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Accounting for the Quantum Capacitance of Graphite in Constant Potential Molecular Dynamics Simulations

Kateryna Goloviznina, Johann Fleischhaker, Tobias Binninger, Benjamin Rotenberg, Heigo Ers, Vladislav Ivanistsev, Robert Meissner, Alessandra Serva, and Mathieu Salanne*

Molecular dynamics (MD) simulations at a constant electric potential are an essential tool to study electrochemical processes, providing microscopic information on the structural, thermodynamic, and dynamical properties. Despite the numerous advances in the simulation of electrodes, they fail to accurately represent the electronic structure of materials such as graphite. In this work, a simple parameterization method that allows to tune the metallicity of the electrode based on a quantum chemistry calculation of the density of states (DOS) is introduced. As a first illustration, the interface between graphite electrodes and two different liquid electrolytes, an aqueous solution of NaCl and a pure ionic liquid, at different applied potentials are studied. It is shown that the simulations reproduce qualitatively the experimentally-measured capacitance; in particular, they yield a minimum of capacitance at the point of zero charge (PZC), which is due to the quantum capacitance (QC) contribution. An analysis of the structure of the adsorbed liquids allows to understand why the ionic liquid displays a lower capacitance despite its large ionic concentration. In addition to its relevance for the important class of carbonaceous electrodes, this method can be applied to any electrode materials (e.g. 2D materials, conducting polymers, etc), thus enabling molecular simulation studies of complex electrochemical devices in the future.

1. Introduction

The electrochemical double-layer (EDL) is the region where the electrolyte contacts the electrode in an electrochemical device. Despite its importance in many fields, ranging from energy storage,^[1] electrocatalysis,^[2] sensing,^[3] water desalination, nanogenerators,^[4] etc., much remains to be known about its structure and properties. In situ experiments provide many useful details,^[5] but they are often hindered by their spatial resolution since the processes at play occur in a liquid region of a few nanometers only. From the theoretical point of view, although textbooks still present the Gouy–Chapman–Stern model as a realistic picture, its relevance for all these applications has largely been questioned due to its drastic approximations.^[6,7] Based on these observations, microscopic simulations appear as the ideal tool for advancing our knowledge about the EDL.^[8]

Yet, the application of microscopic simulations to such complicated interfacial

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systems is not straightforward. Researchers are faced with the conundrum of simulating large-scale systems, over relatively long timescales, while keeping the accuracy of the calculation to the highest level. For example, the characteristic adsorption lifetimes of ionic species inside electrified nanoporous electrodes are of several nanoseconds.^[9] Recent developments have made density functional theory (DFT) accessible to calculate the interactions in MD simulation, but the associated computational costs are very large.^[10] This method was thus so far limited to the important case of metal-water interfaces,^[11] with a special focus on the reactivity of water molecules.^[12] In recent years, several works have intended to account for the presence of ions by including them directly within the electrochemical double-layer. This approach yielded accurate estimates of the interfacial capacitance at the interfaces between gold,^[13] copper or graphene^[14] and simple aqueous solutions, but the limited system size prevents one from controlling the concentration of ions. A similar setup was used in the framework of electrocatalysis applications to simulate the effect of various cations on the CO₂ reduction reaction mechanism.^[15,16] The main example of non-aqueous system simulated using DFT-based MD is the study of the potential drop at the interface formed between a pure ionic liquid and a graphene electrode,^[17] but it is worth noting that the computational cost of the simulations was of the order of 1 million CPU hours to obtain trajectories of ≈1 ps. For applications in which the effects of ion adsorption are important and an efficient sampling of the liquid configurations needs to be performed, the use of a less costly alternative is therefore necessary.

Consequently, the main method currently employed for simulating electrochemical double-layers is classical MD, in which the interactions are represented using a classical force field.^[8] In current electrode models, the polarization by the electrolyte is generally accounted for by allowing the partial atomic charges of the electrode atoms to fluctuate over time, while maintaining a constant potential difference between the negative and the positive electrodes.^[18,19] However, this approach did not include the so-called QC effects. As it has been known since the pioneering work of Gerischer on graphite,^[20] the DOS of the electrode may vary near the Fermi level, which reflects on the electrode's ability to respond to the polarization by the electrolyte. A simple solution to this problem was found by exploiting the fact that the EDL and quantum capacitance contributions are in series, and could thus be obtained by two different calculations.^[21,22] The QC can then be readily obtained at the DFT level from a static calculation; a recent example of this approach is the determination of the capacitance of graphdiynes, a new class of 2D porous materials that display a complex electronic structure.^[23] While it gives accurate capacitances, this approach cannot yield a good structural description of the EDL at a given potential since the interaction of the electrolyte with the electrode is not modified accordingly in the computational workflow. Recently, several attempts to account for the metallicity of the electrode were reported, in particular in the framework of the Thomas–Fermi model.^[24,25] The constant-potential version of this model includes an additional parameter, the Thomas–Fermi length, that accounts for the screening of the potential within the electrode. It was used to systematically study the effect of the electrode metallicity on properties such as the interfacial free energies.^[26] In principle, the Thomas–Fermi model includes intrinsically the QC effects,

yet the use of a single parameter that is material-dependent does not allow to investigate the impact of the variations of the number of charge carriers in the electrode with the applied potential.

In this work, we show how to account explicitly for the DOS of an electrode material in classical molecular dynamics simulations. To this end, we selected graphite as a representative material due to its ubiquitous importance and to its peculiar DOS in the vicinity of the Fermi level. The parameterization requires a simple DFT calculation, and the model can be used in the current implementations of the constant potential method. We show its application by computing the full EDL capacitance, which now contains the quantum contribution, for two different electrolytes, aqueous NaCl (with a concentration of 1 M) and the 1-ethyl-3-methylimidazolium tetrafluoroborate ionic liquid ([EMIM][BF₄] IL). We show that the agreement with experiments is almost quantitative, in particular our simulations reproduce the minimum of capacitance at the PZC. This then allows us to discuss the microscopic details of the systems; in particular, we show that despite its much lower ionic concentration the aqueous system displays a larger capacitance due to a strong contribution from the water molecules. In the future, this approach will enable the study of electrode materials with more complex electronic structure. In particular, beyond the context of carbon materials for which the quantum capacitance importance is now well established,^[23,27] including this effect opens the door for the MD-based modeling of electrolyte–electrode interfaces in lithium-ion batteries, pseudocapacitive materials and electrokinetic energy devices.^[28] For example, MXene electrode materials display a very peculiar DOS that cannot be caught by the conventional constant potential method. They also display many different chemical environment, which makes them very difficult to model accurately. However, further developments remain necessary to model such interfaces since they generally require the use of reactive force fields to account for the chemical reactions occurring on the liquid side of the interface.^[28]

2. Model

2.1. The Electrode Model and its Parameterization

The constant potential electrode model introduced by Siepmann and Sprik^[18] is based on a representation of the electrode charge distribution, ρ_{elec} , as a collection of atom-centered Gaussians:

$$\rho_{\text{elec}}(\mathbf{r}) = \sum_{i=1}^{N_{\text{elec}}} q_i \frac{1}{(2\pi\sigma^2)^{3/2}} \exp\left(-\frac{|\mathbf{r} - \mathbf{r}_i|^2}{2\sigma^2}\right) \quad (1)$$

where N_{elec} is the number of atoms (with positions \mathbf{r}_i) in the electrodes and σ is the width of the Gaussian. The partial charges q_i of the electrode atoms behave as additional degrees of freedom of the simulation, contrary to the electrolyte charges represented using point charges of fixed magnitude. The fluctuating charges should take values such that the potential is uniform within the whole electrode at each timestep. In practice, we define two electrodes with potentials Ψ_+ and Ψ_- (and with the corresponding number of atoms N_+ and N_-). The determination of the fluctuating charges is done within the Born–Oppenheimer approximation, by minimizing the total electrostatic potential energy of the

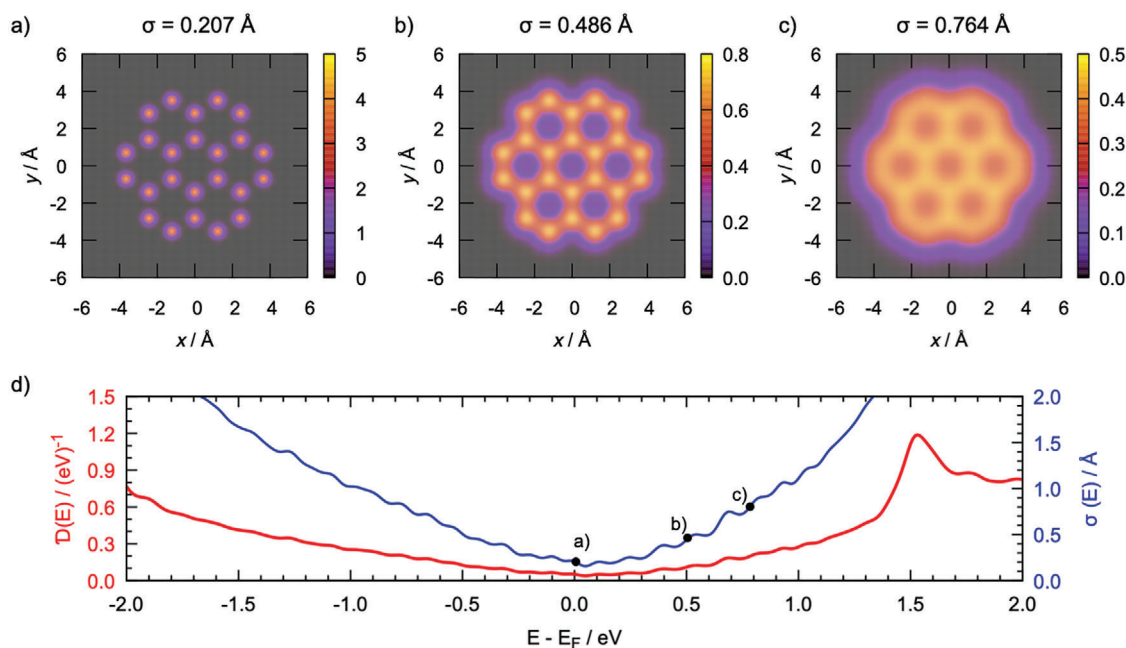


Figure 1. a–c) Illustration of the impact of the Gaussian width (σ) on the charge density of a coronene molecule. For visualization purposes, 3D charge distributions centered on the carbon atoms were used in this plot. The values in the colorbar are given in $e \text{ \AA}^{-2}$. The represented charge density corresponds to a case where all the partial charges of the carbon atoms are equal; in the constant potential model, they fluctuate in response to the electrolyte. d) The density of states of graphite obtained from DFT calculations.

system which includes all the Coulomb interactions, as well as the electric work which is necessary to charge the electrodes to the desired potential. The corresponding expressions are provided in ref. [29].

In the original model, this potential energy term includes only one parameter, which is the width of the Gaussian (taken the same for all the electrode atoms). Tuning its value has two effects. First, larger Gaussians lead to an increased delocalization of the electrode charge. This effect can be appreciated in the example of 2D charge distributions for a coronene molecule in **Figure 1a–c**, in the specific case for which all the atoms would carry the same partial charge. For example, for $\sigma = 0.207 \text{ \AA}$, it is highly localized around the atoms. Some overlap starts to be seen for a value of 0.486 \AA , as a non-negligible charge density starts to be observed along the C–C bonds. Finally, for a larger value of 0.764 \AA , the charge density is now almost equally distributed within the carbon rings. The latter representation is in agreement with pictures extracted from single-molecule imaging of aromatic carbonaceous molecules using scanning tunneling microscopy or atomic force microscopy.^[30] Second, larger Gaussian widths result in an enhanced ability to charge the atom. Indeed, the Gaussian charges self-interact through the term:

$$U_{\text{self}} = \frac{1}{4\pi\epsilon_0} \left(\frac{1}{\sigma\sqrt{4\pi}} \sum_{i \in (N_+ + N_-)} q_i^2 \right) \quad (2)$$

so that the magnitude of this self-interaction corresponds to an energy penalty for charging the electrode.

The extension of the model to include Thomas–Fermi effects simply introduced an additional energy penalty proportional to the Thomas–Fermi screening length l_{TF} :

$$U_{\text{TF}} = \frac{1}{4\pi\epsilon_0} \left(2\pi l_{\text{TF}}^2 \rho \sum_{i \in (N_+ + N_-)} q_i^2 \right) \quad (3)$$

where ρ is the number density of electrode atoms within the material. Note that this term does not impact the interactions between neighbor atoms and the delocalization of the charge within the plane, which remained controlled by the Gaussian width. Quantities such as the interfacial capacitance or the solid–liquid interfacial free energies were shown to vary systematically with l_{TF} in previous works.^[24,26]

Here our objective is to go further and use the DOS of the material ($D(E)$) to parameterize the electrode model. The DOS is defined as a measure of the number of allowed electronic states whose energies lie at a value E . A DOS plot of graphite, calculated at the DFT level (details given in the Supporting Information), is shown in **Figure 1d**. It displays a characteristic minimum around the Fermi level E_{F} , which often leads graphite to be classified as a “semi-metal”. To accurately describe the quantum capacitance effect, the residual value of the DOS at the Fermi level must be correctly quantified, requiring the use of a very dense K -point mesh and a consistent treatment of thermal broadening in the calculation of the DOS (see Supporting Information for details).^[31] The Thomas–Fermi length can be expressed in terms of the DOS at the Fermi level: $l_{\text{TF}} = \left(\frac{e^2}{\epsilon_0} D(E_{\text{F}}) \right)^{-1/2}$ which provides an easy parameterization of the model through Equation (3). This approach

was successfully used in a theoretical model to explain the difference in freezing properties behavior for an ionic liquid in contact with electrodes of varying metallicities.^[32]

Here we extend this approach by using the fact that in the fixed band approximation, the Fermi level is shifted by the bias potential to $E_F + e\Psi$, which leads to an increase in the number of available states. We use this correspondence to determine the DOS at a given electric potential, which in turn gives a new value for the Thomas–Fermi screening length. However, a difficulty arises in choosing the bias potential since in the constant potential electrode model; the fixed quantity is the applied potential $\Delta\Psi$ between the electrodes. A constraint of global electroneutrality is imposed,^[33] so that for $\Delta\Psi = 0$ the two electrodes are equivalent. Their charge must fluctuate around zero, and we can define their potential as the PZC ($\Psi_+ = \Psi_- = \Psi_{PZC}$). For non-null applied voltages, there exists a potential shift that depends on the instantaneous configurations and fluctuates around an average value, so that $\Psi_+ = \frac{\Delta\Psi}{2} + \Psi_{\text{shift}}$ and $\Psi_- = -\frac{\Delta\Psi}{2} + \Psi_{\text{shift}}$. In the following, based on the symmetric shape of the experimental capacitance, we arbitrarily neglect the variation of the electrode potentials ($\Psi_{\text{shift}} = 0$ V, but in future work, it should be possible to allow the Thomas–Fermi length to vary dynamically to account for it).

In principle, l_{TF} and σ play different roles, accounting for screening at long distance and charge delocalization at short distance, respectively. In practice, we chose to use the Gaussian width as the main parameter to control the DOS of the simulation instead of the l_{TF} , but further studies could combine both to tune the properties of the metal. This leads to the following parameterization for σ depending on the potential:

$$\sigma(\Psi) = \frac{1}{4\pi\epsilon_0} \left(\frac{e^2 D(E_F + e\Psi)}{\rho\sqrt{\pi}} \right) \quad (4)$$

The obtained values of Gaussian width are given in Table S1 (Supporting Information). At the PZC, the Gaussian width takes a minimal value of 0.207 Å which corresponds to the situation shown in Figure 1a for coronene. Since the DOS slightly differs depending on the sign of the potential, we considered non-identical Gaussian width for the negative and positive electrodes. However, for the largest applied potential of 2 V, the corresponding Gaussian width would be larger than 1.0 Å. This value corresponds to a situation where the overlap between neighboring Gaussians is too large, and the problem becomes ill-defined mathematically.^[34] In practice, we observed that the chosen σ cannot be larger than 0.764 Å, which corresponds to the case shown in Figure 1c for coronene. This value was thus selected for the 2 V potential simulation for both the positive and negative electrode atoms.

2.2. Non-Coulombic Electrode/Electrolyte Interactions

Additional parameters need also to be defined on the electrolyte side. Although, for the aqueous solution, a well-established non-polarizable force field is employed, it is not the case for the [EMIM][BF₄] ionic liquid, for which we use a polarizable interaction potential. The induced dipoles are calculated at each timestep together with the fluctuating charges, using a preconditioned conjugate gradient method as described in ref. [29]. It

should be noted that in this case, the simulation time depends on the value of the Gaussian width, i.e the performance drops by a factor of three when the Gaussian width increases from 0.207 Å to 0.764 Å. The parameters of the conventional OPLS force field cannot be used in order to avoid double-counting of some interactions as explained briefly in the following. We therefore, use the CL and Pol approach,^[35,36] which consists in scaling down the non-bonded attractive term (in the case of the Lennard Jones potential, the ϵ parameter) by the factor k_{ij} :

$$k_{ij} = \frac{E_{\text{disp}}}{E_{\text{disp}} + E_{\text{ind}}} \quad (5)$$

so that the potential accounts only for dispersion interactions, while the implicit induction contribution is removed. The scaling factor can be evaluated from the symmetry-adapted perturbation theory (SAPT) calculation, which allows isolating the dispersion and induction terms from the total interaction energy of two fragments. IL ions are treated as entire fragments, while a coronene molecule is considered as representative of the carbon nanomaterial. The geometries of corresponding dimers (cation–coronene, anion–coronene, and coronene–coronene) are optimized, and then potential energy scans are performed at different SAPT levels (details given in Supporting Information). An example of the SAPT scan is given in Figure S2 (Supporting Information), and the final k_{ij} values are reported in Table S2 (Supporting Information).

In order to validate the proposed model, we compared the atomic forces (given in the inset of Figure 2; Figure S3, Supporting Information) and density profiles to the ones of a DFT-based MD simulation. We re-used a model (ref. [17]), which represents a single graphite electrode with 3 layers and 200 ion pairs of [EMIM][BF₄]. As regards the forces, a good agreement is obtained, especially taking into account that no force-fitting procedure was applied during the force field development. As discussed above, due to the high computational cost associated with the DFT-MD method (see methods description in Supporting Information), the total simulation time was only ≈ 5 ps. Nevertheless, it allowed us to check whether the equilibrium distances between the ions and the electrodes are well reproduced. An example of density profiles is given in Figure 2. Both in the cases of the N atom of the EMIM⁺ cation and the B atoms of the BF₄⁻ anion, we observe an excellent agreement between two methods in the peak positions and intensities up to ≈ 15 Å from the surface, whether the same configuration as the ab initio one is used or not to perform the 4 ps classical MD simulation. The positions of the two first cationic and anionic layers are also preserved when a long 2 ns run is carried out. The intensities of these peaks then decrease as expected from the better sampling of the configuration space.

3. Results and Discussion

As a first application of our method, we focus on the canonical example of the graphite electrode. By subtracting the EDL capacitance (estimated using analytical contributions associated with the Helmholtz and the diffuse layer) from the experimentally measured differential capacitance, in 1985, Gerischer determined the QC, and hence the DOS around the Fermi level, in

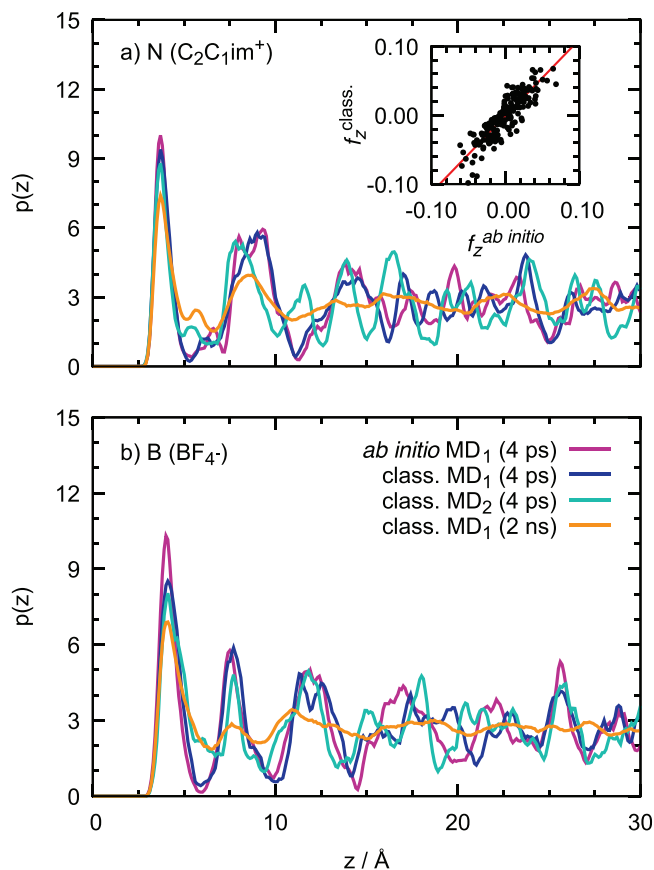


Figure 2. Normalised density profiles of selected atoms of [EMIM][BF₄] as a function of their distance from the electrode. Rose and blue lines correspond to the trajectories of the same length (≈ 4 ps), using the same initial configuration (denoted as MD₁ in the legend), while the orange line corresponds to a long 2 ns run. The aquamarine curve corresponds to a trajectory of ≈ 4 ps but starting from a different initial configuration (MD₂) of electrolyte ions. The inset shows the comparison of the f_z force component of C₂C₁im⁺ N atoms between ab initio MD and classical MD results for a given configuration.

qualitative agreement with the theoretical calculations available at the time.^[20] Here we use an inverse approach, in which the DOS is directly fed into the model, with the aim to determine the differential capacitance without any empirical input (other than the one used to build the force fields).

When using the constant potential method, a direct output of the simulation is the total charge ($Q_{\text{elec}} = \sum_{i \in N_{\text{elec}}} q_i$) on the electrodes. An electroneutrality constraint is imposed so that $Q_+ = -Q_- = Q$ at each timestep. In order to determine the capacitance, it is necessary to extract the average value of Q once the equilibrium has been reached at a given potential. As shown in Figure S4 (Supporting Information), this regime is obtained well before a time of 1 nanosecond of simulation for the aqueous solution, but several nanoseconds are necessary in the case of the ionic liquid due to its much larger viscosity.

The differential capacitance is then calculated by deriving the variation of $\langle Q \rangle$ with respect to the electrode potential, Ψ_+ or Ψ_- . In practice, due to the limited number of points, a fit of the $Q = f(\Psi)$ plot is necessary prior to the derivation. As can be seen in

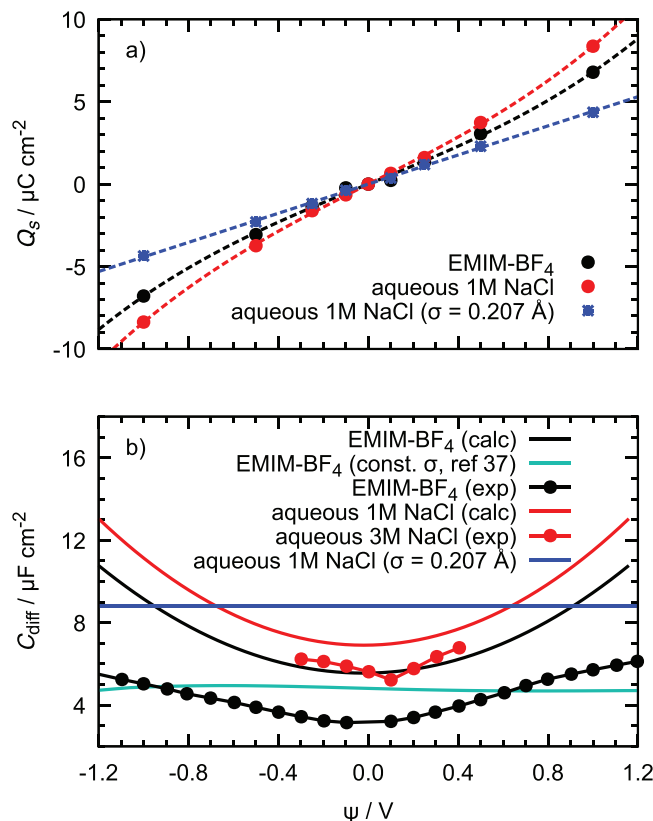


Figure 3. a) Average surface charge on the graphite with respect to the electrode potential (symbol: simulation data, dashed line: fit with a third-order polynomial function). The blue stars are the surface charges obtained using a constant Gaussian width of 0.207 Å, i.e., the Gaussian value obtained from the DOS at 0 V. b) Differential capacitance dependence on the electrode potential. The calculated data for polarisable [EMIM][BF₄] at constant Gaussian width are extracted from ref. [37], while the experimental values from Refs. [38] and [39].

Figure 3a, a third-order polynomial function yields a very good fit of the data. Comparison with the experimental results obtained for the same electrolytes,^[38,39] shown in Figure 3b as dotted lines, gives qualitative agreement. In particular, our simulations reproduce the minimum of capacitance at the point of zero charge (which corresponds by construction to an applied potential of 0 V in our case). Still, they overestimate the value of this minimum by $\approx 2 \mu\text{F cm}^{-2}$. This discrepancy may be attributed to the employed force field or to the fact that we simulate a perfect interface. Another potential source of error in the model is the use of atom-centered Gaussians to represent the charge density of the electrode. As was shown in previous works based on DFT calculations, in contrast to metals, in carbon materials such as graphite the position of the charge image plane lies well inside the surface (by ≈ 1 Å).^[40] Further tests combining our model with a different location of the fluctuating charges should allow to test this effect on the capacitance of graphite. Despite the observed shift in the absolute value, the variation of the differential capacitance with potential is qualitatively very similar between simulations and experiments, which shows that the response to applied voltage is well taken into account. In order to show further the impact of the QC in the simulation, additional simulations

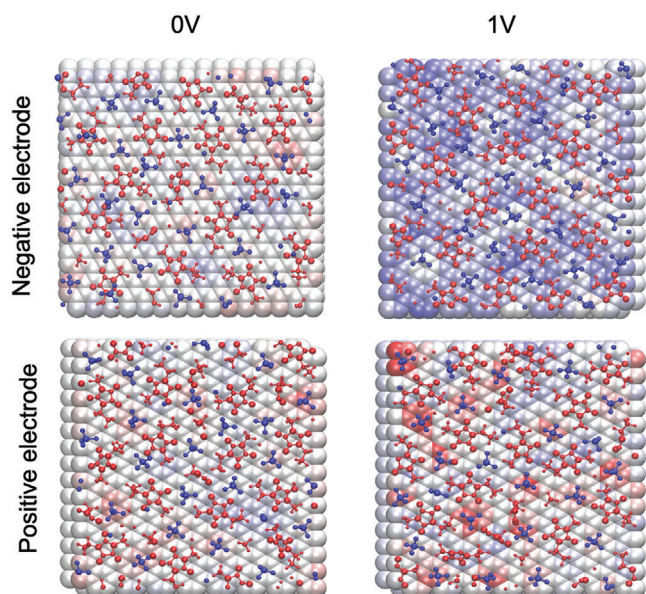


Figure 4. Representative snapshots of the adsorbed layer for [EMIM][BF₄] IL-based simulations: EMIM⁺ is given in red, BF₄⁻ in blue. The surface atoms are colored according to their charge (white: neutral, blue: negative, red: positive).

of the aqueous NaCl solution were performed with a fixed Gaussian width of 0.207 Å (that corresponds to the value taken at rest in our scheme). The accumulated charge and its fluctuations are then much smaller, as can be seen in Figure 3a (blue stars) and Figure S4c (Supporting Information), and its variation is better fitted with a linear function, which corresponds to a constant interfacial capacitance, as shown in Figure 3b. The differential capacitance for [EMIM][BF₄] obtained from constant potential simulations at fixed Gaussian width from ref. [37] (with a box setup very similar to the one used in this work) is also reported. As it can be seen from the Figure, alike the aqueous system the capacitance is constant across the potential range if the DOS are not taken into account.

Notably, as in experiments, when accounting for the QC we observe that the capacitance of the aqueous solution is larger than the one of the ionic liquid. This result may be surprising since the latter is entirely composed of ions, and could thus be expected to induce a larger charge at the surface of the electrode. Nevertheless, this is not the case, and IL ions induce smaller charges than NaCl ions and water molecules, as shown in Figure S5 (Supporting Information), in which we depicted the average charge per carbon atom of an aromatic ring of the top electrode carbon layer as a function of the distance of a given electrolyte species to the closest ring. It is, therefore, interesting to compare the local structure of the two systems and try to understand the microscopic mechanism at play. The structure of the two interfaces was largely documented in various simulation studies,^[37,41–43] but in the absence of an explicit QC effect and no systematic comparison was performed so far. On the one hand, in the ionic liquid, at zero applied electric potential, a “checkerboard-like” structure is obtained in which the first liquid layer is made of well-organized adsorbed ions. In particular, as shown in Figure 4, the imidazolium rings tend to lie flat on the surface^[44] and to arrange collectively

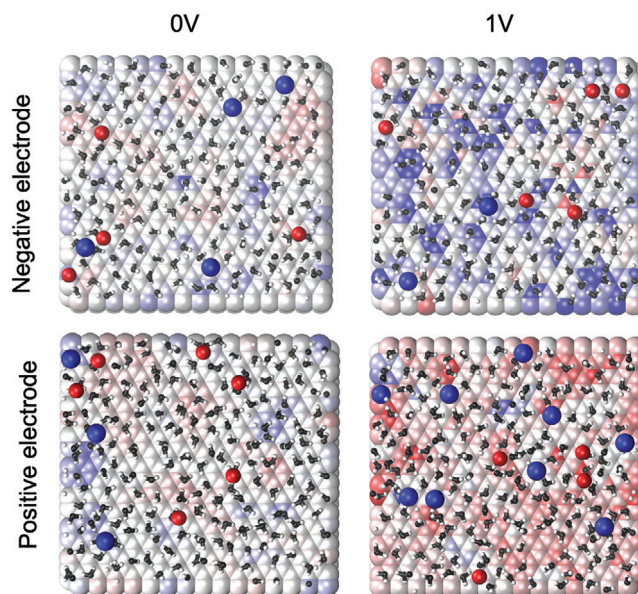


Figure 5. Representative snapshots of the adsorbed layer in the case of 1 M aqueous NaCl (black and white: water molecules, red spheres: Na⁺ ions, blue sphere: Cl⁻ ions). The surface atoms are colored according to their charge (white: neutral, red: positive, blue: negative). The represented Na⁺ and Cl⁻ ions are located beyond the plane of water molecules, and they do not induce any noticeable surface charge on the electrode.

along lines. This is consistent with previous observations based on atomic force microscopy experiments as well as previous MD simulations on similar systems.^[45,46] Upon application of a voltage, due to steric hindrance, the charge can only accumulate via an exchange mechanism. In the case of the negative electrode and for an applied voltage of 2 V, the cations also tend to reorient and become perpendicular to the electrode (Figure S6, Supporting Information), thus allowing a larger ionic density on the surface.

On the contrary, in the case of the aqueous sodium chloride solution, we do not observe any adsorption of the Na⁺ and Cl⁻ in the first liquid layer at low potential (see Figure S7, Supporting Information), even though the interfacial adsorption may differ depending on the force field used as well as the method to treat the surface polarization.^[47,48] Some ions are present in the second layer, but they do not lead to any noticeable surface charge on the electrode surface, as shown in Figure 5. This is consistent with previous simulations performed on gold surfaces, in which it was shown that the degree of metallicity of the gold had to be very high to promote their adsorption.^[49] Similarly, no specific adsorption of Na⁺ and Cl⁻ ions was observed in a previous work when the Thomas–Fermi length was systematically varying.^[24] The polarization of the surface is thus almost entirely due to the behavior of the liquid water molecules. At zero voltage, they adopt a parallel orientation with respect to the electrode surface, and they form a 2D hydrogen bond network inside the plane. Up to one volt, the charging of the electrode does not imply any variation in the composition of the EDL, but it is rather accompanied by a change in the orientation of the water molecules, as shown in Figure S8 (Supporting Information). In line with our results, a recent simulation study of aqueous NaCl at different

concentrations between Pt electrodes has shown that the electrostatic potential of the EDL is almost independent of the ionic concentration due to the molecular influence of water molecules.^[50]

Based on the different charging mechanisms, it is now possible to interpret the differences in capacitance between the two systems. From the energy point of view, charging the ionic liquid EDL requires breaking some strong Coulombic interactions between ions of opposite signs, while in the aqueous NaCl case, it only needs to rearrange the hydrogen bond network, which is substantially easier. Interestingly, this interpretation holds in the case of a planar surface, but it cannot be applied to more complex cases such as porous carbons for which the confinement effects lead to i) an increased screening of the Coulombic interaction, which was termed as a “superionic” effect by Kondrat and Kornyshev^[51,52] and confirmed by in situ X-ray diffraction experiments;^[53] and ii) the impossibility to induce a macroscopic polarization through the water dipoles only, for symmetry reasons. Consequently, in the latter case, the polarization occurs via the adsorption/desorption of the ions, and the capacitances are not larger anymore than for ionic liquids.^[54]

It is worth noting that the mechanisms described above are valid over the whole investigated potential range, and so there is interference from the QC on the overall behavior: It changes the onset potential at which the ion exchange/water reorientations occur. This validates the approaches based on a decoupling of the EDL and QC effect proposed in previous works,^[21–23] but our approach has the advantage of specifying the liquid structure at each potential without any ambiguity.

4. Conclusion and Perspectives

In conclusion, in this work we have shown how the constant potential method can be extended to explicitly account for the DOS of the material in MD simulations of electrochemical double layers. This extension can readily be used in any existing implementation, since it only requires fine-tuning of the Gaussian width of the electrode charge depending on the applied potential. The parameterization is easily made from a calculation of the DOS of the material at rest, using first-principles methods. Our new method, therefore, accounts directly for the QC contribution to the overall capacitance, contrarily to previous works in which it was simply added to the double-layer contribution. The advantage is that it is possible to extract the correct structure of the liquid at the interface for a given applied potential.

As a first application of the method, we focused on the graphite electrode. Despite its importance in energy applications and its apparent simplicity, its peculiar electronic structure is characterized by a minimum of the DOS at the Fermi level. Consequently, a correct description of the double-layer forming at its interface with electrolytes remained lacking so far. In this work, we obtained a qualitative agreement with experimental works for two important electrolytes, an aqueous NaCl solution and a pure ionic liquid. The simulations predict overall too large capacitances (by 1 to 2 $\mu\text{F cm}^{-2}$) for the two systems, but the variation with potential is well reproduced in both cases. The larger capacitance obtained for the aqueous system is interpreted based on the structure of the adsorbed liquid and its effect on the surface-induced charges.

In the future, this work could expand the scope of constant potential MD simulations to a variety of new materials for energy applications. In particular, many porous materials, such as carbon-based materials or metal-organic-frameworks, covalent organic frameworks, conducting polymers,^[55] etc, display a complex electronic structure, that cannot be approximated to the one of a perfect conductor, and could benefit from such a representation. Additional work is however required to discriminate the presence of different chemical elements (with eventually different environments) in such materials. The first steps in that direction were made in the case of 2D nanomaterials such as MoS_2 or MXene,^[56,57] by fixing the electrochemical potential of each element instead of the potential of the whole electrode. However, fixing the DOS cannot be made species by species since it is a material property instead of an individual one. In the future, this difficulty may be overcome by extending machine-learning-based approaches such as the PiNNwall interface,^[58] which was recently applied to the cases of chemically doped graphene and graphene oxide electrodes.

Supporting Information

Supporting Information is available from the Wiley Online Library or from the author.

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Conflict of Interest

The authors declare no conflict of interest.

Data Availability Statement

We will add the data on a public repository before publication of the article.

Keywords

constant potential molecular dynamics, carbonaceous electrodes, density of states, supercapacitors

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[1] M. Salanne, B. Rotenberg, K. Naoi K. Kaneko, P.-L. Taberna, C. P. Grey, B. Dunn, P. Simon, *Nat. Energy* **2016**, *1*, 16070.

- [2] I. Ledezma-Yanez, W. D. Z. Wallace, P. Sebastián-Pascual, V. Climent, J. M. Feliu, M. T. Koper, *Nat. Energy* **2017**, 2, 17031.
- [3] Y. E. Chipangura, B. D. Spindler, P. Bühlmann, A. Stein, *Adv. Mater.* **2023**, 36, 2309778.
- [4] G. Pace, A. Esau del Rio Castillo, A. Lamperti, S. Lauciello, F. Bonaccorso, *Adv. Mater.* **2023**, 35, 2211037.
- [5] A. C. Forse, C. Merlet, J. M. Griffin, C. P. Grey, *J. Am. Chem. Soc.* **2016**, 138, 5731.
- [6] A. A. Kornyshev, *J. Phys. Chem. B* **2007**, 111, 5545.
- [7] J. Wu, *Chem. Rev.* **2022**, 122, 10821.
- [8] G. Jeanmairat, B. Rotenberg, M. Salanne, *Chem. Rev.* **2022**, 122, 10860.
- [9] C. Pean, B. Daffos, B. Rotenberg, P. Levitz, M. Haefele, P.-L. Taberna, P. Simon, M. Salanne, *J. Am. Chem. Soc.* **2015**, 137, 12627.
- [10] A. Groß, *Curr. Opin. Electrochem.* **2023**, 40, 101345.
- [11] A. Groß, S. Sakong, *Chem. Rev.* **2022**, 122, 10746.
- [12] S. Surendralal, M. Rodorova, M. W. Finnis, J. Neugebauer, *Phys. Rev. Lett.* **2018**, 120, 246801.
- [13] C.-Y. Li, J.-B. Le, Y.-H. Wang, S. Chen, Z.-L. Yang, J.-F. Li, J. Cheng, Z.-Q. Tian, *Nat. Mater.* **2019**, 18, 697.
- [14] X.-Y. Li, X.-F. Jin, X.-H. Yang, X. Wang, J.-B. Le, J. Cheng, *J. Chem. Phys.* **2023**, 158, 084701.
- [15] M. C. O. Monteiro, F. Dattila, B. Hagedoorn, R. García-Muelas, N. López, M. T. M. Koper, *Nat. Catal.* **2021**, 4, 654.
- [16] X. Qin, T. Vegge, H. A. Hansen, *J. Am. Chem. Soc.* **2023**, 145, 1897.
- [17] H. Ers, M. Lembinen, M. Mišin, A. P. Seitsonen, M. V. Fedorov, V. B. Ivaništšev, *J. Phys. Chem. C* **2020**, 124, 19548.
- [18] J. I. Siepmann M. Sprik, *J. Chem. Phys.* **1995**, 102, 511.
- [19] S. K. Reed, O. J. Lanning, P. A. Madden, *J. Chem. Phys.* **2007**, 126, 084704.
- [20] H. Gerischer, *J. Phys. Chem.* **1985**, 89, 4249.
- [21] A. A. Kornyshev, N. B. Luque, W. Schmickler, *J. Solid State Electrochem.* **2014**, 18, 1345.
- [22] A. J. Pak, E. Paek, G. S. Hwang, *Phys. Chem. Chem. Phys.* **2013**, 15, 19741.
- [23] T. Mo, Z. Wang, L. Zeng, M. Chen, A. A. Kornyshev, M. Zhang, Y. Zhao, G. Feng, *Adv. Mater.* **2023**, 35, 2301118.
- [24] L. Scalfi, T. Dufils, K. G. Reeves, B. Rotenberg, M. Salanne, *J. Chem. Phys.* **2020**, 153, 174704.
- [25] A. Schlaich, D. Jin, L. Bocquet, B. Coasne, *Nat. Mater.* **2022**, 21, 237.
- [26] L. Scalfi B. Rotenberg, *Proc. Natl. Acad. Sci. USA* **2021**, 118, 210879118.
- [27] J. Zhu, A. S. Childress, M. Karakaya, S. Dandeliya, A. Srivastava, Y. Lin, A. M. Rao, R. Podila, *Adv. Mater.* **2016**, 28, 7185.
- [28] K.-j. Jeong, S. Jeong, S. Lee, C. Y. Son, *Adv. Mater.* **2023**, 35, 2204272.
- [29] A. Coretti, C. Bacon, R. Berthin, A. Serva, L. Scalfi, I. Chubak, K. Goloviznina, M. Haefele, A. Marin-Lafleche, B. Rotenberg, S. Bonella, M. Salanne, *J. Chem. Phys.* **2022**, 157, 184801.
- [30] L. Gross, F. Mohn, N. Moll, P. Liljeroth, G. Meyer, *Science* **2009**, 325, 1110.
- [31] T. Binniger, *Phys. Rev. B* **2021**, 103, L161403.
- [32] J. Comtet, A. Niguès, V. Kaiser, B. Coasne, L. Bocquet, A. Siria, *Nat. Mater.* **2017**, 16, 634.
- [33] L. Scalfi, D. T. Limmer, A. Coretti, S. Bonella, P. A. Madden, M. Salanne, B. Rotenberg, *Phys. Chem. Chem. Phys.* **2020**, 22, 10480.
- [34] T. R. Gingrich M. Wilson, *Chem. Phys. Lett.* **2010**, 500, 178.
- [35] A. A. H. Pádua, *J. Chem. Phys.* **2017**, 146, 204501.
- [36] K. Goloviznina, J. N. C. Lopes, M. C. Gomes, A. A. H. Pádua, *J. Chem. Theory Comput.* **2019**, 15, 5858.
- [37] J. B. Haskins, J. W. Lawson, *J. Chem. Phys.* **2016**, 144, 134701.
- [38] O. Oll, T. Romann, C. Siimenson, E. Lust, *Electrochem. commun* **2017**, 82, 39.
- [39] P. Iamprasertkun, W. Hirunpinoyopas, A. Keerthi, B. Wang, B. Radha, M. A. Bissett, R. A. W. Dryfe, *J. Phys. Chem. Lett.* **2019**, 10, 617.
- [40] N. B. Luque W. Schmickler, *Electrochim. Acta* **2012**, 71, 82.
- [41] P. H. L. Ferreira, A. M. Sampaio, L. J. A. Siqueira, *Electrochimica Acta* **2022**, 410, 139982.
- [42] H. He, J. Zhou, L. Yang, C. Liang, S. Xu, M. Chen, T. Mo, *J. Mater. Chem. A* **2024**, 12, 10279.
- [43] Z. Hu, J. Vatamanu, O. Borodin, D. Bedrov, *Phys. Chem. Chem. Phys.* **2013**, 15, 14234.
- [44] T. Carstens, R. Gustus, O. Höfft, N. Borisenko, F. Endres, H. Li, R. J. Wood, A. J. Page, R. Atkin, *J. Phys. Chem. C* **2014**, 118, 10833.
- [45] E. Hayes, G. G. Warr, R. Atkin, *Phys. Chem. Chem. Phys.* **2010**, 12, 1709.
- [46] J. M. Black, D. Walters, A. Labuda, G. Feng, P. C. Hillesheim, S. Dai, P. T. Cummings, S. V. Kalinin, R. Proksch, N. Balke, *Nano Lett.* **2013**, 13, 5954.
- [47] O. Björneholm, M. H. Hansen, A. Hodgson, L.-M. Liu, D. T. Limmer, A. Michaelides, P. Pedevilla, J. Rossmeisl, H. Shen, G. Tocci, E. Tyrode, M.-M. Walz, J. Werner, H. Bluhm, *Chem. Rev.* **2016**, 116, 7698.
- [48] C. Y. Son Z.-G. Wang, *Proc. Natl. Acad. Sci. USA* **2021**, 118, 2020615118.
- [49] A. Serva, L. Scalfi, B. Rotenberg, M. Salanne, *J. Chem. Phys.* **2021**, 155, 044703.
- [50] A. Limaye, D. Suvlu, A. P. Willard, *Faraday Discuss.* **2024**, 249, 267.
- [51] S. Kondrat A. A. Kornyshev, *J. Phys.: Condens. Matter* **2011**, 23, 022201.
- [52] S. Kondrat, G. Feng, F. Bresme, M. Urbakh, A. A. Kornyshev, *Chem. Rev.* **2023**, 123, 6668.
- [53] R. Futamura, T. Iiyama, Y. Takasaki, Y. Gogotsi, M. J. Biggs, M. Salanne, J. Ségalini, P. Simon, K. Kaneko, *Nat. Mater.* **2017**, 16, 1225.
- [54] M. Simoncelli, N. Ganfoud, A. Sene, M. Haefele, B. Daffos, P.-L. Taberna, M. Salanne, P. Simon, B. Rotenberg, *Phys. Rev. X* **2018**, 8, 021024.
- [55] J. Wang, J. Wang, Z. Kong, K. Lv, C. Teng, Y. Zhu, *Adv. Mater.* **2017**, 29, 1703044.
- [56] S. Bi M. Salanne, *ACS Nano* **2022**, 16, 18658.
- [57] A. M. Sampaio, S. Bi, M. Salanne, L. J. A. Siqueira, *Energy Storage Mater.* **2024**, 70, 103502.
- [58] T. Dufils, L. Knijff, Y. Shao, C. Zhang, *J. Chem. Theory Comput.* **2023**, 19, 5199.