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Assessing the utility of structure in amorphous materials

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ABSTRACT
This paper presents a set of general strategies for the analysis of structure in amorphous materials and a general approach to assessing the utility of any selected structural description. Two measures of structure are defined, “diversity” and “utility,” and applied to two model glass forming binary atomic alloys, Cu_{50}Zr_{50} and a Lennard-Jones A_{80}B_{20} mixture. We show that the change in diversity associated with selecting Voronoi structures with high localization or low energy, while real, is too weak to support claims that specific structures are the prime cause of these local physical properties. In addition, a new structure-free measure of incipient crystal-like organization in mixtures is introduced, suitable for cases where the stable crystal is a compound structure.

I. INTRODUCTION
The explanation of material properties and behavior in terms of the microscopic structure constitutes the modus operandi of the physical sciences—chemistry, physics, and materials science. It seems a natural expectation, therefore, that a science of amorphous materials should eventually be built on analogous structural explanations. While a considerable body of literature records the effort to advance just this program, success has proven elusive. With their periodic repetition of a single unit cell, crystal structures require only a small amount of information to specify the total structure. This is not the case in amorphous solids, where any useful measure of structure (where “useful” refers to a measure that does not involve a complete specification of every particle position) must be seriously incomplete. It follows that each specific structural measure will unavoidably represent a choice regarding what information is retained and what is discarded. Some choices must be more useful than others.

In this paper, we consider how one might assess the utility of a structural measure of an amorphous material. Any definition of amorphous structure requires that the researcher must make a choice of classification scheme. This choice is integral to the requirement that structure be intelligible. The choice of the local classification criteria is unrestricted. Along with Voronoi polyhedra, examples of local classification include common neighbours, coordination geometries of nearest neighbours (roughly, the dual of the Voronoi representation), clusters (e.g., poly-tetrahedrality), local ring lengths, and degree of local centro-symmetry. The question we address in this paper is: how useful is a given structural representation given the arbitrary choice that is made regarding how local structure is classified? Many of the papers reporting on the structure of an amorphous material assume that structure, however defined, is useful by default. The object of this study is to establish how this assumption can be put to a meaningful test. The outcome of such an assessment will depend on the choice of structural measure—common neighbour analysis, order parameter clusters, tetrahedrality, etc. The aim of this paper is not to provide a comprehensive assessment of all the different choices of structural measures that have been proposed. This is a task well beyond the scope of this preliminary study. Nor are we trying to reach a sweeping general
conclusion about the utility of amorphous structure based on any particular choice of structural measure. Our goal is simply to develop a number of operational approaches to quantifying utility, and, to this end, we shall select a single option for structural measure—the frequency of Voronoi polyhedra—as our test case. We stress that we make no claim for the superiority of the Voronoi method in this choice, simply that it is a familiar and widely used approach for describing amorphous structure.

For our purposes, the "structure of an amorphous material" shall be taken to mean the frequencies of some local classification of particle configurations, i.e., nearest neighbour coordination geometry. This is the definition of amorphous structure that is widely used in the literature.\(^1\)\(^-\)\(^2\) There have been a number of studies that have considered structural measures that extend beyond nearest neighbours.\(^3\) The inclusion of this intermediate order must generally improve the capacity to resolve distinctive structural features of disordered materials but at the cost of increasing the complexity of the enumeration of distinct structures. In this study, we have opted for the relative simplicity of the Voronoi resolution to develop the utility analysis and leave the analysis of these more expansive structural measures for future work.

"Useful" is probably an even more treacherous concept to define than "structure." To what uses do we put structural information? In this paper, we shall consider the following three: (i) Structure as information compression. The information required to determine the positions of all of the atoms in a crystal is typically small (i.e., the unit cell structure and lattice parameters) and independent of the size of the sample. This dramatic data compression provided by the structure allows for the structure of crystalline materials to be easily stored, recovered, and used. (ii) Structure as a causal explanation of a physical property. The notion of energetically favored local structures is a common starting point for rationalising liquid structure. Malins et al.\(^9\) used this energetic criterion as means of identifying coordination polyhedra of interest in resolving amorphous structure. If the stability of an amorphous material could be closely correlated with the presence of specific local structures, these structures could, in turn, provide an explanation of a range of material properties arising from configurational stability. (iii) Structure as a measure of proximity to an ordered phase. One of the most cited papers on amorphous structure is a short note by Frank\(^\text{10}\) in 1952 in which he suggested that localicosahedral coordination shells might stabilize a pure liquid. As developed by Mackay,\(^11\) Hoare and Pal,\(^12\) and Kivelson et al.\(^13\) the idea has evolved into a proposition that disorder may be underpinned by a form of geometrically frustrated order.

There is a fourth common usage of structure—Structure as the rules by which the whole is assembled from its parts. While this can be regarded as an example of data compression (i.e., the first point in our list), assembling a structure is a quite specific process and one of the more generic characteristics of a structure (as something that has been assembled). We shall not consider this fourth use in this particular study. By choosing to look at a short range order in the form of nearest neighbour coordination geometries, we have discarded, by default, the information about the correlations between local structures and, hence, how these local environments are assembled over medium range lengths. There have been studies that have sought to extend the structural characterisation to include the spatial arrangement of coordination polyhedra.\(^14\) The interesting challenge to define a measure to assess this fourth aspect of utility—the capacity to reconstitute the whole from some reduced structural measure—is the one we leave for future work. In looking for a more compact description of extended structure, some workers have considered approximating the extended arrangement of local coordination polyhedra as a disordered (plastic) crystal lattice.\(^15\)\(^\text{16}\) Whether such approximants are even stable with respect to the non-periodic reality has been questioned.\(^17\)

II. MODELS AND ALGORITHMS

Access to information about amorphous structure is largely via models. Historically, the initial models were analog: the bubble rafts of Bragg and Nye\(^18\) and Bernal’s randomly close packed ball bearings.\(^19\) With the advent of computer simulations, these analog models were replaced by digital ones. Some important early examples of the application of Voronoi polyhedra to the analysis of simulated liquid structures are the papers by Rahman\(^20\) and Tanemura et al.\(^21\)

In this paper, we shall use two well studied model glass forming liquids, both based on binary atomic mixtures. One is a model of CuZr using a many body potential of the embedded atom type based on the work of Mendeleev et al.\(^22\) The equilibrium crystal phase of the model CuZr\(^23\) is the B2 structure (i.e., a body centred cubic structure with the two species occupying alternating sites similar to that found in CsCl). While the B2 crystal has been observed to grow,\(^24\) the simulated CuZr has proven highly resistant to nucleation during extended simulations. The other model is a mixture of two Lennard-Jones spheres introduced by Kob and Andersen (KA).\(^25\) The interparticle potential has the following functional form:

\[
\phi_{ij}(r) = 4\varepsilon_{ab} \left( \frac{\sigma_{ab}}{r} \right)^{12} - \left( \frac{\sigma_{ab}}{r} \right)^{6},
\]

with potential parameters \(\sigma_{AA} = 1.0; \sigma_{AB} = 0.8 \text{ and } \sigma_{BB} = 0.88; \) and \(\varepsilon_{AA} = 1.0, \varepsilon_{AB} = 1.5. \text{ and } \varepsilon_{BB} = 0.5.\) For the KA model, we shall use temperature units of \(\varepsilon_{AA}/k_B, \) energy units of \(\varepsilon_{AA},\) and length units of \(\sigma_{AA}.\) The model has mainly been studied at the A\(_{80}\)B\(_{20}\) composition (the same composition associated with optimal glass formation in NiP\(^26\) on which the KA model was based). The equimolar mixture crystallizes readily in the same B2 crystal structure as found in the CuZr model.\(^27\)\(^28\) At A\(_{80}\)B\(_{20},\) crystallization is much slower and is driven by crystallization of the face centred crystal of pure A.\(^28\) The structures of the supercooled liquid in CuZr\(^29\) and the KA mixture\(^30\) have both been studied extensively.

The amorphous states studied in this paper were generated by continuous cooling of a liquid initially equilibrated above its melting point down to \(T = 0\) under the constraint of constant pressure. The resulting energy minimum can be characterised by a fictive temperature equal to the temperature at which, on cooling, a property such as the volume deviated from the equilibrium value as a consequence of structural arrest. In Fig. 1, we plot the dependence of the volume \(V\) for the CuZr and A\(_{80}\)B\(_{20}\) alloys as a function of temperature during cooling and indicate the temperature at which the deviation from equilibrium occurs in each case and, hence, the fictive temperature of the respective \(T = 0\) glasses.

The local structure of the \(T = 0\) amorphous states has been characterized using Voronoi polyhedra. We have used a standard 4 digit descriptor for the polyhedra \((n_3, n_4, n_5, n_6)\) where \(n_i\) is the number of facets with \(i\) edges. Note that this Voronoi analysis is purely

FIG. 1. The glass transition temperatures $T_g$ for (a) CuZr and (b) the KA mixture at $A_{80}B_{20}$ for the cooling rates $10^4$ K/s and $1.3 \times 10^5$, respectively, used to generate the $T = 0$ configurations whose structures are reported in this paper. $T_g$ is defined as the temperature at which $dV/dT$ undergoes an abrupt change. This $T_g$ is the fictive temperature of the $T = 0$ configuration. The units in (b) are the Lennard-Jones reduced units.

topological and does not differentiate the two species in the local coordination shell. The Voronoi analysis suffers from a problem common to many forms of structural classification defined in terms of neighbor separations. The identification of a neighbor is all-or-nothing depending on some threshold distance—explicit or implicit—resulting in substantial fluctuations in the topological signature due to the fluctuations of separations close to this threshold. In the case of Voronoi analysis, large separations correspond to small faces, a problem that has been discussed previously. An example of this issue is the Voronoi structure of the FCC crystal at a non-zero temperature. The Voronoi polyhedron around an atom in a perfect FCC lattice is $(0, 12, 0, 0)$, but this polyhedron is not observed at finite $T$—instead the dominant polyhedron is $(0, 6, 0, 8)$—one characteristic of the BCC lattice—simply as a consequence of vibrational motion (i.e., no defects are required to see this structural broadening). An alternative approach has been introduced using Minkowski tensors that weigh neighbor contributions based on their separation from the central particle that promises to reduce these fluctuations. While we shall not explore these more sophisticated measures—in this paper, we seek to frame general questions about structure—readers are encouraged to view the statistics of any local structural description as a combination of real local variability and the noise imparted by the details of the chosen metric.

III. RESULTS

A. Structure as data compression: On quantifying diversity

The statistical structure of a glass, as we have defined it here, takes the form of the fraction $p_i$ of particles in the local polyhedra labeled $i$. In Fig. 2, we present these fractions for CuZr and the KA mixture with composition $A_{50}B_{50}$. Also presented, for comparison, is the analogous structure of the B2 crystal structure as formed from the quenched $A_{50}B_{50}$ KA mixture. This crystal is the equilibrium crystal phase for the model CuZr alloy as well. In both amorphous alloys, we find a broad distribution of local structures, with no one structure exceeding 7% in frequency. This flat distribution of Voronoi structures is a common place observation for amorphous alloys. In the absence of any outstanding structures, we suggest that the most striking feature of distributions like those shown in Fig. 2 is exactly their multiplicity. Indeed, the most straightforward and general structural differentiation between a crystal and a glass is not the presence of any specific structure but the variety of local structures in the latter as compared to the former. Quasicrystals, for example, lack periodicity but are still ordered by virtue of consisting of only a small number of local structures.
The diversity D of structures—i.e., the effective number of distinct structures present—is related to the information S through the relation

\[ D = \exp(S). \]  

This definition of diversity is used in studies of animal and plant populations.\(^\text{35}\) The values of S and D are included on each graph in Fig. 2. For the perfect B2 crystal, there is a single local structure, so S = 0 and D = 1. The difference between this ideal and the values reported for the crystal as formed in Fig. 2(c) is that defects have been captured in the crystal as it was crystallized during the simulation. With values of D = 95 and 129, for amorphous CuZr and A\(_{80}\)B\(_{20}\), respectively, it is clear that the Voronoi classification of the amorphous alloys leaves us with a very large diversity of local structures. To appreciate just how large these amorphous D’s are, it is helpful to consider the range of diversity in crystals. In the A\(_{80}\)B\(_{20}\) cubic crystal, we find D = 3.3 [Fig. 2(c)]. A survey of over 16,000 intermetallic crystal structures\(^\text{36}\) reports that over 92% had 4 or less distinct coordination geometries (i.e., D ≤ 4). Note that the unit cells can be much larger than the number of distinct coordination sites as the same site might appear in different orientations. Based on this statistic, we might tentatively suggest that crystals correspond to a structure class characterised by a low (i.e., D < 10) structural diversity. An analogous observation was been enshrined by Pauling in his “rule of parsimony.”\(^\text{37}\)

The diversity, as defined here, is distantly related to the configurational entropy. Both quantities reflect our efforts to enumerate configurations in terms some sort of imposed resolution. Where diversity depends on a researcher’s choice of structural measure, configurational entropy\(^\text{38}\) depends on the researcher’s definition of a reference configuration (e.g., local potential energy minima and distinct free energy minima). Both measures provide a useful insight as to how the multiplicity of configurations decreases with cooling. The diversity expresses that multiplicity explicitly in terms of the types of structure selected.

The diversity D is introduced here as a useful tool for quantifying structures that might be more usefully characterised by their multiplicity rather than by the frequencies of a few individual structures. Before proceeding to a specific application of D, we need to address the issue of sample size dependence. Let n be the number of individual local geometries that are summed over in calculating D. It is clear that if n is small enough so as to fail to properly sample the distribution of environments, then D will be decreased relative to its true value in the thermodynamic limit. To determine how big n must be to avoid the sample size effect in D, we have calculated D as a function of n for the A\(_{80}\)B\(_{20}\) mixture. The results, plotted in Fig. 3, show that n must be greater than ~10\(^4\) to provide an adequate sampling of the distribution of Voronoi polyhedra.

As an example of how we might employ the diversity D, let us consider how the diversity of structures changes as we increasingly restrict our consideration to those structures associated with some extreme of a property. We shall consider the degree of constraint experienced by particles as measured by \(k_B T/(\Delta r^2)\) where \(\langle \Delta r^2 \rangle\) is calculated by averaging over trajectories at a T well below the glass transition. We are generally interested in the structures corresponding to high constraint. To this end, we shall pick a threshold value \(k_B T/(\Delta r^2)\) and then determine the histogram of structural frequencies for the subpopulation of particles for which \(k_B T/(\Delta r^2) > k_B T/(\Delta r^2)^*\). In Fig. 4, we plot the variation of D with the choice of threshold value \(k_B T/(\Delta r^2)^*\) for the two mixtures. The null hypothesis is that the constraint is independent of structure and so randomly distributed across all local structures. If this were so, then the change in the threshold would not change the diversity of structures at all, at least until the imposed constraint reduced the sample size n to a value below that of the threshold as determined above in Fig. 3. For each mixture, we have identified the value of \(k_B T/(\Delta r^2)^*\) above which n decreases below the sampling threshold value of 10\(^4\) and we disregarded these values of D since any real physical correlation between constraint and diversity will be conflated with the effect of the sampling error. Within the range of statistically significant data, we still find a substantial decrease in the diversity, with D decreasing from 96 to 80 in CuZr and from 129 to 40 in A\(_{80}\)B\(_{20}\). This means that (a) the constraint does exert a significant degree of selectivity on structure and (b) that, within a sampling range, this decrease in diversity still leaves us with a substantial number of contributing structures.

One contributing factor to this reduction in diversity is that the constraint favors one species over the other. In Fig. 5, we plot the change composition of the sub-population as a function of \(k_B T/(\Delta r^2)\). We find that the larger particles are more strongly represented in the high constraint particles than the smaller ones. In the CuZr mixture, the fraction of Zr centred structures for which \(0.65 \leq k_B T/(\Delta r^2)^*\) is 0.77 (as compared to 0.5 for the total system). In the A\(_{80}\)B\(_{20}\) mixture, the fraction of the larger A particles for which \(55 \leq k_B T/(\Delta r^2)^*\) is 0.98 (again, a significant increase over the total value of 0.8). The loss of diversity associated with the complete loss of small particles is given by the difference D\(_\text{total} - D\_\text{large}\) (where D\(_\text{large}\) is the diversity of the large particles measured across all values of constraint) which, as plotted in Fig. 4, is 19 and 40, for CuZr and A\(_{80}\)B\(_{20}\), respectively. Assuming a linear interpolation based on the measured change in composition, we estimate the fraction of the maximum decrease in diversity (i.e., for the maximum values of \(k_B T/(\Delta r^2)\))
FIG. 4. Plot of the diversity D for (a) CuZr and (b) A₈₀B₂₀ glass as a function of \( k_B T / \langle \Delta r^2 \rangle^* \), where \( \langle \Delta r^2 \rangle^* \) is the upper bound on the mean squared displacement. For each glass former, values of D are calculated for the individual chemical components as well as for the two components combined as indicated. The null hypothesis (see text) is also plotted. The shaded region in each plot indicates the values of \( k_B T / \langle \Delta r^2 \rangle^* \) sufficiently restrictive that the sample size has dropped below \( n^* \) (see Fig. 3), thus introducing non-negligible sample size effects.

of \( k_B T / \langle \Delta r^2 \rangle^* \) for which statistical significance is retained) that is due to the change in the composition: 26% and 51% for the CuZr and A₈₀B₂₀ mixtures, respectively. The remaining loss of diversity in each case is due to the explicit structural selectivity associated with the application of the threshold.

This observation—that structural diversity decreases when we focus on increasingly extreme values of some property—is a generic one in amorphous materials and is often cited as evidence that the favored structure must have “caused” the property extreme. As we shall discuss in Sec. III B, this extraction of causal connection from simple correlations is a non-trivial task.

B. Structure and the causal explanation of material properties

The statement that correlations do not imply causation is a basic tenet of statistics. Physical sciences, by contrast, quite routinely see correlations, coupled with some physical insight, employed to establish explanations of material behavior. The question of how this apparent difference is bridged is the basis of a body of literature that starts with the 1921 paper by Wright. In this section, we shall consider how well the expectation that structure can explain properties is met in the case of amorphous materials. Take the relationship between energy and stability. The global ground-state of most many-body systems is crystalline. This observation suggests that certain local structures are lower in energy than others and so expected to more frequent (i.e., favored) on cooling. In general, we propose that the utility of a local structure classification in providing a causal explanation of a property of the material must depend on how well the geometrical classification rules correlate with the property in question.

In Fig. 6, the distribution of potential energy is plotted for the most frequent local structures of the two model liquids. The distribution clearly separates about the two atomic species. This is, in part, a consequence of the difference in the number of neighbors between large and small particles. In considering the utility of the structure to “explain” physical properties, we can again imagine a null hypothesis in which the structure has no bearing on the energy of an atom. In this case, the energy distributions for the different structure would be identical and, hence, the chosen structural classification was of zero utility in accounting for the distribution of local energies. The alternative limiting case is the one characterised by structure-based distributions which are narrow and distinct (i.e., with little overlap). The capability of structure to so effectively resolve some physical properties would provide strong support for the utility of that structural measure. Utility refers to the degree to which the knowledge of the specific local structure changes our ability to predict the associated property value. Utility of structure, therefore, can be measured...
by the overlap $Q_{ij}$ of the distribution $p_i(x)$ and $p_j(x)$ of some property $x$ for two sub-populations characterised by structures $i$ and $j$. We define the overlap $Q_{ij}$ as

$$Q_{ij} = \frac{\int d x p_i(x)p_j(x)}{\sqrt{\int d x p_i^2(x)\int d x p_j^2(x)}}. \tag{4}$$

By the Cauchy-Schwarz inequality, one has

$$0 \leq Q_{ij} \leq 1. \tag{5}$$

If $P_i$ is the relative frequency of structure $i$, then we can define the weighted average overlap $Q$ by

$$Q = \frac{\sum_{i \neq j} P_i P_j Q_{ij}}{\sum_{i \neq j} P_i P_j}, \tag{6}$$

where $0 \leq Q \leq 1$. This inequality follows from the fact that Eq. (6) is a so-called convex combination of numbers between zero and unity [Eq. (5)]. The utility $U_x$ of a given choice of structural characterization in terms of its capacity to differentiate the property $X$ can then be defined as

$$U_x = 1 - Q, \tag{7}$$

reflecting the fact that perfect overlap (i.e., $Q = 1$) would correspond to all $Q_{ij} = 1$, i.e., identical distributions $p_i(x)$, which would correspond to a useless structural resolution, while zero overlap would represent an optimal utility with $U_x = 1$. In the case of the energy, we have calculated values of $U_E$ for the two atomic species separately, i.e., the $Q_{ij}$'s of Eq. (3) are only taken between Voronoi structures centred around a given type of atom. (We again remind the reader that our choice of Voronoi polyhedra to define “structure” does not express any presumption about its suitability for this purpose. Our purpose is to devise a test as to whether this choice, or any other, can “explain” the variation in stability.) The value of $U_E$ presented in Table I is the average value for the two species. For the utility $U_C$ associated with the degree of particle localization, we have considered distributions of the variable $(\Delta r^2)/k_BT$. We find the utility of the Voronoi polyhedra to account for the distribution of local energy is 0.21 and 0.15 for the CuZr and $A_{50}B_{50}$ alloys, respectively. Similar values are obtained for $U_C$. These results for $U_E$ reflect the considerable overlap of distributions that we see in Fig. 6. The Voronoi classification is clearly insufficient, on its own, to explain the range of local energies in these amorphous materials. The utility $U_E$ of the Voronoi analysis is found to systematically decrease as the concentration of the KA mixture approaches the equimolar value. This is probably a generic result arising from the increasing number of distinct polyhedra found as the equimolar concentration is approached. This trend is not observed for $U_C$.

Inspection of the energy distributions in Fig. 6 makes it clear that if we were to look at a subset of particles with sufficiently low energy, we would find them dominated by a small number of specific structures. In Fig. 7, we plot the change in the fraction of the smaller B particle structures in the $A_{50}B_{50}$ mixture when considering different sub-populations defined by an energy threshold value $E'$. We find that the fraction of some structures—(0, 2, 8, 0) and (1, 2, 5, 3)—increase as we focus on the lowest energies, while others—(0, 6, 3, 0), (1, 5, 2, 2), and (1, 3, 3, 3)—are selected against and decrease. This observation is consistent with the notion of favored local structures. The (0, 2, 8, 0) polyhedra, previously identified as

![Image](https://example.com/fig6.png)

**FIG. 6.** The distribution of potential energy for a variety of local structures (as indicated) for the CuZr mixture, with (a) Zr-centred polyhedra and (b) Cu-centred polyhedra, and for the $A_{50}B_{50}$ mixture with (c) A-centred polyhedra and (d) the B-centred polyhedra.

<table>
<thead>
<tr>
<th>TABLE I. Values of the utility $U_E$ and $U_C$, the common structure fraction $f_x$ (defined in Sec. III C below), and the diversity $D$ for CuZr and three concentrations of the KA mixture.</th>
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<tr>
<td>$\text{CuZr}$</td>
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<td>$U_E$</td>
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<td>$D$</td>
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*The data for the equimolar KA mixture were obtained from an instantaneous quench rather than a constant finite cooling rate to avoid crystallization.*
FIG. 7. Plots of the fraction of B-centred structures in A\textsubscript{80}B\textsubscript{20} as a function of the associated subpopulation of particles sampled where the subpopulation is defined by those particles with potential energy below an energy threshold $E^\ast$.

A favored structure in the KA mixture\textsuperscript{42,43} is the bicapped square anti-prism, the basis of the metastable Al\textsubscript{2}Cu crystal structure. The structural selection evident in Fig. 7 is directly related to the decrease in the diversity $D$ that we discussed in Sec. III A. This type of structural selection represents, generally, a common form of evidence presented in the literature to support the proposition that the properties of an amorphous material are governed by the stability of a small set of structures. A number of papers\textsuperscript{42–44} have reported that the $(0, 2, 8, 0)$ structure in A\textsubscript{80}B\textsubscript{20} is associated with a relaxation time that is a factor of 3-6 times slower than the average relaxation time of the liquid. An analogous observation has been reported for the $(0, 0, 12, 0)$ structure around Cu in the CuZr mixture by Cheng, Sheng, and Ma.\textsuperscript{45}

Are these types of correlations sufficient to claim a causal link between the favored structures and the physical effect? In Fig. 8, we sketch three different scenarios regarding structure-property relations. The ideal of a perfect causal resolution [Fig. 8(a)] involves a one-to-one map of each structure to a specific range of property values. This scenario is sufficient to establish that the measured structural property is the cause of the observed property. The opposite of this ideal is that the chosen structural resolution has no correlation with the property (i.e., the null hypothesis), as sketched in Fig. 8(b). A third possibility, one that better describes correlations of the type actually observed (i.e., in Fig. 7 and Refs. 40–43), is the partial resolution depicted in Fig. 8(c). Here an extreme of property values (e.g., lowest mobility and lowest energy) is associated with only one (or a small number) of the structures, while the structures themselves might contribute to a range of property values. This partial resolution as sketched in Fig. 8(c) and as demonstrated in Fig. 7 will tend to have low utility (as defined here), despite exhibiting strong structural selectivity at the extreme of the property distribution, due to the significant global overlap of the distribution of property values (as shown in Fig. 6). So what can we conclude from the observation of a scenario like Fig. 8(c)? We suggest the following: The structures selected for by the property extremae are a component of the structures responsible for the distribution of the property value but they are not a complete description of the structures responsible. In the case of the A\textsubscript{80}B\textsubscript{20} mixture, for example, the $(0, 2, 8, 0)$ B environment does indeed contribute to stability—both mechanical and kinetic—but only when some other, unmeasured and, presumably, nonlocal, structural conditions apply. The utility we have defined can, in this context, be regarded as a rough measure of the degree to which a given structural classification (like the strictly local one provided by the Voronoi analysis) discards important information.

C. Structure and the proximity of a crystalline phase

The most common use proposed for amorphous structures has been to rationalise the absence of crystallization. Exactly how this rationalisation is to be achieved is not clear. (A good example of what is required to establish a connection between structure and crystallization is provided by Taffs and Royall.\textsuperscript{46}) To examine the structure of a glass and try to infer what aspect of the structure might have contributed to the non-observation of ordering is flawed as a logical proposition, akin to trying to explain in cards why one did not get
The essential character of amorphous structure—the large multiplicity of local arrangements—has been presented here as an explicit structural measure in the form of the diversity \( D \). Evaluating \( D \) for glass forming alloys, we find \( D > 100 \), in clear contrast to crystalline states which are roughly limited to \( D < 10 \). The correlation.

In the A\(_{50}\)B\(_{50}\) mixture well before any sign of crystallinity, a non-crystalline structure.

Table I. We find that the A\(_{50}\)B\(_{50}\) KA mixture has a value of \( f_x = 0.55 \), while the large particle rich mixtures have \( f_x \)'s \( < 0.1 \). The difference in \( f_x \) correlates well with the difference in crystallization kinetics, rapid in the case of the A\(_{50}\)B\(_{50}\) mixture and slow in the A\(_{50}\)B\(_{32}\) case. The trend toward shared structures for the two species is prevalent in the A\(_{50}\)B\(_{50}\) mixture well before any sign of crystallinity, a non-trivial result. To conclude, we reiterate that (a) \( f_x \) does not require that we make any selection of specific target structures, and (b) we have established a correlation between slow crystallization and the absence of a crystal-like structure rather than to the presence of some non-crystalline structure.

IV. DISCUSSION

To assess the utility of a particular structure classification of an amorphous material requires that we entertain the possibility of the null hypothesis, i.e., that our measure of structure is not useful. If this hypothesis seems to ignore the history of successes of structure-property correlations in science, we stress that these successes generally refer to the use of the total structure as obtained from crystalline materials. In this paper, we do not challenge this position but emphasize that, in amorphous materials, the total structure is typically inaccessible and, even when available as in colloid microscopy or simulations, is simply too complex to be of intelligible use. It is the incomplete character of structural description in amorphous materials that raises the question of utility. This point is important. There is a growing body of evidence for the existence of heterogeneous dynamical and material properties of an amorphous structure. Such observations provide clear evidence of some sort of structural control without providing any clear indication of which (if any) choice of local structural classification might successfully capture this structural control and the associated structure-property correlation.

The essential character of amorphous structure—the large multiplicity of local arrangements—has been presented here as an explicit structural measure in the form of the diversity \( D \). Evaluating \( D \) for glass forming alloys, we find \( D > 100 \), in clear contrast to crystalline states which are roughly limited to \( D < 10 \). The
quantity D provides an explicit measure of the significance of any particular local structure. The observed values of D raise a number of questions. What manner of materials occupies the intermediate structural diversities, i.e., 10 < D < 100? Should a large diversity be regarded as a fundamental feature of a material or just the signature of a poor choice of structural classification? The latter question can be answered, in part, by testing a range of structural measures so that if D remains relatively constant, then one can assume that the value of D does indeed reflect some intrinsic feature of the structure. How does the diversity D change as we consider the subpopulations characterised by some restricted set of property values? In this paper, we have considered this question in the context of constraint and found a substantial reduction in diversity as we consider only structures with increasing degree of particle localization. Our analysis demonstrated that this loss of diversity could be quantitatively attributed to both species selectivity and structural selectivity. The structural selectivity takes the form of particular Voronoi polyhedra dominating the structure of the most constrained subpopulation—a result qualitatively similar to previous reports of structural significance. Using our utility index, we can qualify this observation by noting that the structural measure can have a low utility even while exhibiting this high selectivity for the extreme of a property. This is a signature of an incomplete structural descriptor. What this means is, for example, while the structure (0, 2, 8, 0) corresponds to a significant fraction of B particles in the KA mixture with low energy and low mobility, the identification of a given particle as being (0, 2, 8, 0) tells us little about its stability relative to particles with other structures. Expanding the structural measure is a non-trivial task. Machine learning has been proposed as a strategy for addressing this incompleteness. In this approach, the weighting of different types of structural data is adjusted to maximise coincidence with some selected properties, e.g., local dynamics. An unresolved issue with such combinatoric approaches is to identify exactly what use (in the sense we discuss in Sec. I) the resulting structure serves.

Based on the analysis presented here, we conclude that the Voronoi analysis is of limited utility in the description of the two alloys selected for this study. We emphasize that this conclusion refers specifically to our choice of Voronoi polyhedra and our glass forming liquids. That said, we would expect that, in liquids with a large diversity, no local measure of topology or geometry is likely to fare much better. In such situations, where do we go with structural analysis? A useful approach is to consider exactly what are the basic consequences of structure. Previously, it has been argued that constraint (i.e., particle localization), rather than the structures responsible for the localization, is sufficient to account for the rigidity of a material, ordered or amorphous. While an amorphous material may have a high diversity of local topologies, the variety of constraints experienced by particles might be much less diverse. Structural measures based on incomplete coordination shells are one choice of local structure aimed to connect constraint and structure.

V. CONCLUSION

In conclusion, we have presented three new measures of structure in amorphous materials that do not rely on intuition or prior assumption regarding the significance of special local arrangements. In treating all aspects of an amorphous structure equally, dependent only on the frequency with which they occur, we have sought to demonstrate how a new class of questions can be put to materials with the goal of moving beyond the demonstration of correlation between structure and property in amorphous materials and establishing those aspects of amorphous structure that provide quantifiable benefit in rationalising the underlying causes of observed material behavior.

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