

The geometric blueprint of perovskites

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Perovskite minerals form an essential component of the Earth’s mantle, and synthetic crystals are ubiquitous in electronics, photonics, and energy technology. The extraordinary chemical diversity of these crystals raises the question on how many and which perovskites are yet to be discovered. Here we show that the ‘no-rattling’ principle postulated by Goldschmidt in 1926, describing the geometric conditions under which a perovskite can form, is much more effective than previously thought, and allows us to predict new perovskites with a fidelity of 80%. By supplementing this principle with inferential statistics and internet data mining we establish that currently known perovskites are only the tip of the iceberg, and we enumerate ninety thousand hitherto-unknown compounds awaiting to be studied. Our results suggest that geometric blueprints may enable the systematic screening of millions of compounds, and offer untapped opportunities in structure prediction and materials design.

I. INTRODUCTION

Crystals of the perovskite family rank among the most common ternary and quaternary compounds and are central to many areas of current research¹⁰. For example silicate perovskites constitute the most abundant minerals on Earth², and synthetic oxide perovskites find applications as ferroelectrics³, ferromagnets⁴, multiferroics⁵, high-temperature superconductors⁶, magnetoresistive sensors⁷, spin filters⁸, superionic conductors⁹ and catalysts¹⁰. Halide perovskites are promising for high-efficiency solar cells, light-emitting diodes and lasers^{11–13}; their double perovskite counterparts are efficient scintillators for radiation detection¹⁴. The unique versatility of the perovskite crystal structure stems from its unusual ability to accommodate a staggering variety of elemental combinations. This unparalleled diversity raises the questions on how many new perovskites are yet to be discovered, and which ones will exhibit improved or novel functionalities. In an attempt to answer these questions, we here begin by mapping the entire compositional landscape of these crystals.

Figure 1a shows the structure of a cubic ABX_3 perovskite. In this structure the A and B elements are cations and X is an anion. B-site cations are six-fold coordinated by anions to form BX_6 octahedra. The octahedra are arranged in a three-dimensional corner-sharing network, and each cavity of this network is occupied by one A-site cation¹⁵. All perovskites share the same network topology, but can differ in the degree of tilting and distortions of the octahedra^{15–19}. The quaternary counterpart of the perovskite crystal is the double perovskite $A_2BB'X_6$. When the B and B' cations alternate in a rock-salt sublattice, the crystal is called elpasolite¹⁴ (Figure S1a). In the following we use the term ‘perovskite’ to indicate both ternary and quaternary compounds.

RESULTS

How many perovskites do currently exist? If we search for the keyword ‘perovskite’ in the inorganic crystal structure database (ICSD) we find 8866 entries², but after remov-

ing duplicates the headcount decreases to 335 distinct ABX_3 compounds. Similarly, the keyword ‘elpasolite’ yields 224 distinct $A_2BB'X_6$ compounds. By including also the extensive compilations of Refs. 3–9, we obtain a grand total of 1622 distinct crystals that are reliably identified as perovskites (database S1.1).

How many perovskites are left to discover? Direct inspection of the elemental composition in database S1.1 indicates that, with the exception of hydrogen, boron, carbon, phosphorous and some radioactive elements, these crystals can host every atom in the Periodic Table. Therefore an upper bound for the number of possible perovskites is given by all the combinations of three cations and one anion. By considering only ions with known ionic radii¹³ we count 3,658,527 hypothetical compounds. Our goal is to establish which of these compounds can form perovskite crystals. Ideally we should like to resort to *ab initio* computational screening²⁹, but these techniques are not yet scalable to millions of compounds. For example, by making the optimistic assumption that calculating a phase diagram required only one hour of supercomputing time per compound, it would take 160 years to complete this task.

An empirical approach to investigate the formability of ABX_3 perovskites was proposed by Goldschmidt almost a century ago¹¹. In this approach the perovskite structure is described as a collection of rigid spheres, with sizes given by the ionic radii r_A , r_B and r_X . These radii can be combined into two dimensionless descriptors, the tolerance factor $t = (r_A + r_X)/\sqrt{2}(r_B + r_X)$, and the octahedral factor $\mu = r_B/r_X$. Goldschmidt postulated that perovskites arrange so that ‘the number of anions surrounding a cation tends to be as large as possible, subject to the condition that all anions touch the cation’¹⁵. This statement constitutes the ‘no-rattling’ principle, and limits the range of values that t and μ can take for a perovskite.

Goldschmidt’s principle was recently tested on larger datasets than those available in 1926. By analysing a few hundred ternary oxides and halides, it was found that in a two-dimensional t vs. μ map perovskites and non-perovskites tend to cluster in distinct regions^{3,4,31}. Based on this observation, much work has been dedicated to identifying a ‘stability range’, either by postulating boundaries for t and μ ^{3,4,19},

or by using machine learning to draw t vs. μ curves enclosing the data points³¹. The merit of these efforts is that they addressed the predictive power of the no-rattling principle in qualitative terms. However, these approaches suffer from relying too heavily on empirical ionic radii. It is well known that the definition of ionic radii is non-unique, and that even within the same definition there are variations reflecting the coordination and local chemistry^{10,11,13,15}. As these uncertainties transfer to the octahedral and tolerance factors, the stability ranges proposed so far are descriptive rather than predictive. In order to overcome these limitations, instead of defining a map starting from empirical data, our strategy is to construct a stability range from first principles, by relying uniquely on Goldschmidt’s hypothesis. This choice allows us to also derive a structure map for quaternary compounds, a step that has thus far remained elusive.

We describe our strategy starting from ternary perovskites, and then we generalize our findings to quaternary perovskites. Figure 1b shows that, for the A cation to fit in the cavity, the radii must satisfy the condition $r_A + r_X \leq \sqrt{2}(r_B + r_X)$, or equivalently $t \leq 1$ (see Supplementary Information). Similarly, Figure 1c shows that the octahedral coordination of the B cation by six X anions is not possible when $\sqrt{2}(r_B + r_X) < 2r_X$, therefore $\mu \geq \sqrt{2} - 1$. We refer to these conditions as the ‘octahedral’ limit and the ‘stretch’ limit, respectively, as shown in Figure 1e. When these conditions are not fulfilled, the lattice tends to distort towards a layered geometry, with edge-sharing or face-sharing octahedra, or lower B-site coordination^{15,19}. These two bounds are well-known^{11,15} but are insufficient for quantitative structure prediction.

When t is smaller than 1, the corner-sharing octahedra exhibit an increased degree of tilting, and the A-site cation is displaced from the central position in the cuboctahedral cavity, as shown Figures 1d and S2^{17,18}. Previous work has shown that the displacement of the A-site cation can be determined by optimizing Coulomb interactions between the perovskite ions, taking into account the bond-valence configuration of each ion¹⁹. Here we explore this scenario using a purely geometric approach, by identifying the ionic positions which achieve the tightest packing of ions in a tilted perovskite configuration.

Figure 1d shows that the octahedra can tilt only until two anions belonging to adjacent octahedra come into contact. In this extremal configuration, to satisfy the no-rattling principle the A-site cation must exceed a critical size. Using the geometric construction described in the Supplementary Information (Figure S2), this condition translates into a lower limit for the tolerance factor, $t \geq \rho_\mu / \sqrt{2}(\mu + 1)$, where ρ_μ is a simple piece-wise linear function of μ (Figure S3) (see Supplementary Information). We refer to this condition as the ‘tilt’ limit (Figure 1e). When this criterion is not met, the perovskite network tends to collapse into structures with edge-sharing or face-sharing octahedra. Along the same lines we must consider the limit of two neighboring A-site cations coming into contact (Figure 1e and Figure S4a), and of the A and B cations touching (Figure S4b). Besides these geometric constraints, we also ought to take into account that the rigid spheres of the model represent chemical elements, therefore we have additional bounds on the size of the ionic radii: the largest toler-

ance factor corresponds to the combination of Cs and F, and the oxidation number of A cannot exceed that of B according to Pauling’s valency rule¹⁵. By considering all the (μ, t) points that satisfy these conditions simultaneously, we obtain the perovskite stability area shown in blue in Figure 1e.

These considerations are readily generalized to double perovskites. In this case we have two different cations B and B’ (Figure S1a), therefore we must consider two octahedral parameters: the average octahedral factor, $\bar{\mu} = (r_B + r_{B'})/2r_X$, and the octahedral mismatch $\Delta\mu = |r_B - r_{B'}|/2r_X$. In the Supplementary Information we derive the generalized tolerance factor, which takes the form $t = (r_A/r_X + 1)/[2(\bar{\mu} + 1)^2 + \Delta\mu^2]^{1/2}$. As we now have three structure descriptors, the generalized stability region is a closed volume in the $(\bar{\mu}, t, \Delta\mu)$ space, as seen in Figure 2a. If we slice this volume through the plane $\Delta\mu = 0$ we revert to the perovskite area of Figure 1d. We emphasize that our present results derive exclusively from Goldschmidt’s principle (barring the chemical limits which are not essential), and make no reference to the definition and values of the ionic radii. The bounds of the perovskite regions are easy to evaluate for any structure, and are described by six inequalities for $\bar{\mu}$, t and $\Delta\mu$ in Table S1.

Can the inequalities in Table S1 be used for structure prediction? To answer this question we analyzed a record of 2291 ternary and quaternary compounds that we collected from the ICSD and from Refs. 3–9. This dataset includes 1622 perovskites (database S1.1), 592 non-perovskites (database S1.2), and 77 compounds which can crystallize either as a perovskite or another structure (database S1.3; see Supplementary Information). Figure 2a shows the distribution of all these compounds in the $(\bar{\mu}, t, \Delta\mu)$ space. We see that our perovskite volume (blue) delimits remarkably well the regions occupied by perovskites (blue markers) and non-perovskite (red markers). A detailed view of these data is provided in Figure 2b–e, where we show slices of the perovskite volumes at fixed octahedral mismatch, and in Figure S5, where data for perovskites and non-perovskites are presented separately. In these panels we see that, as $\Delta\mu$ increases, the stability region decreases in size and moves towards higher octahedral factors. Remarkably, most datapoints from database S1.1 closely follow this trend. Case-by-case inspection reveals several outliers, that is perovskite markers falling outside of the perovskite region or vice-versa. The existence of outliers is to be expected given the simplicity of Goldschmidt’s model, but intriguingly we find many cases where the presence of outliers signals the occurrence of polymorphism. An important example is BaTiO₃: while this compound is mostly known as a ferroelectric perovskite, it is also stable in a hexagonal structure under the same pressure and temperature conditions³⁴.

We now assess the predictive power of the model on quantitative grounds. The simplest way to proceed would be to classify compounds based on whether the corresponding $(\bar{\mu}, t, \Delta\mu)$ point falls inside or outside the stability region. However, this procedure is unreliable as it is very sensitive to small variations in the ionic radii. A better strategy is to replace each point by a rectangular cuboid, with dimensions representing the uncertainty in the ionic radii. The uncertainty calculation is detailed in the Supplementary Informa-

tion. With this choice we define the ‘formability’ as the fraction of the cuboid volume falling within the perovskite region, and we classify the compound as a perovskite if this fraction exceeds a critical value (Figure S6a). In order to quantify the accuracy of this classification procedure and the associated uncertainty, we determine the classification of large subsets of compounds, randomly selected from databases S1.1 and S1.2, and we repeat this operation several thousand times. By the central limit theorem, the average success rates tend to a normal distribution (Figure S7b); the center of this distribution gives the most probable success rate, and the standard deviation yields the statistical uncertainty (Figure S6c).

Our main result is that, for sample sizes of 100 compounds or more, the geometric model correctly classifies $79.7 \pm 4.0\%$ of all compounds with a 95% confidence level. This predictive power is unprecedented among structure prediction algorithms.

For completeness we also assess how our model compares with previous models. To this end, we calculate how many of the known compounds in Databases S1.1 and S1.2 are correctly classified within the original model of Goldschmidt (which considers only the stretching and octahedral limits), within three other empirical models reported in Refs. 3,19,20, and within our present model (see Figure S7). Figure S7e shows that the stretching and octahedral limits correctly categorize nearly all perovskites in Database S1.1, but fail to discriminate against more than half of non-perovskites in Database S1.2. The empirical regions in Figure S7 b-d clearly demonstrate that by setting a lower bound to the tolerance factor, the accuracy of the model improves significantly for both perovskites and non-perovskites. However, up to now, this bound has been described via empirical data fitting. Our model predicts the tilt limit from first principles, all the while retaining a very good accuracy in distinguishing perovskites from non-perovskites. By comparison, the perovskite regions reported in Refs. 3,19,20 can be understood as zeroth- and first-order approximations to our bounds, respectively.

By applying our classification algorithm to all possible 3,658,527 quaternary combinations, we generated a library of 94,232 new perovskites and double perovskites that are expected to form with a probability of 80% (Figure S8). The complete library of predicted perovskites is provided as database S2. Our library of future perovskites dwarves the set of all perovskites currently known, and is comparable in size to the ICSD database of all known inorganic crystals, which contains approximately 188,000 structures².

How many of our predicted perovskites are genuinely new compounds, i.e. have never been synthesized? In order to answer this question we performed a large-scale web data extraction operation by querying an internet search engine about each and every one of the nearly one hundred thousand compounds in database S2 (see Supplementary Information). This procedure revealed that the overwhelming majority of these compounds have never been reported or mentioned before (database S2.1), and that fewer than 1% of the structures were already known, namely 786 out of 94,232 compounds (database S2.2).

The 786 previously-known compounds reported in

database S2.2 were not included in our initial database S1 of known crystals. We use this additional set of known compounds to perform a second blind test of our predictions. According to our inferential analysis we expect 626 compounds of database S2.2 (79.7% of 786) to be perovskites. By carrying out a manual literature search we confirmed that 555 crystals are indeed perovskites (database S2.2.1). This result is remarkably consistent with our prediction. This blind test replaces a validation based on resource-intensive experimental synthesis of hundreds of new compounds with faster and inexpensive data analytics. The success of the blind test clearly demonstrates that, in spite of its simplicity, Goldschmidt’s principle has a considerable predictive power. Naturally, by combining our structure map with experiments and *ab initio* calculations on selected sub-families, the predictive accuracy of the model is bound to improve even further.

What is the topography of our geometric structure map? Figure 3a and Figure S9b show that the majority of predicted perovskites tend to cluster towards the region with the lowest octahedral, tolerance, and mismatch factors. This high density of compounds stems from the occurrence of a large number of lanthanide oxide and actinide oxide perovskites, which tend to have similar geometric descriptors due to the lanthanide contraction³⁶. We also note that the concentration of compounds near the bottom of the map shows that the geometric tilt limit derived in this work (Figure 1d) is essential to accurately predict the formability of perovskites.

Figure 3b shows the relative abundances of predicted perovskites. The majority of compounds are oxides (68%), followed by halides (16%), chalcogenides (12%), and nitrides (4%). Why is the perovskite landscape dominated by oxides, and why nitrides are so rare instead? To answer this question we observe that the -2 oxidation state of O admits as many as ten inequivalent charge-neutral combinations of the oxidation states of the cations (Table S2). Furthermore, the oxygen anion has a small radius (1.3 Å), which is compatible with most transition metals, lanthanides, and actinides; and these elements form the most numerous groups in the Periodic Table. A similar argument could be made for chalcogens, which share the same oxidation state as oxygen. However, the chalcogen radii are too large (1.8-2.2 Å) to accommodate most transition metals and actinides, hence chalcogenide perovskites constitute a much smaller family. Our finding is consistent with recent *ab initio* calculations³⁸. Halide perovskites are even less numerous than chalcogenides, mostly owing to their more restrictive -1 oxidation number: in fact this oxidation state only admits +1 A-site cations and +1/+3 or +2/+2 B-site cations (Table S2). Nitrides constitute an interesting exception to these trends. Indeed, while the ionic radius of N in the -3 oxidation state (1.5 Å) is very similar to that of O and its oxidation number admits as many as seven inequivalent combinations of oxidation states for the cations, in all such combinations at least one B-site cation must have the unusually high +5 oxidation state (Table S2). As the ionic radii tend to decrease with the oxidation number¹⁵, most B-site cations turn out to be too small to be coordinated by six nitrogen anions in an octahedral environment. As a result, if we exclude radioactive elements,

we find fewer than eighty nitride perovskites across the entire Periodic Table. We note that, owing to the lower electronegativity of nitrogen, the ionic character of the chemical bonds in nitride perovskites is reduced, and our geometric model is reaching its limits of applicability¹⁹. However, the scarcity of nitride perovskites predicted by our model is fully consistent with recent *ab initio* calculations³⁹ and with experimental observations (only one ternary nitride perovskite can be found in ICSD), indicating that the rigid sphere approximation can still provide meaningful predictions for nitrides. Among our predicted compounds we also identified many unexpected binary compounds of the type A_2X_3 . One such example is iron oxide, Fe_2O_3 . While this oxide is primarily known in the form of hematite (corundum structure), it was recently found that the crystal undergoes a phase transition to a perovskite at high pressure and temperature¹⁸. The stabilization of Fe_2O_3 as a perovskite under high pressure is in agreement with our *ab initio* calculations (see discussion and Figure S10 of the Supporting Information) and can be associated with the well known phase transition of ilmenites (ternary ordered corundum) into perovskites, observed, for example, for $FeTiO_3$ ²⁰. This finding suggests that several other binary compounds may hide a perovskite phase in their phase diagram, an intriguing possibility that is open to investigation.

To demonstrate the applicability of our model, we take the example of ternary oxide perovskites. Figure 4 shows a comparison between the combinatorial screening of ternary oxides performed using our model, *ab initio* calculations reported in Ref. 37 and experimental data collected from Databases S1.1, S1.3 and S2.2.1. The compositions classified as perovskites by our model include 92% of the experimentally observed oxide perovskites. In particular, when A is a lanthanide and B is a first row transition metal, our model predicts that most compositions can form as a perovskite, in excellent agreement with DFT predictions and experiment. This can be explained by the similar ionic sizes of transition metals and rare earths respectively. The same prediction is made for the case when both A and B are rare earths, however fewer perovskites are found from DFT and experiment. The reason for this discrepancy is that the non-rattling principle effectively probes the dynamical stability of a given chemical composition in the perovskite structure. However, it does not contain information on its stability against decomposition (thermodynamic stability). Of the two criteria, the thermodynamic stability requirement is more stringent, and this explains why generally geometric blueprints tend to predict more perovskites than have actually been made or that are predicted from *ab initio* calculations. Therefore, further theoretical and experimental studies are required to ascertain whether these proposed compositions are also thermodynamically stable. Despite this limitation, Figure 4 demonstrates that the Goldschmidt principle can be used as an efficient and reliable pre-screening tool for the high-throughput combinatorial design of perovskites. In fact, in Figure 4 we show that our model can reduce the number of calculations by more than 70% in the combinatorial screening of ternary oxides. Importantly, the Goldschmidt no-rattling principle becomes increasingly useful in the context of screening all possible perovskites beyond oxides, reducing the total

number of 3.6 million possible compositions by 97%, to fewer than one hundred thousand candidates. Therefore, by leveraging the complementary strengths of Goldschmidt’s empirical no-rattling principle and *ab initio* computational modelling it will be possible to explore the complete chemical landscape of all possible perovskites.

CONCLUSION

In conclusion, we charted the complete landscape of all existing and future perovskites. By combining inferential statistics with large-scale web data extraction, we validated Goldschmidt’s no-rattling principle on quantitative grounds, and developed a structure map to predict the stability of perovskites with a fidelity of 80%. Our model completes the general theory that Goldschmidt proposed almost a century ago, and formalizes the non-rattling hypothesis into a mathematically rigorous set of criteria that can be used in the design and discovery of new perovskites. As an outcome of our study, we were able to generate a library of almost one hundred thousand hitherto-unknown perovskites awaiting discovery (database S2.1). By releasing this library in full, we hope that this work will stimulate much future experimental and computational research on these fascinating crystals. More generally, our findings suggest that geometric blueprints could serve as a powerful tool to help tackle the exponential complexity of combinatorial materials design.

Methods. A full description of the methods, data provenance and statistical analysis used in this manuscript can be found in the Supporting Information.

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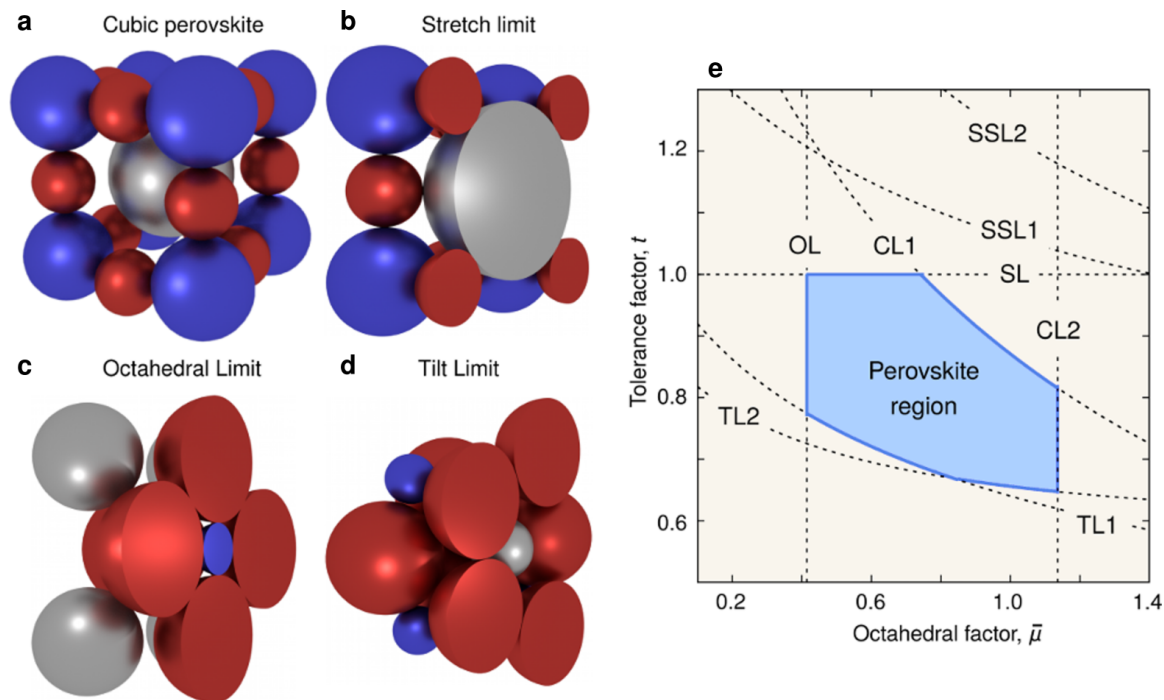


Figure 1. **No-rattling principle for ternary perovskites.** **a**, Rigid-sphere representation of the conventional cell of a cubic ABX_3 perovskite, with A in grey, B in blue, and X in red. **b**, Cross-sectional view of the stretch limit. In this configuration the A cation sits at the center of the cavity and touches twelve nearest-neighbor X anions. **c**, Cross-section of the perovskite structure in the octahedral limit. Here the nearest-neighbor X anions belonging to the same octahedron touch. **d**, Schematic representation of the tilt limit. This configuration can be thought of as obtained from **b** by reducing the size of the A cation (grey), increasing the size of the X anion (red), and tilting the octahedra so that anions of adjacent octahedra touch. In this case the A cation moves away from the center of the cavity so as to optimally fill the available space, and touches five X anions. **e**, Stability area of ternary perovskites (blue), as derived from the no-rattling principle and from additional chemical limits (dashed lines): stretch limit (SL), octahedral limit (OL), tilt limits (TL1 and TL2), chemical limits (CL1 and CL2), secondary stretch limits (SSL1 and SSL2). These boundaries are derived in the Supplementary Information.

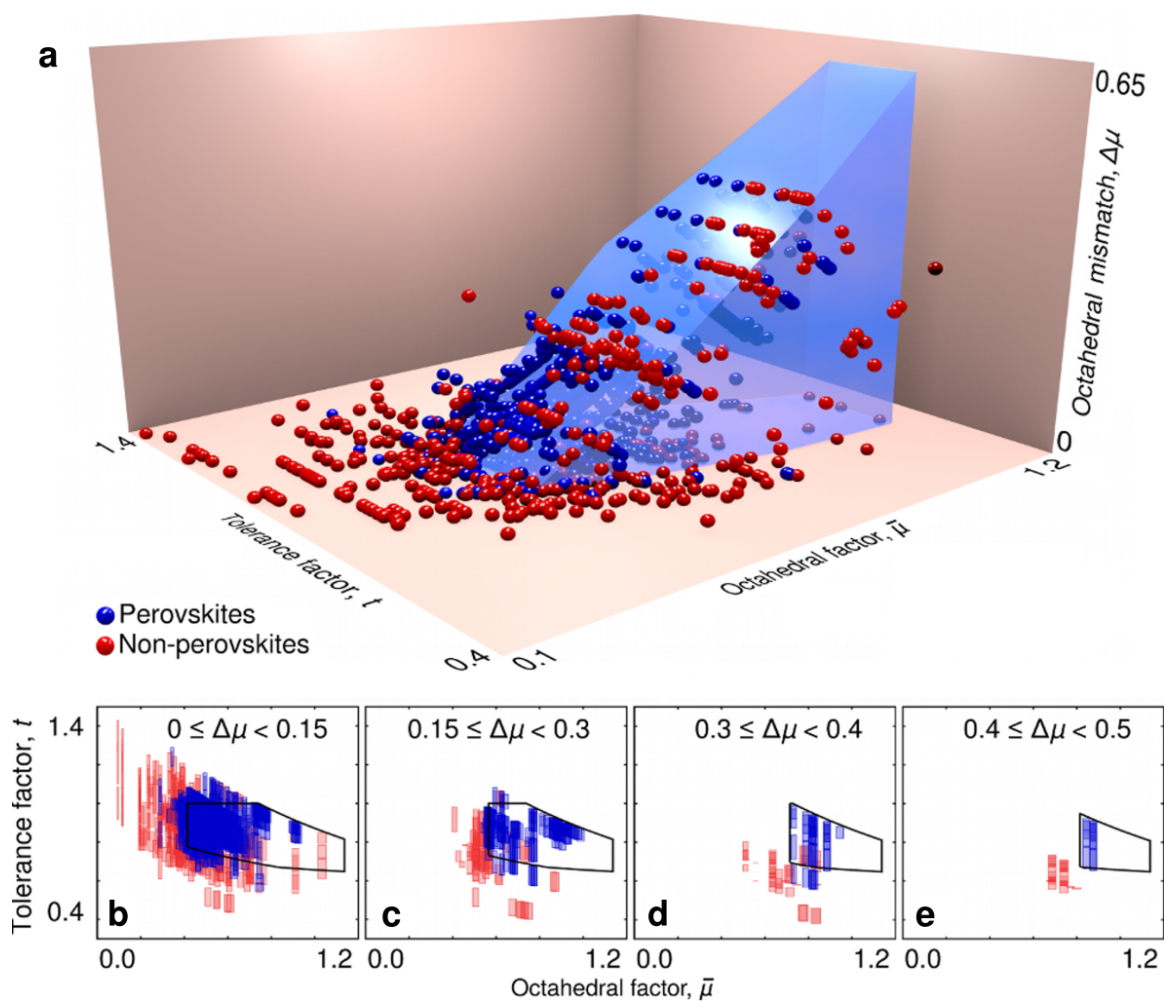


Figure 2. **Stability range of ternary and quaternary perovskites.** **a**, The blue volume represents the stability range of perovskites in the $(\bar{\mu}, t, \Delta\mu)$ space, as derived from Goldschmidt's no-rattling principle. The blue and red markers correspond to perovskites (databases S1.1 and S1.3) and non-perovskites (database S1.2), respectively, calculated for all the compounds in database S1. The location of each marker is the center of the rectangular cuboid defined in the Supplementary Information. **b** to **e**, Slices of the stability volume shown in **a**, reporting all perovskites with octahedral mismatch $\Delta\mu$ in the range indicated at the top of each panel. In these two-dimensional representations the cuboids appear as rectangles, with blue and red indicating perovskites and non-perovskites, respectively.

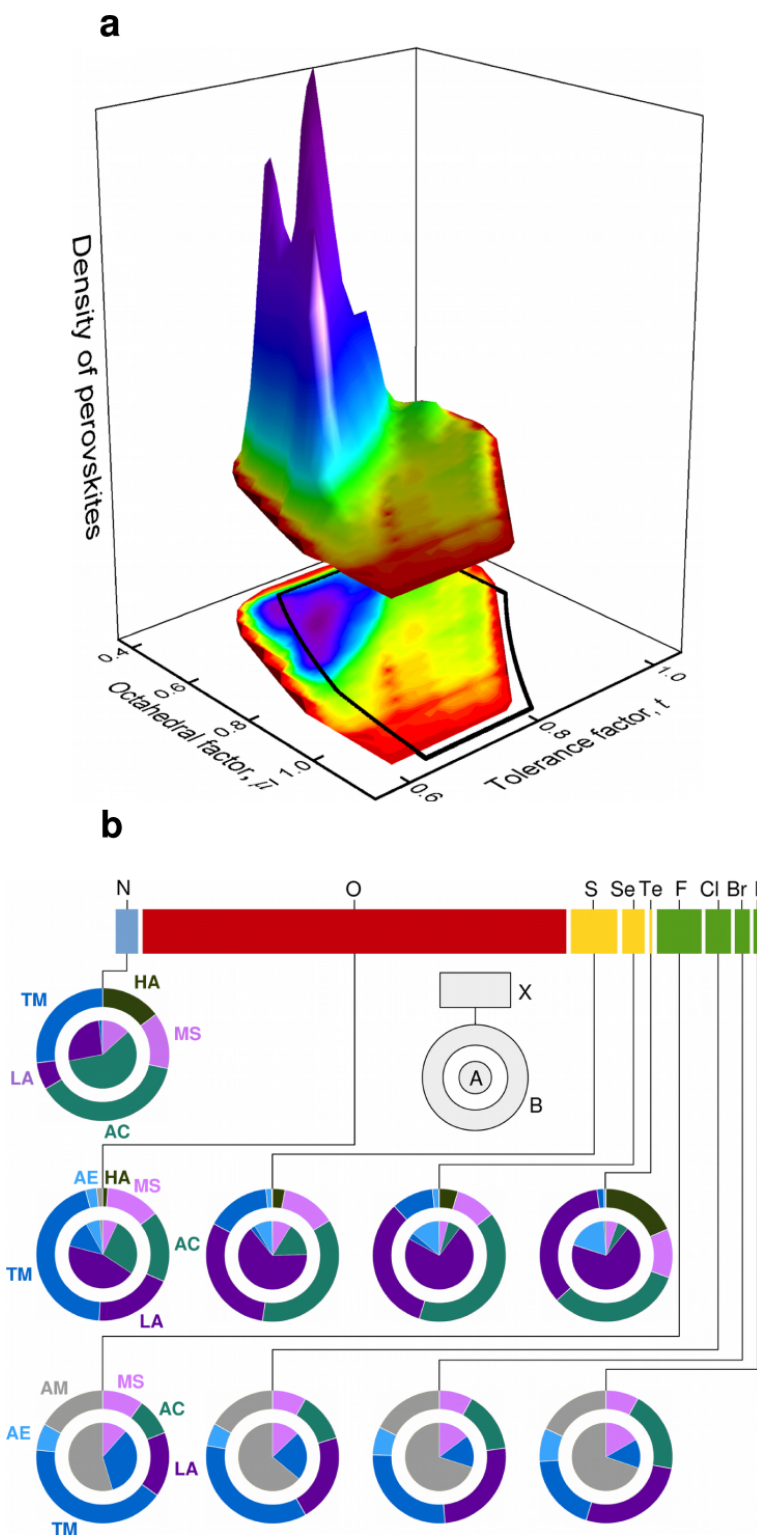


Figure 3. **Topography of the perovskite landscape.** **a**, Density of predicted ternary and quaternary perovskites from databases S1.1, S1.3 and S2. For clarity the density of perovskites has been integrated over all possible values of the octahedral mismatch $\Delta\mu$, so as to obtain a two-dimensional map in the $(\bar{\mu}, t)$ plane. The two plots show the same quantity as a 2D colormap and a 3D surface, respectively. **b**, Crystallographic site preference in databases S1.1, S1.3 and S2. The horizontal bar illustrates the relative abundance of perovskites with a given anion X. For each anion, the rings illustrate the relative abundances of the A-site cation (inner ring) and of the B-site cation (outer ring). Cations are grouped using the standard classification: alkali metals (AM), alkaline earth metals (AE), transition metals (TM), lanthanides (LA), actinides (AC), metals and semiconductors (MS: Al, Ga, In, Sn, Tl, Pb, Bi, B, Si, Ge, As, Sb, Te, Po), and halogens (HA). Database S2 contains 59 binary, 2834 ternary and 90,606 quaternary perovskites.

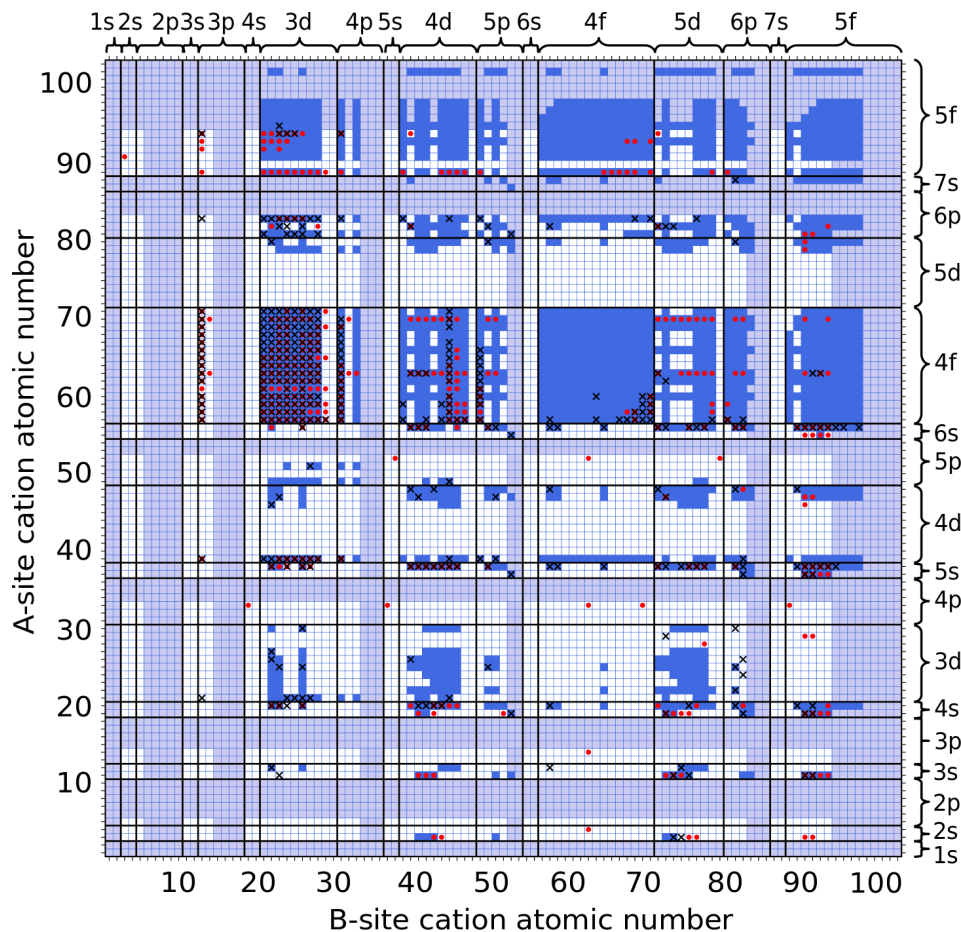


Figure 4. **Combinatorial screening of ternary oxide perovskites.** ABO_3 ternary compounds classified as perovskites by our geometric model (blue squares), *ab initio* calculations within the generalized gradient approximation to density functional theory, DFT/GGA (red disks) reported by Emery et al³⁷ and experimental data from Databases S1.1, S1.3 and S2.2.1 (black crosses). The gray shading highlights the region of the A/B map that is not taken into account in the *ab initio* screening. The curly bracket marks the empty valence shells corresponding to an interval of atomic numbers for A and B, e.g. the 3*d* interval corresponds to $Z = 21-30$ (Sn-Zn). Emery et al reported the study of the thermodynamic stability of 5329 candidate ternary compounds, and found that 382 compositions are stable perovskites. There are 383 ternary oxides in Databases S1.1, S1.3 and S2.2.1, DFT/GGA calculations predict 225 (59%) of them to be stable while our model correctly classifies 354 of them (92%).

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SUPPLEMENTARY INFORMATION

Databases

Data analyzed in this manuscript is reported in two main databases (available on our group repository¹), which are structured as follows:

$$\text{Database S1} = \begin{cases} \text{Database S1.1} & \text{Known perovskites.} \\ \text{Database S1.2} & \text{Known non-perovskites.} \\ \text{Database S1.3} & \text{Known compounds that can be both perovskites and non-perovskites.} \end{cases}$$

$$\text{Database S2} = \begin{cases} \text{Database S2.1} & \text{Predicted compounds that have never been made or mentioned before.} \\ \text{Database S2.2} & \begin{cases} \text{Database S2.2.1} & \text{Predicted compounds that were found on the internet and are perovskites.} \\ \text{Database S2.2.2} & \text{Predicted compounds that were found on the internet and are not perovskites.} \end{cases} \end{cases}$$

Data provenance

The database S1 of known compounds is generated by collecting structures from the Inorganic Crystal Structure Database (ICSD)² and from Refs. 3–9. From Ref. 6 we consider only quaternary compounds. The calculation of $(\bar{\mu}, t, \Delta\mu)$ is performed by using the oxidation states reported in these references. For ternary and quaternary perovskites we search the ICSD by structure type, using the keywords ‘perovskite’, ‘elpasolite’, or ‘LiNbO₃’. Out of the compounds returned by these searches we retain only pure crystals, i.e. compounds with integer site occupation numbers and integer oxidation states. We discard theoretically-predicted compounds that have not been synthesized.

In order to find ternary non-perovskites in the ICSD, we search both by structure type and by ANX formula, following the classification of Refs. 8,10. We consider the following structure types: spinel, ilmenite, aragonite, calcite, olivine, pyroxene, BaNiO₃, BaRuO₃, BaMnO₃; and the following ANX formulae: A₂BX₄, A₃BX₅, AB₂X₅, AB₃X₇, A₂B₃X₈, AB₄X₉ and AB₃X₈. We retain the compounds which fulfil the charge neutrality rule and Pauling’s valency rule for an ABX₃ stoichiometry.

In order to collect quaternary non-perovskites from the ICSD, we first search for quaternary non-perovskites which have the same general formula A₂BB’X₆ as for double perovskites. For the other possible stoichiometric compositions we could not find an established structural classification as for ternary non-perovskites. To overcome this lack of information, we considered crystals with formulas A_mB_nB’_pX_q which fulfil the charge neutrality and valency conditions, 2a + b + b’ - 6x = 0 and a ≤ (b + b’)/2, where a, b, b’, and x are the oxidation numbers of A, B, B’, and X, respectively. This linear system admits infinite solutions, therefore we limit the search to the cases q = 5, 6, 7, so as to span

the neighborhood of q = 6 in A₂BB’X₆ compounds. This choice corresponds to searching the ICSD using the following ANX formulae: ABB’X₅, AB₂B’X₆, AB₂B’X₇, AB₃B’X₆, AB₃B’X₇, AB₄B’X₆, AB₅B’X₇, AB₆B’X₇. A₃BB’X₇, A₂B₃B’X₇, and A₄BB’X₇.

After collecting all data we remove duplicates. In addition, we cross-check the lists of perovskites and non-perovskite to identify those compositions for which both perovskite and non-perovskite compounds are found. We find that 77 compounds (database S1.3) crystallize either as a perovskite or a non-perovskite.

The compounds obtained from these searches form the set of previously-known compounds, and are reported in database S1, together with the source references and the reported oxidation states. The database contains 77 compounds that can be either perovskites or another structure (database S1.3), 345 ABX₃ perovskites and 1277 A₂BB’X₆ double perovskites (database S1.1), and 592 compounds that are neither a perovskite nor a double perovskite (database S1.2).

Geometric bounds resulting from the no-rattling principle: ABX₃ perovskites

As in the main text we indicate by r_A, r_B, and r_X the ionic radii of A, B, and X, respectively. From these values we obtain the tolerance factor as¹¹:

$$t = (r_A + r_X) / \sqrt{2}(r_B + r_X), \quad (1)$$

and the octahedral factor as:

$$\mu = r_B / r_X. \quad (2)$$

Following Goldschmidt’s principle we consider that in ABX₃ perovskites the B cation and the X anion are always in contact. Instead the A cation may or may not touch other anions

or cations. The geometric bounds defining the stability region correspond to extremal packing configurations whereby A-A*, A-B, A-X, or X-X* touch, having indicated by an asterisc a nearest-neighbor ion of the same type. In the following we consider the Platonic model of perovskites¹², whereby the BX₆ octahedra are ideal and corner-sharing. We discuss the geometric limits using the naming convention of Figure 1: stretch limit (SL), octahedral limit (OL), tilt limits (TL1 and TL2), chemical limits (CL1 and CL2), and secondary stretch limits (SSL1 and SSL2).

Stretch limit (SL)

The stretch limit corresponds to the situation where the A cation is so large that it touches all twelve X anions defining the octahedral cavity; this can only happen in a cubic perovskite, as shown in Figure 1b. In this configuration the distances between the centers of A and X and between the centers of B and X are related by $\sqrt{2}(r_B + r_X) = r_A + r_X$. After dividing by the left-hand side we obtain $t = 1$, irrespective of μ . In Figure 1e this boundary is labelled ‘SL’.

Octahedral limit (OL)

The octahedral limit corresponds to the situation where two X anions belonging to the same BX₆ octahedron are in contact, as shown in Figure 1c. In this configuration the distance between the centers of the anions satisfies the condition $\sqrt{2}(r_B + r_X) = 2r_X$. After dividing both sides by r_X we find $\mu = \sqrt{2} - 1$, irrespective of t . This bound is labelled ‘OL’ in Figure 1e.

Tilt limits (TL1 and TL2)

In ABX₃ perovskites, adjacent BX₆ octahedra can tilt until nearest-neighbor anions from each octahedron make contact. In this configuration the octahedral cavity is distorted from a regular cuboctahedron; to satisfy Goldschmidt’s no-rattling principle the A cation must move off-center, as shown in Figure 1e.

In order to investigate this extremal configuration we employ the Platonic model of perovskites developed in Ref. 12. This model considers a orthorhombic perovskite unit cell consisting of four BX₆ octahedra, belonging to the *Pnam* space group; all atomic coordinates are uniquely defined by specifying the B-X bond length, $r_B + r_X$, and two angles, the tilt angle θ of an octahedron with respect to the z axis, and the precession ϕ of this octahedron around the same axis. The coordinate system is shown in Figure S2. The Cartesian coordinates of the anions indicated in Figure S2 can be expressed as follows:

$$X_1 : a(-x_e + 1/2); b(y_e + 1/2); -c(z_e), (3)$$

$$X_2 : a(-x_e + 1/2); b(y_e + 1/2); c(z_e + 1/2), (4)$$

$$X_3 : a(x_e - 1/2); b(-y_e + 1/2); c(-z_e + 1/2), (5)$$

$$X_4 : -a(x_e); b(-y_e + 1); -c(z_e), (6)$$

$$X_5 : -a(x_e); b(-y_e + 1); c(z_e + 1/2), (7)$$

$$X_6 : a(x_a); b(y_a); c(z_a), (8)$$

$$X_7 : a(x_a - 1/2); b(-y_a + 1/2); c(z_a). (9)$$

The fractional coordinates and the lattice parameters appearing in these expressions are given by:

$$x_e = \frac{1}{4}(1 + \cos \theta \tan \phi); \quad y_e = \frac{1}{4}(1 - \tan \phi / \cos \theta); \quad z_e = -\frac{\sqrt{2}}{8} \tan \theta, \quad (10)$$

$$x_a = \frac{\sqrt{2}}{4} \tan \phi \sin \theta; \quad y_a = \frac{\sqrt{2}}{4} \tan \theta; \quad z_a = \frac{1}{4}, \quad (11)$$

$$a = 2\sqrt{2}(r_B + r_X) \cos \phi; \quad b = 2\sqrt{2}(r_B + r_X) \cos \theta \cos \phi; \quad c = 4(r_B + r_X) \cos \theta, \quad (12)$$

where we have obtained x_e, y_e, z_e, x_a, y_a and z_a following the procedure described in Ref. 12, but using the transpose of the rotation matrix. Upon tilting the octahedra, the distance between the anion pairs X₁-X₂, X₂-X₃, X₄-X₅, and X₂-X₆ in Figure S2 decreases. We can ignore the pair X₄-X₅ which is always found at the same distance as X₁-X₂. In this case there are two mutually-exclusive extremal configurations: (1) X₁ and X₂ as well as X₂ and X₃ are in contact; (2) X₁ and X₂ as well as X₂ and X₆ are in contact. In both cases the anions X₁ and X₂ touch, so we examine this condition first. By requiring that the distance between the centers of X₁ and X₂ is $2r_X$ we can determine the tilt angle θ in either extremal configuration:

$$\cos \theta = (2p + \sqrt{3 - 2p^2})/3, \quad \text{with } p = 1/(\mu + 1). \quad (13)$$

By further requiring that the distance between X₂ and X₃ or that the distance between X₂ and X₆ is equal to $2r_X$, we obtain

the two extremal precession angles ϕ_1 and ϕ_2 :

$$\sin 2\phi_1 = (1 - p^2)/\cos \theta, \quad (14)$$

$$\tan \phi_2 = (\cos \theta - s)/(1 - 2p^2), \quad (15)$$

having defined:

$$s^2 = \cos^2 \theta - (1 - 2p^2)(2 \cos^2 \theta - \sqrt{2} \sin 2\theta + 1 - 2p^2). \quad (16)$$

These conditions allow us to determine, for each octahedral factor μ , the extremal configurations corresponding to the maximum possible octahedral tilt. In order to obtain a boundary for the stability region, we need to find the limit of the tolerance factor t in these configurations. To this aim we determine the size of the largest A cation which fits in the cavity and touches as many X anions as possible.

By symmetry the center of this A cation along the c axis must be $z_A = 1/4$. To find the other coordinates x_A and y_A , we require that they allow us to fit the largest possible

A cation in the cavity. This statement corresponds to asking that (x_A, y_A, z_A) be the center of a sphere passing through

$$2a^2x_A(1 - x_a - x_e) + 2b^2y_A(y_a + y_e) = a^2(x_e^2 - x_a^2 + x_a - x_e) + \quad (17)$$

$$+ b^2(y_e^2 - y_a^2 + y_a + y_e) + (1/4 + z_e)^2c^2, \quad (18)$$

$$a^2x_A + b^2y_A(4y_e - 1) = a^2(1/4 - x_e) + b^2(3y_e - 3/4).$$

Furthermore the radius of the sphere is given by:

$$r_\mu = [(x_A - x_a + 1/2)^2a^2 + (y_A + y_a - 1/2)^2b^2]^{1/2}. \quad (19)$$

In order to fulfil the no-rattling principle, the A cation must touch the nearest neighbor X anions. This condition implies that the radius of the sphere must equal the distance between A and X, $r_\mu = r_A + r_X$. As a result we find that the boundary for the tolerance factor is

$$t = r_\mu / \sqrt{2}(r_B + r_X). \quad (20)$$

To proceed we evaluate this boundary numerically as follows:

$$\rho_\mu = 1.366 + 0.442\bar{\mu} \text{ for } \bar{\mu} < 0.8,$$

Using these linear fits inside Eq. (20) yields the tilt limits TL1 and TL2 shown in Figure 1e.

Secondary stretch limits (SSL1 and SSL2)

An additional geometric limit is obtained by considering the extremal configuration wherein two adjacent A-site cations are in contact. The configuration with the largest possible such cations is cubic, as shown in Figure S4a. In this configuration the distance between the centers of A and A*, $2(r_B + r_X)$, equals the diameter of each cation, $2r_A$. From this equality, by using the relations between μ , t and r_A/r_X , r_B/r_X we obtain the boundary $t = (\mu + 2)/\sqrt{2}(\mu + 1)$. This limit is indicated as SSL1 in Figure 1e. Similarly it is straightforward to verify that in the extremal configuration wherein the A and B cations are in contact one has $t = \sqrt{3}/\sqrt{2} - (\mu - 1)/\sqrt{2}(\mu + 1)$ (Figure S4b). This boundary is labelled SSL2 in Figure 1e. Since the boundaries SSL1 and SSL2 lie above the boundary SL in Figure 1e, the upper limit of the tolerance factor t is set by the stretch limit.

Chemical limits (CL1 and CL2)

The largest cation and the smallest anion in the Periodic Table, Cs and F respectively, set an upper bound for the ratio r_A/r_X . Since by definition we have $r_A/r_X = \sqrt{2}t(\mu + 1) - 1$, there will not be any compounds in the perovskite map above the line $t = (r_{Cs}/r_F + 1)/\sqrt{2}(\mu + 1)$. This boundary is indicated as CL1 in Figure 1e. A similar reasoning applies to the octahedral ratio $\bar{\mu} = (r_B + r_{B'})/2r_X$. Among all perovskites and double perovskites in database S1 this ratio is largest for the combination of Fr^+ and Ac^{3+} at the B and

the centers of X_1, X_2, X_4, X_5 and X_7 . Using Eqs. (3)-(12) we obtain the following conditions for x_A and y_A :

we consider a range of octahedral ratios μ ; for each value of μ we calculate the angles θ and ϕ_1, ϕ_2 through Eqs. (13)-(16). Starting from these angles we obtain x_e, y_e, z_e and x_a, y_a, z_a via Eqs. (10)-(12). We then solve Eqs. (17)-(18) to determine x_A and y_A . Finally we use these results inside Eq. (19) to find r_μ .

A plot of $\rho_\mu = r_\mu/r_X$ vs. μ is shown in Figure S3. The two extremal configurations define two approximately straight lines; instead of repeating the above steps for every value of μ used in the structure maps of the main text, we fit the lines using a piece-wise linear function:

$$\rho_\mu = 1.125 + 0.732\bar{\mu} \text{ for } \bar{\mu} > 0.8. \quad (21)$$

B' sites, respectively, and F at the X site ($\bar{\mu} = 1.136$). In Figure 1e this boundary is labelled CL2.

Geometric bounds resulting from the no-rattling principle: $A_2BB'X_6$ double perovskites

In order to derive the geometric bounds for quaternary $A_2BB'X_6$ double perovskites, in this work we introduce generalized Goldschmidt's parameters as follows: (i) the average octahedral factor:

$$\bar{\mu} = (r_B + r_{B'})/2r_X, \quad (22)$$

(ii) the octahedral mismatch:

$$\Delta\mu = |r_B - r_{B'}|/2r_X, \quad (23)$$

and (iii) the generalized tolerance factor:

$$t = \frac{r_A + r_X}{\sqrt{2}\{[(r_B + r_{B'})/2 + r_X]^2 + (r_B - r_{B'})^2/4\}^{1/2}}. \quad (24)$$

It is immediate to verify that when $r_B = r_{B'}$ we obtain $\Delta\mu = 0$, and the tolerance and octahedral factors in Eqs. (24) and (22) reduce to the standard definitions used for ternary perovskites in Eqs. (1) and (2).

Stretch limit

The stretch limit corresponds to the extremal configuration wherein the A cation is in contact with the X anions. At variance with the case of ternary perovskites, in quaternary systems the X anion is no longer located midway between the B

and B' cations, as shown in Figure S1b. By applying Pitago-

ras' theorem to the triangle in Figure S1c we find:

$$(r_A + r_X)^2 = 2 \left(\frac{r_B + 2r_X + r_{B'}}{2} \right)^2 + \left[(r_B + r_X) - \frac{r_B + 2r_X + r_{B'}}{2} \right]^2. \quad (25)$$

After combining this equation with Eq. (22)-(24) we obtain the boundary $t = 1$, irrespective of $\bar{\mu}$ and $\Delta\mu$, precisely as for ternary perovskites.

Octahedral limit

The octahedral limit for double perovskites is found by considering the extremal configuration where two X anions belonging to the smallest octahedron among BX_6 and $B'X_6$ are in contact. The smallest radius among r_B and $r_{B'}$ is given by $r_{\min} = (\bar{\mu} - \Delta\mu)r_X$, therefore the octahedral limit for the corresponding octahedron reads $\sqrt{2}(r_{\min} + r_X) = 2r_X$, as in the case of ternary perovskites. After dividing by r_X we obtain the boundary plane $\Delta\mu = \bar{\mu} + \sqrt{2} - 1$ (irrespective of t). This boundary can be seen in Figure 2a as the surface of the stability region which forms an angle of 45° with the horizontal plane.

Tilt limit

In order to derive the tilt limit it is necessary to generalize the Platonic model of Ref. 12 to the case of double perovskites. However, since the majority of structures reported in database S1 of known perovskites and in database S2 of predicted perovskites have small octahedral mismatch ($\Delta\mu < 0.2$ for 95% of all compounds, Figure S8), and since by construction the tilt limit for $\Delta\mu = 0$ reduces to that of ternary perovskites, it is legitimate to use Eqs. (20) and (21) also for quaternary perovskites. The small error involved in this approximation is absorbed in the optimal formability, as determined in our statistical analysis below.

Chemical limits

The same limits CL1 and CL2 apply to ternary and quaternary perovskites.

Uncertainty quantification, optimum formability, and classification accuracy

To each compound in database S1 we associate a rectangular cuboid in the $(\bar{\mu}, t, \Delta\mu)$ space. The corners of the cuboids correspond to the minima and maxima of t , $\bar{\mu}$, and $\Delta\mu$ for each compound, as calculated using available ionic radii^{13,14}. For B-site cations we consider the ionic radii corresponding to six-fold coordination; for A and X cations we consider all available coordination numbers, since the coordination of these ions is sensitive to the octahedral tilt. When both high-spin and low-spin radii are available, we consider the high-spin state.

For each compound we define the formability f as the fraction of the cuboid volume falling within the stability region. To calculate the formability we discretize each cuboid using a $5 \times 10 \times 5$ mesh of points in the $(\bar{\mu}, t, \Delta\mu)$ space, and count the points that satisfy the geometric criteria set in Table S1. The largest formability is $f = 1$ (the cuboid is entirely inside

the stability region), the smallest formability is $f = 0$ (cuboid outside of the region). We classify a compound as a 'perovskite' if the calculated formability is above a critical value f_c (to be determined), and as 'non-perovskite' otherwise.

In order to determine the critical value f_c we proceed as follows. For each value of f_c we evaluate the classification accuracy of all the perovskites and of all the non-perovskites in database S1. The classification accuracy is defined as the number of compounds correctly classified divided by the number of all compounds of a given type. The variation of the classification accuracy with f_c is shown in Figure S6a. It is seen that the classification accuracy improves towards large f_c for perovskites, and towards small f_c for non-perovskites. The value at which the two curves intersect defines the critical formability, $f_c = 0.34$. Using this value as the dividing line between perovskites and non-perovskites, we achieve a classification accuracy of 79.7% for both perovskites and non-perovskites. Having the same classification accuracy for both groups is important in view of making predictions for datasets where it is not known which and how many compounds belong to either group.

To estimate the statistical uncertainty we make use of the central limit theorem. From all compounds in database S1.1 and S1.2, including both perovskites and non-perovskites, we randomly draw 50,000 subsets of N compounds (with N between 5 and 400), and we evaluate the classification accuracy for each subset. The distribution of these values tends to a Gaussian, as shown in Figure S6b. The standard deviation of this distribution depends on the size N of the subset, as shown in Figure S6c. For sets of 100 compounds or more, the fidelity of our classification scheme is found to be $79.7 \pm 4.0\%$ with a confidence level of 95%. As a result, if we consider sets of 100 ABX_3 or $A_2BB'X_6$ unknown compounds, in 95% of cases our geometric algorithm is expected to correctly classify between 76 and 86 compounds.

High-throughput screening of all possible ABX_3 and $A_2BB'X_6$ compounds

In order to identify all possible perovskites and double perovskites we proceed as shown schematically in Figure S6. We first construct all possible combinations of A, B, B' cations and X anions from the Periodic Table, by considering all elements with associated ionic radii in Shannon's database^{13,14}. We obtain 3,658,527 hypothetical compounds. Based on electrostatic considerations we exclude the combinations that do not satisfy the charge neutrality condition, that is $a + b + 3x = 0$ where a , b , and x denote the oxidation numbers of A, B, and X, respectively (and similarly for double perovskites). After this refinement we are left

Bound type	Abbrev.	Inequality
Stretch limit	SL	$t \leq 1$
Octahedral limit	OL	$\bar{\mu} \geq \sqrt{2} - 1 + \Delta\mu$
Tilt limit	TL1	$t \geq (0.44 \bar{\mu} + 1.37) / \sqrt{2}(\bar{\mu} + 1)$
	TL2	$t \geq (0.73 \bar{\mu} + 1.13) / \sqrt{2}(\bar{\mu} + 1)$
Chemical limit	CL1	$t \leq 2.46 / [2(\bar{\mu} + 1)^2 + \Delta\mu^2]^{1/2}$
	CL2	$\bar{\mu} \leq 1.05$

Table S1. **Geometric limits for the formability of perovskites.** Summary of the inequalities defining the stability region of perovskites and double perovskites according to Goldschmidt’s no-rattling principle (blue volume in Figure 2A). The definitions of the generalized tolerance factor t , the generalized octahedral factor $\bar{\mu}$, and the octahedral mismatch $\Delta\mu$ are given in Eqs. (22) through (24), respectively.

X	A	B	B'	Family		
-1	+1	+1	+3	Halides		
		+2	+2			
-2	+1	+3	+7	Oxides and chalcogenides		
			+4		+6	
			+5		+5	
			+2		+1	+7
			+2		+6	
			+3		+5	
			+4		+4	
			+3		+1	+5
			+2		+4	
			+3		+3	
-3	+2	+7	+7	Nitrides		
			+3		+6	+6
			+5		+7	
			+4		+3	+7
			+4		+6	
	+5	+5				

Table S2. **Cation oxidation states in perovskites and double perovskites.** For each oxidation state of the anion X, the possible oxidation states of the cations A, B, and B' are constrained by the charge neutrality condition and Pauling’s valency rule. The oxidation state of oxygen and chalcogens admits the largest number of combinations of cation oxidation states.

```
w3m "http://duckduckgo.com/?q=\${compound}"
-dump | grep \${compound} | grep -v Search | grep -v 'Limit results to'
```

where ‘ $\${compound}$ ’ is a plain string with the chemical formula of the compound, for example ‘ $\text{Cs}_2\text{BiAgCl}_6$ ’. The compound has been found on the internet if this shell command returns a non-empty string. To test the procedure we executed this command for all the known perovskites in database S1.1, and S1.3 and we correctly found 1254 out of 1699 perovskites. Therefore the success rate of this data mining procedure is 74%.

By executing the above command for the 94,232 compounds of database S2 we found that 92,899 compounds do not have any presence on the internet, and hence can be considered

with 1,439,753 hypothetical compounds, including duplicate structures which differ only by the oxidation states of their elements. We further refine this dataset by retaining only compounds that fulfil Pauling’s valency rule, whereby the oxidation number of A does not exceed that of B¹⁵; after this refinement we are left with 1,131,737 structures. For each of these combinations we calculate its rectangular cuboid in the $(\bar{\mu}, t, \Delta\mu)$ space and the associated formability f . We find 94,232 distinct compounds that pass the test $f > f_c = 0.34$ and which do not belong to the set of known perovskites in databases S1.1 or S1.3. The final set of predicted perovskites and double perovskites is reported in database S2.

Data mining the internet

In order to verify that the 94,232 compounds of database S2 have not hitherto been reported, we performed a high-throughput internet search of all their chemical formulas. We attempted to use Google, but automated searches are considered as web-scraping by this engine, and are not allowed. We therefore opted for DuckDuckGo (DDG), which can be used for high-throughput searches. DDG generates results from many sources including Yahoo!, Wikipedia, and Bing¹⁶. Using a Unix tcsh shell we executed the following search key for each compound:

as new perovskites. 1333 compound formulas were found to have a presence on the internet. We therefore investigated each of these compounds by performing manual internet searches. We retained only compounds which appeared in scientific documents (such as peer-reviewed publications, conference abstracts or proceedings, national laboratory reports or masters and doctoral dissertations) and which contained a minimal Vdiscussion of the crystal structure. Using this procedure we unambiguously identified 786 ternary and quaternary crystals, out of which 555 are perovskites. These 786 compounds were used for our blind test of the classi-

fication accuracy, and are reported in database S2.2.1 (perovskites) and S2.2.2 (non-perovskites) alongside the literature references that we used to confirm their classification. After removing the 786 compounds found from internet mining, we obtain a set of 93,447 predicted perovskites never hitherto reported, listed in database S2.1.

Prediction of Fe₂O₃ perovskite phase

Experimental observations of the transition of Fe₂O₃ into the perovskite phase at high pressure have been previously reported in Refs. 17,18. This transition could be rationalized qualitatively by comparison with the ilmenite to perovskite phase transition which occurs, for example in FeTiO₃²⁰, and involves a rearrangement of the position of the cations types in the alternating layers, as shown in Figure S10b-c. The corundum structure of Fe₂O₃ can be related to the ilmenite structure of FeTiO₃ simply by replacing the Ti with Fe, and it is therefore possible that a similar phase transition mechanism as that of FeTiO₃ occurs for Fe₂O₃ at high pressure.

We have performed a set of preliminary calculations to assess of the stability of Fe₂O₃ in a perovskite crystal structure, with respect to its known stable structure at ambient temperature and pressure condition, the corundum structure. In Figure S10 we show a comparison of the enthalpies calculated within DFT as a function of pressure between 0 and 200 GPa for the corundum and perovskite phases of Fe₂O₃. We note that our calculations do not include the effect of temperature and zero-point motion. As shown in Figure S10, at 0 GPa the corundum structure is more stable than the perovskite structure by approximately 20 kJ/mol, but the formation enthalpies of the two structures become closer as the pressure increases. In the inset of Figure S10 we show that at pressures beyond 140 GPa the perovskite phase becomes more stable than the corundum phase, in agreement with the predictions of our model. We note that transition pressure of 140 GPa is significantly overestimated with respect to the experimental measurement of 30-45 GPa reported in Refs.^{17,18}. To achieve a better agreement with experiment, more refined calculations of the phase diagram of Fe-O are required, which should include a comparison between all possible competing phases of Fe₂O₃, the effect of vibrational entropy, as well as a thorough assessment of the total energy for different exchange correlation functionals and different magnetic configurations of Fe₂O₃.

From our preliminary DFT calculations we estimate that the average ionic size for the Fe₂O₃, $R_0 = \left(3\Omega/4\pi\right)^{1/3}$ where Ω is the unit cell volume, increases by 12% when with pressure increasing from 0 to 200 GPa. On the other hand, the ionic radius of Fe³⁺ takes values between 0.49 Å (4-fold coordination) and 0.78 Å (8-fold coordination), a change of 14%. This observation indicates that our analysis does not distinguish between perovskites which can be synthesized at ambient or high pressures. Therefore, the prediction of Fe₂O₃ perovskite demonstrates that the geometric blueprint can be in general used as a tool for not only predicting novel perovskite compounds, but also novel high-pressure perovskite phases of previously known polymorphs.

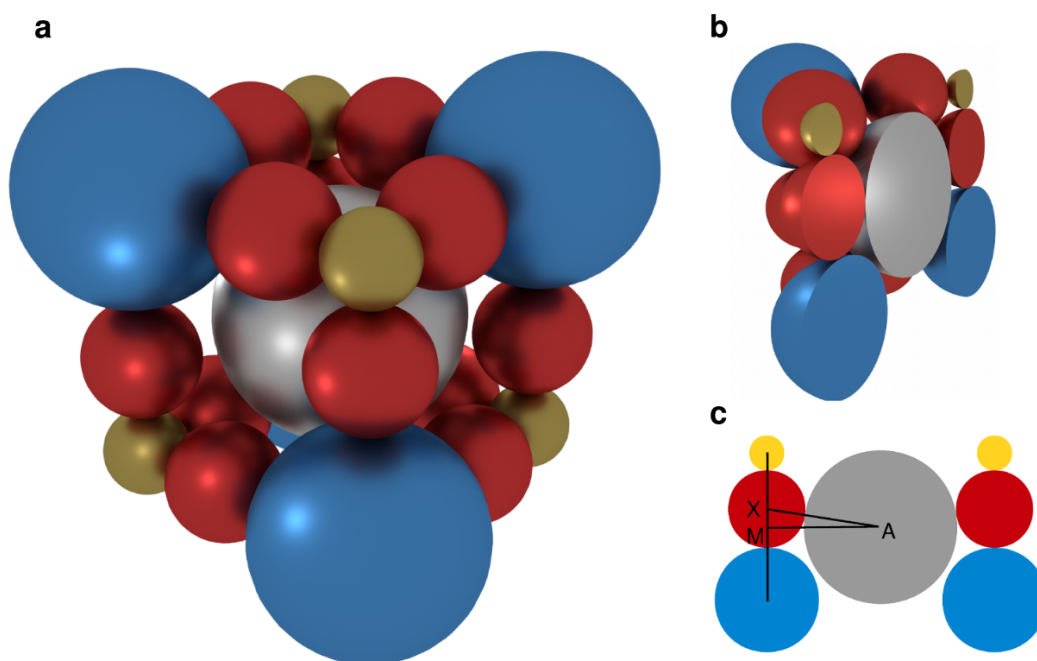


Figure S1. Rigid sphere model of double perovskites. **a**, Illustration of a cubic double perovskite $A_2BB'X_6$ in the rigid-sphere model. The A cation is in grey, B and B' are blue and yellow, respectively, and X is in red. **b**, Schematic representation of the stretch limit for double perovskites. In this section cutting through the centers of A, B, B', and X we see that the A cation is in contact with the X anion. **c**, The triangle used to derive the stretch limit for double perovskites, in the same configuration as in **b**.

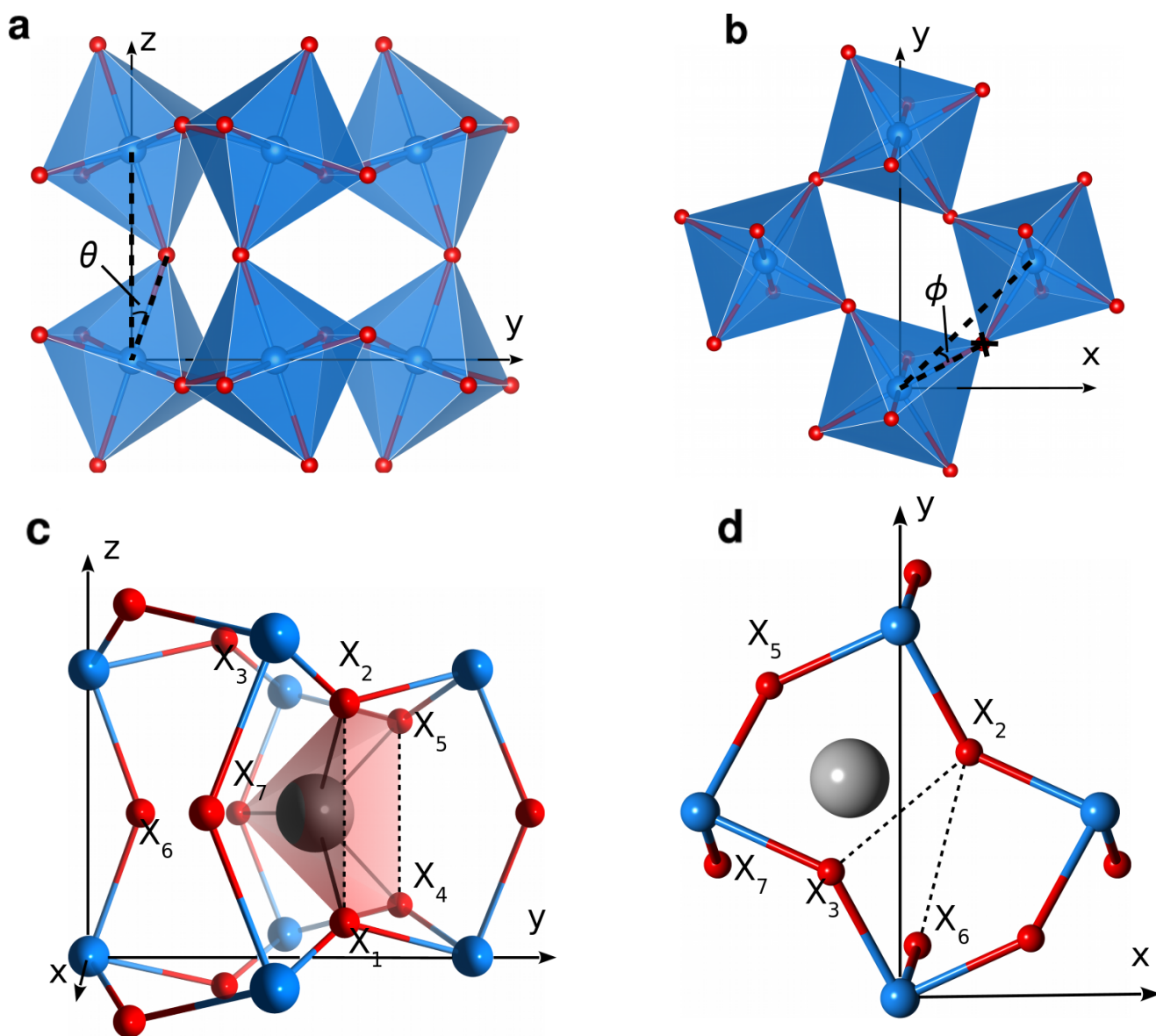


Figure S2. Geometric construction for the tilt limit. Polyhedral models of the perovskite cavity in the tilt limit, **a**, view along the x -axis and **b**, along the z axis. The octahedral tilt angle θ and the precession angle ϕ are indicated, and the A-site cation is not represented in **a** and **b** for clarity. The black cross in **b** indicates that the angle is taken between the adjacent B cations and the projection of the X anion on the (x, y) plane. **c** and **d**, Ball and stick models of the perovskite cavity in the tilt limit, view along the x -axis (**c**) and view along the z axis (**d**). The ions are labelled according to Eqs. (3)–(10). The dotted lines mark the X anions that come in contact in the tilt limit. The coordination of A is highlighted by the shaded polyhedron in **c**. The Cartesian axes refer to an orthorhombic unit cell in the Pnm space group.

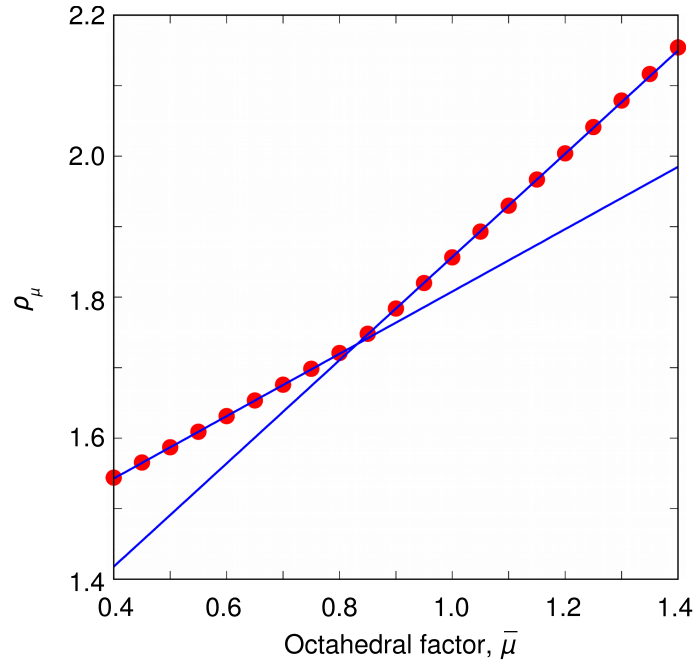


Figure S3. Numerical evaluation of the tilt limit. Plot of the ratio $\rho_\mu = r_\mu/r_X$ as a function of the average octahedral ratio $\bar{\mu}$, as evaluated numerically from Eq. (19) for the tilt limits TL1 and TL2 (disks). We also show the linear fits of these curves, as reported in Eq. (21) (lines).

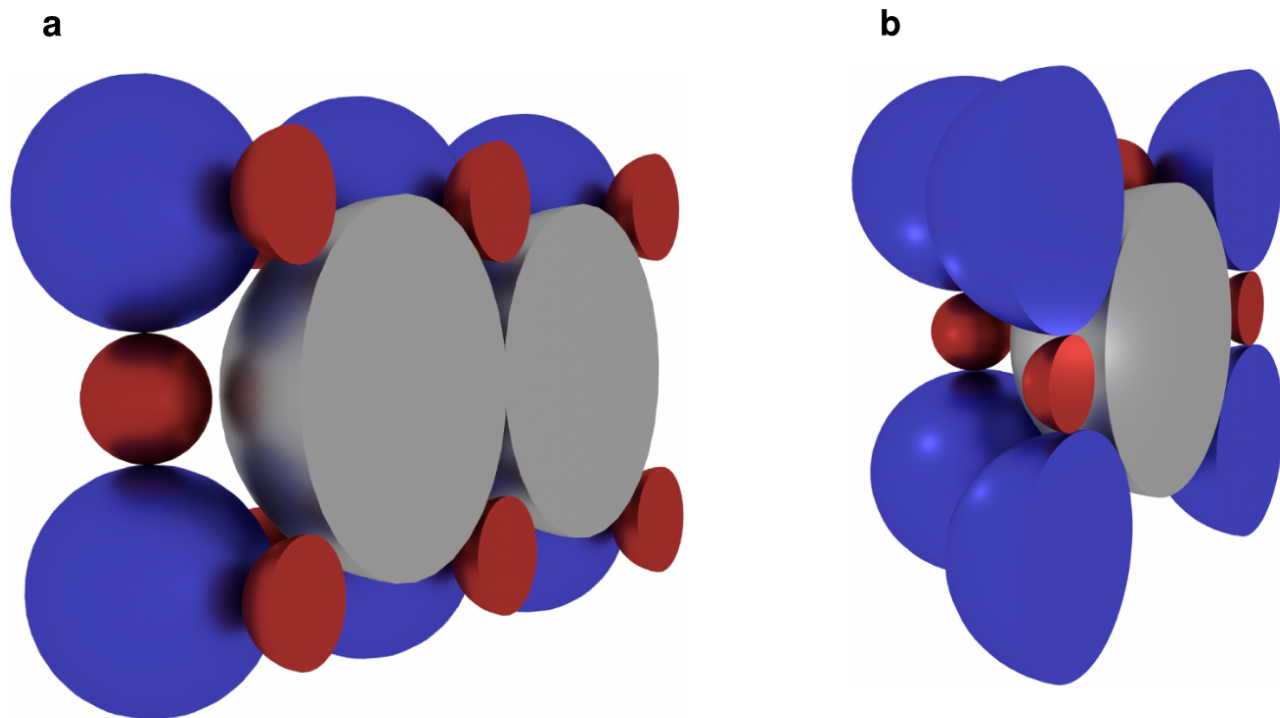


Figure S4. Schematic representation of the secondary stretch limits for ABX₃ perovskites. a, Two adjacent A-site cations are in contact (limit SSL1). **b,** The A and B cations are in contact (limit SSL2). The A, B, and X cations are in grey, blue, and red color, respectively.

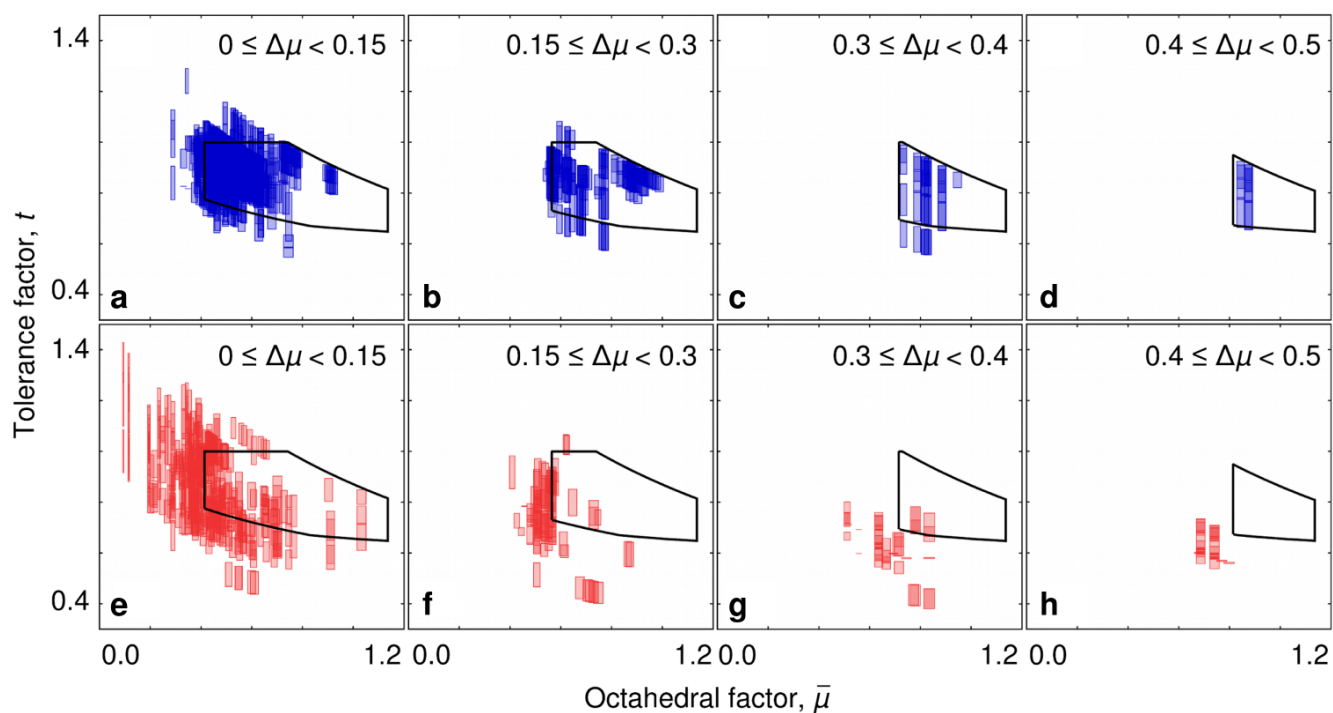


Figure S5. Detailed view of the distribution of known compounds in the $(\bar{\mu}, t, \Delta\mu)$ space. **a-d**, Distribution of known perovskites from databases S1.1 and S1.3 in the $(\bar{\mu}, t)$ plane, for several ranges of the octahedral mismatch $\Delta\mu$ as indicated at the top of each panel. For each compound we show the two-dimensional cross-section of the rectangular cuboid representing the uncertainty in the ionic radii. The black lines represent sections of the stability region for $\Delta\mu$ in the middle of the range indicated. **e-h**, Distribution of compounds from database S1.2 which are not perovskites, for several ranges of $\Delta\mu$. The rectangles have the same meaning as in panels **a-d**.

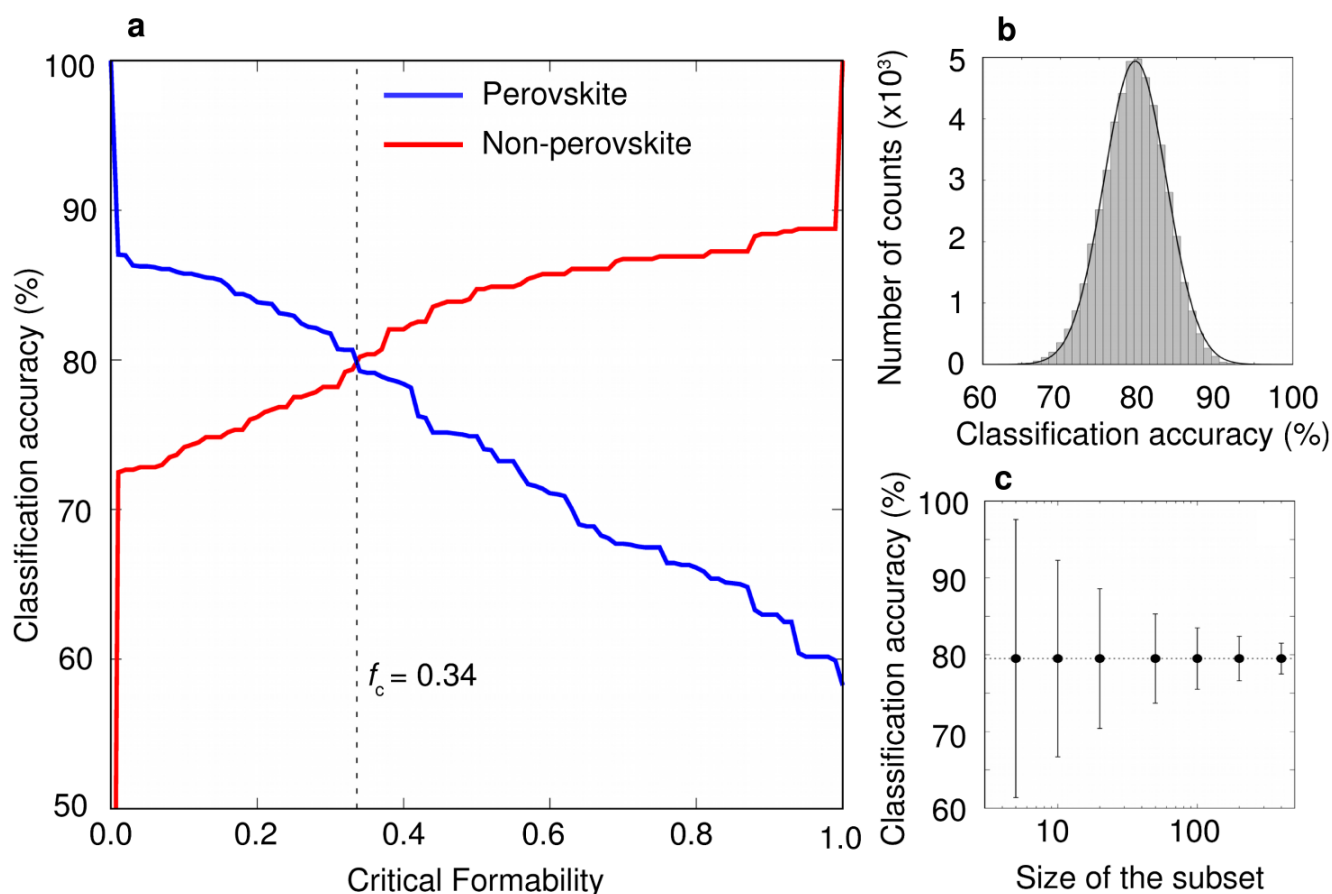


Figure S6. Statistical analysis of the classification accuracy. **a**, Classification accuracy of the geometric model, as a function of the critical formability f_c . The blue line is for perovskites from database S1, the red line is for non-perovskites in the same database. The lines intersect at $f_c = 0.34$, with this choice the classification accuracy is 79.7%. **b**, Analysis of the classification uncertainty. We consider 50,000 subsets of 100 compounds, randomly drawn from database S1, and we plot the distribution of the classification accuracy of each subset (histogram). We obtain an approximately normal distribution, with a standard deviation of 4%. A Gaussian function with the same standard deviation and mean is also shown for comparison. **c**, Mean and standard deviation of the classification accuracy calculated as in **b**, for subsets of varying size. The standard deviation decreases for larger subsets.

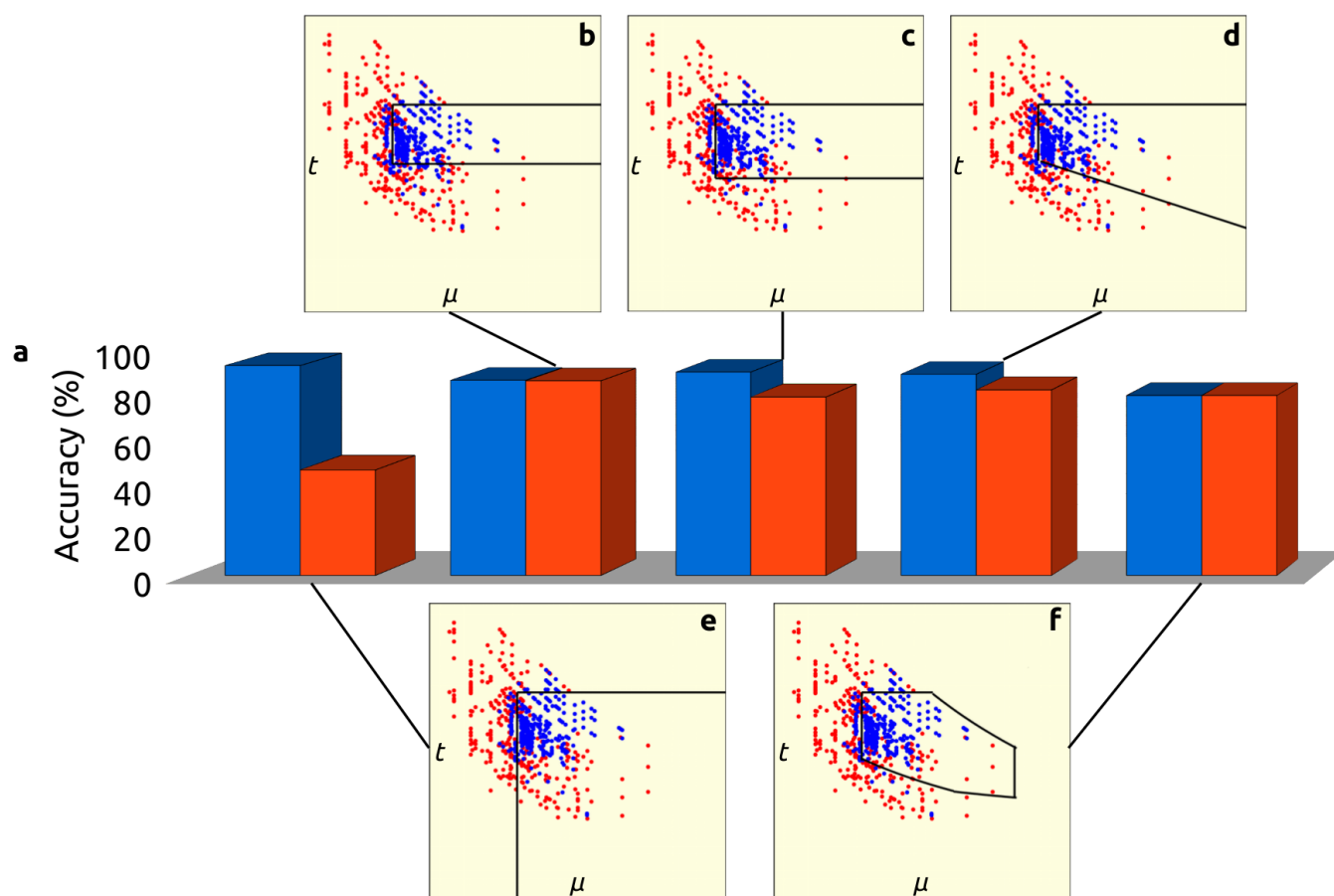


Figure S7. Comparison with other perovskite regions proposed in literature. Accuracy of different perovskite from literature (a): Ref. 19 (b: octahedral limit, stretching limit and $t \geq 0.8$), Ref. 20 (c: octahedral limit, stretching limit and $t \geq 0.75$), Ref. 3 (d: octahedral limit, stretching limit and $t \geq 0.91 - 0.23\mu$), Goldschmidt's model¹¹ (e : octahedral limit and stretching limit). (f) corresponds to the same analysis performed on our model. The bars indicate the accuracy with which a model classifies all compounds in Database S1.1 as perovskites (blue) and Database S1.2 as non-perovskites (red). For (b-e) we calculate the accuracy using the average t , $\bar{\mu}$ and $\Delta\mu$ (as described in the caption of Figure 2 of the main manuscript), while for (f) we take into account the variability of the ionic radii, as described in the Supporting Information. The data points in each maps are only shown for ternary compounds for clarity. The red dots correspond to non-perovskites and the blue dots correspond to perovskites.

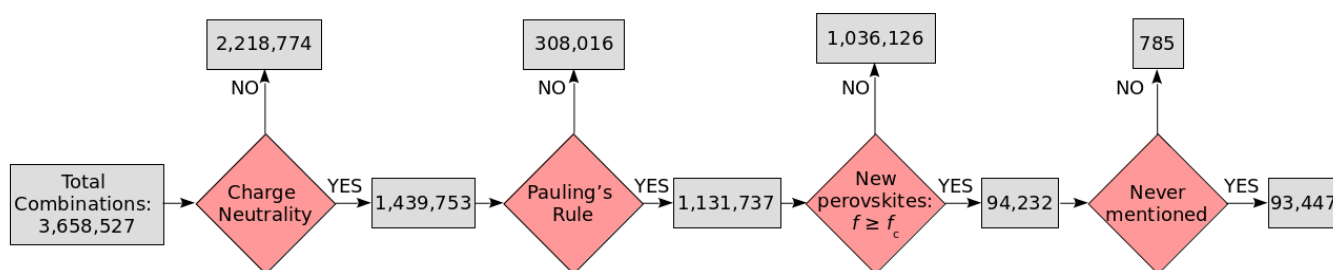


Figure S8. High-throughput screening of perovskites using the geometric model. Schematic flow-chart of the procedure used to predict which compounds, among all 3,658,527 hypothetical combinations of A, B, B', and X ions, will crystallize in a perovskite or double perovskite lattice.

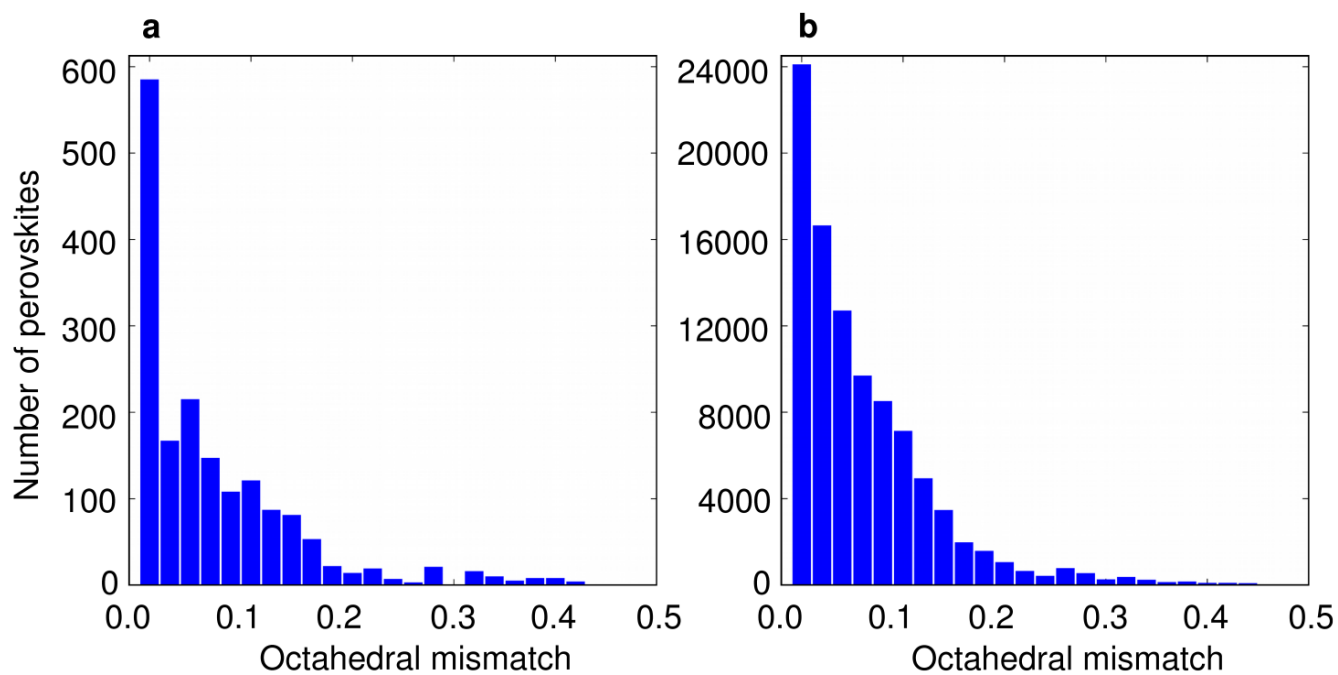


Figure S9. Distribution of known and predicted perovskites as a function of octahedral mismatch. **a**, The histogram shows the number of known perovskites from database S1 (sticks) across the range of the octahedral mismatch $\Delta\mu$. The majority of compounds (94%) is found in the range $\Delta\mu < 0.2$. **b**, Same analysis as in **a**, but for the predicted perovskites from database S2 (sticks). Also in this case the majority of perovskites (96%) is found in the range $\Delta\mu < 0.2$.

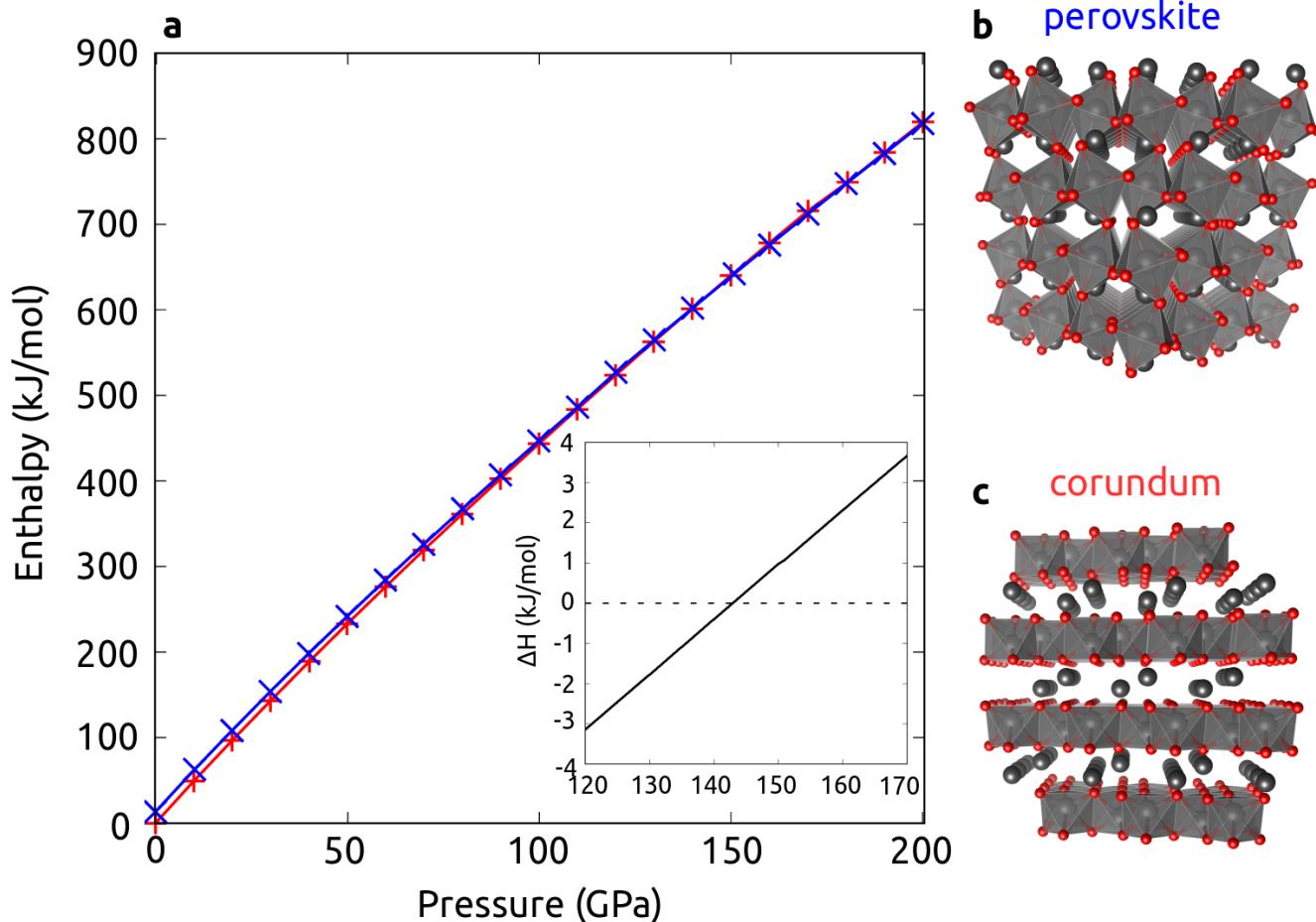


Figure S10. Calculated enthalpies for the perovskite (blue) and corundum phases (red) Fe_2O_3 as a function of pressure. The enthalpies are referred to the total energy of Fe_2O_3 corundum at 0 GPa. The inset shows the difference in the enthalpies of the corundum and perovskite phases as a function of pressure ($\Delta H = H_{\text{hematite}} - H_{\text{perovskite}}$). Enthalpies are calculated from the local density approximation to density functional theory, DFT/LDA, by performing full structural optimizations at pressures between 0 and 200 GPa (with 10 GPa increments). Calculations are performed using ultrasoft pseudopotentials, a plane wave cutoff of 60 Ry, a charge density cutoff of 300 Ry and a \mathbf{k} -point grid of $4 \times 4 \times 4$ centered at Γ . The lines in the (a) are obtained by interpolating through the data points, and the black line in the inset is obtained as the difference of the interpolated enthalpies. (b-c) Polyhedral models of the perovskite and corundum crystal structures. The grey spheres correspond to the Fe ions and the red spheres correspond to O. The structures were obtained by replacing Ti with Fe in the FeTiO_3 ilmenite and perovskite structures, respectively.

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