

Video Article

Multiscale Sampling of a Heterogeneous Water/Metal Catalyst Interface using Density Functional Theory and Force-Field Molecular Dynamics

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Abstract

A significant number of heterogeneously-catalyzed chemical processes occur under liquid conditions, but simulating catalyst function under such conditions is challenging when it is necessary to include the solvent molecules. The bond breaking and forming processes modeled in these systems necessitate the use of quantum chemical methods. Since molecules in the liquid phase are under constant thermal motion, simulations must also include configurational sampling. This means that multiple configurations of liquid molecules must be simulated for each catalytic species of interest. The goal of the protocol presented here is to generate and sample trajectories of configurations of liquid water molecules around catalytic species on flat transition metal surfaces in a way that balances chemical accuracy with computational expense. Specifically, force field molecular dynamics (FFMD) simulations are used to generate configurations of liquid molecules that can subsequently be used in quantum mechanics-based methods such as density functional theory or *ab initio* molecular dynamics. To illustrate this, in this manuscript, the protocol is used for catalytic intermediates that could be involved in the pathway for the decomposition of glycerol (C₃H₈O₃). The structures that are generated using FFMD are modeled in DFT in order to estimate the enthalpies of solvation of the catalytic species and to identify how H₂O molecules participate in catalytic decompositions.

Video Link

The video component of this article can be found at <https://www.jove.com/video/59284/>

Introduction

Modeling molecular phenomena involved in heterogeneous catalysis under liquid conditions is necessary for understanding catalytic function; however, this remains challenging because it requires a fine balance between chemical accuracy and computational expense. In general, since catalysis involves the breaking and forming of chemical bonds, quantum mechanics must be used to at least some degree; however, long simulations are challenging in quantum mechanics, as they require significant computer resources. Since molecules in the liquid phase are under constant thermal motion, simulations must also include configurational sampling, i.e., they must incorporate multiple spatial arrangements of the liquid molecules, as each different spatial arrangement (i.e., each configuration) has a different energy. This means that multiple configurations of liquid molecules must be simulated for each catalytic species of interest. These needs – to use quantum mechanics and to perform multiple calculations per catalytic species – can render modeling in heterogeneous catalysis under liquid phase computationally intractable. The purpose of the method described herein is to enable computationally tractable simulations of phenomena in heterogeneous catalysis under liquid phase.

We are particularly interested in heterogeneously catalyzed reactions that are carried out under liquid water. Water molecules have significant influence on catalytic phenomena, such as interacting with catalytic species (e.g., via dispersion forces and hydrogen bonding)^{1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23}, participating in catalytic reactions^{1,7,8,9,15,21,22,24,25,26,27}, and influencing reaction pathways and/or catalytic rates^{1,11,12,15,18,23,25,27,28,29,30,31}. Modeling of these phenomena has been performed using QM and/or *ab initio* molecular dynamics (AIMD)^{1,2,6,7,14,22,25,27,28,32,33,34}, force field molecular dynamics (FFMD)³⁵, and quantum mechanics/molecular mechanics (QM/MM)¹⁰. In AIMD and FFMD, the atoms in the system are moved pursuant to Newton's equations of motion according to the forces acting upon them. In AIMD, the system energy and forces are calculated with quantum mechanics, whereas in FFMD, the system energy and forces are calculated using force fields, which are algebraic expressions that are parameterized based on experimental or QM data. In QM/MM, the portion of the system where the bond breaking and forming occurs is calculated with QM, and the remainder of the system is calculated with MM, which employs force fields. Because they directly employ QM, AIMD and QM/MM are better suited for capturing the bond breaking and forming that occurs in aqueous phase heterogeneous catalysis; however, FFMD is significantly more computationally tractable and thus better suited for generating the configurations of liquid H₂O molecules. The method presented in this protocol balances chemical accuracy and computational expense by employing a combination of QM and FFMD.

Specifically, this method uses FFMD simulations for generating configurations of liquid H₂O and QM to calculate system energies. FFMD is carried out using LAMMPS.³⁶ The force fields used in FFMD in this work employ Lennard-Jones + Coulomb (LJ+C) potentials, where the LJ parameters have been taken from the TIP3P/CHARMM model³⁷ for H₂O, the universal force field³⁸ (UFF) for Pt, and the OPLS-AA force field³⁹ for catalytic species, and the Coulomb parameters have been taken from the TIP3P/CHARMM³⁷ model for H₂O and the OPLS-AA force field³⁹ for catalytic species. The Coulomb parameters for Pt atoms have been set to 0. QM calculations are performed using the VASP code^{40,41,42}, which is a density functional theory (DFT) code. Water molecule insertions are performed with a code developed in-house called Monte Carlo Plug-in for Quantum Methods (MCPIQ). File conversions from VASP to LAMMPS in this protocol are performed with the Visual Molecular Dynamics (VMD) software⁴³.

The protocol is intended to generate configurations of liquid water molecules around catalytic species on flat transition metal surfaces at low coverage. Coverage is denoted θ and defined as the number of adsorbates per surface metal atom (i.e., the number of surface adsorbates normalized by the number of metal atoms in the topmost layer of the metal slab in the catalyst model). In this manuscript, low coverage is defined as $\theta \leq 1/9$ monolayer (ML), where 1 ML means one catalytic species per surface metal atom. The catalyst models should be placed in periodic simulation boxes. The simulation boxes do not have to be cubes. This manuscript demonstrates the use of the protocol for generating configurations of liquid H₂O that can be used to calculate quantities of interest in aqueous phase heterogeneous catalysis.

This protocol requires that the user has access to installed and working versions of the VASP, MCPIQ, LAMMPS, and VMD software. More information about VASP (<https://www.vasp.at/>), LAMMPS (<https://lammps.sandia.gov/>), and VMD (<https://www.ks.uiuc.edu/Research/vmd/>) are available on their respective websites. The MCPIQ software is documented at https://github.com/getman-research-group/JoVE_article, along with all input files and Python scripts mentioned in this protocol. This protocol assumes that the executables and scripts mentioned within will be run on a high-performance research computer and are installed in a directory that is in the user's \$PATH variable. If an executable or script is placed in a location that is not in the user's \$PATH, then the path to the executable must be included to execute it. Executables and scripts are executed in steps 2.1.2, 2.2.1, 2.2.8, 3.1, 4.2, 5.2, and 6.1.2. For example, to execute the MCPIQ code in step 2.1.2 from a directory that is not in the user's \$PATH, the user would type \$PATHTOMCPLIQ/mcpliq at the command-line interface instead of mcpliq, where \$PATHTOMCPLIQ is the location where the mcpliq executable has been stored (e.g., \$PATHTOMCPLIQ might be ~/bin). Before starting this protocol, all executables and scripts should be given executable permissions (e.g., in Linux, this could be done by typing **chmod +x mcpliq** at the command-line interface from the directory where the mcpliq executable is stored). Further, any modules required by any of the software or scripts should be loaded (these dependencies will be specific to individual installations of the various software and the computer where the simulations will be run).

Protocol

1. Generate the adsorbate structure

1. Create a VASP POSCAR file comprising a supercell with periodic boundary conditions, as you would if you were performing simulations of adsorbates on metal surfaces under vacuum. The supercell should include an initial guess of the adsorbate structure and the metal surface as well as vacuum space above the adsorbate for adding H₂O molecules. Details on how to do this are provided in previous work^{35,44}.
NOTE: It is important that the height of the vacuum space be at least 12 Å above the top of the adsorbate.
2. Relax the structure and minimize its energy using the VASP code. Details on how to do this are provided in previous work^{35,44}. This will produce a file called CONTCAR, which will be used in the next section.

2. Add explicit H₂O molecules

1. Add N water molecules to the vacuum space in the CONTCAR created in step 1.2 using the MCPIQ code, where $N = \rho V$, ρ is the density of water, and V is the volume of the vacuum space above the adsorbate.
NOTE: ρ should be taken as the density of water as determined by the TIP3P/CHARMM water model at the simulation temperature. V will be refined in the next step.
 1. Specify the following information in the MCPIQ master_input.txt file: The number of H₂O molecules to add (N) by changing the first argument in line 28, the path to the water.txt file by changing the second argument in line 28, and the minimum and maximum height of the supercell that can be occupied by water molecules by changing the **Minimum z-coordinate** in line 11 and the **Maximum z-coordinate** in line 12.
 2. Execute the MCPIQ code by typing **mcpliq** from the command-line interface to insert water molecules into the CONTCAR file. The code will output one or more files with the file extension of .POSCAR.
NOTE: If more than one POSCAR files are produced, they will be named POSCAR_n.POSCAR. Select the file where n is largest.
2. Generate the LAMMPS input files for an NPT simulation and equilibrate the cell volume using FFMD in the NPT ensemble in LAMMPS.
 1. Execute the script **lmps_bond_angle.py** on the .POSCAR file generated in step 2.1.2 by typing **lmps_bond_angle.py \$filename.POSCAR** at the command-line interface, where \$filename is the name of the .POSCAR file generated in step 2.1.2. This script creates a file called \$filename.POSCAR.bond_angle_info.txt which lists bonds and angles that will be used in the LAMMPS data file.
 2. Open VMD and select **File > New Molecule** in the main window to open the **Molecule File Browser** window. Select VASP_POSCAR from the **Determine File Type** dropdown menu. Click **Browse** and navigate to the \$filename.POSCAR file. Click **Load** to open the \$filename.POSCAR file.
 3. Open the Tk Console within VMD by selecting **Extensions > Tk Console** from the VMD Main window.
 4. Execute the following command in the Tk console: **topo writelammpsdata \$WDPATH/data.myadsorbate full**, where \$WDPATH is the directory on the computer where VMD will write the LAMMPS data file and data.myadsorbate is the name of the LAMMPS data file.
 5. Delete the **Bonds and Angles** section at the bottom of the data.myadsorbate file. Then, append the bond and angle lists in the file \$filename.POSCAR.bond_angle_info.txt into data.myadsorbate.

NOTE: The indexes for the O-H bond type and H-O-H angle type for the water molecules in the \$filename.POSCAR.bond_angle_info.txt file are both set to 1. Thus, bond and angle types for adsorbates should start counting at 2.

6. Edit the data.myadsorbate file by adding the Lennard-Jones parameters to the **Pair Coeffs** section and the Coulomb parameters to the **Atoms** section. Lennard-Jones and Coulomb parameters for the H₂O molecules, adsorbate atoms, and metal surface atoms need to be added.

NOTE: The Lennard-Jones parameters for Pt atoms, water molecules, and adsorbate atoms in this protocol are obtained from the UFF³⁸, TIP3P/CHARMM³⁷, and OPLS-AA³⁹ force fields, respectively. Coulomb parameters for water molecules and adsorbates atoms are obtained from the TIP3P/CHARMM³⁷ and OPLS-AA³⁹ force fields, respectively. Coulomb parameters for Pt atoms are set to 0 in this protocol. Alternatively, calculated partial charges could be used for the Coulomb parameters for adsorbate atoms and Pt atoms.

7. Copy the LAMMPS input file input.equil into the directory \$WDPATH. Edit the group variable on line 34 to indicate the atom type indexes for the water oxygen and water hydrogen atoms and the group variable on line 35 to indicate the atom type indexes for the Pt and adsorbate atoms.
8. Execute the LAMMPS software by typing `mpiexec -np XX lmp_mpi < input.equil` at the command-line interface, where XX is the number of CPU cores to use, and lmp_mpi is the name of the LAMMPS executable. Doing this will run an energy minimization to refine the H₂O configuration, followed by a FFMD simulation performed at constant number of H₂O molecules (*N*), volume (*V*), and temperature (*T*) to bring the water to the simulation temperature, followed by a FFMD simulation run at constant *N*, pressure (*P*), and temperature (*T*) to determine the physically correct height of the simulation box. Output files that will be used in section 3 are called data.myadsorbate_npt and log.myadsorbate.

NOTE: The duration of the NPT simulation should be long enough to comprise an “equilibration” run, where the volume of the supercell comes to steady state, and a “production” run, which is used to sample the ensemble averages (here, the height of the supercell). During the equilibration run, the volume of the supercell when plotted against time should level off to a steady state value. Once this occurs, the NPT simulation can be said to be in its production run. Verify equilibration of the NPT simulation by ensuring that the fluctuations in the height of the supercell (*lz*) are minimal or have converged to a steady value. If large fluctuations occur, then re-generate a H₂O configuration by decreasing the timestep on line 92 in the input.equil file and repeating step 2.2.8 or starting again from step 2.1.1.

3. Extract the proper height of the supercell

1. Execute the script get_npt_lz.py on the log.myadsorbate file by typing `get_npt_lz.py log.myadsorbate` at the command-line interface. This script outputs the average supercell height from the “production” run portion of the NPT simulation in the avg_lz.txt file.
NOTE: The get_npt_lz.py script assumes that LAMMPS writes the length of the cell z-dimension (*lz*) to the log.myadsorbate file every 1000 fs (customizable at line 20 of the get_npt_lz.py script), which is the default in the provided input.equil LAMMPS input file. The get_npt_lz.py script detects and discards the first 2 ns (customizable at line 19 of the get_npt_lz.py script) worth of *lz* values in the log.myadsorbate file, as they comprise the equilibration portion of the simulation, while the remaining 3 ns comprise the “production” portion and are thus used by the get_npt_lz.py script to compute the average z-dimension length. In addition to the avg_lz.txt file, the get_npt_lz.py script outputs a file called npt_data.txt, which provides values of *lz* as a function of the timestep, as well as a file called npt_plot.png, which plots the same data. The plot can be used to verify equilibration of the NPT simulation.
2. Reconstruct the supercell using the average height determined in NPT.
 1. Copy the data.myadsorbate_npt file into a new directory, referred to here as \$WD2PATH, and rename it data.myadsorbate.
 2. Edit the new data.myadsorbate file so that the *lz* height is equal to the average value output from the get_npt_lz.py script by changing the zlo and zhi arguments in the data.myadsorbate file such that zlo is 0.0 and zhi is the *lz* value from the avg_lz.txt file produced in step 3.1.

4. Generate configurations of H₂O molecules

1. Copy the LAMMPS input file input.prod into \$WD2PATH. Edit the group variable on line 32 to indicate the atom type indexes for the water oxygen and water hydrogen atoms and the group variable on line 33 to indicate the atom type indexes for the Pt and adsorbate atoms.
2. Execute the LAMMPS software by typing `mpiexec -np XX lmp_mpi < input.prod` into the command-line interface, where XX is the number of CPU cores to use, and lmp_mpi is the name of the LAMMPS executable. Doing this will run a constant NVT simulation on the H₂O molecules. The key output file from this simulation is the dump.myadsorbate.lammpstrj file.
NOTE: The duration of the NVT simulation should be long enough to comprise an equilibration run, where the energy of the system comes to steady state, and a production run, from which the ensemble averages (here, the spatial positions of the water molecules) are sampled. During the equilibration run, the energy of the system when plotted against time should level off to a steady state value. Once this occurs, the NVT simulation can be said to be in its production run.

5. Determine the hydrogen bond lifetime for proper time sampling

1. Edit the hb_lifetime_dist.py script to specify: the timestep of the first frame of the dump.myadsorbate.lammpstrj file by changing the actualStart variable on line 22, how often frames are written to the LAMMPS trajectory file by changing the timestep variable on line 23, the first and last timesteps the script should consider (i.e., the production portion of the trajectory) by changing the N_first and N_last variables on lines 24 and 25, whether consecutive frames are considered or frames are skipped by changing the nevery variable on line 26, and the number of lines per frame section of the trajectory file by changing the frameLine variable on line 27. Additionally, edit lines 31 through 35 to specify which atom types within the data.myadsorbate file belong to the adsorbate and which atom types belong to the H₂O molecules.
NOTE: The hb_lifetime_dist.py script analyzes the H₂O configurations in the production run and determines if any H₂O molecules are hydrogen bonded to the adsorbate. It then counts the simulation time that each hydrogen bond remains intact and reports this information as a distribution of hydrogen bond lifetimes in units of ps. The specific version of the script that is provided with this protocol assumes

that LAMMPS writes the configuration of H₂O molecules to the dump.myadsorbate.lammpstrj file every 1000 fs, which is the default in the provided input.prod LAMMPS input file. It detects and discards the first 2 ns worth of configurations in the the dump.myadsorbate.lammpstrj file, as they comprise the equilibration portion of the simulation, and uses the remaining 3 ns to calculate hydrogen bond lifetimes.

2. Execute the script hb_lifetime_dist.py on the dump.myadsorbate.lammpstrj file by typing **hb_lifetime_dist.py** at the command-line interface. Doing this will produce a file called distribution_HB_lifetime.dat.
3. Plot the data in the distribution_HB_lifetime.dat file to view the distribution of hydrogen bond lifetimes that occurred during the NVT simulation.
4. Determine the time increment to use for the time sampling interval based on the calculated hydrogen bond lifetimes. The best choice is the maximum hydrogen bond lifetime; alternatively, a value that would capture the 95% confidence interval can be used.

6. Sample configurations of liquid H₂O molecules

1. Determine the number of configurations from the production run of the NVT FFMD trajectory for further calculations. The number of configurations should be selected so that the minimum time between configurations is equal to or greater than the time sampling interval identified in section 5.
 1. Edit the default value for the num_frames variable on line 21 of the lammps_frames.py script to specify the number of configurations to extract.
 2. Execute the script lammps_frames.py on the file dump.myadsorbate.lammpstrj by typing **lammps_frames.py** at the command-line interface. Doing this will output a list of simulation times corresponding to the configurations that should be extracted from the dump.myadsorbate.lammpstrj file. These configurations can be used as starting structures in AIMD or QM simulations.
- NOTE: 1) The lammps_frames.py script automatically detects the LAMMPS log and dump files as well as the production portion of the trajectory within the dump file and divides the number of configurations within the dump file into 10 groups. Alternatively, the user can specify the log file, the dump file, and the number of configurations from the command-line interface using the -l, -d, and -n options, respectively. To do so, the user should type **lammps_frames.py -n XX -l \$logfile -d \$dumpfilename** at the command-line interface, where XX is the desired number of configurations, \$logfile is the name of the LAMMPS logfile, and \$dumpfilename is the name of the LAMMPS trajectory (dump) file. The simulation times that are output refer to the median times in each group. 2) If the configurations will be calculated in VASP with the LDIPOL flag turned on, a small layer of vacuum space should be added to the top of the supercell above the water layer. This will facilitate convergence of the electronic structure in the VASP calculation. Adding an additional 3 Å of vacuum space above the H₂O molecules has been successful in the simulations discussed below.

Representative Results

One use of this protocol is to calculate the energies of interaction between the liquid water and catalytic species, i.e., ΔE_{int}^{35} :

$$\Delta E_{int} = E_{\text{Catalytic species} + \text{H}_2\text{O}} + E_{\text{Clean catalyst surface}} - E_{\text{Catalytic species}} - E_{\text{Clean catalyst surface} + \text{H}_2\text{O}}$$

where $E_{\text{Catalytic species} + \text{H}_2\text{O}}$ is the energy of a configuration of H₂O molecules around a catalytic species on a metal surface, $E_{\text{Clean catalyst surface}}$ is the energy of the clean catalyst surface in vacuum, $E_{\text{Catalytic species}}$ is the energy of the catalytic species on a metal surface in vacuum, and $E_{\text{Clean catalyst surface} + \text{H}_2\text{O}}$ is the energy of the configuration of H₂O over the catalyst surface with the catalytic species removed. The positions of the H₂O molecules used to calculate $E_{\text{Catalytic species} + \text{H}_2\text{O}}$ and $E_{\text{Clean catalyst surface} + \text{H}_2\text{O}}$ should be identical. All values of E are calculated using the VASP code. The quantity ΔE_{int} includes all of the physical and chemical interactions between all of the molecules in the liquid water structure and the catalytic species and gives a reasonable estimate of the enthalpy of solvation of the catalytic species, which is needed to calculate its free energy of solvation and total free energy. **Table 1** provides values for ΔE_{int} calculated for species on a Pt(111) catalyst surface with chemical formulas equal to C_xH_yO_z in units of eV (1 eV = 96.485 kJ/mol). The values were calculated at coverages $\leq 1/9$ ML.^{35,46} The values reported are the averages taken over 10 configurations of liquid H₂O, and the uncertainties are reported as standard deviations. All the values are negative, indicating favorable interactions with water.

Another application of this protocol is to generate starting structures for AIMD. Movie 1 is a movie of an AIMD trajectory that was started from a configuration generated by this protocol. At the start of this movie, a COH adsorbate is shown on a Pt(111) surface under a structure of liquid H₂O. One H₂O molecule is emphasized, which formed a hydrogen bond with COH. Over the course of the movie, this H₂O molecule abstracts the proton from the COH adsorbate and deposits a second hydrogen atom on the Pt(111) surface. The H₂O molecule thus helps to catalyze the reaction COH* + * → CO* + H*, where the *s indicate catalytic sites. This simulation highlights the main strength and the main purpose of the multiscale sampling method described herein. Numerous configurations of H₂O molecules are generated with FFMD, due to its strength in computational tractability. However, a limitation of FFMD is that it cannot capture bond breaking and forming unless a reactive force field is implemented. AIMD uses quantum mechanics to calculate energies and thus can capture bond breaking and forming. However, AIMD is too computationally demanding to generate all of the configurations of H₂O molecules necessary to ensure sufficient sampling has been achieved. Thus, this protocol combines the two methods.

The structures of liquid H₂O molecules generated by this procedure are dependent on input settings. Setting these improperly can have unintended influences on the water structures. For example, when the intermolecular distances become too small or when other parameters in the molecular dynamics input files are set improperly or take on unphysical values, the water structure can become unreasonable. Under these circumstances, the structure of water will “blow up” unintendedly during the FFMD trajectory. **Figure 1** shows an example of this. The snapshot on the left-hand side is the starting structure for a FFMD run, and the snapshot on the right-hand side is a snapshot taken within 1 ps of starting the simulation. As can be seen, the H₂O molecules have moved far away from the surface. This is caused by improper settings made in the simulation input files and is not a structure that is likely to occur in reality.

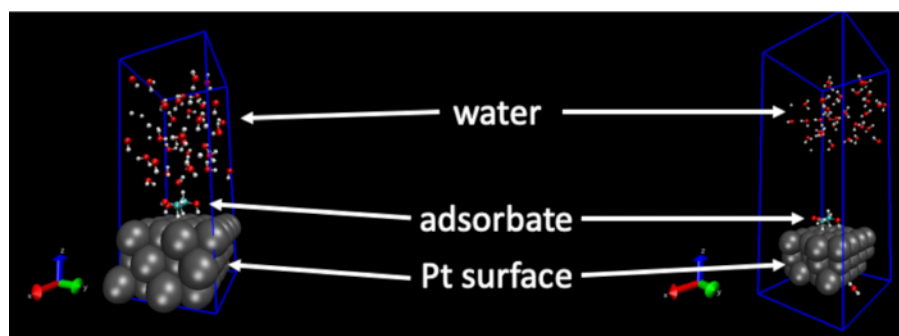
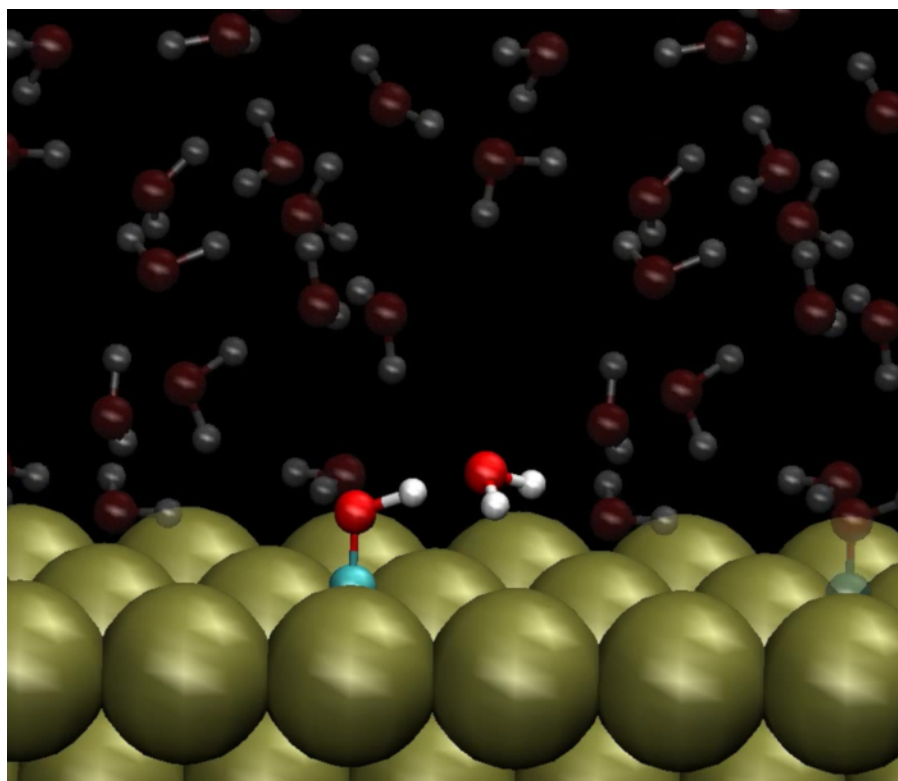


Figure 1: Example of a negative result. The force field molecular dynamics simulation “blows up” due to an unphysical setting or value. Left hand image: The starting geometry of the Pt(111) surface, adsorbate, and liquid water structure. Right hand image: The geometry of the Pt(111) surface, adsorbate, and liquid water structure less than 1 ps later. In the right-hand image, the H₂O molecules have separated from the surface due to unphysically large forces. [Please click here to view a larger version of this figure.](#)



Movie 1: *Ab initio* molecular dynamics (AIMD) simulation initiated from a configuration generated in multiscale sampling. A H₂O molecule that is originally hydrogen bonded to a COH adsorbate on a Pt(111) surface abstracts the proton from COH and deposits a second hydrogen on the Pt(111) surface. This bond breaking and forming event can be captured by AIMD but not with force field molecular dynamics (FFMD) unless a reactive force field is used. The initial configuration of H₂O molecules used in this AIMD simulation was generated using FFMD as described in this manuscript. [Please click here to view this video.](#) (Right-click to download.)

Catalytic Species	ΔE_{int} (eV)
COH	-0.70 ± 0.07
CO	-0.03 ± 0.03
CH ₂ OH	-0.64 ± 0.12
CHO-CHOH-CH ₂ OH	-0.93 ± 0.22
COH-COH-CH ₂ OH	-0.87 ± 0.23
COH-CHOH-COH	-1.72 ± 0.26
CHOH-COH-CO	-1.57 ± 0.25
CHO-CO-CO	-0.31 ± 0.19

Table 1: Water-catalytic species interaction energy results. Interaction energies in eV calculated for eight C_xH_yO_z adsorbates on Pt(111). Values reported are the averages taken over multiple configurations of liquid H₂O. The uncertainties are the standard deviations of the averages. 1 eV = 96.485 kJ/mol.

Discussion

The method as presented was selected for its ease of implementation, but multiple customizations could be made. For one, the force fields used in the FFMD simulations can be modified. Changing the force field parameters and/or potentials can be done by editing the LAMMPS input and data files. Similarly, solvents other than H₂O could be employed. To make this modification, the desired solvent molecule would need to be inserted starting from step 2.1.1, and the LAMMPS input files would need to be edited to incorporate the appropriate potentials and parameters. Inserting the new solvent molecule would also require supplying the internal coordinates of the solvent molecule in a .txt file analogous to the water.txt file.

Another modification that could be made is to modify the area of the surface slab. The results discussed in this manuscript employed 3 Pt x 3 Pt or 4 Pt x 4 Pt surface slabs, which have surface areas less than 120 Å². As the slab surface area increases, the computational expense also increases. Computational expense has the largest impact on section 5 of this protocol. If the data processing steps in section 5 become computationally prohibitive, big data post processing strategies such as those discussed in Li et al. 2018⁴⁵ can be employed.

Possible sources of uncertainty for this procedure include the force field employed, the sampling method, and the sampling frequency. The water structure is determined by the force field that is used, meaning that the choice of force field could influence the specific configurations of H₂O molecules. Our group has assessed how the choices of force field for H₂O molecules and Pt atoms influence the interaction energies calculated in FFMD and found that the choice of force field contributes less than 0.1 eV to this interaction energy. Another source of uncertainty is the sampling method, which influences the specific configurations that are used to calculate a quantity of interest. Our group has compared the performance of the "time sampling" method presented in this protocol with an "energy sampling" method, which is biased to lower energy configurations of H₂O molecules, on the interaction energies calculated in DFT and found both of these sampling methods give statistically equal values^{35,46}. The sampling frequency can also influence the results. We have assessed how increasing the number of configurations from 10 to 30,000 influences the average interaction energies calculated in FFMD for 40 different C₃H_xO₃ adsorbates and found that the sampling frequency contributes less than 0.1 eV to the average interaction energy⁴⁴.

The main limitation to this method is that the adsorbates are approximated by the structures under vacuum during the FFMD simulations. In reality, the adsorbates would exhibit conformational changes (bond stretches, angle bends, torsional motions, etc.) due to normal thermal movements, including interactions with solvent molecules. Attempts to include conformational changes of adsorbates into the FFMD simulations would require detailed development of force fields for catalytic surface adsorbates, i.e., which comprise terms that describe bond stretches, angle bends, and torsional terms, amongst others. As a future direction of this protocol, we are developing such force fields for adsorbates at solid surfaces, which we will use to determine the extent to which using rigid adsorbates influences the results.

Disclosures

The authors disclose no conflicts of interest.

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