Reverse coarsening and the control of particle size distribution through surfactant

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Abstract: The minimization of surface area, as a result of the minimization of (positive) surface energy is a well-known driving force behind the spontaneous broadening of (nano)particle size distribution. We show that surfactant molecules binding to particle surfaces effectively decrease the surface energy and may change its sign. In this case, contrary to the expected broadening behaviour, a minimum of free energy is achieved at maximum surface area for all particles, i.e. when the particles are identical. Numerical simulations based on the classical Lifshitz-Slyozov-Wagner theory with surfactant-induced surface energy renormalisation confirm the collapse of the particle size distribution. As the particle size evolution is much slower than particle nucleation and growth, the manipulation of surface energy with in-situ replacement of surfactant molecules provides a method for controlling particle size distribution with great potential for creating mono-disperse nanoparticles, a key goal of nanotechnology.

Keywords: Free energy; surface energy; surfactants; molecular ligands; mono-disperse nanoparticles; particle coarsening

1. Introduction

The generation of identical nanoparticles [1-7] and fine nanostructures [8-10] is a key requirement for self-assembly processes in bottom-up nanotechnology. However, structure coarsening, or Ostwald ripening [11], is a universal phenomenon that stands in the way of forming identical nano-building blocks through quasi-equilibrium methods. Coarsening represents the spontaneous evolution of precipitated particles, during which smaller particles transfer mass to bigger ones thus broadening the particle size distribution (PSD) [12-15]. This is a common phenomenon taking place at the final stages of many first-order phase transformations following nucleation and growth, for example in two-phase mixtures [11], binary alloys [16], clusters on surfaces [17], oil-water emulsions [18], and during epitaxial growth [19-20] and synthesis of nanoparticles [21]. The main driving force behind coarsening is the minimization of interfacial area between the two phases and as such it depends intimately both on the interface energetics and the system temperature. Much effort has been devoted to find ways to inhibit or even reverse coarsening in order to achieve a narrower PSD. Known examples of PSD narrowing discussed in the literature include the so-called digestive ripening [1-3, 22, 23]; inverse ripening [24-27]; and size focusing during production-controlled growth [28, 29]. Some of these processes take place in the presence of surfactants, which are widely used in nanoparticle production mainly to stabilize particles [30-32] and to tune the growth regimes towards production of particles with required shapes and sizes [33-35]. There are also mechanisms of inhibition, or of arresting Ostwald ripening [36-38], which do
not permit real narrowing of the PSD but rather preserve it from broadening. Digestive ripening can be explained by taking into account electric charges on the particle and by including the electrostatic energy in the free energy analysis [22-23], while inverse ripening has been observed for gold inclusions in an amorphous SiO₂ matrix irradiated with MeV gold ions [25, 27]. The PSD narrowing in the latter case is of transient character. It is affected by mechanical stresses and requires fine-tuning of radiation-annealing cycles. On the other hand, size focusing during production-controlled growth is predicted for effectively open systems where the material deposited on particles is continuously reproduced in the bulk of solution [28, 29]. In summary, there is currently no effective theory describing the phenomenon of PSD narrowing in conservative and charge-neutral systems.

Besides the quasi-equilibrium fabrication methods when the system is thermodynamically driven, narrow PSDs are also successfully generated using discrete size selection by undercooling [39-41] and dewetting [42-43] in non-equilibrium regime. Here, we study the role of surfactants, or surface binding molecular ligands [44-50] in controlling PSDs. We present a thermodynamic mechanism for particle size stabilization and narrowing of a PSD by surfactant molecules binding to particle surfaces. This process is illustrated by considering a simple dilute solution of two species (referred to here as molecules) in a neutral solvent in the presence of precipitated aggregates (called particles) of one of the components. Based on a free energy analysis, we demonstrate that there exists a range of parameters at which reverse coarsening (RC) occurs, and consequently, smaller precipitated particles grow at the expense of bigger ones. This regime is spontaneous, i.e. it is thermodynamically driven and converts any initial distribution of particles into an array of almost identical particles [51]. Notably, the transition into this regime can be obtained by varying the surfactant concentration and/or the system’s temperature.

3. Results and discussion

Thermodynamic theory of reverse coarsening

Following the classical approach [13-15], we use a mean-field approximation in our analysis. This approximation assumes that the particles are sufficiently far from each other so that they only interact through the averaged characteristics of the solution, and that their occupied volume is negligible compared to the entire volume of the system. First, we analyze the thermodynamic stability of a single particle in the solution with a given concentration of identical molecules (we call them A-molecules) in the presence of surfactant molecules (B-molecules). Then, by writing the molecular chemical potential as a function of particle size we obtain the conditions for RC by considering the mass exchange between particles in an ensemble. Second, we confirm our prediction of RC by direct numerical simulations of the particle ensemble evolution using Kinetic Monte Carlo technique.

3.1. Isolated particle in a binary solution

In our thermodynamic analysis we consider a conservative system consisting of a dilute binary A-B solution and a precipitated spherical A-particle of radius R measured in the units of typical intermolecular distance and containing \( N_p = 4\pi R^3 / 3 \) molecules. The system contains \( M_A \) molecules of material A, \( M_B \) molecules of material B and \( M_L \) molecules of a solvent. For simplicity we assume that a molecule A can bind only one molecule B and vice versa with characteristic energy \( \varepsilon_{AB} \). This binding results in the formation of \( N_C \) number of A-B complexes (dimers) in the solution, and in concentration \( n_S \) of B-molecules on particle surfaces. The free energy of the system is

\[
G = -\varepsilon N_p + \gamma S_p (1 - n_S) - \varepsilon_{AB} n_S N_p + TS_p \cdot \left[ n_a \ln(n_a) + (1 - n_a) \ln(1 - n_a) \right] \\
- \varepsilon_{AB} N_c + T \cdot (N_c \ln(N_c) + N_a \ln(N_a) + N_B \ln(N_B) + N_L \ln(N_L) - M_A \ln(M_A) - M_B \ln(M_B)),
\]

where \( \varepsilon > 0 \) and \( \gamma > 0 \) are the cohesive and surface energies per A-molecule in the particle, respectively, \( T \) is the temperature in energy units (eV), \( S_p = 4\pi R^2 \) is the number of molecules on
the particle surface, \( N_4 = M_4 - N_p - N_c \), \( N_b = M_b - S_p n_s - N_c \), \( R \neq 1 \) and
\( M_z = N_4 + N_b + N_c + M_L \). In Eq. (1) we neglected possible interactions between B-molecules (see Appendix A for details) and assumed a simple form for the entropy, which provides a lower limit estimate [52].

The minimization of the free energy \( G \) with respect to the independent variables \( N_p, N_c \) and \( n_s \) in the limit of a dilute solution with \( M_L > N_4 + N_b + N_c \) yields (see Appendix B)

\[
T \ln (n_s) = -\varepsilon + \frac{2}{R} \cdot \Gamma,
\]

\[
n_c = E_0 \cdot n_s \cdot n_b,
\]

\[
n_s = E_1 \cdot n_b \cdot \left(1 + E_1 \cdot n_b\right)^{-1}
\]

where \( E_0 = \exp \left(\varepsilon_{ab} / T\right) \), \( E_1 = \exp \left(\left(\gamma + \varepsilon_{ab}\right) / T\right) \), and \( \Gamma = \gamma - T \cdot \ln \left(1 + n_b E_1\right) \) is the renormalized surface energy of A-particle, \( n_{a(b)} = N_{a(b)} / M_z \) and \( n_c = N_c / M_z \) are the number concentrations of corresponding species in the solution. The first equation in Eq. (2) is a statement of the equality of chemical potentials for A-molecules in the solution \( \mu_a (sol) = T \ln (n_a) \) and in the particle \( \mu_a (R) = -\varepsilon + 2\Gamma / R \) of radius \( R \). The second and third equations of Eq. (2) give the equilibrium concentrations of A-B complexes in the solution and of B-molecules on the particle surfaces, respectively. Note that the surfactant B-molecules always reduce the surface energy, as \( \Gamma < \gamma \) due to \( \ln (1 + n_b E) > 0 \). For high enough values of \( \varepsilon_{ab} \), the effective surface energy \( \Gamma \) can become negative (see Fig. 1) resulting in a narrowing of the size distribution for an ensemble of particles in solution. The same figure also shows the surface concentration of surfactant \( n_s \) as a function of \( \varepsilon_{ab} \) illustrating an increase in surface concentration \( n_s \) and a decrease in \( \Gamma \) with increasing bulk concentration \( n_b \) of the surfactant molecules.

The effect of surface energy renormalization by A-B bonding can be obtained by considering the molecular chemical potential \( \mu_a (R) = -\varepsilon + 2\Gamma / R \) for a particle of radius \( R \) in solution containing other A-particles. The case of positive \( \Gamma \) corresponds to ordinary coarsening when A-molecules are transferred from smaller particles to bigger ones until a single big particle remains. If \( \Gamma = 0 \) the molecular chemical potential according to the first line in Eq. (2) becomes independent of \( R \) hence the PSD should also be time-independent. More interestingly, if \( \Gamma < 0 \) the molecular
Figure 1. Effective surface energy $\Gamma$ (red) and concentration $n_g$ of surfactant molecules (B) on particle surface (blue) as functions of A-B bonding $\varepsilon_{AB}$ for two values of the bulk concentration of surfactant $n_g = 0.002$ (solid) and $n_g = 0.01$ (dotted) plotted according to Eqs (2) for $\gamma = 0.1$ eV and $T = 0.032$ eV.

chemical potential is lower for smaller particles and energy minimization requires the A-molecules to be transferred from bigger to smaller particles. This process of reverse coarsening (“size focusing”) results in a mono-disperse (for thermodynamic limit see [51]) particle ensemble with the equilibrium particle radius $R_{eq}$ determined by the initial PSD function $f(R)$ and the particles’ total volume as

$$R_{eq} = \left[ \int_0^\infty R^3 f(R) dR / \int_0^\infty f(R) dR \right]^{1/3}$$

For instance, if the initial PSD is a Gaussian $f(R) = \frac{1}{\sqrt{2\pi \sigma}} \exp\left(-\frac{1}{2\sigma}(R - \langle R \rangle)^2\right)$ with a mean radius $\langle R \rangle$ significantly higher than the variance $\sigma$, then the value of $R_{eq}$ is very close (within a few percents) to $\langle R \rangle$ and does not depend on $\varepsilon_{AB}$.

3.2. Simulation of reverse coarsening in the particle ensemble

The value of $R_{eq}$ given by Eq. (3) is a good estimate if the PSD before collapse is close to the initial PSD. This estimate implicitly assumes that the surface energy $\gamma$ does not depend on the particle radii or shapes. In reality the shape of small particles may deviate from spherical, which in context of our theory means that $\gamma$ would increase with $R$. In turn, this dependence on the radius implies that $\Gamma > 0$ for small enough particles (depending on $\varepsilon_{AB}$ and $T$) which will dissolve and transfer their mass to bigger particles with $\Gamma < 0$. Such mass transfer modifies the initial PSD and shifts the mean radius, hence $R_{eq}$, towards higher values.

To confirm this effect, we conduct numerical simulations using Kinetic Monte Carlo technique [53]. We start with a distribution of spherical particles nucleated and grown so that their evolution is only possible by exchange of molecules between particles. This initial configuration is chosen so that the vast majority of A-molecules are bound in particles. The particle surfaces are covered with B-molecules, which significantly slows down an exchange of A-molecules between particles and solution. Therefore, it is reasonable to assume that the molecular diffusion in solution is the fastest process and the coarsening process is limited by the attachment-detachment events with effective reaction rate constant $K$. Then, the evolution of the $i$-th particle’s radius $R_i$ (measured in the units of inter-atomic distance) within this Wagner approximation [15] can be described by

$$\frac{dR_i}{dt} = K \cdot (n - n_{GT}(R_i)),$$

where $n_{GT}(R_i)$ is the Gibbs-Thomson concentration (equilibrium concentration) for the $i$-th particle and $n$ is the mean field concentration of free A-molecules in solution. The latter is determined using mass conservation for all particles (valid because of our earlier assumption about the dominant fraction of A-molecules contained in particles)

$$\frac{d}{dt} \sum_{i=1}^N 4\pi R_i^3 = \sum_{i=1}^N 4\pi R_i^3 \frac{dR_i}{dt} = \sum_{i=1}^N 4\pi K R_i^2 \cdot (n - n_{GT}(R_i)) = 0 \rightarrow n = \frac{\sum_{i=1}^N R_i^3 \cdot n_{GT}(R_i)}{\sum_{i=1}^N R_i^3} \quad (5)$$
The Gibbs-Thomson concentration is given by the equality of molecular chemical potentials in the particle and solution
\[ T \ln \left( n_{GT} \right) = -\varepsilon + 2\Gamma / R \] (see Eq. (2)) taking a more realistic expression for surface energy \( \gamma \). This expression is obtained considering the generic expression for particle energy \( \varepsilon(N) \) as a function of the number \( N \) of atoms/molecules it contains [54-57]

\[ -\varepsilon(N) = -\varepsilon + \varepsilon_s \cdot N^{-1/3} + \varepsilon_c \cdot N^{-2/3} + \varepsilon_e \cdot N^{-1} \] (6)

where \( \varepsilon > 0 \) is the bulk contribution to energy (cohesive energy), \( \varepsilon_s > 0 \) is the contribution of facets’ surfaces, \( \varepsilon_c > 0 \) is the contribution of edges, and \( \varepsilon_e \) defines the energy origin (reference point). Adopting this expression for the generic case of spherical particles with radius \( R \) and neglecting the small energy origin \( \varepsilon_e \) we use \( N \sim R^3 \) in Eq. (6) so that the energy per molecule in a particle (or molecular chemical potential) takes the form \( -\varepsilon(R) = -\varepsilon + 2\gamma(R) / R \) with

\[ \gamma(R) = \gamma_0 + \frac{\gamma_1}{R} \] Both parameters \( \gamma_0 \) and \( \gamma_1 \) are positive, as the former corresponds to true surface energy for \( R \gg 1 \), while the latter takes into account non-sphericity effects according to Eq. (6).

With such a generic form, this equation can effectively account also for the contributions to the surface energy coming from possible roughness of particle surfaces. Therefore, the surface energy \( \gamma(R) \) decreases with increasing particle radius making larger particles relatively more stable. With this surface energy, the Gibbs-Thomson concentration is now

\[ n_{GT}(R) = \exp \left[ -\frac{\varepsilon}{T} + \frac{2\gamma_0}{T \cdot R} + \frac{2\gamma_1}{T \cdot R^2} - \frac{2}{R} \ln \left( 1 + n_n E_i \right) \right] \] (7)

Using Eqs (4)-(7) we introduced molecular absorption and emission probabilities by the particle (Eqs (16)-(17) in Ref [51]) and implemented them in the original Fortran code to simulate the evolution of 10^4 particle ensemble. For illustration purposes we considered two values of \( \gamma_1 \): 0 and 0.01 eV. The results shown in Fig. 2 correspond to some realistic values of known model parameters, chosen for illustrative purpose. We see that both the mean \( \langle R \rangle \) and maximum \( R_m \) radii continuously grow with time for \( \varepsilon_{ab} = 0.18 \), which is characteristic for ordinary coarsening. In contrast, for \( \varepsilon_{ab} > 0.197 \) eV the evolution displays reverse coarsening, as the radii \( \langle R \rangle \) and \( R_m \) both converge to the same value \( R_{eq} \) predicted by Eq. (3), which for \( \gamma_1 = 0 \) gives the value of \( R_{eq} \) very close to \( \langle R \rangle \) of the
Figure 2. a) Time evolution of the mean $\langle R \rangle$ and maximum $R_{\text{max}}$ radii calculated for ensemble of 10⁴ particles with Gaussian initial distribution ($\langle R \rangle = 1.5$ and standard deviation $\sigma = 0.25$) obtained via numerical solution of Eqs (4) - (7) for $T = 0.032$ eV (100°C), $\gamma_i = 0.1$ eV, $\varepsilon = 0.4$ eV and $n_g \approx 0.002$. The values of $\varepsilon_{ab}$ and $\gamma_i$, respectively are (eV units): 1) 0.24 and 0; 2) 0.21 and 0; 3) 0.21 and 0.01; 4) 0.196 and 0; 5) 0.18 and 0. b) PSD corresponding to the curves 3) at the time moments indicated by arrows in a).

Initial PSD. For $\gamma_i = 0.01$ eV, the value of $R_{eq}$ is larger than the average radius $\langle R \rangle$ of the initial PSD. As discussed, this effective increase in $\langle R \rangle$ in the early stages is due to the mass transfer from the smaller particles with $\Gamma > 0$ to the bigger particles with $\Gamma < 0$. Clearly, there is a threshold value for $\varepsilon_{ab}$ above which the system displays reverse coarsening. This threshold is determined by the requirement $\Gamma = 0$ and depends on the surfactant concentration $n_g$, the molecular-surfactant interaction energy $\varepsilon_{ab}$, the temperature $T$, and the size dependence of the surface energy determined by $\gamma_i$. In the case of $\gamma_i \approx 0$, the threshold value $\varepsilon_{ab}^{*}$ is estimated for the values in Fig. 1 as

$$\Gamma = \gamma_0 - T \cdot \ln\left(1 + n_g E_i\right) \approx 0 \rightarrow \varepsilon_{ab}^{*} \approx 0.197.$$  

Choosing a surfactant with $\varepsilon_{ab} \approx \varepsilon_{ab}^{*}$ preserves the initial PSD (see curves 4 in Fig. 2). The evolution of the PSD is relatively slow (mind the log scale for $x$-axis in Fig. 2), therefore replacing the surfactant molecules with those having $\varepsilon_{ab} \approx \varepsilon_{ab}^{*}$ or $\varepsilon_{ab} > \varepsilon_{ab}^{*}$ allows to stop or reverse it, respectively, providing a great opportunity for thermodynamic control of the PSD for generated nanoparticles. It is worth noting that the values of $\varepsilon_{ab}$ generating reverse coarsening must also have an upper limit, as at values $\varepsilon_{ab} > \varepsilon_{ab}$, the A-molecules would rather form molecular-surfactant complexes in the solution than stay bound together as a particle. Therefore, in our case the inequality $\varepsilon_{ab} < \varepsilon$ must hold. The value of $\varepsilon_{ab}^{*}$ in our simulations is chosen for illustration purpose without
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reference to any particular material’s surface or binding molecule. Determining $\varepsilon_{\text{AB}}^\ast$ in practice is
difficult, as it requires measuring the particle’s surface energy $\gamma$ in solution. In our case it is
reasonable to demand that $\varepsilon_{\text{AB}}^\ast$ is low enough for the surfactant molecules not to affect the particle
shape during its growth or dissolution [58]. Note that the fine tuning of the coarsening process,
which is determined by the value of the effective surface energy $\Gamma$, can be easier achieved by
varying the bulk concentration of surfactant molecules $n_b$ and the system temperature $T$ rather
than $\varepsilon_{\text{AB}}^\ast$ using different surfactant molecules.

From our simulations, we conclude that the main signature of reverse coarsening is the
dissolution of the largest particles in the ensemble. While both the mean and the maximum particle
radii increase in ordinary coarsening, the maximum radius in reverse coarsening will, at some point
in time, decrease indicating partial dissolution of the biggest particles. This feature is crucial for
experimental identification of the reverse coarsening process.

5. Conclusions

We have presented a free-energy analysis and simulation studies of a dilute binary molecular
solution with precipitated particles of one component and the other component acting as a surface
passivating surfactant. Our analysis predicts that for certain component concentrations and strength
of molecular-surfactant interaction, the system of precipitated particles evolves towards equilibrium
in such a way that the PSD spontaneously narrows, leading eventually to a mono-sized array of
particles. The narrowing of the size distribution is mainly due to a strong enough surface passivation
of particle surfaces by surfactant molecules, which renders the effective surface energy of particles
negative. The effect is confirmed by direct simulations of the PSD evolution in a large ensemble of
poly-dispersed particles. This reverse coarsening effect is a new paradigm for the
thermodynamically controlled solution-processed generation of mono-sized nano-particulate arrays
of various materials.

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Appendix A

The role of interaction between B-molecules

The above analysis is based on an assumption that the interaction between B-molecules can be
neglected. This interaction if high enough may significantly affect the coverage of particle surfaces
by B-molecules. In particular, a significant B-B interaction $\varepsilon_{\text{BB}}$ could promote formation of large
B-clusters on the particle surface and even cause complete coverage if the B-cluster grows over its
critical size. To illustrate the effect consider the simplified situation of the critical 2D cluster
formation of B-molecules on a flat surface of A-crystal. The condition for the critical cluster
formation is given by the equivalence of chemical potentials of B-molecules in the surface gas phase
of concentration $n_B$ and in the critical cluster of radius $r_C$, namely $T \ln \left( n_B \right) = -\varepsilon_{\text{BB}} + \gamma_B / r_C$,
where $\gamma_B$ is the energy per B-molecule on the cluster perimeter. For the critical radius we then
obtain $r_C = \gamma_B / \left( \varepsilon_{\text{BB}} + T \ln \left( n_B \right) \right)$. Using this one can easily obtain the critical cluster nucleation
barrier $\Delta_C$.
Eq. (A1) shows that for the $\varepsilon_{ab}$ values such low that $\varepsilon_{ab} + T \ln (n_s) \leq 0$ no critical cluster can be formed therefore the interaction between B-molecules can be disregarded.

Appendix B

Derivation of main equations

To provide details on the derivation of Eqs (2), consider the free energy of the system

$$\begin{equation}
G = -\varepsilon N_p + \gamma S_p \left(1 - n_s\right) - \varepsilon_{ab} n_s S_p + TS_p \cdot \left[n_s \ln (n_s) + (1 - n_s) \ln (1 - n_s)\right]
- \varepsilon_{ab} N_C + T \cdot \left(N_C \ln (N_C) + N_A \ln (N_A) + N_B \ln (N_B) + N_L \ln (N_L) - M_L \ln (M_L)\right)
\end{equation}$$

where $\varepsilon > 0$ and $\gamma > 0$ are the cohesive and surface energy per A-molecule, respectively, $T$ is the temperature in energy units (eV), $N_p = 4\pi R^3 / 3$ and $S_p = 4\pi R^2$ (particle radius $R$ is measured in the units of typical intermolecular distance) is the total number of molecules in the A-particle and the number of molecules on particle surface, respectively, $N_C$, $N_B$ and $N_L$ are the numbers of A-B complexes, B-molecules and solvent molecules, respectively, in the solution, $n_s$ is the concentration of B-molecules on the surfaces of A-particle, which are assumed to be in a gas-type phase. Introduce the total numbers of molecules $M_i$ such that

$$\begin{align*}
N_A &= M_A - N_p - N_C, \\
N_B &= M_B - S_p n_s - N_C, \\
N_C &= M_C + N_A - N_p - N_C + M_B - S_p n_s + M_L, \\
M_L &= N_L
\end{align*}$$

Taking into account Eqs (A3) we can choose three independent variables: $N_p$, $N_C$ and $n_s$. The free energy given by Eq. (A2) is expressed in these variables as

$$\begin{equation}
G = -\varepsilon N_p + \gamma S_p \left(1 - n_s\right) - \varepsilon_{ab} n_s S_p + TS_p \cdot \left[n_s \ln (n_s) + (1 - n_s) \ln (1 - n_s)\right]
- \varepsilon_{ab} N_C + TN_C \ln (N_C) + T \cdot \left(N_A - N_p - N_C\right) \ln \left(N_A - N_p - N_C\right)
+ T \cdot \left(M_B - S_p n_s - N_C\right) \ln \left(M_B - S_p n_s - N_C\right)
+ T \cdot \left(M_A - N_p - N_C + M_B - S_p n_s + M_L\right) \ln \left(M_A - N_p - N_C + M_B - S_p n_s + M_L\right)
\end{equation}$$

Minimization of $G$ means that $dG = \frac{\partial G}{\partial N_p} dN_p + \frac{\partial G}{\partial n_s} d\ln (n_s) + \frac{\partial G}{\partial N_C} dN_C = 0 \rightarrow \frac{\partial G}{\partial N_p} = \frac{\partial G}{\partial n_s} = \frac{\partial G}{\partial N_C} = 0$.

Applying this minimization to Eq. (A4) gives

$$\begin{align*}
\frac{\partial G}{\partial N_p} &= -\varepsilon + \frac{2}{R} \gamma \left(1 - n_s\right) - \varepsilon_{ab} n_s T + \frac{2}{R} \left[\frac{M_A - N_p - N_C + M_B - S_p n_s + M_L}{M_A - N_p - N_C}\right] \ln \left[M_B - S_p n_s - N_C\right] - \frac{2}{R} \left[\frac{M_A - N_p - N_C + M_B - S_p n_s + M_L}{M_A - N_p - N_C}\right] \ln \left[M_B - S_p n_s - N_C\right] = 0 \\
\frac{\partial G}{\partial n_s} &= -\gamma + \varepsilon_{ab} S_p + TS_p \ln \left(\frac{n_s}{1 - n_s}\right) - S_p T \ln \left[\frac{M_B - S_p n_s - N_C}{M_A - N_p - N_C + M_B - S_p n_s + M_L}\right] = 0 \\
\frac{\partial G}{\partial N_C} &= -\varepsilon_{ab} + T \ln \left[\frac{N_C \cdot \left(M_A - N_p - N_C + M_B - S_p n_s + M_L\right)}{\left(M_A - N_p - N_C\right) \cdot \left(M_B - S_p n_s - N_C\right)}\right] = 0
\end{align*}$$

Suppose that $M_L >> N_A + N_B + N_C$, $M_A >> N_p + N_C$. Then, after simplification, we have
\[ -\varepsilon + \frac{2}{R} \left( \gamma (1 - n_s) - \varepsilon_{AB} n_A + T \left[ n_s \ln \left( \frac{n_s}{n_B} \right) + (1 - n_s) \ln \left( 1 - n_s \right) \right] \right) - T \ln (n_A) = 0 \]

\[ - (\gamma + \varepsilon_{AB}) + T \ln \left( \frac{n_A}{n_B} \right) = 0 \]

The equations represent the equality of chemical potentials of A-molecules in solution and in the particle (first line), of B-molecules in solution and on the particle surfaces (second line), and of the dimers and unbound A- and B-molecules in the solution (third line). The solution to the second and third equations is given by

\[ n_c = n_A n_B E_0, \quad E_0 = \exp \left( \frac{\varepsilon_{AB}}{T} \right), \quad n_s = \frac{n_A E_1}{1 + n_B E_1}, \quad E_1 = \exp \left( \frac{\gamma + \varepsilon_{AB}}{T} \right) \]  

(A7)

Substituting \( n_s \) into the first of Eqs (A6), we obtain

\[ T \ln (n_s) = -\varepsilon + \frac{2}{R} \Gamma, \quad \Gamma = \gamma - T \ln \left( 1 + n_B E_1 \right) \]  

(A8)

Combining Eqs (A7) and (A8) gives Eqs (2).

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