Ab Initio and Force Field Modeling of Sorption Phenomena Associated with Water Treatment

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Molecular Modeling



Quasi-exponential growth in modeling.
Will replace i-Pod as the latest fad by 2009*.

Outline

- Computational methods
- TCE adsorption in adsorbent micropores
- Chemisorption of TCE and PCE on iron surfaces
- Complexation of H₃AsO₃ with ferric hydroxides

Computational Methods

All are based on energy minimization or conformance with a Boltzmann distribution of energies.

Molecular Mechanics - Empirical Force Fields (physical interactions) Monte Carlo Methods (MC) Canonical Monte Carlo Grand Canonical Monte Carlo Molecular Dynamics (MD)

Quantum Mechanics Methods (QM) - Ab Initio (chemical reactions) Molecular Orbital (MO) Density Functional Theory (DFT)

Properties Accessible by Calculation

• molecular properties: bond lengths, bond angles, bond energies, vibrational spectra, partial atomic charges, dipole moments

- arrangements of molecular systems: crystal structure, associations in solution, identification of adsorption or enzyme binding sites
- thermodynamic properties: heat capacities, equilibrium constants and activation energies for reactions, Henry's law constants, solubilities, enthalpies and Gibbs energies of formation, reaction and adsorption
- transport properties: diffusion coefficients, viscosities
- time evolution of molecular systems

Methods Comparison



Molecular Mechanics Force Field



Parameters for each atom type may be empirically measured or determined from *ab initio* calculations.

$$E_{P} = \underbrace{E_{B} + E_{A} + E_{T} + E_{I}}_{\text{bonded}} + \underbrace{E_{vdW} + E_{Elec}}_{\text{nonbonded}}$$

$$E_{B} = \text{bond stretch energy}$$

$$E_{A} = \text{ angle bending energy}$$

$$E_{T} = \text{ dihedral angle torsion energy}$$

$$E_{op} = \text{ out of plane bending energy}$$

$$E_{vdW} = \text{ van der Waals energy}$$

$$E_{elec} = \text{electrostatic energy}$$

$$\begin{aligned} \mathbf{E}_{p}(\mathbf{r}^{N}) &= \sum_{\text{bonds}} \frac{k_{i}}{2} (l_{i} - l_{i,o})^{2} + \sum_{\text{angles}} \frac{k_{i}}{2} (\theta_{i} - \theta_{i,o})^{2} + \sum_{\text{torsions}} \frac{V_{n}}{2} [1 + \cos(n\omega - \gamma)] + \sum_{\text{op-angles}} \frac{k_{i}}{2} \theta_{i}^{2} \\ &+ \sum_{i=1}^{N} \sum_{j=i+1}^{N} \left(4\varepsilon_{ij} \left[\left(\frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left(\frac{\sigma_{ij}}{r_{ij}} \right)^{6} \right] + \frac{q_{i}q_{j}}{4\pi\varepsilon_{o}r_{ij}} \right) \end{aligned}$$

Quantum Mechanics - Molecular Orbital Theory

Schröedinger Equation $\hat{H}\psi = \hat{H}\psi$

$$\widehat{H}\psi$$
 = $E\psi$

$$E\rangle = \int \psi^* \hat{H} \psi \, d\mathbf{r}$$

 $\psi(\mathbf{r}) = \mathbf{N}$ electron wave function for the system



 ψ is unknown but can be approximated as a product of one electron wave functions.

Single electron wave functions can be approximated as a linear combination of single electron atomic orbital functions.

Variational Principle

- Iterative solution required best approximate wave function (ψ_{trial}) is the one that gives the lowest energy.
- Goal is to minimize the calculated energy (E_{trial}) by changing the weighting coefficients in the linear combination of atomic orbitals (LCAO).
- Energy given by the approximate wave function (E_{trial}) will be \geq true ground state energy (E_0) .

$$\left\langle \psi_{trial} \left| \hat{H} \right| \psi_{trial} \right\rangle = E_{trial} \ge E_o = \left\langle \psi_o \left| \hat{H} \right| \psi_o \right\rangle$$

Quantum Mechanics - Density Functional Theory

$$\left[-\frac{1}{2}\nabla_i^2 - \sum_A \frac{Z_A}{r_{iA}} + \int \frac{\rho(\vec{\mathbf{r}_2})}{r_{i2}} d\vec{\mathbf{r}_2} + V_{xc}\right] \psi_i(\mathbf{r}_i) = \varepsilon_i \psi_i(\mathbf{r}_i)$$

Kohn Sham Equation

 $\psi_i(\mathbf{r}_i) = \text{single electron orbital function}$ $\psi_i(\mathbf{r}_i) \sim \text{probability density function for electron } i$

$$\rho(\mathbf{r}) = \sum_{i=1}^{N} |\psi_i(\mathbf{r}_i)|^2$$

- For the same level of accuracy, DFT is much faster than MO methods.
- Hamiltonian is not exact: V_{xc} is unknown and must be approximated.
- LCAO used to approximate ψ_i for each electron.
- Iterative solution based on optimizing the coefficients in the LCAO.
- Best set of coefficients gives the lowest energy for each set of atomic positions.
- Energy (E_{trial}) given by the approximate electron density (ρ_{trial}) will be \geq or \leq true ground state energy (E_{o}) due to errors in V_{xc} .

$$E_{trial}[\rho_{trial}] \ge or \le E_0[\rho_0]$$

Approximation Errors for Quantum Methods

- Correlation error: LCAO approximation of ψ averages the electric field arising from the electrons and ignores instantaneous electron-electron repulsion. Electrons may get too close (Coulomb hole).
- Exchange error: Pauli exclusion principle results in electrons of the same spin being depleted near electrons with the same spatial orbitals. Not a problem with MO theory but DFT uses an empirical function for approximating the exchange energy (Fermi hole).
- Different QM methods use different ways of correcting for these errors.

DFT Geometry Optimization

- Evaluate the energy of the system for different positions of the atoms (nuclei). For each set of atomic positions, determine the set of ψ_i that yield the lowest energy.
- Calculate the gradient of the potential energy with respect to the atomic positions.
- Set the gradient equal to zero and use standard root finding techniques to determine the atomic coordinates for the stationary points.
- Evaluate second derivatives of the potential energy at the stationary points to find the global minimum.
- Activation energies can also be determined from the potential energy surface.



Potential Energy Surface

Canonical Monte Carlo Simulations

Constant: NVT



- 1. Evaluate the energy of the system (E_p) for any initial configuration of the molecules.
- 2. Randomly choose 1 molecule to move (translation, rotation).
- 3. Evaluate the new energy of the system after the move (E'_p)
- 4. Accept the move if: $E'_{p} \le E_{p}$ or if $exp\left[\frac{-(E'_{p} E_{p})}{kT}\right] \ge random_{0 \to 1}$

5. Reject move if
$$exp\left[\frac{-(E'_{p} - E_{p})}{kT}\right] \le random_{0 \to 1}$$

- 6. Repeat steps 2 through 5 until the energy of the system reaches a stable value. Stablilized value not the minimum energy.
- Final configuration is a balance between energy minimizing configuration and randomized positions due to thermal energy (Boltzmann distribution).
- Molecules are usually held rigid and bonds do not stretch or bend.
- Moves are not physical and time evolution of the system is meaningless.
- Displacement distance scaled such that $\sim 50\%$ of the moves are accepted.

Grand Canonical Monte Carlo Constant: µVT

The number of each type of molecule (N_i) may change over time.

1. Evaluate the energy of the system (E_p) for any initial configuration of the molecules.

- 2. Randomly choose 1 molecular move, insertion or deletion.
- 3. Evaluate the energy of the system after the molecular move, insertion or deletion (E'_{p}) .
- 4. For molecular moves, acceptance criteria same as NVT

5. If molecule is inserted, accept if: $E'_{p} \le E_{p}$ or if $exp\left[\frac{-(E'_{p} - E_{p})}{kT} - \ln\frac{(N_{i} + 1)kT}{f_{i}V}\right] \ge random$ 5. If molecule is deleted, accept if: $E'_{p} \le E_{p}$ or if $exp\left[\frac{-(E'_{p} - E_{p})}{kT} + \ln\frac{(N_{i})kT}{f_{i}V}\right] \ge random$

Molecular Dynamics

- Molecules follow Newton's 2nd Law: **F**=m**a**
- $\mathbf{F} = -dE_p/d\mathbf{r}$
- All molecules move at the same time
- Molecules are flexible and bonds may stretch and bend
- Captures true time evolution of the system
- Velocity scaling often used to control temperature
- Time steps of ~10⁻¹⁵ s (fs) constrain simulations to very short time spans (up to 10⁻⁹ s)

Molecular Mechanics Solvation Effects

Explicit water molecules:



Dielectric constant:

$$E_{p}(\mathbf{r}^{N}) = \sum_{\text{bonds}} \frac{k_{i}}{2} (l_{i} - l_{i,o})^{2} + \sum_{\text{angles}} \frac{k_{i}}{2} (\theta_{i} - \theta_{i,o})^{2} + \sum_{\text{torsions}} \frac{V_{n}}{2} [1 + \cos(n\omega - \gamma)] + \sum_{\text{op-angles}} \frac{k_{i}}{2} \theta_{i}^{2}$$
$$+ \sum_{i=1}^{N} \sum_{j=i+1}^{N} \left(4\varepsilon_{ij} \left[\left(\frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left(\frac{\sigma_{ij}}{r_{ij}} \right)^{6} \right] + \frac{q_{i}q_{j}}{4\pi\varepsilon_{o}r_{ij}} \right)$$

Useful for interactions at distances >> the size of a water molecule (~2.5 Å).



Quantum Mechanical Solvation Effects

Explicit Water Molecules



- Convergence problems arise since potential energy surface becomes flattened because many different positions of the water molecules are nearly isoenergetic.
- Greatly increases computational time.

Quantum Mechanical Solvation Effects

Self Consistent Reaction Field Polarized Continuum Models



The electrostatic potential is calculated at the van der Waals surface of each solute and solvent molecule.



$$\Delta G_{sol} = \Delta G_{elec} + \Delta G_{cav} + \Delta G_{vdw}$$
$$\Delta G_{cav} + \Delta G_{vdw} = \gamma A + b$$

Example Applications

Competitive adsorption of water and TCE in mineral micropores

• High energy adsorption in mineral micropores leads to slow desorption of contaminants from sediments.



• Slow diffusion through molecular-sized pores leads to slow desorption rates.



How does the strength of TCE-micropore interactions depend on the pore size and surface properties?

Experimental Approach

TCE Adsorption Isotherms on Zeolite H-Y



Measure the temperature dependence of adsorption isotherms to determine the isosteric heat of adsorption for a range of solids.

$$Q_{iso} = \Delta h_i^{ads} = -RT^2 \left[\frac{\partial \ln f_i}{\partial T}\right]_{n_i, n_j}$$

Modeling Approach





- GCMC simulations: water fugacity= f_w^{pure} and TCE fugacity=0.01 f_{TCE}^{pure} .
- 30 x 30Å x 6-20 Å cell was generated from a silica crystal.
- Periodic boundary conditions eliminate edge effects.
- Silica framework was held rigid.
- Simulations run with different silica surface charges to simulate a range in hydrophilicity. Surface charge controlled by removing H from Si-OH.
- Simulations performed with COMPASS forcefield (Accelrys, Inc.)

Validating the Force Field Parameters for TCE and Water

Constant NPT MD simulations used to calculate the enthalpies of condensation and pure phase liquid densities for TCE and water.

	∆H°(kJ/mole)		Liquid Density (g/ml)	
Compound				
	Calculated	Actual	Calculated	Actual
Water	-44	-45	1.01	0.997
TCE	-39	-34	1.51	1.46

Force field parameters give reasonable results for TCE and water.

NVT Simulation of Water Adsorption

Characterizing the pore hydrophilicity.



μ VT Simulation of Water Adsorption

Adsorption density plot for water adsorption in a silica pore with a 8 Å pore width.



$$Q_{i}^{iso} = \frac{\langle UN_{i} \rangle - \langle U \rangle \langle N_{i} \rangle}{\langle N_{i}^{2} \rangle - \langle N_{i} \rangle^{2}} - kT$$

 $\langle U \rangle$ = system internal energy averaged over a large number of system configurations

 $\langle N_i \rangle$ = number of molecules of type i over averaged over a large number of system configurations

k = Boltzmann's constant = $R/6.022 \times 10^{23}$



μ VT Simulation of TCE & Water Adsorption



Volume fraction occupied by TCE



15

d.

20



TCE at a liquid phase mole fraction of 2 x 10^{-6} (1% x_{sat}) can displace water from both mildly hydrophobic and very hydrophilic micropores.

Isosteric Heats of Adsorption



- Van't Hoff method for evaluating adsorption energies underestimates the true adsorption energies due to ignoring the effects of water adsorption.
- Contaminant-sediment interactions may be stronger than previously believed.

Background Reductive Dechlorination

- Zerovalent iron filings are used for reductive dechlorination of solvents in water.
- Complete dechlorination of chloroethenes but sequential dechlorination of chloroethanes.
- Reaction rates for chloroethenes are much faster than those for analogous chloroethanes.
- Evidence of an inner-sphere electron transfer mechanism associated with chloroethenes.



Do trichloroethylene (TCE) and tetrachloroethylene (PCE) chemically adsorbed on iron surfaces?

Investigate surface complex formation of TCE and PCE with iron slabs using periodic density functional theory

Starting geometries



• iron modeled as a periodic 5 x 5 x 5 layer Fe[100] slab

Di-sigma Complex Formation





- Activation of 2 C-Cl bonds to >3 Å.
- C bond angles and C-C bond lengths indicates sp² hybridization.



Sigma Bonded TCE



- All 3 C-Cl bonds broken to ~ 4.3 Å.
- C-Fe bond length similar to di-sigma complex.
- Reduction in C=C bond length by 0.14 Å, consistent with conversion of double to triple bond.





PCE Pi-Complex Formation





- All 4 C-Cl bonds broken to ~ 4.3 Å.
- Fe-Cl bond distances close to those in FeCl₂ crystal.



Bindings Energies

Binding Energy							
Compound	(kJ/mol)	Structure					
PCE	-915.67	pi					
TCE	-805.21	sigma					
PCE	-609.56	di-sigma					
TCE	-568.93	di-sigma					

- Lower energies of pi and sigma complexes are consistent with complete dechlorination that is most often observed.
- Formation of di-sigma complexes and the breaking of 2 C-Cl bonds is consistent with observations of trace amounts of chloroacetylene and dichloroacetylene^{*}.

*Arnold & Roberts, Environ. Sci. Technol. 2000, 34, 1794.

Chemical Adsorption of As(III) on Ferric hydroxides

Motivation

- Iron oxides are being used as adsorbents for removing arsenic from potable water.
- There is very little experimental information on As(III) complexes with ferric hydroxides.
- Conflicting experimental results on As(III) interactions with ferric hydroxides.
- Unusual hysteretic effects are difficult to resolve experimentally.

Questions to Address

- Why do some studies show oxidation of As(III) upon adsorption to ferric hydroxides and other studies do not?
- How do impurities in the ferric hydroxides affect the binding strength?
- What products form when H₃AsO₃ and HAsO₂ bind to ferric hydroxides?

Chemical Adsorption of As(III) on Ferric hydroxides



Octahedral clusters used to simulate ferric hydroxides.





Binding energies calculated for edge and corner sites and mono- and bi-dentate complexes.



Goethite (α -FeOOH)



Determined the effects of hydration and protonation on binding energies.

Chemical Adsorption of As(III) on Ferric hydroxides



Effects of pH on binding energies and As(III) oxidation investigated by varying the surface charge.



Effects of porosity and As(III) loading on binding energies investigated using molecular mechanics simulations.

Energetically Favorable Binding Modes



Complex A



Complex B





Complex C

Complex	in vacuo	in solution		
	$\Delta \mathbf{E}$	$\Delta \mathbf{E}$	ΔH° (298 K)	ΔG° (298 K)
Α	-814	-629	-629	-667
В	-1208	-349	-337	-366
С	-9 19	-727	-712	-699

Complex A : $HAsO_2^0 + Fe_2(OH)_2(H_2O)_8^{4+} \iff AsO_2 - Fe_2(OH)_2(H_2O)_6^{3+} + H_2O + H_3O^{+}$

Complex B: $H_3AsO_3^0 + Fe_2(OH)_2(H_2O)_8^{4+} \Leftrightarrow HAsO_3 - Fe_2(OH)_2(H_2O)_6^{2+} + 2H_3O^{+}$

Complex C: $H_3AsO_3^0 + Fe_2(OH)_2(H_2O)_8^{4+} \Leftrightarrow H_2AsO_3 - Fe_2(OH)_2(H_2O)_7^{3+} + H_3O^{+}$

New Approaches May be Needed



Use QM/MM methods to better account for hydration.



Use more constrained adsorbents such as lepidocrocite.

Comparison with Experiments



Dixit, S.; Hering, J. G. Environ. Sci. Technol. 2003, 37, 4182-4189.

 $\begin{array}{ll} +4 & H_{3}AsO_{3} + Fe_{2}(OH)_{2}(H_{2}O)_{8}^{4+} \Leftrightarrow HAsO_{3}Fe_{2}(OH)_{2}(H2O)_{6}^{2+} + 2H_{3}O^{+} \\ +2 & H_{3}AsO_{3} + Fe_{2}(OH)_{4}(H_{2}O)_{6}^{2+} \Leftrightarrow HAsO_{3}Fe_{2}(OH)_{2}(H2O)_{6}^{2+} + 2H_{2}O^{+} \\ 0 & H_{3}AsO_{3} + Fe_{2}(OH)_{6}(H_{2}O)_{4}^{0} \Leftrightarrow HAsO_{3}Fe_{2}(OH)_{4}(H2O)_{4}^{0} + 2H_{2}O^{+} \end{array}$

Joys of Modeling

- Can obtain information that is not experimentally accessible.
- Vary properties of the system as desired.
- Inexpensive compared to experimentation. Commercial software packages for PCs typically cost from \$500 to \$10,000.
- With commercial software, don't need a PhD in computational chemistry (any fool can be a modeler).
- Fast compared to experimentation (high research productivity).
- High repeatability between runs (improves sleep habits of PI).
- Good way to make friends (collaborators always welcome).

Woes of Modeling

- Only certain aspects of any system can be modeled at one time.
- The relationship between the model and reality can be difficult to establish.
- A thorough understanding of the system is needed to select which aspects of the system to model.
- Modeling is best used to explain previously collected data.
- Convergence problems for large systems and bad initial guesses.
- Incorporation of solvation effects can be difficult. Convergence problems with too many explicit water molecules.

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