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3	I	Partitioning of Sulfur between Solid and Liquid Iron under Earth's Core Conditions:
4		Constraints from Atomistic Simulations with Machine Learning Potentials
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ABSTRACT

Partition coefficients of light elements between the solid and liquid iron phases are crucial for 23 uncovering the state and dynamics of the Earth's core. As one of the major light element candidates, 24 sulfur has attracted extensive interests for measuring its partitioning and phase behaviors over the last 25 several decades, but the relevant experimental data under Earth's core conditions are still scarce. In 26 this study, using a toolkit consisting of electronic structure theory, high-accuracy machine learning 27 potentials and rigorous free energy calculations, we establish an efficient and extendible framework 28 for predicting complex phase behaviors of iron alloys under extreme conditions. As a first application 29 30 of this framework, we predict the partition coefficients of sulfur over wide range of temperatures and pressures (from 4000 K, 150 GPa to 6000 K, 330 GPa), which are demonstrated to be in good 31 agreement with previous experiments and *ab initio* simulations. After a continuous increase below 32 33 ~250 GPa, the partition coefficient is found to be around 0.75±0.07 at higher pressures and are essentially temperature-independent. Given these predictions, the partitioning of sulfur is confirmed 34 to be insufficient to account for the observed density jump across the Earth's inner core boundary and 35 36 its roles on the geodynamics of the Earth's core should be minor.

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38 Keywords:

Partition coefficient; Sulfur; Earth's core; First principles; Machine learning; Light elements; Density
Functional Theory

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1. Introduction

The Earth's core is constituted of some light elements in addition to the major components of iron 44 and nickel (Allegre et al., 1995; Birch, 1964; Mcdonough and Sun, 1995). The prime candidates for 45 these light elements include sulfur, oxygen, silicon, carbon, and hydrogen (Hirose et al., 2013; Li and 46 Fei, 2007; Poirier, 1994). It is recognized that they redistribute across the Earth's inner core boundary 47 (ICB), with more in the outer core (~5-10 wt%) and less in the inner core (~2-3 wt%). The difference 48 in light element contents between the outer and inner core helps explaining the density jump across 49 the ICB (Cao and Romanowicz, 2004; Masters and Gubbins, 2003), anchoring the temperature profile 50 51 in the whole Earth's core (Morard et al., 2014) and inducing compositional stratification in the inner core (Alboussiere et al., 2010). More importantly, compositional buoyancy created by the 52 redistribution of light elements turns out to be the principal energy source for powering the geodynamo 53 54 in the outer core (Stacey and Stacey, 1999) and may be critical for driving the convection in the inner core (Gubbins et al., 2013). 55

The distributions of light elements across ICB are largely controlled by their partition coefficients 56 57 between the coexisting liquid and solid phases. With several decades' efforts of high-pressure experiments, current knowledge of the partitioning or more broadly phase behaviors of iron alloys has 58 been significantly extended but is still far from adequate (Morard et al., 2014). For sulfur, the focus of 59 this study, the eutectic melting phase relations have been determined by a number of experiments 60 (Chen et al., 2008; Chudinovskikh and Boehler, 2007; Kamada et al., 2012; Kamada et al., 2010; Li et 61 al., 2001; Morard et al., 2014; Morard et al., 2008; Mori et al., 2017; Stewart et al., 2007; Terasaki et 62 al., 2011; Yokoo et al., 2019). Most of these experiments are done at pressures lower than 60 GPa and 63 the data under the real core pressures (>140 GPa) are scarce. Although the highest-pressure record in 64

these experiments has reached 254 GPa (Mori et al., 2017), it is still some distance away from true ICB condition (330 GPa). With these experimental constraints, people find a general trend of decreasing sulfur contents in the eutectic liquids and increasing solubility of sulfur in the solid solutions at higher pressures, which means an increasing tendency for the partition coefficients of sulfur (Kamada et al., 2012; Morard et al., 2014; Yokoo et al., 2019).

Complementary to high-pressure experiments, first principles simulations (mostly based on 70 Density Functional Theory (DFT)) provide an alternative route to unravel the partitioning and phase 71 behaviors of iron alloys under extreme conditions. While early DFT simulations have already predicted 72 73 the partition coefficient of sulfur under ICB conditions (Alfe et al., 2002a, 2003, 2007), further work is needed to establish consistency with the experimentally measured values due to the gap in the T-P74 regime accessible to the experiments and first principles simulations. The reason for the lack of 75 76 extensive first principles simulations for the binary or more complex iron alloying systems is the extremely high computational cost of such simulations, typically taking many orders of magnitude 77 more than corresponding simulations employing empirical atomistic potentials, which however are 78 79 generally not accurate enough to give meaningful predictions at extreme conditions.

In this study, we are taking a different approach and use a new generation of *surrogate models* in place of DFT, interatomic potentials made using machine learning techniques (Rupp, 2015). The basic idea of this approach is to get a relatively small amount of DFT data (total energies, forces and stresses from a short MD trajectory) and construct nonparametric potentials that approximate the true *ab initio* Born-Oppenheimer potential energy surface very closely and then carry out the extensive sampling using the potentials. With newly constructed highly accurate potentials for the Fe-S binary system, combining with rigorous free energy calculations, our efforts in this study not only extend the partition coefficients of sulfur to the overlapping T-P regime of experiments and simulations but also provide a general and extendible framework for effectively predicting the phase behaviors of multi-component iron alloying systems at extreme conditions.

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2. Methods

92 2.1. Machine learning potentials

We use the Gaussian Approximation Potential (GAP) (Bartok et al., 2010), essentially a kernel 93 ridge regression method (Kung, 2014). This is just one of a class of recently popularized machine 94 95 learning methods for creating nonparametric interatomic potentials, which has been shown to be very successful in tackling difficult materials modelling problems, ranging from investigating the structure 96 of amorphous materials (carbon (Deringer et al., 2017, 2019), silicon (Bartók et al., 2018)), the 97 98 mechanics of metals (tungsten (Szlachta et al., 2014), iron (Dragoni et al., 2018)) to molecular liquids such (water (Bart ok et al., 2013a), methane (Veit et al., 2019). There are many alternatives, using other 99 regression frameworks, such as artificial neural networks (Behler and Parrinello, 2007) and even linear 100 101 regression (Shapeev, 2017; Drautz, 2019). All these methods are improvable, since using more input 102 data typically leads to more accurate potentials, due to the nonparametric nature of the functional forms. The theoretical details of the GAP model can be found elsewhere (Bart ók et al., 2013b; Ceriotti et 103 al., 2018), we only give a high-level description of the key formula here. Assuming the Born-104 Oppenheimer potential energy surface of a set of atoms is a smooth function of the atomic coordinates, 105 we write the total energy as a sum of atomic contributions 106

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$$E = \sum_{i} \varepsilon(\boldsymbol{q}_{i}) \qquad (1)$$

108 where the short-ranged local atomic energy ε is assumed to depend explicitly on the positions of the

atoms within a sphere of radius r_{cut} centered on atom *i* and q_i is a vector representing its local atomic environment. In the GAP framework, Gaussian process regression is used to model ε ,

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$$\varepsilon(\boldsymbol{q}) = \sum_{s} \alpha_{s} K(\boldsymbol{q}_{s}, \boldsymbol{q}) = \mathbf{k}(\boldsymbol{q})^{\mathrm{T}} \boldsymbol{\alpha} \quad (2)$$

Given two atoms *i* and *j*, the kernel function $K(\mathbf{q}_i, \mathbf{q}_j)$ is the expected covariance of the respective local atomic energies $\varepsilon(\mathbf{q}_i)$ and $\varepsilon(\mathbf{q}_j)$, and can be interpreted as a measure of similarity of the two local atomic environments. We choose the Smooth Overlap of Atomic Positions (SOAP) kernel (Bart ók et al., 2013b),

$$K(\boldsymbol{q}_i, \boldsymbol{q}_j) = \sigma_w^2 |\widehat{\boldsymbol{q}}_i \cdot \widehat{\boldsymbol{q}}_j|^{\xi} \quad (3)$$

where the descriptor \hat{q}_i is the rotational power spectrum of the atomic neighbor density, which is a 117 smooth and regular function, invariant to rotation and permutation of like atoms. The key advantage 118 of the SOAP representation is that there are very few empirical parameters needed, basically the cutoff 119 120 distance, the Gaussian smearing of atomic positions (essentially band-limiting the spherical Fourier transform of the neighbor density), and small integer power ξ to which the linear SOAP kernel is raised 121 to, controlling the body order of the resulting force field, such that higher exponents result in higher 122 123 order many-body terms. All the hyper-parameters, including those inherent in the definition of the rotational power spectrum are listed in Table 1. 124

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126 2.2. Thermodynamics for predicting the partition coefficient

For the Fe-S binary systems, the partition coefficient of sulfur and the other melting properties are
determined by the following chemical equilibrium between the liquid and solid phase:

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$$\mu_{\rm S}^{\rm liq}(T,P,c_{\rm S}^{\rm liq}) = \mu_{\rm S}^{\rm sol}(T,P,c_{\rm S}^{\rm sol}) \tag{4}$$

130
$$\mu_{\rm Fe}^{\rm liq}(T,P,c_{\rm S}^{\rm liq}) = \mu_{\rm Fe}^{\rm sol}(T,P,c_{\rm S}^{\rm sol})$$
(5)

where $\mu_{\rm S}$ and $\mu_{\rm Fe}$ are the chemical potential of sulfur and iron respectively, the superscripts liq and sol denote liquid and solid phase respectively and $c_{\rm S}$ is the mole fraction of sulfur in the solution.

Since the chemical potential μ_s diverges logarithmically in the low-concentration limit ($c_s \rightarrow 0$), at each temperature and pressure it is useful to express μ_s as (Alfe et al., 2002a)

135
$$\mu_{\rm S} = k_B T lnc_{\rm S} + \overline{\mu_{\rm S}}(c_{\rm S}) \quad (6)$$

where $\overline{\mu_{s}}(c_{s})$ is well behaved for all concentrations. While many models have been proposed for $\overline{\mu_{s}}(c_{s})$ (or its equivalences), such as various symmetric/asymmetric regular solution models (White, 2013) or those proposed by Ma (2001), the following expansion for $\overline{\mu_{s}}(c_{s})$ is simple but practically meaningful

140
$$\overline{\mu_{\rm S}}(c_{\rm S}) = \mu_{\rm S}^{\dagger} + \lambda_{\rm S}c_{\rm S} + O(c_{\rm S}^2)$$

In line with previous studies (Alfe et al., 2007; Gubbins et al., 2013; Labrosse, 2014), we neglect higher order terms ($O(c_s^2)$) in this study. Therefore, only two parameters of μ_s^{\dagger} and λ_s are involved in the calculations and we obtain the following equation for the chemical potential of sulfur

(7)

144
$$\mu_{\rm S} = k_B T ln c_{\rm S} + \mu_{\rm S}^{\dagger} + \lambda_{\rm S} c_{\rm S} \quad (8)$$

According to the Gibbs-Duhem equation, we can straightforwardly get the chemical potential of iron

147
$$\mu_{\rm Fe} = \mu_{\rm Fe}^0 + (k_B T + \lambda_{\rm S}) ln(1 - c_{\rm S}) + \lambda_{\rm S} c_{\rm S}$$
(9)

148 where μ_{Fe}^{0} is the chemical potential of pure iron at the same temperature and pressure, which is the 149 Gibbs free energy per atom of pure iron.

150 Then the Gibbs free energy (G) of the whole system can be expressed by

151
$$G = N_{\rm S}\mu_{\rm S} + N_{\rm Fe}\mu_{\rm Fe} = N[c_{\rm S}\mu_{\rm S} + (1 - c_{\rm S})\mu_{\rm Fe}] = N\{[c_{\rm S}lnc_{\rm S} + (1 - c_{\rm S})\ln(1 - c_{\rm S})]k_{\rm B}T + (1 - c_{\rm S})m_{\rm S}T +$$

152
$$c_{\rm S}$$
) $\mu_{\rm Fe}^0 + c_{\rm S}\mu_{\rm S}^\dagger + [c_{\rm S} + (1 - c_{\rm S})\ln(1 - c_{\rm S})]\lambda_{\rm S}$ (10)

where N_{Fe} is the number of iron atoms, N_{S} is that of sulfur atoms in the system and $N = N_{\text{Fe}} + N_{\text{S}}$.

Now with the simulated free energies of both liquid and solid phases at several concentrations, we can regress the effective values of $\mu_{\rm S}^{\dagger}$ and $\lambda_{\rm S}$ with Eqn. (10) and the chemical potentials of sulfur and iron can be calculated thereafter with Eqns. (8) and (9). Finally, the partition coefficient of sulfur ($D_{\rm S}$) can be derived from the equilibrium compositions from Eqns. (4) and (5), i.e., $D_{\rm S} = c_{\rm S}^{\rm sol}/c_{\rm S}^{\rm liq}$.

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159 2.3. Free energy calculations

160 In this study, we calculated the Helmholtz free energies (F) using thermodynamic integration, 161 which rigorously relate the free energy (F_1) of the target system with that of a reference system (F_0) 162 by

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$$F_1 = F_0 + \int_0^1 \frac{\partial F_\lambda}{\partial \lambda} d\lambda \quad (11)$$

where λ is a coupling parameter relates the two systems and F_{λ} is the free energy of the system with total energy U_{λ} , which is a hybrid total energy potential with the property of being equal to the reference total energy U_0 for $\lambda=0$ and the target total energy U_1 for $\lambda=1$. If we choose $U_{\lambda}=(1-\lambda)U_0+\lambda U_1$, then the integrand in Eqn. (11) becomes the energy difference as a function of λ , $\frac{\partial F_{\lambda}}{\partial \lambda} =$ $\langle U_1 - U_0 \rangle_{\lambda} = \overline{\Delta U}(\lambda)$, where $\langle \rangle_{\lambda}$ means thermal average in the ensemble generated by U_{λ} .

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170 2.3.1. Liquid phases

For the liquids, we choose the ideal gas at the same temperature and composition as the reference system, i.e.,

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$$F_0^{ideal} = -k_B T \ln\left(\frac{V^{N_{\text{Fe}}+N_{\text{S}}}}{N_{\text{Fe}}!N_{\text{S}}!\Lambda_{\text{Fe}}^{3N_{\text{Fe}}}\Lambda_{\text{S}}^{3N_{\text{S}}}}\right)$$
(12)

where $\Lambda = h/(2\pi M k_{\rm B} T)^{1/2}$ (*h* is the Planck constant, $k_{\rm B}$ is the Boltzmann constant, *M* is the atomic mass

of iron or sulfur and *T* is the temperature) is the thermal wavelength of iron or sulfur. Note that in standard statistical mechanics the Stirling approximation is usually written as $\ln N! \approx N \ln N - N$, but since we are dealing with small values of *N*, the natural logarithm of first-order Stirling's series $\ln N! \approx N \ln N - N + \frac{1}{2} \ln(2\pi N)$ was used instead in the calculation of Eqn. (12) to get more accurate free energy (Arfken, 1985).

180 To integrate from ideal gas to the target system, since the energetic changes are large, especially 181 near the end-points, we used the Gauss-Lobatto quadrature by the following variable transformation

182
$$\lambda(x) = \left(\frac{x+1}{2}\right)^{\frac{1}{1-k}} \quad (13)$$

183 with x spans from -1 to 1. Then the integral in Eqn. (11) becomes

184
$$\int_{0}^{1} \frac{\partial F_{\lambda}}{\partial \lambda} d\lambda = \frac{1}{2(1-k)} \int_{-1}^{1} \overline{\Delta U}(\lambda(x))\lambda(x)^{k} dx \approx \frac{1}{2(1-k)} \sum_{i=0}^{n} \omega_{x_{i}} \overline{\Delta U}(\lambda(x_{i}))\lambda(x_{i})^{k}$$
(14)

185 *k* in Eqns. (13) and (14) is selected to be 0.8, which avoids the endpoint singularity since the 186 $\overline{\Delta U}(\lambda(x_i))\lambda(x_i)^k$ can be safely set to be zero when $\lambda=0$ without loss of accuracy (Dorner et al., 2018). 187 In Fig. A.1(a) of Appendix A, we show a typical transformed integrand as a function of the integration 188 variable *x*.

189 The weight functions ω_{x_i} in Eqn. (14) can be calculated by

190
$$\omega_{x_i} = \frac{2}{n(n-1)[P_{n-1}(x_i)]^2} \quad (15)$$

where P_n are the Legendre polynomials. Practically we include 8 abscissas (n=8) in our calculations of Eqn. (14).

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194 2.3.2. Solid phases

For the pure iron systems, we choose the harmonic *hcp*-lattices at the same temperature and composition as the reference systems, whose free energies can be calculated through lattice dynamics and quasi-harmonic approximation theory (Alf è, 2009). From the harmonic system to the target system,
since the energetic changes are small and vary smoothly with the integration variable (as shown in Fig.
A.1(b) of Appendix A), we use the simpler three-point Gauss-Legendre quadrature by the following
linear transformation

6)

$$\lambda(x) = \frac{x+1}{2} \quad (1)$$

with x spans from -1 to 1. Then the integral in Eqn. (11) becomes

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$$\int_{0}^{1} \frac{\partial F_{\lambda}}{\partial \lambda} d\lambda = \frac{1}{2} \int_{-1}^{1} \overline{\Delta U}(\lambda(x)) dx \approx \frac{1}{2} \sum_{i=1}^{3} \omega_{x_{i}} \overline{\Delta U}(\lambda(x_{i}))$$
(17)

For a solid solution of Fe-S alloy under specific T-P condition, we swapped the sulfur atoms into 204 205 irons and calculated the Helmholtz free energy change at the same volume through thermodynamic integration (Eqn. (11)). In this case, the energy barriers for interchanging the atoms at various lattice 206 sites greatly hinder a full sampling of the phase space through simple molecular dynamic simulations 207 208 and the results would inevitably depend on the initial configurations. To circumvent this problem and to avoid exhaustively sampling all the possible configurations with distinct sulfur orderings, which is 209 infeasibe even considering the reducible symmetries, we used the more efficient hybrid Monte 210 Carlo/molecular dynamics simulations (Widom et al., 2014). Since the swapping of species changes 211 the identities and masses of the atoms, the variations of ideal lattice gas contributions to the free 212 energies, which can be calculated through Eqn. (12), need to be counted in the free energy changes. 213

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3. Simulation details

216 3.1. First principles simulations

To train the GAP models, we carried out a number of density functional theory (DFT) simulations
with VASP, with the projector-augmented-wave method (Kresse and Joubert, 1999). We use the PBE

form of Generalized Gradient Approximation (GGA) (Perdew et al., 1996) with valence electrons of 16 (valence configuration $3s^23p^63d^74s^1$) for iron, which has been demonstrated to closely resemble the all-electron potential and be important for obtaining accurate melting properties (Sun et al., 2018), and 6 ($3s^23p^4$) for sulfur.

Extensive molecular dynamics (MD) simulations have been deployed to sufficiently sample the 223 T-P-X space spanning over 0-25 at. % sulfur concentrations, from 1000 K to 7000 K and from about 224 50 GPa to 500 GPa for the solid phases, and from 4000 K to 10000 K and from about 100 GPa to 500 225 GPa for the liquid phases. Overall 243 simulations have been carried out for the solid phases and 215 226 227 simulations for the liquid phases. The solid solutions were initiated with hexagonal close-packed (*hcp*) structures with 96 atoms in the supercells ($4 \times 4 \times 3$ extension of the unit-cell). At high enough 228 temperatures, the solids would always be melted and the final configurations of these runs were 229 230 adopted to initialize liquid simulations. MD trajectories were propagated in the NVT ensemble with the Nos éthermostat for 3~6 ps. At each time step (1 fs interval), the electronic structure is calculated 231 at the Brillouin zone center with an energy cutoff (ENCUT) of 500 eV and iteration convergence 232 criterion (EDIFF) of 10⁻⁶ eV. 233

From these sampled configurations, we extracted independent configurations every 100-150 steps and re-calculated their energies and forces with much higher precisions. The energy cutoff was increased to 800 eV and the Brillouin zone sampling grid of spacing (KSPACING) was set to be 0.3 $Å^{-1}$, which is well converged as revealed by our benchmarks with finer spacing of 0.2 Å⁻¹. The projection operators were evaluated in reciprocal space (LREAL=.FALSE.). As in the MD simulations, the influences of finite temperature on the electronic free energy were accounted through the Mermin functional (Mermin, 1965) with Fermi-Dirac smearing of electron occupancy (ISMEAR= -1) and an

243 3.2. Training of GAP models

The GAP models were trained through the QUIP code (Bartok et al., 2010) based on the high precision DFT simulation results. Since the potential energy surfaces sampled by these simulations are implicitly a function of the electronic temperature, at each temperature we obtained DFT energies that include electronic-entropy, the forces and stresses from over 1000 configurations and organized them into an extended xyz file for the GAP model training, as explained in Appendix B.

249 We list the main hyper-parameters of trainings for the GAP models in Table 1. In these parameters, the zero points of the energy per atom (e0) of iron and sulfur find their contributions mostly from the 250 electronic entropy and therefore are temperature-dependent. They were evaluated by single-step DFT 251 252 calculations with only one atom in vacuum (without the effects of the periodic images). Finally, to keep models robust at very short interatomic distances, which may not be well sampled in the DFT-253 MD simulations, we explicitly calculated the dimer potentials for pairs of Fe-Fe, Fe-S and S-S and 254 255 used their repulsive part as baselines. Details for a typical GAP model training with the QUIP code can be found in Appendix B. 256

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258 3.3. Atomistic simulations with the GAP models

With the derived GAP models, we carried out three types of atomistic simulations in this study: simple molecular dynamics simulations for pure iron and Fe-S liquids, hybrid Monte Carlo/molecular dynamics simulations for Fe-S solid solutions, and lattice dynamics simulations for pure solid iron to get the reference harmonic free energy. The energetics and forces in all these simulations were calculated with the engine of VASP by invoking a custom-made interface to QUIP. External scripts
were utilized to realize the construction of Metropolis Markov chain in Monte Carlo move and relevant
lattice dynamics calculations.

For simple molecular dynamics simulations, at each T and composition X, we carried out an NPT 266 simulation to evaluate the volume and lattice parameters at specific pressure. Parrinello-Rahman 267 dynamics with Langevin thermostat was adopted to control the pressure and temperature in the 268 trajectory. For the liquid phases, the unit cells were constrained to be cubic throughout the simulations; 269 while for the solid iron, the lattice parameters are constrained to be orthorhombic with b/a ratio fixed 270 271 to that of the *hcp*-lattice. In both the liquid and solid phases, we generally used 180 atoms, which is demonstrated to be large enough to get converged equations of state and the free energy changes with 272 respected to the referenced systems (Sun et al., 2018). By discarding the first 5 ps (5000 MD steps) as 273 274 pre-equilibrium stage, we obtained the averaged lattice parameters from the last 20 ps (20000 MD steps) trajectories. Then we carried out an NVT simulation with these lattice parameters, confirming 275 that the system was in a hydrostatic state with deviatoric stresses less than 0.5 GPa. 276

For the solid solutions, at each temperature and composition, with sulfur atoms initially randomly substituted on the *hcp*-Fe lattice sites, we carried out the hybrid Monte Carlo/molecular dynamics simulations (Widom et al., 2014) in a sequence of every 20 MD steps followed by one attempted Monte Carlo (MC) species swap. With a duration of overall 20 ps MD steps and 1000 attempted species swaps, the simulation finds its convergence with well-sampled pressures/volumes and energies and this turns out to be important for the free energy calculations.

Finally, lattice dynamics simulations were deployed with PHON through small displacement method (Alfè, 2009). Based on the final configurations of above-mentioned equilibrated molecular dynamics simulations, with relaxations of the time-averaged ionic positions, the interatomic force constants were calculated by setting a displacement amplitude of 0.01 Å for each atom. A dense

30×30×30 mesh was used for the q-point sampling in the first Brillouin zone to evaluate the vibrational
density of state. For each structure of solid phase, we carefully confirmed that the phonon spectra are
dynamically stable with no imaginary phonon frequencies, as demonstrated in Fig. A.2 of Appendix
A. In the framework of quasi-harmonic approximation theory, the harmonic free energies at finite
temperatures can be calculated.

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4. Results

4.1. Benchmarks of the GAP models

As the fundamental benchmarks of the derived GAP models in this study, we systematically 295 296 compare their atomic forces, stresses and energies with those from DFT simulations. As shown in Fig. 1(a), which includes 344928 data retrieved from both simulations at 6000 K, we find very good 297 agreements between the GAP forces and DFT forces. The deviations of the atomic forces are generally 298 within $\pm 2 \text{ eV/Å}$ and the average error is around 0.2 eV/Å. The GAP model also reproduces the stress 299 tensors very well, with averaged error of the normal components less than 0.8 GPa as illustrated in Fig. 300 1(b). Furthermore, it is remarkable that the GAP energies agree excellently with those from DFT 301 simulations, with most of the deviations in GAP energy within ± 10 meV/atom and the averaged energy 302 error of 3.6 meV/atom as shown in Fig. 1(c). These accurate depictions of microscopic interactions 303 and reproductions of the energies guarantee the almost identical samplings of the phase spaces with 304 the GAP models as compared with those of the DFT simulations. 305

To further demonstrate the accuracy of the GAP models for the free energies, we carried out a

benchmark simulation with the GAP model at 6000 K. The *NVT* simulation trajectory is propagated in a 4.6 cm³/mol cell with 88 iron atoms and 8 sulfur atoms (corresponding to about 237 GPa). We randomly picked out 20 independent configurations from the trajectory and carried out high precision DFT simulations (with parameters mentioned in the second part of Sec. 3.1). Since the energies are sufficiently close between GAP and DFT, through the one-step thermodynamic perturbation method, $\Delta F = k_{\rm B} T ln \langle e^{-\Delta U/k_{\rm B}T} \rangle$, we estimated the free energy deviation to be only about 0.3 meV/atom (0.029 kJ/mol).

In Fig. 2, we show the benchmark of the GAP models for the microscopic structures. It is evident 314 315 that the simulation with GAP models has perfectly sampled the local structures around different atoms as compared with those from the DFT simulation. Just as previous findings from first principles 316 simulations (Alfe and Gillan, 1998; Alfe et al., 2003), we can easily observe the net S-S repulsions and 317 318 Fe-S attractions in the solution. And almost all the structural features (positions of local maximums and minimums, and spreading of the peaks) of the Fe-S system have been reproduced in the GAP 319 model simulation. Since the radial distribution functions are closely related with the free energy (Ben-320 321 Naim, 2006), this benchmark again verifies the robustness of the GAP models in predicting the thermodynamic properties of Fe-S systems. 322

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4.2. Free energy and melting of pure iron

To determine the melting points of pure iron, we firstly carried out some explorative simulations for its equation of states. As shown in Fig. 3(a), squeezing the liquid or expanding the solid inevitably find phase transitions when the pressure deviates far enough from the melting point ($P_{\rm m}$ with uncertainty in the figure, determined below). We carefully inspect the radial distribution functions,

mean square displacements and fluctuations of energies and pressures in the trajectories to verify the 329 stabilities of the simulations within the simulation length scales. Based on these points we get the 330 331 auxiliary Birch-Murnaghan equation of state for the liquid and solid respectively, which are shown in Fig. 3(a) with dashed curves and the numerical data can be found in Appendix C. It is apparent that 332 the melted solids quickly establish new equilibrium within the duration of the simulation and their P-333 V relations match the predictions of the liquid EOS quite well, while the solidifying liquids take much 334 longer time to be fully crystalized and the averaged volumes/pressures are still slightly larger than the 335 solid EOS predictions. These points give us rough estimates of the upper and lower limits of the 336 337 melting pressure. In Fig. 3(a), we also include the curves provided by Komabayashi and Fei (2010) based on existing experimental data and find very good agreements with the simulated EOS. 338 Based on the observations of these simulations, for each temperature we chose a point to calculate 339

its free energy directly with the thermodynamic integration techniques. As listed in Table 2, we got the free energies of eight points, which were used as references for the profiles over the whole pressure range with the aid of the auxiliary equation of state mentioned above. Then through the crossover of these Gibbs free energy profiles of the liquid and solid phases we finally determined the melting points, as demonstrated in Fig. 3(b). The uncertainty of melting pressure was estimated as the half of distance between the upper and lower bound of the crossover by counting the errors of free energies.

Fig. 4 shows the calculated melting points of iron at four temperatures. By interpolating these points with simple second-order polynomial, we obtained the following equation for the simulated melting temperature ($T_{\rm m}$) of pure iron (valid from about 110 GPa to 430 GPa)

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$$T_{\rm m} = 2572 + 14.094 \times P - 0.00891 \times P^2 \quad (18)$$

From this equation and considering the error bars of the four points in Fig. 4, the melting temperature

at 330 GPa is estimated to be 6253±170 K, comparing very well with 6170±200 K as recently estimated by Sun et al. (2018) and 6350±300 K as predicted by Alfe et al. (2002c). In contrast, simulations by Belonoshko et al. (2000) and Laio et al. (2000) give significantly higher (7100 K) or lower (5400 K) melting temperature, respectively, at the inner core boundary. With careful evaluations of the free energy errors from the classical potentials used in their studies, it is possible to correct these results to be in much better accordance with the DFT simulations (Alfe et al., 2002b).

Compared with the experiments, it is remarkable that our results almost perfectly agree with the measurements and extrapolations by Anzellini et al. (2013) over the entire pressure range. Shock wave measurements by Nguyen and Holmes (2004) and Brown and McQueen (1986) also fall into the same trend. On the other hand, diamond anvil measurements by Boehler (1993) and Sinmyo et al (2019) and shock wave measurements by Yoo et al. (1993) give much lower or higher melting temperatures by up to over 1000 K. The exact reasons for these remarkable observed differences in experiments are still under debate due to the extreme technical challenges (Aquilanti et al., 2015; Morard et al., 2018).

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365 4.3. Free energies of Fe-S alloys

For the Fe-S alloys, we deployed simulations with 9-36 sulfur atoms in the 180-atom cells (corresponding to $c_s=0.05\sim0.20$, which is within the stability regime of *hcp*-structured Fe-S solid solutions according to previous studies (Cote et al., 2008; Gavryushkin et al., 2016)) and for each concentration we calculated its free energies through the thermodynamic integration techniques at several *T-P* conditions (as listed in Table 3), based on which the free energies over the pressures can be derived through the auxiliary Birch-Murnaghan equation of states at various temperatures, as listed in Appendix C.

In Fig. 5, we show the simulated Gibbs free energies of Fe-S at 4000 K, 250 GPa and 6000 K, 330 373 GPa. We include the free energies of solid solutions calculated from thermodynamic integrations based 374 on simple MD simulations (open blue circles) to emphasize the importance of enhanced sampling. 375 With quasi-random initial occupations of sulfur atoms on the *hcp*-lattice sites, these free energies agree 376 with those from more sophisticated hybrid MC/MD samplings at low concentrations, but the deviations 377 become more and more severe at high concentrations. The clues for explaining such errors can be 378 found in Fig. 6, which illustratively compares the evolutions of thermodynamic properties and 379 structures in the simulations at 4000 K, 4.3893 cm³/mol and $c_s=0.20$. From Fig. 6(a), obviously the 380 381 hybrid MC/MD sampling quickly finds a more stable state with total energy decreased by over 8.4 kJ/mol within about 4000 steps (involving only 200 attempted MC swapping of the species). The major 382 structural feature of this more stable state is the decreased number of nearest-neighbor S-S pairs (NN_S-383 384 s), as shown in the bottom plot of Fig. 6(a) and demonstrated in Fig. 6(b) by the radial distribution functions. NNs-s decreases from the initial quasi-random arrangement value (around 43 here for a 180-385 atom cell with 36 sulfur atoms, i.e., $NN_{S-S}=0.5zN_Sc_S$, where z=12 for the *hcp*-lattice) to less than half 386 of the initial value (around 19 in Fig. 6), and the S-S coordination number decreases from 2.5 to 1.1. 387 Since the interactions among nearest-neighbored S-S pairs dominate the energetic change of the Fe-S 388 solution as compared with pure iron (Alfe et al., 2002a), the net repulsions of S-S interactions here 389 sampled by the hybrid Monte Carlo/molecular dynamics simulations effectively push the systems to 390 lower energy states. As shown in the middle plot of Fig. 6(a), the accommodations of sulfur atoms in 391 the more appropriate sites result in a smaller pressure by over 4.2 GPa, which would further contribute 392 to the decreasing of free energy. For the readers' reference, the converged configurations of Fe-S solid 393 solutions through hybrid Monte Carlo/molecular dynamics simulations at various conditions can be 394

397 4.4. Partitioning of sulfur

From the data points in Fig. 5, we obtained the free energy profiles over the composition range by 398 linear least-square regressions with Eqn. (10). The partitioning of sulfur in the coexisting liquid and 399 solid iron phases can be firstly inferred from the variations of the two regressed parameters over 400 temperatures and pressures. As listed in Table 4, the difference of $\mu_{\rm S}^{\dagger,{\rm sol}} - \mu_{\rm S}^{\dagger,{\rm liq}}$ is positive. It is almost 401 invariant to pressure but shows clear temperature dependence: it is about 65 kJ/mol at 4000 K, around 402 21 kJ/mol at 5000 K and 13 kJ/mol at 6000 K. These observations demonstrate that sulfur would 403 generally prefer the liquid iron phase, as expected, but this tendency would be decreased at higher 404 temperatures. On the other hand, the difference of $\lambda_{\rm S}^{\rm sol} - \lambda_{\rm S}^{\rm liq}$ keeps increasing from -192 kJ/mol at 405 406 4000 K and 150 GPa to 135 kJ/mol at 6000 K and 330 GPa. This opposite trend would greatly balance the partitioning of sulfur between the two phases. 407

By solving Eqns. (4-5), we quantitatively calculated the partition coefficients of sulfur 408 $(D_{\rm S} = c_{\rm S}^{\rm sol}/c_{\rm S}^{\rm liq})$, as shown in Fig. 7. At 250 GPa and 330 GPa, similar values of $D_{\rm S}$ are predicted over 409 temperature change of 1000 K, which reveals the minor temperature dependence of $D_{\rm S}$ and this is in 410 accordance with the experimental findings (Kamada et al., 2010; Yokoo et al., 2019). For pressures 411 lower than 250 GPa, we find remarkable pressure dependence of $D_{\rm S}$: its value decreases by over 40% 412 from 250 GPa to 150 GPa. The simulated data of $D_{\rm S}$ in this study perfectly match the available 413 experimental measurements and their trends up to 254 GPa (Kamada et al., 2012; Kamada et al., 2010; 414 Li et al., 2001; Mori et al., 2017; Stewart et al., 2007; Yokoo et al., 2019). From 250 GPa to higher 415 pressures, it is notable that $D_{\rm S}$ becomes almost pressure independent with an averaged value of 416

417 0.75 ± 0.07 . This is in good agreement with the early prediction by Alfe et al. (2002a) through DFT 418 simulations. The simple MD simulations with random solid solutions turn out to significantly 419 underestimate $D_{\rm S}$ (~0.53±0.04 as illustrated in Fig. 7), which again emphasizes the importance of 420 sufficient sampling with the hybrid Monte Carlo/molecular dynamics simulations.

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- 422

5. Implications

According to our simulations in this study, the melting of pure iron at Earth's current inner core 423 boundary pressure (330 GPa) results in a density jump ($\Delta \rho = \rho^{\text{sol}} - \rho^{\text{liq}}$) of 0.24 g/cm³, which amounts to 424 1.8% of ρ^{sol} (i.e., $\Delta \rho / \rho^{\text{sol}} \approx 1.8\%$) and compares very well with those predicted by previous simulations 425 (as listed in the Table II of Sun et al. (2018)). This density jump is far smaller than the seismologically 426 observed value of 0.6 g/cm³ (Dziewonski and Anderson, 1981) or 0.8±0.2 g/cm³ (Masters and Gubbins, 427 2003). Assuming Fe-S binary model for the Earth's core and the temperature around 6000 K, we would 428 need around 20% sulfur ($c_{\rm S}^{\rm liq} \approx 0.20$) to match the outer core density ($\rho^{\rm sol}=12.166$ g/cm³, according to 429 Dziewonski and Anderson (1981)) at ICB. With smaller amount of sulfur coexisting in the inner core, 430 $c_{\rm S}^{\rm sol} = D_{\rm S} \times c_{\rm S}^{\rm liq} = 0.15$, the density jump $\Delta \rho$ increases to 0.39 g/cm³. So the partitioning with sulfur 431 alone is insufficient to account for the observed density jump across the Earth's inner core boundary 432 and the Earth's core composition should be more complex than the simple Fe-S binary, which supports 433 the findings of previous studies (Alfe et al., 2002a, 2003, 2007). 434

Although the exact core compositions are still elusive, geochemical studies often place around 2 wt% sulfur in the Earth's core (Allegre et al., 1995; McDonough, 2003; Wood et al., 2006). With this amount of sulfur, according to the comprehensive explorations by Badro et al (2014), the outer core would additionally need about 2.6 wt% oxygen and 1.8 wt% silicon to best fit the densities and seismic

velocities at ICB and CMB (core mantle boundary). This leads to an outer core composition with 439 $c_0^{\text{liq}} = 0.083$, $c_{\text{Si}}^{\text{liq}} = 0.033$ and $c_{\text{S}}^{\text{liq}} = 0.032$. Since $D_{\text{Si}} \approx 1$, $D_0 < 0.01$ (calculated from the parameters 440 provided by Alfe et al (2002a)) and $D_S = 0.75$, we would estimate the coexisting inner core composition 441 with about 1.7 wt% silicon and 1.4 wt% sulfur (i.e., $c_0^{sol} \approx 0$, $c_{Si}^{sol} \approx c_{Si}^{liq} = 0.033$ and $c_s^{sol} = 0.033$ 442 $D_{\rm S} \times c_{\rm S}^{\rm liq} = 0.024$). Now with the solid solution model provided by Li et al (2018), the obtained inner 443 core density would agree well with the PREM model by Dziewonski and Anderson (1981) with a 444 relative error of 0.5% and the density jump would be around 0.66 g/cm³. Note that we have not included 445 hydrogen and carbon here since the properties of their alloys with iron are much more unclear, although 446 they may be potentially important for explaining the seismic observations under certain circumstances 447 (Li et al., 2018; Li et al., 2019; Mashino et al., 2019; Umemoto and Hirose, 2020). 448

If the above core composition is plausibly in accord with geochemical and geophysical constraints, 449 450 then the roles of sulfur on the geodynamics of Earth's core should be minor. This may be inferred from its contribution to the density jump (0.66 g/cm³ as mentioned above), which is important for evaluating 451 the gravitational energy to drive the geodynamo (Stacey and Stacey, 1999; Gubbins et al., 2004). By 452 "turning on" the partitioning coefficients stepwise (i.e., we inspect the differences by switching the D_0 453 and $D_{\rm S}$ from 1.0 to the expected values), the exclusion of oxygen in the inner core and partitioning of 454 sulfur would account for about 0.44 g/cm³ and 0.04 g/cm³ respectively, in addition to 0.18 g/cm³ from 455 the assumed congruent freezing (i.e., the solid is assumed to be in the same composition with the 456 coexisting liquid). It seems that the 6% contribution here from sulfur cannot be entirely neglected, but 457 this is likely to be an upper bound, since sulfur is recently found to be less siderophile in the Earth's 458 core (Suer et al., 2017). 459

460 Finally, the almost invariant partition coefficient from 250 GPa to higher pressures is interesting.

It would imply a constant partitioning of sulfur since the advent of the solid inner core to the time when 461 its radius expands to at least 1000 km larger as compared with the present size. This would support a 462 stable stratification of sulfur in the Earth's inner core, since more sulfur precipitates in the inner core 463 as the outer core sulfur concentration increases over time (Cottaar and Buffett, 2012; Deguen and 464 Cardin, 2011). By contrast, without counting the T-P dependence of the relevant parameters in the 465 equations of chemical potentials (as listed in Table 4 and analyzed in the previous section), recent 466 numerical simulations propose continuously decreasing partition coefficient with time and find its 467 destabilizing buoyancy effects on the stratification of the Earth's inner core (Gubbins et al., 2013; 468 469 Labrosse, 2014; Lythgoe et al., 2015). Our results in this study show that these simulations may need to be re-evaluated at least for sulfur based on our results. 470

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6. Concluding remarks

In this study, we derive new generation nonparametric interaction potentials for Fe-S systems 473 applicable under Earth's core conditions. Based on machine learning techniques, these Gaussian 474 Approximation Potentials are shown to reproduce the first principles simulation results with 475 unprecedented accuracies, including the interatomic forces, local structures and, most importantly, the 476 free energies that fundamentally govern all thermodynamic properties. With a similar approach, we 477 will be able to derive accurate potentials for more complex systems (e.g., multicomponent systems 478 including elements of Ni, O, S, Si, C, H), which are very difficult to investigate solely with first 479 principles techniques due to the increase in the size of the phase space. 480

The substantial initial efforts of training the machine learning potentials provide a return in the remarkable efficiency in sampling the phase spaces of iron and its alloys under various temperature

and pressure conditions. It is then possible for us to simulate free energies and predict phase behaviors 483 with fundamentally rigorous thermodynamic integration method within affordable computational cost. 484 485 In fact, to thoroughly sample the phase space around the liquidus and solidus of Fe-S solutions, we have carried out over 500 independent atomistic simulations in this study, each with 180 atoms and at 486 least 20000 steps. The current implementation of the GAP models takes about 12 CPU seconds for 487 each step in each run. With parallel acceleration of 24 CPU cores for each run, we have managed to 488 accomplish the simulations within two months. As a comparison, the direct high precision DFT 489 simulations are about three orders of magnitudes slower than the GAP simulations. 490

As a first application of the framework mentioned above, we focus in this paper on the partition coefficients of sulfur between the solid and liquid iron under Earth's core conditions. While the results at ICB are in good agreements with early DFT simulations, we obtained the melting and partitioning behaviors over the entire relevant T-P regime of the Earth's core. In particular, the invariance of partition coefficients from 250 GPa to higher pressures found in this study provides new constraint on the compositions and dynamics of Earth's inner and outer core.

497 Finally, since the phase behaviors of iron alloys are comprehensively complex, it should be noted that much more endeavors are needed beyond our current efforts of predicting sulfur partitioning in 498 Fe-S binary systems under core conditions. The interplays of different impurities, the possible 499 stabilization of face-cubic-centered (fcc) or even body-cubic-centered (bcc) structures, the 500 immiscibility of liquid iron-alloying systems, the heterogeneities of the Earth's inner core, etc., can all 501 be important to estimate the roles of light elements in real Earth's core. Interests in these issues would 502 imply quickly growing demands of computations over broader phase spaces. The high accuracy and 503 efficiency gained by the framework proposed in this study would benefit providing new constraints 504

505 over all these issues.

506

507 Supplementary materials

508 Supplementary materials associated with this article can be found in: Appendix A that includes 509 several supporting figures for the main text; Appendix B that has the details for a GAP model training 510 with the QUIP code; Appendix C that includes the complementary research data and goes to the 511 Electronic Annex.

512

513 **Research data for this article**

514 Most data used in this article are given in Tables 2-4, complementary research data are included 515 in the Electronic Annex.

516

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Table 1. Hyper-parameters for training the GAP models

GAP soft	ware version	1527075646				
Atomic envi	ronment kernel	SOAP				
	<i>r_{cut}</i>	5.0 Å				
	r_{Δ}	1.0 Å				
σ^{energy}	^y default	0.002 eV/atom				
σ^{force}	default	0.05 eV/Å				
σ^{virial}	default	0.05 eV/atom				
	σ_{w}	1.0 eV				
c	Datom	0.5 Å				
	ξ	2				
ľ	ı _{max}	8				
l	lmax	8				
Representative environments		4000				
Sparse method		CUR				
	4000 K	Fe:-2.70299745:S:-1.33262117				
	5000 K	Fe:-3.36146432:S:-1.66173158				
$e_0(ev)$	6000 K	Fe:-4.02404024:S:-1.99085328				
	7000 K	Fe:-4.69558982:S:-2.32005321				

Table 2. Simulated Helmoholtz free energies (F) of pure iron

$T(\mathbf{V})$	Р	Liquid		Solid		
<i>I</i> (K)	(GPa)	$V(\text{cm}^3/\text{mol})$	F (kJ/mol)	V (cm ³ /mol)	F (kJ/mol)	
4000	100	5.4277	-1040.49 ± 1.16	5.2783	-1024.35±0.46	
5000	190	4.8171	-1061.59 ± 1.25	4.7146	-1041.26±0.48	
6000	330	4.2473	-1024.00±1.27	4.1721	-1001.65±0.59	
7000	400	4.0714	-1069.38±1.29	4.0072	-1041.69±0.94	

Table 3. Simulated Helmoholtz free energies (F) of Fe-S alloys

<i>T</i> (K)	Phase $c_{\rm S}$ V		$V (\text{cm}^3/\text{mol})$	P (GPa)	F (kJ/mol)	
	Liquid	0.05	5.4239	100.07±0.15	-1028.15±0.81	
4000		0.10	5.4245	100.12±0.14	-1012.87±0.77	
4000		0.15	5.4330	99.64±0.30	-996.13±1.11	
		0.20	5.4462	99.80±0.24	-978.22±0.84	
		0.05	4.3603	249.61±0.05	-858.26±0.42	
4000	Calid	0.10	4.3896	241.92±0.06	-847.21±0.42	
4000	Solid	0.15	4.3819	243.42±0.11	-826.18±0.50	
		0.20	4.3893	243.87±0.16	-804.96±0.47	
		0.05	4.8086	189.94±0.20	-1047.60±0.92	
5000	Liquid	0.10	4.7990	190.30±0.17	-1030.67 ± 1.04	
3000	Liquid	0.15	4.7968	190.14±0.14	-1012.67 ± 1.10	
		0.20	4.7955	189.88±0.20	-992.56±0.95	
	Solid	0.05	4.1225	329.36±0.05	-878.78±0.30	
5000		0.10	4.1131	330.06±0.07	-859.37±0.31	
5000		0.15	4.1139	328.59±0.12	-840.38±0.49	
		0.20	4.1153	327.80±0.09	-820.03±0.42	
	Liquid	0.05	4.2395	329.51±0.20	-1010.93 ± 1.67	
6000		0.10	4.2268	330.44 ± 0.37	-992.56±1.03	
0000		0.15	4.2195	330.32±0.17	-973.51 ± 1.21	
		0.20	4.2134	329.73±0.19	-951.78±1.26	
		0.05	4.1636	329.27±0.10	-987.78±0.54	
6000	Solid	0.10	4.1807	322.99±0.10	-977.19±0.43	
		0.15	4.1692	325.74±0.20	-954.92±0.47	

Table 4. Parameters for chemical potential and free energy at various *T-P* conditions (Eqns. (8-10))

Т	Р	Liquid				Solid	
(K)	(GPa)	$\mu_{\rm Fe}^0$ (kJ/mol)	$\mu_{\rm S}^{\dagger}$ (kJ/mol)	$\lambda_{\rm S}$ (kJ/mol)	$\mu_{\rm Fe}^0$ (kJ/mol)	$\mu_{\rm S}^{\dagger}$ (kJ/mol)	$\lambda_{\rm S}$ (kJ/mol)
4000	150	-237.94±1.16	116.43±1.11	463.67±12.04	-243.08±0.46	182.49±7.23	271.38±92.00
4000	200	3.66±1.16	349.75 ± 1.97	496.38±21.30	-6.50±0.46	415.01±5.81	317.29±73.86
4000	250	231.91 ± 1.16	566.66±2.19	525.74±23.73	217.31±0.46	629.14±7.19	405.41±91.53
5000	250	132.88 ± 1.25	518.53±6.73	322.69±72.17	128.11±0.48	542.03±2.73	365.32±31.77
5000	300	353.11±1.25	726.27±6.97	361.62±74.72	344.14±0.48	746.94±2.80	442.20±32.65
5000	330	480.79±1.25	844.73±6.68	393.97±71.60	469.52±0.48	865.41±2.84	478.25±33.07
6000	330	377.61±1.27	766.37±5.51	375.25±61.15	375.48±0.50	779.01±6.12	510.14±88.08

Figure Captions

- Figure 1. Comparisons of atomic forces (a), stresses (b) and energies (c) between DFT and the GAP model at 6000 K. 1178 energies, 3534 normal stress components and 344928 forces are included in these comparisons. The red dashed lines are guides for perfect matches. In the inset plots, we show the cumulative probability distribution of force component errors and energy errors (relative to reference DFT calculations).
- Figure 2. Comparisons of the radial distribution functions of a Fe-S liquid (6000 K, 237 GPa and $c_{\rm S}$ =0.21) from the DFT simulation (full lines) and that with the GAP models (broken lines).
- Figure 3. Equation of state (a) and free energy profiles (b) of pure iron at 5000 K. In (a), the filled symbols are simulation results with GAP models without phase transition over the whole trajectories, while the open symbols are the statistical averages after the phase changes. The dashed lines are Birch-Murnaghan equations regressed from the filled symbols. The solid lines are those provided by Komabayashi and Fei (2010) based on experimental data. In the upper plot of (b), the line is the Gibbs free energy change ($\Delta G=G_{solid}-G_{liquid}$) and light-blue shadow marks the uncertainty.
- Figure 4. Melting of iron from different approaches. Symbols and lines in blue are the experimental
 data, whole those in black are the previous simulation results. The red squares are the
 melting points from this study with GAP models.
- Figure 5. Gibbs free energies of Fe-S solutions under 4000 K, 250 GPa (a) and 6000 K, 330 GPa (b).
- The filled black and red symbols are the free energies for liquids and solids, respectively.
 The lines are regressions with Eqn. (10) based on these data (relevant parameters are listed in Table 4). Open blue circles are the results for solid solutions through simple molecular

dynamics samplings, as compared with the filled red circles through hybrid Monte
Carlo/molecular dynamics samplings (see the text). The uncertainties of the free energies
are generally smaller than the sizes of the symbols.

- Figure 6. Effects of Monte Carlo swapping on the samplings of thermodynamic properties and 737 structures of the Fe-S solid solution. In (a), we show the propagations of total energy (U), 738 pressure (P) and number of nearest-neighbored S-S pairs (NN_{S-S}) during the trajectories of 739 the two simulations (one with simple MD, in green; the other with hybrid Monte Carlo and 740 MD, or MD+MC, in black) at 4000 K, 4.3893 cm³/mol, $c_s=0.20$. The broken lines are 741 742 equilibrium averaged values (i.e., discarding the first 5000 fs as pre-equilibrium stage) from both simulations (MD, in blue; MD+MC, in red) for better guides of the fluctuations. In (b), 743 we compare the radial distribution functions of different pairs from the two simulations. 744
- Figure 7. Partition coefficients of sulfur between solid and liquid iron. Filled symbols are those from
 this study at different temperatures and pressures. The additional red crossed circle at 6000
 K and 330 GPa is the result from simple MD simulations with random solid solutions. Open
 symbols are data from previous experiments (Kamada et al., 2012; Kamada et al., 2010; Li
 et al., 2001; Mori et al., 2017; Stewart et al., 2007; Yokoo et al., 2019) and simulations (Alfe
 et al., 2002a). Note that all the experimental data have been carefully converted to the ratio
 of mole faction to be consistent with simulation results.
- 752
- 753





(b)









/01

762 763

Figure 1



Figure 2









Figure 4











Figure 6



Figure 7