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Recasting a model atomistic glassformer as a system of icosahedra
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I. INTRODUCTION

Among the challenges of the glass transition is to understand how solidity emerges with little apparent change in structure. Indeed, whether the glass transition has a thermodynamic (implying structural) or dynamical origin remains unclear.1,2 It has been proposed that upon cooling, icosahedral arrangements of atoms might form in supercooled liquids3 and that dynamical arrest may be related to a (geometrically frustrated) transition to a phase of such icosahedra.4 It is now possible to identify geometric motifs such as icosahedra and related locally favored structures (LFSs) using computer simulation5–15 and particle-resolved studies in colloid experiments.16–19 Further evidence of increasing numbers of LFSs upon cooling is also found in metallic glassformers.20,21

The discovery of dynamic heterogeneity, i.e., locally fast and slow regions,22,23 has spurred attempts to correlate LFS such as icosahedra with the dynamically slow regions. This has met with some success;7,8,11–14,24 however, such correlation does not by itself demonstrate a mechanism for arrest24,25 and in any case is dependent on the model system under consideration.14 A key limitation here is that direct detection of LFS and dynamic heterogeneity is only possible in the first 4-5 decades of dynamic slowing accessible to particle-resolved experiments on colloids16 and computer simulation. This compares to 14 decades of slowing corresponding to the glass transition in molecular systems and divergence of relaxation time at a putative thermodynamic transition. Clearly, any discussion of the nature of the glass transition itself requires extensive extrapolation of data. Significantly, the limit of this accessible 4-5 decades corresponds roughly to the mode-coupling temperature $T_{\text{MCT}}$ which in $d = 3$ leads to a crossover to a regime where relaxation is believed to occur in a qualitatively different fashion through cooperative behavior.2,26,27

Recently, it has become possible to access certain properties of this deeply supercooled regime. One approach is to vapor deposit onto a substrate cooled below the temperature at which the system can usually be equilibrated.28,29 Alternatively, by immobilizing or “pinning” a subset of particles, a transition to an “ideal glass” can be induced that is accessible to simulation30,31 and experiments with colloids.32 Other methods include the observation of a transition in distributions of overlaps in configuration space33,34 and so-called large deviations where trajectory sampling of moderately supercooled liquids indicates a dynamical transition to a state rich in LFS with very slow dynamics.35,36 It is also possible to decompose the system into a range of geometric motifs. Such an approach indicated that there is no thermodynamic transition.9,10 However, obtaining dynamical information in the deeply supercooled regime beyond that accessible to simulation remains a challenge.

Here, we consider a binary Lennard-Jones glassformer whose dynamics are strongly correlated with LFSs, which are icosahedra8,13,14. In particular, the occurrence of super-Arrhenius dynamics coincides with the emergence of a population of icosahedra (Figs. 1 and 2).8,13 We build on this observation and use the population of icosahedra and its dynamics to predict the behavior in the regime beyond that accessible to computer simulation. To do this, we introduce a stochastic model for the population dynamics of icosahedra which we parameterize in the regime accessible to simulation. We then use the model to obtain the population of icosahedra for all temperatures. In the simulation accessible regime, we show that the population and lifetime of the domains of icosahedra which we term mesoclusters are correlated with the super-Arrhenius dynamics. Using the calculated

Recasting a model atomistic glassformer as a system of icosahedra

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We consider a binary Lennard-Jones glassformer whose super-Arrhenius dynamics are correlated with the formation of icosahedral structures. Upon cooling, these icosahedra organize into mesoclusters. We recast this glassformer as an effective system of icosahedra which we describe with a population dynamics model. This model we parameterize with data from the temperature regime accessible to molecular dynamics simulations. We then use the model to determine the population of icosahedra in mesoclusters at arbitrary temperature. Using simulation data to incorporate dynamics into the model, we predict relaxation behavior at temperatures inaccessible to conventional approaches. Our model predicts super-Arrhenius dynamics whose relaxation time remains finite for non-zero temperature. © 2015 AIP Publishing LLC. [http://dx.doi.org/10.1063/1.4938424]
population of icosahedra, we predict the dynamical behavior of the system at all temperatures. In particular, we make the following assumptions: the dynamical behaviour of the system is democratically represented by particles in different sized mesoclusters of icosahedra and those not in icosahedra. The dynamics of particles not in icosahedra are assumed to be Arrhenius. The dynamics of particles in mesoclusters are determined from parameterisation of mesocluster lifetimes using simulation. The model predicts a super-Arrhenius dynamics comparable to the simulation data, however it indicates there is no divergence of the relaxation time at finite temperature.

Our model captures the super-Arrhenius behavior of the system with reasonable accuracy. The largest discrepancy between the predictions of the model and the simulation data occurs during the first few decades of super-Arrhenius dynamics around $0.7 \geq T \geq 0.6$. These first few decades of arrest are well-described by mode-coupling theory (MCT). MCT takes as its input two-point correlations. These are entirely neglected in our analysis which focuses on higher order correlations, and we attribute the discrepancy predominantly to our neglect of MCT.

This paper is divided as follows: we discuss the simulation details in Section II and describe our model in Section III. Results are shown in Section IV, and we conclude with a summary and discussion in Section V.

II. SIMULATION DETAILS

We simulate the Wahnström equimolar additive binary Lennard-Jones model. The size ratio is 5/6 and the well depth between all species is identical. The mass of the large particles is twice that of the small. We use molecular dynamics simulations of $N = 1372, 10976, 87808$ particles. We equilibrate for at least 100$t_\alpha$ in the NVT ensemble before sampling in the NVE ensemble. Here, $t_\alpha$ is the structural relaxation time which is determined from a stretched exponential fit to the intermediate scattering function. We identify icosahedra with the topological cluster classification (TCC) and consider those which last longer than 0.1$t_\alpha$ (the distributions of which can be seen in Fig. 3(a)) to suppress the effects of thermal fluctuations. Our simulation and analysis protocol are detailed in Refs. 13 and 38.

The Wanström mixture has been shown to crystallise to a Frank-Kasper phase. Indeed, some simulations of $N = 1372$ particles crystallised at temperatures $T \leq 0.6$; clearly evidenced by a substantial increase in the population of icosahedra at fixed temperature. These simulations were discarded. No crystallisation was observed in the larger systems.

We plot the structural relaxation time $t_\alpha$ with an Arrhenius form for $T \geq 1$ and a Vogel-Fulcher-Tamman (VFT) form for lower temperatures in Fig. 2. The VFT form reads $t_\alpha = t_{\alpha 0} \exp[\Delta / (T - T_0)]$, where the fragility parameter $\Delta = 0.799$ and the temperature at which our fit predicts a divergence of $t_\alpha$ at a temperature $T_0 = 0.456$. In Fig. 2, we also indicate the mode-coupling temperature $T_{MCT} = 0.57$, fitted across the region $0.58 \leq T \leq 1$. 

![FIG. 2. “Angell” plot of relaxation time as a function of inverse temperature. Circles: Simulation data. Dashed red line: Arrhenius. Dashed green line: Population dynamics model prediction. Solid navy line: As described by geometric frustration. Thinner, solid pale blue line: MCT fit to the data across the region $0.58 \leq T \leq 1$. Thin, solid red line: VFT fit to data $T \leq 1$. Inset: Proportion of particles identified as being part of icosahedra as a function of temperature. Circles: Simulation data. Solid red line: Fitted model. Simulations are limited to $T \geq 0.56$ (blue shaded region indicates inaccessibility). Black dashed lines correspond to $T_{VFT} = 0.456$ and $T_{MCT} = 0.57$.](image-url)
We make the assumption that any change in mesocluster size is a change in \( m \) of \( \pm 1 \). This assumption means that even large mesoclusters (which in practice might break in two) can only decrease incrementally. Thus, mesocluster scission, or indeed coalescence, is not a feature of our model. The rates at which the mesocluster size increases or decreases are given by \( g \) and \( r \), respectively. The master equation for the population dynamics model reads

\[
\begin{align*}
\dot{p}_1 &= g_0 p_0 + r_2 p_2 - [g_1 + r_1] p_1, \\
\dot{p}_m &= g_{m-1} p_{m-1} + r_{m+1} p_{m+1} - [g_m + r_m] p_m, \\
\dot{p}_M &= g_{M-1} p_{M-1} - r_M p_M,
\end{align*}
\]  

(1)

where \( p_m \) is the probability that a particle is in a mesocluster of size \( m \). We set a limit at \( m = M \) which is the largest possible mesocluster that can be formed given the relative population of particles in icosahedra, \( \phi \). Considering geometric frustration, we expect \( \phi < 1 \) and set a maximum value for the relative population \( \phi_{\text{max}} = 0.75 \). This in turn constrains the largest number of icosahedra in a mesocluster, \( M \). While the limit of the parameter \( \phi \) is chosen, rather than determined, results obtained in the range \( 0.6 \leq \phi_{\text{max}} \leq 0.9 \) exhibit only slight quantitative differences and have no impact on our conclusions.

The steady state solution for the master equation (solution when \( \dot{p}_m = 0 \) for all \( m \)) is

\[
p_m = \frac{g_{m-1}}{r_m} p_{m-1} = \frac{g_1 \cdots g_{m-1}}{r_2 \cdots r_m} p_1.
\]

(2)

If \( g_m, r_m \) are assumed constant across all \( m \) (for a given \( T \)), \( g_m = g \), and \( r_m = r \), we can denote the ratio as a “decay parameter” \( \alpha = g/r \). This results in \( p_m = a^{m-1} p_1 \), so all \( p_m \) can be determined from just two parameters. We impose \( \sum_{m=0}^{M} p_m = 1 \) and \( \sum_{m=1}^{M} p_m = \phi(T) \), where \( \phi(T) \) is the expected proportion of particles to be in icosahedra at that temperature as shown in the inset of Fig. 2 [and Fig. 5(a)]. In the high temperature Arrhenius regime \( T \geq 1 \) and at slightly lower temperatures \( (T \geq 0.7) \), the mesocluster size distribution is well described by a decaying exponential \( p_m = a^{m-1} p_1 \). The parametrisation is discussed in Section III B.

Upon cooling, at around \( T \approx 0.6 \), the number of icosahedra is sufficiently large that the mesoclusters form a percolating network (see Fig. 3; the change in slope indicates percolation at mesoclusters with \( \geq 500 \) particles, which corresponds to \( m \geq 70 \)). Now this percolation does not correspond to arrest, because the icosahedra have a limited lifetime.\(^{13} \) Indeed, the Angell plot in Fig. 2 shows no significant feature when the LFSs begin to percolate. However, percolation leads to a peak in the mesocluster size distribution, which necessitates some explicit considerations for the population dynamics model. We introduce a Gaussian-like weighting function, \( W_m(T) \), to account for the peak that forms in the distribution when percolation occurs. \( W_m(T) \) is a system-size dependent parameter that controls the location and width of the distribution peak constrained such that the largest mesocluster does not exceed \( M \). The steady state solution in the percolated regime is then \( p_m = a(T) W_m(T) p_{m-1} \).

To describe the dynamics, given the population of icosahedra, we proceed as follows. From the mesocluster size distributions, we can determine the super-Arrhenius contribution to \( \tau_\alpha \),

\[
\tau_\alpha = \tau_\alpha^{\text{Arr}} \sum_m I_m(T) p_m(T).
\]

(3)

Here, \( \tau_\alpha^{\text{Arr}} \) is the relaxation time assuming Arrhenius behaviour, extrapolated from the high-temperature \( T > 1 \) value. Each icosahedron is categorised according to the largest mesocluster it joins during its lifetime. The lifetime of an icosahedron is determined by the amount of (simulation) time \( \tau_\alpha \) during its lifetime. The lifetime of an icosahedron is determined by the amount of (simulation) time \( \tau_\alpha \).

The expression above is based on the following assumptions: (1) the dynamics of each particle in the system is represented democratically and (2) particles not in icosahedra are contained in the number of icosahedra that comprise it, i.e., how many

\[1 \equiv \text{mesocluster size increases or decreases are given by } g \text{ and } r, \text{ respectively.} \]

\[\text{The master equation for the population dynamics model reads} \]

\[\dot{p}_1 = g_0 p_0 + r_2 p_2 - [g_1 + r_1] p_1, \]

\[\dot{p}_m = g_{m-1} p_{m-1} + r_{m+1} p_{m+1} - [g_m + r_m] p_m, \]

\[\dot{p}_M = g_{M-1} p_{M-1} - r_M p_M, \]

(1)

where \( p_m \) is the probability that a particle is in a mesocluster of size \( m \). We set a limit at \( m = M \) which is the largest possible mesocluster that can be formed given the relative population of particles in icosahedra, \( \phi \). Considering geometric frustration, we expect \( \phi < 1 \) and set a maximum value for the relative population \( \phi_{\text{max}} = 0.75 \). This in turn constrains the largest number of icosahedra in a mesocluster, \( M \). While the limit of the parameter \( \phi \) is chosen, rather than determined, results obtained in the range \( 0.6 \leq \phi_{\text{max}} \leq 0.9 \) exhibit only slight quantitative differences and have no impact on our conclusions.

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\[\tau_\alpha = \tau_\alpha^{\text{Arr}} \sum_m I_m(T) p_m(T). \]

(3)

Here, \( \tau_\alpha^{\text{Arr}} \) is the relaxation time assuming Arrhenius behaviour, extrapolated from the high-temperature \( T > 1 \) value. Each icosahedron is categorised according to the largest mesocluster it joins during its lifetime. The lifetime of an icosahedron is determined by the amount of (simulation) time that has elapsed between the first and last instances of an icosahedron being identified by the TCC. The mesocluster lifetimes, \( I_m \), are the average lifetimes of the icosahedra in the corresponding size category.

The expression above is based on the following assumptions: (1) the dynamics of each particle in the system is represented democratically and (2) particles not in icosahedra have Arrhenius dynamics. We therefore attribute all super-
Arrhenius behavior to the emergence of the icosahedra. As noted above, this is motivated by the correlation between icosahedra and fragility in model glassformers. Given the populations of the mesoclusters as a function of temperature and extrapolating the dynamical trends we see, we predict $\tau_\alpha$ at temperatures far beyond those accessible to simulation (see Fig. 2).

Since the emergence of the population of icosahedra is associated with super-Arrhenius dynamics, we consider the dynamical behavior of the system by comparison with an Arrhenius relaxation time $\tau_\alpha^{\text{Arr}}$, which we assume the system would have if there were no icosahedra to form. Fig. 3(b) shows the lifetime distribution of all icosahedra whose lifetimes are $\geq 0.1\tau_\alpha$ across a range of temperatures ($0.58 \leq T \leq 2.5$) scaled by the Arrhenius time scale $\tau_\alpha^{\text{Arr}}$. The lifetime distributions collapse onto each other at temperatures $T \geq 1$ but spread out at $T \leq 1$.

The mesocluster lifetimes are modeled as a function of size (number of participating icosahedra) for a range of temperatures. We fit the mesocluster lifetimes with two linear expressions, a $T$-dependent expression for small $m$ and a $T$-independent expression for large $m$,

$$I_m = \begin{cases} 10^{k_1(T)m + k_2(T)} & \text{for} \quad m \leq m^* \\ 10^{h_1m + h_0} & \text{for} \quad m > m^*. \end{cases}$$

Here, $m^*$ is the point at which the two expressions are equal, i.e., the (feasible) solution to $(k_1(T)m + k_2(T)) - (h_1m + h_0) = 0$ ($h_0$ and $h_1$ are listed in Table I and $k_0(T)$ and $k_1(T)$ are plotted in Fig. 4 and described in Sec. III A). At high temperatures, the mesocluster lifetimes are dominated by the temperature-dependent expression corresponding to $m \leq m^*$. At low temperatures, the mesocusters are able to grow large enough to pass this “threshold” into a regime where the mesocluster lifetimes are no longer related to the temperature and are functions of size only.

### A. Mesocluster lifetime parameters

Throughout the model descriptions, we utilise a smoothing function with the following form:

$$B_i(j) = 0.5 \left( 1 + \tanh[Y(j - X)] \right).$$

where $i$ is a function indicator, $j$ is the variable, and $X,Y$ are fitted values. All functions and values are listed in Table II.

The temperature-dependent regime of the lifetime model has two parameters described as follows:

$$k_0(T) = d_1T^{-3} + d_2T^{-2} + d_3T^{-1} + d_0,$$

$$k_1(T) = \begin{cases} 0.0393B_i(\frac{1}{T}) & \text{for} \quad k_0(T) < h_0 \\ 0 & \text{for} \quad k_0(T) \geq h_0. \end{cases}$$

The coefficients for $d_i$ in the expression for $k_0$ have different values for the two ranges $T > 0.7$ and $T \leq 0.7$. These are listed in Table III.

### B. Mesocluster population dynamics parameters

Our model for $\phi$ is shown in Fig. 5(a) and described as follows (coefficients $d_i$ listed in Table IV),

$$\phi(T) = B_\phi\left(\frac{1}{T}\right) \exp\left(d_2T^{-2} + d_3T^{-1} + d_0\right) + \phi_{\text{max}}\left[1 - B_\phi\left(\frac{1}{T}\right)\right].$$

<table>
<thead>
<tr>
<th>Region</th>
<th>$d_0$</th>
<th>$d_1$</th>
<th>$d_2$</th>
<th>$d_3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$k_0(T), T &gt; 0.7$</td>
<td>-1.0256</td>
<td>0.5033</td>
<td>0.0733</td>
<td>0</td>
</tr>
<tr>
<td>$k_0(T), T \leq 0.7$</td>
<td>-202.77</td>
<td>398.73</td>
<td>-262.4</td>
<td>57.8</td>
</tr>
</tbody>
</table>
where $B_\phi(1/T)$ controls the transition to a plateau. Using simulation data on the number of particles that comprise a typical mesocluster of (arbitrary) size $m$, we infer a linear relation between the proportion of particles in icosahedra, $\phi$, and the size of mesocluster that could be formed if all the particles in icosahedra were to aggregate into a single mesocluster. We use this linear relation as our model for $M$,

$$M(\phi) = \frac{1}{7.66} \left( N\phi - 7.1 \right), \quad \text{(9)}$$

where $N$ is the total number of particles in the system (1372, 10 976, 87 808).

Our model for $p_1$ is shown in Fig. 5(b) and described below (coefficients $d_i$ listed in Table IV),

$$p_1(T) = \exp \left( d_3 T^{-3} + d_2 T^{-2} + d_1 T^{-1} + d_0 \right). \quad \text{(10)}$$

The decay parameter $a$ is plotted in Fig. 6(b) and described as follows (coefficients $d_i$ listed in Table IV):

$$a(T) = B_a(T) \exp \left( d_2 T^{-2} + d_1 T^{-1} + d_0 \right) + 0.6 \left( 1 - B_a(T) \right), \quad \text{(11)}$$

where $B_a(T)$ is a smoothing function (Eq. (5)).

The maximum in $a(T)$ occurs at percolation. Beyond this ($T < 0.6$), the system is dominated by the large percolating mesocluster (large $m$) and small-mesocluster (small $m$) effects that become increasingly negligible. In this simulation inaccessible regime, we assign a fixed value to $a(T)$ for simplicity.

The function $W_m(T)$ has a Gaussian-like form

$$W_m(T) = 1 + B_w(m) \exp \left( -\frac{(m - \mu)^2}{2\sigma^2} \right). \quad \text{(12)}$$

### Table IV. Fitted parameter values for all population model components.

<table>
<thead>
<tr>
<th>Fctn.</th>
<th>$N$</th>
<th>$d_0$</th>
<th>$d_1$</th>
<th>$d_2$</th>
<th>$d_3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\phi(T)$</td>
<td>1 372</td>
<td>-14.706</td>
<td>10.543</td>
<td>-1.696</td>
<td></td>
</tr>
<tr>
<td></td>
<td>10 976</td>
<td>-14.922</td>
<td>11.075</td>
<td>-1.938</td>
<td></td>
</tr>
<tr>
<td></td>
<td>87 808</td>
<td>-14.697</td>
<td>10.482</td>
<td>-1.627</td>
<td></td>
</tr>
<tr>
<td>$p_1(T)$</td>
<td>1 372</td>
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<td>1.588</td>
<td>9.737</td>
<td>-4.673</td>
</tr>
<tr>
<td></td>
<td>10 976</td>
<td>-12.932</td>
<td>3.053</td>
<td>8.079</td>
<td>-4.109</td>
</tr>
<tr>
<td>$a(T)$</td>
<td>1 372</td>
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<td>13.104</td>
<td>-4.091</td>
<td></td>
</tr>
<tr>
<td></td>
<td>10 976</td>
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<td>12.914</td>
<td>-3.991</td>
<td></td>
</tr>
<tr>
<td></td>
<td>87 808</td>
<td>-10.750</td>
<td>13.182</td>
<td>-4.129</td>
<td></td>
</tr>
<tr>
<td>$\mu(\phi)$</td>
<td>1 372</td>
<td>18.546</td>
<td>-165.35</td>
<td>794.51</td>
<td></td>
</tr>
<tr>
<td></td>
<td>10 976</td>
<td>16.780</td>
<td>-99.817</td>
<td>3439</td>
<td></td>
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<tr>
<td></td>
<td>87 808</td>
<td>-173.14</td>
<td>2537</td>
<td>1013.9</td>
<td></td>
</tr>
<tr>
<td>$\sigma(\phi)$</td>
<td>1 372</td>
<td>4.5</td>
<td>0.18</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>10 976</td>
<td>5.3</td>
<td>0.17</td>
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</tr>
<tr>
<td></td>
<td>87 808</td>
<td>7</td>
<td>0.1</td>
<td></td>
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</tr>
</tbody>
</table>
where $\mu$ and $\sigma$ are temperature-dependent parameters that control the location and width of the peak in the distribution, and $G$ is chosen for each $T$ in order to satisfy $\sum_{m=1}^{M} p_m = \phi(T)$. The parameters $\mu$ and $\sigma$ only apply to $T < 0.7$ and are only defined in this range.

The parameters $\mu$ and $\sigma$ are plotted in Figs. 6(c) and 6(d) respectively and given by (coefficients $d_i$ listed in Table IV)

$$\mu(\phi) = B_\mu(\phi) \left( d_2 \phi^2 + d_1 \phi + d_0 \right) + \left( \frac{0.75N - 7.1}{7.66} \right) \left( 1 - B_\mu(\phi) \right), \quad (13)$$

$$\sigma(\phi) = \exp \left( d_0 B_\sigma(\phi) + d_1 \right), \quad (14)$$

with smoothing functions $B_\mu(\phi)$ and $B_\sigma(\phi)$, and $N$ the total number of particles in the system.

**IV. RESULTS**

Our approach is based on the observation that at high temperature, there is Arrhenius behavior in the dynamics and very few icosahedra, but at the onset temperature ($T_{on} \approx 11^3$), there is a crossover to super-Arrhenius behavior which is accompanied by the emergence of a population of icosahedra found in mesoclusters which grows upon supercooling.\(^7,8,13\)

These two dynamical regimes are indicated in the Angell plot in Fig. 2 and the population of icosahedra is shown in the inset.

Motivated by these observations, we propose that the correlation between icosahedra and super-Arrhenius dynamics continues to lower temperature. This gives us a means to predict the relaxation behavior of the system at arbitrary temperature under the assumptions made in constructing the model (Section III). The dynamical behavior is predicted from measurements of cluster lifetimes. The global dynamical behavior is shown in the Angell plot in Fig. 2 and discussed in Section V. Below we consider some further dynamical features of the model.

In Fig. 7(a), we compare the results of the population dynamics model ($p_m = a(T) W_m(T) p_{m-1}$) for the size distribution of mesoclusters with simulation data. In the inset, we show the predicted mesocluster distributions for temperatures inaccessible to simulations. The population distribution model results for different system sizes at a fixed temperature ($T = 0.58$) are shown in Fig. 7(b). Larger systems allow for larger mesoclusters, resulting in the location of the peak having system-size dependence. However, the systems still percolate at the same temperature ($T \approx 0.6$).

In the simulation accessible regime, we can directly measure the dynamical properties of the mesoclusters. These are shown in Fig. 8(a) which plot the mean lifetime of icosahedra as a function of mesocluster size. This increases strongly with $m$, while retaining some temperature dependence. The mesocluster lifetime model [Eq. (4)] for different system sizes (for fixed $T = 0.58$) is shown in Fig. 8(b). This shows the behaviour of the model predictions. Note the cusp-like feature shown in this log-log representation which is due to the meeting point $m^*$ between the functions for large and small $m$ in Eq. (4) is described in Section III.

The mesocluster lifetimes differ with system size which we explain as follows. Let us assume that in the thermodynamic limit, each mesocluster size has a fixed lifetime. We imagine that in a simulation box, percolating mesoclusters of a given size correspond to a distribution of larger mesoclusters in the thermodynamic limit, rather than being a fixed size. So, in the thermodynamic limit, these larger mesoclusters would have a distribution of lifetimes, but below we assign a single lifetime for each sampled size. Crucially, this effect varies with system size.

In Fig. 9, we show the structural relaxation time our model predicts for different system sizes across the whole range of $T$ (Eq. (3)). Despite the strong system size dependence in both the mesocluster population distributions and lifetimes, the resulting relaxation times are scarcely effected by the system size.

**V. SUMMARY AND DISCUSSION**

We have presented an approach to predict the dynamics of a glass forming liquid at arbitrary temperatures. This we have done by decomposing a model glassformer into an
Mesocluster lifetimes. Simulation data (points) fitted with the lifetime model (lines) as described in Equation (4). Small mesoclusters (lower values of m) have highly T-dependent lifetimes, but at some large enough mesocluster size, the lifetimes become T-independent and are determined only by the mesocluster size. Inset: The predicted mesocluster lifetimes for some temperatures inaccessible to simulations. (b) The resulting lifetime model for a fixed temperature, T = 0.58, for different system sizes. The sudden increase in slope in (b) occurs at the point m*, where the T-dependent description crosses to the T-independent curve.

Effective system of mesoclusters of locally favored structures whose population is described by a population dynamics model which we parameterize with results from simulations. The lifetime of mesoclusters of LFS are also parameterized with simulations. Under the assumptions above and that the super-Arrhenius dynamics can be attributed to the population of particles in mesoclusters of icosahedra, our model predicts dynamical behaviour at arbitrary temperature. In its present form, the model predicts that there is no thermodynamic glass transition. We have considered different system sizes and find that while the population and lifetime components of the model are strongly system size dependent, the resulting relaxation time is scarcely dependent on system size.

Whether or not there is a thermodynamic transition, in the sense of a divergence in relaxation time scales at finite temperature, boils down to whether the lifetime of icosahedra in mesoclusters, \( l_m \), diverges. In Fig. 8(a), we see that it does not. Although our data are compatible with dynamical divergence of \( l_m \), — that is to say \( l_m \) can be fitted in the regime accessible to simulation such that it diverges at finite temperature — better fits are obtained with non-divergent behaviour.

Now since the population dynamics model itself does not exhibit a phase transition, perhaps one could argue that it is natural that we do not find a divergence in relaxation time. One might imagine that population dynamics models in which a phase transition is encountered would lead to dynamical divergence.\(^{41}\) It is also possible that further refinement may fit the relaxation time data better than our current approach [Fig. 2], which might provide further insight into the question as to whether there is a thermodynamic glass transition. At this stage, we observe that our model which predicts no thermodynamic transition actually over-estimates the super-Arrhenius nature of the dynamics. This would lend support to the observation that within this framework, there should be no transition as also found in other treatments of LFS.\(^{4,9,10}\)

We also plot in Fig. 2 predications from geometric frustration.\(^{4}\) Here \( \tau(T) = \tau_\alpha \exp (\Delta E(T) + E_\infty / k_B T) \), where \( E_\infty \) is the Arrhenius contribution. Below the onset temperature \( T_{\text{on}} \), the super-Arrhenius contribution \( \Delta E = 0 \), for \( T < T_{\text{on}} \), \( \Delta E(T) = B k_B T c (1 - T / T_c)^\psi \), where \( B = 650, \psi = 8 / 3, \) and \( T_c = 0.65 \). We see that the results, also predicated on icosahedra, seem to describe the dynamical behavior in a similar way to our model. It is possible that certain aspects of geometric frustration are captured by our approach.

One explanation for the rather strong increase in \( \tau_\alpha \) exhibited in Fig. 2 might be that our model does not include mesocluster scission or coalescence, because the mesocluster size increases/decreases only by one. Larger changes in mesocluster size might lead to closer agreement with simulation data, but would not change the picture in a qualitative fashion. System sizes for simulations which represent certain properties of deeply supercooled systems are small,\(^{33,35,36}\) but it is tantalizing to consider parameterizing the model with such data. Alternatively, one can consider how the network geometry might be influenced by certain scaling properties near an assumed transition.\(^{42}\)

Our model underestimates the initial increase in structural relaxation time in the dynamical regime where it is well described by MCT (Fig. 2). It is tempting to suggest that this is related to our emphasis on icosahedra in describing the dynamical slowdown. In the temperature regime in question, LFSs (icosahedra) are relatively few in number and there is
no percolating network. Moreover, it is possible that in this regime, dynamical slowdown may be dominated by lower-order correlations such as the 2-body and 3-body. These arguments are supported by Banerjee et al.43 and Nandi et al.44 who have suggested that two-point based relaxation times strongly increase before higher-order contributions. In Fig. 2, we show the MCT fit which describes the simulation data accurately for 0.7 ≥ T ≥ 0.6 but diverges at $T_{\text{MCT}} = 0.57$.45 One imagines that better agreement might be obtained by including contributions from MCT in this dynamical range, noting that these can be systematically extended at least to 4th order.46 At deeper supercoolings where MCT diverges, our population dynamics model presumably captures other dominant relaxation pathways absent from MCT.27 at least for the Wahnström model. Thus, it is in the deeply supercooled regime beyond MCT whose dynamics remain inaccessible to the particle-resolved techniques of computer simulation and colloid experiment where our approach has the most to contribute. Other possibilities to explain the discrepancy at weak supercooling ($T > T_{\text{MCT}}$) include amorphous order distinct from icosahedra. The correlation of the dynamics with the icosahedra is high in the Wahnström model, but it is not perfect.14 Considerations from other work 24,47 suggest that other contributions from the structure may also contribute to the dynamics.20

Before closing, we comment on the generality of our results. Recently,14 a number of models have been compared. Of those, the Wahnström model considered here exhibits the strongest correlations between structure and dynamics, so one might expect this to be most likely to exhibit a structure-based transition at finite temperature. That our analysis hints towards no such transition therefore suggests the same should hold at least for the range of models considered.14 Our approach may also be used to optimise metallic glassformers such as Cu$_2$Zr$_{1-x}$. Like the Wahnström model, these materials are well known to exhibit correlations between non-Arrhenius dynamics and the emergence of icosahedra.21 These models also exhibit networks of icosahedra, whose emergence seems similarly related to the crossover to super-Arrhenius dynamics as the Wahnström model we consider.28,49

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