Bound-Preserving Finite-Volume Schemes for Systems of Continuity Equations with Saturation

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Abstract

We propose finite-volume schemes for general continuity equations which preserve positivity and global bounds that arise from saturation effects in the mobility function. In the case of gradient flows, the schemes dissipate the free energy at the fully discrete level. Moreover, these schemes are generalised to coupled systems of non-linear continuity equations, such as multispecies models in mathematical physics or biology, preserving the bounds and the dissipation of the energy whenever applicable. These results are illustrated through extensive numerical simulations which explore known behaviours in biology and showcase new phenomena not yet described by the literature.

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1 Introduction

Systems of aggregation-diffusion equations are ubiquitous in science and engineering, particularly in mathematical biology, where population dynamics models have a distinctive importance. These systems faithfully reproduce attraction effects, usually modelled by non-local terms; repulsion effects, typically encoded by non-linear diffusion and cross-diffusion; and global constraints, usually due to volume exclusion effects within the populations. The balance and interplay between these terms, which leads interesting phenomena in single population models [10, 9], can produce even richer behaviours in systems of multiple species [23, 14]. For instance, cell-sorting phenomena, triggered by differential adhesion, has been reported in aggregation-reaction-diffusion systems [6, 28]; this biological behaviour is sharply captured by these models, matching experimental data well [14].

Many of these models take the form of a system of gradient flows:

$$\partial_t \rho = \nabla \cdot \left( M(\rho) \psi(\sigma) \nabla \left( \frac{\delta E}{\delta \rho} \right) \right) = \sum_{l=1}^{d} \partial x_l \left( M(\rho) \psi(\sigma) \partial x_l \left( \frac{\delta E}{\delta \rho} \right) \right)$$

for a given energy functional $E$. Here, $\rho$ is a vector $\rho = (\rho_1, \rho_2, \cdots, \rho_P)\top$ of the densities of the $P$ species, and $\rho_p(t, x) : \mathbb{R}^+ \times \Omega \to \mathbb{R}^+$. $M(\rho)$ is a $P \times P$ positive semi-definite matrix, possibly non-symmetric. Additionally, $\psi(\sigma) : \mathbb{R}^+ \to \mathbb{R}$ is a saturation: a decreasing function of the total density of the system, $\sigma(t, x) := \sum_{p=1}^{P} \rho_p(t, x)$, such that $\psi(\alpha) = 0$ at a given saturation level $\alpha$, viz.
Definition 2.1. The energy functional is key to the analysis of this problem, since it provides a formal dissipation estimate:

\[
\frac{d}{dt} E[\rho] = \int_\Omega \frac{\delta E}{\delta \rho} \cdot \partial_t \rho = - \sum_{l=1}^{d} \int_\Omega \psi(\sigma) \partial_{x_l} \left( \frac{\delta E}{\delta \rho} \right) \cdot M(\rho) \partial_{x_l} \left( \frac{\delta E}{\delta \rho} \right) dx \leq 0,
\]

since \( M(\rho) \) is positive semi-definite. The form of the energy functional can lead to a large range of phenomena, including linear/non-linear diffusion and local/non-local drifts [12].

The analysis of cross-diffusion systems which include potential or non-local terms is nevertheless quite challenging. We highlight a series of recent works [19, 20, 30] which deal with a family of cross-diffusion systems without drift terms; they exploit the gradient-flow structure of the systems to define global solutions with global pointwise bounds. The entropy dissipation techniques used there are not only essential to define the solutions, but also to determine their long-time asymptotics. Similar results have only been achieved for systems with drift terms in a handful of specific cases [6, 5].

This work is concerned with the design of numerical schemes that preserve the structural properties of the solutions to (1.1), and thus, that are suitable to explore their rich qualitative behaviours. To that end, two crucial aspects beg consideration. The first one is the saturation effect, due to \( \psi(\sigma) \), which arises in these models due to volume exclusion effects, see for instance [1]; the total density of the system has a natural global bound, which must be preserved by the numerical scheme, along with the non-negativity of the density of each species. The second one is the gradient-flow structure, present in many applications; in these cases, a scheme which captures the dissipation at the fully discrete level (a structure-preserving scheme) is desirable. We emphasise that the saturation terms affect the rate of energy dissipation of the system, and therefore their proper discretisation is required for this goal.

While numerical works for continuity equations are abundant ([8, 24, 26, 2, 15] among others), few deal with the saturation effects as well systems [27, 22]. The work [1] proposes an upwind scheme for scalar Fokker-Planck-like equations in chemotaxis, where the saturation effects only act on the drift term. A series of works by Jüngel and collaborators ([21] and the references therein) propose finite-volume schemes for cross-diffusion systems without drifts; there, a hard volume-filling constraint is imposed by expressing one solution variable in terms of the rest.

In this work we will propose finite-volume schemes for the general formulation (1.1) which preserve both the positivity and the global bounds induced by the saturation effects, and they also guarantee the dissipation of the free energy at the fully discrete level. Based on our previous work [2], we consider implicit schemes where the numerical fluxes are carefully chosen to preserve the lower and upper bounds, a property which later plays a crucial role in preserving the gradient-flow structure. Our numerical methods combine techniques from systems of hyperbolic conservation laws [17, 16], such as upwinding, with a careful discretization of the velocity fields [8, 2] in order to capture the desired dissipative structure.

We will demonstrate the prowess of our methods through a series of numerical examples. We remark that no general theoretical results for equations or systems with saturation are available in the literature, except for a few cases [11, 17, 16], making our numerical explorations very interesting in terms of new phenomena.

Our work is organised in a constructive way. We begin by discussing the case of scalar gradient flows with saturation in Section 2, where we design an implicit numerical scheme, and prove that it preserves the bounds and dissipation structure of the equation unconditionally. The scheme is generalised in Section 3 to the case of systems of gradient flows, again demonstrating the unconditional structure-preserving properties at the fully discrete level. To conclude, Section 4 is devoted to numerical experiments which validate the numerical schemes and showcase their use. These experiments illustrate novel phenomena which arises from the saturation effects, not yet described by the available literature.
2 Scalar Gradient Flows with Saturation

We begin this work by studying a family of scalar general flow equations of the form

\[
\begin{aligned}
\partial_t \rho &= \nabla \cdot \left[ \rho \psi(\rho) \nabla (H'(\rho) + V + W^* \rho) \right], \\
\rho(0, x) &= \rho_0(x),
\end{aligned}
\]

(2.1)

for \( t > 0 \) and \( x \in \Omega \subseteq \mathbb{R}^d \). This equation describes the evolution of a density of particles \( \rho(t, x) : \mathbb{R}^+ \times \Omega \rightarrow \mathbb{R}^+ \), an unknown non-negative function with a prescribed initial datum \( \rho_0(x) : \Omega \rightarrow \mathbb{R}^+ \).

The internal energy density \( H(\rho) \), a convex function, models linear or non-linear diffusion on \( \rho \).

The confining potential \( V(x) \) models external forces acting on the density. The interaction potential \( W(x) \), a symmetric function (typically radial), models the attraction or repulsion between particles.

Equation (2.1), which can be interpreted as a non-linear continuity equation, includes the applications discussed in the introduction which exhibit saturation effects. The term \( \psi(\rho) \) is a saturation of the density \( \rho \) in the following sense:

**Definition 2.1 (Saturation).** A saturation is a continuous function \( \psi(s) : \mathbb{R}^+ \rightarrow \mathbb{R} \) with the properties:

- \( \psi \) is non-increasing (\( s_1 \leq s_2 \) implies \( \psi(s_1) \geq \psi(s_2) \)); and
- there exists \( \alpha > 0 \), called the saturation level, such that \( \psi(\alpha) = 0 \) and \( (\alpha - s) \psi(s) > 0 \) for \( s \neq \alpha \).

Typical examples of saturations are \( \psi(s) = \alpha - s \) and \( \psi(s) = (\alpha - s)|\alpha - s|^{m-1} \), \( m > 1 \), commonly used in population models in mathematical biology [18, 14], as well as the social sciences [7, 4]. The case without saturation, \( \psi(\rho) = 1 \), given by

\[
\begin{aligned}
\partial_t \rho &= \nabla \cdot \left[ \rho \nabla (H'(\rho) + V + W^* \rho) \right], \\
\rho(0, x) &= \rho_0(x),
\end{aligned}
\]

(2.2)

for \( t > 0 \) and \( x \in \Omega \subseteq \mathbb{R}^d \), possesses an interesting structural property: it is the gradient flow of the energy functional

\[
E[\rho] = \int_{\Omega} [H(\rho) + V \rho + \frac{1}{2} (W^* \rho) \rho] \, dx,
\]

(2.3)

see [12, 13]. In the 2-Wasserstein sense, this means that Eq. (2.2) can be written as the non-linear continuity equation

\[
\partial_t \rho + \nabla \cdot (\rho u) = 0, \quad u = -\nabla \xi, \quad \xi = \frac{\delta E}{\delta \rho},
\]

where \( \frac{\delta E}{\delta \rho} \) is the first variation of the energy functional. The variation is given by

\[
\frac{d}{dx} E[\rho + \varepsilon h] \bigg|_{\varepsilon=0} = \int_{\Omega} \frac{\delta E}{\delta \rho} h \, dx,
\]

for any \( h \) such that \( \int_{\Omega} h \, dx = 0 \). This structure leads to the dissipation of the energy along solutions of the equation,

\[
\frac{dE}{dt} = -\int_{\Omega} \rho |\nabla \xi|^2 \, dx \leq 0,
\]

(2.4)

and thus \( E \) is a Lyapunov functional for the problem. Due to the relevance of Eq. (2.2) in applications, numerical solutions which preserve this dissipation property are desirable; indeed [2] introduces schemes to that effect.
Equation (2.1) can no longer be obtained from the usual Wasserstein setting, but can nevertheless be formally cast as a gradient flow by writing
\[ \partial_t \rho + \nabla \cdot (\rho \psi(\rho) u) = 0, \quad u = -\nabla \xi, \quad \xi = \frac{\delta E}{\delta \rho}, \]
yielding the dissipation rate
\[ \frac{dE}{dt} = -\int_{\Omega} \rho \psi(\rho) |\nabla \xi|^2 \, dx \leq 0 \] (2.5)
for the energy (2.3). Under certain conditions on the saturation \( \psi(\rho) \), Equation (2.1) can be understood as a gradient flow with a suitably modified distance between probability measures [11]. As in the previous case, numerical schemes that preserve this dissipation structure are desirable; these will be introduced in Section 2.2.

An interesting property of the gradient-flow structure of these equations is that the energy dissipation rate characterises their steady states. In the Wasserstein case (2.2), the stationary solutions \( \rho_{\infty} \) correspond to \( \nabla \xi_{\infty} \equiv 0 \) on their support. Likewise, in the case with saturation (2.1), the steady states \( \rho_{\infty} \) verify \( \nabla \xi_{\infty} \equiv 0 \) on the support of \( \rho_{\infty} \psi(\rho_{\infty}) \).

### 2.1 Bounds on the Solution

An important aspect in the study of Eq. (2.1) are the bounds on the solution \( \rho \). If the datum satisfies the bounds \( 0 \leq \rho_0(x) \leq \alpha \), we might hope that the solution would satisfy them too, provided sufficient assumptions on \( \psi, H, V \), and \( W \). This, in fact, is crucial to establishing any gradient structure, as the dissipation rate (2.5) hinges on the non-negativity of \( \rho \psi(\rho) \) (just as, in the case without saturation, the dissipation (2.4) relies on the non-negativity of \( \rho \)). Unfortunately, proving such bounds is no trivial matter, and it often requires a case-by-case analysis. Moreover, when we consider the extension of Eq. (2.1) to systems in Section 3, we will find that the existing literature is very sparse.

While a rigorous analysis is beyond the scope of this work, we will invoke the comparison principle for conservation laws of [17, 16] to justify the bounds in certain settings. For Eq. (2.1), the bounds \( 0 \leq \rho(t, x) \leq \alpha \) are satisfied if all of the following conditions hold:

1. Eq. (2.1) is posed on \( \mathbb{R}^d \), or on a domain \( \Omega \) with periodic boundary conditions;
2. the datum satisfies the bounds \( 0 \leq \rho_0(x) \leq \alpha \);
3. \( \rho \psi(\rho) u \) is continuous at \( \rho = 0 \) and \( \rho = \alpha \);
4. \( \nabla \cdot (\rho \psi(\rho) u) \) exists.

The third and fourth conditions depend on the choice of \( \psi, H, V \), and \( W \). The common examples of saturations discussed above, \( \psi(s) = \alpha - s \) and \( \psi(s) = (\alpha - s)|\alpha - s|^{m-1} \), are both compatible. In applications, the diffusion term is typically \( H(\rho) = \rho(\log \rho - 1) \) (the Boltzmann entropy, often associated with the heat equation) or \( H(\rho) = \rho^m / (m - 1) \) for some \( m > 1 \) (associated with the porous-medium equation). These forms present no difficulty at \( \rho = \alpha \), but require attention at \( \rho = 0 \). In the \( H(\rho) = \rho(\log \rho - 1) \) case, it is preferable to rewrite the equation as \( \partial_\rho \rho = \partial_\rho (\psi(\rho) \partial_\rho \rho) \) and to apply parabolic theory in order to obtain lower bounds. In the \( H(\rho) = \rho^m / (m - 1) \) case, the question of regularity at the origin is quite involved, see [29] for an exhaustive survey. This complexity permeates also the analysis of numerical schemes, as explored in [3], which establishes the convergence of the scheme of [2]. Concerning the potentials, it is sufficient to assume \( V, W \in C^2(\Omega) \). This is stringent and could be relaxed in some cases; for instance, if the form of \( H \) leads to establishing some regularity on the solution, it might be possible to interpret \( \partial_\xi (-\partial_\xi (W * \rho)) \) as \( -\partial_\xi W * \partial_\xi \rho \), requiring milder assumptions on \( W \).
2 SCALAR GRADIENT FLOWS WITH SATURATION

2.2 Numerical Schemes

If the bounds have been established on the solutions of Eq. (2.1), we can improve the numerical schemes for gradient flows from [2] to include the problem with saturation by constructing a suitable discretisation of \( \psi \). The crucial question we address here is the choice of the upwinding of the flux which preserves the saturation bounds. Two schemes were given in [2]: schemes S1 and S2. In the interest of brevity, we describe the extension of S2 only, though the S1 can be similarly improved.

We will prescribe a full discretisation in the finite-volume setting in one dimension. The spatial domain \((-L, L)\) is divided into \(2M\) uniform volumes of size \( \Delta x = L/M \). We shall define the points \( x_i = \Delta x(i - 1/2) - L, \ i = 1, \cdots, 2M \) to construct the \( i^{th} \) cell \( C_i = (x_{i-1/2}, x_{i+1/2}) \) with centre \( x_i \).

The time domain \((0, T)\) is discretised through equispaced points with separation \( \Delta t = T/N \), with the \( n^{th} \) point given by \( t^n = n\Delta t, \ n = 0, \cdots, N \).

The solution \( \rho(t, x) \) is to be approximated by a function defined at each time step and constant over each cell; we denote by \( \rho_i^n \) its value at time \( t^n \) over the cell \( C_i \). The new scheme reads:

\[
\frac{\rho_i^{n+1} - \rho_i^n}{\Delta t} + \frac{F_i^{n+1} - F_i^{n+1/2}}{\Delta x} = 0, \quad (2.6a)
\]

\[
F_i^{n+1/2} = \rho_i^{n+1}(\psi_{i+1}^{n+1} + (\psi_{i+1/2}^{n+1})^+ + \rho^{n+1}((\psi_{i+1/2}^{n+1})^+ - (\psi_{i+1/2}^{n+1})^-), \quad (2.6b)
\]

\[
u_i^{n+1} = -\frac{\sum_{j=1}^{k} \psi_s^{n+1} - \psi_s^n}{\Delta x}, \quad \psi_s^{n+1} = H(\rho_i^{n+1}) + V_i + (W \ast \rho^*)_i, \quad (2.6c)
\]

where \((s)^+ := \max\{0, s\}\) and \((s)^- := \min\{0, s\}\). The saturation terms are given by \( \psi_s = \psi(\rho_s) \). The confining potential terms are defined as \( V_i = V(x_i) \) or \( V_i = \frac{1}{\Delta x} \int_{C_i} V(s) \, ds \). The convolution is given by \((W \ast \rho^*)_i = \sum_k W_{i-k} \rho_{k}^* \Delta x \), where \( W_{i-k} = W(x_i - x_k) \) or \( W_{i-k} = \frac{1}{\Delta x} \int_{C_k} W(x_i - s) \, ds \), and \( \rho^* = (\rho_i^{n+1} + \rho^* \mu^2) / 2 \); in practice, the sum will be evaluated through a fast Fourier transform. This scheme is first-order accurate in both time and space.

We will often consider no-flux boundary conditions for the scheme, given by \( F_{1/2} = 0 \) and \( F_{2M+1/2} = 0 \). These are beyond the scope of the theory of Section 2.1, and therefore the bounds on the solution to the continuous problem are not justified; nevertheless, if the no-flux problem approximates the problem on the whole space well, a similar behaviour can be expected.

We begin the analysis of the scheme by proving it preserves the \( 0 \leq \rho \leq \alpha \) bounds unconditionally:

**Proposition 2.2 (Boundness and non-negativity).** For a given \( n \), suppose \( 0 \leq \rho_i^n \leq \alpha \) for all \( i \). Then scheme (2.6) satisfies \( 0 \leq \rho_i^{n+1} \leq \alpha \) bounds unconditionally for all \( i \).

**Proof.** We will prove non-negativity followed by boundness. Seeking a contradiction, we will assume \( \rho_i^{n+1} < 0 \) or \( \rho_i^{n+1} > \alpha \) for some values of \( i \). Without loss of generality we may assume that these “pathological” values lie on contiguous cells; if there are two or more separate pathological clusters, the proof we present below may be applied several times to deal with each one individually. Thus we may simply assume the values in question are \( \rho_{j^{n+1}}, \rho_{j+1}^{n+1}, \cdots, \rho_{k}^{n+1} \). Summing scheme (2.6) over the corresponding cells yields

\[
\sum_{i=j}^{k} (\rho_i^{n+1} - \rho_i^n) \frac{\Delta x}{\Delta x} = -F_{k+1/2}^{n+1} + F_{j-1/2}^{n+1}. \quad (2.7)
\]

To prove non-negativity, we assume \( \rho_i^{n+1} \) is strictly negative for \( j \leq i \leq k \). Since \( \rho_i^n \geq 0 \), the left hand side of Eq. (2.7) is strictly negative also. The right hand side is comprised of

\[-F_{k+1/2}^{n+1} + F_{j-1/2}^{n+1} = -\rho_i^{n+1}(\psi_{i+1}^{n+1} + u_i^{n+1} - \rho_k^{n+1}(\psi_{k}^{n+1})^+ u_{k+1/2} + \rho_{n+1}^j(\psi_{j}^{n+1})^+ u_{j-1/2}.
\]

The first and fourth terms, \(-\rho_i^{n+1}(\psi_{i+1}^{n+1} + u_i^{n+1} - \rho_k^{n+1}(\psi_{k}^{n+1})^+ u_{k+1/2} \) and \( \rho_{n+1}^j(\psi_{j}^{n+1})^+ u_{j-1/2} \), are non-negative, since \( \rho_i^{n+1}, \rho_j^{n+1} < 0 \); the remaining terms are also non-negative. This follows because, outside the range
boundedness one shows \( \rho \) and \( \psi \) because the saturation is non-positive. Furthermore, the first and fourth terms, \( -\rho^{n+1}_k(\psi^{n+1}_k)^+u_{k+1/2}^+ \) and \( \rho^{n+1}_j(\psi^{n+1}_j)^+u_{j-1/2}^- \), are identically zero because the saturation is non-positive. Furthermore, the first and fourth terms, \( -\rho^{n+1}_k(\psi^{n+1}_k)^+u_{k+1/2}^+ \) and \( \rho^{n+1}_j(\psi^{n+1}_j)^+u_{j-1/2}^- \), are non-positive. Thus, the right hand side is also non-positive, yielding a contradiction.

**Remark 2.3** (Strict bounds). It is possible to show a strict inequality version of Proposition 2.2: for a given \( n \), suppose \( 0 < \rho^n_i < \alpha \) for all \( i \); then scheme (2.6) satisfies \( 0 < \rho^n_i < \alpha \) unconditionally for all \( i \). This can be accomplished through an M-matrix argument, as was done for the positivity results in [2]; for positivity, one proves \( A\rho^{n+1} = \rho^n \) where \( A \) is an inverse-positive matrix, whereas for boundedness one shows \( B[\alpha1-\rho^{n+1}] = \alpha1-\rho^n \), where \( B \) is also inverse positive and \( 1 = (1, \ldots, 1)^T \). However, for the latter, one needs the additional assumptions on the saturation; namely, that the quantity \( (\alpha-\rho)/\alpha\psi(\rho) \) is bounded away from zero in the \( \rho \rightarrow \alpha \) limit. This is satisfied by the typical saturations mentioned above, \( \psi(s) = \alpha-s \) and \( \psi(s) = (\alpha-s)(\alpha-s)^{m-1} \) for \( m > 1 \).

Once the proof of bounds has been accomplished, we are ready to show the most important property of the scheme: that it preserves the energy dissipation structure of Equation (2.1) unconditionally. We will define the discrete counterpart of the energy (2.3) as

\[
E_\Delta[\rho^n] = \sum_{i=1}^{2M} H(\rho^n_i) \Delta x + \sum_{i=1}^{2M} V_i \rho^n_i \Delta x + \frac{1}{2} \sum_{i=1}^{2M} \sum_{k=1}^{2M} W_{i-k} \rho^n_i \rho^n_k \Delta x^2, \tag{2.8}
\]

and show, as a straightforward corollary to [2, Theorem 3.9], the dissipation result:

**Theorem 2.4** (Energy dissipation). Scheme (2.6) satisfies a fully discrete version of the energy dissipation property (2.4) unconditionally; namely, the discrete energy (2.8) satisfies

\[
\frac{E_\Delta[\rho^{n+1}] - E_\Delta[\rho^n]}{\Delta t} \leq -\sum_{i=1}^{2M-1} \min\{\rho^n_{i+1}(\psi^{n+1}_{i+1})^+, \rho^n_{i+1}(\psi^{n+1}_i)^+\} |u^{n+1}_{i+1/2}|^2 \Delta x \leq 0.
\]

**Proof.** The proof of [2, Theorem 3.9] remains valid until the last two lines, where the inequality becomes

\[
\frac{E_\Delta[\rho^{n+1}] - E_\Delta[\rho^n]}{\Delta t} \Delta x \leq -\sum_{i=1}^{2M-1} \left( \rho^n_{i+1}(\psi^{n+1}_{i+1})^+ u^{n+1}_{i+1/2}^+ + \rho^n_{i+1}(\psi^{n+1}_i)^+ u^{n+1}_{i+1/2}^- \right) u^{n+1}_{i+1/2},
\]

\[
\leq -\sum_{i=1}^{2M-1} \min\{\rho^n_{i+1}(\psi^{n+1}_{i+1})^+, \rho^n_{i+1}(\psi^{n+1}_i)^+\} |u^{n+1}_{i+1/2}|^2 \leq 0,
\]

by virtue of Proposition 2.2. \( \square \)

**Remark 2.5.** Theorem 2.4 also characterises the steady states of the numerical scheme. By looking at the discrete dissipation rate, we see that the stationary condition is consistent to first-order with the continuous condition stated at the beginning of this section.

**Remark 2.6** (Second Order Schemes). It is also possible to develop structure-preserving schemes for Equation (2.1) with second-order accuracy in space, as was done for Equation (2.2) in [8] (a fully explicit scheme) and [2] (a semi-implicit scheme). Both approaches will preserve the solution bounds,
provided the CFL condition

\[ \Delta t \leq \Gamma \frac{\Delta x}{2 \max|u_{i+1/2}|}, \quad \Gamma = \min \left\{ \frac{1}{\psi(0)^+}, \gamma \right\}, \quad \gamma = \inf_{s \in [0,\infty]} \frac{\alpha - s}{\alpha \psi(s)} \]

is met. The explicit scheme will, however, not preserve the dissipative structure; dissipation can only be proven for a continuous-in-time, discrete-in-space scheme, as done in [8]. The semi-implicit scheme will preserve the energy dissipation property, but an implicit scheme with a CFL condition quickly becomes computationally impractical. The numerical example in Section 4.2 demonstrates the advantages of the fully implicit scheme.

**Remark 2.7 (Higher Dimensions).** Scheme (2.6) may be generalised directly to higher dimensions and, mutatis mutandis, analogues of Proposition 2.2 and Theorem 2.4 can be similarly proven. However, the computational cost of an implicit numerical scheme with non-local terms in multiple dimensions is impractically high. On the other hand, a direct dimensional splitting generalisation will capture the positivity and boundedness properties, but will not preserve the energy dissipation whenever the non-local terms are present.

Yet, as was demonstrated in [2], there is a special form of dimensional splitting (sweeping dimensional splitting) which generalises the one-dimensional scheme to the higher-dimensional setting and which retains the energy dissipation property even in the presence of the convolution term. That approach can be immediately adapted to our scheme, and we will employ it in all the examples of Section 4 to perform the computations efficiently.

### 3 Systems of Gradient Flows with Saturation

We now turn our attention to the generalisation of (2.1); the system of multiple species

\[
\begin{align*}
\frac{\partial \rho}{\partial t} &= \nabla \cdot \left( M(\rho) \psi(\sigma) \nabla \left( \frac{\delta E}{\delta \rho} \right) \right) = \sum_{l=1}^{d} \frac{\partial}{\partial x_{l}} \left( M(\rho) \psi(\sigma) \frac{\delta E}{\delta \rho} \partial_{x_{l}} \right), \\
\rho(0, x) &= \rho(x),
\end{align*}
\]

(3.1)

for \( t > 0 \) and \( x \in \Omega \subseteq \mathbb{R}^d \). This equation describes the evolution of the density \( \rho = (\rho_1, \rho_2, \cdots, \rho_P)^\top \), a vector with entries \( \rho_p(t, x) : \mathbb{R}^+ \times \Omega \to \mathbb{R}^+ \) corresponding to the density of the \( p \)th species. \( M(\rho) \) is a \( P \times P \) positive semi-definite matrix (possibly non-symmetric). Additionally, \( \psi(\sigma) : \mathbb{R}^+ \to \mathbb{R} \) is once again a saturation, viz., Definition 2.1; in this case, the saturation depends on the total density of the system, \( \sigma(t, x) := \sum_{p=1}^{P} \rho_p(t, x) \).

Once an energy functional \( E[\rho] \) is prescribed, the system can be formally cast as a gradient flow by writing, as in the scalar case,

\[ \frac{d}{dt} E[\rho] = - \int_{\Omega} \psi(\sigma) \partial_{x_{l}} \left( M(\rho) \frac{\delta E}{\delta \rho} \right) \cdot \frac{\delta E}{\delta \rho} \, dx \leq 0, \]

since \( M(\rho) \) is positive semi-definite.

Just as in the scalar case, the dissipation rate characterises the steady states of (3.1). The steady states \( \rho_{\infty} \) verify \( M(\rho_{\infty}) \partial_{x_{l}} \xi_{\infty} \equiv 0 \) for every \( l \) on the support of \( \psi(\sigma_{\infty}) \).

For brevity, we will consider here the case of two species, though the general case can be handled identically, and all the results shown in this section generalise immediately. Letting \( \rho = (\rho, \eta)^\top \), and
\( \sigma = \rho + \eta, \) we define the energy functional as

\[
E[\rho] = \mathcal{H}[\rho] + \mathcal{V}[\rho] + \mathcal{W}[\rho],
\]

\[
\mathcal{H}[\rho] = \int_{\Omega} [H_\rho(\rho) + H_\eta(\eta) + H_\sigma(\sigma)] \, dx,
\]

\[
\mathcal{V}[\rho] = \int_{\Omega} [V_\rho(\rho) + V_\eta(\eta)] \, dx,
\]

\[
\mathcal{W}[\rho] = \int_{\Omega} \left[ \frac{1}{2} \rho(W_\rho * \rho) + \frac{1}{2} \eta(W_\eta * \eta) + \rho(W_\sigma * \eta) \right] \, dx.
\]

The internal energy densities \( H_\rho(\rho), H_\eta(\eta), \) and \( H_\sigma(\sigma) \) are convex functions which model diffusion, respectively, on the first species, on the second, and on the total density of the system. The confining potentials \( V_\rho, V_\eta \) and \( W_\sigma \) are bounded away from zero by considering the case without cross-diffusion \((M(\rho) = \text{diag}(\rho))\), or by establishing that the densities are bounded away from zero a priori. The key to the upper bound is to replace \( \sigma \) by \( \alpha - \theta \), where \( \alpha \) is the saturation level and \( \theta \) is a new, independent variable. Letting \( \tilde{\psi}(\theta) = \psi(\alpha - \theta) \), we define a new system

\[
\begin{align*}
\partial_t \rho + \nabla \cdot (\rho \tilde{\psi}(\sigma) v) &= 0, \\
\partial_t \eta + \nabla \cdot (\eta \tilde{\psi}(\sigma) w) &= 0,
\end{align*}
\]

where \( \begin{pmatrix} v \\ w \end{pmatrix} = -\text{diag}(\rho)^{-1}M(\rho)\nabla \left( \frac{\delta E}{\delta \rho} \right) \]

it is important to ensure that these velocities are well-defined, either by making assumptions on \( M \), by considering the case without cross-diffusion \((M(\rho) = \text{diag}(\rho))\), or by establishing that the densities are bounded away from zero a priori. The key to the upper bound is to replace \( \sigma \) by \( \alpha - \theta \), where \( \alpha \) is the saturation level and \( \theta \) is a new, independent variable. Letting \( \tilde{\psi}(\theta) = \psi(\alpha - \theta) \), we define a new system

\[
\begin{align*}
\partial_t \rho + \nabla \cdot [\rho \tilde{\psi}(\theta) v] &= 0, \\
\partial_t \eta + \nabla \cdot [\eta \tilde{\psi}(\theta) w] &= 0, \\
\partial_t \theta - \nabla \cdot [\tilde{\psi}(\theta)(\rho v + \eta w)] &= 0.
\end{align*}
\]

Crucially, any weak solution \((\rho, \eta, \theta)\) with datum \( \rho_0 + \eta_0 + \theta_0 \equiv \alpha \) satisfies \( \rho + \eta + \theta \equiv \alpha \), establishing a one-to-one correspondence between solutions of Eq. (3.4) and solutions of Eq. (3.5) via \( \sigma \equiv \alpha - \theta \).

The new system is in the correct form to apply the comparison principle of [17]. We therefore recover, for any datum \( \rho_0, \eta_0, \theta_0 > 0 \), a viscosity solution \( \rho, \eta, \theta > 0 \), understood here as the \( \varepsilon \to 0 \) limit of the solutions to the family of parabolic systems

\[
\begin{align*}
\partial_t \rho + \nabla \cdot [\rho \tilde{\psi}(\theta) v] &= \varepsilon \Delta \rho, \\
\partial_t \eta + \nabla \cdot [\eta \tilde{\psi}(\theta) w] &= \varepsilon \Delta \eta, \\
\partial_t \theta - \nabla \cdot [\tilde{\psi}(\theta)(\rho v + \eta w)] &= \varepsilon \Delta \theta.
\end{align*}
\]

In particular, solutions to Eq. (3.6) with datum \( \rho_0 + \eta_0 + \theta_0 \equiv \alpha \) correspond to solutions \((\rho, \eta)\) to Eq. (3.4) which satisfy \( 0 < \rho, \eta \) and \( \rho + \eta < \alpha \). Moreover, these are also viscosity solutions of (3.4), which can be verified by establishing a similar correspondence between Eq. (3.6) and the parabolic
As a result, provided the assumptions stated in Section 2.1 are met, the solution to Eq. (3.1) satisfies

\[ 0 \leq \rho, \eta \text{ and } \sigma \leq \alpha. \]

### 3.2 Numerical Schemes

We can now extend scheme (2.6) to the case of systems. The new scheme reads:

\[
\begin{align*}
\rho_i^{n+1} - \rho_i^n + & \frac{F_{n+1}^{i+1/2} - F_{n-1}^{i+1/2}}{\Delta t} = 0, \\
F_{i+1/2}^{n+1} = & \text{diag}(\rho_i^{n+1}) (\psi_i^{n+1})^+ (u_i^{n+1})^+ + \text{diag}(\rho_i^{n+1}) (\psi_i^{n+1})^- (u_i^{n+1})^-,
\end{align*}
\]

(3.7a)

\[
\begin{align*}
u_i^{n+1} = & D_{n+1}^{i+1/2} v_i^{n+1}, \\
\mathbf{g}_i^{n+1} = & \frac{\xi_i^{n+1} - \xi_i^n}{\Delta t},
\end{align*}
\]

(3.7b, 3.7c, 3.7d)

where \( \rho_i^{n+1} = (\rho_1^{n+1} + \rho_2^{n+1})/2 \). The terms \((s)^+\) and \((s)^-\) are as described in Section 2.2, and defined entry-wise when applied to a vector. The saturation terms are given by \( \psi_i^n = \psi(\sigma_i^n) \), where \( \sigma_i^n = \sum_{p=1}^P (\rho_i^n)^p \). The diagonal matrix \( D_{n+1}^{i+1/2} \) is defined as

\[
D_{n+1}^{i+1/2} = \text{diag}(\rho_i^{n+1})^{-1} \text{He}(v_i^{n+1}) + \text{diag}(\rho_i^{n+1})^{-1} \text{He}(-v_i^{n+1}),
\]

(3.7e)

where He is the Heaviside step function, applied to the vector \( v_i^{n+1} \), entry-wise.

In the two species case, the entropy variable \( \xi_{n+1}^{i+1} \) is given by

\[
\xi_i^{n+1} = (\xi_{n+1}^{i+1}, \xi_{n+1}^{n+1})^T,
\]

(3.7f)

\[
\xi_{n+1}^{i+1} = H_\rho(\rho_i^{n+1}) + H_\sigma(\sigma_i^{n+1}) + V_{n+1}, \quad (W_\sigma * \rho^*)_{ij}, \quad (W_\sigma * \eta^*)_{ij},
\]

(3.7g)

\[
\xi_{n+1}^{i+1} = H_\rho(\rho_i^{n+1}) + H_\sigma(\sigma_i^{n+1}) + V_{n+1}, \quad (W_\sigma * \rho^*)_{ij}, \quad (W_\sigma * \eta^*)_{ij},
\]

(3.7h)

The terms of \( V_{n+1}, \), \( (W_\rho * \rho^*)_{ij}, \), \( (W_\rho * \eta^*)_{ij}, \) \( (W_\sigma * \rho^*)_{ij}, \) \( (W_\sigma * \eta^*)_{ij}, \) \( \rho^* \) and \( \eta^* \) are defined analogously to the terms of scheme (2.6), and the spatial discretisation is identical. These entropy terms can be immediately generalised to any number of species, and all the results presented in this section hold.

**Remark 3.1.** In the absence of cross-diffusion effects, i.e. if \( M(\rho) = \text{diag}(\rho) \), the scheme can be simplified to bypass the matrix \( D_{n+1}^{i+1/2} \) altogether:

\[
\begin{align*}
\rho_i^{n+1} - \rho_i^n + & \frac{F_{n+1}^{i+1/2} - F_{n-1}^{i+1/2}}{\Delta t} = 0, \\
F_{i+1/2}^{n+1} = & \text{diag}(\rho_i^{n+1}) (\psi_i^{n+1})^+ (u_i^{n+1})^+ + \text{diag}(\rho_i^{n+1}) (\psi_i^{n+1})^- (u_i^{n+1})^-,
\end{align*}
\]

(3.7a, 3.7b, 3.7c)

\[
\begin{align*}
u_i^{n+1} = & D_{n+1}^{i+1/2} v_i^{n+1}, \\
\mathbf{g}_i^{n+1} = & \frac{\xi_i^{n+1} - \xi_i^n}{\Delta t}.
\end{align*}
\]

Nevertheless, Proposition 3.3 and Theorem 3.5 will be proven for the general version.

**Remark 3.2.** The definition of \( D_{n+1}^{i+1/2} \) can be altered slightly to handle the cases with vacuum:

\[
D_{n+1}^{i+1/2} = \text{diag}(\max(\rho_i^{n+1}, \varepsilon))^{-1} \text{He}(v_i^{n+1}) + \text{diag}(\max(\rho_i^{n+1}, \varepsilon))^{-1} \text{He}(-v_i^{n+1}),
\]

(3.7e)
where the maximum is applied entry-wise, and where $\varepsilon$ is a small parameter. In the simulations of Section 4.1, we take $\varepsilon$ as the machine precision.

We begin the analysis by showing that this new scheme preserves the $0 \leq \rho$ and $\sigma \leq \alpha$ bounds unconditionally:

**Proposition 3.3** (Boundedness and non-negativity). *For a given $n$, suppose $0 \leq \rho_i^n$ (entry-wise) as well as $\sigma_i^n \leq \alpha$, for all $i$. Then scheme (3.7) satisfies $0 \leq \rho_i^{n+1}$ as well as $\sigma_i^{n+1} \leq \alpha$, unconditionally for all $i$.

*Proof.* We will argue by contradiction, proving non-negativity followed by boundedness. As in the proof of Proposition 2.2, we will only deal with one cluster of contiguous pathological values, $j \leq i \leq k$. Summing scheme (3.7) over the corresponding cells yields

$$\sum_{i=j}^{k} (\rho_i^{n+1} - \rho_i^n) \frac{\Delta x}{\Delta t} = -F_{k+1/2}^{n+1} + F_{j-1/2}^{n+1}. \quad (3.8)$$

We prove non-negativity one species at a time. Given $p$, suppose $(\rho_i^{n+1})_p$ is strictly negative for $j \leq i \leq k$, and non-negative otherwise. Whether the entries outside the pathological range satisfy the boundedness property is, for this part, irrelevant, as are the values of the other species. The $p^{th}$ entry of Eq. (3.8) is given by

$$\sum_{i=j}^{k} [(\rho_i^{n+1})_p - (\rho_i^n)_p] \frac{\Delta x}{\Delta t} = -(F_{k+1/2}^{n+1})_p + (F_{j-1/2}^{n+1})_p. \quad (3.9)$$

Since $(\rho_i^n)_p \geq 0$, the left hand side of this equation is strictly negative. The right hand side is comprised of

$$-(F_{k+1/2}^{n+1})_p = -(\rho_{k+1}^{n+1})_p (\psi_{k+1}^{n+1} + (u_{k+1/2})_p \leq \rho_{k+1}^{n+1})_p (\psi_{k+1}^{n+1} + (u_{k+1/2})_p, \quad (3.10)$$

$$F_{j-1/2}^{n+1})_p = (\rho_{j-1}^{n+1})_p (\psi_{j-1}^{n+1} + (u_{j-1/2})_p + (\rho_j^{n+1})_p (\psi_j^{n+1} + (u_{j-1/2})_p. \quad (3.11)$$

Just as in the proof of Proposition 2.2, the first term of (3.10) and the second term of (3.11) are non-negative, since $(\rho_{k+1}^{n+1})_p, (\rho_j^{n+1})_p < 0$; the other terms are also guaranteed to be non-negative. Therefore, the right hand side of Eq. (3.9) is shown to be non-negative, producing a contradiction.

Boundedness is now proven in a similar fashion, by considering instead the sum of Eq. (3.8) over all species:

$$\sum_{i=j}^{k} (\sigma_i^{n+1} - \sigma_i^n) \frac{\Delta x}{\Delta t} = -\sum_{p=1}^{P} (F_{k+1/2}^{n+1})_p + \sum_{p=1}^{P} (F_{j-1/2}^{n+1})_p. \quad (3.12)$$

Here, we assume that $\sigma_i^{n+1}$ is strictly larger than $\alpha$ for $j \leq i \leq k$, resulting on $\psi_i^{n+1} \leq 0$, and rendering the left hand side of (3.12) strictly positive. The right hand side now comprises four terms, corresponding to the sums over $p$ of Eqs. (3.10) and (3.11). Two of the terms, $-\sum_p (\rho_{k+1}^{n+1})_p (\psi_{k+1}^{n+1} + (u_{k+1/2})_p$ and $\sum_p (\rho_j^{n+1})_p (\psi_j^{n+1} + (u_{j-1/2})_p$, are identically zero; the other terms are guaranteed to be non-positive. Thus, the right hand side is also non-positive, yielding a contradiction. \(\square\)

**Remark 3.4** (Strict Bounds). As in Remark 2.3, strict bounds on the solution can be found if the initial datum also satisfies them. This, in particular, ensures that the matrix $D_{i+1/2}$ is well-defined for cross-diffusion systems without vacuum.
Armed with the bounds, we arrive at the crux of the analysis: the unconditional dissipation structure of the scheme. We will define the discrete counterpart of the energy (3.3) as

\[ E_\Delta[\rho^n] = H_\Delta[\rho^n] + V_\Delta[\rho^n] + W_\Delta[\rho^n], \]

(3.13)

\[ H_\Delta[\rho^n] = \sum_{i=1}^{2M} [H_\rho(\rho_i^n) + H_\eta(\eta_i^n) + H_\sigma(\sigma_i^n)] \Delta x, \]

\[ V_\Delta[\rho^n] = \sum_{i=1}^{2M} [V_\rho_\sigma(\rho_i^n) + V_\eta(\eta_i^n)] \Delta x, \]

\[ W_\Delta[\rho^n] = \frac{1}{2} \sum_{i=1}^{2M} 2 M \sum_{k=1}^{\Delta x} [\rho_i^\rho_k^\rho W_{\rho,i-k} + \eta_i^\eta_k^\eta W_{\eta,i-k} + 2 \rho_i^\eta_k^\sigma W_{\sigma,i-k}] \Delta x^2, \]

and show:

**Theorem 3.5 (Energy dissipation).** Scheme (3.7) satisfies a fully discrete version of the energy dissipation property (3.2) unconditionally; namely, the discrete energy (3.13) satisfies

\[ \frac{E_\Delta[\rho^{n+1}] - E_\Delta[\rho^n]}{\Delta t} \leq - \sum_{i=1}^{2M-1} g_{i+1/2}^n \cdot M(\rho_{i+1/2}) \min\{\psi_{i+1}^n, \psi_{i+1}^{n+1}\} g_{i+1/2}^n \Delta x \leq 0. \]

**Proof.** We multiply the scheme by the entropy variable \( \xi_i^{n+1} \) and sum over the spatial domain to find

\[ \sum_{i=1}^{2M} \xi_i^{n+1} \cdot (\rho_{i}^{n+1} - \rho_{i}^{n}) = - \frac{\Delta t}{\Delta x} \sum_{i=1}^{2M} \xi_i^{n+1} \cdot (F_{i+1/2}^{n+1} - F_{i-1/2}^{n+1}), \]

which can be rewritten as

\[ \sum_{i=1}^{2M} V_i \cdot (\rho_{i}^{n+1} - \rho_{i}^{n}) \]

\[ = - \frac{\Delta t}{\Delta x} \sum_{i=1}^{2M} \xi_i^{n+1} \cdot (F_{i+1/2}^{n+1} - F_{i-1/2}^{n+1}) - \sum_{i=1}^{2M} H'_\rho(\rho_i^{n+1})(\rho_i^{n+1} - \rho_i^{n}) \]

\[ - \sum_{i=1}^{2M} H'_\eta(\eta_i^{n+1})(\eta_i^{n+1} - \eta_i^{n}) - \sum_{i=1}^{2M} H'_\sigma(\sigma_i^{n+1})(\sigma_i^{n+1} - \sigma_i^{n}) \]

\[ - \sum_{i=1}^{2M} (\rho_i^{n+1} - \rho_i^{n})(W_{\rho} + \rho^{**})_i - \sum_{i=1}^{2M} (\rho_i^{n+1} - \rho_i^{n})(W_{\sigma} + \sigma^{**})_i \]

\[ - \sum_{i=1}^{2M} (\eta_i^{n+1} - \eta_i^{n})(W_{\eta} + \eta^{**})_i - \sum_{i=1}^{2M} (\eta_i^{n+1} - \eta_i^{n})(W_{\sigma} + \sigma^{**})_i, \]

where \( V_i = (V_{\rho,i}, V_{\eta,i}, V_{\sigma,i})^T \). On the other hand, the evolution of the discrete energy, \( E_\Delta[\rho^{n+1}] - E_\Delta[\rho^n] \), is equal to the sum of \( H_\Delta(\rho^{n+1}) - H_\Delta(\rho^n) \), \( V_\Delta(\rho^{n+1}) - V_\Delta(\rho^n) \), and \( W_\Delta(\rho^{n+1}) - W_\Delta(\rho^n) \). The contribution corresponding to \( H \) is

\[ H_\Delta(\rho^{n+1}) - H_\Delta(\rho^n) \]

\[ = \sum_{i=1}^{2M} [H_\rho(\rho_i^{n+1}) - H_\rho(\rho_i^n) + H_\eta(\eta_i^{n+1}) - H_\eta(\eta_i^n) + H_\sigma(\sigma_i^{n+1}) - H_\sigma(\sigma_i^n)] \Delta x; \]
the contribution of $\mathcal{V}$ is

$$
\mathcal{V}_\Delta(\rho^{n+1}) - \mathcal{V}_\Delta(\rho^n) = \sum_{i=1}^{2M} \left[ V_{\rho,i}(\rho_i^{n+1} - \rho_i^n) + V_{\eta,i}(\eta_i^{n+1} - \eta_i^n) \right] \Delta x
$$

$$
= \sum_{i=1}^{2M} \mathcal{V}'_i \cdot (\rho_i^{n+1} - \rho_i^n);
$$

finally, the contribution of $\mathcal{W}$ is

$$
\mathcal{W}_\Delta(\rho^{n+1}) - \mathcal{W}_\Delta(\rho^n) = \frac{1}{2} \sum_{i=1}^{2M} \sum_{k=1}^{2M} (\rho_i^{n+1} \rho_k^{n+1} - \rho_i^n \rho_k^n) W_{\rho,i-k} \Delta x^2
$$

$$
+ \frac{1}{2} \sum_{i=1}^{2M} \sum_{k=1}^{2M} (\eta_i^{n+1} \eta_k^{n+1} - \eta_i^n \eta_k^n) W_{\eta,i-k} \Delta x^2
$$

$$
+ \frac{1}{2} \sum_{i=1}^{2M} \sum_{k=1}^{2M} (\rho_i^{n+1} \eta_k^{n+1} - \rho_i^n \eta_k^n) W_{\sigma,i-k} \Delta x^2.
$$

The second contribution, that pertaining to $\mathcal{V}$, corresponds to the identity Eq. (3.14). Through its substitution, we may rewrite the evolution of the discrete energy as the sum of three terms: $E_\Delta[\rho^{n+1}] - E_\Delta[\rho^n] = I + II + III$.

The first energy term involves only entropies:

$$
I = \sum_{i=1}^{2M} \left[ H_\rho(\rho_i^{n+1}) - H_\rho(\rho_i^n) - H'_\rho(\rho_i^{n+1})(\rho_i^{n+1} - \rho_i^n) \right] \Delta x
$$

$$
+ \sum_{i=1}^{2M} \left[ H_\eta(\eta_i^{n+1}) - H_\eta(\eta_i^n) - H'_\eta(\eta_i^{n+1})(\eta_i^{n+1} - \eta_i^n) \right] \Delta x
$$

$$
+ \sum_{i=1}^{2M} \left[ H_\sigma(\sigma_i^{n+1}) - H_\sigma(\sigma_i^n) - H'_\sigma(\sigma_i^{n+1})(\sigma_i^{n+1} - \sigma_i^n) \right] \Delta x.
$$

This quantity is immediately controlled, using the convexity of $H_\rho$, $H_\eta$, and $H_\sigma$; for instance, $H_\rho(\rho_i^{n+1}) - H_\rho(\rho_i^n) - H'_\rho(\rho_i^{n+1})(\rho_i^{n+1} - \rho_i^n) \leq 0$. Applying this to every term, we find $I \leq 0$.

The second energy term involves only potentials:

$$
II = \frac{1}{2} \sum_{i=1}^{2M} \sum_{k=1}^{2M} (\rho_i^{n+1} \rho_k^{n+1} - \rho_i^n \rho_k^n - (\rho_i^{n+1} - \rho_i^n)(\rho_k^{n+1} + \rho_k^n)) W_{\rho,i-k} \Delta x^2
$$

$$
+ \frac{1}{2} \sum_{i=1}^{2M} \sum_{k=1}^{2M} (\eta_i^{n+1} \eta_k^{n+1} - \eta_i^n \eta_k^n - (\eta_i^{n+1} - \eta_i^n)(\eta_k^{n+1} + \eta_k^n)) W_{\eta,i-k} \Delta x^2
$$

$$
+ \frac{1}{2} \sum_{i=1}^{2M} \sum_{k=1}^{2M} (\rho_i^{n+1} \eta_k^{n+1} - \rho_i^n \eta_k^n - (\rho_i^{n+1} - \rho_i^n)(\eta_k^{n+1} + \eta_k^n)) W_{\sigma,i-k} \Delta x^2
$$

Exploiting the symmetric character of the potentials, e.g. $W_{\sigma,i-k} = W_{\sigma,k-i}$, this term is shown to be identically zero.

The last energy term involves only fluxes:

$$
III = -\Delta t \sum_{i=1}^{2M} \mathcal{E}_i^{n+1} \cdot (\mathcal{F}_{i+1/2}^{n+1} - \mathcal{F}_{i-1/2}^{n+1}) = \Delta t \sum_{i=1}^{2M-1} (\mathcal{E}_i^{n+1} - \mathcal{E}_i^{n+1} \cdot \mathcal{F}_{i+1/2}^{n+1}).
$$
performing summation by parts and invoking the zero-flux boundary conditions. Applying the definition of the numerical flux, each summand is

\[(\xi_{i+1}^{n+1} - \xi_{i}^{n+1}) \cdot \left( \text{diag}(\rho_{i}^{n+1}) v_{i+1/2}^{n+1} + \text{diag}(\rho_{i+1}^{n+1}) v_{i+1/2}^{n+1} \right).\]

Note that we write \(v_{i+1/2}^{n+1}\) and \(v_{i+1/2}^{n+1}\), rather than \((v_{i+1/2}^{n+1})^+\) and \((v_{i+1/2}^{n+1})^+\), because the bounds of Proposition 3.3 imply that these terms cannot be negative. The summand equals

\[(\xi_{i+1}^{n+1} - \xi_{i}^{n+1}) \cdot (v_{i+1/2}^{n+1} (v_{i+1/2}^{n+1})^+ + \psi_{i}^{n+1} (v_{i+1/2}^{n+1})^-),\]

using the definition of \(v_{i+1/2}^{n+1}\), and the fact that the \(p\)th entry of the vectors \(u_{i+1/2}^{n+1}\) and \(v_{i+1/2}^{n+1}\) have the same sign by construction. Employing, in turn, the definitions of \(v_{i+1/2}^{n+1}\) and \(g_{i+1/2}^{n+1}\), and the positive-semi-definiteness of \(M\), this term is bounded above by

\[-g_{i+1/2}^{n+1} \cdot M(\rho_{i+1/2}^{n+1}) \min\{\psi_{i}^{n+1}, \psi_{i+1}^{n+1}\} g_{i+1/2}^{n+1} \Delta x \leq 0,\]

calling the sum:

\[III \leq -\Delta t \sum_{i=1}^{2M-1} g_{i+1/2}^{n+1} \cdot M(\rho_{i+1/2}^{n+1}) \min\{\psi_{i}^{n+1}, \psi_{i+1}^{n+1}\} g_{i+1/2}^{n+1} \Delta x \leq 0.\]

This finally yields

\[E_{\Delta}[\rho^{n+1}] - E_{\Delta}[\rho^n] = I + II + III \leq 0.\]

\[\square\]

**Remark 3.6** (Higher Dimensions). Scheme (3.7) will be extended to the higher-dimensional setting through the process described in Remark 2.7.

### 4 Numerical Experiments

This section is devoted to showcasing the properties of our numerical schemes in several challenging problems. We will use the implicit scheme (3.7) for all examples, except for Section 4.2, where we compare the performance of both the implicit scheme (2.6) and its explicit version. We shall prioritise numerical experiments involving systems of equations, as these are both more challenging and more interesting.

We first verify the numerical accuracy of our implicit scheme on systems with and without saturation in Section 4.1. Section 4.2 describes the effect of saturation in a gradient flow with an external potential, a case not directly covered by the theory of gradient flows with non-linear mobility [11]. We numerically explore the qualitative properties of the solutions, showing that saturation leads to the convergence towards stationary states with kinks at the free boundary of the fully saturated region. Section 4.3 is devoted to another novel qualitative effect for systems due to saturation, what we call the freeze-in-place phenomenon. Because of the upper bound on the density, certain populations, whose fate was to segregate in the absence of saturation effects, do not achieve complete segregation, as the total population reaches the saturation level, leading to a “frozen” configuration. Finally, Section 4.4 showcases cell-cell adhesion sorting phenomena, based on the different strengths of attraction between two subpopulations. The different attractions between the species lead to segregation, partial engulfment, or complete engulfment, as was reported in [14] for compactly supported attractive kernels.
4 NUMERICAL EXPERIMENTS

Figure 1: SKT model. $L^1$, $L^2$, and $L^\infty$ errors of the numerical solution of scheme (3.7) with respect to the manufactured solution (4.3). Left: regular model (4.2). Right: model with saturation (4.4). $\Delta x = 2^{-k}\pi$ and $\Delta t = 2^{-k}/10$ for $k = 1, \cdots, 8$.

4.1 Cross-Diffusion Experiment

The first experiment aims to verify the numerical accuracy of the scheme (3.7) on a problem with cross diffusion, a feature not present in the scalar tests found in [2]. To that end, we shall employ the SKT population model [25], given by

$$
\begin{align*}
\partial_t \rho &= \nabla \cdot [(2 \rho + \eta) \nabla \rho + \rho \nabla \eta] + s_\rho, \\
\partial_t \eta &= \nabla \cdot [\eta \nabla \rho + (\rho + 2 \eta) \nabla \eta] + s_\eta,
\end{align*}
$$

(4.1)

where $s = (s_\rho, s_\eta)^T$ is a given source term. We will consider the problem on $\Omega = (-\pi, \pi)^2$, posed with periodic boundary conditions. Equation (4.1) can be cast in an analogous form to (3.1) with right-hand sides, written as

$$
\partial_t \left( \begin{array}{c}
\rho \\
\eta
\end{array} \right) = \nabla \cdot \left[ M(\rho, \eta) \nabla \left( \begin{array}{c}
\log \rho \\
\log \eta
\end{array} \right) \right] + \left( \begin{array}{c}
s_\rho \\
s_\eta
\end{array} \right),
$$

(4.2)

where

$$
M(\rho, \eta) = \begin{pmatrix}
\rho(2\rho + \eta) & \rho \eta \\
\rho \eta & (\rho + 2\eta)\eta
\end{pmatrix} = \begin{pmatrix}
\rho & 0 \\
0 & \eta
\end{pmatrix} \begin{pmatrix}
2\rho + \eta & \eta \\
\rho & \rho + 2\eta
\end{pmatrix}.
$$

If $s = 0$, then this is a gradient flow with respect to the energy

$$
E = \int_{\Omega} \rho(\log \rho - 1) + \eta(\log \eta - 1) \, dx.
$$

In order to perform the validation, we will use manufactured solutions. We assume the solution is of the form

$$
\begin{align*}
\rho(t, x, y) &= \frac{1}{4}[1 + \sin(x + t)], \\
\eta(t, x, y) &= \frac{1}{4}[1 + \cos(y + t)],
\end{align*}
$$

(4.3)
4 NUMERICAL EXPERIMENTS

and compute the corresponding source terms,

\[ s_\rho = \frac{1}{4} \cos(t+x) + \frac{3}{16} \sin(t+x) + \frac{1}{16} \cos(t+y) \]
\[ - \frac{1}{8} \cos^2(t+x) + \frac{1}{8} \sin^2(t+x) + \frac{1}{8} \sin(t+x) \cos(t+y), \]

and

\[ s_\eta = \frac{1}{16} \sin(t+x) + \frac{3}{16} \cos(t+y) - \frac{1}{4} \sin(t+y) \]
\[ + \frac{1}{8} \cos^2(t+y) - \frac{1}{8} \sin^2(t+y) + \frac{1}{8} \sin(t+x) \cos(t+y). \]

We will also perform the validation on a version of Equation (4.2) with a saturation term:

\[ \partial_t \begin{pmatrix} \rho \\ \eta \end{pmatrix} = \nabla \cdot \left[ (1 - \rho - \eta)M(\rho, \eta)\nabla \left( \frac{\log \rho}{\log \eta} \right) \right] + \begin{pmatrix} s_\rho \\ s_\eta \end{pmatrix}. \tag{4.4} \]

Assuming the same solution, (4.3), yields the sources

\[ s_\rho = \frac{1}{4} \cos(t+x) + \frac{3}{32} \sin(t+x) + \frac{1}{32} \cos(t+y) - \frac{1}{64} \cos^2(t+x) + \frac{1}{64} \sin^2(t+x) \]
\[ - \frac{1}{32} \sin^3(t+x) - \frac{1}{64} \cos^2(t+y) + \frac{1}{64} \sin^2(t+y) + \frac{1}{16} \cos^2(t+x) \sin(t+x) \]
\[ - \frac{1}{32} \cos^2(t+y) \sin(t+x) + \frac{1}{64} \sin^2(t+y) \sin(t+x) \]
\[ + \frac{3}{64} \cos^2(t+x) \cos(t+y) - \frac{1}{16} \sin^2(t+x) \cos(t+y), \]

and

\[ s_\eta = \frac{1}{32} \sin(t+x) + \frac{3}{32} \cos(t+y) - \frac{1}{4} \sin(t+y) + \frac{1}{64} \cos^2(t+x) - \frac{1}{64} \sin^2(t+x) \]
\[ + \frac{1}{64} \cos^2(t+y) - \frac{1}{32} \cos^3(t+y) - \frac{1}{64} \sin^2(t+y) - \frac{1}{16} \cos^2(t+y) \sin(t+x) \]
\[ + \frac{3}{64} \sin^2(t+y) \sin(t+x) + \frac{1}{64} \cos^2(t+x) \cos(t+y) \]
\[ - \frac{1}{32} \sin^2(t+x) \cos(t+y) + \frac{1}{16} \sin^2(t+y) \cos(t+y). \]

We solve Eqs. (4.2) and (4.4) over the interval \( t \in (0, 0.1) \), choosing \( \Delta x = \Delta y = 2^{-k} \pi \) and \( \Delta t = 2^{-k}/10 \) for \( k = 1, \ldots, 8 \), and compute the \( L^p \) error \( (p = 1, 2, \infty) \) at the final time with respect to the solution (4.3). To handle the points where the solutions touch zero, we proceed as described in Remark 3.2. Figure 1 shows the results of the validation; scheme (3.7) clearly approximates the manufactured solution (4.3) with first-order accuracy, despite the vacuum modification.

4.2 Saturation Experiment

The second experiment concerns the saturation effects. In order to demonstrate the boundedness property of the numerical schemes, we consider the equation

\[ \partial_t \rho = \nabla \cdot \left( \rho (\alpha - \rho) \nabla \left( D \log(\rho) + \frac{C}{2} |x|^2 \right) \right). \tag{4.5} \]

This problem exhibits different steady states which depend on \( m := \|\rho_0\|_{L^1} \), the (conserved) mass of the solution. If \( m \) is small (\( m < m_c \) for some critical mass \( m_c \)), the steady state is simply \( \rho_\infty = A \exp \left( - \frac{C}{2D} |x|^2 \right) \), where \( A \) is a positive constant such that \( \|\rho_\infty\|_{L^1} = m = \|\rho_0\|_{L^1} \). However, if
Figure 2: Stationary solutions to Eq. (4.5) in one dimension. In the cases $M < M_c$ and $M = M_c$, the solution is a half-Gaussian. For $M > M_c$, the solution is made of a straight saturated segment and a Gaussian tail.

Figure 3: Stationary solution of Eq. (4.5) in one dimension with mass $M = 1.66$ corresponding to $l = 1$. $\Omega = (0, 4)$, $\alpha = 1$, $D = 1$, $C = 1$. Explicit: $\Delta t = \Delta x^2/4$. Implicit: $\Delta t = \Delta x$. 
Figure 4: Solution of Eq. (4.5) in one dimension with mass $M = 1.66$ corresponding to $l = 1$. $\Omega = (0, 4)$, $\alpha = 1$, $D = 1$, $C = 1$, $\Delta t = \Delta x = 2^{-7}$.

Figure 5: Stationary solution of Eq. (4.5) in two dimensions with mass $M = 4.71$ corresponding to $l = 2$. $\Omega = (0, 4)^2$, $\alpha = 1$, $D = 1$, $C = 1$, $\Delta t = \Delta x = 2^{-7}$. 
the mass is beyond the critical value, the steady state consists of two segments:

$$\rho_\infty = \begin{cases} 
\alpha & \text{if } |x| \leq l, \\
B \exp\left(-\frac{C}{2D}|x|^2\right) & \text{if } |x| > l, 
\end{cases}$$

where $B$ is again a normalisation constant, and $l$ is to be determined, see Fig. 2 for a diagram. Observing that $\rho_\infty = \alpha$ whenever $|x| = l$, we determine $B = \alpha \exp\left(\frac{C^2}{2D}\right)$; thus the steady state is rewritten as

$$\rho_\infty(x) = \alpha \exp\left(-\frac{C}{2D}(|x|^2 - l^2)^+\right), \quad (4.6)$$

where $(s)^+ = \max\{s, 0\}$ for any $s \in \mathbb{R}$. The value of $l$ is an increasing function of $m$, and can be determined form the initial datum. It is left to the reader to check (4.6) are indeed weak stationary solutions to (4.5).

We may further explore the normalization condition. We first consider Eq. (4.5) in one dimension, posed on $\Omega = (0, \infty)$; by virtue of the conservation of mass, we arrive at the self-consistency equation

$$m = \int_0^\infty \alpha \exp\left(-\frac{C}{2D}(x^2 - l^2)^+\right) dx = \alpha \left[l + \int_l^\infty \exp\left(\frac{C}{2D}(l^2 - x^2)\right) dx\right] = \alpha \left[l + \sqrt{\frac{\pi D}{2C}} \exp\left(\frac{C}{2D}l^2\right) \left[1 - \text{erf}\left(l\sqrt{\frac{C}{2D}}\right)\right]\right].$$

In particular, this expression yields the value of the critical mass, $m_c = \alpha \sqrt{\frac{\pi D}{2C}}$, which corresponds to $l = 0$. In the two dimensional case, we pose Eq. (4.5) on the domain $\Omega = (0, \infty)^2$. We find

$$m = \int_0^\infty \int_0^\infty \alpha \exp\left(-\frac{C}{2D}(x^2 + y^2 - l^2)^+\right) dx dy = \int_0^{\pi/2} \int_0^\infty \alpha r \exp\left(-\frac{C}{2D}(r^2 - l^2)^+\right) dr d\theta = \frac{\alpha \pi}{2} \left[\int_0^l r dr + \int_l^\infty r \exp\left(-\frac{C}{2D}(r^2 - l^2)\right) dr\right] = \frac{\alpha \pi}{2} \left[\frac{C}{2} - \frac{D}{C}\right],$$

yielding $m_c = \frac{\alpha \pi D}{2C}$.

We will solve Eq. (4.5) in one and two dimensions, over sufficiently large but finite domains, with supercritical masses, in order to validate the behaviour of the schemes in the presence of saturation. The problems are initialised with a constant density $\rho = m/|\Omega|$, where $|\Omega|$ is the Lebesgue measure of $\Omega$; the solution is computed for $t \in (0, 15)$, and the final state is taken as an approximation of the steady state. We will employ Scheme (3.7), as well as an explicit counterpart found by evaluating the numerical fluxes at the $n$th time.

In one dimension, we let $\Omega = (0, 4)$, prescribe no-flux boundary conditions, and choose the supercritical mass $m = 1.66$, which corresponds to $l = 1$. Figure 3 shows the numerical steady state, computed with both the explicit and implicit schemes, each on a coarse ($\Delta x = 2^{-3}$) and a fine ($\Delta x = 2^{-7}$) mesh. Both schemes capture the upper bound of the solution, but the implicit scheme permits the use of a much larger time step. Figure 4 shows the evolution in time of the solution with the implicit scheme, as well as the evolution of the variation $\xi$. As expected, as the solution approaches the steady state, $\xi$ tends to a constant on the support of $\rho_\infty \psi'(\rho_\infty)$.

In two dimensions, we set $\Omega = (0, 4)^2$, again with no-flux boundary conditions, and choose the supercritical mass $m = 4.71$, found by letting $l = 2$. Figure 5 shows the numerical steady state computed with the implicit scheme.
We would like to emphasise that there is no general weak solution theory yet that establishes the kink solutions (4.6) as the global asymptotic profile for generic initial data to (4.5) with supercritical mass, though this behaviour was observed numerically for diverse initial data. The Cauchy theory for these general gradient flows (4.5) with non-linear mobility and external interaction potentials, not included in [11], is challenging and will be studied elsewhere.

4.3 Freeze-In-Place Experiment

This test will explore the appearance of infinitely many steady states as a result of the saturation effects. To that end, we will consider the two species system

$$\begin{align*}
\partial_t \rho &= \nabla \cdot \left[ \rho (\alpha - \sigma) \nabla (D\sigma + \frac{C_1}{2} |x|^2) \right], \\
\partial_t \eta &= \nabla \cdot \left[ \eta (\alpha - \sigma) \nabla (D\sigma + \frac{C_2}{2} |x|^2) \right].
\end{align*}$$

(4.7)

There are two effects present here: diffusion and confinement. The non-linear diffusion acts on the sum, $\sigma = \rho + \eta$. This can be seen by ignoring the $|x|^2$ terms; then, the two equations sum to

$$\partial_t \sigma = D \nabla \cdot (\sigma (\alpha - \sigma) \nabla \sigma).$$

Hence, the diffusion terms act to smooth irregularities on $\sigma$, but not necessarily on the individual profiles. Meanwhile, the quadratic potentials serve to confine each species, driving their densities towards the origin.

Crucially, we will choose $C_1 = 2C_2$; the confinement effects will therefore act more strongly on $\rho$ than on $\eta$. This difference in strengths, combined with the diffusion, will cause $\rho$ to displace $\eta$ near the origin. Below the saturation level, the species are expected to segregate completely; however, this might not occur if the species sum reaches the saturation level. As $\sigma$ approaches $\alpha$, the mobility reduces drastically, which will cause the solution to “freeze” in whatever configuration it may find itself.

We will let $D = 0.1$, $C_1 = 4$, $C_2 = 2$, and $\alpha = 1$. In one dimension, we solve the problem on $\Omega = (-1,1)$, with the initial datum

$$\begin{align*}
\rho_1(0,x) &= f(x) \left(1 - \cos(\omega x)\right)^+ + f(x) \left(1 + \cos(\omega x)\right)^+,
\end{align*}$$

where $f(x) = \frac{4}{5} \left(1 - \left(\frac{4x}{3}\right)^2\right)$, (4.8)

and $\omega = 16\pi$. We will solve the system for $t \in (0,30)$ and take the final state as an approximation to the asymptotic steady state which corresponds to this datum. Figure 6 shows the datum, the “frozen-in-place” steady state, as well as the segregated steady state of the analogous system without saturation, all computed with the implicit scheme (3.7).

In two dimensions, we solve the problem on $\Omega = (-1,1)^2$. The datum now reads

$$\begin{align*}
\rho_1(0,x,y) &= f\left(\sqrt{x^2 + y^2}\right) \left(1 - \cos(\omega x) \cos(\omega y)\right)^+ + f\left(\sqrt{x^2 + y^2}\right) \left(1 + \cos(\omega x) \cos(\omega y)\right)^+,
\end{align*}$$

where $\omega = 4\pi$ and $f$ is defined as above. Once again, we take the solution at $t = 30$ as an approximation to the asymptotic steady state. Figure 7 shows the datum as well as the steady states with and without saturation. As in the one-dimensional case, the saturation induces a “freezing-in-place” effect, preventing the complete segregation of the species.

4.4 Cell-Cell Adhesion Experiment

To conclude the experiments section, we explore a version of the two-species model for cell-cell adhesion introduced in [14]. The model for the evolution of the densities of two different species,
Figure 6: Stationary solutions of Eq. (4.7) in one dimension. $\Omega = (-1, 1)$, $\alpha = 1$, $D = 0.1$, $C_1 = 4$, $C_2 = 2$, $\Delta x = 2^{-n}$, and $\Delta t = 0.1$. **Top:** datum (4.8). **Middle:** “frozen-in-place” stationary state with saturation effects. **Bottom:** segregated stationary state without saturation.
Figure 7: Stationary solutions of Eq. (4.7) in two dimensions. \( \Omega = (-1, 1)^2 \), \( \alpha = 1 \), \( D = 0.1 \), \( C_1 = 4 \), \( C_2 = 2 \), \( \Delta x = 2^{-7} \), and \( \Delta t = 0.1 \). **Top:** datum (4.8). **Middle:** “frozen-in-place” stationary state with saturation effects. **Bottom:** segregated stationary state without saturation.
Figure 8: Steady state of Eqs. (4.9) and (4.10) in two dimensions. $\Omega = (-2, 2)^2$, $\alpha = 1$, $\varepsilon = 0.1$, $c_{\rho\rho} = 2c_{\rho\eta} = 2c_{\eta\eta} = 1$, $\Delta x = \Delta y = \Delta t = 2^{-6}$. **Left**: Eq. (4.10) (saturation). **Right**: Eq. (4.9) (no saturation).
Figure 9: Steady state of Eqs. (4.9) and (4.10) in two dimensions. \( \Omega = (-2, 2)^2 \), \( \alpha = 1 \), \( \varepsilon = 0.1 \), \( 4c_{\rho \rho} = 2c_{\rho \eta} = 2c_{\eta \rho} = c_{\eta \eta} = 1 \), \( \Delta x = \Delta y = \Delta t = 2^{-6} \). **Left:** Eq. (4.10) (saturation). **Right:** Eq. (4.9) (no saturation).
\[ \begin{align*}
\frac{\partial}{\partial t} \rho & = \nabla \cdot \left[ \rho \nabla (\varepsilon \sigma + c_{\rho \rho} K * \rho + c_{\rho \eta} K * \eta) \right], \\
\frac{\partial}{\partial t} \eta & = \nabla \cdot \left[ \eta \nabla (\varepsilon \sigma + c_{\eta \rho} K * \rho + c_{\eta \eta} K * \eta) \right],
\end{align*} \]  
(4.9)

in two spatial dimensions, for a positive value of \( \varepsilon \), non-negative values of \( c_{\rho \rho} \), \( c_{\rho \eta} \), \( c_{\eta \rho} \), and \( c_{\eta \eta} \), and where \( K(x) = |x|^2/2 \). The parameter \( \varepsilon \) regulates the strength of non-linear diffusion on the total density of the system, which in this model accounts for localised repulsion effects. Each of the coefficients \( c_{\alpha \beta} \) correspond to the strength of attraction of the species \( \alpha \) to species \( \beta \).

This model exhibits different asymptotic behaviours, regulated by the values of \( c_{\alpha \beta} \). As discussed in [14], different values of the parameters may lead to the mixing of the two species, a complete segregation, or the engulfment of one of the species (either partial or complete). We will concentrate here on the latter case, and will observe how the inclusion of saturation effects affect these behaviours, through the modified model

\[ \begin{align*}
\frac{\partial}{\partial t} \rho & = \nabla \cdot \left[ \rho(\alpha - \sigma) \nabla (\varepsilon \sigma + c_{\rho \rho} K * \rho + c_{\rho \eta} K * \eta) \right], \\
\frac{\partial}{\partial t} \eta & = \nabla \cdot \left[ \eta(\alpha - \sigma) \nabla (\varepsilon \sigma + c_{\eta \rho} K * \rho + c_{\eta \eta} K * \eta) \right].
\end{align*} \]  
(4.10)

We will prescribe an initial datum consisting of two touching disks:

\[ \begin{align*}
\rho_0 & = 0.95 \mathbb{1}_{D_{\rho}}, \\
\eta_0 & = 0.95 \mathbb{1}_{D_{\eta}},
\end{align*} \]

where \( D_{\rho} \) is a disk centred at \((-0.5, 0)\) with radius 0.5, \( D_{\eta} \) is a disk centred at \((0.4, 0)\) with radius 0.4, and \( \mathbb{1} \) indicates the characteristic function. We will let \( \varepsilon = 0.1 \), and \( \alpha = 1 \) when applicable. We shall solve Eqs. (4.9) and (4.10) on \( \Omega = (-2, 2)^2 \), over the interval \( t \in (0, 45) \), and take their final state as an approximation of the asymptotic steady state. We choose \( \Delta x = \Delta y = \Delta t = 2^{-6} \) to ensure high accuracy.

We first explore the partial engulfment case, given by \( c_{\rho \rho} = 2c_{\rho \eta} = 2c_{\eta \rho} = c_{\eta \eta} = 1 \), where the two species are expected to partially surround each other in a lunar shape. Figure 8 shows the numerical steady states. In the system with saturation, the steady state exhibits a saturated, flat region. Nevertheless, both solutions are relatively similar.

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**Figure 10:** Dissipation of the free energy in the solutions to Eqs. (4.9) and (4.10) shown in Figs. 8 and 9. **Left:** partial engulfment, corresponding to Fig. 8. **Right:** complete engulfment, corresponding to Fig. 9.
We turn to the complete engulfment case, given by \(4c_{\rho \rho} = 2c_{\rho \eta} = 2c_{\eta \rho} = c_{\eta \eta} = 1\). Here, one species is expected to completely surround the other; in the long-time limit, one species will be supported on a disk, and the other, on an annulus. Figure 9 shows the numerical steady states. This time, the effects of the saturation are very visible: the initial concentration of the density creates a region of saturation where the two species remain mixed for all time, in the same vein as the example from Section 4.3. This prevents the full segregation seen in the unsaturated case, and yields a cardioid-like free boundary between the species.

We conclude by showing the evolution of the (relative) free energy corresponding to the both the partial engulfment and the complete engulfment scenarios in Fig. 10. The energy is clearly dissipated by the numerical scheme, and the dissipation rates are affected by the presence of the saturation terms. In the partial engulfment case, the energies with and without saturation evolve very differently, but eventually reach a comparable value; this is consistent with the similar (though not equal) steady states reached in either setting. Meanwhile, the energies of the complete engulfment case demonstrate similar evolutions, but the dissipation rate is higher in the unsaturated setting.

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