

## Spatial correlations of mobility and immobility in a glass-forming Lennard-Jones liquid

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Using extensive molecular dynamics simulations of an equilibrium, glass-forming Lennard-Jones mixture, we characterize in detail the local atomic motions. We show that spatial correlations exist among particles undergoing extremely large (“mobile”) or extremely small (“immobile”) displacements over a suitably chosen time interval. The immobile particles form the cores of relatively compact clusters, while the mobile particles move cooperatively and form quasi-one-dimensional, stringlike clusters. The strength and length scale of the correlations between mobile particles are found to grow strongly with decreasing temperature, and the mean cluster size appears to diverge near the mode-coupling critical temperature. We show that these correlations in the particle displacements are related to equilibrium fluctuations in the local potential energy and local composition. [S1063-651X(99)06309-6]

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### I. INTRODUCTION

The bulk dynamical properties of many cold, dense liquids differ dramatically from what might be expected from extrapolation of their high-temperature behavior [1]. For example, many liquids cooled below their melting temperature exhibit rapid non-Arrhenius increases of viscosity and relaxation times with decreasing temperature, and two-step, stretched exponential decay of the intermediate scattering function  $F(\mathbf{q}, t)$ . Such behavior is often discussed as a “signature” of the approach to the glass transition. It has long been a central goal of theories of the glass transition to account for these bulk phenomena in terms of the microscopic dynamical motions of the molecules of the liquid. As a consequence, computer simulations of supercooled liquids, in which this microscopic information is immediately available, are increasingly used to complement theoretical and experimental efforts. In particular, simulations in recent years have focused on the study of “dynamical heterogeneity” to understand the microscopic origin of slow dynamics and stretched exponential relaxation in glass-forming liquids [2–11].

Recently we reported the observation of dynamical heterogeneity [10] and also cooperative molecular motion [11] in extensive molecular dynamics simulations of a supercooled Lennard-Jones (LJ) mixture. These spatially correlated dynamics were observed in a regime of temperature  $T$ , density  $\rho$  and pressure  $P$  for  $T$  above the dynamical critical temperature  $T_c$  obtained [12,13] from fits by the ideal mode coupling theory (MCT) [14] to other data on the same system. The principal goals of the present paper are twofold: (1)

To test directly for spatial correlations of particles assigned (according to their displacement over some time) to subsets of extreme mobility or immobility and (2) to establish connections between this “dynamical heterogeneity” and local structure.

This paper is organized as follows. In Sec. II we present relevant background information and in Sec. III we describe the model and computer simulation techniques. In Sec. IV we group particles into subsets according to the maximum displacement they achieve on the time scale defined in Sec. IV, and show that particles of extremely high or low displacement are spatially correlated. From this we are able to identify a length scale that grows with decreasing  $T$ . In Sec. V we show that fluctuations of the local mobility are correlated to fluctuations of the potential energy, or alternatively to fluctuations in the local composition of the liquid. In Sec. VI we examine certain time dependent quantities associated with the observed dynamical heterogeneity, and finally in Sec. VII we conclude with a discussion.

### II. BACKGROUND

It has been proposed that the stretched exponential behavior exhibited by the long time relaxation of  $F(\mathbf{q}, t)$  can be attributed to a sum of many independent local exponential relaxations with different time constants, i.e., to a distribution of relaxation times [15,16]. This interpretation is one form of the so-called “heterogenous” scenario for relaxation [6,15,17–21]. A number of recent experiments [17–19] have shown that in liquids such as orthoterphenyl and polystyrene within 10 K of their glass transition temperature  $T_g$ , subsets of molecules rotate slowly relative to the rest of the molecules on time scales long compared with collision times, but shorter than the relaxation time of density fluctuations. These liquids were thus termed “dynamically heterogenous.” None of these experiments were able to explicitly demonstrate whether slow molecules are spatially correlated, but

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typical distances over which slow molecules may be correlated were inferred [17].

There have been numerous attempts to indirectly measure a characteristic length scale over which molecular motions are correlated at the glass transition both in experiments [22–25] and in simulations [3,26]. Donth [22] relates the distribution of relaxation times in systems approaching their glass transition to equilibrium thermodynamic fluctuations having a characteristic size of  $\approx 3$  nm at  $T_g$ . Thermodynamic measurements on orthoterphenyl [23], and dielectric measurements on salol [24], *N*-methyl- $\epsilon$ -caprolactan and propylene glycol [25], showed a shift in  $T_g$  due to confinement in pores up to 10 nanometers in diameter. Mountain [3] showed that the size of regions that support shear stress in a simulation of a glass-forming mixture of soft spheres grows with decreasing temperatures. Monte Carlo simulations of polymer chains in two dimensions demonstrated strong finite size effects on diffusion [26]. A number of experiments and simulations on polymers confined to thin films all found a shift of  $T_g$  due to confinement [27–31]. These effects have all been attributed to the presence of cooperatively rearranging regions that grow with decreasing  $T$ . However, the origin of this characteristic length has never been shown explicitly. In particular, the connection of the characteristic length to a cooperative mechanism of molecular motion has not been experimentally demonstrated.

The intuitively appealing picture of cooperative molecular motion was proposed in 1965 by Adam and Gibbs [32]. In their classic paper, they proposed that significant molecular motion in a cold, dense fluid can only occur if the molecules rearrange their positions in a concerted, cooperative manner. They postulated that a glass-forming liquid can be viewed as a collection of independently relaxing subvolumes within which the motion of the particles is cooperative. As the temperature of the liquid is lowered, the number of particles involved in cooperative rearrangements increases. If structural relaxation occurs through the cooperative rearrangement of groups of molecules, the liquid observed over a time scale shorter than the structural relaxation time will appear as a collection of regions of varying mobility. These predictions can be tested by selecting subsets of molecules that relax slower (or faster) than the average, and determining whether the molecules in a subset are randomly scattered through the sample or tend to cluster in a characteristic way.

The explicit connection between dynamical heterogeneity and cooperative motion is only recently being investigated experimentally in detail [24]. However, there have been a number of recent computational investigations addressing these issues. For example, Muranaka and Hiwatari [2] showed that displacements of particles measured over a time scale of the order of 5 collision times are correlated within a range of about two interparticle distances in a two-dimensional binary mixture of soft disks below the freezing point. Wahnström [33] showed that hopping processes in a strongly supercooled binary mixture are cooperative in nature. Hurley and Harrowell [4] identified fluctuating local mobilities in a supercooled two-dimensional (2D) soft-disk system, and showed an example of correlated particle motion on a time scale of the order of 20 collision times. Mountain [3] demonstrated similar correlated particle motion in a 2D supercooled Lennard-Jones mixture. By examining the time

at which two neighboring particles move apart in 2D and 3D simulations of a supercooled soft-sphere mixture, Yamamoto and Onuki demonstrated the growth of correlated regions of activity [5]. They further studied the effect of shear on these regions [5], and showed that the size of the regions diminished in high shear. The clusters of “broken bonds” (denoting pairs of neighboring particles that separate beyond the nearest neighbor distance) identified in that work are similar in some respects to the clusters of highly mobile particles in a 3D binary Lennard-Jones liquid reported previously by us [10], and described in detail in the present paper. The connection between the clusters of Ref. [10], which demonstrate a form of dynamical heterogeneity, and cooperative particle motion, was shown in Ref. [11].

### III. SIMULATION DETAILS

We performed equilibrium molecular dynamics (MD) simulations of a binary mixture (80:20) of  $N = 8000$  particles in three dimensions. The simulations are performed using the LAMMPS molecular dynamics code [34] which was designed for use on distributed memory parallel machines. LAMMPS partitions particles (atoms or molecules) across processors via a spatial decomposition [35] whereby each processor temporarily “owns” particles in a small fixed region of the simulation box. Each processor computes the motion of its particles and exchanges information with neighboring processors to compute forces and allow particles to migrate to new processors as needed.

The 6400 particles of type A and 1600 particles of type B interact via a 6-12 Lennard-Jones potential,

$$V_{\alpha\beta}(r) = 4\epsilon_{\alpha\beta} \left[ \left( \frac{\sigma_{\alpha\beta}}{r} \right)^{12} - \left( \frac{\sigma_{\alpha\beta}}{r} \right)^6 \right], \quad (3.1)$$

where  $\alpha\beta \in \{A, B\}$ . The interaction forces between particles are zero for all  $r > r_c = 2.5\sigma_{AA}$ . Both types of particles are taken to have the same mass  $m$ . The Lennard-Jones interaction parameters  $\epsilon_{\alpha,\beta}$  and  $\sigma_{\alpha,\beta}$  for this mixture are  $\epsilon_{AA} = 1.0$ ,  $\epsilon_{AB} = 1.5$ ,  $\epsilon_{BB} = 0.5$ ,  $\sigma_{AA} = 1.0$ ,  $\sigma_{AB} = 0.8$ ,  $\sigma_{BB} = 0.88$ . Both the relative concentration of particle types and the interaction parameters were chosen to prevent demixing and crystallization [12]. Throughout this paper, lengths are defined in units of  $\sigma_{AA}$ , temperature  $T$  in units of  $\epsilon_{AA}/k_B$ , and time  $t$  in units of  $\sqrt{\sigma_{AA}^2 m / \epsilon_{AA}}$ .

The simulations for each state point  $(P, T, \rho)$  are performed in three stages. First, a constant NPT adjustment run is performed by coupling the system to stochastic heat and pressure baths to bring the system from a nearby state point (usually the previously simulated state point) to the desired state point [36]. Second, a constant NVT equilibration run is performed to test for unwanted drifts in pressure  $P$  or potential energy  $U$  [37]. If no drift is observed, the final state of the system is considered to represent an equilibrium state of the system. Third, a constant NVE data-gathering run is performed using the final equilibrated state obtained from the second stage, and snapshots containing the particle coordinates and velocities are taken at logarithmic time intervals during the run. In this stage, the equations of motion are integrated using the velocity Verlet algorithm with a step size of 0.0015 at the highest temperature, and 0.003 at all

TABLE I. Temperature  $T$ , pressure  $P$ , and density  $\rho$  of the nine state points simulated.

Run	$T$	$P$	$\rho$
1	0.5495	0.4888	1.0859
2	0.5254	1.0334	1.1177
3	0.5052	1.4767	1.1397
4	0.4899	1.8148	1.1553
5	0.4795	2.0488	1.1651
6	0.4737	2.1746	1.1705
7	0.4685	2.2959	1.1757
8	0.4572	2.5490	1.1856
9	0.4510	2.6800	1.1910

other temperatures. All quantities presented here are calculated in this third stage. The analysis is performed by post-processing the snapshot files, which number as many as several thousand for the lower temperatures.

We simulated nine state points along a path in  $P, T, \rho$  that is linear when projected in the  $(P, T)$  plane. This path was chosen so that the system would approach, from high temperature, the mode-coupling critical point  $T_c = 0.435$ ,  $P_c = 3.03$ ,  $\rho_c = 1.2$  [12] along a path different from that used in Ref. [12]. Table I shows the values of  $P, T$ , and  $\rho$  for each state point.

For state points far above  $T_c$  (e.g. runs 1–5), the data-gathering runs required more CPU time than the equilibration. For state points nearer  $T_c$ , the equilibration stage was the most time consuming. In these cases the NVT stage of the simulations showed a slight drift of the pressure over time. To shorten the time required for equilibration, we estimated a new volume or temperature to create a nearby state point that we expected would be very nearly equilibrated with the current particle configuration. Then we instantaneously scaled the positions or velocities of the system (i.e., adjusted the volume or temperature) and began another constant NVT run to test for equilibration. By iterating this procedure a few times we were able to find an equilibrated state point within 0.03% of the desired  $P$  and  $T$  at low temperatures. At the lowest  $T$  studied ( $T = 0.4510$ ), the total run time following equilibration is  $1.2 \times 10^4$  time units. Thus, assuming Argon values for the parameters in Eq. (3.1), the data presented here extend up to 25.8 ns.

In previous papers [12,13], the mode-coupling theory developed for supercooled liquids by Götze and Sjögren was tested using the same LJ potential in a regime of  $P, T$ , and  $\rho$  similar but not identical to that simulated here. The simulations performed in the present work extend from a point in the phase diagram where two-step relaxation begins to emerge, down to a state point that is within approximately 4% of  $T_c$ . Over this range, we find that the structural ( $\alpha$ ) relaxation time  $\tau_\alpha$  increases by 2.4 orders of magnitude. In the present system, we find that excellent power law fits are obtained for both  $\tau_\alpha$  and the self-diffusion coefficient  $D$  with singular temperature  $T_c \approx 0.435$  and exponents  $\gamma_\tau \approx -2.8$  and  $\gamma_D \approx 2.13$ .

#### IV. ANALYSIS OF SPATIAL CORRELATIONS OF PARTICLE DISPLACEMENTS

In this paper, we are interested in whether spatial correlations exist between particles that exhibit either extremely

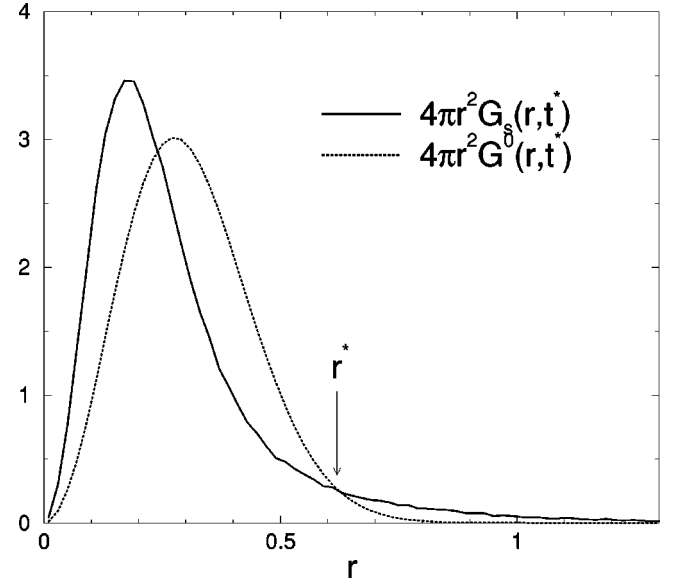


FIG. 1. Solid line:  $4\pi r^2 G_s(r, t)$  of the  $A$  particles at  $t = 155.5$  at  $T = 0.4510$ . Dashed line: Gaussian approximation calculated using the measured  $\langle r^2(t) \rangle$  for the same time.

large or extremely small displacements over some time interval. To determine this, we must first define the time interval over which the particle displacements will be monitored. Obviously, displacements may be monitored over any time interval, from the ballistic regime to the diffusive regime. To see whether there is a natural time scale on which the particle displacements might exhibit a particularly strong correlation we have studied the deviations from the Gaussian form of the self-part of the time-dependent van Hove correlation function  $G_s(r, t)$  [38].  $G_s(r, t)$  is Gaussian both in the short time (for times shorter than one collision time) and in the long time regime, when diffusion sets in. Between these two limits an intermediate time interval exists during which  $G_s(r, t)$  has a tail extending to (and beyond) one interparticle distance, that greatly exceeds what is expected on the basis of the Gaussian approximation. The quantity  $4\pi r^2 G_s(r, t)$ , which gives the number of particles located a distance  $r$  from their original position at time  $t$ , is shown in Fig. 1 for the  $A$  particles at the lowest  $T$  for a time chosen in this intermediate time regime. Also shown is the Gaussian approximation  $4\pi r^2 G_0(r, t)$ .

Among all the possible measures of the degree of deviation from the Gaussian form of  $G_s(r, t)$ , the “non-Gaussian” parameter [39],

$$\alpha_2(t) = \frac{3\langle r^4(t) \rangle}{5\langle r^2(t) \rangle^2} - 1, \quad d=3 \quad (4.1)$$

involves the lowest moments of  $G_s(r, t)$ . In Eq. (4.1),  $\alpha_2(t)$  is defined so as to be zero if  $G_s(r, t)$  is Gaussian. It has been shown [10] that  $\alpha_2(t)$  in this system is zero at short times, then becomes positive, exhibits a maximum, and finally goes to zero at long times. As  $T$  decreases, the position of the maximum  $t^*$  shifts towards longer times, and the height of the maximum  $\alpha_2^*$  increases. In Fig. 2 we show the mean square displacement (MSD)  $\langle r^2(t) \rangle$  and  $\alpha_2$  as a function of  $t$ . For all  $T$ , we find that  $t^*$  corresponds to times in the late-

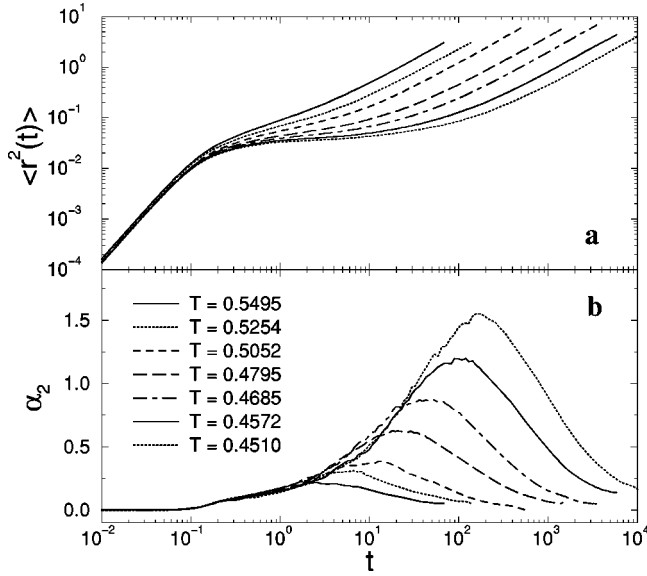


FIG. 2. (a) Mean square displacement  $\langle r^2(t) \rangle$  of the A particles vs time for several values of  $T$ . (b) Non-Gaussian parameter  $\alpha_2(t)$  vs time for the same values of  $T$  as in (a).

$\beta$ /early- $\alpha$  relaxation regime, at the end of the plateau in the mean square displacement. Note that  $t^*$  is orders of magnitude larger than the microscopic “collision time” [40]  $\tau$ ; for example, at  $T=0.4510$ ,  $t^*=155.5$  and  $\tau=0.09$ . The MSD and non-Gaussian parameter for the B particles (not shown) exhibit qualitatively the same time dependence as shown in Fig. 2, but the diffusive regime is reached at shorter times, and the diffusion constant is larger, than for the A particles [12]. This difference can be explained by the different sizes of the A and B particles and by the fact that the interaction constant  $\epsilon_{BB}$  is smaller than  $\epsilon_{AA}$ .

The interval from zero to  $t^*$  provides a convenient choice over which to monitor the particle displacements and study their correlations because (i) since  $t^*$  is the time at which the distribution of particle displacements is broadest, it may also be when the liquid is likely to be most “dynamically heterogeneous” and (ii)  $t^*$  is well defined and easily calculated. Thus, throughout this paper we will use the time window from zero to  $t^*$  as the time interval over which the particle displacements are calculated, and over which we investigate dynamical heterogeneity.

In Refs. [10] and [11], we defined the magnitude of the displacement  $\mu_i(t, t^*) \equiv |\mathbf{r}_i(t^* + t) - \mathbf{r}_i(t)|$  of particle  $i$  in a time interval  $t^*$ , starting from its position at an arbitrarily chosen time origin  $t$ , as a measure of the mobility of the  $i$ th particle. At  $t^*$ , the distribution of  $\mu_i$  values is given by the self part of the van Hove correlation function  $G_s(r, t^*)$ , where  $r \equiv \mu$  (see Fig. 1). A subset of “mobile” particles was defined by selecting all the particles that in the interval  $[0, t^*]$  had traveled beyond the distance  $r^*$  where  $G_s(r, t^*)$  exceeds  $G^0(r, t^*)$ . With this definition, “mobile” particles are those that contribute to the long tail of the van Hove distribution function at the time  $t^*$  (see Fig. 1). In Refs. [10,11], it was shown that mobile particles selected according to this rule tend to cluster [10], and move cooperatively [11]. This definition of mobility given by the magnitude of

particle displacement is thus sufficient to establish the phenomena of both dynamical heterogeneity and cooperative motion.

Intuitively, we think of immobile particles as those particles that are trapped in cages formed by their neighbors. Nevertheless, particles do not sit at one position; they essentially “oscillate” back and forth within the cage formed by their neighbors. To study correlations between the most immobile particles, we need a definition of mobility that allows us to select the particles for which the amplitude of this oscillation (i.e., the maximum displacement of the particle) is the smallest. In this paper, we therefore define the mobility  $\mu_i(t, t^*)$  of the  $i$ th A particle as the *maximum* distance reached by that particle in the time interval  $[t, t + t^*]$ :

$$\mu_i(t, t^*) \equiv \max_{t' \in [0, t^*]} \{ |\mathbf{r}_i(t' + t) - \mathbf{r}_i(t)| \}. \quad (4.2)$$

This new definition of mobility, which we use throughout this paper, allows us to examine different subsets of particles, from the most to the least mobile, in the same way. As a compromise between examining the most extreme behavior and including enough particles to obtain good statistics when examining their spatial correlation (i.e., maximizing the signal-to-noise ratio), we will examine the 5% most mobile and 5% least mobile particles. Thus, we define as “mobile” the 5% of all particles having the highest values of  $\mu(t, t^*)$ , and “immobile” the 5% having the lowest value. Note that this new definition of mobility does not qualitatively change the results obtained previously in [10,11], provided that the new definition selects approximately the same fraction of the sample as the definition previously used (approximately 5.5% in Ref. [10]).

The subsets of mobile particles selected using the definition of Ref. [10] and that used here have a large overlap, since particles that have moved relatively far at some time in the interval  $[0, t^*]$  are likely to remain relatively far at the end of the interval. However, subsets of immobile particles selected with the two different rules do not have as large an overlap, since a particle with a small displacement at some time may have previously traveled far, and then returned to its original position. The probability distribution of  $\mu$  values  $4\pi\mu^2 P(\mu, t)$  at  $t=t^*$  is shown in Fig. 3. For comparison, the probability distribution  $4\pi r^2 G_s(r, t^*)$  is also shown. Note that, although at  $t^*$  particles can be found arbitrarily close to their position at  $t=0$ ,  $P(\mu, t^*)$  is zero for  $\mu < 0.17$ .

The tendency for particles of extremely high or low mobility to be spatially correlated may be measured by calculating static pair correlation functions between particles contained within these subsets. Let us define the pair correlation function  $g_{\alpha\beta}(\mathbf{r})$  ( $\alpha \neq \beta$ ) between particles of species  $\alpha$  and particles of species  $\beta$  as

$$g_{\alpha\beta}(\mathbf{r}) = \frac{V}{N_\alpha N_\beta} \left\langle \sum_{i \in \alpha} \sum_{j \in \beta} \delta(\mathbf{r} + \mathbf{r}_j - \mathbf{r}_i) \right\rangle \quad (4.3)$$

and for like species

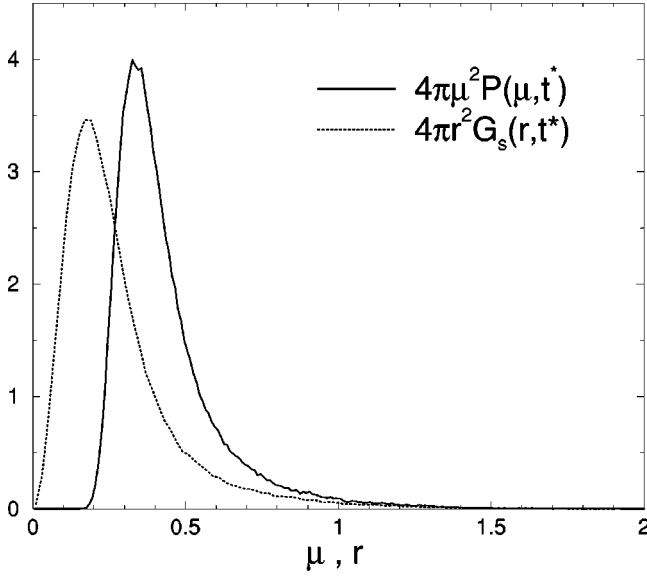


FIG. 3. Probability distribution  $4\pi\mu^2P(\mu, t^*)$  (solid line) of a particle having a maximum displacement of magnitude  $\mu$  at  $t^*$ . For comparison, the distribution  $4\pi r^2G_s(r, t^*)$  is also shown (dotted line).

$$g_{\alpha\alpha}(\mathbf{r}) = \frac{V}{N_\alpha(N_\alpha - 1)} \left\langle \sum_{i=1}^{N_\alpha} \sum_{\substack{j=1 \\ j \neq i}}^{N_\alpha} \delta(\mathbf{r} + \mathbf{r}_j - \mathbf{r}_i) \right\rangle. \quad (4.4)$$

$N_\alpha(N_\beta)$  is the total number of particles of species  $\alpha$  ( $\beta$ ). With this normalization,  $g_{\alpha\beta}(\mathbf{r})$ ,  $g_{\alpha\alpha}(\mathbf{r})$  converge to unity for  $r \rightarrow \infty$  in the absence of long range correlations. Assuming rotational invariance, the correlation functions do not depend on the direction of the vector  $\mathbf{r}$ , but only on the distance  $r = |\mathbf{r}|$ . We have calculated  $g_{AA}(r)$ ,  $g_{AB}(r)$ , and  $g_{BB}(r)$  and the corresponding structure factors for the present liquid, and find that they are qualitatively identical to those calculated in Refs. [12,13]. In particular, these bulk functions show no tendency for long range density or composition fluctuations over the range of temperatures and densities investigated.

In contrast, Fig. 4 shows the pair correlation function  $g_{MM}(r)$  between mobile  $A$  particles for four different temperatures.  $g_{MM}(r)$  is defined by Eq. (4.4) with the sum restricted to the mobile  $A$  particles. For all  $T$ ,  $g_{MM}(r)$  is appreciably higher than the average  $g_{AA}(r)$  (not shown) for all  $r$ . The ‘‘excess’’ correlation given by the ratio  $\Gamma(r) = [g_{MM}(r)/g_{AA}(r)] - 1$  is also plotted as a function of  $r$  in Fig. 4. With the exception of the excluded volume sphere of the LJ potential,  $\Gamma(r) > 0$  at intermediate distances and converges to zero for large  $r$ . It is clear from the figure that the total excess correlation, given by the area under the curve, increases with decreasing  $T$ .

We can obtain an estimate of the typical distance over which mobile particles are correlated by identifying clusters of nearest-neighbor mobile particles [41]. To do this, we use the following rule: two particles belong to the same cluster if their distance at  $t=0$  is less than  $r_{nn}$ , the radius of the nearest neighbor shell, which is defined by the first minimum in  $g_{AA}(r)$  (not shown) and has a weak temperature dependence. In our hottest run  $r_{nn} = 1.45$ , while in the coldest run  $r_{nn}$

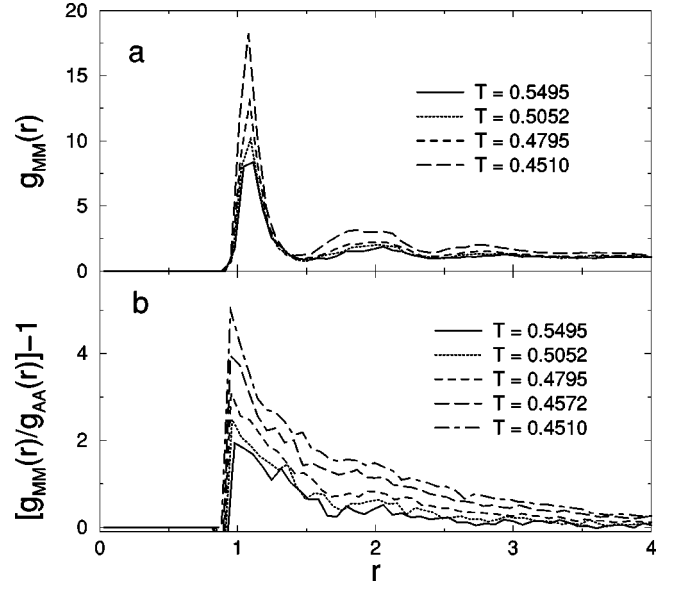


FIG. 4. (a) Pair correlation function  $g_{MM}(r)$  between mobile  $A$  particles for four different temperatures. (b)  $\Gamma(r) = [g_{MM}(r)/g_{AA}(r)] - 1$  vs  $r$  for five different temperatures.

$= 1.40$ . The probability distribution  $P(n)$  of clusters of size  $n$  is shown in Fig. 5. Although most of the clusters have only a modest size, the data show that a significant fraction of the mobile particles, which themselves make up only 5% of the sample (320 particles), are part of big clusters. For instance, at  $T = 0.4510$ , there is typically at least one cluster in each configuration that contains  $\approx 100$  particles. For that  $T$ ,  $P(n) \sim n^{-\tau}$  with  $\tau = 1.86$ . In the inset we show the mean cluster size  $S = \sum n^2 P(n) / \sum n P(n)$  [42], plotted log-log versus  $T - T_c$ , where  $T_c = 0.435$  is the fitted critical temperature of the mode coupling theory [12,13]. Although there is less than a decade on either axis, the figure shows that the temperature dependence of  $S$  is consistent with a divergence at  $T_c$  of the form  $S \sim (T - T_c)^{-\gamma}$ , with  $\gamma \approx 0.62$ . These expo-

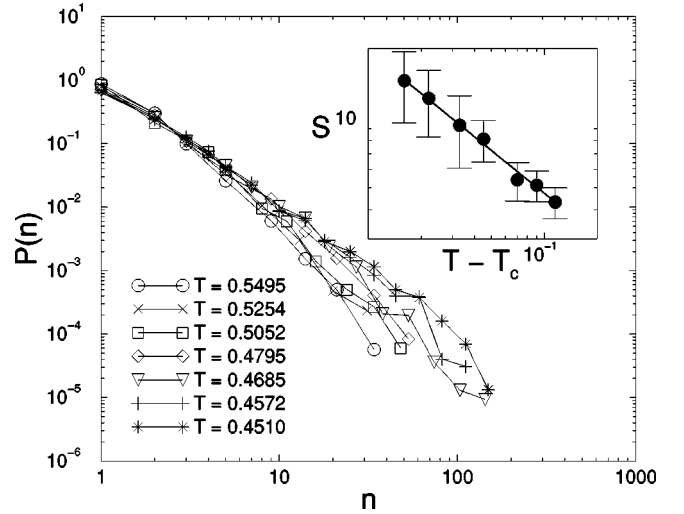


FIG. 5. Probability distribution of the size  $n$  of clusters of mobile particles for seven values of  $T$ . Inset: Mean cluster size  $S$  plotted vs  $T - T_c$ , with  $T_c = 0.435$  as determined in independent studies [13]. The straight line is a power law fit  $S \sim (T - T_c)^{-\gamma}$ , with  $\gamma = 0.618$ .

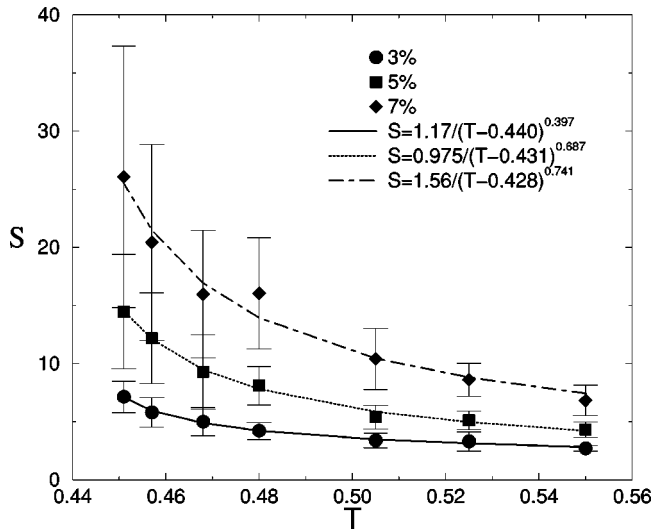


FIG. 6. Mean cluster size  $S$  plotted vs  $T$ , for subsets containing 3, 5, and 7 % of the most mobile particles. The data for the 5 % are the same as those shown in the inset of the previous figure. The lines are power law fits  $S \sim (T - T_p)^{-\gamma}$ . Best fit parameters are  $T_p = 0.440, 0.431,$  and  $0.428,$  respectively, and  $\gamma = 0.397, 0.687,$  and  $0.741,$  respectively.

nents do not fall into any of the usual percolation universality classes (in particular, note that  $\gamma < 1$ ) [43,44]. Note that MCT makes no predictions about clustering or the divergence of any length scales as the critical point is approached [45].

To test the sensitivity of the apparent percolation transition at  $T_c$ , we repeat the cluster size distribution analysis for the 3 and 7 % most mobile particles. For each subset, the mean cluster size  $S$  is shown vs  $T - T_c$  in Fig. 6. The best fit of  $S \sim (T - T_p)^{-\gamma}$  to each set of data gives  $T_p = 0.440$  for the set containing the 3% most mobile particles,  $T_p = 0.431$  for the set containing the 5% most mobile particles, and  $T_p = 0.428$  for the set containing the 7% most mobile particles. However, within the accuracy of the data the three sets are also consistent with a divergence at  $T_c$ . If we further increase the fraction of mobile particles beyond the fraction corresponding to a random close-packed percolation transition [46], the mobile particles percolate and most of the mobile particles are found in a single cluster that spans the whole simulation box. A power law fit to the data with  $T_p$  substantially different than  $T_c$  gives a worse fit than with  $T_p$  close to  $T_c$ , but the large size of the error bars, which is a consequence of our relatively small system size and large sample-to-sample fluctuations, prevents us from rigorously excluding one value of  $T_p$  in favor of another in the present simulation.

In Fig. 7 we show one of the largest clusters of mobile particles found in our coldest simulation. It is evident from the figure that these clusters cannot be described as compact, as often supposed either implicitly or explicitly in phenomenological models of dynamically heterogeneous liquids [17,47]. Instead, the clusters formed by the mobile particles appear to have a disperse, stringlike nature. As discussed in Ref. [11], a preliminary calculation of the fractal dimension of the clusters, although hampered by a lack of statistics,

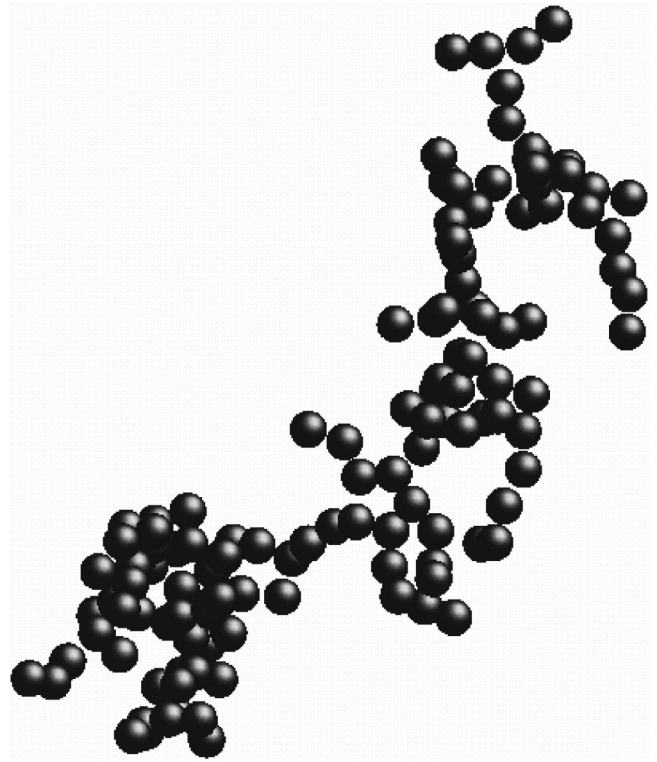


FIG. 7. One of the largest clusters of mobile A particles found at  $T=0.4510$ . The cluster is composed of 125 particles, which are represented here as spheres of radius  $r=0.5\sigma_{aa}$ .

indicates that the clusters have a fractal dimension close to 1.75, similar to that for both self-avoiding random walks and the backbone of a random percolation cluster in three dimensions [43].

In Ref. [11], it was shown that this quasi-one-dimensionality appears to arise from the tendency for mobile particles to follow one another. This is demonstrated in Fig. 8, where we plot the time-dependent pair correlation function for the mobile particles,  $g_{MM}(r, t^*)$  for different temperatures, where

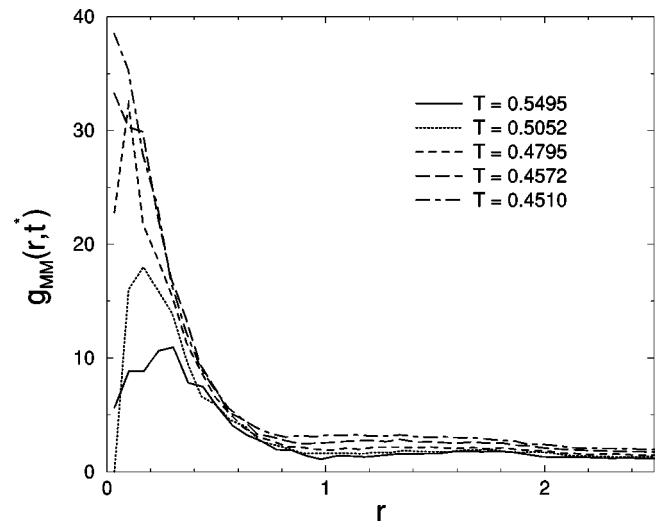


FIG. 8. Time-dependent pair correlation function  $g_{MM}(r, t^*)$  vs  $r$ , for several different temperatures.

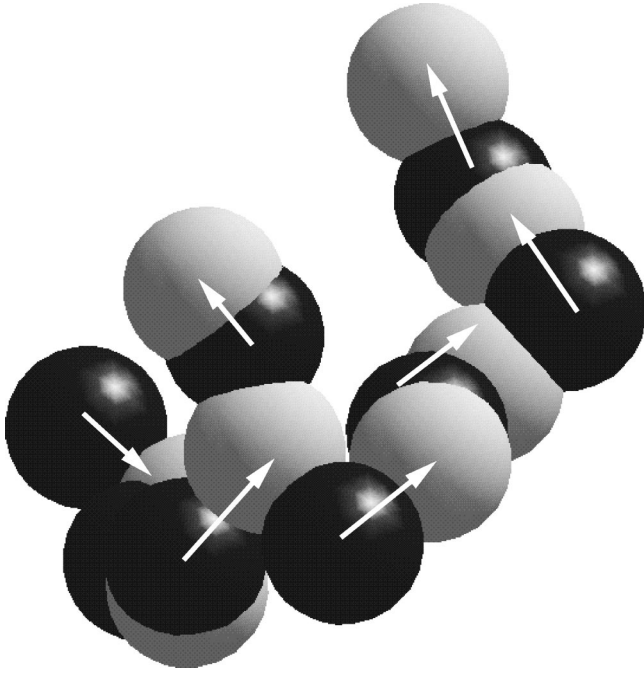


FIG. 9. A cluster of mobile particles at  $t=0$  (dark spheres) and  $t=t^*$  (light spheres), for  $T=0.4510$ . This cluster is composed of two “strings.”

$$g_{MM}(\mathbf{r}, t) = \frac{V}{N_M(N_M-1)} \left\langle \sum_{i=1}^{N_M} \sum_{\substack{j=1 \\ j \neq i}}^{N_M} \delta[\mathbf{r} - \mathbf{r}_j(t) + \mathbf{r}_i(0)] \right\rangle, \quad (4.5)$$

and where the sums are restricted to the total number  $N_M$  of mobile particles. At  $t=0$ , this function coincides with  $g_{MM}(r)$  in Fig. 4. For  $t>0$ , the nearest neighbor peak moves toward  $r=0$ , demonstrating that a mobile particle that at  $t=0$  is a nearest neighbor of another mobile particle tends to move toward that particle at later times. We find that the peak at  $r=0$  is highest near  $t=t^*$ , and decreases for later times. A small but discernable peak at  $r=0$  is also present in  $g(r, t^*)$  [48]. Figure 9 shows a cluster of mobile particles at two different times,  $t=0$  and  $t=t^*$ , to demonstrate the cooperative, stringlike nature of the particle motion. As explained in Refs. [11,49], each cluster of mobile particles may be decomposed into smaller, cooperatively rearranging “strings” of particles. Although the mean mobile-particle cluster size grows rapidly with decreasing  $T$ , the mean string length appears to grow much more slowly.

In a manner identical to our analysis of the mobile particles, we define as immobile the 5% of the A particles that have the lowest value of  $\mu$ . The pair correlation function  $g_{II}(r)$  between immobile particles shown in Fig. 10 shows that these particles also tend to be spatially correlated. It is interesting to note that while the maxima in  $g_{II}(r)$  are higher at all  $T$  than the corresponding maxima in  $g_{AA}(r)$ , the depth of the minima does not change appreciably for the lowest temperatures. Figure 10 shows also the ratio  $\Gamma(r) = [g_{II}(r)/g_{AA}(r)] - 1$  as a function of  $r$ . In contrast to what we find for the most mobile particles, the correlation between immobile particles does not show any evidence of singular behavior as  $T$  decreases. Instead, the correlation appears to grow and then “saturate” to some limiting behavior

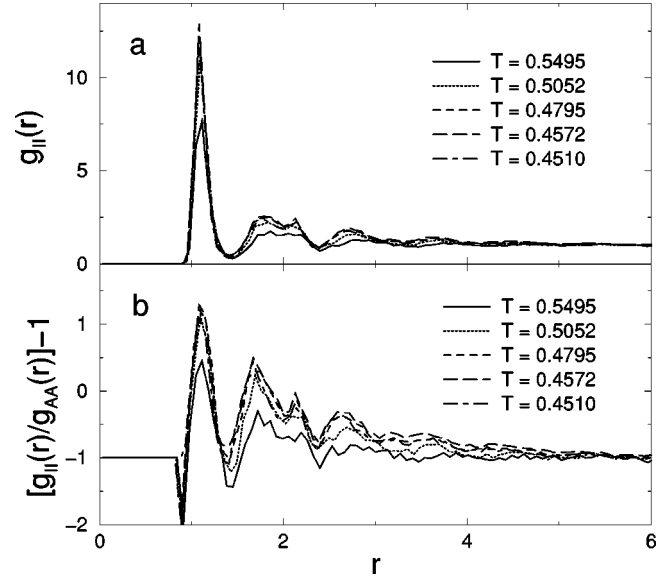


FIG. 10. (a) Pair correlation function  $g_{II}(r)$  between immobile particles for several different temperatures. (b)  $\Gamma(r) = [g_{II}(r)/g_{AA}(r)] - 1$  vs  $r$  for several different temperatures.

for all  $T < 0.468$ . Moreover, Fig. 10 shows that the local structure of the liquid appears to be more ordered in the vicinity of an immobile A particle than in the vicinity of a mobile A particle.

In Fig. 11 we show the size distribution of the clusters of immobile particles, formed with the same rule used for the mobile ones. One of the largest clusters found at  $T = 0.4510$  is shown in Fig. 12. In the inset of Fig. 11 we show the mean cluster size  $S$  versus  $T - T_c$ . We find that the mean cluster size of immobile particles is relatively constant with  $T$ . This may be because immobile particles are relatively well packed, and cannot grow beyond some limiting size [50]. Or, these clusters may be the “cores” of larger clusters of particles with small displacements, that may grow with decreasing  $T$ . To elucidate this, more particles (e.g., the next 5% higher mobility) should be included in the analysis. We

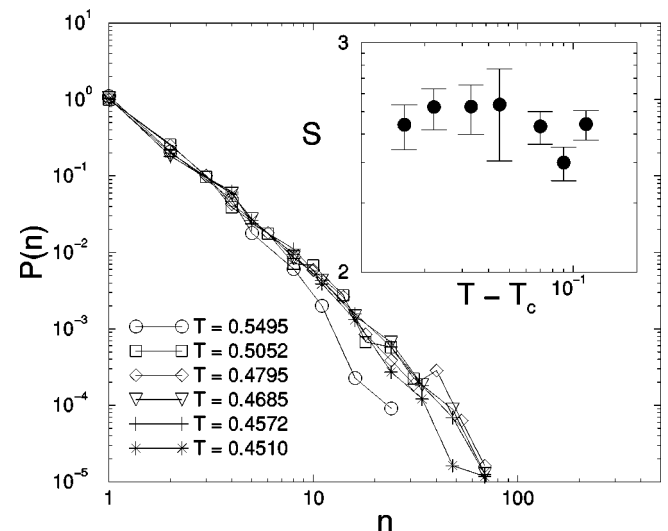


FIG. 11. Probability distribution of the size  $n$  of clusters of immobile particles for six values of  $T$ . Inset: mean cluster size  $S$  plotted vs  $T - T_c$ .

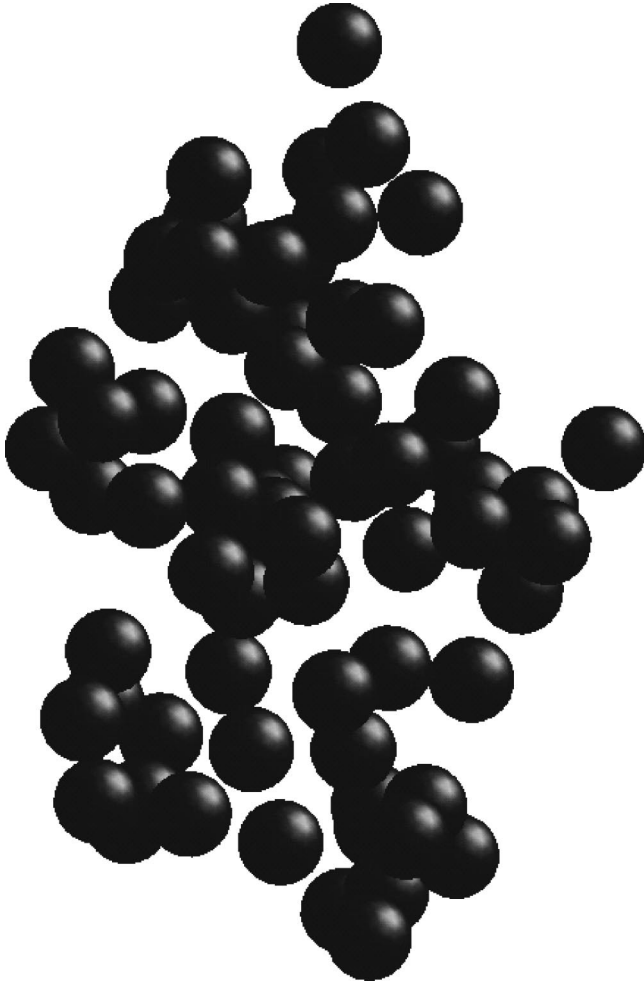


FIG. 12. One of the largest clusters of immobile A particles found at  $T=0.4510$ . The cluster is composed of 70 particles, which are represented here as spheres of radius  $r=0.5\sigma_{aa}$ .

will return to this important point and provide further relevant data in the next section.

The correlation between mobile and immobile particles, measured by the pair correlation function  $g_{MI}(r)$  (Fig. 13), shows that mobile and immobile particles are anticorrelated. A comparison between  $g_{MI}(r)$  and  $g_{AA}(r)$ , also shown in Fig. 13, demonstrates that, over several interparticle distances, the probability to find an immobile A particle in the vicinity of a mobile one is lower than the probability to find a generic A particle. The figure also shows that the characteristic length scale of the anticorrelation grows with decreasing  $T$ . This length scale does not show a tendency to diverge as  $T_c$  is approached. In particular, the curves for the two coldest runs (and closest to  $T_c$ ) are almost coincident.

### V. LOCAL ENERGY AND LOCAL COMPOSITION VERSUS MOBILITY

We have seen in the previous section that despite the lack of a growing static correlation, a growing *dynamical* correlation — characterizing spatial correlations between particles of similar mobility — does exist. These correlations must therefore arise from subtle changes in the local environment that are not completely captured by the usual static pair cor-

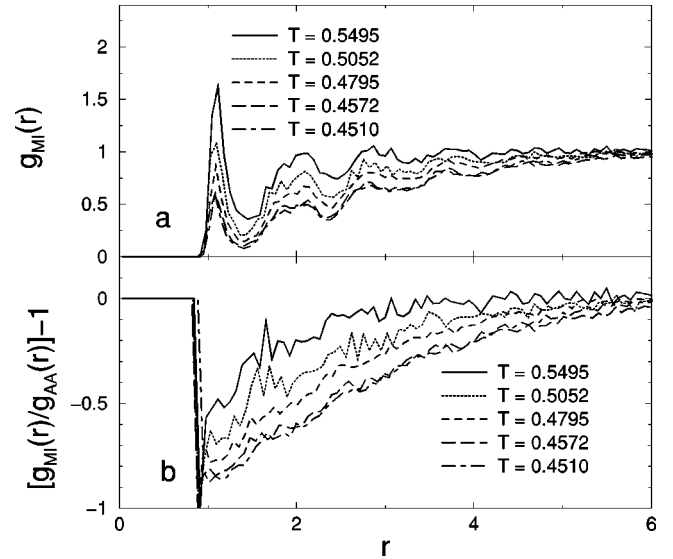


FIG. 13. (a) Pair correlation function  $g_{MI}(r)$  between mobile and immobile A particles for several different temperatures. (b)  $\Gamma(r)=[g_{MI}(r)/g_{AA}(r)]-1$  vs  $r$  for several different temperatures.

relation functions. In this section, we calculate several quantities to elucidate whether the mobility of a particle is related to its potential energy, and to the composition of its local neighborhood.

In Fig. 14 we show the distributions of the potential energies of the 5% most mobile, 5% least mobile, and all particles at  $T=0.4510$ , calculated at the beginning of an arbitrary time interval  $[t, t+t^*]$ . The distributions have been normalized such that the area under each curve is 1. The distributions differ by a small relative shift of the mean value, approximately 3% for the high mobility distributions and somewhat less for the low mobility distribution. We find that the magnitude of the shift increases with decreasing  $T$ , but the *relative* shift appears to be *independent* of  $T$ . Since the liquid is in equilibrium, this shift will vanish for  $t \rightarrow \infty$ . Thus, not surprisingly, mobile particles are those that in a

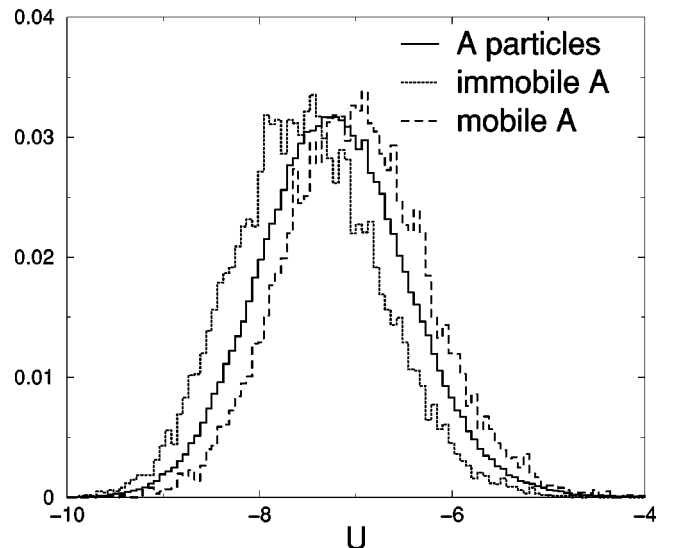


FIG. 14. Distribution of the potential energy of all the A particles, of the mobile A particles, and of the immobile A particles for  $T=0.4510$ .



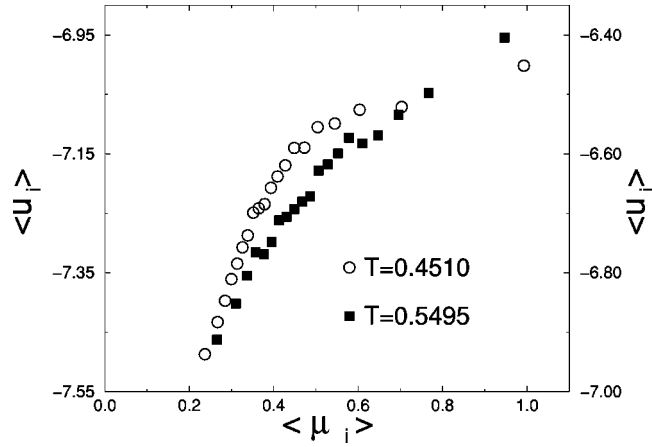


FIG. 15. Potential energy  $\langle U_i \rangle$  as a function of the mobility  $\langle \mu_i \rangle$  for the A particles. The A particles have been divided into 20 subsets according to their mobility at  $t^*$ . Each subset is represented by a point in the graph. The energy scale for  $T=0.4510$  is on the left hand side y axis, while the energy scale for  $T=0.550$  is on the right hand side y axis.

time  $t^*$  are able to rearrange their position so as to lower their potential energy. It is worth noting that the mobility does not show any correlation with the kinetic energy of the particles measured at  $t=0$ . The kinetic energy distributions of the subsets with different mobility coincide exactly with the average distribution, showing that the mobility cannot be related to the presence of “hot spots” in the liquid.

We next divide the entire population of A particles into 20 subsets, each composed of 5% of the particles. In the first subset we put the 5% of the particles with the highest values of  $\mu$  (the mobile particles defined above), in the second subset the next 5%, and so on. The last subset thus contains the 5% most immobile particles. In Fig. 15 we plot (on the x axis) the average mobility of each subset versus (on the y axis) the average potential energy of that subset at  $t=0$ . We find that the subset with the lowest mobility is also the one with the lowest potential energy. We also find that as the potential energy increases, the mobility increases. We see from the figure that the mobile particles are the subset with the highest average potential energy at  $t=0$ .

Two more points are worth noting in Fig. 15. First, at all  $T$  the mobile particles move, on average, approximately one interparticle distance in the time interval  $[0, t^*]$ . Second, for all  $T$  the difference in both mobility and potential energy between the 5% most mobile particles and the next subset is significantly larger than between any other two consecutive subsets. This observation suggests that the choice of 5%, while arbitrary, is a reasonable one. As shown in the figure, the separation between the 5% most mobile particles and the next subset shows a tendency to grow with decreasing  $T$ . Note however, that the distance between the lowest mobility subset and the next subset *decreases* with decreasing  $T$ , making it very difficult in the current approach to define an appropriate subset containing particles whose mobility is distinctly lower than the rest. This, together with the result that the mean cluster size of immobile particles is relatively constant over the range of temperatures studied, suggests that our analysis of the lowest subset is inadequate to fully char-

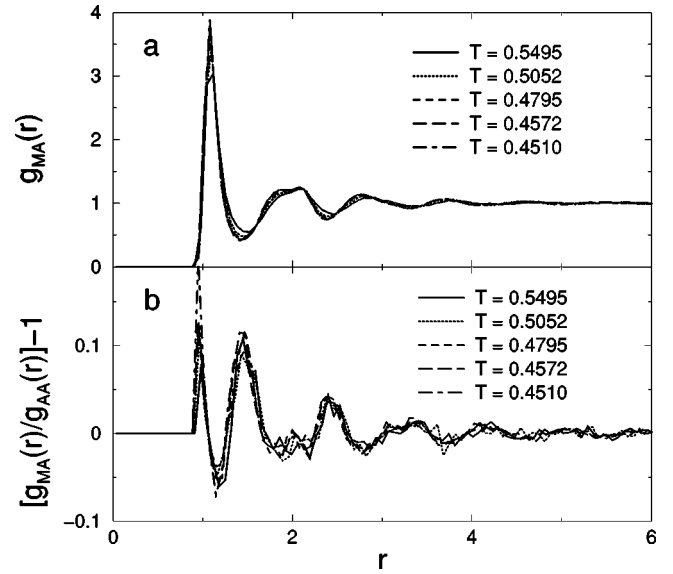


FIG. 16. (a) Pair correlation function  $g_{MA}(r)$  between mobile A and generic A particles for several different temperatures. (b)  $\Gamma(r) = [g_{MA}/g_{AA}(r)] - 1$  vs  $r$  for several different temperatures.

acterize clusters of particles that do not move a substantial distance [51].

Thus, we see that the gross structural information contained in the potential energy is sufficient to establish a general correlation between energy and mobility. However, as seen in Fig. 14, the distribution of potential energies of mobile particles overlaps for most of the range of the abscissa with the distribution for the generic A particles. Thus, it is not possible to decide if a certain particle is mobile on the basis of that particle’s instantaneous potential energy alone. Other factors, such as defects in the local packing, and the relative potential energy of neighboring particles, must contribute as well.

The relation between mobility and potential energy suggests a relation between mobility and local composition. Indeed, a calculation of the pair correlation functions  $g_{MA}(r)$  and  $g_{MB}(r)$  (see Figs. 16, 17) between a mobile particle and a generic A or B particle, respectively, shows that, on average, a mobile A particle tends to have less B particles, and more A particles, in its nearest neighbor shell than a generic A particle.

A correlation can also be found between *immobility* and small composition fluctuations of the mixture. A comparison of  $g_{AB}(r)$  to the pair correlation function  $g_{IB}(r)$ , which measures the number of B particles a distance  $r$  from a test immobile A particle, shows that an immobile A particle has, on average, more B particles in its nearest neighbor shell than does a generic A particle. As shown in Fig. 18, where the ratio  $[g_{IB}/g_{AB}(r)] - 1$  is plotted as a function of  $r$ , this enhanced correlation is independent of  $T$ , and therefore does not suggest any evidence of A–B phase separation (recall that the chosen energy parameters preclude A–B phase separation).

From these results it is clear how a correlation between mobility and local composition causes a correlation between mobility and potential energy. Since the attractive interaction between A and B particles is stronger than either the attractive AA or BB interaction, the presence of a B between two

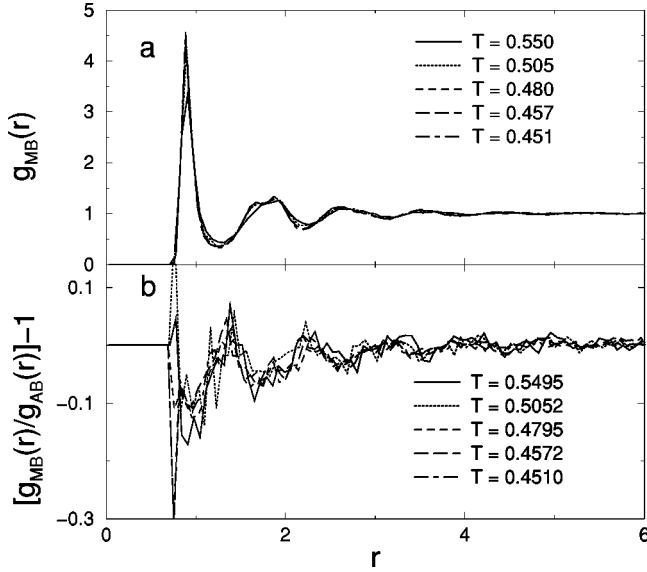


FIG. 17. (a) Pair correlation function  $g_{MB}(r)$  between mobile  $A$  and generic  $B$  particles for several different temperatures. (b)  $\Gamma(r) = [g_{MB}(r)/g_{AB}(r)] - 1$  vs  $r$  for several different temperatures.

$A$  particles reduces their potential energy.  $A$  particles in a  $B$ -rich region can thus be expected to have a reduced mobility.  $A$  particles in a  $B$ -poor region, however, will have a higher potential energy, resulting in a higher mobility.

## VI. STRUCTURAL RELAXATION OF PARTICLE SUBSETS AND DYNAMICAL HETEROGENEITY

We have shown that it is possible to select subsets of particles according to their maximum displacement over a timescale in the region of the late  $\beta$ -early  $\alpha$  relaxation. We have also shown that the particles belonging to subsets selected at the extrema of the mobility spectrum are spatially correlated, and are related to small fluctuations in the local potential energy, and, consequently, in the local composition of the mixture. All of the data presented here suggest that this supercooled liquid contains fluctuations in local mobility, with diffuse, quasi-one-dimensional regions of high mo-

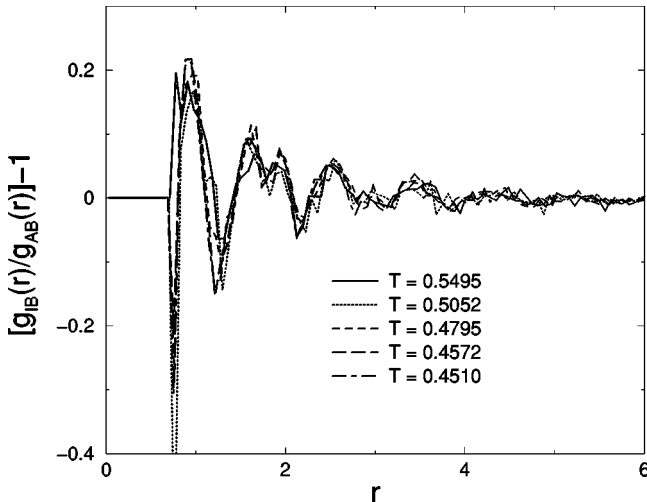


FIG. 18.  $\Gamma(r) = [g_{IB}(r)/g_{AB}(r)] - 1$  vs  $r$  for several different temperatures.

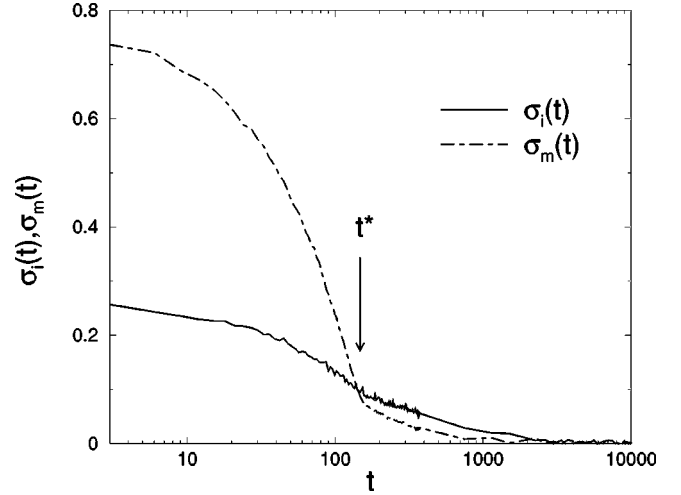


FIG. 19.  $\sigma_M(t)$  (dot-dashed line) and  $\sigma_I(t)$  (solid line) for  $T = 0.4510$ .

bility, and relatively compact regions of low mobility.

To measure how long a mobile particle will continue to be mobile, we define a variable  $\nu_i^M(t)$  as 1 if the  $i$ th particle belongs to the subset of the 5% most mobile particles in the interval  $[t, t+t^*]$ , and 0 otherwise. The function  $\sigma_M(t)$

$$\sigma_M(t) = \frac{1}{n_M - n_M^2/N_A} \left( \sum_i \langle \nu_i^M(t) \nu_i^M(0) \rangle - \frac{n_M^2}{N_A} \right), \quad (6.1)$$

measures the fraction of particles that are mobile in the interval  $[0, t^*]$  and still mobile in the interval  $[t, t+t^*]$ , when the time origin is shifted by  $t$ . Here  $n_M$  is the number of mobile particles (320 in the present case), and  $N_A$  is the total number of  $A$  particles (6400). The second term on the right-hand side of Eq. (6.1) is the number of particles that by random statistics would be classified as mobile in both time intervals. The normalization of  $\sigma_M(t)$  is chosen so that  $\sigma_M(0) = 1$ . The results for  $\sigma_M(t)$  for the coldest  $T$  are shown in Fig. 19.

We have also measured the fraction of particles that are immobile in the interval  $[0, t^*]$  and still immobile in the interval  $[t, t+t^*]$ . Analogous to the case for the mobile particles, we define  $\sigma_I(t)$  as

$$\sigma_I(t) = \frac{1}{n_I - n_I^2/N_A} \left( \sum_i \langle \nu_i^I(t) \nu_i^I(0) \rangle - \frac{n_I^2}{N_A} \right), \quad (6.2)$$

where  $n_I$  is the number of immobile particles (320) and  $\nu_i^I(t)$  is a function that is 1 if the  $i$ th particle is an immobile one in the interval  $[t, t+t^*]$ , and 0 otherwise. The function  $\sigma_I(t)$  is also shown in Fig. 19.

The functions  $\sigma_M(t)$  and  $\sigma_I(t)$  are memory functions of mobility. When they have decayed to zero, there are no particles that have retained memory of their mobility in the initial time interval. Because a particle's mobility is based upon a criterion that depends on  $t^*$ , certain time-dependent functions measured for these subsets will have some ‘‘kink’’ at  $t^*$ . If a different  $t^*$  is chosen, the kink will move to the

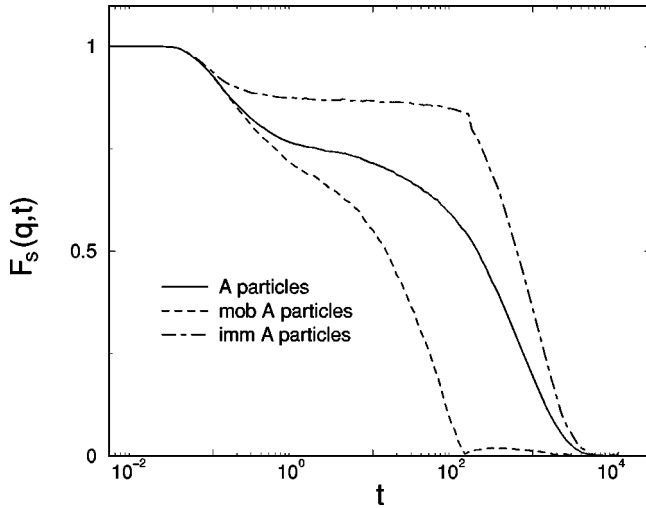


FIG. 20. Self-intermediate scattering function for  $T=0.4510$  for all the A particles (solid line), for the mobile particles (dashed line), and for the immobile particles (dot-dashed line).

new  $t^*$ . In this respect, there is no “natural” lifetime for these clusters — by definition, they survive for a time  $t^*$  [49,52].

Nevertheless, we can obtain information from the form of the decay both before and after  $t^*$ . Because the data was stored not less than every 3 time units, we are unable to calculate  $\sigma_M(t)$  and  $\sigma_I(t)$  for  $t < 3$ . However, we see that these functions decay substantially before this time, since already at  $t=3$  both functions are significantly smaller than one. After this initial short-time relaxation, a second decay of both functions is observed up to  $t=t^*$ . At this time, a third decay process appears for the mobile particles, and possibly also for the immobile particles. The main point of Fig. 19 is that beyond  $t^*$ , both functions are less than 0.1. Hence, there is only a small tendency for particles to retain memory of their mobility beyond the initial time interval [53].

Thus, mobile and immobile regions do not persist beyond the time  $t^*$  over which the particle mobility is monitored. After the observation time, mobile and immobile particles retain little memory of their previous state. Therefore the strong correlations found between particles must arise from the motion itself. If, for instance, the mobility of a particle and its spatial correlation with other particles of similar mobility could be explained solely by local fluctuations in quantities like density or composition, the mobility should persist until these fluctuations decay to zero. Instead, the dependence of the lifetime on the observation time can be explained if one assumes that particles can move only in a cooperative manner. Indeed, as was shown in Ref. [11], clusters of mobile particles such as that shown in Fig. 7 can be decomposed into numerous, smaller stringlike clusters (“strings”) of particles that follow one another in a cooperative fashion.

Figure 20 shows for  $T=0.4510$  the intermediate scattering functions  $F_s^M(q,t)$  and  $F_s^I(q,t)$ , defined as the spatial Fourier transform of the self part of the van Hove correlation function  $G_s^M(r,t)$  or  $G_s^I(r,t)$  of the mobile and immobile particles, respectively. As shown in the figure, both functions are identical to the bulk  $F_s(q,t)$  for the A particles for times less than the “collision time”  $\tau=0.09$  [40]. The figure

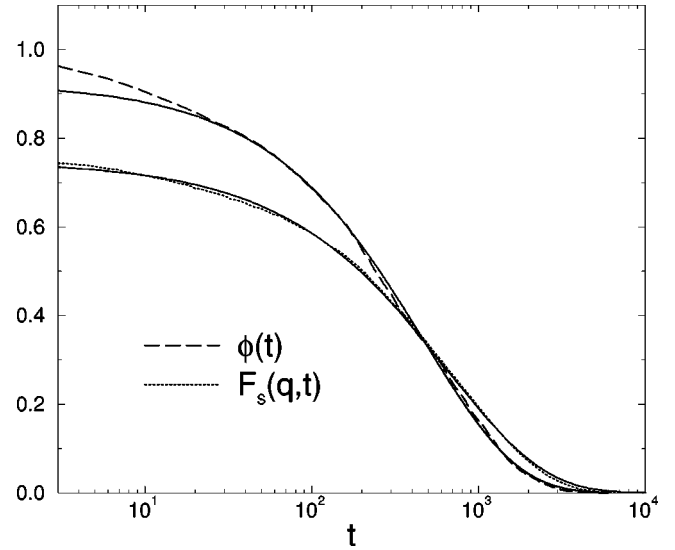


FIG. 21.  $\phi(t)$  (dashed line) and  $F_s(q,t)$  (dotted line) for  $T=0.4510$ . The solid lines are fits to a stretched exponential (see text).

shows that a two-step relaxation process occurs for the mobile particles, although the height of the plateau is smaller than for the bulk. The presence of the plateau in  $F_s^M(q,t)$  indicates that the mobile particles are subject to the same “cage effect” experienced by the other particles, although the effective cage “size” and “lifetime” are different. Consequently, clusters of mobile particles should not be thought of as “fluidized” regions of the liquid in the simple sense that those regions might behave similar to high-temperature or low density liquids. Instead, the difference between the mobile particles and the rest of the sample appears to be, from the point of view of the single particle dynamics, that they “escape” the cage earlier than the other particles.

We also see that the three curves in Fig. 20 cannot be superimposed by scaling the time axis in the same way as one can superimpose  $F(q,t)$  curves for different temperatures. Again, this indicates that the mobile and immobile subsets are not simply “hotter” or “colder” subsets of the sample, in agreement with the perfect superposition of the kinetic energy distributions (not shown).

In contrast to the bulk average  $F_s(q,t)$ ,  $F_s^M(q,t)$  is not a monotonically decreasing function of time. For times longer than  $t^*$ , a small but clearly detectable increase of the function can be noticed in Fig. 20. This behavior can be interpreted as a tendency of a small fraction of the particles that we have selected to return towards their position at the beginning of the selection interval. These particles may also be those that contribute to the small memory effect observed in  $\sigma_M(t)$  in Fig. 19, but further analysis is required to establish this connection.

Finally, we show in Fig. 21 for  $T=0.4510$  the fraction  $\phi$  of particles that at time  $t$  have not yet been labeled mobile. This function is calculated by labeling the mobile particles in the first interval  $[0,t^*]$ , and then shifting the interval by  $t$  and reassigning the particle mobilities. Thus at  $t=0$ , 95% of the particles have not been labeled mobile. In the interval  $[t,t+t^*]$ , more particles will have been labeled mobile, so  $\phi$  will decrease. We have normalized  $\phi(t)$  such that  $\phi(0)=1$ . Also shown in Fig. 21 is the long-time  $\alpha$  relaxation part of

the bulk  $F_s(q, t)$  for  $q = q_{\max}$ . Fits to both functions are also shown. Both functions fit well to a stretched exponential  $y(t) = A \exp[-(t/\tau)^\beta]$  with  $\beta = 0.75$  and  $\tau_\alpha = 655$  for the intermediate scattering function, and with  $\beta = 0.78$  and  $\tau_\alpha = 475$  for  $\phi$ . That both functions have a similar form (similar  $\beta$ ), and similar time constants, suggests that the process by which immobile particles become mobile governs the long time structural relaxation of density fluctuations at wavevectors corresponding to the peak of the static structure factor. Moreover, it demonstrates that the ‘‘arbitrary’’ choice of 5% represents a physically meaningful fraction of the system.

## VII. DISCUSSION

In this paper, we have described an investigation of the individual particle dynamics of a cold, dense Lennard-Jones mixture well above the glass transition in an effort to discover if the liquid is dynamically heterogenous, and if so to determine the extent and nature of the dynamical heterogeneity. Since there were no quantitative theoretical predictions regarding this matter, the approach we have taken is exploratory; particle trajectories were saved during the course of the simulation and then analyzed and visualized in numerous ways. We find that this supercooled liquid is ‘‘dynamically heterogeneous’’ because particles with similar mobility are spatially correlated. Note that our definition of heterogeneity is different from the one used, for instance, in 4D NMR experiments, where the system is defined as heterogeneous if a slow subset remains slow for times longer than the average relaxation time [18]. We further find that highly ramified clusters of mobile particles grow with decreasing  $T$  and appear to percolate at the mode-coupling temperature. This evidence for a percolation transition coincident with  $T_c$  is very

different from the type of percolation transition proposed in free volume theory [54]. It is especially interesting since MCT does not make any predictions regarding clusters or diverging length scales. We also find that particles of low mobility form relatively well-ordered, compact clusters which do not appear to grow with decreasing  $T$  if the number of immobile particles included in the subset is kept constant. Although mobile and immobile clusters are anticorrelated, there is no tendency towards bulk phase separation of mobile and immobile regions because of the highly ramified, extended nature of the mobile regions.

In our analysis, we find no evidence to support a picture in which the system can be thought of as a collection of subvolumes that each relax independently and simultaneously with their own time constant. Instead, it appears that at any given time, most particles are localized in cages and a small percentage of particles form large clusters of smaller, cooperatively rearranging ‘‘strings.’’ After rearranging, these mobile particles become caged themselves, and others become mobile. This process repeats until, on the time scale of the  $\alpha$  relaxation, each particle has rearranged at least once. Thus the structural relaxation of the liquid appears to be highly cooperative in the spirit of Adam and Gibbs, but where different subvolumes of the liquid are able to relax only after other subvolumes relax. This sequential relaxation process will be further explored in a separate publication.

## ACKNOWLEDGMENTS

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