

Nanobubble Nucleation and Dissolution Near the Anatase (101)–Water Interface

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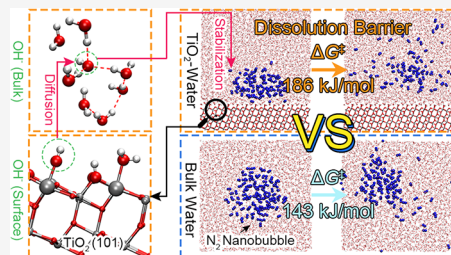


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ABSTRACT: In gas-involving (photo)electrochemical systems, nanoscale bubbles generate and enrich near the electrode–liquid interface, influencing interfacial transport and reactivity. However, it remains unclear how the solid–liquid interfacial microenvironment governs nanobubble evolution at the microscopic level. In this work, we perform deep potential molecular dynamics simulations with enhanced-sampling to investigate nucleation, dissolution, and detachment of nitrogen nanobubbles near the anatase (101)–water interface under neutral, acidic, and alkaline conditions. Our results show that the undercoordinated titanium and oxygen sites on the anatase (101) surface promote water dissociation, changing local ionic microenvironments. The resulting free hydroxide ions accumulate near the nanobubble surface, yielding a system-dependent negative zeta potential. The zeta potential of the nanobubble in the anatase–saline system is less negative than in other systems, due to the screening of locally paired sodium and chloride ions near the nanobubble surface. The dissolution barrier of nanobubbles shows a good linear positive correlation with the magnitude of zeta potential. This finding is further supported by the modeling with the Epstein–Plesset equation and the simulated bubble surface charge, as well as the experimental observations from nanoparticle tracking analysis and dynamic light scattering. The nucleation barriers are increased in systems with the anatase (101) surface compared to the anatase-free systems but are less sensitive to the acid–base strength. A significantly lower nucleation barrier in the anatase–saline system is attributed to the salting-out effect. The present study provides insights into nanobubble evolution near solid–liquid interfaces, with implications for bubble management in energy conversion systems.



INTRODUCTION

Bubble management has recently received special attention as an important approach of improving energy and chemical conversion systems.^{1,2} In both gas-evolving and gas-consuming (photo)electrochemical systems, bubbles that populate near the micro(nano)structured electrode surface may modify mass-transfer pathways, alter catalytic surface area, and reshape local ionic strength and pH, thereby altering efficiency and selectivity of (photo)electrochemical processes.^{1,2} Understanding bubble nucleation,^{3–5} growth,⁶ coalescence,^{7,8} detachment,⁹ and dissolution^{10–14} at the solid–liquid interfaces is therefore crucial for improving system performance.

Titanium dioxide (TiO₂), a prototypical oxide semiconductor, is one of the most widely used electrode materials in (photo)electrochemical systems, for example, the (photo)electrochemical nitrogen reduction reaction (NRR) to produce ammonia^{15,16} or water (photo)electrolysis to produce hydrogen.^{17,18} In practical TiO₂-based NRR systems, a continuous stream of N₂ is often bubbled near the electrode, inevitably forming solid–liquid–gas three-phase contact regions populated by micro- and nanoscale bubbles.¹⁹ In water electrolysis, gas bubbles are generated in situ at TiO₂-containing electrode surfaces. The adhesion of gas bubbles to the electrode surface reduces the electrochemically active area and hinders ion transport, leading to increased overpotentials and lower

efficiency.¹ These observations underscore the broader importance of understanding and controlling gas bubble dynamics at TiO₂ electrode–electrolyte interfaces. Therefore, in this work, we focus on the evolution dynamics of N₂ gas nanobubbles near the TiO₂–water interface.

The basic properties of gas bubbles, in particular the stability of nanobubbles, remain the subject of active discussion. Classical theory suggests short-lived nanobubbles. According to the Epstein–Plesset description, the nanobubbles dissolve on millisecond time scales due to rapid diffusive outflux driven by Laplace pressure.²⁰ However, in apparent contradiction to this expectation, both surface nanobubbles and bulk nanobubbles have been observed to persist for hours or longer under ambient conditions.^{3,21–28} For the stability of surface nanobubbles, experiments and theory have largely converged on a pinning–supersaturation mechanism, that is, the bubble is stabilized by the contact-line pinning in the condition of gas

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supersaturation.^{29–32} Relatively, bulk nanobubbles are less well understood, and their observed longevity is often associated with interfacial charging.^{25–27,33} A recurring observation is that nanobubble suspensions exhibit a negative zeta potential over a broad pH range from 4 to 12,^{25,27,34–36} and that the zeta potential becomes more negative with increasing pH.^{37,38} In addition, the stability of bulk nanobubbles is also sensitive to the electrolyte composition.¹¹ This sensitivity mainly originates from the dual role of electrolytes: electrolytes not only shift gas solubility in the bulk water but also regulate interfacial charge and mass transport at gas–water interfaces. From a thermodynamic perspective, electrolytes often reduce gas solubility via the salting-out effect,^{39,40} which can increase the degree of supersaturation at a fixed dissolved-gas content and thereby favor bubble nucleation. In addition, recent studies on bulk nanobubbles suggested that nanobubble surface charges arising from ion adsorption can slow gas diffusion across the bubble surface,^{27,33,41} thereby enhancing nanobubble stability.

It is well recognized that the TiO₂–water interface is not electrostatically inert. A body of evidence has shown that the TiO₂ surface carries a pH-dependent surface charge through amphoteric hydroxylation reactions.^{42–46} Molecular simulations at atomic levels^{42–44} and surface-specific experiments^{45,46} further revealed that TiO₂ surfaces in contact with water typically exhibit mixed adsorption of water molecules and ions.

Thus, it is natural to ask how nanobubbles nucleate, evolve, detach, and ultimately dissolve in the vicinity of a charged TiO₂–water interface. In particular, it remains unclear how local ionic environments shaped by the TiO₂–water interface chemistry feed back on the nucleation barrier and the dissolution kinetics of nanobubbles.

Addressing this question requires a framework that can treat gas–liquid interfacial physics and solid–liquid interfacial chemistry on the same footing. Although experiments can reveal variations of bubble populations with electrolyte composition and pH, they do not readily provide molecular-level insights into the complex mechanisms that control nanobubble evolution.^{26–28,47} In this regard, complementary molecular simulations are essential for identifying microscopic evolution pathways and for quantifying the underlying thermodynamic driving forces.

Molecular dynamics (MD) simulations with classical force fields, including coarse-grained models, have been widely used to study the nanobubble behavior.^{3,7,8,10–14,28,48} However, empirical force fields are not designed to reliably describe bond-breaking and bond-forming processes that are central to oxide–water interface reactions, such as interfacial water dissociation and proton transfer via the Grotthuss mechanism.^{49–51} Recently, machine learning interatomic potentials have provided a promising solution. In particular, the deep potential (DP) molecular dynamics (DPMD) approach^{52,53} offers a near-*ab initio* accuracy at a computational cost closer to the classical MD, allowing chemically faithful simulations of aqueous oxide surfaces. Nevertheless, generally, direct DPMD trajectory simulations are hardly able to capture events like nanobubble nucleation, dissolution, and detachment, because these processes are activated events that occur too rarely on the time scales accessible to be observed by unbiased simulations. Enhanced sampling techniques^{54–57} are therefore required to simulate these processes and to reconstruct the corresponding free energy landscape.

In this work, we investigate the evolution of N₂ nanobubble near the anatase TiO₂ (101)–water interface (hereafter referred to as the TiO₂–water interface) primarily through enhanced-sampling DPMD simulations, combined with experimental validation. Based on this combined investigation, we quantify the thermodynamic and kinetic information associated with nanobubble nucleation, dissolution, and detachment near the TiO₂ surface in various environments, including pure water and acidic, basic, and salt solutions. The present study links the interfacial ionic microenvironment to dynamically controlled nanobubble evolution and provides a guideline for regulating gas accumulation and release near TiO₂–water interfaces in gas-involving electrochemical environments.

METHODS

Deep Potential Model Training

Considering that the energy component of the deep potential model depends on local atomic environments, it is crucial to construct representative subsystems (Table S1 in the Supporting Information) that capture the relevant structural diversity. Part of the configurations used in this study were adapted from previously developed data sets, including N₂ molecules in bulk water containing self-ions^{33,51} and configurations representing the TiO₂–water interface.⁴⁴ To accurately describe the dynamic evolution of N₂ nanobubbles near the TiO₂–water interface, additional configurations were generated. These new data incorporate the local atomic environments of N₂ molecules in neutral, acidic, and basic aqueous solutions adjacent to the TiO₂ surface.

The final DP model was trained using the DeePMD-kit package^{53,58} based on a data set containing 144,215 configurations (Figure S1a) and validated against an independent test set. All reference energies and forces were obtained from SCAN⁵⁹ functional-based DFT calculations performed in VASP.^{60,61} To enable efficient large-scale molecular dynamics simulations, the trained DP model was further compressed.⁶² Root-mean-square errors (RMSEs) of the compressed⁶² DP model are 1.21 meV/atom for energies and 43 meV/Å for forces relative to the SCAN results (Figure S1b,c), which confirm its accuracy.

Enhanced Sampling

The collective variable (CV) based umbrella sampling (US)⁶³ or on-the-fly probability enhanced sampling (OPES)^{64–66} method was used in DPMD simulations with the PLUMED package⁶⁷ to investigate the dynamics of nanobubble, including the formation, growth, and detachment from the TiO₂ surface. We designed two CVs to capture the slow collective motions governing the N₂ nanobubble evolution. Specifically, we used CV Δz to describe the vertical distance between the nanobubble centroid and the TiO₂ surface in US simulations and CV s to characterize the formation and growth of N₂ molecular cluster in OPES simulations.⁶⁸

This CV s is formulated based on an adjacency matrix $\mathbf{M}$, in which each matrix element M_{ij} quantifies the degree of connection between particles. Here, each particle corresponds to the centroid of a N₂ molecule, which is defined as the midpoint between its two nitrogen atoms. The CV s is then expressed as the total coordination number of all centroids that belong to the largest connected molecular cluster:

$$s = \sum_{i=1}^{N_l} \sum_{j=1}^N M_{ij} \quad (1)$$

where the first summation runs over the N_l centroids that form the largest cluster, and the second summation extends over all N centroids present in the system.

DPMD Simulation

Deep potential molecular dynamics simulations with the US or OPES method were performed using the LAMMPS package⁶⁹ to investigate

the dynamic behavior of N_2 nanobubbles near the anatase (101)–water interface in neutral, acidic, and alkaline aqueous environments. For simulations of the TiO_2 –pure water system, an initial nanobubble containing 102 N_2 molecules was embedded in a periodic simulation box containing 7,401(16,144) H_2O molecules and an anatase $Ti_{1920}O_{3840}$ slab. The systems TiO_2 –NaOH, TiO_2 –HCl, and TiO_2 –NaCl were constructed by introducing 0.24 M solutes into the aqueous phase of the TiO_2 –(pure) water system. For comparison and validation, DPMD simulations were also conducted in bulk Water and NaOH solution systems without the TiO_2 slab. The initial atomic arrangements were generated using PACKMOL,⁷⁰ and a summary of all simulation systems is provided in Table S2.

All simulations were carried out in the isothermal–isobaric (NPT) ensemble at a temperature of 330 K and a pressure of 1 atm. The simulation temperature was set 30 K above ambient, because prior studies showed this slight temperature rise in SCAN-based molecular simulations helped to approach the liquid water properties at 300 K.⁷¹ Temperature and pressure were controlled using the Nosé–Hoover thermostat and barostat,^{72,73} with damping constants of 0.1 and 1 ps, respectively. Independent trajectories were generated for each system, and each trajectory was run over at least 10 ns with a time step of 1 fs.

Further details of the simulations can be found in the Supporting Information (SI).

Experimental Details

Three types of aqueous samples were prepared in this study: (i) bulk nanobubble suspensions, (ii) TiO_2 nanoparticle suspensions, and (iii) bulk nanobubble suspensions first prepared and then mixed with TiO_2 nanoparticles. Bulk nanobubbles were generated using an ultrasonic cavitation system (VCX 750 W, Sonics & Materials, USA), with each batch prepared at a volume of 80 mL. The ultrasonication duration and total energy input were maintained at 20 min and 18.5 kJ, respectively. Anatase-phase TiO_2 nanoparticles (purity $\geq 99.99\%$ metals basis, nominal size ≤ 50 nm) were purchased from Shanghai Aladdin Biochemical Technology Co., Ltd. The TiO_2 concentration was fixed at 5 $\mu g/mL$ for suspensions prepared in both ultrapure water and nanobubble water. After generating the nanobubble suspension, the nanoparticle dispersion was added and mixed to obtain the final nanobubble–nanoparticle mixture for subsequent characterization.

Nanobubble size distributions were determined using nanoparticle tracking analysis, implemented via a custom dark-field imaging system built around an Olympus IX73 inverted microscope.⁷⁴ A 20 $\times$ objective lens (depth of field: 6.526 μm) and a CMOS camera (Basler acA2440-35uc, 2448 $\times$ 2048 pixels; spatial resolution: 172.5 nm/pixel; 30 frames per second) were used for image and video acquisition. A 532 nm continuous-wave laser (SUM-EMERP-3, China) was coupled through a prism-edged optical flat to illuminate nanosubstances within a quartz visualized chamber. For each sample, at least three video sequences (≥ 45 s each) were recorded. Number concentrations of nanobubbles were estimated by counting scattering features in the recorded frames, and particle trajectories were analyzed using the NanoTrackJ plugin in ImageJ/Fiji to calculate size distribution from diffusion-based tracking.^{28,75} In the mixed nanobubble and TiO_2 system, nanoparticle tracking analysis shows number concentration and size distribution of dispersed objects. Since NTA cannot distinguish nanobubbles from solid particles, these results should be interpreted as indicating an overall characterization of the dispersed population.

The pH values of samples were measured using a multiparameter system (S975-uMix, Mettler Toledo, Switzerland). Zeta potential of aqueous samples was measured based on the electrophoretic mobility with a ZEN3700 Zetasizer Nano ZSE instrument (Malvern Instruments, UK). Before each measurement, the suspension was injected into a clean U-shaped capillary cell and equilibrated at room temperature.⁷⁶ For each sample, six replicate measurements were performed, and the standard deviation was calculated from the resulting data.

RESULTS AND DISCUSSION

Propensity of Nanobubble toward the Anatase (101)–Water Interface

Since the present study focuses on the dynamic behavior of a single N_2 nanobubble near the anatase (101)–water interface, we first investigate the preference of the nanobubble toward the interface. Figure 1 presents the free energy surface (FES)

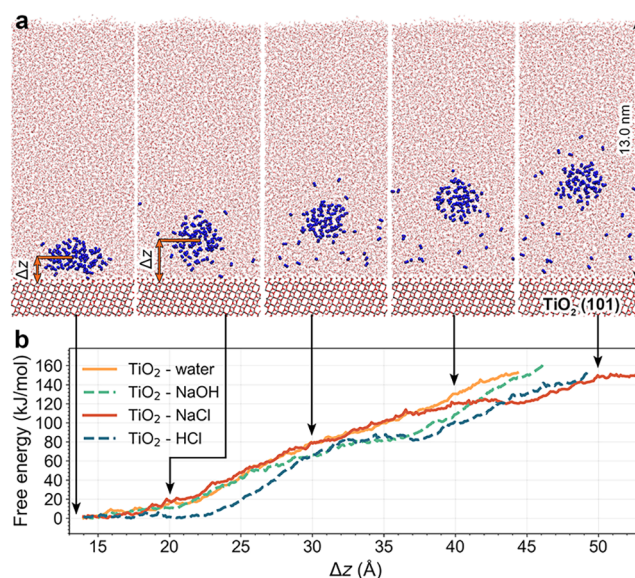


Figure 1. Nanobubble detachment dynamics and free energy surfaces. (a) Representative snapshots showing the detachment of an N_2 nanobubble from the TiO_2 surface into the solution. (b) One-dimensional free energy profiles of nanobubble detachment under different solution conditions, projected onto the CV Δz , which represents the vertical distance between the nanobubble centroid and the TiO_2 surface.

of a N_2 nanobubble (about 3 nm in diameter) departure from the TiO_2 surface toward the aqueous phase as a function of the normal distance to the surface (z), as well as representative snapshots of the dynamic process, which show that the nanobubble remains compact when its center of mass is pulled upward (Figure 1a). The free energies of the systems increase nearly monotonically from $z \approx 14$ Å (the approximate position of the TiO_2 –water interface) to $z \approx 45$ –50 Å, and the increases of the free energies are about 150 kJ/mol (Figure 1b). This monotonic rise indicates that for the nanobubble, the interfacial adsorption state is thermodynamically preferred, and the detachment into the bulk is strongly disfavored. Notably, the profiles show weak sensitivity to the solution. The four investigated systems (TiO_2 –water, TiO_2 –NaOH, TiO_2 –HCl, and TiO_2 –NaCl as listed in Table S2), display similar FESs (including both trend and magnitude) along z . Because these distinct solution environments can affect the interfacial charge and water structure differently, the observed similarity of energy profiles suggests that the detachment thermodynamics is mainly dominated by generic interfacial stabilization; that is, the free energy gain is obtained through the interaction of the nanobubble with the interface and is less affected by specific ionic conditions.

In addition, we observe an interfacial water film with a thickness of only a few angstroms that separates the nanobubble from the TiO_2 surface at the minimum of free

energy (zero point of free energy) in Figure 1. This is due to the hydrophilic nature of the anatase (101) surface, and with the approaching of nanobubble toward the surface, the gradually formed ultrathin water film between the bubble and the surface can be sustained by a disjoining pressure arising from van der Waals interactions across the film.⁷⁷ The water film prevents the nanobubble from direct contact with the TiO₂ surface, that is, avoids the formation of surface nanobubble.

Note that although we observed a strong preference of the nanobubble staying near the interface, this does not mean that bubble detachment cannot occur. The nanobubble addressed here may be too small for buoyancy to provide a substantial driving force, and so the departure is not expected to follow the same pathways as larger bubbles. In practical (photo)-electrochemical systems, bubble detachment from the electrode surface is commonly observed only after bubbles have grown or coalesced to larger sizes (micrometer scale), and the detachment of micrometer-size bubbles is promoted by macroscopic driving forces, including buoyancy, shear, and Marangoni stresses,^{1,9} which largely determine the diameters of the detached bubbles and release frequencies.¹

Evolution of the N₂ Nanobubble near the Anatase (101)–Water Interface

We now examine the nucleation and dissolution dynamics of N₂ nanobubble near the TiO₂–water/electrolyte interface. Figure 2 shows one-dimensional FES profiles of nanobubbles in various environments projected along the CV *s*. For aqueous systems without the TiO₂ surface (Figure 2a), the profiles clearly resolve three characteristic states of nanobubble along *s*: A fully dissolved state (labeled as DS) at *s* ≈ 0, a critical nucleus (labeled as CN) at 120 < *s* < 220, in which a small nanobubble coexists with dissolved gas, and an intact nanobubble state (labeled as IS) at *s* > 400. The representative snapshots of the three states are also provided in Figure 2a. The pure water system yields a nucleation barrier of 141 kJ/mol and a dissolution barrier of 144 kJ/mol. The NaOH solution shows a lower nucleation barrier of 108 kJ/mol and a higher dissolution barrier of 191 kJ/mol. Furthermore, the nucleation free energy (ΔF_{IS-DS}), defined as the free energy difference between the fully dissolved and intact bubble states, changes significantly from about −3 kJ/mol in Water to about −83 kJ/mol in NaOH (Figure 2a and Table 1). These results indicate that not only do alkaline conditions dynamically promote the formation of nanobubble by reducing the activation free energy barrier, but they also stabilize the nanobubble better from a thermodynamic perspective. This agrees with the experimental observation of faster nucleation and greater persistence of nanobubbles under basic conditions.⁷⁸ The present bulk water results are also consistent with our previous unbiased DPMD simulations.³³ Specifically, the unbiased trajectories reproduced the qualitative pH-dependent evolution trend, in which nanobubbles show limited dissolution and enhanced persistence under alkaline conditions, whereas acidic conditions promote dissolution. The present enhanced-sampling simulations further provide a quantitative characterization in free energy along the whole evolution process that is not accessible from unbiased runs, by resolving the critical nucleus state and quantitatively determining both the nucleation and dissolution barriers.

Figure 2b presents the FES plots of nanobubbles near the TiO₂–water/electrolyte interfaces and the associated snap-

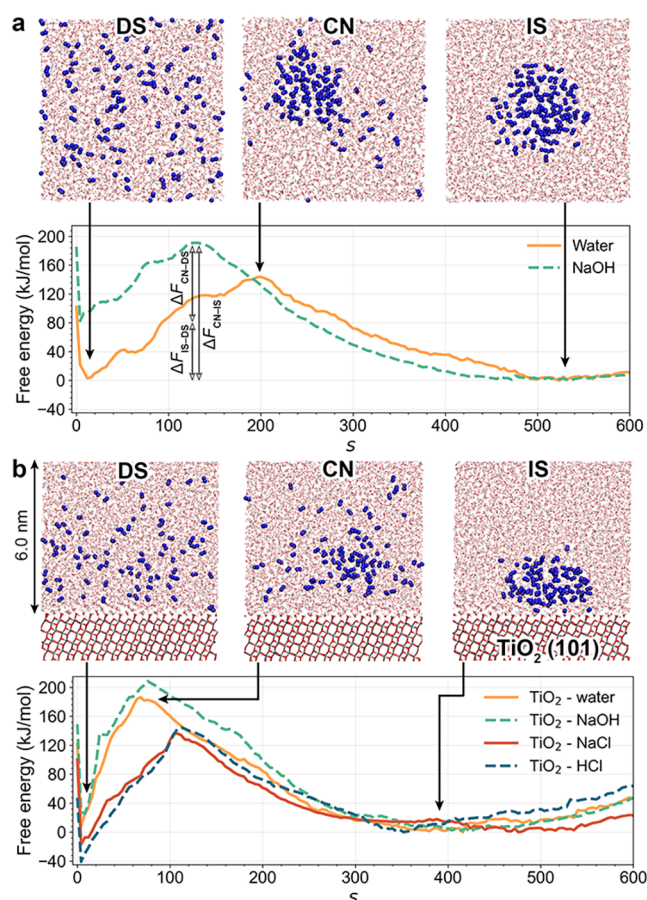


Figure 2. Dynamic snapshots and free energy surfaces of the N₂ nanobubble nucleation and dissolution. Representative snapshots and corresponding one-dimensional free energy profiles of the N₂ nanobubble along the CV *s* are presented for two systems: (a) aqueous systems without TiO₂, and (b) the TiO₂-containing systems. The snapshots illustrate three characteristic stages of nanobubble evolution: the fully dissolved state (DS), the critical nucleus (CN), and the intact state (IS). The nucleation free energy (ΔF_{IS-DS}), nucleation barrier (ΔF_{CN-DS}), and dissolution barrier (ΔF_{CN-IS}) are illustrated with the NaOH aqueous system as a representative example for clarity. Specific values are given in Table 1.

Table 1. Nucleation Free Energy (ΔF_{IS-DS}), Nucleation Barrier (ΔF_{CN-DS}), and Dissolution Barrier (ΔF_{CN-IS}) of Nanobubble Evolution

system	free energy (kJ/mol)		
	nucleation free energy	nucleation barrier	dissolution barrier
TiO ₂ –water	−8	178	186
TiO ₂ –NaOH	−25	183	208
TiO ₂ –HCl	40	185	145
TiO ₂ –NaCl	17	153	136
water	−3	141	144
NaOH	−83	108	191

shots. Near the TiO₂–water interface, the fully dissolved state of nanobubble is also less stable than the intact bubble state with a free energy difference of 8 kJ/mol, and the barriers are 178 kJ/mol for nucleation and 186 kJ/mol for dissolution, both larger than those in bulk Water. Near the TiO₂–NaOH interface, the same relative stability is observed, with a nucleation free energy of −25 kJ/mol, a nucleation barrier of

183 kJ/mol, and a dissolution barrier of 208 kJ/mol. These results indicate that in both systems, the interaction of nanobubble with the interface reduces the thermodynamic stability of nanobubble and slows the dynamic interconversion between the dissolved state and intact nanobubble state. Unlike the four systems mentioned above, the TiO_2 –HCl and TiO_2 –NaCl interface systems favor the dissolved state both thermodynamically and dynamically. The nucleation free energies are 40 and 17 kJ/mol, respectively; the dissolution barriers are 145 and 136 kJ/mol, respectively.

For convenience of comparison, Table 1 further tabulates the nucleation free energies, the nucleation free energy barriers, and the dissolution free energy barriers of all these investigated systems. We notice that the nucleation barriers are larger in all systems containing TiO_2 than in the TiO_2 -free systems, which is consistent with the point of view that hydrophilic surfaces suppress heterogeneous bubble nucleation.⁷⁹ Across the TiO_2 -containing systems, the nucleation barriers follow the order of TiO_2 –HCl $\gtrsim$ TiO_2 –NaOH $\gtrsim$ TiO_2 –water > TiO_2 –NaCl, whereas the dissolution barriers follow TiO_2 –NaOH > TiO_2 –water > TiO_2 –HCl $\gtrsim$ TiO_2 –NaCl. Thus, relatively, TiO_2 –NaOH interface stabilizes the nanobubbles both kinetically and thermodynamically, as evidenced by a higher dissolution free energy barrier and a more negative nucleation free energy. In contrast, chloride-containing interface environments thermodynamically favor the dissolved state and lower the free energy barrier for nanobubble dissolution.

TiO_2 Surface-Modulated Microenvironments

To understand the different thermodynamics and dynamics of nanobubble nucleation and dissolution observed in the investigated systems, we investigated the microscopic structures near the anatase (101) surface in various solutions, in particular focusing on the formation and distribution of ions at the anatase (101)–water/electrolyte interface.

As illustrated in Figure 3a, undercoordinated atoms on the anatase (101) surface, specifically 5-fold coordinated titanium (Ti_{5c}) and 2-fold coordinated oxygen (O_{2c}), can adsorb water molecules and facilitate their dissociation, leading to the coadsorption of OH^- and H^+ on the surface. Figure 3a(I) shows that the yielded OH^- and H^+ ions are bound to Ti_{5c} and O_{2c} sites, respectively. Previous studies^{42–44} have described this process as a concerted, hydrogen-bond-mediated proton-transfer pathway that proceeds via a Grotthuss mechanism.^{49,50}

Although water dissociation on the anatase (101) surface is well established, our simulations observed the presence of OH^- ions in the bulk water of the nominally TiO_2 –pure water system. This observation motivates a mechanistic analysis of how bulk OH^- is generated from interfacial reactions.

Figure 3b summarizes the species transition matrix in the TiO_2 –water system among adsorbed water molecules (H_2O (surf)) and adsorbed OH^- ions (OH^- (surf)) on the TiO_2 surface and free OH^- ions in bulk water (OH^- (bulk)), where the species identity is tracked by the oxygen index (see more details about the transition matrix in the SI). The matrix shows that all three species largely preserve their identities with probabilities exceeding 0.97, indicating that the interfacial speciation is stable. A small but non-negligible fraction (0.028) of surface hydroxide oxygens are protonated and reassigned from OH^- (surf) to H_2O (surf), which signals proton transfer at the interface (Figure 3c). Decomposing this 0.028 probability reveals two distinct PT pathways. In the first pathway, denoted PT (surf), a neighboring H_2O (surf) donates

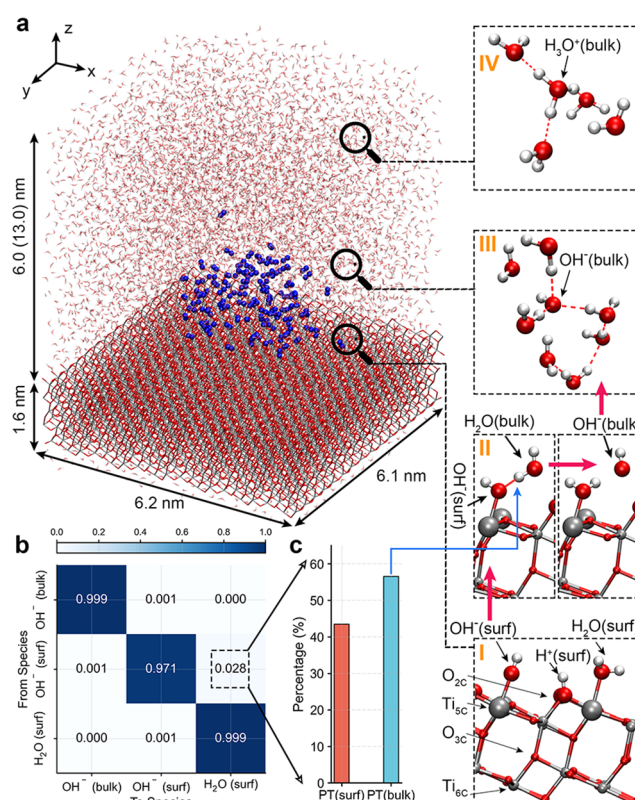


Figure 3. Schematic structure of the TiO_2 system and ion transition mechanism. (a) Three-dimensional schematic of an N_2 nanobubble located near the anatase (101)–water interface. The magnified subfigures from bottom to top illustrate: (I) the interfacial region contains undercoordinated surface Ti_{5c} and O_{2c} atoms that adsorb OH^- , H^+ , and H_2O species, whereas Ti_{6c} and O_{3c} atoms are the fully coordinated atoms in the inner layer of TiO_2 . (II) the proton-transfer process from H_2O (bulk) to OH^- (surf), resulting in the exchange of their identities to OH^- (bulk) and H_2O (surf); (III) OH^- and (IV) H_3O^+ ions in the bulk region and their surrounding water molecules. (b) Visualization of the transition matrix describing the conversion probabilities among H_2O (surf), OH^- (surf), and OH^- (bulk) species in the TiO_2 –water system. The color bar indicates the normalized transition probabilities, where each row is normalized independently. (c) Schematic illustration of the proton-transfer (PT) pathways from OH^- (surf) to H_2O (surf). PT (surf) and PT (bulk) represent proton transfers involving H_2O molecules located at the surface and in the bulk region, respectively.

a proton to the OH^- (surf), exchanging the adsorption sites of OH^- (surf) and H_2O (surf). In the second pathway, denoted PT (bulk), the OH^- (surf) abstracts the proton of a nearby free water molecule that is unbound to the TiO_2 surface. This step neutralizes the surface OH^- (surf) to form a H_2O (surf) and generates a free OH^- (Figure 3a(II)). Once generated, the free OH^- can migrate away from the surface region toward the bulk water along hydrogen-bonded water networks through a Grotthuss-like mechanism, as illustrated in Figure 3a(III). This PT (bulk) channel therefore provides a direct route for producing bulk OH^- . H_3O^+ ions are also occasionally detected in the bulk (Figure 3a(IV)) of the pure water system, which are mainly from the self-ionization reaction of bulk water.

When the production and interconversion of ions in the interface reach a dynamic steady state, the overall ionic composition becomes stable. Figure 4a shows the equilibrium ionic concentrations of the investigated systems, as well as

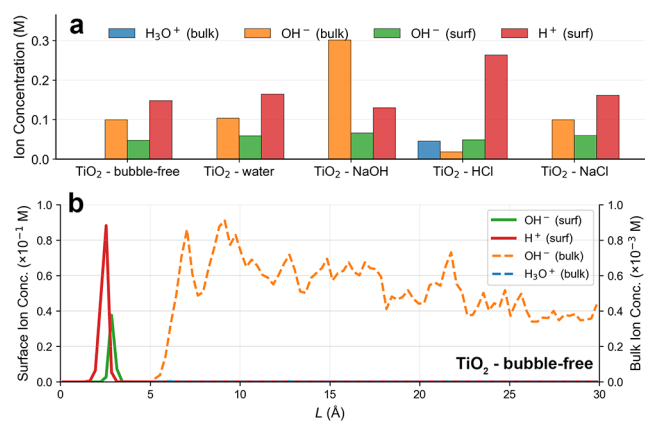


Figure 4. Concentration distribution of ions. (a) Concentration profiles of H^+ , H_3O^+ and OH^- ions in the five TiO_2 -containing systems. (b) Concentration profiles of H^+ , H_3O^+ , and OH^- ions as a function of the vertical distance $L = z_{\text{ion}} - z_{\text{surface}}$ from the TiO_2 surface of the TiO_2 -bubble-free system, where z_{surface} represents the mean position of the surface O_{2c} sites.

those of a bubble-free TiO_2 -water system (which removes the nanobubble from the TiO_2 -water system and will be called TiO_2 -bubble-free) for comparison. The TiO_2 -bubble-free, TiO_2 -water and TiO_2 -NaCl systems exhibit nearly identical ionic concentrations, each containing approximately 0.15–0.16 M of surface H^+ , 0.05–0.06 M of surface OH^- , and 0.10 M of bulk OH^- , while bulk H_3O^+ remains negligible. It indicates that the presence of nanobubble has negligible influence on the water dissociation reactions on the anatase (101) surface, and the Na^+ and Cl^- ions have no influence either. For the electrolyte-containing TiO_2 systems, the initial solute concentration is 0.24 M. The anatase (101) surface changes the acid–base property of solutions, making the neutral aqueous solution become alkaline. In the TiO_2 -NaOH system, the alkaline environment markedly enhances the OH^- concentration to about 0.30 M in the bulk and 0.07 M on the TiO_2 surface, accompanied by a slight reduction of surface H^+ to 0.13 M. In the acidic TiO_2 -HCl system, the surface H^+ concentration increases substantially to 0.26 M, along with a low bulk H_3O^+ concentration of 0.05 M and a low surface OH^- concentration of 0.05 M. Interestingly, a very low concentration of bulk OH^- (0.02 M) is observed in the acidic system, and these bulk OH^- ions are generated by the PT (bulk) reaction mentioned earlier.

We also calculated the depth-resolved concentration profiles to further characterize the spatial distribution of ions, along the normal distance (L) from the TiO_2 surface for the bubble-free system (Figure 4b). The H^+ (surf) concentration reaches a maximum at $L \approx 2.5$ Å, while the OH^- (surf) concentration exhibits a prominent peak around 2.9 Å. OH^- (bulk) ions are predominantly distributed at $L > 5$ Å, whereas bulk H_3O^+ remains nearly absent.

Taken together, these findings indicate that the presence of anatase (101) surface changes ionic microenvironments, making the TiO_2 -water and TiO_2 -NaCl systems alkaline, the TiO_2 -NaOH system more alkaline, and the TiO_2 -HCl system close to neutral. The experimental pH results qualitatively support this conclusion. Pure water and water with bulk nanobubbles have nearly identical pH values of 5.960 and 5.958. After adding TiO_2 nanoparticles, the pH increases to 7.097 for TiO_2 suspensions and to 6.687 for TiO_2 and

nanobubble suspensions. This restructuring of ionic microenvironments and change of the acid–base equilibrium might influence nanobubble nucleation and dissolution processes.

Ion-Modulated Nanobubble Evolution

The TiO_2 surface-reconstructed ionic microenvironments have a direct influence on the ionic distribution near the nanobubble surface and consequently affect the stability and evolution of nanobubble.

Figure 5 reports volume-normalized number-density profiles of bulk H_3O^+ , OH^- , Na^+ , and Cl^- as a function of the normal

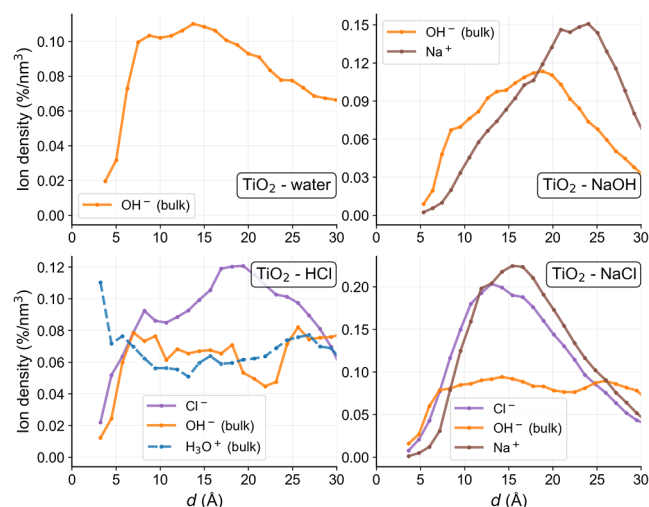


Figure 5. Spatial distribution of ions surrounding the nanobubble in the TiO_2 -containing systems. Volume-normalized number density profiles of bulk H_3O^+ , OH^- , Na^+ , and Cl^- ions are shown as a function of the radial distance (d) from the N_2 nanobubble surface. Each profile is volume-normalized by averaging over concentric spherical shells of increasing mean radius from the nanobubble center.

distance d from the N_2 nanobubble surface. Here, d is defined as the minimum distance between the ionic center of mass and the outermost nitrogen atoms of the largest N_2 cluster. The cluster is identified through a connectivity analysis.

Introducing TiO_2 into pure water creates an alkaline microenvironment for the nanobubble, so that in the TiO_2 -water system, a negatively charged shell forms near the nanobubble surface with bulk OH^- as the dominant ions, peaking at $d \approx 9$ –16 Å. Compared with TiO_2 -water, the bulk OH^- distribution peak shifts to larger distances ($d \approx 19$ Å) in TiO_2 -NaOH, which may arise from electrostatic attraction between OH^- and Na^+ , given that Na^+ ions mainly enrich at $d \approx 24$ Å. In the region of $d \approx 5$ –18 Å, the number density of OH^- remains higher than that of Na^+ . This OH^- -rich shell yields a negative interfacial potential and stabilizes the local hydrogen-bond network,³³ suppressing molecular exchange across the boundary and slowing gas outflux. Based on the Debye–Hückel approximation,^{51,80} the cumulative zeta potentials ζ at a distance of $d = 2$ nm from the N_2 nanobubble surface for the TiO_2 -NaOH and TiO_2 -water systems are -4.0 and -2.7 mV, respectively. It well explains the larger dissolution barrier of nanobubble in TiO_2 -NaOH (Table 1) than in TiO_2 -water, and these results are consistent with prior simulations³³ and experimental trends^{25,27,41,81} that nanobubbles are more persistent under more alkaline conditions.

Similar to the TiO_2 -water system, there are some bulk OH^- ions in TiO_2 -NaCl systems produced by the surface reaction

of TiO_2 which also show a distribution peak at $d \approx 9\text{--}16$ Å (Figure 5). The key differences in nanobubble stability and evolution between these two neutral TiO_2 systems are therefore the presence of NaCl. Figure 5 shows that the Na^+ and Cl^- in $\text{TiO}_2\text{--NaCl}$ system mainly populate near the nanobubble with distinct peaks at $d \approx 15$ Å for Na^+ and at $d \approx 13$ Å for Cl^- . This ionic layer of Na^+ and Cl^- spanning $d \approx 9\text{--}16$ Å screens the interfacial charges,^{41,82} weakens the net interfacial potential with a zeta potential ζ (-0.6 mV) less negative than that in the $\text{TiO}_2\text{--water}$ system, leading to a remarkably lower dissolution barrier (Table 1). In addition, the elevated local ionic strength induces salting-out and lowers gas solubility,^{39,40} which favors nanobubble nucleation and explains the much lower nucleation barrier (approximately 30 kJ/mol lower) than those of the other TiO_2 -containing systems.

In the $\text{TiO}_2\text{--HCl}$ system, the maximum H_3O^+ number density appears at $d \approx 3\text{--}5$ Å (Figure 5), consistent with our previous observation.^{33,51} This interfacial enrichment of H_3O^+ perturbs the local hydrogen-bond network and thereby promotes nanobubble dissolution.³³ For $d > 5$ Å, bulk H_3O^+ and OH^- are both less abundant and nearly compensate for each other. The Cl^- ions are also populated near the nanobubble surface with a relatively wide distribution at $d \approx 5\text{--}30$ Å (Figure 5), which neutralizes the weak positive charges of the interfacial accumulated H_3O^+ , leading to a zeta potential ζ of -1.6 mV. Accordingly, as shown in the free-energy profile (Figure 2b), the dissolved state of nanobubble in the $\text{TiO}_2\text{--HCl}$ system is remarkably preferred, as well as with a lower dissolution barrier.

Figure 6a clearly shows the linear correlation between the cumulative zeta potentials ζ and the dissolution barriers ΔF of nanobubbles in all the TiO_2 -containing systems. In particular, a more negative zeta potential corresponds to a higher barrier

for nanobubble decay. These results support that bulk OH^- and H_3O^+ exert dominant yet opposite effects on nanobubble evolution: OH^- stabilizes the bubble, whereas H_3O^+ promotes dissolution.³³ By contrast, Na^+ and Cl^- can cooperatively influence the evolution dynamics by tuning the interfacial electrostatic environment and by the salting-out effect. Such cooperative behavior is not expected when only one of Na^+ or Cl^- is present, as in alkaline or acidic systems. In the $\text{TiO}_2\text{--NaCl}$ system, the number-density profiles of Na^+ and Cl^- are similar, presenting local $\text{Na}^+\text{--Cl}^-$ pairing around the nanobubble. The ion-pair separations (~ 2.8 to 4.5 Å) are energetically favorable,⁸³ which maintains spatial correlation between cations and anions near the interface and enhances charge screening. This behavior differs from $\text{TiO}_2\text{--NaOH}$, where Na^+ is mainly distributed at $d > 20$ Å, and from $\text{TiO}_2\text{--HCl}$, where Cl^- is more broadly distributed (Figure 5).

We further analyzed the structural characteristics of the critical nucleus (CN) state connecting the intact bubble and the dissolved state by quantifying the fraction of N_2 molecules in the CN ensemble for the TiO_2 -containing systems, and the results are shown in Figure 6b. For $\text{TiO}_2\text{--water}$, the CN distribution peaks at $\sim 43\%$ of the total N_2 population with a probability of ~ 0.09 . In $\text{TiO}_2\text{--NaOH}$, the peak shifts slightly to $\sim 45\%$, and the probability increases to ~ 0.12 . By contrast, the CN peaks in $\text{TiO}_2\text{--HCl}$ and $\text{TiO}_2\text{--NaCl}$ correspond to a larger cluster fraction ($\sim 65\%$), accompanied by lower probabilities (~ 0.06) and broader distributions (Figure 6b). These CN characteristics are consistent with the order of dissolution barriers. The smaller sizes of CN in $\text{TiO}_2\text{--water}$ and $\text{TiO}_2\text{--NaOH}$ correspond to a higher free-energy cost for barrier crossing.

Classical Kinetic Modeling

The enhanced-sampling DPMD simulations and the above analysis revealed that interfacial charges (zeta potential) still play a dominant role in the evolution of nanobubble near the TiO_2 surface, which are regulated by the TiO_2 surface reactions in various solutions and the ionic propensity toward the nanobubble surface. We therefore investigated the time evolution of nanobubble dissolution by a classical kinetic modeling using the Epstein–Plesset relation²⁰ and the interfacial charge (Q) of the bubble extracted from the DPMD trajectory simulations for bulk water and $\text{TiO}_2\text{--water}$ systems. Here, the use of the Epstein–Plesset relation is justified by the fact that the hydrophilic TiO_2 surface is separated from the nanobubble by a thin interfacial water film, so that the bubble remains in a bulk-like liquid environment.

Following our earlier derivation,³³ the dissolution rate is estimated as

$$\frac{dR}{dt} = \frac{D}{\rho}(c_1 - Hp_2)\left(\frac{1}{R} + \frac{1}{\sqrt{\pi Dt}}\right) \quad (2)$$

where R is the nanobubble radius, t is time, D is the diffusivity of N_2 in water, ρ is the gas density inside the nanobubble, c_1 is the dissolved N_2 concentration in the liquid, H is the Henry coefficient, and p_2 is the gas pressure inside the bubble obtained from mechanical equilibrium:⁸⁰

$$p_1 + \frac{2\gamma}{R} = p_2 + \frac{Q^2}{32\pi^2\epsilon R^4} \quad (3)$$

where p_1 is the liquid pressure, γ is the surface tension, ϵ is the permittivity of the medium, and Q , taken as a trajectory-

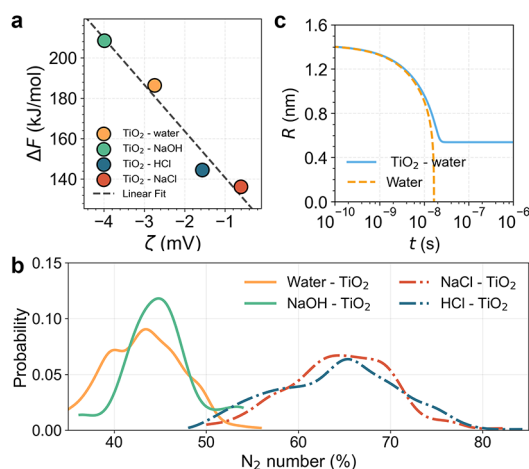


Figure 6. Zeta potential, critical nucleus ensemble, and time-evolution of the N_2 nanobubble in different systems. (a) Correlation between the Gibbs free energy barrier (ΔF) of dissolution and the cumulative zeta potential (ζ) evaluated at a distance of $d = 2$ nm from the N_2 nanobubble surface. The dashed line indicates the linear fit. (b) Probability distribution of the fraction of N_2 molecules within the critical nucleus ensemble of the nanobubble. Configurations included in this ensemble were identified based on the value of the CV s corresponding to the critical nucleus shown in Figure 2. (c) Theoretical dynamic evolution of N_2 nanobubble radius R with time in water with(out) TiO_2 .

averaged value, is the net ionic charge within $d = 2$ nm of the nanobubble surface.

The predicted evolution results are shown in Figure 6c. In bulk Water, the radius of the nanobubble decreases monotonically with time and vanishes at ~ 16 ns. In TiO_2 –water, the decay is slower and reaches an arrested state at ~ 32 ns, with a finite radius of ~ 0.54 nm (a six- N_2 cluster). This kinetic modeling result is qualitatively consistent with the enhanced-sampling DPMD simulations, in the sense that both approaches show nanobubble dissolution to be less favorable in TiO_2 –water than in bulk water. The DPMD simulations predict a larger dissolution barrier (Figure 2b), whereas the kinetic modeling gives a slower radius decay. The good agreement between the modeling which introduces the electrostatic contribution (Equation 3) and the DPMD simulation supports well the dominant role of interfacial charge in stabilizing nanobubble near the TiO_2 surface, which is mainly from bulk OH^- produced via the TiO_2 surface reactions and reduces the driving force for gas outflux.

Experimental Validation of the Anatase-Surface Effect on Nanobubble Evolution

We further experimentally validated the effect of the anatase-surface on nanobubble evolution. Figure 7a,b present representative dark-field snapshots showing the scattering features of bulk nanobubbles in pure water and of the mixed system containing both nanobubbles and anatase TiO_2 nanoparticles, respectively. As quantified in Figure 7c, the equilibrium number concentrations of bulk nanobubbles and TiO_2 nanoparticles in their individual suspensions are 0.997×10^8 and $0.210 \times 10^8 \text{ mL}^{-1}$, respectively. After introducing the nanoparticles into the nanobubble suspension, the number concentration of the mixed system increases markedly to $1.655 \times 10^8 \text{ mL}^{-1}$, which significantly exceeds the simple arithmetic sum of the concentrations measured in the isolated nanobubble and nanoparticle systems.

As illustrated in Figure 7e, in the presence of anatase, the mean diameter (d_{50}) of the dispersed substances increased from 302 to 327 nm. The pronounced enhancement of the number concentration together with the observed larger mean diameter for nanobubble indicates that TiO_2 nanoparticles may reduce nanobubble decay rather than merely contributing their own scattering signatures. This experimental trend aligns well with our simulation results shown in Figure 2 and in Table 1. The computed free-energy barrier for nanobubble dissolution is higher by approximately 42 kJ mol^{-1} near the anatase (101)–water interface than in bulk water. This substantially higher barrier in the vicinity of TiO_2 indicates enhanced dynamic stability of nanobubbles once formed, which is also qualitatively consistent with the theoretical kinetics analysis shown in Figure 6c.

Figure 7d shows that the measured zeta potential of bulk nanobubbles is -24.4 mV in pure water and -29.6 mV in the $\text{NB}+\text{TiO}_2$ suspensions. The simulation predicts a zeta potential of -2.7 mV for the N_2 nanobubble in the TiO_2 –water system (Figure 6a), which is about 1 order of magnitude lower in absolute value than the experimental measurement. This difference might arise partially because the simulated nanobubbles are only about 3 nm in diameter, whereas the experimental nanobubbles exceed 100 nm. We attribute the more negative zeta potential in the TiO_2 suspensions to ionic contributions from both the anatase surface and the bulk water. The simulated TiO_2 –water system shows the presence of OH^-

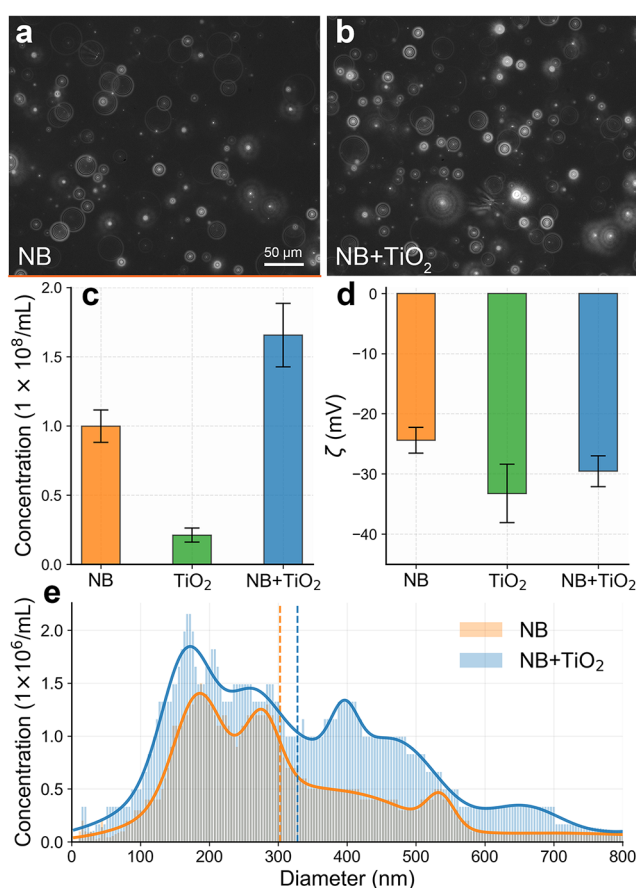


Figure 7. Experimental characterization of nanobubble and nanoparticle suspensions. (a) Snapshots of bulk nanobubbles (NBs) in pure water, and (b) in anatase TiO_2 (labeled as $\text{NB}+\text{TiO}_2$) suspensions. (c) Number concentrations and (d) zeta potentials (ζ) of NBs and of the combined nanobubble–nanoparticle ($\text{NB}+\text{TiO}_2$) suspensions in water, with TiO_2 -only suspensions as the control group. (e) Number concentration histogram as a function of diameter for bulk NB and $\text{NB}+\text{TiO}_2$ systems, where the dashed line marks the mean diameter (d_{50}) and the solid line shows the fitted distribution.

(surf) as well as OH^- (bulk). The OH^- (bulk) ions preferably enrich near the nanobubble surface, as illustrated in Figure 5, and increase the negative surface charge of nanobubble.³³ In addition, the zeta potential of pure TiO_2 nanoparticles is -33.3 mV , more negative than the zeta potential of nanobubble in the TiO_2 suspensions, consistent with the ionic distributions shown in Figure 3.

In summary, the present experimental evidence supports the main conclusion of simulations that the presence of TiO_2 surface improves nanobubble stability through ion-mediated modulation.

CONCLUSIONS

The present study provides an atomistic understanding of how the interfacial ionic environment and thermodynamic–dynamic coupling collectively determine the evolution of N_2 nanobubble near the anatase (101)–water interface. By enhanced-sampling DPMD simulations for nanobubbles in different anatase (101)–water/electrolyte interface systems, we show that the interfacial adsorption state of nanobubbles is thermodynamically favored, whereas detachment into the bulk is unfavorable. The anatase (101) surface promotes the water dissociation through undercoordinated titanium and oxygen

sites, which changes the local ionic environment and makes the neutral systems alkaline, the alkaline system more alkaline, and the acidic system close to neutral, thus influencing the stability and evolution dynamics of nanobubble near the anatase-water interface. Compared with anatase-free systems, the nucleation barriers of nanobubbles are increased in all systems with the anatase (101) surface, but are less relevant for the acid–base strength. The nucleation barriers follow the qualitative order $\text{TiO}_2\text{--HCl} \gtrsim \text{TiO}_2\text{--NaOH} \gtrsim \text{TiO}_2\text{--water} > \text{TiO}_2\text{--NaCl}$. The much lower nucleation barrier in the anatase-saline system is a result of the salting-out effect. Because free OH^- ions generated by the surface reaction of anatase accumulate in the water layer near the nanobubble surface, we observe an approximate pH-dependence of negative zeta potentials of nanobubbles in the anatase-containing systems, except for the anatase-saline system. In the $\text{TiO}_2\text{--NaCl}$ system, Na^+ and Cl^- ions, which also populate mainly near the nanobubble surface, screen the interfacial OH^- layer, so that a less negative zeta potential is obtained than that in the $\text{TiO}_2\text{--HCl}$. The dissolution barrier of nanobubble is linearly correlated with the magnitude of the zeta potential, that is, a more negative zeta potential corresponds to a higher barrier to nanobubble dissolution. These results are further supported by Epstein–Plesset modeling and experimental observations.

Despite these mechanistic insights, several limitations should be noted. The simulated nanobubbles consist of a limited number of molecules and have a diameter of just a few nanometers, much smaller than typical experimental nanobubbles (~ 100 nm). Moreover, because nucleation is simulated in a finite system, the free energy profile can decrease beyond the barrier to a minimum at a finite cluster size, rather than approaching the asymptotic behavior expected from macroscopic classical nucleation theory.⁸⁴ In addition, nuclear quantum effects were not included, and the deep potential model employed short-range approximations without explicit long-range electrostatics. While these simplifications may lead to modest quantitative differences, they do not affect the central mechanistic insights or the core conclusions of this work.

■ ASSOCIATED CONTENT

Data Availability Statement

The requisite input files for replicating the research can be accessed at <https://github.com/Zhang-pchao/research/tree/main/OPES-DPMD-Bubble-TiO2>.

SI Supporting Information

The Supporting Information is available free of charge at <https://pubs.acs.org/doi/10.1021/jacs.6c05480>.

Details of potential model generation, including building data sets, DFT calculations and DP model training; details of enhanced sampling settings, including bias potential and collective variables, and free energy calculations; details of molecular dynamics simulations, identification of interfacial ionic species, construction of the transition matrix, as well as other discussions and information (PDF)

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Notes

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