles) determined from the extension histograms do not vary with filtering window size until the window size exceeds the lifetime of I.

Since F and I are well separated, the lifetimes for F (figure S2 a) obtained from dwell-time analysis are relatively insensitive to the filtering window size; as expected, it is also independent of the I-U threshold location. Due to the overlap of I and U, however, the lifetimes for these two states (figures S2 b and c) as determined from the dwell-time analysis depended strongly on both the filtering window size and the I-U threshold location, producing results which varied by well over an order of magnitude. In contrast, the lifetimes determined from signal-pair correlation analysis were robust against changes in the filter window size, varying little for all three states over the range 0.1-1 ms (although the lifetime for I calculated at a window size of 2 ms starts to increase, because the filtering window becomes too large compared to the lifetime and begins to alter the results of the analysis). This demonstrates the robustness of the signal-pair correlation method against extra

noise in the measurement. Whereas the lifetimes for F determined by the two methods agree fairly well, the lifetimes for I and U do not, because of the strong dependence of the dwell-time analysis on filtering window and threshold position.

To illustrate further how extra noise affects the signal-pair analysis, correlations calculated from data filtered at 0.1 ms and 1ms are shown in figure S3, together with the corresponding fits. The global fit to all six correlation functions is robust despite the change in noise, as all the fits appear very similar for both filtering levels.



**Figure S3.** Signal-pair correlation analysis of the trajectory for the three-state-folding hairpin: unfiltered (gray) and filtered in a 1 ms window (red). The respective global fits in black and orange are essentially the same. The center of the pair of signal ranges for each plot is denoted  $(e_1, e_2)$ .

To compare the results of the signal-pair analysis to a standard autocorrelation analysis, we also calculated the autocovariance of the entire trajectories of figures 1 and 2. Since the covariance is directly proportional to the autocorrelation function, both will reveal the same decay rates<sup>[7]</sup>. The autocovariance functions (figure S4) are fit very well by single-exponential decay functions. In case of the two-state hairpin (figure S4 a), the fit reveals a decay rate of 50 s<sup>-1</sup>, in good agreement with both the thresholding results ( $k_{F,U} + k_{U,F} = 52 \pm 4 \text{ s}^{-1}$ ) and signal-pair correlation results ( $k_{F,U} + k_{U,F} = 50 \pm 4 \text{ s}^{-1}$ ). In the case of the three-state hairpin, it is obvious that the single decay rate (49 s<sup>-1</sup>) cannot be used to extract either the four microscopic rates of the system or the kinetic scheme, information which is readily available from signal-pair correlation analysis.

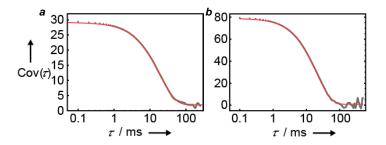
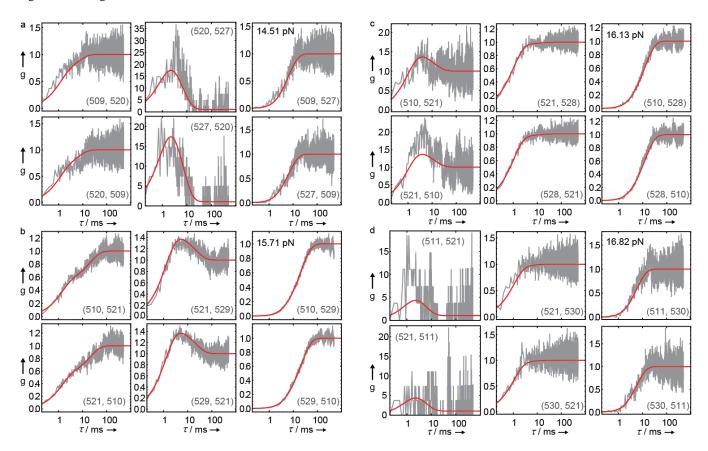


Figure S4. Autocovariances of the trajectories from (a) figure 1 and (b) figure 2. Covariance functions are shown in gray and single exponential fits in red.

#### Additional correlation functions

Correlations calculated from six cross-peaks in the 2D histograms were used for global fitting of the three-state kinetic models. For the three-state-folding hairpin, these are shown in figure S5 at four different forces: 14.5 pN, the lowest force measured (figure S5 a), 15.7 pN (figure S5 b), 16.1 pN (figure S5 c), and 16.8 pN, the highest force measured (figure S5 d). Note that for each force, all six correlations were globally fitted with only two free parameters, due to the constraints from the state occupancies (detailed balance). At the extreme ends of the force range, state occupancies were very low: at the lowest force, the intermediate (2) and unfolded (3) states are occupied only 1% of the time each; at the highest force, the folded (1) and intermediate states are occupied respectively 0.7% and 3.4% of the time. Measurement length and filtering was the same for all forces.



**Figure S5.** Correlation analysis of the three-state-folding hairpin (figure 2) at four different forces, including the lowest (a) and highest (d) force used in the study. The red line represents the global fit to the sequential-folding scheme. The force is shown in the upper-right plot of each set of correlations. The signal ranges used were 1 nm wide, with the midpoints of the range denoted as  $(e_1, e_2)$  for each plot.

The full set of fitted correlations for the data in figure 3 is shown in figure S6.

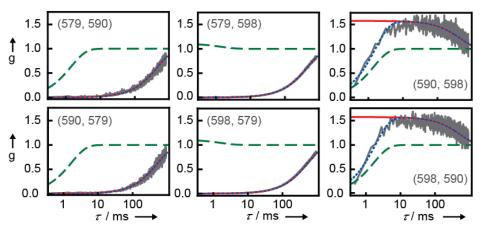


Figure S6. Correlation analysis of the prion protein (compare figure 3) showing all six cross-correlation functions used for the global fit. For each plot (e<sub>1</sub>, e<sub>2</sub>) denotes the midpoints of the 1-nm wide signal ranges used for the correlation function. Red: sequential folding model; green: off-pathway intermediate entered from folded state; blue: off-pathway intermediate entered from unfolded state.

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