




National Institutes of.....

Advantages (+) and disadvantages (-) of working with the nine types of principal investigators

<h3>Big Talker</h3> <p>These results have clear implications for curing cancer in our lifetime</p>  <ul style="list-style-type: none">(+) Makes your data seem really important(-) Doesn't really understand what you do	<h3>Slave Driver</h3> <p>You know, 60 hours a week just isn't go to cut it in this lab</p>  <ul style="list-style-type: none">(+) You get lots done(-) You forget your spouse's name	<h3>Demi God</h3>  <ul style="list-style-type: none">(+) Power, prestige, better job prospects(-) You never see them
<h3>Control Freak</h3> <p>Why didn't you use 25 mM NaCl in the second wash?</p>  <ul style="list-style-type: none">(+) Knows exactly what experiment you are doing(-) Knows exactly what experiment you are doing	<h3>Science Wonk</h3> <p>Why don't you try this new reverse gyoprismatic amplifying DCR technique?</p>  <ul style="list-style-type: none">(+) Knows everything about science(-) He's a total geek	<h3>Laid-Back</h3> <p>Make it quick. I've got a 2:00 tee-time</p>  <ul style="list-style-type: none">(+) Leaves you alone(-) Doesn't care about your results
<h3>Psycho</h3> <p>WHAT DO YOU MEAN YOU MADE A MISTAKE!?</p>  <ul style="list-style-type: none">(+) Keeps you on your toes(-) Scary	<h3>Small Town Grocer</h3>  <ul style="list-style-type: none">(+) Happy with his own little niche(-) Little ambition	<h3>Rising Star</h3>  <ul style="list-style-type: none">(+) Exciting ride.(-) Not much room for you

KITP
June 7, 2011

“Watching individual protein molecules fold and unfold
using fluorescence spectroscopy”
(Progress toward observing transition paths)

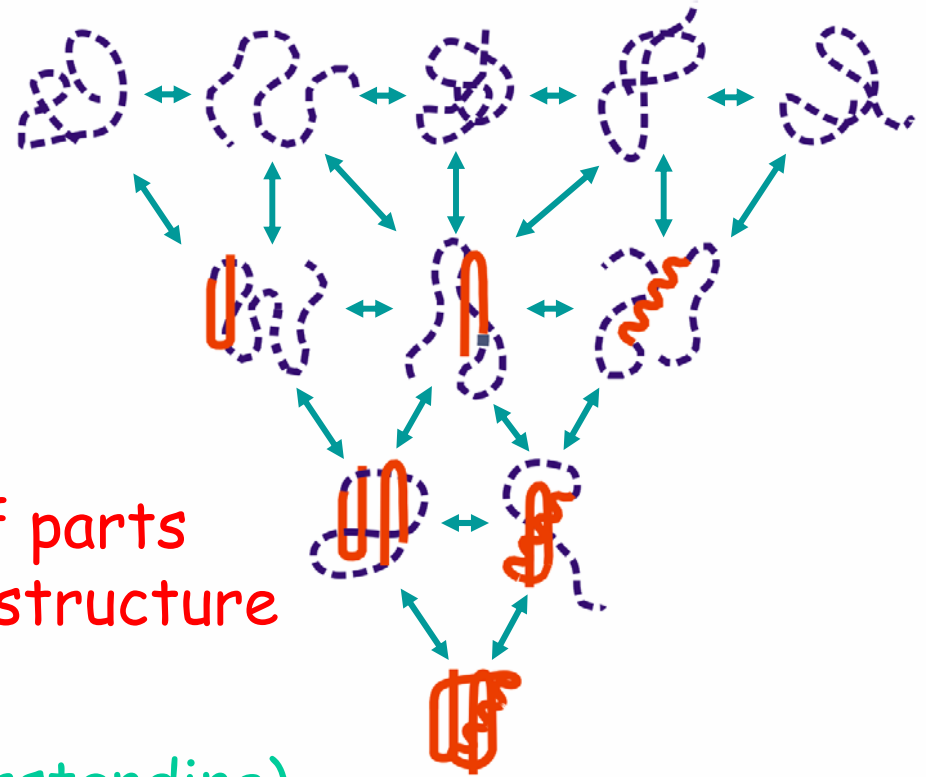
William A. Eaton.
Laboratory of Chemical Physics
NIDDK, National Institutes of Health (NIH)
Bethesda, Maryland

The Question: *How* does a protein fold?

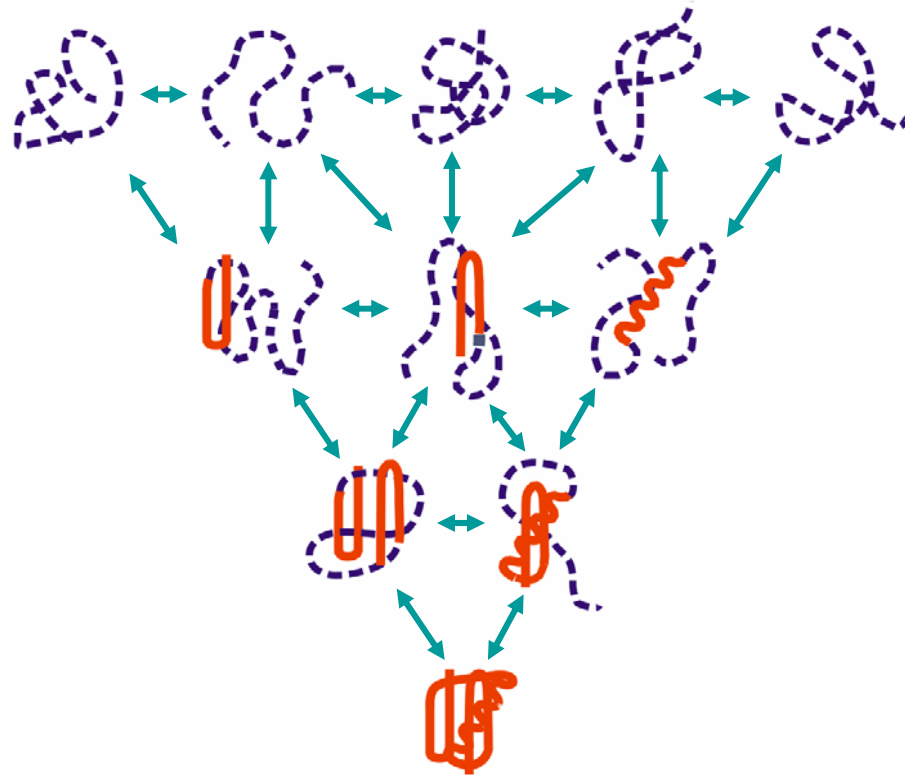
Biologist:
spontaneously!
(Anfinsen Nobel Prize)

Biochemist:
picture of order of assembly of parts
correlation between order and structure

Biophysicist (quantitative understanding)
theoretical model (i.e. partition function,
master equation) - quantitative predictions
of experimental measurements.
universal principles



Folding is heterogeneous with many microscopic pathways connecting the folded and unfolded states



The distribution of microscopic pathways is predicted by theory and simulations, what about experiments?

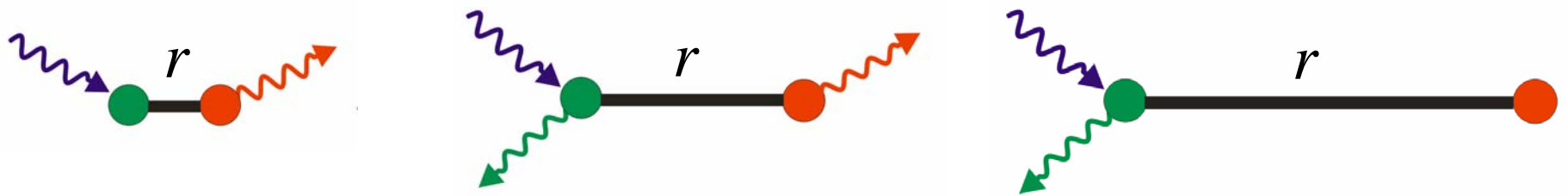
SINGLE MOLECULES!

Motivations for current experiments

Determining folding heterogeneity requires observing transition paths (**a uniquely single molecule property**), which is challenging and not yet observed for any system.

For an experimentalist that can be **fun!!**
(if it works)

Förster resonance energy transfer (FRET)

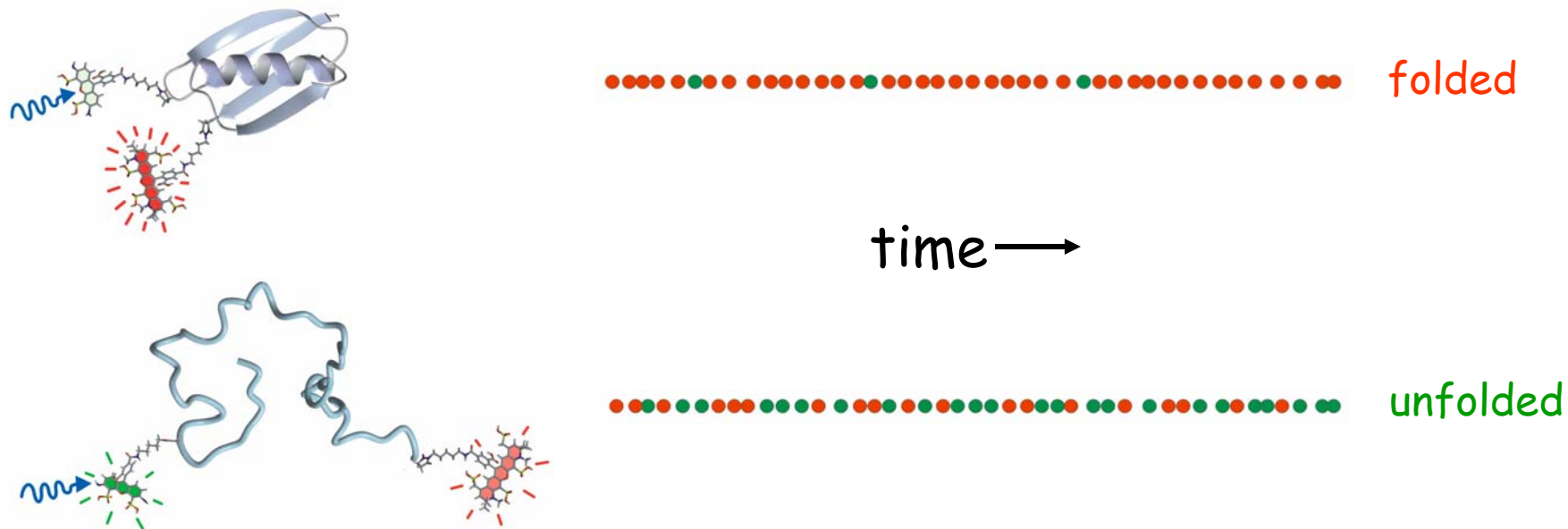


$$k_T = \frac{1}{\tau_D} \left(\frac{R_0}{r} \right)^6$$

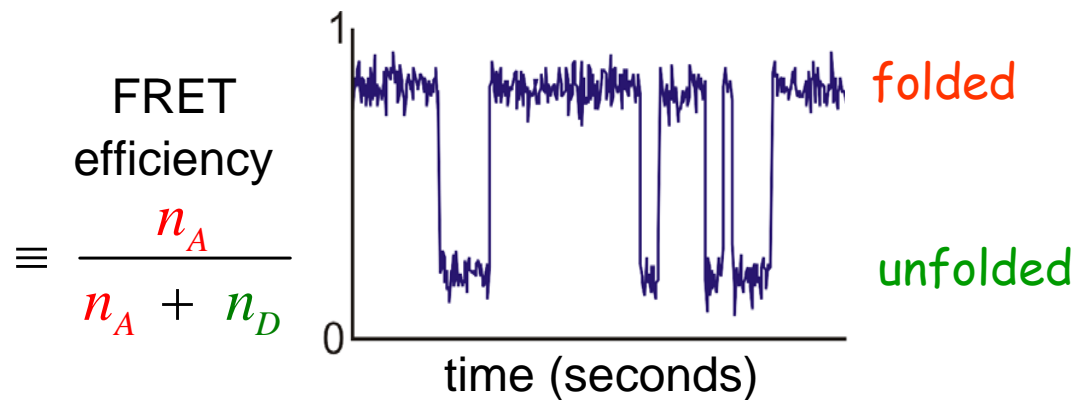
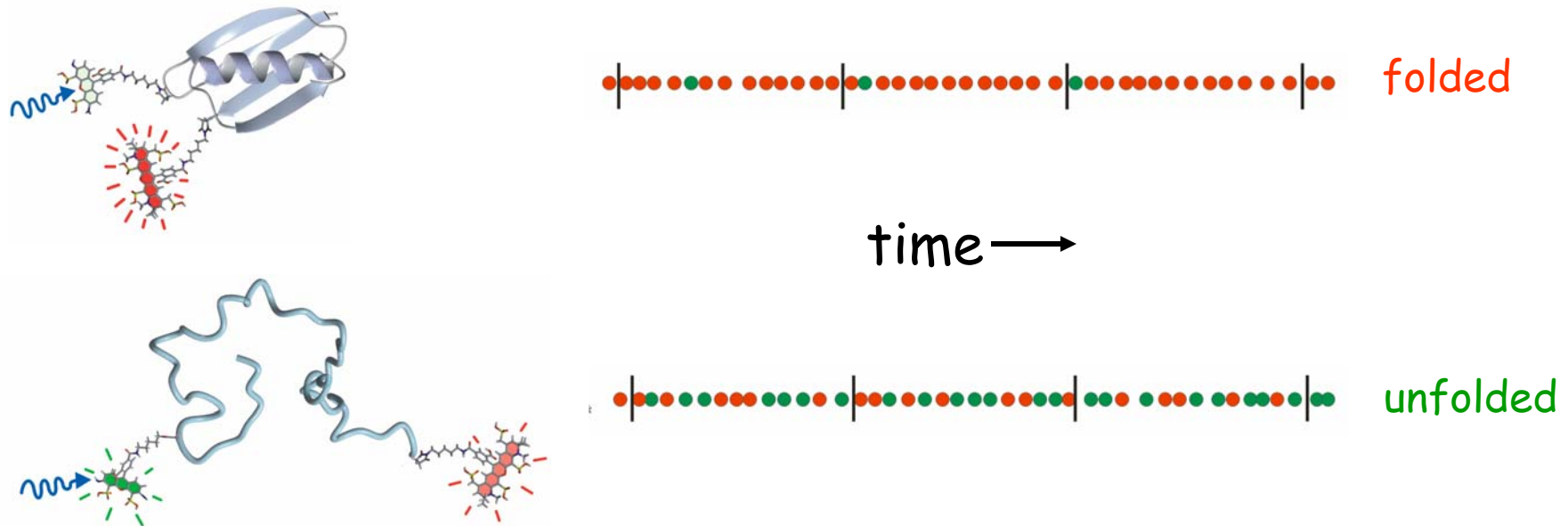
$$\text{FRET efficiency} = \frac{k_T}{k_T + 1/\tau_D} = \frac{1}{1 + (r/R_0)^6}$$

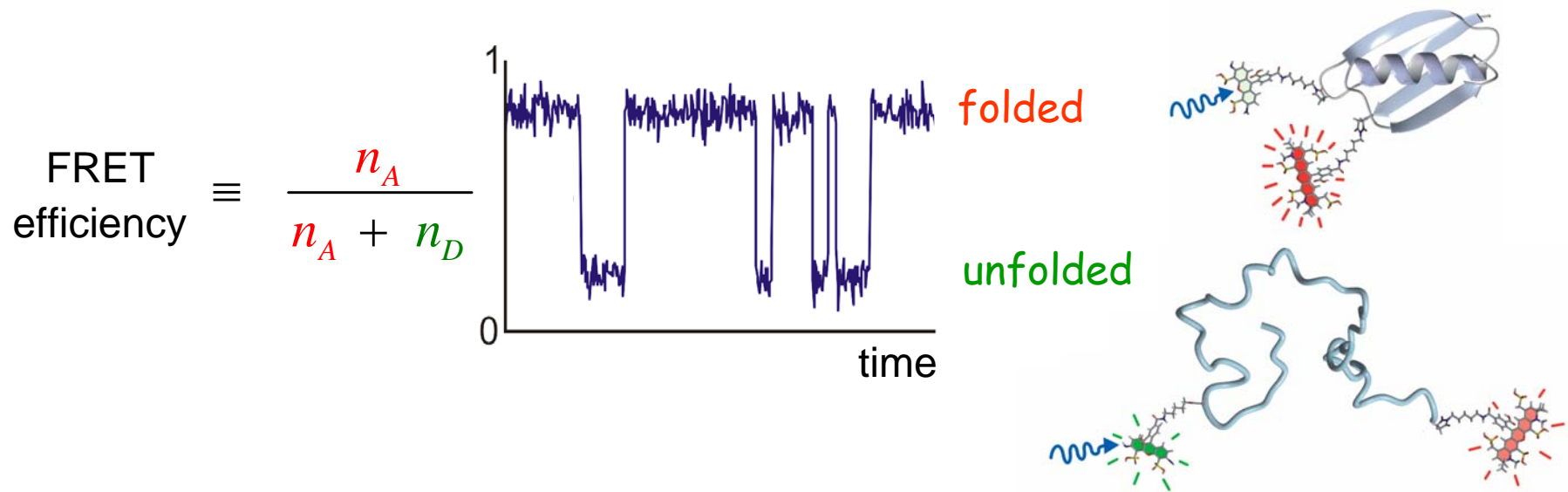
$$= \frac{n_A}{n_A + n_D}$$

Photon trajectory for "two-state" protein



FRET efficiency trajectory for "two-state" protein





All mechanistic information contained in the “jumps”:

The transition path: a property unique to single molecules

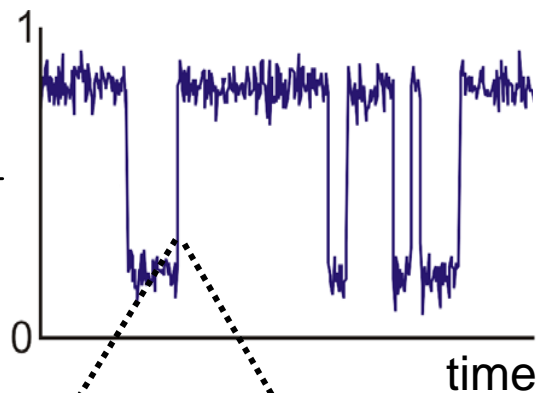
Goal for single molecule experiments: observe transition paths

Transition paths can be obtained from MD simulations for ultrafast ($< 100 \mu\text{s}$) folders, or from clever theoretical methods for slower folding proteins (e.g. H. Orland), but have never been observed *experimentally* for any system

MAJOR EXPERIMENTAL CHALLENGE

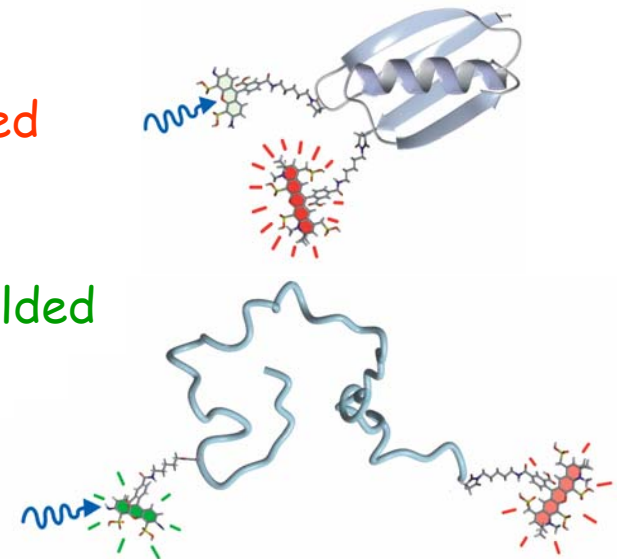
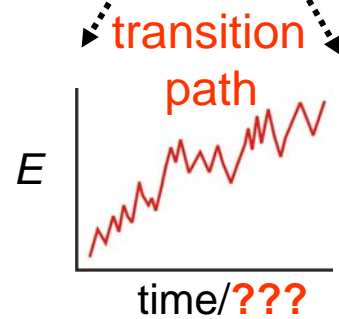
FRET efficiency \equiv

$$\frac{n_A}{n_A + n_D}$$



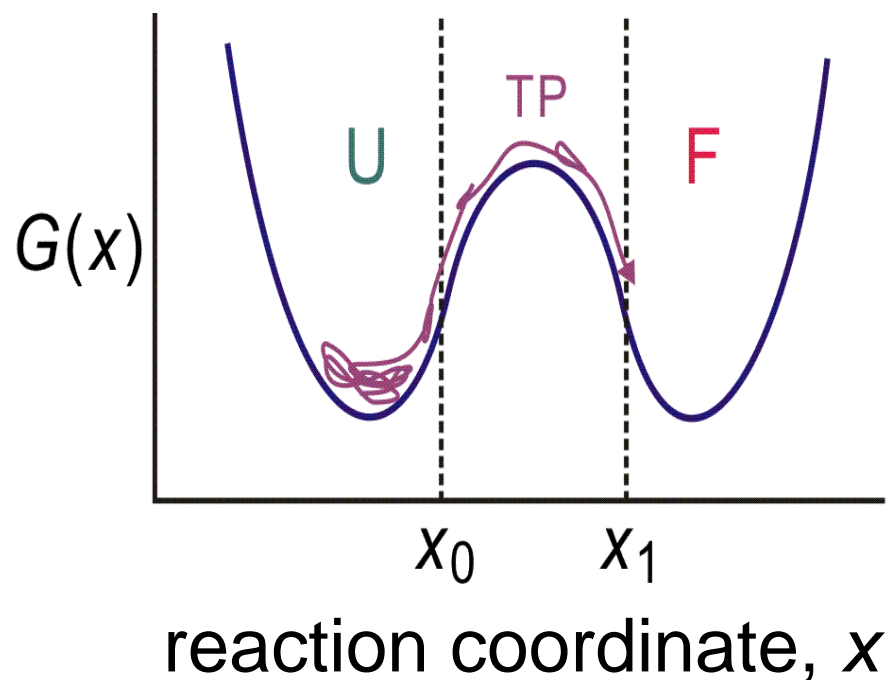
folded

unfolded



Zeroth order question about transition paths:
what is duration?

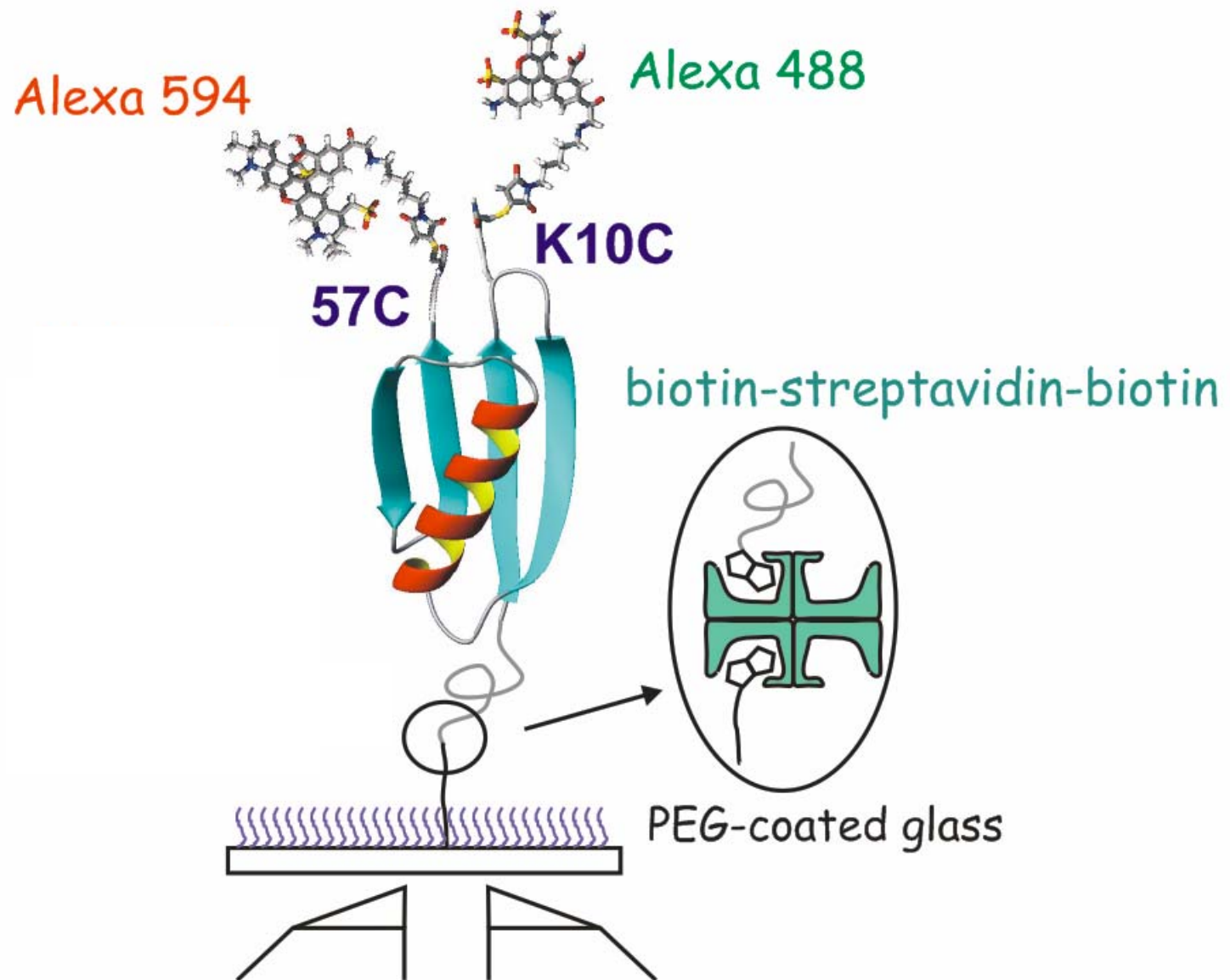
Wolynes Energy Landscape Theory (Socci, Onuchic, and Wolynes, JCP 1996)



Definition of transition path time (TPT):

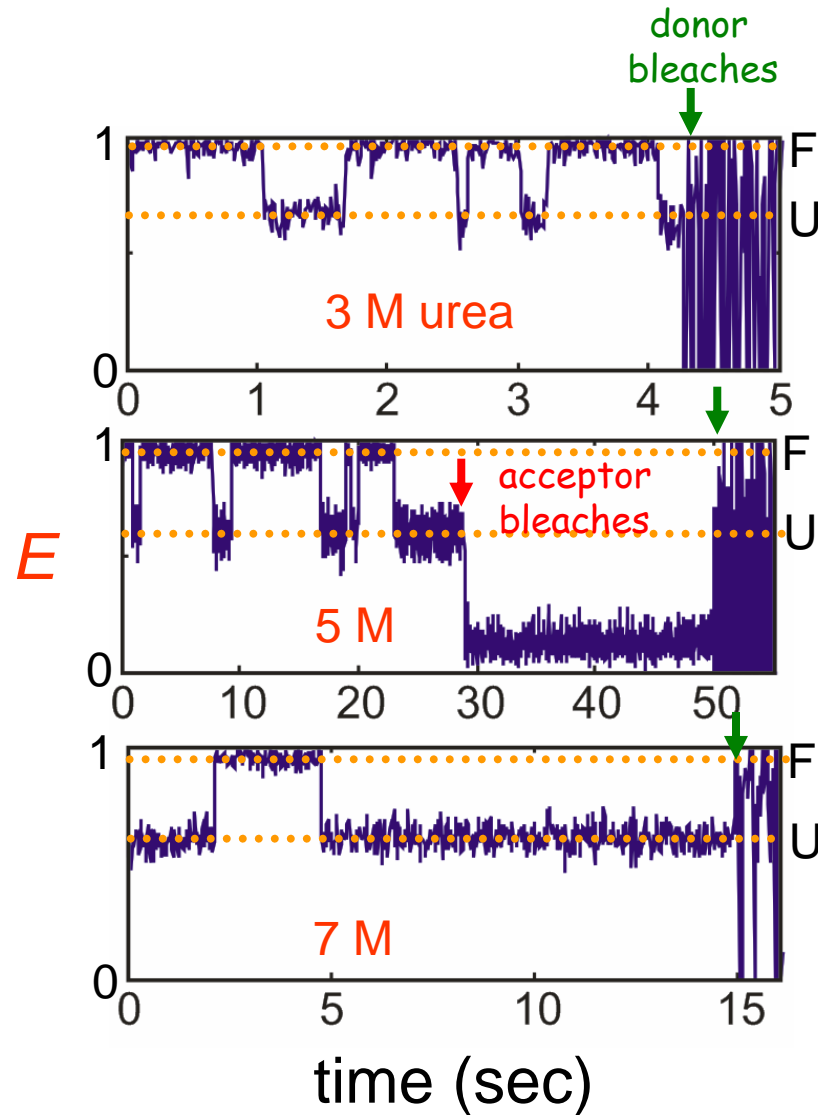
Trajectories that cross x_0 and reach x_1
without ever recrossing x_0

The well-studied two-state protein
56-residue protein G labeled with donor and acceptor dyes



Explained almost every one of 2,000 trajectories in detail

Representative (65%) folding/unfolding trajectories



Almost every trajectory explained in detail.

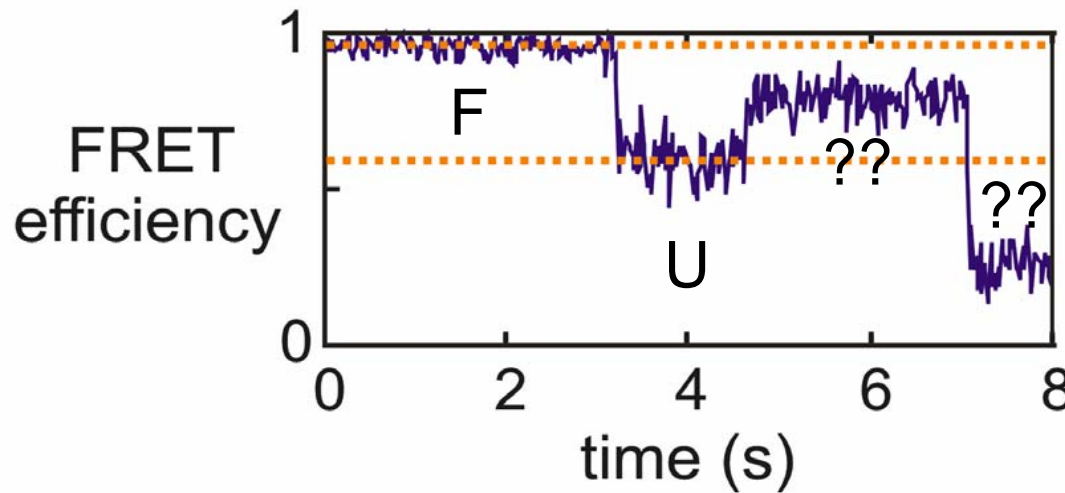
65% folding/unfolding

21% "blinking"

5% dye sticking to surface or linker

8% shifted donor dye spectrum

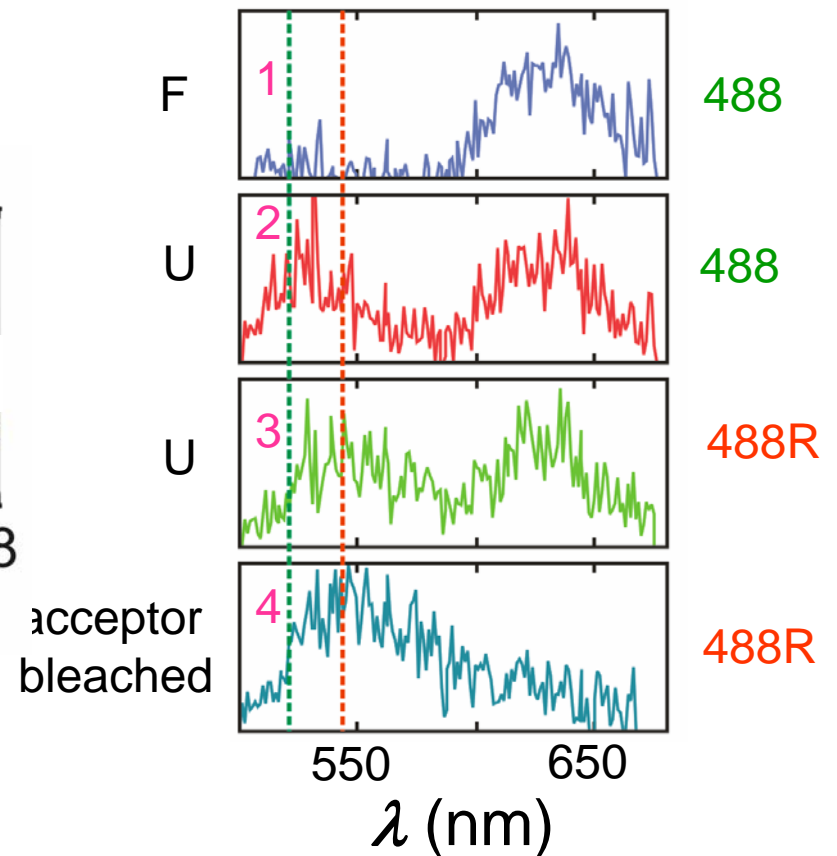
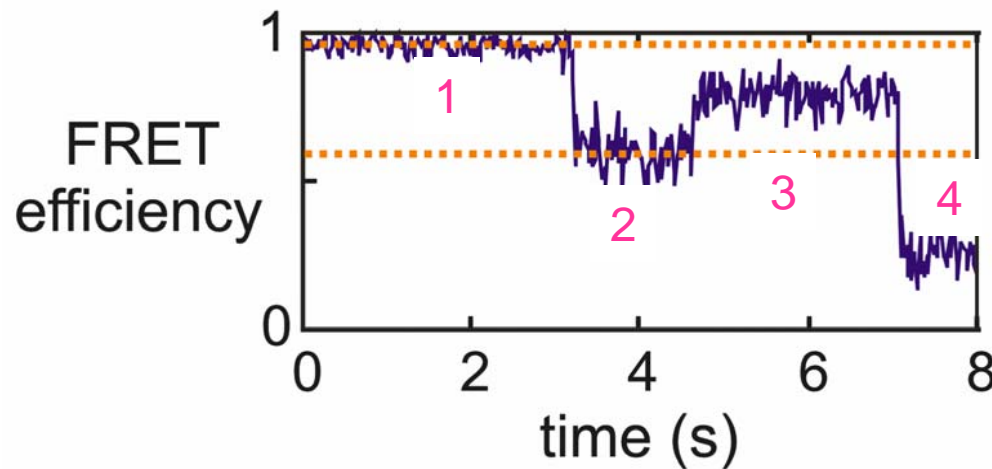
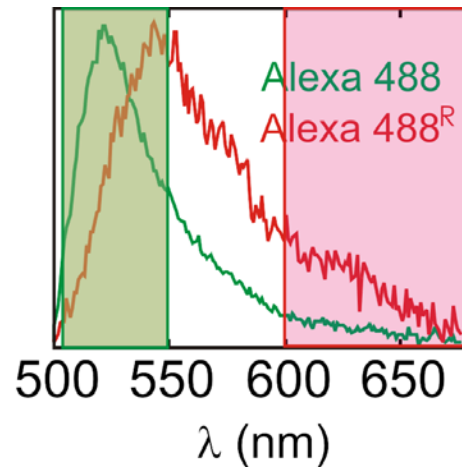
initially puzzling trajectory



folding intermediates????

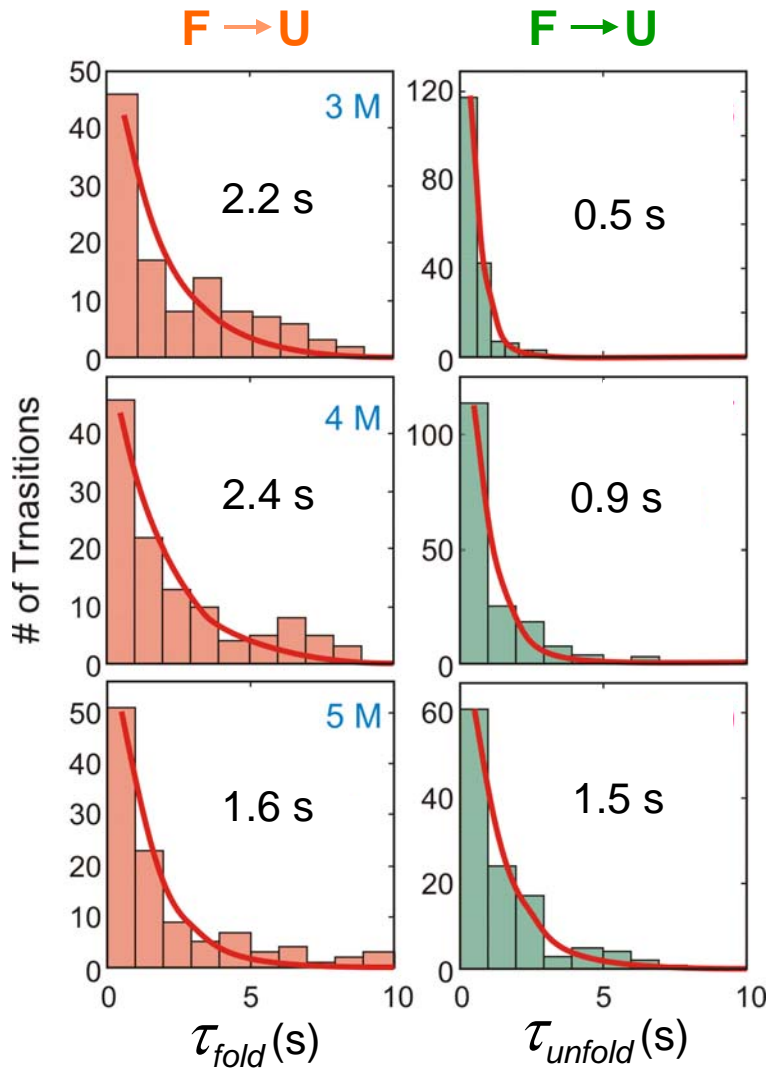
Almost every trajectory explained
in detail.

- 65% folding/unfolding
- 21% "blinking"
- 5% dye sticking to surface or linker
- 8% shifted donor dye spectrum

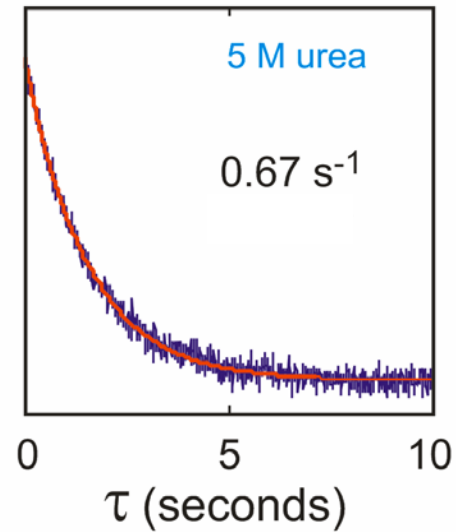


Comparison of Kinetics

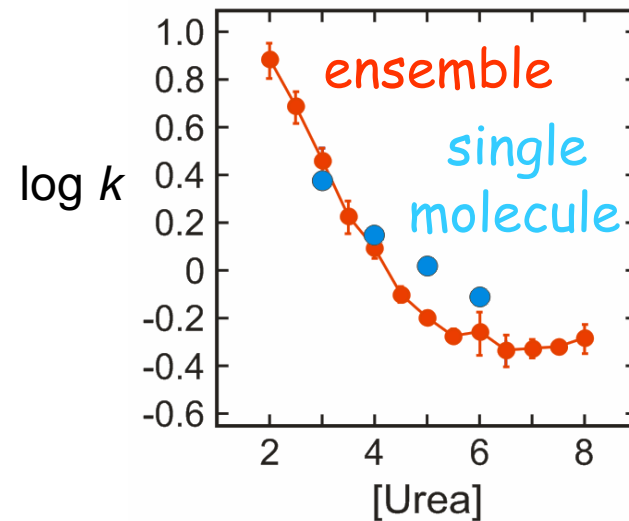
Waiting times are exponentially distributed



stopped-flow

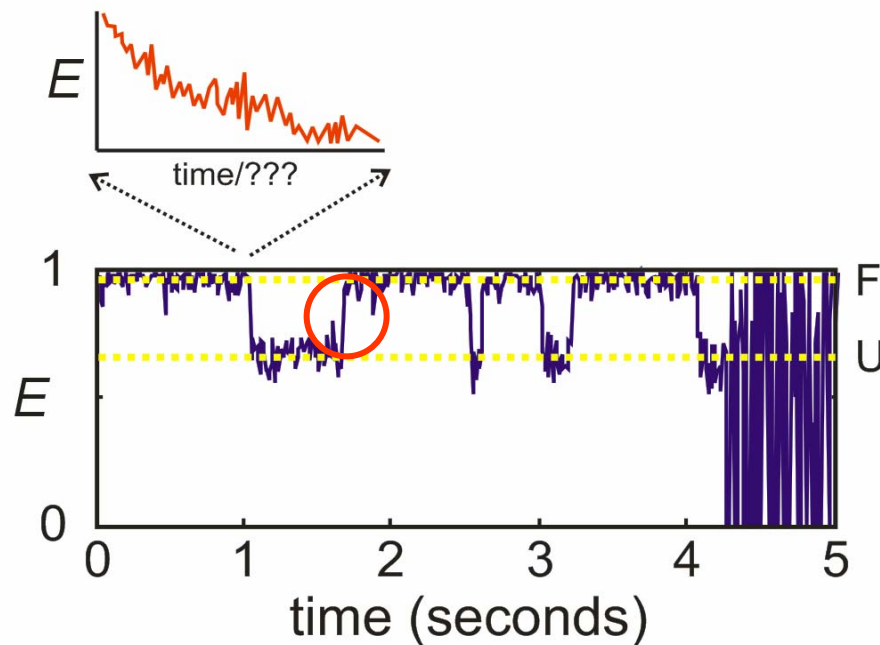


relaxation rate, $k = k(U \rightarrow F) + k(F \rightarrow U)$

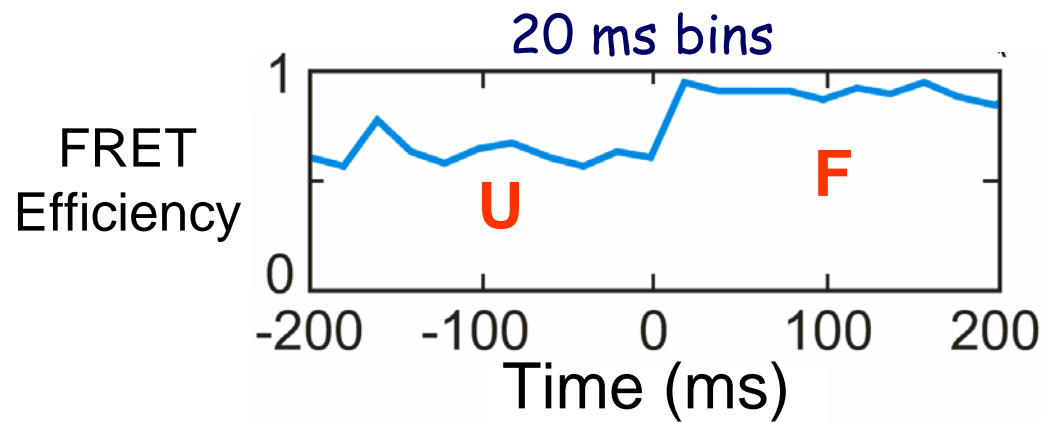


Small effect of immobilization

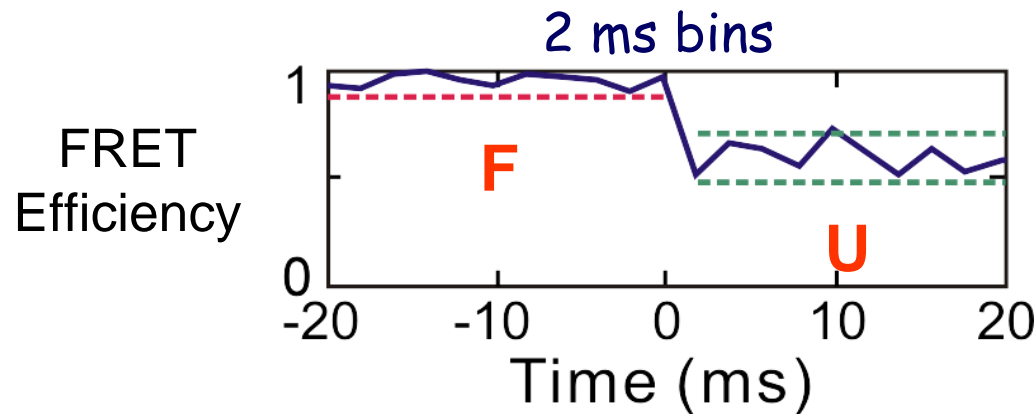
All mechanistic information contained
in "jumps" (transition paths)



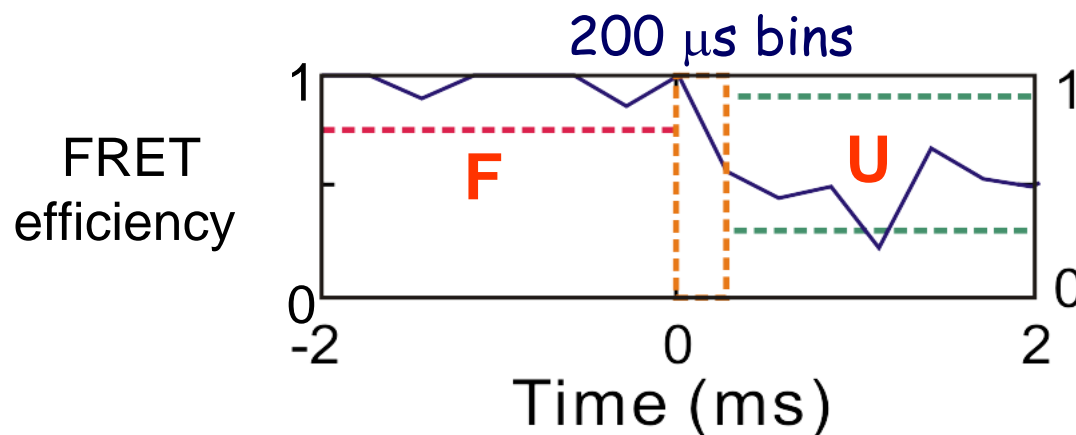
What can we say about transition path times?



5 photons/ms
 Bleach in ~ 10 s
 ~ 1000 transitions
 ~ 2000 trajectories



50 photons/ms
 Bleach in ~ 0.1 s
 46 transitions in
 1500 trajectories



Transition still in
 single bin, \therefore
 $TPT < 200 \mu$ s

TPT < 200 μs (>10,000x shorter than folding time)
(Chung, Louis, & Eaton, *PNAS* 2009)

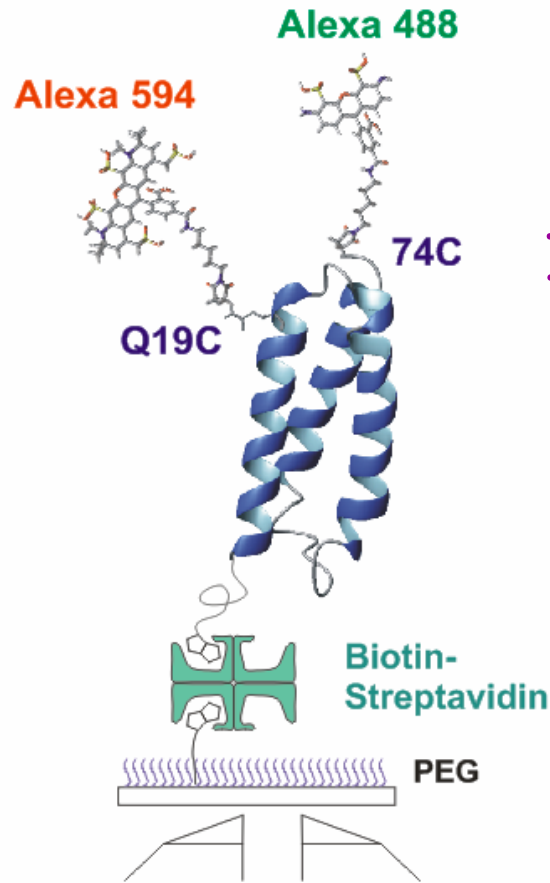
Can we measure a transition path time?

YES, but we will have to

more intense excitation to increase photons
(shortens bleaching time)

observe a very large number of transitions
(automate data acquisition, faster folding protein)

analyze these transitions *collectively*
(maximum likelihood method)

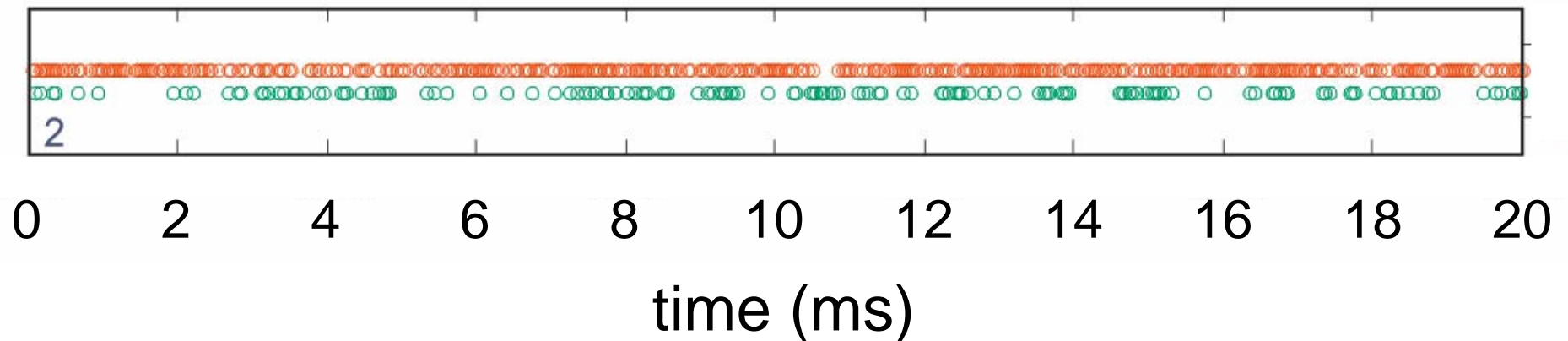


$\alpha_3\text{D}$, two-state folder, $\tau_F = 1 \text{ ms}$

Instead of 1 transition/2s for protein G,
1000 transitions/s for $\alpha_3\text{D}$

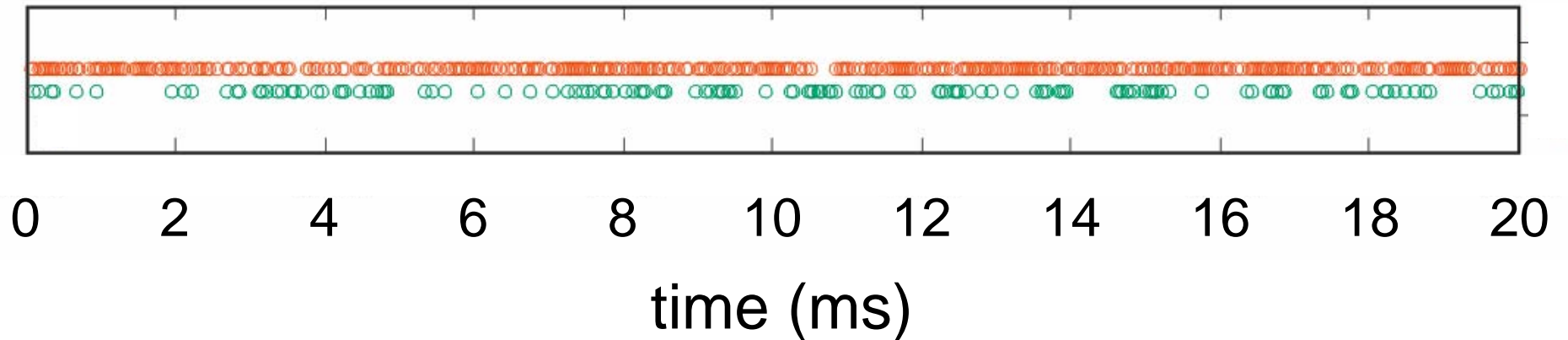
$\sim 1,000$ transitions at each denaturant
concentration, ~ 50 photons/ms,
 ~ 30 ms bleaching time

photon trajectory

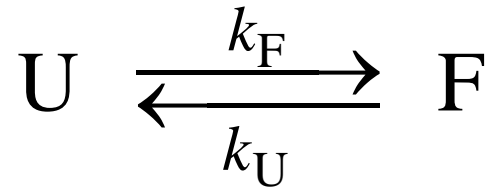


Gopich/Szabo Maximum Likelihood Function (J Phys Chem B, 2009)

Given a model, the method yields the most likely model parameters consistent with the photon trajectories



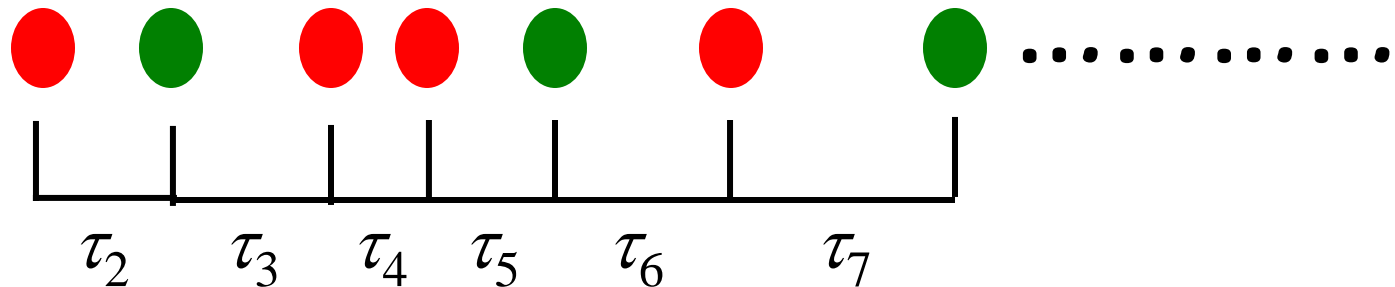
Gopich/Szabo Maximum Likelihood Function for Two-State Model

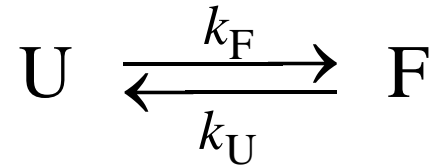


$$L_j = (1 \ 0) \prod_{k=2}^{N_j} \left[\Phi(c_k) \begin{pmatrix} 1 & 0 \\ 0 & e^{-(k_F + k_U)\tau_k} \end{pmatrix} \right] \Phi(c_1) \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

● $\Phi(\text{acceptor}) = \begin{pmatrix} \varepsilon_F p_F + \varepsilon_U p_U & (\varepsilon_U - \varepsilon_F) p_U \\ (\varepsilon_U - \varepsilon_F) p_F & \varepsilon_F p_U + \varepsilon_U p_F \end{pmatrix}$

● $\Phi(\text{donor}) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} - \Phi(\text{acceptor})$



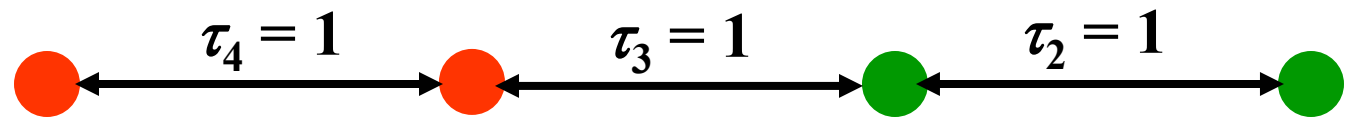


$$L_j = (1 \ 0) \prod_{k=2}^{N_j} \left[\Phi(c_k) \begin{pmatrix} 1 & 0 \\ 0 & e^{-(k_F + k_U)\tau_k} \end{pmatrix} \right] \Phi(c_1) \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

● $\Phi(\text{acceptor}) = \begin{pmatrix} \varepsilon_F p_F + \varepsilon_U p_U & (\varepsilon_U - \varepsilon_F) p_U \\ (\varepsilon_U - \varepsilon_F) p_F & \varepsilon_F p_U + \varepsilon_U p_F \end{pmatrix}$

● $\Phi(\text{donor}) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} - \Phi(\text{acceptor})$

$p_U = p_F = 1/2, \ \varepsilon_U = 0, \ \varepsilon_F = 1, \ \tau_k = 1, \ k_U = k_F = ??$

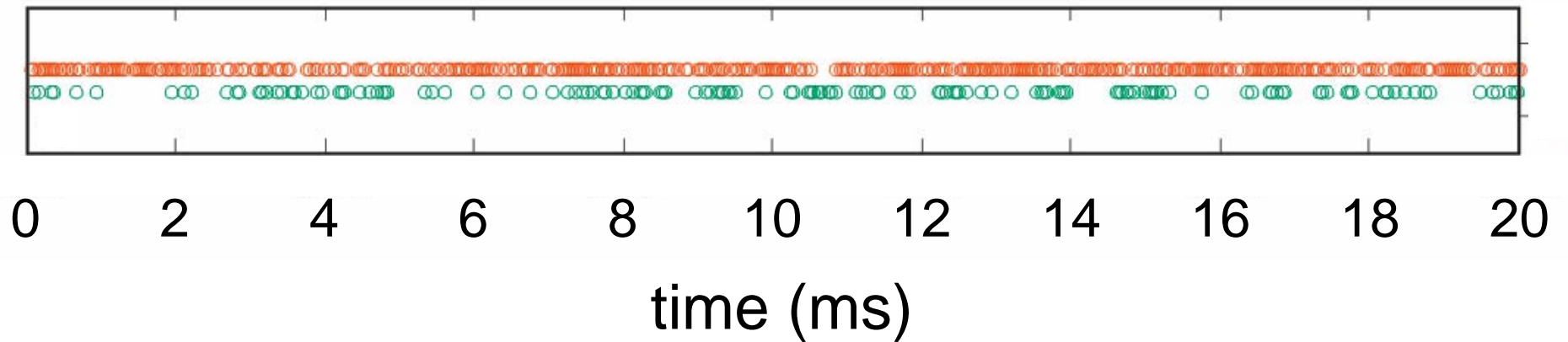


$$L(k_F = 0.5) = (1 \ 0) \left[\begin{pmatrix} 1/2 & -1/2 \\ -1/2 & 1/2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1/e \end{pmatrix} \right] \left[\begin{pmatrix} 1/2 & -1/2 \\ -1/2 & 1/2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1/e \end{pmatrix} \right] \left[\begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1/e \end{pmatrix} \right] \begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$= 0.074$

k_F	L
0.25	0.063
1	0.070

photon-by-photon analysis to obtain k_F , k_U , ε_F , ε_U

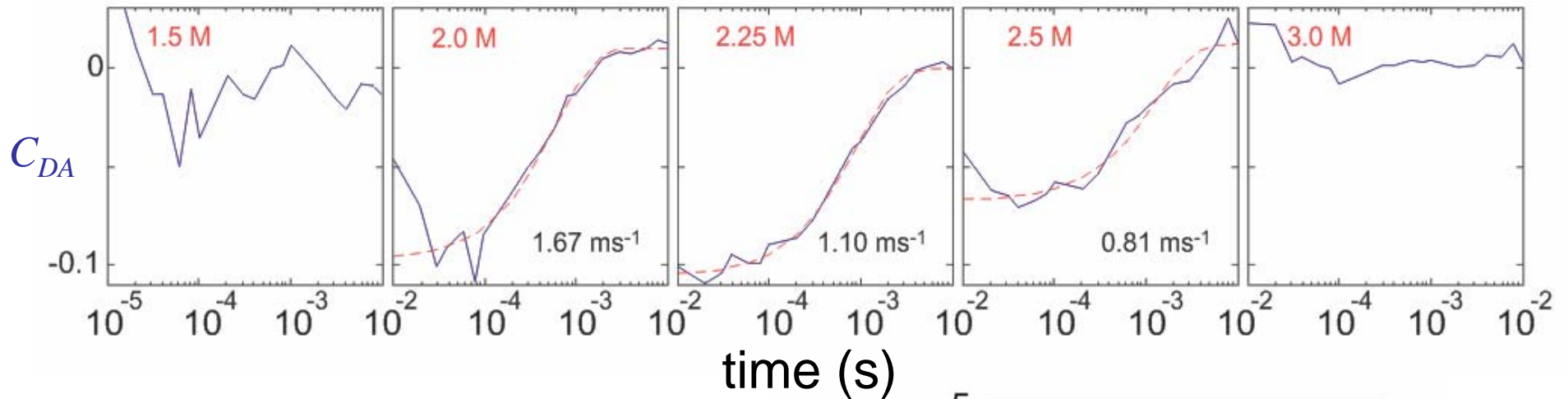


GdmCl (M)	ε_F	ε_U	k (ms ⁻¹)	p_F^-
1.5	0.94	0.64	2.64 (±0.23)	0.92
2	0.93	0.64	1.29 (±0.05)	0.67 (±0.01)
2.25	0.93	0.61	0.99 (±0.04)	0.43 (±0.01)
2.5	0.93	0.62	1.06 (±0.04)	0.32
3	0.92	0.56	1.22 (±0.08)	0.05

Are the results accurate?

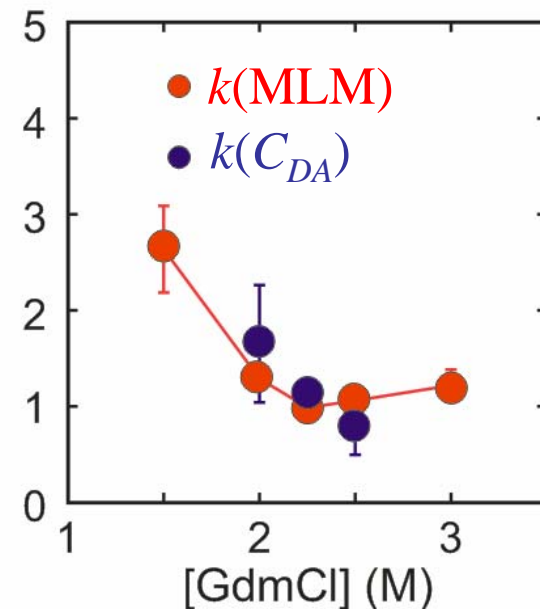
Compare with the donor-acceptor cross-correlation function decay

$$C_{DA}(\tau) = \frac{\langle n_D(t)n_A(t+\tau) \rangle}{\langle n_D(t) \rangle \langle n_A(t) \rangle} - 1$$



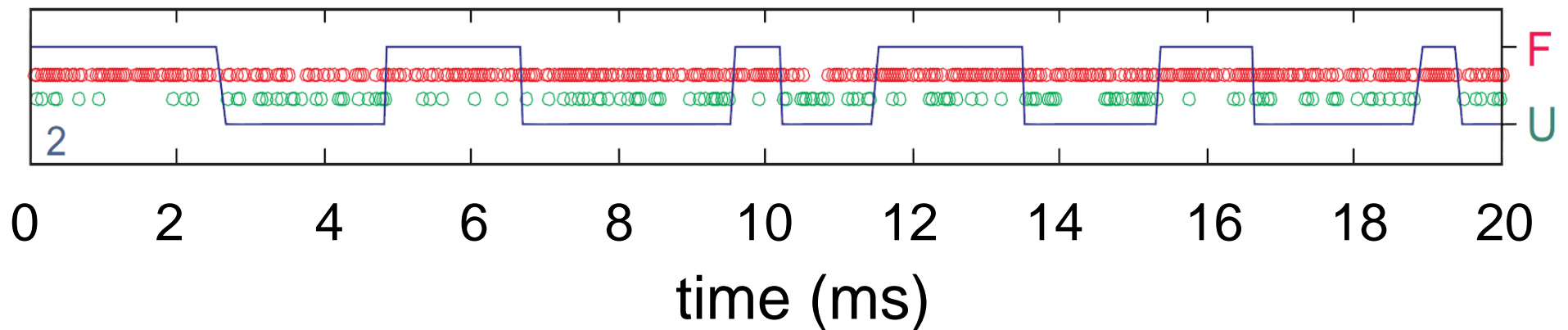
$$k(C_{DA}) = k(\text{MLM})$$

$$k(\text{MLM}) = k_F(\text{MLM}) + k_U(\text{MLM}) \quad k \quad (\text{ms}^{-1})$$



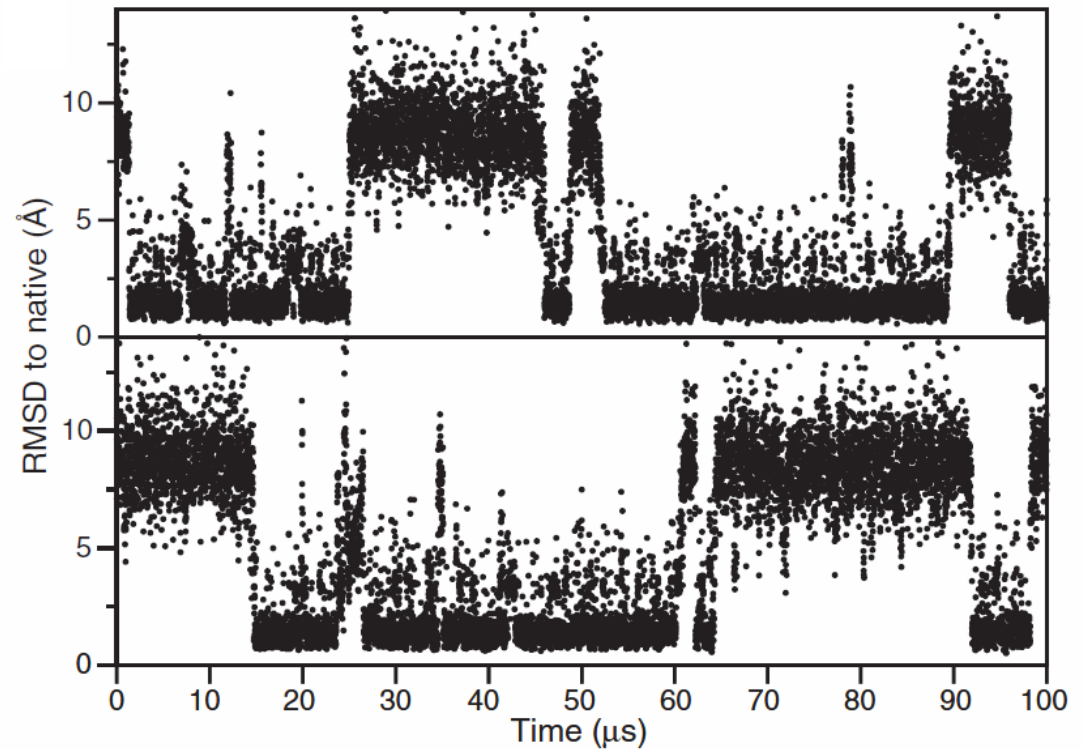
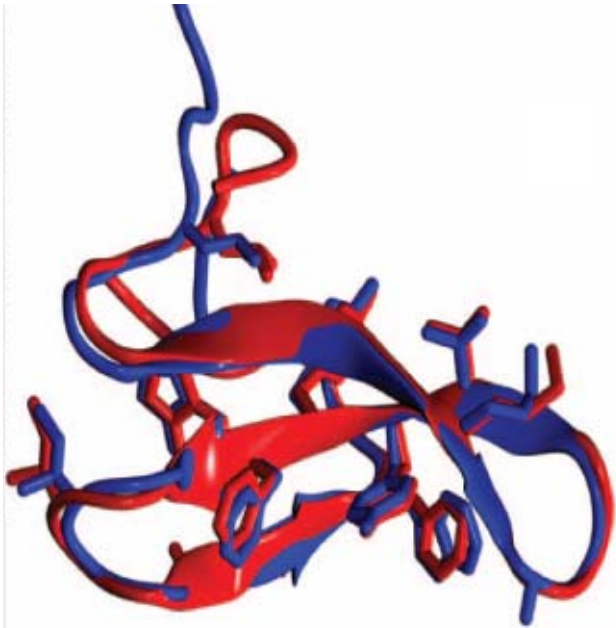
Strategy

Apply Viterbi algorithm to produce most probable state trajectory to locate transition region



and only analyze the trajectory in the vicinity of the transition

David E. Shaw et al.: *Science* October 15, 2010 issue,
fully atomistic MD simulation of ultrafast folder using
"Anton", hard-wired computer for MD calculations
- 35 residue WW domain



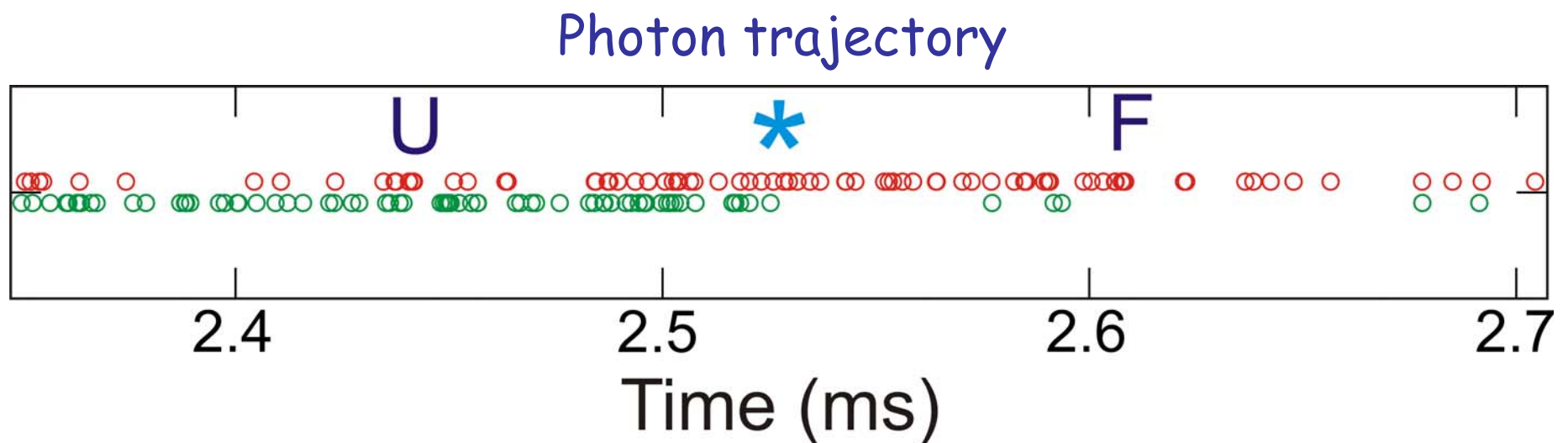
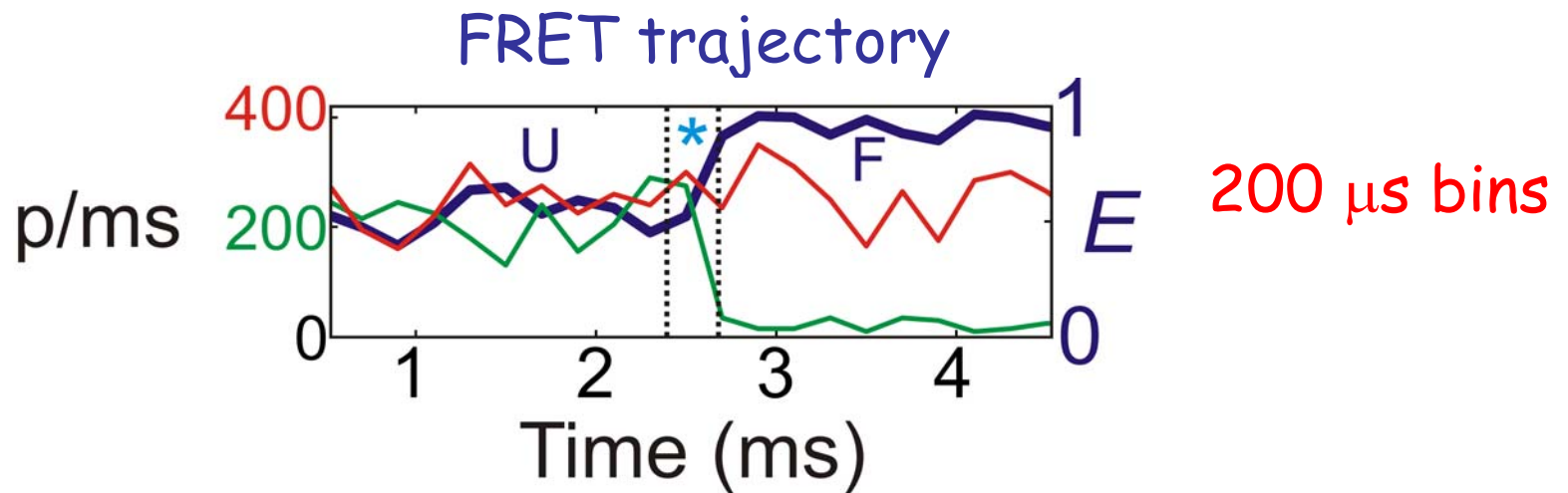
simulation: $\tau_F = 10 \pm 3 \mu s$, experiment: $\tau_F = 14 \pm 2 \mu s$
 $\langle TPT \rangle \approx 2 \mu s$ at $20^\circ C$ (after viscosity correction)

We now have a good estimate of the average transition path time for an ultrafast (10 microsecond)-folding protein

What can we say about our 2 second-folding protein *G* - inaccessible even to "Anton."

Hoi Sung got ambitious!

Measured 46,932 trajectories for protein G, $\tau_F = 2$ s
350 photons/ms; bleaching time ~ 10 ms
observed 151 transitions

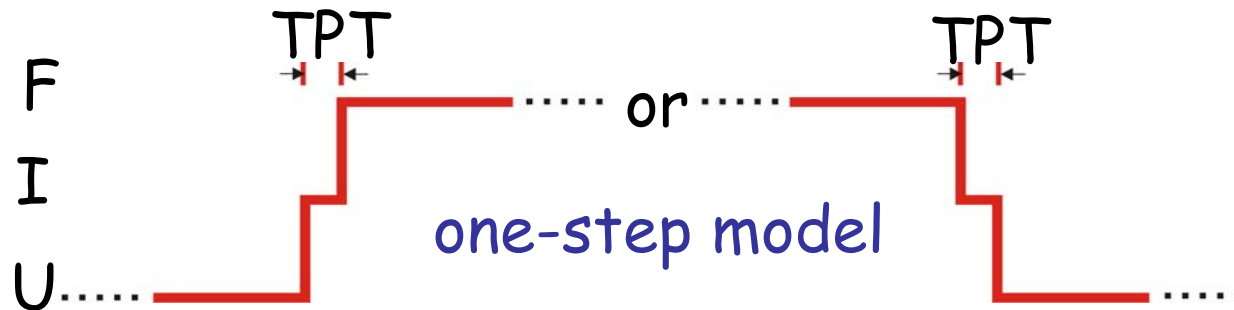


Simulate trajectories with model to answer question:

Given the current data (number of transitions, count rate, FRET efficiencies,), how short a transition path time we can expect to be able to determine with the current data ?

151 transitions, 350 photons/ms, $\epsilon_F = 0.95$ $\epsilon_U = 0.60$

Simulations of trajectories for one-step model:



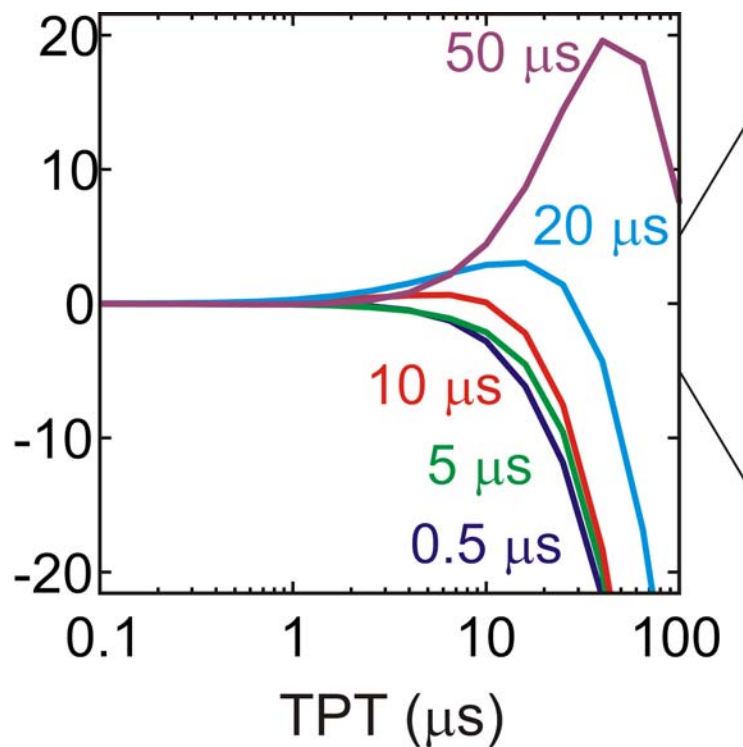
$$\langle TPT \rangle = 1/k_{I \rightarrow F} = 1/k_{I \rightarrow U} \quad \varepsilon_I = \frac{\varepsilon_U + \varepsilon_F}{2}$$

$$\mathbf{K} = \begin{pmatrix} -k_{U \rightarrow I} & k_{I \rightarrow U} & 0 \\ k_{U \rightarrow I} & -(k_{I \rightarrow F} + k_{I \rightarrow U}) & k_{F \rightarrow I} \\ 0 & k_{I \rightarrow F} & -k_{F \rightarrow I} \end{pmatrix}$$

Analyse simulated trajectories using Gopich/Szabo function with and without transition path, and compare likelihoods

(TPT is only variable parameter in likelihood function)

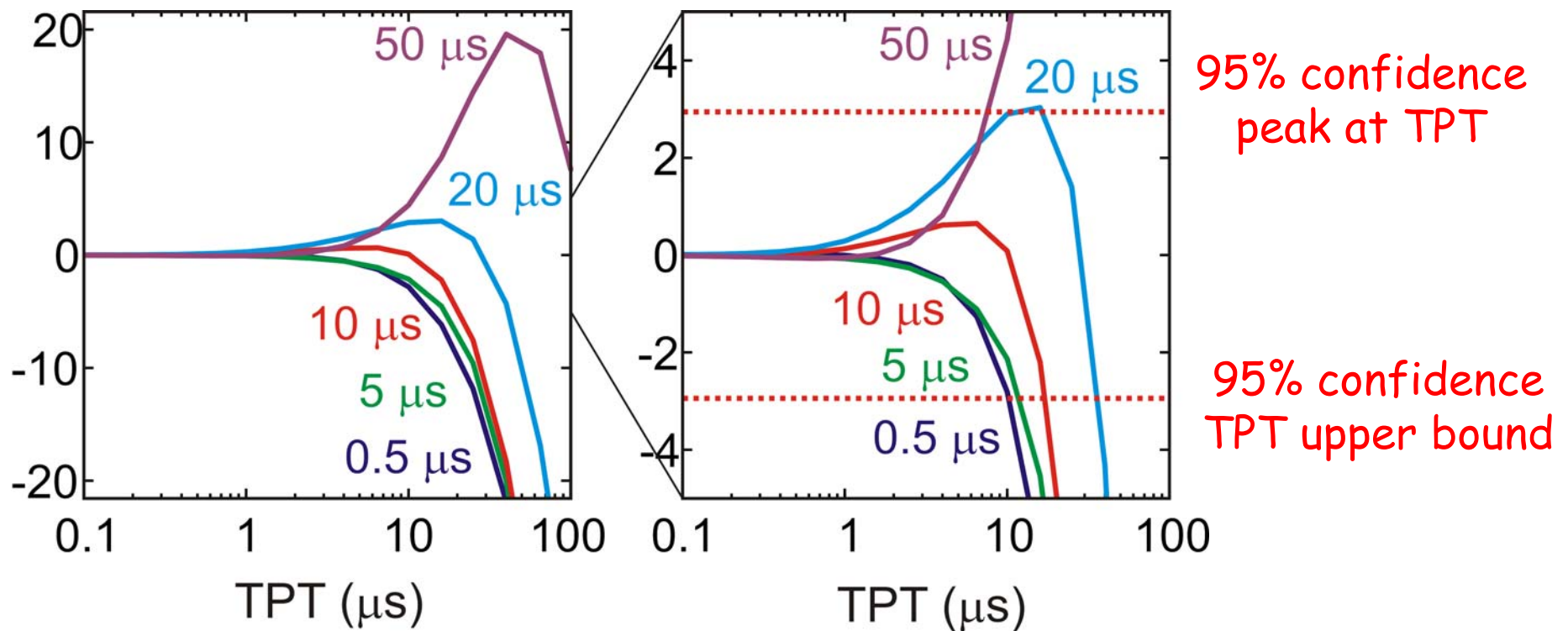
$\ln L(\text{TPT}) - \ln L(0)$



Analyse simulated trajectories using Gopich/Szabo function with and without transition path, and compare likelihoods

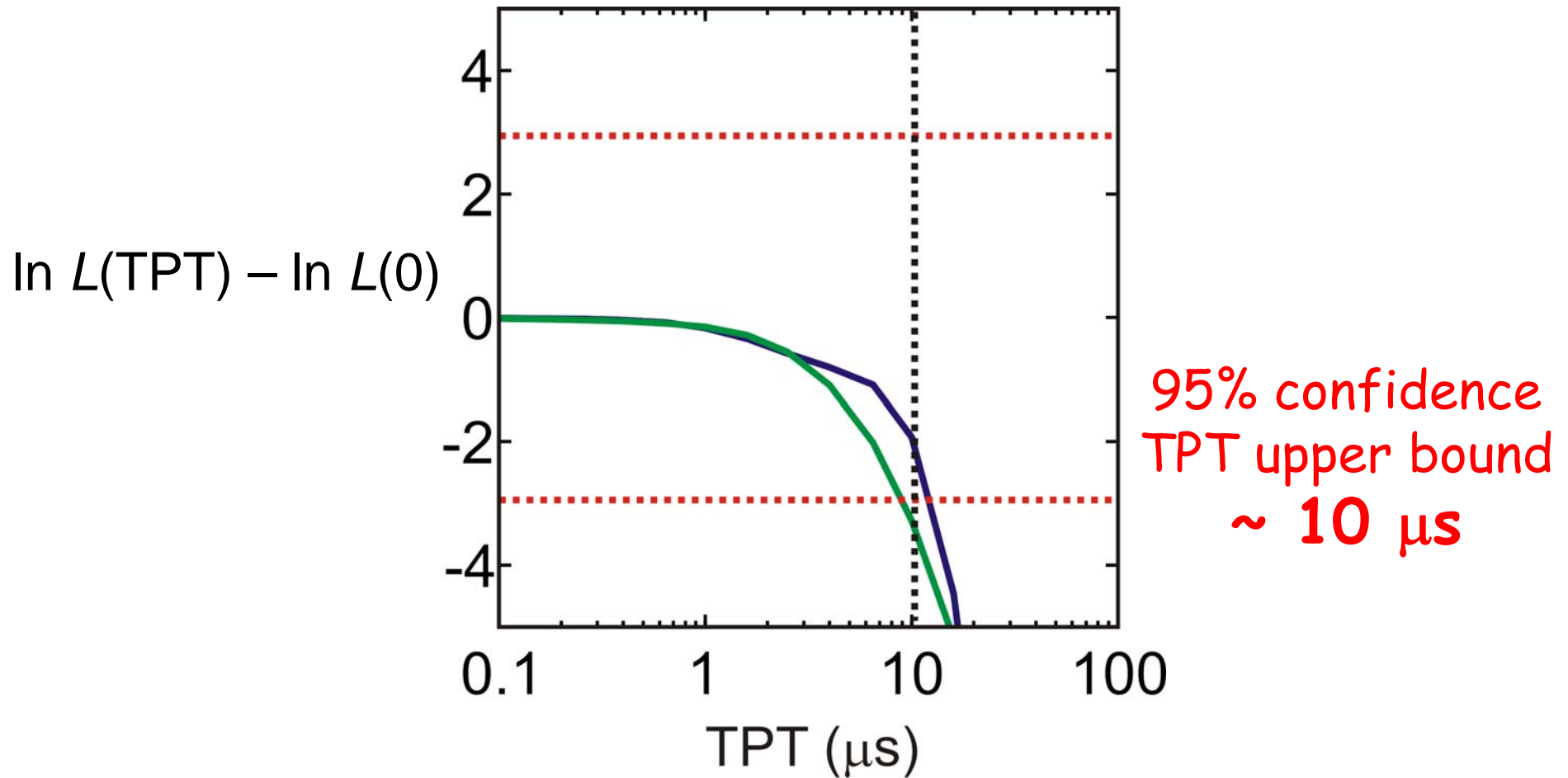
(TPT is only variable parameter in likelihood function)

$\ln L(\text{TPT}) - \ln L(0)$



Experiment

111 unfolding transitions and 40 folding transitions
for Protein G with folding time of ~ 2 s



TPT 200,000-fold < 2 s folding time

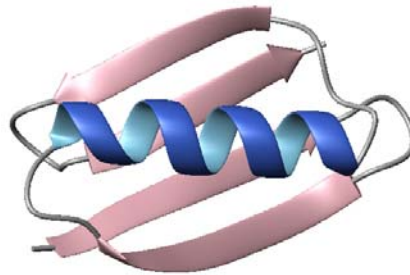
WW domain (Shaw, Gruebele)



$$\langle \tau_F \rangle = 10 \mu\text{s}$$

$$\langle \text{TPT} \rangle \sim 2 \mu\text{s}$$

Protein G



$$\langle \tau_F \rangle = 2 \text{ s}$$

$$\langle \text{TPT} \rangle < 10 \mu\text{s}$$

Transition path times differ by less than 5-fold for folding times that differ by 200,000-fold

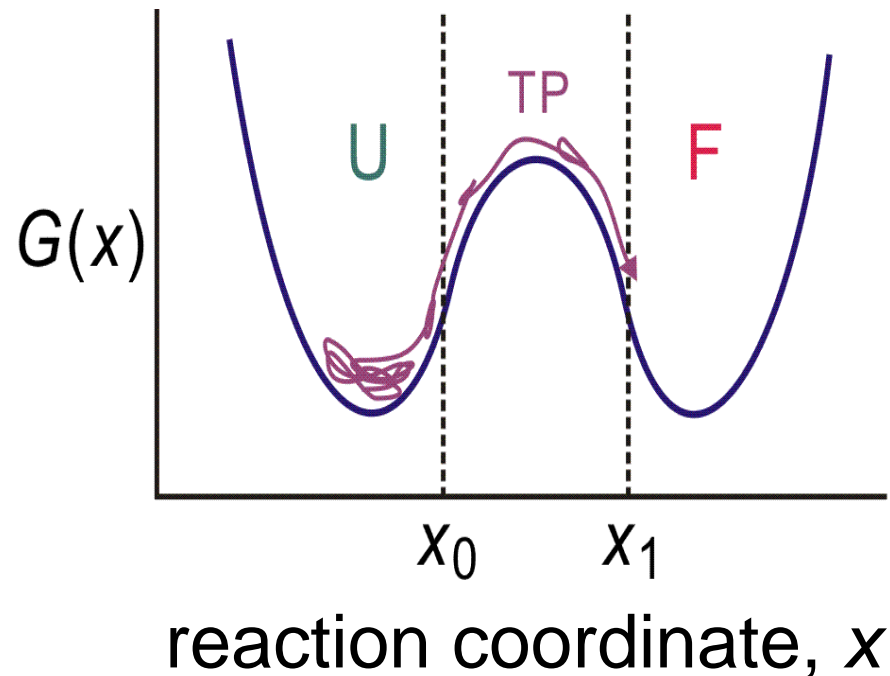
Conclusion (with caveats)

A slow folder and an ultrafast folder take almost the same time to fold when it actually happens!!

Transition path times differ by less than 5-fold for folding times that differ by 200,000-fold

How do we explain this result?

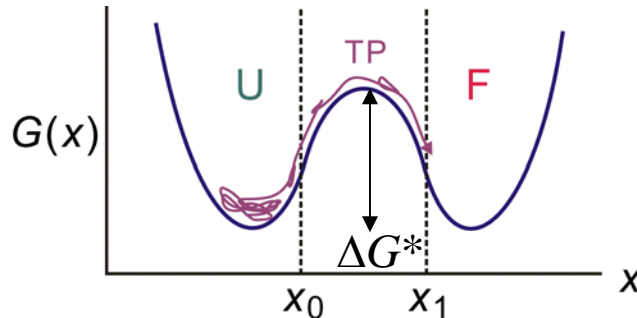
Wolynes Energy Landscape Theory (Socci, Onuchic, and Wolynes, JCP 1996)



Definition of transition path time (TPT):

Duration of trajectory that cross x_0 and reaches x_1
without ever recrossing x_0

Prediction from energy landscape theory



$$\beta = 1/k_B T$$

ΔG^* : barrier height

ω^2 : well curvature

$(\omega^*)^2$: barrier curvature

D : diffusion coefficient

$$1/k_F = \langle \tau_F \rangle = \tau_0 \exp(\beta \Delta G_F^*) \stackrel{\text{Kramers}}{=} \frac{2\pi}{D^* \beta \omega \omega^*} \exp(\beta \Delta G_F^*)$$

For diffusive crossing of harmonic barrier from x_0 to x_1 ,

$$\langle TPT \rangle \stackrel{\text{Szabo}}{\approx} \frac{\ln(3\beta \Delta G^*)}{D^* \beta (\omega^*)^2}$$

$$\text{For } \omega \omega^* = (\omega^*)^2$$

$$\langle TPT \rangle \approx \tau_0 \frac{\ln(3\beta \Delta G^*)}{2\pi}, \quad \langle TPT \rangle \approx \tau_0 \frac{\ln(3 \ln(\tau_F / \tau_0))}{2\pi}$$

Transition path time is insensitive to barrier height

Experiment

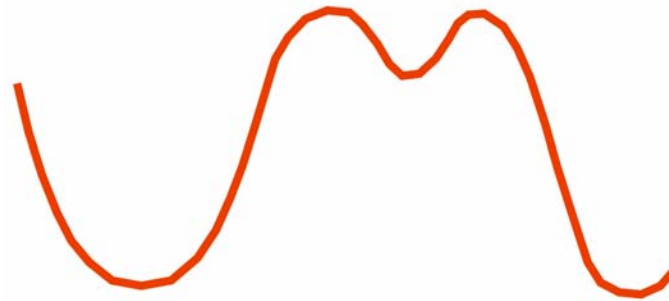
$$\frac{\tau_{F, \text{protein G}}}{\tau_{F, \text{WW domain}}} = 200,000 \frac{\langle TPT \rangle_{\text{protein G}}}{\langle TPT \rangle_{\text{WW domain}}} < 5$$

Landscape Theory (same D , ω , ω^*)
($\tau_0 \sim 1 \mu\text{s}$)

$$\frac{\langle TPT \rangle_{\text{protein G}}}{\langle TPT \rangle_{\text{WW domain}}} \approx \frac{\ln \left(3 \ln \left(\tau_{\text{protein G}} / \tau_0 \right) \right)}{\ln \left(3 \ln \left(\tau_{\text{WW domain}} / \tau_0 \right) \right)} \approx 2$$

Caveats to theoretical estimate

1. High-lying **free energy** minima will slow TPT



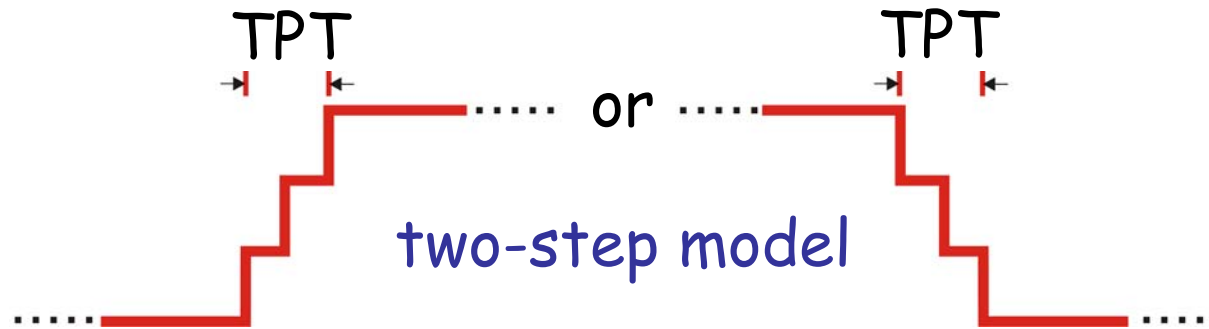
2. Very different diffusion coefficients, i.e. much rougher underlying **energy** landscape in slow folders

Caveat to experiment

FRET may only monitor part of the transition path.

Caveats

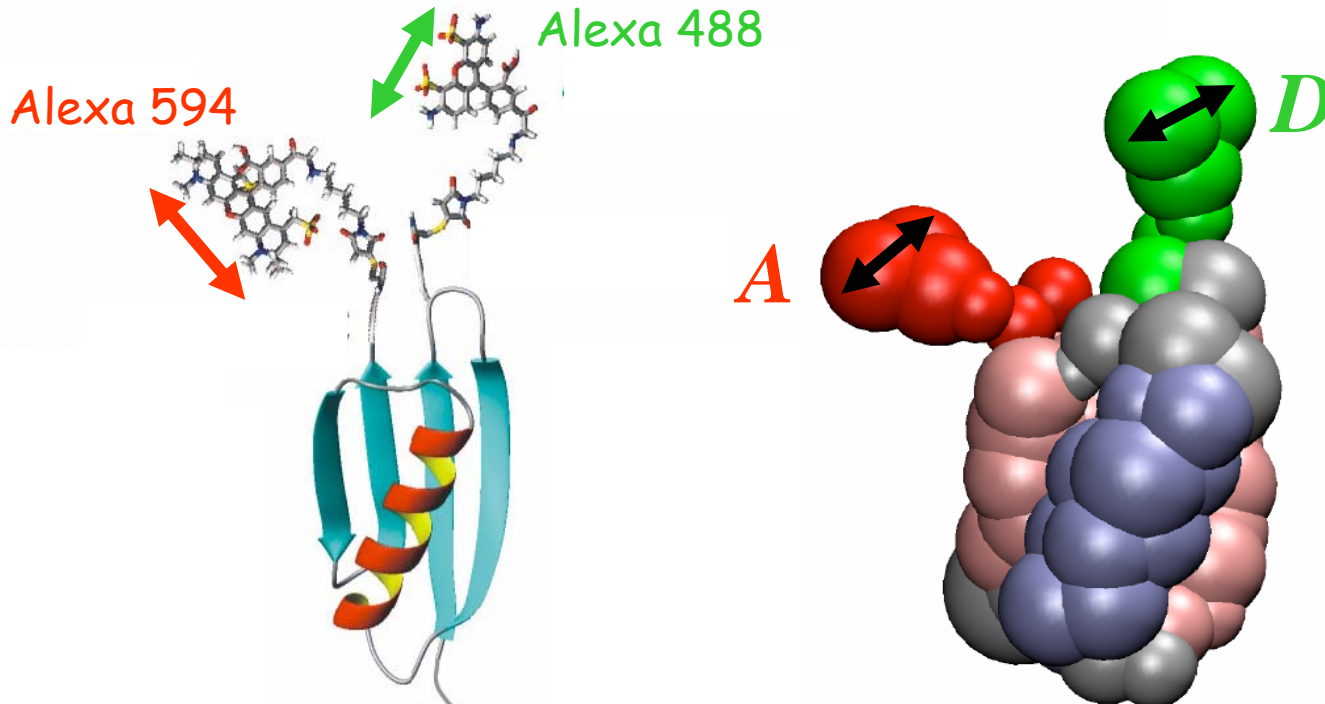
Oversimplified model



FRET may only monitor part of the transition path

Robert Best (U. Cambridge) Langevin simulations
of (Brooks, native interactions only) bead model

Calculated $Q(t)$, $R_g(t)$, $r_{DA}(t)$, $E(t)$



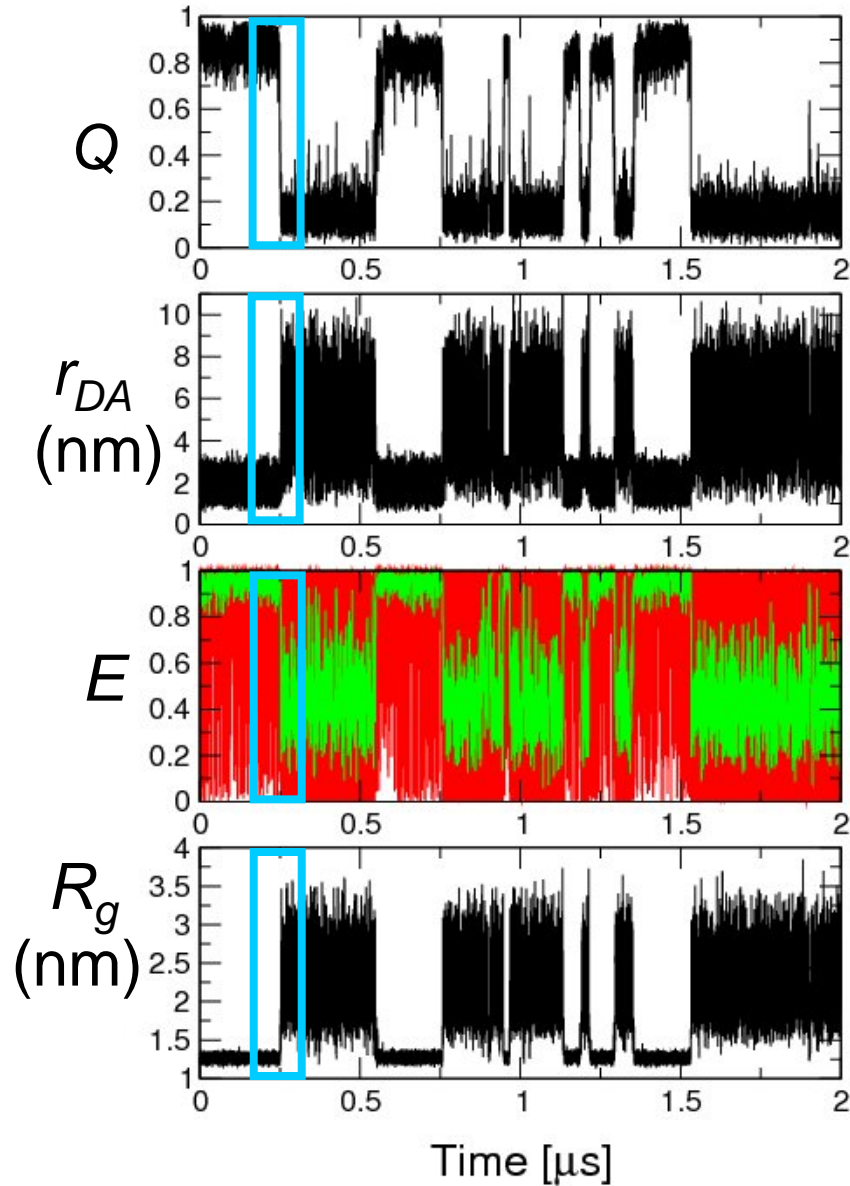
$$\kappa(t) = \hat{\mu}_D(t) \cdot \hat{\mu}_A(t) - 3(\hat{r}_{DA}(t) \cdot \hat{\mu}_D(t))(\hat{r}_{DA}(t) \cdot \hat{\mu}_A(t))$$

$$k_T(t) = k_D R_0^6 \frac{3(\kappa(t))^2}{2(r_{DA}(t))^6}$$

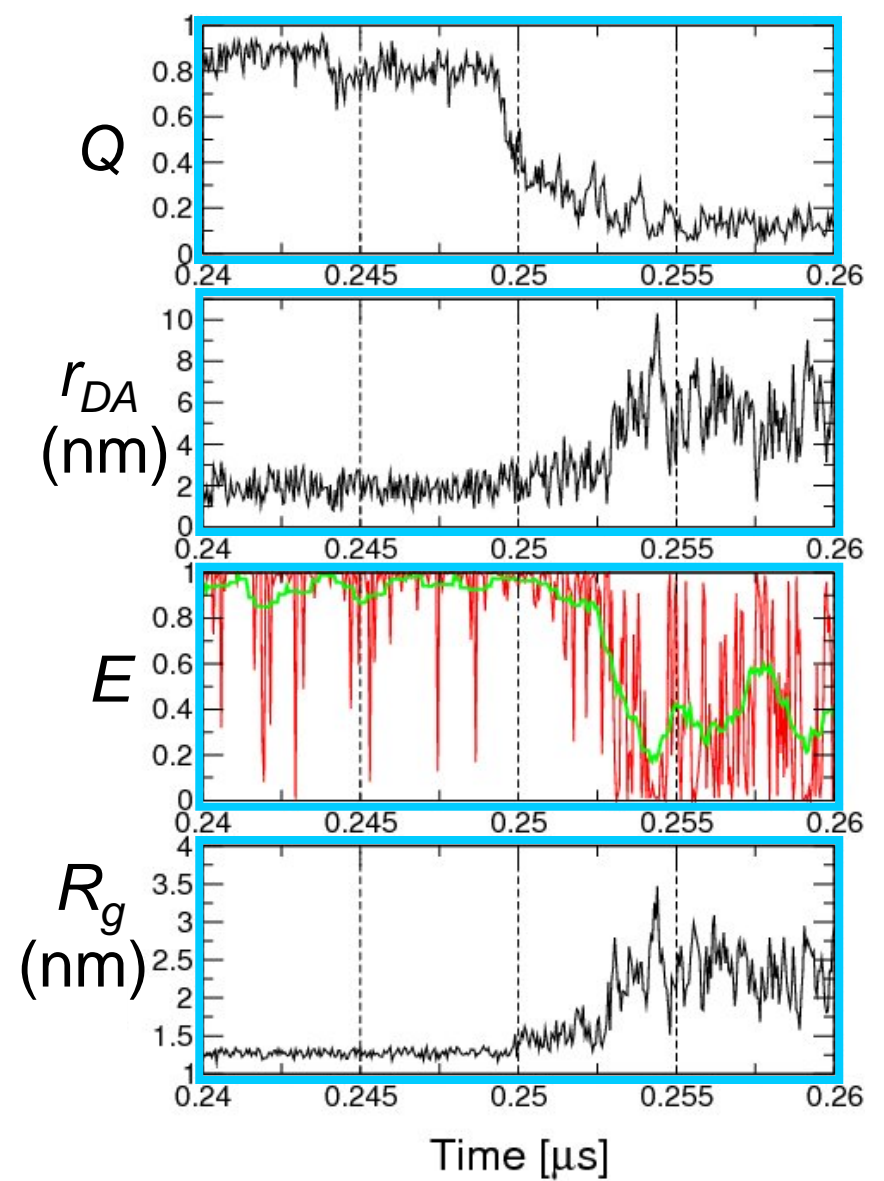
$$E(t) = \frac{k_T(t)}{k_T(t) + k_D}$$

Simulation at 1/500th water friction

2 microsecond segment



20 ns segment



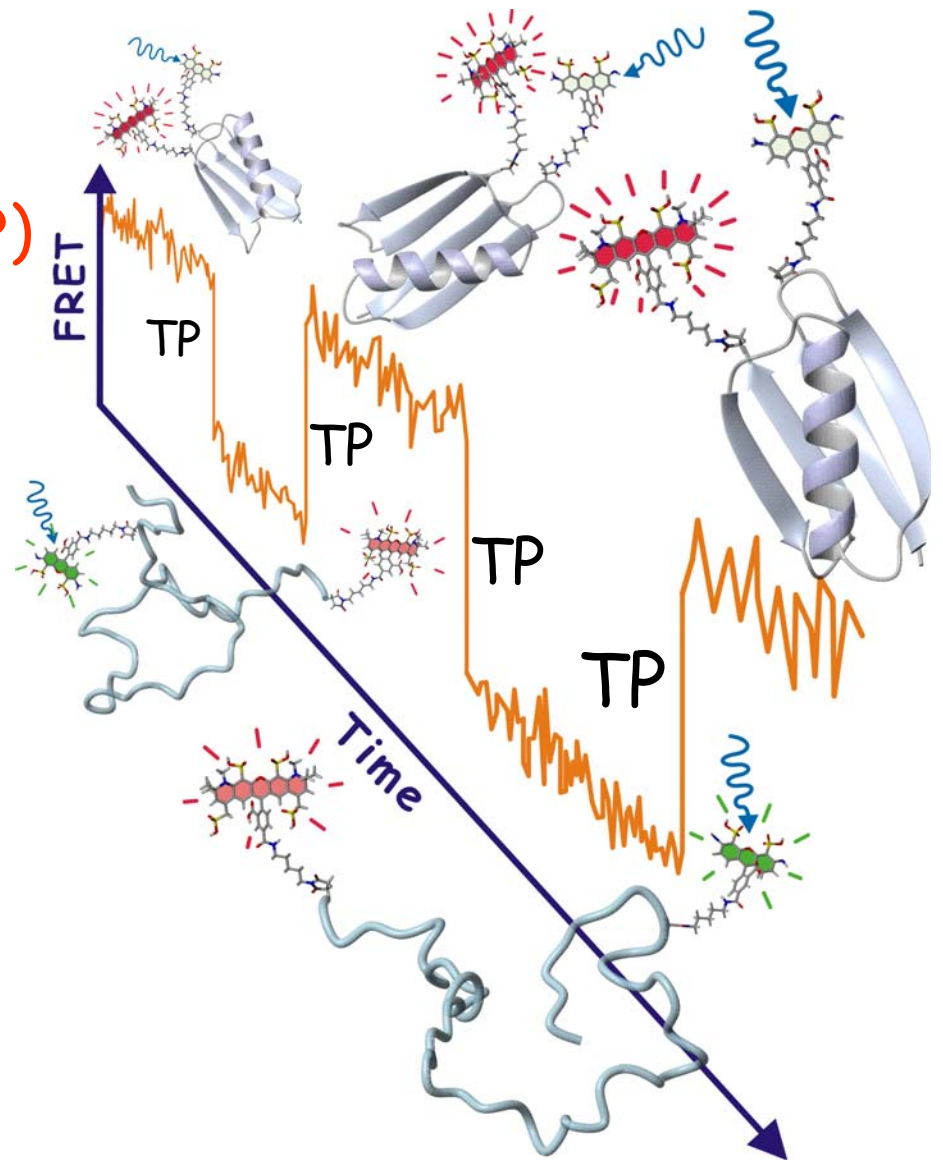
Future Directions

Measure transition path time for WW domain and protein G (?)

Distance versus time measurements during transition path for WW domain (increase viscosity)

Label with > 2 fluorophores: multiple simultaneous distances versus time places major constraints on possible folding mechanisms

Looking forward to lots of interesting results !!



Coworkers in Laboratory of Chemical Physics, NIH, Bethesda



Hoi Sung Chung



Theory of single molecule expts.
Attila Szabo and Irina Gopich



Protein engineering
John M. Louis



Automation
Kevin McHale



Langevin simulations
Robert Best, U. Cambridge