Computational Spectroscopy of Reverse Micelles

Philipp Honegger

Department of Computational Biological Chemistry

University of Vienna



Crowding and encapsulation

 Chemical environment of an actual cell differs from that of a diluted buffer



Impacts both thermodynamic equilibria and kinetic reaction rates





Images taken from A. P. Minton, J. Cell Sci. 2006 119, 2863-2869

Macromolecular confinement

- Impact on:
 - Molecular mobility
 - Folding rates
 - Aggregation behaviour
 - Biomolecular equilibria
 - Reaction rates



Diagram taken from R. John Ellis, Trends Biochem. Sci. 2001, 26, 2863-2869

Reverse micelles as cell mimics



hydrophobic exterior

Example surfactant molecules:



bis(2-ethylhexyl)-sulfosuccinate (aerosol-OT, AOT)



65 1-decanoyl-rac-glycerol (DMAG) / 35 lauryldimethylamine-N-oxide (LDAO)

Spectroscopy: A molecule-level microscope

The umbrella term spectroscopy encompasses various kinds of measurements of the interaction between matter and electromagnetic waves.

Depending on the latter's frequency, different constituents of physical matter will resonate with the incoming radiation.



M. Schmollngruber







time step t

Atom count n Atom positions **r**(n,x,y,z,t) Atomic masses m(n) Atomic charges q(n) Nonbonded parameters Bonds, Angles, Dihedrals

→ evaluate potentials





time step t

Atom count n Atom positions **r**(n,x,y,z,t) Atomic masses m(n) Atomic charges q(n) Nonbonded parameters Bonds, Angles, Dihedrals

→ evaluate potentials

forces as the negative gradient of the potential

 \rightarrow calculate forces





→ calculate

r(n,x,y,z,t+1)

Atom count n Atom positions **r**(n,x,y,z,t) Atomic masses m(n) Atomic charges q(n) Nonbonded parameters Bonds, Angles, Dihedrals

time step *t*

→ evaluate potentials

5







δł

electrostatic repulsion, electrostatic attraction

time step <i>t</i>	potential function V(n)	forces f(n,x,y,z,t)	time step <i>t</i> +1
Atom count n Atom positions r (n,x,y,z,t) Atomic masses m(n) Atomic charges q(n) Nonbonded parameters Ronds, Anglos, Dibodrals	Calculate the atomic forces as the negative gradient of the potential → calculate forces	Integrate the equations of motion to get the new atomic positions	Iterate this t _{max} times to obtain the trajectory: r(n,x,y,z,t) v(n,x,y,z,t)
→ evaluate potentials		→ carculate r(n,x,y,z,t+1)	5

Why MD simulations?

... and not Quantum Mechanics (QM)?



"The only way to calculate spectroscopic observables correctly is to use QM"

Why Molecular Dynamics Simulations?

- QM is size-limited, hence limited to some classes of matter.
 Soft matter systems:
 - Condensed like solids
 - Dynamics like fluids
 - Complicated dynamics with manifold processes



Why Molecular Dynamics Simulations?

- QM is size-limited, hence not always applicable Soft matter systems:
 - Condensed like solids
 - Dynamics like fluids
 - Complicated dynamics with manifold processes
- QM not always better at reproducing observables than MD!

QM bound to fail if the statistic stability of an observable

- Requires many time frames
- Requires many particles
- Requires large spatial extension (boundary conditions)

Why MD simulations?



→ Use the right tool for the right job!

MD simulations of reverse micelles



MD simulations of reverse micelles



9

MD simulations of reverse micelles

Computational spectroscopy – Link between experiment and simulation!









In theory, any spectral function can be calculated as the Fourier transform of the correlation function of the corresponding molecular property.

A wide range of such properties are accessible from molecular dynamics simulations.





some observable spectral feature $\tilde{A}(\omega)$



In theory, any spectral function can be calculated as the Fourier transform of the correlation function of the corresponding molecular property.

A wide range of such properties are accessible from molecular dynamics simulations.





M. Schmollngruber



100

1000

10

v /MHz

0.1

1

10000

In theory, any spectral function can be calculated as the Fourier transform of the correlation function of the corresponding molecular property.

A wide range of such properties are accessible from molecular dynamics simulations.





M. Schmollngruber

Fast Field Cycling NMR dispersion R(ω)



In theory, any spectral function can be calculated as the Fourier transform of the correlation function of the corresponding molecular property.

A wide range of such properties are accessible from molecular dynamics simulations.





Forward Fourier Transformation M. Schmollngruber

Fast Field Cycling NMR dispersion R(ω)



In theory, any spectral function can be calculated as the Fourier transform of the correlation function of the corresponding molecular property.

A wide range of such properties are accessible from molecular dynamics simulations.





M. Schmollngruber

Infrared spectroscopy Τ(ύ)



In theory, any spectral function can be calculated as the Fourier transform of the correlation function of the corresponding molecular property.

A wide range of such properties are accessible from molecular dynamics simulations.





M. Schmollngruber



Terahertz

12

In theory, any spectral function can be calculated as the Fourier transform of the correlation function of the corresponding molecular property.

A wide range of such properties are accessible from molecular dynamics simulations.





M. Schmollngruber

Ionic conductivity $\Sigma(\omega)$



t = 0



 $\langle \mathrm{A}(0) {ullet} \mathrm{A}(0)
angle$











 $\langle \mathrm{A}(0) {ullet} \mathrm{A}(0)
angle \ \langle \mathrm{A}(0) {ullet} \mathrm{A}(\Delta \mathrm{t})
angle$











 $\tilde{A}(\omega) = \int_{-\infty}^{+\infty} \langle A(0)A(t)\rangle e^{-i\omega t} dt$

Water under confinement

Difference single-particle vs. collective dynamics



- Motion of single molecular dipoles
- Average



- Motion of sample sum dipole
- Sum
- Contains cross correlations

Connected via mutual orientational structure (?)

Water under confinement





DRS absorption spectra of aqueous reverse micelles



DRS absorption spectra of aqueous reverse micelles



Collective dynamics include cross-correlations!

			₩(t)				
	•		$\vec{\mu_1}(t)$	$\vec{\mu}_2(t)$	$\vec{\mu}_{3}(t)$	$\vec{\mu}_4(t)$	
			*	*			
, (0)	,µ₁(0)		$\vec{\mu}_1(0) \cdot \vec{\mu}_1(t)$	$\vec{\mu}_1(0) \cdot \vec{\mu}_2(t)$	$\vec{\mu}_1(0) \cdot \vec{\mu}_3(t)$	$\vec{\mu}_1(0) \cdot \vec{\mu}_4(t)$	
	,µ2(0)	×	$\vec{\mu}_2(0) \cdot \vec{\mu}_1(t)$	$\vec{\mu}_2(0) \cdot \vec{\mu}_2(t)$	$\vec{\mu}_2(0) \cdot \vec{\mu}_3(t)$	$\vec{\mu}_2(0) \cdot \vec{\mu}_4(t)$	
	, µ₃(0)		μ ₃ (0)•μ₁(t)	μ ₃ (0)•μ ₂ (t)	μ ₃ (0)•μ ₃ (t)	μ ₃ (0)•μ ₄ (t)	
	μ _₄ (0)		$\vec{\mu}_4(0)\cdot\vec{\mu}_1(t)$	$\vec{\mu}_4(0)\cdot\vec{\mu}_2(t)$	$\vec{\mu}_4(0)\cdot\vec{\mu}_3(t)$	$\vec{\mu}_4(0)\cdot\vec{\mu}_4(t)$	

Mutual dipole orientation: Kirkwood g-factor

g

Why does structure modulate dynamics? A thought experiment

... but is this a real thing?

Experimental setup: Phosphatidylcholines as surfactants, onionlike vesicles, dielectric spectroscopy

Gun-Sik Park, Seoul National University

but is this a real thing?

Gun-Sik Park, Seoul National University Micelle, unconfined water

Pure water

Micelle

40

30

20

10

n

Dielectric loss

0

0

0

Reverse micelle, confined water

Can this be generalized to biological cells?

- Size beyond fully atomistic MD simulation
- Idea: Solve dielectric equations directly using a concentric spherical model

Can this be generalized to biological cells?

\rightarrow Embedding mechanisms

Thank you for your attention!

