Modelling the evaporation of thin films of colloidal suspensions using dynamical density functional theory

M J Robbins, A J Archer and U Thiele

Department of Mathematical Sciences, Loughborough University, Leicestershire LE11 3TU, UK

Received 28 March 2011, in final form 15 August 2011
Published 23 September 2011
Online at stacks.iop.org/JPhysCM/23/415102

Abstract

Recent experiments have shown that various structures may be formed during the evaporative dewetting of thin films of colloidal suspensions. Nanoparticle deposits of strongly branched ‘flower-like’, labyrinthine and network structures are observed. They are caused by the different transport processes and the rich phase behaviour of the system. We develop a model for the system, based on a dynamical density functional theory, which reproduces these structures. The model is employed to determine the influences of the solvent evaporation and of the diffusion of the colloidal particles and of the liquid over the surface. Finally, we investigate the conditions needed for ‘liquid–particle’ phase separation to occur and discuss its effect on the self-organized nanostructures.

(Some figures in this article are in colour only in the electronic version)

1. Introduction

Surface patterns resulting from structure formation occur naturally in many different systems and are extensively studied in various scientific fields. A classic example are branched patterns, e.g. found in river networks [1] or formed by bacterial colonies [2], that sometimes form fractals. Other examples are the labyrinth patterns such as those formed via calcification and mineralization processes [3]. The spontaneous self-assembly and self-organization of atoms, molecules and nanoparticles at interfaces is a widely researched topic not only because the resulting structures are interesting but also because they may be used in the manufacturing of nanostructures, e.g. assembling colloids to form photonic bandgap crystals [4]. A large variety of intricate structures can be formed even during the dewetting process of films of non-volatile fluids on solid substrates. The dewetting process starts with the rupture of the initially homogeneous film caused either by a surface instability (spinodal dewetting [5]) or by the nucleation of holes which often occurs at surface defects [6–12]. The holes then grow [13] and the rims meet to form a global network, drop or labyrinth pattern [14–16].

The evaporative dewetting of polymer/macromolecule solutions [17–19] and colloidal (nanoparticle) suspensions [20–24] can produce a wide variety of patterns and has been intensely studied in various experimental settings over recent years. Although the motivation for the work presented here is mainly drawn from the latter, we believe that our results also explain the basic features in the case of evaporative dewetting of solutions. The particular experiments that directly motivate our theoretical work presented here are those described in [22–26] that use a suspension of thiol-passivated gold nanoparticles dispersed in an organic solvent. A drop of the suspension is spin-coated onto a flat silicon substrate to form a thin film over the surface. The solvent evaporates during the spin-coating and leaves a nanoparticle pattern on the surface. In another experimental setup a drop of suspension is placed on the surface within a teflon ring [27]. The evaporative dewetting is slower than in the case of spin-coating and the structuring is observed using video-microscopy [23]. What these experiments show is that branched structures are formed by transverse instabilities of the receding mesoscopic contact line. However, more intriguing are the patterns that are formed in an ultrathin layer that is left behind this contact line. Three different types of
structures have been observed: a labyrinth pattern formed during spinodal dewetting, a two-scaled network structure formed via the nucleation and growth of holes and a branched structure formed by a fingering instability that occurs at the dewetting front of nucleated holes. The diffusive mobility of the nanoparticles and the interaction energies between the particles can be altered by changing the thiol chain length. The fingering instability is found for relatively low nanoparticle mobilities. Because these structures form in the ultrathin layer, the height of the film is of the same order of magnitude as the diameter of the colloids.

To model these processes, simple two-dimensional (2d) lattice kinetic Monte Carlo (KMC) models were proposed [28, 26, 23, 29]. Vancea et al [29] made a detailed investigation of the characteristics of the branched structures and their dependence on the model parameters. They observed that as in the experiments the fingering instability becomes stronger the smaller the nanoparticle mobility gets. A pseudo three-dimensional KMC model has also been considered [26, 30] which can reproduce dual-scale network patterns.

An alternative approach that may be used for modelling such systems is based on thin film hydrodynamical models, which are derived by making a long-wave approximation [31]. Recently, for example, line-pattern formation has been observed in a simple long-wave model for a thin film of colloidal suspension evaporatively dewetting from a surface [32]. Such thin film equation based models provide a good description of the system on mesoscopic length scales and reproduce the experimental results [33–36]. However, they are unable to describe the dynamics of the system at the microscopic (single particle) level. In [37] a more detailed account of the different approaches that may be used for modelling the evaporative dewetting of colloidal suspensions is given, so we do not re-review the subject here.

In the present work, we give a detailed account of an alternative model for this system that has been briefly discussed before [37, 38]. It is based on dynamical density functional theory (DDFT) [39–42] and goes beyond the 2d KMC model by also allowing us to investigate the influence of liquid diffusion over the surface. This paper is laid out as follows: first, in section 2, we present the coarse-grained model for the Hamiltonian and the resulting approximation for the free energy used in our model. This is followed in section 3 by an introduction to the dynamical equations of the DDFT. In section 4 we discuss the phase diagram and perform a linear stability analysis of homogeneous steady films, whereas in section 5 we present fully nonlinear simulation results. In section 6 we summarize our findings and draw some conclusions.

2. Free energy for the system

We consider a simplified coarse-grained two-dimensional model for the system. The surface of the substrate is divided into a square array of discrete lattice sites. We choose the cell size so that each cell may be occupied by at most one nanoparticle, i.e. the lattice spacing $\sigma$ is roughly equal to the diameter of the nanoparticles. Note that previous studies of related models in the context of granular segregation have shown that having larger nanoparticles in the lattice model, i.e. having nanoparticles which have a diameter larger than $\sigma$, inevitably changes the thermodynamics of the system [43, 44]. However, in [29] it was discussed that setting the size of the nanoparticles to be $3\sigma$ in diameter has no qualitative effect on the structures that are formed.

Each lattice site must be in one of three possible states: (i) occupied by a nanoparticle, (ii) occupied by liquid or (iii) unoccupied (as displayed in figure 1). To characterize the state of the system, we introduce two occupation numbers for each lattice site $i$: $n_i$ for nanoparticles and $l_i$ for liquid. These occupation numbers are binary values which describe the state of each site. The occupation numbers for (i) a lattice site containing a nanoparticle would be $n_i = 1$, $l_i = 0$, for (ii) a site occupied by a film of liquid we have $n_i = 0$, $l_i = 1$ and for (iii) a vacant site we have $n_i = 0$, $l_i = 0$. This type of lattice model has the following Hamiltonian [28]:

$$E = -\epsilon_1 \sum_{\langle ij \rangle} l_i l_j - \epsilon_n \sum_{\langle ij \rangle} n_i n_j - \epsilon_{nl} \sum_{\langle ij \rangle} n_i l_j - \mu \sum_i l_i + \sum_i \phi^\ell_l l_i + \sum_i \phi^n_l n_i,$$

where $\sum_{\langle ij \rangle}$ denotes a sum over pairs of nearest neighbours. $\phi^\ell_l$ and $\phi^n_l$ are external potentials acting on the liquid and nanoparticles respectively, at lattice site $i$. Three interaction terms are included which determine the strength of attraction between neighbouring cells. $\epsilon_1$ is the interaction energy between two adjacent cells containing films of liquid, $\epsilon_n$ is for adjacent cells which both contain nanoparticles and $\epsilon_{nl}$ is the energy between a cell containing a nanoparticle and a cell containing liquid. The amount of liquid on the surface is not a conserved quantity because it can evaporate to and condense from a reservoir of vapour above the surface. $\mu$ is the chemical potential of this reservoir. The Hamiltonian (1) has a symmetry between droplets (liquid layer) and holes (nearly dry substrate) ($l_i \rightarrow (1 - l_i)$) which is not present in the real fluid. While this has negligible effect on the qualitative behaviour of the model it is important to note...
that for any quantitative analysis the free energy derived below (equation (8)) should be replaced by a more reliable approximation.

From the Hamiltonian (1) we can derive an expression for the free energy of the system. The probability that a lattice site \(i\) is covered by a liquid film in an equilibrium configuration is given by the following integral over time \(t\):

\[
\rho_i^l = \lim_{\tau \to \infty} \frac{1}{\tau} \int_0^\tau l(t) \, dt. \tag{2}
\]

Similarly, the probability that a lattice site \(i\) contains a nanoparticle is given by the following expression:

\[
\rho_i^n = \lim_{\tau \to \infty} \frac{1}{\tau} \int_0^\tau n_i(t) \, dt. \tag{3}
\]

By choosing the grid spacing \(\sigma\) to be our unit of length, equal to one, these probabilities are also equal to the local number densities for the liquid and nanoparticles. Following the approach described in [45, 46], we make a Bragg–Williams mean-field approximation for the (semi-grand) free energy of the system:

\[
F = k_B T \sum_i \left[ \rho_i^l \ln \rho_i^l + (1 - \rho_i^l) \ln(1 - \rho_i^l) + \rho_i^n \ln \rho_i^n + (1 - \rho_i^n) \ln(1 - \rho_i^n) \right] - \epsilon f_i \sum_{\langle ij \rangle} \rho_i^l \rho_j^l - \epsilon n_i \sum_{\langle ij \rangle} \rho_i^n \rho_j^n
\]

\[+ \epsilon_{nl} \sum_{\langle ij \rangle} \rho_i^n \rho_j^l - \mu_i \sum_i \rho_i^n \]

\[+ \mu_j \sum_i \rho_j^l + \sum_i \phi_i^l \rho_i^l + \sum_i \phi_i^n \rho_i^n, \tag{4}
\]

where \(k_B\) is Boltzmann’s constant and \(T\) is the temperature. This is a semi-grand free energy because the liquid is treated in the grand canonical ensemble (the reservoir of vapour fixes the chemical potential \(\mu\)), whereas the number of nanoparticles in the system is a conserved quantity (so these are treated canonically). Equation (4) is derived using a ‘zeroth-order’ mean-field approximation and thus higher order terms are omitted (e.g. the terms involving \(\ln(1 - \rho_i^n - \rho_i^l)\) which describe the excluded area correlations between the nanoparticles and the liquid). This approximation allows the sum of the two density values \(\rho^n\) and \(\rho^l\) at site \(i\) to be greater than one. However, in the work presented here we consider cases where initially \(\rho_n\) is small and so the effect of omitting such excluded area correlations and the other higher order terms should not qualitatively affect the behaviour of the model. If we assume that the density values only vary on length scales \(\gg \sigma\), then we can define a gradient operator for this discrete system:

\[
\nabla \rho_{i,x,y}^l = (\rho^l_{i+1,x,y} - \rho^l_{i,x,y}, \rho^l_{i,x+1,y} - \rho^l_{i,x,y}),
\]

\[
\nabla \rho_{i,x,y}^n = (\rho^n_{i+1,x,y} - \rho^n_{i,x,y}, \rho^n_{i,x+1,y} - \rho^n_{i,x,y}), \tag{5}
\]

where each lattice site \(i\) is now represented by the coordinates \((x, y)\). Using the operators (5) we can express the summation over pairs of nearest neighbours as

\[
\sum_{\langle ij \rangle} \rho_i^n \rho_j^l = 4 \sum_i \rho_i^n \rho_i^l - \sum_i (\nabla \rho_i^l) \cdot (\nabla \rho_i^n), \tag{6}
\]

where \(\alpha, \beta = n, l\). Substituting (6) into our lattice free energy (4) we obtain

\[
F = k_B T \sum_i [\rho_i^l \ln \rho_i^l + (1 - \rho_i^l) \ln(1 - \rho_i^l) + \rho_i^n \ln \rho_i^n + (1 - \rho_i^n) \ln(1 - \rho_i^n)]
\]

\[+ \epsilon f_i \sum_{\langle ij \rangle} \rho_i^l \rho_j^l - \epsilon n_i \sum_{\langle ij \rangle} \rho_i^n \rho_j^n - \epsilon_{nl} \sum_{\langle ij \rangle} \rho_i^n \rho_j^l - \mu_i \sum_i \rho_i^n - \mu_j \sum_i \rho_j^l + \sum_i \phi_i^l \rho_i^l + \sum_i \phi_i^n \rho_i^n,
\]

where the factor of \(\frac{1}{2}\) is include to avoid double counting. Taking the continuum limit, so that \(\sum_i \to \int d\mathbf{r}, \rho_i^n \to \rho_n(\mathbf{r}), \rho_i^l \to \rho_l(\mathbf{r}), \phi_i^n \to \phi_n(\mathbf{r})\) and \(\phi_i^l \to \phi_l(\mathbf{r})\) where the vector \(\mathbf{r} = (x, y)\) is a continuous variable, we obtain for the free energy of the system:

\[
F[\rho_l, \rho_n] = \int d\mathbf{r} \left[ f(\rho_l(\mathbf{r}), \rho_n(\mathbf{r})) + \frac{\epsilon f_1}{2} (\nabla \rho_l(\mathbf{r}))^2 + \frac{\epsilon f_2}{2} (\nabla \rho_n(\mathbf{r}))^2 + \epsilon_{nl} (\nabla \rho_n(\mathbf{r})) \cdot (\nabla \rho_l(\mathbf{r})) \right]
\]

\[+ \int d\mathbf{r} \rho_l(\mathbf{r}) \phi_n(\mathbf{r}) - \mu \int d\mathbf{r} \rho_n(\mathbf{r}) \phi_l(\mathbf{r}), \tag{8}
\]

where

\[
f(\rho_l, \rho_n) = k_B T [\rho_l \ln \rho_l + (1 - \rho_l) \ln(1 - \rho_l) + \rho_n \ln \rho_n + (1 - \rho_n) \ln(1 - \rho_n)] - 2\epsilon f_1 \rho_l^2 - 2\epsilon f_2 \rho_n^2 - 4\epsilon_{nl} \rho_n \rho_l. \tag{9}
\]

This free energy functional may be employed to determine the phase diagram of the system—i.e. the state of the system in the thermodynamic limit (see section 4). However, the observed patterns are often non-equilibrium structures that are ‘dried in’, i.e., that evolve towards the equilibrium state on a timescale that is much longer than the typical observation times. To model the non-equilibrium processes that result in the observed self-organized structures, one needs kinetic equations for the time evolution of the densities. They are developed in the following section.

### 3. Modelling the dynamics of the system

The chemical potential of the nanoparticles may be calculated using the following functional derivate [47–49]:

\[
\mu_n = \frac{\delta F[\rho_l, \rho_n]}{\delta \rho_n}. \tag{10}
\]

In equilibrium systems the chemical potentials take a uniform value throughout the system. However, this is not the case for non-equilibrium configurations that the system takes during its time evolution. There, the chemical potential varies temporally and spatially over the surface. In particular, non-equilibrium density profiles \(\rho_n(\mathbf{r}, t)\) and \(\rho_l(\mathbf{r}, t)\) give, via equation (10), a non-equilibrium chemical potential for the nanoparticles \(\mu_n(\mathbf{r}, t)\). Thus, the time-dependent densities are coarse-grained ‘average’ quantities. Assuming that locally the system is in equilibrium, we may define these non-equilibrium
density fields in a similar way as the equilibrium densities (see equations (2) and (3)):

\[ \rho^i = \frac{1}{\tau_M} \int_0^\infty \bar{l}_i(t) \, dt, \]

\[ \rho^n = \frac{1}{\tau_M} \int_0^\infty n_i(t) \, dt, \]

where \( \tau_M \) is now a finite time that is large compared to the solvent molecular collision time, but is small compared to the timescale for a nanoparticle to move from one lattice site to a neighbouring lattice site. If we now assume that the driving force which causes a flux of the nanoparticles over the surface of the substrate is given by the gradient of the chemical potential \( \mu_n \), then the nanoparticle current is given by

\[ j_n = -M_n(\mathbf{r}, t) \nabla \frac{\delta F[\rho_n, \rho_l]}{\delta \rho_n(\mathbf{r}, t)}. \]

We expect that when the liquid density is uniform throughout the system (\( \rho(\mathbf{r}) = \rho \)) and the density of the nanoparticles is small everywhere (\( \rho_n(\mathbf{r}) \ll 1 \)) then the nanoparticle dynamics given by equation (14) must reduce to the diffusion equation (i.e. equation (13) becomes Fick’s law). When we apply this to equation (10), for small \( \rho_n \), we get the leading order term:

\[ \mu_n(\mathbf{r}) \approx k_B T \ln \rho_n(\mathbf{r}) + C + O(\rho_n), \]

where \( C = C(\rho) \) represents constant terms. Substituting equation (15) into equation (14) we see that in order for equation (14) to reduce to the diffusion equation the mobility \( M_n \) has to be proportional to \( \rho_n(\mathbf{r}, t) \). Therefore, we write the nanoparticle mobility as:

\[ M_n = \rho_n(\mathbf{r}, t) m(\rho(\mathbf{r}, t)), \]

where \( m(\rho(\mathbf{r}, t)) \) is a function of the local liquid density. We assume that the diffusion of the nanoparticles over the surface is caused by the Brownian ‘kicks’ from the molecules in the liquid. Therefore when the liquid density is low (on the dry substrate) the nanoparticles are almost immobile. However, when the density of the liquid on the surface is high (where the substrate is covered by the liquid film), the nanoparticles are much more mobile. We model this behaviour by a function \( m(\rho(\mathbf{r}, t)) \) which switches between a very small value, when \( \rho_l \) is small and \( \alpha \), which is the mobility coefficient for the nanoparticles in a liquid film, when \( \rho_l > 0.5 \). The precise form of \( m(\rho_l) \) has a negligible effect on the qualitative behaviour of the system. Here we use

\[ m(\rho_l) = \frac{\alpha}{2} \left\{ 1 + \tanh \left[ 30 \left( \rho_l - \frac{1}{2} \right) \right] \right\}. \]

Note that dynamical equations of the form in equation (14), taken together with various expressions for the free energy, have been developed. A notable early example is the Cahn Hilliard equation [50]. In various recent studies, equations of this form with alternative mobility functions are used. These are chosen in order to model various physical effects. For example, the mobility function may take the form

\[ m(\rho) = (1 - \rho/\rho_c)^\delta \] \[ \rho_c \]

where \( \rho_c \) is a critical density value. As the local density approaches this critical value the mobility goes to zero, making \( \rho_c \) an upper bound of the density field. A mobility function of this type is used to model glassy transitions, in systems where jamming of the nanoparticles occurs as the density is increased. Another example is [53], in which a density-dependent mobility was used in order to model the effect of hydrodynamic interactions between colloidal particles.

Next we consider the time evolution of the liquid density \( \rho(\mathbf{r}, t) \). The dominant process governing the dynamics of the liquid is the evaporation and condensation of the liquid between the surface and the vapour reservoir above the substrate. We define two different chemical potentials: \( \mu \) is the chemical potential of the liquid in the reservoir (cf equation (4)) and \( \mu_S(\mathbf{r}, t) = \frac{\delta F}{\delta \rho_l} + \mu \) denotes the local chemical potential of the liquid on the substrate. We assume that the evaporative contribution to the time evolution of the liquid density is proportional to the difference between \( \mu_S(\mathbf{r}, t) \) and \( \mu \). This gives us the following expression:

\[ \frac{\partial \rho_l(\mathbf{r}, t)}{\partial t} = -M^nc \frac{\delta F[\rho_n, \rho_l]}{\delta \rho_l(\mathbf{r}, t)} + \frac{\delta F[\rho_n, \rho_l]}{\delta \rho_l(\mathbf{r}, t)}, \]

where the dynamical coefficient \( M^nc \) is assumed to be a constant. The value of \( M^nc \) determines the rate of the non-conserved (evaporation) dynamics of the liquid. We should also allow for (conserved) diffusive motion of the liquid over the surface. We assume from DDFT that the diffusion of the liquid takes a similar form as the diffusion of the nanoparticles given in equation (14). We therefore model the full liquid dynamics by combining the diffusive and the evaporative terms

\[ \frac{\partial \rho_l(\mathbf{r}, t)}{\partial t} = \nabla \cdot \left[ M^l c \nabla F[\rho_n, \rho_l] \right] = M^nc \frac{\delta F[\rho_n, \rho_l]}{\delta \rho_l(\mathbf{r}, t)} + \frac{\delta F[\rho_n, \rho_l]}{\delta \rho_l(\mathbf{r}, t)}, \]

The mobility coefficient \( M^l c \) for the conserved part of the dynamics is assumed to be constant. The ratio between the conserved and non-conserved mobility coefficients determines the influence that the diffusive/evaporative terms have on the overall dynamics of the liquid (i.e. \( M^nc/M^l c \gg 1 \) corresponds to the case when the liquid dynamics are strongly dominated by evaporation and \( M^nc/M^l c \ll 1 \) corresponds to the case when liquid diffusion plays an important role in the dynamics). Thus, equations (8), (9), (14) and (19), taken together, define our model equations, which govern the dynamics of the system.

Note that when the liquid density \( \rho(\mathbf{r}, t) \) is a constant, the theory reduces to the DDFT developed by Marconi and Tarazona [39], that may be obtained by approximating the
Fokker–Planck equation for a system of Brownian particles with overdamped stochastic equations of motion [39–42, 45]. Equations similar to (19) can also be derived in the context of hydrodynamics. The resulting mesoscopic hydrodynamic thin film equations contain different mobilities and local energies [54, 55]. The combination of the diffusive and the evaporative terms can also be seen as a combination of a conserved Cahn–Hillard-type dynamics with a non-conserved Allen–Cahn-type dynamics [56–58].

4. Phase behaviour

4.1. One component fluid

It is important to understand the equilibrium behaviour of the fluid in our system as this gives us some insight into how the system behaves when it is out of equilibrium. Of particular importance is to determine what phases may we observe and the stability of these phases. Since we are modelling the evaporative dewetting of the liquid we initially seek parameter values which lead to a high density liquid phase (liquid film) coexisting with a low density phase (‘dry’ substrate). Employing a linear stability analysis, we calculate the spinodal curve, i.e., the limit of linear stability for an infinitely extended system. The spinodal curve is defined as the locus of points where the curvature of the free energy is zero, \( \frac{\partial^2 F}{\partial \rho^2} = 0 \), which is equivalent to the isothermal compressibility being zero [47]. Note that in this section we set \( \rho_i = \rho \), for simplicity. We also calculate the binodal curve, i.e., the coexisting density values for a system in equilibrium, by equating the chemical potentials, temperature and pressure in each of the coexisting phases. The area outside the binodal curve is a stable region where we see no phase separation. Inside the binodal curve we have phase separation in the thermodynamic limit. However, the linear stability of the fluid depends on whether the curvature of free energy is positive or negative. When we have positive curvature (outside the spinodal curve), the system at this state point rests within a local minimum of the free energy, i.e., it is linearly but not absolutely stable. There is a free energy barrier that must be traversed to cross into the (absolutely stable) equilibrium phase. This is known as the metastable region, where local fluctuations in the density (if sufficient in size) create nucleation points for the phase transition to occur. When the curvature of the free energy is negative (inside the spinodal curve) there is no free energy barrier. This is the unstable region where we have spontaneous phase separation, i.e. where fluctuations in the densities spontaneously grow. One may also say that the homogeneous fluid layer is linearly unstable to harmonic perturbations with certain wavenumbers.

We first consider the phase behaviour of the reduced case where we have a single component fluid with no nanoparticles (i.e. \( \rho_n = 0 \)), as shown in figure 2. We set the external potential \( \phi_0(\mathbf{r}) = 0 \) which results in a uniform fluid density \( \rho_0(\mathbf{r}) = \rho = \frac{N}{A} \), where \( N \) is the number of ‘particles’ of liquid (i.e. filled lattice sites) and \( A \) is the area of the system. From equation (9) we find that the Helmholtz free energy per unit area of the uniform system is

\[
f = \frac{F}{A} = k_B T \left[ \rho \ln \rho + (1 - \rho) \ln(1 - \rho) \right] - 2 \epsilon_1 \rho^2. \tag{20}\n\]

We define the Helmholtz free energy per ‘particle’ as

\[
a = \frac{F}{N} = k_B T \left[ \ln \rho + \frac{(1 - \rho)}{\rho} \ln(1 - \rho) \right] - 2 \epsilon_1 \rho. \tag{21}\n\]

From this, we may calculate other thermodynamic quantities: the pressure \( P \) and chemical potential \( \mu \), which are given by the following relations [47]:

\[
P = \rho^2 \left( \frac{\partial a}{\partial \rho} \right), \quad \tag{22}\n\]

\[
\mu = a + \rho \left( \frac{\partial a}{\partial \rho} \right). \tag{23}\n\]

We can calculate the spinodal for the one component fluid from the free energy equation (20) and the definition of the spinodal \( \frac{\partial^2 F}{\partial \rho^2} = 0 \), giving us the following equation:

\[
\frac{k_B T}{\epsilon_1} = 4 \rho (1 - \rho). \tag{24}\n\]

Equations (21) and (22) give the following expression for the pressure in the system:

\[
P = -k_B T \ln(1 - \rho) - 2 \epsilon_1 \rho^2. \tag{25}\n\]

In order to simplify the task of determining the phase diagram, we may use the symmetry between filled and empty lattice...
sites, which is present in the Hamiltonian (1). This means that for the one component fluid, the phase diagram is symmetric around the line \( \rho = \frac{1}{2} \), i.e. for a phase on the binodal with a density of \( \rho = \rho_1 \) the coexisting phase has a density of \( \rho_2 = (1 - \rho_1) \). Using equation (25) and this symmetry of the Hamiltonian, we obtain the following expression for the density along the binodal, by equating the pressure in the two phases \( (P_1 = P_2) \):

\[
\frac{k_B T}{\epsilon_1} = \frac{2(2\rho - 1)}{\ln[\rho/(1 - \rho)]}.
\]

The maximum on this curve is at \( \rho = \frac{1}{2} \) and corresponds to the critical temperature \( k_B T/\epsilon_1 = 1 \). Below the critical temperature there are two solutions; these are the coexisting densities. The binodal and spinodal curves for the pure liquid are plotted in figure 2(a). Figure 2(b) shows the value of the chemical potential along these curves.

The spinodal region can also be calculated from the dynamical equations (14) and (19) employing a linear stability analysis. This allows us to determine the typical length scales of the density fluctuations in the liquid film which might exist during the evaporation process. To perform the linear stability analysis we consider a liquid density which varies in space and time \( \rho_l = \rho(r, t) \). The free energy for the single component fluid \( (\rho_0 = 0) \) is given by:

\[
F = \int d\mathbf{r} \left[ f(\rho) + \frac{\epsilon}{2} (\nabla \rho)^2 - \mu \rho \right].
\]

where the subscript on the liquid interaction variable \( \epsilon_1 \) is dropped for simplicity. The steady state solutions of the liquid dynamical equation (19) represent the equilibrium density configurations. There are several steady states for this system, (e.g. a density profile containing a free interface between two coexisting densities with \( \mu = \mu_{\text{coex}} \)) but here we consider the simplest steady state: a flat homogeneous film with a density \( \rho = \rho_0 \) which is defined by:

\[
\frac{\delta F}{\delta \rho} \bigg|_{\rho_0} = 0.
\]

We consider small amplitude perturbations \( \delta \rho \) from \( \rho_0 \) of the form \( \rho = \rho_0 + \delta \rho = \rho_0 + \phi e^{ik \cdot \mathbf{r}} e^{\beta t} \), where the amplitude \( \phi \ll 1 \) is a small positive constant, \( k = |k| \) is the wavenumber and \( \beta \) is the rate of growth/decay with time (for positive/negative values) of the perturbation. We substitute this expression for \( \rho \) into the dynamical equation (19) and expand in powers of \( \delta \rho \).

Then taking just the leading order terms allows us to derive a simple expression for \( \beta \) which can be solved analytically.

A Taylor series expansion of the functional derivative of the free energy equation (27) yields:

\[
\frac{\delta F}{\delta \rho} = \frac{\partial f}{\partial \rho} - \epsilon \nabla^2 \rho - \mu
\]

\[
= \frac{\partial f}{\partial \rho} \bigg|_{\rho_0} + \frac{\partial^2 f}{\partial \rho^2} \bigg|_{\rho_0} \delta \rho + \epsilon k^2 \delta \rho - \mu + O(\delta \rho^2).
\]

Substituting this approximation for the functional derivative (29) into the dynamical equation (19) we obtain

\[
\beta \delta \rho = M_c \nabla \cdot \left[ \rho_0 \left( \frac{i k}{\partial \rho^2} \delta \rho + i e k^3 \delta \rho \right) \right]
\]

\[
- M_{nc} \left( \frac{\partial f}{\partial \rho} \bigg|_{\rho_0} + \frac{\partial^2 f}{\partial \rho^2} \bigg|_{\rho_0} \delta \rho + \epsilon k^2 \delta \rho - \mu \right) + O(\delta \rho^2).
\]

Using the definition of \( \rho_0 \) (equation (28)) gives \( \frac{\partial f}{\partial \rho} \bigg|_{\rho_0} - \mu = 0 \) and neglecting second order terms \( O(\delta \rho^2) \) we arrive at the expression for the growth rate

\[
\beta = -\left( M_c \rho_0 k^2 + M_{nc} \right) \left( \frac{\partial^2 f}{\partial \rho^2} \bigg|_{\rho_0} + \epsilon k^2 \right).
\]

When \( \beta \) is positive, small perturbations from the steady state \( \rho_0 \) with the wavenumber \( k \) grow in amplitude over time. Conversely, if \( \beta \) is negative then small perturbations decay.

Since \( M_c, \rho_0, k^2, M_{nc} \) and \( \epsilon \) are all positive quantities, \( \beta \) is always negative (i.e. the fluid is stable for all wavenumbers \( k \)).

When the second derivative \( \frac{\partial^2 f}{\partial \rho^2} \bigg|_{\rho_0} \) is positive. However when \( \frac{\partial^2 f}{\partial \rho^2} \bigg|_{\rho_0} \) is negative, we find that \( \beta \) is positive for small values of \( k \) and negative when \( k \) is large—i.e. the fluid is unstable against long wavelength (small wavenumber) fluctuations in density. This corresponds to the thermodynamic definition of the spinodal as previously discussed. We may define a critical wavenumber \( k_c \), as the wavenumber at which \( \beta(k_c) = 0 \).

When \( \frac{\partial^2 f}{\partial \rho^2} \bigg|_{\rho_0} < 0 \) the relevant critical wavenumber \( k_c \) is given by:

\[
k_c = \pm \sqrt{-\frac{1}{\epsilon} \frac{\partial^2 f}{\partial \rho^2} \bigg|_{\rho_0}}.
\]

The real system we are modelling is very large \( L \gg 2\pi/k_c \), which means fluctuations can occur on the full spectrum of wavenumbers. The mode with wavenumber \( k = k_m \), which has the largest positive value for \( \beta \) grows the fastest. The wavenumber \( k_m \) corresponds to a typical length scale \( 2\pi/k_m \) which is visible during the early stages of spinodal decomposition. However, at later stages (beyond the linear stage) the length scale of the modulations is likely to deviate from this value as the pattern coarsens. For the purely evaporative case when \( M_c = 0 \), the maximum value of \( \beta(k) \) occurs at \( k = k_m = 0 \), which means there is no typical length scale in the early stages of the evaporation process. However, in the purely diffusive case \( M_c = 0 \) and \( \frac{\partial^2 f}{\partial \rho^2} \bigg|_{\rho_0} < 0 \), \( \beta(k = 0) = 0 \) due to mass conservation, and the maximum value of \( \beta(k) \) occurs at a non-zero value of \( k_m \), so the typical length scale \( 2\pi/k_m \) is visible during the spinodal decomposition–evaporation. In all other cases the following expression is both a necessary and sufficient condition for the existence of the typical length scale (i.e. \( k_m \neq 0 \)):

\[
\frac{M_{nc}}{M_c} < -\frac{\rho_0}{\epsilon} \frac{\partial^2 f}{\partial \rho^2} \bigg|_{\rho_0}.
\]
Binodal curves for the binary mixture for varying fixed values of the chemical potential of the nanoparticles $\mu_n$, for the case when $\epsilon_n/\epsilon_l = 0.43$ and $\epsilon_{nl}/\epsilon_l = 0.57$.

If a typical length scale does exist, then the corresponding wavenumber $k_m$ is given by:

$$k_m = \left[ -\frac{1}{2} \left( \frac{1}{\epsilon} \frac{\partial^2 f}{\partial \rho^2} |_{\rho_0} + \frac{M_{nl}}{\rho_0 M_c} \right) \right]^{1/2}.$$

(34)

Figure 3 displays (a) typical $\beta(k)$ curves for a one component fluid and (b) the values of $1/\epsilon$ and $\rho_0$ where each type of curve is found (for the case when $M_{nl}/M_c = 1$ and $k_B T = 1$). Three distinct cases are displayed: case (i) the solid line (red online) in (a) and the striped area (red online) in (b) display the case when one observes the growth of density fluctuations with a typical length scale. Case (ii), the dashed line (blue online) in (a) and the hashed area (blue online) in (b) correspond to the situation when the fluid is unstable for density fluctuations corresponding to small wavenumbers $k$ but no typical length scale is observed, because the fastest growing mode is the $k = 0$ fluctuation. Case (iii), the dotted line (green online) in (a), corresponding to the (green) solid area in (b), displays the situation when the liquid is linearly stable against density fluctuations with all wavenumbers $k$.

4.2. Binary mixture

To determine the binodal and spinodal curves for the binary mixture is more challenging than for the one component system. There are many more parameters defining the model. In particular, the behaviour of the system strongly depends on the ratios between the different interaction strengths $\epsilon_1$, $\epsilon_n$ and $\epsilon_{nl}$. Reference [59] gives a good overview of the equilibrium bulk fluid phase behaviour for binary fluid mixtures and the different classes of phase diagrams that may be observed.

Here, we only describe the influence of the chemical potential of the nanoparticles $\mu_n$ on the densities in the coexisting phases and on the critical point.

For two phases of a binary mixture to coexist in thermodynamic equilibrium, there are four conditions that must be satisfied. We denote these two phases as (i) the low density phase (LDP) or ‘dry’ substrate and (ii) the high density phase (HDP) or substrate covered by a colloidal film:

$$T_{LDP} = T_{HDP}, \quad \mu_{LDP} = \mu_{HDP}, \quad \rho_{LDP} = \rho_{HDP}, \quad \rho_{LDP}^n = \rho_{HDP}^n.$$

(35)

(36)

(37)

(38)

where $T$ is the temperature, $\mu_1$ and $\mu_n$ are the chemical potentials of the liquid and nanoparticles respectively, $P$ is the pressure and the superscript denotes the phase. The first of these equations is trivial to satisfy in our model. We may then fix the chemical potential of the nanoparticles, to some value $\eta$, and then solve equations (36) and (38), $\mu_{LDP}^n = \eta$ and $\mu_{HDP}^n = \eta$ for the four density values: $\rho_{LDP}^1$, $\rho_{LDP}^n$, $\rho_{HDP}^1$ and $\rho_{HDP}^n$. The curves of the coexisting density values (binodals) for the parameters $\epsilon_{nl}/\epsilon_l = 0.57$ and $\epsilon_n/\epsilon_l = 0.43$ are displayed in figure 4. The density values calculated for the two phases meet at a critical temperature to form binodal curves similar to the one found for the one component system, displayed in figure 2(a). However, we find that the reflection symmetry w.r.t. the liquid density $\rho_1 = 1/2$ seen for the one component fluid is broken. In particular, the critical point of the liquid is no longer at $\rho_1 = 1/2$. The liquid binodal reduces to the one for the pure liquid as $\mu_n \to \pm \infty$. The curves for $\mu_n \to -\infty$ and $\mu_n \to \infty$ are identical within our model due to
the symmetry of the Hamiltonian (1). We observe that the density of the nanoparticles becomes very small \((\rho_n \to 0)\) as the chemical potential decreases \((\mu_n \to -\infty)\). Conversely, the density becomes very large \((\rho_n \to 1)\) as the chemical potential increases \((\mu_n \to \infty)\). The critical point on the nanoparticle binodal curve shifts in a similar manner to that of the liquid binodal.

We now consider the linear stability of the fluid for the two component case. There are many different possible steady states for the binary system including some with interfaces between coexisting phases. However, here we limit ourselves to investigating the linear stability of the simplest steady state: where both components have a uniform constant density over the surface. The uniform density of the nanoparticle film is denoted by \(\rho_n = \rho_n^0\) and the density of the liquid is given by \(\rho_l = \rho_l^0\) where \(\rho_l^0\) is determined from the condition:

\[
\delta F \bigg|_{\rho_l^0} = 0. \tag{39}
\]

We consider small amplitude perturbations in the density of both components from this steady state. We assume these perturbations to take the form:

\[
\rho_l = \rho_l^0 + \delta \rho = \rho_l^0 + \phi e^{ikr} e^{\beta t}, \tag{40}
\]

\[
\rho_n = \rho_n^0 + \chi \delta \rho = \rho_n^0 + \chi \phi e^{ikr} e^{\beta t}, \tag{41}
\]

where the amplitude \(|\phi| \ll 1\) and the parameter \(\chi\) is the ratio between the perturbation in the densities of the two components. The sign of \(\chi\) determines whether any instabilities are in-phase (positive) or anti-phase (negative) between the two coupled density fields. From the magnitude of \(\chi\) we can determine whether an instability is driven by the liquid component \((|\chi| \ll 1)\), the nanoparticles \((|\chi| \gg 1)\) or stems from the interaction between the two components \((|\chi| = O(1))\). Constructing a Taylor series expansion of the functional derivative of the free energy with respect to the liquid density \(\partial^2 F / \partial \rho_l^2\), keeping only terms linear in \(\delta \rho\) and substituting this expression into the liquid dynamical equation (19), we obtain

\[
\beta = - (M_l^0) k^2 + (M_{nl}^0) \times \left( \frac{\partial^2 f}{\partial \rho_l^2} \bigg|_{\rho_l^0, \rho_n^0} + \chi \frac{\partial^2 f}{\partial \rho_l \partial \rho_n} \bigg|_{\rho_l^0, \rho_n^0} \right) + k^2 \epsilon_l + k^2 \chi \epsilon_{nl}. \tag{42}
\]

Turning our attention now to the nanoparticles: in order to linearize equation (14), we must first examine the nanoparticle mobility function \(M_f^0(\rho_l, \rho_n) = \rho_n m(\rho_l)\), given by equations (16) and (17) in our model. Making a Taylor series expansion of the function \(m(\rho_l)\) we obtain

\[
\frac{m}{\alpha} = \gamma_0 + \gamma_1 \delta \rho + O(\delta \rho^2), \tag{43}
\]

where \(\gamma_0 \approx 0\) for small values of \(\rho_l \) \((\rho_l < 0.45)\) and \(\gamma_0 \approx 1\) for large values of \(\rho_l \) \((\rho_l > 0.55)\). Substituting a Taylor series expansion of the functional derivative, together with equation (43), into the time evolution equation for the nanoparticles equation (14) we find

\[
\chi \beta = - \alpha \rho_n^0 \rho_l^2 k^2 \left( \frac{\partial^2 f}{\partial \rho_l^2} \bigg|_{\rho_l^0, \rho_n^0} + \frac{\partial^2 f}{\partial \rho_n \partial \rho_l} \bigg|_{\rho_l^0, \rho_n^0} \right) + k^2 \chi \epsilon_n + k^2 \epsilon_l. \tag{44}
\]

When \(\rho_l\) is small, \(\gamma_0 \approx 0\). In consequence, \(\chi = 0\) and the equations (42) and (44) reduce to the one of the one component fluid (with a local free energy that depends also on \(\rho_n^0\)). We now address the case when \(\rho_l > 0.55\), and we therefore assume \(\gamma_0 = 1\). The expressions for the time dependency of the amplitudes of the two density fields (equations (42) and (44)) can be solved simultaneously to determine \(\beta\) and \(\chi\) as a function of the wavenumber \(k\). This allows us to determine the stability of the fluid for different values of the system parameters. A fact that simplifies the analysis is that these two equations may be written in matrix form (similar to the case of two coupled mesoscopic hydrodynamic equations for dewetting two-layer films [60]):

\[
\beta \left( \begin{array}{c} \chi \\ 1 \end{array} \right) = M \cdot G \left( \begin{array}{c} \chi \\ 1 \end{array} \right), \tag{45}
\]

where,

\[
M = \begin{pmatrix} -\alpha \rho_n^0 k^2 & 0 \\ 0 & -(M_k^0) k^2 + (M_{nl}^0) \end{pmatrix},
\]

\[
G = \begin{pmatrix} k^2 \epsilon_n + f_n k^2 \epsilon_{nl} + f_l \\ k^2 \epsilon_{nl} + f_n k^2 \epsilon_n + f_l \end{pmatrix},
\]

and we have used the shorthand notation \(f_i \equiv \frac{\partial^2 f}{\partial \rho_l \partial \rho_n} \bigg|_{\rho_l^0, \rho_n^0}\), where \(i, j = n, l\) and where \(n\) denotes the nanoparticles and \(l\) denotes the liquid. We can determine \(\beta(k)\) using the following expression for the eigenvalues of a \(2 \times 2\) matrix:

\[
\beta(k) = \frac{\text{Tr}(M \cdot G)}{2} \pm \sqrt{\frac{\text{Tr}(M \cdot G)^2}{4} - |M \cdot G|}. \tag{46}
\]

We define critical wavenumbers \(k = k_c\) for the density fluctuations in the two component fluid as the wavenumbers at which one of the two solutions of equation (46) is equal to zero. Below (equations (49) and (50)) we derive explicit expressions for \(k_c\) which can be used to determine the conditions for the linear stability of the two component fluid (i.e. the fluid is stable when there is no solution for \(k_c\) and the function \(\beta(k) < 0\) for all wavenumbers \(k\)). Since the matrix \(M\) is diagonal and all diagonal elements are non-zero the inverse \(M^{-1}\) exists, allowing us to rewrite equation (45) as a generalized eigenvalue problem

\[
(G - M^{-1} \beta) \left( \begin{array}{c} \chi \\ 1 \end{array} \right) = 0. \tag{47}
\]

Setting \(\beta = 0\) in this equation in order to find the critical wavenumbers \(k_c\), we find that the determinant \(|G| = 0|\).
where (a)–(f) Dispersion relations for the case when

\[ \ln \rho \]

\[ \ln k \]

\[ \ln \alpha \]

\[ \ln f \]

\[ \ln m \]

\[ \ln n \]

\[ \ln t \]

\[ \ln T \]

\( \rho_0 = 0.648, \rho_0 = 0.3, \epsilon_1 = 1.25, \epsilon_n = 0.5, \mu = -3.35, M^0 = 1, M^0 = 0 \) and \( M^0 = 1 \). In (b) and (c) the parameters are:

\[ \rho_0 = 0.648, \rho_0 = 0.3, \epsilon_1 = 1.25, \epsilon_n = 0.5, \mu = -3.35, M^0 = 1, M^0 = 0 \) and \( M^0 = 9 \). In (e) and (f) the parameters are:

\[ \rho_0 = 0.901, \rho_0 = 0.3, \epsilon_1 = 1.4, \epsilon_n = 0.6, \mu = -3.8, M^0 = 1, M^0 = 0 \) and \( M^0 = 1 \). (g) shows how the spinodal line shifts for increasing \( \rho_0 \), where \( \epsilon_n \epsilon_1 - \epsilon^2_n = 0.25, \epsilon_n = \epsilon_1 \) and \( k_b T = 1 \). (h) shows how the spinodal line shifts as the value of \( a = \epsilon_n \epsilon_1 - \epsilon^2_n \) increases, when \( \rho_0 = 0.3, \epsilon_n = \epsilon_1 \) and \( k_b T = 1 \).

implying that

\[ k_c^2 (\epsilon_n \epsilon_1 - \epsilon^2_n) + k_c^2 (\epsilon_n fll + \epsilon f_m - 2 \epsilon f_m) + f_m fll - f^2_n = 0. \]  

In the special case when \( \epsilon_n \epsilon_1 = \epsilon^2_n \), the critical wavenumber is given by:

\[ k_c = \frac{f^2_n - f_m fll}{\epsilon_n fll + \epsilon f_m - 2 \epsilon f_m}. \]

However, more generally, when \( \epsilon_n \epsilon_1 \neq \epsilon^2_n \), we have

\[ k_c = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}. \]

where

\[ a = \epsilon_n \epsilon_1 - \epsilon^2_n, \]

\[ b = \epsilon_n fll + \epsilon f_m - 2 \epsilon f_m, \]

\[ c = f_m fll - f^2_n. \]

Note that the locus \( c = 0 \) is the spinodal curve \([61, 62]\). We categorize the linear behaviour of the system by the signs of \( a \) and \( b \) in equation (51) as they have a profound impact on the shape of the dispersion curves. In figure 5 we display for the case when \( a > 0 \) and \( b > 0 \) all the different possible \( \beta(k) \)

\[ \text{Figure 5. (a)–(f) Dispersion relations for the case when } \epsilon_n \epsilon_1 > \epsilon^2_n, \epsilon_n fll + \epsilon f_m - 2 \epsilon f_m > 0 \text{ (i.e. in equation (51) } a > 0, b > 0 \text{ and } k_b T = 1 \text{. The graphs in the top row show } \beta(k) \text{ and the graphs in the middle row show } \chi(k). \text{ In (a) and (d) the parameters are: } \rho_0 = 0.648, \rho_0 = 0.3, \epsilon_1 = 1.25, \epsilon_n = 0.5, \mu = -3.35, M^0 = 1, M^0 = 0 \) and \( M^0 = 1 \). In (b) and (c) the parameters are:

\[ \rho_0 = 0.648, \rho_0 = 0.3, \epsilon_1 = 1.25, \epsilon_n = 0.5, \mu = -3.35, M^0 = 1, M^0 = 0 \) and \( M^0 = 9 \). In (e) and (f) the parameters are:

\[ \rho_0 = 0.901, \rho_0 = 0.3, \epsilon_1 = 1.4, \epsilon_n = 0.6, \mu = -3.8, M^0 = 1, M^0 = 0 \) and \( M^0 = 1 \). (g) shows how the spinodal line shifts for increasing \( \rho_0 \), where \( \epsilon_n \epsilon_1 - \epsilon^2_n = 0.25, \epsilon_n = \epsilon_1 \) and \( k_b T = 1 \). (h) shows how the spinodal line shifts as the value of \( a = \epsilon_n \epsilon_1 - \epsilon^2_n \) increases, when \( \rho_0 = 0.3, \epsilon_n = \epsilon_1 \) and \( k_b T = 1 \).

\[ k_c^2 (\epsilon_n \epsilon_1 - \epsilon^2_n) + k_c^2 (\epsilon_n fll + \epsilon f_m - 2 \epsilon f_m) + f_m fll - f^2_n = 0. \]

curves, together with the corresponding \( \chi(k) \). From equation (46) we see that there are two branches for \( \beta(k) \), which we denote \( \beta^+ (k) \) and \( \beta^- (k) \). The corresponding \( \chi(k) \) curves are denoted \( \chi^+ \) and \( \chi^- \), respectively. The \( \beta^+ \) branch (red solid lines in figures 5(a)–(c)) corresponds to the second term in equation (46) being positive and the \( \beta^- \) branch (dashed lines) corresponds to the second term being negative. Due to mass conservation of the nanoparticles, there is always one \( \beta(k) \) branch that is zero at \( k = 0 \), and \( \chi = 0 \) at \( k = 0 \) for the curve that corresponds to the \( \beta(k) \) that is not zero at \( k = 0 \). If \( f_m \geq 0 \) then \( \beta^+(k = 0) = 0 \) and \( \chi^-(k = 0) = 0 \). Alternatively, if \( f_m \leq 0 \) then \( \beta^-(k = 0) = 0 \) and \( \chi^+(k = 0) = 0 \).

Inspecting figure 5, we find that there are three possible forms for \( \beta(k) \), similar to the one component fluid case (figure 3). In figure 5(a) there is a maximum in \( \beta^+ \) at \( k \neq 0 \). This indicates that the fluid is unstable and that one will observe the growth of density fluctuations with a typical process. Here \( k_m^0 \) refers to the wavenumber at the maximum of \( \beta^+ \). In the second case (figure 5(b)) the fluid is linearly unstable for density fluctuations with small wavenumbers \( k \) as in figure 5(a), but the maximum in \( \beta^+(k) \) occurs at \( k_m^0 = 0 \), i.e. there is no typical length scale visible in the density profiles. In the last case, shown in figure 5(c), the fluid is stable for all wavenumbers \( k \). We can use equation (50) for the critical wavenumber \( k_c \) to determine the stability of the
system. Since $a > 0$ and $b > 0$ we observe that there is at most one positive root of $\beta(k)$, which only exists if $c < 0$ (i.e. inside the spinodal). The spinodal curve when $a > 0$ and $b > 0$ is plotted in figures 5(g) and (h). In (g) we show how the spinodal line moves upwards in $1/\epsilon_l$ as the nanoparticle density $\rho^0_n$ is increased. In (h) we show how the shape of the spinodal changes as the value of the parameter $a$ is increased. We find that larger values of $a$ make the spinodal curve flatter.

In the reduced case when $a = 0$, one can determine from equation (51) that $b$ is always positive. We find the same three types of dispersion relations for this reduced case as we do for the $a > 0$, $b > 0$ case. For the case when $a > 0$ and $b < 0$ we observe that the fluid is unstable for all possible combinations of parameter values. When we have $a < 0$ (where at least one of the components of the mixture is more attracted to the other component than to itself), we find that $b$ must be positive. In this regime, we observe that as $k \to \infty$, $\beta^+(k) \to \infty$ and $\beta^-(k) \to -\infty$. This indicates that the density fluctuations with an infinitesimally small typical length scale will grow fastest. This behaviour corresponds to a mixture that exhibits micro-phase separation. This behaviour is common in block copolymer systems, where chemical bonding prevents macroscopic demixing and instead demixing on the nanoscale is witnessed [63, 64]. Micro-phase separation is also observed in certain colloidal suspensions [65–69].

An important consideration which must be made with the binary system is whether ‘liquid–particle’ phase separation can occur as well as the ‘liquid–gas’ (low density–high density) phase separation that we have already discussed. The former corresponds to the coexistence of two phases, both having a high density. In terms of our system this would mean that we have coexisting phases with a high liquid density (i.e. $\rho_l \gtrsim 0.6$) in each phase. The coexisting values depend upon the temperature $T$, chemical potential $\mu$ and the interaction energies $\epsilon_l$, $\epsilon_n$ and $\epsilon_{nl}$. It is known that the following condition must be satisfied for ‘liquid–particle’ demixing to occur [59]:

$$\epsilon_{nl} < \frac{\epsilon_n + \epsilon_l}{2}. \quad (52)$$

The existence of liquid–particle phase separation in addition to the gas–liquid phase separation implies that for certain parameter values we may have three coexisting phases. This situation has the potential to lead to dramatic consequences for the pattern formation in our dynamical system. We return to this issue in section 5.4 below.

5. Nonlinear dynamics

We now go beyond the linear analysis presented above and discuss numerical results for the fully nonlinear time evolution.

5.1. Numerical setup

Recall that the dynamics of our model is governed by the coupled partial differential equations (14) and (19), together with the free energy functional equation (8). We numerically solve these nonlinear partial differential equations using a finite difference scheme on a square lattice with grid spacing $\Delta x = 1$. The time step size $\Delta t$ varies between simulations, as the stability of our numerical scheme depends strongly on the values of the parameters in the model. Central difference approximations are made for the partial derivatives with respect to space and forward difference approximations are made for the partial derivatives with respect to time. The Laplacian terms $(\nabla^2)$ are approximated using the eight-neighbour discretization [70]:

$$\nabla^2 \rho = \frac{1}{(\Delta x)^2} \left( \sum_{NN} \rho_{NN} + \frac{1}{2} \sum \rho_{NN} - 6\rho \right). \quad (53)$$

where $\sum \rho_{NN}$ denotes a sum over the nearest neighbour lattice sites and $\sum \rho_{NN}$ denotes a sum over the next nearest neighbour lattice sites. Alternative approximations may be used [71]. However, the choice of Laplacian approximation has little effect on the qualitative behaviour of the system.

Our numerical results show that as the various parameters in the model are varied, several different patterns are formed. We begin in section 5.2 by considering the effect of changing the chemical potential $\mu$ of the vapour reservoir. We then focus on the fingering instability. In section 5.3 we discuss the effect of varying the parameter $\alpha$ in the nanoparticle mobility and also the role which the conserved part of the liquid dynamics plays in the overall dynamics of the system, by varying the mobility coefficient $M_1$. Finally, in section 5.4 we discuss the influence of ‘liquid–particle’ demixing at the receding front and how this affects the fingering mechanism.

For the simulation results shown in sections 5.2 and 5.3 we set the interaction energies to $\epsilon_l = 1.4$, $\epsilon_n = 0.6$ and $\epsilon_{nl} = 0.8$. Using the linear stability analysis we have shown that it is possible to obtain stable phases with these parameter values (figure 5(c), $a > 0$ and $b > 0$ in equation (51)).

5.2. Influence of the vapour chemical potential $\mu$

In section 4.1 we have discussed how changing the value of $\mu$, the chemical potential of the vapour reservoir above the surface, affects the structures displayed by the system as the pure liquid evaporatively dewets from the surface. Recall that as $\mu$ is decreased below its coexistence value $\mu_{coex}$, the dewetting mechanism is at first via the nucleation of holes and then when $\mu$ is further decreased, via spinodal dewetting. This sequence as $\mu$ is decreased is also observed when there are nanoparticles dispersed in the liquid although, as previously discussed, the phase boundaries are shifted and there is the possibility of other (liquid–particle) phase transitions. Choosing a binary mixture with the parameter values $k_B T = 1.0$, $\epsilon_l = 0.6$, $\epsilon_n = 1.4$ and $\epsilon_{nl} = 0.8$ gives a phase with a high liquid density coexisting with a low density of the liquid. We also set $M_1 = 0$ and $M_0^p = 1$ to allow us to initially focus on just the evaporative non-conserved dynamics of the liquid. We set the initial density profiles corresponding to a (high density) uniform film of liquid with density $\rho(x, t = 0) = 1 - 10^{-6}$ mixed with nanoparticles having an average density of $\rho_{av}^n = 0.3$. In order to allow for the growth of density fluctuations when the system is (linearly) unstable, we
Density profiles displaying evaporation via spinodal decomposition. The top row shows the liquid density profiles and the bottom row shows the nanoparticle profiles at times $t/n = 7$ (left), $t/n = 8$ (centre) and $t/n = 9$ (right), where $n = \frac{N}{V}$. The system parameters are: $k_BT = 1$, $\epsilon_l = 1.4$, $\epsilon_n = 0.6$, $\epsilon_{nl} = 0.8$, $M_1^\alpha = 0$, $M_1^{\mu} = 1$, $\alpha = 0.5$, $\beta\mu = -4.08$ and $\lambda = 0.005$.

Figure 6. Density profiles displaying evaporation via spinodal decomposition. The top row shows the liquid density profiles and the bottom row shows the nanoparticle profiles at times $t/n = 7$ (left), $t/n = 8$ (centre) and $t/n = 9$ (right), where $n = \frac{N}{V}$. The system parameters are: $k_BT = 1$, $\epsilon_l = 1.4$, $\epsilon_n = 0.6$, $\epsilon_{nl} = 0.8$, $M_1^\alpha = 0$, $M_1^{\mu} = 1$, $\alpha = 0.5$, $\beta\mu = -4.08$ and $\lambda = 0.005$.

add a small random noise amplitude to the density profile of the nanoparticles. Thus, the initial nanoparticle density profile is $\rho_n(x, t = 0) = \rho_n^{av} + 2\lambda(Y - 0.5)$, where $Y$ is a random real number uniformly distributed between 0 and 1 and $\lambda$ is the magnitude of the noise. Without these random density fluctuations, the density profiles would remain uniform under the evolution of the DDFT with a film of liquid remaining. Our boundary conditions are periodic in all directions.

The $c = 0$ spinodal curve (as calculated in section 5.1, figure 5) defines the limit of stability, i.e. inside this line the fluid becomes linearly unstable. Thus, the fluid is unstable when $\beta\mu < -3.869$, where $\beta = 1/k_BT$. The speed of the process increases with decreasing values of $\beta\mu$. For very low values of $\beta\mu$, the evaporation process is so fast that we do not see any pattern formation in the nanoparticles—the liquid evaporates too quickly for the nanoparticles to diffuse. For the parameters $\epsilon_l = 1.4$, $\epsilon_n = 0.6$, $\epsilon_{nl} = 0.8$, $\alpha = 0.5$, $M_1^\alpha = 0$ and $M_1^{\mu} = 1$ this occurs when $\beta\mu \lesssim -4.2$. Figure 6 shows the particular case when $\beta\mu = -4.08$. We see that the liquid behaves in a similar manner to that of a single component fluid by spontaneously dewetting everywhere. Now the evaporation is slow enough for the nanoparticles to move into areas with a high density of liquid during this evaporative process which creates a fine network structure (as shown in figures 6(e) and (f)). However, this diffusion of the nanoparticles is limited as it is still a much slower process than the evaporation of the liquid. We observe that towards the end of the process small heaps of nanoparticles are formed, where the density is significantly larger. This effect is enhanced by the attraction between the nanoparticles ($\epsilon_n > 0$). Increasing the value of $\mu$, we move from the linearly unstable (spinodal) region into the metastable region of the phase diagram (cf figure 2). Note, however, that the actual values of $\beta\mu$ on the binodal and spinodal now differ from those in figure 2 because of the inclusion of nanoparticles, cf figure 4.

On increasing the chemical potential into the range $-3.869 < \beta\mu \lesssim -3.8$, the liquid film becomes metastable but may still evaporate through the nucleation and growth of holes. Figure 7 shows the case when $\beta\mu = -3.86$. The nucleation is caused by the random fluctuations in the density distributions (the initial density profiles are defined in a similar manner to the previous case: $\rho_l(x, t = 0) = 1 - 10^{-6}$, $\rho_n(x, t = 0) = \rho_n^{av} + 0.4(Y - 0.5)$). The amount of noise used and the free energy ‘barrier’ for forming a hole determine the probability of a nucleation event occurring. There is a critical hole radius $R^c$ which can be determined from the free energy of the system. If a hole is smaller than this critical radius then it will shrink and the liquid density will return to its bulk high density value in this region. However, if the size of the hole is larger than this critical value then this hole will begin to grow. We can apply classical nucleation theory to calculate an estimate for the critical hole radius $R^c$ by determining the change in free energy $\Delta F$ when a low density (thin film) circular ‘hole’ with a radius of $R$ is inserted into the metastable liquid film. The radius $R$ which corresponds to the maximum in $\Delta F$ is the critical hole radius $R^c$. We approximate the change in free energy using the formula:

$$\Delta F = \pi R^2 \Delta P + 2\pi R\gamma,$$  \hspace{1cm} (54)
where $\Delta P$ is the pressure difference between the two phases (the hole and the fluid film). $\gamma$ is the interfacial tension (excess free energy) for creating a straight interface between the two phases at coexistence $\mu = \mu_{\text{coex}}$. It is important to note that the density values at coexistence are different to the density values out of coexistence. The density of the thin film of fluid inside the hole is such that its chemical potential is equal to that of the bulk film of fluid surrounding the hole. The critical hole radius $R_c$ is given by the maximum of equation (54)—i.e. when $\frac{\partial \Delta F}{\partial R} = 0$. Thus, the critical hole radius is $R_c = -\frac{\gamma}{\Delta P}$.

Figure 8 shows how $R_c$ depends on the chemical potential $\mu$ for different nanoparticle densities $\rho_n$. The lines start at the lowest value of $\mu$ for which the system is linearly stable. The dotted lines show the value of the chemical potential at coexistence $\mu_{\text{coex}}$, which the curves approach asymptotically. Here $k_B T = 1, \epsilon_l = 1.4, \epsilon_n = 0.6, \epsilon_{nl} = 0.8$. This analysis only applies to the metastable region of the phase diagram (cf figure 2) and therefore the critical hole radius $R_c$ curves are bounded on the left by the spinodal curve and on the right by the binodal curve. Note that classical nucleation theory incorrectly predicts a finite value for $R_c$ at the spinodal, due to the fact that the theory assumes a sharp interface as one approaches the spinodal. For further discussion on this see e.g. [72]. The probability for a hole to be nucleated by random thermal fluctuations is proportional to $e^{-\beta \Delta F}$. The curves in figure 8 show that as we approach $\mu_{\text{coex}}$ the size of the critical hole increases. Hence, the probability of nucleation is greater nearer the limit of stability (spinodal curve) and decreases greatly as we approach coexistence (binodal curve). We observe that increasing the density of the nanoparticles $\rho_n$ shifts the metastable region to lower chemical potential values and increases the range of the metastable region. This is due to the increase in the critical temperature associated with the increase in the nanoparticle density $\rho_n$, as previously discussed.

In the case shown in figure 7, we are near the limit of stability where the critical radius of a hole is very small. This results in many nucleation points where holes are formed and begin to grow. The nanoparticles are picked up by these growing holes which creates a rim around each hole with a high density of nanoparticles in the rim. The holes in the liquid film continue to grow until their rims meet, creating
a random polygonal network pattern of nanoparticles. The liquid wets the surface of the nanoparticles, which means the liquid remains on the surface in areas with a high density of nanoparticles. This is due to the positive interaction energy between the liquid and the nanoparticles ($\epsilon_{nl} > 0$).

If we increase the chemical potential further into the metastable range $-3.8 \lesssim \beta \mu < \mu_{\text{coex}}$, then the probability of a hole being nucleated becomes much smaller. In an experiment, in this parameter range, all holes that are formed are normally nucleated at defects or impurities of the surface (heterogeneous nucleation). Any interfaces between a high density liquid phase and a low liquid density phase will recede ($\lambda t^3$ $\simeq$ $n l$).

Recall that one of the goals of our work is to develop an understanding of how the different self-organized structures of nanoparticles observed in the experiments [22, 25, 26, 23] are formed. Distinct observed structures are (a) network structures and (b) branched structures. Results from our model have shown how two different types of network structures can develop: (i) a fine network structure created by a spinodal evaporation process (figure 6) in which the nanoparticle density varies over a small range $0.27 \lesssim \rho_n \lesssim 0.45$, (ii) a large well defined network structure created by the nucleation and growth of holes in the liquid (figure 7), in which the nanoparticle density varies over a large range $0.05 \lesssim \rho_n \lesssim 0.9$. Our model also shows how instabilities at the evaporative dewetting front can create branched structures for certain parameter values (figure 9). Note that there is also evidence of early stages of the fingering instability in the nucleation case shown in figure 7. There one can observe that small bumps begin to develop in the edges of some of the larger holes. We now consider the formation of the branched structures in more detail. In particular, we investigate the dependence of these ‘fingered’ structures on the parameters of the model.
5.3. Influence of mobilities on the fingering

To make a detailed investigation of the branched finger structures it is important to maximize the distance a front can recede. This allows us to obtain better statistics which is important due to the fact that solving the DDFT in the fingering regime on a large grid can be time consuming. To achieve this objective we create a straight dewetting front along the bottom edge of the system (i.e. we set the two density values to \( \rho_l = \rho_h = 10^{-6} \) for the first five horizontal lines). We also set no-flux boundary conditions at the top and the bottom, to prevent a dewetting front forming at the top. The periodic boundary conditions on the left and the right side of the system domain remain. This setup also allows for easier analysis of the branched structures since we begin with an initially straight front. We define a measure for the average number of fingers \( \langle f \rangle \) to be the average number of branched structures per unit length in the final density profile, after the dewetting front has reached the top of the system. To calculate this quantity we implemented an algorithm which counts the number of transitions between a high density of nanoparticles and a low density of nanoparticles on each horizontal line of the system. We then determine the number of fingers on a given horizontal line by dividing this value by two. We set a minimum and maximum line for a given set of final density profiles and calculate the average number of fingers between these two lines. This value is then divided by the size of the system to give a value that is independent of the system size. We have investigated how this measure is affected by the different parameters of the system.

We begin by discussing the effect that varying the parameter \( \alpha \) has on \( \langle f \rangle \) (recall that \( \alpha \) determines the mobility of the nanoparticles in the liquid film). We use the same parameter values as above, with \( \beta \mu = -3.8 \) and \( \alpha \) varying between 0.1 and 3. Figure 10 shows final nanoparticle density profiles for several values of \( \alpha \) and also a plot of \( \langle f \rangle \) versus \( \alpha \), which is calculated from the average of five runs. We see that the value of \( \alpha \) has a significant influence on the average finger number. Increasing the mobility of the nanoparticles results in fewer fingers being developed, which is in (qualitative) agreement with the experiments [23] and the KMC model results [29]. The mobility of the nanoparticles directly influences the speed of the receding front, so evaporation is much slower when \( \alpha \) is small. When \( \alpha \) is very small (\( \alpha \lesssim 0.002 \)) we find that the two density fields become practically decoupled, with the liquid evaporating at high speed leaving the nanoparticles behind as a homogeneous film of the initial density \( \rho_0^{\text{av}} \).

The effect of liquid diffusion over the surface has also been investigated by varying the liquid conserved mobility \( M_l^c \). Using the same parameter values as above, and setting \( \alpha = 1 \), we display in figure 11 final nanoparticle density profiles for varying \( M_l^c \) from 0 to 15 and also the average finger number \( \langle f \rangle \) versus \( M_l^c \) averaged over five runs. We see that the diffusive mobility of the liquid \( M_l^c \) does affect the average number of fingers but to a much smaller extent than the mobility of the nanoparticles. The average finger number generally increases as \( M_l^c \) is increased.

5.4. Influence of liquid–particle demixing on the fingering

We now discuss the effect of possible liquid–particle phase separation on the front instability. Such a phase separation may occur near the front even for nanoparticle concentrations inside the liquid film that are far smaller than the binodal value for liquid–particle phase separation. This occurs because as a
As we increase $\epsilon_n$ further, for $\epsilon_n > 1$ we observe that the dynamics at the front remains similar, however, now all the finger structures are very thin and therefore quickly break up into droplets. One also notices that the tendency to form side branches decreases with increasing $\epsilon_n$ whereas the orientation of the branches becomes increasingly perpendicular to the receding front. We also observe an increase in the density of the nanoparticles within the fingers/droplets as $\epsilon_n$ is increased. This is due to the increased attraction between the nanoparticles. These results agree qualitatively with the KMC model [29]. The KMC results also show an initial increase in the number of fingers followed by a transition from fingers to droplets as the interaction energy between the nanoparticles $\epsilon_n$ is increased. Figure 17 of [29] displays a plot of $\langle f \rangle$ versus $\epsilon_n$ which shows a similar trend as the DDFT results displayed in figure 12(e). However, as it is a discrete stochastic model, the details of the transition in the way the branching occurs are less discernible than in the present DDFT model.

6. Concluding remarks

We have presented a DDFT based model for the evaporative dewetting of an ultrathin film of a colloidal suspension. We have derived an expression for the free energy of the system using a mean-field approximation for a coarse-grained Hamiltonian model (1) for the system. We have also derived dynamical equations which describe the diffusive dynamics of the solvent and of the colloids as well as the evaporation of the solvent. We have considered the equilibrium phase behaviour of the pure solvent and of the two component fluid and identified parameter ranges where unstable, metastable and stable phases exist. We then solved the coupled dynamical equations numerically to investigate

Figure 11. Nanoparticle density profiles for (a) $M_f^i = 0$, (b) $M_f^i = 5$, (c) $M_f^i = 10$ and (d) $M_f^i = 15$. In (e) we display a plot showing how the number of fingers $\langle f \rangle$ depends on the value of $M_f^i$. The parameter values are: $k_B T = 1$, $\epsilon_1 = 1.4$, $\epsilon_2 = 0.6$, $\epsilon_{ai} = 0.8$, $M_f^\infty = 1$, $\alpha = 1$, $\beta \mu = -3.8$ and $\lambda = 0.1$. 

Dewetting front recedes it collects nanoparticles (as previously discussed) and therefore increases the value of $\rho_l$ near the front. For certain parameter values we find that if $\rho_l$ increases above a certain threshold value, then liquid–particle phase separation occurs in the front region. The liquid separates into two liquid phases, a mobile one poor in colloids and a less mobile one rich in colloids. To investigate the resulting effects we set the interaction energies to $\epsilon_l = 1.7$, $\epsilon_{ai} = 1$ and vary $\epsilon_n$ from 0 to 1.2. Equation (52) indicates that we should observe liquid–particle phase separation for $\epsilon_n > 0.3$. We set the average nanoparticle density to be low, $\rho_n^\infty = 0.1$, so that there is no liquid–particle phase separation in the bulk of the fluid film. For the value of the chemical potential that we use $\mu = -4$, the fluid is linearly stable for all values of $\epsilon_n$ and leads to a relatively fast dewetting front.

In figures 12(a)–(d) we display typical final nanoparticle density profiles for various $\epsilon_n$ and the average finger number $\langle f \rangle$ versus $\epsilon_n$, averaged over five runs. As $\epsilon_n$ is increased from $\epsilon_n = 0$, we initially see a linear increase in the average number of fingers $\langle f \rangle$. This reaches a peak at $\epsilon_n \approx 0.6$, after which we begin to see the development of droplets. When $0.6 \lesssim \epsilon_n \lesssim 0.9$ we observe separation between regions of high density of nanoparticles and low density of nanoparticles occurring locally at the dewetting front. The areas with a lower density of nanoparticles form very thin fingers which quickly rupture into a series of droplets, whereas the areas with a higher density form thicker fingers which are much more stable as shown in figure 12(c). The sections of the front with a lower density of nanoparticles recede faster than the rest of the front. This results in a ‘doublon’ pattern which has been observed in many different systems, e.g. in the thin film directional solidification of a nonfaceted cubic crystal [73]. As we increase $\epsilon_n$ further, for $\epsilon_n > 1$ we observe that the
Figure 12. Nanoparticle density profiles for (a) $\epsilon_n = 0.3$, (b) $\epsilon_n = 0.6$, (c) $\epsilon_n = 0.8$ and (d) $\epsilon_n = 1.1$. In (e) we display a plot showing how the mean number of fingers $\langle f \rangle$ depends on the parameter $\epsilon_n$. When $\epsilon_n > 0.6$ we begin to observe droplets being deposited along with the branched structures (as shown in (c)). The parameter values for these calculations are: $k_B T = 1$, $\epsilon_l = 1.7$, $\epsilon_{nl} = 1.0$, $M_c = 0$, $M_{nc} = 1$, $\alpha = 0.5$, $\mu = -4$, $\rho_{av} = 0.1$ and $\lambda = 0.025$.

the different dynamical pathways of the phase transition and the resulting self-organized patterns of the nanoparticles.

The model successfully describes the various self-organized structures found in experiments [23] and is in qualitative agreement with the discrete stochastic KMC model [29]. Our numerical results show how nanoparticle network structures can form either from a spinodal processes (figure 6) or through the nucleation and growth of holes (figure 7). We have also observed how branched structures develop from a fingering instability of the receding dewetting front (figure 9).

The transverse front instability results from a build-up of the nanoparticles close to the front as the solvent evaporates, when diffusion is too slow to disperse them. This slows down the front and renders it unstable. As a result, density fluctuations along the front grow into an evolving fingering pattern. This transverse front instability can be considered to be a self-optimization process which maintains the mean front velocity constant [29] (see also the discussion of this in the context of a similar front instability occurring in the dewetting of non-volatile polymer films [74]). One may also say that the constant average front velocity is maintained by depositing some of the nanoparticles onto the dry substrate creating the branched structures. Experimental observations show that the branched structures found in the ultrathin film behind the mesoscopic dewetting front are initiated from random nucleation sites. The holes which are nucleated then grow, initially creating circular dewetting fronts. We subsequently observe that the fingering instabilities and the development of branched structures form on the circular interfaces. Figure 9 shows how these circular branched patterns develop from a single nucleation point; our numerical results bare a striking resemblance to the experimental AFM images of this phenomenon [23].

We have studied the branched structures in greater detail using a planar geometry, i.e., by creating initially straight dewetting fronts. We have considered how the different mobilities affect the fingering. The nanoparticle mobility in liquid films has a significant effect on the average number of fingers in the branched structure (figure 10). The finger number decreases rapidly with increasing mobility in agreement with earlier KMC results [29]. This behaviour can be attributed to the lower mobility of the nanoparticles that hinders re-distribution by diffusion and also reduces the speed of the dewetting front. For the system to attain a higher front speed it must deposit nanoparticles onto the surface at a greater rate. Therefore, if the mobility of the nanoparticles is low this leads to the creation of more fingers because in this case the average distance a nanoparticle has to travel to reach a finger is smaller. Increasing the mobility for the conserved diffusive dynamics of the liquid has the opposite effect on the average number of fingers (figure 11). The complex relationship between the diffusion of the liquid and the average finger number is not yet fully understood. Our hypothesis is that increasing the mobility of the liquid results in an effective increase in the speed of the dewetting dynamics of the liquid for fixed nanoparticle mobility. Thus, the mobility of the nanoparticles becomes lower in comparison. The increased finger number then results from the increased mobility contrast, in agreement with the general instability mechanism laid out above.

The basic front instability as described above is a purely dynamic effect and does not depend on particle–liquid and
particle–particle attractive interactions that favour demixing of the liquid and the nanoparticles. However, beside this regime (which we call the ‘transport regime’) we have investigated how interactions that favour demixing influence the instability (figure 12). In general, when increasing the interaction energy between the nanoparticles one increases the tendency towards liquid–particle demixing. However, this has no practical effect as long as the nanoparticle concentration is low, so that it is outside of the two-phase region. This is normally the case for our initial densities. However, in the course of the evaporative dewetting the density increases close to the receding front. Increasing the interaction energy between the nanoparticles causes demixing to occur close to the front (but not in the bulk film). The demixing makes the fingering instability stronger (we call this the ‘demixing regime’). At first, one finds a linear increase in the average finger number with increasing interaction energy. At higher values of $\epsilon_{\text{nn}}$, when the localized phase separation sets in, the fingers become straight with less side branches, before finally lines of drops are emitted directly at the front. In this regime, the mean number of fingers is determined by the dynamics and the energetics of the system.

The results we have obtained with our DDFT model confirm that jamming of discrete particles (as already discussed in [37]) is not a necessary factor for the fingering instability to occur. Our model is a continuum model with a diffusion constant that is independent of the nanoparticle concentration. The present two-dimensional DDFT model has several advantages over the two-dimensional KMC model [28, 29]: in particular, the early instability stages are more easy to discern without the background noise of the KMC. Furthermore, the underlying free energy may be employed to analyse the equilibrium phase behaviour in detail, in a similar manner to [59]. Many standard tools for the analysis of partial differential equations can be applied to the coupled evolution equations, such as, e.g., the linear stability analysis of the homogeneous films. In the future, one may perform a linear stability analysis for the receding straight front and also investigate steady state solutions as has been done for evaporating films of pure liquids [55]. There are many details that would merit further investigations such as, for example, the doublon structure mentioned in section 5.4 and its relation to such structures formed in directional solidification [75].

The present DDFT model does not include the effect of surface forces, i.e., wettability effects (substrate–film interactions). Therefore, a non-volatile liquid film would not dewet the substrate. This implies that an important avenue for future improvement is to incorporate wettability effects into the model. This could be done by making a mean-field approximation to derive an expression for the free energy for a fully three-dimensional KMC model [76, 77] (after incorporating substrate–particle and substrate–liquid interactions). The resulting three-dimensional DDFT could then either be used directly or be averaged perpendicularly to the substrate employing e.g., a long-wave approximation. Another possible option consists of combining a mesoscopic hydrodynamic approach, e.g., a thin film evolution equation (see [31, 78, 55]) with elements of DDFT. For a brief discussion of such an approach see [52].

As a final remark, we recall that in the present work we have only considered dewetting from homogeneous substrates. However, it is straightforward to include surface heterogeneities in our model via the external potentials $\phi(r)$ in equation (8). As future work, it would be interesting to study the influence of surface patterning on the finger formation displayed by the present system.

Acknowledgments

This work was supported by the EU via the ITN MUL-TIFLOW (PITN-GA-2008-214919). MJR also gratefully acknowledges support from EPSRC and AIA thanks RCUK for support.

References

[1] Giacometti A, Maritan A and Banavar J R 1995 Phys. Rev. Lett. 75 577
[2] Mimura M, Sakaguchi H and Matsushita M 2000 Physica A 282 283
[3] Yochelis A, Tintut Y, Demer L L and Garfinkel A 2008 New J. Phys. 10 055002
[4] Vlasov Y A, Bo X-Z, Sturm J C and Norris D J 2001 Nature 414 289
[5] Mitlin V S 1993 J. Colloid Interface Sci. 156 491
[6] Reiter G 1992 Phys. Rev. Lett. 68 75
[7] Xie R et al 1998 Phys. Rev. Lett. 81 1251
[8] Seemann R, Herminghaus S and Jacobs K 2001 Phys. Rev. Lett. 86 5534
[9] Thiele U, Vélarde M G and Neuffer K 2001 Phys. Rev. Lett. 87 016104
[10] Thiele U 2003 Eur. Phys. J. E 12 409
[11] Becker J et al 2003 Nature Mater. 2 59
[12] Beltrame P and Thiele U 2010 SIAM J. Appl. Dyn. Syst. 9 484
[13] Redon C, Brochard-Wyart F and Rondelez F 1991 Phys. Rev. Lett. 66 715
[14] Sharma A and Reiter G 1996 J. Colloid Interface Sci. 178 383
[15] Sharma A and Khanna R 1998 Phys. Rev. Lett. 81 3463
[16] Bestehorn M and Neuffer K 2001 Phys. Rev. Lett. 87 046101
[17] Thiele U, Mertig G and Pompe W 1998 Phys. Rev. Lett. 80 2869
[18] Xu R, Raghavan D, Douglas J F and Karim A 2002 J. Polym. Sci. B 40 2825
[19] Xu L, Shi T, Dutta P K and An L 2007 J. Chem. Phys. 127 144704
[20] Maillard M, Motte L, Ngo A T and Pileni M P 2000 J. Phys. Chem. B 104 11871
[21] Ge G L and Brus L 2000 J. Phys. Chem. B 104 9573
[22] Moriarty P, Taylor M D R and Brust M 2004 Phys. Rev. Lett. 92 248303
[23] Pauliucci-Vajour E et al 2008 Phys. Rev. Lett. 100 176102
[24] Stannard A 2011 J. Phys.: Condens. Matter 23 083001
[25] Martin C P, Blunt M O and Moriarty P 2004 Nano Lett. 4 2389
[26] Martin C P et al 2007 Phys. Rev. Lett. 99 116103
[27] Pauliucci-Vajour E and Moriarty P 2007 J. Phys. Chem. C 111 16255
[28] Rabani E, Reichmann D R, Geissler P L and Brus L 2003 Nature 426 271
[29] Vazquez I et al 2008 Phys. Rev. E 78 041601
[30] Stannard A et al 2008 J. Chem. Phys. C 112 15195
[31] Oron A, Davis S H and Bankoff S G 1997 Rev. Mod. Phys. 69 931
[32] Frastia L, Archer A J and Thiele U 2011 Phys. Rev. Lett. 106 077801
[33] Yabu H and Shimomura M 2005 Adv. Funct. Mater. 15 575

J. Phys.: Condens. Matter 23 (2011) 415102 M J Robbins
