

Integral Equation Study of the Hydrophobic Interaction between Graphene Plates

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Abstract: The hydrophobic association of two parallel graphene sheets is studied using the 3D-RISM HNC integral equations with several theoretical methods for the solvent distribution functions. The potential of mean force is calculated to study the effects of the aqueous solvent models and methods on the plates as a function of distance. The results of several integral equations (IE) are compared to MD simulations for the same model. The 3D-IEs are able to qualitatively reproduce the nature of the solvent effects on the potential of mean force but not quantitatively. The local minima in the potential of mean force occur at distances allowing well defined layers of solvent between the plates but are not coincident with those found in simulation of the same potential regardless of the theoretical methods tested here. The dewetting or drying transition between the plates is generally incorrectly dependent on steric effects with these methods even for very hydrophobic systems without solute–solvent attractions, in contradiction with simulation.

I. Introduction

Since the conception and implementation of the three-dimensional - reference interaction site model (3D-RISM) integral equation (IE) method, the theory has been applied to large anisotropic models to obtain structural details not possible using the analogous one-dimensional theory.^{1–3} The solutions to the IEs are used to predict the thermodynamic properties of the solution as well as the spatial distribution of the solution species.^{4,5} Studies of the conformational stability of molecular solutes based on thermodynamic considerations have shown the structural stability to depend on a delicate balance between competing forces among the solution species.⁶ This is especially true for biomolecules where the folding mechanism and stability are determined by thermodynamic interactions within and between the solute and solvent which help to direct the folding of the native structure along a converging path.^{7–10}

The phenomenological forces responsible for stabilizing the folded macromolecule and its association with other biomolecules usually include electrostatic interactions, covalent cross-linking, and hydrophobic interactions.⁶ Many models of the stability for macromolecules have implicated effective hydrophobic interactions in the interior of the folded structure consisting mainly of constituents of varying hydrophobicity.^{7,8} The effective interactions between individual hydrophobic moieties are relatively small, but the collective effects of these interactions and the consequences of desolvation in large regions in the interior or at an interface provide the stability seen in biomolecules containing a hydrophobic core. The degree to which the 3D-IEs predict these forces for different solution mixtures is somewhat ambiguous, and the study of more basic models is warranted to ascertain the qualitative predictions of the theory. These types of studies are needed for two reasons: to gain an understanding of the predictive quality of the present theory and to provide a path for relevant advancements to the theory.

A great deal of recent work has been applied to understanding the nature of hydrophobic effects at a fundamental

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level.^{11–16} The nature of the shorter ranged attractions, less than a few molecular diameters, of hydrophobic species has been described with multiple mechanisms over the years.^{17–23} Authors have attributed the nature of the hydrophobic interactions to arise from either a gain of entropy due to release of solvent structured at the hydrophobic interface,²² disruption of the H-bonding network of the aqueous solvent,²¹ the entropic cost of forming the cavity with respect to the size of the solvent,^{19,20,23} or combinations of these. A more detailed discussion of these theories is given in the recent review by Ball.¹⁷ Biochemically, hydrophobicity refers to a variable with a significant range and is not a property which is binary, i.e. on or off as defined by, for instance, a single contact angle criterion.

In addition to these shorter ranged interactions, a much longer less well understood interaction has been observed at length scales up to a few thousand molecular diameters.^{24,25} Current explanations for such forces include the existence of microscopic bubbles causing a collapse of the intersolute region^{24,26} and an electrostatic mechanism due to an induced dipole–dipole state from the slow reorientation of the polar solvent at the liquid–solid interface.^{27,28} There are still differences in how different fields of science define hydrophobicity and the hydrophobic effect.²⁹ Here we concentrate on nanoscopic plates larger than atoms but still on the nanoscopic scale which have surface areas in the range of protein–protein contacts.

Simulations and theory have generated literature on a variety of solutes with varying hydrophobicity in polar solvents discussing the factors effecting the magnitude, the length scales of the interactions, and the time dependent kinetics.^{15,16,30–32} For solutes in close proximity the factors contributing to the crossover from the solvent separated state to the contact states have been extensively studied.^{15,18,33} The length scale of the crossover has been shown to be highly dependent on the solute–solvent attractive interactions.^{15,33} For models with a repulsive or quite weak attractive interaction the intersolute region has been shown to go through a drying phase at larger distances of two or three solvent diameters.^{33,34} The region between the solutes becomes depleted of the polar solvent and ultimately collapses through capillary evaporation. This intersolute collapse following dewetting, whether for hard sphere systems or for systems where dewetting is purely steric, is thermodynamically characterized by a large positive change in entropy due to the solvent being able to recoup some of the translation and rotational entropy due to the region being released to the bulk solvent.

Simulations of models having realistic solute–solvent attractive terms behave qualitatively differently than their purely repulsive models.^{33–35} The state of hydration in the intersolute regions in simulations has been shown to be highly dependent on a number of factors characterized by the solute properties such as shape and interaction strength. For example, water was shown to exist in a stable one-dimensional hydrogen bonded chain in the interior of a carbon nanotube with sufficient attractive solute–solvent interactions but less so with lower attractions.^{36,37} Other studies with small attractive interaction terms have shown a

two-dimensional layer of water to be unstable when confined in a hydrophobic region at close contact.^{15,38,39} A systematic analysis³³ used the calculated potential of mean force as a function of solute–solvent attraction and distance to determine the full range of hydrophobic effects on the association of 2-D sheets of atoms in aqueous solvent. Those authors showed the hydration state corresponding to the PMF at contact depends critically on the potential. They also observed that the stability of the monolayer of water between the plates at a separation consistent with the atomic diameters of the constituents of the plate–water–plate layers is strongly interaction potential strength dependent.

Simulations of solvation effects are considered the most accurate means to study such model systems at the molecular level, but they are computationally intensive. Once the solvent structure is determined via the pair distribution functions (PDF), the thermodynamic properties may be calculated. Many less expensive methods exist for such calculations, which vary in their qualitative and quantitative accuracy, but one of the most promising methods which still accounts for the atomistic nature of the solution species is based on integral equations.⁴ Integral equations based on Ornstein–Zernike (OZ)-like constructs are attractive as an economical method for calculating the approximate pair distribution functions for the constituents of a solution.^{40–42} The distribution functions predicted by the IEs can be used to describe much of the thermodynamics of the solution mixture but remain inherently approximate. Statistical thermodynamics provides convenient analytical expressions for the solution thermodynamics using the site–site distribution functions.^{43,44} Some recent advances in IE theory offer more rigor and accuracy but have yet to be extended to aqueous systems.⁴¹

Extensions of the traditional one-dimensional (1D) radial methods to three-dimensional (3D) grid calculations show promise for the popular interaction site models^{1–3,45} to obtain the solute molecule–solvent site pair distribution functions for a solute molecule at infinite dilution. The 3D-RISM theory, which is an extension of 1D-RISM theory,⁴⁶ requires the solvent–solvent correlations from a one-dimensional, radial theory and are used on a 3D grid to calculate the solute molecule–solvent site distributions. The distributions obtained from 3D-RISM provide an approximate angular distribution description of the solvent structure around solute molecules compared to the simple radial distribution functions (RDF) from 1D methods.

Some recent applications of the 3D-IEs have focused on elucidating the thermodynamic role the solvent plays on the conformational stability of large biomolecules such as proteins.^{9,47,48} These studies considered the origin of the energetic and entropic contributions to the free energy and confirmed how the entropic part is a significant driving force in protein folding. An insightful discussion by those authors demonstrated the qualitative and quantitative nature of the results based on modern day solvation mechanisms. Other applications include the prediction of the spatial positioning of solvent species throughout possible cavities in the interior of large biomolecules, where cavities are not always accessible during MD simulation time scales, and the results may

be expected to show ensemble dependence.^{49–51} In principle this is one of the advantages such methods have over molecular dynamics simulations outside of the grand canonical ensemble; they are able to sample all of configuration space, whereas NVE and NPT MD simulations can suffer from sampling errors due to time constraints.

The IEs and their closures are only approximations, and little has been discussed in the literature on the deficiencies of the 3D-IE theory as compared to volumes written on 1D theories compared with other methods (simulation or experiment). Comparison with the same Hamiltonian or potential model used in simulation for a nonpathological system eliminates the ambiguity of comparison with experiment. Once the accuracy of the theory for a given model is known, comparison with experiment can establish the precision of the model with confidence.

Here we address the question as to how accurately 3D-RISM equations describe a particular well studied system of interest, namely the 2D plates considered by this group earlier via simulation.^{11,30,33,35} Some of the primary focus is on the dewetting transition seen in the simulation results and how the 3D-IEs coupled with the HNC closure will handle this transition. Expectations are limited since the HNC closure, like many other closures, is incapable of predicting coexisting phases.⁵² This is mathematically shown in density functional theory to do with the second order truncation of the expansion in density of the thermodynamic potential and the reality that only a single minimum with respect to density occurs in a quadratic function.⁵² The solvent behavior as described by IEs for small hydrophobic solutes has been studied in numerous papers.^{12,14,53–55} The IE approach provides convenient analytical expressions for the solution thermodynamics using the site–site distribution functions.^{43,44} The IEs in these studies generally provided qualitative descriptions of the solvation of small species. We also wish to quantitatively compare the effect of the solvent–solvent correlations from several IEs on the results for 2D plates. We will quantitatively compare the free energy from IE results for the state of hydration for various models of water near 2D plates of varying hydrophobicity with the exhaustive simulation data.

The IE approach is many orders of magnitude more computationally efficient than simulations, but questions about the accuracy and precision of the numerical solutions are still outstanding.⁵ Numerical solutions for larger solutes which require a larger spatial grid can be difficult to obtain and can exhaust large amounts of memory. The convergence and memory issues are the dominating factors in considering which numerical method to use. Direct inversion of the iterative subspace (DIIS)^{56–59} methods are known as being well balanced between memory requirements and convergence of the solutions. However the method is not an exact minimization routine. It is only as good as the solutions being used in its iterative subspace. Methods based on a Newton–Raphson type iterative routine are exact in that they will minimize the residual if the starting solution is within the radius of convergence. The negative aspect of these routines is dealing with a potentially large Jacobian matrix which has to be calculated, stored, and inverted.⁶⁰ In this paper we will

present a method based on an exact iterative routine which does not require the storage of the full Jacobian or the costly matrix inversion.

In section II of this paper an introduction to the integral equations, the thermodynamic equations, and the numerical method used in this study are given. Section III describes the model interactions used in this study and in the corresponding simulation. In section IV the hydration structure and thermodynamics predicted by 3D-IEs is compared with simulation results to show the qualitative nature of the IE methods. Section V contains our conclusion.

II. Theory and Methods

A. Equations. The three-dimensional molecular solute–solvent site distributions were calculated using the 3D-RISM IEs^{1,3,45} for a multicomponent fluid using various approximations for the solvent distributions. For a solute molecule at infinite dilution the equation can be expressed as

$$h^{uv}(\mathbf{r}) = c^{uv}(\mathbf{r}) * (w^{vv}(\mathbf{r}) + \rho h^{vv}(\mathbf{r})) \quad (1)$$

where $h^{uv}(\mathbf{r})$ is the 3D solute–solvent pair total correlation function, $c^{uv}(\mathbf{r})$ is the 3D solute–solvent direct correlation function, ρ is the particle number density of the solvent, $w^{vv}(\mathbf{r})$ is the intramolecular correlation function of the solvent, and $h^{vv}(\mathbf{r})$ is the solvent site–site distribution functions which we obtained from and will compare among various theories including RISM,⁴⁶ DRISM,^{61,62} and closure optimized PISM variants.⁴⁰ The matrix convolution integral in real space is represented by an *. The hypernetted chain (HNC) equation was used in the 3D calculations as a closure. The HNC equation, although not exact, is computationally convenient and gives more reliable results for charged and polar species over a broader range of solution species than some other closure relations.⁶³ The HNC equation is defined for each component as

$$c_{ij}(r) = \exp(-\beta u_{ij}(r) + t_{ij}(r)) - t_{ij}(r) - 1 \quad (2)$$

where $t(r)$ is the indirect correlation function, $t(r) = h(r) - c(r)$, $u(r)$ is the pair potential between the molecular solute and solvent site, and β is the inverse product of the solution temperature and Boltzmann's constant. We use this in the traditionally renormalized form.

The solvent site–site distributions in eq 1, h^{vv} , constitute a completely separate problem and are solved independently of the 3D calculations. The RISM equation for the solvent–solvent distributions is

$$h^{vv} = w^{vv} * c^{vv} * w^{vv} + w^{vv} * c^{vv} * \rho h^{vv} \quad (3)$$

where the correlation functions have the same meaning as in eq 1. Solutions based on XRISM theory are obtained from using eq 3 closed with the HNC (eq 2) in renormalized form.^{64–66} For solution mixtures containing charged sites the correct screening due to the solvent is more accurately represented using dielectrically consistent RISM theory (DRISM).^{61,67} DRISM theory is similar to RISM theory, but an ad hoc bridge term correcting the long-range behavior of the solvent is included in the closure. Methods based on RISM are not formally exact in that the graphical expansion

of the RISM equations includes incorrect terms and excludes other correct ones.⁶³ The proper interaction site method (PISM) is a formal improvement over RISM and more accurately represents the graphical expression but generally provides less accurate results.^{68,69}

The hydration free energy (HFE) was calculated using the Morita-Hiroike HNC formula^{43,44} on the 3D grid as

$$\Delta\mu = \rho k_B T \sum_a \int \left[\frac{1}{2} h_a(\mathbf{r})^2 - c_a(\mathbf{r}) - \frac{1}{2} h_a(\mathbf{r}) c_a(\mathbf{r}) \right] d\mathbf{r} \quad (4)$$

The chemical potential can be decomposed into its partial molar enthalpic, $\Delta\epsilon$, and entropic, Δs , parts using

$$\Delta\mu = \Delta\epsilon^{uv} + \Delta\epsilon^{vv} - T\Delta s \quad (5)$$

where the entropic part can be calculated by taking the temperature derivative of the chemical potential (eq 6).⁵³ The entropy is calculated using a finite difference method with a temperature step of 5 K.

$$-\frac{\partial \Delta\mu}{\partial T} = \Delta s \quad (6)$$

B. Numerical Methods. The correlation functions ultimately must be computed on a grid of sufficient resolution to capture the details of the finest physical length scales defined by the problem. With atomic diameters and bond lengths defined by angstroms, the grid size needs to be a fraction of an angstrom, and the extent of the grid must be well beyond the correlation lengths induced by the solute in the solvent. Given an appropriate grid, coarse solutions to eqs 1 and 2 could be calculated using a Newton–Raphson type numerical scheme on a smaller or reduced grid followed by Picard iterations on the full grid to refine the solution to the desired accuracy. Obtaining solutions in this manner however is usually not feasible due to memory constraints and the time required for large matrix inversions.⁵⁶ We have dealt with both of these issues to make these types of algorithms more tractable. The method presented here is an extension of the multigrid idea developed by Gillian⁷⁰ where a coarse solution is represented by a subset of points on a finer grid. The projection of the fine grid onto the coarse grid represents a significant reduction in the number of points used to describe the major features of the fluid. The coarse grid does not describe all the fine details in the solutions but can aid in the reduction of the number of Picard iterations needed on the fine grid to refine the solutions. On returning to the coarse grid values are calculated by averaging the functions at points in the fine grid.⁶⁰

To reduce the computationally exhaustive task of inverting large matrices we used an iterative routine to approximate solutions for the linear equations. The algorithm we implemented for the solution to the linear equation, $Jx=F$, where J is the Jacobian matrix, x is our desired solution, and F is the residual error, is the GMRes method.⁷¹ Methods of this type, based on solving Newton's equations

$$\begin{aligned} 0 &= F(x) + \frac{\partial F(x)}{\partial x} \Delta x \\ \Delta x &= \left(-\frac{\partial F(x)}{\partial x} \right)^{-1} \\ F(x) &= (-J)^{-1} F(x) \end{aligned}$$

are more stable than Picard or relaxation iterations and when implemented correctly accelerate the convergence of the solution.⁷² The Generalized Minimal Residual (GMRes) routine iteratively forms an approximate solution to x using $x_0 + z$, where x_0 is an initial starting guess and z is the incremental solution in Krylov space. The approximating solution converges to the exact solution x as the Krylov subspace approaches the rank of the transformation matrix A representing the full set of equations. At each iteration a new approximation to x is formed by minimizing the residual norm in the Krylov subspace, $K_k = \text{span}(v_1, Av_1, \dots, A^{k-1}v_1)$.

$$\min_{z \in K_k} \|b - A[x_0 + z]\| = \min_{z \in K_k} \|r_0 - Az\|$$

The k^{th} iterative solution to x using z is of the form $x_k = x_0 + V_k y_k$, where $z = V_k y_k$. The matrix V_k consists of k columns, which are the l_2 -orthonormal basis (v_1, v_2, \dots, v_k) of the Krylov subspace, and y_k consists of the coefficients for the basis set. For each iteration the size of the Krylov subspace is increased by one, and the basis set coefficients are calculated to minimize the residual norm. This is continued until the residual norm meets some predetermined convergence criterion. The number of operations needed for each iteration is $O(n^2)$, where n is the number of grid points. The number of iterations needed to converge to a solution depends on the initial basis set and the predetermined convergence criterion. An advantage to this routine is that the Jacobian matrix does not have to be stored, if one can efficiently generate the elements of the Jacobian (Appendix). However, the matrix consisting of the l_2 -orthonormal basis should be stored and increases linearly in size with the number of iterations. A common variant of the GMRes routine is one that allows restarting. Once the number of basis vectors gets too large, the GMRes routine is restarted with the last approximate solution to x as the new initial starting solution, x_0 .

C. Numerical Procedure. After situating the solute coordinates in the approximate center of the 3-D grid, the steps used in our implementation are as follows:

1. Calculate the potential energy interactions between solute sites and solvent grid points.
2. Guess initial correlation functions.
3. If necessary perform a few initial Picard iterations.
4. Decompose correlation functions into their basis set coefficients and their fine displacements.
5. Calculate the residual vector.
6. Calculate the Jacobian elements on the coarse grid.
7. Perform GMRes routine until predetermined convergence criterion is met.
8. Calculate new basis coefficients.
9. Check residual vector to determine if another iteration on the coarse grid is needed (if so return to step 4) or to continue to fine grid (next step).
10. Calculate fine grid values from the basis set representation and the fine displacements.
11. Use Picard iterations to refine direct correlation, $c(\mathbf{r})$, solution in the fine grid.
12. Calculate fine grid residual vector $f^{j+1}(\mathbf{r})$. If this value meets our convergence criteria, the solution to the desired error has been obtained.

Numerical methods similar to the one presented here, which are based on the GMRes method, have previously been successfully applied to 1D site–site IEs.^{72,73} A final thought on these types of numerical routines is that the responsibility of calculating the solution to the desired criterion is up to the fine grid iterations. For solutions to nonstiff problems a coarser grid can be used to decrease the amount of time needed per iteration, and for stiff problems where convergence issues are the main concern more points should be used in the coarse grid.

III. Model

In this study the integral equations described above are used to study the effective interactions between two planar hydrophobic solutes in a polar solvent of H₂O. The results are then compared with simulation results for the same model.³³ The solute and solvent parameters are identical to the simulation study. The pair potential between the sites is modeled with the Lennard-Jones (LJ) plus Coulomb potential. The potential for the solute–solvent interactions in this study is calculated using

$$u_{ab}(r) = 4\epsilon_{ab} \left[\left(\frac{\sigma_{ab}}{r} \right)^{12} - \left(\frac{\sigma_{ab}}{r} \right)^6 \right] + \frac{q_a q_b}{r} \quad (7)$$

The subscripts specify the species of the solvent defined on a grid for the 3D calculations. The symbols σ , ϵ , and q are the usual LJ plus Coulomb parameters: the diameter, the well depth, and the charge on the atomic sites, respectively. The H₂O molecules are represented by the SPC/E model.⁷⁴ The LJ parameters for this SPC/E H₂O model have the values of $\sigma_{\text{OO}} = 3.1655$ Å, $\epsilon_{\text{OO}} = 0.1554$ kcal mol^{−1}, and $q_{\text{O}} = -0.8476$ for the sites characterizing the oxygen atom and $\sigma_{\text{HH}} = 0.4$ Å, $\epsilon_{\text{HH}} = 0.02$ kcal mol^{−1}, and $q_{\text{H}} = 0.4238$ for the sites characterizing the hydrogen atoms. The attractive part of the Lennard-Jones potential is excluded for interactions involving the solvent hydrogen atoms as is consistent with previous usage. The two molecular plates are represented by flat graphene sheets of carbon atoms, the same used by Choudhury and Pettitt.³³ The sheets consisted of 60 carbon atom sites each and measured $\sim 11 \times 12$ Å. The carbon atoms have been placed in a rigid hexagonal graphite-like pattern with distances of 1.4 Å between nearest neighbors. The LJ parameters for the carbon atoms are $\sigma_{\text{CC}} = 3.4$ Å and $\epsilon_{\text{CC}} = 0.3598$ kJ mol^{−1}, and the sites are uncharged. These parameters represent sp² carbon atoms from the AMBER 96 force field from the hydrophobic aromatic peptide residues. The Lorentz–Berthelot combination rules, $\sigma_{\text{OC}} = (\sigma_{\text{OO}} + \sigma_{\text{CC}})/2$ and $\epsilon_{\text{OC}} = (\epsilon_{\text{OO}} \epsilon_{\text{CC}})^{1/2}$, are used to calculate the LJ parameters for site–site interactions between different species. For some calculations the solute–solvent attractions are turned off by using this same potential with a Week–Chandler–Anderson decomposition.⁵ The number density of the bulk H₂O was 0.03334 Å^{−3}, and the temperature is 298.15 K.

The site–site radial distributions for the H₂O solvent, h^{vv} , are calculated using XRISM, DRISM, and optimized PISM theories.^{40,61,66,67} The results between XRISM and DRISM are quantitatively indistinguishable for the charge free solutes used here (see Figure 1), so only the distributions from

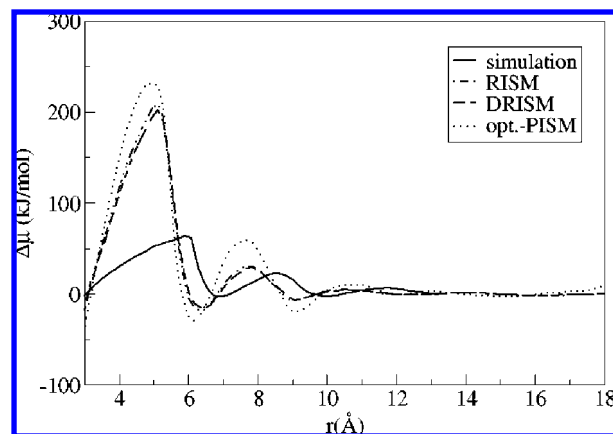


Figure 1. Solvent contributions (HFE) to the PMF for different h^{vv} H₂O theories.

DRISM theory are shown in the results after Figure 1. The radial solvent–solvent distributions are calculated on a fine grid of 16,364 points over a range of 40 Å ($\Delta r = 0.00244$ Å) to minimize errors when transferred to the 3D grids. To limit the errors in the authenticity of the solution and the thermodynamic variables calculated from 3D distributions the calculations are routinely converged to a relative residual error of 10^{-12} . The 3D calculations are performed in a cubic box measuring 40 Å to a side. The fine grid is defined by a linear grid consisting of 128^3 points, and the coarse grid consisted of 19^3 points. We note that some poor initial guesses for thermodynamic values not in a stable part of the phase diagram result in solutions that initially appear to start to converge down to a relative tolerance of 10^{-3} or 10^{-4} . These solutions then diverge on further attempts to refine them and so are rejected in our work as unphysical unless stable solutions at our most stringent criteria are met.

IV. Results and Discussion

The 3D distribution functions for the parallel graphene plates in solvent H₂O were calculated for intersolute distances ranging from 1.0–18.4 Å in increments of 0.1 Å. The HNC hydration free energies (HFE) from the 3D-RISM results shown in Figure 1 were calculated using the chemical potential formula⁴⁴ from eq 4. The value for the infinitely separated plates was subtracted from each curve for each solvent theory. The sensitivity of RISM-like absolute free energies or chemical potentials is well-known.

The three sets of data in Figure 1 correspond to the different solvent–solvent distributions obtained from XRISM, DRISM, and optimized PISM theory. All of the IE results show qualitative features but are quantitatively disappointing in both the magnitude and phase of the oscillations. As mentioned above, the RISM and DRISM results are similar since graphite has no atomic site charges, so only the DRISM results are displayed throughout the remainder of the analysis. Optimized PISM theory is a recent method for calculating the distribution functions which compares to simulation better than DRISM for pure solvents.⁴⁰ Optimized PISM theory has distributions which are more consistent with the simulation data for the solvent–solvent correlations. However, in this usage it appears to over-represent the free energy

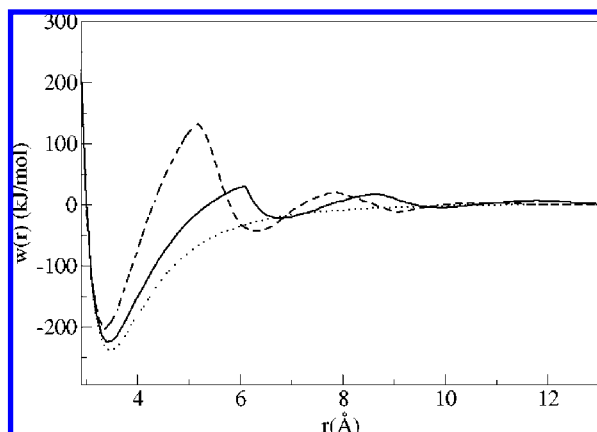


Figure 2. Potential of mean force for DRISM (dashed line) and MD (solid line) results plotted with the direct potential (dotted line) interaction between the plates.

contributions or features from the solvent when compared to the simulation and DRISM results. It should be mentioned that no separate solute–solvent closure optimization was done here as would be more consistent with that method.⁴⁰

The HFE contains the contributions to the PMF from direct solvent interactions. The PMF obtained by the addition of the HFE and the direct solute–solvent potential is shown in Figure 2 for the DRISM solvent.

The minimum in the potential of mean force for the solutes in the solute contact configuration (no intervening water layers) predicted by the 3D-RISM theory is 3.4 Å. This distance is ~ 0.1 Å less than the distance at which the minimum occurred in the PMF obtained from the MD simulation study.³³ Both distances are slightly less than the minimum shown in the direct potential. An explanation for these slightly shorter distances is due to the slope of the cavity potential as the plates are forced together which has been seen in RISM style calculations before.⁷⁵ The generally incorrect placement of the HFE to the direct potential for RISM-like theories has been noted previously and has a variety of consequences for the resulting PMF and properties.⁷⁵

The PMF shown in Figure 2 for the DRISM solvent has three significant solvent stabilized minima occurring at distances of 6.5, 9.3, and 12.5 Å. These solvent stabilized minima occur at distances allowing well defined solvent layers between the plates as shown in Figure 3 and expected from simulation. The first, second, and third solvent separated minima occur at configurations allowing one, two, and three intervening water layers between the plates, respectively.

The same characteristics are seen in the results from the simulation study.³³ However, the simulation results predicted these minima at slightly larger distances consistent with the phase shift in the PMF noted earlier. The DRISM and MD results³³ for the solute–solvent oxygen site distributions perpendicular to the plate surfaces for the three minima corresponding to the distances at which the minima occur in the simulations are shown in Figure 4.

The distributions shown in Figure 4 from DRISM are structurally similar to the distributions obtained from MD simulations with some noted differences. For the intersolute

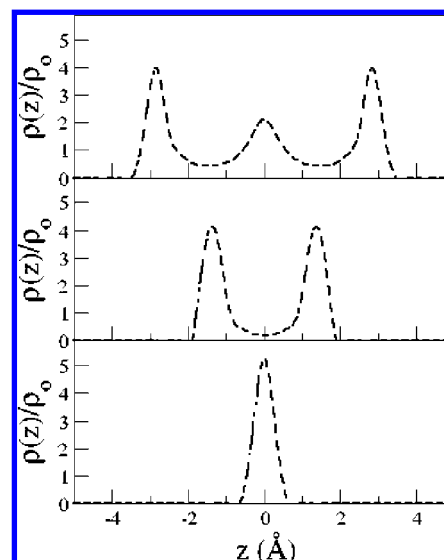


Figure 3. Solute–solvent (oxygen) distributions at the solvent stabilized solute configurations for DRISM at plate separations of 6.5 (bottom), 9.3 (middle), and 12.5 (top).

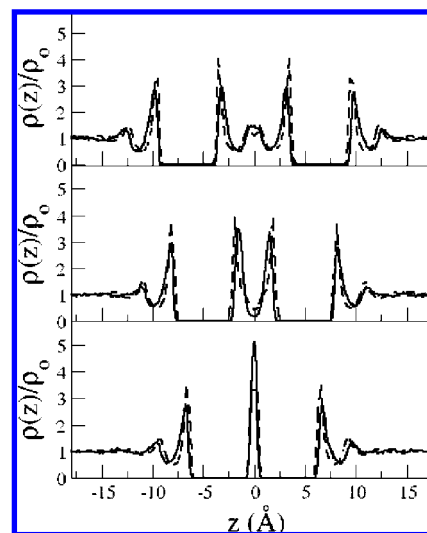


Figure 4. Oxygen site distributions for simulation and DRISM at distances of 6.8 (bottom), 9.8 (middle), and 13 (top), corresponding to the solvent stabilized minima observed in the PMF from simulation.

distances displayed in Figure 4 all theories predicted well defined peaks for the solvent in contact with the outer surface of the plates. Comparisons with simulation are most favorably made when considering the predicted PMF minima; however, the minima are not found at identical distances.

The solvent distributions at the intersolute distance corresponding to the first solvent separated minimum are characterized by a large single peak between the plates. This represents a well ordered monolayer between the plates. However, this minimum in the PMF for the MD simulation occurs at 6.8 Å which is larger than the DRISM results by about 0.3 Å. The distributions for the second solvent separated minimum are also qualitatively similar between the two methods and represent an ordered bilayer of water between the plates. The distance between the peaks in the solvent distributions between the plates is 2.9 Å, which

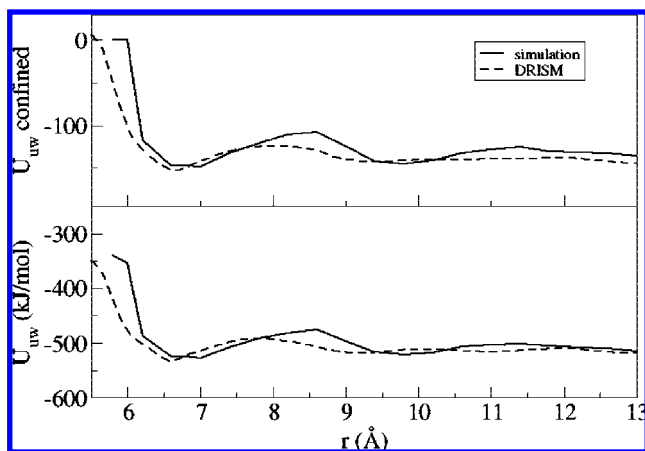


Figure 5. The solute–solvent total potential energy (bottom) and the potential energy between the plates and solvent in the intersolute region (top) in kJ mol^{-1} for IEs and simulation.

roughly corresponds to the peak for O–O distributions of water. However, the interplate distance of this feature in the DRISM calculations is about 0.5 \AA less than the plate distance from the MD simulation.

There is a third solvent stabilized configuration corresponding to three layers of water between the plates which is also replicated by the IE method. The inconsistency in the locations of the solvent separated minima between IEs and simulation are more pronounced and appear to be due to differences in the effective widths of the solvent structure layers. The structure of the solvent layers for the configurations corresponding to the free energy barriers—the regions between the solvent separated minima—in the PMF are characterized by the expected zigzag layering of water between the plates, which is also seen in the MD simulations.⁷⁶ The consequence of this layering is a loss of solute–solvent energetic interactions at each surface facing the interior region.

The most prominent characteristic of the PMF from the IEs is the large free energy barrier between the contact state and the first solvent separated state. The IE approach estimates the magnitude of this free energy barrier to be significantly larger than the barrier seen in simulations. The other obvious feature observed in the PMF for both the theories and simulation is a cusp that occurs at the start of the steric drying distance. The solute–solvent interaction energy helps explain the changes occurring in the free energy (Figure 2).

The qualitative energetic features of the solute–solvent interactions in the interior region are reproduced by the 3D-IE theory when compared to MD simulations. As the interplate distance decreases from the first solvent separated state the rise in the PMF is mainly attributed to the loss of solute–solvent interactions in the interior region. The cusp in the PMF for MD simulations is sharper because of the well defined, steric-induced loss of these favorable energetic interactions as the monolayer of water is effectively excluded from the interior region. This feature can be seen in Figures 2 and 5 at $\sim 6.0 \text{ \AA}$. The cusp is not as sharp in the PMF from the IE theory because the IEs allow some probability

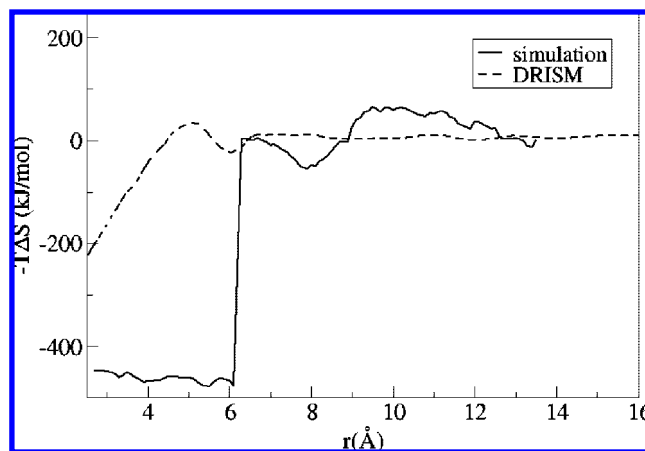


Figure 6. Entropic contribution to the HFE ($-T\Delta S$).

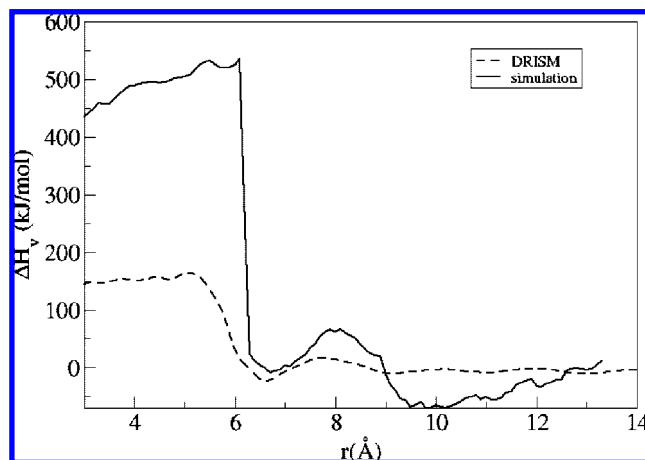


Figure 7. Enthalpic contributions to the HFE, $H^v = \Delta\epsilon^{uv} + \Delta\epsilon^{vv}$.

of water molecules to remain in sterically forbidden positions located above the center of the hexagonal solute atom configuration (see below).

To better understand the factors contributing to the HFE and to investigate the reason for the large free energy barrier in the PMF from IEs, the HFE was separated into its enthalpic and entropic contributions. The partial molar entropy and enthalpy are shown in Figures 6 and 7.

The nature of the larger free energy barrier is obvious when comparing the factors contributing to the PMF from simulation and IEs. Both methods predict a large positive change in solute–solvent interactions as the solvent is forced from the intersolute region upon decreasing the plate distance. However, in the simulation result there is a substantial favorable response in the entropy of the solvent which counteracts the energy lost in the solute–solvent interactions. This rapid change in entropy is not seen in the IE results; there is nonetheless a gradual gain in entropy as the intersolute distance decreases, which is most likely attributed to the gain in translational and rotation entropy as the excluded volume regions overlap or shrink.

As stated earlier, one of the more difficult features of hydrophobic interactions is the length scale at which drying between the plates occurs. In the MD simulations with full solute–solvent interactions³³ the smallest interplate distance

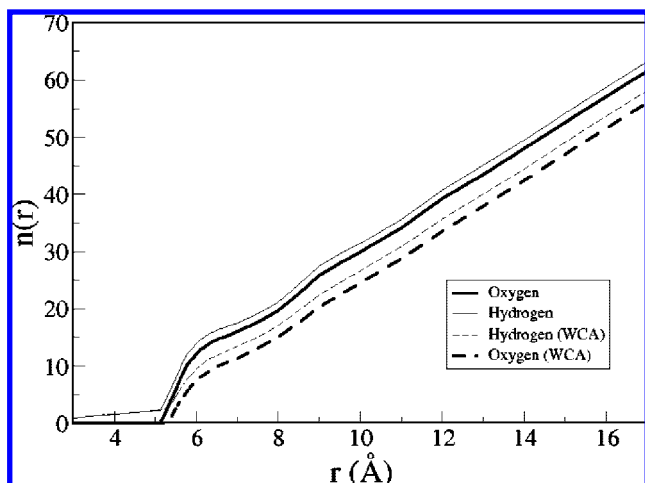


Figure 8. Water oxygen and the sum of hydrogen solvent sites between the plates from DRISM for the model presented and a model represented with the WCA interaction potential.

at which water was allowed between the plates is 6.2 Å; any shorter a distance and all the water molecules were excluded due to steric effects. The sharp change in the number of water molecules between the plates was not duplicated by the 3D-IEs with any solvent theory used here. An observation of the distribution functions at these shorter distances shows that there is some nontrivial probability for the water molecules to be located in the sterically forbidden region. Although not explicitly observed by the number of water molecules between the plates, there is evidence that the IEs have a rough qualitative sense of the beginning of the steric drying transition as seen in the MD simulations. For the interplate region to be hydrated by water molecules in a hydrogen bonded network forming a two-dimensional sheet ~ 13 water molecules would be needed to cover the area. The distance at which the number of water molecules falls below 13 in the IEs is about 6.1 Å, as shown in Figure 8, which is near the same length of steric drying in the MD simulations. It is however, unphysically gradual, and so the cavity contains significant density when just steric effects would have it be dry.

The number of solvent sites between the plates for a model with no solvent–solute attraction (via the WCA decomposition)⁷⁷ is also shown in Figure 8. Here we expect to demonstrate the drying effects due to the lack of attraction of the plates. In contradiction to the finding from simulations, the lack of attraction with the IEs does not lead to drying starting at over 10 Å as has been seen and verified by simulation.^{33,34} In fact the dramatic effect of solute–solvent attractions seen in simulations is effectively suppressed. The lack of a transitional state for the coexistence of phases has been noted before for integral equation methods.⁵² This brings into question the reliability of IEs in situations where a molecular sized cavity might not have sufficient interactions to stabilize a water molecule e.g. on the hydrophobic interior of a protein.

Figure 8 also shows the inconsistency in the stoichiometry of the water oxygen and hydrogen sites expected from all XRISM-type theories.⁶³ Only half the hydrogen site densities

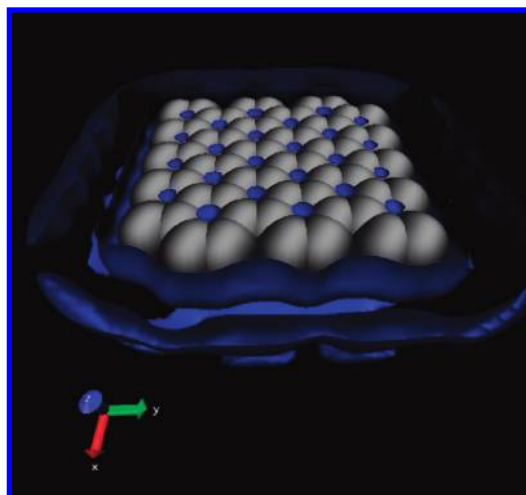


Figure 9. Water hydrogen sites between the plates at a separation of 3.0 Å. Shows the existence of finite probabilities of hydrogen between the plates when no oxygens are present, i.e. in carbon–carbon contact.

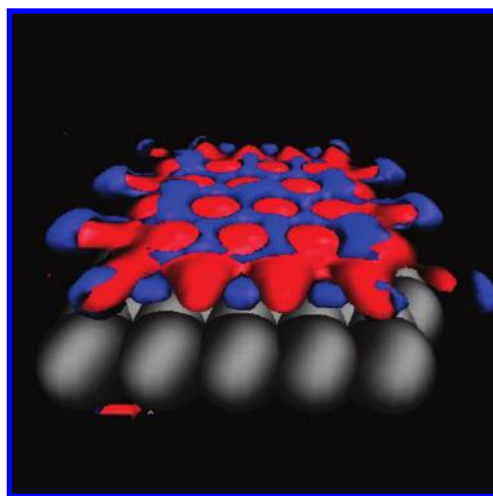


Figure 10. Isosurface of the water oxygen (red) and hydrogen (blue) sites between the plates at a separation of 6.8 Å.

are shown so the ratio should be 1:1, but there are consistently more hydrogen atoms between the plates. This behavior continues past the point where all water oxygen sites are excluded from the intersolute region—the number of hydrogen sites observed should then be zero by stoichiometry. Figure 9 shows the location of the nonzero probability of these rogue hydrogen atoms between the plates. Note that no dissociation is possible for the model, and thus this is a pure artifact of the methods.

This stoichiometric problem has been noted for XRISM class theories for some time.⁷⁸ Another property of the solvent in the interplate region observed in the MD simulation, which is also computable in the 3D-IE theories, is the orientation of water and the static hydrogen bonding occurring between the plates with a single water layer. The isosurface plot shown in Figure 10 shows the oxygen atoms (red) tend to occupy the positions directly above the center of the hexagonal pattern. There is both a steric hole as well as an attractive maximum in the potential in these positions.

The hydrogen atoms (blue) tend to occupy the positions between the oxygen atoms forming a hydrogen bonding network. The overall orientation of the water molecules is for the plane through HOH to be just above or below the surface of the oxygens which is the same orientational preference observed in MD simulations and in good accord.

V. Conclusion

In this study we used the 3D-RISM IEs with XRISM, DRISM, and optimized PISM solvent–solvent distributions to calculate the free energies and distribution functions of the solvent water sites around large graphene plates. The graphene sheets were modeled by Lennard-Jones potentials from Amber describing sp^2 carbon atom sites with and without the WCA decomposition. The SPC/E water model was used for the solvent. The PMF as a function of interplate distance was calculated to determine the ability of the IEs to predict the physical behavior of the solute/solvent combination types presented. The thermodynamic values and structural properties obtained by the IEs were compared to the same from MD simulation results for the same model. The gross structural details of the solvent in contact with the hydrophobic surfaces were consistent between the theories and in reasonable accord with simulation with some notable exceptions. The structural layering of water between the sheets seen in the 3D-IEs was somewhat shifted in position and included stoichiometric deviations compared to the MD simulations.

The free energy barrier for the transition from the first solvent separated configuration to contact was qualitatively replicated by all the IE theories but was off quantitatively. This barrier was significantly larger for the IE results compared to the MD simulation results. The barrier was shown to be dependent on the expected translation-rotation entropy effects for solvent release (increasing intersolute distance) and solute–solvent interaction energies for association (decreasing intersolute distance). The difference in the magnitude of the first free energy barrier from IEs and simulations was shown to arise from the inability of the IEs to accurately account for changes in the solvent entropy and the compensating solvent enthalpy when the solvent was excluded from the interplate region.

The abrupt change in the number of water molecules between the plates at the transition distance for simulation was not seen by the IE theory and may also reflect the standing controversy in the two-state versus glassy debate for such systems. However, the distance at which the number of water molecules between the plates predicted by the IE theory would not completely hydrate the surface was similar to the distance at which the MD simulations saw dramatic dewetting in the intersolute region. The IEs were also shown to predict a reasonable orientation of water molecules in contact with the plate surfaces as compared with the MD simulations.

The calculations at high precision shown in this work were done with a combination of Picard/Newton–Raphson based method to obtain the solutions to the 3D-IEs. We showed how the problem of calculating and storing a large Jacobian

can be replaced with a method to generate the analytically calculated Jacobian elements as they are needed for a smaller system of equations. Although this numerical method will not completely replace DIIS methods, it can help to converge solutions which appear to be unstable with DIIS, especially in the early iterations.

The accuracy of the probability of water in confined spaces and on the interior of macromolecules predicted by this class of theories may suffer from problems found with our more ideal systems. The lack of sharp (first order) hydrophobic drying and inconsistent stoichiometry are problematic. Grand canonical simulations would more accurately quantify this in biomolecule interiors where entrance and escape by diffusion can be problematic.

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Appendix

To calculate the Jacobian elements we use the following equation, where the second equation is shown in Fourier space and in matrix form.

$$F(c(r)) = \exp(-\beta u(r) + t(r)) - t(r) - c(r) - 1 \quad (A1)$$

$$\hat{h}(k) = \hat{c}(k)\hat{w}(k) + \hat{c}(k)p^v\hat{h}(k) \quad (A2)$$

The elements of the Jacobian matrix in this implementation are defined as

$$J_{b,j}^{a,i} = \frac{\partial F^{ua}(c^{ua}(r_i))}{\partial c^{ub}(r'_j)} = -\delta_{(i,j)}\delta_{(a,b)} + \frac{\partial F^{ua}(c^{ua}(r_i))}{\partial t^{ua}(r_i)} \frac{\partial t^{ua}(r_i)}{\partial c^{ub}(r'_j)} \quad (A3)$$

where the superscripts denote the functions between the solute and the solvent sites, and the subscripts on r denote the position in coordinate space.

The partial derivatives can be calculated using the site–site representation of the HNC closure and the OZ equation in real space. The partial derivative can be calculated as

$$\begin{aligned} \frac{\partial F^{ua}(c^{ua}(r_i))}{\partial t^{ua}(r_i)} &= \exp(-\beta u^{ua}(r_i) + t^{ua}(r_i)) - 1 \\ &= h^{ua}(r_i) \end{aligned} \quad (A4)$$

To calculate the second partial derivative in the last term we start with the OZ equation in momentum space, A.2, and expand into the components for all constituents

$$\begin{aligned} \hat{h}^{ua}(k) &= \sum_b \hat{c}^{ub}(k)\hat{w}^{ba}(k) + p^v \sum_b \hat{c}^{ub}(k)\hat{h}^{ba}(k) \\ \hat{h}^{ua}(k) &= \sum_b \hat{c}^{ub}(k)(\hat{w}^{ba}(k) + p^v\hat{h}^{ba}(k)) \end{aligned} \quad (A5)$$

Next we can rearrange and collect the constant terms into χ^{ba} , which is calculated in Fourier space

$$\hat{t}^{ua}(k) = \hat{c}^{ub}(k) p^v \hat{h}^{aa}(k) + \sum_{b \neq a} \hat{c}^{ub}(k) (\hat{w}^{ba}(k) + p^v \hat{h}^{ba}(k)) \quad (\text{A6})$$

$$\hat{t}^{ua}(k) = \hat{c}^{ub}(k) p^v \hat{h}^{aa}(k) + \sum_{b \neq a} \hat{c}^{ub}(k) \hat{\chi}^{ba}(k) \quad (\text{A7})$$

This equation may be transformed back into real space to give

$$t^{ua}(\vec{r}) = \rho^v \int_0^\infty c^{ua}(\vec{r}') * h^{aa}(\vec{r} - \vec{r}') d\vec{r}' + \sum_{b \neq a} \int_0^\infty c^{ub}(\vec{r}') * \chi^{ba}(\vec{r} - \vec{r}') d\vec{r}' \quad (\text{A8})$$

where the integrals run over all space.

The equation on a discretized grid in Cartesian coordinates, letting $r = (x, y, z)$ and $r' = (x', y', z')$ can be recast as

$$t^{ua}(x, y, z) = \rho^v \sum_{x', y', z'} c^{ua}(x', y', z') * h^{aa}(x - x', y - y', z - z') \Delta x' \Delta y' \Delta z' + \sum_{b \neq a} \sum_{x', y', z'} c^{ub}(x', y', z') * \chi^{ba}(x - x', y - y', z - z') \Delta x' \Delta y' \Delta z' \quad (\text{A9})$$

$$\frac{t^{ua}(x, y, z)}{c^{ua}(x', y', z')} = \rho^v h^{aa}(x - x', y - y', z - z') \Delta x' \Delta y' \Delta z' \quad (\text{A10})$$

$$\frac{t^{ua}(x, y, z)}{c^{ub}(x', y', z')} = \chi^{ba}(x - x', y - y', z - z') \Delta x' \Delta y' \Delta z' \quad (\text{A11})$$

for $a = b$, where $i = (x, y, z)$ and $j' = (x', y', z')$

$$J_{a, j'}^{a, i} = \frac{\partial F^{ua}(c^{ua}(i))}{\partial c^{ub}(j')} = -\delta_{(i, j')} + (\exp(-\beta u^{ua}(i)) + t^{ua}(i)) - 1) \rho^v h^{aa}(i - j') \Delta x' \Delta y' \Delta z' \quad (\text{A12})$$

$$= -\delta_{(i, j')} + h^{ua}(i) \rho^v h^{aa}(i - j') \Delta x' \Delta y' \Delta z' \quad (\text{A13})$$

and for $a \neq b$

$$J_{b, j'}^{a, i} = \frac{\partial F^{ua}(c^{ua}(i))}{\partial c^{ub}(j')} = (\exp(-\beta u^{ua}(i)) + t^{ua}(i)) - 1) \chi^{ab}(i - j') \Delta x' \Delta y' \Delta z' \quad (\text{A14})$$

$$= h^{ub}(i) \chi^{ab}(i - j') \Delta x' \Delta y' \Delta z' \quad (\text{A15})$$

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