Bypassing sluggishness: SWAP algorithm and glassiness in high dimensions

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The recent implementation of a swap Monte Carlo algorithm (SWAP) for polydisperse glass forming mixtures bypasses computational sluggishness and closes the gap between experimental and simulation timescales in physical dimensions d=2 and 3. Here, we consider suitably optimized systems in $d=2,3,\ldots,8$ to obtain insights into the performance and underlying physics of SWAP. We show that the speedup obtained decays rapidly with increasing the dimension. SWAP nonetheless delays systematically the onset of the activated dynamics by an amount that remains seemingly finite in the limit $d \to \infty$. This shows that the glassy dynamics in high dimensions d > 3 is now computationally accessible using SWAP, thus opening the door for the systematic consideration of finite-dimensional deviations from the mean-field description.

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Introduction. A glass emerges when a supercooled liquid past its crystallization point becomes so sluggish that it falls out of equilibrium. Upon cooling or increasing packing fraction, the dynamics of glass formers exhibits a marked slowdown beyond the dynamical onset, thus making this outcome inescapable [1,2]. In mean-field descriptions, the structural relaxation time exhibits a power-law divergence at the dynamical transition [3]. In any finite dimension, although activated processes wash out this transition, the rapid growth of the associated relaxation time nonetheless impedes equilibration of low-temperature or high-density liquids. Standard simulation protocols, in particular, do not easily explore the regime beyond the dynamical transition, because structural relaxation is already too sluggish.

The application of the swap Monte Carlo algorithm (SWAP), which exchanges the identity of pairs of particles, to complex mixtures sidesteps this difficulty [4-6]. By considering systems with, for instance, a continuous size polydispersity, one can follow the equilibrium liquid up to unprecedented high packing fractions or low temperatures. Tuning the range and functional form of polydispersity provides systems for which the sampling efficiency of swap moves is maximal within the liquid state, while remaining robust against crystallization and fractionation [7], and keeping equilibrium properties unaffected. For properly chosen polydispersities in d=3this procedure has recently provided a speedup of at least 10¹⁰ compared to standard dynamics, matching the experimental timescales [7,8], and in d = 2 it has given access to timescales that are truly cosmological [9]. This computational progress has triggered the exploration of new glass physics in computer simulations, notably low-temperature anomalies [10,11], the Gardner transition [10,12], the rheology of glasses [13], the extension of the jamming line [14], and the ultrastability of vapor-deposited glasses [15].

The efficiency of SWAP has also triggered theoretical activity aimed at better understanding its physical origin and its physical implications for the glass transition [16]. Ikeda et al. [17] present a replica calculation of a mean-field glass model proposing that SWAP and physical dynamics are ruled by distinct dynamical transitions. A qualitatively similar result is obtained by Szamel who obtains two dynamical transitions for the two dynamics [18]. Brito et al. [19] obtain a similar result, and interpret the dynamical transition as an onset of mechanical rigidity that is again shifted by SWAP. Finally, Berthier et al. [20] argue that the onset of thermal activation past the dynamical transition is also considerably affected by SWAP. There is thus a general consensus that SWAP can delay the (avoided) dynamical transition by an amount that is system dependent, and can speed up the dynamics even past the dynamical transition.

However, because dynamical transitions are avoided in any finite d [21], other physical processes might also explain the dramatic change in dynamics. In particular, structural imperfections closely tied to local geometry [22], which are putatively important in the dynamics of low-dimensional glass formers, could impact SWAP efficiency. Distinguishing one contribution from the other can be achieved by considering how SWAP performance evolves with increasing d. A nonvanishing SWAP efficiency in the limit of $d \to \infty$ or a perturbative correction in 1/d would suggest that the meanfield dynamical transition is indeed shifted, while an exponential suppression would suggest that nonperturbative features associated with geometry dominate. Because numerical work on SWAP has thus far only been concerned with physical dimensions, d = 2 and 3, distinguishing between these scenarios is not currently possible.

Resolving this question would not only shed light on the physical origin of the glassy slowdown, but help devise novel algorithms that further bypass it. Interestingly, sidestepping the mean-field dynamical threshold could also be key to general algorithmic improvements in hard problems, such as

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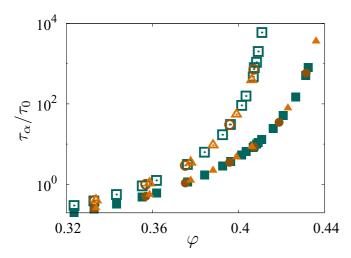


FIG. 1. Structural relaxation time of standard (open symbols) and SWAP dynamics (solid symbols) for various particle size distributions $P(\sigma)$ with $\Delta=10\%$ in d=4: flat (circles), $1/\sigma^3$ (squares), and $1/\sigma^4$ (triangles). For a given Δ both dynamics are essentially unaffected by the functional form of $P(\sigma)$.

statistical inference, high-dimensional optimization, and deep learning [23]. A fundamental grasp of the effectiveness of SWAP dynamics could thus bolster advances far beyond the problem at hand. More immediately, if one could generically push the current limitations of high d simulations, crucial questions in glass physics could be tackled [21,24,25]. In this Rapid Communication, we study the dynamics of suitably optimized polydisperse mixtures of hard spheres in various spatial dimensions, so as to systematically approach the meanfield, $d \to \infty$ description, and provide microscopic insight into the underlying physics and computational efficiency across a broad range of dimensions.

Simulation model. We consider size polydisperse systems with N hard spheres in a hypercubic box of constant volume V, under periodic boundary conditions in $d = 2, 3, \dots, 8$. We choose N = 2000 for $d \le 6$ and N = 7000 in d = 8 [26]. The size distribution function has the form $P(\sigma) = K/\sigma^3$, with normalization constant K for $\sigma \in [\sigma_{\min}, \sigma_{\max}]$, where σ_{\min} and $\sigma_{\rm max}$ are the minimum and the maximum diameter values, respectively. The average diameter $\bar{\sigma}=\int_{\sigma_{\rm min}}^{\sigma_{\rm max}} P(\sigma)\sigma d\sigma$ sets the unit of length, and the standard deviation of the size distribution, Δ , quantifies the degree of polydispersity (see simulation details and model parameters in Ref. [27]). For a fixed Δ , this specific choice of size distribution function does not significantly affect the system dynamics. Figure 1, which explicitly compares the dynamics at fixed Δ and various $P(\sigma)$ in d = 4, confirms that Δ is the most relevant variable. Our analysis should therefore be reasonably independent of the specifics of the model studied.

Standard and SWAP simulations are run for different Δ and d. Both dynamical protocols include basic single-particle translational moves along a vector randomly drawn within a d-dimensional hypercube of side $\delta \ell$; SWAP includes additional diameter exchanges between two randomly chosen particles, attempted with probability p=0.2 (setting p=0 recovers standard dynamics). While $0 monotonically increases sampling efficiency, for <math>p \gtrsim 0.2$ efficiency

saturates, and hence additional swap moves wastefully slow down simulations [7]. For each volume fraction φ , the pressure P is measured using pair correlations [28,29], to compute the unitless reduced pressure $Z = \beta P/\rho$ for the number density $\rho = \varphi/\bar{V}_{\rm d}$, with $\bar{V}_{\rm d}$ being the average volume of a d-dimensional hypersphere.

Equilibration is assessed by the complete decay of the selfpart of the particle-scale overlap function

$$Q(t) = \frac{1}{N} \sum_{i=1}^{N} \Theta(a - |\mathbf{r}_i(t) - \mathbf{r}_i(0)|), \tag{1}$$

where Θ is a step function and $a=0.3\bar{\sigma}$ is a microscopic length chosen to be close to the typical particle cage size; Q(t) thus represents the fraction of particles with displacement smaller than a after t Monte Carlo (MC) sweeps, each consisting of N moves. The associated structural relaxation time τ_{α} is defined such that $Q(\tau_{\alpha})=e^{-1}$. We define the relaxation time for both the standard ($\tau_{\alpha}^{\rm std}$) and SWAP ($\tau_{\alpha}^{\rm swap}$) dynamics. In all dimensions studied, SWAP equilibrates systems far beyond what is computationally accessible with standard Monte Carlo. After achieving equilibration with SWAP, the system is evolved using standard dynamics.

Results. In physical dimensions, crystallization competes with equilibration of deeply supercooled liquids [30]. For instance, for $\Delta \lesssim 8\%$ in d=3 crystallization at high φ is unavoidable. For d > 3, by contrast, crystallization does not interfere with the metastable fluid phase even for arbitrarily low Δ . The nucleation time at finite Δ in d > 3 is thus as equally out of computational reach as it is for monodisperse systems ($\Delta = 0$) [31–33]. In all d, however, size fractionation may take place at high Δ and φ . In d=3, fractionation appears at $\Delta \gtrsim 10\%$, which helps crystallization [34,35]. In practice, this only happens when SWAP is used [20], because composition fluctuations leading to fractionation are then much faster. SWAP thus not only accelerates the sampling of the metastable fluid, but also changes the glass-forming ability of the system and forces the use of $\Delta > 20\%$ in d = 3. In d=4, by contrast, fractionation only appears at $\varphi \gtrsim 0.43$ for $\Delta \gtrsim 15\%$, and is further suppressed at higher Δ (see dynamic and static observables in Ref. [27]). For each d, a Δ window, within which SWAP efficiency is reasonably good and fractionation (with or without crystallization) does not interfere, can thus be found. Qualitative and even quantitative aspects of the standard Monte Carlo dynamics are otherwise not remarkably affected by changing Δ , as expected from previous studies of naturally polydisperse systems, such as colloidal suspensions [36].

A strong dependence of the SWAP dynamics on Δ is observed in the dynamically sluggish regime, beyond the onset of slow diffusion at φ_0 [Figs. 2(a)–2(d)]. As an illustration, we consider the evolution of the SWAP efficiency ratio $\tau_\alpha^{\rm std}/\tau_\alpha^{\rm swap}$ measured at a fixed $\tau_\alpha^{\rm std}/\tau_0$ with $\tau_0 \equiv \tau_\alpha(\varphi_0)$. In Fig. 3(a), we specifically consider $\tau_\alpha^{\rm std}/\tau_0 = 5 \times 10^3$, but the results are qualitatively robust for $\tau_\alpha^{\rm std}/\tau_0 > 1$ (see dynamical and static observables in Ref. [27]). At low Δ , SWAP dynamics is indistinguishable from standard dynamics and its efficiency increases monotonically. This efficiency, however, essentially saturates beyond a certain Δ , resulting in its overall sigmoidal growth. We empirically fit the results to a generalized lo-

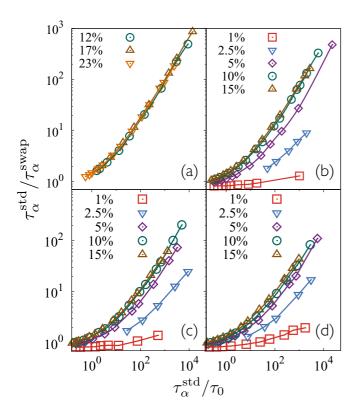


FIG. 2. SWAP efficiency $\tau_{\alpha}^{\rm std}/\tau_{\alpha}^{\rm swap}$ as a function of the relaxation time of the standard dynamics (representing the sluggishness) for different polydispersities Δ in (a) d=3, (b) d=4, (c) d=5, and (d) d=6. Sluggish dynamics at low Δ cannot be reached in d=3 because crystallization interferes. In all d, SWAP performs better as Δ increases, and saturates at larger Δ .

gistic function, $S(\Delta) = A \exp(a\Delta)/[B + \exp(b\Delta)]$, with fit parameters A, a, b, and B (where A is the plateau height, B controls the onset of saturation, and a, $b \approx 1$), to quantify the crossover polydispersity Δ_0 defined such that $S(\Delta_0) = 0.9A$. We obtain $\Delta_0 \approx 10\%$ in d = 3, $\approx 7.5\%$ in d = 4, $\approx 7\%$ in d = 5,

and $\approx 6.5\%$ in d = 6. In d = 2 and 3, the trend is almost hidden by crystallization, and had gone unnoticed in previous work. The shrinking of Δ_0 with increasing d is nonetheless very clear. No theoretical framework formally predicts the saturation with Δ and the associated scaling with dimension. Physically, we interpret these results as follows. The amplitudes of particle size fluctuations, which help uncage particles in SWAP dynamics, increase with Δ , which accounts for the initial growth of efficiency with Δ . The diffusion of particle diameters beyond a typical size, however, itself becomes slower than the structural relaxation when Δ is large, because the diameter and position dynamics are intimately coupled [7]. Increasing Δ thus no longer improves SWAP efficiency, and this saturation develops earlier in larger d, where the vibrational dynamics (or, loosely speaking, caging) itself occurs over a length scale decreasing with d.

The most remarkable feature of the efficiency results is the weakening of SWAP performance with increasing d. Figure 3(b) shows that the efficiency decays rapidly with increasing d (nearly exponentially, at least up to d=8) for various $\tau_{\alpha}^{\rm std}/\tau_0$. The decay of SWAP performance becomes more prominent when estimated beyond the accessible regimes of the standard dynamics, such as where $\tau_{\alpha}^{\rm swap}/\tau_0 = 5 \times 10^3$ —see Fig. 3(c) (and dynamical and static observables in Ref. [27]).

In order to examine explicitly whether or not this strong suppression is due to nonperturbative effects, we consider how SWAP impacts the avoided mean-field dynamical transition φ_d . We estimate φ_d for both standard dynamics and SWAP by fitting the growth of the relaxation time to the critical scaling form $\tau_{\alpha} \propto (\varphi_d - \varphi)^{-\gamma}$ [21] (see the mode coupling analysis in Ref. [27]). As expected [37], this scaling form captures the data increasingly well as d increases. In d=2, it does not have a good regime of validity, but its validity eventually reaches up to three decades in the computationally accessible regime. We find that γ is fairly insensitive to both dimension [21] and polydispersity [38]. Three features of the results are particularly noteworthy. First, collapsing $\tau_{\alpha}^{\rm std}/\tau_0$ by rescaling $\varphi/\varphi_{\rm d}^{\rm std}$ clearly reveals that SWAP postpones the pu-

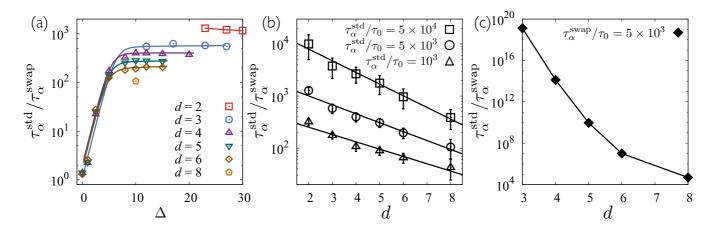


FIG. 3. (a) Evolution of SWAP efficiency $\tau_{\alpha}^{\rm std}/\tau_{\alpha}^{\rm swap}$ with Δ , measured at $\tau^{\rm std}/\tau_0=5\times10^3$. Saturation to a plateau height, A, occurs at a lower Δ and at a lower A as d increases. (b) Same quantity (at saturation) measured at $\tau_{\alpha}^{\rm std}/\tau_0=5\times10^4$, 5×10^3 , and 10^3 as a function of d. Exponential fits are denoted as solid lines. (c) Estimated SWAP efficiency at $\tau_{\alpha}^{\rm swap}/\tau_0=5\times10^3$, which is roughly the edge of the computationally accessible regime using SWAP. The dramatic speedup in d=3 decreases rapidly with d, but remains larger than four decades in d=8.

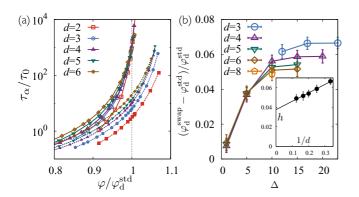


FIG. 4. (a) The structural relaxation time $\tau_{\alpha}^{\rm std}$ (open symbols) collapses for different d upon rescaling $\varphi/\varphi_{\rm d}^{\rm std}$. The gap between SWAP and standard dynamics shrinks and saturates with increasing d. For d=2 and 3, we set $\Delta=23\%$ and for d>3, $\Delta=10\%$. (b) The relative gap between the dynamical transition densities for the two different dynamics $(\varphi_{\rm d}^{\rm sw}-\varphi_{\rm d}^{\rm std})/\varphi_{\rm d}^{\rm std}$ also saturates at high Δ (inset) and the saturation height asymptotically approaches a nonzero value (≈ 0.037) as $\sim 1/d$.

tative dynamical transition in all dimensions [Fig. 4(a)]. Second, while φ_d monotonically grows with polydispersity [39], its relative impact $(\varphi_d^{swap} - \varphi_d^{std})/\varphi_d^{std}$ eventually plateaus on a scale consistent with the estimates for Δ_0 [see Fig. 4(b)]. This suggests that the shift of dynamical transition is directly correlated with the SWAP efficiency, as both quantities evolve similarly with Δ . Third, the plateau height, $h \equiv (\varphi_d^{swap} - \varphi_d^{std})/\varphi_d^{std}$ at the maximum polydispersity considered in Fig. 4(b), decays to a nonzero value (\approx 0.037) with a correction that scales with dimension as \sim 1/d. Our results thus suggest that the gain in SWAP efficiency survives in the limit $d \to \infty$, and that perturbative corrections survive all the way down to d = 3, independently of nonperturbative effects.

How can one explain the relatively rapid suppression of swap efficiency despite the slow decay of the density gap $(\varphi_d^{\text{swap}} - \varphi_d^{\text{std}})/\varphi_d^{\text{std}}$ to a nonzero value? While the relative increase of φ_d is qualitatively consistent with mean-field treatments in d=3 [17,18], the saturation and the asymptotic behavior of the gap with d were not anticipated. Plugging this result into the critical scaling forms $\tau_{\alpha} = A(\varphi_d - \varphi)^{-\gamma}$, and we obtain an approximate expression for the efficiency ratio

$$\tau_{\alpha}^{\text{std}}/\tau_{\alpha}^{\text{swap}} \approx \tau_{\alpha}^{\text{std}} (h\varphi_d^{\text{std}})^{\gamma}.$$
 (2)

For a given value of $\tau_{\alpha}^{\rm std}$, the key contribution to the efficiency gain therefore arises from the term $(h\varphi_{\rm d}^{\rm std})^{\gamma}$. Because asymptotically $\varphi_{\rm d}^{\rm std} \sim d \, 2^{-d}$ [21], this gain decreases rapidly with increasing d—qualitatively consistent with Figs. 3(b), 3(c)

and 4(b). Because τ_{α} diverges upon approaching $\varphi_{\rm d}$ in high dimension, however, one should always be able to identify sufficiently sluggish systems for SWAP to speed up sampling. In intermediate dimensions, the approach remains sufficiently productive to obtain equilibrium configurations much beyond the dynamical transition of the standard dynamics. Figure 3(c) provides a rough estimate of how useful SWAP can be in accessing regimes that are not accessible by the standard dynamics in high dimensions. For instance, in d=8 a speedup of roughly 10^4 should remain computationally achievable.

Conclusion. We have shown that SWAP improves sampling in dimensions $d \ge 2$ by generically delaying the dynamical transition that indicates the emergence of activated dynamics in the standard dynamics. This finding in itself does not directly reveal the microscopic nature (dynamical or thermodynamic) of the standard dynamics in the regime $\varphi_d^{\rm std} < \varphi < \varphi_d^{\rm swap}$, where SWAP provides most of its dynamical speedup, but offers a platform for assessing this question in the future. Because the gap between the dynamical transition of the standard and the SWAP dynamics remains finite in the limit $d \to \infty$, SWAP can efficiently be used to study pure glass physics in reasonably large dimensions, far from the regime in which significant local structure [22] or orientational ordering [40] might interfere. In other words, although caging imperfections go away exponentially quickly with increasing dimension, SWAP can still break cages in high d. Even within this analysis, the two-dimensional speedup is remarkably large, and techniques specifically tailored to identify local structural weaknesses, (e.g., Refs.[41–44]) might thus help obtain additional microscopic insights. More generally, our observations suggest that the standard dynamical transition might not be as strong an algorithmic constraint as previously conceived in problems ranging from physics to information theory. If a proper sampling scheme can be devised and exploited in those problems, other stunning algorithmic advances might thus be within reach.

Data associated with this work are available from the Duke Digital Repository [47].

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- of the (avoided) dynamical transition. Because the dynamical quantities studied in this work do not couple strongly to this length, no significant finite-size effects are expected.
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