

Simulations of a lattice model of two-headed linear amphiphiles: Influence of amphiphile asymmetry

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Using a 2D lattice model, we conduct Monte Carlo simulations of micellar aggregation of linear-chain amphiphiles having two solvophilic head groups. In the context of this simple model, we quantify how the amphiphile architecture influences the critical micelle concentration (CMC), with a particular focus on the role of the asymmetry of the amphiphile structure. Accordingly, we study all possible arrangements of the head groups along amphiphile chains of fixed length $N = 12$ and 16 molecular units. This set of idealized amphiphile architectures approximates many cases of symmetric and asymmetric gemini surfactants, double-headed surfactants, and bolofom surfactants. Consistent with earlier results, we find that the number of spacer units s separating the heads has a significant influence on the CMC, with the CMC increasing with s for $s < N/2$. In comparison, the influence of the asymmetry of the chain architecture on the CMC is much weaker, as is also found experimentally. © 2011 American Institute of Physics. [doi:10.1063/1.3593404]

I. INTRODUCTION

Amphiphile molecules having two solvophilic head groups and the overall topology of a linear chain (or at least approximating that of a linear chain) represent an important class of surfactant. Most prominently, this class includes the symmetric gemini surfactants, in which two conventional single-headed amphiphiles are joined at or near the head groups by a linear spacer.^{1,2} Other examples of two-headed quasi-linear amphiphiles are certain species of double-headed³ and bolofom surfactants.^{4,5}

The relationship between the architecture of amphiphilic molecules and their bulk properties in solution has long been a central theme in the study of surfactants. This is certainly true for gemini surfactants, for which numerous studies (reviewed in Ref. 2) have demonstrated the influence of the length of the solvophobic tails, head group size, and the length of the spacer, in controlling important properties of the solution, for example, the critical micelle concentration (CMC), and the typical size of the micelles formed (often referred to as the “aggregation number”).

Although much of the experimental work on two-headed amphiphiles has focussed on the symmetric gemini surfactants, there are of course a large number of distinct two-headed linear architectures that are asymmetric. Indeed, over the last decade there has been an emerging interest in the properties of “dissymmetric gemini surfactants” in which the two solvophobic tails are of unequal length.^{6–18} In particular, Thomas and co-workers^{10,14} found that the CMC decreases by ~35% as the asymmetry of the amphiphile increases. However, the magnitude of this decrease is small compared to the

effect on the CMC of other structural factors of gemini surfactants, such as head spacing, which can change the CMC by an order of magnitude or more.^{2,19} At the same time, changes in micelle size and morphology have been reported for dissymmetric gemini surfactants as a function of asymmetry.^{6,9,12} It would be useful to quantify the degree to which the purely geometrical influence of asymmetry is responsible for these changes or if their origins are more subtle.

Several simulation studies have been carried out on specific cases of two-headed linear amphiphiles.^{20–28} These studies have confirmed the dominant influence of tail length and head spacing on the properties of the solution.^{21,22,26} However, to our knowledge no simulation studies have focussed specifically on the question of the extent to which the asymmetry of the amphiphile architecture influences, e.g., the CMC. This is somewhat surprising given that symmetry, or lack thereof, is a fundamental property of any molecule and is often an important one for determining bulk properties, e.g., the structure and properties of the crystalline phases.

In this work, we present computer simulation results that examine the properties of every member of a family of two-headed linear amphiphiles for a given fixed N , in the context of a simple 2D lattice model. We idealize a two-headed linear amphiphile as a flexible, linear chain made up of N units. $N - 2$ of these units are solvophobic tail units, while the remaining two are solvophilic head units. These head units can be separately located at any point along the chain. Viewed in this purely geometric way, dozens of distinct architectures of such amphiphiles are possible even for the modest values of $N \leq 16$. At the same time, each amphiphile architecture (for fixed N) can be precisely specified in terms of only two independent parameters, e.g., the positions along the chain of the two heads. From the standpoint of theory and computer

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modelling, this family of amphiphiles thus provides an interesting case for studying the influence of structural parameters on solution properties, in the sense that it is rich enough to be interesting, but simple enough to be tractable.

Our goals are twofold: (i) We wish to characterize a complete set of amphiphile structures in order to assess, in a comprehensive way, the influence of architectural parameters on the behavior of the resulting solutions. (ii) In particular, we wish to quantify the influence of asymmetry on system properties, relative to other architectural parameters. To achieve these goals, we use a simple 2D lattice model of an amphiphile solution, of the type introduced by Larson,^{29–33} and sample its equilibrium properties using Monte Carlo (MC) dynamics. Simulation models of this type have been shown to reproduce a wide range of qualitative behavior observed for surfactant solutions and continue to play an important role in the exploration of these systems.^{21,22,24,34–50} To our knowledge, an exhaustive examination via simulations of all possible amphiphile architectures of a given class, on the scale presented here, has not been carried out to date. As shown below, our use of a simple 2D lattice model allows such a survey to be completed on a reasonable computational time scale. While our results will necessarily be qualitative in nature, a simple lattice model allows us to focus on effects that are purely geometrical in origin, including those due to amphiphile asymmetry.

II. METHODS

A. Model

Our model consists of n_s linear amphiphile molecules on a two-dimensional $L \times L$ square lattice. Unless otherwise noted, all data presented are for $L = 200$. Each amphiphile is represented as a linear chain of N connected sites. In this work, we present results for $N = 12$ and 16 . Each site along an amphiphile is either solvophobic (labelled T for “tail”) or solvophilic (labelled H for “head”). Note that any spacer units occurring between the two heads are considered to be equivalent to tail units. All other sites of the lattice are occupied by solvent molecules (labelled S). The total energy of the system is given by

$$\mathcal{H} = \sum_{\alpha\beta} \epsilon_{\alpha\beta} n_{\alpha\beta}, \quad (1)$$

where the sum is taken over all possible nearest-neighbor pairs (α, β) of the species types T, H, and S. $\epsilon_{\alpha\beta}$ is the interaction energy of two nearest-neighbor sites on the lattice occupied by species α and β and $n_{\alpha\beta}$ is the number of nearest-neighbor (α, β) contacts occurring in the system.

Numerous simulations of micelle-forming systems have been carried out using a Hamiltonian of the form of Eq. (1).^{21,22,24,34–50} A wide range of choices for the interaction parameters have been shown to give systems exhibiting stable micelles. In this work, we set $\epsilon_{TS} = +1$, $\epsilon_{TH} = +1$, $\epsilon_{HS} = -1$, and $\epsilon_{HH} = +2$, with all other interaction energies set to zero. The positive HH interaction is meant to model a screened electrostatic repulsion of the head groups. The positive TS and TH interactions model the solvophobic na-

ture of the tails, while the negative HS interaction models the solvophilic character of the head groups. Our choice of parameters is a simplified version of the parameter set used by Kapila *et al.*⁴³ who chose parameters equivalent to $\epsilon_{TS} = +1$, $\epsilon_{TH} = +1$, $\epsilon_{HW} = -5.77$, and $\epsilon_{HH} = +5.77$, with all others being zero. We choose parameters close to those of Ref. 43 because the purpose of their study was similar to ours, that is, to evaluate the properties of several amphiphile architectures in 2D on a square lattice. The chain lengths studied in Ref. 43 are also similar to ours; they studied $N = 13$ and 19 , while we study $N = 12$ and 16 . However, it must be emphasized that when using a Larson-type model, a wide range of parameter choices yield systems in which micelle formation occurs. Our choice is, therefore, not necessarily unique and other parameter sets may work equally well. At the same time, we have taken care to confirm that our choices are appropriate for our purposes, and in Sec. II C we present several tests of our model to confirm that it exhibits the behavior expected of a micelle-forming system.

In the following, we define the amphiphile concentration as $X = n_s/(L^2 - n_s N)$, that is, the ratio of the number of amphiphile molecules to the number of solvent molecules. Clusters of amphiphiles are defined as contiguous groups of adjacent amphiphiles that are connected by nearest-neighbor contacts involving either chain element (i.e., H or T units). “Free monomers” are isolated amphiphiles that are completely surrounded by solvent. The free monomer concentration is defined as $X_1 = n_1/(L^2 - n_s N)$, where n_1 is the number of free monomers.

In this work, we examine all distinct amphiphile architectures for linear chains having two head units. The architecture of two-headed linear amphiphiles is often specified as a triplet of integers: m - s - k . Here, m is the number of tail units between one end of the chain and the first head unit; s is the number of spacer units occurring between the two head units; and k is the number of tail units between the second head unit and the other end of the chain. For fixed N , only two of these integers are independent, since $N = 2 + m + s + k$. Hence we specify each architecture by m and s . Some example architectures, and their corresponding (m, s) values, are shown in Fig. 1 for $N = 12$. Most architectures can be specified by two combinations of (m, s) , e.g., both $(m = 0, s = 0)$ and $(m = 10, s = 0)$ correspond to the same double-headed amphiphile for $N = 12$. In such cases, we only consider the (m, s) pair having the lower value of m . If N is even, there are $M = \sum_{i=1}^{N/2} (2i - 1)$ distinct architectures of amphiphiles containing N units. For $N = 12$, $M = 36$; for $N = 16$, $M = 64$. Figure 1 illustrates that the set of these amphiphiles contains double-headed amphiphiles ($m = s = 0$), symmetric gemini surfactants ($m = k$, with $m > 0$ and $k > 0$), asymmetric gemini surfactants ($m \neq k$, with $m > 0$ and $k > 0$), as well as boloform amphiphiles ($m = k = 0$).

B. Simulation protocol

We sample the equilibrium configurations of the system in the canonical ensemble using a MC dynamics in which both amphiphile reptation and translation moves are

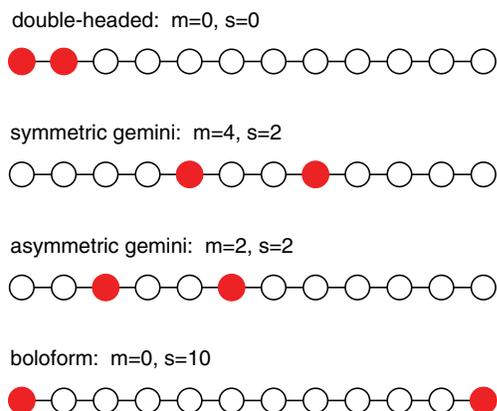


FIG. 1. Four characteristic examples of amphiphile architectures for linear chains of $N = 12$ having two solvophilic head units (filled circles), with all other units being solvophobic tail units (open circles). For each architecture the corresponding m and s values are given.

attempted.^{43,45} In the reptation move, an attempt is made to move one end of the chain onto a solvent-occupied site, with the rest of the amphiphile following in train and the displaced solvent unit moving to the site vacated by the other end of the chain. Reptation serves to both relax the chain shape and diffuse chains through the lattice. The translation move, in which an attempt is made to move a chain to a randomly chosen location, accelerates the diffusion of chains throughout the system volume, especially at low concentration.

We use the following procedure to equilibrate the system at a given X and temperature T . First, a starting configuration is generated by distributing the required number of amphiphiles at random throughout the lattice. The initial configuration of each amphiphile is a straight line-segment oriented along the y axis. The system is then evolved using only reptation attempts until $n_s N^2$ or 30 000 moves are accepted, whichever is greater. Reptation causes a single site on a surfactant molecule to execute a random walk along the path length explored by the molecule. After N^2 accepted reptation moves, a given molecule will have, on an average, moved a distance N along its path length, which is sufficient to achieve a preliminary relaxation of its shape from the initially straight starting configuration. Consequently, after $n_s N^2$ accepted reptation moves, every molecule in the system will have, on average, relaxed in shape.

We then continue the MC trajectory by choosing reptation and translation attempts with equal probability. The run is equilibrated for as many MC steps as are required to accumulate $n_s (10N)^2$ accepted reptation moves. This criterion ensures that regardless of how many translation attempts are made, a sufficient number of reptation attempts have been accepted for each molecule to have (on average) diffused a path length equal to ten times its own length N .

After this period of equilibration is over, an identical production phase is carried out, again selecting translation and reptation attempts with equal probability, and for as many MC steps as are required to accumulate $n_s (10N)^2$ accepted reptation moves. All averages reported here are accumulated over this production phase. In addition, unless noted other-

wise, all runs have been carried out for three independent starting configurations and the results averaged. Also, unless otherwise noted, we have evaluated the system properties for concentrations from $X = 0.0005$ to $X = 0.02$ in steps of $\Delta X = 0.0005$.

C. Properties of the model

We have chosen our model parameters to allow for the observation of micellelike aggregation over a range of chain lengths and architectures. In this section, we illustrate this behavior for two distinct test cases. The first case (denoted in the following as “ H_1T_6 ”) models an amphiphile consisting of a single H unit attached to one end of a chain of six T units; this corresponds to a relatively simple, single-headed surfactant. The second case (denoted “6-2-6”) models a symmetric gemini surfactant with $N = 16$ and architecture $(m-s-k) = (6-2-6)$, that is, two T units separating two H units, and two symmetric tails each of six T units. Note that the 6-2-6 amphiphile is equivalent to two H_1T_6 chains joined at the heads by two T units.

Figure 2 shows the dependence of X_1 on X for both the H_1T_6 and 6-2-6 cases, at several different T . These curves display the shape consistent with a micelle-forming system, in that X_1 increases linearly with X for small X , but saturates to a nearly constant value for larger X .^{33,41} The region of this crossover in the behavior of X_1 approximately coincides with the onset of amphiphile aggregation in the system. The T dependence of these curves, in which the saturation value of X_1 increases with T , is consistent with the behavior observed in numerous other simulations of amphiphile systems (see, e.g., Ref. 38).

Aggregation of the amphiphiles is also reflected in the behavior of $P(n)$, the probability that a randomly chosen amphiphile is part of a cluster of amphiphiles of size n .^{34,38,48} We compute this probability as $P(n) = nC(n)/n_s$, where $C(n)$ is the average number of clusters of size n . As shown in Fig. 3, the emergence as T decreases of a shoulder or a peak in $P(n)$

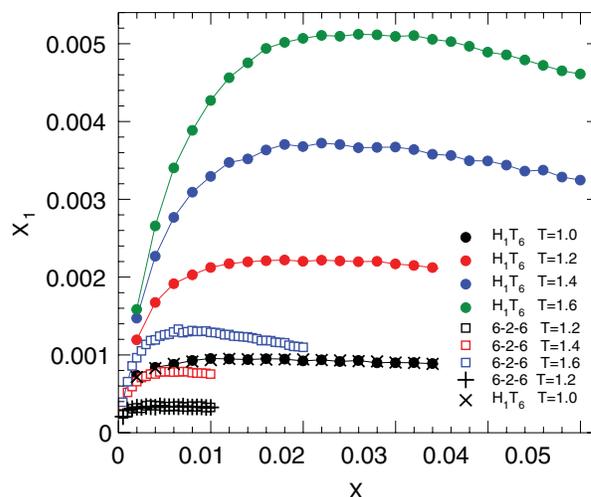


FIG. 2. X_1 versus X for the H_1T_6 and 6-2-6 systems, at several T . All data are for $L = 200$, except for the data represented by the + and \times symbols, which are for $L = 400$.

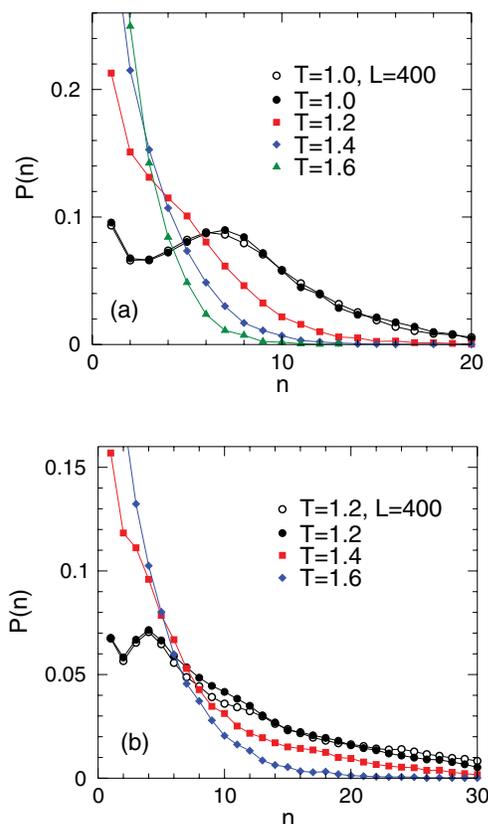


FIG. 3. $P(n)$ for the (a) H_1T_6 and (b) 6-2-6 systems, at several T . All curves are for $L = 200$ except as indicated. In (a) $X = 0.01$ and (b) $X = 0.005$.

at non-zero n is an indication that the aggregation process is generating clusters of a defined size. Snapshots of both the H_1T_6 and 6-2-6 systems are shown in Fig. 4 for the same T at which a peak is observed in $P(n)$ in Fig. 3.

In 3D, gemini surfactants have been observed to form cylindrical and branched micellar aggregates.⁵¹ In 2D, the analog behavior is the formation of elongated or wormlike aggregates.⁴⁰ Figures 3 and 4 illustrate this trend: While the single-tailed H_1T_6 architecture generates many relatively compact and circular aggregates, the double-tailed 6-2-6 architecture by comparison forms many aggregates that are distinctly irregular and elongated (Fig. 3). This trend is reflected in $P(n)$ by the fact that the distribution of aggregate sizes is much broader for the 6-2-6 architecture, and extends to much larger n , than for the H_1T_6 architecture (Fig. 4). Hence, even though the most-probable aggregate size is smaller for the 6-2-6 amphiphiles than for the H_1T_6 case, the number of large aggregates (e.g., those of size $n > 20$) is much greater for the 6-2-6 architecture.

In carrying out simulations of amphiphile systems it is important to distinguish between conditions in which aggregation occurs due to stable micelle formation and aggregation that occurs as a result of the onset of macroscopic phase separation of the amphiphiles from the solvent.⁴² We, therefore, conduct several tests to confirm the validity of the model and our simulation algorithm, as well as to ensure that a regime of micellelike aggregation is observed:

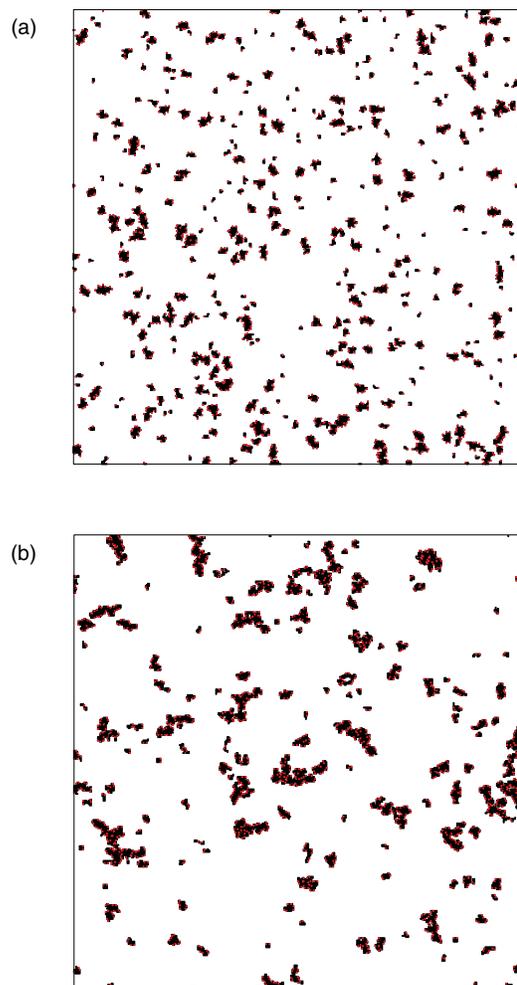


FIG. 4. Snapshots of the (a) H_1T_6 and (b) 6-2-6 systems, both for $L = 400$. In (a) $T = 1.0$ and $X = 0.01$; in (b) $T = 1.2$ and $X = 0.005$. Tail units are rendered as black line segments and head units are red dots.

- (i) We have tested our simulation algorithm and equilibration protocol by reproducing results from a number of previous works. In particular, by appropriate variation of the model parameters, we have reproduced the results for X_1 and $P(n)$ described in Refs. 34, 41, and 46.
- (ii) In all the cases reported here, we confirm that the runtime criteria described in Sec. II B yield stationary time series during the production phase for the system energy and X_1 . We also monitor the number of clusters and the size of the largest cluster as a function of time during the production phase and find no overall drift that would suggest that the system is undergoing macroscopic phase separation.
- (iii) Finally, we test for finite-size effects in two ways. First, we evaluate X_1 as a function of X for systems with both $L = 200$ and $L = 400$ (Fig. 2). The curves for these two system sizes coincide within statistical error, indicating that a system of size $L = 200$ is large enough to be free of significant finite-size effects. Second, we show in Fig. 3, $P(n)$ at the lowest T for both $L = 200$ and 400. Again, the curves coincide supporting the absence of finite-size effects for the $L = 200$ system. This second test is also consistent with the absence of

macroscopic phase separation: For a finite system undergoing phase separation, the n value for the peak in $P(n)$ will increase as L increases since larger clusters are possible in a larger system.⁵² However, for a micelle-forming system, $P(n)$ is independent of system size, consistent with the behavior observed here.

D. Evaluating the CMC

We determine X_{CMC} from a plot of X_1 versus X . There are a number of procedures that have been used to estimate X_{CMC} from such data. The commonly used definition of Tanford^{33,53} assumes that X_1 saturates to a constant value X_1^∞ for $X \gg X_{\text{CMC}}$ and so defines X_{CMC} as the value of X at which the curves $X_1 = X_1^\infty$ and $X_1 = X$ intersect.

However, as is widely observed in both simulations and experiments, our curves for X_1 versus X pass through a maximum value X_1^{max} and then slowly decrease as X grows larger (see Fig. 2). To take this behavior into account, some studies have modified Tanford's definition by fitting a straight line to the large X behavior of X_1 , and defining the CMC as the value of X at which this fitted line and the line $X_1 = X$ intersect.^{36,43} (In the following, we refer to the value of the CMC obtained in this way as X'_{CMC} .) While this definition is reasonable, it requires that a threshold be chosen for the range of X over which the straight-line fit is to be carried out. In a study, such as ours, in which a large number of architectures are examined, our analysis is complicated by the need to choose this threshold for each dataset.

By comparison, the maximum in X_1 versus X is a well-defined feature in all of our datasets and locating its position is straightforward. To simplify the estimation of the CMC, we therefore make the approximation that X_1^{max} can be used as a proxy for X_1^∞ in Tanford's original definition, and so define X_{CMC} as the value of X at which the curves $X_1 = X_1^{\text{max}}$ and $X_1 = X$ intersect. That is, $X_{\text{CMC}} = X_1^{\text{max}}$. We have compared the values so obtained for X_{CMC} with values of X'_{CMC} for a number of important cases central to our overall conclusions. In all cases, we find that the variation of X_{CMC} and X'_{CMC} with both T and the amphiphile architecture are the same, with the only difference being that X'_{CMC} is never more than 10% greater than X_{CMC} . Hence, none of our conclusions concerning the dependence of the CMC on the amphiphile architecture are affected by which of these two definitions we choose.

In this regard, we also note that we have not used the CMC definition of Isrealachvili, in which the CMC is defined as the value of X at which the X_1 curve intersects the line $X_1 = X/2$.⁵⁴ We find that for most of our cases, the slopes of these two curves are very similar in the region where they intersect, resulting in larger statistical errors for the CMC than are found using the above definitions.

To compute X_1^{max} , we smooth the data by averaging X_1 over successive groups of five data points in the interval $X - 0.001 \geq X \leq X + 0.001$, for each value of X in the vicinity of the maximum. We estimate $X_1^{\text{max}} = X_{\text{CMC}}$ as the largest value of the smoothed X_1 data; see Fig. 5.

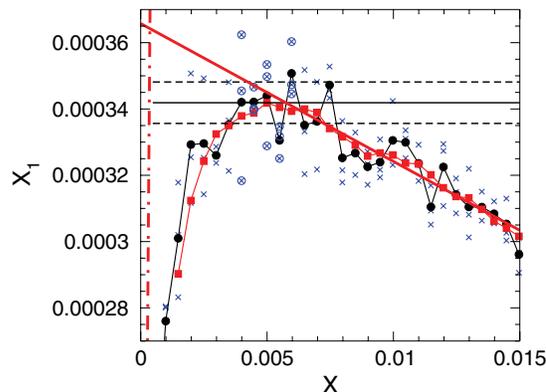


FIG. 5. X_1 as a function of X for the 6-2-6 system for $T = 1.2$. We compute the value of X_{CMC} (thin solid horizontal line) as the maximum value of the smoothed dataset for X_1 as a function of X (red squares), as described in the text. Also shown are the values of X_1 obtained from individual runs (blue crosses) as well as the average of X_1 over the three runs conducted at each X (black circles). The error in X_{CMC} (horizontal dashed lines) is twice the standard deviation of the mean of the 15 data points (circles with crosses) occurring in the vicinity of the maximum value of X_1 . The thick solid line is a linear fit to the smoothed data for X_1 in the range $0.0075 < X < 0.015$. The dotted-dashed line describes $X_1 = X$. The value of X at which the thin solid line intersects the dotted-dashed line defines the value of X_{CMC} ; the intersection of the thick solid line and the dotted-dashed line defines the value of X'_{CMC} .

We estimate the statistical error of X_{CMC} from the scatter in the independent runs used to compute X_1 in the vicinity of the maximum (Fig. 5). Let X_{max} be the value of X at which the maximum occurs in the smoothed data for X_1 . For each X we have three independent runs, and therefore three independent evaluations of X_1 at each X near X_{max} . Further, the variation of X_1 in the vicinity of X_{max} is weak (because it is a maximum), and so we assume that the X_1 measurements for the five values of X in the interval $X_{\text{max}} - 0.001 \geq X_{\text{max}} \leq X_{\text{max}} + 0.001$ are all estimates of X_{CMC} . This provides us with a set of 15 independent estimates of X_{CMC} , for which we compute the standard deviation σ . The error in X_{CMC} is taken as $\pm 2\sigma/\sqrt{15}$, that is, twice the standard deviation of the mean for a sample size of 15. As shown in the following, we find that the computed error for X_{CMC} is smaller than the symbol size used in our plots.

III. RESULTS

We now present our results for the CMC of all distinct amphiphile architectures containing two head units, for fixed $N = 12$ and 16. We carry out simulations of all these architectures using the computational protocol described above. All simulations are conducted for $T = 1.2$. As we will see below, at this T we find that all the architectures give systems in which some degree of micellelike aggregation is occurring.

A. Influence of the number of spacer units

Our results for the CMC of all the distinct architectures are shown in Fig. 6. In Fig. 6, we plot X_{CMC} as a function of s and group points corresponding to constant values of m

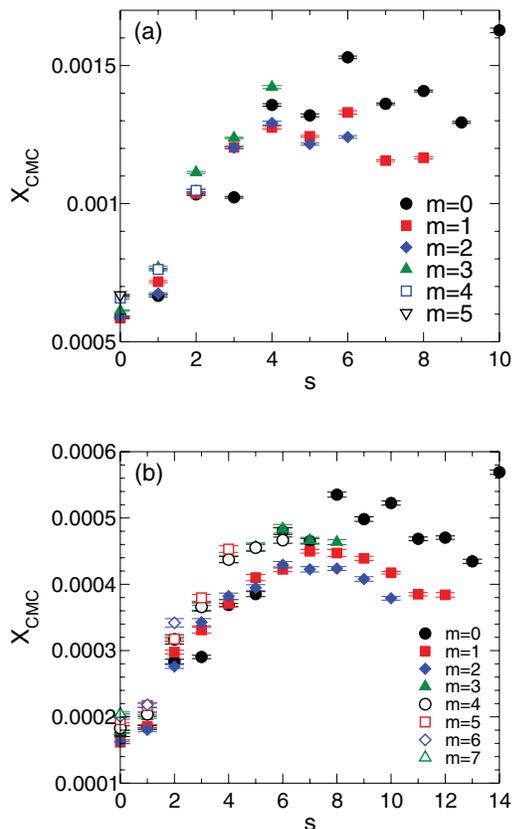


FIG. 6. X_{CMC} for all distinct combinations of m and s for (a) $N = 12$ and (b) $N = 16$. For all points, $T = 1.2$.

by using the same symbol type. Note that in all cases, the statistical error in X_{CMC} is smaller than the symbol size.

For $s < N/2$, the CMC increases with s . The scale of the increase is a factor of ~ 2.5 relative to the lowest CMC at $s = 0$. While the data are more scattered for $s > N/2$, there is a trend for the CMC values to pass through a maximum and begin to decrease as s increases further. This is most evident in the $m = 1$ and $m = 2$ curves for $N = 16$ [Fig. 6(b)]. This behavior, where the CMC initially increases with s and then passes through a maximum, is consistent with several experimental studies,^{15,19,55,56} as well as earlier computer simulations.²¹ Most of the experimental work reporting this phenomenon has been conducted for symmetric gemini surfactants having a fixed tail length where the size of the spacer segment is progressively increased; hence, the total length N of the amphiphile chain is increasing. It is interesting to note that the same behavior occurs in our system, where N is fixed regardless of how s changes and where both symmetric and asymmetric cases are considered.

The single exception to the trend in the CMC as a function of s occurs for the boloform amphiphiles, which have the largest possible value of $s = N - 2$ and a head unit located exactly at each end of the chain. X_{CMC} for the boloform amphiphiles is the largest of all the architectures for both $N = 12$ and 16. The behavior of X_{CMC} for the boloform case is therefore anomalous, and we return to this issue in Sec. III C below.

In Fig. 7(a), we quantify the changes in the morphology of the amphiphile aggregates by plotting $P(n)$ for several val-

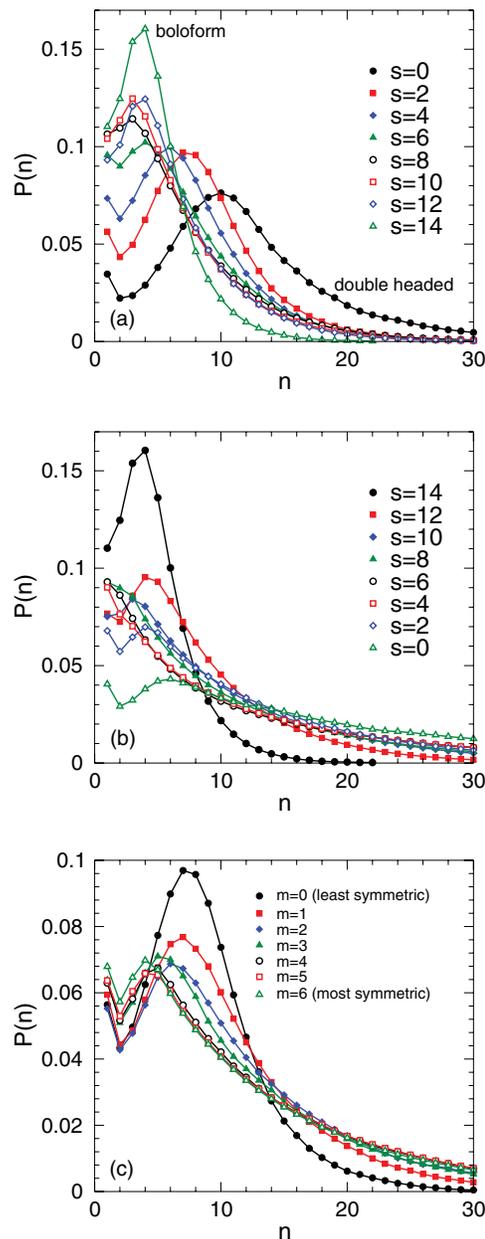


FIG. 7. $P(n)$ for $T = 1.2$, $N = 16$, and $X = 0.005$. These results are averages over 100 equilibrium configurations, each generated from a distinct starting configuration. (a) Influence of s on $P(n)$ for fixed $m = 0$. $s = 0$ corresponds to a double-headed amphiphile, while the largest value of s corresponds to a boloform amphiphile. (b) Influence of s on $P(n)$ for symmetric amphiphiles only. (c) Influence of amphiphile symmetry on $P(n)$ for fixed $s = 2$. The smallest value of m is the least symmetric, while the largest value corresponds to a symmetric gemini surfactant.

ues of s , all at constant $m = 0$. (In Fig. 7, we only show the data for $N = 16$ since the corresponding plots for $N = 12$ are qualitatively the same.) This series corresponds to fixing one head unit on one end of the chain and moving the second head unit along the chain. These $P(n)$ curves, therefore, progress from the double-headed architecture ($s = 0$) through the boloform case ($s = N - 2$). We find that the double-headed architecture produces the largest aggregates, as characterized by the value of n at the maximum in $P(n)$; as well as the widest distribution of aggregate sizes, as characterized by the width of the peak.

As s increases up to approximately $s = N/2$, the size of the aggregates decreases, and the peak of $P(n)$ becomes less distinct, indicating that the micellelike character of the amphiphile aggregates is degrading. This regime coincides with the range of s in which X_{CMC} is increasing with s . For $N/2 < s < N - 2$, the distinctness of the peaks in $P(n)$ shown in Fig. 7(a) recovers somewhat, though the aggregates remain small compared to $s = 0$. This modest recovery of the micellelike morphology is consistent with the trend for X_{CMC} to decrease over this range of s , with the exception of the bolofom case noted above. The $P(n)$ curve for the bolofom amphiphiles ($s = N - 2$) has the sharpest peak of all the curves but the typical aggregate size remains small.

As shown in Fig. 7(b), the same general trends for the influence of s on the CMC and aggregate morphology are observed if we restrict our attention solely to symmetric amphiphiles. These symmetric architectures include all the distinct gemini surfactants having equal tail lengths, as well as the bolofom architecture. For $s < N/2$, increasing the number of spacer units, at the expense of the length of the tails, increases the CMC and degrades the formation of aggregates with a distinct size. In addition, as s decreases, we observe that the magnitude of $P(n)$ increases at large n , reflecting the formation of wormlike and branched structures, as expected for gemini amphiphiles. Indeed, for $4 < s < 8$, we find that $P(n)$ is monotonically decreasing for all n , indicating a continuous distribution of aggregate sizes, characteristic of irregular and wormlike aggregates. However, a peak in $P(n)$ is observed for both small and large s , demonstrating the occurrence of distinct finite aggregates at both ends of the range of s . In addition, here again we see that the bolofom case is a special one, in which the CMC is highest while the aggregates have a very distinct but small size.

Experimentally, it has been observed for symmetric gemini surfactants that the typical size of aggregates decreases as s increases.^{2,55,57} These experiments were conducted using m - s - m architectures in which m was held constant and s increased, and hence the total amphiphile length was increased. Our results show that it is also possible to realize similar changes in aggregate morphology by varying s at constant N , although the effect in the case of symmetric amphiphiles [Fig. 7(b)] is much weaker than for the asymmetric architectures [Fig. 7(a)].

To better understand why the CMC depends on s in the manner shown in Fig. 6, we have analyzed the structure of the aggregates formed in our $N = 16$ simulations at a fixed value of $X = 0.005$. At this concentration, all the architectures studied are near to or just above their respective CMC values. In order to quantify how well-structured the micelles are, we define f as the fraction of head units that are entirely surrounded by tail units. From an energetic standpoint, a head unit prefers to sit at the interface between tail units and the solvent. A head unit entirely surrounded by tail units is an energetically unfavorable local structure, and a higher frequency of occurrence of such structures indicates poorly structured micelles in which head units have been (entropically) drawn into the interior of the aggregates. The unfavorable energy of these structures may reduce the free energy difference that drives an isolated amphiphile to become part of an aggregate. If this

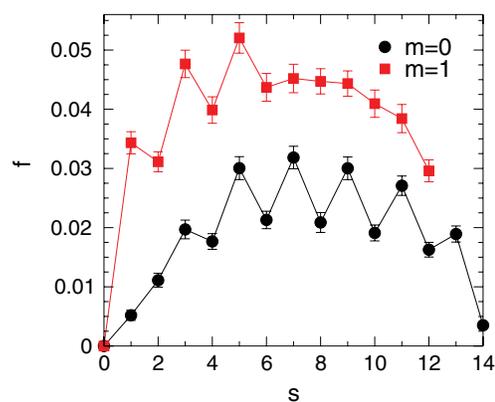


FIG. 8. f as a function of s for $T = 1.2$, $N = 16$, and $X = 0.005$. Curves for both $m = 0$ and $m = 1$ are shown. These results are averages over 100 equilibrium configurations, each generated from a distinct starting configuration. The error bars give the standard deviation of the mean over the 100 configurations.

driving force for aggregation decreases, then the CMC should rise; hence, larger f should correspond to larger X_{CMC} . This expectation is confirmed in Fig. 8, where we show f as a function of s for fixed $m = 0$ and $m = 1$. Similar to the behavior of X_{CMC} in Fig. 6, we find that f also passes through a maximum in the vicinity of $N/2$. (The saw-tooth behavior of the curves in Fig. 8 is almost certainly a lattice effect.) The behavior of f demonstrates that head units on amphiphiles with intermediate spacer lengths are more effectively drawn into the interior of the aggregates. For such an amphiphile, at least one head unit is chemically bonded on both sides to significant lengths of solvophobic tail units. Consequently, this head unit may be more susceptible to being “entrapped” by its own adjacent tail units and thus pulled into the aggregate interior.

B. Influence of amphiphile symmetry

Implicit in the results shown in Fig. 6 is the finding that there is little dependence of X_{CMC} on m , especially for small values of s . Since m determines the placement of the first head unit along the chain, it controls the symmetry of the architecture for a given value of s . However, the degree of symmetry is not immediately apparent from a given (m, s) pair. As noted in Sec. I, Thomas and coworkers^{10,12} investigated m - s - k architectures of dissymmetric gemini surfactants and quantified the asymmetry of the surfactants using the ratio k/m . They studied surfactants having values of k/m from 1 to 3 and found that the CMC decreased by $\sim 35\%$ over this range.

To facilitate a direct comparison with the experimental data of Thomas and co-workers, we also quantify the asymmetry in terms of k/m . Accordingly, in Fig. 9, we plot our CMC data versus k/m for several values of s , for architectures having $m > 0$. As in Refs. 10 and 12, we also find a decrease in the CMC as k/m increases, although the effect is somewhat weaker, despite the fact that our data extends up to $k/m = 13$. At fixed s , the variation of X_{CMC} with k/m never exceeds a factor 0.2. This is small compared to the relative changes found as a function of s , which are a factor of ~ 2.5 .

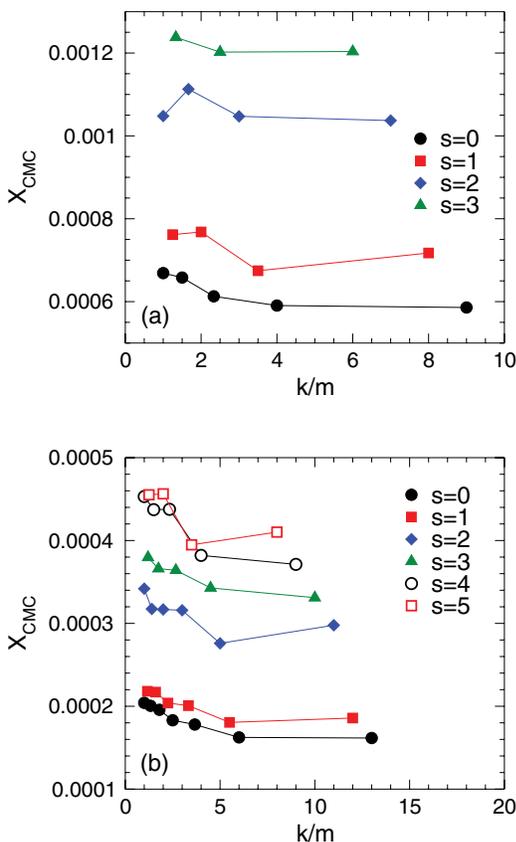


FIG. 9. Influence of the asymmetry of gemini amphiphiles on X_{CMC} at $T = 1.2$, for (a) $N = 12$ and (b) $N = 16$. The asymmetry is quantified by k/m , where k is the number of tail units in the longer of the two tails, while m is the number in the shorter tail. For example, a symmetric gemini surfactant has $k/m = 1$, while the most asymmetric gemini surfactant for $N = 16$ and $s = 0$ has $k/m = 13$. Note that amphiphiles with $m = 0$ are excluded from this plot since in this case k/m diverges.

Our results, thus, quantify the relative importance of spacer size and asymmetry for amphiphiles of fixed length and highlight the dominant influence of spacer size relative to that of the other architectural properties.

The behavior of $P(n)$ as a function of asymmetry at fixed s is consistent with the variation of the CMC noted above. As shown in Fig. 7(c), while there are changes in the height and position of the peak in $P(n)$, these are smaller changes than those observed due to changes in s [Fig. 7(a)]. For the small value of $s = 2$ depicted in Fig. 7(c), the peak is present for all values of m suggesting that the degree of aggregate formation is not qualitatively disrupted as the symmetry of the amphiphile architecture is varied. This is in contrast to the suppression of the peak in $P(n)$ that we observe for intermediate values of s in Figs. 7(a) and 7(b). While the change in the typical size of the aggregates is relatively smaller in Fig. 7(c), the direction is the same as found in experiments: Reference 12 reports that the size of aggregates increases as the degree of asymmetry increases, as is found here. At the same time, examination of the large- n behavior of $P(n)$ in Fig. 7(c) shows that increasing symmetry favors the occurrence of larger wormlike aggregates.

C. Behavior of boliform amphiphiles

As noted above, the behavior of the CMC for the boliform architecture studied here does not follow the trend established at smaller values of s (Fig. 6). Our results for the CMC of the boliform architecture is also inconsistent with the experimental results. Based on a comparison of the experimental value of the CMC for an 8-6-8 gemini surfactant⁵⁸ with that of a boliform amphiphile of approximately equal overall length,⁵⁹ we would conclude from published data that the boliform architecture should have a CMC that is smaller by almost an order of magnitude. This is clearly at odds with our result that the boliform amphiphile has the highest CMC of any architecture studied here.

It may be that the anomalous behavior of our boliform amphiphiles is due to a lattice effect. Of all the architectures we study, only for the boliform case do both head units have the potential to participate in three energetically favorable HS interactions; in all other cases, at least one, and usually both head units can have at most two HS interactions. The increased potential for HS interactions increases the solubility of the boliform amphiphiles in the solvent and so may contribute to the large value of X_{CMC} that we observe.

Despite this discrepancy, the trend in the morphology of the aggregates formed by the boliform amphiphiles in our simulations is consistent with the experiments. Boliform surfactants have long been recognized for producing unusually small micellar aggregates in comparison to other amphiphiles.^{4,5,60} This is also true here. The typical sizes of our boliform aggregates are the smallest of all the architectures we examine (Figs. 7).

IV. CONCLUSIONS

Our model amphiphiles are highly idealized, and real surfactants are obviously more complex. For example, we have ignored the effect of chain stiffness or restrictions in bending angles that exist, especially in the bonds that connect unlike segments of the structure, such as the bonds linking head groups to the hydrocarbon chains. Also, many of the real surfactants that motivate our work are not truly linear chains. For example, in many gemini surfactants the spacer does not connect directly to the head groups, but rather at some other point along the tails. Restricting the dimensions of the lattice to 2D almost certainly introduces serious distortions into the results as compared to 3D, although we note that the case of micelle formation in quasi-2D systems, such as on surfaces and in thin films, may be interesting in its own right.⁶¹

Nonetheless, the qualitative trends in most of our results are in line with the experimental findings. Our main result is to show for a complete family of two-headed linear amphiphiles, all of the same overall length and analyzed in the same way, that both the size of the spacer and the amphiphile asymmetry play a role in determining the CMC and the aggregate morphology; however, the influence of the spacer size is about an order of magnitude more than that due to asymmetry. This result is in line with the findings reported in the experimental literature on symmetric and dissymmetric gemini surfactants. The fact that we observe a relative influence of

spacer length and asymmetry similar to that found in experiments, using such a simple and highly idealized 2D model, strongly suggests the fundamental geometric origin of this relationship.

Finally, our work illustrates that an exhaustive examination of amphiphile architectures (in the spirit of “combinatorial chemistry”) is computationally possible using models of the kind studied here. Our results demonstrate that this approach can help elucidate and quantify structure-property relationships that otherwise must be deduced from an analysis of multiple studies, often conducted using different measurement techniques and chemically distinct species.

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