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Stable Solid Molecular Hydrogen above 900 K from a Machine-Learned Potential Trained with Diffusion Quantum Monte Carlo

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We survey the phase diagram of high-pressure molecular hydrogen with path integral molecular dynamics using a machine-learned interatomic potential trained with quantum Monte Carlo forces and energies. Besides the HCP and *C2/c-24* phases, we find two new stable phases both with molecular centers in the *Fmmm-4* structure, separated by a molecular orientation transition with temperature. The high temperature isotropic *Fmmm-4* phase has a reentrant melting line with a maximum at higher temperature (1450 K at 150 GPa) than previously estimated and crosses the liquid-liquid transition line around 1200 K and 200 GPa.

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Introduction.—Experimental methods for probing the phase diagram of high-pressure hydrogen are limited. At room temperature and below, the diamond anvil cell (DAC) has allowed exploration for pressures up to roughly 450 GPa [1–3]. However, the small size of the cell and fragility of the sample limit experimental probes to low-power optics such as infrared and Raman spectroscopy [4]. Hydrogen weakly scatters x rays [5], making structural determination difficult. Recently, direct measurement of the structure of solid molecular hydrogen has been achieved up to pressures of 254 GPa [5,6]. At temperatures below 100 K, pressures above 400 GPa can be achieved in DAC, but x-ray structural determination is not yet available. At higher temperatures, shock wave compression methods have achieved higher pressures, but due to the transient nature of these experiments, acquiring and analyzing shock-wave data is challenging. Notably, one cannot directly measure temperature, which may cause difficulty interpreting results [7–9]. Given the experimental difficulties, accurate simulations are necessary to inform and complement experimental works [10].

Simulation of high-pressure hydrogen requires accurate methods both in the description of the electronic ground-state Born-Oppenheimer (BO) potential energy surface (PES) and the inclusion of nuclear quantum effects beyond the harmonic approximation. Many calculations have been performed on structures found with density functional theory (DFT) based random structure searches [11]. The Perdew-Burke-Ernzerhof (PBE) exchange-correlation functional often used in DFT studies incorrectly predicts

that the molecular structures in phase III are metallic [12]. Benchmarks using diffusion Monte Carlo (DMC) [13] have established that the van der Waals density functional (vdW-DF1) is the best compromise for molecular hydrogen at pressures above 100 GPa. But exploring large pressure and temperature ranges using first-principles methods is so time consuming that establishing convergence with respect to supercell size and trajectory length is difficult to achieve.

Calculations using the most accurate simulation technique, coupled electron ion Monte Carlo (CEIMC), which calculates electronic energies using explicitly correlated wave functions, find that solid molecular hydrogen remains stable at higher temperatures than indicated by simulations based on PBE forces and at higher temperatures than experimental estimates. However, the number of atoms and the length of trajectories accessible by CEIMC is limited by computer resources. This is particularly worrying for disoriented molecular phases: alternative algorithms for studying the melting line are needed.

In recent years, machine-learned (ML) interatomic potentials have emerged as a promising tool providing a balance between accuracy and efficiency thus allowing for accurate but less expensive calculations [14–16] that can address the temporal and spatial limitations of first-principle simulations. ML methods have recently been applied to dense hydrogen. However, there are several conflicting theoretical results regarding dissociation, melting, and the critical point obtained with the various simulation methods [17–20].

In this Letter, we perform large-scale simulations of molecular hydrogen using quantum protons with a ML

approximation to the DMC-BO-PES for pressures from 50 to 220 GPa.

Training data.—Our training data consist of about 10^5 hydrogen configurations, each with 96 protons, obtained from classical molecular dynamics (MD), path integral molecular dynamics (PIMD) and CEIMC simulations, the majority of which are of molecular hydrogen. Their pressures and temperatures are shown in Supplemental Material (SM) [21]. The total energy, atomic forces, and cell stresses are evaluated with DFT using both PBE and vdW-DF1 functionals. For a subset of approximately 20 000 configurations the energy and forces are estimated by DMC using the Chiesa-Ceperley-Zhang estimator [27] to a statistical error bar of 0.1 meV and 130 meV/Å, respectively. As we show in SM [21], the DMC configurations are overcomplete so that the statistical error in the forces of the ML models is about 80 meV/Å.

Hierarchical Δ learning.—The potential energy model is a sum of three terms:

$$E = E_{\text{pair}} + \Delta E_{\text{DFT}} + \Delta E_{\text{DMC}}, \quad (1)$$

where E_{pair} is the contribution from a proton pair potential using only the repulsive part of the exact BO energy of a H_2 molecule [35]. This allows an accurate description of the internal motion of a single H_2 without having to learn it from the training data.

The next two terms in Eq. (1) are determined by a ML model. ΔE_{DFT} is a fit to DFT energy, forces, and stresses minus the pair potential contribution over all of the configurations. This fit should capture the overall band structure and electron correlation effects as described by the density functional. Finally, ΔE_{DMC} is determined by fitting $E - E_{\text{pair}} - \Delta E_{\text{DFT}}$ to the subset of configurations that have quantum Monte Carlo (QMC) forces. The difference between the DFT energy and the QMC energy is expected to be small and a smooth function of positions and thus requires a simpler network. This procedure makes optimal use of the computationally expensive and noisy QMC forces. Final results of the DMC trained model are fairly independent of the DFT functional used whether PBE or vdW-DF1. The results which we present here used vdW-DF1 functional.

To construct the two ML models in Eq. (1), we use the deep potential molecular dynamics (DPMD) [14–16] framework: the energy is a sum of atomic contributions, similar to the embedded atom method (EAM) [36,37]. The contribution from one atom is calculated from its local atomic environment, represented by a real-valued vector that is invariant with respect to translation, rotation, and permutation. See SM [21] for further model and hyperparameter details.

PIMD simulation.—The nuclear quantum effects of protons are substantial, thus PIMD simulations using a NPT ensemble, where the number of particles N , pressure

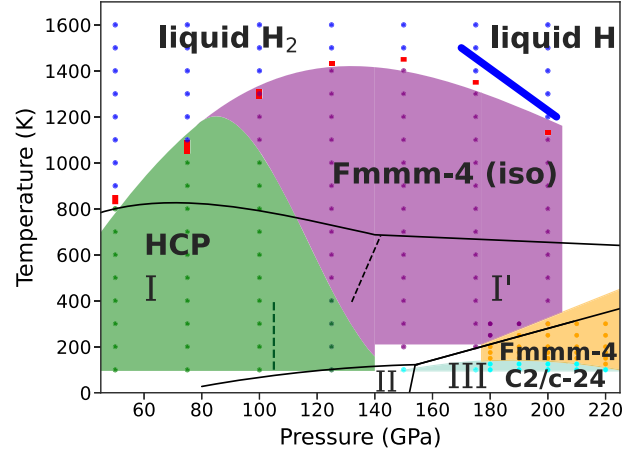


FIG. 1. Phase diagram of dense hydrogen. The dots indicate the (P, T) values where we ran PIMD simulations using a DMC-trained DPMD interatomic potential. Colors indicate the identified phase: dark blue (liquid), green (HCP), purple ($Fmmm-4$ isotropic), orange ($Fmmm-4$ oriented), and cyan ($C2/c-24$). The vertical green dashed line at 105 GPa indicates a crossover within HCP to an in-plane orientation. The thick blue line is the estimate of the transition from liquid H_2 to liquid H from Ref. [10]. Red bars are estimates of the melting temperature from two-phase coexistence simulation. The black lines are experimental estimates of phase boundaries, solid lines for the melting, dashed line for I-I' transition [39].

P , and temperature T are controlled during the MD simulation, are used to explore the phase diagram of molecular hydrogen with the DMC-trained potential. The melting line shown in Fig. 1 was obtained using a two-phase method [38]: the PIMD simulations started from a half-liquid half-solid configuration consisting of 3072 protons. See SM [21] for details of the analysis and comparison between MD and PIMD results.

Equation of state.—Comparison of our results with x-ray scattering experiments of molecular hydrogen up to 225 GPa at room temperature [5] is shown in Fig. 2. The overall comparison of the equation of state is excellent. Experiment identified the structure as hcp, but we find the hexagonal symmetry is broken above 150 GPa as discussed next. The c/a ratio decreases with increasing pressure in both simulation and experiment. We find an anisotropy developing in the molecular orientation as the c/a ratio deviates from the closed packed limit.

Solid structures.—To identify the stable crystal structures in the PIMD trajectories we use several techniques: the molecular center of mass structure factor, the protonic structure factor, and the molecular bond orientational distribution. Snapshots are also quenched to determine the ideal crystal structure and then the crystal symmetries are found using the FINDSYM package [41].

As shown in Fig. 3 we find the dominant structure above 400 K and 120 GPa to be the isotropic $Fmmm-4$ with 2 molecules per unit cell without preferential orientation.

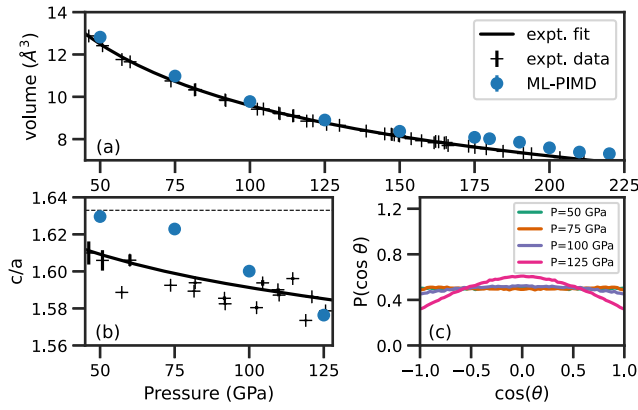


FIG. 2. (a) Unit cell volume and (b) c/a ratio, vs pressure in GPa both at 300 K. Blue dots are the results of PIMD simulation. The crosses are experimental XRD measurements Ref. [40] and the lines are fits to experimental data. (c) The distribution of the molecular angle with the basal plane at various pressures within HCP at 300 K from PIMD.

In the HCP structure, a molecular center of the second layer lies above the center of the equilateral triangle formed by three molecular centers in the first layer, while in $Fm\bar{m}m-4$, the molecular center of the second layer lies above the edge of that triangle. The hexagonal symmetry of the basal plane is broken so that the angle between the two primitive vectors becomes $70.5^\circ \pm 0.2^\circ$ throughout this phase. For $T \leq 400$ K the molecules become oriented in the basal plane as shown in Fig. 3. For $T < 200$ K we find the structure to be $C2/c-24$, one of the candidate structures predicted in [11]. This is again an HCP lattice of molecular centers with oriented molecules; there are four distinct layers with a total of 12 molecules per unit cell.

Phase diagram.—Figure 1 shows the phase diagram of dense hydrogen using the DMC-trained ML model

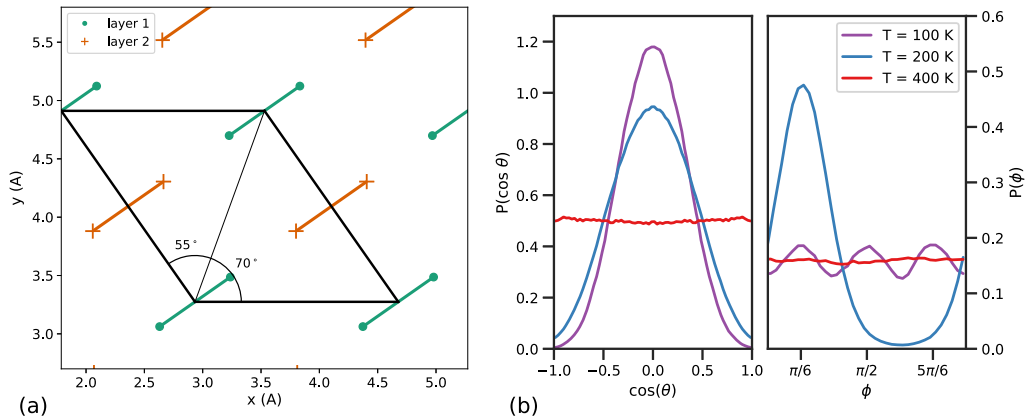


FIG. 3. (a) The oriented $Fm\bar{m}m-4$ structure with layer 1 (green) and layer 2 (orange). The primitive cell (shown in black) contains one molecule per layer all lying in the basal plane. Layer 2 is shifted by half a primitive lattice vector from the HCP site. Also the hexagonal symmetry within the basal plane is broken. (b) Histograms of molecular orientation at $P = 200$ GPa, where θ is the angle with respect to the basal plane and ϕ is the azimuthal angle. There are three distinct in-plane orientations in the $C2/c-24$ structure at 100 K, one orientation in the $Fm\bar{m}m-4$ structure at 200 K. Orientation order is absent in the isotropic $Fm\bar{m}m-4$ phase at 400 K and above.

compared to experimental results of Ref. [39] based on Raman measurements. At 100 K, we find the HCP phase to be stable up to 150 GPa. Below 100 GPa, the molecules are isotropic, but as the pressure increases their orientation begins to favor the basal plane as shown in Fig. 2(c). The azimuthal orientation remains isotropic.

We find a solid transition to an orientationally ordered $C2/c-24$ structure at around 150 GPa. This change has been seen in a recent experiment [4]. The $C2/c-24$ structure is absent from our training set, so its spontaneous emergence demonstrates the predictive power of the model. A previous ground-state QMC of a perfect crystal puts the transition to the $C2/c-24$ at 250 GPa [12] while for $150 \text{ GPa} \leq P \leq 250 \text{ GPa}$ $P2_1/c-24$ is the ground state. We attribute this difference to treatment of the nuclear quantum effect: we observe large-amplitude librations of the H_2 molecules [see Fig. 3(b)] which are inadequately treated within the vibrational self-consistent field method underlying the prediction of Ref. [12].

We find that for $T > 150$ K and $P > 160$ GPa, the $C2/c-24$ crystal transitions into an oriented $Fm\bar{m}m-4$ structure. Even in this phase, the molecules have large librations as shown in Fig. 3(b). At even higher temperature, under conditions similar to the experimental phase I'-III boundary, the orientational order is lost in favor of an isotropic $Fm\bar{m}m-4$ phase (see Fig. 3). We see a sharp orientational transition in temperature and observe hysteresis, signaling the first-order nature of the transition.

Previous structure searching [11] did not favor the $Fm\bar{m}m-4$ phases because they are stable only above 200 K, the molecules are highly disoriented, so that estimates of the zero point energy based on a perturbation expansion about a perfect crystal is difficult, and density functionals may not be precise enough to distinguish between the hydrogen structures. Note that in this region

of temperature and pressure (200 K and 150 GPa), *five* different crystal phases are relevant. Further studies will be needed to predict their stability and understand the experimental observations.

Molecular melting curve.—The ML model finds the molecular HCP solid structure melts at 800 K at 50 GPa, consistent with experiment [39] and the melting temperature increases to about 1100 K at about 90 GPa. Above this pressure, we find that the HCP structure transitions to the isotropic *Fmmm-4* phase before melting, with a maximum melting temperature 1450 K at 150 GPa. Our result suggests the presence of a triple point HCP-*Fmmm-4*(iso)-liquid at around 90 GPa and 1100 K and a HCP-*Fmmm-4*(iso) transition line declining rapidly with pressure. Our prediction of melting is in striking disagreement with experiments. The experimental estimates of melting for pressures above 100 GPa are indirect: they do not show absence of long-range density order or shear stress, but rather rely on a change in the Raman spectra [39,42,43], in the rate of heating of the sample [44], or of its dielectric properties [45]. The increase in melting temperature that we find can be attributed to the higher stability of the isotropic *Fmmm-4* structure with respect to the liquid.

Figure 4 shows the predictions of the melting line from DMC-trained and PBE-trained models using both classical and quantum proton simulations. Above 100 GPa the melting temperature is very sensitive to the underlying electronic PES or the mass of the protons. Closed symbols represent quantum protons, while open symbols classical protons. Figure 4 also reports previous melting calculations

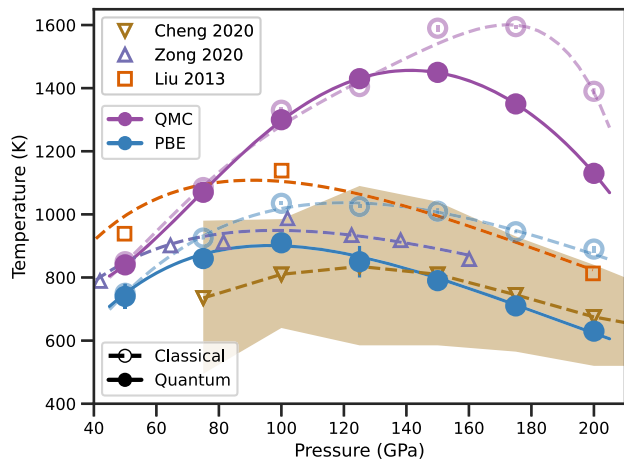


FIG. 4. Comparison of melting lines from various calculations. Circles are our two-phase MD results using ML constrained by data from either PBE forces (blue) or QMC forces (purple). Open symbols represent classical protons, closed symbols quantum protons. Previous PBE-DFT predictions, all using classical protons, are reported: gold downward triangles Ref. [18], light purple upward triangles Ref. [17], orange squares Ref. [46]). Lines (either continuous or dashed) are guides for the eyes. The shaded region is the uncertainty estimated in Cheng *et al.* [18].

based on the PBE functional, either by DFT-MD [46] or by DFT-trained ML potential [17,18]. Our ML predictions of classical protons based on the PBE functional (blue open circles) are in reasonable agreement with previous studies giving us confidence in the reproducibility of the computational procedure. We observe a divergence of the classical and quantum melting curves around 120 GPa consistent with the DFT calculations of the D_2 and H_2 melting curves by Caillabet *et al.* [47].

The melting maximum is at about 150 GPa for the DMC-trained model and at about 100 GPa for the PBE-trained model with either quantum or classical protons. Melting based on vdW-DF1 trained model (see SM [21]) lies in-between the PBE and DMC models. The nuclear quantum motion lowers the melting temperature by roughly 100-200 K above 100 GPa. These large differences highlight the importance of accurately treating both the electronic structure and nuclear quantum effect in dense hydrogen [48].

Figure 1 includes the predicted liquid-liquid phase transition (LLPT) line from CEIMC [10]: the line intersects the melting line of the new isotropic phase at 1200 K and 200 GPa. Both the ML model and CEIMC predict dissociation as the LLPT line is crossed. However, the older CEIMC calculations, though the PES is more accurate (CEIMC does not make the assumptions that ML does), were done with simulations of 54 protons in a fixed orthorhombic volume: the boundary conditions and number of iterations did not allow the system to find the *Fmmm-4* phase. A more complete comparison of the present results with those of CEIMC for the LLPT will be presented elsewhere. The present calculation corrects these biases and illustrates the power of combining QMC calculations with ML techniques. However, additional research is required to make a definitive calculation of the hydrogen phase diagram. In particular, the QMC forces have been evaluated for systems with 96 protons and the ML descriptors are short ranged (see SM [21]). Examination of the effect of the long-ranged interactions is needed. Of course, neither the ML description nor the QMC calculations are exact but both can be systematically improved. Finally, simulations were done without taking into account proton antisymmetry. This is expected to be a small effect at temperatures above 500 K. Well-established techniques are available for its inclusion [49].

The data for this Letter are publicly available at [50].

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