

# Maximum Flux Transition Paths of Conformational Change

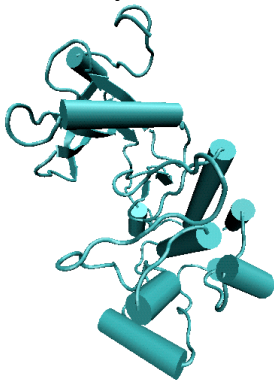
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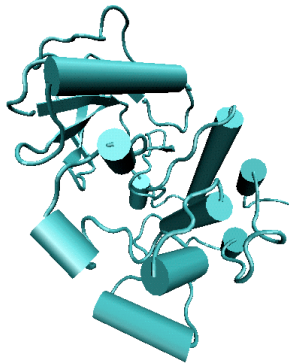
Acknowledgments: He Huang, Carol Post, NIH  
Voichita Dadarlat

# Src tyrosine kinase

active catalytic domain



inactive catalytic domain



## Message

We can do better than compute a minimum free energy path:  
find a path which intersects each isocommittor  
at that point through which there is  
the highest number of crossings of **distinct** reactive trajectories.

# Outline

- I. What is the problem?
- II. Three uncontrolled approximations
- III. An algorithm
- IV. Comparison

## What to compute

Given two metastable states  $A$  and  $B$  in configuration space, the problem is to find one or several “representative” reaction paths connecting them.

### Motivation:

calculating free energy differences,

finding intermediate meta-stable states

(targets for inhibitors of enhanced specificity)

# Problems vs. Algorithms

Two steps:

1. define the problem,
- 
2. construct an algorithm.

We follow the approach of Vanden-Eijnden, E, Ren, Ciccotti, . . .

# Dynamical equations

Consider a molecular system with potential energy function  $U(x)$

Assume Newtonian dynamics with mass matrix  $M$  and

initial values from a Boltzmann-Gibbs distribution:

initial  $x$  from probability density  $\rho(x) = \text{const } e^{-\beta U(x)}$

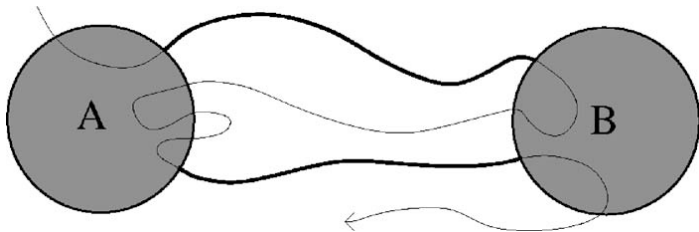
and  $(d/dt)x$  from a Maxwell distribution.

# An ensemble of paths

How to define an ensemble of transition paths from  $A$  to  $B$ :

Imagine an extremely long trajectory.

The trajectory enters and leaves  $A$  and  $B$  many times yielding a huge set of reactive paths from  $A$  to  $B$ , shown in dark in the figure below:



from Metzner, Schütte, and Vanden-Eijnden (2006)



## Defining a path

Rather than generate an ensemble of transition paths, which would have to be clustered anywhere, one might directly determine a concise description of the paths.

Specifically, if the paths cluster into one or several distinct isolated channels, one might compute the “center” of each cluster.

## Collective variables

Transition paths might not cluster adequately  
—in full configuration space.

Assume, however, there is a smaller set of *collective variables*,  
functions of the configuration  $x$ ,

$$\zeta_1 = \xi_1(x), \zeta_2 = \xi_2(x), \dots, \zeta_k = \xi_k(x), \quad \text{abbreviated as } \zeta = \xi(x),$$

such that in  $\zeta$ -space,  
paths cluster into one or several distinct isolated channels.

Else, there is little of interest to compute.

Our alanine dipeptide tests use phi and psi angles.

## Choice of collective variables

We want a minimal set of collective variables subject to two conditions:

- ▶ Coordinates  $\zeta$  must suffice to describe states  $A_\zeta$ ,  $B_\zeta$  in  $\zeta$ -space corresponding to  $A$ ,  $B$ .
- ▶ Coordinates  $\zeta$  must also be rich enough to “express the mechanism of conformational change” along the transition path.

To make the second condition more precise, introduce ...

# The committor

To measure the progress of a transition, there is a natural reaction coordinate, known as the *committor*.

For each point  $\zeta$ , consider a trajectory starting with random initial values **conditioned on**  $\xi(x) = \zeta$  and define the committor  $q(\zeta)$  to be the probability of reaching  $B_\zeta$  before  $A_\zeta$ :

$$q(\zeta) = \Pr(X(t) \text{ reaches } B_\zeta \text{ before } A_\zeta \mid \xi(X(0)) = \zeta).$$

# Expressing mechanism of change

The variables  $\zeta = \xi(x)$  are rich enough to express the mechanism of conformational change if the committor  $q(\zeta)$  has no local minima or maxima.

Else, there is some unexpressed DOF important to the transition.



from Dickson, Warmflash, and Dinner (2009)

## Defining a path

How to define the “center” of a cluster of paths in  $\zeta$ -space:

- most probable path

  - swarm-of-trajectories string method

- maximum flux path

  - our choice

- center of flux path

  - finite temperature string method

## Maximum flux path

A hypersurface  $\{\zeta \mid q(\zeta) = p\}$  of equal probability  $p$  is called an **isocommittor**. On each isocommittor consider the distribution  $j(\zeta)$  of crossing points for distinct reactive trajectories (last hitting points).

Seek the path  $\zeta = Z(s)$ ,  $0 \leq s \leq 1$ , which (locally) maximizes  $j(\zeta)$  on each isocommittor through which it passes.

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## Short Anecdote

...

Is it offensive  
to suggest that computational scientists are not in control  
of the errors that they are introducing?

Intractible.

## Uncontrolled approximation #1:

**separation of time scales.** Suppose there is some time interval  $\tau_{\text{rlx}}$   
(i) over which the collective variables change only a little, but  
(ii) during which all other degrees of freedom almost fully relax.

Hence, evolve the dynamics of  $\zeta(t) \stackrel{\text{def}}{=} \xi(x(t))$  as follows:  
Choose  $x(t)$  at random from  $\rho(x)$  conditioned on  $\xi(x(t)) = \zeta(t)$ .  
Choose  $(d/dt)x(t)$  at random from a Maxwell distribution.  
Determine  $x(t + \tau_{\text{rlx}})$ , from Newtonian dynamics.  
Set  $\zeta(t + \tau_{\text{rlx}}) = \xi(x(t + \tau_{\text{rlx}}))$ .

## Before stating the result

Define

$$\exp(-\beta F(\zeta)) = \text{const} \langle \delta(\xi(x) - \zeta) \rangle,$$

$$\langle O(x) \rangle_{\xi(x)=\zeta} = \frac{\langle \delta(\xi(x) - \zeta) O(x) \rangle}{\langle \delta(\xi(x) - \zeta) \rangle},$$

$$D(\zeta) = \frac{\tau_{\text{rlx}}}{2\beta} \langle \xi_x(x) M^{-1} \xi_x(x)^T \rangle_{\xi(x)=\zeta},$$

and  $D_{1/2} D_{1/2}^T = D$ .

Assumptions (i) and (ii) imply that approximately

$$\begin{aligned}\zeta(t + \tau_{\text{rlx}}) &= \zeta + \sqrt{2\tau_{\text{rlx}}} D_{1/2}(\zeta) N(0, 1)^k \\ &\quad + \tau_{\text{rlx}} (-\beta D(\zeta) \nabla F(\zeta) + (\nabla \cdot D(\zeta))^T) + \mathcal{O}(\tau_{\text{rlx}}^{3/2})\end{aligned}$$

where  $\zeta = \zeta(t)$ . This is the Euler-Maruyama discretization for stochastic dynamics and assumption (i) implies that  $\zeta(t)$  approximately satisfies Brownian dynamics (BD) equations

$$\frac{d}{dt} \zeta = -\beta D(\zeta) \nabla F(\zeta) + (\nabla \cdot D(\zeta))^T + \sqrt{2} D_{1/2}(\zeta) \frac{d}{dt} W(t).$$

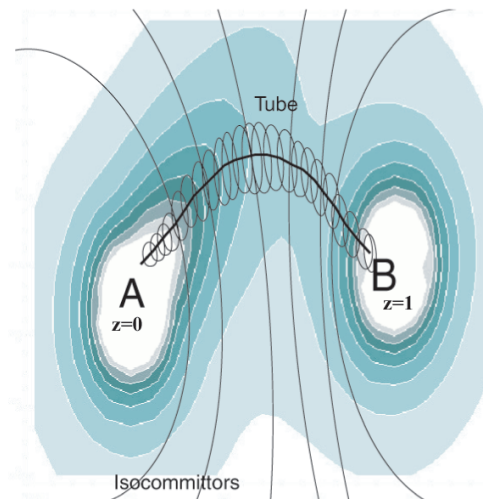
Validity of the assumptions might be checked a posteriori by comparing committor values of the Brownian dynamics to those of actual dynamics.

## The path of most probable points

It can be shown that *on an isocommittor* the distribution of last hitting points of reactive trajectories, as well as the net normal reactive flux, is given by

$$j(\zeta) = \text{const } e^{-\beta F(\zeta)} \nabla q(\zeta) \cdot D(\zeta) \nabla q(\zeta) / |\nabla q(\zeta)|.$$

An illustration follows.



The BD committor minimizes the functional

$$I(q) = \int e^{-\beta F(\zeta)} \nabla q(\zeta) \cdot D(\zeta) \nabla q(\zeta) d\zeta$$

subject to  $q(\zeta) = 0$  on the boundary of  $A_\zeta$  and  $q(\zeta) = 1$  on the boundary of  $B_\zeta$ .

## Uncontrolled approximation #2:

localized tube assumption.

Assume that regions of low  $F(\zeta)$  constitute a tube and that isocommittors are nearly planar there and that  $D(\zeta)$  is nearly constant on each plane.



## (Approximating the isocommittor)

Take for  $q(\zeta)$  an approximation constructed  
from  $q(Z(s))$  and  $\nabla q(Z(s))$ ,  $0 \leq s \leq 1$ ,  
by extrapolation.

Need solve only for  $k + 1$  functions of  $s$  to get committor.

## Uncontrolled approximation #3:

narrow tube assumption.

Assume that on each isocommittor

the probability is strongly peaked around path.

Then the probability flux of reactive trajectories is tangent to the path

$$\text{const } e^{-\beta F(Z)} D(Z) \nabla q(Z) \parallel Z_s.$$

where  $Z = Z(s)$  and  $Z_s = (d/ds)Z(s)$ .

result is a

## Maximum flux transition path

$$Z_s \parallel g, \quad g = -D(Z)\nabla F(Z) + \frac{1}{\beta} \frac{D(Z)(D(Z)^{-1}Z_s)_s}{Z_s^\top D(Z)^{-1}Z_s}.$$

## Uncontrolled approximation #4:

zero temperature assumption.

Neglect the term  $\frac{1}{\beta} \frac{D(Z)(D(Z)^{-1}Z_s)_s}{Z_s^\top D(Z)^{-1}Z_s}$ .

result is a

## Minimum free energy path

$$Z_s \parallel -D(Z)\nabla F(Z).$$

Free energy is minimized “orthogonal” to the path.

We can prove that the MFEP

has cusps at some intermediate local minima.

This undermines the localized tube assumption.

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# Controlled approximations

- ▶ discretization of path
- ▶ solution of nonlinear discrete equations
- ▶ sampling

## Discretization of path

Sequence of replicas for  $\zeta = Z_j$ ,  $j = 0, 1, \dots, J$ .

Upwinded differencing for  $(Z_s)_j$

based on direction of modified mean force  $g_j$

Normalization:  $(|Z_s|)_s = 0$ .

MFEP would have cusps at some intermediate local minima,  
which requires adaptive discretization methods.



## Solution of nonlinear discrete equations

For large systems, targeted MD has been used to get initial path. Simplified string method is good for refining it:

1.  $Z_j^* = Z_j + \tau g_j$
2. choose the  $Z_{j+1}$  to be equidistant along the resulting curve

$$(\tau_{\text{rlx}}\tau)^{1/2} = 48.89 \text{ fs}$$

Number of iterations = 50.

# Sampling

Strong harmonic restraints are good for constrained sampling.

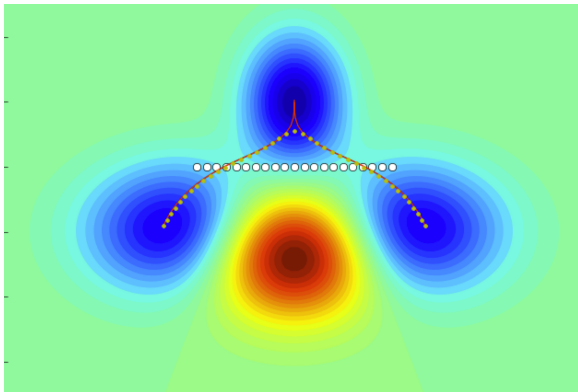
Our alanine dipeptide simulations use  
force constant  $K = 1000 \text{ kcal/mol/rad}^2$ ,  
Langevin dynamics with friction coefficient  $10/\text{ps}$  on all atoms,  
timestep =  $1 \text{ fs}$ ,  
 $10 \text{ ps}$  equilibration,  $100 \text{ ps}$  production.

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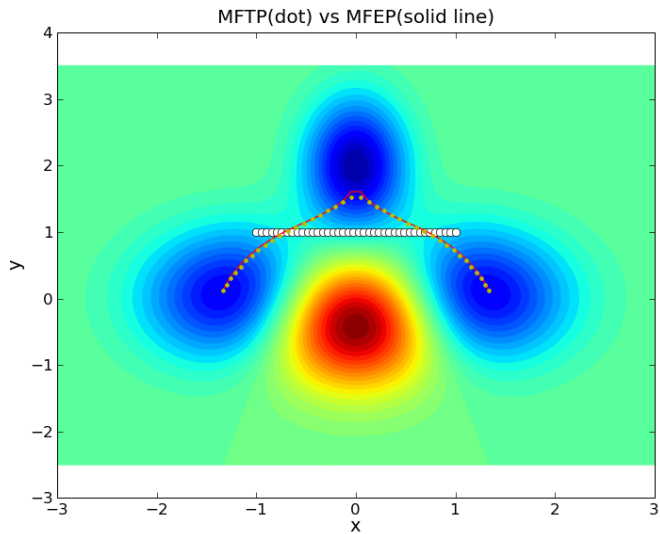
The following figure compares MEP (having a cusp) and MFTP for the potential energy function

$$U(x, y) = -4 \exp(-4x^2 - (y - 2)^2) - 5 \exp(-(x - 1)^2 - y^2) \\ - 5 \exp(-(x + 1)^2 - y^2) + 8 \exp(-x^2 - (y + \frac{1}{4})^2).$$



Contour plot of potential energy, white circles are initial string, yellow dots are MFTP, and red line is MEP.

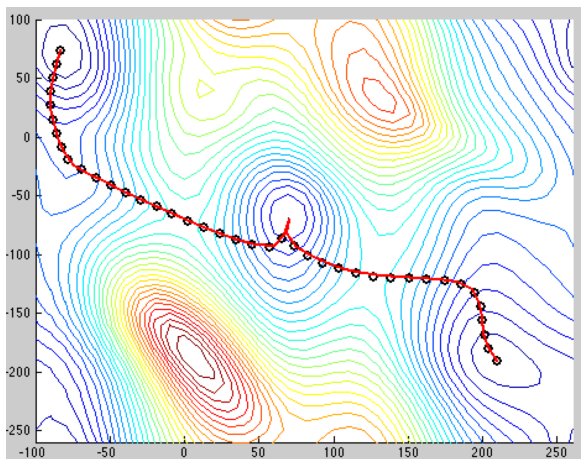
The cusp of MEP/MFEP is hard to compute. For example, the cusp will be missed if there are 40 replicas along the string rather than 41 as shown below:



The next figure compares MFEP and MFTP for alanine dipeptide in vacuo at  $T = 300$  using CHARMM22 force field.

MPI for Python + CHARMM  
hours of CPU time on 8 cores





Contour plot of potential energy in  $\varphi$  and  $\psi$  torsion angles, black circles are MFTP, and red line is MFEP.

## Conclusion

The maximum flux transition path (MFTP) involves one less approximation than the minimum free energy path (MFEP).

The MFEP has cusps, which makes it

- unsuitable for defining an isocommittor,
- unsuitable for defining a reaction coordinate, and
- harder to compute.