

Lecture 3: Ideal chain distributions & Free Energy

A. The equivalent FJC:

- * In the last lecture, we saw that all polymer models have the same size behavior at large n :

$$\langle R_e^2 \rangle = C_\infty n l^2$$

- * The fact that different models give rise to equivalent behavior is called the principle of universality. The principle of universality is a corollary to the PPT.

PPT - Conformational entropy (& energy) dominates the behavior of macromolecules

Universality - Key properties of polymers are independent of their chemical identity (or polymer model) in the large- n limit. Only a few parameters contain chemical dependence, e.g. C_∞ .

- * If all models behave the same, then let's use the simplest one: the FJC!

- * Let's map the other models to an "equivalent FJC".

- Postulate: $\langle R_e^2 \rangle = \langle R_c^2 \rangle$ and contour length L are universal in the long n limit.

- Result:

Polymer model

Equiv. FJC

$$R_e^2 = \cos n l^2$$

$$L = n l \cos(\theta/2)$$

$$Nb^2$$

$$Nb$$

map: solve 2 EQ, 2 unknowns

$$\boxed{b = \frac{\langle R_e^2 \rangle}{L} = \frac{\cos l}{\cos(\theta/2)} \quad |}$$

$$N = \frac{L^2}{\langle R_e^2 \rangle} = \frac{L}{b} = \frac{n \cos^2(\theta/2)}{\cos}$$

- we get 2 parameters out of the equivalent FJC.

- b : the statistical segment length or kuhn length.

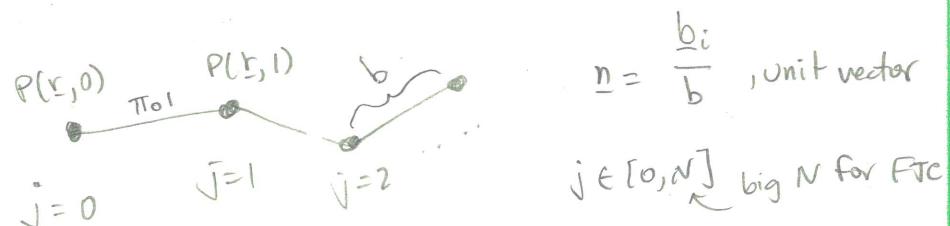
b is a measure of chain flexibility. It is in units of length, unlike \cos which is in units of repeat units (monomers)

- N : the number of kuhn lengths or number of statistical segments. N is a normalized number of monomers. Sometimes: # of kuhn monomers.

<u>polymer</u>	<u>$b(\text{\AA})$</u>	
PEO	6.0	
PI	6.5	
PS	6.7	
DNA	1000	100 nm

B. End-to-End Distribution (see GHTF & Doi & Edwards)

- * We have a pretty good handle on estimating R for ideal chains. Can we say more? We started out thinking we could get any property if we knew $p(r_0, \{b_i\})$.
- * As we said before, it is not easy to get p . This involves doing $(3n+1)$ integrals. But, for an ideal chain we can make some progress.
- * Suppose we knew $P(r, 0)$ the probability of the first monomer.



If we know the probability to go from 0 to r , then we could know $P(r, 0)$.

- $\pi_{0,1}$: a transition probability
- $P(r, j)$ is not the full $p(r_0, \{b_i\}) = p(\{r_i\})$
 $i \in [0, N]$

If it is a reduced distribution.

$$P(r, j) = \int_{r_0} r_1 \dots \int_{r_{j-1}} r_{j+1} \dots \int_{r_N} P(\{r_i\})$$

Only the probability of monomer j at point r .

* How could we get $P(r, i)$?

$$P(r, i) = \int d\mathbf{n} \pi_{0i} p(r - b\mathbf{n}, 0)$$

convolution
 $\Rightarrow P_i = \pi_{0i} * P_0$

"Chapman-Kolmogorov" Equation

(see in stochastic processes)

$$= \int d\theta \int d\phi \pi_{0i} p(r - b\mathbf{n}, 0) \sin\theta d\theta d\phi$$

\curvearrowleft
integral over unit sphere

* Let's find π_{0i} :

$$\pi_{0i} = \text{const} \quad (\text{uniform distribution})$$

$$\int d\mathbf{n} \pi_{0i} = 1 \quad (\text{to be a probability})$$

$$\int_0^\pi d\theta \int_0^{2\pi} d\phi \sin\theta \pi_{0i} = 1 \Rightarrow 4\pi \pi_{0i} = 1$$

$\pi_{0i} = \frac{1}{4\pi}$

* Therefore we get:

$$P(r, i) = \int d\mathbf{n} \frac{1}{4\pi} p(r - b\mathbf{n}, 0)$$

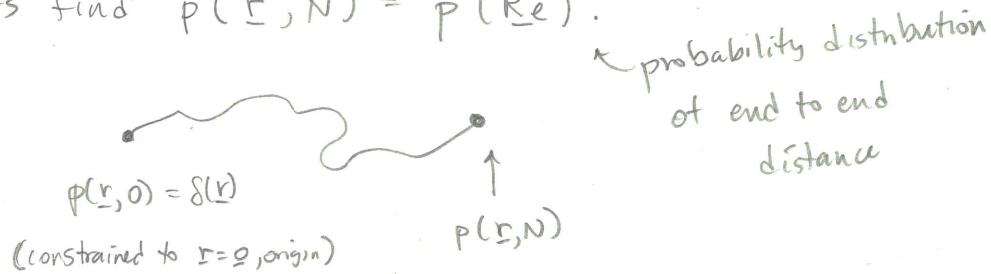
* This applies at all steps along the chain

therefore :

$$p(\underline{r}, j) = \frac{1}{4\pi} \int d\underline{n} p(\underline{r} - b\underline{n}, j-1)$$

we can get all of the monomers!

* Let's find $p(\Sigma, N) = p(\underline{R}_e)$.



we "simply" need to chain the CK equations together.

* This is hard w/ lots of integrals. However, it is easy using a Fourier Transform:

$$\text{Fourier Transform: } \hat{f}(\underline{k}) = \int d\underline{r} f(\underline{r}) e^{-i\underline{k} \cdot \underline{r}}$$

$$\text{Inverse Fourier transform: } f(\underline{r}) = \frac{1}{(2\pi)^3} \int d\underline{k} \hat{f}(\underline{k}) e^{i\underline{k} \cdot \underline{r}}$$

$$F \left[\int d\underline{r}' g(\Sigma - \underline{r}') h(\underline{r}') \right] = \hat{g}(\underline{k}) \hat{h}(\underline{k})$$

convolution theorem.

$$F[p(\underline{r}, j)] = F \left[\frac{1}{4\pi} \int d\underline{n} p(\underline{r} - b\underline{n}, j-1) \right]$$

$$\hat{p}(\underline{k}, j) = j_0(b|\underline{k}|) \hat{p}(\underline{k}, j-1)$$

$$\uparrow j_0(x) = \frac{\sin x}{x}, \text{ Spherical Bessel function}$$

(yay transport!)

$$\hat{p}(\underline{k}, 1) = j_0(b|\underline{k}|) \hat{p}(\underline{k}, 0)$$

$$\hat{p}(\underline{k}, 2) = j_0(b|\underline{k}|) \hat{p}(\underline{k}, 1)$$

$$\hat{p}(\underline{k}, 3) = j_0(b|\underline{k}|) \hat{p}(\underline{k}, 2)$$

$$\hat{p}(\underline{k}, N) = j_0(b|\underline{k}|) \hat{p}(\underline{k}, N-1)$$

$$= [j_0(b|\underline{k}|)]^N \hat{p}(\underline{k}, 0)$$

$$\uparrow \quad \quad \quad \text{3D Dirac Delta}$$

$$F[\delta(\underline{r})] = 1$$

$$p(r, N) = F^{-1}[\hat{p}(\underline{k}, N)]$$

$$p(r, N) = \frac{1}{(2\pi)^3} \int d\underline{k} \left[j_0(b|\underline{k}|) \right]^N e^{i\underline{k} \cdot \underline{r}}$$

* Hard to do this integral, but we can approximate it easily when $b|\underline{k}| \ll 1$. $\rightarrow |\underline{k}| \gg b$ $\underline{k} = |\underline{k}|$

- when \underline{k} is large, at larger length scales. This makes sense, we only care about stuff much larger than monomers

$$j_0(bk) = \frac{\sin(bk)}{bk}$$

$$\frac{\sin(x)}{x} \approx 1 - \frac{x^2}{6}$$

(Taylor series, wolframAlpha)

$$j_0(bk) \approx 1 - \frac{(bk)^2}{6}$$

- Recall $\ln(1-x) \approx -x$ for $x \approx 1$

$$\text{so } N \ln(1-x) \approx -Nx$$

$$\exp[(1-x)^N] \approx \exp(-Nx)$$

$$\Rightarrow [j_0(bk)]^N \approx \exp\left(-\frac{Nb^2 k^2}{6}\right)$$

$$\Rightarrow p(r, N) = \frac{1}{2\pi^3} \int dk \exp\left(-\frac{Nb^2 k^2}{6}\right) e^{ik \cdot r}$$

• This integral, we can do!

(F.T. of a Gaussian is a Gaussian)

$$p(r, N) = \left[\frac{3}{2\pi Nb^2} \right]^{3/2} \exp\left(-\frac{3|r|^2}{2Nb^2}\right)$$

* Comments:

• This is a Gaussian distribution (bell curve).

$$\text{mean : } \bar{r} = 0$$

$$\text{variance : } \frac{Nb^2}{3} = 2R_g^2$$

$$p(r, N) = \left(\frac{1}{4\pi R_g^2} \right)^{3/2} \exp\left(-\frac{|r|^2}{4R_g^2}\right)$$

• This distribution represents conformational entropy! We added them up in our convolutions.

• In spherical coordinates :

$$p(R, N) \cdot 4\pi R^2 dR = 4\pi \left(\frac{3}{2\pi Nb^2} \right)^{3/2} \exp\left(-\frac{3}{2} \frac{R^2}{Nb^2}\right) R^2 dR$$

no vector

(see plot) \longrightarrow very small R and very large R is unlikely.

- For this to work, we need $N \gg 1$ and $R \ll L = Nb$. The distribution is unphysical at large R .

C. Free Energy of an Ideal Chain

*Remember from last time:

$$F = -kT \ln Z$$

Free energy as a function of r .

$$F(r, N) = -kT \ln Z(r, N)$$

Now, $P(r, N) = \frac{z(r, N)}{\int dr z(r, N)}$

over all but r, N

integral over them all.

$$Z(r, N) = P(r, N) \int dr z(r, N)$$

$$F(r, N) = -kT \ln P(r, N) - kT \ln \int dr z(r, N)$$

$$= -\frac{3}{2} kT \left(\frac{3}{2\pi Nb^2} \right) + \frac{3}{2} kT \frac{|r|^2}{Nb^2}$$

$\nearrow -kT \ln \int dr z(r, N)$

don't depend on $r \Rightarrow F(0, N)$

$$F(r, N) = F(0, N) + \frac{3}{2} kT \frac{|r|^2}{Nb^2}$$

or

$$\Delta F(R) = \frac{3kT}{2} \frac{R^2}{Nb^2}$$

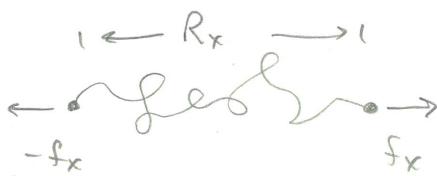
- * The chain acts like an "entropic spring" with spring constant:

$$F = \frac{1}{2} kx^2 \quad \frac{\partial F}{\partial x} = kx$$

$$F = \frac{3kT}{2} \frac{R^2}{Nb^2} \quad \frac{\partial F}{\partial R} = \frac{3kT}{Nb^2} \cdot R$$

$$\boxed{k = \frac{3kT}{Nb^2}}$$

- $T \uparrow$, chain gets stiffer.



- metals $T \uparrow$, get softer

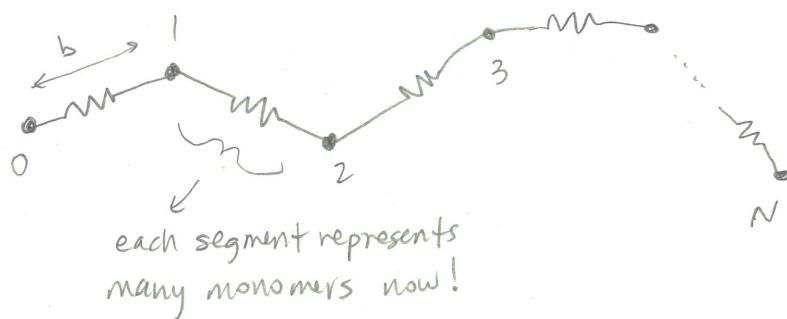
$$|f| \propto \frac{3kT}{Nb^2} \cdot R_x$$

- Entropic vs. Enthalpic elasticity

- * Conformational entropy resists the chain from being stretched or compressed.

D. The Gaussian chain

- * Now we see that if we only care about universal behavior at large N , we get something that looks like a Hookean spring. With this in mind, why even model a complex chain with bond and torsion angles. Instead, we can model a chain as a "bead-spring" model with flexible bonds.



$$U_0 = \sum_{i=1}^N \frac{3kT}{2b^2} |r_{i+1} - r_i|^2$$

* This model gives

$$\langle \underline{R} \cdot \underline{R} \rangle = Nb^2 \quad \underline{R} = \underline{r}_N - \underline{r}_0$$

$$P(\Sigma, N) = \left(\frac{3}{2\pi Nb^2} \right)^{3/2} \exp\left(\frac{-3|\Sigma|^2}{2Nb^2} \right) \leftarrow \text{exact}$$

exactly the same as the equivalent FJC
in the limit $N \gg 1$, $|r_i| \ll Nb$.

* In fact:

$$\hat{\Pi}_{ij}(k) = \exp(-R_g^2 k^2) \leftarrow \begin{array}{l} \text{each transition} \\ \text{probability is} \\ \text{a Gaussian chain.} \\ (\text{like the } N \gg 1 \text{ limit of} \\ \text{many FJC links}) \end{array}$$