Aging and relaxation in glass-forming systems

Valery Ilyin, Itamar Procaccia, Ido Regev, and Nurith Schupper Department of Chemical Physics, The Weizmann Institute of Science, Rehovot 76100, Israel (Received 16 March 2008; published 19 June 2008)

We propose that there exists a generic class of glass-forming systems that have competing states (of crystalline order or not) which are locally close in energy to the ground state (which is typically unique). Upon cooling, such systems exhibit patches (or clusters) of these competing states which become locally stable in the sense of having a relatively high local shear modulus. It is in between these clusters where aging, relaxation, and plasticity under strain can take place. We demonstrate explicitly that relaxation events that lead to aging occur where the local shear modulus is low (even negative) and result in an increase in the size of local patches of relative order. We examine the aging events closely from two points of view. On the one hand we show that they are very localized in real space, taking place outside the patches of relative order, and from the other point of view we show that they represent transitions from one local minimum in the potential surface to another. This picture offers a direct relation between structure and dynamics, ascribing the slowing down in glassforming systems to the reduction in relative volume of the amorphous material which is liquidlike. While we agree with the well-known Adam-Gibbs proposition that the slowing down is due to an entropic squeeze (a dramatic decrease in the number of available configurations), we do not agree with the Adam-Gibbs (or the Volger-Fulcher) formulas that predict an infinite relaxation time at a finite temperature. Rather, we propose that generically there should be no singular crisis at any finite temperature: the relaxation time and the associated correlation length (average cluster size) increase at most superexponentially when the temperature is lowered.

DOI: 10.1103/PhysRevE.77.061509

I. INTRODUCTION

In this paper we attempt to provide generic answers to some long-standing questions regarding the spectacular increase in relaxation times in typical structural glass formers as the temperature is reduced $\begin{bmatrix} 1-3 \end{bmatrix}$. In particular we address the relation between structural and dynamical phenomena, the question of existence of a typical length scale that is associated with the slowing down [4-7], the nature of aging in such systems [8], and the issue of existence of a "true" glass transition temperature $T_{o} > 0$ where the relaxation time becomes infinite [9]. We will base most of our comments on a careful analysis of one model system on which we performed extensive simulations, but will provide supporting evidence also for a second model system and an additional experimental system for which theory had been proposed recently. We will argue that these very different systems share some important characteristics, and we will risk a conjecture that these characteristics are generic for structural glass formers. Most importantly, these systems share the characteristic of having, in addition to a crystalline ground state, other states whose energy does not differ much from the ground state, at least for relatively small patches. The existence of such states leads first and foremost to the frustration of crystallization when the temperature is lowered. Glass formation is accompanied by the creation of local patches, or clusters, of different nature, and these patches are locally stable in the precise sense of having high local shear moduli. To make this point crystal clear we will introduce and study the notion of a local shear modulus for systems of this kind. The existence of the inhomogeneity (or patchiness) is characterized by a typical scale, and we will demonstrate how this typical scale increases rapidly upon decreasing the temperature. Aging and relaxation events occur generically in the remainder of the system, in regions of low local shear moduli, as will be shown below. The squeezing out of the PACS number(s): 64.70.Q-, 61.43.Fs

regions where relaxations occur is the fundamental reason for the slowing down, and we connect the increase of typical scale with the increase of relaxation time. We will analyze individual aging events and show that they can be understood either as localized "excitation chains" [10] in real space or transitions between local minima of the potential surface [11]. Finally we will argue that our picture does not predict the existence of a "true" glass transition, in the sense that infinite relaxation times are expected only as $T \rightarrow 0$ (unless a very unlikely catastrophic crystallization event intervenes at a finite temperature).

In Sec. II we present results for the binary mixture glassforming model, studying in a variety of ways, qualitative and quantitative, the spatial inhomogeneity that is crucial for glass formation. In Sec. III we introduce the notion of "local shear modulus," explain how to compute it in numerical simulations, and demonstrate the utility of this notion. We show that in the clusters of relative order the local shear modulus is typically high, whereas in between the clusters the local shear modulus can be low, even negative, signifying local mechanical instability. Aging events take place spontaneously in the latter regions. In Sec. IV we consider the anatomy of the aging events and show that they are very localized in real space. In terms of the potential energy landscape the aging events are mapped to transitions between local minima crossing saddles of order 1 (not necessarily always reducing the potential energy). In Sec. V we discuss the generality of the picture emerging here with the help of other examples of glass formers. Section VI offers a summary and a discussion.

II. CLUSTERS AND HETEROGENEITY IN A BINARY GLASS FORMER

The two-dimensional (2D) system consists of an equimolar mixture of two types of point particles interacting via an inverse power potential [12,13]:



FIG. 1. (Color online) A T1 process in the Voronoi tessellation (in red lines).

$$\phi_{ab}(r) = \epsilon \left(\frac{\sigma_{ab}}{r}\right)^n, \quad a, b = 1, 2, \tag{1}$$

where *r* is the separation distance between particles of types *a* and *b* and *n*=12. The particles have the same mass *m*, but half of the particles are "small" with "diameter" $\sigma_{11} = \sigma$, and half of the particles are "large" with "diameter" $\sigma_{22}=1.4\sigma$; the interaction between different kinds of particles is defined by $\sigma_{12}=(\sigma_{11}+\sigma_{22})/2$. We choose the units of mass, length, and time as *m*, σ , and $\tau=\sigma\sqrt{m/\epsilon}$, respectively. Other reduced units are the system energy $U^*=U/\epsilon$, enthalpy $H^*=H/\epsilon$, pressure $P^*=P\sigma^2/\epsilon$, reduced temperature $T^*=k_BT/\epsilon$, with k_B being Boltzmann's constant, and density $\rho^*=N/A^*$, where *N* is the particle number in the simulation box of dimensionless area $A^*=A/\sigma^2$.

We performed molecular dynamics simulations in the canonical (*NVT*) ensemble with N=1024 particles in a square simulation box of side $L^* = \sqrt{A^*}$ with periodical boundary conditions. The equations of motion were integrated using a third order Gear predictor-corrector algorithm [14] with a time step $\delta \tau = 0.005 \tau$. A constant temperature was preserved using a velocity rescaling method [15]. At each temperature the density was chosen in accordance with the simulations results in an *NPT* ensemble as described in [13] with the pressure value fixed at $P^*=13.5$.

A. Clusters and heterogeneity: Qualitative picture

1. Voronoi diagrams

The Voronoi tessellation is an effective tool for the structural analysis of many-body systems. In the particular case of two-dimensional systems the Voronoi polygon construction associates with every configuration of the particles a subdivision of position space into polygons, one per particle. The polygon associated with any particle contains all points closer to that particle than to any other particle. The edges of such a polygon are the perpendicular bisectors of the vectors joining the centers of neighboring particles.

Any Voronoi construction obeys Euler's theorem: $\chi = V$ -*E*+*F*, where χ is the Euler characteristic, *V*, *E*, and *F* are, respectively, the number of vertices, edges, and faces. In an infinite two-dimensional system with periodic boundary conditions (on the torus $\chi=0$), the average number of sides of the polygons is exactly six. This fixed average should be conserved under a local rearrangement of the polygons. The elementary collective movement of particles which leads to local topological rearrangement of polygons, and known as the *T*1 process [16], is shown in Fig. 1. In a binary mixture there are two types of particles (small and large, or blue and



FIG. 2. (Color online) Upper panel: a Voronoi polygon construction for the hexagonal structure of a one-component system. Lower panel: another phase of the binary mixture that can be stable at low temperatures, made from a Voronoi tessellation of heptagons with large particles and pentagons with small particles.

red, respectively). Thus to achieve a mapping of the particle positions and the number of sides of the polygons in the Voronoi tessellation, we distinguish between polygons having small or large particles in their center. Thus a coloring scheme of cells will take into account not only the number of sides, but also the type of the particle in each cell.

2. Competing phases

At this point we would like to explain that the present system has one ground state (lowest energy state), that of phase separated hexagons of small and large particles, but nearby there is another homogeneous state which is not too far in energy. We expect that the ground state of the system under consideration should be homogeneous. The obvious candidate structure is phase-separated large and small particles (an example of Voronoi tessellation for a subsystem of small or large particles is shown in the upper panel of Fig. 2); but this is not the only possible homogeneous phase. It was shown [5,17,18] that the phase shown in the lower panel of Fig. 2 is possible for our binary systems in the range $1.18 < \sigma_{22}/\sigma < 2$. This phase consists of small particles in



FIG. 3. (Color online) A Voronoi tesselation of a typical glassy state. Note the existence of clusters of large particles in hexagons, small particles in hexagons, and patches of the phase shown in Fig. 2. Our point of view is that glass formers generically form clusters of such competing ground and close to ground states.

pentagons and large particles in heptagons; these are the polygons which in the theory of two-dimensional melting of one-component hexagonal crystals are referred to as "defects" [19]. This phase is characterized by pairs of small particles in pentagons lying in parallel lines [17]. In one line all the pairs have the same orientation but can change it from line to line. Therefore this phase is periodic in one direction and can be ordered or disordered in the other one. It is interesting to compare these two homogeneous states in terms of their energy and enthalpy, since we will argue below that they compete in the glassy state.

The ground state of this model (at T=0) is defined by the minimum of the enthalpy $H^*=U^*+P^*A^*$ or the energy U^* in an *NPT* or *NVT* ensemble respectively. In the Appendix A we show that the ground state of our system is indeed the phase separated hexagons of small and large particles. The phase consisting of pentagons and heptagons is, however, close in energy, and certainly once formed will not be easy to deform to the phase separated ground state.

The crucial comment is that in the supercooled state this model exhibits clusters of all these phases. In Fig. 3 we show snapshots of the system at the temperature $T^*=0.2$. In the remainder of this paper we connect between the heterogeneity that is caused by the existence of such clusters and the phenomenon of slow aging in the glassy state. Reference [13] found, using molecular dynamics simulations in the isothermal-isobaric (NPT) ensemble, that for temperature T >0.5 the system is liquid and for lower temperatures dynamical relaxation slows down. A precise glass transition had not been identified in [13]. In [5,6] it was argued on the basis of statistical mechanics that there exists a typical length scale that grows superexponentially fast when the temperature decreases. Associated with this fast growing scale there exists a fast growing relaxation time, such that below a certain temperature the system is jammed for all practical purposes. Here we will shed further light on this phenomenon, relating it to the inhomogeneity seen with the bare eye in Fig. 3.

III. CLUSTERS AND HETEROGENEITY-QUANTITATIVE PICTURE USING THE LOCAL SHEAR MODULUS

In an NVT ensemble a given configuration can be characterized by components of the microscopic stress tensor which is defined by (see, e.g., [20])

$$\sigma_{\alpha\beta} = \frac{1}{A^*} \left(\sum_i p_i^{\alpha} p_i^{\beta} - \frac{1}{2} \sum_{j \neq i} \frac{\partial \phi_{ab}(r_{ij})}{\partial r_{ij}} \frac{r_{ij}^{\alpha} r_{jj}^{\beta}}{r_{ij}} \right), \qquad (2)$$

where p_i^{α} is the α component of the dimensionless momentum of particle *i* and r_{ij}^{α} is the α component of the vector joining particles *i* and *j*.

The diagonal components of this tensor correspond to density fluctuations and its trace determines the pressure, in accordance with the virial theorem:

$$P^* = \rho^* T^* - \rho^* \frac{1}{2dN} \sum_{j \neq i} r_{ij} \frac{\partial \phi_{ab}(r_{ij})}{\partial r_{ij}}, \qquad (3)$$

where d is the spatial dimension [h]. For interactions of the form (1) the pressure is related to the total potential energy of the system (A1):

$$P^* = \rho^* T^* + \rho^* \frac{n}{d} \frac{U^*}{N}.$$
 (4)

The point is that an economic designation of a "state" is given in the *NVT* ensemble by the pair of quantities energy and shear stress σ_{xy} .

In Fig. 4 we exhibit the actual trajectory of the molecular dynamics simulations in the stress-energy plane for three different temperatures and for the same simulation time. We see that at T=0.2 the trajectory hovers mostly around two distinct states with infrequent transitions between them (only one transition in this run). This is a general observation for low temperatures: the system fluctuates around one "solidlike" state, but then jumps to another such "solid-like" state [21]. The ease of transitions increases with temperature but also with the number of particles. Here we have 1024 particles; in previous simulations using 256 particles [22] we did not observe any transition at T=0.2. Changing the temperature to T=0.3 but keeping the same simulation time one resolves between seven to eight "solid-like" states. For T =0.4 the trajectory now fills up an extended region in the stress-energy plane. As stated in [22], it appears that between "real" solid and "real" liquid the system is locally not a "viscous fluid." Rather, the trajectory concatenates a relatively long period of time where the system behaves like a solid, interconnected by a relatively short period of times where the system flows between these states. Clearly, a viscous fluid behaves very differently, responding to stress by a viscous flow, be it as sluggish as one wishes. Here, most of the time, the system does not respond to stress, except in the narrow corridors of flow which become rare when the temperature goes down and more common when the temperature warms up. Of course this does not mean that in the sense of long time averaging the notion of viscosity cannot be resurrected, but locally in time this is impossible.

Since we have the system trajectory at our disposal, we can examine more closely the structural transformation that



FIG. 4. Trajectories from molecular dynamics simulations in the stress-energy plane, for the three temperatures T=0.2, 0.3, and 0.4, respectively. U^*/N is the potential energy of the system per particle. Each point represents an average over a duration of 5τ , connected with straight lines. The time of simulation is $10^4\tau$. Note the dramatic increase in the scale of the variations when the temperature increases.

is taking place when the system jumps from one solid-like state to the other. In Fig. 5 we show what happens by monitoring three characteristics of the system. One is simply the energy per particle as a function of time, cf. upper inset. We see that the energy fluctuates around a given value and then jumps to a new state where the energy fluctuates at a lower value. A second characteristic is the average shear stress $\langle \sigma_{xy} \rangle$ (cf. lower inset) which exhibits a similar jump, except that the variance of its fluctuations before the transition ap-



FIG. 5. (Color online) Details of an event of the type seen in Fig. 4 leading to a jump from one solidlike state to another. In the inset we show the energy per particle and the average shear stress as a function of time. Every point in these graphs is an average over 10^4 simulation steps. The figure itself is a window in the simulation box showing that the event is a circular motion of particles replacing each others position. The color code outside the event is red particles are large and black are small. In the circular event the red particle is large and all the rest are small.

pears considerably larger than the same variance after the jump. Most interestingly, the figure shows a window into the particle configuration itself, focusing on the local event that is responsible for the jumps in the insets. We see that the event is a concerted motion of six particles that change positions along a circular path; such events are referred to as "excitation chains" [10]. At this point we want to explore what is the significance of this event, why it takes place where it does, and what is it accomplishing in terms of aging and relaxation. To this aim we will study local mechanical properties.

A. Local shear moduli

For inhomogeneous systems such as amorphous materials it is important to study local properties in connection with thermodynamical and structural characteristics. To measure local values we take our simulation box and subdivide it into smaller squares of length $l_k = L^*/2^k$ ($2 \le k \le 4$). We calculated the local stress and the local shear modulus following the definition of these quantities in [23]. In the frame of this approach the local stress tensor ascribed to a subdomain m($1 \le m \le k^2$ in the *k*th level) is given by [23,24]

$$\sigma_{\alpha\beta}^{m} = \rho_{m}^{*} T^{*} \delta_{\alpha\beta} - \frac{1}{2l_{k}^{2}} \sum_{j \neq i} \frac{\partial \phi_{ab}(r_{ij})}{\partial r_{ij}} \frac{r_{ij}^{\alpha} r_{jj}^{\beta}}{r_{ij}} \frac{q_{ij}^{m}}{r_{ij}}, \qquad (5)$$

where ρ_m^* is the particle number density in the subdomain *m*, and q_{ij}^m is the length of the line segment of the vector joining

particles *i* and *j* which is located inside a subdomain *m* (if this vector does not pass through this subdomain, $q_{ij}^m=0$). The term q_{ij}^m/r_{ij} determines how different pairwise interactions are apportioned to the local stress of a subdomain *m*, and includes contributions from pairs of segments located outside the subdomain. An average of $\sigma_{\alpha\beta}^m$ over the entire area of the simulated system yields the usual bulk stress tensor (2).

The local elastic modulus tensor is related to the internal stress fluctuations. The expression for the shear modulus per subdomain is defined by [23,24]

$$\mu_m = \mu_m^{\infty} - \frac{L^{*2}}{T^*} (\langle \sigma_{\alpha\beta}^m \sigma_{\alpha\beta} \rangle - \langle \sigma_{\alpha\beta}^m \rangle \langle \sigma_{\alpha\beta} \rangle), \qquad (6)$$

where angular brackets mean averaging over configurations and the "infinite frequency shear modulus" μ_m^{∞} [25,26] per subdomain is given by

$$\mu_{m}^{\infty} = \rho_{m}^{*} T^{*} + \frac{1}{2l_{k}^{2}} \left\langle \sum_{i \neq j} \left\{ r \frac{\partial}{\partial r} \left[\left(\frac{1}{r} \frac{\partial \phi_{ab}(r)}{\partial r} \right) \left(\frac{r^{\alpha} r^{\beta}}{r^{2}} \right)^{2} \frac{q^{m}}{r} \right]_{ij} \right\} \right\rangle.$$

$$\tag{7}$$

B. Results

The results for the local shear modulus can be exhibited in a number of forms, each shedding light on another aspect of this useful measure. In Fig. 6 we show histograms of the local shear modulus, for different levels of box resolution, from coarse to fine as one goes from the upper panel to the lowest. In general we see that as long as the temperature is high such that the system is fluid, the histograms peak around zero and are quite narrow. With T=0.4 the histogram widens, but the mean still moves only a bit from zero. Not so for T=0.2 which is deep in the glassy domain, where the histograms move to exhibit a positive average, but still showing wide variations in the values of the local shear modulus. Note that with the increase of resolution we find more negative values of the shear modulus, and these are indicative of sensitive regions in the system which are mechanically ready to exhibit aging events. To clarify this point further we turn now to another way of presenting the same results, color coding real space to exhibit the high degree of spatial inhomogeneity of this system.

Typical spatial results for the local shear modulus with spatial resolution k=3 are shown in Fig. 7. The local shear modulus is color coded, showing positive and negative values of the local shear modulus. Obviously, in the liquid (high temperatures) the distribution of local values averages to zero when summed over all the cells, whereas when the system is at low temperatures the average is positive, in agreement with the results of Ref. [22]. Also, for higher temperatures the spread of values of the shear modulus (or the variance of its distribution) is smaller, whereas the variance also increases upon decreasing the temperature.

We can then expect that spontaneous aging events will take place where the local shear modulus is low. Indeed, for the system shown in Fig. 7 the first aging event took place in the square (4,5), which is a region where the local shear



FIG. 6. (Color online) Histograms of the local shear moduli for different temperatures (see the line type in the inset) and for different partitions, from gross (k=2, upper panel) to fine (k=4, lower panel). Note the tendency to move the average to positive values with lowering the temperature, with nevertheless a tail at T=0.2 going into negative values of the local shear modulus.

modulus attains its lowest value. We checked over many simulations that this is indeed typical, and that looking at the distribution of local shear moduli one can guess very reliably where the next aging event will take place. Of course, once the aging event occurred, the distribution of local shear moduli changes dramatically. In Fig. 8 we show the distribution of local shear moduli after the event in square (4,5) took place. We see that the local shear moduli went up consider-



FIG. 7. (Color online) Color-coded local shear modulus before the aging transition. Note that the local values can be either positive or negative, with positive values indicating local mechanical stability, whereas local negative values indicate mechanical instability. We expect relaxation (aging) events to occur in the blue regions.

ably in the region of the event; this is shown quantitatively in Fig. 9 where the difference between the local shear modulus in Figs. 7 and 8 is presented. We certainly need to understand what is precisely happening here. A clue to that aim is obtained when we superpose on Fig. 7 the clusters of hexagonal patches of big particles, see Fig. 10. We see that most of the changes in local shear modulus occur outside the clusters which are locally stable in the mechanical sense.

IV. ANATOMY OF AGING EVENTS

In this section we examine the same localized events of aging, or further relaxation steps toward equilibrium, but change the point of view from real space to the potential



FIG. 8. (Color online) Color-coded local shear modulus after the local relaxation event occurred in box (4,5).



FIG. 9. (Color online) Color-coded difference in the local shear modulus between the two last figures, computed after the local relaxation event occurred in box (4,5).

energy landscape. We start with introducing the basic notions.

A. Potential energy landscapes

One popular way of discussing the dynamics of many particle systems is to visualize the state of the system as a point on the surface that the potential energy function $U = f(\mathbf{x})$ draws in a Nd+1-dimensional space, where N is the number of particles and d the dimension of space (d=2 in our examples) [27]. The evolution in time is described by the path of the coordinate vector \mathbf{x} along the energy surface. Adding the thermal energy to the potential energy elevates the point above the energy surface, but we will be interested



FIG. 10. (Color online) The same as Fig. 9 but with the clusters of large particles superposed to show that the main changes occur outside the clusters which are locally stable in the mechanical sense.



FIG. 11. (Color online) Aging process. Shown is the potential energy per particle measured in the direct numerical simulation. Note the existence of five relaxation events. We analyze carefully the two events marked with ellipses.

in locating the paths associated with the relaxation events on the potential energy surface itself. At low temperatures the system is trapped in a minimum of this surface. In a crystallized system this minimum corresponds to the global minimum of the system-the ground state. In an amorphous system, however, this minimum is typically not a global one, but rather a local minimum which is separated from other minima of the system by moderate potential barriers. At temperatures greater than zero, we expect that, due to thermal fluctuations, the system will pass from one minimum to another via saddles. These saddles can be classified by their "order," which by definition is the number of negative eigenvalues of the Hessian matrix of U at the saddle. At low temperatures one encounters typically saddles of order one [28]. Our aim here is to focus on spontaneous aging events and map them onto trajectories on the energy surface. To do so we will take data from our numerical simulations, locate the aging events, and for each point in time before and after the event find the nearest extremum of the potential. The methods used to find the extrema are standard and are summarized for convenience in Appendix B. Unfortunately the numerics involved in this procedure is rather heavy, and for that reason in this section we will refer to simulations with 256 rather than 1024 particles and also somewhat higher temperatures where there is a larger number of aging events, cf. Fig. 4. The insights achieved are of course independent of this numerical restriction.

As an example consider Fig. 11 depicting the potential energy per particle as a function of time during an aging process in a system of 256 particles. This figure contains evidence for five aging events, and below we analyze in detail transitions occurring in the temporal proximity of the two events marked in red. These transitions are localized, in the sense that the irreversible change in the configuration of the system involves only a few particles. In Fig. 12, we can see the actual changes in configuration caused during these events. Here the particles' positions are plotted before, during, and after the event. The changed area consists of a mix of small and large particles. A temporary displacement occurs in a larger area around this region. This collective stringlike process [29] allows for the permanent change to take place despite the high pressure.



FIG. 12. (Color online). Configurational changes during transitions 1 and 2 in Fig. 11, respectively. Large symbols refer to large particles and small symbols to small particles. In circles we denote the positions of the particles before the transition, in x the positions at the saddle and in + the positions after the transition.

Figure 13 shows the energy of the extrema found in the vicinity of these transitions. To find those we use a timedependent trajectory from the molecular dynamics simulation and for each configuration analyzed we seek the closest extremum of the potential energy surface. The vertical scale shows the potential energy of the system at the nearest extremum found using the algorithm explained in Appendix B. Most of the time the method locates a particular extremum with a specific energy, starting with a minimum, going through a saddle, and ending at another minimum. We make sure that these extrema are indeed minima or saddles by finding the smallest eigenvalue of the Hessian matrix at each extremum; it is positive for the minima and uniquely negative in the saddle, meaning that the saddle is of order unity.

Note that during the long residence near the first minimum in Fig. 13 there are instances when the system fluctuates to a point near a saddle; However, the system remains trapped in the first minimum until the aging transition takes place.

The same method showed similar results for the first transition marked in Fig. 11, i.e., a transition between two minima, with the transition state being a *first order* saddle



FIG. 13. (Color online) Energy of nearest extrema as a function of time for the first and the second transitions in Fig. 11, respectively. Note that in the upper panel we focus on an event in which the energy is increasing, and compare with Fig. 14.

point. It is therefore of interest to find the "reaction coordinate" for the transition, or the path on the potential surface which underlies the aging process. It is very difficult to do so using the molecular dynamics simulation, and to achieve this we used the method of "eigenvector following" (see the Appendix B for details). Starting from the saddle, one takes a small step in one of the directions of the negative eigenvalue and then uses the eigenvector following algorithm to trace the path leading to the minimum lying in this direction. This method works very well, resulting in the reaction coordinates shown in Fig. 14. It is important to notice that the energies of the minima are indeed in agreement with the results of the previous algorithm. We thus believe that the result shown connect indeed between the picture in time and the picture in energy landscape, and the reaction coordinate is the one corresponding to the temporal events shown in Fig. 11.

V. OTHER MODELS

The emerging insight from the analysis presented in the pervious section is that the reason for the slowing down in



FIG. 14. "Reaction coordinate" for the two marked transitions in Fig. 11. These graphs are obtained by starting at the saddle point and moving in the direction of the (unique) unstable eigenvector. Note that in general saddles are not guaranteed to be of order 1, and may have more than one unstable direction.

glass-forming systems is the generation of clusters of ground and close to ground states. These clusters have relatively high local shear moduli, since they are mechanically stable and can support locally a significant amount of shear without flowing. In between the clusters is where the aging events take place, and indeed in those regions the local shear moduli are small, even negative, depending on the resolution of the calculation of the local shear modulus. When the temperature gets lower, the clusters increase in size, but there is no point of crisis where the relaxation time diverges unless T=0 or for some reason the system crystallizes. Note that the lack of such finite temperature crisis was shown rigorously for this particular model of binary mixture in [9]. One question that we need to ask is whether this behavior is generic to glass formers or is it a peculiarity of systems in two dimensions or even to the present model of binary mixture in two dimensions.

In the context of our own work we have considered recently two other, very different, models of glass formers, i.e., the two-dimensional Shintani-Tanaka model [30] and the three-dimensional experimental system of dry glycerol [31]. In both systems it was demonstrated, using quantitative theory, that the scenario discussed here appears general, independent of the very different details and the different dimensionality characterizing these systems [7,32]. Thus, for example, we could show in the context of the Shintatni-Tanaka model that there exists a typical sale ξ that dominates the mean relaxation time as a function of the temperature. In other words, the τ_{α} relaxation time in this model could be quantitatively determined by a formula reading

$$\tau_{\alpha} = \tau_0 \exp(A\xi/T), \tag{8}$$

where A is a temperature-independent dimensional constant and τ_0 is the microscopic attempt time. A similar result was

TABLE I. NPT ensemble and NVT ensemble.

Phase	NPT	NPT	NVT	NVT
	hexagons of small and large particles	IIIIXture	hexagons of small and large particles	mixture
<i>r</i> ₁₁	1.031	1.003	1.039	0.995
r_{22}	1.376	1.473	1.386	1.462
r_{12}	_	1.187	_	1.179
P^*	13.5	13.5	12.2	14.9
$ ho^*$	0.781	0.77	0.77	0.781
U^*/N	2.881	2.922	2.648	3.179
H^*/N	20.168	20.453	18.752	22.257

demonstrated for the present binary model in [5]. The non-Arrhenius behavior of this formula is due to the strong increase of ξ when the temperature decreases, due to the increase in the size of the clusters, but without any singularity at any temperature T > 0. For the glycerol one could go even further, predicting not only the mean relaxation time but also computing the functional form of the dielectric response function in good relation to experiments. Here one argued that each cluster of s molecules contributes an exponential decay of its dipole moment with an s-dependent decay constant. Once averaged over all the clusters, the resulting response function is strongly nonexponential as is indeed observed. The main point is that the long relaxation times are again due to the large clusters (a smaller cluster may contribute β peaks when the conditions are right, cf. [32]). As long as the system does not crystallize there is no mechanism for a crisis of the type predicted by the Vogel-Fulcher formula or the Kauzmann argument. For a more detailed discussion of this important issue the reader is referred to [9].

VI. SUMMARY AND DISCUSSION

In summary, we focused on generic aging events in our glassy system to understand the relation between slowing down and spatial inhomogeneity. We have shown that the inhomogeneity can be seen in two complementary ways, one by observing the clusters of competing ground and close to ground states, and the other by measuring the local shear modulus. The latter quantity fluctuates from place to place, tending to be high and positive in the presence of clusters and low and even negative outside the clusters. Aging events occur in these mechanically unstable regions, and they result in a local reorganization, usually increasing the area spanned by clusters. We examined closely the nature of these aging events and showed again that they can be described in two complementary forms. One is a localized chain of movements, a so-called "excitation chain" (accompanied by a larger reversible collective motion) and the second is a transition between two inherent states, or local minima in the potential surface of the system, with the transition crossing a saddle point of order unity. Note that we could differentiate between the very localized permanent change and the "string-like" event that is in part reversible.

We proposed, on the basis of the present model of a binary mixture and other systems that were analyzed recently in the same spirit, that the results obtained here are generic, and that there exists a wide class of glass-forming systems whose theoretical understanding calls for an analysis of its distribution of clusters of competing ground and close to ground states; in a future paper we will show that the temporal properties of such systems can be completely understood in these terms.

APPENDIX A: GROUND STATE

The average potential energy per particle in a binary mixture with the interactions defined by the potential (1) is given by

$$\frac{U^*}{N} = \frac{1}{2N} \sum_{i \neq j} \left(\frac{\sigma_{ab}}{r_{ij}} \right)^{12},\tag{A1}$$

where r_{ij} is the distance between particles *i* and *j*.

For the equimolar system of $N \rightarrow \infty$ separated small and large particles (see the upper panel of Fig. 2) the interactions between the two species is negligible $[o(1/\sqrt{N})]$ and the total energy (A1) can be written as

$$\frac{U^*}{N} = \frac{3}{2} \left[\left(\frac{\sigma_{11}}{r_{11}} \right)^{12} + \left(\frac{\sigma_{22}}{r_{22}} \right)^{12} \right], \tag{A2}$$

where r_{11} and r_{22} are the distances between small and large particles, respectively.

The area of a hexagon is $\frac{\sqrt{3}}{2}r^2$, where *r* is the distance between particles. The dimensionless total area $A^* = 1/\rho^*$ per particle of the system of separated small and large particles is given by

$$\frac{A^*}{N} = \frac{\sqrt{3}}{4} (r_{11}^2 + r_{22}^2).$$
(A3)

The enthalpy depends on two variables r_{11} and r_{22} , so the minimum of this function is defined by two equations $\partial H^* / \partial r_{11} = 0$ and $\partial H^* / \partial r_{22} = 0$. Solutions of these equations and evaluated values for the energy, enthalpy, and density are displayed in Table I.

In the configuration where small and large particles are mixed (the lower panel of Fig. 2) each small particle interacts with one small particle and four large particles. In turn, each large particle interacts with three neighboring large particles and four neighboring small particles. Therefore the general expression (A1) is simplified to

$$\frac{U^*}{N} = \frac{1}{4} \left[\left(\frac{\sigma_{11}}{r_{11}} \right)^{12} + 8 \left(\frac{\sigma_{12}}{r_{12}} \right)^{12} + 3 \left(\frac{\sigma_{22}}{r_{22}} \right)^{12} \right], \quad (A4)$$

where in addition to Eq. (A2) the contribution of interactions between small and large particles separated by r_{12} are taken into account.

Any configuration of homogeneously mixed small and large particles can be constructed using a "boat" unit [18] (see the lower panel of Fig. 2) which consists of one "thin" and three "fat" rhombi. There are 1/4N "thin" rhombi and 3/4N "fat" rhombi in the system, so the area per particle is given by

$$\frac{A^*}{N} = \frac{1}{4}(A_T^* + 3A_F^*), \tag{A5}$$

where A_T^* is the area of a thin rhombus and A_F^* that of a fat one.

It follows from the boat design that the distances between particles are not independent. It is useful to introduce a new variable α , the small angle of the fat rhomb, so that the large angle of the thin rhomb is $2\pi - 3\alpha$. All rhombi have sides of the same size r_{12} and the distances between small-small and large-large particles can be expressed as

$$r_{11} = -2r_{12}\cos\left(\frac{3}{2}\alpha\right),$$

$$r_{22} = 2r_{12}\sin\left(\frac{1}{2}\alpha\right).$$
(A6)

The area per particle of the mixed system (A5) as a function of these variables is given by

$$\frac{A^*}{N} = \frac{1}{4}r_{12}^2[-\sin(3\alpha) + 3\sin(\alpha)] = r_{12}^2\sin^3(\alpha).$$
 (A7)

Substitution of Eq. (A6) into Eq. (A4) yields the following expression for the energy:

$$\frac{U^*}{N} = \frac{1}{4} \left(\frac{1}{2r_{12}}\right)^{12} \left[\left(\frac{\sigma_{11}}{\cos\left(\frac{3}{2}\alpha\right)}\right)^{12} + 8(\sigma_{12})^{12} + 3\left(\frac{\sigma_{22}}{\sin\left(\frac{1}{2}\alpha\right)}\right)^{12} \right].$$
(A8)

The minimum of the enthalpy in this case is defined by equations $\partial H^* / \partial r_{12} = 0$ and $\partial H^* / \partial \alpha = 0$. The solution for the angle α depends only on the ratio $\sigma_{22} / \sigma_{11}$ and for the system under consideration $\alpha = 76.65^{\circ}$. The interparticle distances, energy, enthalpy, and density are shown in Table I.

In the *NVT* ensemble it is necessary to find the minimum of the energy given by Eq. (A2) or Eq. (A8) under the constraints (A3) or (A7), respectively. The calculation results are also presented in Table I.

APPENDIX B: LOCATING EXTREMA

Finding a local minimum of a function of several variables is a common problem in applied math. There are several methods for doing this; some of them like the Newton-Raphson method are quite simple. Finding a saddle point, on the other hand, is a more subtle problem since it is not clear if the appropriate direction to go is up or downhill. In our work we have used two different methods to overcome this problem.

1. Square gradient minimization

One way of finding saddle points is by finding a local minimum of the squared modulus of the gradient of the original potential function:

$$W(\mathbf{x}) = |\nabla U(\mathbf{x})|^2 = \sum_{i=1}^{N} \sum_{\alpha=1}^{2} \left(\frac{\partial U}{\partial x_{i,\alpha}}\right)^2.$$
 (B1)

Since $W(\mathbf{x})$ is non-negative, at the absolute minima $W(\mathbf{x}) = 0$, which implies $\nabla U = 0$ (a saddle point). An efficient way to minimize such functions is the Newton-Raphson method. At each algorithm step $x_{n+1} = x_n + \delta x_n$ the following equation should be satisfied:

$$\nabla U(x_n + \delta x_n) = 0. \tag{B2}$$

This equation is solved to first order:

$$\nabla U(x_n) + \delta x_n \nabla^2 U(x_n) = 0.$$
 (B3)

Inverting for δx_n one finds

$$\delta x_n = -\nabla^2 U(x_n) \nabla U(x_n) = -\hat{H} \nabla U(x_n).$$
(B4)

We have used a publicly available numerical algorithm—the LBFGS algorithm ([33,34]) which uses an approximate version of the Newton method. The biggest drawback in minimizing the function $W(\mathbf{x})$ instead of minimizing $U(\mathbf{x})$ directly is that the minimization often converges to a local minimum of $W(\mathbf{x})$, which is not a saddle point of $U(\mathbf{x})$. Therefore, when minimizing a molecular dynamics realization of a system traveling in the energy landscape, not all of the instantaneous states converge to a minimum or a saddle of $V(\mathbf{x})$ and some information is lost.

2. Eigenvector following

In order to demonstrate the topological connectedness of two minima separated by a saddle we have used the eigenvector following the algorithm as described by Wales and co-workers [35–37] (the original idea for the algorithm was proposed by Cerjan and Miller [38]). At each iteration a step Δx is proposed, which in the base that diagonalizes the Hessian (at the specific point in the algorithm path) is [36,37]

$$\Delta x_{\mu} = S_{\mu} \frac{2g_{\mu}}{|h_{\mu}|(1 + \sqrt{1 + 4g_{\mu}^2/h_{\mu}^2})},$$
 (B5)

where h_{μ} are the eigenvalues of the Hessian and g_{μ} are the components of the gradient in the diagonal base (Δx_{μ} is set to 0 for the directions where $h_{\mu}=0$; i.e., uniform displacements). The sign $S_{\mu}=\pm 1$ is chosen as explained below. Note that as $g_{\mu}\rightarrow 0$,

$$\Delta x_{\mu} = -\frac{g_{\mu}}{h_{\mu}} + O(g_{\mu}^2), \quad g_{\mu} \to 0,$$
 (B6)

where the first term is the Newton-Raphson step.

If S_{μ} =-1 the algorithm converges to a minimum along μ . Therefore starting from a saddle of order one, setting all of the S_{μ} to -1 we were able to reach the two minima separated by this saddle.

- For a balanced review in which many of the common concepts and terminology are well spelled out see, M. D. Ediger, C. A. Angell, and S. R. Angell, J. Phys. Chem. **100**, 13200 (1996).
- [2] See E. Donth, *The Glass Transition* (Springer, Berlin, 2001), and references therein.
- [3] J. C. Dyre, Rev. Mod. Phys. 78, 953 (2006).
- [4] L. Berthier, G. Biroli, J.-P. Bouchaud, L. Cipelletti, D. El Masri, D. L'Hôte, F. Ladieu, and M. Pierno, Science 310, 1797 (2005).
- [5] E. Aharonov, E. Bouchbinder, H. G. E. Hentschel, V. Ilyin, N. Makedonska, I. Procaccia, and N. Schupper, Europhys. Lett. 77, 56002 (2007).
- [6] H. G. E. Hentschel, V. Ilyin, N. Makedonska, I. Procaccia, and N. Schupper, Phys. Rev. E 75, 050404(R) (2007).
- [7] V. Ilyin, E. Lerner, T.-S. Lo, and I. Procaccia, Phys. Rev. Lett. 99, 135702 (2007).
- [8] L. Buisson, S. Ciliberto, and A. Garcimartín, Europhys. Lett. 63, 603 (2003).
- [9] J.-P. Eckmann and I. Procaccia, e-print arXiv:0802.4346.
- [10] J. S. Langer, Phys. Rev. Lett. 97, 115704 (2006).
- [11] F. H. Stillinger and T. A. Weber, Phys. Rev. A 25, 978 (1982).
- [12] D. Deng, A. S. Argon, and S. Yip, Philos. Trans. R. Soc. London, Ser. A **329**, 549 (1989); **329**, 575 (1989); **329**, 595 (1989); **329**, 613 (1989).
- [13] D. N. Perera and P. Harrowell Phys. Rev. E 59, 5721 (1999), and references therein; J. Chem. Phys. 111, 5441 (1999).
- [14] M. P. Allen and D. J. Tildesley, Computer Simulations of Liquids (Clarendon, Oxford, 1987).
- [15] D. W. Heerman, Computer Simulation Methods in Theoretical Physics (Springer-Verlag, Berlin, 1990).
- [16] J. Stavans, Rep. Prog. Phys. 56, 733 (1993).
- [17] S. Goncalves and J. R. Iglesiast, J. Phys.: Condens. Matter 7, 2395 (1995).

- [18] H. K. Lee, R. H. Swendsen, and M. Widom, Phys. Rev. B 64, 224201 (2001).
- [19] J. Tobochnik and G. V. Chester, Phys. Rev. B 25, 6778 (1982).
- [20] C. A. Croxton, *Liquid State Physics—A Statistical Mechanical Introduction* (Cambridge University Press, Cambridge, England, 1974).
- [21] T. B. Schrøder, S. Sastry, J. C. Dyre, and S. C. Glotzer, J. Chem. Phys. **112**, 9834 (2000).
- [22] V. Ilyin, N. Makedonska, I. Procaccia, and N. Schupper, Phys. Rev. E 76, 052401 (2007).
- [23] J. Lutsko, J. Appl. Phys. 64, 1152 (1988).
- [24] K. Yoshimoto, T. S. Jain, K. Van Workum, P. F. Nealey, and J. J. de Pablo, Phys. Rev. Lett. 93, 175501 (2004).
- [25] R. Zwanzig and R. M. Mountain, J. Chem. Phys. 43, 4464 (1965).
- [26] D. C. Wallace, Phys. Rev. E 62, 3077 (2000).
- [27] M. Goldstein, J. Chem. Phys. 51, 3728 (1969).
- [28] T. S. Grigera, A. Cavagna, I. Giardina, and G. Parisi, Phys. Rev. Lett. 88, 055502 (2002), and references therein.
- [29] C. Donati, J. F. Douglas, W. Kob, S. J. Plimpton, P. H. Poole, and S. C. Glotzer, Phys. Rev. Lett. 80, 2338 (1998).
- [30] H. Shintani and H. Tanaka, Nat. Phys. 2, 200 (2006).
- [31] A. Puzenko, Y. Hayashi, Y. E. Ryabov, I. Balin, Y. Feldman, U. Kaatze, and R. Behrends, J. Phys. Chem. B 109, 6031 (2005).
- [32] H. G. E. Hentschel and I. Procaccia, Phys. Rev. E 77, 031507 (2008).
- [33] D. C. Liu and J. Nocedal, Math. Program. 45, 503 (1989).
- [34] http://www.netlib.org/opt/index.html.
- [35] D. J. Wales and T. R. Walsh, J. Chem. Phys. 105, 6957 (1996).
- [36] D. J. Wales, J. Chem. Soc., Faraday Trans. 89, 1305 (1993).
- [37] D. J. Wales, J. Chem. Phys. 101, 3750 (1994).
- [38] C. Cerjan and W. H. Miller, J. Chem. Phys. 75, 2800 (1981).