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Phys. Rev. Lett. **116**, 075503 — Published 19 February 2016

DOI: [10.1103/PhysRevLett.116.075503](https://doi.org/10.1103/PhysRevLett.116.075503)

Sampling polymorphs of ionic solids using random superlattices

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(Dated: February 2, 2016)

Polymorphism offers rich and virtually unexplored space for discovering novel functional materials. To harness this potential approaches capable of both exploring the space of polymorphs and assessing their *realizability* are needed. One such approach devised for partially ionic solids is presented. The structure prediction part is carried out by performing local DFT relaxations on a large set of random superlattices (RSLs) with atoms distributed randomly over different planes in a way that favors cation-anion coordination. Applying the RSL sampling on MgO, ZnO and SnO₂ reveals that the resulting probability of occurrence of a given structure offers a measure of its realizability explaining fully the experimentally observed, metastable polymorphs in these three systems.

Keywords: polymorphs, metastability, structure prediction, random structure sampling

The discovery of polymorphism in the late 18th and early 19th century [1, 2] revealed the significance of structural degrees of freedom in determining physical properties of solids. The best known example is probably elemental carbon with markedly different mechanical, optical and electronic properties between its graphite and diamond forms [3]. Other notable cases include white and grey tin, which also exhibit significant differences in electronic and mechanical properties [3]; or enhanced photocatalytic activity of anatase TiO₂ compared to the ground state rutile polymorph [4]; or elemental Silicon, an indirect band-gap semiconductor in the ground state diamond structure predicted to become a direct gap material in a number of higher energy structures [5] including the experimentally realized clathrate structure [6].

However, the development of rational approaches to explore the space of polymorphs and (desirably) assist in their experimental realization faces significant challenges. First, the complexity of the potential energy surface (PES) of periodic systems, evidenced by the exponential increase in the number of local minima with the system size [7], limits our ability to systematically explore the spectrum of possible structures.

A related problem of finding the ground state structure attracted attention, especially with the development of first-principles total energy methods, resulting in a number of structure prediction techniques [8, 9]. These include simulated annealing [10, 11], methods based on evolutionary algorithms [12–15], metadynamics [16, 17], basin and minima hopping [18, 19], random structure searching [20], methods based on data mining and machine learning [21], structure prototyping [22–24], etc. Although focused on finding the ground state structure some of these methods were also used in exploring the space of polymorphs (see for example Refs. [5, 24–26]) with the energy above the ground state as the main quantifier of their potential for experimental realization.

This brings us to the second major challenge, the assessment of the likelihood for experimental realization of

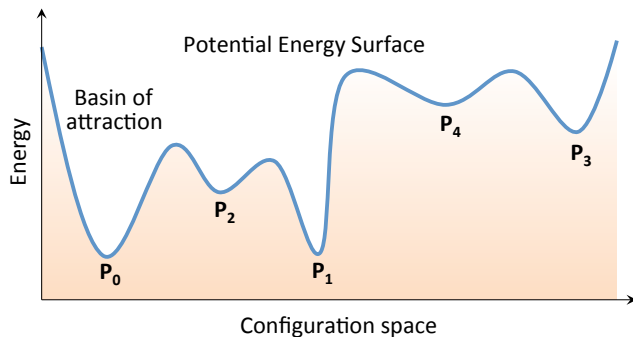


FIG. 1. (color online) A sketch of the potential energy surface (PES) of solids with different polymorphs corresponding to different PES local minima.

different polymorphs. While certainly being an important quantity, the energy above the ground state alone is insufficient to explain observations based on available experimental data. For example, in the case of MgO, despite predictions [25] only the ground state rocksalt structure, and no other, is experimentally realized as reported in the Inorganic Crystal Structure Database [27]. In the case of ZnO, only the ground-state wurtzite and two other structures, zincblende and rocksalt, are experimentally realized [27–29]. Another important example is SnO₂, which undergoes a series of phase transitions under pressure [30], but all of the high pressure phases relax to either the ground state rutile or the metastable α -PbO₂ structure type upon releasing the pressure [31]. These facts indicate that for a given composition there seems to exist a finite set of structures that have higher likelihood for realization than the rest.

If this is true, then the realizability of a given polymorph can be thought of as determined by a combination of three factors: (i) the energy above the ground state, (ii) the energy barrier to escape from a given PES minimum, and (iii) the volume of configuration space occupied by the PES minimum. The (i) and (ii) describe

the principles of energy minimization and kinetic trapping. The factor (*iii*) on the other hand, measures the probability of getting into a given PES minimum. More precisely, as shown in Fig. 1 every local minimum on the PES defines its basin of attraction, or the region of configuration space that has that minimum as its "center of gravity". Hence, the probability of "falling" into a certain structure has to be proportional to the total volume of configuration space occupied by its basin of attraction including all of the symmetry equivalent basins. Similar arguments were presented recently by Sandip De et al.[32] in the context of experimental realizability of finite size systems (i.e. clusters).

Motivated by the general features of the potential energy surfaces of solids that are discussed in more details in Ref. [20] (and the references therein) it is demonstrated here, using MgO, ZnO and SnO₂ as case examples, that the factor (*iii*) is actually critical in establishing a ranking of realizability with (*i*) and (*ii*) providing additional constraints. This is done by pursuing the idea that the total volume occupied by various basins of attraction can be estimated using a large number of random structures (random unit cell vectors and random atomic positions) that are relaxed to the closest PES local minimum utilizing density functional theory (DFT). The frequency of occurrence of a given structure would then provide an estimate of the probability to "fall" into its local minimum.

To do this and, at the same time, to overcome in part the difficulties posed by the already mentioned complexity of the PES together with the fact that the volume of attraction basins is ill defined in truly infinitely periodic systems, the size of the simulation cell is constrained and a structure prediction method is proposed to bias the random sampling toward the region of the PES more relevant for ionic systems. Because of the charge transfer only the structures that have cations preferentially coordinated by anions and vice-versa are relevant. The method adopted here favors the cation-anion coordination by distributing different types of ions in a random fashion over two interpenetrating grids of points. The grids are constructed using the alternating planes of a superlattice defined by a randomly chosen reciprocal lattice vector (see Fig. 2). Constructed in this way, these random superlattice (RSL) structures exhibit dominant cation-anion coordination.

For each MgO, ZnO, and SnO₂ a total of 2000 RSL structures with sizes varying between 1–20 formula units are constructed and DFT-relaxed to the closest local minimum. The relaxed structures are sorted into classes of equivalence and for the classes with largest occupancies (frequencies of occurrence) additional phonon calculations are performed with the purpose of providing the information on the dynamic stability. The analysis of the resulting frequencies of occurrence shown in Fig. 3 and 4 reveals that the experimentally observed polymorphs are exclusively the ones with the highest occurrence.

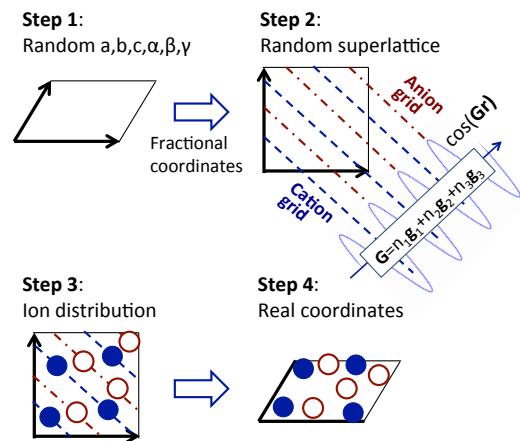


FIG. 2. (color online) Steps in the random superlattice (RSL) structure generation (see text for details).

RSL Sampling. The details of the RSL structures generation are shown in Fig. 2. It is a modification of the Ab Initio Random Structure Searching method [20] and similarly starts with the random choice of unit cell parameters $a, b, c, \alpha, \beta, \gamma$. In the second step the cation and anion grids are constructed in the following way. First, a transformation to the fractional (crystal) coordinates is performed to provide a cubic-like representation of the unit cell. Then, a reciprocal lattice vector $\mathbf{G} = n_1\mathbf{g}_1 + n_2\mathbf{g}_2 + n_3\mathbf{g}_3$ with random n_1, n_2, n_3 is constructed. \mathbf{G} defines a plane wave $\cos(\mathbf{G}\mathbf{r})$ and an associated superlattice. The two grids are constructed by discretizing the planes corresponding to the minima (cation grid) and the maxima (anion grid) of the plane wave. In the third step the ions are distributed over the two grids. To ensure homogeneous distribution and that no two ions of the same kind are too close, the probability distribution is constructed by placing a gaussian centered at each occupied grid point. The next ion is then placed on a grid point chosen randomly among those that have low probability. Finally, the structure is converted from fractional back into the real coordinates and the scaling factor is adjusted such that the minimal distance between any two atoms is larger than a certain threshold. The example of an RSL structure of ZnO shown in the Supplemental Material clearly displays the dominant cation-anion coordination.

A total of 6000 RSL structures are generated (2000 per system) with the following parameters: a, b , and c randomly chosen between 0.6 and 1.4 (in units of *scale*); α, β , and γ random in the $(30^\circ, 160^\circ)$ range; n_1, n_2 , and n_3 also random between 4 and 10, the range which ensures that sufficient, but not too large, number of planes in the unit cell; and the *scale* is adjusted such that the shortest distance between the atoms is not shorter than 1.8 Å. Different unit cell sizes are sampled by creating the RSL structures with one through 20 formula units

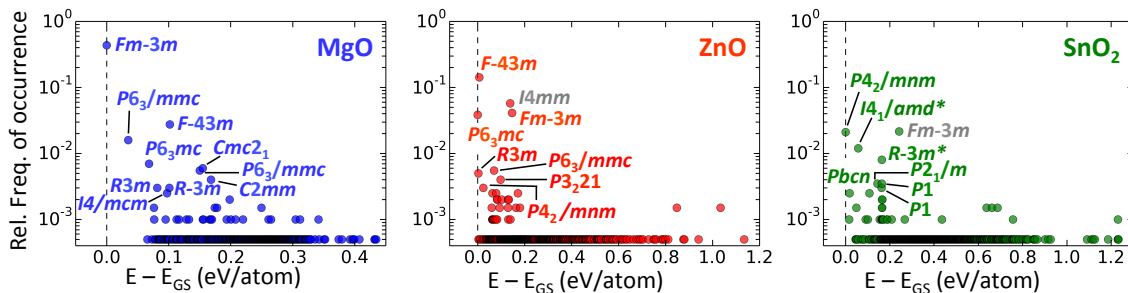


FIG. 3. (color online) Relative frequencies of occurrence of structures resulting from the RSL sampling shown against the energy above the ground state. Only the top occurring structures have their space groups explicitly marked. Space groups shown in grey represent the structures that are predicted to be dynamically unstable. The $I4_1/amd$ (anatase) and $R-3m$ SnO_2 structures are marked with asterisks as they appear only for small cell sizes (see text for discussion).

and 100 RSLs per size. Alternative would be to fix the cell size, but this would bias the sampling only to structures with sizes compatible with the chosen one. Another important reason for sampling over different cell sizes is that some PES minima may appear large in low dimensions, but are actually small when the number of dimensions increases. These will likely not have high chances for experimental realization because the nucleation and growth of bulk phases typically start at the nanometer scale. The ranges of cell sizes within which structures appear (provided in Tables I, II and III in supplemental materials) are important indicators of this effect.

DFT Calculations. Full relaxations, including volume, cell shape and atomic positions, are performed on all RSL structures. This is done by employing standard DFT approach [33] with the PBE form of the exchange-correlation functional [34] and the projector augmented wave (PAW) method [35] as implemented in the VASP code [36]. The employed numerical setup (\mathbf{k} -points, various cutoffs) results in absolute total energies that are converged to within 3 meV/atom. All relaxations (volume, shape, atomic positions) are conducted using the conjugate gradient algorithm [37]. For numerical reasons, the relaxation procedure (both volume and ions) has been restarted at least four times followed by a self-consistent run. For all structures with total final pressure exceeding 3 kbar and/or forces exceeding 10^{-4} eV/Å additional restarts have been performed until these criteria are achieved. Construction of the workflows, management of large number of calculations, and analysis of results is carried out using the *pylada* software [38].

Structure sorting. Sorting of the resulting, DFT-relaxed, structures into the classes of equivalence is done based on four criteria. Two structures are considered equivalent if: (1) their total energies are within 10 meV/atom, (2) their space groups match, (3) their volumes per atom are within 0.5 %, and (4) the coordination of atoms up to the 4th neighbor is the same. Because of very low symmetry of the starting RSL structures as well as numerical inaccuracies that remain after DFT relax-

ation, the tolerances for both the space group and coordination shells determination were set to a relatively generous 0.1 Å. Increasing the tolerance factors to 0.3 Å does not affect final results. After extensive testing these criteria were proven robust in establishing equivalence between the structures.

Note that because of the size and coordination constraints imposed in the sampling procedure as well as some dependence of the final structure on the relaxation algorithm, the population of each class is an estimate, rather than an accurate measure, of the basin volume. Furthermore, because DFT relaxations could potentially drive the system across very small local minima and assign parts of configuration space to larger nearby basins, the sampling procedure adopted here actually estimates the volume of configuration space that “funnels” toward a given larger minimum. To indicate this difference, the term *funnel of attraction* will be used from here on instead of the term *basin of attraction*.

Results. The RSL sampling procedure combined with DFT relaxations results in: 904 distinct structure types (classes of equivalence) for MgO, 1306 for ZnO and 1740 for SnO_2 . Fig. 2 in supplemental material shows for all three systems the number of distinct structures within various energy windows above the ground state as a function of the total number of structures used in sampling. Two observations can be made. First, there are clear differences between the PESs of MgO, ZnO and SnO_2 , evidenced by a different number of distinct structures present in the corresponding windows. Second, for all three systems the number of distinct structures grows monotonically with little, or no sign of convergence. This implies that even within a relatively narrow energy range a very large number of structures can potentially be found including low energy structures with defects and/or interfaces between low energy structures.

Fortunately however, a vast majority of structures that result from the RSL sampling are actually irrelevant for their very low probability (frequency) of occurrence. As shown in Fig. 3, except for a relatively small number of

is used as an estimate (indirect) of the volume of configuration space occupied by a given structure (its *funnel of attraction*). Application of the RSL sampling on MgO, ZnO, and SnO₂ reveals that the experimentally observed polymorphs are exclusively the ones that have highest frequency (probability) of occurrence, consistently as the cell size used in sampling increases, explaining the physical reasons behind the experimental observations.

This work was supported as part of the Center for the Next Generation of Materials by Design, an Energy Frontier Research Center funded by the U.S. Department of Energy, Office of Science, Basic Energy Sciences. The research was performed using computational resources sponsored by the Department of Energy's Office of Energy Efficiency and Renewable Energy and located at the National Renewable Energy Laboratory.

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