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Can microsolvation effects be estimated from vacuum computations? A case-study of alcohol decomposition at the $H_2O/Pt(111)$ interface †

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Converting biomass into sustainable chemicals and energy feedstocks requires innovative heterogeneous catalysts able to efficiently work in aqueous conditions. Computational chemistry is a key asset in the design of these novel catalysts, but it has to face two challenges: the large reaction networks and the potential role of hydration. They can be addressed using scaling relations such as Brønsted-Evans-Polanyi (BEP) and solvation models, respectively. In this study, we show that typical reaction and activation energies of alcohol decomposition on Pt(111) are not strongly modified by the inclusion of the water solvent as a continuum model. In contrast, adding a single water molecule strongly favors O-H and C-OH scission while it prevents C-O and to a lesser extent C-C scissions. The resulting BEP relationships partially reflect these changes induced by the solvent. Predicting Pt-catalyzed alcohol decomposition in water thus should account for the influence of the solvent on thermodynamics and kinetics. In addition, we found that the reaction energy obtained in presence of an explicit water molecule scales with the ones obtained in vacuum. Hence, we reveal that vacuum computations in combination with corrections based on our linear regressions are able to capture the important H-bonding effect.

1 Introduction

The necessity for sustainable chemical and energy ressources urges the scientific community to develop renewable processes. One of the many trails is the lignocellulosic biomass conversion, for energy and chemicals. ^{1–3} The cellulosic fraction that can be extracted contains mainly cellulose and hemicellulose at different relative amounts depending on the source material. ⁴ These sugars and polyols cannot easily be processed in gas phase but are water soluble. Thus, their valorisation necessitates the development of novel heterogeneous catalysts, active and stable in water, sustained by an improved knowledge of the reactivity of alcohol at the water/catalyst interface.

Density functional theory (DFT) has proven itself relevant to bring understanding in a variety of domains, ⁵ including photoin-duced phenomenons, ⁶ bio-systems, ⁷ and our field of interest, heterogeneous catalysis, ⁸ at reasonable computational cost. DFT can now provide accurate energies of large systems, ⁹ and adsorption energies on metal surfaces are in good agreement with experimental data when using state-of-the-art functionals. ^{10,11} Cou-

Several approaches can be used to investigate the impact of the water solvent at the metal/water interface as reviewed recently. 14 A brute force approach consists in including a full description of the liquid water. While this approach is more and more feasible when considering oxides such as silica 15 or alumina, 16,17 it is still too costly to be systematically implemented in the case of metallic systems, 18-20 and liquid water/metal interfaces are typically modeled using ice-layers. 21 Therefore, improved force fields 22 and approximate QM/MM schemes are being actively developed. 23,24 A much cheaper approach is to utilize a polarizable continuum model (PCM). 25-32 It has been used recently to investigate hydrodeoxygenation of propanoic acid 33 and methyl propionate³⁴ on Pd catalysts or the reforming of ethylene glycol on Pt³⁵ and for modelling electrocatalytic reactions. ³⁶ However, it is well-known that not all the effects of H-bonding can be described with a continuum solvent. ^{25,37} An elegant way to palliate this shortcoming is to explicitly introduce a few selected water molecules as shown not only in homogeneous phases 25,38,39 but

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pled with kinetic simulations, it can help designing efficient metal supported catalysts. ¹² Modeling realistic heterogeneous catalysis in liquid water remains highly challenging ¹³ since it faces two main issues: how to take into account the hydration? how to identify optimal reaction pathways among extensive and complex reaction networks?

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also on metal/water interfaces. ^{40–42} For instance, on metal surfaces, the OH scission in alcohol is rather sensitive to H-bonding while the CH scission is not. ^{41–44} Similarly, one could expect a larger impact on C-O scission than on the C-C one, but this has not been investigated yet in details with an explicit inclusion of water molecules.

A reasonable way of dealing with large and complex reaction networks *in silico* is to follow a multiscale approach. ^{45,46} Starting with density functional theory (DFT) on smaller reactions networks, one can build linear scaling relationships and group additivity schemes for estimating reaction energies, ^{47–49} and Brønsted-Evans-Polanyi (BEP) relationships for estimating reaction barriers. ^{50–52} Together, these fast approximations allow to perform micro-kinetic simulations for extended reaction networks to identify the most important pathways. ⁵³

In the quest of a universal BEP relationship to describe bond scissions at transition-metal surfaces, ^{54–58} several variants of the BEP relationships have been proposed over the last decade. ^{59–61} We have recently shown that the original formulation, ^{62,63} that simply relates the activation energy with the reaction energy of the elementary step, performs very well to describe the C-H and O-H scissions in alcohols at metallic surfaces. ⁶¹ However, very few works focused on the influence of the solvent on these relationships for bond scission at metallic surfaces. We recently investigated the impact of a continuum water model on group additivity schemes ⁴⁹ and the explicit inclusion of water on C-H and O-H bonds scissions on various metallic surfaces. ^{43,44,64}

In this study, we will address these two issues focusing on alcohol decomposition catalyzed by a Pt catalyst immersed in water. This choice is motivated by the work of Dumesic and co-workers that have identified Pt as a reasonable catalyst for aqueous phase reformation of mono and polyols such as sorbitol or glycerol. 65 To assess the role of the solvent model, we systematically compare the continuum approach (PCM) with the combination of PCM and micro-solvation, that is called cluster-continuum model (CCM), ³⁸ on a set of typical scissions in alcohols (over 40 reactions). The corresponding BEP relations are established and their sensitivity to the solvent model is discussed. Last, we demonstrate that the reaction energy of the elementary steps under consideration scales with the ones obtained in vacuum or PCM. As a result, the BEP established here including the influence of the solvent can be used with reaction energies computed in vacuum, considerably speeding up the exploration of extensive reaction networks.

2 Computational methods

2.1 Periodic DFT

Restricted periodic density functional theory (DFT) computations were performed using the Vienna Ab-initio Simulation Package, ^{66–69} VASP 5.3.5. The ion-electron interactions are described using the projector augmented wave (PAW) method. ^{70,71} The electronic structure is described within the generalized gradient approximation (GGA), using the PBE exchange correlation functional, ^{72,73} with a plane wave basis set cut off at 400 eV. Dispersion interactions are included with the recently implemented

dDsC dispersion correction, that computes electronic structure dependent dispersion coefficients 74 and couples the dispersion correction to the density functional via a bond-index based damping function. 75 This scheme has been shown to perform excellently for adsorption energies on Pt(111). 10 Integration of the Brillouin zone is performed with a Monkhorst-Pack mesh 76 of 3x3x1 k-points for surface species, and the gamma point for gaseous species. Convergence criteria are set to 10^{-6} eV for the electronics and 0.02 eV Å $^{-1}$ for the forces. Dipole corrections were applied when a polarizable continuum solvent model was used, in order to avoid an overestimation of dipolar interaction between the solvent model and the electronic system.

The platinum surface is modeled using a p(3x3) cell of a 4 layers slab, with 9 layers of void. The optimal bulk position was obtained with a Pt-Pt distance of 2.80 Å , which is used as a constraint for the two bottom layers of the slab. Gaseous species are computed in a cubic box of 20 Å in length. To save computational time, the reported energies do not include zero point energy correction.

2.2 Continuum solvent model

Solvation effects are taken into account using a Polarizable Continuum Model (PCM) available for VASP as the VASPSol package. ^{29,30} This allows to compute a solvation free energy as in eq

$$\Delta G_{inter}^{PCM} = \Delta G_{elec} + \Delta G_{cavitation} \tag{1}$$

where ΔG_{inter}^{PCM} represents the interaction energy between the continuum solvent and the solute, which is composed of the electrostatic interaction ΔG_{elec} and the non-electrostatic term $\Delta G_{cavitation}$ which depends on the area of the cavity and captures an average of positive and negative contributions (Pauli repulsion and dispersion interactions). This second term is often called cavitation energy and was set to 0 for our study to increase computational efficiency (see ref 49 for a discussion of the small effect of the cavitation energy). Therefore, the solvent model only assess dielectric interactions. We used default values for the empirical parameters and dielectric constant, which are representative of water.

2.3 Transition states

The transition states were optimized using the same level of theory, with either a quasi Newton algorithm when the structures were close enough to the saddle point, or the dimer method ⁷⁷. Each transition state is confirmed by a normal mode analysis showing a unique imaginary frequency following the expected displacement mode.

2.4 BEP relations

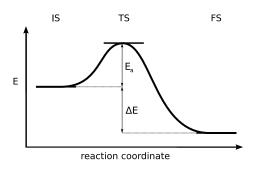


Fig. 1 Definition of the reaction energy and the activation energy

Brønsted-Evans-Polanyi^{50–52} (BEP) relations consist in linearly correlating a reaction energy and its activation energy. The critical point is to define the proper reaction set, which should be general enough, but without loosing too much system specificity. For instance, we could not establish reliable BEP relationships when using a large set of hydrogenation or C-C bond formation reactions of hydrocarbons on Pd⁷⁸. We define the BEP relations as follows:

$$E_a = \alpha \Delta E + \beta \tag{2}$$

with α and β as the fitting parameters, E_a the activation energy and ΔE the energy difference between the final state (FS) and the initial state (IS) (see Figure 1):

$$\Delta E = E_{FS} - E_{IS} \tag{3}$$

We will consider energies as $E = E_{DFT} + \Delta G_{PCM}^{solv}$ for calculations including PCM solvation.

2.5 Set of reactions

Following our previous study, ⁴⁴ 11 typical C-H and the O-H scissions reactions are investigated. In this set, all C-H are in α position to a hydroxyl or alkoxy group. 17 reactions from the ethanol reforming network were considered for the C-C scissions and 15 for the C-O. ⁷⁹ All C-C bonds belong to a fragment with an oxygen atom. In the Cluster Continuum Model (CCM), a water molecule has been added. Starting from the PCM optimized structure, several configurations have been tested where this water molecule is H-bonded with the molecular fragment and is chemisorbed on a neighbor Pt (top site) through its lone pair. A detailed list of the reactions can be found in supplementary materials together with the corresponding energies and structures.

3 Results and discussion

Reaction and activation energies of 43 typical alcohol decomposition steps on Pt(111) were computed to compare two models of solvation: polarizable continuum model (PCM) and polarizable continuum model supplemented with the explicit inclusion of a water molecule in a Cluster Continuum Model (CCM). To assess the impact of solvation on quantities of interests such as the reaction energy ΔE and the activation energy E_a , we have introduced the solvation enhancement 49 Λ^{sol} . For instance, $\Lambda^{sol}_{\Lambda E}(PCM)$ repre-

sents the effect of the inclusion of the PCM on the reaction energy and is defined as:

$$\Lambda_{\Lambda F}^{sol}(PCM) = \Delta E^{PCM} - \Delta E^{vacuum} \tag{4}$$

A negative value of $\Lambda_{\Delta E}^{sol}(PCM)$ means that the reaction is favored by the PCM, and a positive the opposite.

Before comparing these two solvation models and their impact on BEP relationships, we have taken advantage of this study to analyse the impact of dispersion corrections on BEP relationships.

3.1 Dispersion effects

Since most of the BEP relationships provided in the literature were obtained using GGA energies that were not corrected for dispersion interactions, we have analyzed the effect of the inclusion of the dDsC dispersion correction on the reaction and activation energies and on the resulting BEP relationships. Detailed values are available in the SI. The dispersion correction increases the reaction energies for C-C and C-O scissions by a mean of 0.05 eV and 0.03 eV respectively, while the reaction barriers are hardly affected. The impact is similar on reaction energies and barriers for $C-H_{\alpha}$ with a mean of -0.03 eV for the reaction energies and -0.04 eV for the activation energies. As a consequence, the corresponding BEP relationships are barely affected (see Table S1) by the inclusion of dispersion effects. For O-H scissions, reaction energies are slightly more modified than barriers as illustrated by the O-H cleavage of CH3C(OH)CH3 to CH3C(O)CH3, Figure 2, where the reaction barrier only decreases from 0.82 to 0.81 eV, while the reaction energy goes up from -0.02 to 0.14 eV. In this extreme case, the reactant is adsorbed through an sp3 α carbon, bringing the methyl groups close to the surface, which is stabilized upon inclusion of the dispersion. After the reaction, the steric hindrance does not allow the C=O bond to be chemisorbed in a π configuration. Therefore, the product is adsorbed by the oxygen lone pairs, with the methyl groups at a longer distance from the surface, reducing the dispersion interaction with the surface. All in all, the reaction becomes endothermic ($\Delta E = 0.14 \text{ eV}$) upon the inclusion of a dispersion correction while it is slightly exothermic ($\Delta E = -0.02$ eV) without it. However, the impact of the dispersion correction on reaction and activation energies is limited on average and the corresponding BEP relationships are barely affected as shown in Table S1.

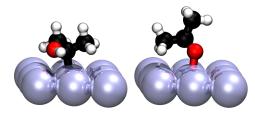


Fig. 2 Structures of $CH_3C(OH)CH_3$ and $CH_3C(O)CH_3$ adsorbed on Pt(111)

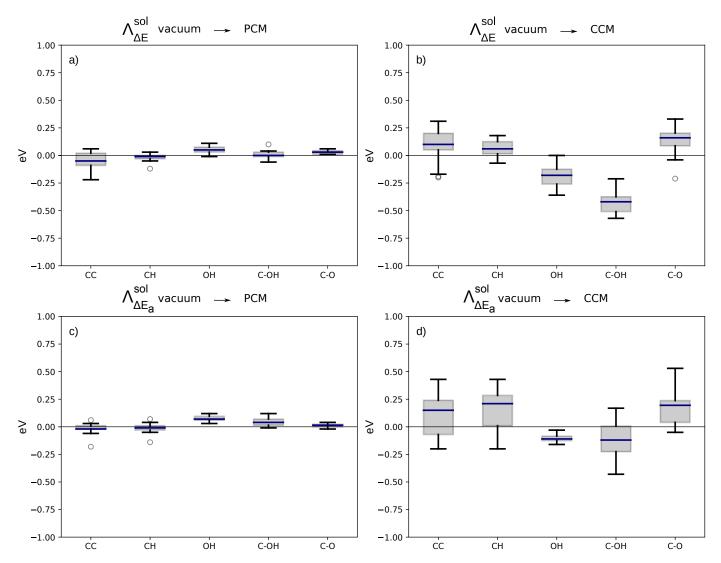


Fig. 3 Solvation enhancement for reaction and activation energies of C-C, C-H, O-H and C-O scissions, the later being separated in two subsets (C-OH & C-O). a) Solvation enhancement on reaction energies using PCM; b) Solvation enhancement on reaction energies using CCM; c) Solvation enhancement on activation energies using PCM; d) Solvation enhancement on activation energies using CCM. The boxes represent the first and the third quartiles of the data distribution, the blue band is the median (second quartile) and the whiskers are the minimum and the maximum, outliers are represented as individual points.

3.2 Polarisable Continuum Model (PCM)

To assess the influence of the water PCM on the reaction energies and activation energies in alcohol decomposition, $\Lambda^{sol}_{\Lambda E}(PCM)$ and $\Lambda_{E_{-}}^{sol}(PCM)$ were collected for the entire set of elementary steps. Their distribution are represented as box plots in Figure 3, separating the various types of bonds under consideration: C-H, O-H, C-C and C-O(H). For the sake of consistency with the rest of the discussion, the C-O(H) set was split into two subsets (C-O and C-OH), but this is clearly not necessary at this stage as shown by the similarity of the box plots. To start, it is worth noting that the values of $\Lambda^{sol}(PCM)$ span a limited range of energy, with the largest value obtained for a C-C scission at $\Lambda_{\Delta E}^{sol}(PCM)$ =-0.22 eV (CH₃-COH scission). This is in the range of what has been found in previous studies. 14,32,37,49

When considering more specifically the O-H scission, the PCM water solvent shifts up the reaction barriers and the reaction energies by an average of 0.05 eV. This can be explained by the high affinity of O-H groups for the PCM solvent: the system is loosing a well solvated group by breaking this bond weakening the solute/solvent interaction along the reaction. 49 Conversely, the C- H_{α} bond dissociation generally induces only small modification of the orientation of the hydroxyl group. As a consequence, the hydration energy does not vary much and Λ^{sol} is almost zero for most C-H scissions.

While the activation energy of C-C dissociation is generally not affected, the reaction energy is more exothermic for most reactions upon the inclusion of the water solvent as a PCM $(\Lambda_{AE}^{sol}(PCM) < 0)$. That the PCM favors C-C cleavage can be rationalized bearing in mind that all our C-C reactions involve an oxygen atom attached at one of the carbon atom. Hence, breaking the reactant can lead to products with an oxygenated group pointing towards the solvent and therefore enhancing the interaction with the implicit model through dipole/continuum interaction. ⁴⁹ Conversely, the dissociation of the C-O(H) bond is generally slightly slowed down by the PCM hydration $((\Lambda^{sol}_{\Delta E_a}(PCM) > 0)$ and the corresponding reaction energies are more endothermic.

3.3 Cluster Continuum Model (CCM)

A clear limitation of the PCM is the correct description of H-bonding between the solvent and the solute. To circumvent this issue, the cluster continuum model (CCM) includes one or a few water molecules, H-bonded to the solute, to capture this contribution to the solvation energy (or its variation). Here, we have chosen to systematically add a water molecule to adsorbates with an oxygen atom. When possible, the water molecule was placed as to form an H-bond with the adsorbed organic fragment and on atop sites, so that the oxygen lone-pair optimally interacts with the Pt surface, which can lead to significant geometric distortions. In Figure 3, the Λ^{sol} box plots obtained for CCM are shown together with the ones obtained with PCM. Clearly, the addition of a water molecule influences the reaction and activation energies of alcohol decomposition much more than the PCM.

Let us focus first on a typical case. The largest solvent effect is found for the activation energy of the C-O scission in CH₃CH₂-O (Λ_E^{sol} = 0.53 eV). Figure 4 shows the reaction energy profile for the C-O scission, illustrating that the PCM only leads to a global stabilization of \sim 0.1 eV compared to vacuum. CCM, however, stabilizes the initial state more than the final state (difference of ~0.2 eV), while it leaves the TS almost unaffected, increasing the activation energy. The corresponding geometries in vacuum and using CCM are depicted in Figure 5 and 6 respectively. CH₃CH₂O in presence of an explicit water molecule illustrates the typical case where the adsorbate is the H-bond acceptor. The O lone pair of CH₂CH₂O interacts indeed strongly with the hydrogen of the chemisorbed water in a very short H-bond (1.39 Å), with a concomitant increase of the Pt-O bond (from 1.99 Å in vacuum to 2.07 Å in CCM). This cumulation of water chemisorption and H-bonding has a strong synergy as already detailed in similar cases 80, which explains the strong stabilization of the initial state seen in Figure 4. In the transition state, the H-bonding is almost lost (2.07 Å) and water does not interact anymore in an ideal configuration with Pt(111), explaining the little stabilization observed from vacuum to CCM in Figure 4 and the strong $\Lambda_{E_a}^{sol}$ of 0.53 eV. In the final state, water is chemisorbed in a neighboring position, with no specific interaction with the chemisorbed oxygen. The gain in energy compared with vacuum or PCM corresponds to the chemisorption of water with no extra-stabilization of the chemisorbed oxygen. Thus, the reaction energy becomes endothermic (+0.24 eV) and $\Lambda_{\Lambda E}^{sol}$ =0.33 eV.

After having discussed the most affected case, we now turn to the general trends that can be deduced from Figure 3. The contrasting behavior of $\Lambda^{sol}(CCM)$ for the C-OH and C-O scissions leads us to separate them into two subsets. The C-OH gathers the cases where the oxygen is involved in an hydroxyl group while the C-O set collects the other fragments.

While the C-O scissions are clearly hindered with $\Lambda_{\Lambda E}^{sol}(CCM)$

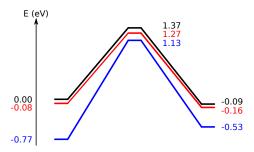


Fig. 4 Reaction pathway for the C-O scission of $\mathrm{CH_3CH_2}$ –O to $\mathrm{CH_3CH_2}$ and O in vacuum in black, in PCM in red and in CCM in blue. The reference is the pristine slab, the chemisorbed initial state and an isolate water molecule, all in vacuum. Corresponding structures are shown in Figure 5 and Figure 6

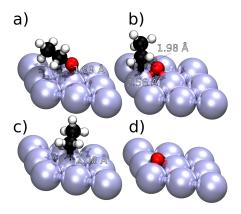


Fig. 5 Vacuum optimized structures a) Initial state $\rm CH_3CHO$ b) Transition state $\rm CH_3CH_2-O$ c) Final state $\rm CH_3CH_2$ d) Final state O

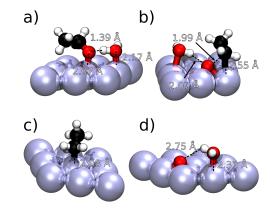


Fig. 6 CCM optimized structures a) Initial state CH_3CH_2O b) Transition state CH_3CH_2-O c) Final state CH_3CH_2 d) Final state O

and $\Lambda_{E_a}^{sol}(CCM)$ being strongly positive (0.12 eV and 0.18 eV on average respectively), the C-OH scissions are facilitated by the explicit presence of a H-bonded molecule, with an average $\Lambda_{\Delta E}^{sol}(CCM)$ of -0.43 eV. This opposite effect of the micro-solvation can be traced back to the solvation of the resulting OH fragment (C-OH set) vs. atomic O (C-O set) that are generated from the bond breaking. OH-H₂O exhibits a very stabilizing interaction (-0.72 eV gain in energy) to be compared with the adsorption of water from bulk PCM to Pt/water interface (-0.35 eV). Conversely, the atomic O fragment has no interaction with the added water molecule since the explicit description of water leads to a gain of only -0.37 eV.

For C-H $_{\alpha}$ scissions, the average $\Lambda_{E_a}^{sol}(CCM)$ and $\Lambda_{\Delta E}^{sol}(CCM)$ of 0.07 and 0.16, respectively, indicate that the inclusion of a water molecule to solvate the hydroxyl or the oxygen group, hampers this bond breaking process. As already discussed in previous studies, ^{43,44,64} this effect can be related to a stronger constraint on the fragment, limiting the access to the ideal geometry to break the C-H bond when it is in α position to a hydroxyl group.

Splitting the C-C bond responds very similarly to the inclusion of an explicit water molecule as the C-H $_{\alpha}$ scissions. Especially on $\Lambda_{\Delta E}^{sol}(CCM)$ the larger distance between the whiskers of the box plot evidence that the variability is larger than for the other sets. This can be rationalized by the competition between the geometric constraints, in analogy to breaking the C-H $_{\alpha}$ and an electronic stabilization of the oxygenated fragments by CCM.

Last, the O-H rupture was slightly but clearly disfavored by PCM compared with the situation in gas phase. With the inclusion of a water molecule that is H-bonded to this hydroxyl group, the situation drastically changes, with reaction energies and activations energies that are noticeably more favorable (average $\Lambda_{E_a}^{sol}(CCM)$ and $\Lambda_{\Delta E}^{sol}(CCM)$ of -0.21 and -0.11 respectively), in line with previous studies. ^{42–44,64}

3.4 BEP relations

We now turn to the important question of the impact on the BEP relationships of the observed modifications in bond scissions induced by the solvent model. The regression parameters of the BEP relations are collected in Table 1. We will mainly base our discussion on the mean average error (MAE) and maximal error (MAX). R^2 is not always the most suitable figure of merit for the prediction using BEP relations. For instance, for early transition states such as O-H scissions, the activation energy does not depend much on the reaction energy (low α). As a consequence, the activation energy is rather a constant than a function of the reaction energy, which results in a low R^2 . Nevertheless, the prediction of the activation energies is particularly simple, since it is a constant (β).

To start with, let us consider the O-H bond scission. Figure 7 gathers the BEP relationships in vacuum, PCM and CCM. The three lines are almost parallel and clearly separated. The corresponding equations are provided in Table 1. The apparent large scattering of the data is due to the fact that the deviations from the linear model (MAX of \sim 0.05 eV) is of the order of magnitude as the range of activation energies (\sim 0.1 eV). As expected

based on the previous discussion, the line matching the PCM data is above the vacuum one, while the line matching the CCM is below. However, this behavior is not general. Despite the increase in the activation and reaction energies observed for C-H $_{\alpha}$ bond scissions with CCM (Figure 3), the BEP is only slightly tilted, but not significantly shifted compared to vacuum or PCM (Figure 8). Provided the dispersion of the data especially in the low activation energy region, the three lines can even be considered identical. In other words, both solvent models on average influence the reaction and the activation energy of a C-H $_{\alpha}$ to the same extent. The same conclusion can be drawn regarding the other dissociations as shown in Figure S3 and Figure S4. Nevertheless, since the thermodynamics can still be influenced by the solvation model, this does not imply that solvation can be ignored for these reactions, as indicated earlier (Figure 3).

The O-H scission BEP shows the best quality (see Table 1), whatever the environment taken into consideration (vacuum, PCM, CCM). The mean absolute error (MAE) is lower than 0.03 eV and the maximal error (MAX) is under 0.1 eV. The slope α is lower than 0.5, indicating that the O-H rupture is an early process, with a transition state that is closer to the initial state than the final state. Hence, for these steps, the stability of the product does not significantly influence the catalytic activity. The C-H dissociation shows a larger scattering of the data (Figure 8), increasing the MAE and the MAX (> 0.1 eV and > 0.3 eV respectively). The steps with a low activation energy are liable for those large errors. They correspond to the C-H α scission in alkoxy species, e.g., ethoxy, CH₂CH₂O. Since the formation of these alkoxy species on Pt is rather unfavorable, accurately describing their decomposition through fast C-H scission is not critical in a micro-kinetic modeling. Removing these species leads to a BEP relation with a clearly improved correlation and a MAE lower than 0.1 eV (Figure S1), a relation that could be used to predict the C-H scission in alcohols with a good accuracy while activation of the C-H scission in alkoxy could be assigned a low activation barrier.

The regressions for C-O dissociations are overall of lower quality compared with the ons for O-H or C-H. With a MAE of around 0.30 eV and MAX higher than 0.75 eV, it shows a larger intrinsic uncertainty. However, the relative error is still around 20% since the average activation energy is around 1.9 eV for the scission of C-O bonds while it is around 0.7 eV for ones of O-H and C-H bonds. Similarly, the correlation coefficient is in the same ballpark as for the other BEPs. Since the C-O scission exhibited contrasted sensitivity to CCM depending on the oxygen fragment generated (OH or atomic O, see Figure 3), two separate regressions were performed for the C-OH and C-O subset (see definition earlier). They are represented in Figure S4. The quality is noticeably improved for the C-O subset compared with the original C-O(H) set that includes also the C-OH (MAE around 0.15 eV and MAX lower than 0.5 eV, see Table 1). Yet, the C-OH scissions are still widely dispersed and show a greater sensitivity to the inclusion of an explicit H-bonded molecule. Last, the C-C scissions exhibit a regression quality similar to C-O, with an average error of \sim 0.25 eV and a maximal error \sim 0.55 eV on the prediction of activation barriers that range from 0.4 eV to 2 eV. The quality of the BEP relation is the same for the vacuum, PCM and CCM.

Table 1 BEP regression parameters using the data obtained in vacuum, PCM and CCM separately: α , the slope ; β , the intercept in eV ; Mean Absolute Error (MAE, eV) ; Maximal error (Max, eV).

Bond	Solvent model	α	β	R ²	MAE	Max
О-Н	vacuum	0.20	0.75	0.47	0.03	0.05
0 11	PCM	0.22	0.80	0.63	0.02	0.05
	CCM	0.24	0.67	0.49	0.03	0.08
$\overline{\text{C-H}_{lpha}}$	vacuum	1.36	0.98	0.56	0.12	0.29
	PCM	1.26	0.95	0.61	0.12	0.33
	CCM	0.98	0.93	0.33	0.16	0.42
C-C	vacuum	0.77	1.44	0.50	0.24	0.52
	PCM	0.74	1.45	0.43	0.24	0.56
	CCM	0.47	1.47	0.28	0.24	0.56
C-O(H)	vacuum	0.71	1.30	0.65	0.29	0.76
	PCM	0.73	1.29	0.68	0.28	0.73
	CCM	0.81	1.37	0.78	0.27	0.53
C-O subset	vacuum	0.79	1.44	0.92	0.15	0.25
	PCM	0.80	1.42	0.93	0.15	0.24
	CCM	0.82	1.50	0.89	0.17	0.39
C-OH subset	vacuum	0.66	1.12	0.53	0.35	0.54
	PCM	0.69	1.13	0.57	0.35	0.53
	CCM	0.66	1.29	0.68	0.24	0.56

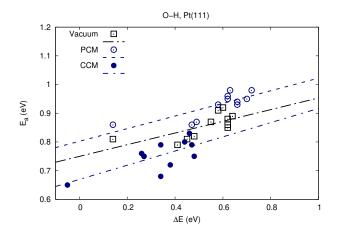


Fig. 7 BEP relationships for the O-H scission in vacuum, PCM and CCM.

Here again, the quality is improved removing the fragments that contain an hydroxyl function (C-COH scission), with a decreased mean absolute error of 0.2 eV and and maximal error of 0.4 eV (see Figure S2). However, prediction of the complementary set (C-CO scission) is worse with a mean absolute error of 0.32 eV and a maximal error of 0.60 eV. Thus, we propose to keep all C-C scissions in a unique set.

3.5 Predicting the solvation enhancement

The inclusion of the water solvent can strongly modify the predicted reaction energies and activation energy, especially when using CCM. However, placing an explicit solvent molecule increases the computational and human burden and remains arbitrary to a certain extent. Thus, it would be highly desirable to predict the reaction energies and activations energies in water from the ones obtained in vacuum or at least from the ones obtained using PCM, which is much easier to set up. In Figure 9, we show that the reaction energies in water using CCM nicely correlate with the ones in vacuum or PCM for the C-OH scissions, opening the door to a simplified prediction of improved reaction

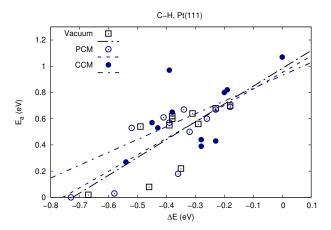


Fig. 8 BEP relationships for the C-H $_{\alpha}$ scission in vacuum, PCM and CCM.

Table 2 Regression parameters for the prediction of the CCM reaction energy (ΔE^{CCM}) from the one obtained using vacuum (ΔE^{vac}): s, the slope ; t, the intercept in eV ; Mean Absolute Error (MAE, eV) ; Maximal error (Max, eV).

		\mathbb{R}^2	MAE	Max
0.77	-0.07	0.55	0.09	0.18
0.95	0.05	0.72	0.06	0.14
1.03	0.09	0.86	0.12	0.29
0.88	0.20	0.96	0.09	0.32
1.08	-0.49	0.98	0.09	0.14
	0.95 1.03 0.88	0.95 0.05 1.03 0.09 0.88 0.20	0.95 0.05 0.72 1.03 0.09 0.86 0.88 0.20 0.96	0.95 0.05 0.72 0.06 1.03 0.09 0.86 0.12 0.88 0.20 0.96 0.09

energies. This behavior is typical and was found for all the other sets (see Figures S3-S14). The corresponding regressions parameters are given in Table 2 and lead to MAE of about 0.1 eV.

This scaling relation between vacuum and CCM can be used to predict ΔE^{CCM} . In combination with the BEPs we have established, it can predict activation energies that include the microsolvation effect. This approach results in an hybrid BEP that predicts ΔE_a^{CCM} from ΔE^{vacuum} . The regression parameters of this hybrid BEP are reported in Table 3. The quality is remarkably good, i.e., on par or better than the original BEPs of Table 1 at a minimal computational cost.

4 Conclusions

In this study, we have investigated the impact of continuum solvation model (PCM) and the explicit inclusion of a water molecule (CCM) on more than 40 typical bond scissions in alcohols on Pt(111). While the PCM has a limited impact on the activation and reaction energies, the CCM strongly favors O-H and C-OH scissions while it prevents C-O and to a lesser extent C-

Table 3 Hybrid BEP $(E_a^{CCM}=\alpha_H.\Delta E^{vac}+\beta_H)$ regression parameters: α_H , the slope; β_H , the intercept in eV; Mean Absolute Error (MAE, eV); Maximal error (Max, eV)

Bond	α_H	β_H	\mathbb{R}^2	MAE	Max
O-H	0.18	0.66	0.38	0.03	0.07
$C-H_{\alpha}$	1.34	1.13	0.50	0.13	0.36
C-C	0.64	1.53	0.42	0.21	0.54
C-O	0.71	1.68	0.82	0.21	0.39
C-OH	0.73	0.95	0.71	0.23	0.53

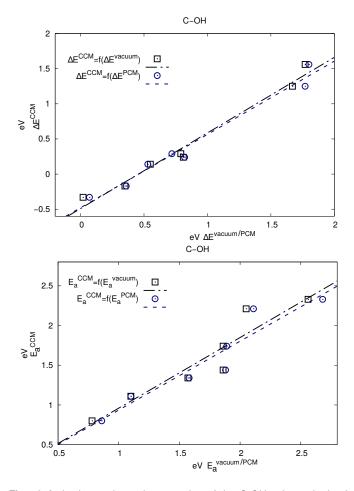


Fig. 9 Activation and reaction energies of the C-OH subset obtained with the CCM model in function of the activation and reaction energies obtained in vacuum (squares) and activation and reaction energies obtained in PCM (dots) and the corresponding regression lines. The regression parameters are reported in Table 2

C and C-H α . In other words, the presence of water may strongly modify the reaction paths followed during alcohol decomposition at the Pt(111)/water interface compared with the Pt(111)/gas interface. Noticeably, the slight effects induced by the PCM are opposite to those obtained with CCM, suggesting that the implicit solvation model is not adapted to capture the effect of an H-bonding water molecule. The impact of the solvent can also be seen in the resulting BEP relationships. For instance we obtained three parallel lines for the O-H scission in vacuum, PCM and CCM. The quality of these BEP relationships is not strongly affected by the inclusion of the solvent, whatever the model. Somewhat surprisingly, the reaction energy at the CCM level can be estimated through a linear regression based on vacuum DFT values. Hence, vacuum computations in combination with simple corrections are able to capture the important H-bonding effect. This lead us to propose a hybrid approach were reaction energies of interest can be easily evaluated at the Pt(111)/vacuum interface, then extrapolated at the CCM level and used in a hybrid BEP that relates the activation energy at the Pt(111)/water interface and the reaction energy at the Pt(111)/vacuum interface. The quality of the later is on par and sometimes even better than the BEP relationships obtained at the CCM level. This is a promising strategy to improve the quality of the prediction of activation and reaction energies of extensive reaction networks such as in the aqueous phase reforming of alcohols.

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