Unfolding times for proteins in a force clamp

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The escape process from the native valley for proteins subjected to a constant stretching force is examined using a model for a β-barrel. For a wide range of forces, the unfolding dynamics can be treated as one-dimensional diffusion, parametrized in terms of the end-to-end distance. In particular, the escape times can be evaluated as first passage times for a Brownian particle moving on the protein free-energy landscape, using the Smoluchowski equation. At strong forces, the unfolding process can be viewed as a diffusive drift away from the native state, while at weak forces thermal activation is the relevant mechanism. An escape-time analysis within this approach reveals a crossover from an exponential to an inverse Gaussian escape-time distribution upon passing from weak to strong forces. Moreover, a single expression valid at weak and strong forces can be devised both for the average unfolding time as well as for the corresponding variance. The analysis offers a possible explanation of recent experimental findings for ddFLN4 and ubiquitin.

Single-molecule pulling experiments have become an important and widely used tool for examining mechanical properties of proteins. These experiments have stimulated a renewed interest in the escape processes from metastable potential wells in the presence of a biasing force. Traditionally, the dependence of the escape rate k on the stretching force F has often been modeled using the phenomenological Bell formula $k(F) = k_0 e^{β F x_u}$, where $x_u$ is the distance from the native to the transition state and assumed constant ($β = 1/k_BT$, with $k_B$ being Boltzmann’s constant and $T$ the temperature). The zero-force rate $k_0$ satisfies $k_0 \propto e^{-β ΔG_0}$, where $ΔG_0$ is the escape free-energy barrier at zero force. There are, however, uncertainties about how to extract the zero-force properties $k_0$, $x_u$, and $ΔG_0$ from observed escape rates at non-zero force. One problem is the unknown constant of proportionality in the expression for $k_0$. Another difficulty is that the distance $x_u$ between the native free-energy minimum and the unfolding barrier, which is assumed constant in the Bell formula, generally depends on the applied force.

To address these problems, several generalizations of the Bell formula have recently been proposed. Most of the extensions are based on the same underlying picture as for the Bell formula; the protein is viewed as a Brownian particle moving in a tilted one-dimensional potential, $G(x) = G_0(x) - F x$, where $G_0(x)$ is the equilibrium free-energy profile. Using different approximations and parametrizations of $G_0(x)$, key properties of the escape process have been analyzed, like the mean and variance of the rupture force at constant velocity pulling. It was further shown that the approach of Dudko, Hummer and Szabo (DHS) is able to describe experimentally observed deviations from the Bell formula for the protein ddFLN4.

These extensions based on Kramers theory assume that the escape barrier is high compared to $k_BT$, leading to single-exponential kinetics. Very recently, Yew et al. analyzed deviations from single-exponential kinetics in unfolding simulations based on a Cα model. By including the next-to-leading term in an eigenfunction expansion, they obtained an improved description of the unfolding dynamics at strong force. However, a comprehensive picture describing $k(F)$ and the full escape-time distribution at both weak and strong force is still missing. A key parameter when describing the force dependence is the critical force $F_c$, at which the escape barrier disappears. In the DHS approach, one has $F_c = ΔG_0/x_u$, where $x_u$ is a model parameter ($ν = 1$ corresponds to the Bell formula). The above-mentioned ddFLN4 analysis (with $ν = 1/2$ or $2/3$) suggests that $F_c \sim 80–110$ pN for this protein. The titin module I27, on the other hand, $F_c$ appears to be significantly larger ($ΔG_0/x_u \sim 640$ pN). Due to different $F_c$, when analyzing experimental data, the strong-force regime $F > F_c$ may or may not be relevant, depending on the protein.

In this Letter we investigate the response of a model protein to a wide range of constant pulling forces. We show that, once the free-energy landscape is known with sufficient accuracy, the usual Smoluchowski equation in one dimension is sufficient to obtain a good estimate of the average escape time from the native valley and the associated variance. Two force regimes, separated by the critical force $F_c$, are observed. For $F < F_c$, unfolding occurs through a thermally activated escape process. For $F > F_c$, the unfolding dynamics can instead be interpreted as pure diffusion with an external bias. The transition from the weak- to the strong-force regime is accompanied by a drastic change in the shape of the escape-time distribution, from exponential to inverse Gaussian. The applicability of this approach to real proteins, at forces studied experimentally, is addressed using recently reported data for ddFLN4 and ubiquitin.
The protein model we consider is the 3D off-lattice BPN model \[14, 15, 16\], where each residue is represented by a single point and is one of the following three types: hydrophobic (B), polar (P) or neutral (N). We study a 46-residue sequence which is known to form a four-stranded β-barrel in its native state. The folding \[14, 15, 16, 17\] and mechanical unfolding \[18, 19\] of this sequence have been extensively studied. We analyze via Langevin dynamics the response of this model protein to external forces acting on the chain ends in proximity of its folding temperature, namely at \(T = 0.3\) \[19\]. Parameters values are as in Ref. \[19\] and all model quantities are dimensionless; for a comparison with physical units, see Ref. \[16\].

A typical unfolding trajectory begins with a waiting phase, where the end-to-end distance \(ζ\) stays close to its native value. This phase is followed by a sudden increase in \(ζ\). A fundamental question is whether the escape from the native valley can be effectively described as one-dimensional diffusion, parametrized in terms of \(ζ\). Based on this assumption the unfolding process is commonly described as the motion of a point-like Brownian particle in the potential \(G(ζ) = G_0(ζ) - Fζ\), where \(G_0(ζ)\) is the equilibrium free-energy profile. The average first passage time \(τ(ζ)\) at a threshold \(ζ_s\) for a particle with initial position \(x \in [ζ_0, ζ_s]\) can be obtained by solving the Smoluchowski equation. One finds that \[9\]

\[
τ(x) = \frac{βMγ}{γ_ζ} \int_{ζ_0}^{ζ_s} dy \frac{e^{βG(y)}}{γ_ζ} \int_{ζ_s}^{ζ} dz \ e^{-βG(z)} \tag{1}
\]

where \(M\) is the particle mass and \(γ\) the damping constant. The boundaries at \(ζ_0\) and \(ζ_s\) are reflecting and absorbing, respectively. When using Eq. (1) to calculate the escape time from the native valley, \(ζ_0\) is the native \(ζ\) and \(ζ_s\) is that of the saddle, or barrier, to be crossed. The escape time is obtained as \(τ_s = τ(ζ_0)\). In our simulations, escape times are measured using a threshold slightly larger than \(ζ_s\), to avoid saddle recrossing \[9\].

We begin by testing the escape-time prediction \(τ_s\) directly against simulation results for the BPN protein, without making any further assumption on the form of \(G(ζ)\). For this purpose, we determine \(G(ζ)\) numerically, using methods described in Ref. \[14\]. Fig. 1 shows the calculated free-energy profile at zero force, \(G_0(ζ)\), which exhibits a pronounced native minimum at \(ζ_0 \approx 2.0\) and a barrier at \(ζ_s \approx 5.25\). The height of the barrier is \(ΔG_0 = G_0(ζ_s) - G_0(ζ_0) \approx 5.62\). The application of a stretching force \(F\) tilts the free-energy landscape to \(G(ζ) = G_0(ζ) - Fζ\) and reduces the barrier height \(ΔG\). As shown in the inset of Fig. 1, \(ΔG\) decreases almost linearly with \(F\). The barrier finally disappears at \(F_c \approx 1.83\).

Knowing \(G(ζ)\), the escape-time prediction \(τ_s\) can be obtained by numerically evaluating the double integral in Eq. (1). In Fig. 2 we compare \(τ_s\) with simulated escape times. The agreement is very good for strong forces \((F \gtrsim 3)\) as well as at weak forces \((F \lesssim 1.2)\). Due to computational limitations, it was impossible to investigate forces \(< 0.6\). The regime in which the simulated escape times are most difficult to reproduce is around the critical force \(F_c\), where there is no clear free-energy gradient either towards or away from the native state. In this regime, the details of the free-energy profile matter. It is remarkable, however, that this simple picture, without employing any fitting parameter, is able to describe the behavior at both strong and weak forces, despite escape-time differences of almost six orders of magnitude.

This analysis, based on the full profile \(G(ζ)\), addresses in a direct manner the question of whether or not the system can be described in terms of one-dimensional diffusion. In unfolding experiments, \(G(ζ)\) is unknown, and the challenge
In the latter regime, it is known that first-passage times follow a so-called inverse Gaussian distribution \[24\]. This distribution is given by

\[
P(t) = \frac{t}{\sqrt{2\pi \tau}} e^{-(t-\tau)^2/(2\tau^2)}
\]

is to extract the main features of the free-energy landscape from measured escape times. This task is greatly facilitated if the free energy can be linearly approximated in the interval \([\zeta_0, \zeta_s]\), as \(G(\zeta) = (F_c - F)(\zeta - \zeta_0)\) (up to an additive constant). With this approximation, the integrals in Eq. (1) can be evaluated analytically. The resulting expression, for the average escape time of a diffusive particle in one dimension in the presence of a bias \((F\text{ in the present context})\), is \[1\]

\[
\tau_L = \frac{M\gamma a}{F - F_c} - \frac{M\gamma k_B T}{(F - F_c)^2} \left[ 1 - e^{-(F-F_c)a} \right]
\]

where \(a = \zeta_s - \zeta_0\) is the distance between the reflecting and absorbing boundaries. Unlike the result reported in Ref. \[1\], Eq. (2) includes the effect of a zero-force barrier, represented by the term \(F_c a\). The singular terms at \(F = F_c\) in Eq. (2) cancel out, as they should.

The assumption that \(G(\zeta)\) is linear between the native state and the saddle is quite well satisfied for the BPN protein (see Fig. 1). Actually, the escape times obtained using this approximation, \(\tau_L\), essentially coincide with the estimated \(\tau_S\) obtained using the full \(G(\zeta)\), as can be seen from Fig. 2. Note that Eq. (2), like Eq. (1), has no parameter that needs to be fitted, because we can use the value of \(F_c\) previously determined. While Eq. (2) well describes the escape time down to the lowest forces that could be studied, one should still be cautious in using this expression to extrapolate to zero force, because a “turnover” to a force-independent process is likely to occur at weak force \[21\]. The extent of this weak-force regime might be non-negligible if the temperature is high \[21\].

The variance of the escape time is in the Smoluchowski approach given by \(\tau_S = \tau_{2,S} - \tau_{2,L}\) where the second moment \(\tau_{2,S}\) reads \[22\]

\[
\tau_{2,S} = 2 \frac{M\gamma}{k_B T} \int_{\zeta_0}^{\zeta_s} dy e^{yG(y)} \int_{\zeta_0}^{y} dx e^{-yG(x)} \tau(x)
\]

Like the mean, the variance can be obtained analytically if \(G(\zeta)\) depends linearly on \(\zeta\). This estimate of the variance, \(\tau_{2,L}\), can be found in Eq. (S1) \[23\]. The inset of Fig. 2 shows our simulation results for the variance of the escape time for the BPN protein, along with the estimates \(\tau_S\) and \(\tau_L\). For \(F < F_c\), \(\tau_S\) and \(\tau_L\) are almost identical; while for \(F > F_c\), \(\tau_L\) is slightly larger than both \(\tau_S\) and the simulation results, although the corresponding three average times are very similar in this regime. Overall, both \(\tau_S\) and \(\tau_L\) agree well with the simulation results.

It is informative to go beyond the first and second moments and also study the full probability distribution of the escape time. For \(F > F_c\), we find that the escape-time distribution of the BPN protein to a very good approximation is exponential, \(P(t) = \tau^{-1}e^{-t/\tau}\), with \(\tau\) being the mean (see Fig. 3b). This observation confirms that at weak forces, where a free-energy barrier is still present, the main escape mechanism is thermal activation. At \(F \sim F_c\), the escape process changes in character, from a thermally activated process to a diffusive process driven by an external bias (force). In the latter regime, it is known that first-passage times follow a so-called inverse Gaussian distribution \[24\]. This distribution is given by

\[
P(t) = \frac{t}{\sqrt{2\pi \tau^3}} e^{-(t-\tau)^2/(2\tau^2)}
\]
where $\tau$ is the mean and $\Gamma = V/\tau$, $V$ being the variance. This expression indeed provides a very good description of our simulation results at strong forces, as illustrated in Fig. 3b. It should be noticed that this comparison does not involve any parameter fitting, because $\tau$ and $V$ are determined directly from the simulations.

Previous studies have used a log-normal distribution, rather than the inverse Gaussian, to describe the escape-time distribution at strong forces [25, 26]. While the log-normal distribution is similar to the inverse Gaussian (see Fig. 3b), there is no theoretical background to justify its use in the present context. The inverse Gaussian distribution is, by contrast, known to arise from biased Brownian motion [24], which provides a simple physical picture of the unfolding dynamics at strong forces.

Having seen that our approach provides a good description of the unfolding dynamics of the BPN protein, we now turn to two real proteins, ddFLN4 and ubiquitin. Two results of the above analysis are particularly useful when comparing with experimental data. The first is Eq. (2), which provides an approximate closed-form expression for the average escape time $\tau(F)$ at both weak and strong forces. The second result is that the onset of the non-exponential strong-force behavior of $\tau(F)$ is accompanied by a change of shape of the escape-time distribution, from exponential to inverse Gaussian.

Experimental unfolding times for ddFLN4 show, as mentioned earlier, clear deviations from the Bell formula [3, 4]. It has been demonstrated [3] that the DHS approach [4] describes the data well. In Fig. 4, we show a fit of our Eq. (2) to the same data. The fit is good, and the fitted values $a = 1.1$ nm (corresponding to $x_u$) and $\Delta G_0 = F_c a = 9.50$ kcal/mol are consistent with the results of Ref. [3].

![FIG. 3: (Color online) Escape-time distribution $P(t)$ for the BPN protein at two different forces. Large (black) dots are simulation results. a) $F = 0.6$ ($< F_c$). The (red) curve is an exponential ($\tau = 7.02 \cdot 10^5$). b) $F = 2.2$ ($> F_c$). The (red) curve is an inverse Gaussian ($\tau = 71.1, V = 1.35 \cdot 10^3$), whereas small (blue) dots represent a log-normal fit to data.](image)

![FIG. 4: (Color online) Average unfolding time versus force for ddFLN4. The symbols (explained in detail in Ref. [23]) represent experimental data, originally obtained at constant velocity [4] and transformed to constant-force conditions in Ref. [3]. The (black) curve is a fit of Eq. (2) ($M \gamma = 0.012$ pN s/nm, $a = 1.1$ nm and $F_c = 60$ pN). The inset shows ubiquitin data from AFM force-clamp experiments reported in Ref. [13].](image)
very interesting to see whether the escape-time distribution is inverse Gaussian at, say, 100 pN, but this distribution has not been evaluated, as far as we know. For ubiquitin, the escape-time distribution has been measured experimentally at 110 pN \[12\]. The data were found to be well described by a log-normal distribution \[12\], which is very similar to the inverse Gaussian one found above at strong forces. Our approach thus offers an explanation of the shape of the observed distribution. This explanation requires that \( F_c < 110 \text{ pN} \). Very recent experimental \( \tau(F) \) data for ubiquitin \[13\] show signs of deviations from the Bell formula (see inset of Fig. 4). However, it was found that the data could not discriminate between the Bell and DHS formulas \[13\]. Neither are the data sufficient to permit a stable fit of Eq. (2), which would have given us an independent estimate of \( F_c \). The assumption that \( F_c < 110 \text{ pN} \) seems, however, fully consistent with the experimental \( \tau(F) \) data.

In this Letter we have shown for a model protein that the unfolding process from the native valley under force-clamp conditions can be modeled as a Brownian motion in a tilted one-dimensional free-energy landscape. Moreover, it turned out that this description could be further simplified with a surprisingly small loss of accuracy, by adopting a linear approximation for the free energy. This analysis links deviations from the Bell formula for \( k(F) \) for \( F > F_c \) to an altered shape of the escape-time distribution, from exponential to inverse Gaussian. Comparison with experiments indicates that the strong-force regime might set in at relatively weak force \( (F_c \lesssim 100 \text{ pN}) \) for both ddFLN4 and ubiquitin.

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The variance of the unfolding time as obtained in the Smoluchowski approach is given in the Letter, for a general free energy $G(\zeta) = G_0(\zeta) - F\zeta$. The integrals in this expression can be evaluated analytically if the free energy is assumed to be linear in the interval $[\zeta_0, \zeta_s]$, $G(\zeta) = (F_c - F)(\zeta - \zeta_0)$. At external force $F$, the variance is then given by

$$V_L = \frac{2ak_B T (M\gamma)^2}{(F - F_c)^3} \left(1 + 2e^{-\beta(F-F_c)a}\right) + \frac{(M\gamma k_B T)^2}{(F - F_c)^4} \left(e^{-2\beta(F-F_c)a} + 4e^{-\beta(F-F_c)a} - 5\right)$$

where the symbols are as in the Letter. This result extends the one reported in equation (2.2.31) of Ref. [1], for a particle diffusing in one dimension in the presence of a bias ($F$ in our context), to the case where a barrier (represented by the term $F_c a$) is present also at zero bias.

**Strong force limit**

The expressions for the average unfolding time and the associated variance, in the linear approximation, can be simplified for strong forces [2],

$$\tau_L \rightarrow \tau_{SF} = \frac{M\gamma a}{F - F_c} \quad V_L \rightarrow V_{SF} = \frac{2k_B T (M\gamma)^2 a}{(F - F_c)^3}$$

Using these expressions, it is, in principle, possible to extract the critical force ($F_c$) and the distance to the transition state ($a$), and thereby also the zero-force barrier ($\Delta G_0 = F_c \times a$), from escape-time measurements at strong forces.

**Description of the experimental data for ddFLN4**

Fig. 4 in the Letter shows unfolding times at different forces for the ddFLN4 protein. The data have been extracted from Fig. 2b in Ref. [3]. The results in Ref. [3] were based on constant-velocity AFM pulling experiments by Ref. [4]. Rupture-force histograms obtained by Ref. [4] were transformed in Ref. [3] into force-dependent unfolding times measurable at constant force. The symbols in Fig. 4 are as in Ref. [3] and correspond to different pulling velocities in the original experiments: $v = 200\text{ nm/s}$ (blue squares), $v = 400\text{ nm/s}$ (green triangles); $v = 2,000\text{ nm/s}$ (yellow diamonds), and $v = 4,000\text{ nm/s}$ (red stars).

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