

SIMULATION TECHNIQUES FOR SOFT MATERIALS

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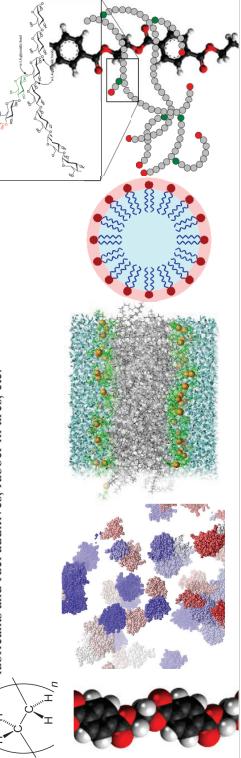


- 1) What is a soft material?
- 2) Stability.
 - Double Layer \rightarrow NVT MC
 - Effect of pH \rightarrow Semi-Grand Canonical (SGCMC)
- 3) Flexible materials.
- 4) Water/Ion permeability
 - Effect of Charge/Ionic strength \rightarrow Molecular Dynamics (MD)
- 5) Diffusion and Reactivity
 - Effect of crowding/aggregation \rightarrow Brownian Dynamics (BD)

Simulation Techniques for Soft Materials

Soft Material

- These materials share an important common feature in that predominant physical behaviors occur at an energy scale comparable with room temperature thermal energy.
- Distinctive physics: soft matter self-organizes into mesoscopic physical structures that are much larger than the microscopic scale
- Examples: polymers (natural, synthetic), biological membranes, biomaterials, colloids, gels, microemulsions, proteins, surfactants,...
- Applications: wide range of technological applications. They may appear as structural and packaging materials, foams and adhesives, detergents and cosmetics, paints, food additives, lubricants and fuel additives, rubber in tires, etc.



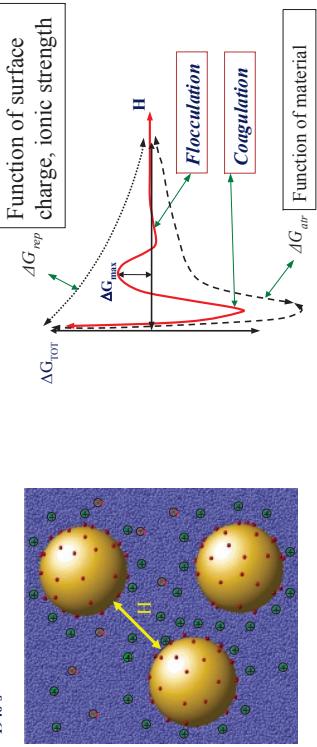
Colloids in Solution. What Theory says?

COLLOIDAL STABILIZATION: ELECTROSTATIC

$$\Delta G_{\text{tot}} = \Delta G_{\text{attract}} + \Delta G_{\text{rep}}$$

VdW Attraction + Electrostatic Repulsion

DLVO Model
(Derjaguin, Landau, Verwey, Overbeek)
1940's



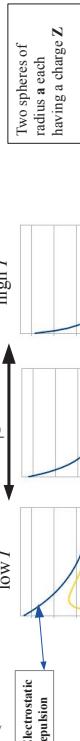
DLVO Model: Effect of the ionic strength

*Van der Waals attraction and the electrostatic repulsion due to the so called electric double layer (EDL) of counterions. The electrostatic part of the DLVO interaction is computed in the mean field approximation in the limit of low surface potentials

$$\text{Electrostatic repulsion: } \beta U(r) = Z^2 \lambda_B \left(\frac{\exp(kr)}{1 + \kappa a} \right)^2 \frac{\exp(-kr)}{r}$$

$$\lambda_B = \frac{e}{4\pi \epsilon_0 \epsilon_r k_B T}$$

$$\kappa^{-1} = \sqrt{\frac{\epsilon_r \epsilon_0 k_B T}{2 N_A e^2 I}}$$



- Uniformly charged nanoparticles
- Mean field approximation for electrostatic interactions: no prediction of the overcharging phenomena
- Approximate prediction of the critical coagulation concentration (ccc)

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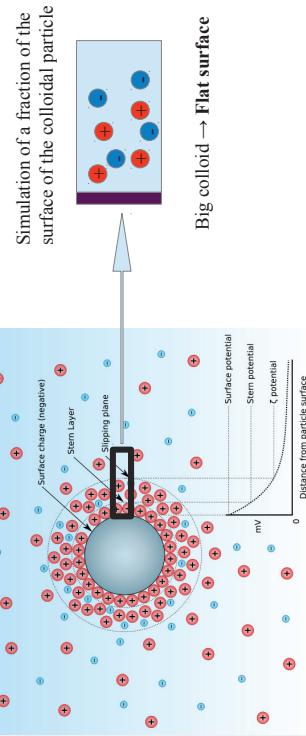
DLVO

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The electrical double layer (EDL) around charged colloids

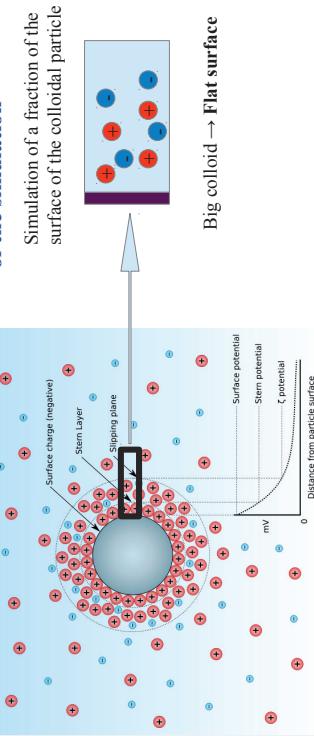
The electrical double layer (EDL) is a key concept in physical chemistry referring to the ionic distribution formed by an electrolyte around a charged colloid. Equilibrium and transport properties of colloidal systems are determined by the specific structure of the EDL. Related with experimental determined Zeta Potential.

Selection of the Geometry of the simulation



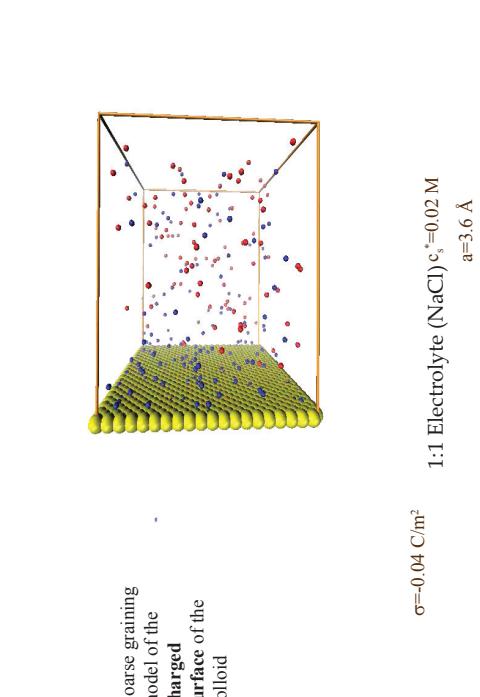
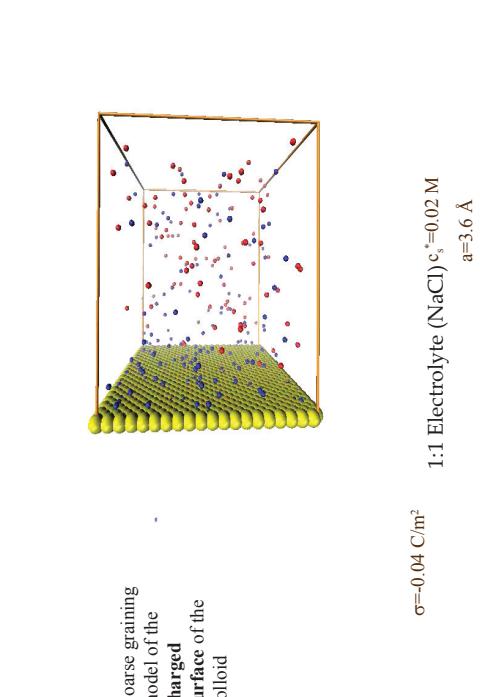
Simulation Model

MC simulations: Canonical ensemble



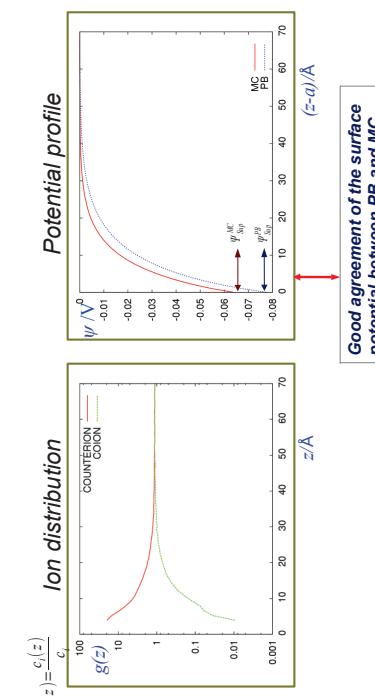
Simulation Model

MC simulations: Canonical ensemble



1:1 Electrolyte (NaCl)

$c_s^* = 0.1 \text{ M}$
 $\sigma = 0.08 \text{ C/m}^2$
 $a = 3.6 \text{ \AA}$



2:1 Electrolyte (CaCl₂)

$c_s^* = 0.05 \text{ M}$
 $\sigma = 0.24 \text{ C/m}^2$
 $a = 4.2 \text{ \AA}$

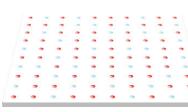
3:1 Electrolyte (LaCl₃)

$c_s^* = 0.1 \text{ M}$
 $\sigma = 0.16 \text{ C/m}^2$
 $a = 4.8 \text{ \AA}$

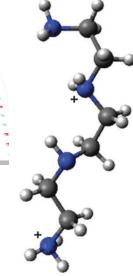
Semi-Grand Canonical Monte Carlo

- Appropriate to study the effect of pH
- Appropriate to simulate systems with fixed number of sites.

For example:



a) Surfaces of polyacids



b) Linear polyacids:

Semi-grand Canonical Ensemble

Ensemble to study mixtures (phase behaviour, interconversion reactions)



$$n_{Total} = n_A + n_B$$

- The chemical potential of one component is fixed (H^+).

Thermodynamic equilibrium:

$$\mu_A = \mu_B + \mu_{H^+}$$

- The chemical potential of other components (A,B) can be imposed by trial MC identity changes.

The probability of accepting a trial move is:

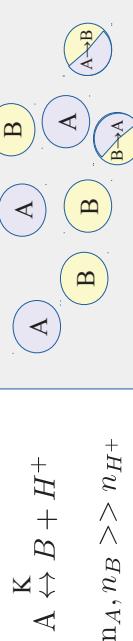
$$acc(\xi_i \rightarrow \xi'_i) = min\{1, \frac{\xi'_i}{\xi_i} \exp[-\beta \Delta U(s^N)]\}$$

$\Delta U(s^N)$ Is the change in potential energy of the system if we change the identity of a randomly selected particle from i to i' .

Understanding Molecular Simulations. From algorithms to Applications. D. Frenkel & B. Smit

Semi-grand Canonical Ensemble

Ensemble to study mixtures (phase behaviour, interconversion reactions)



- Fixed number of particles of the system
- The chemical potential of one component is fixed (H^+).

Each particle of the system

$$\mu_i(P, T, x_i) = \mu_i^0(T) + k_B T \ln(f_i)$$

$$\xi_i = \frac{f_i}{\sum_{j=1}^n f_j}$$

fugacity fraction
fugacity

One of the N particles is selected at random and with equal probability assign it one of the n possible identities.

The probability of accepting a trial move is:

$$acc(\xi_i \rightarrow \xi'_i) = min\{1, \frac{\xi'_i}{\xi_i} \exp[-\beta \Delta U(s^N)]\}$$

$\Delta U(s^N)$ Is the change in potential energy of the system if we change the identity of a randomly selected particle from i to i' .

Understanding Molecular Simulations. From algorithms to Applications. D. Frenkel & B. Smit

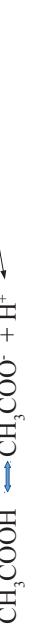
Accepted/Rejected trial movements with SGCMC:

$$\text{Energy contributions for SGCMC} = \begin{cases} \Delta F = \Delta U_{el} + k_B T (\ln 10) (\text{pH} - \log K_0) \\ \quad \text{for protonation process} \\ \Delta F = \Delta U_{el} - k_B T (\ln 10) (\text{pH} - \log K_0) \\ \quad \text{for deprotonation process} \end{cases}$$

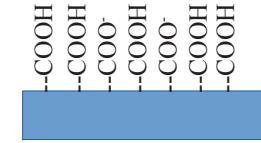
Apply ΔF for Metropolis MC algorithm.
Accepted movement with probability:

$$\min \{1, \exp(-\Delta F)\}$$

Acid/base in solution:



Acid/base in surface:



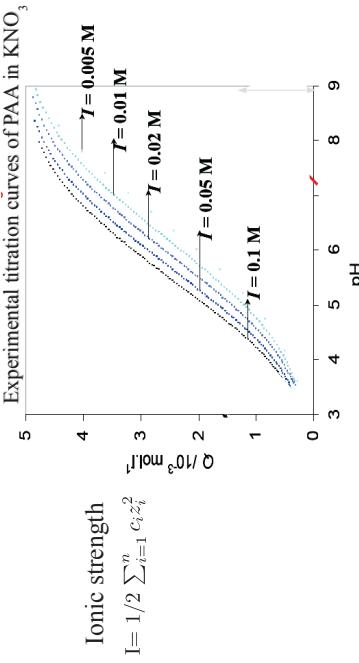
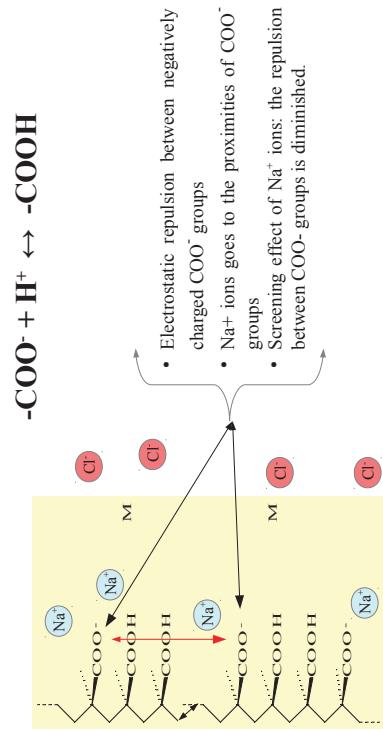
Effect of inert ions



But, K_C is not constant!
Depend on concentration of ions

System: H⁺ - PAA (polyacrylic acid)

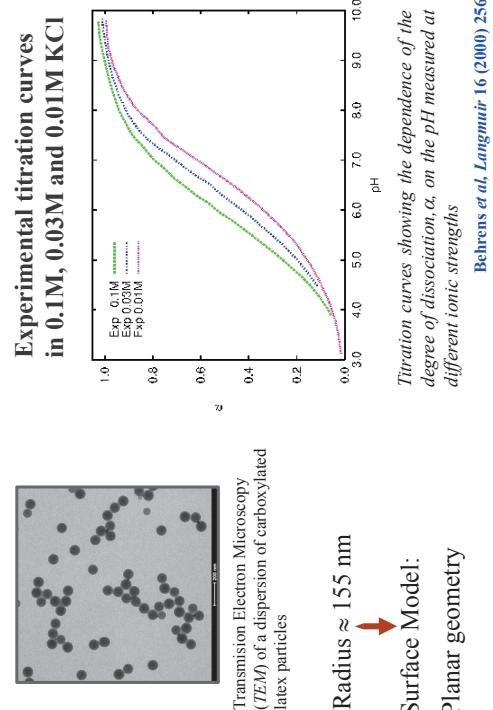
Polyelectrolytic effect
Effect of the ionic strength



At fixed pH, the increase of I makes a increase of the charge of the polyelectrolyte. Charge screening effect.

CARBOXYLATED LATEX PARTICLES: pH EFFECT

Experimental titration curves
in 0.1M, 0.03M and 0.01M KCl

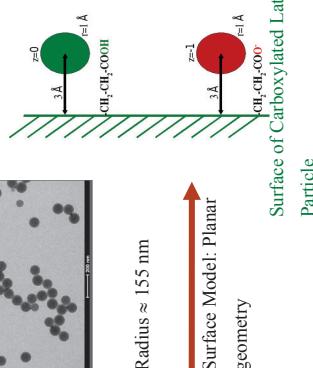
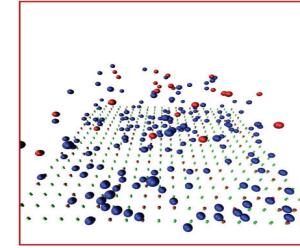


Theriation curves showing the dependence of the degree of dissociation, α , on the pH measured at different ionic strengths

Behrens et al, *Langmuir* 16 (2000) 2566

CARBOXYLATED LATEX PARTICLES

Transmission Electron Microscopy (TEM) of a dispersion of carboxylated latex particles
Behrens et al, *Langmuir* 16 (2000) 2566



SEMI-GRAND CANONICAL MONTE CARLO SIMULATIONS

*Canonic Monte Carlo criterion is used to sample the configurations of salt solution.

*Macrocanonic Monte Carlo criterion is used to attempt for the ionization and deionization process of each ionizable group of the polyelectrolyte.

$$\Delta F^{SGC} = \Delta E_c \pm k_B T \ln 10 (pH \cdot pK_0)$$

$$\min \{1, \exp(-\Delta F)\}$$

KCl Electrolyte:

$$c_{KCl} = 0.3-0.01 \text{M}$$

$$\sigma_{max} = 0.098 \text{ C/m}^2$$

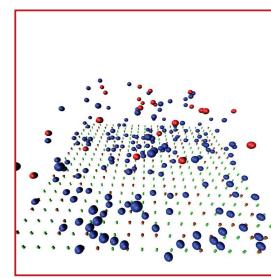
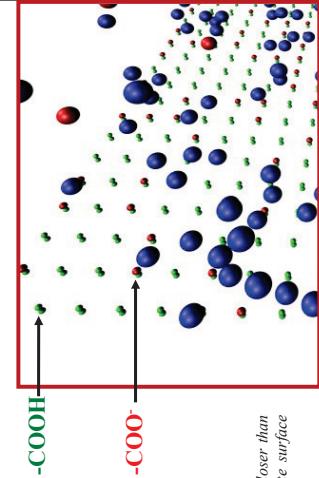
Using hydrated ionic radius of K⁺ and Cl⁻:

$$r_h(K^+) = r_h(Cl^-) = 3.3 \text{ \AA}$$

$$\text{to}$$

$$r_i(K^+) = 1.5 \text{ \AA} \quad r_i(Cl^-) = 1.8 \text{ \AA}$$

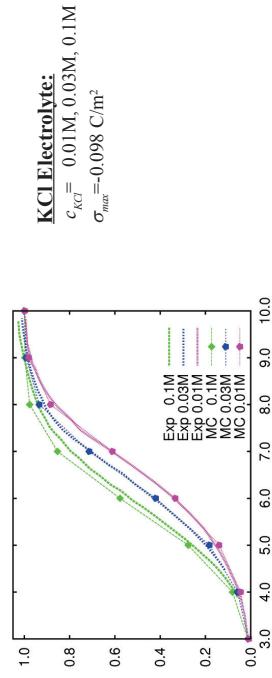
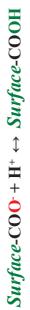
Total number of ions:
238-2780 depending I and pH



Snapshot for a solution of KCl 0.01M with a 0.098 C/m² charged polyelectrolyte surface.

Only counterions (blue) and coions (red) closer than a distance of 30 Å to the polyelectrolyte surface charges (green and red) are shown.

Degree of surface ionization (α) vs pH for a carboxylic functionalized surface in presence of inert salts



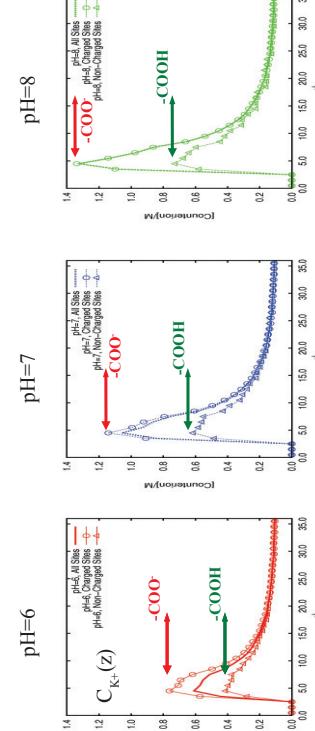
Titration curves showing the dependence of the degree of dissociation, α , on the pH were calculated using different radii values

Titration profiles α vs pH are in good correspondence with the experimental profiles for the MC profiles for all c_{KCl} concentrations.

Layering of Counterions around -COOH and -COO-

[K⁺] vs distance to -COOH or -COO-

KCl 0.1 M

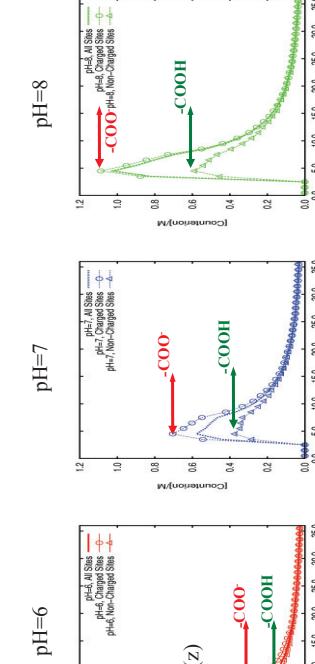


S. Madurga, et al., *J. Chem. Phys.* 135, 184103 (2011)

Layering of Counterions around -COOH and -COO-

[K⁺] vs distance to -COOH or -COO-

KCl 0.01 M



S. Madurga, et al., *J. Chem. Phys.* 135, 184103 (2011)

Simulation Techniques for Soft Materials

SGCMC for conformation and ionization

Case Study: conformational and ionization properties of linear poly(ethyleneimine) (LPEI)

1) What is a soft material?

2) Stability.

– Double Layer \rightarrow NVT MC

– Effect of pH \rightarrow Semi-Grand Canonical (SGCMC)

3) Flexible materials.

– Conformational vs pH \rightarrow Semi-Grand Canonical (SGCMC)

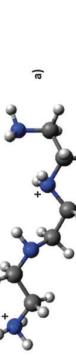
4) Water/Ion permeability

– Effect of Charge/Ionic strength \rightarrow Molecular Dynamics (MD)

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LPEI: $\text{NH}_2-(\text{CH}_2)_2-\text{[NH}-(\text{CH}_2)_2-\text{NH}_2-(\text{CH}_2)_2-\text{NH}_2$



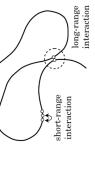
Description of a linear polymer:

- Site binding model (SB) with the rotational isomeric state (RIS): only three angles, those corresponding to energy minima of the bonds (i.e. trans, gauche+ and gauche-)
- Ionization state (charged/neutral) of the amine groups.

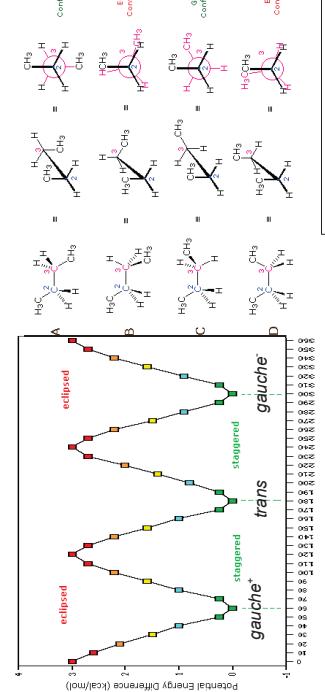
Fundamental concepts

In general, internal angle rotation energy is similar to $k_b T$. As a consequence, macromolecules can easily change their conformational state (i.e. they are flexible).

- Short range interactions
- Long range interactions

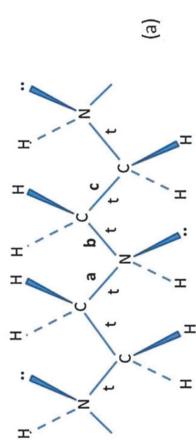


Conformational energy profile



Relevant conformations: trans, gauche⁺, gauche⁻

Site binding/rotational isomeric state (SBRIS) model



Three different dihedral angles:

a: C-C-N-C

b: C-N-C-C

c: N-C-C-N

Accurate description of the conformational properties requires different energetic parameters for **a**, **b** and **c** bonds.

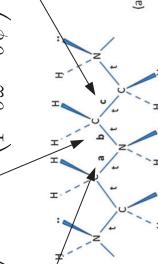
Transfer matrices for **a**, **b** and **c** of LPEI:

$$\Xi_{ref} = qU_{c,end}(U_{a,end}U_bU_c)(U_aU_bU_c)(U_aU_bU_c)...(U_aU_{b,end}U_{c,end})p^T$$

$$U_i : \text{Correlation of two consecutive dihedral angles}$$

$$\Phi_{ij} = \frac{t}{g^+} \begin{pmatrix} t-t & t-g^+ & t-g^- \\ g^+-t & g^+-g^+ & g^+-g^- \\ g^- & g^- - g^+ & g^- - g^- \end{pmatrix}$$

$$U_a = \begin{pmatrix} 1 & \sigma & \sigma \\ 1 & \sigma\psi & \sigma\omega \\ 1 & \sigma\omega & \sigma\psi \end{pmatrix} \quad U_b = \begin{pmatrix} 1 & \sigma & \sigma \\ 1 & \sigma\psi' & \sigma\omega' \\ 1 & \sigma\omega' & \sigma\psi' \end{pmatrix} \quad U_c = \begin{pmatrix} 1 & \sigma' & \sigma' \\ 1 & \sigma'\psi & \sigma'\omega \\ 1 & \sigma'\omega & \sigma'\psi \end{pmatrix}$$



$$\text{Debye length is: } \kappa^{-1} = \left(\frac{k_B T_c}{2 N_A e^2 I}\right)^{1/2}$$

$$\text{Ionic strength: } I = 1/2 \sum_{i=1}^n c_i z_i^2$$

$$F_{DH} = \sum_{j>i+1} \frac{e^2}{4\pi r_{ij}} \exp(-\kappa r_{ij})$$

Long-range Interactions

• Excluded Volume:

Each group has a hard-sphere radius:
Amine group: 1.55 Å
Methylene group: 1.70 Å

• Electrostatic interactions:

The electrostatic interactions between distant amine groups are mediated by solvent and the ions in solution.

$$\text{Debye-Hückel potential:}$$

$$F_{DH} = \sum_{j>i+1} \frac{e^2}{4\pi r_{ij}} \exp(-\kappa r_{ij})$$

SGCMC steps

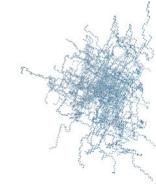
- Two configurational MC changes:
 - 1) Rotational state of a bond.
 - One bond is selected at random. The rotational change of a bond involves a ± 20 degree rotation of its dihedral angle.
 - * ... \leftrightarrow Gauche \leftrightarrow trans \leftrightarrow gauche \leftrightarrow ...
 - If some overlap takes place, the new configuration is rejected.
 - Atoms of the polyamine has a hard sphere repulsion.
- 2) Binding state of amines

$$\begin{cases} \Delta F = \Delta U_{\text{el}} + k_B T (\ln 10) / (\text{pH} - \log K_0) & \text{for protonation process} \\ \Delta F = \Delta U_{\text{el}} - k_B T (\ln 10) / (\text{pH} - \log K_0) & \text{for deprotonation process} \end{cases}$$

Apply ΔF for Metropolis MC algorithm. Accepted movement with probability: $\min \{1, \exp(-\Delta F)\}$

SGCMC of Polyamine

- Study of the conformational and ionization properties of linear polyamine

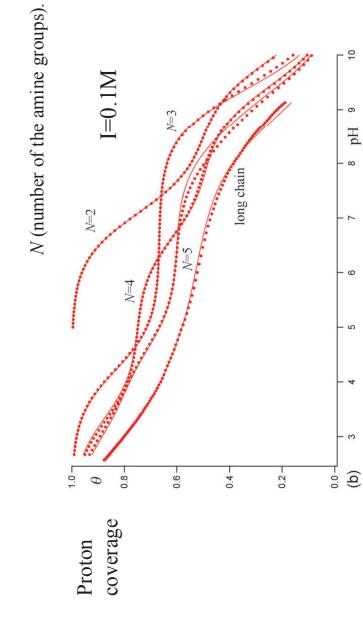


100 different polymer configurations



JL. Garcés, S. Madurga, M. Borkovec, *Phys.Chem.Chem.Phys.*, 2014, 16, 4626.

Titration curves for different LPEI



Dotted lines: Experimental binding curves.

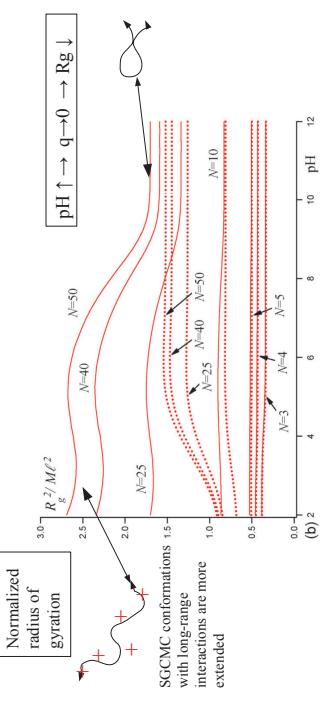
Continuous lines: Simulation results.

JL. Garcés, S. Madurga, M. Borkovec, *Phys.Chem.Chem.Phys.*, 2014, 16, 4626.

SGCMC simulations.

Equilibration: $2 \cdot 10^6$ configurations

Sampling: $8 \cdot 10^8$ configurations



Continuous lines: Simulation: SGCMC. (the excluded volume + long range electrostatic interactions)

Dotted lines: Theory (without long-range interactions).

JL. Garcés, S. Madurga, M. Borkovec, *Phys.Chem.Chem.Phys.*, 2014, 16, 4626.

Simulation Techniques for Soft Materials

Modeling the spherical colloid

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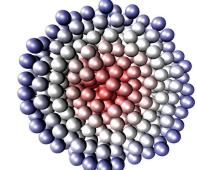
- Conformational vs pH \rightarrow Semi-Grand Canonical (SGCMC)

4) Water/Ion permeability

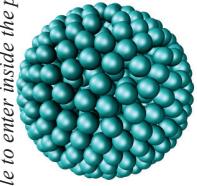
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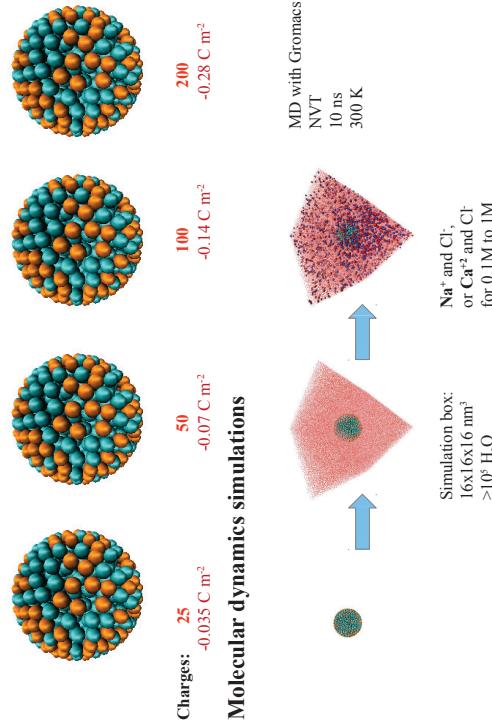
6 concentric layers



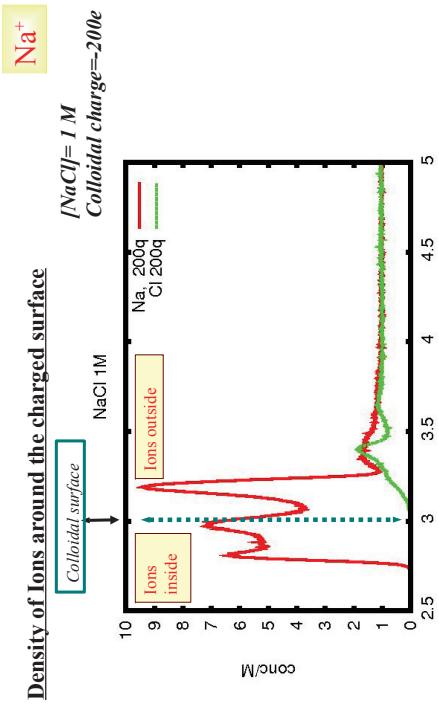
6 nm

M. Nedylkova, S. Madurga, et.al, *J. Chem. Phys.*, 137 (2012) 174701

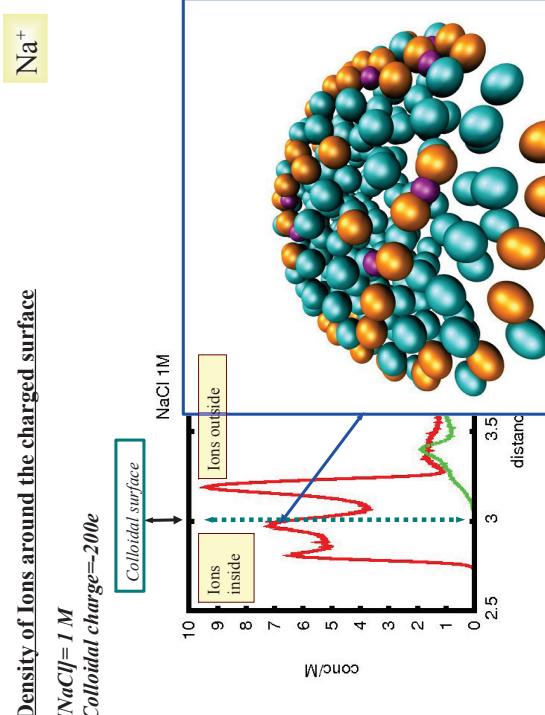
Modeling the spherical colloid



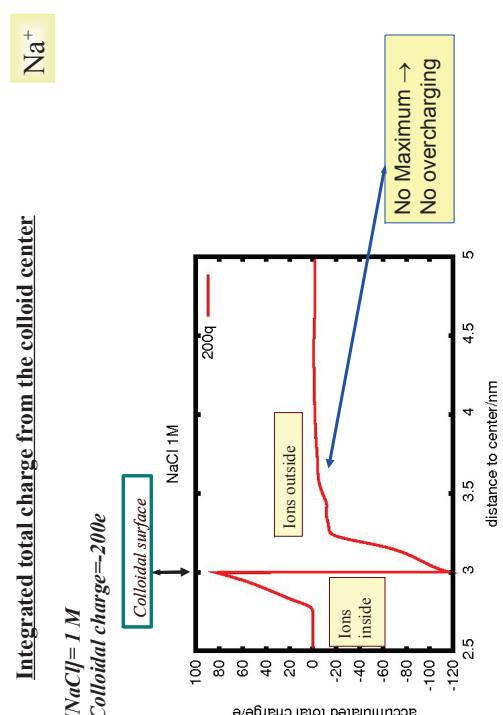
MD SIMULATIONS WITH MONOVALENT SALT



M. Nedyalkova, S. Madurga, et al. *J. Chem. Phys.* 137 (2012) 174701

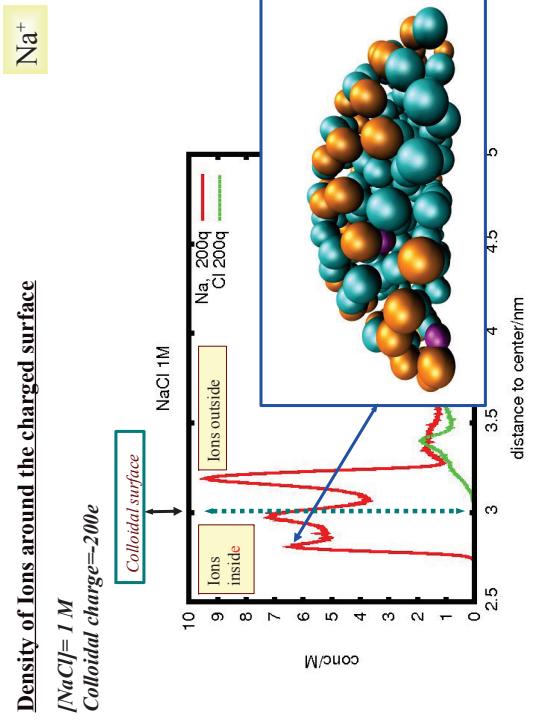


M. Nedyalkova, S. Madurga, et al. *J. Chem. Phys.* 137 (2012) 174701

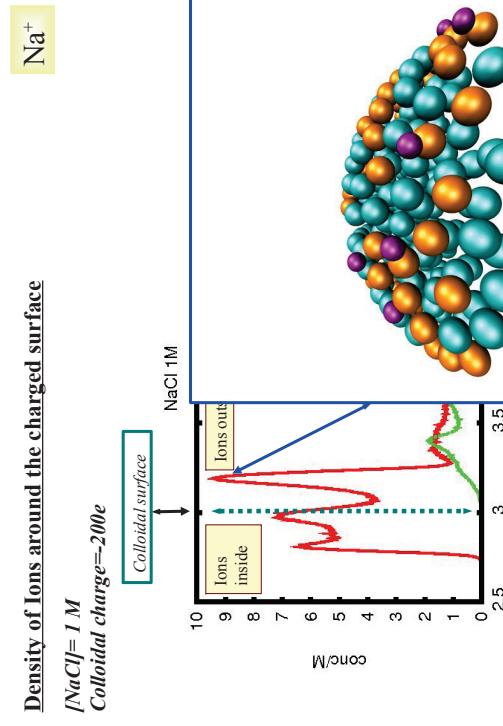


No Maximum → No overcharging

M. Nedyalkova, S. Madurga, et al. *J. Chem. Phys.* 137 (2012) 174701



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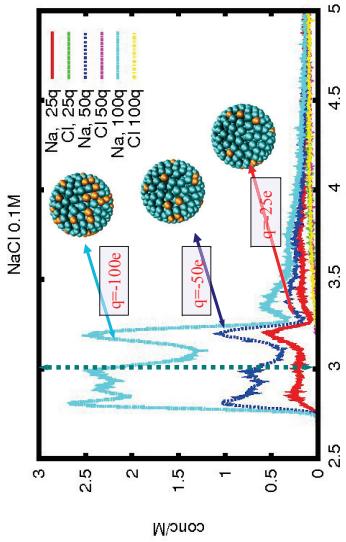


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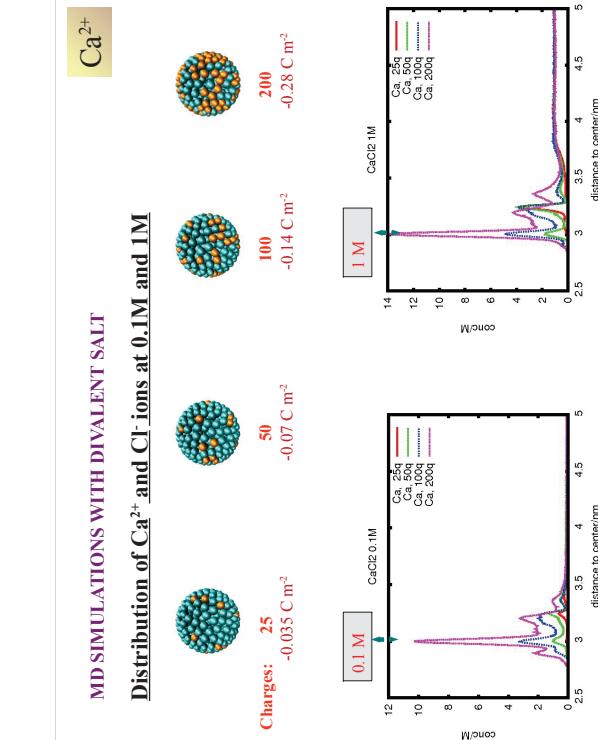
Comparison of different cases for Na at 0.1 M

- Greater amount of Na^+ are accumulated in the surface of the colloidal particle with the greater charge.
- Ions are accumulated between the two layers of methanes, between methanes of the last layer and outside the colloid.



MD SIMULATIONS WITH DIVALENT SALT

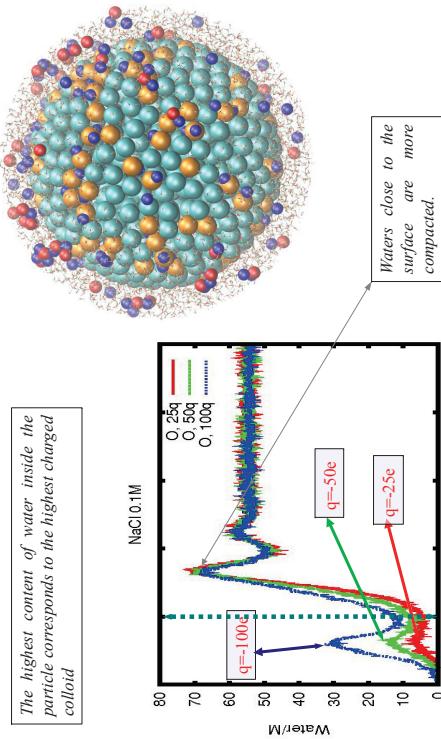
Distribution of Ca^{2+} and Cl^- ions at 0.1M and 1M



Na^+

Distribution of water molecules around the colloidal particle

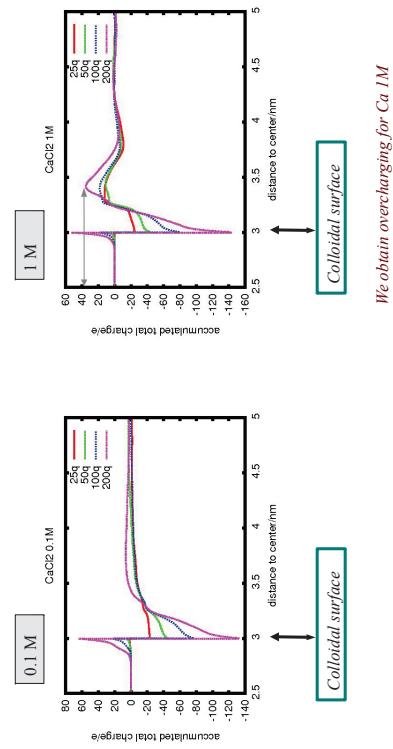
The highest content of water inside the particle corresponds to the highest charged colloid



M. Nedalkova, S. Madurga, et.al, *J. Chem. Phys.*, 137 (2012), 174701

Integrated total charge from the colloid center

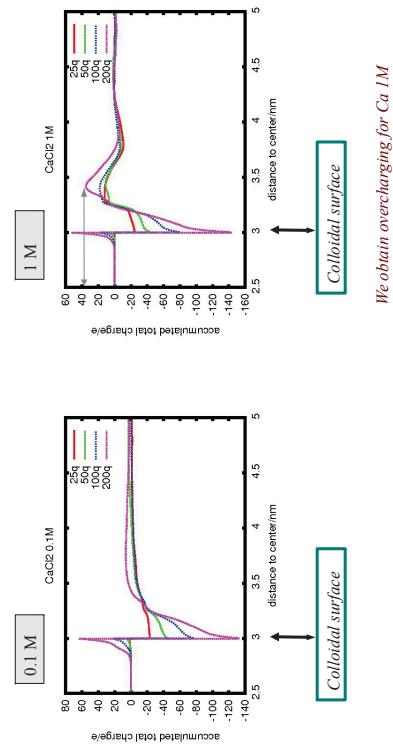
$[\text{CaCl}] = 0.1 \text{ and } 1 \text{ M}$
Colloidal charge = -25, -50, -100 and -200e



We obtain overcharging for $\text{Ca} 1 \text{ M}$

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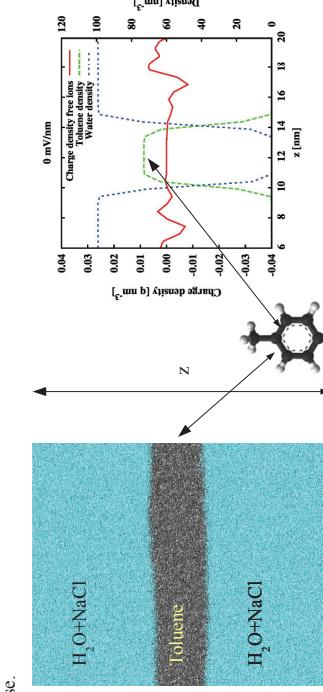


We obtain overcharging for $\text{Ca} 1 \text{ M}$

Ca^{2+}

Layer of Toluene. Effect of Voltage

Electro-induced instability of a thin toluene emulsion film in contact with saline aqueous phase.

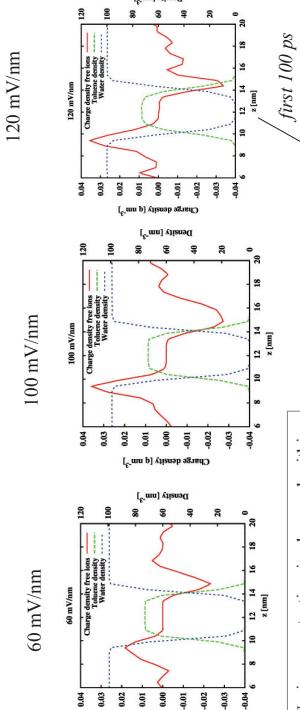


Then, external electric field applied perpendicularly to the toluene film

D. Dimitrova, S. Pisov, N. Panchev, M. Nedalkova, S. Madurga, A. Proykova, *J. Chem. Phys.*, (2017)

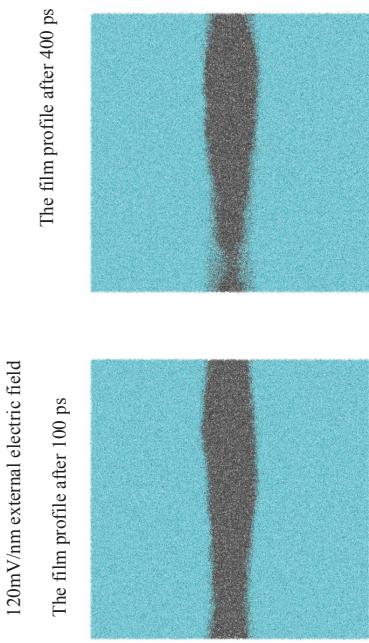
Effect of applied Voltage

Electro-induced instability of a thin toluene emulsion film in contact with saline aqueous phase.



Great accumulation of ions
After 400-500 ps the film is

MD (NVT/NPT)+Voltage

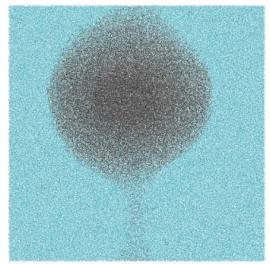


Startup of rupturing of the thinnest part of the film - between 400 ps and 500 ps

MD (NVT/NPT)+Voltage

120mV/mm external electric field

Breakdown of the film and formation of a toluene drop at about 2000 ps



- When critical electric field is reached within a certain time after the field application, **electric discharge** occurs indicating the beginning of the rupturing process.
- Molecular dynamics simulations demonstrate the **role of charge accumulation** in the toluene-film rupture induced by a DC electric field.

Simulation Techniques for Soft Materials

- 1) What is a soft material?
- 2) Stability.
 - Double Layer → NVT MC
 - Effect of pH → Semi-Grand Canonical (SGCMC)
- 3) Flexible materials.
 - Conformational vs pH → Semi-Grand Canonical (SGCMC)
- 4) Water/Ion permeability
 - Effect of Charge/Ionic strength → Molecular Dynamics (MD)
- 5) Diffusion and Reactivity
 - Effect of crowding/aggregation → Brownian Dynamics (BD)

Diffusion and Reactivity

Study effect of Mobility and Reactivity in crowding media

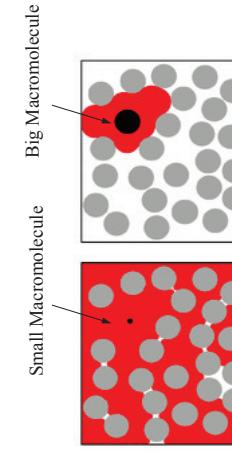
Brownian Dynamics (BD)

- Introduction:
 - What is macromolecular crowding?
 - Diffusion in a crowded media
- Brownian Dynamics (BD) simulations:
 - How to model Dextran?
 - Hydrodynamic Interactions (HI)
 - Reactivity

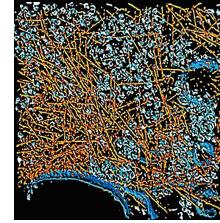
What is macromolecular crowding?

→ Environment of High concentration of macromolecules.

→ Crowding affect the properties (Diffusion & Reactivity).



The accessible volume (red) for two molecules of different sizes (black circles)



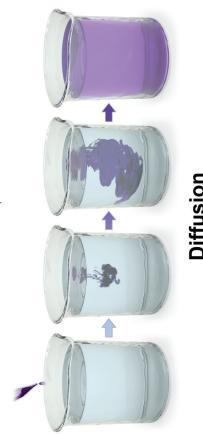
Diffusion in a crowded media

Diffusion in a crowded media

Normal Diffusion:

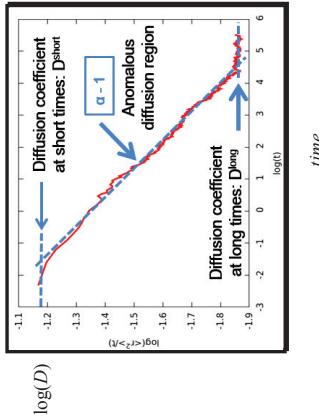
$$\text{Einstein-Smoluchowski equation: } \langle r^2 \rangle = (2d)Dt \quad \langle r^2 \rangle > = (2d)Dt$$

$$\text{Stokes-Einstein equation: } D = \frac{k_B T}{6\pi\eta R_H} \quad D(t) = \frac{1}{2d} t^{\alpha-1} \quad \begin{cases} \alpha = 1 & \text{Normal Diffusion} \\ \alpha > 1 & \text{Superdiffusion} \\ \alpha < 1 & \text{Subdiffusion} \end{cases}$$



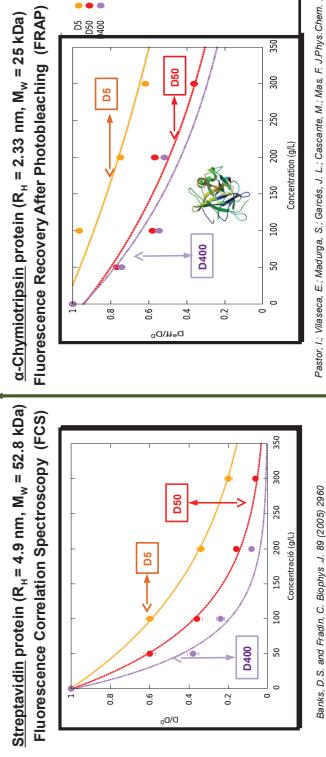
Diffusion

Off lattice simulations (Brownian Dynamics):



Vilaseca, E.; Isvoran, A.; Madurga, S.; Pastor, I.; Garcés, J.L.; Mas, F. *Phys. Chem. Chem. Phys.*, 13 (2011) 7396

Experimental evidences: Diffusion studies in crowded media



Banks, D.S. and Freire, C. *Biophys. J.* 89 (2005) 2980
Pastor, I.; Vilaseca, E.; Madurga, S.; Garcés, J.L.; Casamitja, M.; Mas, F. *J. Phys. Chem. B*, 114 (2010) 12148-12152
Vilaseca, E.; Isvoran, A.; Madurga, S.; Pastor, I.; Garcés, J.L.; Mas, F. *Phys. Chem. Chem. Phys.*, 13 (2011) 7396

Crowding agents

Dextran	M_w (kDa)	$R_{\text{effective}}$ (nm)
D5	5.2	1.23
D50	48.6	2.93
D400	409.8	6.69

Diffusion coefficient reduction depends not only on the obstacle concentration but on protein size!

Brownian Dynamics simulations:

- The Langevin equation allows to simulate implicitly the solvent adding a stochastic force $F^R(t)$:

$$m \frac{dv(t)}{dt} = F_T(t) - F^D(t) + F^{\text{fric}}(t) + F^R(t)$$

- In the over-damped limit, this equation can be integrated. This gives the BD equation of motion:

$$\mathbf{r}(t + \Delta t) = \mathbf{r}(t) - \frac{\Delta t D \nabla V(\mathbf{r})}{K_B T} + \sqrt{2 D \Delta t} \xi(t)$$

- In our model, a coarse-graining of macromolecules is done and each macromolecule is modeled as a single sphere

- An harmonic pairwise repulsion potential is used in order to avoid macromolecule overlapping:

$$V_{ij} = \begin{cases} \frac{1}{2} k (d_{ij} - R_C)^2 & d_{ij} < R_C \\ 0 & d_{ij} \geq R_C \end{cases} \quad R_C = R_{C,i} + R_{C,j}$$

P.M. Blanco, M. Vía, J.L. Garcés, S. Madurga, F. Mas, *Entropy* 105 (2017)

Brownian Dynamics: Hydrodynamic Interactions

- Hydrodynamic interactions (HI) emerge from the fact that Brownian motion of a particle generates solvent fluxes which disturb the other particles motion.

- Procedures to include HI:
 - Ermak and McCammon using the Diffusion tensor (precise but costly).

$$\mathbf{r}(t + \Delta t) = \mathbf{r}(t) - \frac{\Delta t}{K_B T} \mathbf{D} \cdot \mathbf{F}(\mathbf{r}) + \sqrt{2 \Delta t} \mathbf{B} \cdot \xi(t)$$

$3N \times 3N$ Diffusion tensor

→ Tokuyama (mean field, fast procedure).

$$\mathbf{r}(t + \Delta t) = \mathbf{r}(t) - \frac{\Delta t D^{\text{short}} \nabla V(\mathbf{r})}{K_B T} + \sqrt{2 D^{\text{short}} \Delta t} \xi(t) \quad D^{\text{short}}(\phi) = \frac{D_0}{[1 + H(\phi)]}$$

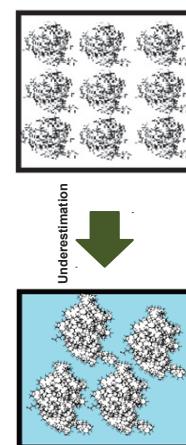
$$H(\phi) = \frac{2b^2}{1 - b} - \frac{c}{1 + 2c} - \frac{bc(2 + c)}{(1 + c)(1 - b + c)}$$

D.L. Ermak and J.A. McCammon *J. Chem. Phys.* 69 (1978) 352

How to model Dextrans?



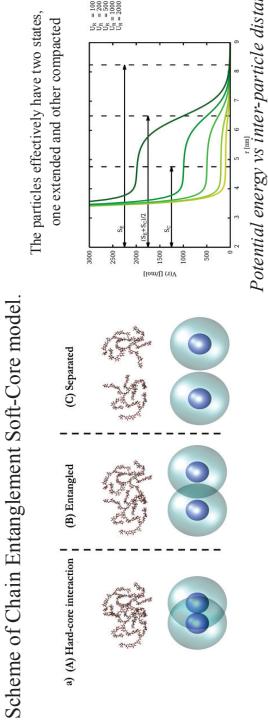
Compact radius:
Radius that would have the macromolecule fit was in solid state



Effective radius
Experimental hydrodynamic radius:
Effective hydrated radius of a hard sphere that diffuses at the same rate as the macromolecule in dilute solution

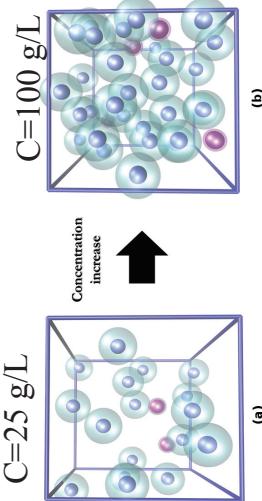
Compact radius

How to model Dextrans



- Macromolecules are modelled as spheres with two different shells.
- The outer shell (transparent blue) accounts for the soft-core repulsion between macromolecular branches.
- The inner shell (opaque blue) takes into account the hard-core steric repulsion due to the increasing steric repulsion.

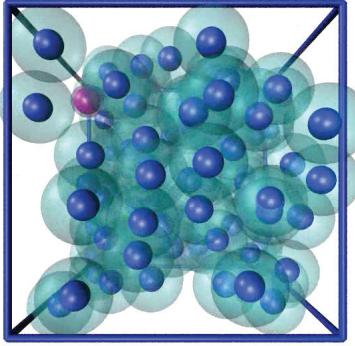
BD Simulation



Simulation box of two BD simulations of streptavidin protein (purple) diffusing among D50 dextran obstacles (blue). As the concentration increases, steric compression promotes macromolecules to become entangled.

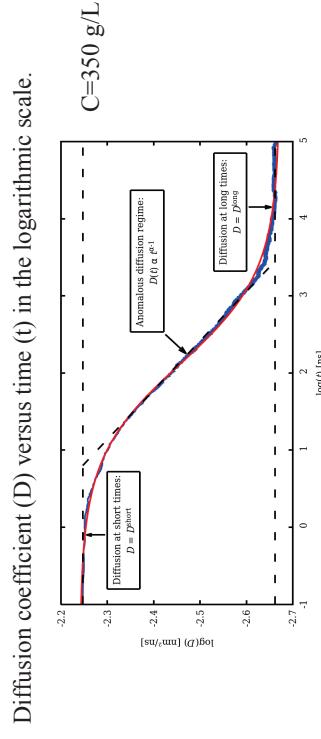
BD Simulation

C=100 g/L



Size: 43.2 nm
Enzymes: 5
Dextrans: 100
radii E.: 3.04/2.45 nm
radii D.: 5.2/2.3 nm

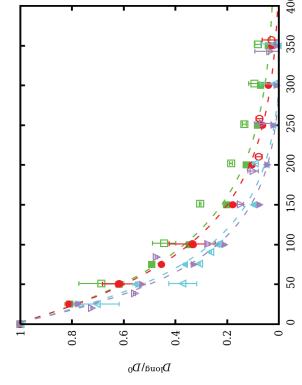
Diffusion in a crowded media



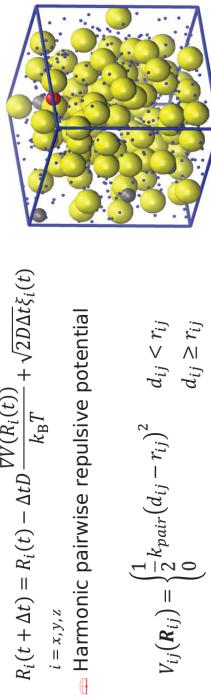
3000 BD simulations in a cubic simulation box with a side of 28.466 nm. 5 tracer particles (streptavidin) diffusing among 100 obstacle particles (D50 dextran). Macromolecular crowding slows down the tracer diffusion coefficient from D_{short} to D_{long} .

Effect of crowding

D_{long} for streptavidin versus macromolecular concentration.



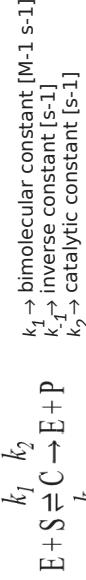
Reactivity – Brownian Dynamics



Michaelis-Menten mechanism

Reactivity

Michaelis-Menten mechanism



Probability of Reaction:

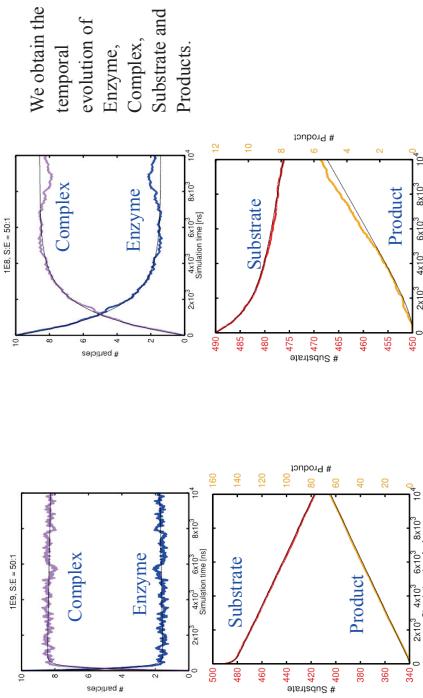
$$p_{\text{react}} = k_{\text{micro}} * \Delta t \quad (0 < p_{\text{react}} < 1)$$

So we can use a Monte Carlo criterion

If random > p_{react} → **no reaction**

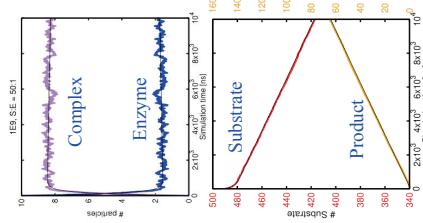
If random ≤ p_{react} → **reaction**

Slow Enzyme



We obtain the temporal evolution of Enzyme, Complex, Substrate and Products.

Fast enzyme



Results: Reactivity



Research Group

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