

Lecture 6: Entropy Rules! (contd.)

Lecturer:
Brigita Urbanc
Office: 12-909
(E-mail: *brigita@drexel.edu*)

Course website:

www.physics.drexel.edu/~brigita/COURSES/BIOPHYS_2011-2012/

Boltzmann distribution (by counting: Textbook, pages 230-231)

Information theory:

based on the constraints (for example the total energy) *derive the least biased probability distribution*

Shannon entropy:

$$S(p_1, p_2, p_3 \dots, p_N) = - \sum_{i=1}^N p_i \ln p_i$$

where p_i is the probability of the system to be in the i -th microstate.

For example, if nothing is known about the system except that there is N microstates, then for all $p_i = 1/N$ and $S = \ln N$ (the maximal value).

How do we formally derive this result?

Maximize Shannon entropy S' using Lagrange multiplier method for each constraint:

$$S' = - \sum_{i=1}^N p_i \ln p_i - \lambda [\sum_{i=1}^N p_i - 1]$$

Maximization equations:

$$\partial S' / \partial \lambda = 0 \quad \rightarrow \quad - \sum_{i=1}^N p_i + 1 = 0$$

$$\partial S' / \partial p_i = 0 \quad \rightarrow \quad - \ln p_i - 1 - \lambda = 0$$

$$p_i = \exp(-1-\lambda)$$

Note that the probabilities p_i do not depend on i !

$$\sum_{i=1}^N p_i = 1 \rightarrow \sum_{i=1}^N \exp(-1-\lambda) = 1 \rightarrow \exp(-1-\lambda) = 1/N$$

$$p_i = 1/N$$

Boltzmann distribution is a maximum entropy distribution with a fixed average energy:

$$S' = -\sum_i p_i \ln p_i - \lambda [\sum_i p_i - 1] - \beta [\sum_i p_i E_i - \langle E \rangle]$$

$$-\ln p_i - 1 - \lambda - \beta E_i = 0 \rightarrow p_i = \exp(-1-\lambda) \exp(-\beta E_i)$$

$$\sum_i p_i = 1 \rightarrow \exp(1+\lambda) = \sum_i \exp(-\beta E_i) = Z$$

$$p_i = \exp(-\beta E_i) / \sum_i \exp(-\beta E_i)$$

Ideal gas approximation: the interaction range is small in comparison to the mean spacing between molecules

For a system with N independent variables ($\mathbf{x}_1, \mathbf{x}_2, \dots \mathbf{x}_N$) the probability distribution can be factorized:

$$P(\mathbf{x}_1, \mathbf{x}_2, \dots \mathbf{x}_N) = P(\mathbf{x}_1) P(\mathbf{x}_2) \dots P(\mathbf{x}_N)$$

– uniform spatial distribution of ideal gas molecules.

Microscopic state of the system is described by $(\mathbf{X}, \mathbf{p}_x)$ so the key information that we need to derive is $P(\mathbf{p}_x)$ knowing the kinetic energy of $E = \mathbf{p}_x^2/(2m)$.

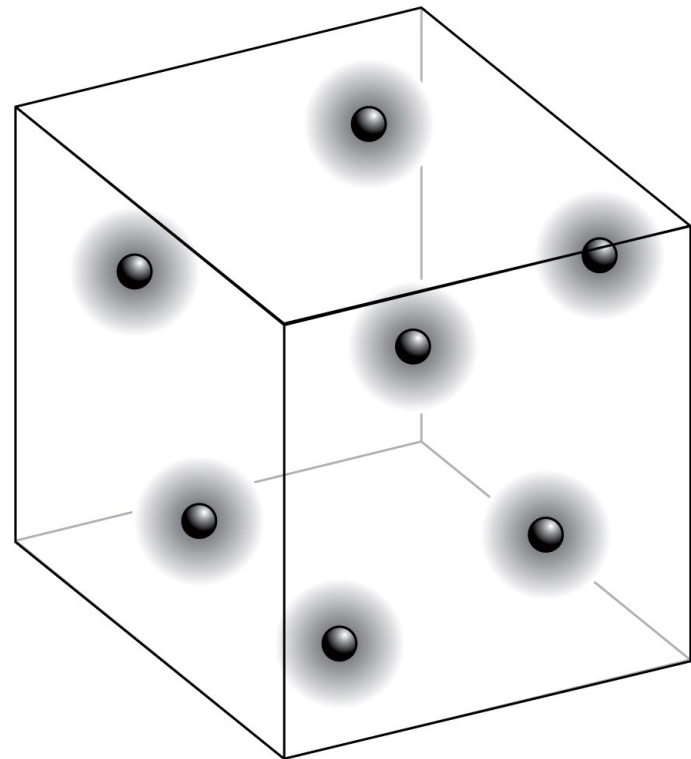


Figure 6.19 Physical Biology of the Cell (© Garland Science 2009)

Probability distribution for a system with average energy is:

$$\exp[-\beta p_x^2/(2m)]$$

$$P(p_x) = \frac{\exp[-\beta p_x^2/(2m)]}{\sum_{\text{states}} \exp[-\beta p_x^2/(2m)]}$$

Instead of a sum, we use an integral over a continuous variable p_x :

$$\sum_{\text{states}} \rightarrow \int_{-\infty}^{\infty} dp_x$$

where we will use the integral:

$$\int_{-\infty}^{\infty} \exp(-\alpha p_x^2) dp_x = \sqrt{\pi/\alpha}$$

According to the equipartition theorem, each degree of freedom is associated with :

$$\langle E \rangle = \frac{1}{2} k_B T$$

Calculate $\langle E \rangle$ using the Boltzmann distribution:

$$Z = \int_{-\infty}^{\infty} \exp[-\beta p_x^2/(2m)] dp_x = \sqrt{2m\pi/\beta}$$

$$\langle E \rangle = \int_{-\infty}^{\infty} p_x^2/(2m) \exp[-\beta p_x^2/(2m)] dp_x$$

We use the following trick:

$$\langle E \rangle = - Z^{-1} \partial Z / \partial \beta = \frac{1}{2} \beta^{-1}$$

so we showed that the Lagrange multiplier $\beta = (k_B T)^{-1}$.

Free energy and chemical potential of a *dilute solution*: Application of a lattice model (and ideal gas approx.)

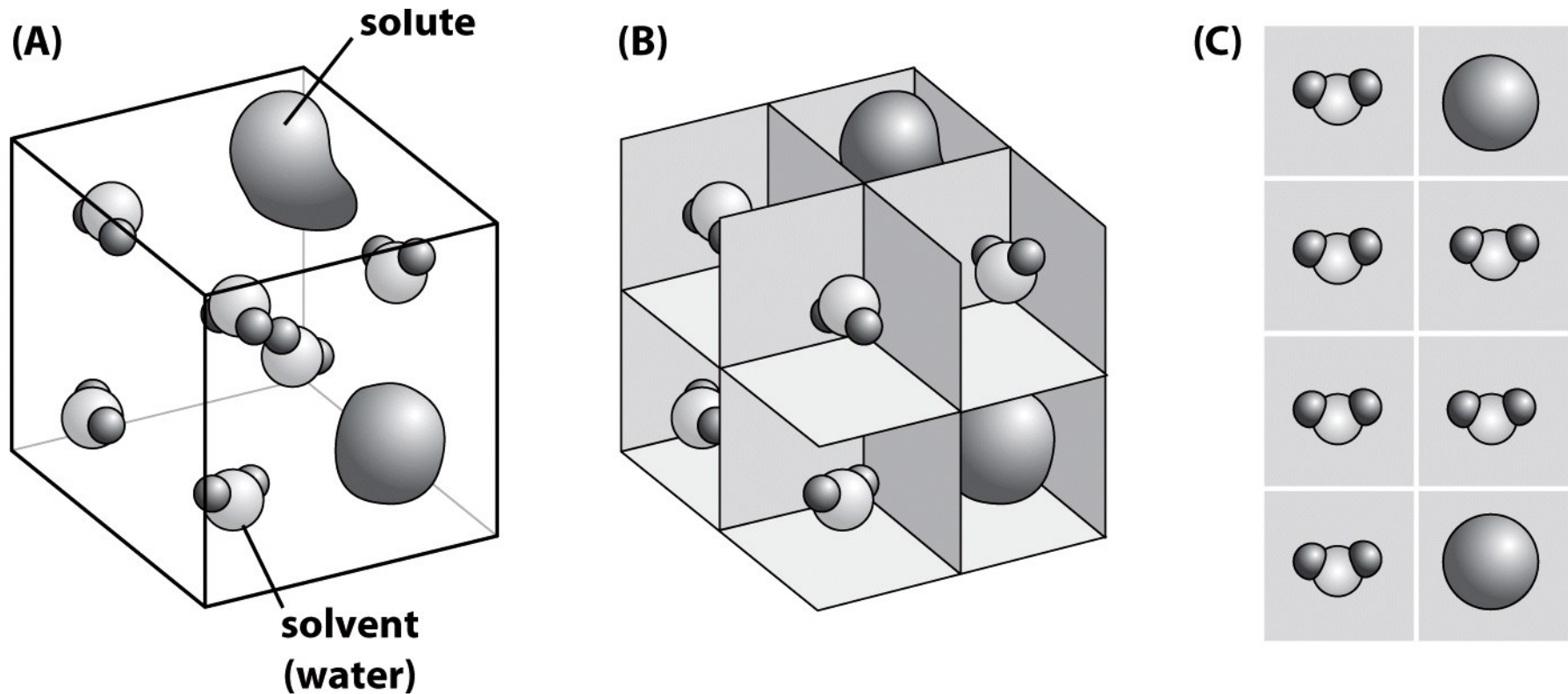


Figure 6.20 Physical Biology of the Cell (© Garland Science 2009)

**Configurational entropy of N objects placed into W available spots:
 Ω !**

$$W(N, \Omega) = \frac{\Omega!}{N! (\Omega - N)!} \quad S = k_B \ln W$$

How to calculate the chemical potential of a dilute solutions?

$$\mu_{\text{SOLUTE}} = (\partial G_{\text{TOT}} / \partial N_S)_{T,p}$$

$$G_{\text{TOT}} = N_W \mu_W^0 + N_S \epsilon_S - T S_{\text{MX}}$$

water G + solute energy + mixing entropy contribution

Mixing entropy contribution:

-independent solute molecules (ideal gas)

-lattice model: $N_W + N_S$ is a total number of lattice sites

$$W(N_W, N_S) = (N_W + N_S)! / (N_W! N_S!)$$

$$\begin{aligned}
 S &= k_B \ln W = \ln[(N_w + N_s)! / (N_w! N_s!)] \approx \\
 &- k_B \{N_w \ln[N_w/(N_w + N_s)] + N_s \ln[N_s/(N_w + N_s)]\} = \\
 &- k_B [N_w \ln(1 - N_s/N_w) + N_s \ln(N_s/N_w)]
 \end{aligned}$$

Taking into account the Taylor expansion of $\ln(1+x) \approx x$, we get:

$$S_{MX} = - k_B [N_s \ln(N_s/N_w) - N_s]$$

$$\begin{aligned}
 G_{TOT}(T,p,N_w,N_s) &= N_w \mu_w^0 + N_s \epsilon_s(T,p) + \\
 &\quad k_B T (N_s \ln(N_s/N_w) - N_s)
 \end{aligned}$$

$$\mu_s(T,p) = \epsilon_s(T,p) + k_B T \ln(c/c_0)$$

or in a general form expressed in concentration $c=N/V$:

$$\mu_i = \mu_{i0} + k_B T \ln(c_i/c_{i0})$$

Osmotic pressure Is an Entropic Effect

→ consider a cell in an aqueous environment exchanging material with a solution (intake of food, excretion of waste)

→ chemical potential difference proportional to ΔG :

$$\Delta G = (\mu_1 - \mu_2) dN \leq 0 \text{ (spontaneous)}$$

→ cell with a crowded environment of biomolecules: tendency of almost all components to move out causes a mechanical pressure called *osmotic pressure*

→ lipid membranes with ion channels to regulate ion concentration

→ calculate osmotic pressure due to a dilute solution of N_s molecules

Osmotic pressure on a semipermeable membrane, which only allows water molecules through

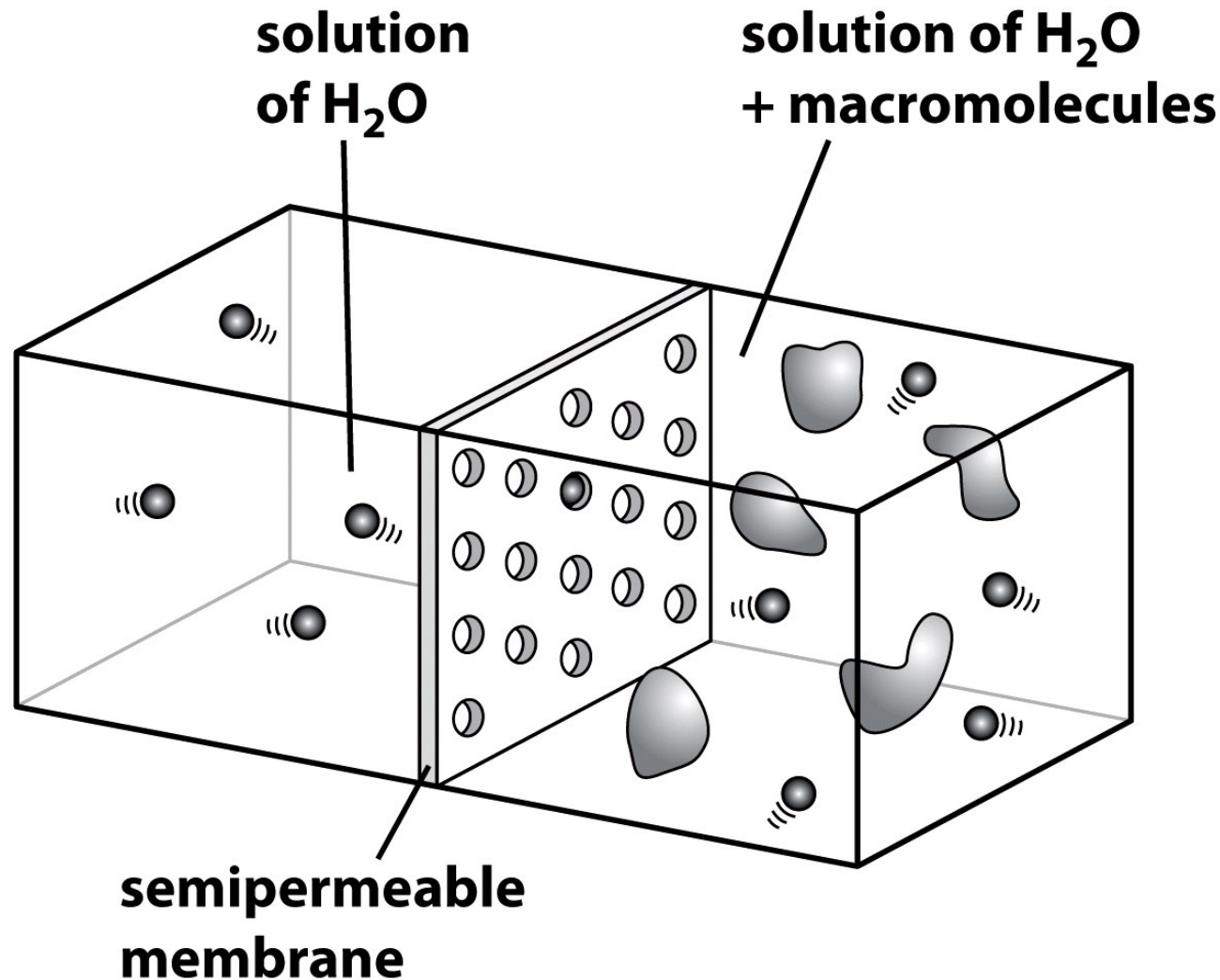


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$$\mu_w(T,p) = (\partial G_{\text{TOT}} / \partial N_w)_{T,p} =$$

$$\mu_w^0(T,p) - k_B T N_s / N_w$$

For both sides of the membrane in equilibrium:

$$\mu_w^0(T,p_1) = \mu_w^0(T,p_2) - k_B T N_s / N_w$$

Expand the chemical potential at p_2 around the p_1 value:

$$\mu_w^0(T,p_2) \approx \mu_w^0(T,p_1) + (\partial \mu_w^0 / \partial p) (p_2 - p_1)$$

and consider that $(\partial \mu_w^0 / \partial p) = v = N_w / V$ is volume per water molecule so that

$$(p_2 - p_1) = k_B T N_s / V$$

Measuring interstrand interactions in DNA using osmotic pressure

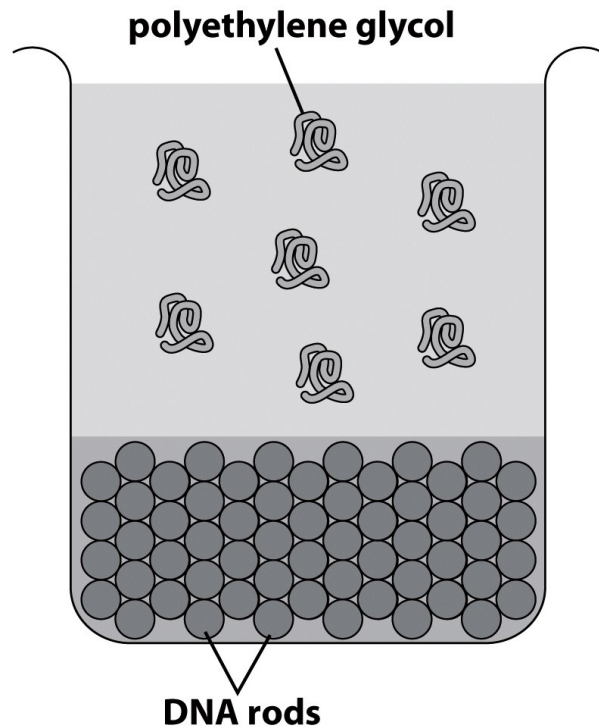


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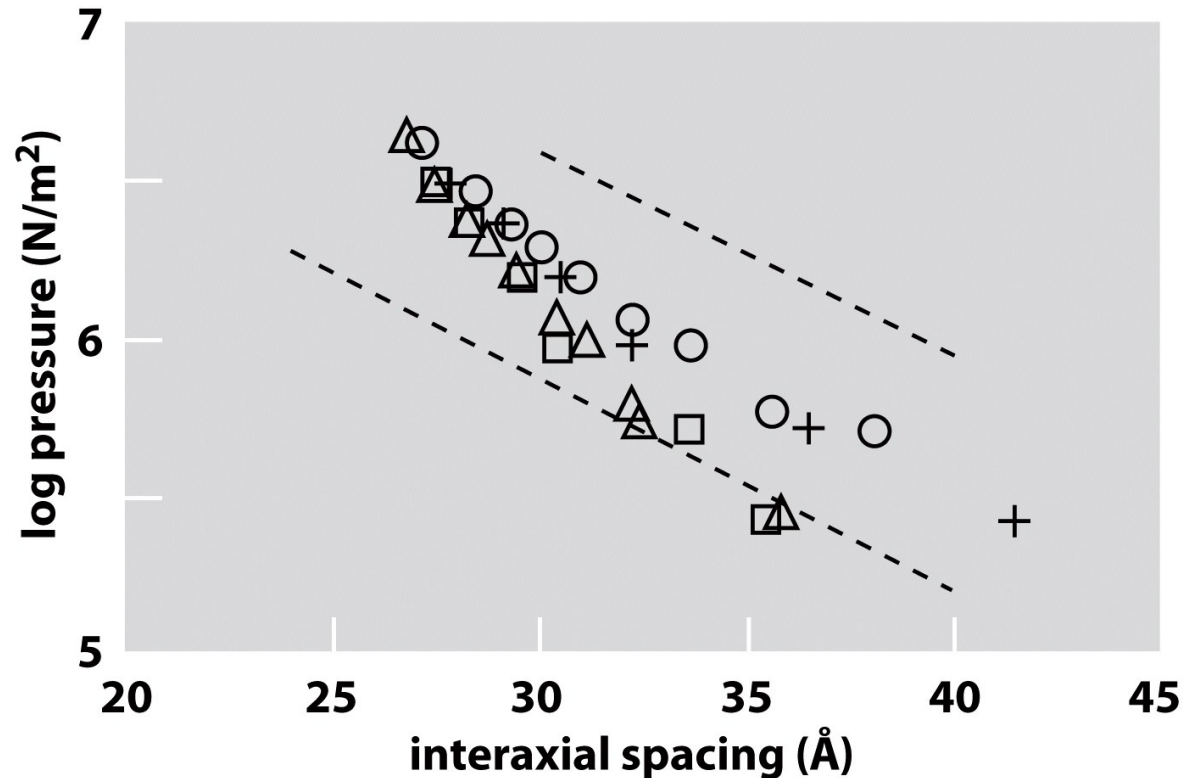


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$$\text{pressure: } p(d_s) = F_0 \exp(-d_s/c)$$

d_s ... interstrand spacing; F_0 depends on the ionic solution

Law of Mass Action and Equilibrium Constants (Chemical Reactions)

→ chemical equilibrium between A, B and their complex AB:



→ final equilibrium independent of whether we start with only A and B, or with a high concentration of AB and no A or B

→ N_A , N_B , N_{AB} ... number of A, B, and AB molecules

→ In equilibrium: $dG = 0$

$$0 = (\partial G / \partial N_A) dN_A + (\partial G / \partial N_B) dN_B + (\partial G / \partial N_{AB}) dN_{AB}$$

→ A more convenient and general expression:

$$\sum_i^N \mu_i dN_i = 0$$

Stoichiometric coefficients for each of the reactants are defined as:

$$\mathbf{v}_i = \pm 1$$

depending on whether the number of particles of the i-th type increases or decreases during the reaction:

$$\sum_{i=1}^N \mu_i \mathbf{v}_i = 0$$

$$\sum_{i=1}^N \mu_{i0} \mathbf{v}_i = -k_B T \sum_{i=1}^N \ln(c_i/c_{i0})^{v_i}$$

or

$$-\beta \sum_{i=1}^N \mu_{i0} \mathbf{v}_i = \ln[\prod_{i=1}^N (c_i/c_{i0})^{v_i}]$$

or

$$\prod_{i=1}^N c_i^{v_i} = (\prod_{i=1}^N c_{i0}^{v_i}) \exp(-\beta \sum_{i=1}^N \mu_{i0} \mathbf{v}_i)$$

where we define *the equilibrium constant* \mathbf{K}_{eq} :

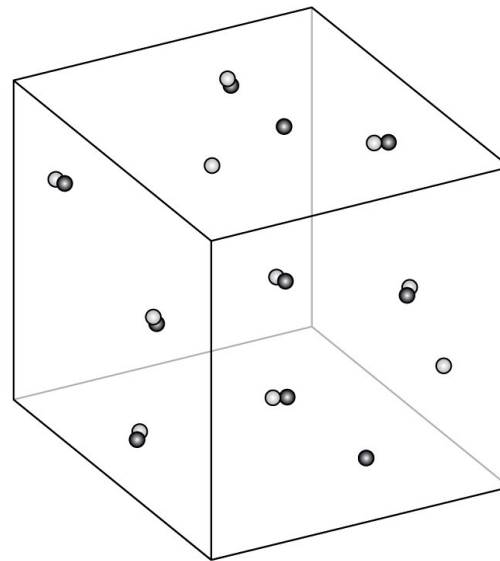
$$\mathbf{K}_{\text{eq}} = (\prod_{i=1}^N c_{i0}^{v_i}) \exp(-\beta \sum_{i=1}^N \mu_{i0} \mathbf{v}_i)$$

$K_d = 1/K_{eq}$... dissociation constant

In our case of the reaction $A + B \leftrightarrow AB$, we can express K_d as:

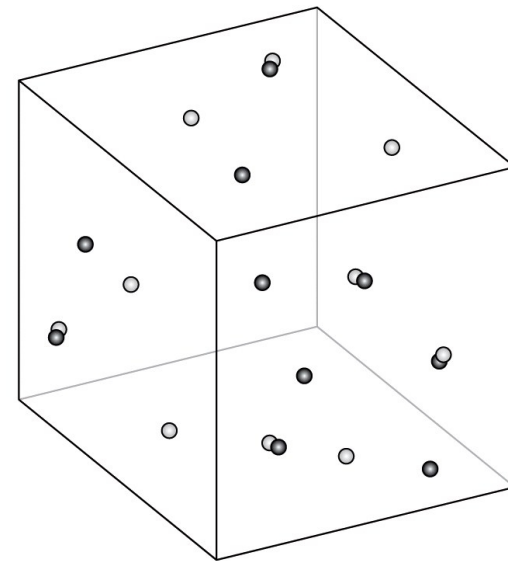
$$K_d = \prod_i c_i^{-\nu_i} = c_A c_B / c_{AB}$$

**Example: total
concentration:
50 μM**



$K_d = 2.5 \mu\text{M}$

80% bound



$K_d = 25 \mu\text{M}$

50% bound

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Application to Ligand-Receptor Binding:



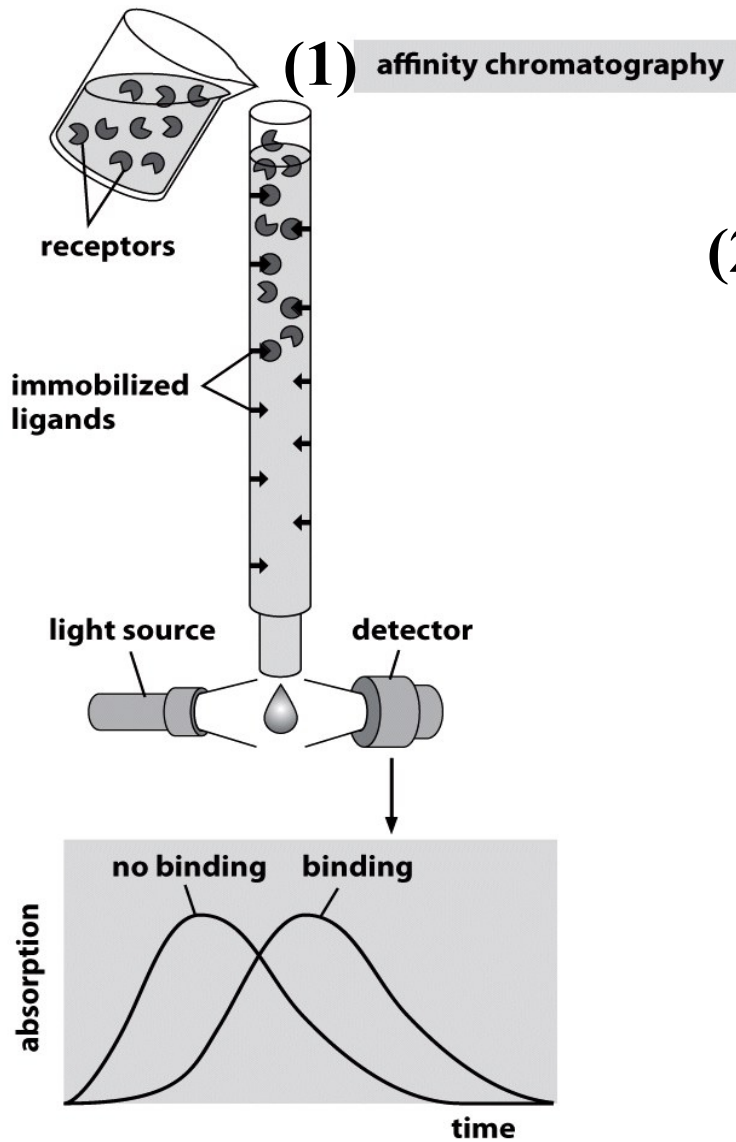
$$\mathbf{K_d = [L][R]/[LR] \text{ or } [LR] = [L] [R]/K_d}$$

Binding probability:

$$\mathbf{p_{\text{bund}} = \frac{[LR]}{[LR] + [R]} = \frac{[L]/K_d}{1 + [L]/K_d}}$$

A natural interpretation of K_d : K_d is the concentration at which the receptor has a probability of $\frac{1}{2}$ of being occupied by a ligand. Based on our prior result, we can express it in terms of lattice model parameters as: $K_d = v^{-1} \exp(\beta \Delta \epsilon)$

Important: K_d depends on the concentration of *free* ligands not their *total* concentration!



(2) equilibrium dialysis

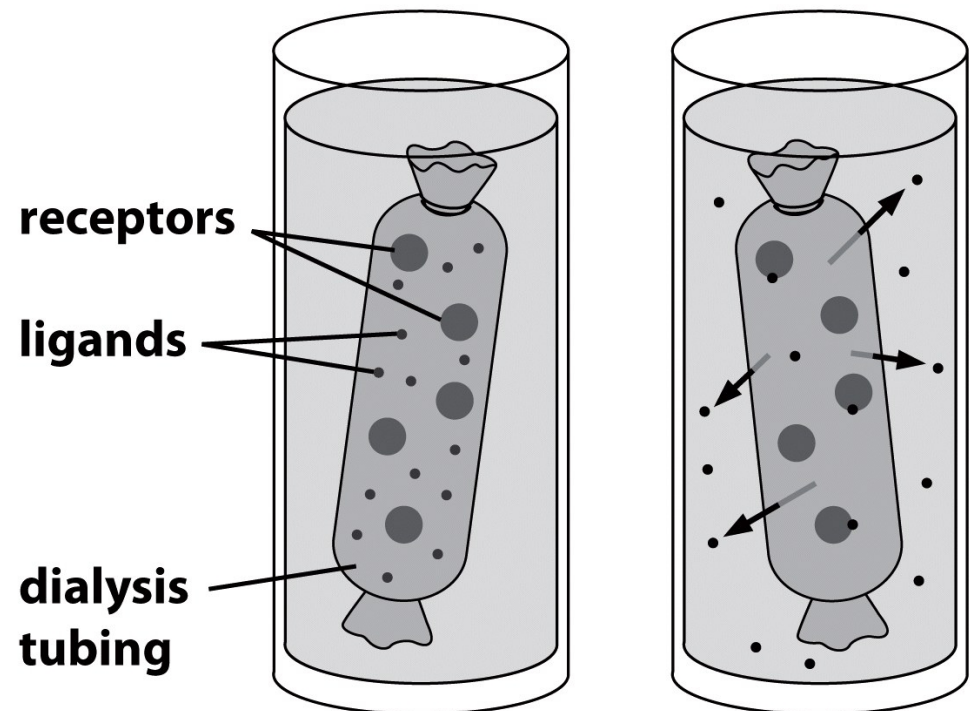
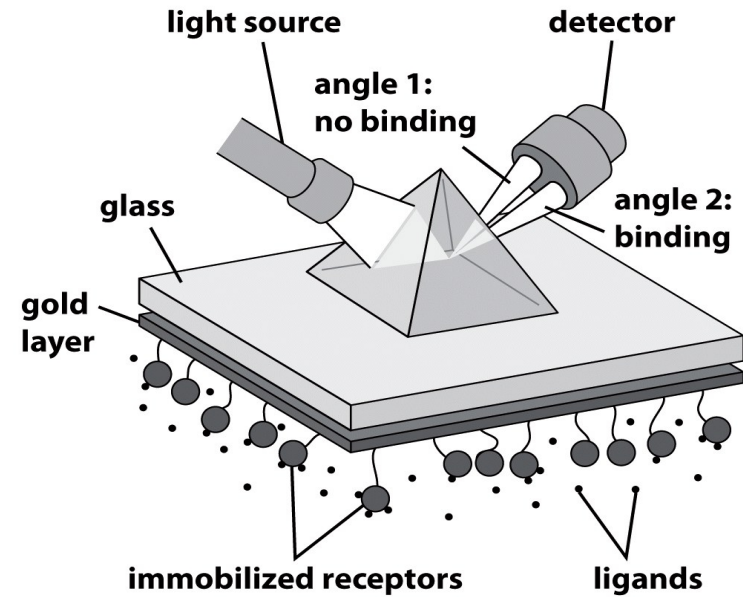


Figure 6.25b Physical Biology of the Cell (© Garland Science 2009)

Figure 6.25a Physical Biology of the Cell (© Garland Science 2009)

(C) surface plasmon resonance



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(D)

isothermal titration calorimetry

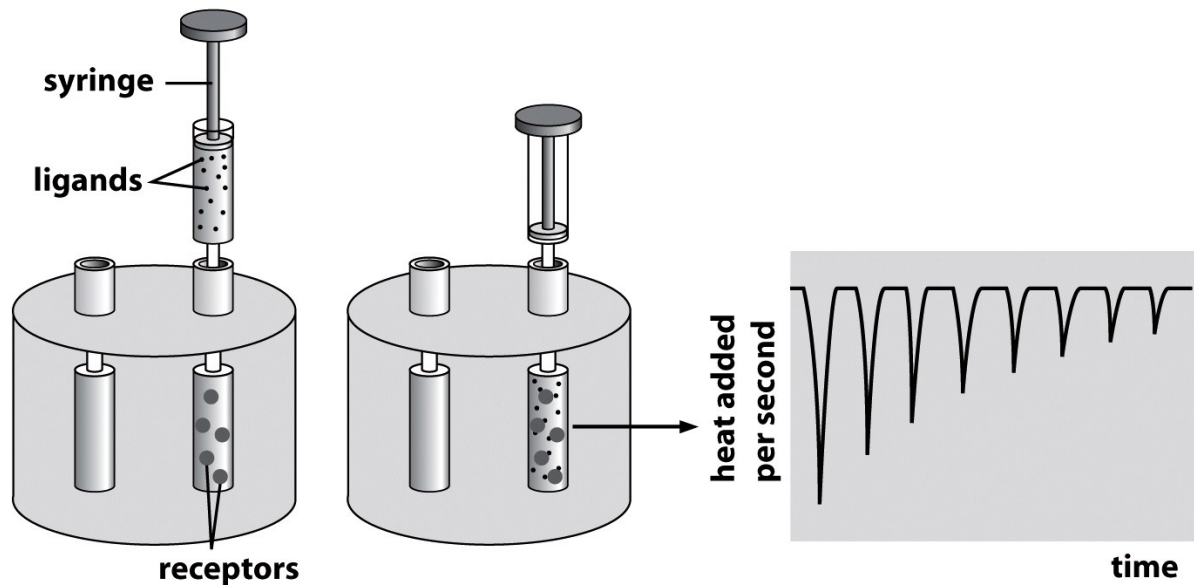


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Cooperative Ligand-Receptor Binding: The Hill Function

Biological function: an on-off switch behavior triggered by Binding of a ligand to a receptor involves a cooperative (all-or-none) mechanism:



$$K_d^2 = [\text{L}]^2[\text{R}]/[\text{L}_2\text{R}]$$

$$p_{\text{bound}} = \frac{([\text{L}]/K_d)^2}{1 + ([\text{L}]/K_d)^2} = \frac{([\text{L}]/K_d)^n}{1 + ([\text{L}]/K_d)^n}$$

The larger the n , the sharper the binding curve (probability of binding versus ligand concentration)

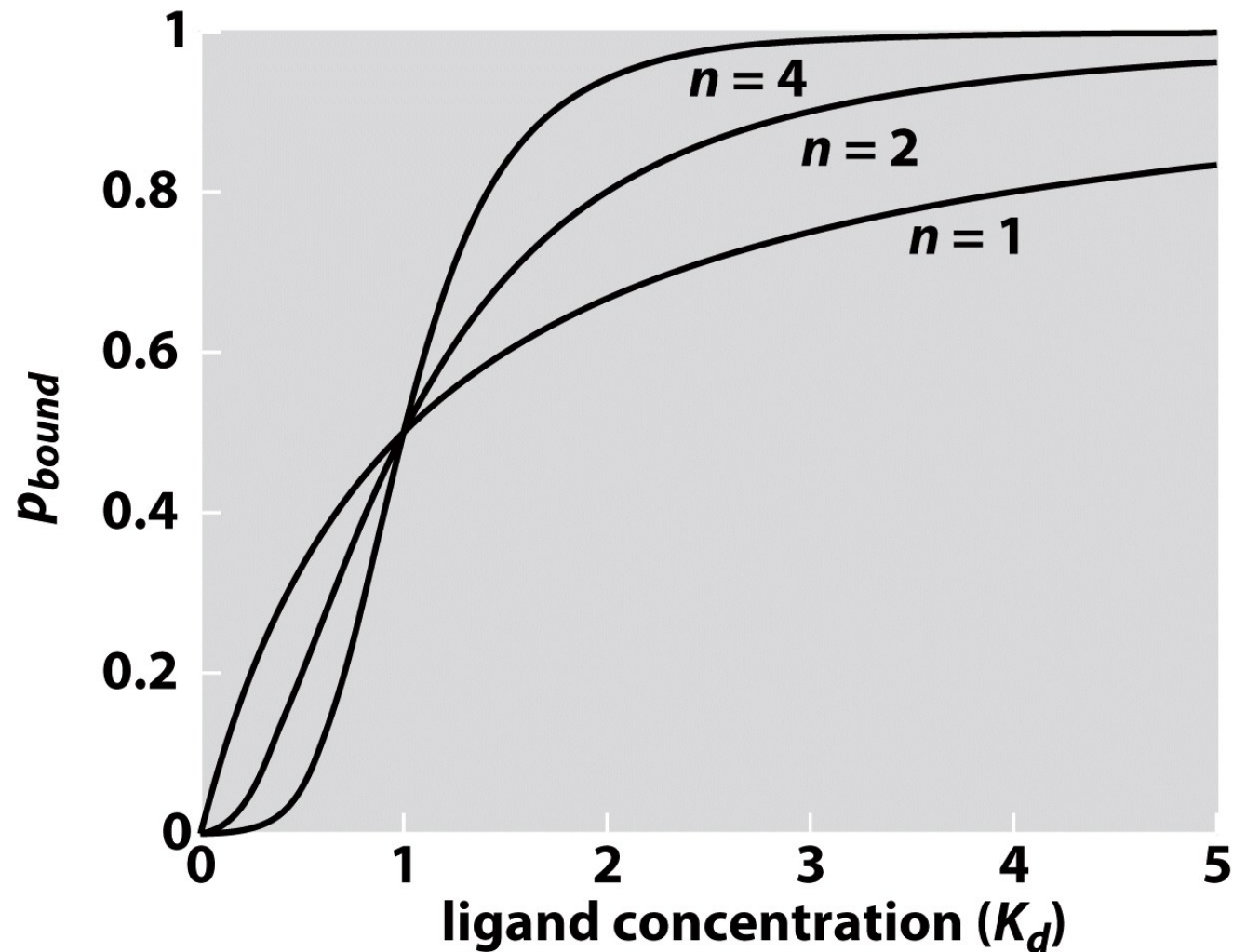


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