

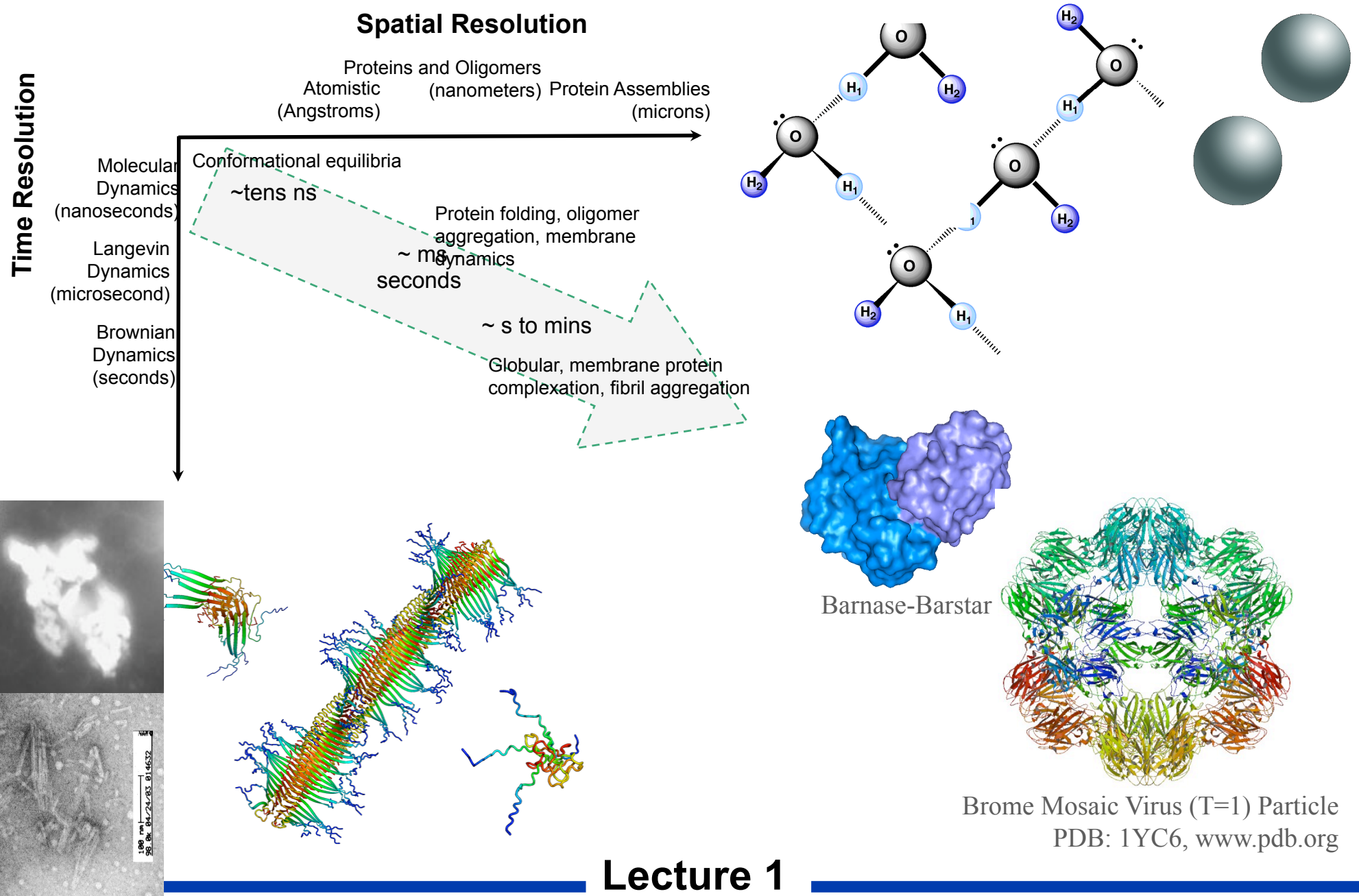
Coarse-Graining Methods in Biology and Materials

School on Multiscale Modeling and
Simulations of Hard and Soft Materials

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University of California, Berkeley



Coarse-Graining and MultiScale

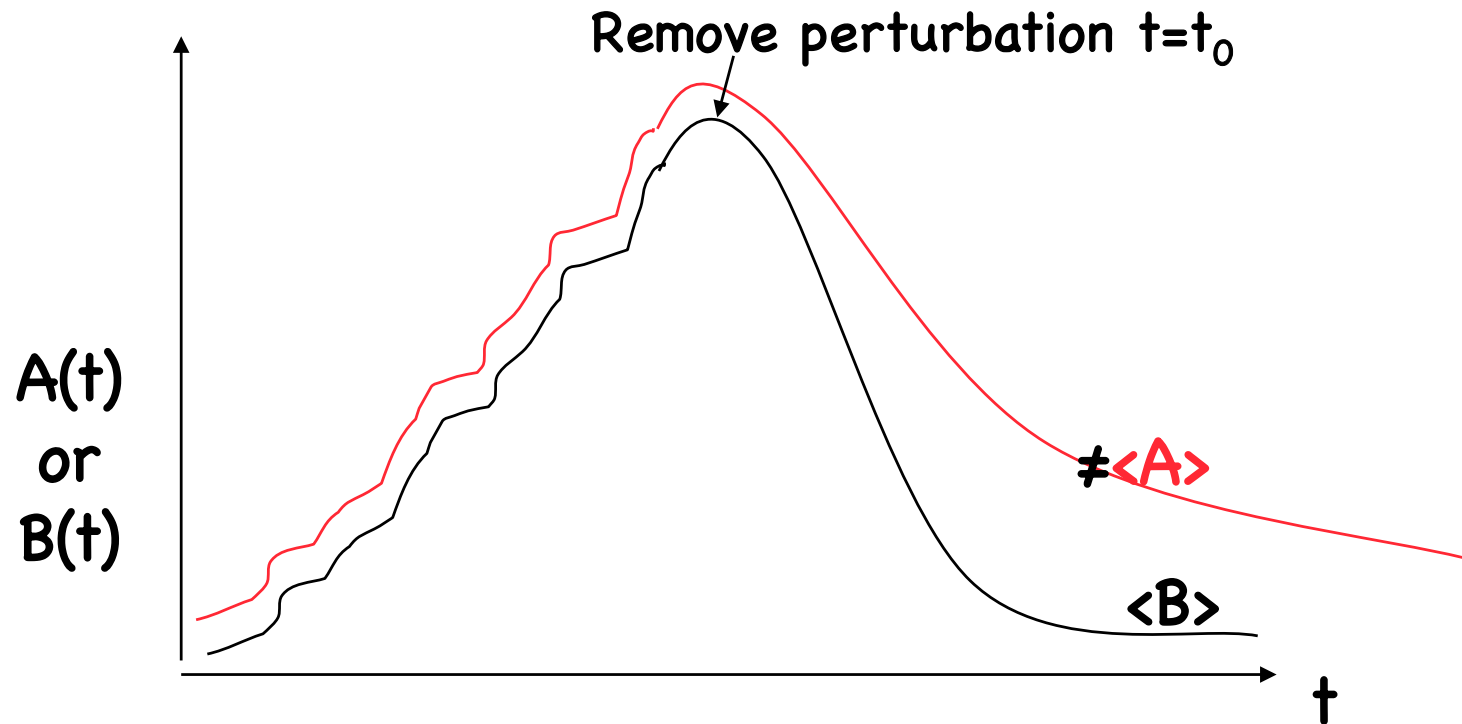


Separation of Scales in Time and Space

Certain degrees of freedom (position, momentum, etc) of different components of a system (protein, water) come to equilibrium more quickly than do others.

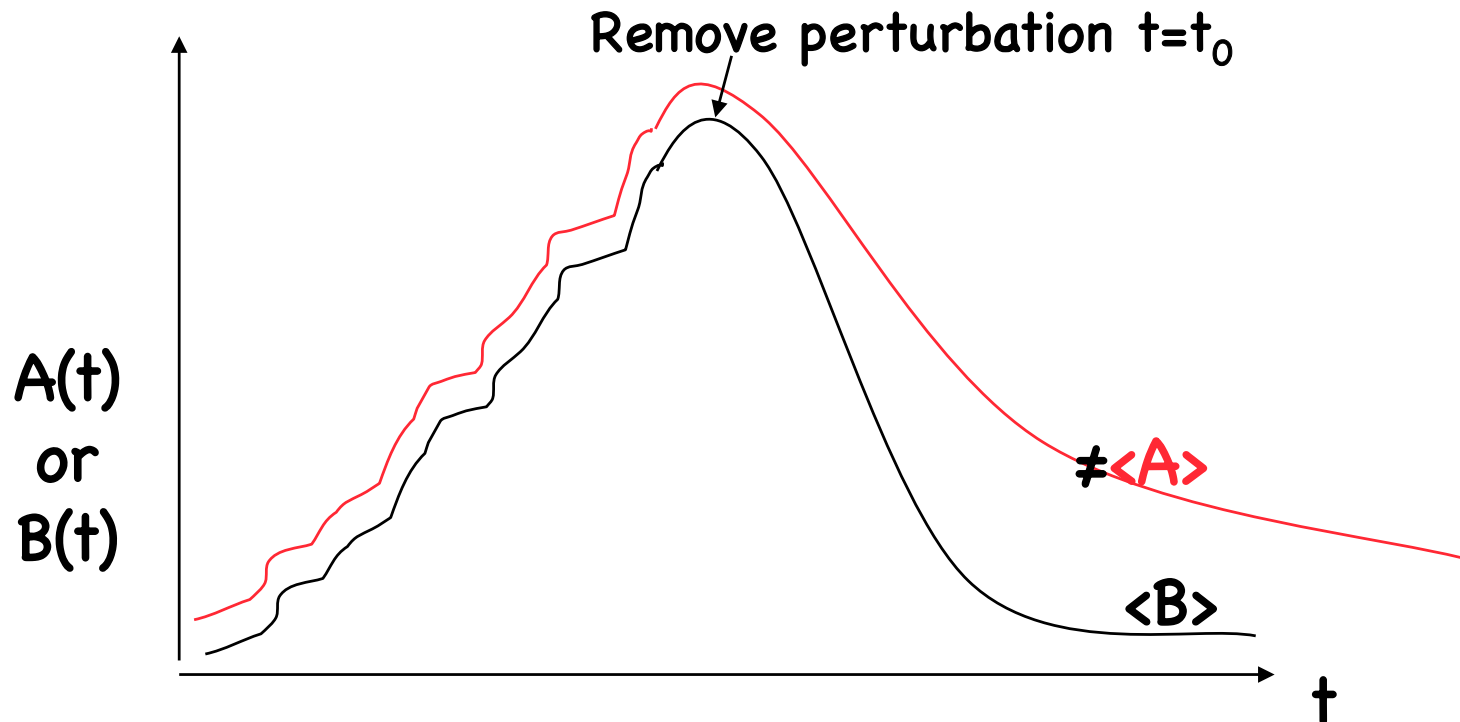
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This difference in rapidly equilibrated regions, and regions whose equilibrium properties are more slowly established, allows for a physically correct separation of spatial scales and time scales.

Coarse-Graining

Given protein (A) and solvent (B) degrees of freedom, let's follow equilibrium property, O , of protein:

$$\langle O \rangle = \frac{\int_{-\infty}^{\infty} d\vec{r}_A^N \int_{-\infty}^{\infty} d\vec{r}_B^N O(\vec{r}_A^N, \vec{r}_B^N) \exp[-\beta H(\vec{r}_A^N, \vec{r}_B^N)]}{(Q =) \int_{-\infty}^{\infty} d\vec{r}_A^N \int_{-\infty}^{\infty} d\vec{r}_B^N \exp[-\beta H(\vec{r}_A^N, \vec{r}_B^N)]}$$

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Our effective Hamiltonian is now in terms of an equilibrium free energy

$$H_{eff} = H_{AA} - kT \ln Q_B(\vec{r}_A^N; N, V, T)$$

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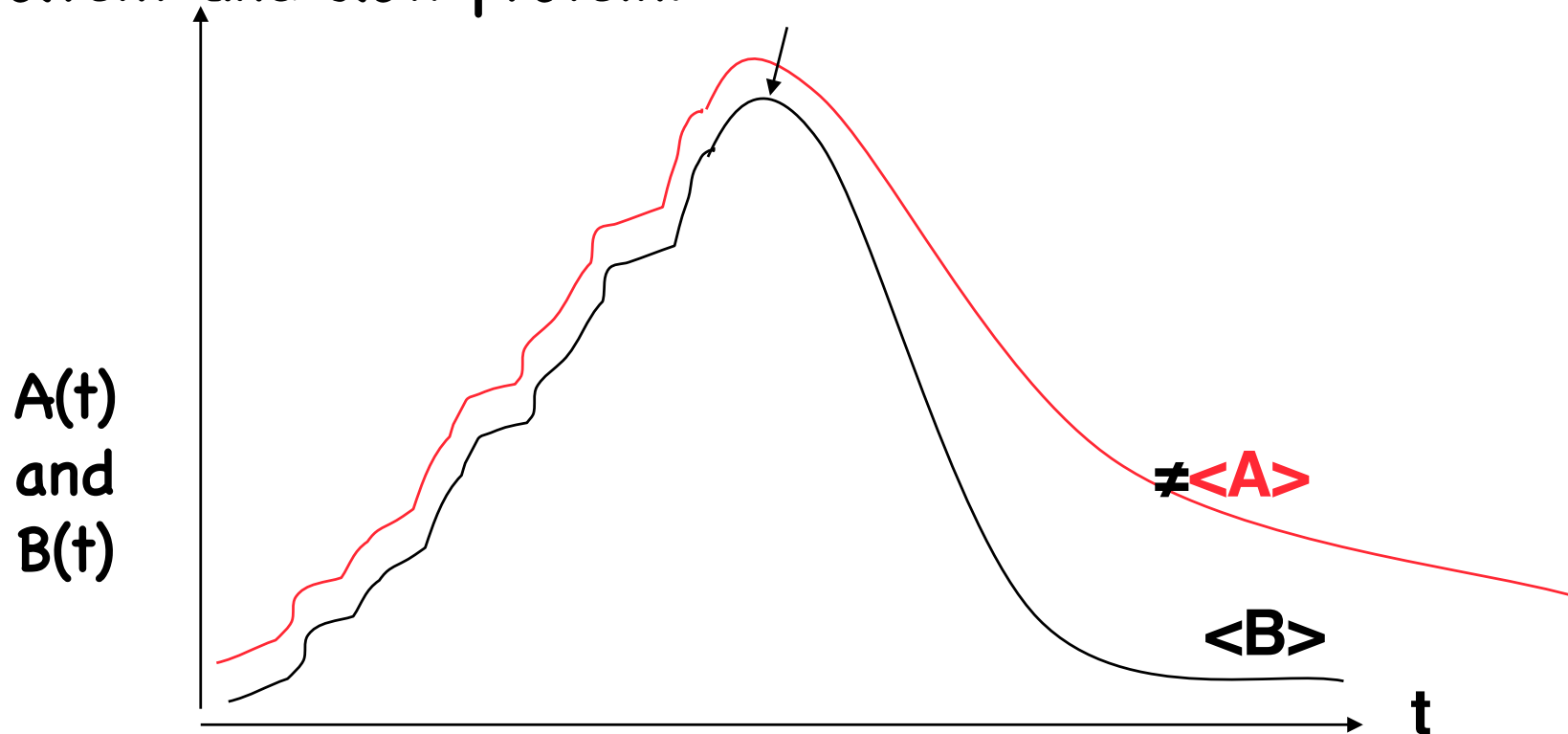
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I.e. degrees of freedom belonging to A have not come to equilibrium but B degrees of freedom have!

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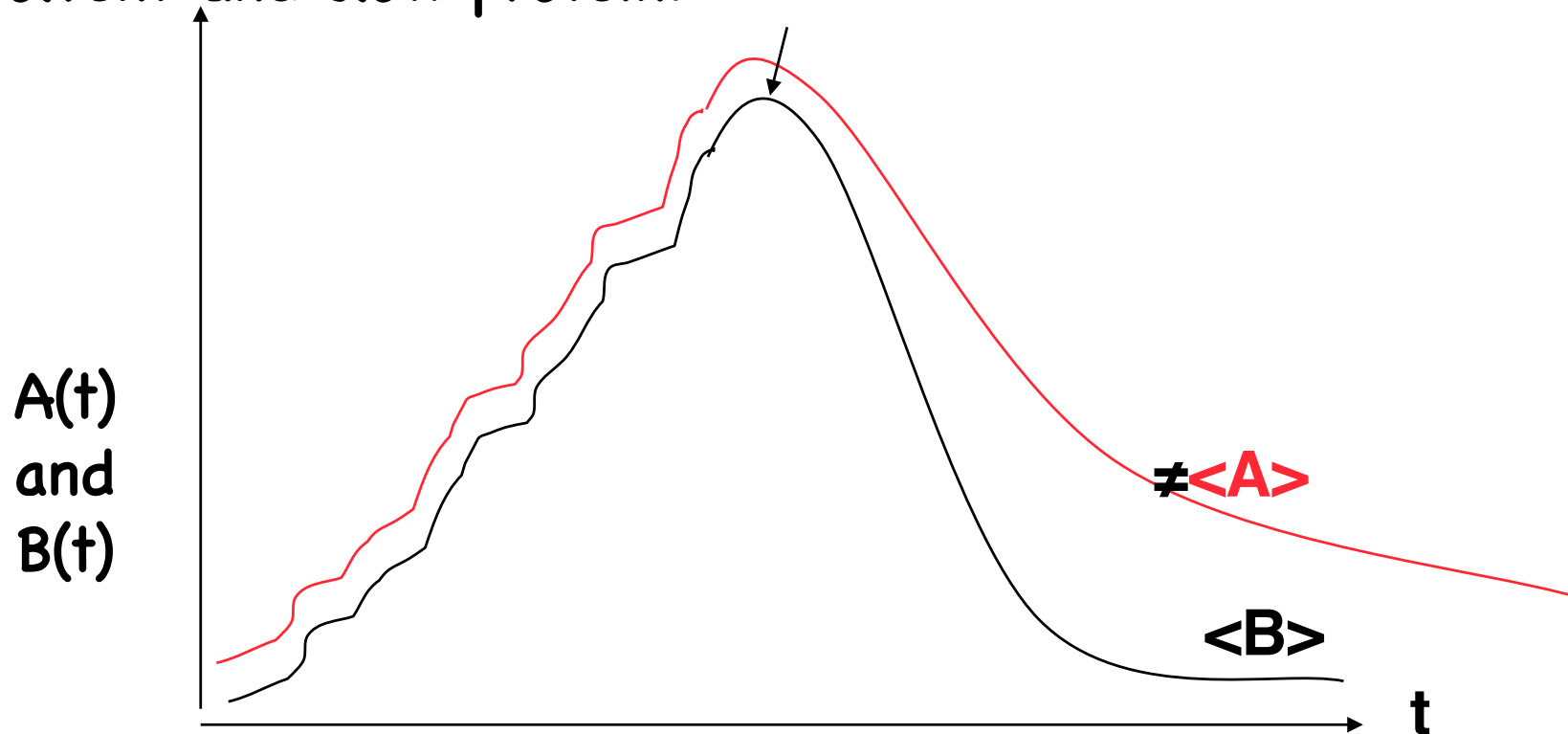
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However our numerical MD must follow the fastest timescale, which is not only costly due to its small Δt , but which is also the least interesting.

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Note that the limiting velocity goes to zero at $t \rightarrow \infty$

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However, the protein will eventually relax back to equilibrium, and the limiting velocity fluctuations should be governed by the usual Maxwell-Boltzmann (equipartition of energy),

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Apparently we are missing an additional force due to solvent. Since solvent is at equilibrium, its motion is only equilibrium fluctuations about its mean.

Fluctuation-Dissipation Theorem

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Solve ordinary first order differential equation by multiplying with exponential integrating factor:

$$\exp(-\gamma t / m) \frac{dv}{dt} - \exp(-\gamma t / m) \frac{\gamma v}{m} = \frac{\delta Q(t)}{m} \exp(-\gamma t / m)$$

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The relationship between friction and random force can be determined in limit as $t \rightarrow \infty$ for which the fluctuations of the velocity should conform to the equipartition of energy

Fluctuation-Dissipation Theorem

Continue solving, and many tedious steps later we arrive at:

$$\langle v^2 \rangle = v(0)^2 \exp(-2\gamma t / m) + \frac{B}{\gamma m} [1 - \exp(-2\gamma t / m)]$$

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The essence of fluctuation-dissipation theorem is that it relates equilibrium thermal fluctuations to out-of-equilibrium quantities.

It states that the dissipative effects needed to restore a non-equilibrium system to equilibrium are derived from equilibrium fluctuations of our system correlated at different times.

Newtonian to Langevin Dynamics

Newton's equation of motion of all degrees of freedom:

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(2) Random fluctuating force described as "white noise" or Gaussian distribution of the random walk

$$\vec{Q}(t) \quad \rightarrow \quad \langle \vec{Q} \rangle = 0 \quad \langle \vec{Q}^2 \rangle \propto D$$

Langevin Dynamics

Let's examine this separation of into fast and slow timescales

$$m\dot{\vec{v}} = -\frac{I}{D^*}\vec{v} + \vec{Q} \quad \text{Ignore systematic force.}$$

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$$m\dot{\vec{v}} = -\frac{l}{D^*}\vec{v} + \vec{Q}$$

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Rewrite in terms of position variable

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$$m \frac{d^2 \langle \vec{r} \cdot \vec{r} \rangle}{dt^2} - 2m \langle \dot{\vec{r}} \cdot \dot{\vec{r}} \rangle = -\frac{1}{D^*} \frac{d \langle \vec{r} \cdot \vec{r} \rangle}{dt} + 2 \langle \vec{r} \cdot \vec{Q} \rangle$$

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Assume that slow particle is at thermal equilibrium with solvent bath so equipartition holds

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$$\frac{d^2 \langle \vec{r} \cdot \vec{r} \rangle}{dt^2} + \frac{1}{mD^*} \frac{d \langle \vec{r} \cdot \vec{r} \rangle}{dt} = 2 \langle \dot{\vec{r}} \cdot \dot{\vec{r}} \rangle + \frac{2}{m} \langle \vec{r} \cdot \vec{Q} \rangle$$

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And now this partial differential can be solved

$$\langle \vec{r} \cdot \vec{r} \rangle = 6kTmD^{*2} \left\{ \frac{t}{mD^*} - \left(1 - e^{-t/mD^*} \right) \right\}$$

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$$\langle \vec{r} \cdot \vec{r} \rangle = \frac{6kTt\tau}{m} = 6kTD^* t = 6Dt \qquad D = ktD^* = \frac{kT}{6\pi\eta a}$$

Brownian Dynamics

Brownian dynamics is a next level of coarse-graining. We don't see the instantaneous acceleration of the Brownian particle: i.e. we are on an effectively longer timescale.

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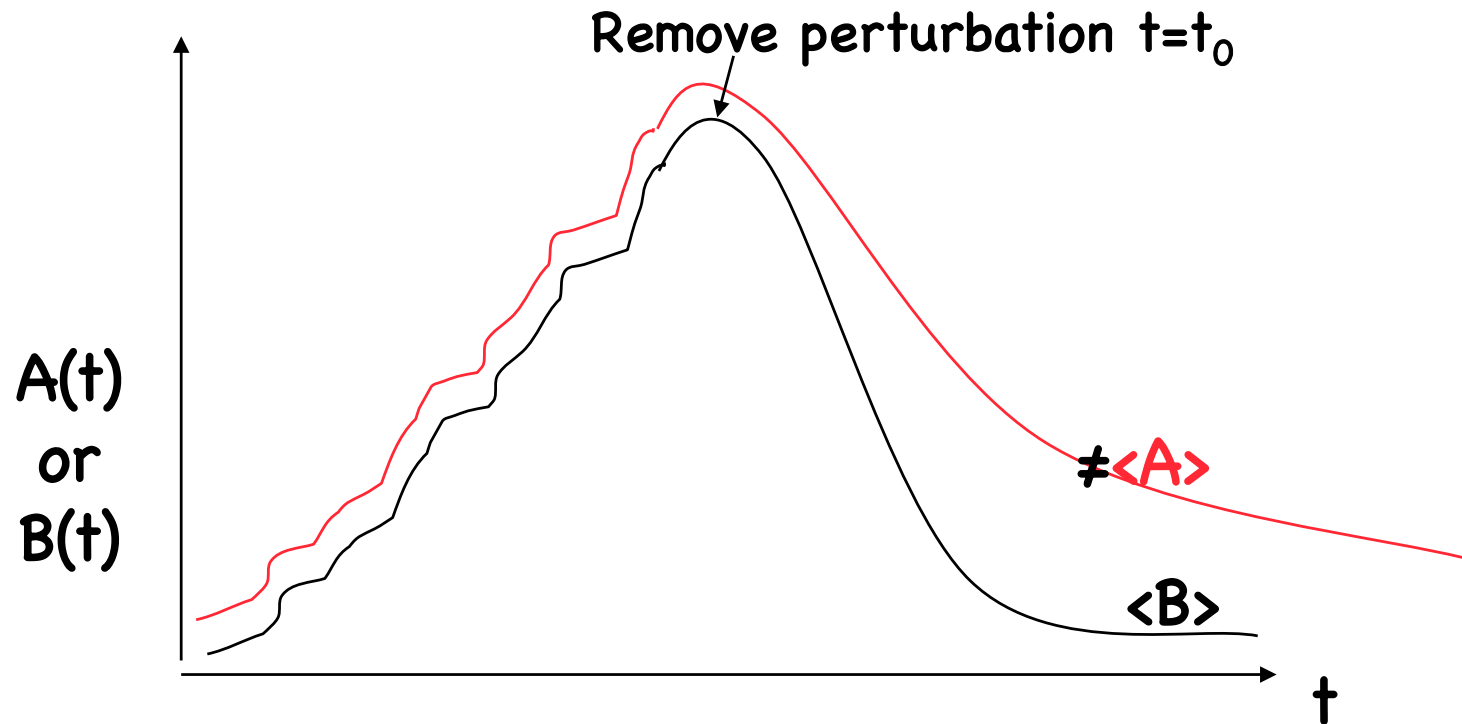
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$\langle \vec{Q}'(t) \rangle = 0$ $\langle \vec{Q}'(t)^2 \rangle = 6D\Delta t$ Draw from a Gaussian distribution of this 2nd moment

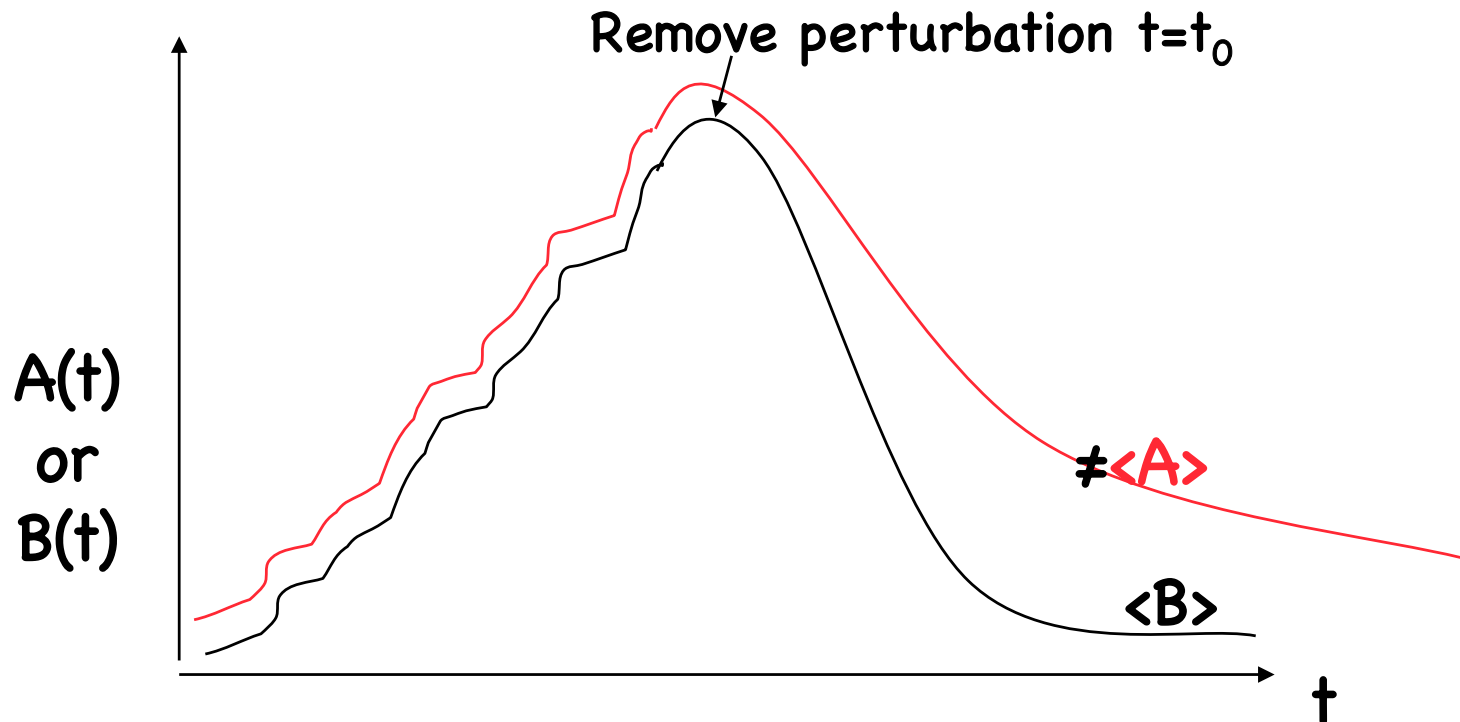
Summary of Coarse Graining

Certain degrees of freedom (position, momentum, etc) of different components of a system (protein, water) come to equilibrium more quickly than do others.



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This difference in rapidly equilibrated regions, and regions whose equilibrium properties are more slowly established, allows for a physically correct separation of spatial scales and time scales.

Separation of Lengthscales

How do we decide what to integrate out?

The work in this paper clearly shows that there is an art in knowing what needs to be preserved in the coarse-graining procedure in order to correctly render the key underlying physical processes one is trying to emulate

M. E. Johnson, T. Head-Gordon, A. A. Louis (2007). Representability problems for coarse-grained water models. *J. Chem. Phys.* 126, 144509-144519.

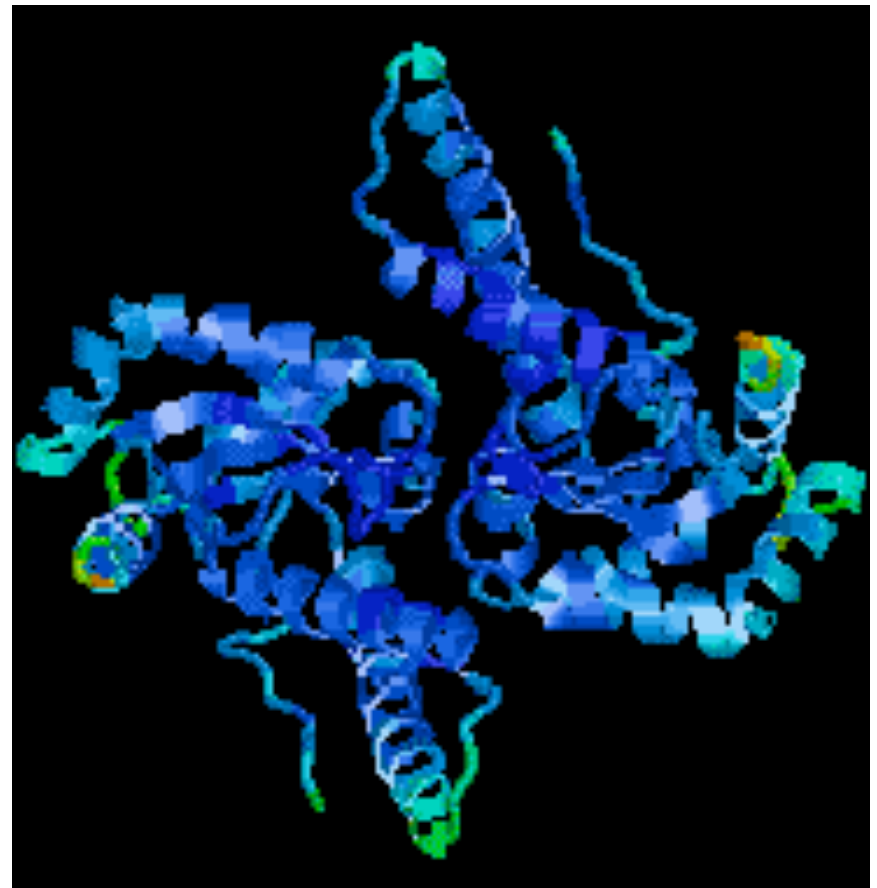
CG electrostatic models for long-range recognition in protein-ligand and protein-protein association

Structural coarse-graining, Henderson uniqueness theorem, and Thermodynamic-dynamic relation for water

Principle of minimal frustration: protein folding, protein aggregation, and protein-membrane interactions

Protein-Ligand Binding: SOD/Superoxide

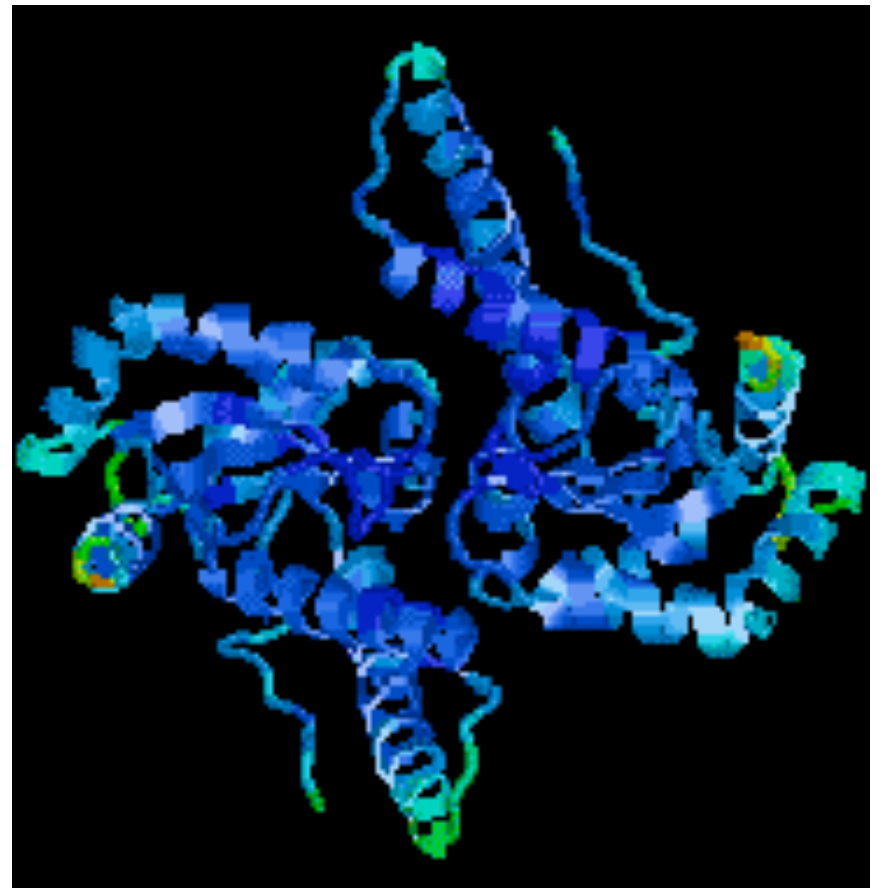
“CuZn superoxide dismutase (CuZnSOD) catalyses the dismutation reaction of the toxic superoxide radical to molecular oxygen and hydrogen peroxide and thus forms a crucial part of the cellular antioxidant defense mechanism”. <http://srs.dl.ac.uk/mbg/sod.html>



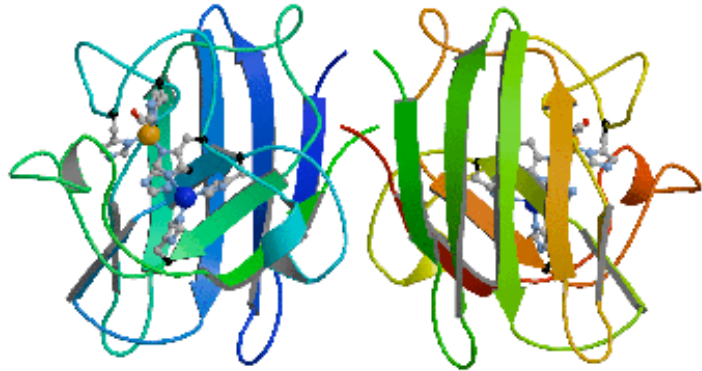
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Single point mutations in SOD are associated with the development of a form of motor neuron disease known as ALS or Lou Gehrig's disease



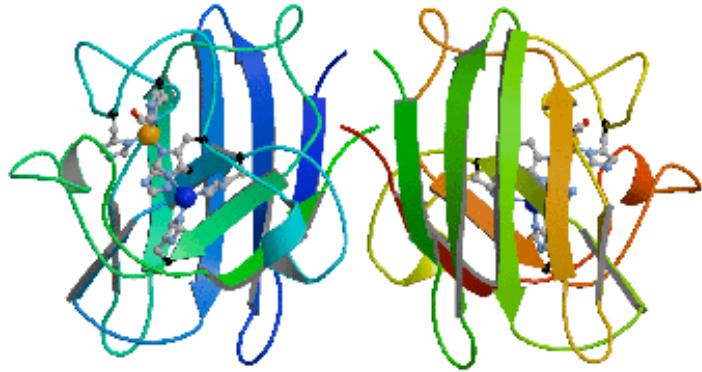
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CuZnSOD is a protein dimer. Each 151 amino acid monomer folds as an eight stranded greek-key beta-barrel connected by three external loops. The overall charge of SOD is $-4e$.

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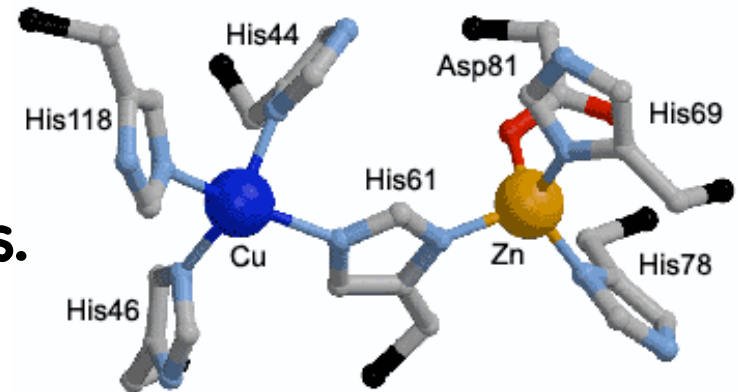
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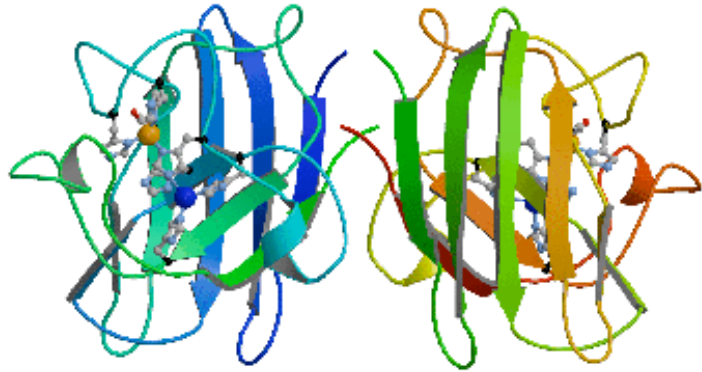
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The active site of SOD lies at base of 15Å deep cavity formed by two loops containing charged residues. The active site consists of 1 Cu and 1 Zn ion bridged by imidazole ring of His61, a feature unique to enzymes of this class.



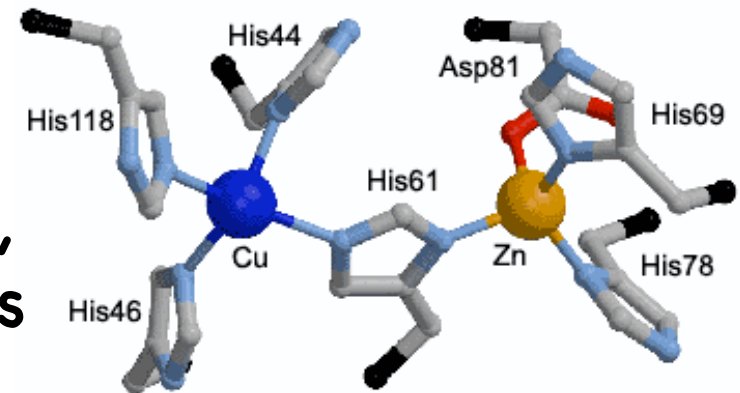
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Diffusion Controlled Chemical Reactions

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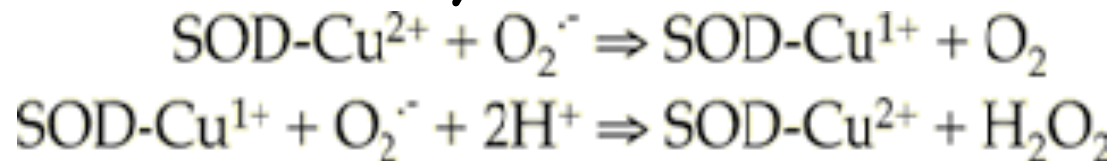
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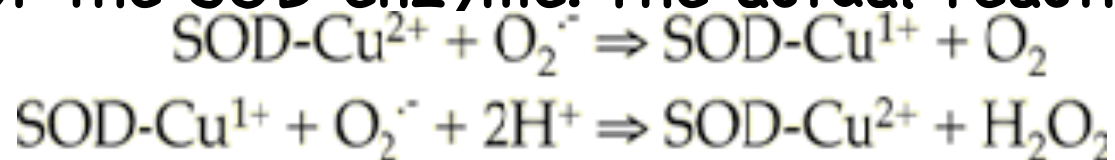


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takes very little time to complete, but instead the rate limiting step is diffusion of the $\text{O}_2^{\cdot-}$ radical to the active site.

SOD/ O_2^- Diffusion Controlled Rates

Thus the SOD catalysis rate of conversion of superoxide to hydrogen peroxide is close to diffusion-controlled. Diffusion controlled rates imply that there is no systematic force between SOD and O_2^- (its all random collisions and friction forces).

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- (2) The rate decreases with increasing salt concentration
- (3) The active site sits at the bottom of a deep cleft that occupies 1/150th of the total protein surface area.

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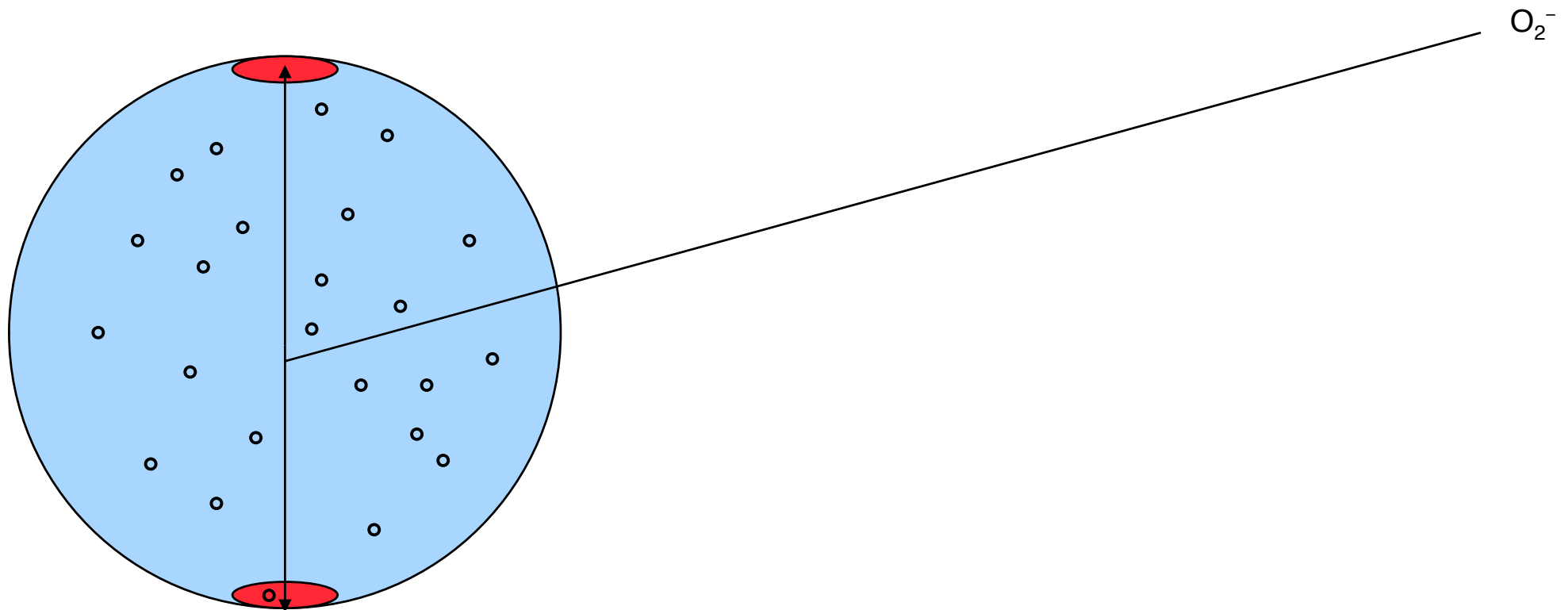
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We propose a possible explanation for poor function for ALS (although recently it has been classified as an aggregation disease)

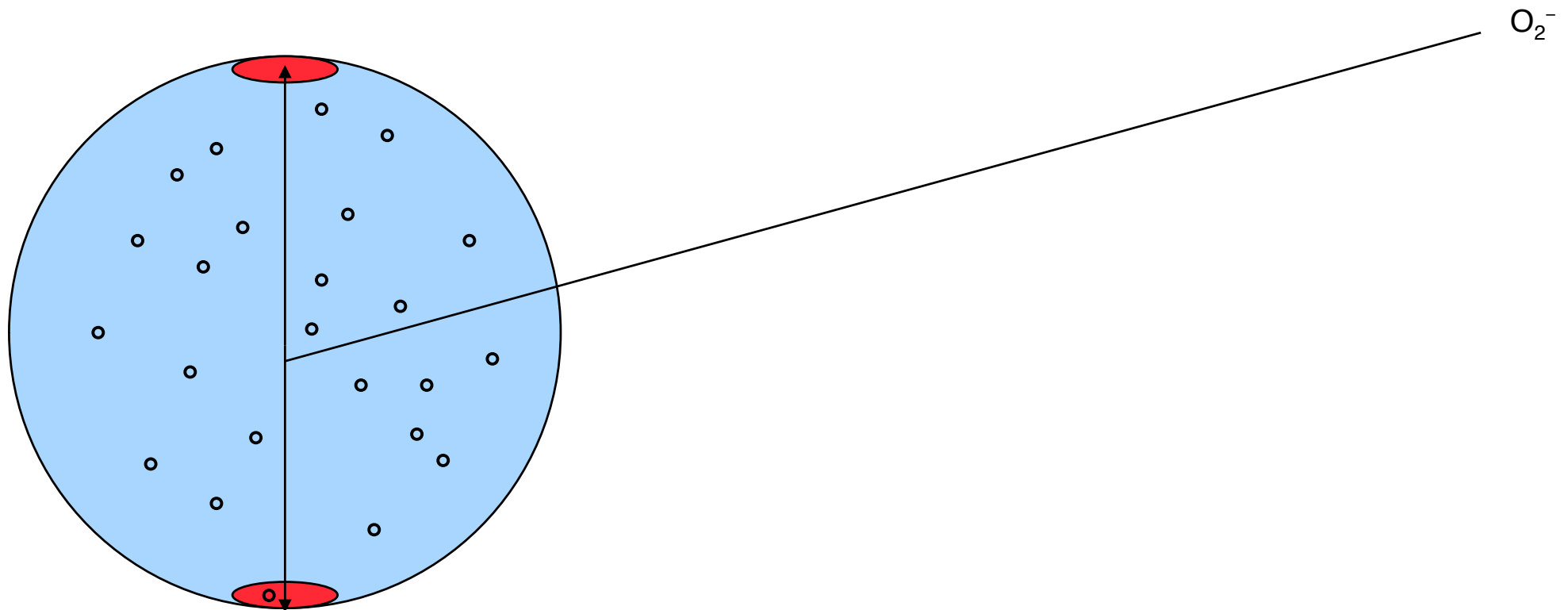
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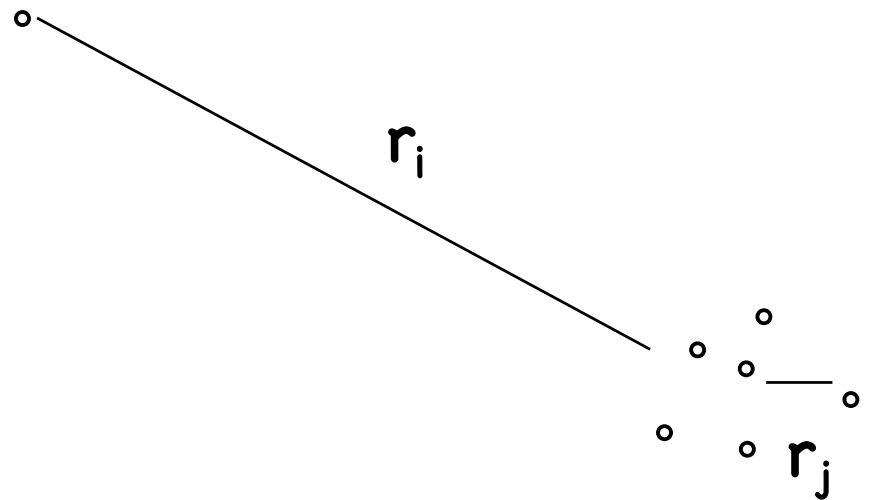
The earliest molecular recognition events are controlled by long-range forces. The electrostatic potential due to 76 charges of SOD can be described as a multipole expansion about the origin:



Multipole Expansion

The electrostatic potential energy at some point r_i due to a distribution of charges near some origin can be described as a multipole expansion

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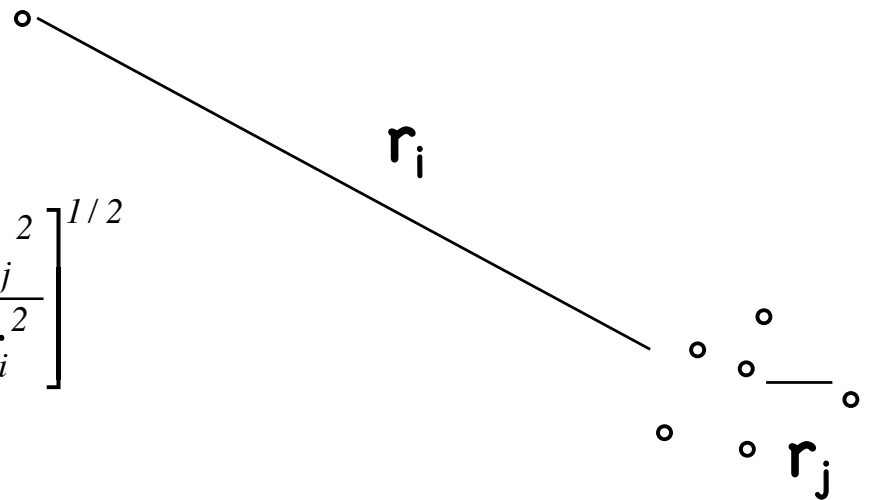


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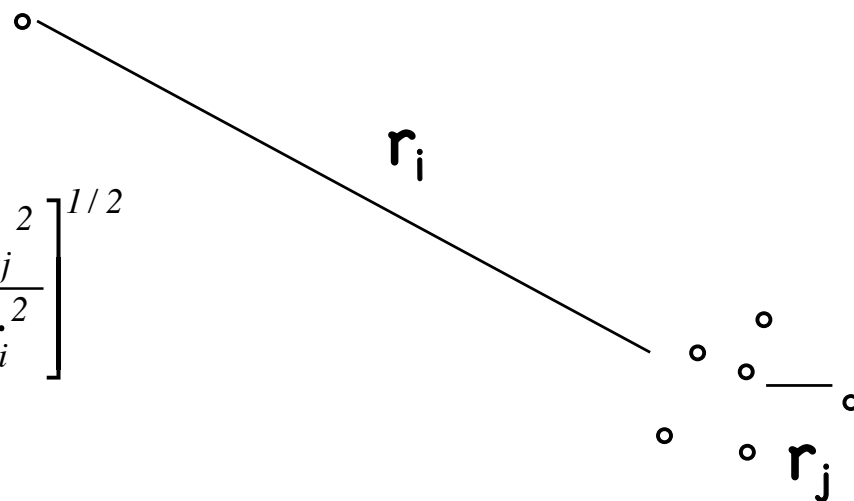
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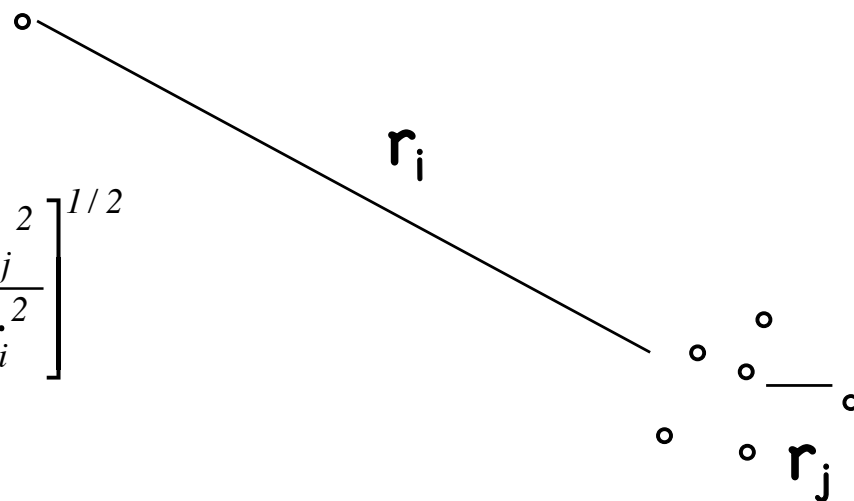
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Multipole Expansion

Since $r_j \ll r_i$, we can reexpress the $1/|r_i - r_j|$ as a binomial expansion (i.e x is small)

$$\left[1 - \frac{2\vec{r}_i \cdot \vec{r}_j}{r_i^2} + \frac{r_j^2}{r_i^2} \right]^{-1/2} = [1 - x]^{-1/2} = \left[1 + \frac{1}{2}x + \frac{3}{8}x^2 + \frac{15}{48}x^3 + \dots \right]$$

$$V_{elect} = q_i \sum_j \frac{q_j}{|\vec{r}_i - \vec{r}_j|} = \frac{q_i}{r_i} \sum_j q_j \left[1 + \frac{\vec{r}_i \cdot \vec{r}_j}{r_i^2} - \frac{r_j^2}{2r_i^2} + \frac{3(\vec{r}_i \cdot \vec{r}_j)^2}{2r_i^4} - \frac{3r_j^2(\vec{r}_i \cdot \vec{r}_j)}{2r_i^4} + \frac{3r_j^4}{8r_i^4} \dots \right]$$

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Since $r_j \ll r_i$, last 2 terms are ignored

Multipole Expansion

Reexpress previous equation:

$$V_{elect} = q_i \sum_j \frac{q_j}{|\vec{r}_i - \vec{r}_j|} = \frac{q_i}{r_i} \sum_j q_j \left[1 + \frac{\vec{r}_i \cdot \vec{r}_j}{r_i^2} + \frac{1}{2r_i^4} \left[3(\vec{r}_i \cdot \vec{r}_j)^2 - (\vec{r}_i \cdot \vec{r}_i)(\vec{r}_j \cdot \vec{r}_j) \right] \right]$$

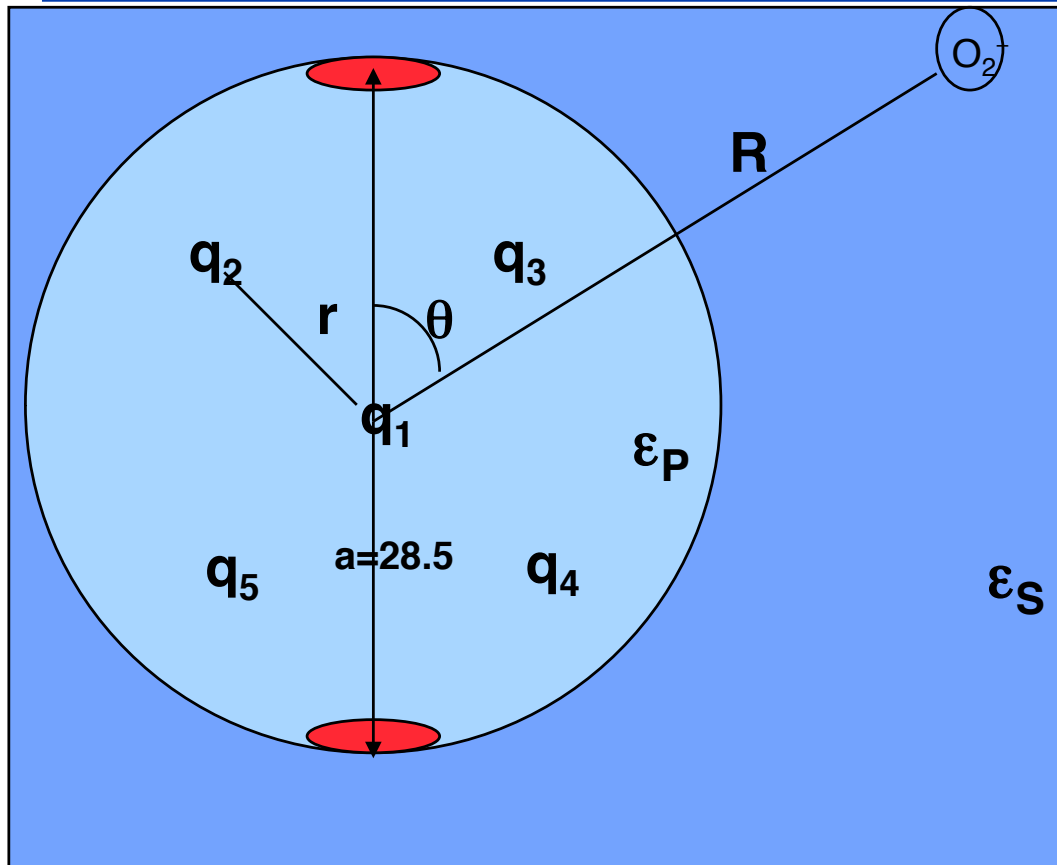
Reexpress this equation in terms of vector components:

$$V_{elect} = q_i \sum_j \frac{q_j}{|\vec{r}_i - \vec{r}_j|} = \frac{q_i}{r_i} \sum_j q_j \left[1 + \frac{\vec{r}_i \cdot \vec{r}_j}{r_i^2} + \frac{1}{2r_i^4} \sum_{\alpha=1}^3 \sum_{\beta=1}^3 \left[r_{i\alpha} r_{i\beta} (3r_{j\alpha} r_{j\beta} - \delta_{\alpha\beta} r_j^2) \right] \right]$$

Which gives us following multipole coefficients:

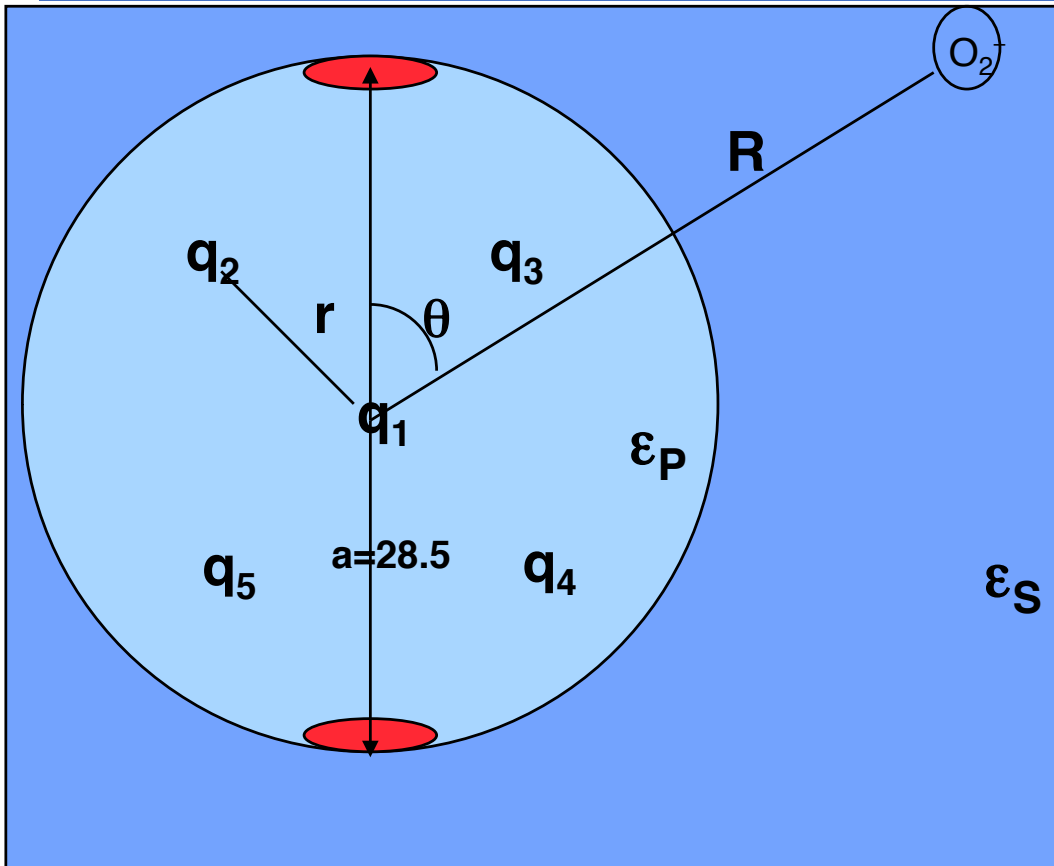
$$Q_M = \sum_j q_j \quad Q_D = \sum_j q_j \vec{r}_i \cdot \vec{r}_j \quad Q_{Q_{\alpha\beta}} = \sum_j \sum_{\alpha} \sum_{\beta} q_j \left[3r_{j\alpha} r_{j\beta} - \delta_{\alpha\beta} r_j^2 \right]$$

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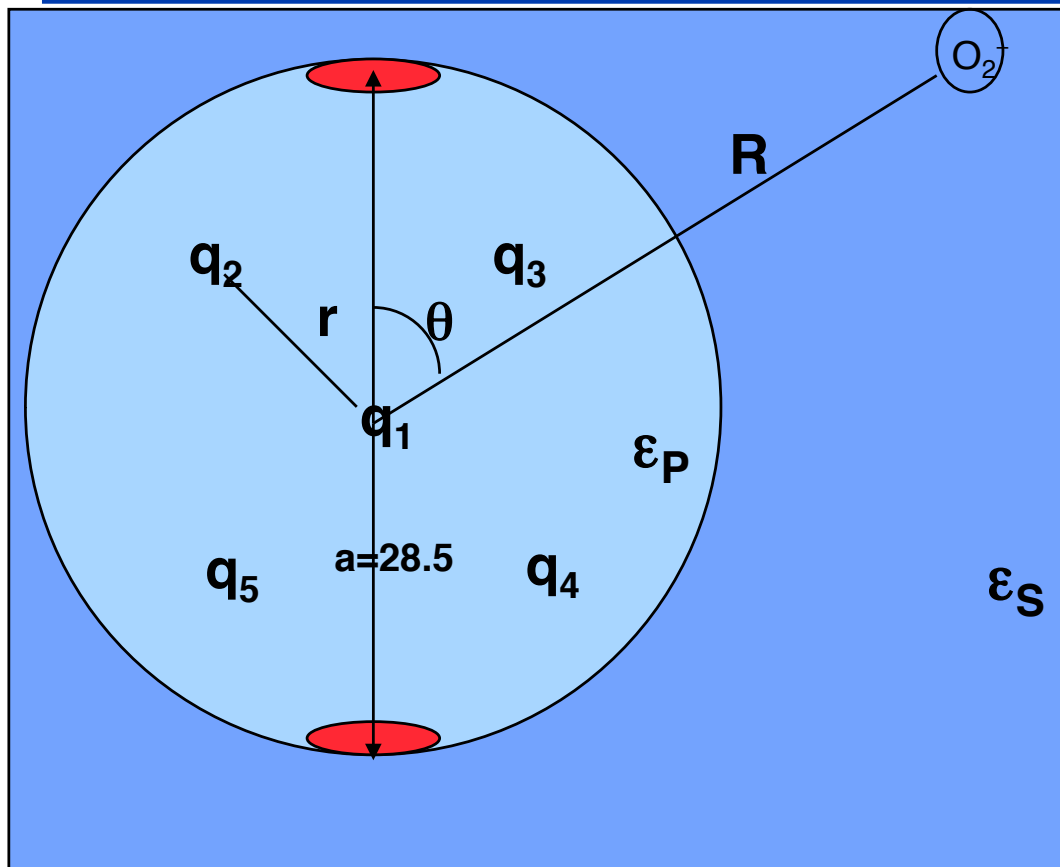
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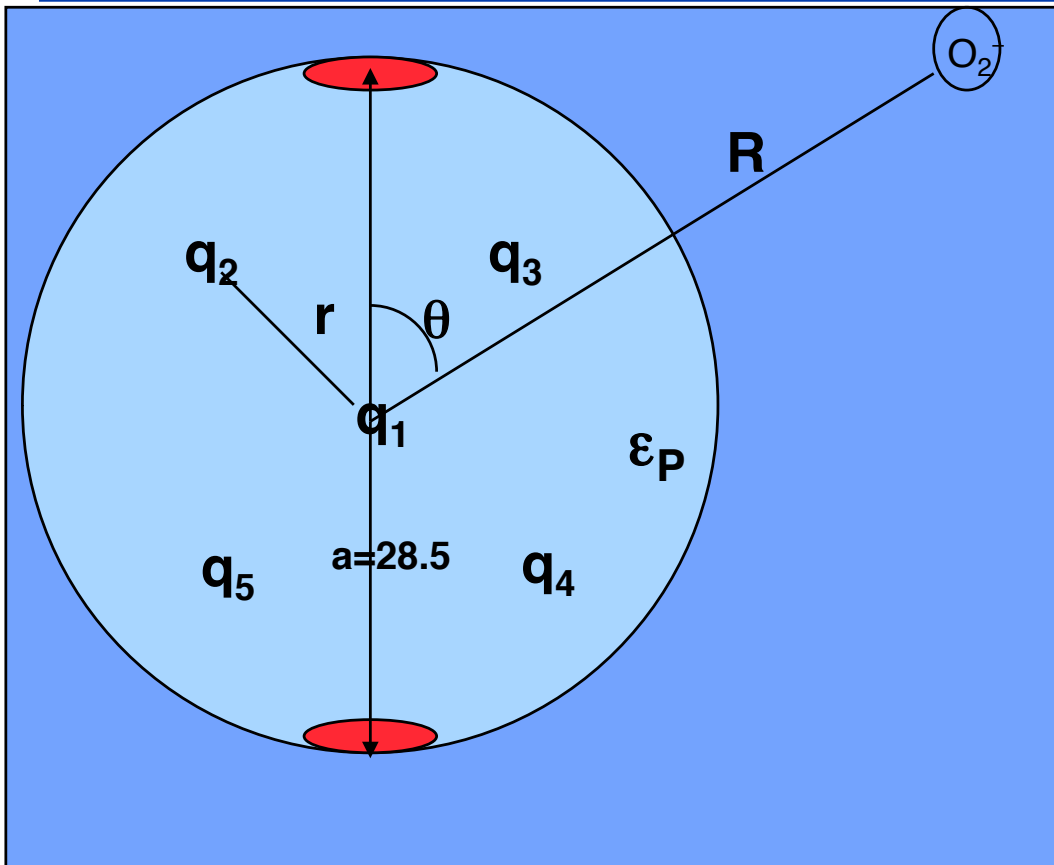
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Now just place five charges in such a way that there multipole expansion reproduces same moments:

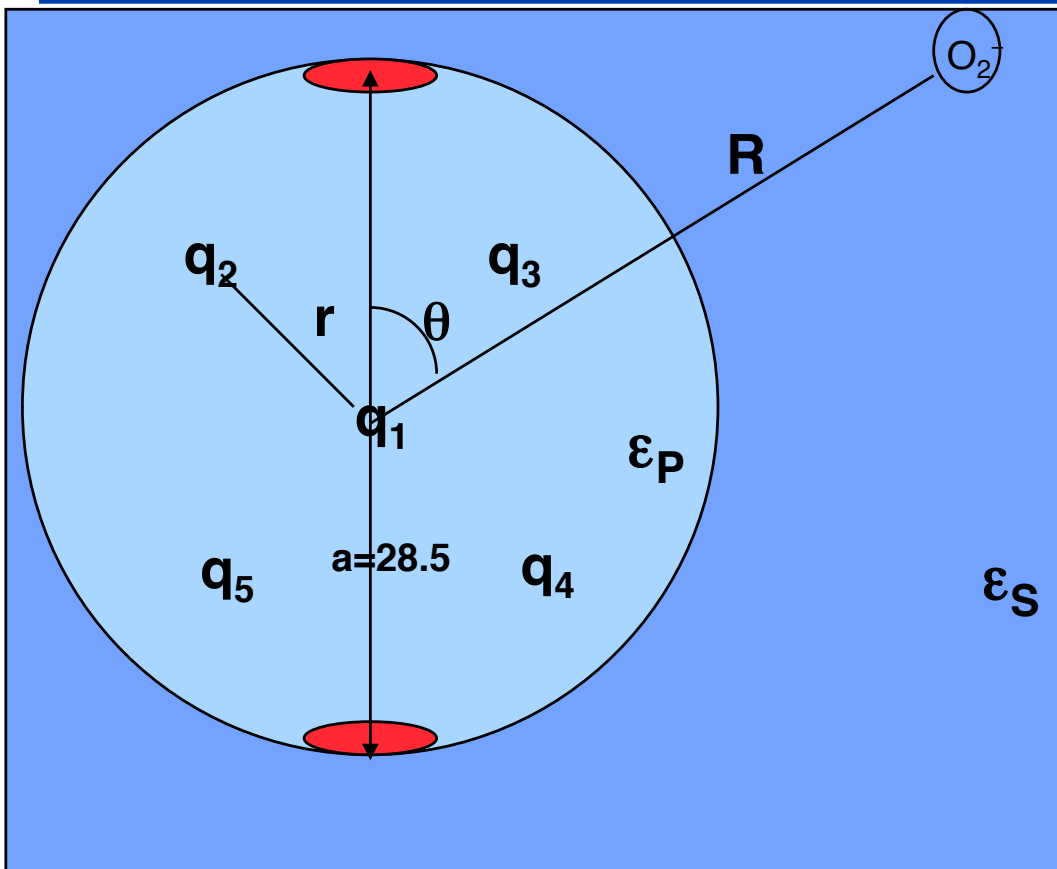
This is equivalent to 5 charges arranged as shown

Coarse-Grained Model of SOD System



Charges are enclosed in sphere of $r=28.5\text{\AA}$ of low dielectric (ϵ_p)

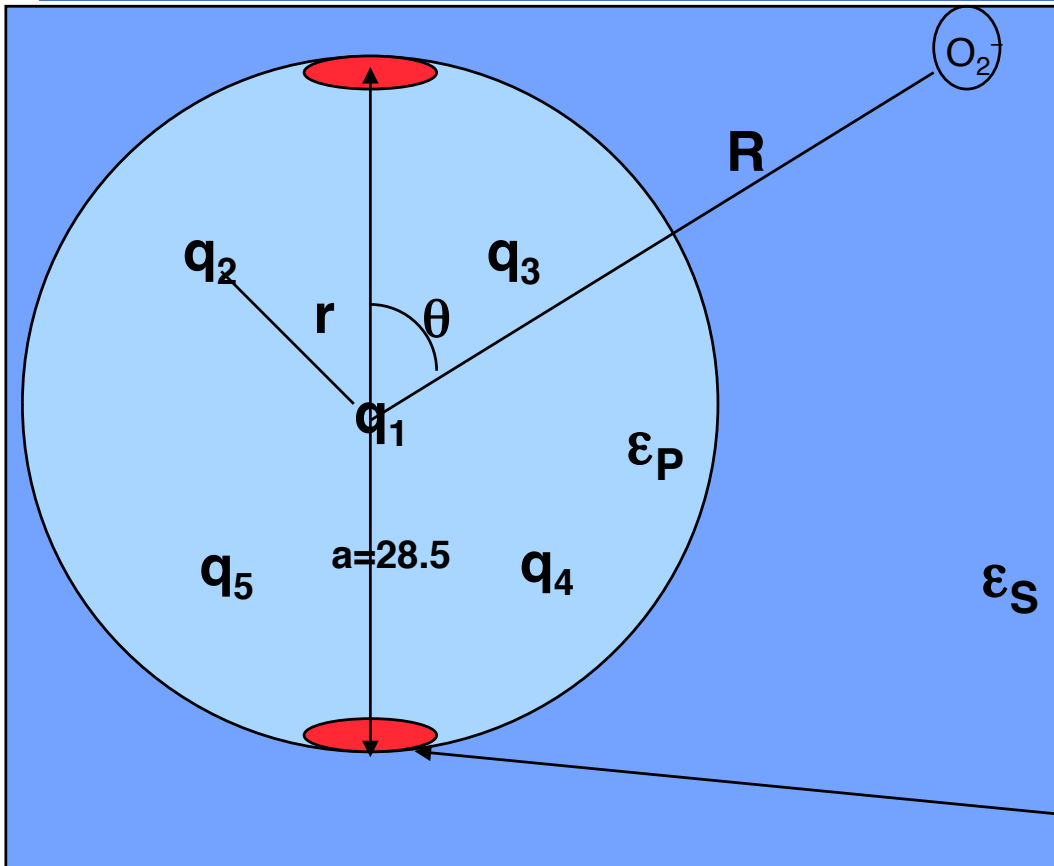
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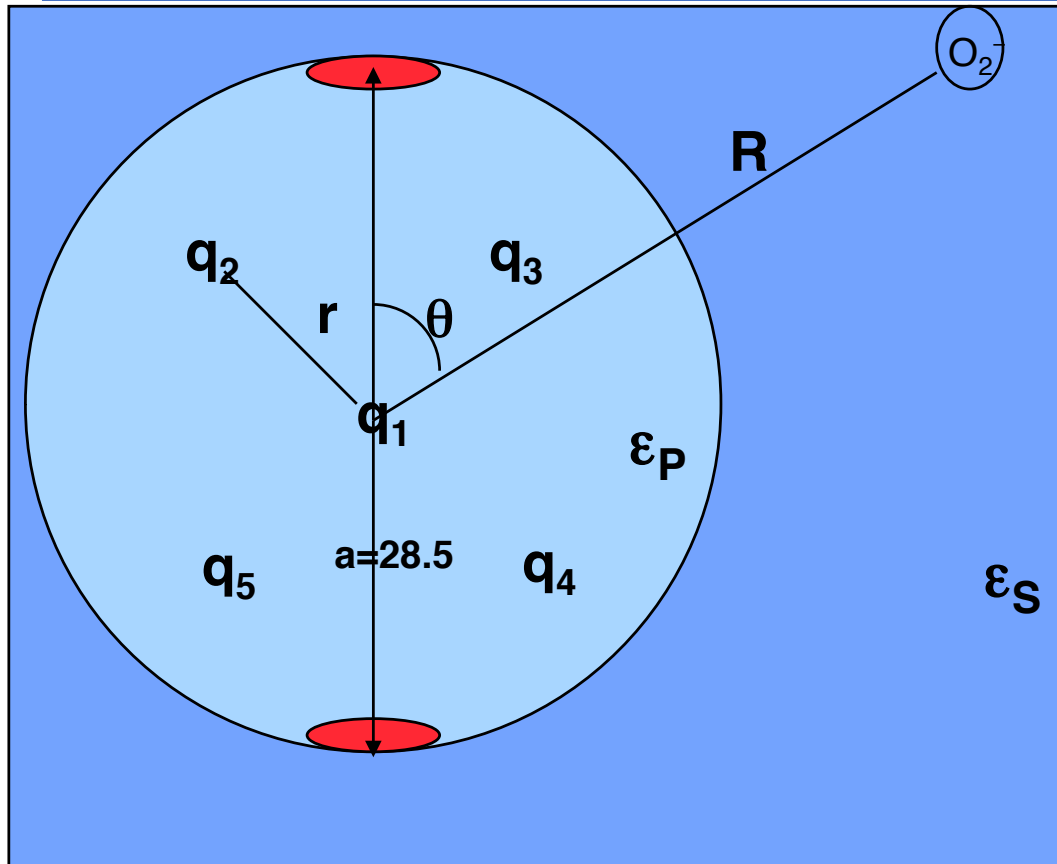


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How to model the salty solution? More explicit ions? Since they diffuse around, we are only interested in their average spatial distributions. Use Debye Huckel theory

The Poisson-Boltzmann Equation

Poisson's Equation:
$$-\nabla[\varepsilon(\mathbf{r})\nabla\Phi(\mathbf{r})] = \rho(\mathbf{r}) = \rho_{fixed}(\mathbf{r}) + \rho_{ions}(\mathbf{r})$$

Mean Field approximation:
$$\rho_{ions}(\mathbf{r}) = \sum_{j=1}^K c_j q_j \exp(-q_j \Phi(\mathbf{r})/kT)$$

Non-linear PB
$$-\nabla[\varepsilon(\mathbf{r})\nabla\Phi(\mathbf{r})] + \underbrace{\sum_{j=1}^K c_j q_j \exp(-q_j \Phi(\mathbf{r})/kT)}_{\text{If 1:1 salt; } q\Phi/kT \ll 1} = \rho_{fixed}(\mathbf{r})$$

Linearized PB

$$-\nabla[\varepsilon(\mathbf{r})\nabla\Phi(\mathbf{r})] + \kappa^2\Phi(\mathbf{r}) = \rho_{fixed}(\mathbf{r})$$

Inverse Debye Length

$$\kappa = \sqrt{\frac{8\pi e^2 C}{\varepsilon_s kT}}$$

Debye-Huckel Theory

What do solutions of the mean field equations look like?

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$$\frac{1}{r^2} \frac{d}{dr} \left(r^2 \frac{d\Phi}{dr} \right) = -\frac{4\pi}{\epsilon} \rho(r)$$

Poissons equation in spherical coordinates

Debye-Huckel Theory

What do solutions of the mean field equations look like?

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Poissons equation in spherical coordinates

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For low [monovalent salts], $\Phi \ll kT$
and do expansion of exponential

Debye-Huckel Theory

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κ is the DH screening length. It is what we will use to control salt concentration in the model.

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$$\Phi_j = \frac{A}{R} \exp(\kappa R)$$

With a solution that will look something like this...

Electrostatic Potential Energy

The general form of the potential for charges inside low dielectric sphere and for outside high dielectric solvent

$$\Psi_{in} = \sum_{n=1}^{\# \text{ charges}} \sum_{m=1} \left(\frac{q_n r^m}{R^{m+1}} + B_m R^m \right) P_m(\cos\theta) \quad 0 < R < a$$

$$\Psi_{out} = \sum_{m=1} \frac{A_m \exp(-\kappa R) K_m(\kappa R) P_m(\cos\theta)}{R^{m+1}} \quad R > a$$

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$P_m(\cos \theta)$ Legendre polynomials

$K_m(\kappa R)$ Modified spherical bessel functions

Electrostatic Potential Energy

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and series expansions converges to $m \sim 10$.

Brownian Motion: Brownian Dynamics

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Solve the numerical BD equation of motion of O_2^- to SOD active sites

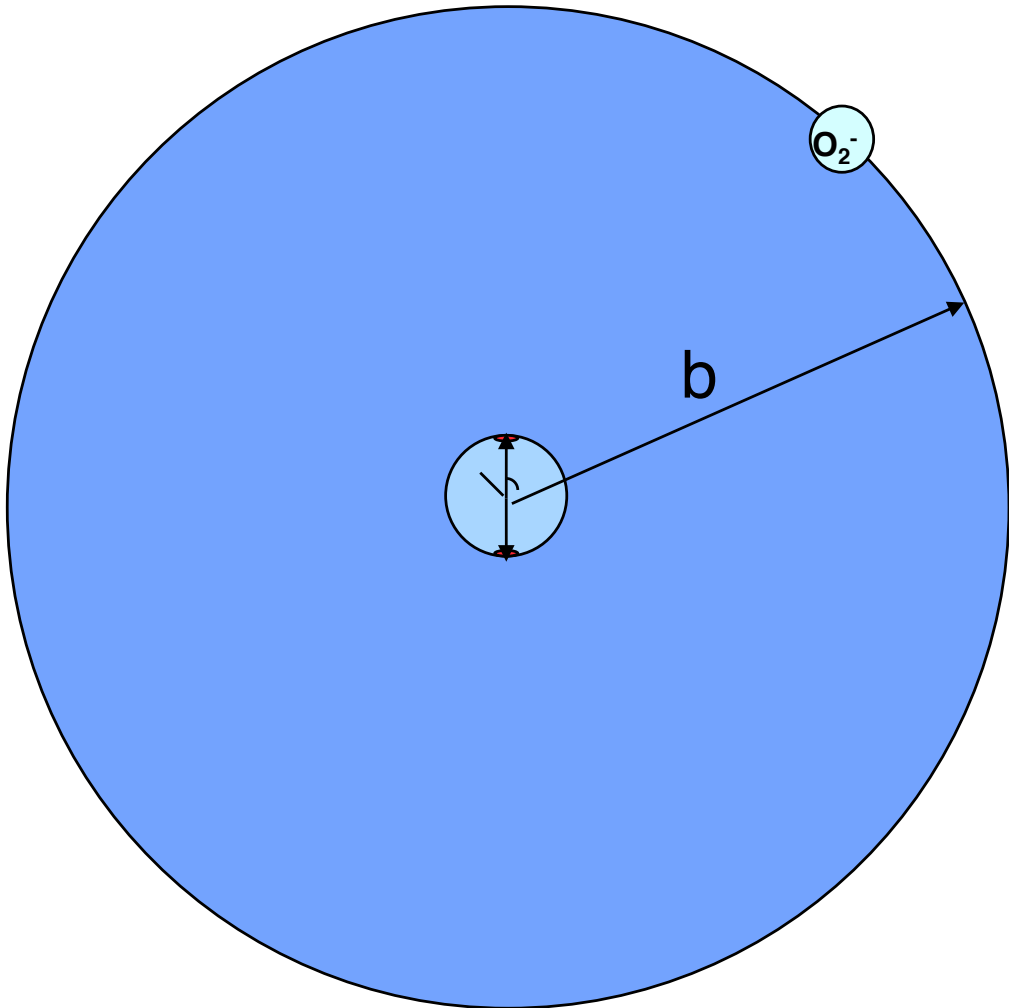
$$R(t + \Delta t) = R(t) + \frac{Dm}{kT} f(t) \Delta t + Q(t)$$

$$D = \frac{(a_{SOD} + a_{O_2^-}) kT}{4\pi\eta a_{SOD} a_{O_2^-}}$$

$$\langle Q(t) \rangle = 0 \quad \langle Q(t)^2 \rangle = 6D\Delta t$$

Simulating association rates of SOD/ O_2^-

O_2^- starts at a distance b from the origin where the potential and therefore the flux at b is isotropic.

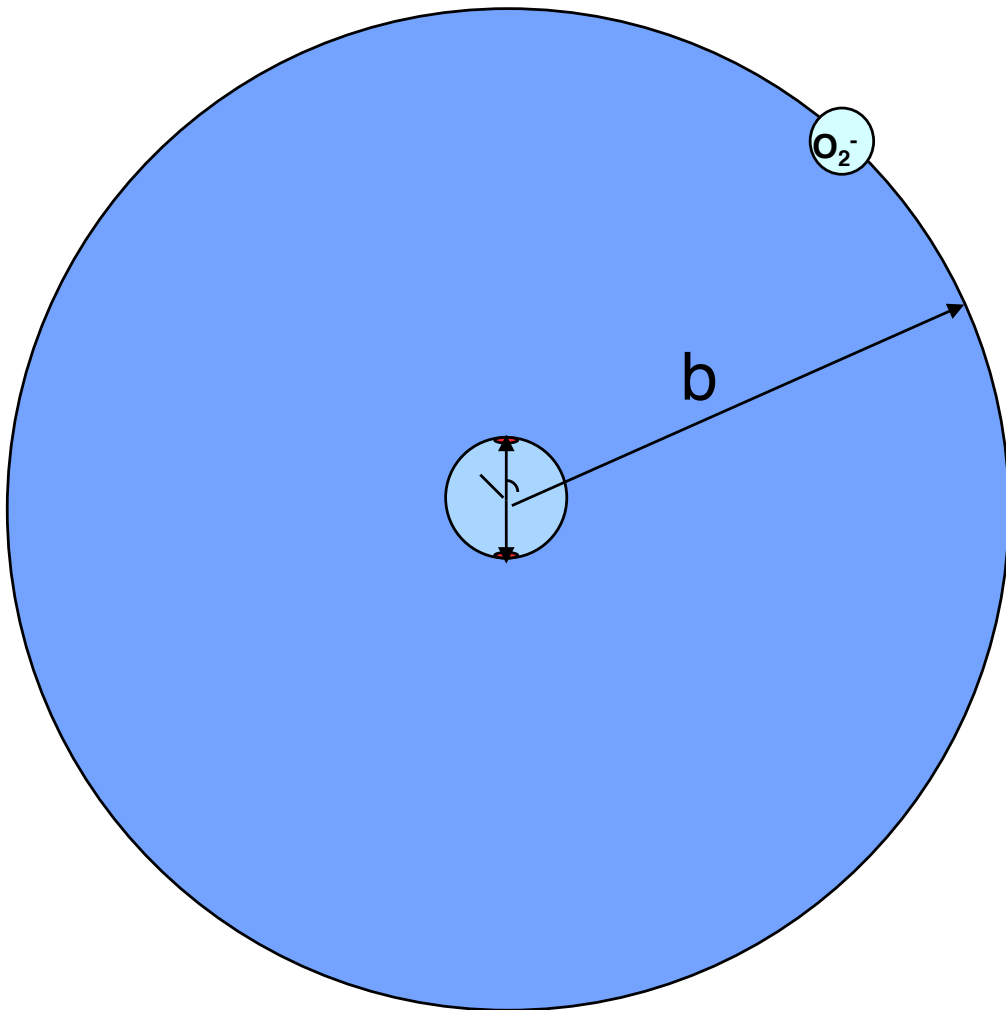


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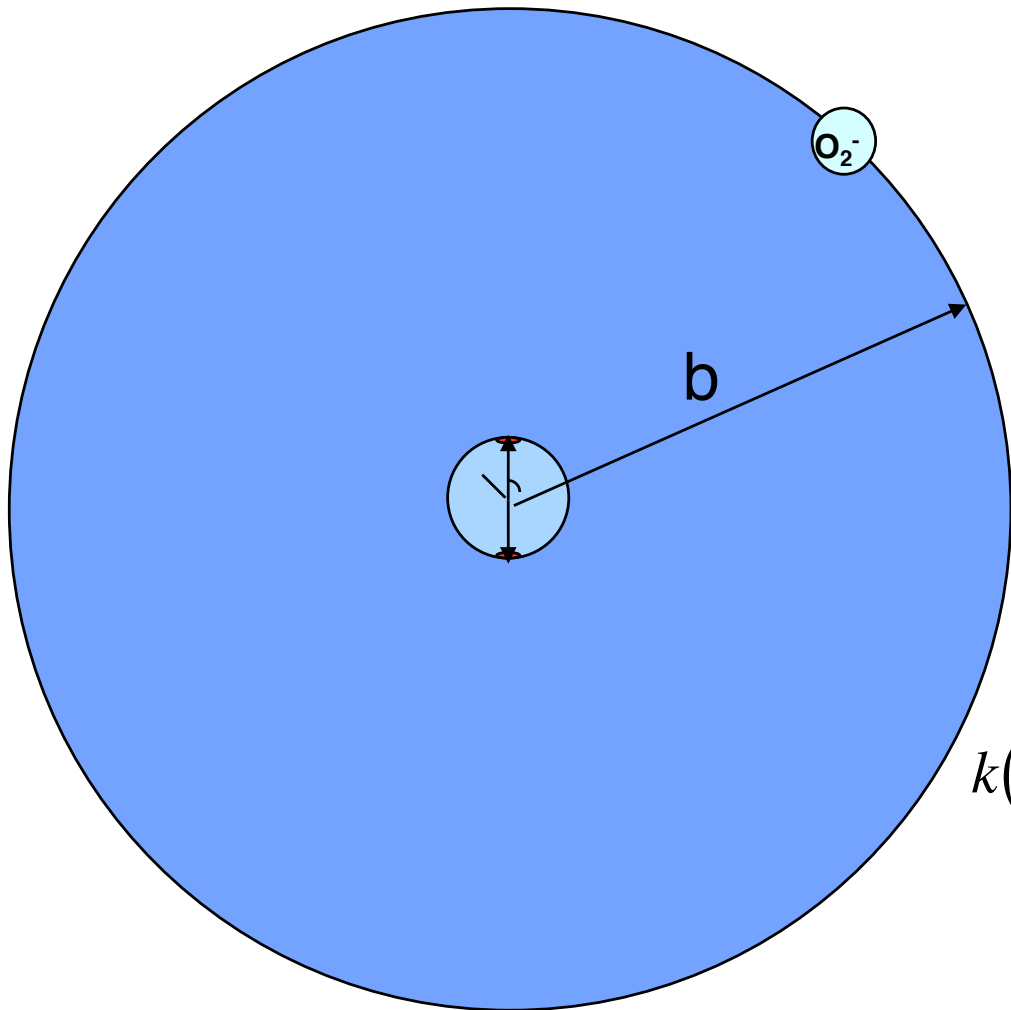
The rate can be expressed as

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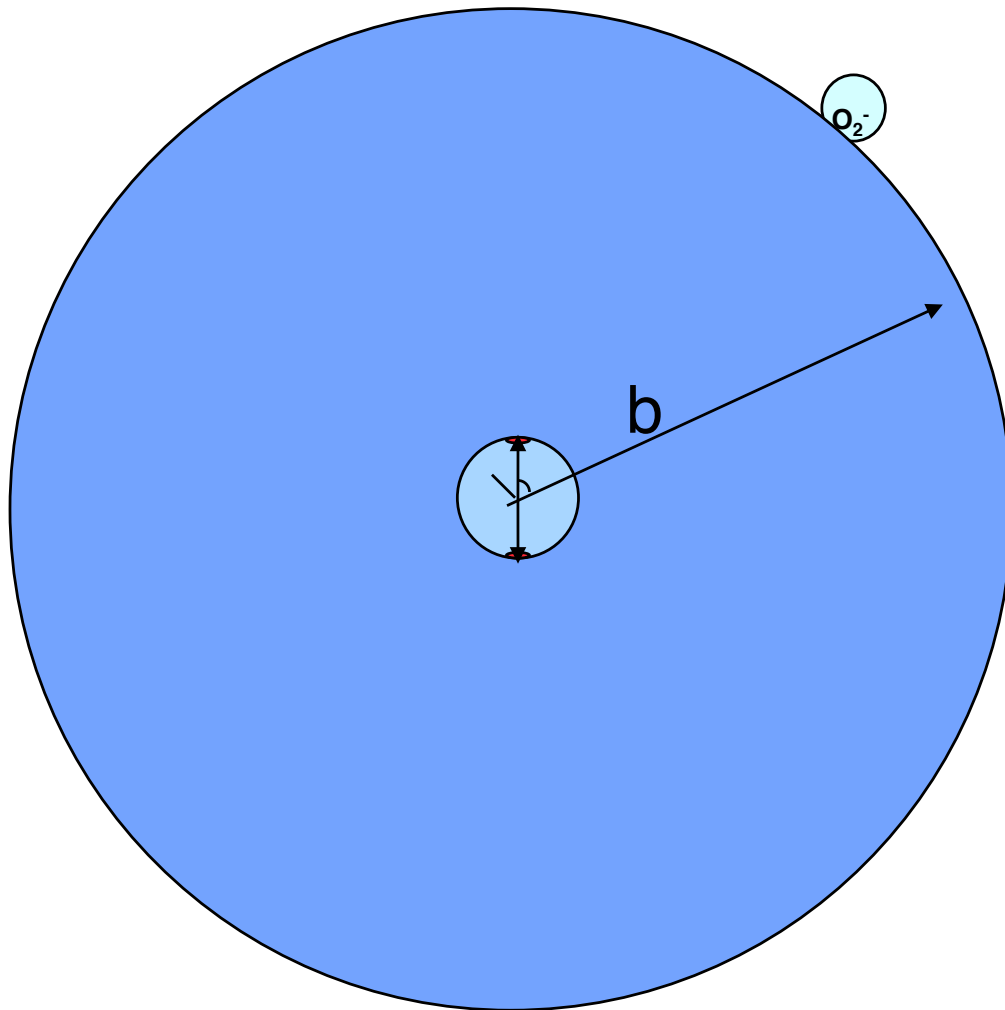
$$k = k(b)P$$

$k(b)$ is rate for reaching b after diffusing from infinity and can be solved analytically

$$k(b) = \left[\frac{1}{4\pi D} \int_b^{\infty} \frac{\exp\left(\beta q_{O_2^-} q_{SOD}^{monopole} / R\right)}{R^2} dR \right]^{-1}$$

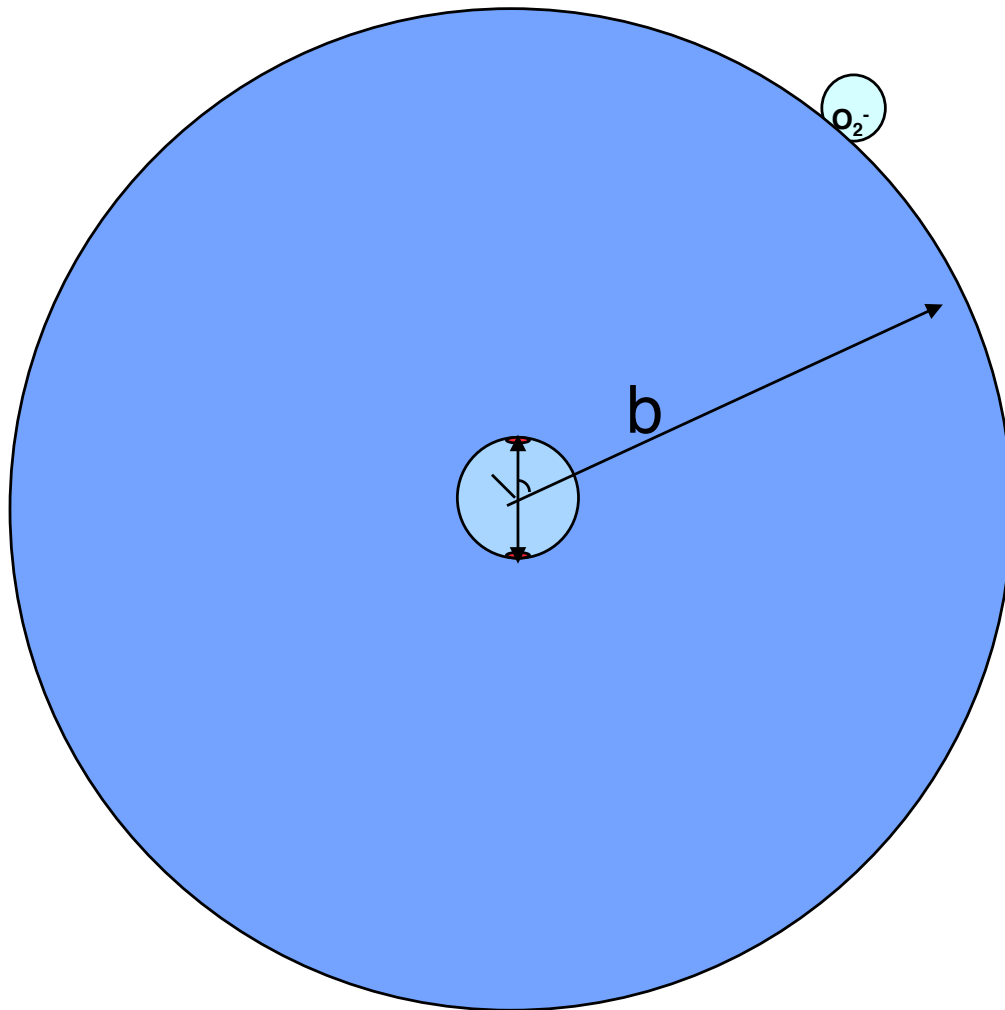
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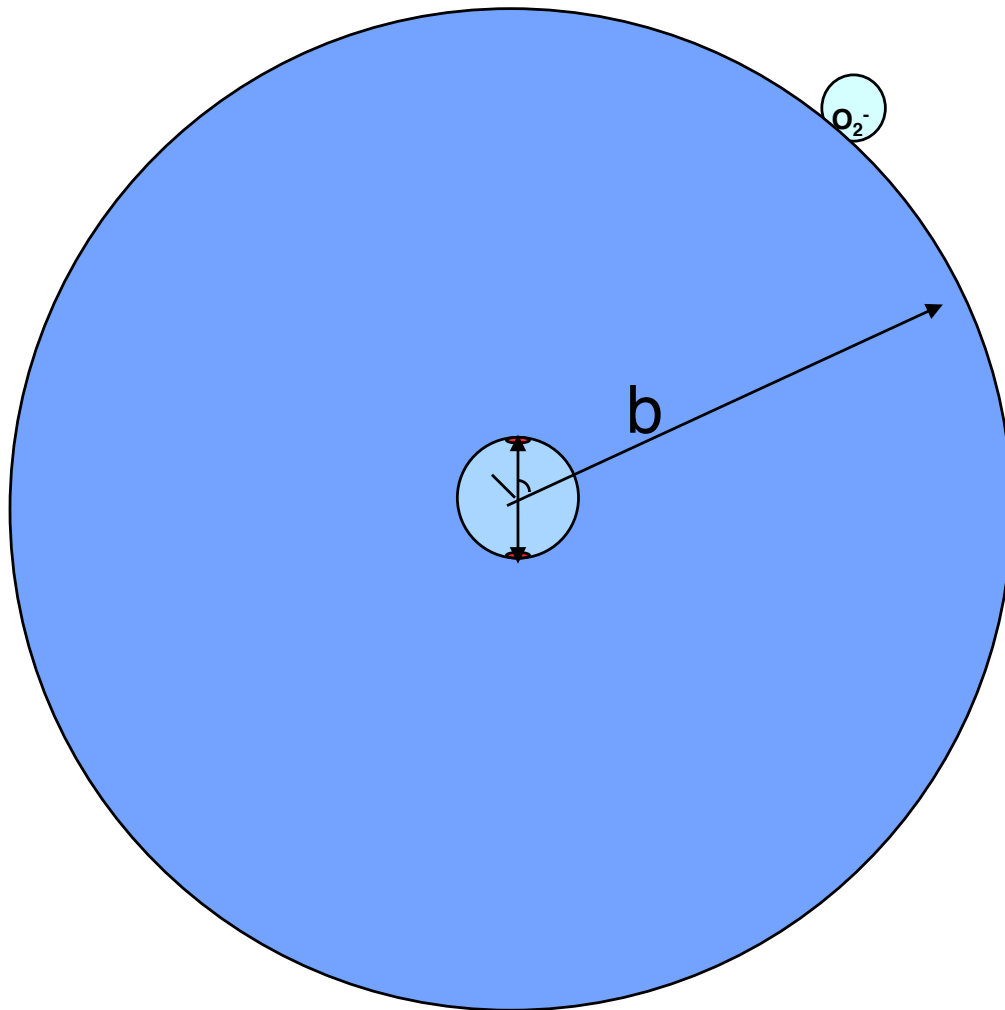
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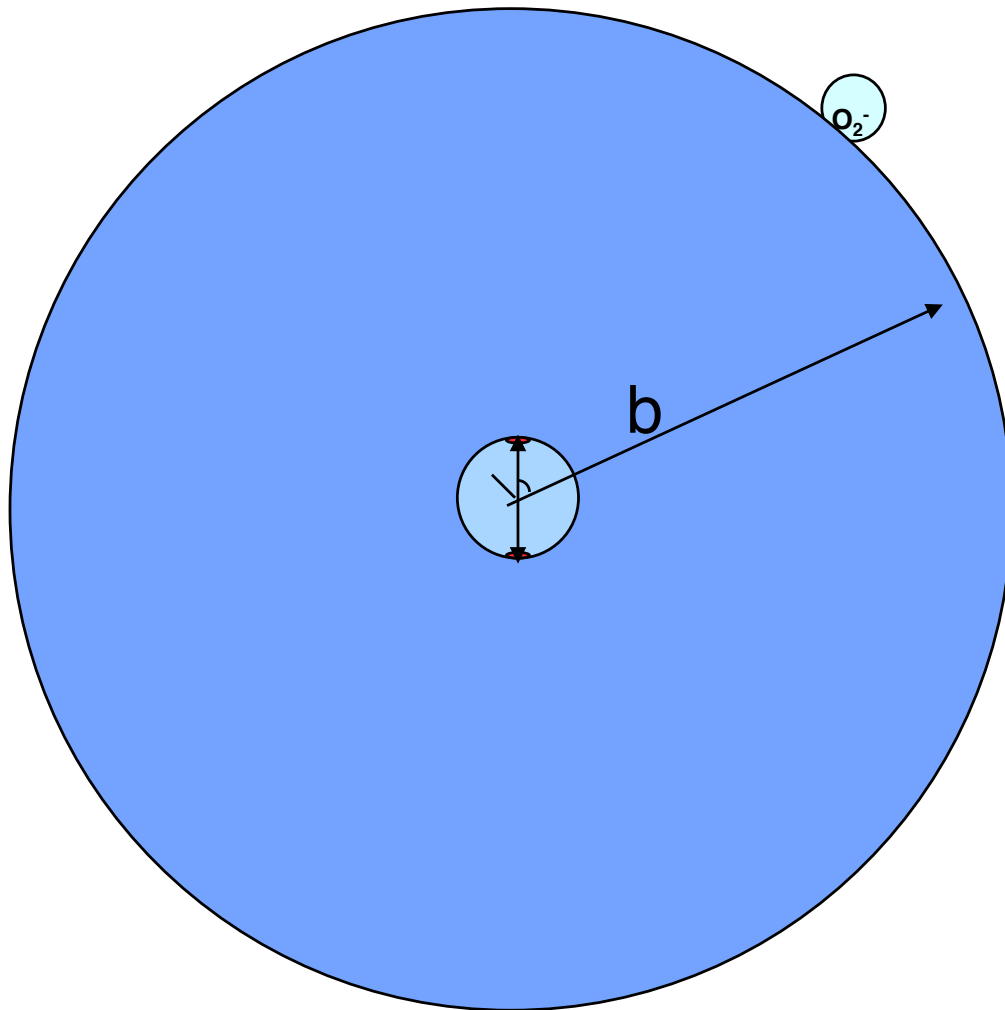


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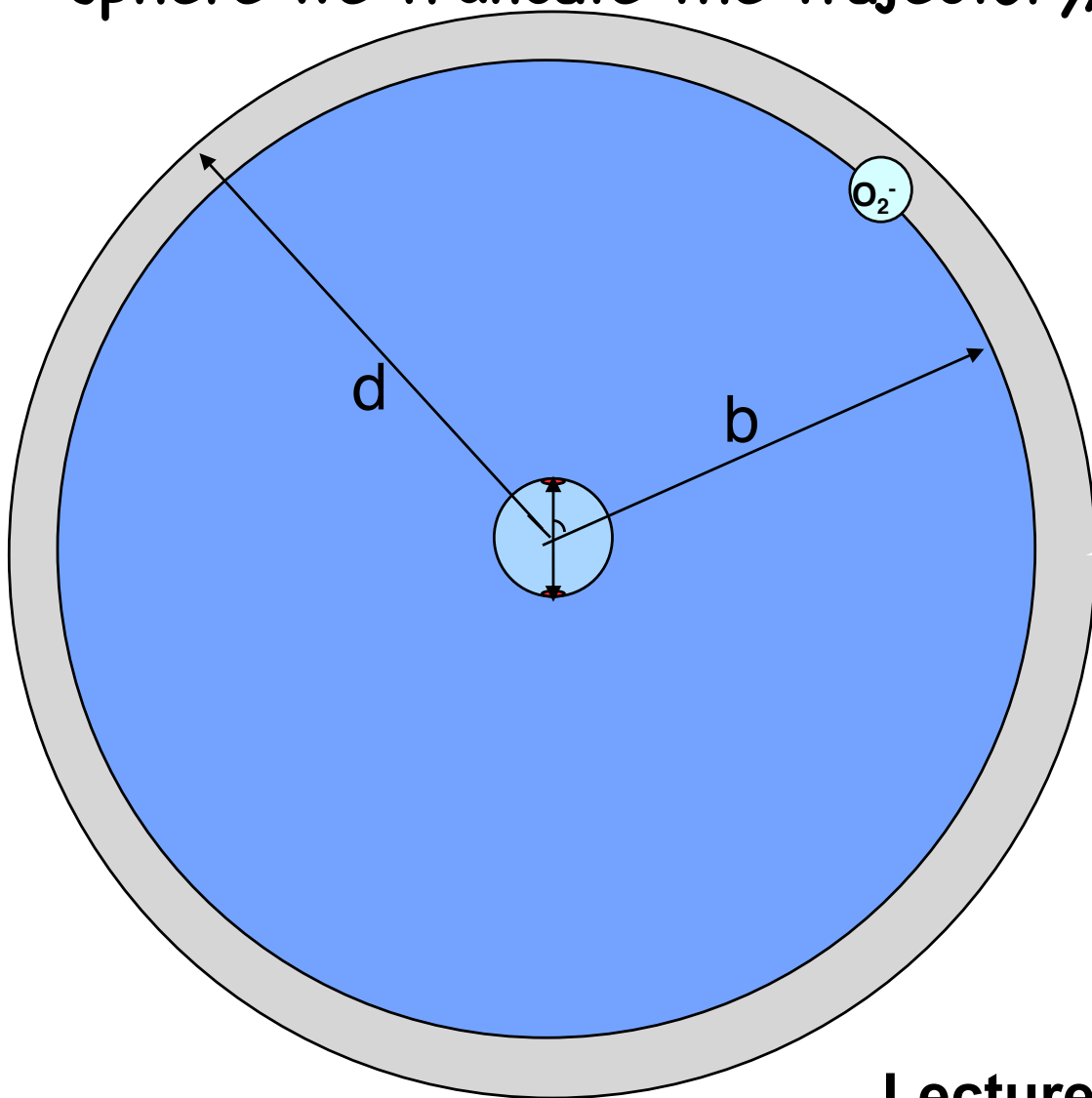
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But given all of these simplifications, it still would be impossible to simulate in an infinite domain.

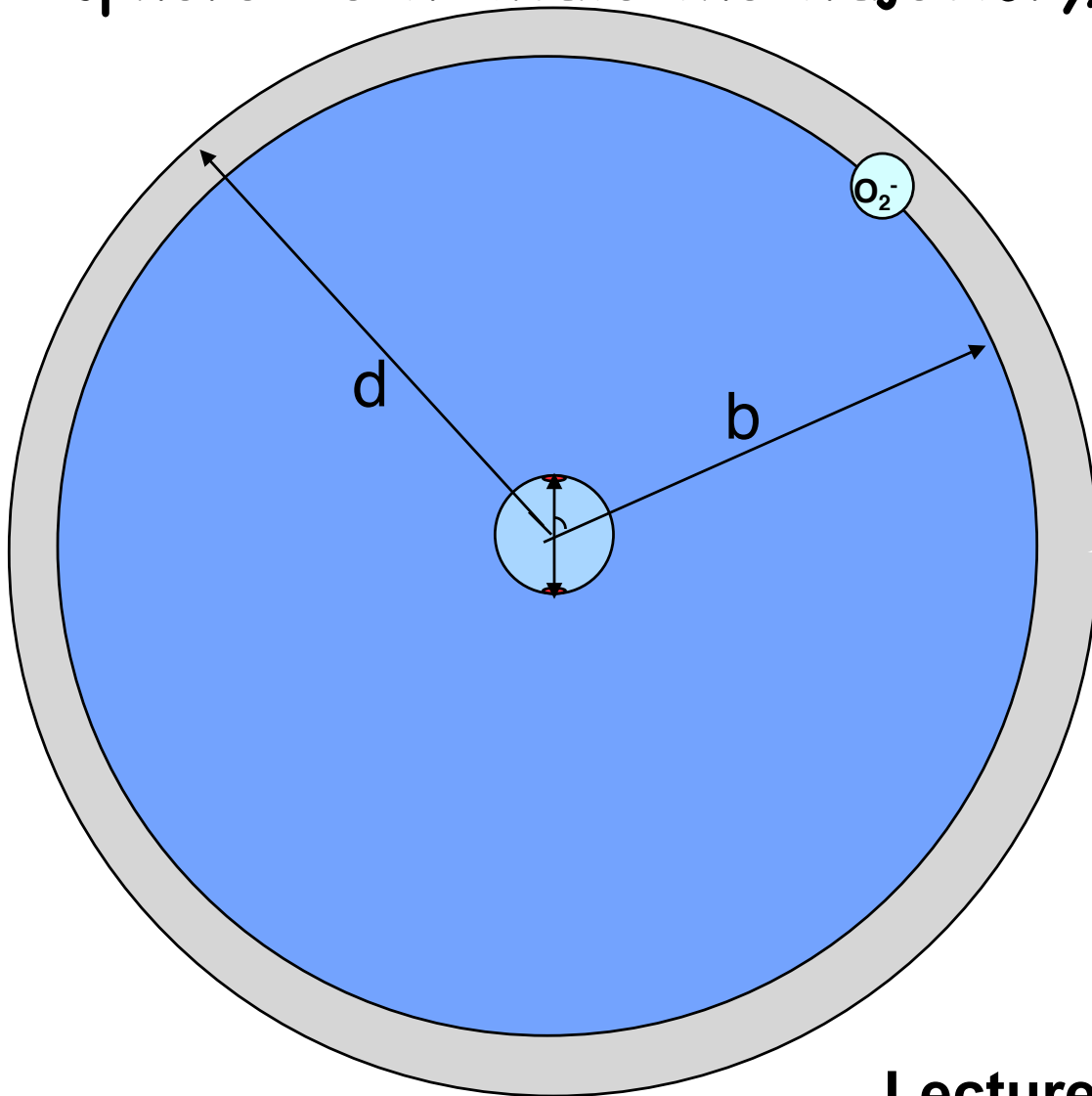
Simulating association rates of SOD/ O_2^-

We define another sphere at d , and if O_2^- reaches this outer sphere we truncate the trajectory, and count the trajectory as non-reactive



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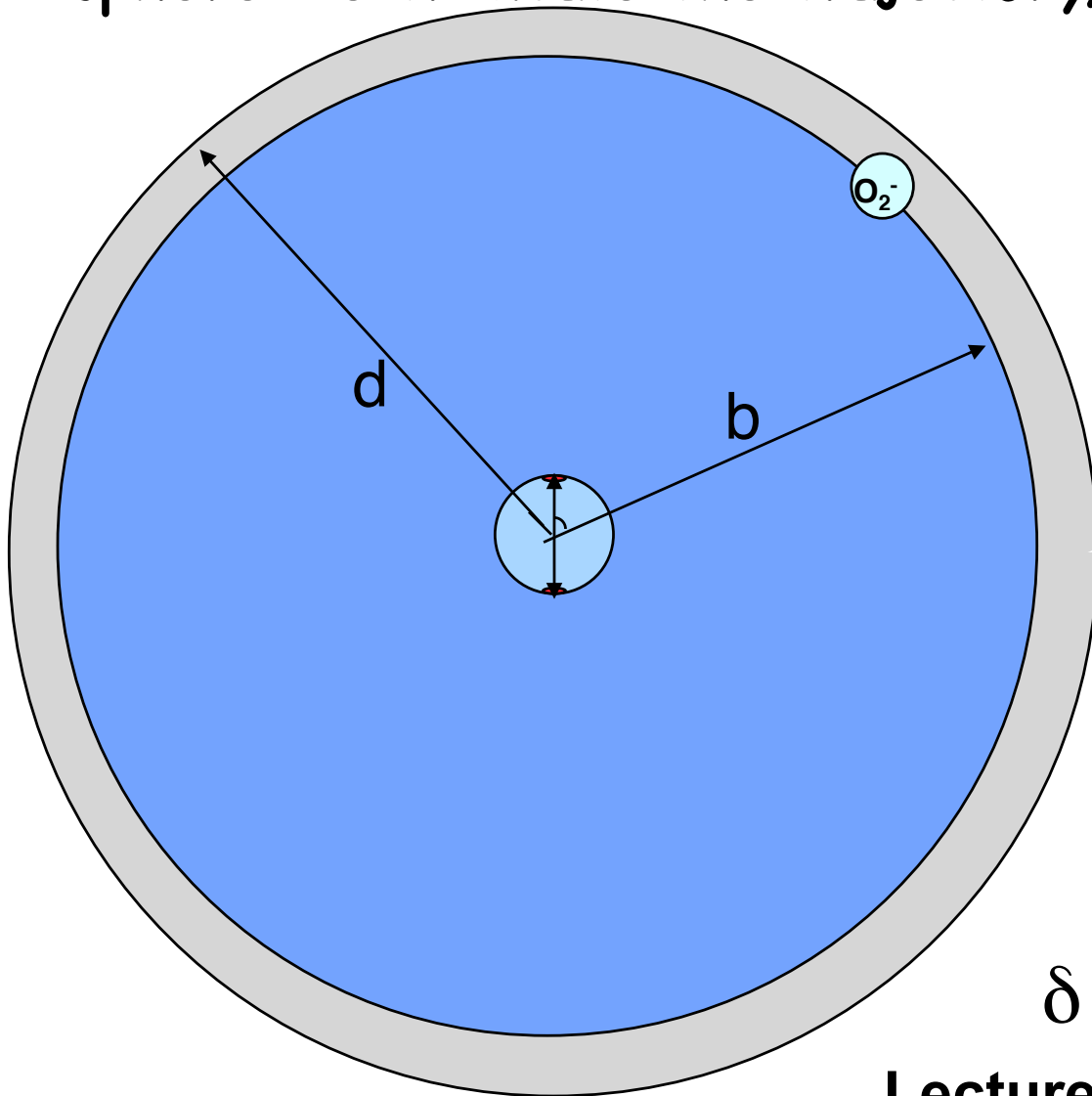
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Where $(1-\delta)$ is the probability that particle reaches separation $R=d$.

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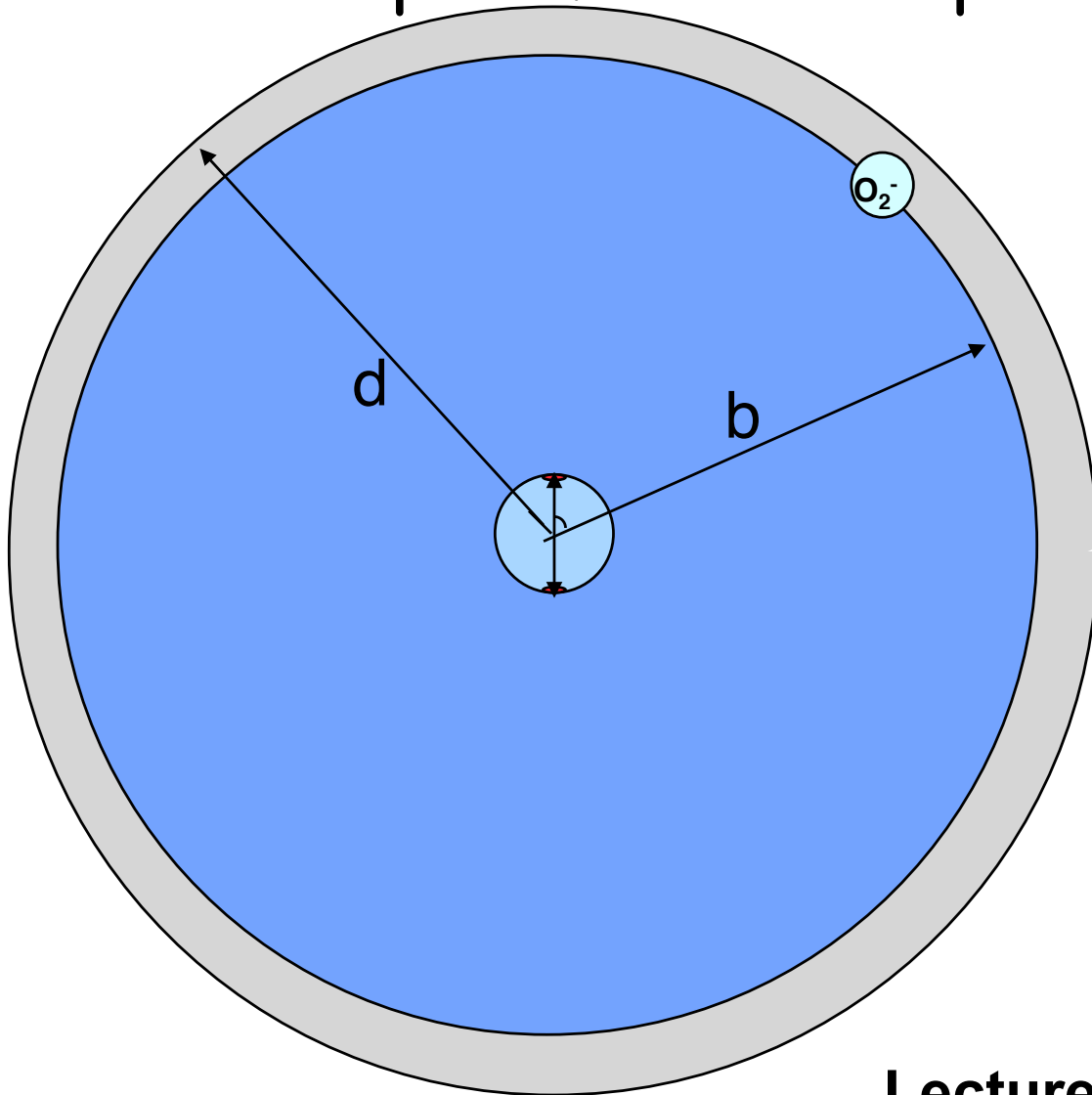
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δ is given by our simulation:

$$\delta = \frac{\# \text{ trajectories reaching active site}}{\text{total \# trajectories}}$$

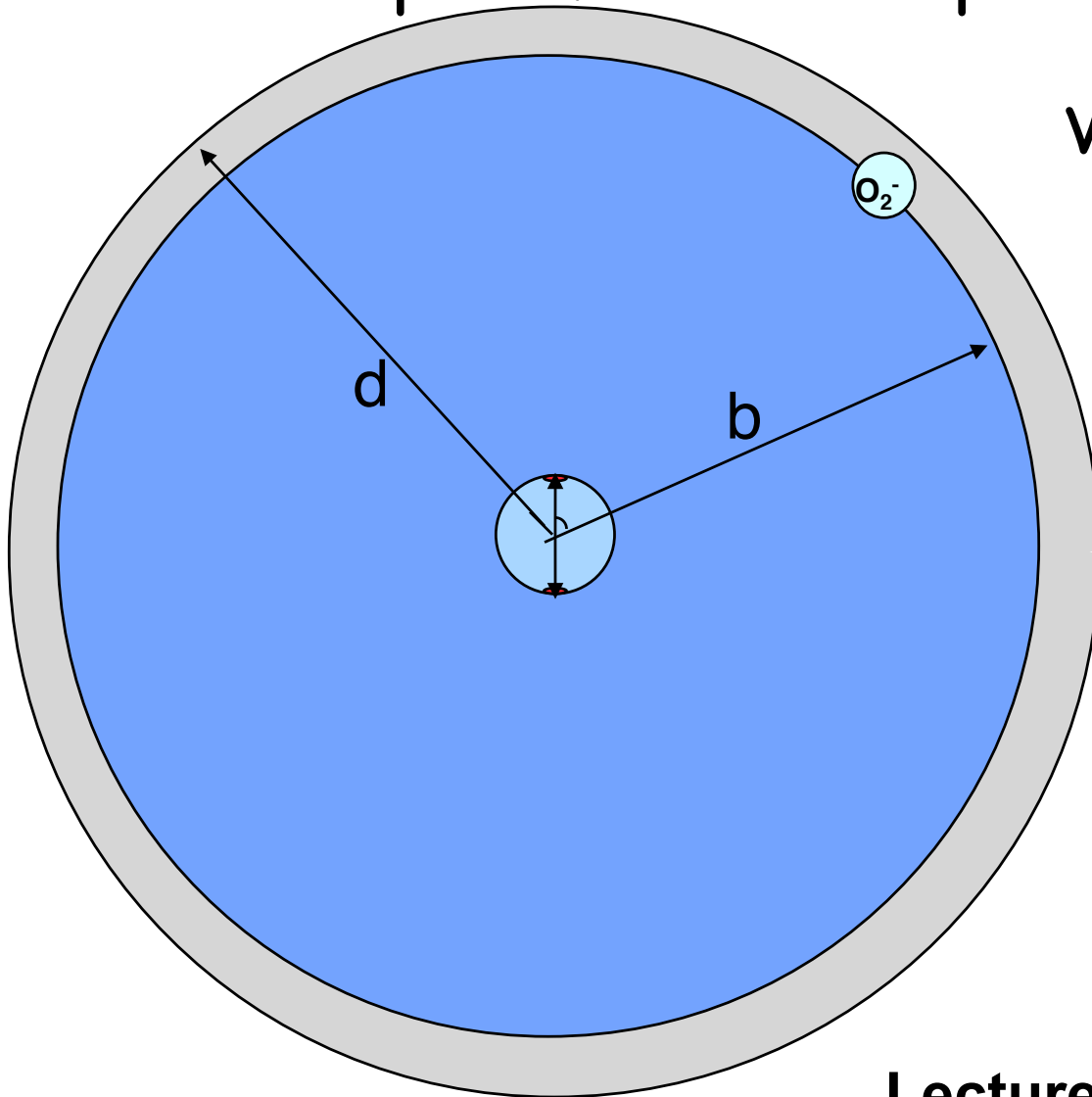
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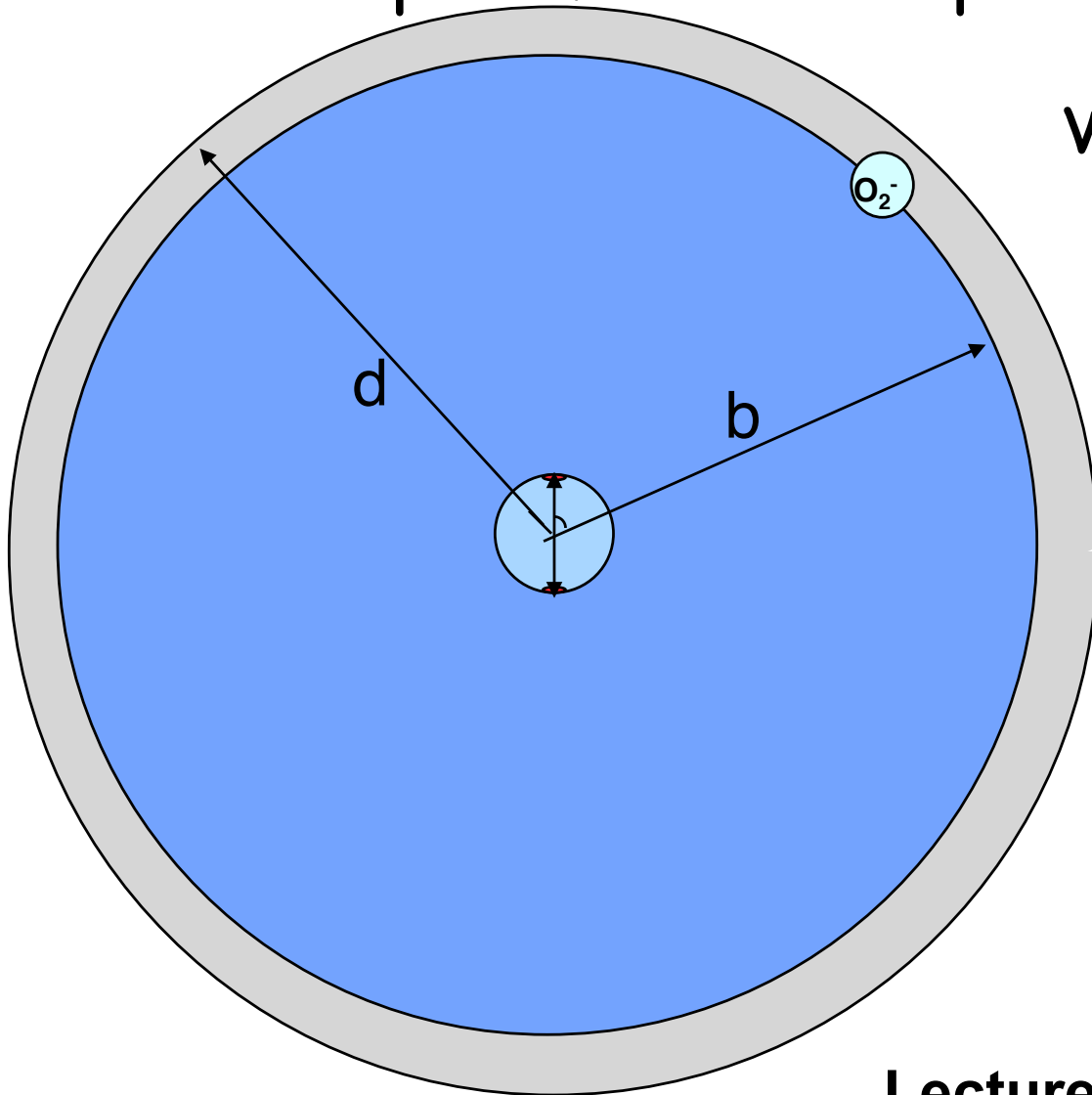


We can solve for γ analytically

$$\gamma = \frac{k(b)}{k(d)}$$

Simulating association rates of SOD/O₂⁻

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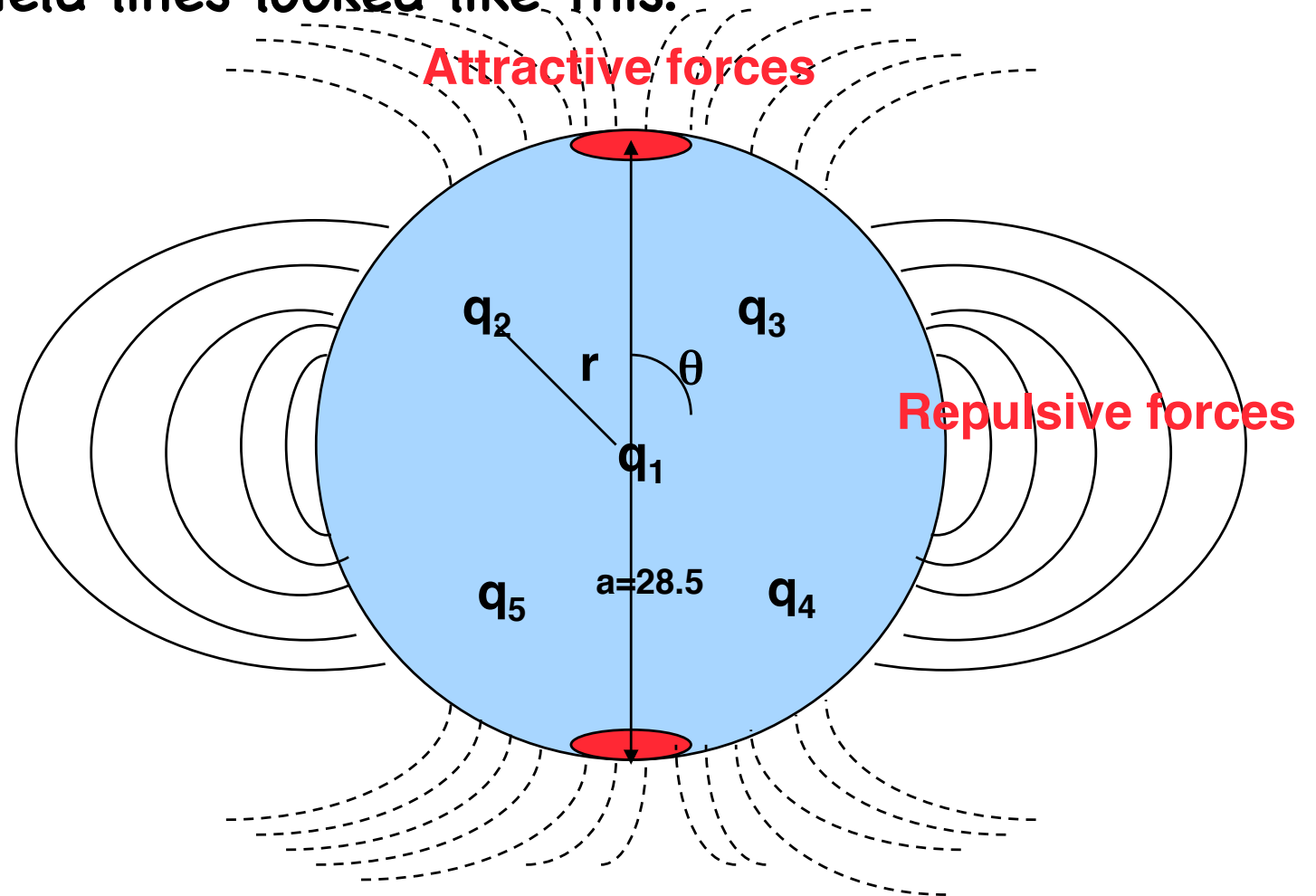
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However you can have multiple recrossings across the boundary d , so that final expression for B_∞ .

$$B_\infty = \frac{\delta}{[1 - (1 - \delta)\gamma]}$$

Summary of Results for SOD/O₂⁻

The electrostatic field lines looked like this:

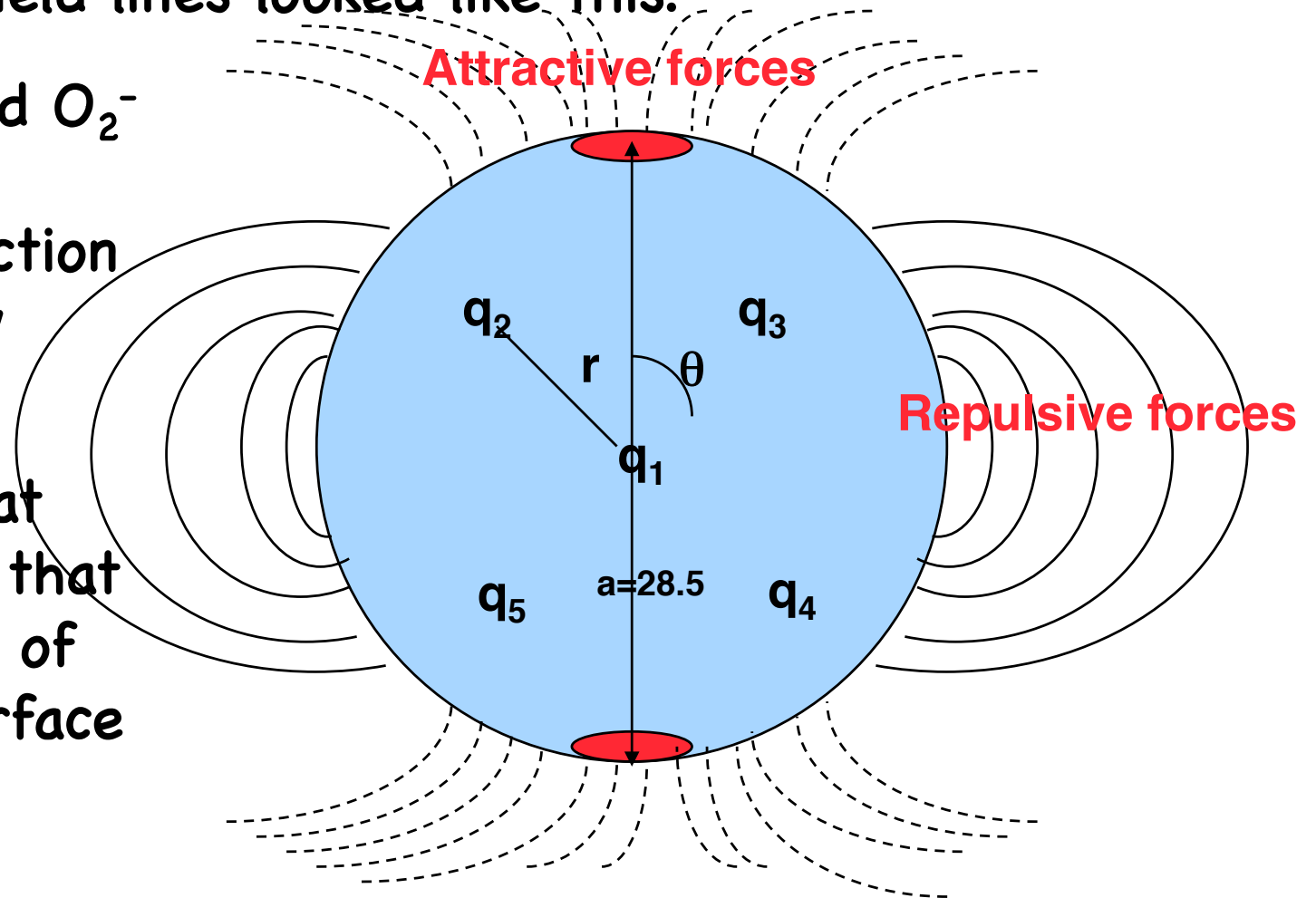


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(1) Protein ($-4e$) and O_2^- ($-1e$) suggest a repulsive interaction that would slow down the rate.

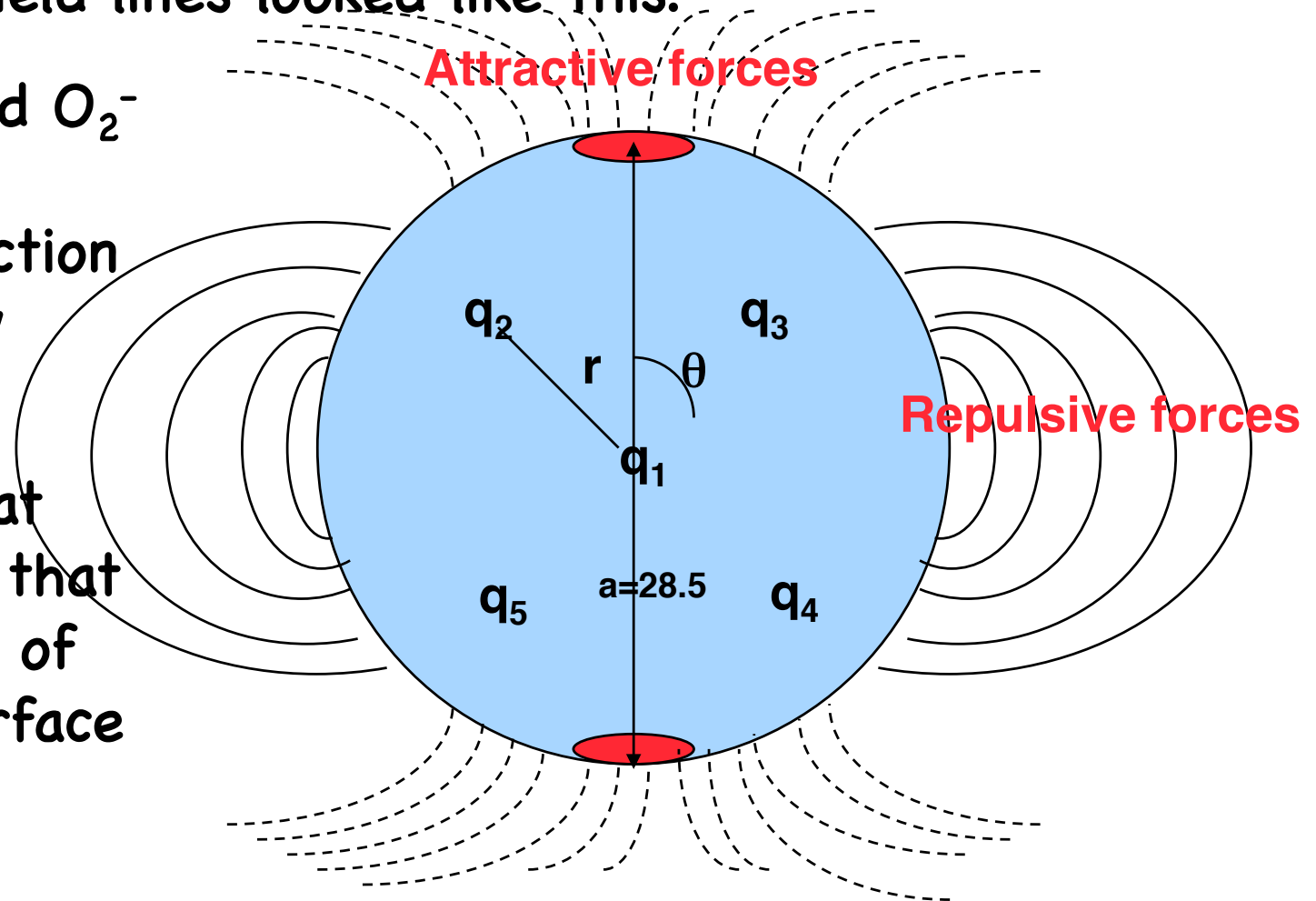
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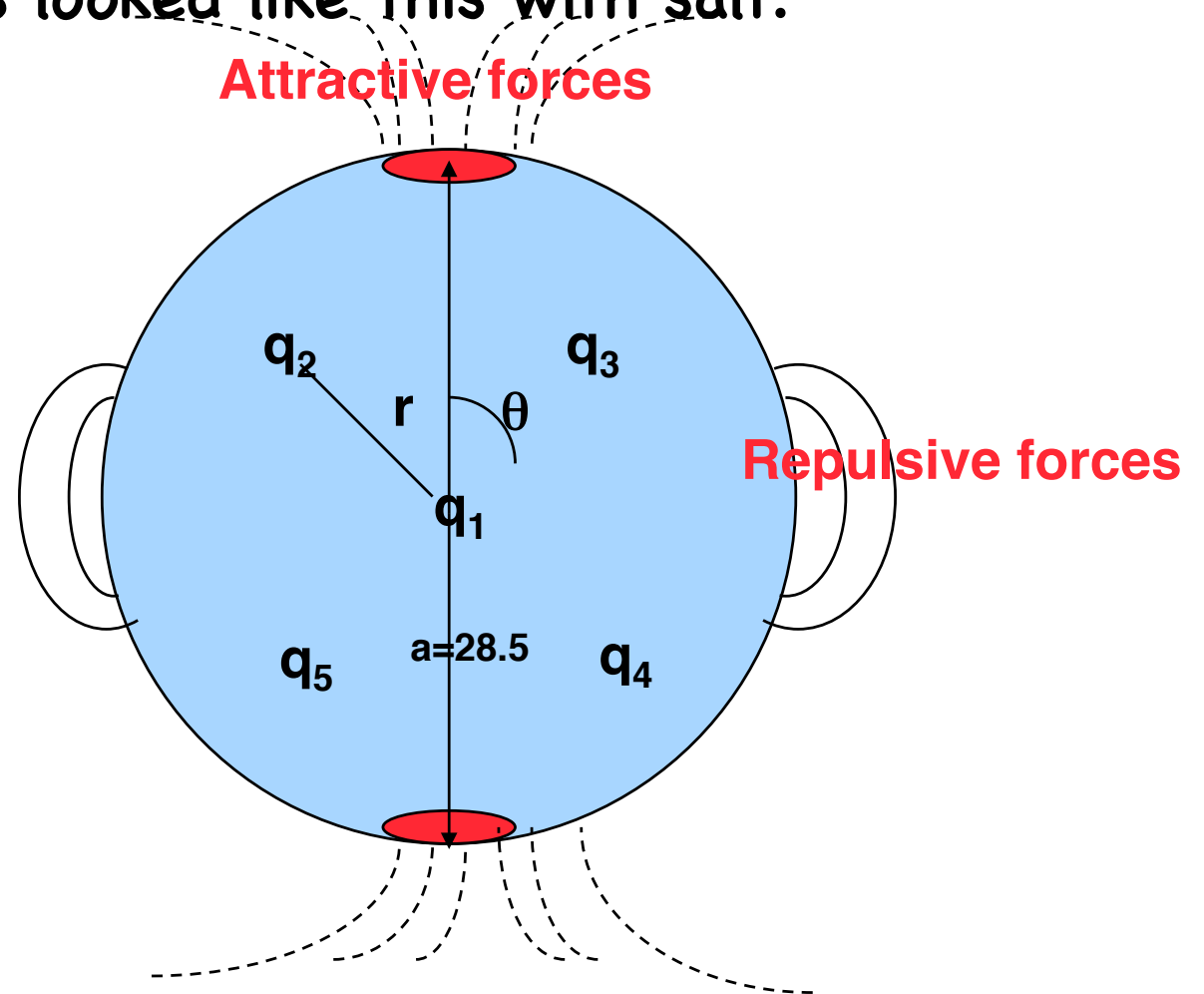
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- (2) Active site sits at bottom of cleft that occupies 1/150th of total protein surface area.



Effectively the potential steers superoxide toward the protein active sites.

Summary of Results for SOD/ O_2^-

The electrostatic field lines looked like this with salt:

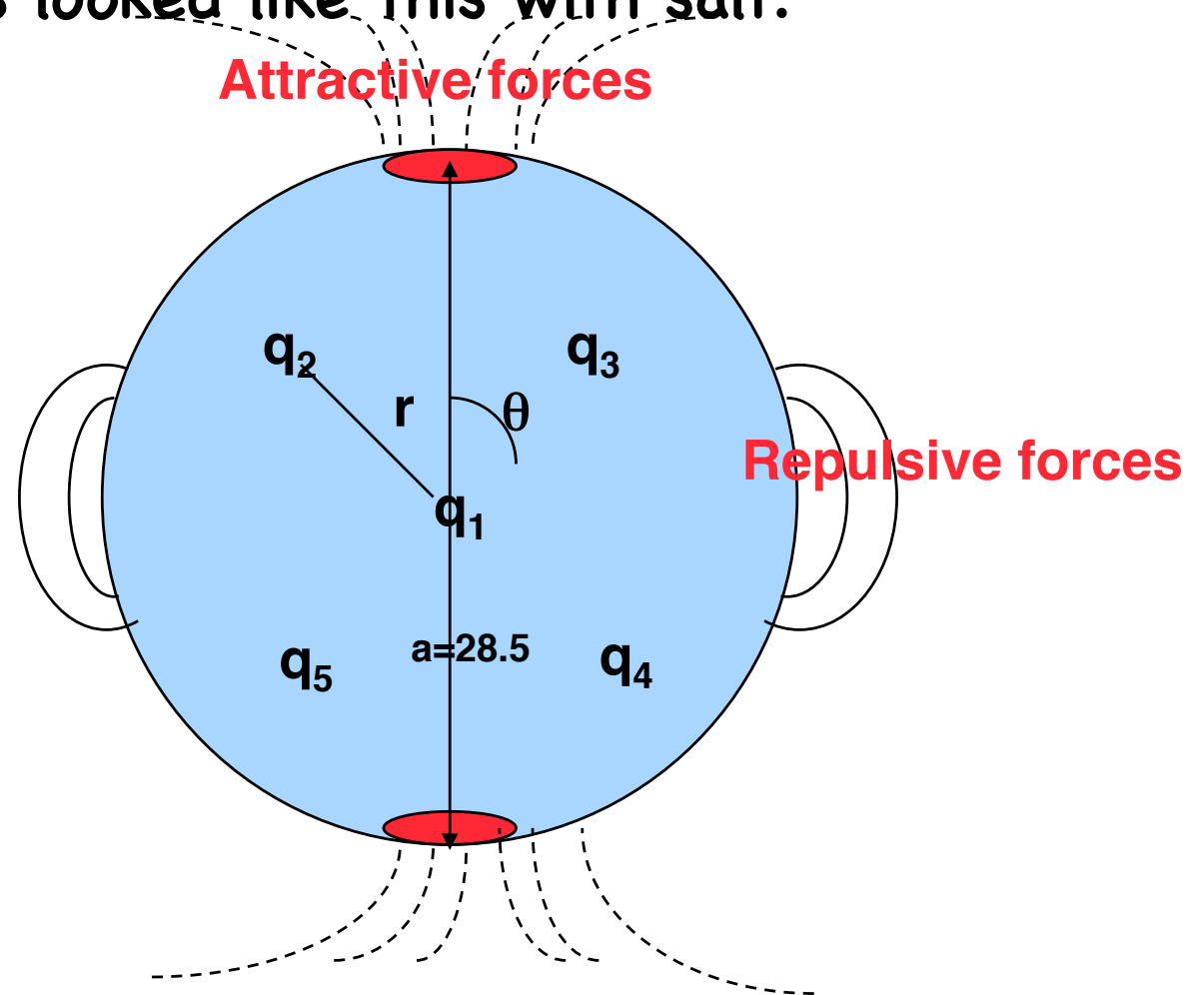


Summary of Results for SOD/O₂⁻

The electrostatic field lines looked like this with salt:

(3) The rate decreases with increasing salt concentration

The steering effect by SOD to guide superoxide toward protein active site is screened out. Salt decreases rate!



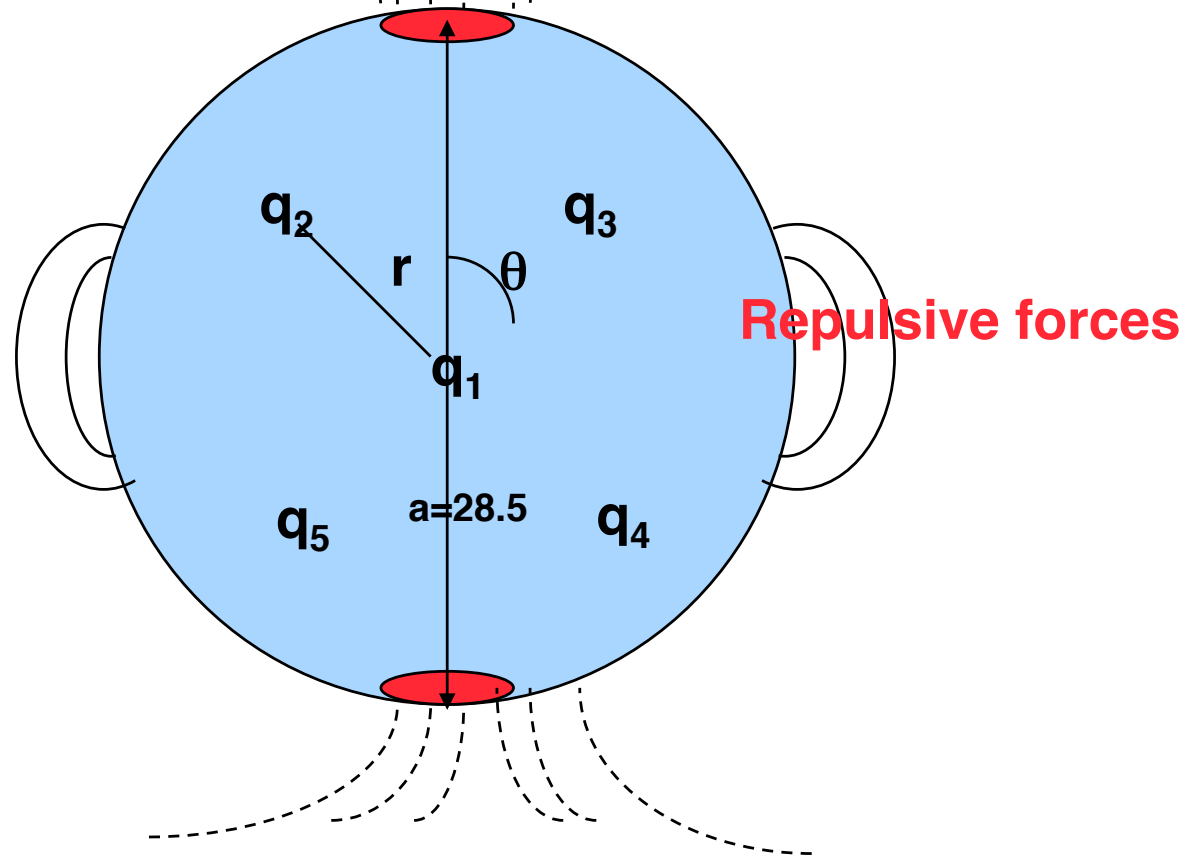
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Attractive forces

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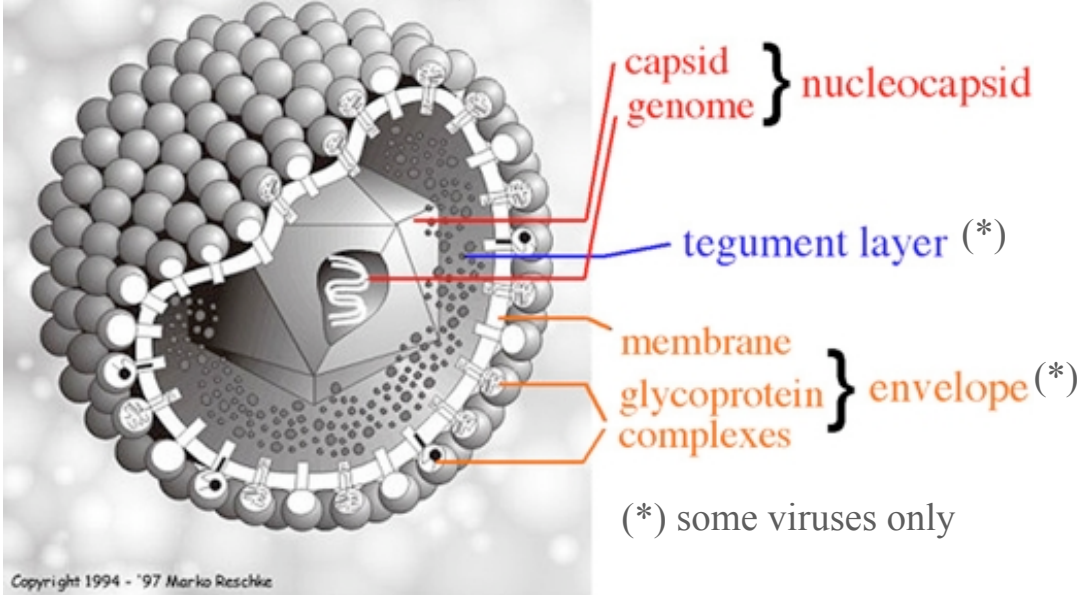
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Mutations that destroy this charge distribution have the potential to destroy this steering effect

Virus Structure and Assembly

PRV virus, Dr. Marko Reschke, 1994

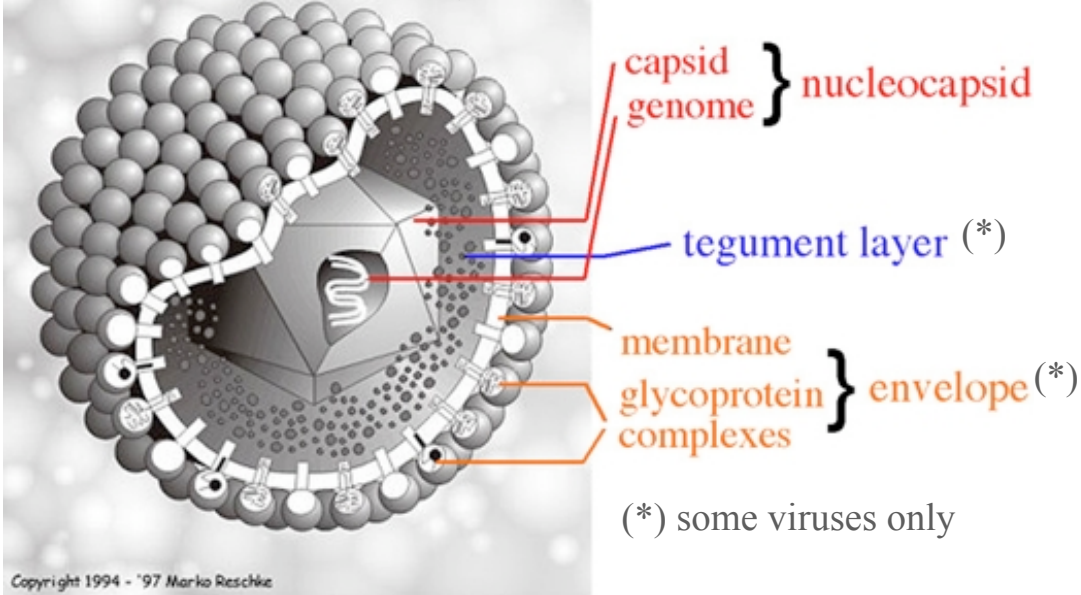


Anti-viral drugs and vaccines and predicting new strains of viruses that arise from mutation are of tremendous clinical and fundamental importance

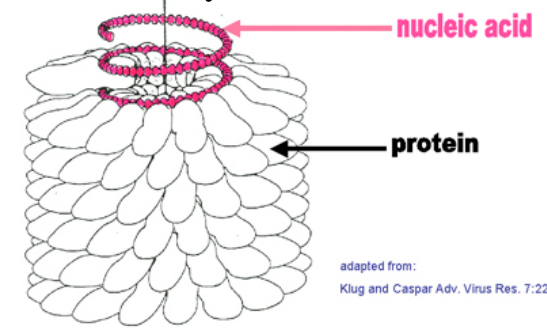
Simulating their assembly from their constituent proteins is of importance for designing better gene delivery systems

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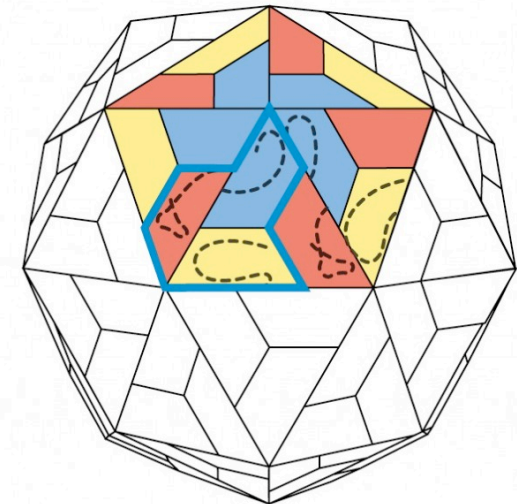


Helical capsid



Icosahedral capsid

- 20 facets



Each facet:

- 3-fold symmetry
- comprised of 3T monomers

T = 'Triangulation number'

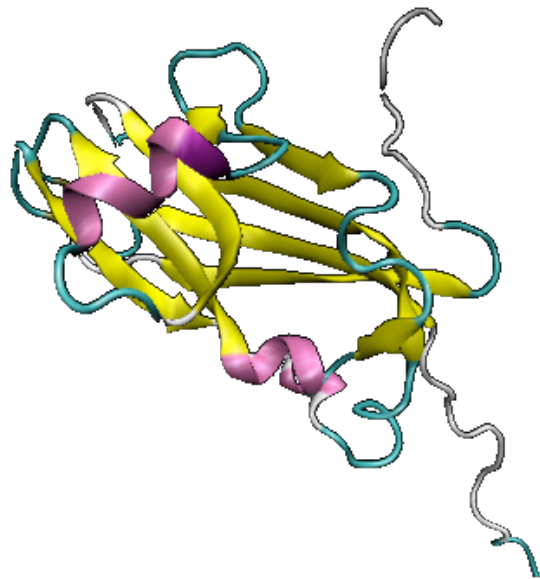
➤ $T=1$: $20 \times 3 \times 1 = 60$ monomers

➤ $T=3$: $20 \times 3 \times 3 = 180$ monomers

Brome Mosaic Virus

- plant virus: infects grass
- forms empty capsid *in vitro*
- potential drug delivery vehicle

Biological Assembly Image for 1YC6
Crystallographic Structure of the T=1 Particle of Brome Mosaic Virus



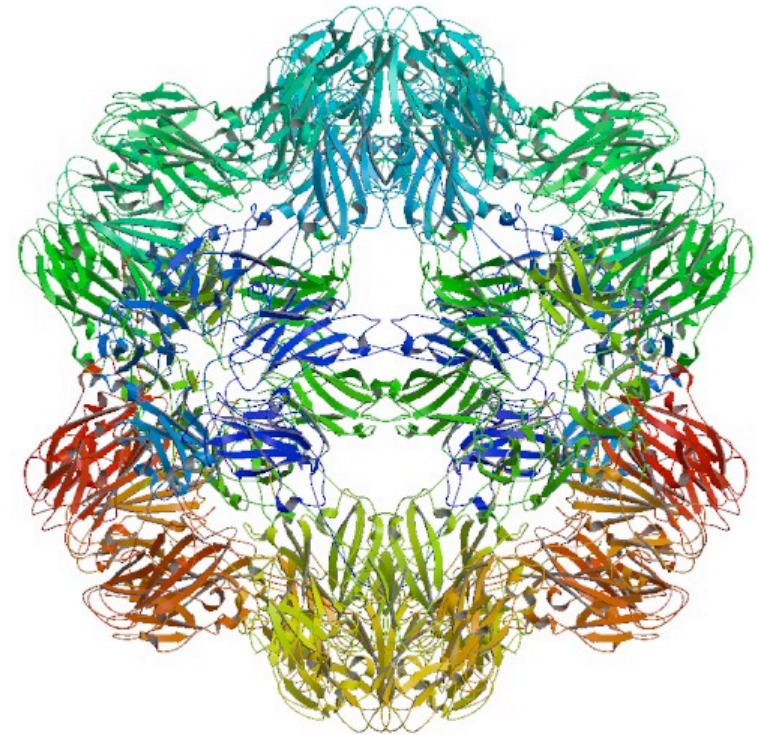
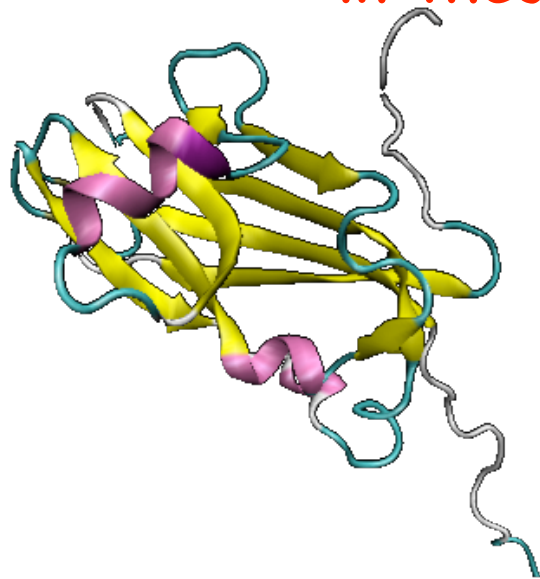
WT forms T=3 particle (189 residues/
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Mutant (1YC6) forms T=1 particle (154
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Brome Mosaic Virus

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Can we assemble the Brome Mosaic virus from its monomers in these two forms?



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Solving the Linearized PB Equation

- Analytical (Multipole) Methods
 - Debye-Huckel; Kirkwood (1934): 1 molecule

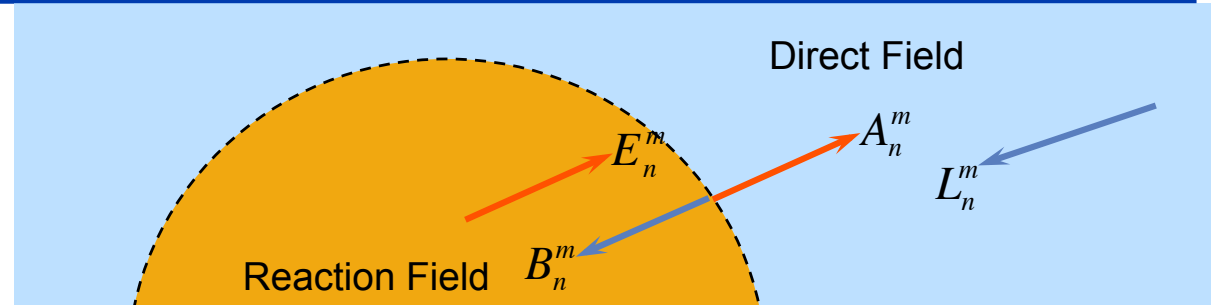
Solving the Linearized PB Equation

- **Analytical (Multipole) Methods**
 - Kirkwood (1934): 1 molecule
 - Lotan & THG (2006): N molecules! Generalized theory of Kirkwood (1934) to arbitrary numbers of molecules through advances in multipole methods for Yukawa potential (Greengard and co-workers)
 - Fast and accurate
 - Restricted to idealized geometries

Solving the Linearized PB Equation

- **Numerical Methods (Arbitrary dielectric geometries)**
 - **Finite Difference Methods (FD):** Delphi, UHBD, APBS
 - **Finite Element Methods (FE):** APBS
 - Realistic geometries; sparse matrix solvers available
 - More recent (APBS) are very fast
 - **Mesh does not follow dielectric boundary exactly**
 - **does not include von Neumann boundary condition**
 - **forces estimated from finite difference of potential**
 - **Boundary Element Methods (BEM)**
 - Surface (2D) instead of volume (3D): less unknowns
 - both Dirichlet and Von Neumann b.c. satisfied by construction
 - **Memory Intensive; computationally intensive, can handle only few macromolecules**

Analytical Solution to PBE for N Molecules

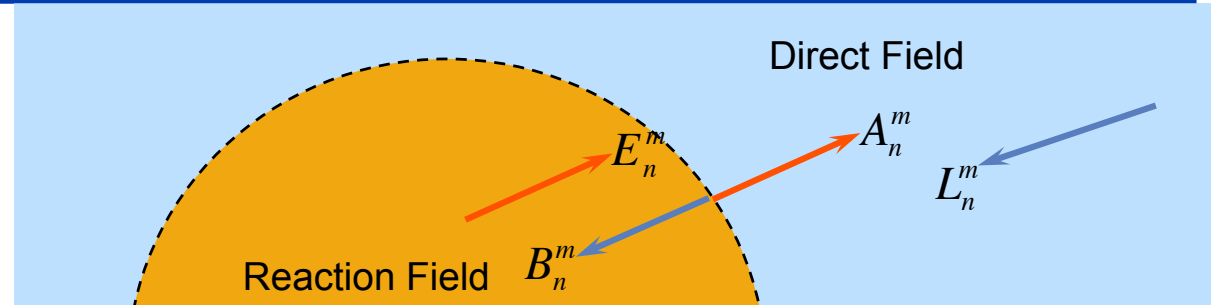


Multipole expansion of charge:
$$E_{n,m}^{(i)} = \sum_{j=1}^{M_i} q_j^{(i)} \left(\rho_j^{(i)} \right)^n Y_{n,m} \left(\vartheta_j^{(i)}, \varphi_j^{(i)} \right)$$

Potential inside cavity:
$$\Phi_{in}^{(i)}(\mathbf{t}) = \sum_{n=0}^{\infty} \sum_{m=-n}^n \left(\frac{E_{n,m}^{(i)}}{\epsilon_p r^{n+1}} + B_{n,m}^{(i)} r^n \right) Y_{n,m}(\theta, \phi)$$

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Linear re-expansion operator:
$$\mathbf{L}^{(i)} = \sum_{\substack{j=1 \\ j \neq i}}^N \mathbf{T}^{(i,j)} \cdot \mathbf{A}^{(j)}$$

Multipole methods for Yukawa potential (Greengard et al)

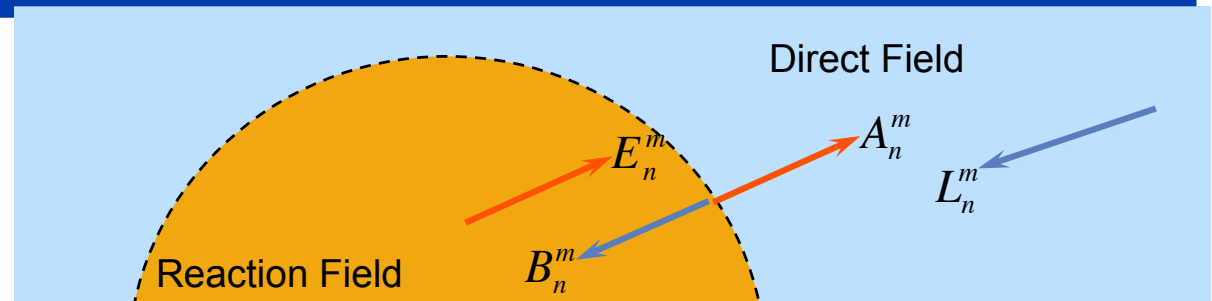
Transforms a multipole expansion at $\mathbf{c}^{(j)}$ to a local (Taylor) expansion at $\mathbf{c}^{(i)}$.

Analytical Solution to PBE for N Molecules

Solving for $\mathbf{A}^{(i)}$ and $\mathbf{B}^{(i)}$:

$$A_{n,m}^{(i)} = \gamma_n^{(i)} \delta_n^{(i)} L_{n,m}^{(i)} + \gamma_n^{(i)} E_{n,m}^{(i)}$$

$$B_{n,m}^{(i)} = \frac{1}{\epsilon_s} \left(\frac{A_{n,m}^{(i)}}{a_i^{2n+1}} e^{-\kappa a_i} \hat{k}_n(\kappa a_i) + L_{n,m}^{(i)} \hat{i}_n(\kappa a_i) \right) - \frac{E_{n,m}^{(i)}}{\epsilon_p a_i^{2n+1}}$$

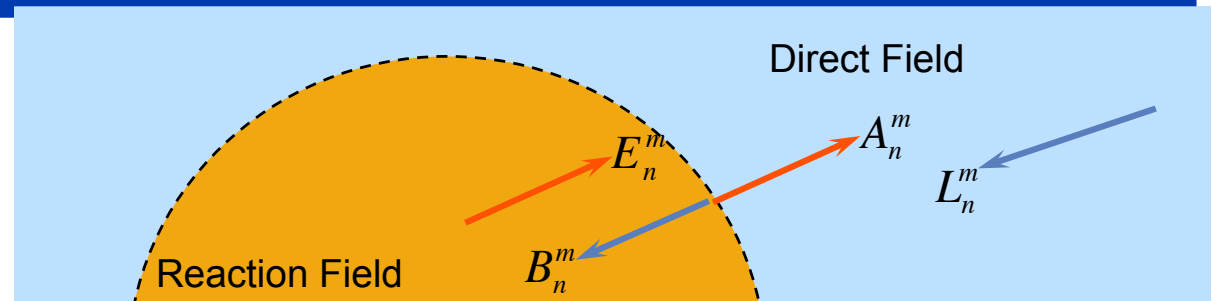


If I solve $\mathbf{A}^{(i)}$, then $\mathbf{B}^{(i)}$ is straightforward; focus on solving $\mathbf{A}^{(i)}$

Analytical Solution to PBE for N Molecules

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If I solve $\mathbf{A}^{(i)}$, then $\mathbf{B}^{(i)}$ is straightforward; focus on solving $\mathbf{A}^{(i)}$

Define matrix $\Gamma^{(i)}$

$$\gamma_n^{(i)} = \frac{(2n+1)e^{\kappa a_i}}{(2n+1)\hat{k}_{n+1}(\kappa a_i) + n\hat{k}_n(\kappa a_i)(\epsilon_p/\epsilon_s - 1)}$$

and matrix $\Delta^{(i)}$

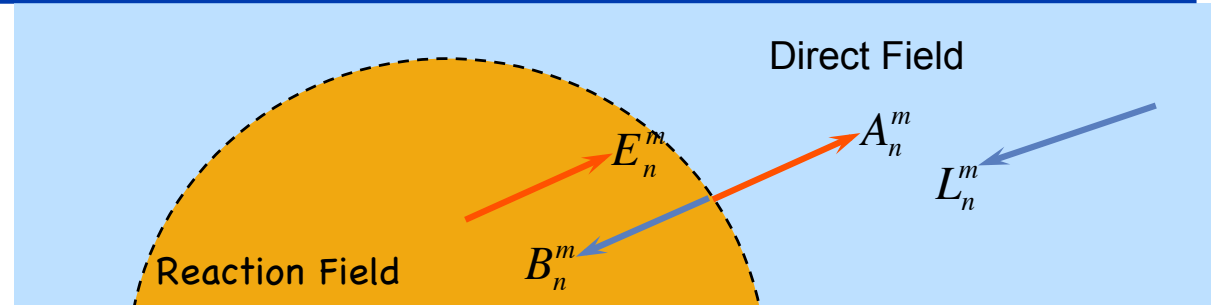
$$\delta_n^{(i)} = \frac{a_i^{2n+1}}{2n+1} \left[\kappa^2 a_i^2 \frac{\hat{i}_{n+1}(\kappa a_i)}{2n+3} + n\hat{i}_n(\kappa a_i)(1 - \epsilon_p/\epsilon_s) \right]$$

Note that these matrices are merely constants of the system we are solving (and we solve once in a CG scheme)

Analytical Solution to PBE for N Molecules

Solution to $A^{(i)}$ can be written as

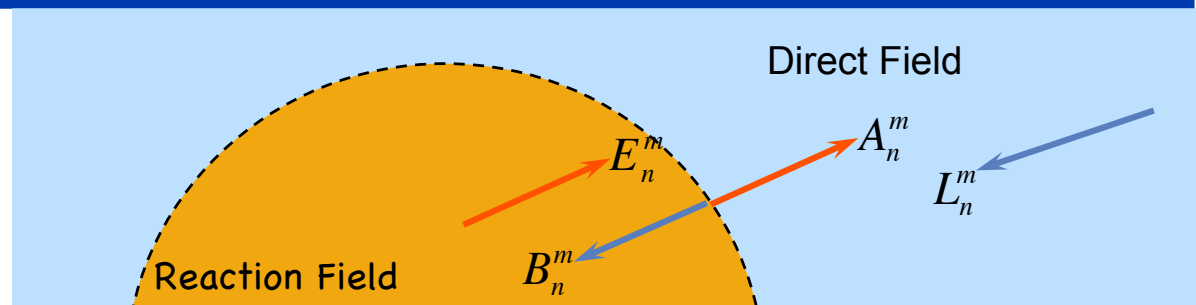
$$A = \Gamma(\Delta \cdot T \cdot A + E)$$



Analytical Solution to PBE for N Molecules

Solution to $A^{(i)}$ can be written as

$$A = \Gamma(\Delta \cdot T \cdot A + E)$$



$\Delta^{(i)}$ Cavity-polarization operator from external charges

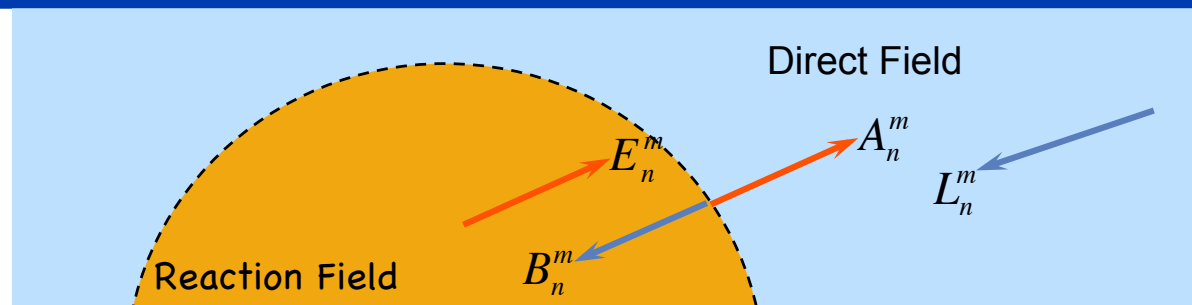
$\Gamma^{(i)}$ Dielectric-boundary-crossing operator for a molecule

$A^{(i)}$ *Effective* multipole expansion: transformed charges due to dielectric boundary and polarization of other empty dielectric cavities

Analytical Solution to PBE for N Molecules

Solution to $A^{(i)}$ can be written as

$$A = \Gamma(\Delta \cdot T \cdot A + E)$$



$\Delta^{(i)}$ Cavity-polarization operator from external charges

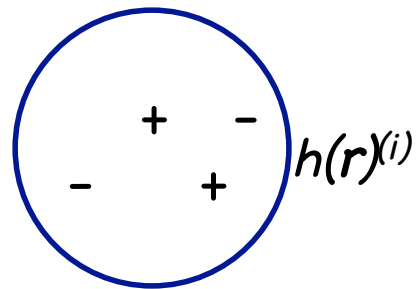
$\Gamma^{(i)}$ Dielectric-boundary-crossing operator for a molecule

$A^{(i)}$ *Effective* multipole expansion: transformed charges due to dielectric boundary and polarization of other empty dielectric cavities

The potential field induced by a molecule is due to the sum of contribution of its free charges ($E^{(i)}$) and the contribution of polarization charges induced by other molecules, transformed by the effect of its dielectric boundary.

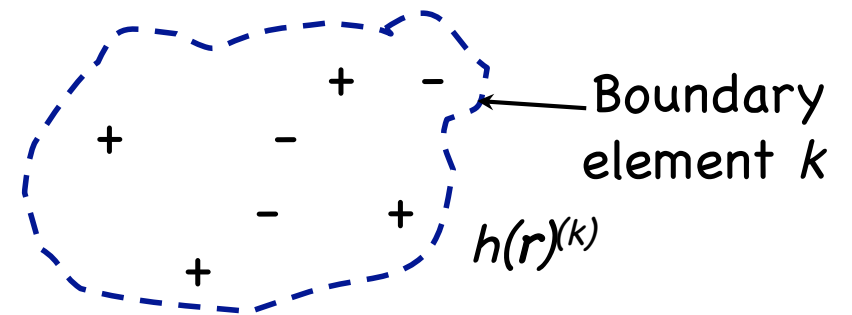
Semi-analytical PBE: N Realistic Shapes

Multipole Method



Solve for entire molecule i
analytically

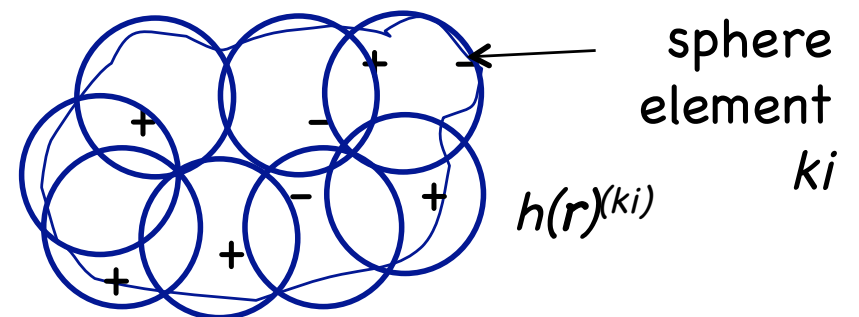
Boundary Element Method



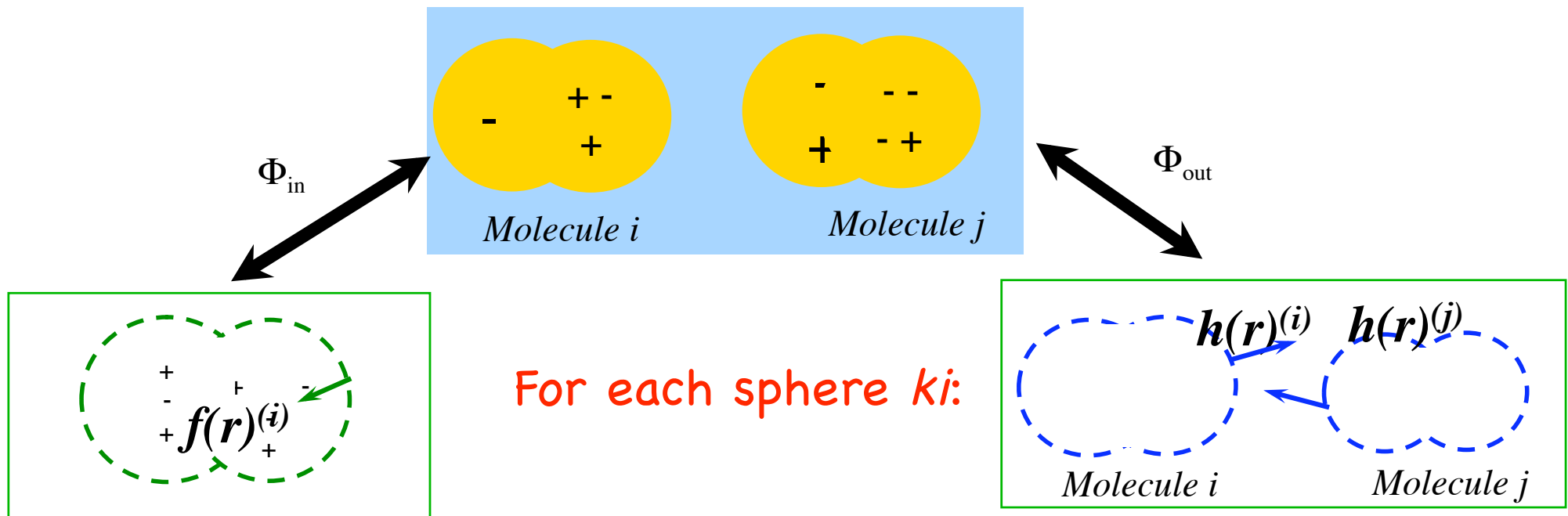
Solve for N_k system of equations
numerically $N_k \sim 10^6$

Multipole Method for arbitrary shapes

Discretize molecule into N_{ki} spheres
Solve for each sphere ki analytically,
Iterate over N_{ki} spheres, $N_{ki} \sim 100$



Semi-analytical PBE: N Realistic Shapes



Express Φ_{in} and Φ_{out} in terms of multipoles $F^{(ki)}$ and $H^{(ki)}$

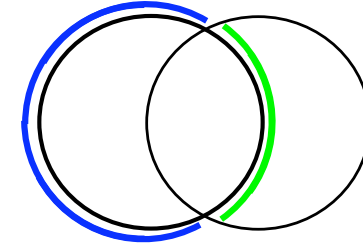
Impose boundary equations and solve for $F^{(ki)}$ and $H^{(ki)}$

Update contribution from $F^{(ki)}$ and $H^{(ki)}$ to other spheres

Repeat for all spheres until convergence criteria is reached

Semi-analytical PBE: N Realistic Shapes

On the EXPOSED sphere surface (a, θ_E, ϕ_E) :



$$\sum_{n=0}^{\infty} \sum_{m=-n}^n \left(-H_{nm}^{(ki)} e^{-\kappa a_{ki}} \hat{k}_n(\kappa a_{ki}) + F_{nm}^{(ki)} + X_{H_{nm}}^{(ki)} \right) Y_{nm}(\theta_E, \phi_E) = 0$$

$$\sum_{n=0}^{\infty} \sum_{m=-n}^n \left(e^{-\kappa a_{ki}} \left[n \hat{k}_n(\kappa a_{ki}) - (2n+1) \hat{k}_{n+1}(\kappa a_{ki}) \right] H_{nm}^{(ki)} - n \varepsilon F_{nm}^{(ki)} + X_{F_{nm}}^{(ki)} \right) Y_{nm}(\theta_E, \phi_E) = 0$$

On the BURIED sphere surface (a, θ_B, ϕ_B) :

$$\sum_{n=0}^{\infty} \sum_{m=-n}^n \frac{(2n+1)}{4\pi} F_{nm} Y_{nm}(\theta_B, \phi_B) = 0$$

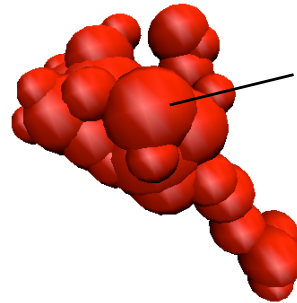
$$\sum_{n=0}^{\infty} \sum_{m=-n}^n \frac{2n+1}{4\pi} \frac{H_{nm}}{\hat{i}_n(\kappa a_{ki})} Y_{nm}(\theta_B, \phi_B) = 0$$

Now we can solve for $F^{(ki)}$ and $H^{(ki)}$

Semi-analytical PBE: N Realistic Shapes

Each *protein* modeled as a cluster of overlapping Spheres

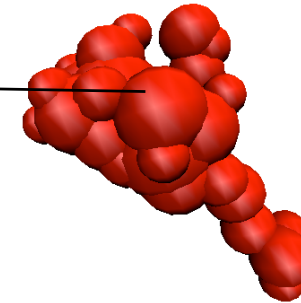
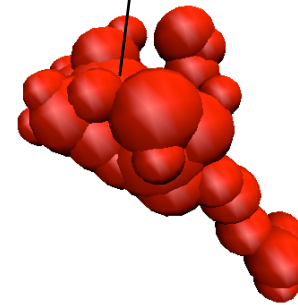
Each *sphere* contains fixed partial charges



Protein Interior:

- $\epsilon_p = 4, \kappa = 0$

- Partial Charges



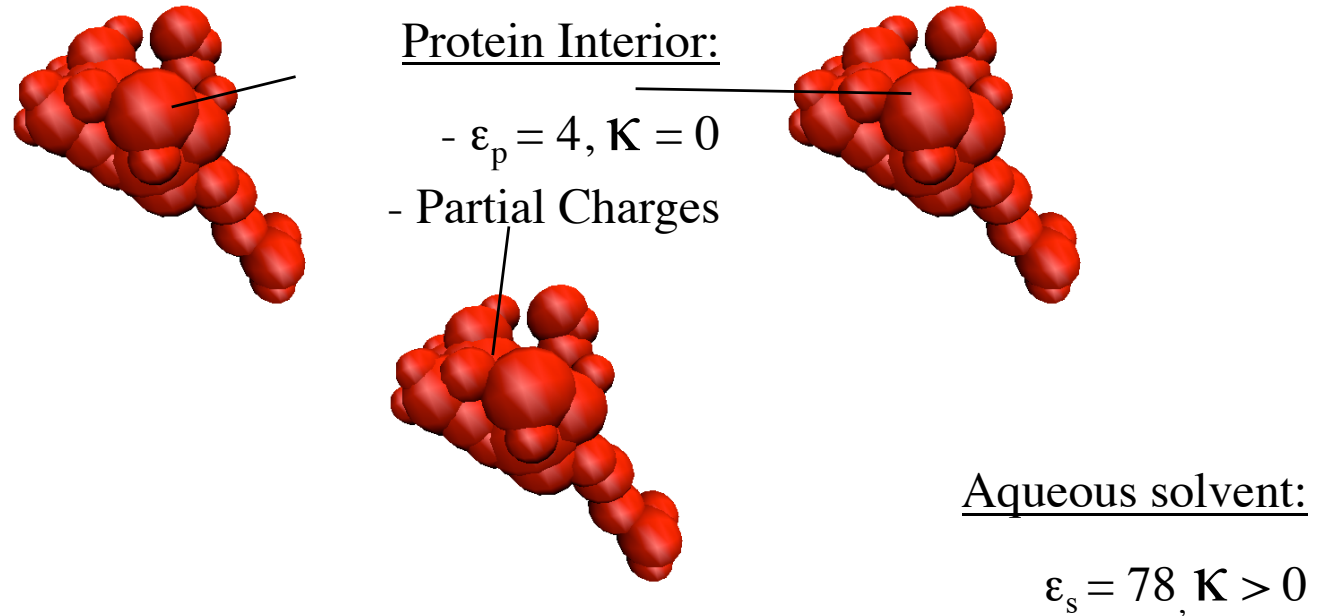
Aqueous solvent:

$\epsilon_s = 78, \kappa > 0$

We are currently investigating the assembly of the capsid



Semi-analytical PBE: N Realistic Shapes



We are currently investigating the assembly of the capsid

Multiscale models

Proposed transitions between models

