

Molecular Biophysics III - DYNAMICS

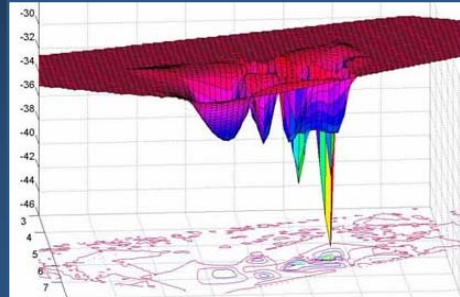
References

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- ***Computational Molecular Dynamics: Challenges, Methods, Ideas*** LNCSE (Lecture Notes In Comp Sci & Eng) Series, Eds: Griebel, Keyes, Neimien, Roose, Schlick, Springer-Verlag, 1999.
- ***Protein Physics*** A.V. Finkelstein and O. G. Ptitsyn, Academic Press, 2002.
- ***A Theoretical Perspective of Dynamics, Structure, and Thermodynamics***. C. L. Brooks, III, M. Karplus, and B.M. Pettitt. Wiley Interscience, New York, 1988.
- ***Biophysical Chemistry***. C.R. Cantor and P.R. Schimmel. Vol.1,2,3. W.H. Freeman and Company, San Francisco, 1980.
- ***Structure and Mechanism in Protein Science: A Guide to Enzyme Catalysis and Protein Folding***. A. Fersht. W. H. Freeman and Company, New York, 1999.
- ***Molecular Modelling. Principles and Applications***. A. R. Leach. Addison Wesley Longman, Essex, England, 1996.
- ***Dynamics of Proteins and Nucleic Acids***. J.A. McCammon and S.C. Harvey. Cambridge University Press, Cambridge, 1987.
- ***Physics With Illustrative Examples from Medicine and Biology – Statistical Physics*** Benedek & Villars, Springer-Verlag, 2000

‘Protein folding problem’

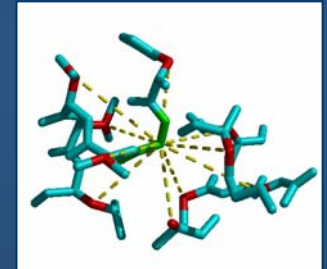
Sequence -----> Structure

Bioinformatics. Sequence alignments



```
PROTEIN Sequence in any format:
>mouse protein sequence
MQIIEPQVQY NYVYDEEYH IQEEVDRDL LLDPAWEKQQ
RKTPTANCNS HLRKAGTQIE NIEEDFRNL KMLLLEVIS
GERLPRPDG KMRFKIANV NKALDYASK GVKLVSIGAE
EIVDGNVKMT LGHIWTLIR FAIQDISVEE TSAKEGLLLW
CQRKTAPEYN VNIQNFHTSW KDGLGLCALI HRHRPDLIDY
SKLNKDDPIG NINLAMEIAE KHLDIPEMLD AEDIVNTPKP
DERAINTYVS CFYHAFAGAE QAETAANRIC KGLAVNQENE
RLMEEVERLA SELLEWIRRT IPWLENRTP EKTQAMQKKL
EDFRDYRRKH KPPKVQEKCO LEINFNTLQT KLRISNRAAF
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Modelling and simulations



Function

Fundamental paradigm: Sequence encodes structure; structure encodes function

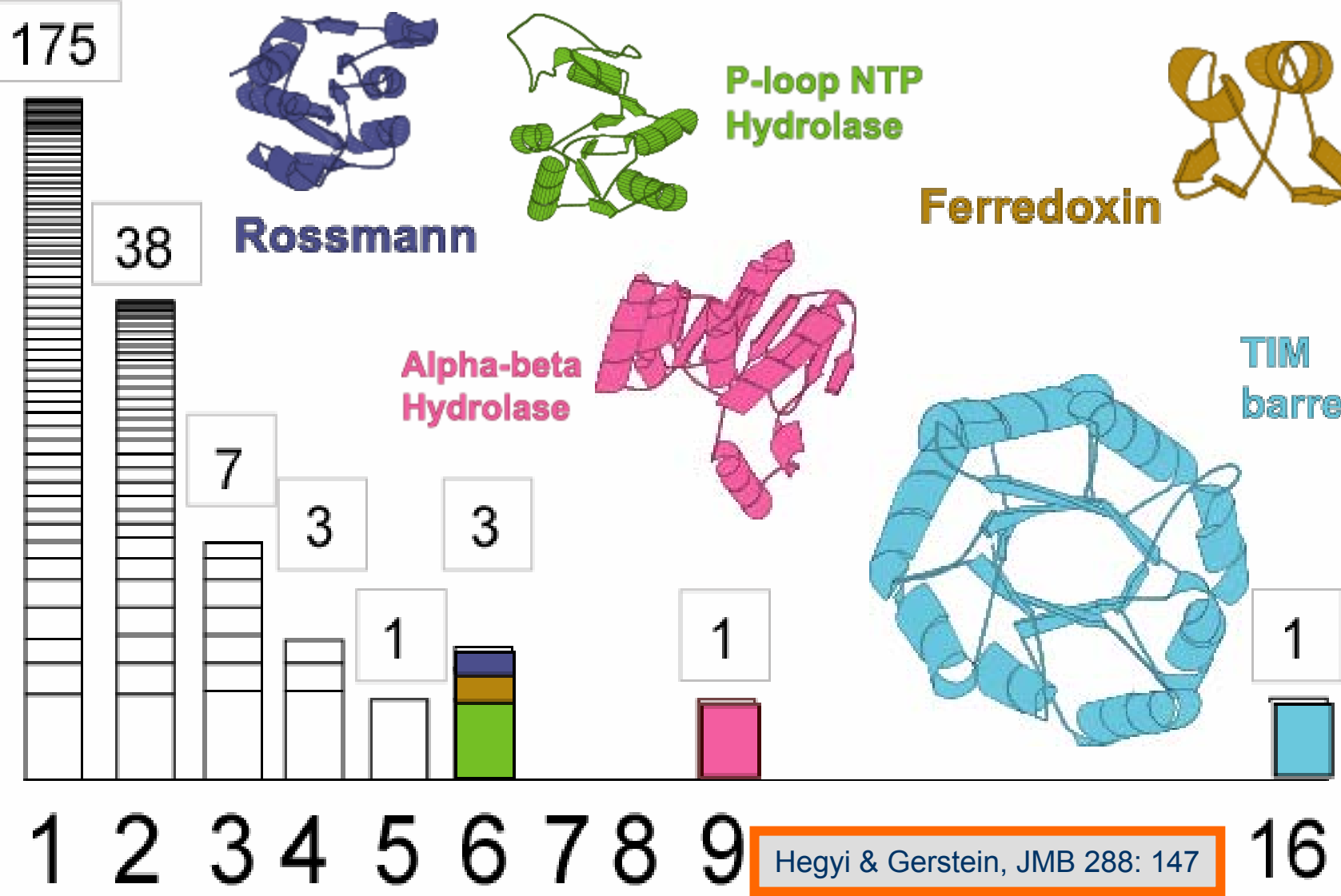
Structure → Function

**To what degree is fold
associated with function?**

Structure → **Dynamics** → Function

Is there a unique relation between fold and function?

Frequency in database of 229 folds



Number of functions associated with a fold

Hegy & Gerstein, JMB 288: 147

Knowledge of sequence or structure does not permit us to

- Understand the mechanism of function
- Devise methods of controlling/inhibiting function
- Predict the behavior in different forms, different environments
- Answer the questions 'how' or 'why'!

Biological function is a dynamic process

Structure → **Dynamics** → Function

To understand the principles that underlie the passage

from structure,



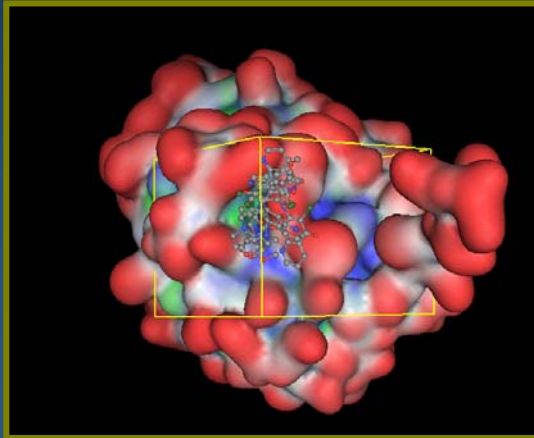
to function...

We need to examine the conformational dynamics.



Dynamics \leftrightarrow Function

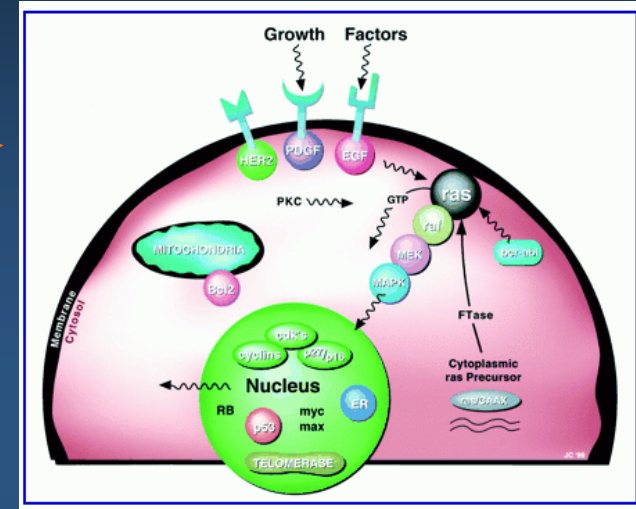
State-of-the-art in computational/mathematical biology



← ----- 25 Å ----- →

Molecular computations

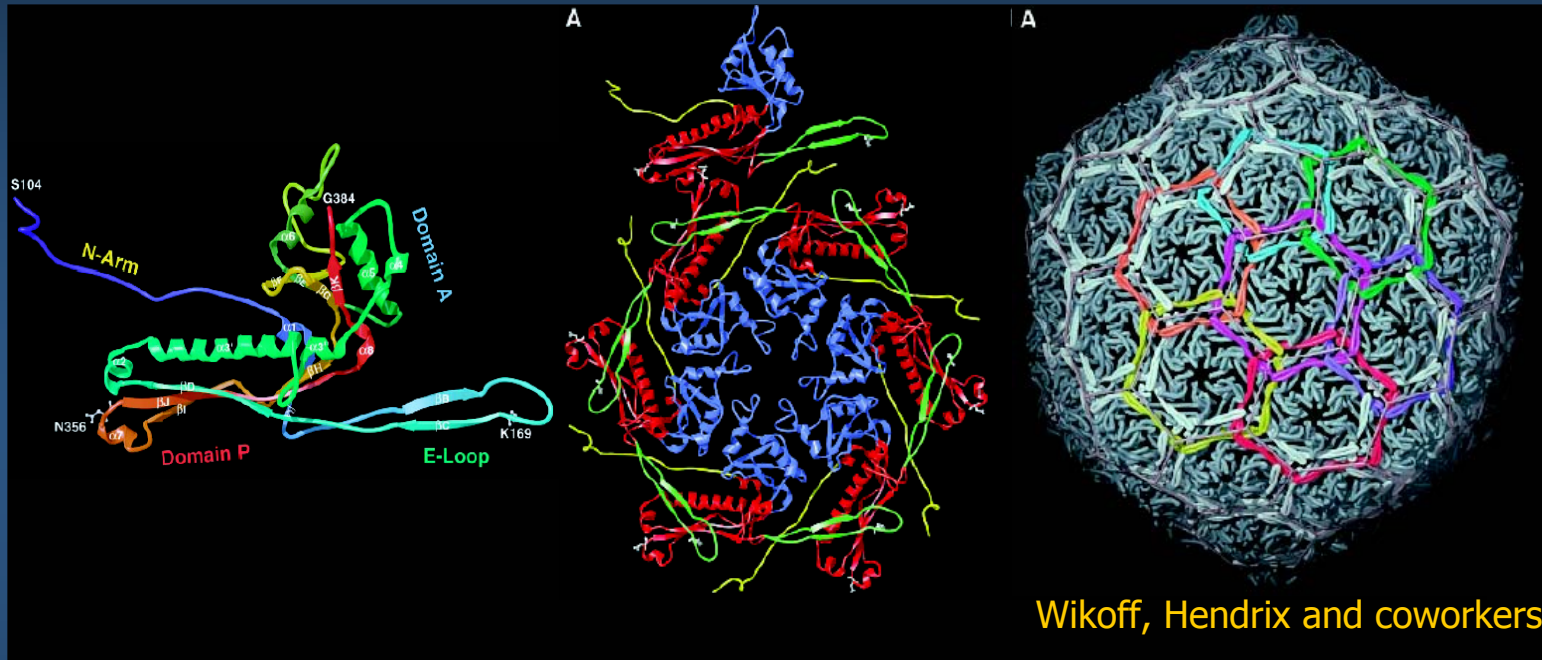
- Limited to small systems (one macromolecule) or short times (~ ns)
- Dependent on force field
- Solvent effect – a problem



Subcellular/cellular computations

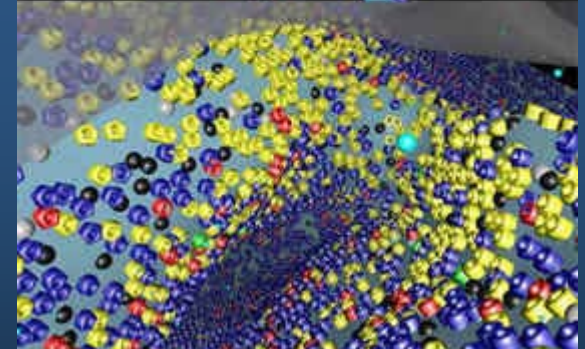
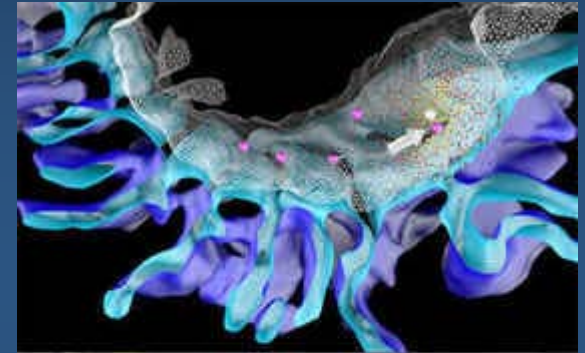
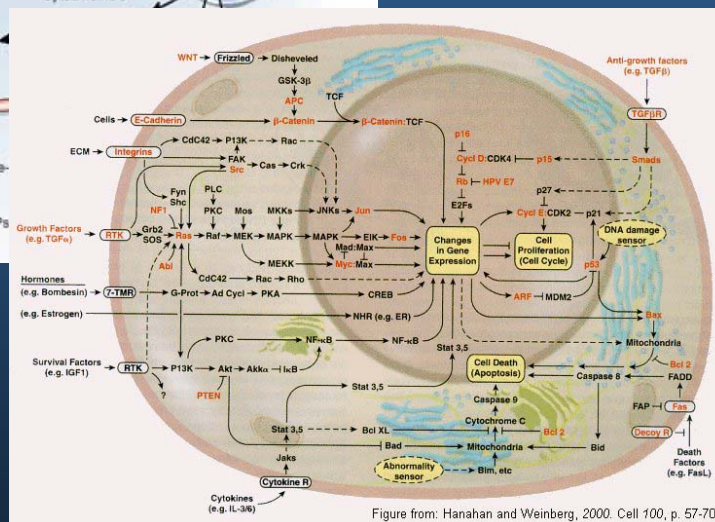
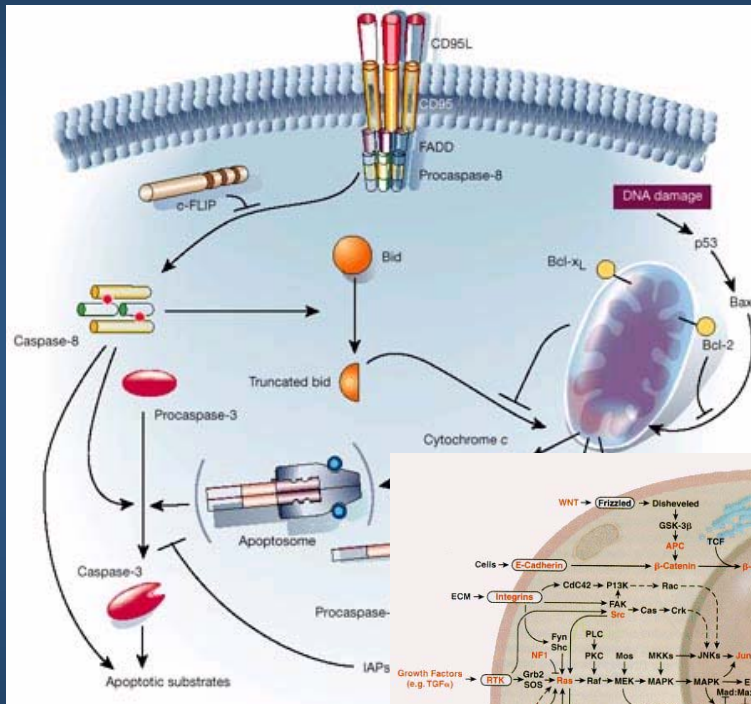
- Simple mass-action kinetics
- No spatial-structural realism
- Lack of data for model parameters

Supramolecular dynamics



Multiscale modeling – from full atomic to multimeric structures

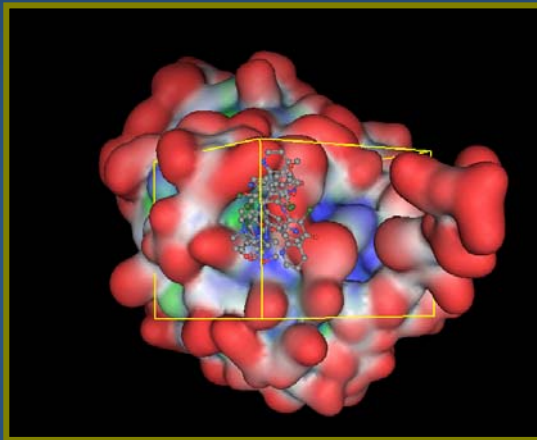
Subcellular and cellular simulations



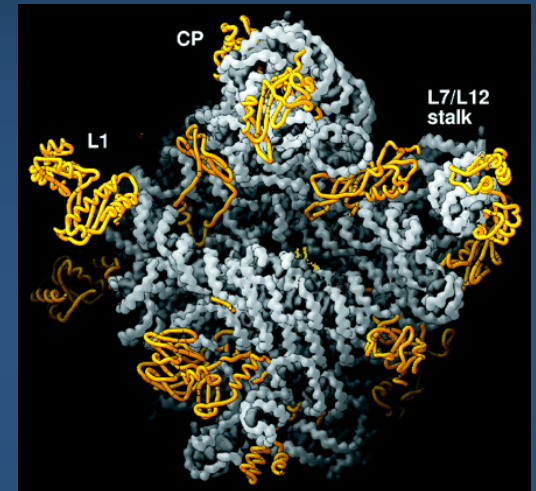
Monte Carlo simulations using MCell and DReaMM - Stiles and coworkers (PSC)

Dynamics systems - Mathematical modeling using ODEs

Progresses in molecular approaches: Coarse-grained approaches for large complexes/assemblies

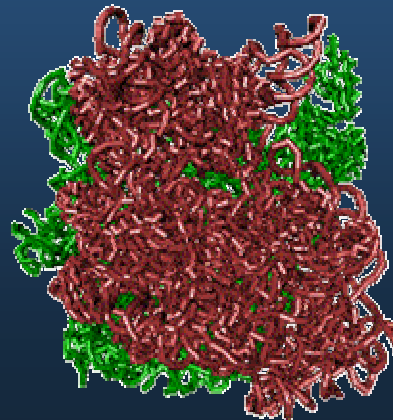


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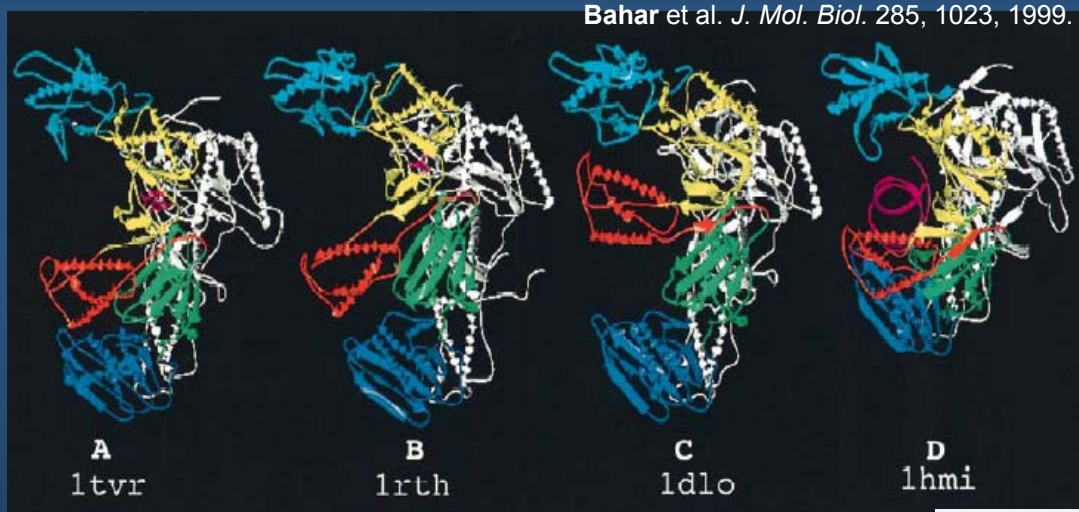
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Example: EN models for modeling
ribosomal machinery (Frank et al,
2003; [Rader et al., 2004](#))

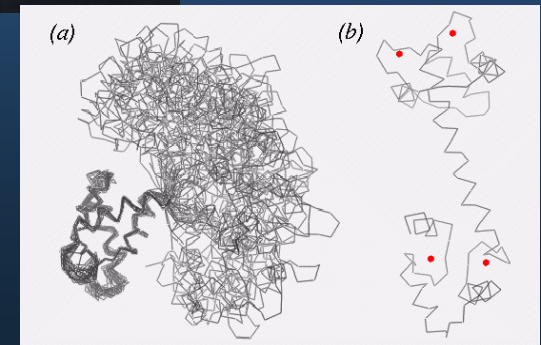


Structures suggest mechanisms of function

A. **Comparison of static structures** available in the PDB for the same protein in different form has been widely used as an *indirect* method of inferring dynamics.



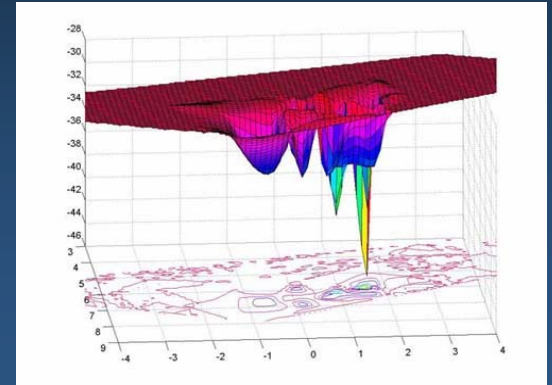
B. **NMR structures** provide information on fluctuation dynamics



Protein dynamics

● Folding/unfolding dynamics

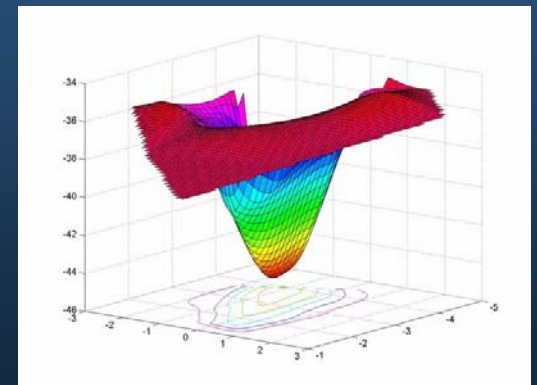
- Passage over one or more energy barriers
- Transitions between infinitely many conformations



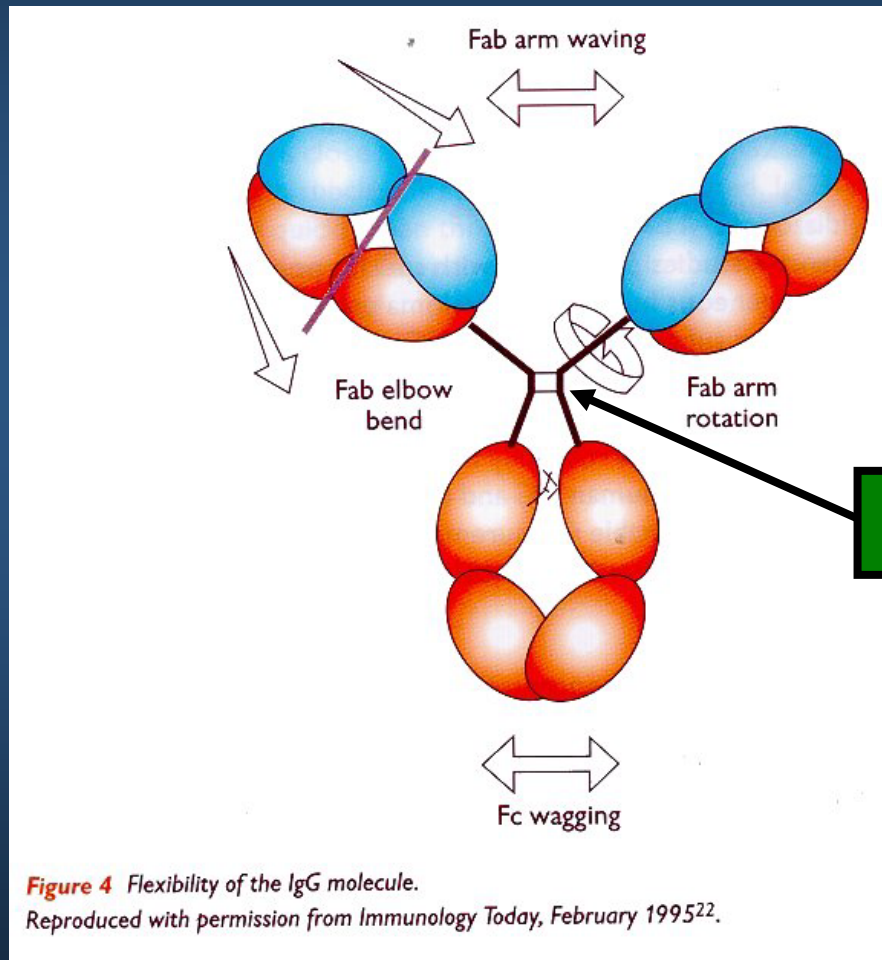
B. Ozkan, K.A. Dill & I. Bahar, *Protein Sci.* 11, 1958-1970, 2002

● Fluctuations near the folded state

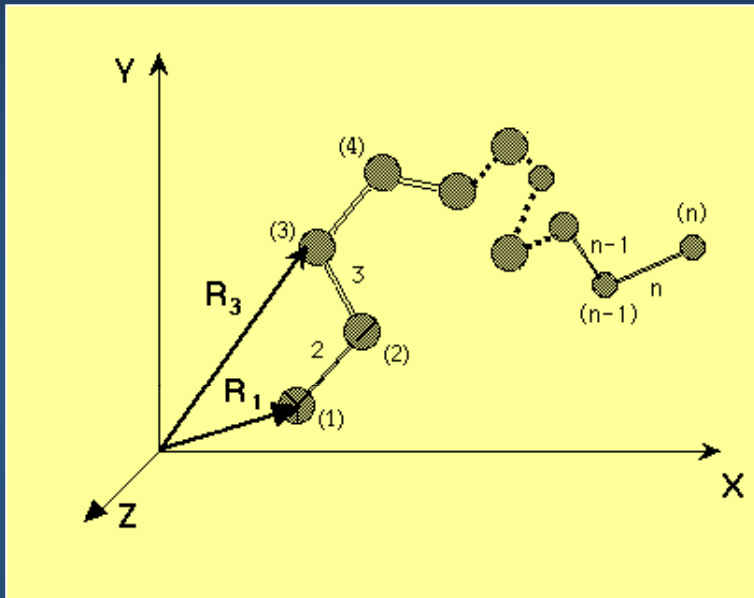
- Local conformational changes
- Fluctuations near a global minimum



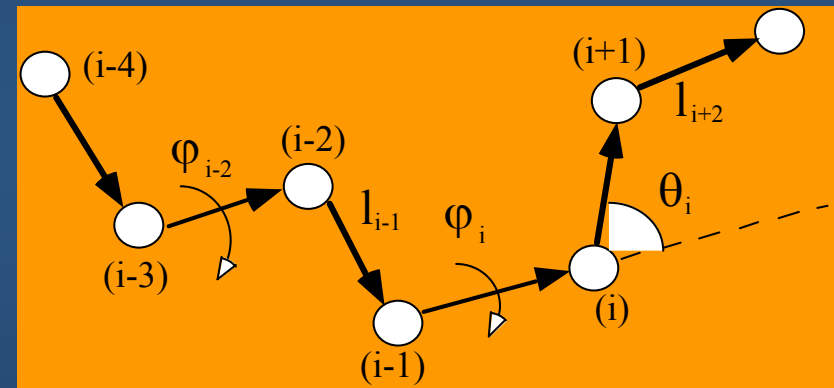
Several modes of motions in native state



Macromolecular Conformations



Schematic representation of a chain of n backbone units. Bonds are labeled from 2 to n , and structural units from 1 to n . The location of the i th unit with respect to the laboratory-fixed frame OXYZ is indicated by the position vector R_i .



Schematic representation of a portion of the main chain of a macromolecule. l_i is the bond vector extending from unit $i-1$ to i , as shown. φ_i denotes the torsional angle about bond i .

How/why does a molecule move?

Among the $3N-6$ internal degrees of freedom, **bond rotations** (i.e. changes in dihedral angles) are the softest, and mainly responsible for the functional motions

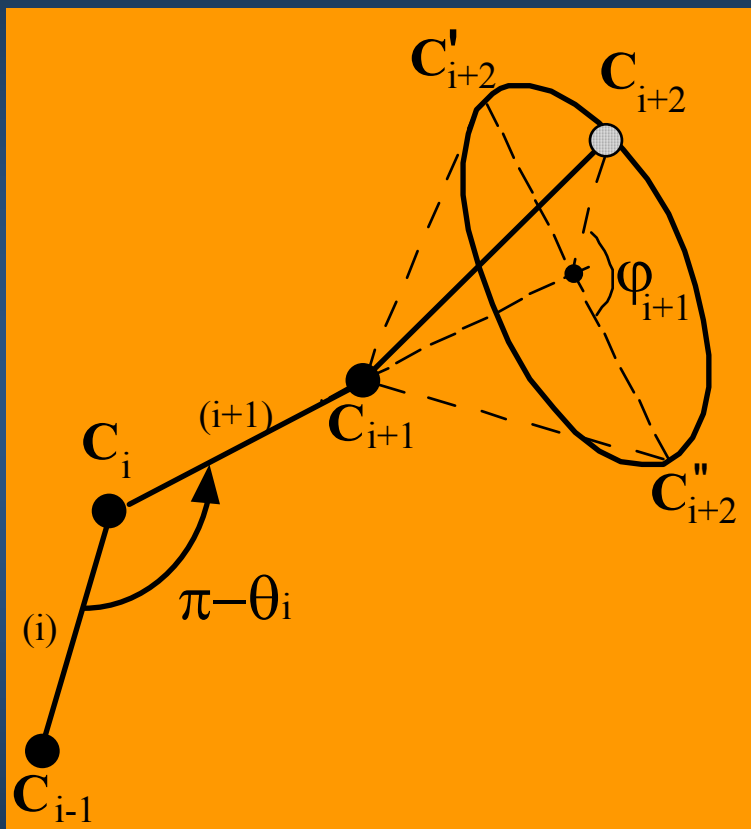
Two types of bond rotational motions

- Fluctuations around isomeric states
- Jumps between isomeric states

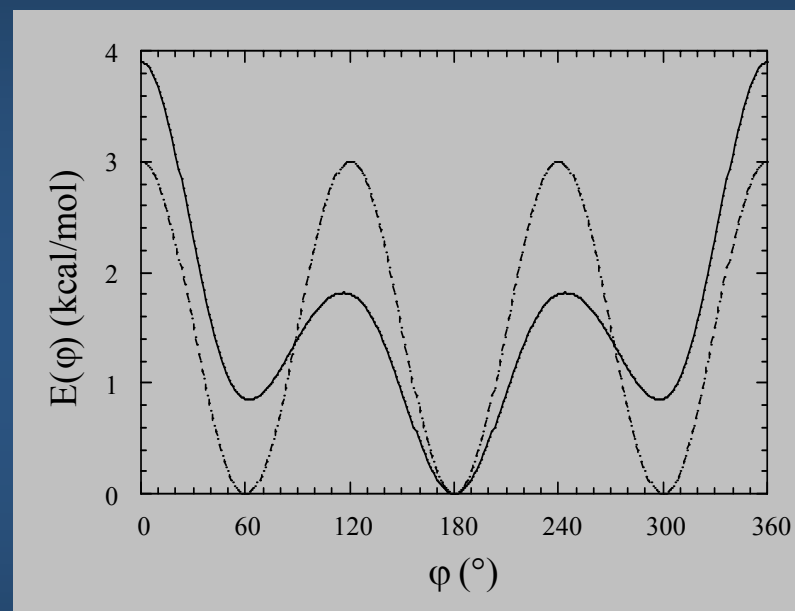
Most likely near native state



Definition of dihedral angles

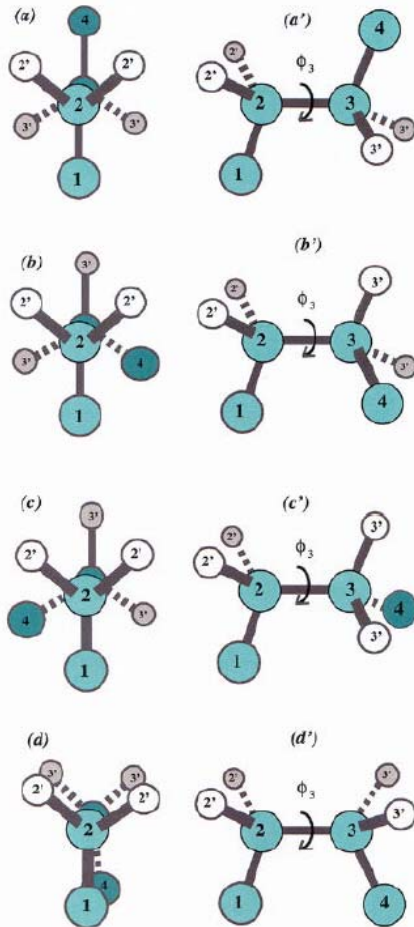


Spatial representation of the torsional mobility around the bond $i+1$. The torsional angle ϕ_{i+1} of bond $i+1$ determines the position of the atom C_{i+2} relative to C_{i-1} . C'_{i+2} and C''_{i+2} represent the positions of atom $i+2$, when ϕ_{i+1} assumes the respective values 180° and 0° .



Rotational energy as a function of dihedral angle for a threefold symmetric torsional potential (dashed curve) and a three-state potential with a preference for the *trans* isomer ($j = 180^\circ$) over the *gauche* isomers (60° and 300°) (solid curve), and the *cis* (0°) state being most unfavorable.

Rotational Isomeric States (Flory – Nobel 1974)



c. Calculation of generalized coordinates from known position vectors.

In structural analyses, it is often necessary to transform known *Cartesian* coordinates $\{x_2, x_3, y_3, \dots, x_n, y_n, z_n\}$ into generalized coordinates $\{l_2, l_3, \dots, l_n, \theta_2, \theta_3, \dots, \theta_{n-1}, \phi_3, \phi_4, \dots, \phi_{n-1}\}$ or vice versa. To this aim, it is convenient to define the bond vectors \mathbf{l}_i , pointing from atom $i-1$ to atom i . The following equations are conveniently used for transforming the Cartesian into the generalized coordinates

$$l_k = |\mathbf{r}_k - \mathbf{r}_{k-1}| \quad (1)$$

$$\theta_k = \theta_k(\mathbf{r}_{k-1}, \mathbf{r}_k, \mathbf{r}_{k+1}) = \cos^{-1} \left[\frac{\mathbf{l}_k \cdot \mathbf{l}_{k+1}}{|\mathbf{l}_k| |\mathbf{l}_{k+1}|} \right] \quad (2)$$

$$\phi_k = \phi_k(\mathbf{r}_{k-2}, \mathbf{r}_{k-1}, \mathbf{r}_k, \mathbf{r}_{k+1}) = \text{sign}[\sin(\phi_k)] \cos^{-1}(\mathbf{n}_{k-1} \cdot \mathbf{n}_k) \quad (3)$$

where \mathbf{n}_k is the unit normal vector, perpendicular to the plane spanned by \mathbf{l}_k and \mathbf{l}_{k+1} , found from

$$\mathbf{n}_k = [(\mathbf{l}_k \times \mathbf{l}_{k+1}) / |\mathbf{l}_k \times \mathbf{l}_{k+1}|] \quad (4)$$

Bond-based coordinate systems

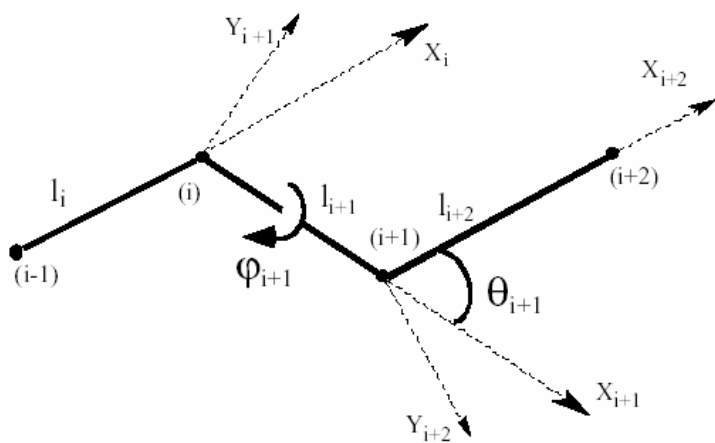
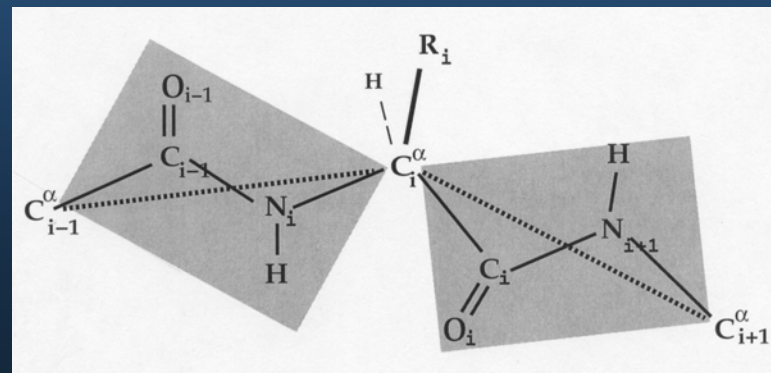


Figure 5. Schematic representation of a chain segment of four bonds. Atomic serial indices are indicated in parentheses. The i th bond connects atoms $i-1$ and i along the main chain, and its torsion angle is denoted as φ_i . θ_i is the supplemental bond angle defined by bonds i and $i+1$. The X_{i+1} and Y_{i+1} axes of the bond-based coordinate system $X_{i+1} Y_{i+1} Z_{i+1}$ appended to the bond $i+1$ are shown. Y_{i+1} lies in the plane defined by bonds i and $i+1$, and makes an acute

Transformation matrix between frames $i+1$ and i

$$T_i(\theta_i, \varphi_i) = \begin{bmatrix} \cos \theta_i & \sin \theta_i & 0 \\ -\sin \theta_i \cos \varphi_i & \cos \theta_i \cos \varphi_i & -\sin \varphi_i \\ -\sin \theta_i \sin \varphi_i & \cos \theta_i \sin \varphi_i & \cos \varphi_i \end{bmatrix}$$

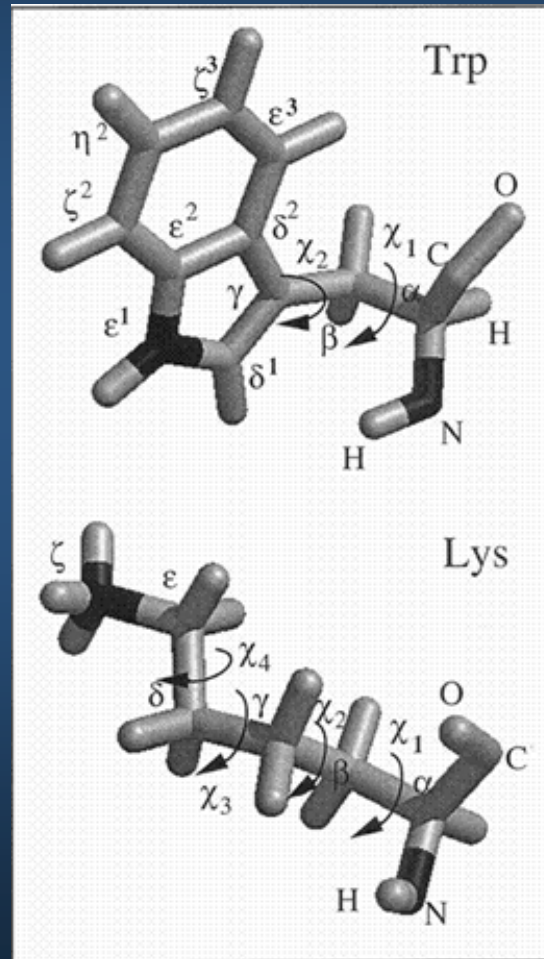
Virtual bond representation of protein backbone



Homework 1: Passage between Cartesian coordinates and generalized coordinates

- Take a PDB file. Read the position vectors (X-, Y- and Z-coordinates – Cartesian coordinates) of the first five alpha-carbons
- Evaluate the corresponding generalized coordinates, i.e. the bond lengths l_i ($i=2-5$), bond angles θ_i ($i=2-4$), and dihedral angles ϕ_3 and ϕ_4 using the Flory convention for defining these variables.
- Using the PDB position vectors for alpha-carbons 1, 2 and 3, generate the alpha carbons 4 and 5, using the above generalized coordinates and bond-based transformation matrices. Verify that the original coordinates are reproduced.

Side chains enjoy additional degrees of freedom



Harmonic Oscillator Model

- Rapid movements of atoms about a valence bond
- Oscillations in bond angles
- Fluctuations around a rotational isomeric state
- Domain motions – fluctuations between open and closed forms of enzymes

Harmonic Oscillator Model

$$F = -kx$$

A linear motion: Force scales linearly with displacement

The corresponding equation of motion is of the form

$$m \frac{d^2x}{dt^2} + kx = 0$$

The solution is the sinusoidal function $x = x_0 \sin(\omega t + \phi)$ where ω is the *frequency* equal to $(k/m)^{1/2}$, x_0 and ϕ are the original position and velocity.

Energy of a harmonic oscillator

- Kinetic energy: $E_K = \frac{1}{2} m v^2$

where $v = dx/dt = d [x_0 \sin(\omega t + \phi)]/dt = x_0 \omega \cos(\omega t + \phi)$

$$\rightarrow E_K = \frac{1}{2} m x_0^2 \omega^2 \cos^2(\omega t + \phi) = \frac{1}{2} m \omega^2 (x_0^2 - x^2)$$

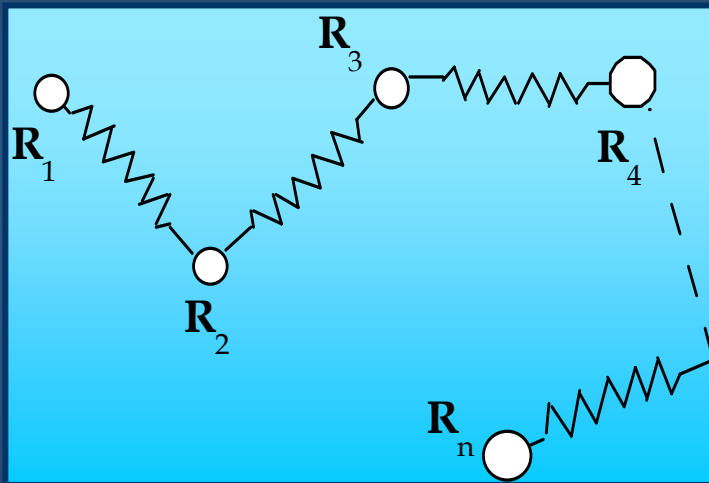
(because $x = x_0 \sin(\omega t + \phi)$ or $x^2 = x_0^2 [1 - \cos^2(\omega t + \phi)] \rightarrow x_0^2 \cos^2(\omega t + \phi) = x_0^2 - x^2$)

- Potential energy: $E_P = \frac{1}{2} k x^2$

- Total energy: $E_P + E_K = \frac{1}{2} k x_0^2$

Always fixed

Rouse chain model for **macromolecules**



Connectivity matrix

$$\Gamma = \begin{bmatrix} 1 & -1 & & & & \\ -1 & 2 & -1 & & & \\ & -1 & 2 & -1 & & \\ & & & \dots & \dots & \\ & & & & -1 & 2 & -1 \\ & & & & & -1 & 1 \end{bmatrix}$$

$$V_{\text{tot}} = (\gamma/2) [(\Delta R_{12})^2 + (\Delta R_{23})^2 + \dots + (\Delta R_{N-1,N})^2]$$

$$= (\gamma/2) [(\Delta R_1 - \Delta R_2)^2 + (\Delta R_2 - \Delta R_3)^2 + \dots] \quad (1)$$

Homework 2: Potential energy for a system of harmonic oscillators

- (a) Using the components ΔX_i , ΔY_i and ΔZ_i of $\Delta \mathbf{R}_i$, show that Eq 1 (Rouse potential) can be decomposed into three contributions, corresponding to the fluctuations along x-, y- and z-directions:

$$V_{\text{tot}} = V_x + V_y + V_z.$$

where

$$V_x = (\gamma/2) [(\Delta X_1 - \Delta X_2)^2 + (\Delta X_2 - \Delta X_3)^2 + \dots] \quad (2)$$

and similar expressions hold for V_y and V_z .

- (b) Show that eq 2 can alternatively be written as

$$V = \frac{1}{2} \Delta X^T \Gamma \Delta X \quad (3)$$

where $\Delta X^T = [\Delta X_1 \ \Delta X_2 \ \Delta X_3 \dots \Delta X_N]$, and ΔX is the corresponding column vector.

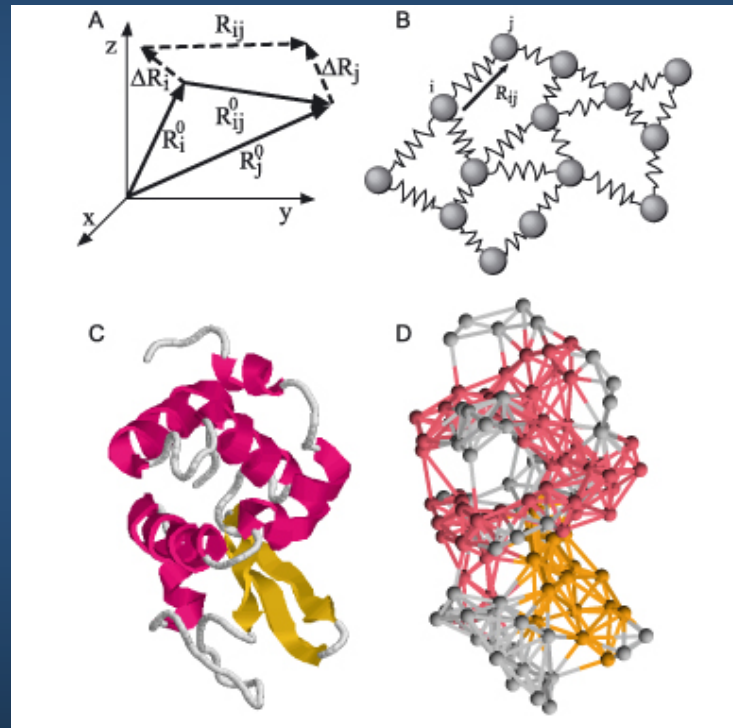
Hint: start from eq 3, obtain eq 2.

Harmonic oscillators → Gaussian distribution of fluctuations

- Consider a network formed of beads/nodes (residues or groups of residues) and springs (native contacts)
- Residues/nodes undergo Gaussian fluctuations about their mean positions – similar to the **elastic network (EN)** model of polymer gels (Flory)

$$W(\Delta\mathbf{R}_i) = \exp\left\{-\frac{3}{2} \frac{(\Delta\mathbf{R}_i)^2}{\langle(\Delta\mathbf{R}_i)^2\rangle}\right\}$$

Proteins can be modeled as an ensemble of harmonic oscillators



Gaussian Network Model - GNM

Molecular Movements

Physical properties of gases – a short review (Benedek & Villars, Chapter 2)

$$\text{Ideal gas law: } PV_M = RT$$

$$PV = NkT$$
$$PV = nRT$$

where V_M is the molar volume, T is the absolute temperature, R is the gas constant (1.987×10^{-3} kcal/mol or 8.314 J/K), k is the Boltzmann constant, N is the number of molecules, n is the number of moles = N/N_0 , N_0 is the Avogadro's number.

Mean kinetic energy of a **molecule** of mass m and its mean-square velocity:

$$\langle \frac{1}{2} mv^2 \rangle = (3/2) kT \rightarrow \langle v^2 \rangle = (3kT/m)$$

$$v_{\text{rms}} = \langle v^2 \rangle^{1/2} = (3kT/m)^{1/2}$$

Physical kinetics – Kinetic theory of gases

Root-mean-square velocities

$$v_{\text{rms}} = \langle v^2 \rangle^{1/2} = (3kT/m)^{1/2}$$

Molecule	M (g/mol)	v_{rms} (m/s)
H ₂	2	1880
O ₂	32	474
Macromolecules	10 ⁴ - 10 ⁶	2.6 - 26
Viruses (e.g. tobacco mosaic virus)	10 ⁸ - 10 ¹⁰ (5 × 10 ⁷ g/mol)	0.026 - 0.26 (35 cm/s)

Brownian motion
(Brown, 1827)

These numbers provide estimates on the time/length scales of **fluctuations** or Brownian motions

Equipartition law

An energy of $\frac{1}{2} kT$ associated with each degree of freedom

For a diatomic molecule, there are three translational (absolute), two rotational degrees of freedom, and the mean translational energies are

$$\langle \frac{1}{2} m v_x^2 \rangle = \langle \frac{1}{2} m v_y^2 \rangle = \langle \frac{1}{2} m v_z^2 \rangle = \frac{1}{2} kT$$

And the mean rotational energy is kT . For non interacting single atom molecules (ideal gases), there are only translational degrees of freedom such that the total internal energy is

$$U = \frac{3}{2} kT \quad \text{and specific heat is } C_v = \frac{\partial U}{\partial T} = \frac{3}{2} k$$

Random Walk

$$P_N(R, L) = (1/2^N) N! / R! L!$$

Probability of R steps to the right and L steps to the left in a random walk of N steps

$$\begin{matrix} R + L = N \\ R - L = m \end{matrix} \rightarrow P_N(m) = (1/2^N) N! / [(N + m/2)! (N - m/2)!]$$

Probability of ending up at m steps away from the origin, at the end of N steps

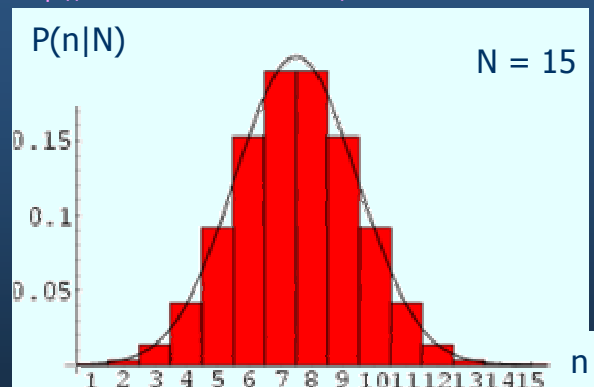
Binomial (or Bernoulli) Distribution

$$P_p(n|N) = \binom{N}{n} p^n q^{N-n} = \frac{N!}{n!(N-n)!} p^n (1-p)^{N-n},$$

Properties of Binomial Distribution

Mean	Np
Variance	Npq
Standard deviation	(Npq) ^{1/2}

<http://mathworld.wolfram.com/BinomialDistribution.html>



$$P(n) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{(n - \bar{n})^2}{2\sigma^2}\right], = \frac{1}{\sqrt{2\pi Npq}} \exp\left[-\frac{(n - Np)^2}{2Npq}\right].$$

Gaussian form of Bernoulli distribution


$$P_N(m) = (1/2^N) N! / [(N + m/2)! (N - m/2)!]$$

As m increases, the above distribution may be approximated by a continuous function

$$P_N(m) = (2/\pi N)^{1/2} \exp \{-m^2/2N\}$$

Gaussian approximation

Length of
Each step



Examples of Gaussianly distributed variables:

- Displacement (by random walk) along x -direction $\rightarrow W(x) \approx \exp \{-x^2/2Nl^2\}$ where $m=x/l$
- Fluctuations near an equilibrium position $\rightarrow W(r) \approx \exp \{-3(\Delta r)^2/2\langle(\Delta r)^2\rangle_0\}$
- Maxwell-Boltzmann distribution of velocities $\rightarrow P(v_x) = (m/2\pi kt)^{1/2} \exp \{-1/2mv_x^2/kT\}$
- Time-dependent diffusion of a particle $\rightarrow P(x,t) = \sqrt{[4\pi Dt]} \exp\{-x^2/4Dt\}$

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