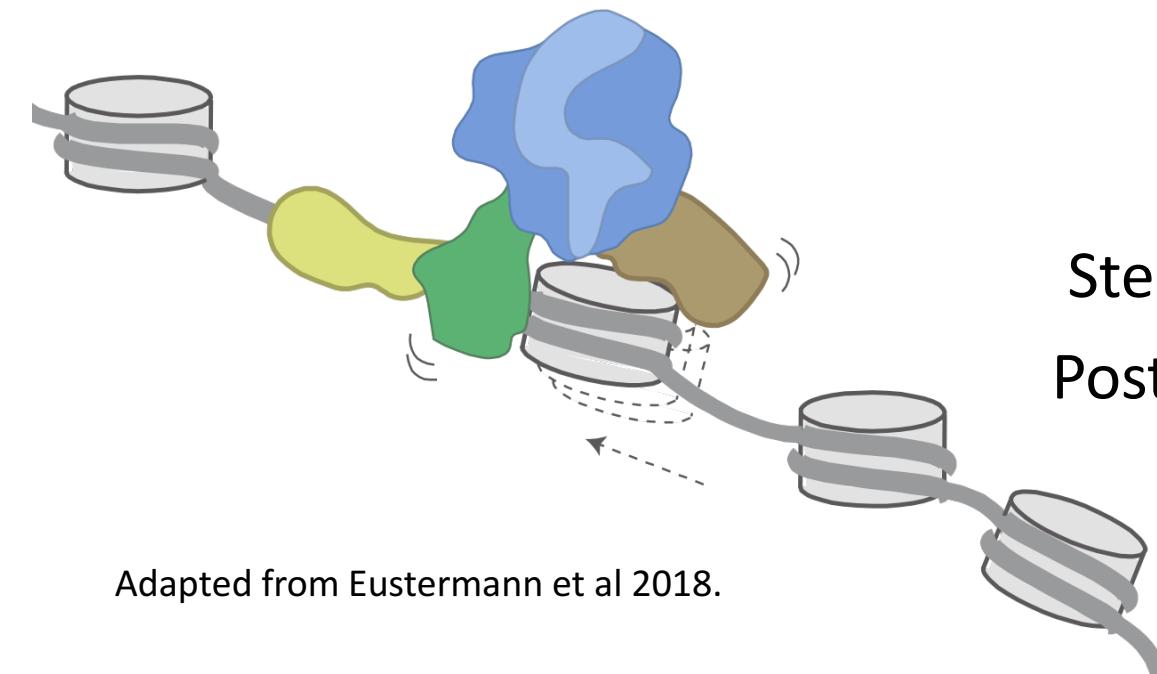


Remodeling of genome structure by ATP-dependent molecular motors:

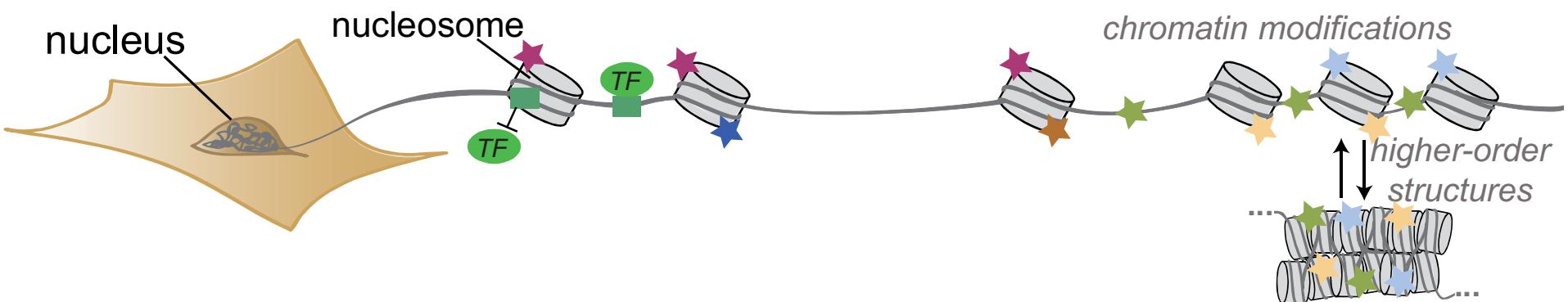
Insights from Bayesian modeling of time-
dependent microscopy data with single
molecule resolution



Stephanie Johnson, Ph.D.
Postdoctoral scholar, UCSF

Adapted from Eustermann et al 2018.

Chromatin structure is a major regulator of genome function and integrity



Nucleosomes represent a fundamental form of genome regulation by modulating DNA accessibility

AND by creating a regulatory platform containing “annotations” from:

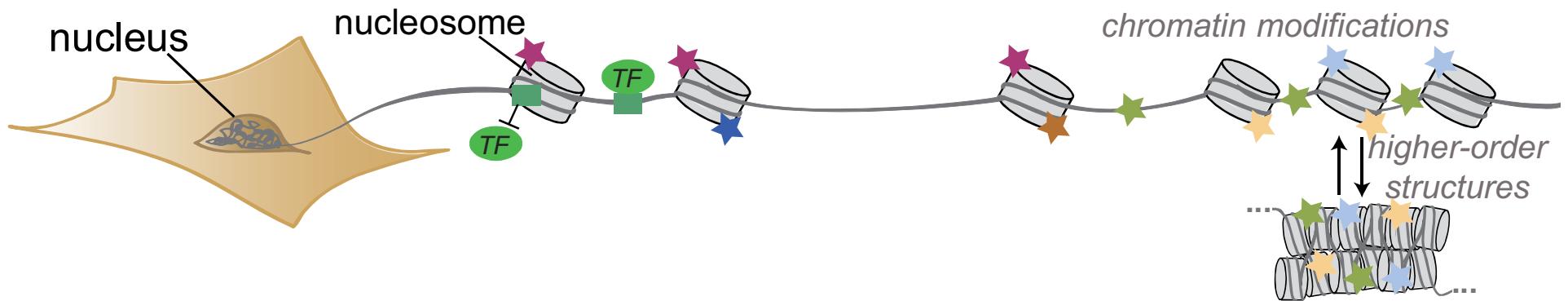
cell type	exercise
age	history of social interactions
diet	<i>etc</i>

Bludau et al. 2019

Luger et al. 2012; Chen et al. 2017;

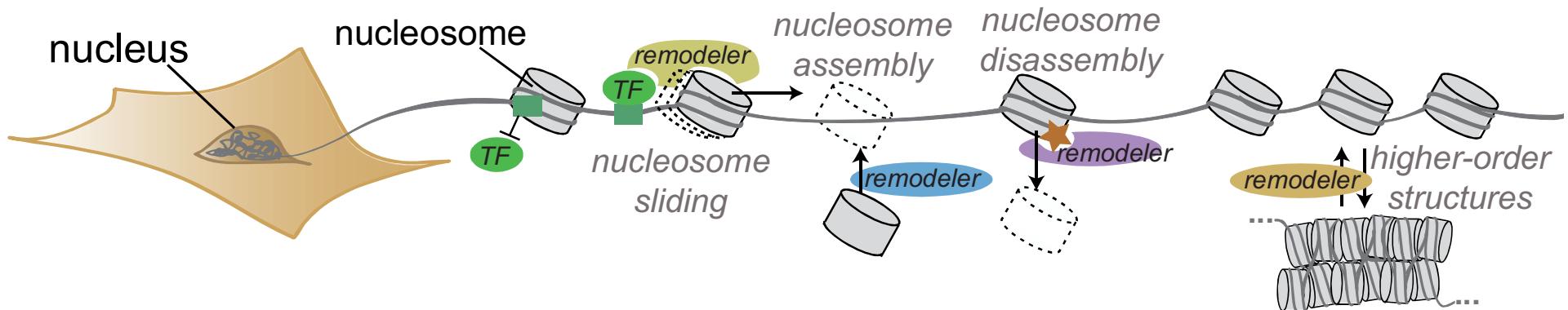
Valencia and Kadoch 2019; Molina-Serrano et al. 2019.

This regulatory platform must be highly *dynamic*



Chromatin remodelers: ATP-dependent enzymes that catalyze a wide variety of transformations to their nucleosomal substrates

This regulatory platform must be highly dynamic

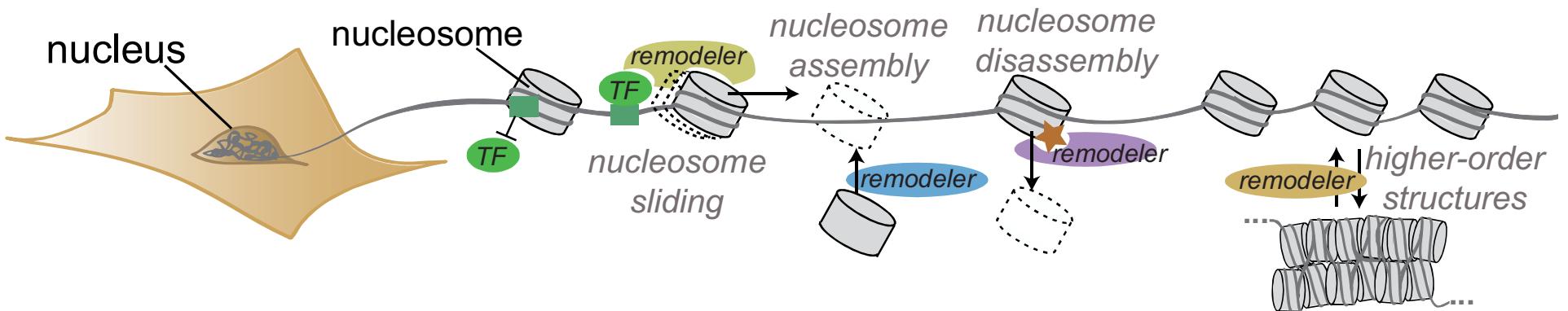


Chromatin remodelers: ATP-dependent enzymes that catalyze a wide variety of transformations to their nucleosomal substrates

Essential to all DNA-based processes (transcription, DNA damage repair, etc)

mutations to/mis-regulation of remodelers =
aging, cancer, neurodegeneration, developmental disorders, other diseases

Remodelers are specialized for specific in vivo processes



Chromatin remodelers: ATP-dependent enzymes that catalyze a wide variety of transformations to their nucleosomal substrates

10's-100's of remodelers, grouped into 4 major families

For example:

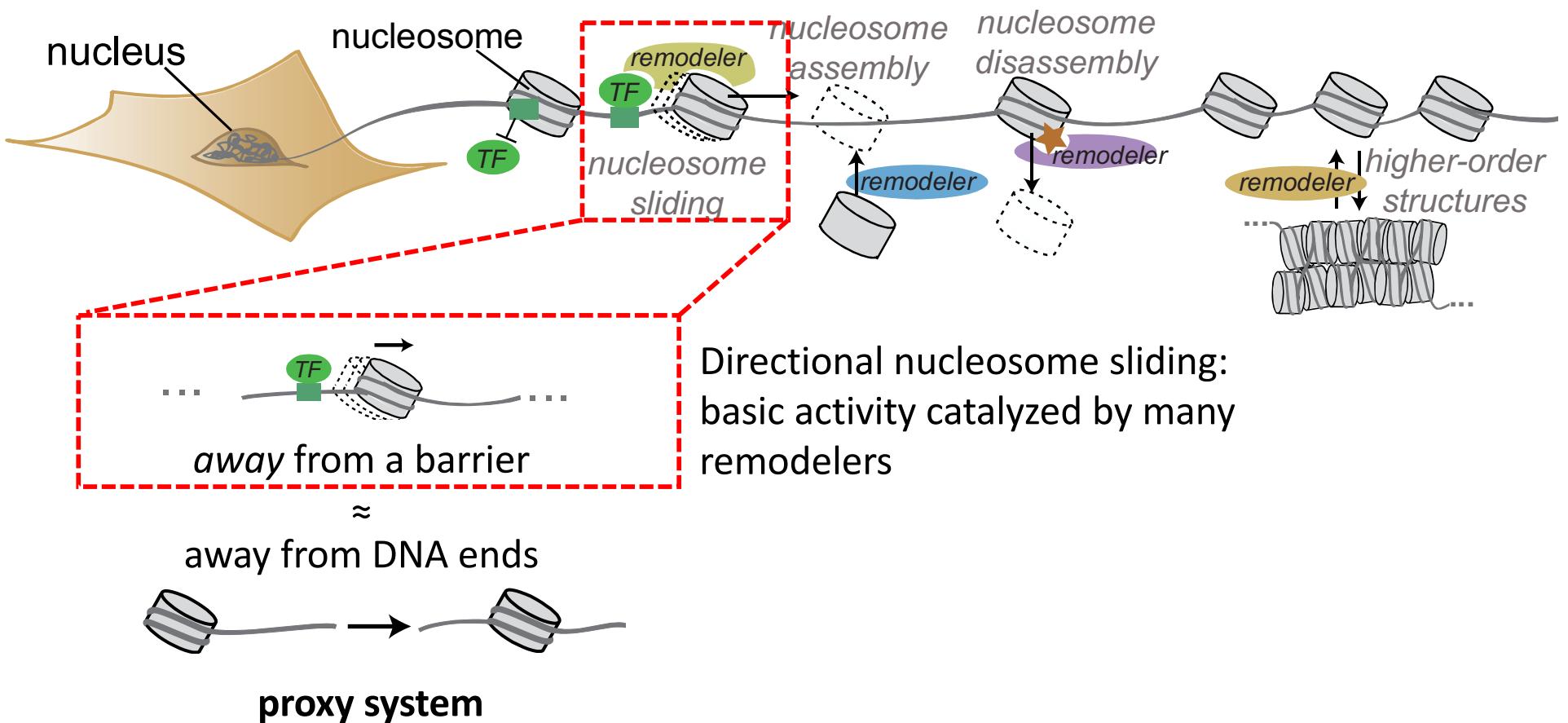
SWI/SNF family: primarily in cancer (Kadoch et al. 2013)

CHD family seems equally important in aging and cancer

(Pegoraro et al. 2009, Riedel et al. 2013, McCormick et al. 2015)

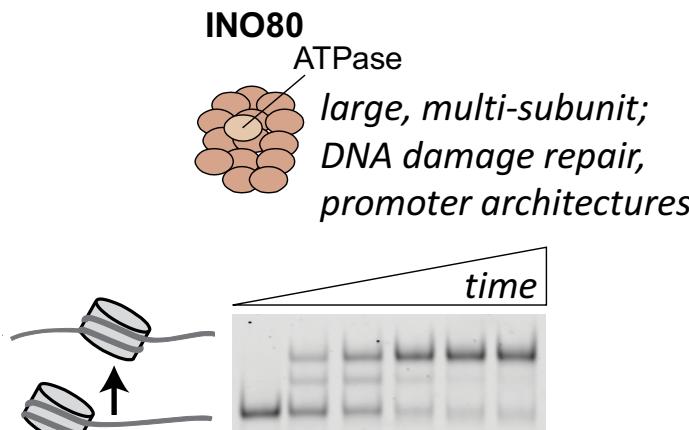
Remodelers are specialized for specific in vivo processes. But identifying mechanistic and regulatory differences has been hard!

Focusing on a particular enzymatic reaction:

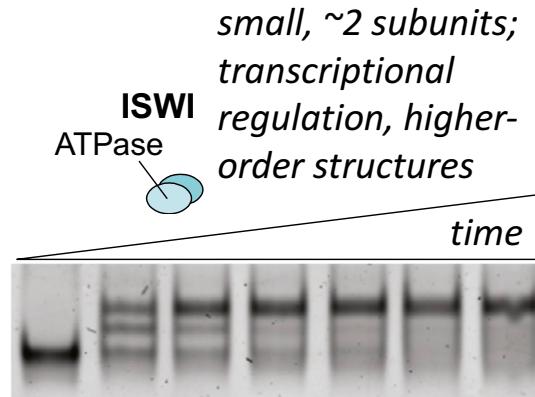


Conventional assays are limited in what they can tell us about this activity

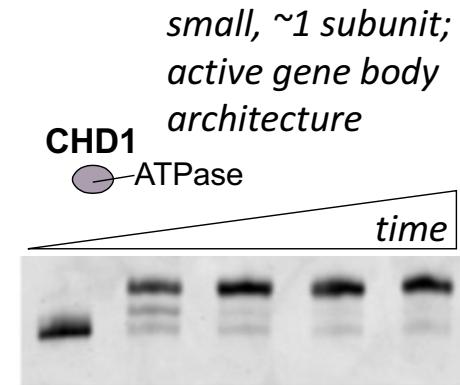
Problem: Directional nucleosome sliding by very different enzymes often looks very similar in standard biochemical assays!



Zhou*, Johnson*, et al. 2018
(*equal contribution)



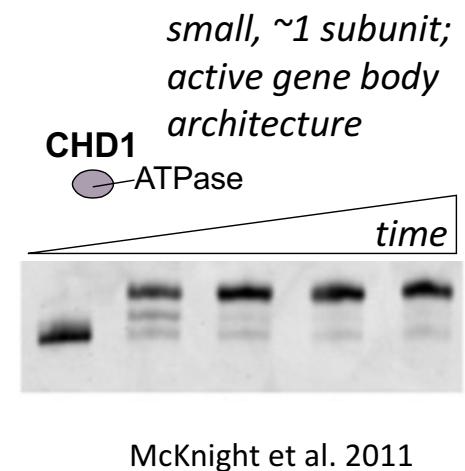
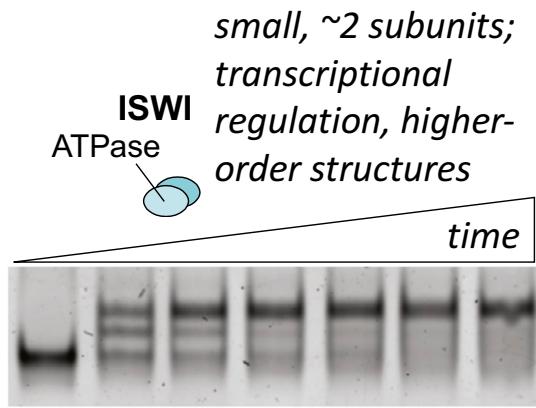
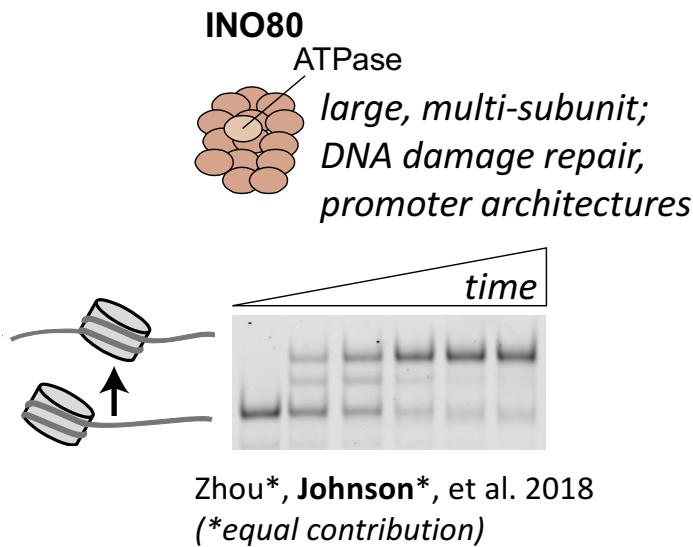
Leonard and Narlikar 2015



McKnight et al. 2011

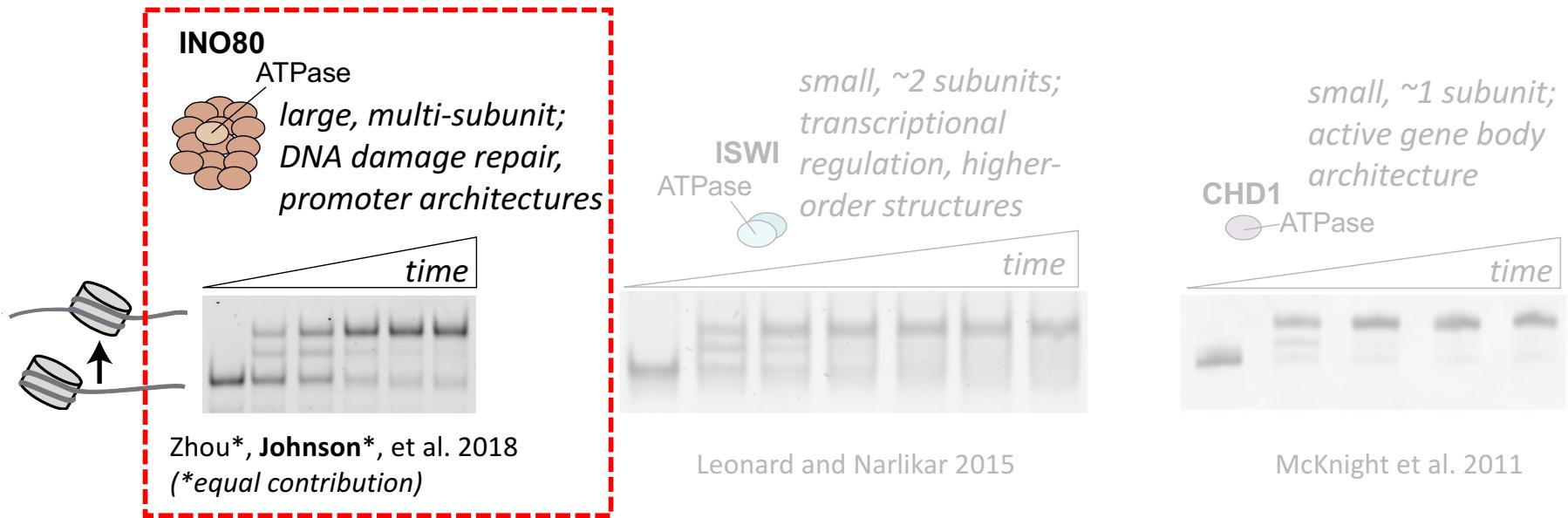
Hypothesis: there are differences here, we just can't see them in these assays

Outline



- A microscopy-based assay for generating time course data that gives a different window on directional nucleosome sliding
- Quantifying fast, non-instantaneous transitions in noisy microscopy data with sub-camera frame resolution: Slopey
- What Slopey tells us about INO80's mechanism and regulation

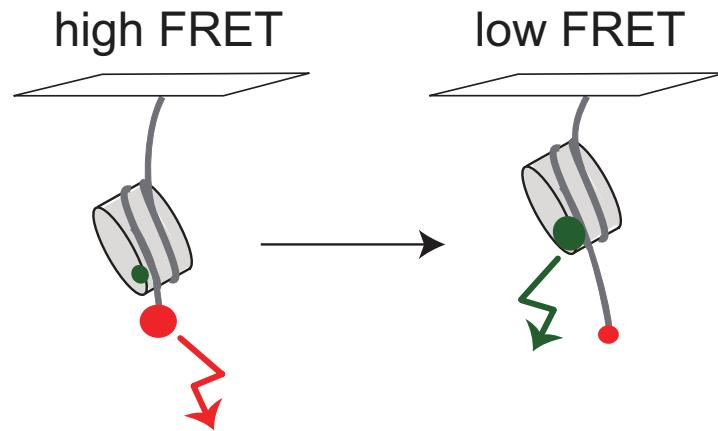
Outline



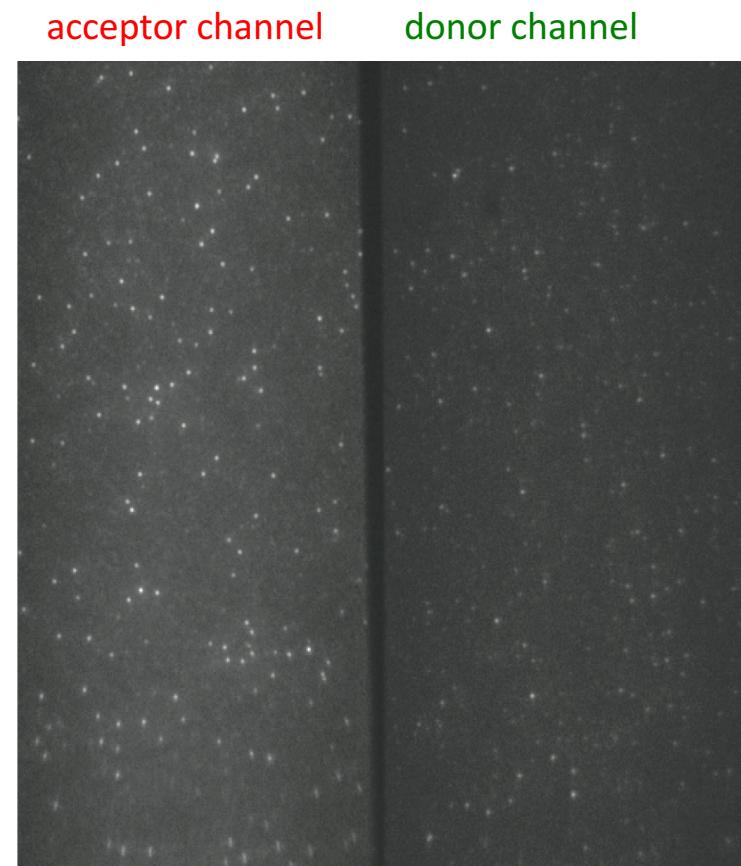
- A microscopy-based assay for generating time course data that gives a different window on directional nucleosome sliding
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Watching the sliding of individual nucleosomes by single molecule FRET

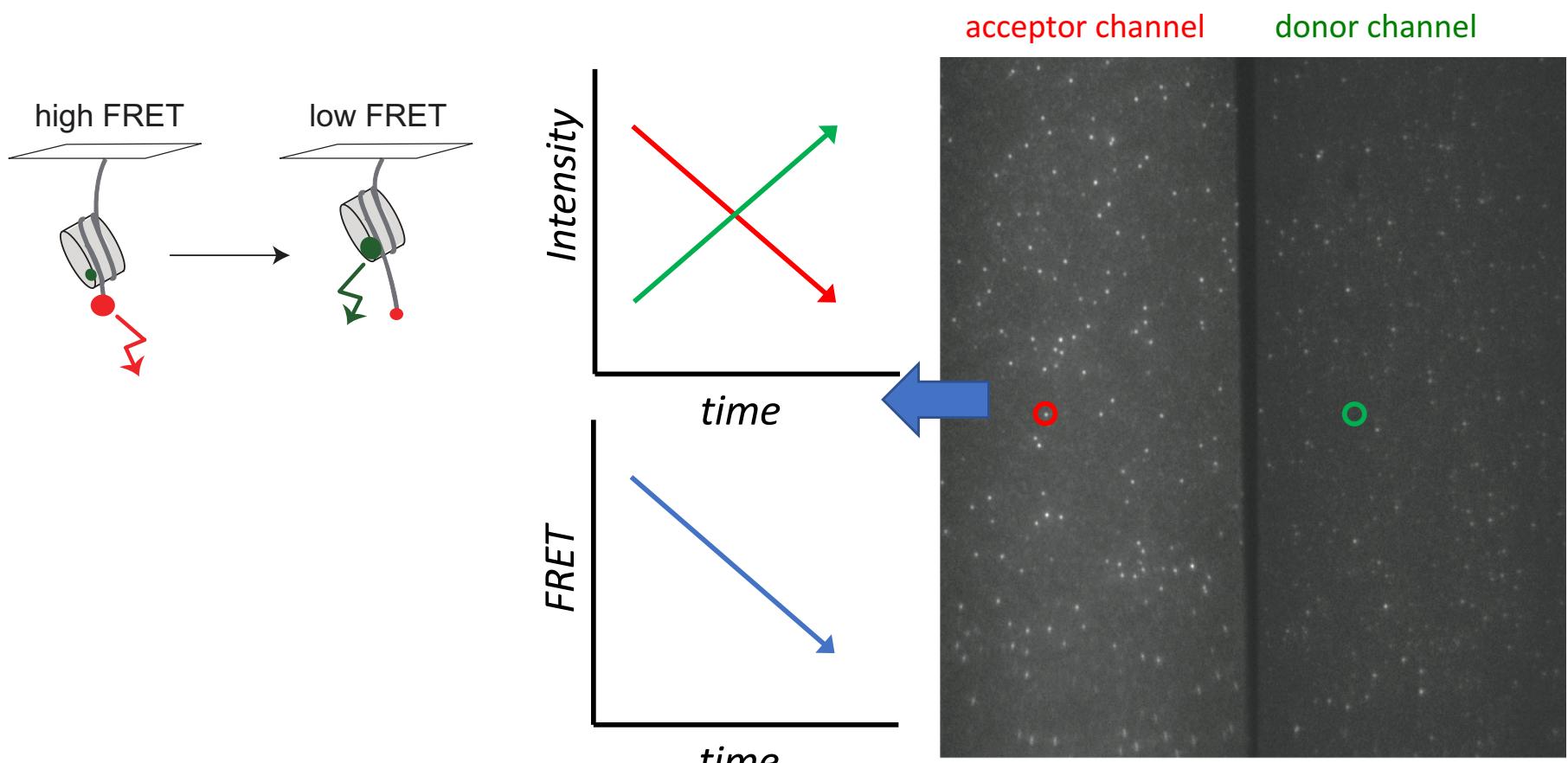
“It is very easy to answer many of these fundamental biological questions; you just look at the thing!” – *Richard Feynman*



Red (Cy5) fluorescent ACCEPTOR dye
Green (Cy3) fluorescent DONOR dye

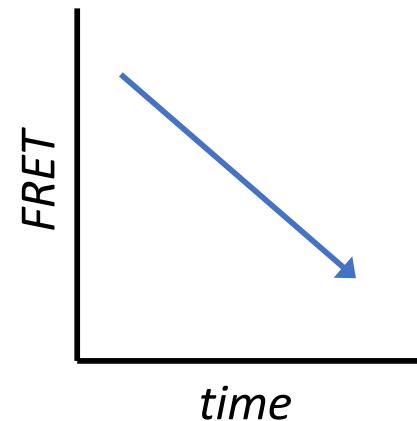
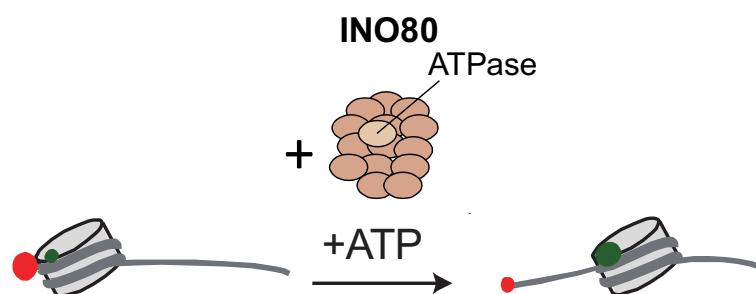


Watching the sliding of individual nucleosomes by single molecule FRET

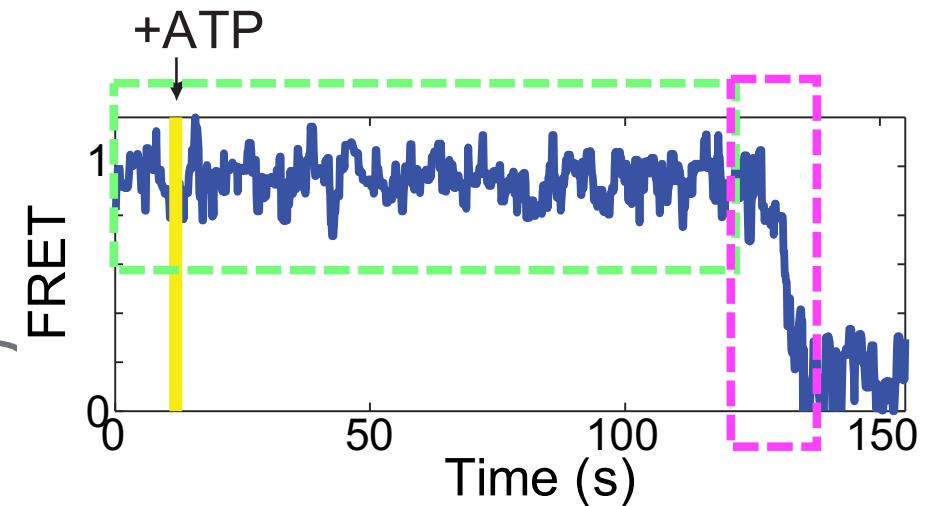
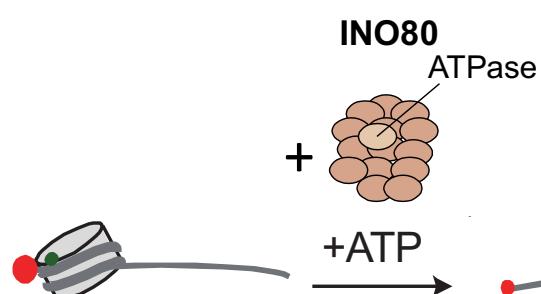


Traces: github.com/stephlj/Traces

What does INO80-mediated sliding look like at the single nucleosome level?



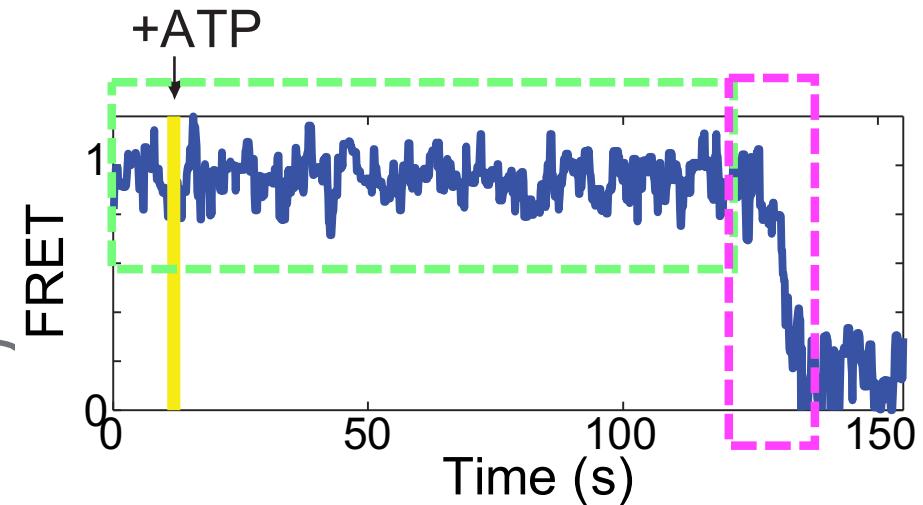
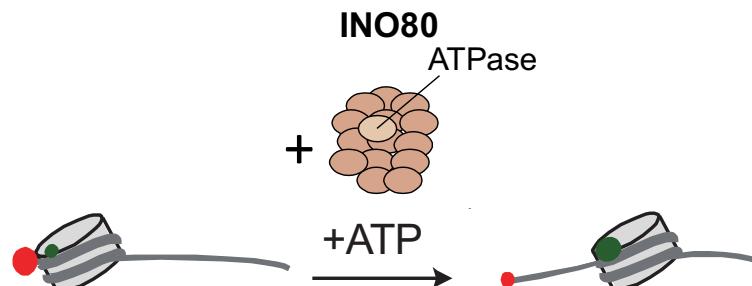
What does INO80-mediated sliding look like at the single nucleosome level?



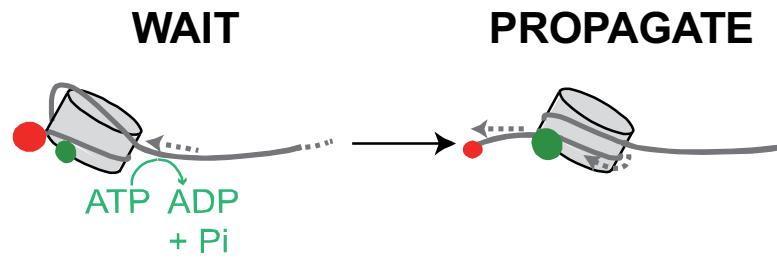
Two reaction phases, with very different timescales.

What physical processes do these two phases of the reaction correspond to?

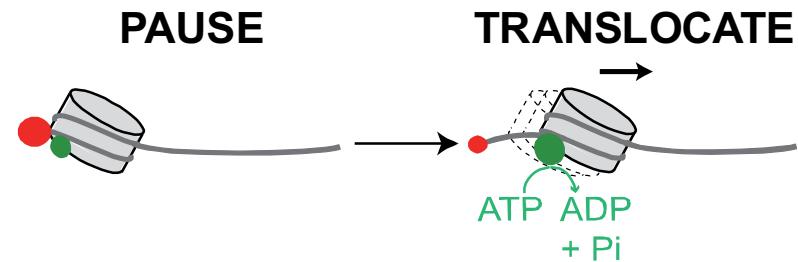
What's happening in the two reaction phases?



“Bulge” propagation?

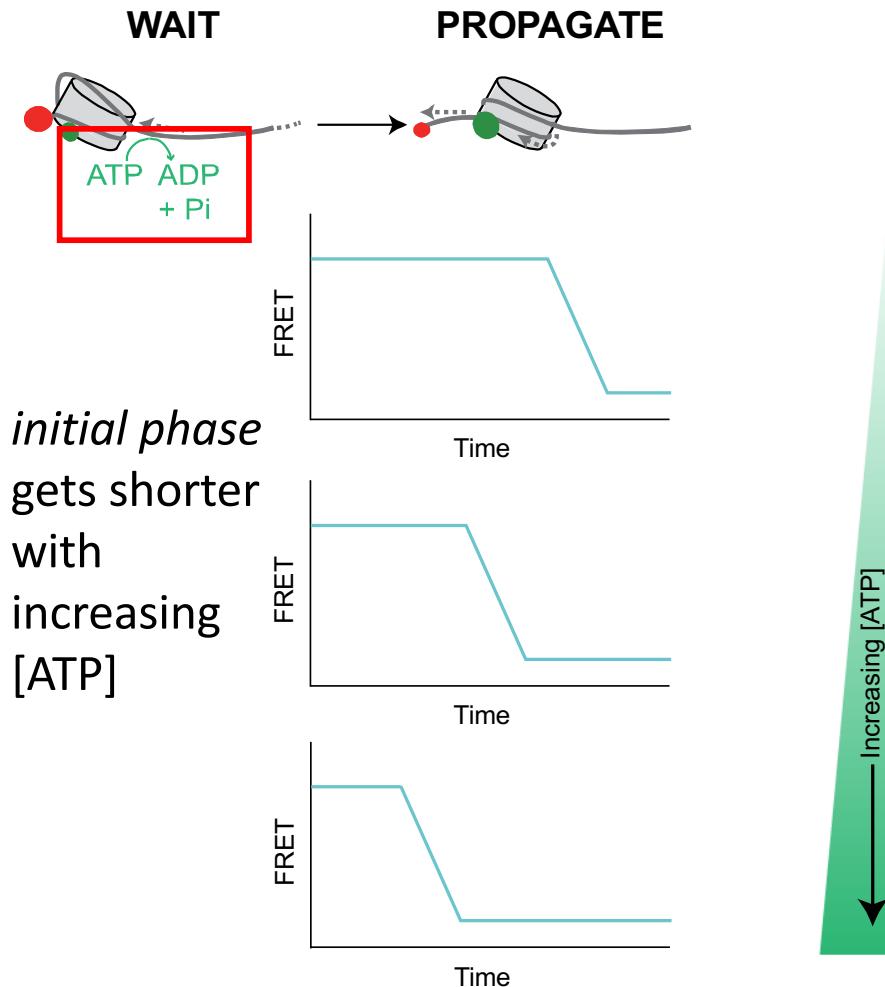


ATP-dependent translocation?



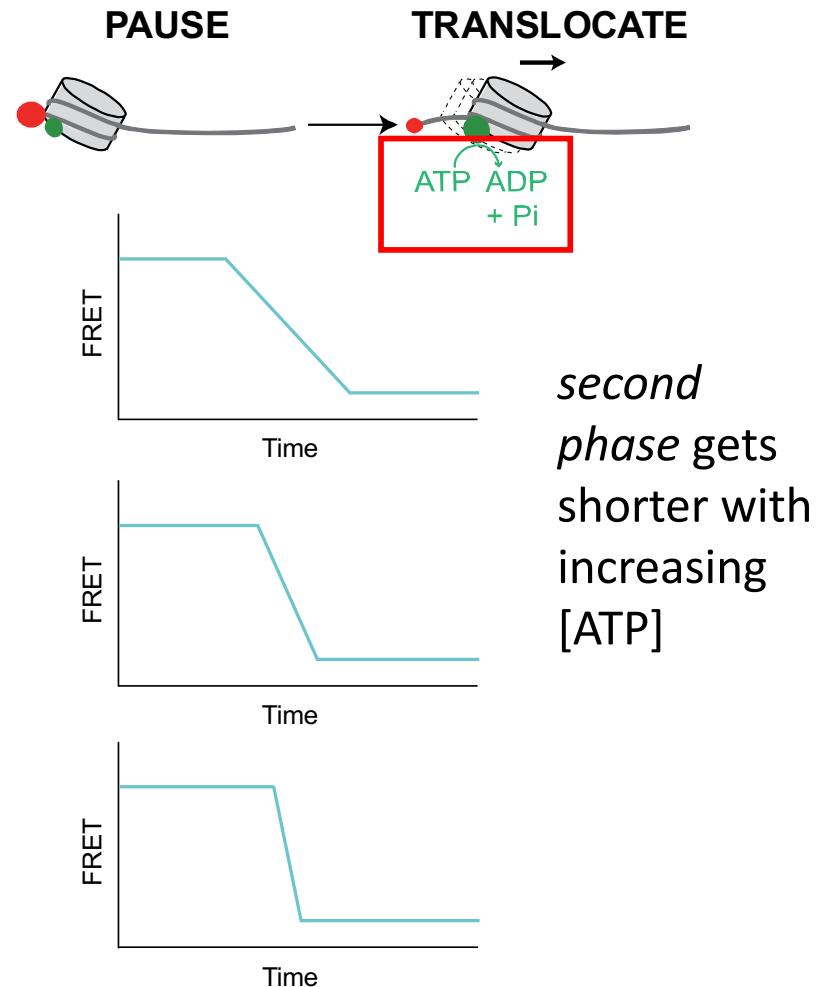
Bulge propagation or pause-translocation?

“Bulge” propagation?



*initial phase
gets shorter
with
increasing
[ATP]*

ATP-dependent translocation?

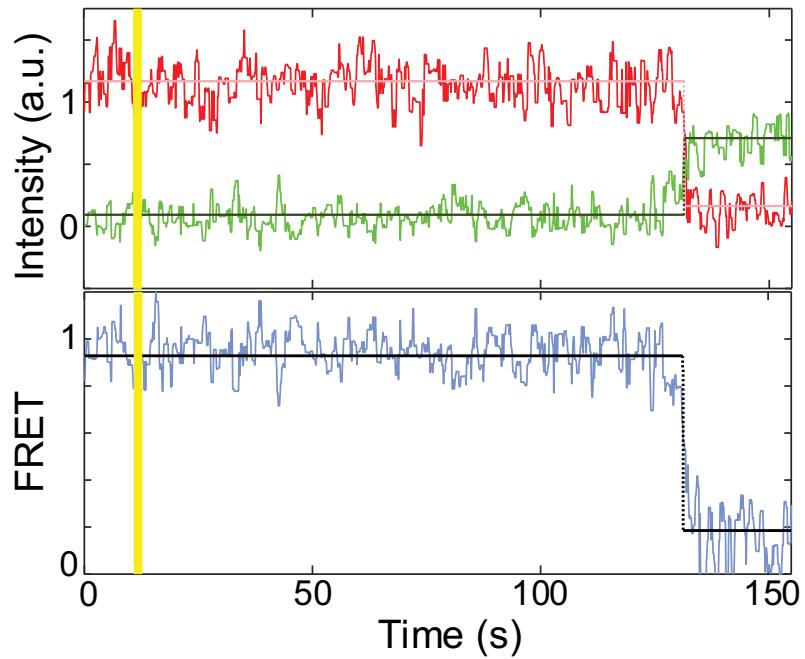


*second
phase gets
shorter with
increasing
[ATP]*

Need a way to quantify the two phases!

A conventional HMM can quantify the long initial phase

Discrete-time HMM



Hidden Markov model with Gaussian emissions

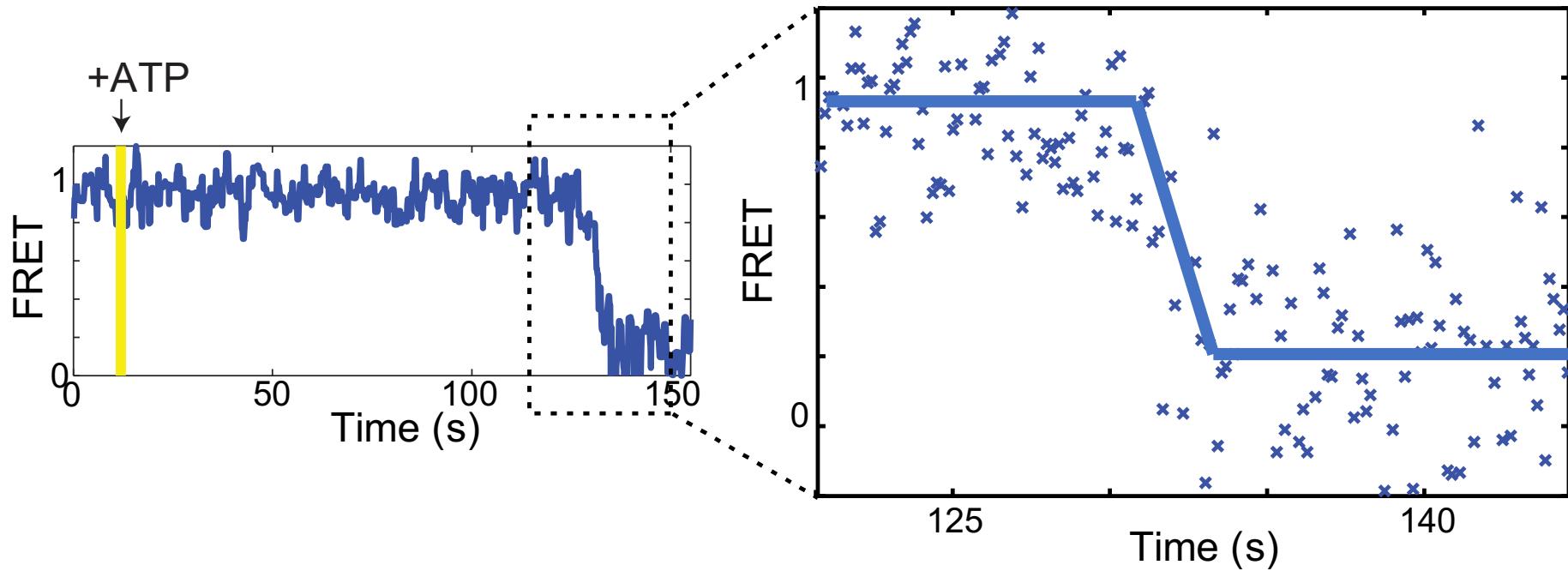
Fit using Gibbs sampling

Based on the pyhsmm Python package (github.com/mattjj/pyhsmm),
available as part of the Traces package at github.com/stephlj/Traces

Armache*, Gamarra*, **Johnson** et al. 2019,
Gamarra, **Johnson**, et al. 2018,
Zhou*, **Johnson***, et al 2018 (*equal contribution).

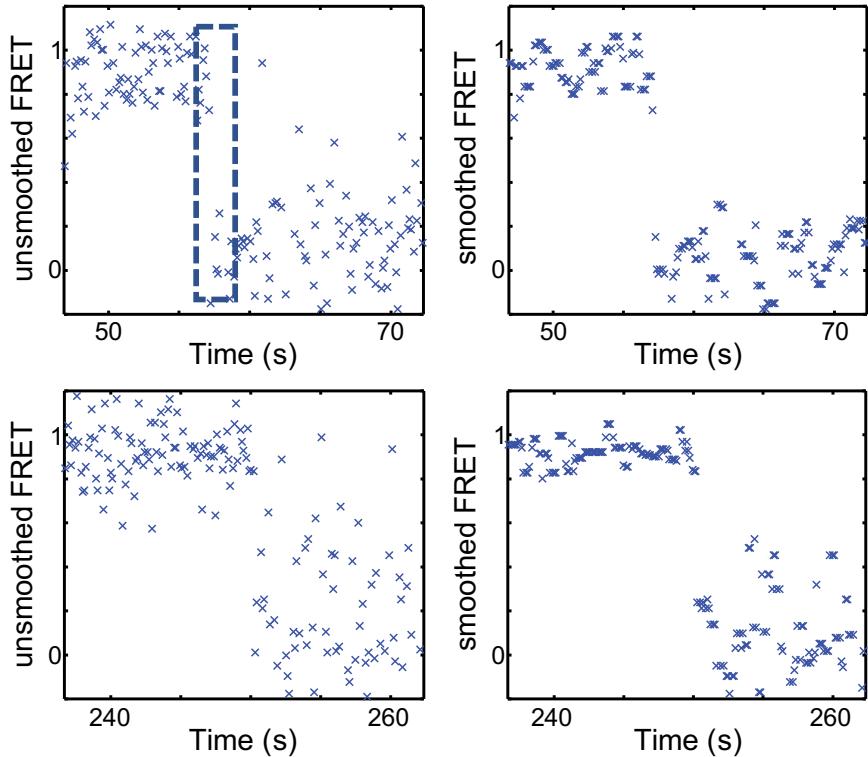
But how to quantify the rapid decrease in FRET?

It's really fast! But not instantaneous.

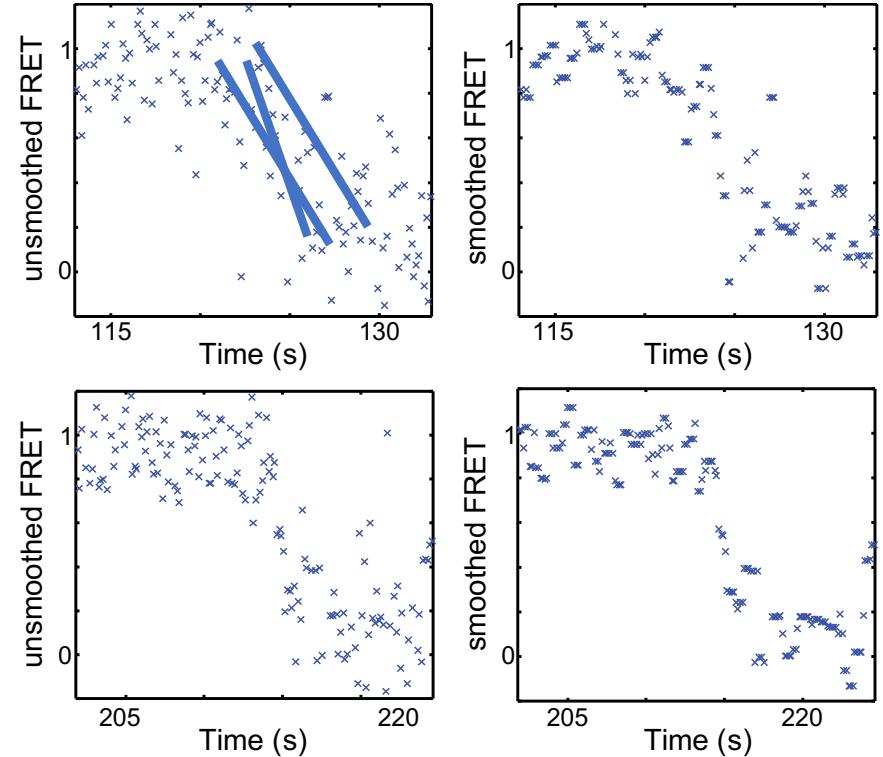


The second phase is not instantaneous

Photobleaching (no INO80):



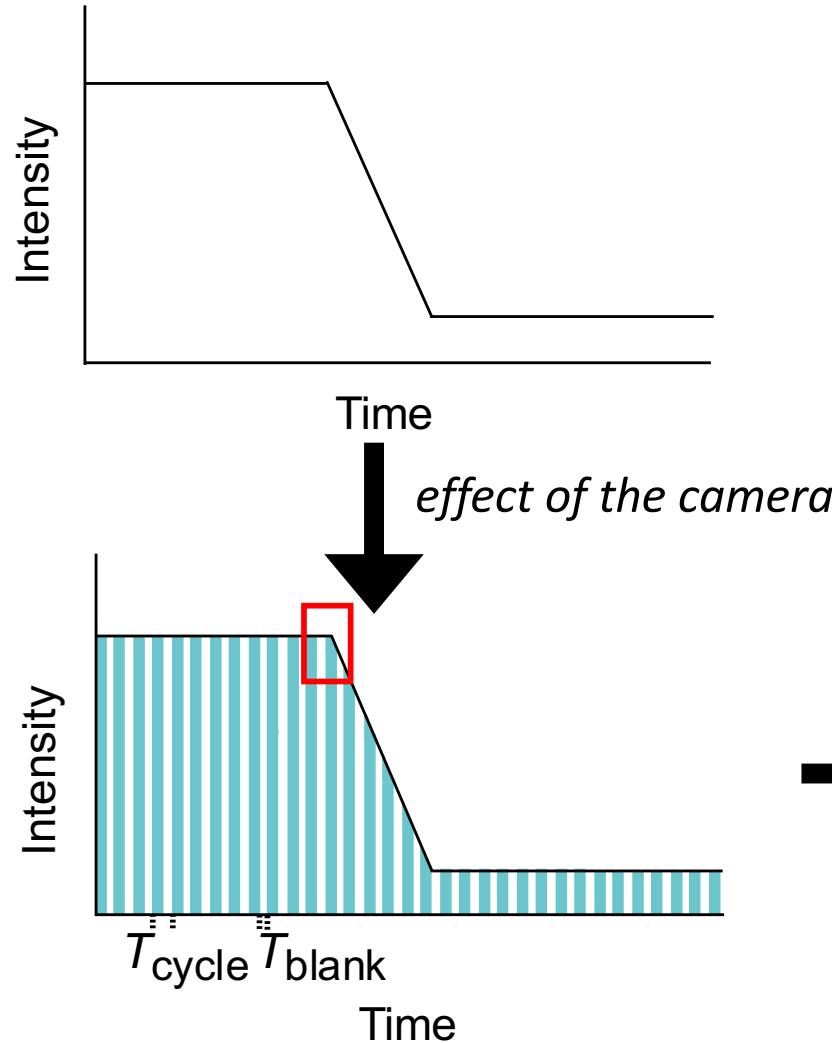
INO80 second phase:



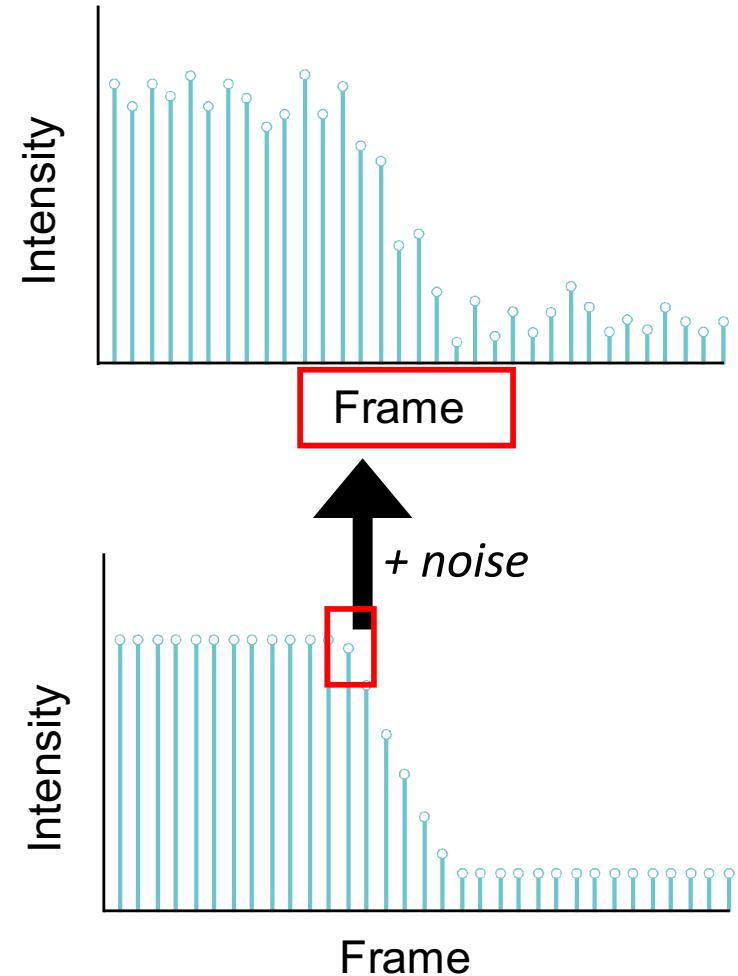
Slopey: a continuous-time generative model with explicit camera modeling
github.com/stephlj/slopey (with Matt Johnson at Google Brain)

Explicit camera modeling allows sub-frame inferences

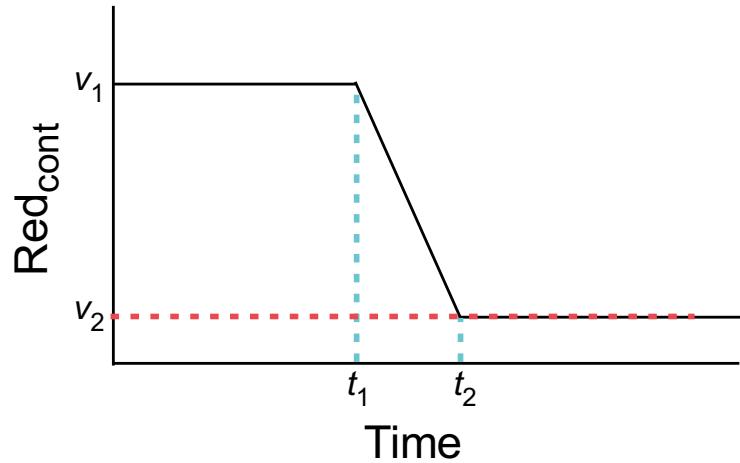
Model for the underlying (continuous) reality:



But what we observe is:



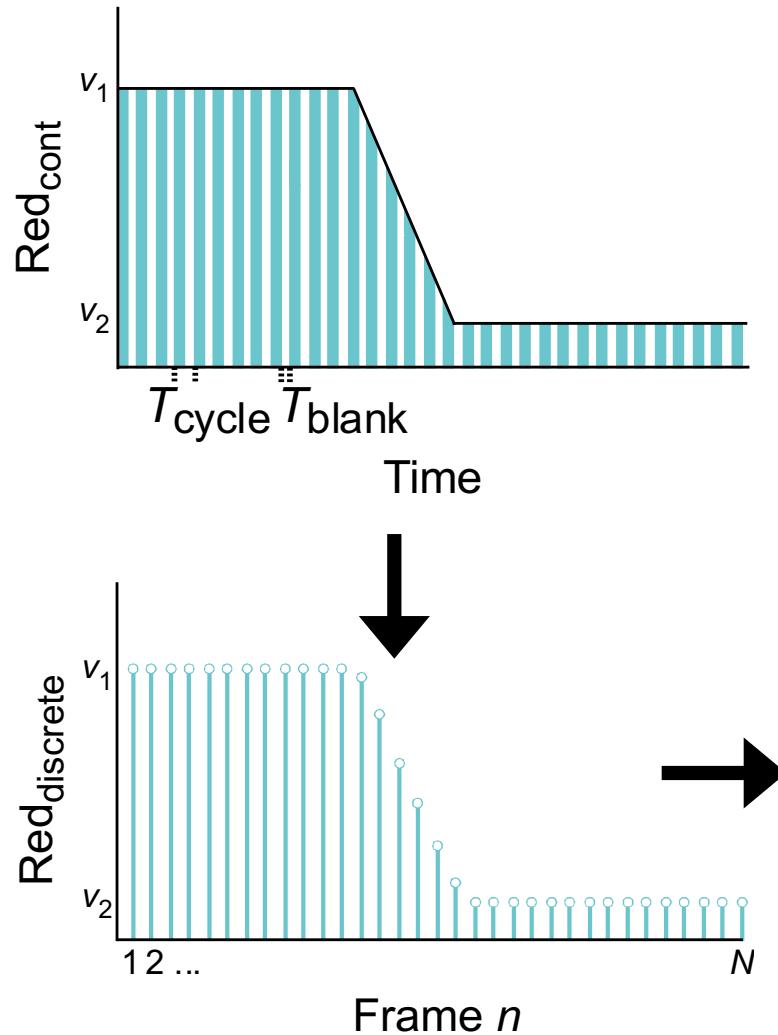
A generative model from continuous time slopes to discrete time observations



Red_{cont} is piecewise linear in time, with

$$\text{Red}_{\text{cont}}(t) = \begin{cases} v_1, & t < t_1 \\ \frac{v_2 - v_1}{t_2 - t_1}t + v_1 - t_1 \frac{v_2 - v_1}{t_2 - t_1}, & t_1 \leq t \leq t_2 \\ v_2, & t > t_2 \end{cases}$$

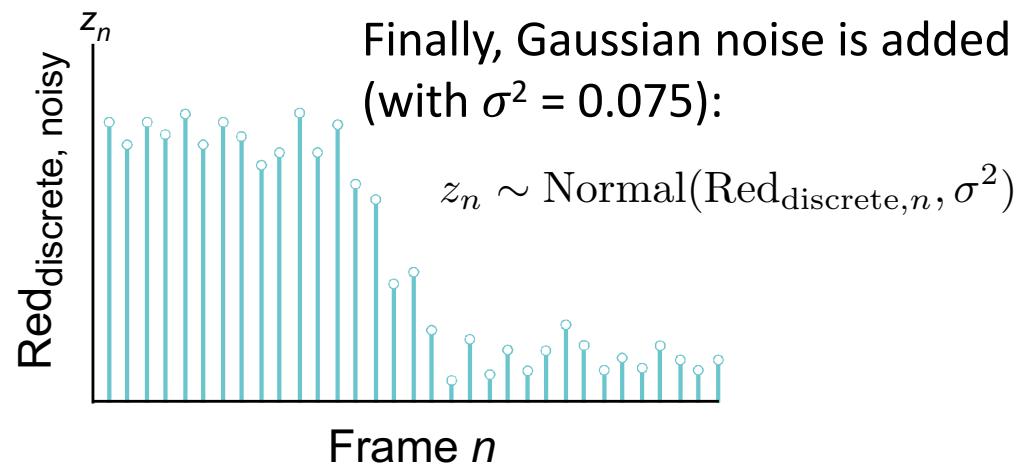
A generative model from continuous time slopes to discrete time observations



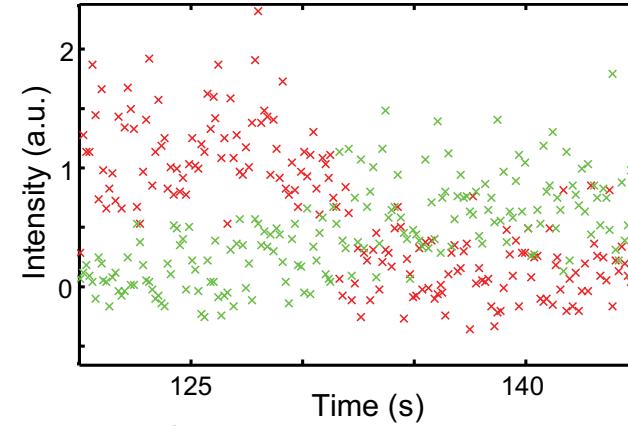
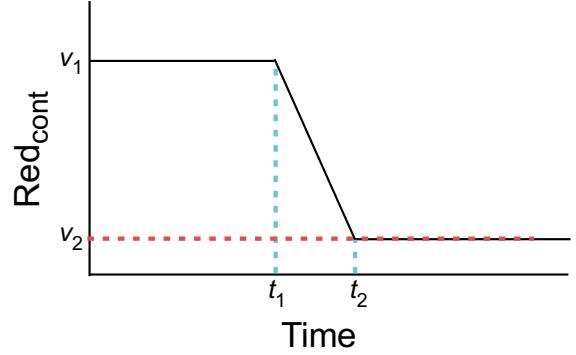
$\text{Red}_{\text{discrete}}$ is related to Red_{cont} by taking into account the integration performed by the camera:

$$\text{Red}_{\text{discrete},n} = \int_{t_{n-1} + T_{\text{blank}}}^{t_n} \text{Red}_{\text{cont}}(t) dt$$

where $t_n = T_{\text{cycle}} \cdot n$ for $n = 1, 2, \dots, N$.



Inference: the inverse problem



We have a model that generates data, given parameters.
But we have data, and want parameters!

Bayesian inference:

Priors: $t_1 \sim \text{Exp}(a)$

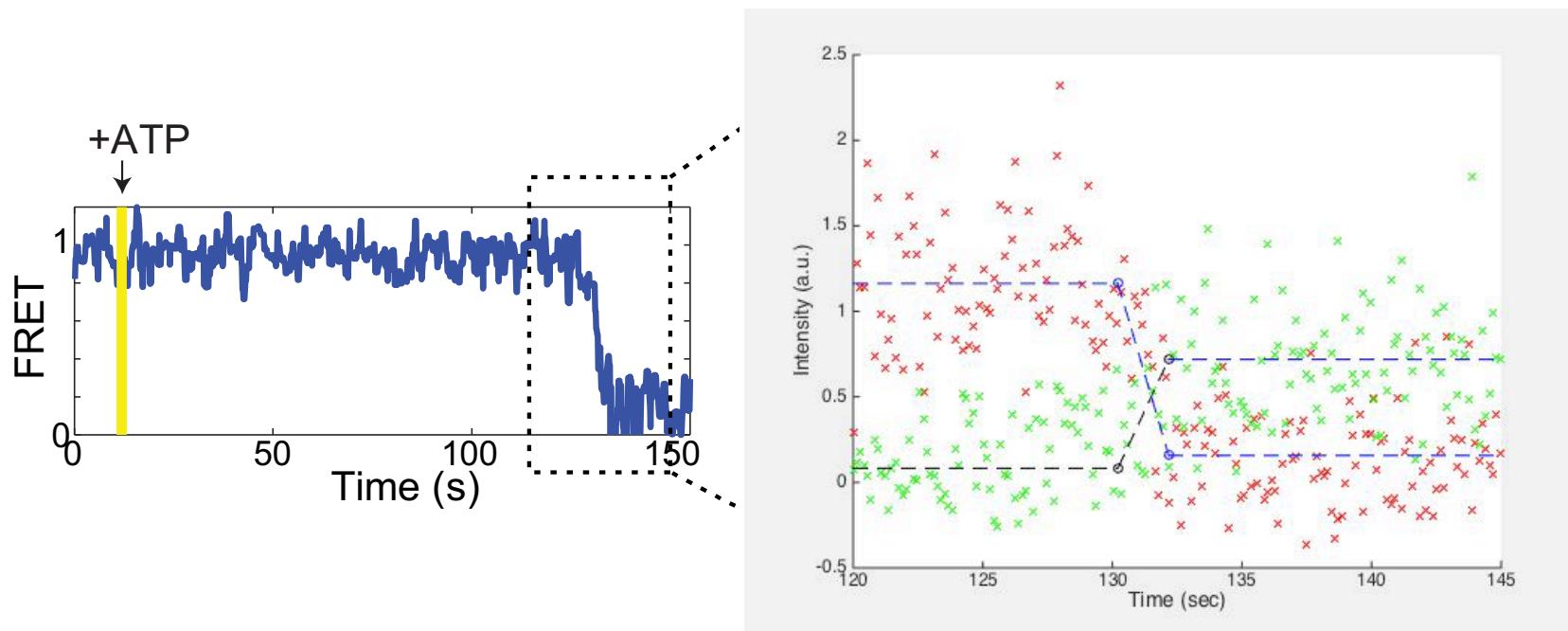
$(t_2 - t_1) \sim \text{Uniform}(0.14, 4.5)$

$v_k \sim \text{Exp}(b)$

Likelihood: camera model

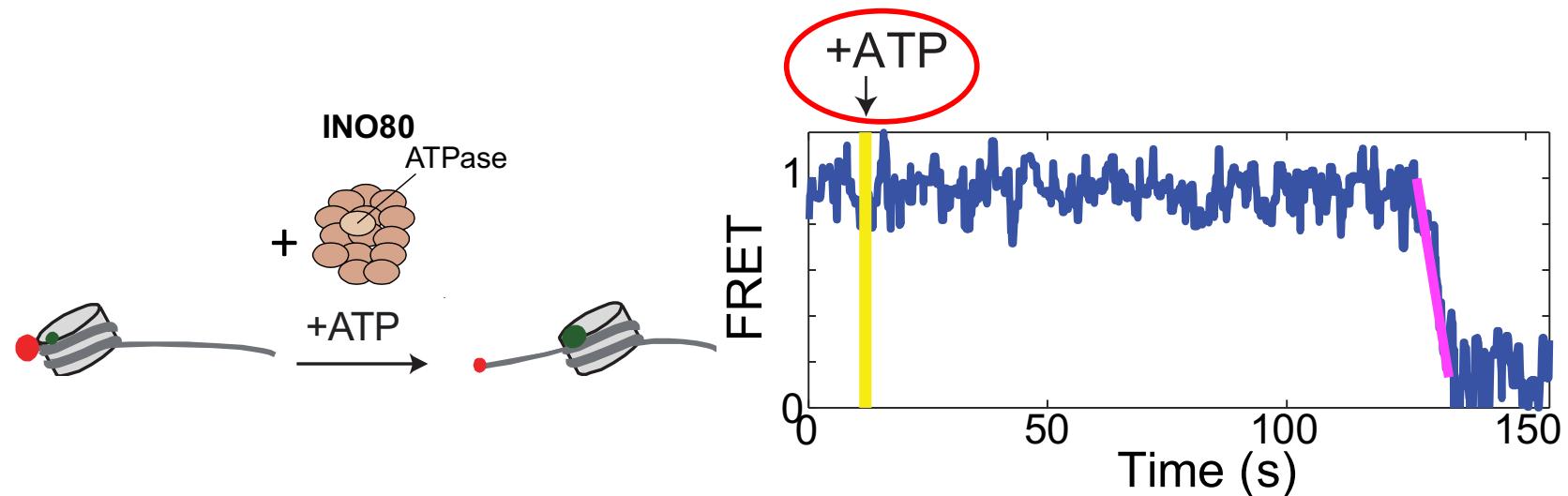
Posterior distribution \propto prior \times likelihood

We can now make inferences about the slope of the second phase of the reaction!

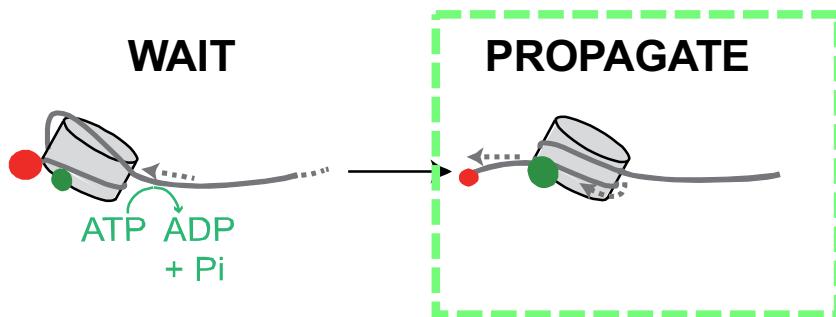


With this tool in hand: do the slopes of the “slopey bits” change with [ATP]?

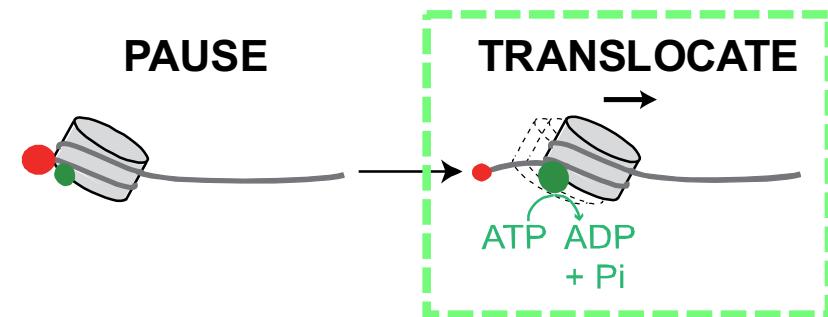
How does ATP concentration affect the rapid decrease in FRET?



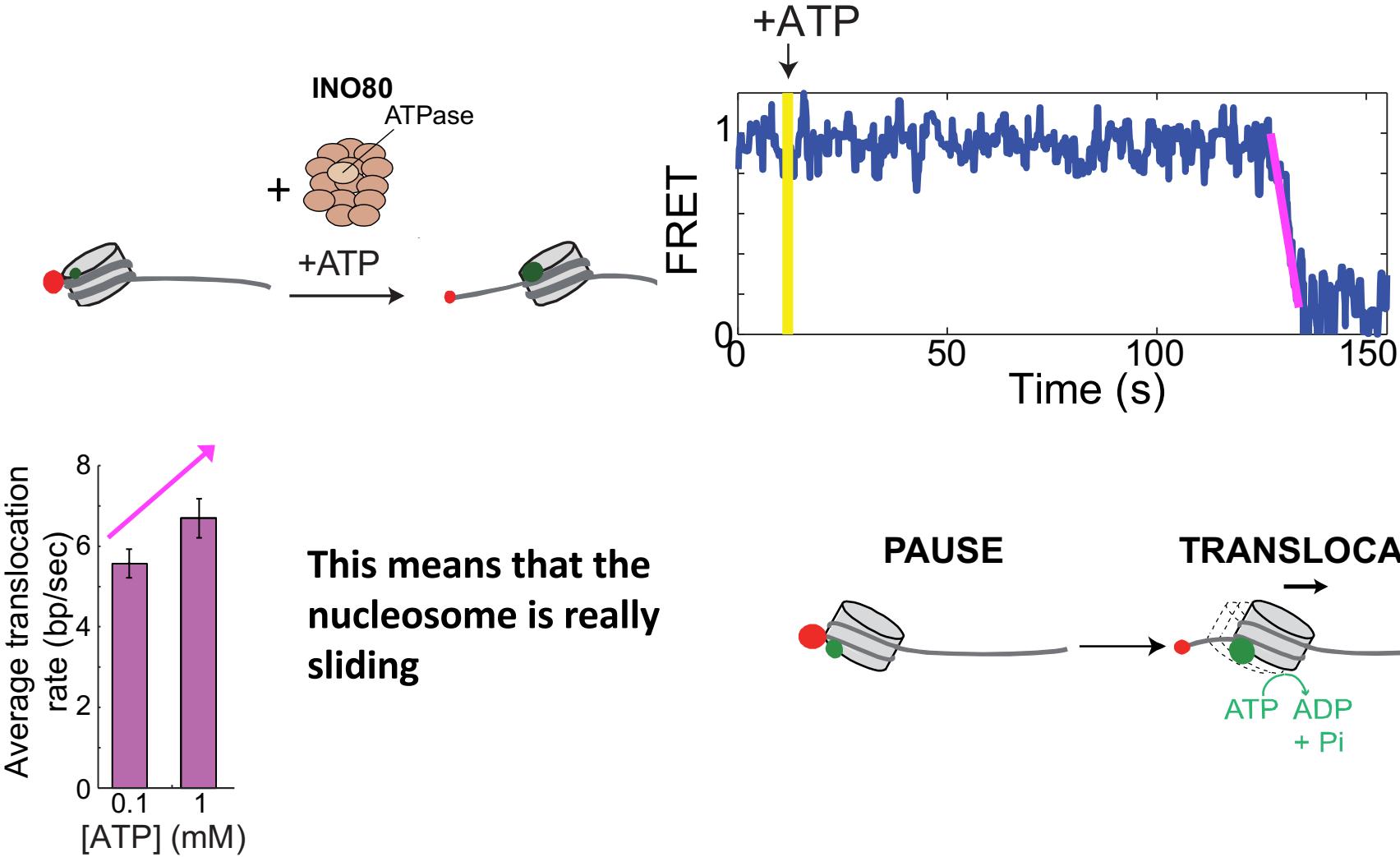
"Bulge" propagation?



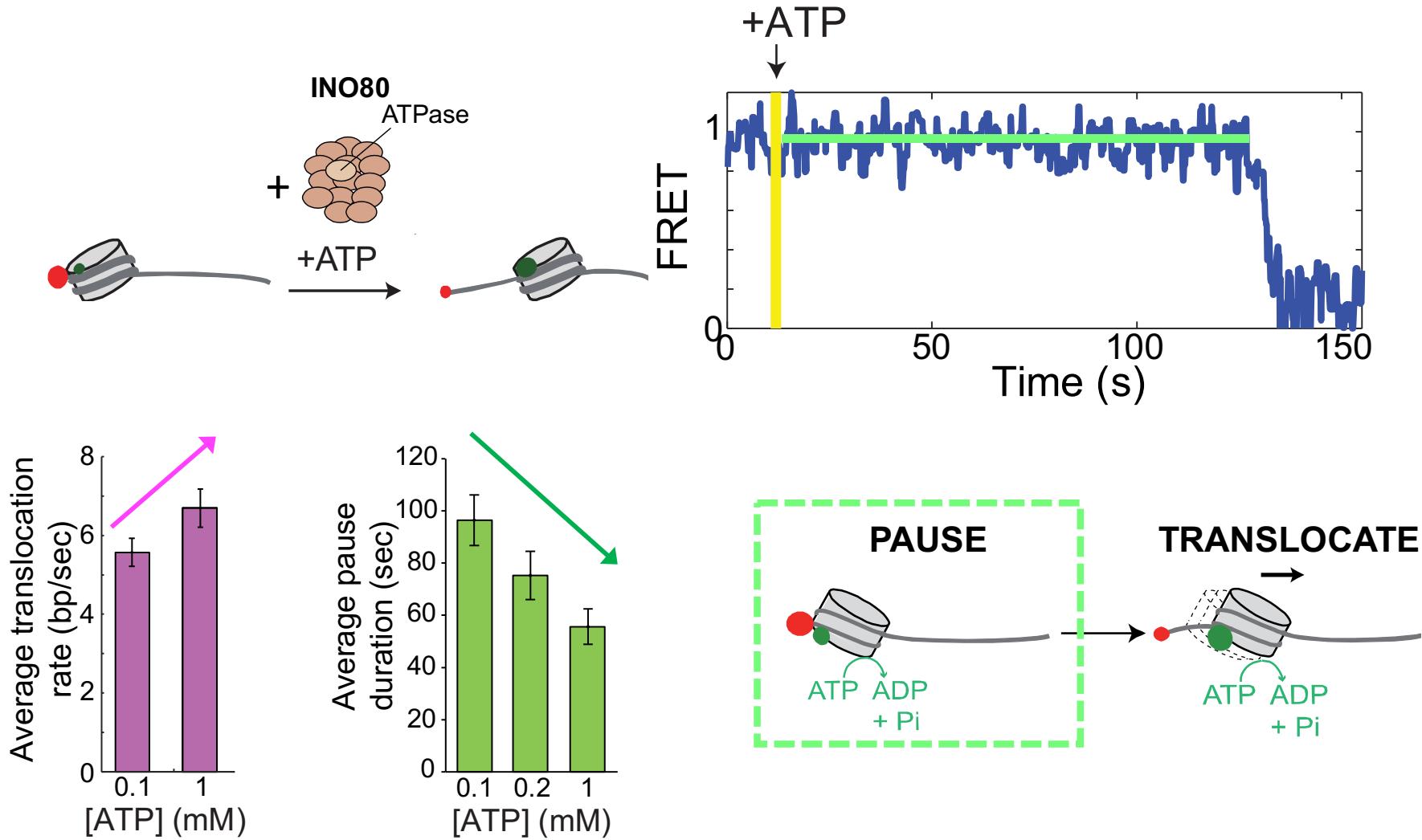
ATP-dependent translocation?



The rate of the translocation phase is ATP concentration dependent

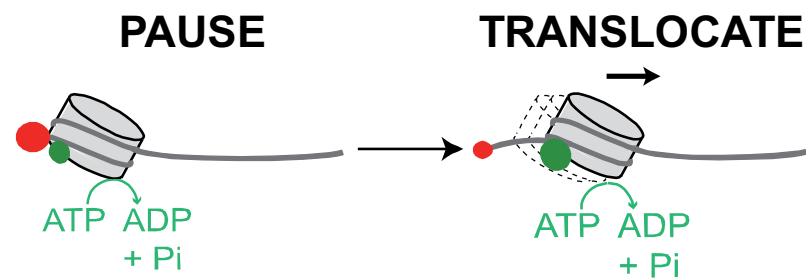
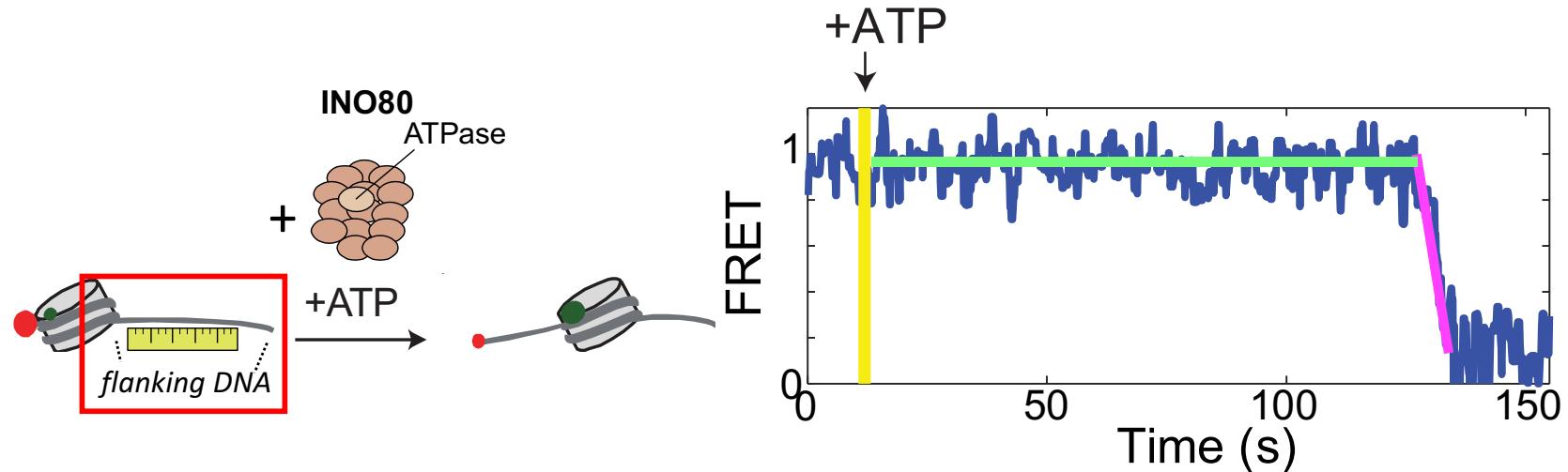


But the pause phase is also ATP-dependent!



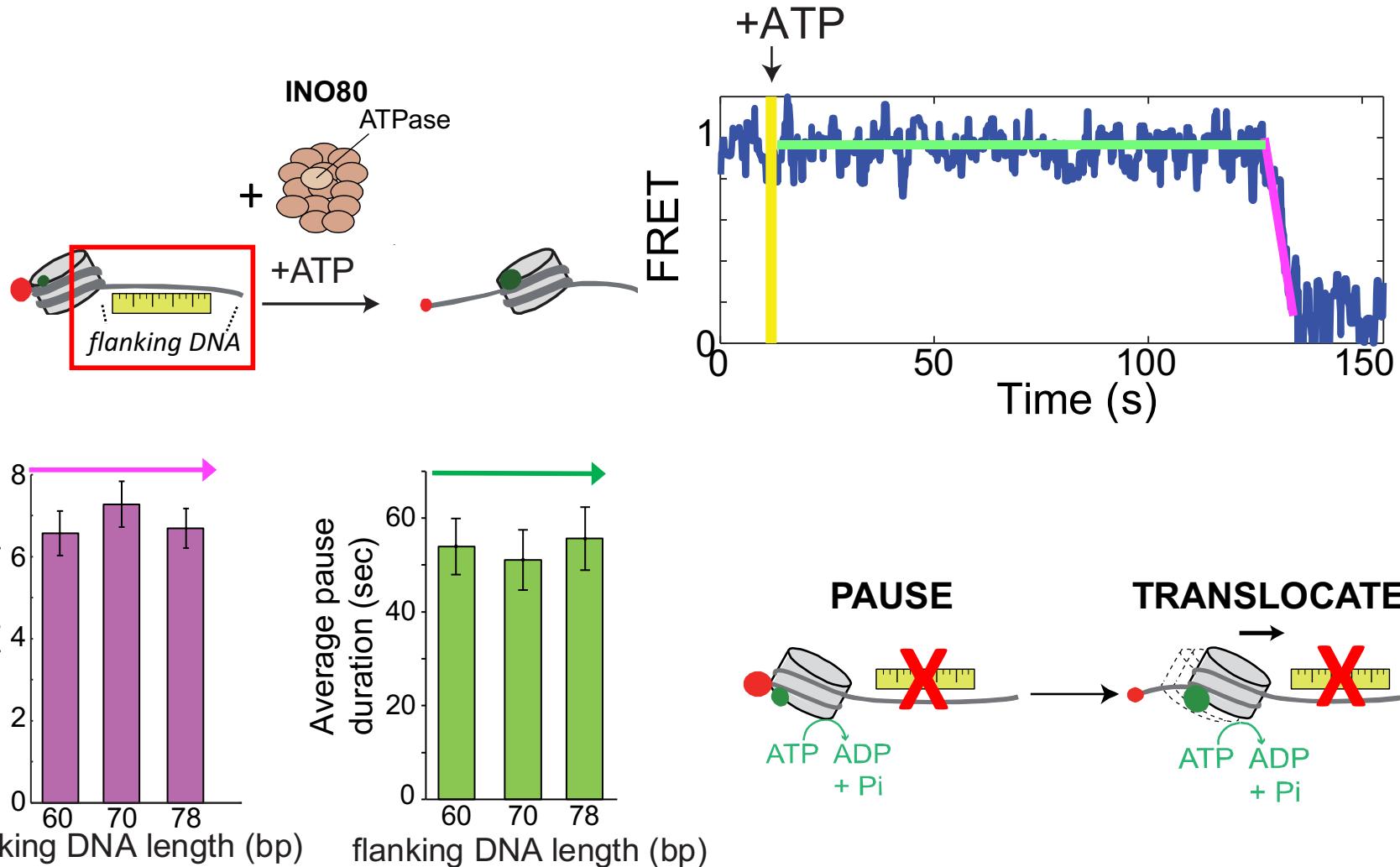
Is the initial pause a regulatory event?

If so, maybe it is sensitive to substrate cues that should regulate the reaction:



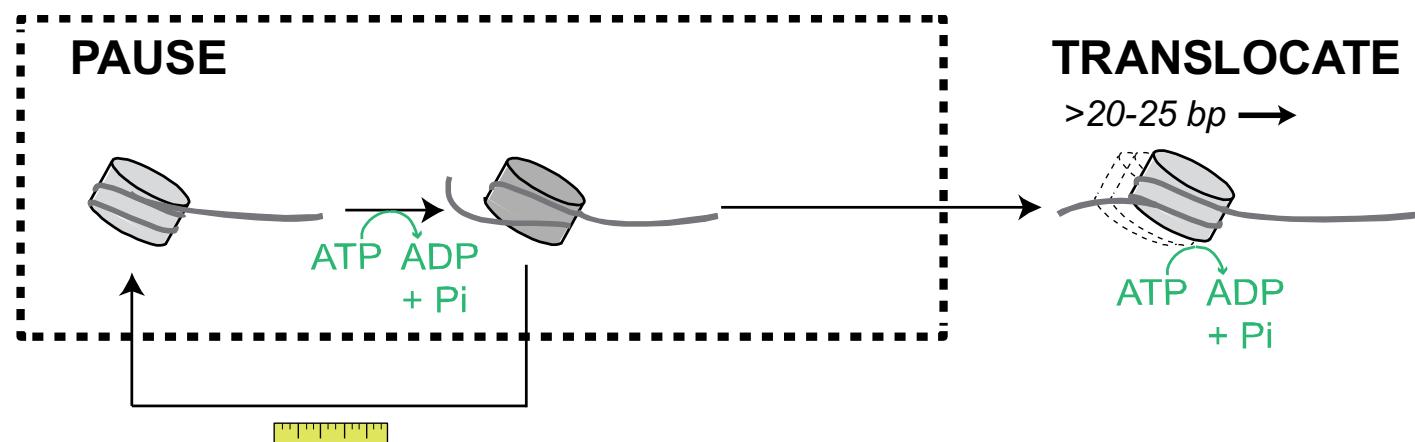
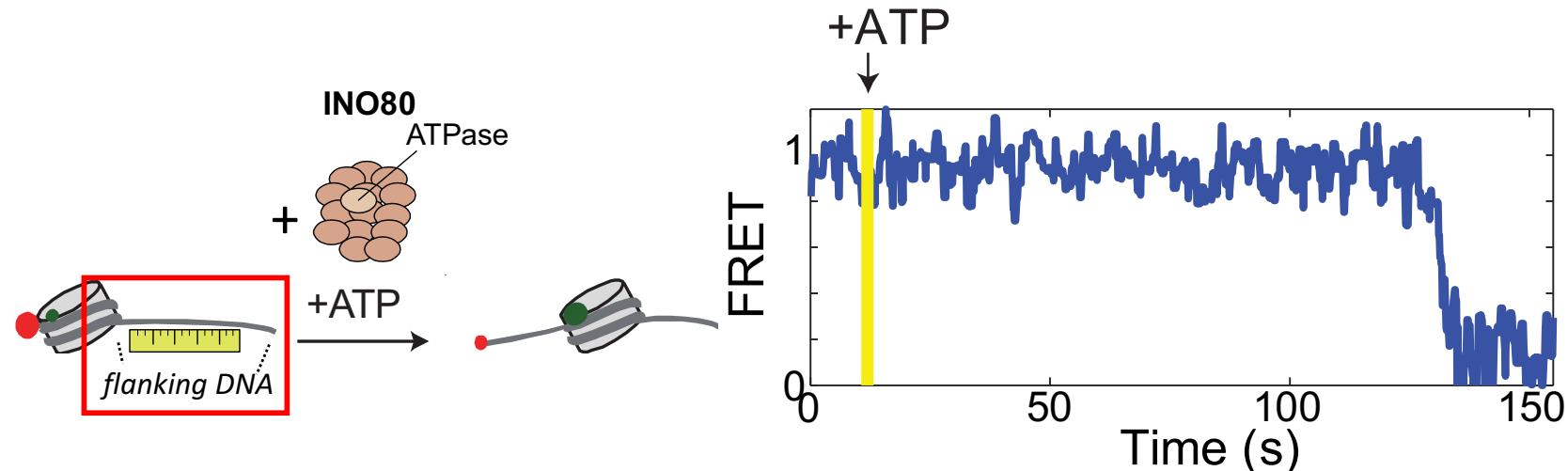
Is the initial pause a regulatory event?

Neither the pause nor the translocation phase is sensitive to flanking DNA length.



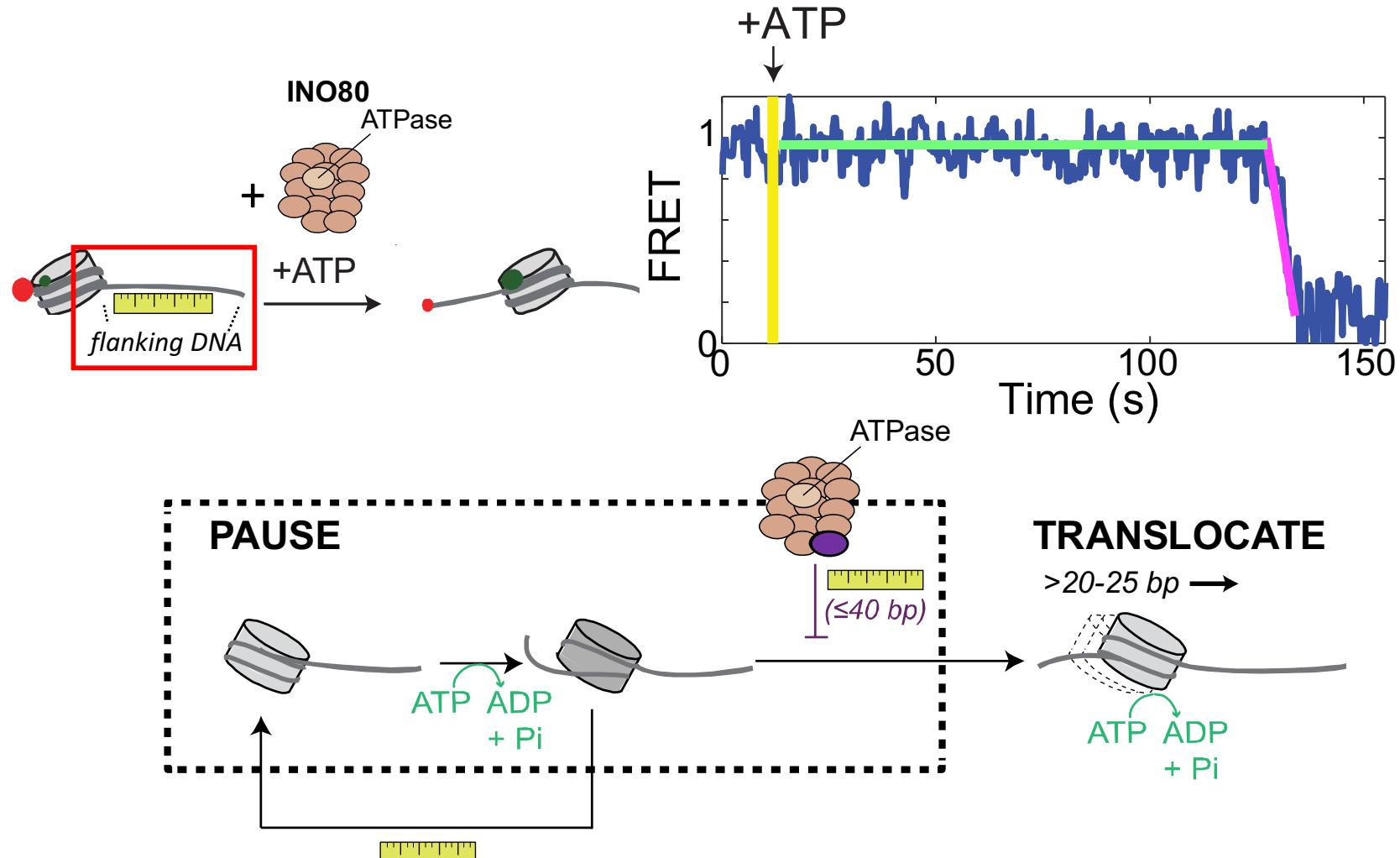
The initial pause is actually a regulatory event

But with a more complicated relationship to substrate cues than we thought!

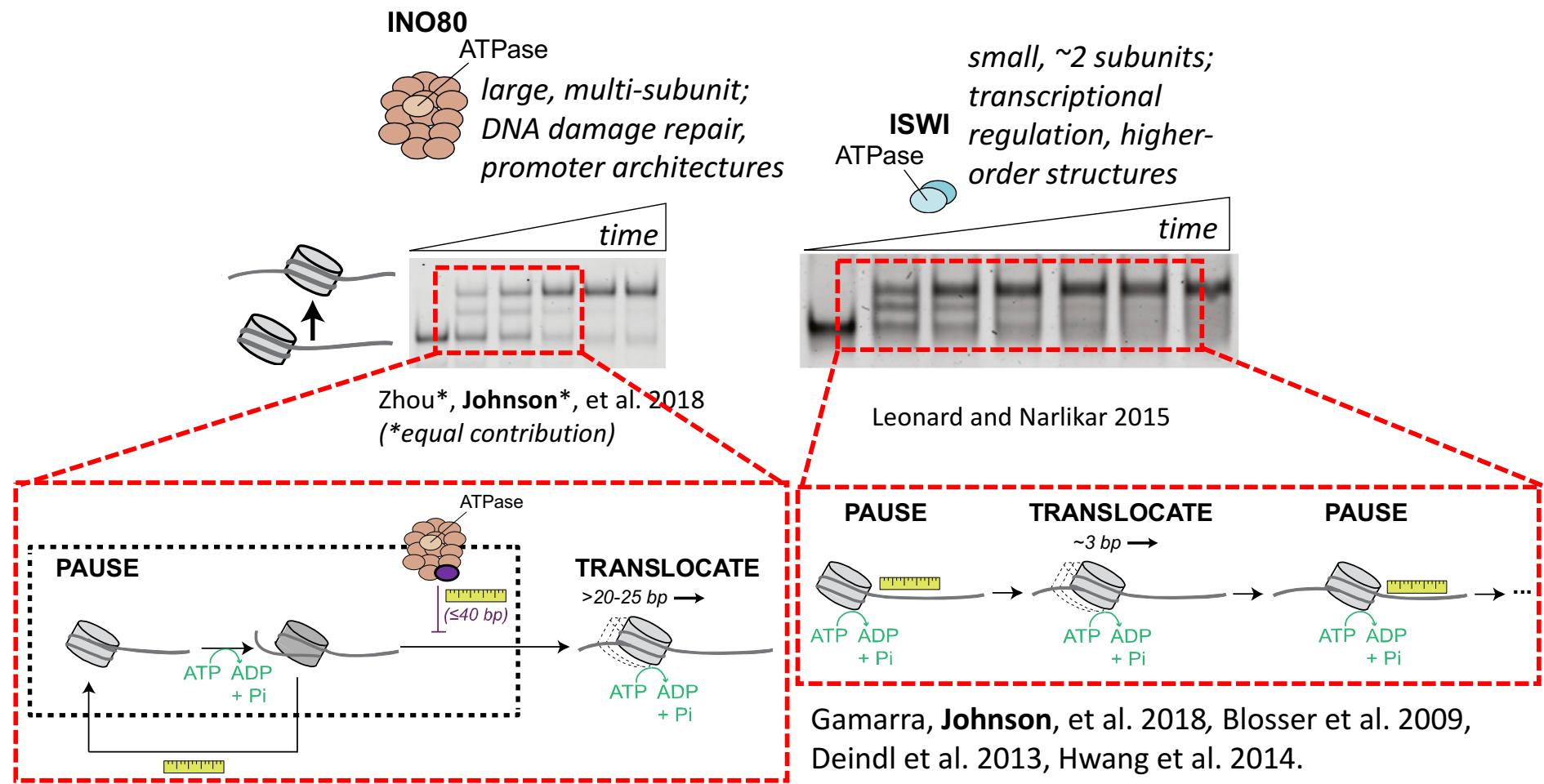


The initial pause is actually a regulatory event

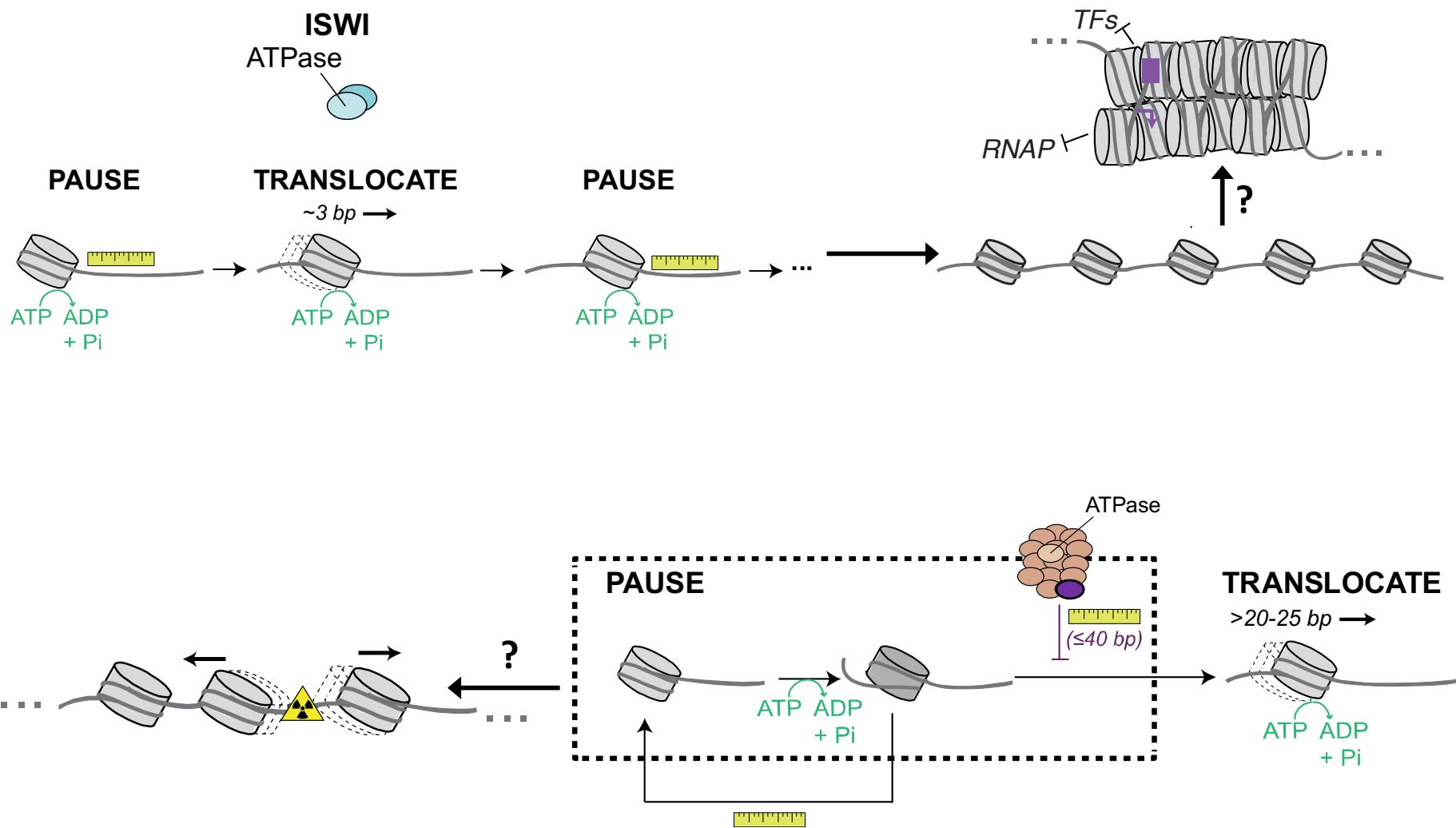
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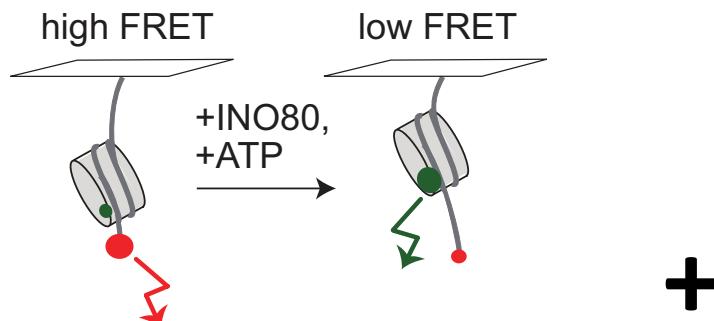
INO80's mechanism differs significantly from previously described sliding mechanisms



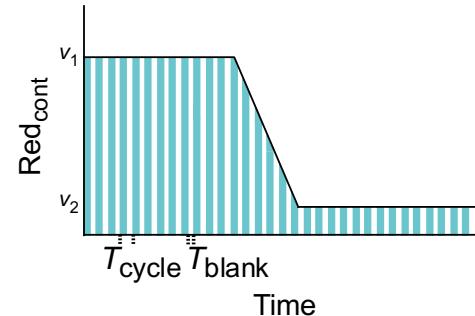
Two distinct mechanisms = specialization for particular in vivo roles?



Conclusions

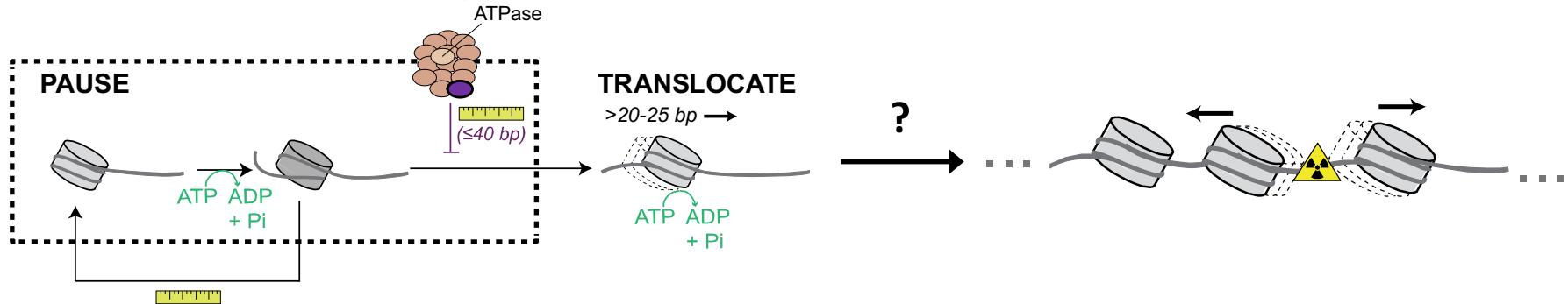


Time series data
from microscopy assay
with single molecule resolution



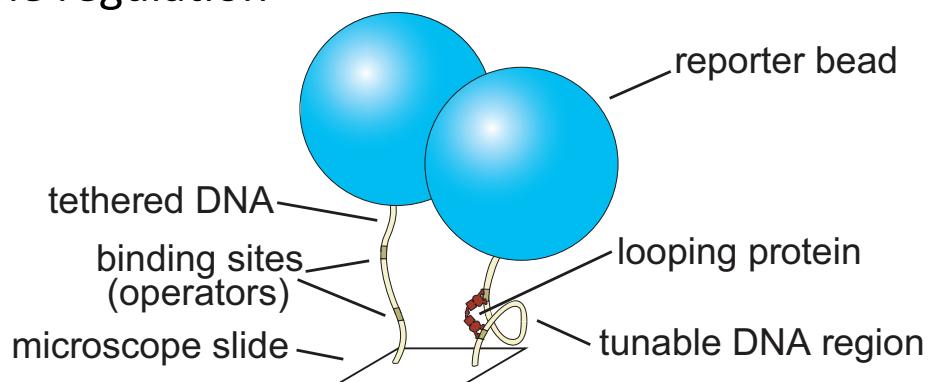
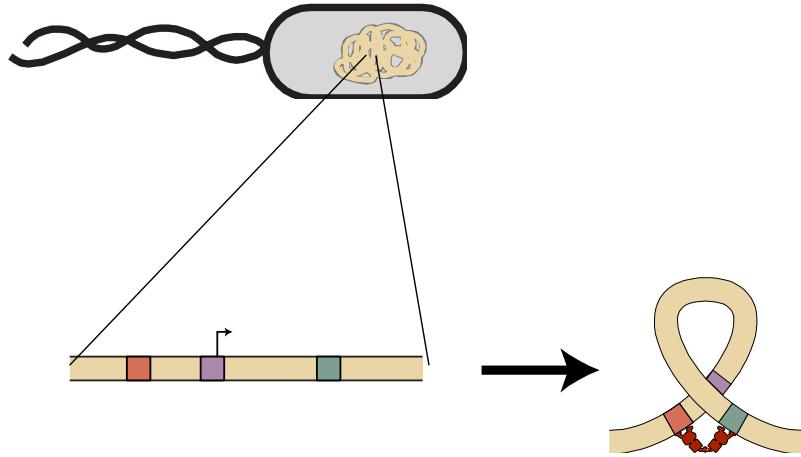
Bayesian inference
with explicit camera
modeling

= mechanistic insights into an enigmatic remodeler that were not obtainable by conventional biochemistry



Bayesian modeling of other microscopy-derived time series data

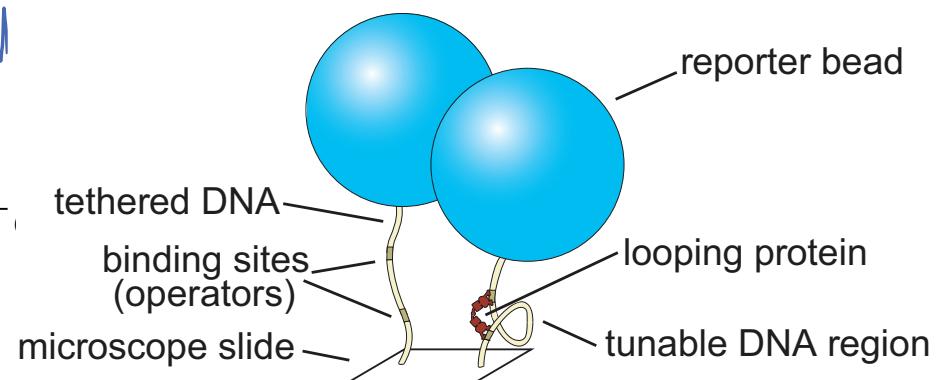
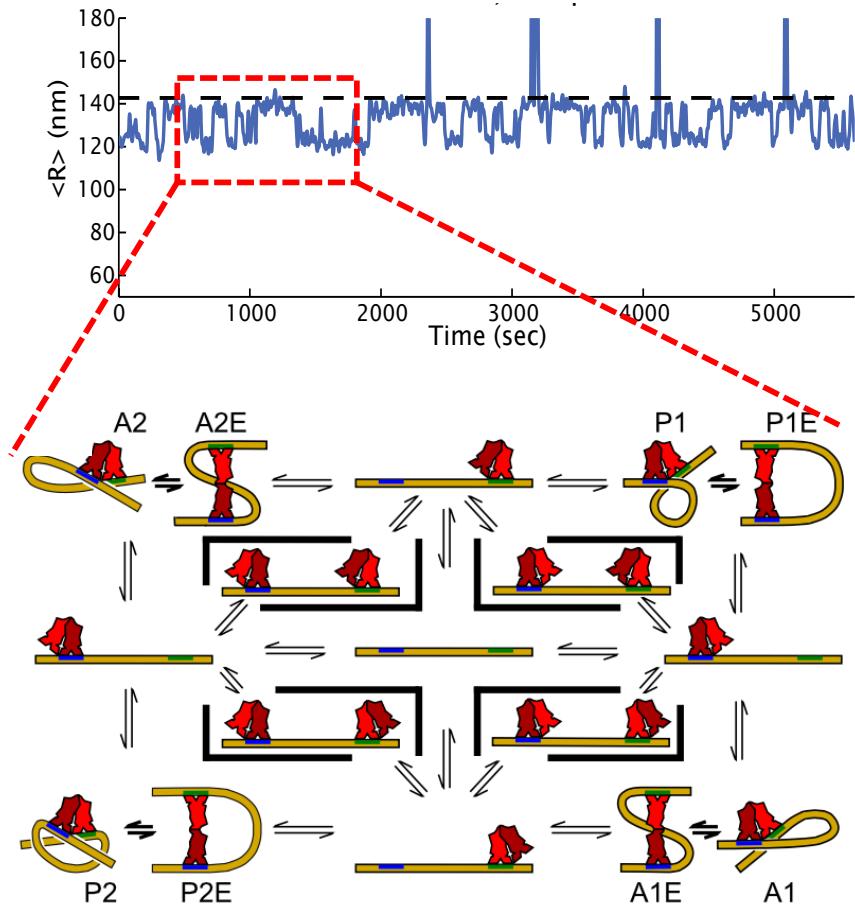
Ph.D. work with Rob Phillips at Caltech:
dynamics of DNA looping in bacterial gene regulation



DNA looping plays essential roles at many bacterial promoters,
but it's another hard process to study biochemically

Johnson et al. 2014; Johnson et al. 2012;
Boedicker, Garcia, **Johnson**, and Phillips 2013;
Johnson*, Chen*, and Phillips 2013;
Chen*, **Johnson***, and Phillips 2013 (*equal contribution (*equal contribution)).

Bayesian modeling of other microscopy-derived time series data



An HMM implemented with Bayesian inference revealed (many!) more loop configurations than previously anticipated



Narlikar lab

Geeta Narlikar

Coral Zhou

Laura Lee

Julia Tretyakova

Caitlin Stoddard

Serena Sanulli

John Leonard

Tracy Lou

Stefan Isaac

Madeline Keenan

Adam Larson

Kalyan Sinha

Nathan Gamarra

Elise Munoz

Muryam Gourdet

Emily Wong

Lucy Pill

Israel Saucedo

Ahmad Nabahn



University of California
San Francisco

Collaborators

Matthew Johnson (Google Brain)

Adam Longhurst (UCSF)

Sean Beckwith, Ashby Morrison
(Stanford)

