

A "slow" protein folds quickly in the end

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Until just a few years ago, it was not clear whether it would be possible to fold proteins using all-atom molecular dynamics simulations with explicit solvent molecules, despite the insights that these simulations had yielded into other biological problems. This was not just because of the computational challenge of reaching folding time scales, typically microseconds to seconds or longer, but also because it was not generally accepted that the empirical energy functions ("force fields") used were sufficiently accurate to locate the folded state as a global free-energy minimum. This has changed dramatically in the last 2 y, with the development by Shaw and coworkers of a special-purpose supercomputer, Anton, capable of running biomolecular MD simulations on a microsecond or even millisecond timescale (1). This group showed that, with only minor adjustments to existing force fields (2–4), it was possible to fold 12 small, "fast-folding" proteins (1), which adopt their native structure in microseconds. At the time, it was still not clear, however, whether it would be possible to do the same for larger, slower-folding proteins (5, 6). In PNAS, Piana et al. report simulations of ubiquitin folding, which occurs in milliseconds (7). Their results not only extend the computationally accessible time scale for calculating equilibrium folding trajectories by 2–3 orders of magnitude but have a number of implications that can only be deduced from a comparison of fast and slow folding proteins.

Protein folding is now often described in terms of an energy landscape primarily determined by the formation of contacts present in the folded state (8), the key assumption underlying a number of theoretical models and certain coarse-grained simulations (9). For proteins to fold in a reasonable time, there should not be significant formation of stable nonnative structures that introduce "roughness" in the energy landscape. Although the results of the earlier study of 12 small fast-folding proteins were consistent with the above picture (4), the greater complexity of ubiquitin folding presents a potentially stronger test. In fact, Piana et al. do find that the folding can be captured reasonably

well by a 1D reaction coordinate based on the formation of native contacts, which is one of the expectations for a "funneled" energy landscape. Furthermore, there is a strong correlation between the enthalpy of the system and the fraction of native contacts formed as the native state is approached. Interestingly, however, a number of stable nonnative or misfolded structures, persisting for microseconds, are also populated by ubiquitin, including a near-native conformation with Cα root mean square distance (RMSD) of

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less than 2 Å from the native state (7), indicating the hazards of using RMSD as a metric for comparing structures. However, it appears that transient visits to these traps are sufficiently short-lived that the folding still appears approximately twostate on a millisecond time scale—although it may be possible to detect transiently populated species by sufficiently highresolution experiments.

The finding that a simple coordinate based on native contacts is sufficient to capture much of the folding mechanism bodes well for the effectiveness of enhanced sampling methods based on reaction coordinates to reduce the cost of folding simulations (10). More generally, the results of this work, and the earlier all-atom folding simulations of fastfolding proteins (4), serve as an invaluable benchmark for developing novel methods for computational studies of protein folding (and other biomolecular processes) using enhanced sampling methods. Until the advent of Anton, the only way to fold a protein on a computer with all-atom simulations was to use some form of enhanced sampling. For example, in a distributed computing approach, many very short trajectories are run on a large number of computers (11–13), in some cases, using crowd-sourced computing

resources (13). These short trajectories can be used to study folding by assembling them using an appropriate theoretical framework such as a master equation or Markov state model. Alternative enhanced sampling methods include replica-exchange molecular dynamics (14, 15) and path-based methods that focus on sampling "reactive" barriercrossing events, which may potentially be much shorter than the waiting times in the unfolded or folded free-energy minima (10, 16, 17). The public availability of very long equilibrium folding simulations allows other methods to be tested against realistic and challenging problems, rather than using the relatively simple model systems often used for this purpose.

Experimental Benchmarks Of course, it is also essential to test, as far as possible, the results of the simulations against experiment. As a favorite protein of biophysicists, ubiquitin is an excellent choice because of the wealth of experimental data available for comparison, and the agreement with experiment is generally very good. For example, in the unfolded state, Piana et al. find that the N-terminal hairpin is ∼60% formed, which is consistent with the known stability of this hairpin (18). An intermediate identified just after the transition state appears to be consistent with one inferred from Tjump infrared kinetics experiments (19). Particularly convincing is the comparison with Φ values, derived from kinetic and stability data of single-point mutants, which provide information about structure formation in the folding transition state (20). Piana et al. compute Φ values by approximating the effect of mutations on a 1D freeenergy surface. The agreement with two independent sets of experimental data (21, 22) is excellent. However, the simulations also help to interpret the data: although the simulation transition state is structured in the regions suggested by the regions of high Φ values, other low-Φ-value regions, such as the fourth β-strand, are also forming nativelike secondary structure while still lacking many of their native contacts.

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A topic of much recent experimental and theoretical interest is touched on at the end of the paper (7). The authors compare the duration of barrier-crossing events, or "transition paths," obtained from their ubiquitin simulations with transition-path durations computed from fast-folding proteins. Remarkably, the average transition-path duration for ubiquitin is only ∼1.7 μs, very similar to that for similar-sized proteins using the same energy function: ∼0.9 μs for the 73-residue–designed helical protein α3D, and 3.1 μs for the 80-residue λ repressor (4). However, the folding time, which is the total time the protein takes to fold, not just to cross the barrier, is very different for these proteins: folding times for α3D and λ repressor in the simulations are two orders of magnitude shorter than for ubiquitin, 31 and 13 μs, respectively (4). These findings are consistent with recent estimates of the transition-path time for the WW domain and protein G obtained from single-molecule fluorescence experiments: both proteins had transition-path times of 2–10 μs but folding times differing by four orders of magnitude (23). The lack of correlation between folding time and transitionpath time may be explained by a theoretical model for diffusion over a 1D free-energy barrier that predicts the transition-path time (in contrast to the folding time) to depend only very weakly on the barrier height (24).

The larger separation between transitionpath time and folding time for ubiquitin has implications for studying the folding of other proteins by molecular dynamics. Although in the present context, the millisecond folder ubiquitin appears "slow," it is worth bearing in mind that ubiquitin is still faster than many other single-domain proteins (25). Clearly, sampling folding events for such slow folders by long equilibrium molecular dynamics simulations is still some way in the future. However, the separation of the transition-path time and folding time suggests that, in these cases, methods based on enhanced sampling of transition paths (10, 16) might be profitably used in conjunction with recent advances in computing power, exemplified by Anton to extend further the folding time scale accessible to atomistic molecular dynamics simulations.

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