The interplay among gas, liquid and solid interactions determines the stability of surface nanobubbles†

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Surface nanobubbles are gaseous domains found at immersed substrates, whose remarkable persistence is not fully understood. Recently, it has been observed that the formation of nanobubbles is often associated with a local high gas oversaturation at the liquid–solid interface. Tan, An and Ohl have postulated the existence of an effective potential attracting the dissolved gas to the substrate and producing a local oversaturation within 1 nm from it that can stabilize nanobubbles by preventing outgassing in the region where gas flow would be maximum. It is this effective solid–gas potential – which is not the intrinsic, mechanical interaction between solid and gas atoms – its dependence on chemical and physical characteristics of the substrate, gas and liquid, that controls the stability and the other characteristics of surface nanobubbles. Here, we perform free energy atomistic calculations to determine, for the first time, the effective solid–gas interaction that allows us to identify the molecular origin of the stability and other properties of surface nanobubbles. By combining the Tan–An–Ohl model and the present results, we provide a comprehensive theoretical framework allowing, among others, the interpretation of recent unexplained experimental results, such as the stability of surface nanobubbles in degassed liquids, the very high gas concentration in the liquid surrounding nanobubbles, and nanobubble instability in organic solvents with high gas solubility.

1 Introduction

Surface nanobubbles are spherical-cap gas bubbles attached to immersed substrates1,2 (Fig. 1A). They are noted for their counter-intuitive properties, foremost of all their puzzling longevity. Despite numerous efforts over the years, no single model for their stability is widely accepted and able to yield a comprehensive explanation of all their properties.

The fundamental challenge in rationalizing the stability of nanobubbles is well illustrated by the early theoretical calculations of Epstein and Plesset,3 which indicate that spherical bubbles can never achieve stationarity—they either grow catastrophically or dissolve into bulk liquid in a subsecond timescale due to their small size and high curvature. Recently, this difficulty has been overcome by the insight that the contact line of surface nanobubbles may be pinned at nanoscale surface heterogeneities, thus modifying their diffusive dynamics such that they can attain a stable equilibrium when the bulk liquid is oversaturated with gas.4,5 Here, oversaturation is defined as \( \zeta = c_a/c_{sat} - 1 \), where \( c_a \) is the dissolved gas concentration in the bulk liquid and \( c_{sat} \) is the saturation concentration. Unfortunately, the pinning-oversaturation model cannot be a definitive explanation of the long-term stability of surface nanobubbles. The model requires that \( \zeta > 0 \), i.e. that gas concentration in the bulk liquid is higher than the saturation value, which is contradicted by the well-established experimental observation of surface nanobubbles in equilibrated open systems (\( \zeta = 0 \))6 and even under undersaturated conditions (\( \zeta < 0 \)).7–9 Indeed, results reported in ref. 8 led the authors to conclude that “the pinning mechanism alone cannot explain the long-term stability of surface nanobubbles”.

In parallel to theoretical progress, numerous simulation studies have been performed to unveil the mechanistic processes governing the stability of surface nanobubbles.10–12
Fig. 1 (A) Sketch of a nanobubble pinned at a solid surface forming a gas side equilibrium contact angle \( \theta_e \). Due to the small radius of curvature of the bubble and the corresponding large Laplace pressure, estimated to be several atmospheres, a rapid outgassing and, eventually, dissolution of the bubble is expected. The gas outflow (blue arrows, referring to an entire slice) is maximum at the bottom of the bubble, where the liquid/gas surface per bubble slice (horizontal gray lines) is maximum. This intuitively justifies why a large local gas oversaturation within \( \sim 1–2 \) nm from the surface is efficient at stabilizing nanobubbles: large local oversaturation near the bottom of the nanobubble (red arrows) can balance the gas outflow in the region where it is maximum. In the inset is shown the Young contact angle, \( \theta_Y \), the angle formed by the tangent to a liquid droplet at contact point with the surface. (B) Snapshot of a portion of the computational sample. Grey spheres represent carbon atoms of the graphite-like slab, the blue dumbbell \( \text{N}_2 \) and the white and red spheres connected by rod water molecules. The surface enclosing the volume occupied by \( \text{H}_2\text{O} \) is also shown.

Although these studies were successful in upholding the importance of contact line pinning to stability, they gave a limited contribution in identifying other key elements to achieve nanobubble stability. Indeed, perhaps influenced by the theoretical hypothesis that nanobubbles are stable only under oversaturated conditions, but in contrast with experimental evidence, these works focused only on \( \zeta \geq 0 \) conditions. Moreover, standard molecular dynamics techniques adopted in these pioneering studies allow to span a limited timescale, typically few tens of nanoseconds, 11–12 orders of magnitude shorter than the experimentally reported lifetime of nanobubbles, hours to days. This limits the predictive power of these simulations. To cope with these long timescales, special simulation techniques are needed, some of which have been adopted in the present study.

To resolve the difficulties of the pinning-oversaturation model, Tan, An and Ohl suggested that an attractive effective potential \( \phi \) of depth \( \sim k_B T \) (\( k_B \) is the Boltzmann’s constant and \( T \) is the temperature) exists between the substrate and gas molecules, which leads to the formation of a localized gas-oversaturated layer at the surface. \( \phi \) would rationalize the existence of a permanent local oversaturation sustaining surface nanobubbles, while relaxing the unrealistic requirement that the bulk water is indefinitely oversaturated with gas. Combined with pinning, local oversaturation can explain the stability of nanobubbles: (i) pinning prevents the shrinking of their footprint and (ii) local oversaturation counteracts the loss of gas from nanobubbles driven by their high Laplace pressure associated with their small curvature. In the original article, it was shown that a gas-rich water layer of \( \sim 1–2 \) nm at the liquid/solid interface would be sufficient to stabilize nanobubbles. This is possible because at the bottom of nanobubbles, where slices of the spherical cap have the largest area, the gas exchange is more intense (Fig. 1A). Thus, a local oversaturation in this region can effectively counteract the gas outflux. A number of recent experiments have shown that, indeed, one can produce a large local gas oversaturation at the solid/water interface and when the oversaturation reaches a critical value, of the order of \( \zeta \sim 150–250 \), nanobubbles are formed. Among the others, Zhou et al. have performed near-edge X-ray absorption fine structure experiments showing a large gas oversaturation in the liquid surrounding nanobubbles. Overall, these results call for a new comprehensive microscopic theory addressing the open questions concerning the stability of nanobubbles in a broader range of experimental conditions and their relationship with the possible existence of a permanent local oversaturation.

The stability of surface nanobubbles in water is not their only unique characteristic. Although stable in water, surface nanobubbles are unstable in selected protic (e.g. ethanol) and aprotic liquids (organic solvents). Moreover, surface nanobubbles present small contact angles, well below their equilibrium value, that changes significantly, well above the experimental accuracy, among experiments performed under the same nominal conditions (same solid, liquid, gas, degree of oversaturation \( \zeta \) and temperature). Finally, surface nanobubbles exhibit a remarkable resilience to harsh conditions, such as the near complete degassing of the liquid hosting them.

The TAO model alone does not provide a direct answer to these questions. Indeed, a key conclusion of the TAO model is that the question “what determines the stability of nanobubbles?” must be recast into “what determines the existence of a suitably attractive effective solid–gas potential \( \phi \)?”. The form and strength of \( \phi \), which depends on the nature of the solid, liquid and gas species, determines the properties of nanobubbles summarized above. However, within the TAO model, the effective potential \( \phi \) is an input, i.e., it must be obtained from an independent source. By combining the TAO model and the effective potential \( \phi \), determined here from atomistic simulations and its dependence on the chemical and physical characteristics of the solid–liquid–gas three phase system –
one obtains (i) a comprehensive theory of surface nanobubbles and (ii) a multiscale model with quantitative predictive capabilities.

Here, the form, strength and range of the effective potential is determined using restrained molecular dynamics (RMD), a simulation technique introduced in 2006 by Maragliano and Vanden-Eijnden\cite{Mara,vdEj} extending and simplifying previously well established approaches, e.g. the Blue-Moon ensemble.\cite{BM1,BM2}

This leads us to estimate the local gas oversaturation of the liquid near the substrate and the stability and contact angle of surface nanobubbles hosted in this environment. For solids immersed in water, our results reveal, for the first time, the existence of a potential well of few k_BT's at subnanometric distances from the wall, capable of attracting dissolved gases to the substrate, thus endorsing the viability of local oversaturation as a stabilization mechanism for surface nanobubbles in this liquid. By a systematic variation of the liquid–gas–solid interactions, type of liquid and solid our simulations provide a mechanistic understanding of a number of counterintuitive, previously unexplained, phenomena. For example, we were able to explain, within a unified framework, the large variability of nanobubbles' contact angle at highly ordered pyrolytic graphite immersed in water, why the contact angle of nanobubbles is di

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**2 Computational methods**

**2.1 Theoretical background**

In this work, we use restrained molecular dynamics, RMD, to compute the (Landau) free energy profile as a function of the distance of the center of mass of N_2 or O_2 from a graphite-like surface, z. This free energy is the microscopic analogue of the effective potential \( \phi \). We remark that the (Landau) free energy is different from the intrinsic interaction potential between N_2 and the solid, i.e. the force field; for example, the free energy embodies the effect of the liquid intervening between the solid and the gas molecule on the interaction potential, which is a key aspect in determining the stability and instability of nanobubbles in different liquids.

In statistical mechanics, any thermodynamic potential is related to the logarithm of a suitable probability density function in the relevant ensemble. In the present case, the relevant probability density function is \( M_z(z^*) \), the probability density that the center of mass of N_2 (or O_2), the function \( z(t_{N_2}) \) of the atomic positions of nitrogen (or oxygen) and surface atoms \( t_{N_2,s} \) is at a distance \( z^* \) from the graphite slab, regardless of the position of all the other atoms in the system. For the sake of simply of notation, in the following, we will write the distance between the center of mass of N_2 and the solid surface as a function of the (3N dimensional) atomic position vector \( r \).

In terms of the ensemble distribution \( m(r) \), \( M_z(z^*) \) reads:

\[
M_z(z^*) = \frac{\int dr m(r) \delta(z(r) - z^*)}{\int dr m(r)}
\]

where \( \delta(z) \) is the Dirac delta function, which has the role of selecting only those atomic configurations satisfying the condition \( z(r) = z^* \). In other words, \( M_z(z^*) \) is the probability to find the system in any atomic configuration \( r \) satisfying the condition \( z(r) = z^* \).

One can associate a thermodynamic potential to \( M_z(z^*) \), here the (Landau) free energy \( G(z^*) \), (see, e.g., ref. 30 and 31).

\[
G(z^*) = -k_BT \log M_z(z^*)
\]

One notices that this definition has the properties of usual thermodynamics potentials: \( G(z^*) \) has a minimum at the stable state, i.e. in correspondence with the maximum of the probability density function, and is expressed in energy units. One may notice that this definition is consistent with the definition of the Boltzmann entropy, apart the sign, which is opposite, consistently with the properties of entropy to be maximum at the stable state, and for the units of measure, which is energy divided by temperature for entropy. Finally, it can be shown that \( \Delta G = G(z^*) - G(z^*) \) between two states can be expressed as the reversible work operated by a generalized force to move the system from \( z_1^* \) to \( z_2^* \), which is consistent with the definition of free energies in thermodynamics.

To compute \( M_z(z^*) \) – hence \( G(z^*) \) – in atomistic simulations one can in principle run a long molecular dynamics trajectory and determine the histogram of \( z \) along it; this histogram is a discrete approximation of \( M_z(z^*) \) from which one can compute the free energy \via eqn (2). However, it is well known that this approach is computationally inefficient (see, e.g., ref. 30 and references cited therein). The key problem is that if the effective interaction is attractive, the molecule spends most of the time near the surface, if it is repulsive, the molecule is all the simulation far from the substrate; in both cases, it is impossible to collect statistics at all relevant \( z \) values within the timescale accessible by atomistic simulations (maximum hundreds of ns).

To overcome this problem, we use RMD,\cite{Mara,vdEj,DRMD} in which one introduces a controlled bias forcing the system to explore configurations, in which the distance \( z(r) \) of a gas molecule from the substrate is fixed at the value \( z^* \).†

† Indeed, RMD forces the system to visit atomic configurations almost satisfying the condition \( z(r) = z^* \). As explained in the main text, the departure from this condition is determined by the coupling parameter \( k \) of the auxiliary potential \( k/2(z(r) - z^*)^2 \) added to the physical interaction potential. Maragliano and Vanden-Eijnden have proven that the estimate of \( dG(z^*)/dz^* \) converges smoothly with \( k \), i.e. one can use a finite value of the coupling parameter to determine the derivative of the mean force within the accuracy of one or few k_BT's. On the other hand, releasing the strict condition of the constraint solves the problem of the implicit – and undesired – additional constraint \( z(r) = 0 \), which makes the computational determination of \( dG(z^*)/dz^* \) more complex.‡

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In other words, within RMD, one forces the system to visit only those configurations consistent with the condition \( z(r) = z^* \). As we will show in the following, this allows us to efficiently estimate the derivative of the free energy \( G(z^*) \), i.e. \( \frac{dG(z^*)}{dz^*} \), which can then be numerically integrated on \( z^* \) to obtain the entire profile of the thermodynamic potential as a function of the distance of \( N_2 \) from the graphite slab. The estimation of \( \frac{dG(z^*)}{dz^*} \) can be expressed as a local average of a suitable observable on the hyperplane \( z(r) = z^* \) and restrained MD can be used to compute this average.

In more detail, consider the (Landau) free energy of eqn (2) and for the sake of simplicity assume that the ensemble is canonical, hence \( m(r) = \exp[-V(r)/k_BT] \int dr \exp[-V(r)/k_BT] \), where \( V(r) \) is the (physical) interacting potential, the so-called force field. We remark that the arguments developed below can be straightforwardly extended to other ensembles, e.g. isothermal–isobaric. Within the canonical ensemble, the probability density function of eqn (1) reads:

\[
M_z(z^*) = \frac{\int dr \exp[-V(r)/k_BT] \delta(z(r) - z^*)}{\int dr \exp[-V(r)/k_BT]}
\]

from which one obtains that the derivative of the free energy is:

\[
\frac{dG(z^*)}{dz^*} = -k_BT \frac{\int dr \exp[-V(r)/k_BT] d\delta(z(r) - z^*)/dz^*}{\int dr \exp[-V(r)/k_BT] \delta(z(r) - z^*)}
\]

In eqn (4), one can replace the Dirac delta functions with a smooth Gaussian approximation, \( \delta(z(r) - z^*) \sim g_z(z(r) - z^*) = \sqrt{2\pi k_BT/k} \exp[-k(2k_BT)(z(r) - z^*)^2] \), \( k_BT/k \) is the variance of the Gaussian function, the parameter determining its width and the accuracy of the approximation to the corresponding Dirac delta function. The reason to express the variance of the gaussian ensemble by the apparently more complex form \( k_BT/k \) will be clear shortly. Here, we just remark that \( T \), the temperature, is determined by the experimental conditions (\( T = 300 \) K, in our case); \( k \) is the only parameter of the RMD method, whose suitable value can be determined as explained in the following. By replacing \( \delta(z(r) - z^*) \) with \( g_z(z(r) - z^*) \) in eqn (4), one can obtain an explicit, though approximated, formula for the derivative of the free energy:

\[
\frac{dG(z^*)}{dz^*} \approx \frac{\int dr k(z(r) - z^*) \exp[-V(r) + k/2(z(r) - z^*)^2]/k_BT}{\int dr \exp[-V(r) + k/2(z(r) - z^*)^2]/k_BT}
\]

Eqn (5) shows that, within the Gaussian approximation of the Dirac delta function, the derivative of the free energy can be computed as the expectation value of \( k(z(r) - z^*) \) over the canonical ensemble of a system driven by the so-called augmented potential \( V(r) + k/2(z(r) - z^*)^2 \). In practice, one computes \( \frac{dG(z^*)}{dz^*} \) at the current value of \( z^* \) by averaging the observable \( k(z(r) - z^*) \) along the trajectory of a constant number of particles/volume/temperature molecular dynamics, restrained by the potential \( k/2(z(r) - z^*)^2 \). The operation is repeated for several values of \( z^* \) in the relevant range, here between 0.4 nm and 1.6 nm every 0.02 nm, then the so-obtained \( \frac{dG(z^*)}{dz^*} \) is numerically integrated by the trapezoidal rule.

The suitable value of \( k \) is determined by studying the convergence of the approximate \( \frac{dG(z^*)}{dz^*} \), eqn (5), with increasing \( k \) at a set of \( z^* \) values. In the present case, \( k = 0.7 \) kcal mol\(^{-1}\) Å\(^{-2}\).

### 2.2 Simulation details

The sample we considered in our simulation consists of a 5-layer thick slab of graphite with an ABAB pattern\(^{14} \) in contact with TIP4P/2005 water\(^{35} \) containing a single \( N_2 \) or \( O_2 \) molecule. The molecule interacts with the surface and the liquid through a Lennard-Jones (LJ) potential \( v(r) = 4\epsilon_{\alpha\beta}((\sigma_{\alpha\beta}/r)^{12} - (\sigma_{\alpha\beta}/r)^6) \), where \( r \) is the distance between an atom of the molecule and one of the surface or water. \( \epsilon_{\alpha\beta} \) sets the strength of the interaction between atoms of type \( \alpha \) and \( \beta \) and \( \sigma_{\alpha\beta} \) the corresponding range. The values of \( \epsilon_{\alpha\beta} \) and \( \sigma_{\alpha\beta} \) are determined using the Lorentz–Berthelot combination rules: \( \sigma_{\alpha\beta} = 1/2(\sigma_{\alpha\alpha} + \sigma_{\beta\beta}) \) and \( \epsilon_{\alpha\beta} = \epsilon_{\alpha\alpha} \epsilon_{\beta\beta}/ \epsilon_{\alpha\alpha} \), where \( \epsilon_{\alpha\alpha} \) and \( \epsilon_{\beta\beta} \) are the parameters characterizing two atoms of the same species. For the purely repulsive gas–solid potential, we use the so-called Weeks–Chandler–Andersen potential. This potential is obtained from the \( N_2 \)-graphite one by cutting off the LJ potential at \( r = 2^{1/6} \epsilon_{\alpha\beta} \), i.e. at its minimum, and keeping it constant for \( r > 2^{1/6} \epsilon_{\alpha\beta} \). For \( \epsilon_{\alpha\alpha} \) and \( \epsilon_{\beta\beta} \), we used the values reported in Table 1. To model a liquid with a higher air solubility, we used a value of \( \epsilon_{\alpha\beta} \) three times larger than that used for water. The relationship between force field parameters and gas solubility is explained in the ESI\(^*\).

Consistently with the TIP4P/2005 model\(^{35} \), interaction between graphite and water is modeled by a modified LJ potential, \( v(r) = 4\epsilon_{\mathrm{ECOL}}((\sigma_{\mathrm{ECOL}}/r)^{12} - (\sigma_{\mathrm{ECOL}}/r)^6) \), in which the attractive term is scaled by a factor \( c.\quad^\text{36} \) \( c \) can be tuned to control the hydrophobicity of the surface. In practice, we compute the Young contact angle \( \theta_Y \) formed by a cylindrical droplet deposited on the graphite slab (Fig. ESI2†) for 26 values of \( c \) in the range 0.5–0.75 with a spacing of 0.01. Over this range, we observed a linear relationship between \( \theta_Y \) and \( c \) (Fig. ESI2†); thus, from the fitting of the \( \theta_Y-c \) pairs, we estimate the values of \( c \) corresponding to the Young contact angles of water used in free energy calculations: 70°, 80°, 90°, 100° and 110°.

Once again, consistent with the TIP4P/2005 model\(^{35} \), the gas/water interaction is also modeled by a LJ potential in which the parameters are obtained by literature data via the Lorentz–Berthelot combination rules.

### Table 1  LJ parameter for the chemical species composing our sample

| Species | \( \epsilon \) [kcal mol\(^{-1}\)] | \( \sigma \) [nm] |
|---------|---------------------------------|-----------------|
| C\(^{37,38}\) | 0.105067 | 0.3851 |
| N\(^{39}\) | 0.072332 | 0.332 |
| O\(^{18}\) | 0.103332 | 0.299 |
| O\(^{35}\) | 0.185200 | 0.31589 |
To mimic aprotic organic solvents, we use a LJ liquid with the same parameters of oxygen in the TIP4P/2005 water model. Solid–liquid interaction is modeled using the same modified LJ potential used for water, in which the “ε” parameter has been optimized to obtain the same Young contact angle values used for H₂O. For the liquid–gas interaction, we maintained the values used for water, hence the same affinity between the gas and the liquid.

The approximation of considering a single gas molecule dissolved in the liquid, which neglects possible interactions among gas molecules, is justified by the low solubility of air in water. In fact, as we will show in sec. 3, even considering the highest computed local oversaturation and the bulk liquid in equilibrium with air (no degassing), the highest value of local oversaturation is consistent with recent experimental results. More difficult is the comparison with computational results available in the literature, which also predicts a large oversaturation at the solid–liquid interfaces. In fact, in these works, the authors use a different computational approach consisting of preparing a sample with a very high gas oversaturation and determining the gas local oversaturation profile as a function of the distance from the surface. In our case, in constraint, thanks to the use of RMD to compute the free energy, we can consider more realistic bulk oversaturation conditions.

In general, the free energy exhibits two properties that are common for all values of the Young contact angle investigated. Firstly, $G(z)$ presents a characteristic minimum at ~0.4 nm from the surface, with a well depth ranging from 4 to 7.5$k_BT$; the well depth $G_{\text{min}}$ increases with $\gamma_\theta$ (Fig. 2E), i.e., the more hydrophobic the surface, the more it attracts nitrogen dissolved in water (Fig. 2F). Secondly, the free energy reaches its bulk value over a distance of ~1 nm. We remark that the range of this effective interaction is not determined by the cutoff of the solid–gas intrinsic interaction (only), which is 1.6 nm, but by the complex interplay between solid–gas, solid–liquid and liquid–gas interactions. This will be shown more in detail in the following. The free energy profile also presents additional features, such as a tiny maximum at ~0.7 nm that becomes more pronounced for more hydrophilic surfaces. The origin of the peculiar shape of $G(z)$ and its dependence on the characteristics of the solid, liquid and gas will be discussed in detail in a forthcoming article.

It should be stressed that the general features of $G(z)$ are not specific to nitrogen. In the ESI for $O_2$ dissolved in water, illustrating the similarities of the free energy profiles of nitrogen and oxygen (Fig. ESI4†).

Next, we connect atomistic results to experimental data, namely nanobubbles’ contact angles, using the TAO model, in which the free energy curves of Fig. 2 represent the key input (see the ESI† for a brief summary of the TAO model). The model is summarized by eqn (6), which is the condition that must be satisfied for the gas flow at the liquid–gas interface of the bubble to be zero (stationary) in the pinning-local oversaturation theory:

$$\int_0^{\pi/2} \left( \frac{2\gamma}{L_P} \sin \theta - \zeta(z) \right) dz = 0. \quad (6)$$

Here, the local oversaturation is (inset of Fig. 2A):

$$\zeta(z) = \frac{c_w}{c_{\text{sat}}} \exp \left( -\frac{\phi(z)}{k_BT} \right) - 1. \quad (7)$$
where $\phi(z)$ is replaced by a constrained B-spline of its microscopic counterpart $G(z)$. Note that the height of the nanobubbles $h(\theta)$ is related to the contact angle by the geometric identity $h = L \sqrt{(1 - \cos \theta)(1 + \cos \theta)}$. The solution of eqn (6) is the nanobubble contact angle $\theta_e$ for which the net gas flux in and out of the bubble is zero, i.e., the equilibrium contact angle $\theta_e$. Fig. 2G shows that the effective solid–gas attraction of Fig. 2A is sufficient to produce stable nanobubbles with equilibrium contact angles within the typical experimental range. In particular, consistent with recent experimental findings, our results show that nanobubbles can persist on nominally hydrophilic substrates.

The present results allow us to address two puzzling questions about nanobubbles at HOPG in water: (i) why is the contact angle much lower than the value one could predict from $\theta_Y$ (i.e., $\theta_e \neq 180^\circ - \theta_Y$, considering the opposite convention for the Young and nanobubble contact angles)? (ii) why is the experimental value of the contact angle of nanobubbles at HOPG scarcely reproducible? We remark that the TAO model provides a connection between the microscopic properties of the system, namely $G(z)$, and its macroscopic characteristics, the nanobubbles contact angle $\theta_e$, but the TAO model itself does not provide any estimation of the effective solid–gas interaction. Concerning the first question, our calculations show that given the strength of the solid/gas interaction and the typical Young contact angle of HOPG, which is slightly hydrophilic, one can achieve a local oversaturation that can compensate the outflow gas from the bubble only when $\theta_e$ is rather small, 20° or lower (Fig. 2G). This is, indeed, a combined effect of the pinning and local oversaturation model. Concerning the limited reproducibility of $\theta_e$ values, it is known that the variation among data reported by different research groups exceeds the experimental accuracy of contact angle measurements. To answer this question, one has to take into account that the preparation history of the HOPG substrate influences the effective affinity between water and graphite. In particular, the Young contact angle $\theta_Y$ of water on HOPG depends sensitively on how quickly the liquid is deposited onto the substrate. When deposited on freshly cleaved graphite, a droplet of water has $\theta_Y$...
Can the effective air attraction explain the survival of surface nanobubbles under (bulk) non-saturated conditions, \( i.e., \) in open \((\zeta = 0)^\text{,}^,\) or degassed systems \((\zeta < 0)^\text{,}^7\) ? In Fig. 3A, we report the degree of undersaturation \( \zeta* = \epsilon*_w*/\epsilon*_sat - 1 \) (<0) that nanobubbles can withstand before their equilibrium contact angle becomes negligible according to eqn (6). A negative value of \( \zeta* \) means that the gas concentration in the liquid bulk, \( \epsilon*_w^\text{,} \), is lower than the saturation value, \( \epsilon*_sat^\text{,} \). \( \zeta* \) is a convenient measure of the tolerance of nanobubbles to degassing. The equilibrium contact angle decreases monotonically with decreasing air concentration in the bulk liquid \(^16\) until it reaches a critical value \( \zeta* \) at which \( \theta_e = 0^\circ \) (Fig. 3B). One notices that for the HOPG/water/air, nanobubbles can sustain any level of degassing (blue line). Of course, the \( \zeta* = 0^\circ \) limit must be understood in the sense that the liquid still contains enough air that can accumulate at the surface to counteract nanobubble degassing. Under these conditions, the bulk oversaturation \( \zeta^\text{,} \) is negligible with respect to saturation conditions \( \zeta = 1^\text{,} \) virtually corresponding to complete degassing. Although the precise value of \( \zeta* \) might be critically affected by the limited accuracy of the empirical potentials used in our simulations, our general conclusions are consistent with the results reported in a very recent article \(^8\) showing that nanobubbles resisted to the highest level of degassing the authors were able to achieve (~80% of air dissolved in water was removed). As mentioned in the Introduction, their results led Qian et al. to conclude that the pinning-supersaturation mechanism is insufficient at explaining the long-term stability of surface nanobubbles. The present work allows the reconciliation of experiments and theory, adding the missing ingredient – local gas supersaturation at the wall – to the pinning model, which can explain the stability of nanobubbles under degassing conditions.

We remark that the TAO model and the present calculations do not allow the complete determination of the lifetime of nanobubbles. For example, the TAO model assumes contact line pinning but, indeed, the triple line can depin (see, \( e.g., \) ref. 45) and, as a consequence, nanobubbles can possibly be destabilized. Indeed, the relationship between contact line pinning and nanobubble lifetime has been experimentally confirmed (see, \( e.g., \) ref. 9). The characteristic times of contact line depinning and other processes determining the lifetime of nanobubbles are beyond the aim of the TAO model and of the present work. The theoretical determination of the time-scale of these processes requires the use of other advanced simulation techniques.\(^13,46,47\)

3.1 Effect of solid–liquid–gas interactions on the stability of nanobubbles

In the following, through a systematic variation of the solid–liquid–gas interactions, we investigate (i) the dependence of the stability of nanobubbles on the characteristics of the three phases and (ii) we show that our theory can explain a number of counterintuitive properties observed in experiments.

3.1.1 Interplay between solid–gas and solid–liquid interactions. Nanobubbles have been observed on many substrates including HOPG, glass, mica, talc, molybdenum disulphide, octadecyltriethylchlorosilane or \( 1H,1H,2H,2H\text{-perfluorodecyl}-\)trichlorosilane coated Si, decanethiol-coated gold and many more.\(^1\) The nature of the solid affects both the value of the Young contact angle and the strength of the intrinsic solid–gas interaction. Certainly, a stronger intrinsic solid–gas interaction increases the local oversaturation and, thus, the stability of nanobubbles. Here, we address the opposite question, whether weaker solid–gas intrinsic interactions still allow obtaining...
stable nanobubbles. In other words, we investigate how critical is the intrinsic solid–gas interaction for the stability of nanobubbles.

In addition to the standard sample modeling HOPG in water, we considered two alternative cases: first, a weaker solid–gas interaction, half the strength of HOPG/N₂ (Fig. 2B), and second, a purely repulsive solid–gas interaction, via the so-called Weeks–Chandler–Andersen potential 26 (Fig. 2C). The standard case shows that – except for the most hydrophilic case \( \theta_e = 70^\circ \) – the presence of water strengthens the effective solid–gas interaction beyond its intrinsic value, i.e. the well of \( G(z) \) in the presence of water is deeper than that of a substrate exposed to nitrogen gas only (dashed line in Fig. 2A). We stress that this does not mean that the density of gas at the substrate is higher in the liquid than in the nanobubble as \( G(z) \) is computed taking as a reference the value in the liquid and the pure gas, respectively; our results indicate that there might be a different gas enrichment at the substrate, which for the HOPG/H₂O/N₂ system at \( \theta_e > 70^\circ \) is higher in the liquid than in nanobubbles. Generally speaking, the higher the Young contact angle, the stronger the effective solid–gas attraction, which results in larger nanobubble contact angles (Fig. 2G). A similar trend is also observed for the weaker (halved) intrinsic solid–gas interaction (Fig. 2B), where the free energy is attractive for all Young contact angles. Comparing results in panels A and B of Fig. 2, one notices that the main effect of a weaker solid–gas interaction is a shift of all curves toward less attractive values, with the corresponding reduction in the local oversaturation (Fig. 2F) and nanobubbles’ contact angles (Fig. 2G). The weaker solid–gas attraction reduces the tolerance to undersaturation (Fig. 3), which, however, remains high.

It is worth remarking that despite the significant reduction in the solid–gas intrinsic interaction, the nanobubble still presents a non-negligible contact angle also at hydrophilic substrates. This is due to the enhancement in the effective gas attraction associated with the presence of water and explains the relative independence of the nanobubble contact angle on the chemical nature of the substrate (see ref. 27 and references cited therein). It is worth stressing that the TAO model alone is not sufficient to predict the dependence of nanobubbles’ contact angle up to \( \theta_e \leq 20^\circ \).1

Starting from the HOPG/N₂/H₂O case (standard case), we increase gas solubility by enhancing the strength of the N₂–H₂O interaction by a factor 3, corresponding to a 5-fold increase in solubility 49 (see the ESI† for additional information). This results in a substantially shallower well (Fig. 2D), more difficult to detect experimentally. Also, in this case, the tolerance to degassing is reduced, but nanobubbles at purely gas-repulsive surfaces can withstand \( \zeta < 0 \) provided that the substrate is hydrophobic (\( \theta_r \geq 90^\circ \)).

The case of the purely repulsive intrinsic solid–gas interaction illustrates the relevance of the gas–liquid interaction in determining the effective potential acting on the gas. In fact, one notices that \( G(z) \) is non-negligible well beyond the distance of N₂ from the surface, where the intrinsic solid–gas interaction is zero. To explain the effect of the liquid and the value of its Young contact angle on the free energy, we propose that the effective solid–gas interaction is due to the balance between its intrinsic strength, i.e. the strength in the absence of liquid (dashed line in Fig. 2A–C), and the gas solubility. The intuitive argument behind this hypothesis is that if the gas is insoluble, \( i.e. \) if hosting it in the liquid is energetically disfavored, it will preferentially accumulate at the solid/water interface, where the hydrogen bond network of the liquid is already compromised and the energetic cost of insertion of a molecule is reduced. The more hydrophobic the surface, the less energetically expensive it is to replace a molecule of water at the solid/water interface with one of N₂ (or O₂). Moreover, due to the depletion of liquids at the interface with solids (see, e.g., ref. 48), \( G(z) \) reaches its plateau value at a \( z \) distance where the liquid has fully recovered its bulk density. In fact, for larger distances from the wall, the gas molecule is embedded in a liquid environment that is independent on the position \( z \) and the effective interaction potential becomes flat. This explains why the range of \( G(z) \) is longer than that of the purely repulsive intrinsic solid–gas interaction, as the range of the effective potential is determined by the width of the depletion zone. The effect of gas solubility in the liquid is discussed in more detail below.

Summarizing, we found that nanobubbles with a contact angle up to \( \sim 10^\circ \) can be formed also at surfaces with a weak intrinsic solid–gas attraction; even purely gas-repulsive surfaces can produce stable nanobubbles, although their small contact angle might prevent their experimental identification. For surfaces with low gas attraction, the stabilization of nanobubbles requires a complementary high hydrophobicity, which enhances the strength of the effective solid–gas attraction. In contrast, relatively strong intrinsic solid–gas interactions (few \( k_BT/s \)) can lead to stable nanobubbles also at hydrophilic surfaces. By revealing the key and active role of the liquid in determining the strength of the effective solid–gas interaction, our simulations allow the explanation of the experimental observation that in water surface the contact angle of nanobubbles seemingly ceases to be dependent on the chemical nature of the substrate, typically falling in the range \( 5^\circ \leq \theta_e \leq 20^\circ \).1

3.1.2 Effect of the gas solubility. With the insight that the stability of nanobubbles depends also on the liquid–gas interaction, we investigate in detail the effect of gas solubility. Starting from the HOPG/N₂/H₂O case (standard case), we increase gas solubility by enhancing the strength of the N₂–H₂O interaction by a factor 3, corresponding to a 5-fold increase in solubility 49 (see the ESI† for additional information).
Correspondingly, both the predicted oversaturation (Fig. 2F, cyan triangles) and the equilibrium contact angle (Fig. 2G, cyan triangles) are greatly reduced. Also, in this case, $\theta_e$ might be too low for nanobubbles to be experimentally observed. Tolerance to degassing is reduced as well, although the TAO model predicts that stable nanobubbles are found in liquid with $\zeta < 1$ for all Young contact angles considered (Fig. 3).

The effect of gas solubility on the stability of surface nanobubbles has been observed in experiments. Surface nanobubbles are known to be very stable in water and are insensitive even to the addition of high concentrations of acids, bases or salts. However, a notable exception to their reputed ‘super-stability’ is that, when surface nanobubbles in water have their liquid replaced with ethanol, they destabilize and dissolve.\textsuperscript{22} Systematic experiments have further shown that the surface coverage of nanobubbles incubated in water–ethanol mixtures monotonously drops to zero as the mole fraction of the alcohol is progressively increased from 0 to 20%. Note that ethanol, which, similar to water, presents strong hydrogen bonding among its molecules as confirmed by a relatively high boiling point ($\sim 79 ^\circ$C), has double the gas solubility of water; thus, consistently with experiments, our theory predicts that surface nanobubbles are unstable in a liquid with an air solubility as high as that in ethanol across the entire range of $\theta_e$, which is again consistent with experiments.

### 3.2. Nanobubbles in aprotic liquids

The fact that experiments are nearly exclusively performed in water has, over time, raised questions about possible specific properties of this liquid in stabilizing surface nanobubbles. Can surface nanobubbles exist in non-aqueous liquids? Is hydrogen bonding responsible for the persistence of surface nanobubbles in water? These questions will be addressed in this section. An influential experimental study by An et al.\textsuperscript{23} positively answers both questions. The authors used solvent exchange various non-aqueous liquids to nucleate surface nanobubbles on HOPG. They found surface nanobubbles to exist in protic solvents but not in aprotic ones, leading them to propose that a hydrogen bonding network is necessary for the stability of surface nanobubbles.

To isolate the effect of the hydrogen bonding in our simulations, we replaced water with a (mono-atomic) Lennard-Jones liquid with the same LJ parameters as in the TIP4P/2005 model.\textsuperscript{35} It is worth recalling that the TIP4P/2005 model is characterized by two contributions: (i) the electrostatic interactions among point charges centered on H and O atoms, which result in the formation of hydrogen bonding among H$_2$O molecules reproducing the typical tetrahedral structure of water and (ii) the non-directional interaction among LJ centers positioned slightly off of oxygen atom, determining the “size” of the molecule. Thus, by replacing the TIP4P/2005 water model with the corresponding LJ allowed us to effectively investigate the effect of hydrogen bonding. To make the comparison stringent, the solid–liquid interaction is tuned to form the same Young contact angle as for water on HOPG considered above: $70^\circ < \theta_e < 110^\circ$. Liquid–gas and solid–gas interactions are the same as that in the water case (see sec. Simulation details for additional information).\textsuperscript{26} Thus, the only difference with previous simulations is the absence of hydrogen bonding within the liquid. This setup is not meant to model any specific solid/liquid/gas system; here, we exploit the flexibility of simulations to model systems with ad hoc characteristics to identify the role of hydrogen bonding to stabilize nanobubbles.

Fig. 4A, B and C show $G(z)$ for a surface immersed in the aprotic, LJ liquid when the solid–gas intrinsic interaction is standard, weakened (halved) and purely repulsive. These panels are equivalent to the corresponding panels of Fig. 2 for water. Our results show that strong attractive effective solid–gas interactions may still be present but this is due to the intrinsic characteristics of the solid–gas pair: aprotic solvents do not enhance this interaction. Thus, in an aprotic liquid, one cannot obtain stable nanobubbles at purely gas-repulsive surfaces even if they are highly solvophobic. The present results are consistent with the explicit molecular dynamics simulations by Maheshwari et al.,\textsuperscript{11} which show that a pinned surface nanobubble in a classic LJ liquid could be stable, albeit in the original article these conclusions have been drawn only for a short timescale of 30 ns.

The fact that aprotic solvents do not deepen the solid–gas well seems related to the lack of directional intermolecular interactions, which make the liquid–liquid cohesion stronger in the protic case. Thus, at a variance with water and other protic solvents, there is no or only a minor energetic gain to move the solute gas from the bulk to the solid/liquid interface, and one expects a lower local oversaturation. In the case of purely repulsive surfaces, the presence of an aprotic liquid increases the repulsion because there is an energetic penalty to have an interaction with a gas molecule rather than a (or more) liquid particle, and this is not balanced by the penalty to move the gas molecule in the liquid bulk, like in the case of hydrogen bonded liquids. In short, at a variance with water, an aprotic liquid does not enhance the intrinsic solid–gas attraction, which must be strong enough to stabilize nanobubbles.

Another effect one has to take into account is that non-polar gases often have a higher solubility in aprotic liquids. In the previous section, we have already shown that a higher gas solubility reduces the effective solid–gas attraction, which destabilizes nanobubbles. To prove that the same concept holds also in the case of aprotic liquids, the above calculations were repeated for the case of a stronger liquid–gas interaction.

As for the case of water, we increased the liquid–gas interaction strength by a factor 3. Also in this case, a higher solubility weakens the effective solid–gas attraction; moreover, since hydrophobicity plays a minor role in the case of aprotic liquids, here the solid–gas effective interaction is too weak to stabilize nanobubbles over the entire range of $\theta_e$ considered.

The principle we have just identified, that gas solubility is key at determining the stability of nanobubbles, provides a theoretical framework to explain recent experimental results on aprotic liquids. Consider, for example, dimethyl sulfoxide (DMSO); nitrogen has a solubility in DMSO that is about one order of magnitude higher than that in water\textsuperscript{21} and An et al.\textsuperscript{23}
have shown that nanobubbles are unstable in the former liquid. Similarly, propylene carbonate (PC), another aprotic liquid that An et al. have shown to be unable to form nanobubbles, has a nitrogen solubility which is more than twice that of water. Additionally, both do not form hydrogen bonds and have a lower contact angle at HOPG than water (DMSO $\theta_Y = 45^\circ$, PC $\theta_Y = 31^\circ$, water $\sim 70^\circ$), which further weakens the effective gas–solid interaction.

Summarizing, our results show a more subtle picture of stability of nanobubbles than previously hypothesized: aprotic liquids can still form attractive wells for the dissolved gas and hence form stable nanobubbles of sizable contact angle (Fig. 4G), provided that the solid–gas interaction is strong enough. However, another effect competes, and might overcome the strength of solid–gas attraction, i.e. the gas solubility in the liquid, which for air is usually higher in aprotic solvents.

4 Conclusions

In this work, we have computed the effective interaction potential between a solvated gas molecule and a planar substrate, with a view towards understanding how the effective, liquid-mediated solid–gas interaction affects the stability and other properties of surface nanobubbles. These calculations explain why surface nanobubbles have small contact angles, obtaining theoretical $\theta_e$ values in quantitative agreement with experimental results. Moreover, our calculations allow the explanation why nanobubbles are stable in water on a surprisingly wide variety of substrates with disparate hydrophobicities and gas affinities, under undersaturated conditions, and yet also catastrophically destabilize in organic solvents. The ability of our calculations, in conjunction with the TAO model, to simultaneously account for the vast majority of experimental observations on surface nanobubbles suggests that it effectively and comprehensively captures the chemistry and physics of the stability of surface nanobubbles and its dependence on the chemical nature of the solid, liquid and gas species.

Conflicts of interest

There are no conflicts to declare.

Acknowledgements

The research received funding from the European Research Council under the European Union’s Seventh Framework
Programme (FP7/2007-2013)/ERC grant agreement no. 339446 and under European Union's Horizon 2020 Research and Innovation Programme grant agreement no. 803213. S. M. acknowledges financial support from the Sapienza University of Rome, grant no. RG11715C81D4F43C, and from the project HPC-EUROPA3 (INFRAIA-2016-1-730897), with the support of the EC Research Innovation Action under the H2020 Programme, grant no. HPC174LQM. B. H. T. acknowledges financial support from the SMART Fellowship.

C. D. O. acknowledges support from the Joint Research Programme of the National Natural Science Foundation of China (11861131005) and the Deutsche Forschungsgemeinschaft (OH 75/3-1). The authors acknowledge PRACE for awarding access to resource MARCONI based in Italy at Casalecchio di Reno.

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