

High-throughput crystal structure solution using prototypes

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Databases of density functional theory (DFT) calculations, such as the Open Quantum Materials Database (OQMD), have paved the way for accelerated materials discovery. DFT calculations require crystal structure information as input; however, due to inherent challenges in solving a compound's structure from powder diffraction data alone, there are thousands of experimentally synthesized compounds whose structures remain unsolved. We present a rapid DFT-based structure solution method capable of resolving numerous outstanding structure solution problems at low computational cost. The method involves (1) searching inorganic compound databases for all prototypes that match known structural characteristics of the compound, such as stoichiometry, space group, and number of atoms per cell, (2) performing DFT calculations of the target composition in each of the structural prototypes, and (3) evaluating these prototypes as candidates using a combination of DFT energy and match between calculated and experimental diffraction pattern. As this approach is straightforward and inexpensive, we employ it to solve 521 previously unsolved compounds from the Powder Diffraction File, resulting in a 1.4% expansion of the set of all experimental compounds in the OQMD. DFT calculations of these compounds could yield valuable properties.

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I. INTRODUCTION

Crystal structure is a fundamental descriptor of inorganic compounds and is necessary input for first principles calculations. Indeed, the composition and crystal structure of a compound, comprising unit cell vectors and atomic coordinates, are the only inputs required for a density functional theory (DFT) calculation of the compound's energetic, electronic, and magnetic properties. Thanks to knowledge of crystal structures obtained by experiment, databases of high-throughput DFT calculations, such as the Open Quantum Materials Database (OQMD), Materials Project, and Automatic Flow (AFLOW), have enabled the calculation of phase diagrams, screening of materials for future applications, and prediction of novel materials [1–8]. However, due to challenges in extracting crystal structure from experimental diffraction data, there are many known compounds with unknown crystal structures. For example, there are thousands of diffraction patterns in the Powder Diffraction File (PDF) without an associated crystal structure, meaning that compounds have been synthesized, diffraction patterns measured, and yet there is still no solved structure [9]. Identifying the structure of these materials would enable DFT calculations of their properties and open the door to a full exploration of their potential.

There are several reasons why a complete crystal structure is not always obtained in a diffraction experiment. For instance, while the unit cell parameters, space group, formula units per unit cell, and elemental composition can often be determined from high quality data by indexing the diffraction peaks, the determination of atomic coordinates, known

as structure solution, is especially challenging because the process of measuring a diffraction intensity does not capture the phase of diffracted waves, complicating the inversion from reciprocal space to real space [10]. Another common reason is that the compound either is part of a multiphase material, is impure, or contains elements that only weakly scatter x rays, hindering the ability to capture relevant information in the diffraction pattern. When attempting to solve structures, crystallographers routinely use structure optimization algorithms, in which atomic coordinates are optimized to match the diffraction pattern, i.e., minimize the R factor. When only the R factor is used for the objective function, structure optimization algorithms are sometimes challenged by the existence of multiple solutions with similar R factors. A promising workaround is to supplement the R factor with DFT calculations in order to rule out candidates that are unphysically high in energy [11]. For example, the first-principles-assisted structure solution (FPASS) method, which uses a genetic algorithm with the R factor and DFT energy as a combined objective function, has been used to effectively resolve several long-standing problems [12–15]. Other DFT-based structure optimization algorithms that can be constrained using experimental input include USPEX [16], CALYPSO [17], XtalOpt [18], PEGS [19], and AIRSS [20]. DFT provides a highly accurate estimate of the energetic stability of candidate structures; however, DFT is computationally expensive to use in structure optimization algorithms like FPASS, USPEX, and others, where as many as hundreds or thousands of structures are considered over the course of optimization.

On the other hand, a simpler and cheaper way to solve structures is to search existing databases. The OQMD contains DFT calculations of over 800 000 compounds, including experimentally observed compounds from a 2011 version of the inorganic crystal structure database (ICSD) as well as many

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hypothetical compounds. As we will show, structures from the ICSD portion of the OQMD can be grouped into 10 203 prototypes, distinguished by space group, stoichiometry, and Wyckoff site occupancies. This grouping allows us to identify a relatively small number of symmetrically distinct prototypes as candidates for a given unsolved structure, according to the experimentally determined stoichiometry, number of formula units, and space group. Furthermore, we find that 83% of distinct compounds in the ICSD share a common prototype with at least one other ICSD compound, giving us confidence that we can solve many (but not all) new structures using a “prototype searching” method. In this prototype searching method, we search for candidate prototypes in the OQMD, select a representative structure for each prototype, decorate the structures with the experimental composition, and evaluate them by computing R factors and DFT energies. A related prototype searching method has been used to predict new compounds for hydrogen storage applications [21–24]; however, without experimental input to constrain the search, the prototype searching method is still computationally expensive. On the other hand, when used for structure solution with experimental input, our prototype searching method usually requires evaluating up to three prototypes, far fewer than what is needed for other structure solution methods, allowing us to solve structures at low cost. We note that, while structure optimization algorithms like FPASS, USPEX, and others can leverage experimental input to constrain the search, as optimization algorithms they still typically require DFT calculations of many structures over the search space, including highly unphysical structures whose atomic coordinates are consistent with the experimental space group, stoichiometry, and Wyckoff site occupancies. On the other hand, our prototype searching method gets us straight to the answer with just a few DFT calculations for prototypes that are known to exist in the ICSD. Since the prototype searching method is inexpensive, it can be used to quickly solve numerous unsolved compounds and expand crystal databases with a limited computational budget.

In this paper, we leverage the low computational cost of the prototype searching method to solve the structures of 521 compounds from experimental diffraction patterns in the PDF. All 521 compounds were missing from the ICSD and OQMD and constitute a 1.4% expansion of all experimentally known compounds in the OQMD. We have provided the solved structures as VASP-formatted files in the Supplemental Material [25], and they are available in the latest release of the OQMD. Confident that we have identified structures that both match experimental input and are energetically stable, we open the door to analyzing the properties of these materials and consider their use in a wide range of future applications.

II. PROTOTYPE SEARCHING METHOD

In this section, we detail the prototype searching method to solve the structure of a compound given experimental data.

A. Searching for candidate structures

A completely solved structure is one where we know all the descriptive details, in particular, the unit cell parameters

and the coordinates of all atoms in the unit cell. For the compounds we address in this paper, we have from experimental data the unit cell parameters, elemental composition, space group, and number of formula units per unit cell, but we do not have the atomic coordinates, suggesting that the diffraction peaks were successfully indexed but the structure solution step was not completed. Our approach to solve for the atomic coordinates of the structure is to take the stoichiometry, space group, and the number of formula units per unit cell, and search the OQMD for prototypes with the same attributes. We define the prototype of a crystal structure as the set of the following attributes: (i) stoichiometry, e.g., ABC_3 ; (ii) space group; (iii) set of Wyckoff site occupancies in the unit cell. For example, the calcite prototype ($CaCO_3$) has ABC_3 stoichiometry, $R\bar{3}c$ space group, six atoms on the “6a” (0,0,1/4) Wyckoff site, six atoms on the “6b” (0,0,0) Wyckoff site, and 18 atoms on the “18e” ($x,0,1/4$) Wyckoff site. Leveraging this definition allows us to classify 33 compounds within the OQMD with these attributes as having the prototype of calcite, irrespective of the elements comprising $\{A,B,C\}$, value of x , or unit cell parameters. This classification allows us to treat this group as one, symmetrically unique candidate solution to an experimental structure.

Having identified a relatively small number of prototypes as possible solutions to the experimental structure, we proceed to generate candidate structures by populating the prototypes with the experimentally determined unit cell parameters and elements from the composition. We consider all possible arrangements of elements in the structure, e.g., the two distinct ways to swap Ca and C onto the Wyckoff sites of the calcite prototype. We must also account for the fact that a single prototype can produce multiple structures that, while symmetrically identical, have different local geometries. For example, the structures C23, C25, C29, and C37 ($PbCl_2$, $HgCl_2$, SrH_2 , and Co_2Si , respectively) have the same stoichiometry, space group, and Wyckoff site occupancies (AB_2 , Pnma, {4c, 4c, 4c}), but are distinct structures. In order to decide which of these structures to select as a candidate, we compute the R factor of all compounds with this prototype in the OQMD but with the target composition and experimental lattice parameters substituted in. Since the calculation of R factor is very fast, we can quickly select the structure with the lowest R factor as the candidate (see Sec. II B 2 for an explanation of the R factor calculation). By the end of our procedure, we have generated a set of candidate structures, usually no more than seven structures across three prototypes (see Results section), as possible solutions for the experimental structure.

We note that we initially assume the experimental structure does not have any partially occupied sites. In some cases, this assumption will be inevitably incorrect. We can justify the assumption if we obtain a structure that has a satisfyingly low energy and R factor; otherwise, we say that none of our candidate structures are valid solutions.

B. Evaluating energy and R factor of candidate structures

1. Calculating formation energy using density functional theory

We use DFT to compute the formation energy of all candidate structures. All DFT calculations are performed using the

Vienna Ab-Initio Simulation Package (VASP) v5.4.4 [26,27], using the PBE exchange correlation functional [28], and potentials supplied by VASP with the projector augmented-wave method [29]. We use the same DFT settings as those used for hundreds of thousands of compounds in the OQMD, so that the energetic stability of our candidate structures can be directly compared to other existing compounds. To measure the energetic stability of a candidate structure, we compute the convex hull of formation energies of all OQMD compounds in the relevant phase space and compute the difference between the candidate structure's formation energy and the convex hull energy at the target composition [30,31]. If this difference is less than zero, then the candidate structure is energetically stable and a new convex hull that includes this structure can be constructed. The reader is referred to Ref. [32] for a complete discussion of DFT settings used as well as how formation energies and the convex hull are computed for all compounds in the OQMD.

2. Calculating match to diffraction pattern, or R factor

In addition to measuring the energetic stability of candidate structures using DFT, we also evaluate how well they match the experimental diffraction pattern. We do this by generating a hypothetical diffraction pattern associated with each candidate structure, computing the R factor as a metric for the distance between the hypothetical and experimental diffraction patterns, and slightly refining the atomic coordinates to minimize the R factors. Our methods of computing the R factor are implemented in the Materials Interface (MINT) software [33] and are described in detail in Ward *et al.* [14]. Here, we will provide a brief summary.

For each candidate structure, we search for the locations of all peaks that would be expected based on crystal symmetry and lattice parameters. We then use the following equation [from Pecharsky and Zavalij, Eq. (8.41)] to compute the intensity of a peak located at (hkl) [10]:

$$I_{hkl} = K \times m_{hkl} \times LP(\theta) \times T_{hkl} \times |F_{hkl}|^2, \quad (1)$$

where K is the scaling factor, m_{hkl} is the number of lattice planes corresponding to the peak, $LP(\theta)$ is the Lorentz-polarization factor at the peak's diffraction angle, T_{hkl} is a March-Dollase function used to describe the grain orientation distribution [34], and F_{hkl} is the structure factor, which involves atomic positions. In order to compare this diffraction pattern to the experimental pattern, we use the integrated peak intensities provided by the PDF4+ software to compute the R factor:

$$R = \frac{\sum_{\text{peaks}} (I_{\text{calc}} - I_{\text{obs}})^2}{\sum_{\text{peaks}} I_{\text{obs}}^2}, \quad (2)$$

where I_{calc} and I_{obs} are the candidate structure's calculated peak and experimentally reported peak intensities, respectively. Peaks between the two patterns are paired together if they are within 0.15° of each other; if there are multiple peaks within this range, then they are added together, and if there are no peaks within this range, then a peak of zero intensity is used. While MINT also has the capability of performing Rietveld refinement to obtain the R

factor for continuous patterns, we do not utilize this feature, since many of the PDF entries we attempted to solve only provided integrated diffraction peaks. Prior to reporting the diffraction pattern matches for any candidate structure, we first refine the atomic coordinates of the structure in order to minimize the R factor. For refinement, we use the Broyden-Fletcher-Goldfarb-Shanno (BFGS) algorithm implemented in Dlib [35], during which we optimize the free parameters of Eq. (1), including atomic positions, thermal factors, and texturing.

3. Choosing a structure as the solution

After computing the energy and R factor of each candidate structure, we select the best performing structure as the final solution. To do so, we take all candidate structures with an R factor within 0.2 above the lowest R factor found, and then select the lowest-energy structure among these. We then decide whether the final solution is valid, based on the values of energy and R factor; we provided a detailed description of the validation procedure in the Results section.

A schematic diagram of our prototype searching method is given in Fig. 1 for an example PDF entry, VI_3 (PDF no. 00-023-0719), that contained a diffraction pattern, space group, and unit cell parameters, but no atomic coordinates. We obtain three candidate prototypes (FeF_3 -type, PdF_3 -type, BiF_3 -type) from the OQMD, generate one structure of VI_3 for each prototype, and evaluate their DFT formation energies and R factors. The BiI_3 -type structure has both the lowest formation energy and the lowest R factor and is thus the best-performing prototype of the three. The BiI_3 -type structure also has sufficiently low energy and R factor according to our validation criteria (see the Results section), and so we declare it to be the solution of the VI_3 measurement.

III. RESULTS

A. Prevalence of prototypes among known inorganic compounds

The OQMD contains DFT calculations of experimentally observed inorganic compounds from a 2011 version of the ICSD, excluding those with partial occupancy or very large unit cells. Using the definition of a prototype outlined in Sec. II A, we build a database of all prototypes that exist among 36 807 nonduplicate, stoichiometric, and inorganic compounds in the ICSD portion of the OQMD. An exhaustive database like this one can be compared to existing prototype databases such as the one built from AFLOW [36,37]. The AFLOW prototype database distinguishes prototypes in a similar manner, i.e., by space group, stoichiometry, and Wyckoff sites, but also distinguishes prototypes with different local geometries, e.g., C23, C25, C29, and C37. A key distinction of our prototype database is that it is exhaustive and contains many more prototypes than the 1100 prototypes in AFLOW. From 36 807 nonduplicate, stoichiometric compounds, of which 7852 are binary, 18 482 are ternary, and 8076 are quaternary, we identify a total of 10 203 prototypes, of which 1617 are binary, 4120 are ternary, and 3062 are quaternary. Although this implies that there is an average of

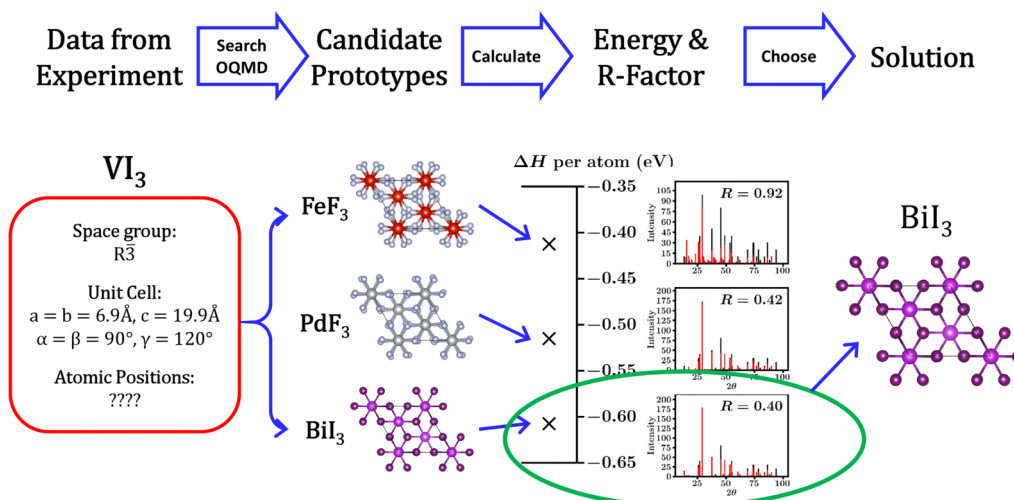


FIG. 1. Flow chart of the prototype searching method to solve structures. The compound VI_3 (PDF no. 00-023-0719) is presented here as an example. Using experimentally determined structure attributes absent atomic coordinates, we search the OQMD for all prototypes with the space group ($R\bar{3}$), stoichiometry (AB_3), and 6 formula units per cell. We then evaluate each of the three prototypes found (FeF_3 -type, PdF_3 -type, BiI_3 -type) using DFT and R factor. We find that the BiI_3 prototype is the most plausible solution because it has the lowest formation energy and R factor.

3.6 compounds per prototype, some prototypes are shared by hundreds of compounds. In Fig. 2, we plot the sorted number of compounds per prototype. There are 77 prototypes with 50 or more compounds, accounting for 27% of the total number of compounds in the ICSD set; these prototypes are listed in Table I. Such prototype sharing reflects that compounds with similar chemistries tend to arrange in the same or similar geometries. For example, binary compounds containing a cation and an anion most commonly have $NaCl$, $PbCl_2$, CaF_2 , and CdI_2 prototypes, while half-Heusler and related prototypes are often observed for compounds with metals and metalloids where the sum of valence electrons equals 8 or 18 [38,39].

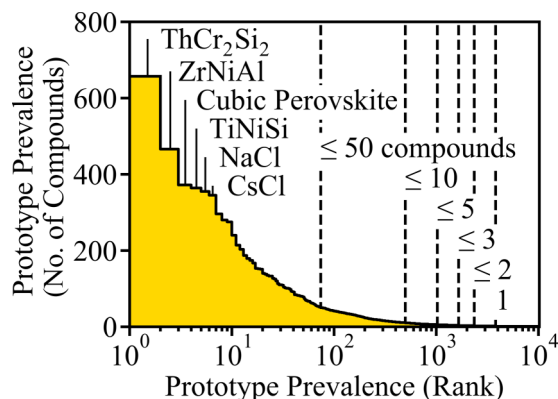


FIG. 2. Sorted numbers of compounds associated with prototypes in the 2011 version of the ICSD, present in the OQMD. The total number of compounds in this ICSD set is 36 807. The most prevalent (rank 1) prototype is $ThCr_2Si_2$, with 657 compounds in the ICSD; and the second-most prevalent (rank 2) prototype is $ZrNiAl$, with 466 compounds. Beginning at rank 10 or so, the trend in prototype prevalence smoothly decays with a wide tail.

While most compounds share common prototypes, there are also 6394 prototypes, of which 981 are binary, 2428 are ternary, and 1928 are quaternary, that are associated with just a single compound in the ICSD. These “one-hit wonders” highlight a shortcoming of the prototype searching method for solving crystal structures: some structures, $\sim 17\%$ of the ICSD, are unique and cannot be solved by searching for already-known prototypes. We can acquire an insight about the one-hit wonders by investigating their statistics. For instance, we note that hydrogen is disproportionately represented among the one-hit wonders: 37% of compounds containing H are one-hit wonders, much higher than the average of 17%. Other elements that are disproportionately represented are N (34%), F (31%), and Xe (56%). Nonmetals and alkali metals in general are disproportionately represented ($\geq 20\%$), while lanthanide and actinide elements rarely occur by themselves ($< 10\%$), except for La (15%) and U (18%). Previous studies have identified which element pairs commonly appear together in compounds of the same prototype [40,41]. In addition, many of the one-hit wonders have unique stoichiometries, such as $Fe_{107}O_{125}$. We find that 941 compounds do not share the same stoichiometry with any other compound in the ICSD. Many-component compounds tend to be unique as well: 955 of 2156 compounds (44%) with five or more components are one-hit wonders. One-hit wonders tend to also have larger unit cells: 29% of compounds with 40 or more atoms are one-hit wonders, compared to 16% of compounds between 20 and 40 atoms and just 6% of compounds with fewer than 20 atoms. Furthermore, most space groups are rarely observed. We find that 159 of 230 space groups have an above average proportion of one-hit wonders ($> 17\%$), and 11 space groups are not observed at all. On the other hand, a select few space groups account for a much larger proportion of ICSD compounds. One such space group is $Fm\bar{3}m$, which is found in 1464 compounds, of which only 33 (2%) are one-hit wonders.

TABLE I. Prototypes shared by 50 or more unique compounds sourced from ICSD in the OQMD. Notation under “Typical chemistry”: MT = metal (groups 1–12), NMT = nonmetal, AA = alkali or alkaline earth metals, TM = transition metal (groups 3–12), PTM = post-transition metal, MTO = metalloid, LT = lanthanide, AT = actinide, PG = pnictogen, CG = chalcogen, HG = halogen. “+” denotes AND and “{...}” denotes OR. The “Strukturbericht” designations are a commonly used classification of crystal structures that is similar to our classification based on prototype. Not every prototype as we have defined it has a corresponding Strukturbericht designation, and some prototypes have multiple Strukturbericht designations.

Prototype	Other name	Typical chemistry	Strukturbericht	Space group	Wyckoff sites	Number of compounds
Binary						
NaCl	rocksalt	MT + NMT	B1	Fm $\bar{3}$ m	4a, 4b	355
CsCl		MT + MT	B2	Pm $\bar{3}$ m	1a, 1b	345
Cu ₃ Au	bogdanovite	MT + MT	D _{0h} , L1 ₂	Pm $\bar{3}$ m	1a, 3c	296
MgCu ₂	laves	MT + MT	C9, C15	Fd $\bar{3}$ m	8b, 16c	240
Mn ₅ Si ₃	mavlyanovite	MT + {PTM,MTO}	D _{8h}	P6 ₃ /mcm	4d, 12g	167
CrB	herzenbergite, westerveldite	variable	B33, B _c , B _f	Cmcm	8c	135
GeS		variable	B14, B16, B27, B29, B31, B _a , B _m	Pnma	8c	120
MoSi ₂		variable	C11 _a , C11 _b	I4/mmm	2a, 4e	110
PbCl ₂	cotunnite	variable	C23, C25, C28, C29, C37	Pnma	12c	109
CaF ₂	fluorite	MT + NMT	C1, C1	Fm $\bar{3}$ m	4a, 8c	104
CaCu ₅	hexagonal ω	MT + MT	D _{2d}	P6/mmm	1a, 2c, 3g	102
AlB ₂		MT + {PTM,MTO}	C32	P6/mmm	1a, 2d	101
NiAs		{LT,AT} + {PG,CG}	B8 ₁	P6 ₃ /mmc	2a, 2c	84
ZnS	zinc blende	MT + {PTM,MTO,NMT}	B3	F $\bar{4}$ 3m	4a, 4c	83
CdI ₂	khayrkite	Variable	C6	P $\bar{3}$ m1	1a, 3d	82
Th ₃ P ₄		{LT,AT} + {PG,CG}	D7 ₃	I $\bar{4}$ 3d	12a, 16c	81
MgZn ₂	laves	MT + MT	C14	P6 ₃ /mmc	2a, 4f, 6h	80
Cu ₂ Sb		{LT,AT} + {PG,CG}	C38	P4/nmm	2a, 4c	69
CuAu		MT + {MT,PTM,MTO}	L1 ₀	P4/nmm	1d, 1d	67
Cr ₃ Si		{groups 4–6} + {groups 8–12,PTM,MTO}	A15	Pm $\bar{3}$ n	2a, 6c	67
Fe ₃ C	cementite	variable	D0 ₁₁ , D0 ₂₀	Pnma	8c, 8d	63
FeS ₂	pyrite	MT + {PG,CG}	C2	Pa $\bar{3}$	4a, 8c	61
Th ₂ Ni ₁₇		{LT,AT} + {Mg,Co,Fe,Ni,Al}		P6 ₃ /mmc	2b, 2d, 4f, 6g, 12j, 12k	60
CuAl ₂	khayrkite	MT + {MT,PTM,MTO}	C16	I4/mcm	4a, 8h	58
Ni ₂ In		MT + {PTM,MTO}	B8 ₂	P6 ₃ /mmc	2a, 2c, 2d	54
Ternary						
ThCr ₂ Si ₂		MT + {MT,PTM} + {PTM,MTO}		I4/mmm	2a, 4d, 4e	657
ZrNiAl		MT + MT + {PTM,MTO}		P6 ₂ m	1b, 2c, 3f, 3g	466
CaTiO ₃	cubic perovskite	MT + {MT,PTM,MTO} + NMT	E2 ₁	Pm $\bar{3}$ m	1a, 1b, 3c	372
TiNiSi		MT + MT + {PTM,MTO}		Pnma	12c	364
Cu ₂ MnAl	full Heusler	MT + MT + {PTM,MTO}	L ₂	Fm $\bar{3}$ m	4a, 4b, 8c	280
PbFCI	matlockite	mixed anion, MT + {MT,PTM,MTO} + {PTM,MTO,NMT}	E0 ₁	P4/nmm	2a, 4c	275
CrNaS ₂	caswellsilverite	MT + {MT,PTM,MTO} + CG	F5 ₁	R $\bar{3}$ m	3a, 3b, 6c	214
CaTiO ₃	orthorhombic perovskite	MT + {MT,PTM,MTO} + NMT		Pnma	4a, 8c, 8d	203
LiBC		MT + MT + {PTM,MTO,NMT}		P6 ₃ /mmc	2a, 2c, 2d	187
Mo ₂ FeB ₂		MT + MT + {PTM,MTO}		P4/nmbm	2a, 4g, 4h	174
Ce ₂ SO ₂		MT + {MT,PTM,MTO} + {PTM,MTO,NMT}		P3m1	1a, 4d	153
MgAgAs	half Heusler	MT + MT + {PTM,MTO}	C1 _b	F $\bar{4}$ 3m	4b, 4c, 4d	152

TABLE I. (Continued.)

Prototype	Other name	Typical chemistry	Strukturbericht	Space group	Wyckoff sites	Number of compounds
SrCuO ₂		MT + MT + {Si,Ge,Sn}		Cmcm	16c	140
AlMg ₂ O ₄	spinel	MT + {MT,PTM,MTO} + CG	H1 ₁	Fd $\bar{3}$ m	8b, 16c, 32e	138
Cr ₂ AlC	MAX phase	MT + {MT,PTM,MTO} + NMT		P6 ₃ /mmc	2a, 2d, 4f	133
CaBe ₂ Ge ₂		MT + LT + {PTM,MTO}		P4/nmm	2a, 2b, 6c	127
FeSb ₂ S ₄	berthierite	MT + LT + CG	E3 ₃	Phma	28c	125
Ca ₂ Nb ₂ O ₇	pyrochlore	MT + {MT,PTM,MTO} + O	E8 ₁	Fd $\bar{3}$ m	8a, 16c, 16d, 48f	113
K ₂ PtCl ₆		AA + {MT,PTM,MTO} + {HG,H}	J1 ₁	Fm $\bar{3}$ m	4a, 8c, 24e	110
LiGaGe		MT + {MT,PTM} + {PTM,MTO}		P6 ₃ /m	2a, 4b	101
YCrB ₄		{LT,AT} + TM + B		Pbam	8g, 16h	100
Mg ₆ Cu ₁₆ Si ₇		{groups 3-4} + {Ni,Cu,Al,Ga,Zn} + {groups 8-14}		Fm $\bar{3}$ m	4b, 24d, 24e, 64f	98
CuHf ₅ Sn ₃		{Bi, groups 3-7} + {Ti,Mn,Fe,Co,Ni,Cu,Zn,B,N,P,O} + {MT,PTM,MTO}		P6 ₃ /mcm	2b, 4d, 12g	93
Co ₃ Gd ₂ B ₂		{LT,AT} + {groups 8-13} + {Ni,Pd,Cu,B,Al,Ga,Si}		P6/nmm	1a, 2c, 3g	84
MgCu ₄ Sn	friauf laves	{LT,AT} + {Pd,Ag,Au,Zn,Cd,In,Sn} + {Co,Ni,Cu,Pt}		F43m	4a, 4d, 16e	82
CeMn ₄ Al ₈		{LT,AT} + {Cu,Cr,Mn,Ni,Fe} + {Mn,Al,Co,Fe}		I4/mmm	2a, 8f, 8i, 8j	80
Zr ₂ Fe ₁₂ P ₇		{LT,AT} + {Mn,Fe,Co,Ni} + {P,As}		P $\bar{6}$	1a, 1d, 1e, 9j, 9k	79
FeO(OH)	goethite, diaspore, chalcocarbonate	metal hydroxides, MT + {MT,PTM,MTO} + {groups 13-16}	F5 ₆	Phma	16c	78
KCuO ₂		MT + MT + {MT,PTM}		Cmcm	8c, 8f	74
CePO ₄	monazite	MT + {MT,PTM,MTO} + {O,H,HG}		P2 ₁ /c	24c	71
Sr ₂ TiO ₄	ruddlesden-popper	AA + MT + {CG,HG}		I4/mmm	2a, 4c, 8e	69
Y ₃ Rh ₄ Ge ₁₃		{LT,AT} + group 9 + {Si,Ge,Sn,Pb}		Pm $\bar{3}$ n	2a, 6d, 8e, 24k	65
K ₂ SO ₄	arcanite	AA + {MT,PTM,MTO} + {H ₁ CG,HG}		Phma	20c, 8d	62
KAsF ₆		MT + {MT,PTM,MTO} + F		R $\bar{3}$	3a, 3b, 18f	62
ZrSiO ₄	zircon	{groups 3-5} + O-ate	S1 ₁	I4 ₁ /amd	4a, 4b, 16h	59
CuFeS ₂	chalcocopyrite	MT + {PTM,MTO} + {MTO,PG,CG}	E1 ₁	I42d	4a, 4b, 8d	56
[NH ₄]CdCl ₃		MT + {MT,PTM} + {CG,HG}		Phma	20c	55
BaSO ₄	barite	MT + {PTM,MTO,NMT} + NMT	H0 ₂	Phma	16c, 8d	53
K ₂ UF ₆		MT + {MT,PTM} + {PTM,MTO}		P $\bar{6}$ 2m	4a, 2b, 4d, 8e	52
CaWO ₄	scheelite	MT + O-ate	H4	I4 ₁ /a	4a, 4b, 16f	52
U ₂ Co ₃ Si ₅		{LT} + {groups 8-10} + {Si,Ge}		Ibam	4a, 4b, 8g, 24j	51
YNi ₅ Si ₃		MT + {MT,PTM,MTO} + {MTO,NMT}		Phma	36c	50
BaNiSn ₃		{groups 1-3} + {groups 8-10} + {Si,Ge,Sn}		I4mm	6a, 4b	50
Cu ₄ Gd ₃ Ge ₄		{LT,AT} + {Ni,Pd,Cu,Ag} + {Si,Ge,Sn}		Immm	2a, 4h, 4i, 4j, 8l	50
Quaternary						
K ₂ NaAlF ₆	elpasolite	AA + MT + {MT,PTM} + {F,Cl,O}		Fm $\bar{3}$ m	4a, 4b, 8c, 24e	179
La ₃ CuSi ₅ 7		LT + TM + {PTM,MTO} + {S,Se}		P2 ₁	2a, 4b, 18c	151
La ₂ LiSbO ₆		AA + {LT,AT} + {MT,PTM} + O		P2 ₁ /c	2b, 2c, 16e	133
KAlP ₂ O ₇		MT + MT + pyro-O-ate		P2 ₁ /c	44e	92
KCuZrS ₃		AA + {LT,AT} + {groups 11-12} + {S,Se,Te}		Cmcm	4b, 12c, 8f	91
CuZrSiAs		MT + MT + 2 Anions		P4/nmm	2a, 2b, 4c	67
K ₃ NaFeCl ₆	rinneite	AA + MT + {MT,PTM} + O		R $\bar{3}$ c	6a, 6b, 18e, 36f	50
Quinary						
[NH ₄]Mg[SO ₄] ₂ · [H ₂ O] ₆		AA + MT + H + {S,Se} + O		P2 ₁ /c	2a, 60e	54

B. High-throughput structure solution by prototype searching

1. Description of target compounds from the powder diffraction file

As the prototype searching method is cheap, often costing only a few DFT calculations, we leverage this approach to perform “high-throughput” structure solution for numerous entries from the International Centre for Diffraction Data (ICDD) database within the PDF for which the atomic coordinates are missing but other structure details are known. We start with 80 624 entries missing atomic coordinates in the 2018 version of the PDF4+ software. We screen for entries that satisfy the following criteria:

(i) The entry is “primary” status as identified by the PDF4+ software, i.e., it is not an alternative to another similar entry.

(ii) The diffraction experiment was performed under ambient conditions. We note that the enthalpy of high-pressure compounds can be accounted for in DFT by supplying external pressure to the stress tensor. Furthermore, the enthalpies of high-pressure compounds can be compared to those of other compounds in the OQMD [42].

(iii) The compound is binary, ternary, or quaternary.

(iv) The compound is inorganic and does not contain noble gases, actinides, or radioactive elements.

(v) The diffraction data quality is listed as “star,” “good,” or “indexed,” indicating that the diffraction pattern represents a single-phase crystal with minimal impurities. While structures with poorer quality diffraction patterns can still be solved, their R factors may be less useful.

(vi) The space group and number of formula units per unit cell are already known.

(vii) Reduced cell volumes are less than 3000 \AA^3 and unit cells contain few enough atoms to be cheaply assessed by high-throughput DFT: cubic, hexagonal, trigonal, and tetragonal cells contain 80 or fewer atoms; orthorhombic cells contain 40 or fewer atoms; monoclinic and triclinic cells contain 20 or fewer atoms.

(viii) The structure does not evidently contain partially occupied sites, i.e., the listed composition contains only natural numbers and it is possible to generate a structure with a set of fully occupied Wyckoff sites given the listed space group and number of formula units per unit cell. We note that a PDF entry satisfying these conditions may not necessarily represent a fully occupied structure. We can justify the validity of our prototype structures based on DFT energy and R factor.

(ix) There is no existing OQMD nor ICSD compound with the same composition and space group.

(x) There is at least one prototype in the OQMD matching the known stoichiometry, space group, and number of formula units per unit cell.

We find 603 PDF entries that satisfy the above constraints. We additionally find hundreds of entries that satisfy all the above constraints except for the last one, i.e., there is no prototype in the OQMD that matches the provided stoichiometry, space group, and number of atoms per unit cell. However, it is possible that the listed space group is incorrect, and hence we attempt to solve these by searching within the crystal system, e.g., tetragonal space groups, instead of the listed space group.

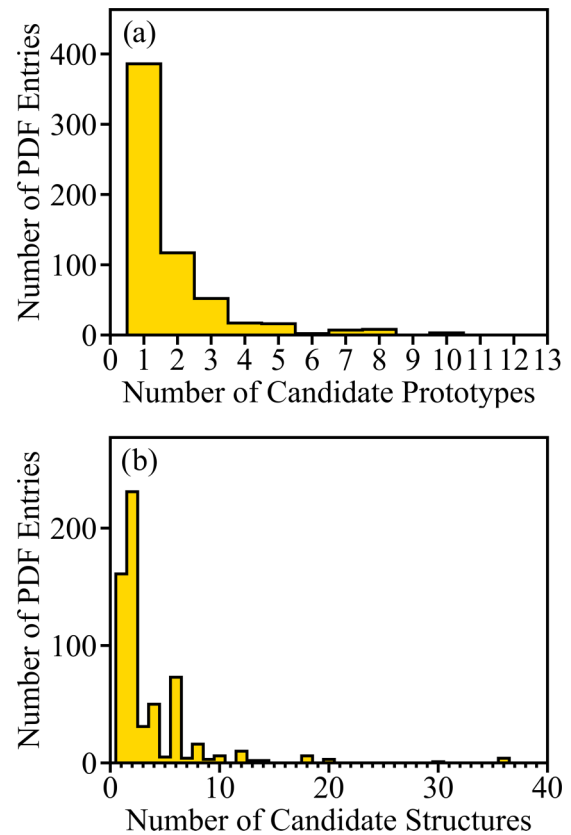


FIG. 3. Distribution of the number of (a) OQMD prototypes and (b) candidate structures matching stoichiometry, space group, and number of formula units per unit cell of 603 PDF entries with missing atomic coordinates.

2. Summary of structures obtained by prototype searching

For 603 PDF entries with diffraction data but no structure, we find at least one prototype in the OQMD that matches the provided space group, stoichiometry, and number of formula units per unit cell. In Fig. 3(a), we plot a distribution of the number of prototypes found per PDF entry. The highest number of prototypes is only ten. In most cases (386, or 64%), only one prototype is found. Although the number of candidate prototypes is almost always very few, each prototype can produce multiple structures representing the possible ways to arrange elements onto the prototype’s Wyckoff sites. Despite this, there are rarely more than a dozen structures to evaluate [see Fig. 3(b)].

After computing the DFT energy and R factor of all candidate structures, we select the structure with the lowest DFT energy among all candidates within 0.2 of the lowest R factor. We are thus left with 603 structure candidates, each one outperforming other candidates for every attempted PDF entry. For ten of the 603 entries, we find a candidate with a different space group within the same crystal system that outperforms all candidates with the reported space group. In these ten cases, the structure with the same space group fails our validation checks of energy and R factor (described in the following section), while the structure with a different space group passes these checks; we thus opt to present the ten structures with a different space group. In addition, we find that for 21 of

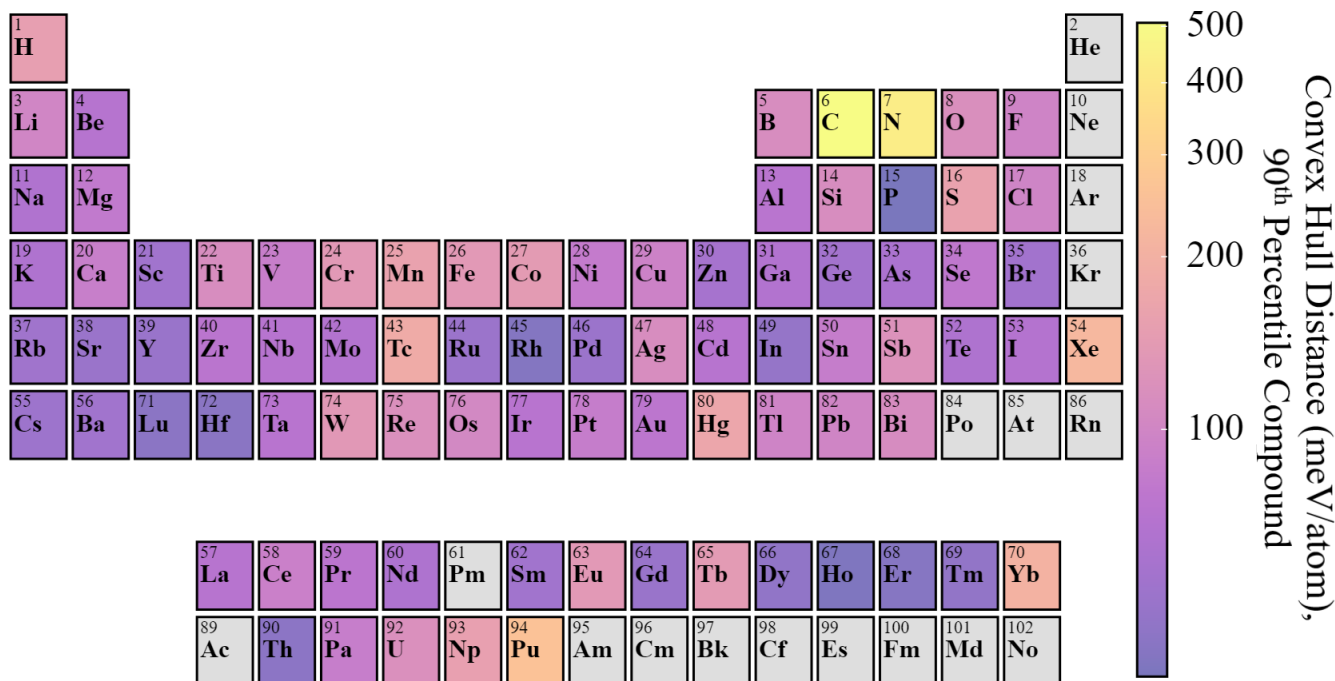


FIG. 4. DFT-computed 90th percentile convex hull distance (meV/atom) of ICSD compounds containing each element on the periodic table. The metastability of materials is chemistry dependent, with carbides and nitrides standing out as being particularly high in energy. Gray shaded elements are excluded from this analysis.

the PDF entries, while there is no prototype in the OQMD that matches the reported space group, stoichiometry, and number of formula units per cell, there is a candidate with a different space group within the same crystal system that passes our validation checks. In total, we present 624 structures (603 + 21) in the following analysis. Of these, 521 pass our validation checks of energy and R factor, and we thus declare them to be solved.

3. Analysis of structures obtained by prototype searching

After selecting the 624 best-performing candidate structures, one for each PDF entry, we proceed to assess their validity by examining the values of energy and R factor. As for energy, we are interested in the difference between the structure's formation energy and the OQMD convex hull at the relevant composition with this structure included. If this value is 0 meV/atom, then our candidate structure is stable and thus highly plausible. However, metastable compounds with nonzero convex hull distances are also common in nature. Although not all hypothetical structures with nonzero convex hull distances can be synthesized, they should be considered as potentially valid solutions in our structure search. Analyses of experimentally known metastable compounds calculated by DFT have revealed that, while most metastable compounds are within 100 meV/atom of the convex hull, the values of convex hull distance are highly dependent on chemistry [43,44]. In Fig. 4, for each element, we plot the 90th percentile convex hull distance for ICSD compounds containing that element. There is a stark contrast in the convex hull distances as a function of element; carbides and nitrides are especially metastable [45–47]. We thus opt to use these values of con-

vex hull distance as cutoff values in determining whether the structures we obtain from the prototype searching method are valid based on energy. Specifically, for a compound of interest, e.g., $\text{Ba}_2\text{CeSnO}_6$ (PDF no. 00-056-0332) solved in this work, we use the highest of the four 90th percentile convex hull distance values as the cutoff: 116 meV/atom for oxygen. Since our best-performing candidate structure for $\text{Ba}_2\text{CeSnO}_6$ is 102 meV/atom above the convex hull, we deem this structure valid based on energy.

In Fig. 5(a), we plot the convex hull distances of 624 compounds with structures obtained by prototype searching in this work, along with those of ICSD compounds. If all 624 of these compounds were correctly solved, then we would expect that their convex hull distances would line up well with the ICSD distribution. Although the proportion of our compounds that lie on the convex hull is high (275 compounds within 5 meV/atom of the hull), this proportion is shy of the ICSD, where 61% of compounds are within 5 meV/atom of the convex hull. We find that 544 compounds (87%) pass our validation criterion for energy, compared to 93% of the ICSD. However, we also expect that the prototype searching method will fail to solve compounds that are “one-hit wonders,” i.e., compounds that do not share a prototype with any other in the ICSD (described in Sec. III A). Since as many as 17% of known compounds are one-hit wonders, this inevitable shortcoming of our approach might explain why our convex hull distances are higher than those of ICSD compounds, on average.

R factors of all 624 compounds are plotted in Fig. 5(b). To give context to our R factor values, we overlay a distribution of R factors for 136 randomly selected already-solved ICSD compounds with diffraction patterns stored in the PDF. With

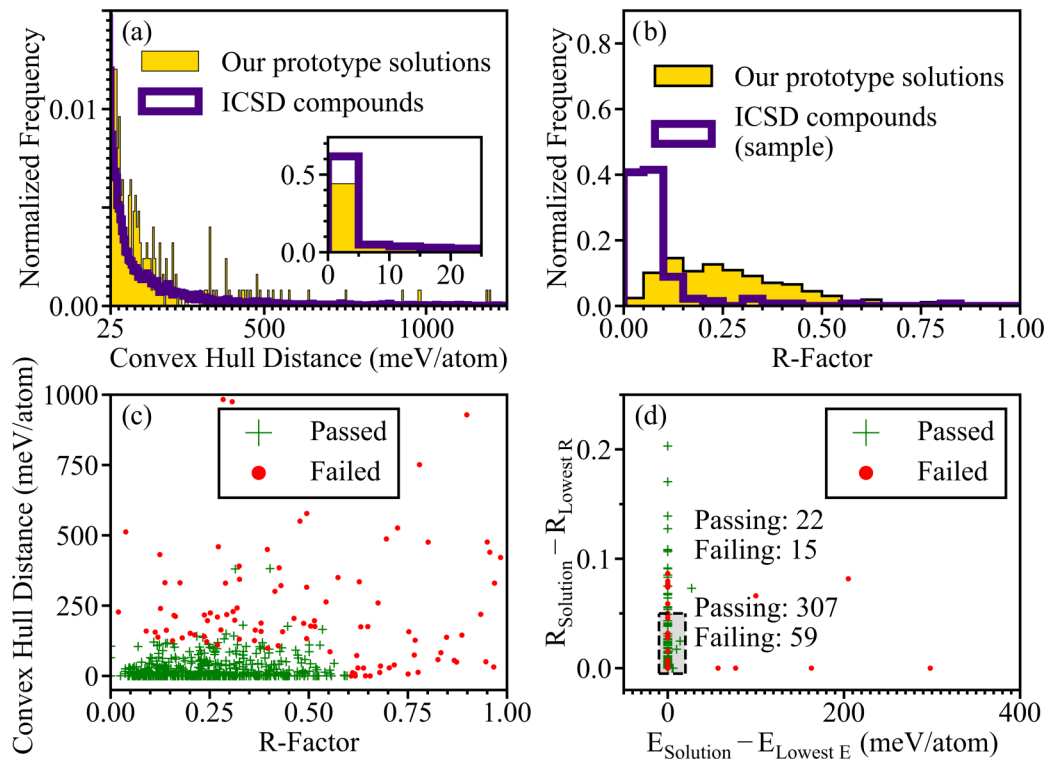


FIG. 5. (a) Distribution of convex hull distances of 624 compounds with structures obtained by prototype searching in this work, compared to 23 247 ICSD compounds that have been calculated in the OQMD. Inset is the same distribution between 0 and 25 meV/atom; almost half of the 624 compounds lie within 5 meV/atom of the convex hull, somewhat shy of the ICSD. (b) Distribution of R factors of 624 compounds with structures obtained by prototype searching in this work, compared to 136 randomly chosen solved compounds from the PDF. (c) Convex hull distances and R factors of 624 compounds with structures obtained by prototype searching. Green pluses and red dots are compounds that passed and failed validation checks, respectively. (d) Discrepancies in the best-performing candidate structure energies and R factors with the lowest-energy and lowest- R -factor candidate structures. Only cases with multiple candidate structures are shown (403 of 624 PDF compounds). The vast majority (91%) of cases lie within the shaded region; in other words, the best-performing candidate structure usually is close to the lowest energy and lowest R factor of all candidate structures. Cases with low discrepancy are also more likely to pass our validation checks (84%) than cases with high discrepancy (59%).

a median value of 0.25, our structures have higher R factors overall than the already-solved structures (median of 0.06). We argue that this discrepancy does not suggest an issue with our prototype solutions, because many of our solutions with R factor greater than 0.05 are clearly the right answers by inspection. For example, 44 of our compounds evidently have the elpasolite (K_2NaAlF_6) prototype, since they have A_2BCD_6 stoichiometry, space group of $\text{Fm}\bar{3}m$, and four formula units per unit cell. The only other possible prototype is typically very high in energy. Indeed, we find that 17 of the elpasolite compounds lie on the convex hull, despite R factors ranging from 0.05 to 0.52. The high R factors are not due to any issue with our refinement code either; despite elpasolite having only one degree of freedom to refine (the x coordinate of the 24e site), we still obtain high R factors. We argue that the high R factors highlight an issue with the diffraction patterns, not with our prototype searching approach. Because we cannot impose a strict R factor validation criterion, we look to the relationship with energy values to decide on a cutoff R factor value. Stable compounds tend to have low R factors: 50% of compounds with an R factor below 0.1 lie on the convex hull, 53% with an R factor between 0.1 and 0.2 lie on the convex hull, 50% between 0.2 and 0.3. Following these intervals, we

have 41%, 36%, 36% between 0.5 and 0.6, 24% 0%, 0%, and 0% between 0.9 and 1.0. As the proportion of stable compounds begins dropping off at 0.6, we opt to use an R factor of 0.6 as the cutoff value for validation. This works out to be a generous cutoff value: 580 of our compounds (93%) have an R factor less than 0.6.

Combining our validation checks, we declare that 521 of 624 (83%) of our compounds are “solved” based on low convex hull distance and an R factor less than 0.6. The convex hull distances and R factors of all 624 compounds are plotted in Fig. 5(c). Although most of our compounds simultaneously pass both energy and R factor validation criteria, there are cases that pass only one of the criteria. Compounds with high energy and low R factor might have structures that happen to exhibit a close match to diffraction data while being theoretically unphysical. On the other hand, compounds with low energy and high R factor could be polymorphs of the “true” structure observed in experiment. It is also possible that compounds with low energy and high R factor are, in fact, correctly solved; indeed, we are using an atypically high cutoff for R factor. Despite the high R factors, we argue that the R factors are helpful in distinguishing structures that best match experimental data. In Fig. 5(d), we demonstrate that

even though many of our structures have a high R factor, they are most often both the lowest-energy and lowest- R -factor candidate out of all possible candidates. Considering 403 cases where more than one possible candidate structure exists, we find that 366 (91%) of our best-performing candidates lie within the shaded region, i.e., are within 20 meV/atom of the lowest-energy candidate and 0.05 of the lowest- R -factor candidate. Compounds that pass our validation criteria are even more likely to lie within the shaded region (93%) than failing compounds (80%). This result demonstrates that even when all candidate structures have a high R factor, we can still use the R factor to distinguish the best structure from other candidates; however, DFT energy is often helpful in determining which candidates are physical.

Upon inspecting our prototypes selected for the PDF compounds, we noticed that they are quite often chemically similar to other ICSD compounds with the same prototype. For example, the solution to Ag_7SbS_6 (PDF no. 00-021-1333) is the prototype of Ag_7AsS_6 , found in ICSD. We can quantify “chemical similarity” by taking advantage of a data mined Pettifor chemical scale developed by Glawe and co-workers [41]. They computed a chemical similarity metric P_{AB} for pairs of elements A and B on the periodic table. To compute the chemical similarity between two compounds, e.g., Ag_7SbS_6 and Ag_7AsS_6 , we take the product $P = \prod P_{AB}$ of chemical similarities of the closest-matching element pairs in the two compounds, setting P_{AB} to 1 when the elements are identical and 0 if the element pairs rarely or never occur in the ICSD. For all of our chosen prototypes, we searched for the ICSD compound of the same prototype with the highest chemical similarity; the results are plotted in Fig. 6. The trends in the plots demonstrate that compounds that pass our validation criteria are more likely to be chemically similar to ICSD compounds than compounds that fail. The chemical similarities we find here give us an extra layer of confidence in our solutions.

All 521 compounds solved in this work are provided in the Supplemental Material [25], along with a complete tabular summary of all 624 attempts. In addition, all compounds can be found in the OQMD, which can be accessed via the web or directly downloaded (see Ref. [48]). As there are 36 807 unique ICSD compounds already in the OQMD, we have expanded the set of all experimentally observed compounds in the OQMD by 1.4%. The simplicity and efficiency of the prototype searching method presented in this paper has thus enabled us to significantly expand the set of experimentally observed compounds accessible to DFT. It will be of interest to further study the properties of these materials. For example, as shown in Fig. 7, 284 of our solved compounds have nonzero band gaps within 4 eV, making them potential candidates for semiconductor applications. In addition to the 521 recently solved compounds, we find 33 PDF “unsolved” compounds where there is either no matching prototype in the OQMD or no prototype matching the reported space group that passes our validation checks, but there is a solution with a different space group within the same crystal system that not only passes our validation checks but also already exists in the ICSD. We provide these 33 solutions in a separate table in the Supplemental Material [25].

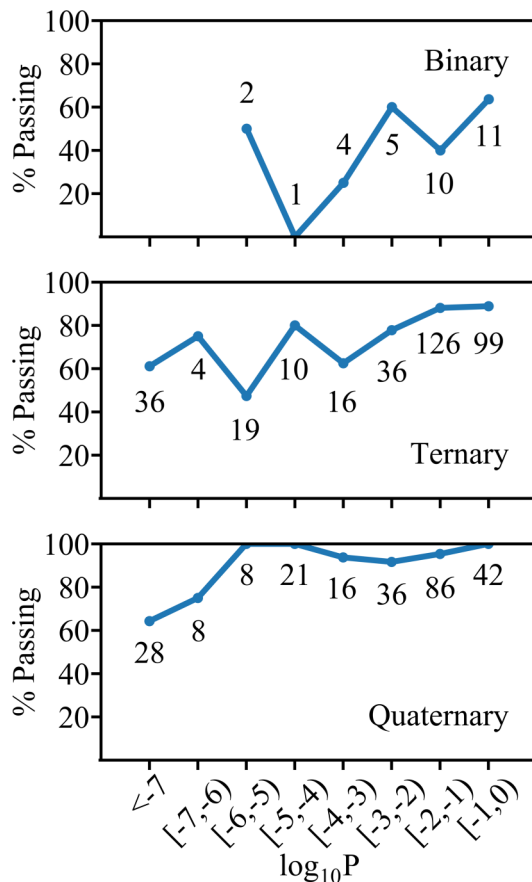


FIG. 6. The percentage of the prototypes chosen for each of 624 PDF compounds that pass our validation criteria, plotted against the chemical similarities P of these compounds to ICSD compounds of the same prototype. The chemical similarities P are binned by decades on a log scale; see the text of Sec. III B 3 for the definition of P between a pair of compounds. The top, middle, and bottom plots focus on binary, ternary, and quaternary compounds, respectively. The numbers of compounds that fall within each range of chemical similarities are shown beside the data points. The trends demonstrate that compounds that pass our validation criteria are more likely to be chemically similar to ICSD compounds than compounds that fail.

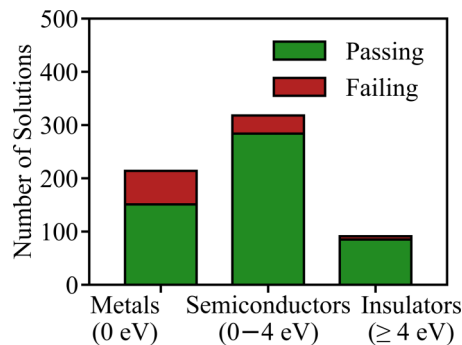


FIG. 7. 151 of our solutions that pass validation criteria are metals (0 eV), 284 are semiconductors (0–4 eV), and 85 are insulators (≥4 eV); band gap was not determined for one solution.

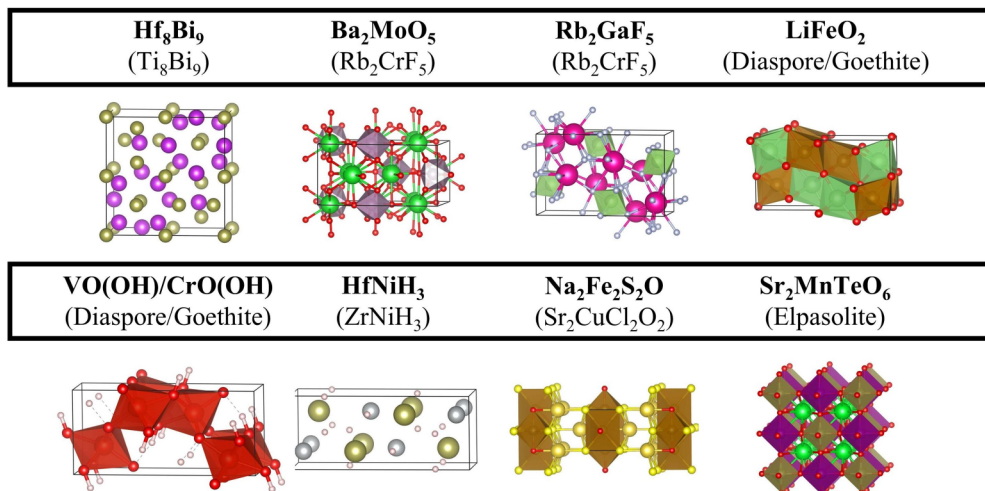


FIG. 8. Crystal structures of 9 of the 521 materials solved using prototypes. The compositions of the solved materials are in bold, and the prototypes are in parentheses. Note that some of the solutions presented here have the same prototype, specifically Ba₂MoO₅ and Rb₂GaF₅ as well as LiFeO₂, VO(OH), and CrO(OH).

4. Examples of solutions obtained by prototype searching

In this section, to demonstrate the prototype searching method at work, we discuss nine PDF compounds that we solved. An illustration of the solved compounds is shown in Fig. 8. All of the nine compounds in this section pass our validation criteria of energy and R factor and are chemically similar to other compounds in the ICSD with the same prototype. For some of these compounds, the paper describing the diffraction experiment stated the name of the prototype that matches our solution but did not present atomic coordinates. Although the prototypes of these compounds were already known, our prototype searching method enabled us to obtain atomic coordinates for all structures and expand the OQMD.

(a) *Hf₈Bi₉*: In the PDF entry for Hf₈Bi₉ (no. 00-051-0679), a diffraction pattern is supplied along with a space group (P4/nmm), unit cell, and formula units ($Z = 2$), but atomic coordinates are missing [49]. Because the atomic coordinates are missing, this compound did not previously exist in the ICSD nor OQMD and has thus been excluded from DFT studies. However, in the reference for this entry, the authors presented the prototype Ti₈Bi₉, complete with atomic coordinates, and stated that Hf₈Bi₉ has the same prototype as Ti₈Bi₉. As Ti₈Bi₉ is indeed already in the OQMD, we use the prototype searching method to complete the structure of Hf₈Bi₉. Specifically, our crystal structure for Hf₈Bi₉ consists of the unit cell parameters provided by the PDF entry for Hf₈Bi₉, and the DFT-relaxed atomic coordinates of Bi plus the atomic coordinates of Hf substituted for Ti in the already-solved compound Ti₈Bi₉. We find that this structure matches the diffraction pattern well [R factor = 0.21; see Fig. 9(a)] and is on the convex hull [Fig. 9(b)].

(b) *Ba₂MoO₅ and Rb₂GaF₅*. The reference for Ba₂MoO₅ provided by the PDF (no. 00-025-0011) for this diffraction pattern describes the structure as isostructural with K₂VO₂F₃ with Pnma symmetry and four formula units per unit cell but does not provide atomic coordinates [50]. The OQMD prototype Rb₂CrF₅ is indeed isostructural with K₂VO₂F₃ in

that it has the same space group and Wyckoff site occupancies (though with different stoichiometry), and, with the elements Ba, Mo, and O substituted in, lies on the Ba-Mo-O convex hull and has an R factor of 0.16, indicating it is a highly plausible solution. An existing ICSD compound, Ba₂WO₅, has the same prototype and is highly chemically similar to Ba₂MoO₅ ($P = 0.25$). We also considered three other candidate prototypes: BaSi₂O₅ (convex hull distance = +19 meV/atom, R factor = 0.28), KPd₂F₅ (hull distance = +102 meV/atom, R factor = 0.42), and CsN₂H₅ (DFT failed to converge, R factor = 0.19). Since the Rb₂CrF₅ prototype has both the lowest energy and lowest R factor out of all candidates and passes our validation criteria, we deem it to be the solution to Ba₂MoO₅.

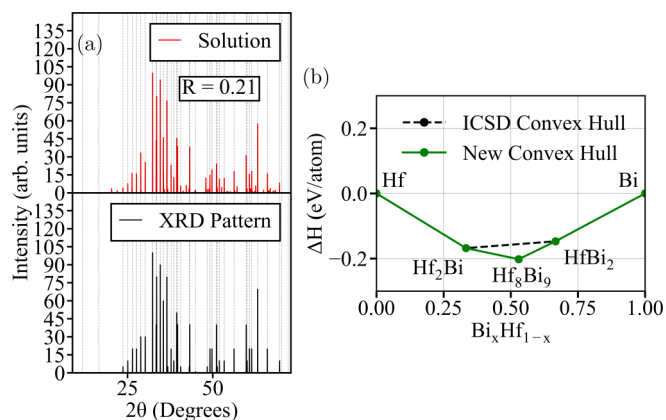


FIG. 9. (a) Diffraction pattern describing the prototype solution Ti₈Bi₉ for the compound Hf₈Bi₉ from the PDF (top), compared to the experimental XRD pattern reported in the PDF for Hf₈Bi₉ (bottom). The R factor is 0.21, highlighting a close match between the two patterns. (b) “New” Hf-Bi convex hull including Hf₈Bi₉ solution generated from DFT calculations stored in the OQMD, plotted against the “old” convex hull found in the OQMD not including Hf₈Bi₉. Since our Hf₈Bi₉ solution was lower in energy by 46 meV/atom than the old convex hull at that composition, we construct a new convex hull to include our solution.

The compound Rb_2GaF_5 from the PDF (no. 00-032-0914) has the same story [51]. The prototype Rb_2CrF_5 is the solution because it is on the convex hull and has an R factor of 0.43, lower than other candidates. The ICSD compound Rb_2FeF_5 , with the same prototype, is the most chemically similar to Rb_2GaF_5 ($P = 0.04$).

(c) *LiFeO₂ polymorph, VO(OH), and CrO(OH)*. While several polymorphs of LiFeO_2 are known, to our knowledge there are no reports of the atomic coordinates of the Pnma polymorph of LiFeO_2 listed in the PDF (no. 00-052-0698), and consequently its structure was not previously in the OQMD. The reference listed in the PDF reported that the goethite polymorph of LiFeO_2 is rechargeable in lithium cells [52]. We find that goethite, or FeO(OH) , is the correct prototype for this polymorph of LiFeO_2 when Li atoms are substituted for H, since the convex hull distance is only +36 meV/atom and the R factor is 0.13. We reject another candidate, YPd_2Si (convex hull distance = +272 meV/atom, R factor = 0.29), because it is highly unstable.

Single crystals of VO(OH) (PDF no. 00-011-0152), found in montroseite, were examined by x-ray crystallography in 1953 and found to be isostructural with diaspore, or AlO(OH) [53]. Diaspore and goethite are the same prototype (as is chalcostibite). An incomplete structure for VO(OH) having only V and O positions can be found in the ICSD [54]; hydrogen positions are missing, presumably since they cannot be detected in the x-ray pattern, and as a result, the properties of VO(OH) have not been studied with DFT. We obtain a complete structure for VO(OH) , including H positions, by substituting V, O, and H into the sites of the diaspore structure and find it to be nearly stable (convex hull distance = +8 meV/atom, R factor = 0.48). We similarly apply our prototype searching method to fill in the H coordinates of the CrO(OH) structure (PDF no. 00-025-1497), which was previously found to resemble diaspore [55]. Our structure for CrO(OH) is close to the convex hull (+8 meV/atom), but has a poor match to the diffraction pattern (R factor = 0.62).

(d) *HfNiH₃*. We report several stable hydrides in this work, including four lanthanide hydrogen chalcogenides. It is tricky to solve the hydrogen positions from x-ray diffraction data since hydrogen scattering is too weak to detect in an x-ray diffraction pattern. In the case of HfNiH_3 (PDF no. 00-047-1412), the peak indices could be matched to those of space group Cmcm . The authors inferred that the H atoms situate within the HfNi structure (space group Cmcm , eight atoms per unit cell) [56]. Separate DFT studies of HfNiH_3 utilized the assumption that H atoms occupy octahedral and tetrahedral interstices between Hf and Ni atoms in order to estimate the positions of H [57,58]. We find ten unique prototypes having Cmcm space group and 20 atoms per cell, but the best performing prototype is that of ZrNiH_3 (convex hull distance = 0, R factor = 0.45). This is indeed a superstructure of HfNi , in which nine Hf-H bonds constitute edge-sharing polyhedra. Notably, the other nine prototypes with much higher energy are not hydrides. The ZrNiH_3 structure in the OQMD, complete with H positions, was obtained using neutron diffraction [59]; we utilize the solution from this past neutron diffraction study to complete the structure of HfNiH_3 .

(e) *Na₂Fe₂S₂O*. The diffraction pattern for the mixed anion compound $\text{Na}_2\text{Fe}_2\text{S}_2\text{O}$ was obtained through an ICDD Grant-

In-Aid (PDF no. 00-065-0329) [9]. The atomic positions are missing from the entry, but the space group and number of formula units were reported to be $I4/mmm$ and $Z = 2$, respectively. We conclude that the $\text{Sr}_2\text{CuCl}_2\text{O}_2$ prototype is a convincing solution. Since there are $3! = 6$ unique ways to arrange the elements Na, Fe, and S onto the 4c, 4e, and 4e Wyckoff sites of the $\text{Sr}_2\text{CuCl}_2\text{O}_2$ prototype, we check each one individually and find that the best arrangement is on the convex hull and has an R factor of 0.21. Interestingly, this arrangement has cation Na^{1+} occupying the anion Cl^{-1} site of $\text{Sr}_2\text{CuCl}_2\text{O}_2$, and likewise has anion O^{2-} occupying the cation Cu^{2+} site. Such an arrangement could be a direct consequence of the balancing of oxidation states in $\text{Na}_2\text{Fe}_2\text{S}_2\text{O}$. Other arrangements are significantly higher in energy, so they are ruled out.

(f) *Elpasolites*. Many materials presented in this work share the same prototypes with one another. Forty-four of the materials in this work have the elpasolite structure, or K_2NaAlF_6 , which is an ordered double perovskite. Elpasolite is one of two prototypes that are possible given the experimentally known $\text{Fm}\bar{3}\text{m}$ space group, ABC_2D_6 stoichiometry, and 40 atoms per unit cell. The other possibility is the same as elpasolite but with the D_6 atoms occupying the “24d” Wyckoff site rather than the “24e” Wyckoff site; this prototype is rare in the ICSD and is typically higher in energy by 1000–2000 meV/atom. Elpasolite is the most common quaternary prototype in nature, with 179 examples from the ICSD subset of the OQMD. All of the elpasolite-type compounds we present here are within +114 meV/atom of the OQMD convex hull (22 are on the hull), and have an R factor below 0.52 (28 had R factors below 0.20), indicating that they were all stable or metastable and had reasonable pattern matches. For the metastable cases, the ground state is often a distortion of double perovskite; in the case of $\text{Sr}_2\text{MnTeO}_6$ (PDF no. 00-029-0897), the ground state is monoclinic ($\text{P}21/c$) double perovskite, which is 24 meV/atom lower in energy than the elpasolite decoration. Recently there has been interest in identifying more elpasolite compounds. It is difficult to perform high-throughput DFT calculations of elpasolite structures using elemental substitution, since there are millions of permutations. Faber *et al.* developed a machine learning model to predict the energies of elpasolite compounds, and found 90 structures on the convex hull, after considerable model training and DFT calculations of 2133 candidates [60]. We note that one of our 44 elpasolites is in their set of 90: Cs_2KGaF_6 (PDF no. 00-021-0849).

IV. DISCUSSION

Structure solution is a challenging roadblock to materials discovery. Thankfully, crystal structures are rarely unique, and a successful structure solution can often be obtained by searching among a relatively small number of prototypes as valid candidates. We apply this simple and inexpensive strategy to solve 521 structures taken from the PDF. Utilizing the OQMD as an exhaustive database of prototypes as well as to validate the energetic stability of candidates along with the R factor, we have identified potential solutions to these materials, and we have a high degree of confidence in our solutions.

The prototype strategy employed in this work can be improved upon in many ways. One way is to tweak the definition of a prototype to distinguish different structures more effectively. In our approach, we define the prototype of a structure as the combination of its stoichiometry, space group, and Wyckoff site occupancies. All structures from the OQMD sharing these characteristics are grouped into one prototype. However, within these constraints, there can be many degrees of freedom in atomic coordinates and lattice parameters, and it is possible for two structures with the same prototype, as defined in this paper, to in fact have very different local geometries, a problem described at length by Trimarchi *et al.* [61]. Our workaround is to choose the OQMD compound whose structure, with its elements replaced by the target elements, gives the lowest R factor, since the calculation of R factor is nearly instantaneous compared to DFT. A more reliable workaround would be to devise a stricter prototype definition capable of properly distinguishing structures with different local geometries. For instance, some definitions apply additional restrictions on unit cell axial ratios and angles [62]. One could also quantify the difference between structures using a distance metric, such as one devised from radial distribution functions [63] or atomic/molecular matching algorithms [64–66]. Moreover, if a given prototype has many internal degrees of freedom, one could conceivably develop an algorithm to optimize DFT and the R factor within the search space of that prototype.

Another way to improve the performance of the prototype searching method is to recommend the most plausible prototypes first, prior to evaluating them with DFT. There was no need to do so for this work, since constraining the search to the PDF-provided space group, composition, and number of atoms per unit cell of all solved materials reduced the number of candidate prototypes fewer than 3 in most cases. If, on the other hand, we could not constrain the search as much, there would have been too many candidates to evaluate. Existing techniques for recommending prototypes as candidates for an unsolved compound involve machine learning [67] as well as data-mined ion substitution [40].

Furthermore, we suggest incorporating prototypes as initial guesses to structural optimization algorithms as a way to improve their performance. If an existing prototype is indeed the correct answer, as is the case for most compounds in nature, then optimization algorithms would converge immediately without wasting computational resources.

V. CONCLUSION

In this work, we outline a prototype searching method and use it to solve the structures of 521 PDF diffraction patterns. For each diffraction pattern, we obtain all prototypes in the OQMD satisfying the known stoichiometry, space group, and number of atoms per unit cell that are provided by the PDF, and select a structure based on DFT energy and R factor. We then validate each structure by assessing its energetic stability with respect to competing phases in the OQMD as well as the R factor. The 521 solved compounds, along with a table of descriptive details, can be found in the Supplemental Material [25], and the compounds are also available in the latest release of OQMD. Identifying structures for these experimentally observed materials enables us to explore their properties from first-principles and unveil their potential for a wide variety of future applications. To allow others to take advantage of the low cost of our prototype searching method, we plan to update the FPASSMGR software package, currently available under an open-source license [14,68]. Currently, the FPASSMGR package can be used to perform FPASS calculations along with validation checks, including evaluating the energetic stability of candidate solutions against OQMD competing phases. Our update will automate the process of searching for and evaluating candidate prototypes from the OQMD for many unsolved compounds in parallel.

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