

POSITIVE AND FREE ENERGY SATISFYING SCHEMES FOR DIFFUSION WITH INTERACTION POTENTIALS

HAILIANG LIU AND WUMAIER MAIMAITIYIMING

ABSTRACT. In this paper, we design and analyze second order positive and free energy satisfying schemes for solving diffusion equations with interaction potentials. The semi-discrete scheme is shown to conserve mass, preserve solution positivity, and satisfy a discrete free energy dissipation law for nonuniform meshes. These properties for the fully-discrete scheme (first order in time) remain preserved without a strict restriction on time steps. For the fully second order (in both time and space) scheme, we use a local scaling limiter to restore solution positivity when necessary. It is proved that such limiter does not destroy the second order accuracy. In addition, these schemes are easy to implement, and efficient in simulations over long time. Both one and two dimensional numerical examples are presented to demonstrate the performance of these schemes.

1. INTRODUCTION

This paper is concerned with efficient numerical approximations to the following problem,

$$\begin{cases} \partial_t \rho = \nabla \cdot (\nabla \rho + \rho \nabla (V(\mathbf{x}) + W * \rho)), & \mathbf{x} \in \Omega \subset \mathbb{R}^d, \quad t > 0, \\ \rho(\mathbf{x}, 0) = \rho_0(\mathbf{x}), & \mathbf{x} \in \Omega \subset \mathbb{R}^d, \end{cases} \quad (1.1)$$

subject to zero flux boundary conditions. Here Ω is a bounded domain in $\mathbb{R}^d$, $\rho = \rho(\mathbf{x}, t)$ is the unknown density, $V(\mathbf{x})$ is a confinement potential, and $W(\mathbf{x})$ is an interaction potential, which is assumed to be symmetric.

Such problems appear in many applications. If W vanishes, this model includes heat equation ($V(\mathbf{x}) = 0$) and the Fokker–Planck equation ($V(\mathbf{x}) \neq 0$, see e.g. [41]). With interaction potentials, the equation can model nematic phase transition of rigid rod-like polymers [14], chemotaxis [39], and aggregation in biology (see [17, 21, 43] and references therein). For chemotaxis, a wide literature exists in relation to the Patlak-Keller-Segel system [22, 38], and for rod-like polymers, the Doi-Onsager equation [13, 14, 30, 34] is a well studied model.

Main properties of the solution to (1.1) are non-negativity, mass conservation and free energy dissipation, i.e.,

$$\rho_0(\mathbf{x}) \geq 0 \implies \rho(\mathbf{x}, t) \geq 0, \quad t > 0, \quad (1.2)$$

$$\int_{\Omega} \rho(\mathbf{x}, t) d\mathbf{x} = \int_{\Omega} \rho_0(\mathbf{x}) d\mathbf{x}, \quad t > 0, \quad (1.3)$$

$$\frac{dE(\rho)}{dt} = - \int_{\Omega} \rho |\nabla (\log(\rho) + V(x) + W * \rho)|^2 d\mathbf{x} = -I(\rho) \leq 0, \quad (1.4)$$

1991 *Mathematics Subject Classification.* 35K20, 35R09, 65M08, 82C31.

Key words and phrases. Drift-diffusion equations, Implicit-explicit scheme, Energy dissipation, Positivity preserving.

where the free energy associated to (1.1) is given by

$$E(\rho) = \int_{\Omega} \rho \log(\rho) d\mathbf{x} + \int_{\Omega} V(\mathbf{x}) \rho d\mathbf{x} + \frac{1}{2} \int_{\Omega} \int_{\Omega} W(\mathbf{x} - \mathbf{y}) \rho(\mathbf{y}) \rho(\mathbf{x}) d\mathbf{y} d\mathbf{x}. \quad (1.5)$$

This energy functional is a sum of internal energy, potential energy, and the interaction energy. The functional I is referred to as the entropy dissipation. The nice mathematical features (1.2)-(1.4) are crucial for the analytical study of (1.1), while free-energy dissipation inequality (1.4) is particularly important to understand the large time dynamics of solutions of (1.1) (see e.g., [6, 7, 32]). There have been many studies about the connection between the free energy, the Fokker-Planck equation, and optimal transportation in a continuous state space (see e.g., [3, 16, 20, 35, 44]).

One way of obtaining a structure-preserving numerical scheme is the minimizing movement approximation (see [1] and the references therein), also named Jordan-Kinderlehrer-Otto (JKO) scheme (Jordan et al. [20]), which is given by

$$\rho^{n+1} = \operatorname{argmin} \left\{ \frac{1}{2\tau} W^2(\rho^n, \rho) + E(\rho) \right\}$$

Here, at each time step, the distance of the solution update acts as a regularization to the free energy. Yet such problems involving the Wasserstein distance $W(\rho^n, \rho)$ are computationally demanding, see, e.g., [5, 10, 15, 31] for some recent advances.

The second way of obtaining a structure-preserving numerical scheme is by a direct discretization of (1.1) so that these solution properties are preserved at the discrete level. This way has gained increasing attention in recent years, some closely related works include [8, 25–29, 42]. In [25], second order implicit numerical schemes designed for linear (yet singular) Fokker-Planck equations satisfy all three solution properties without any time step restriction. In [28], the authors extended the idea in [25] to a system of Poisson-Nernst-Planck equations using the explicit time discretization. For a more general class of nonlinear nonlocal equations,

$$\partial_t \rho = \nabla \cdot (\rho \nabla (H'(\rho) + V(\mathbf{x}) + W * \rho)), \quad (1.6)$$

where H is a smooth convex function, a second order finite-volume method was constructed in [8], where positivity is enforced by using piecewise linear polynomials interpolating interface values. Structure preserving schemes based on the Chang-Cooper scheme [9] have been constructed in [37] to numerically solve nonlinear Fokker-Planck equations. Note that in [8, 28, 37] different time step restrictions are imposed in order to preserve the desired solution properties.

The construction of higher order schemes using the discontinuous Galerkin (DG) framework has recently been carried out for Fokker-Planck-type equations. We refer to [27] for entropy satisfying DG schemes of arbitrary high order, and to [26] for a DG scheme of third order to satisfy the discrete maximum principle for linear Fokker-Planck equations. In [29], the authors designed free energy satisfying DG schemes of any high order for Poisson-Nernst-Planck equations, but positive cell averages are shown to propagate in time only for special cases. While in [42], a high order nodal DG method for (1.6) was constructed using $k + 1$ Gauss-Lobatto quadrature points for degree k polynomials in order to preserve both the entropy dissipation and the solution positivity; somehow degeneracy of accuracy in some cases was reported. Despite some well-known advantages

of the DG method, structural properties of the above fully discrete DG schemes are verified under some CFL conditions. It would be interesting to explore some explicit-implicit strategies for DG schemes.

In this paper we extend the idea in [25] to construct explicit-implicit schemes which are proven to preserve three main properties of (1.1) without a strict restriction on time steps. This therefore has improved upon the work [28]. Our main results include the scheme formulation, proofs of mass conservation, solution non-negativity, and the discrete free-energy dissipation law for both semi-discrete and fully discrete methods. In particular, the fully-discrete scheme (first order in time) is shown to satisfy three desired properties without strict restriction on time steps, in both one and two dimensional cases with nonuniform meshes. For the fully second order (in both time and space) scheme, we apply a local scaling limiter to restore solution positivity, such limiter was first introduced in [24], in this paper we rigorously prove that such limiter does not destroy the second order accuracy.

More precisely, our scheme construction is based on a reformulation

$$\partial_t \rho = \nabla \cdot \left(M \nabla \left(\frac{\rho}{M} \right) \right), \quad (1.7)$$

where $M = e^{-V(\mathbf{x})-W*\rho}$, motivated by the fact that the equilibrium solutions of (1.1) may be expressed as $\rho = C e^{-V(\mathbf{x})-W*\rho}$. For linear Fokker-Planck equations, such reformulation with $M = e^{-V(\mathbf{x})}$ (so called non-logarithmic Landau form) has been used in [25], as well as in earlier works (see e.g., [4]). We note that for the general nonlinear nonlocal model (1.6), our scheme construction remains valid if we take $M = \rho e^{-H'(\rho)-V(\mathbf{x})-W*\rho}$ in the reformulation (1.7).

The advantage of formulation (1.7) can be seen from both spatial and temporal discretization. The symmetric spatial discretization of the one-dimensional version of (1.7) yields the semi-discrete scheme

$$h_j \frac{d}{dt} \rho_j = h_{j+1/2}^{-1} M_{j+1/2} \left(\frac{\rho_{j+1}}{M_{j+1}} - \frac{\rho_j}{M_j} \right) - h_{j-1/2}^{-1} M_{j-1/2} \left(\frac{\rho_j}{M_j} - \frac{\rho_{j-1}}{M_{j-1}} \right), \quad (1.8)$$

in which the evaluation of M at cell interfaces $\{x_{j+1/2}\}$ and cell centers $\{x_j\}$ is easily available as defined in (2.4). Here ρ_j approximates the cell average of $\rho(x, t)$ on j -th computational cell $[x_{j-1/2}, x_{j+1/2}]$ of size h_j , and $h_{j+1/2} = (h_j + h_{j+1})/2$.

For time discretization of (1.8), we adopt an implicit-explicit approach to obtain

$$h_j \frac{\rho_j^{n+1} - \rho_j^n}{\tau} = h_{j+1/2}^{-1} M_{j+1/2}^n \left(\frac{\rho_{j+1}^{n+1}}{M_{j+1}^n} - \frac{\rho_j^{n+1}}{M_j^n} \right) - h_{j-1/2}^{-1} M_{j-1/2}^n \left(\frac{\rho_j^{n+1}}{M_j^n} - \frac{\rho_{j-1}^{n+1}}{M_{j-1}^n} \right), \quad (1.9)$$

where ρ_j^n approximates $\rho_j(t)$ at time $t = n\tau$, see (3.1). This scheme is easy to implement, and is shown to preserve all three desired properties without a strict time step restriction. However, the scheme (1.9) is only first order in time. We further propose a fully second order scheme:

$$\begin{aligned} h_j \frac{\rho_j^* - \rho_j^n}{\tau/2} &= h_{j+1/2}^{-1} M_{j+1/2}^* \left(\frac{\rho_{j+1}^*}{M_{j+1}^*} - \frac{\rho_j^*}{M_j^*} \right) - h_{j-1/2}^{-1} M_{j-1/2}^* \left(\frac{\rho_j^*}{M_j^*} - \frac{\rho_{j-1}^*}{M_{j-1}^*} \right), \\ \rho_j^{n+1} &= 2\rho_j^* - \rho_j^n, \end{aligned} \quad (1.10)$$

based on the predictor-corrector methodology, where M_j^* and $M_{j+1/2}^*$ are given in (5.1). This scheme is second order in both time and space, and it preserves solution positivity for small time steps. For large time steps, we use a local scaling limiter to restore the solution positivity.

Although we derive the schemes for the model equation (1.1), the methods can be easily applied to a larger class of problems where the solution depends on additional parameters and the PDE is of drift-diffusion type; see [23].

The rest of the paper is organized as follows. In section 2, we present a semi-discrete scheme for one dimensional problems. Theoretical analysis of three properties is provided. In section 3, we present fully discrete implicit-explicit schemes for one dimensional case and prove the desired properties. Section 4 is devoted to numerical schemes for two dimensional problems. In section 5, we extend the scheme to a fully second order (in both time and space) scheme, a mass conserving local limiter is also introduced to restore solution positivity. Numerical examples for one and two dimensional problems are presented in section 6. Finally, concluding remarks are given in section 7.

2. NUMERICAL METHOD: ONE DIMENSIONAL CASE

We begin with

$$\begin{cases} \partial_t \rho = \partial_x(\partial_x \rho + \rho \partial_x(V(x) + W * \rho)), & x \in \Omega, \quad t > 0, \\ \rho(x, 0) = \rho_0(x), & x \in \Omega, \\ \partial_x \rho + \rho \partial_x(V(x) + W * \rho) = 0, & x \in \partial\Omega, \quad t > 0. \end{cases} \quad (2.1)$$

and reformulate (2.1) as

$$\begin{cases} \partial_t \rho = \partial_x(M \partial_x(\rho/M)), & x \in \Omega, \quad t > 0, \\ \rho(x, 0) = \rho_0(x), & x \in \Omega, \\ M \partial_x(\rho/M) = 0, & x \in \partial\Omega, \quad t > 0, \end{cases} \quad (2.2)$$

where $M = e^{-V(x) - W * \rho}$. We propose a finite volume scheme for (2.2) over the interval $\Omega = [a, b]$. For a given positive integer N , we partition domain Ω into computational cells $I_j = [x_{j-\frac{1}{2}}, x_{j+\frac{1}{2}}]$ with mesh size $h_j = |I_j|$ and cell center at $x_j = x_{j-\frac{1}{2}} + \frac{1}{2}h_j$, $j \in \{1, 2, \dots, N\}$, we set $h_{j+1/2} = (h_j + h_{j+1})/2$.

2.1. Semi-discrete scheme. We integrate on each computational cell I_j to obtain

$$\frac{d}{dt} \int_{I_j} \rho(x, t) dx = M \partial_x(\rho/M)|_{x_{j+1/2}} - M \partial_x(\rho/M)|_{x_{j-1/2}}.$$

Let $\rho(t) = \{\rho_1, \dots, \rho_N\}$ be the numerical solution approximating all cell averages and $C_{j+1/2}$ be an approximation to $M \partial_x(\rho/M)|_{x_{j+1/2}}$, then one has the following semi-discrete scheme,

$$\frac{d}{dt} \rho_j = \frac{C_{j+1/2} - C_{j-1/2}}{h_j}, \quad j = 1, 2, \dots, N, \quad (2.3)$$

we define

$$C_{j+1/2} = \frac{M_{j+\frac{1}{2}}}{h_{j+1/2}} \left(\frac{\rho_{j+1}}{M_{j+1}} - \frac{\rho_j}{M_j} \right) \quad \text{for } j = 1, 2, \dots, N-1,$$

$$C_{1/2} = 0, \quad C_{N+1/2} = 0.$$

Here $M_{j+1/2} = Q_1(x_{j+1/2}, \rho)$ and $M_j = Q_1(x_j, \rho)$ with

$$Q_1(x, v) = e^{-V(x) - \sum_{i=1}^N h_i W(x_i - x) v_i}, \quad \text{for } x \in \mathbb{R}, \quad v \in \mathbb{R}^N. \quad (2.4)$$

Note that the zero flux boundary conditions have been weakly enforced.

2.2. Scheme properties. We investigate three desired properties for this semi-discrete scheme. For the energy dissipation property, we define a semi-discrete version of the free energy (1.5) as

$$E_h(t) = \sum_{j=1}^N h_j \left(\rho_j \log(\rho_j) + V_j \rho_j + \frac{1}{2} g_j \rho_j \right), \quad (2.5)$$

where $g_j = \sum_{i=1}^N h_i W(x_i - x_j) \rho_i$ is a second order approximation of the convolution $(W * \rho)(x_j)$.

The following theorem states that the semi-discrete scheme (2.3) is conservative, positive, and energy dissipating.

Theorem 2.1. The semi-discrete scheme (2.3) satisfies the following properties:

(1) Conservation of mass: for any $t > 0$ we have

$$\sum_{j=1}^N h_j \rho_j(t) = \sum_{j=1}^N h_j \rho_j(0). \quad (2.6)$$

(2) Positivity preserving: if $\rho_j(0) \geq 0$ for all $j \in \{1, \dots, N\}$, then $\rho_j(t) \geq 0$ for any $t > 0$.

(3) Entropy dissipation: $\frac{dE_h(t)}{dt} \leq -I_h$, where

$$I_h = \sum_{j=1}^{N-1} C_{j+1/2} \left(\log \left(\frac{\rho_{j+1}}{M_{j+1}} \right) - \log \left(\frac{\rho_j}{M_j} \right) \right) \geq 0. \quad (2.7)$$

Proof. (1) Summing all equations in (2.3), we have

$$\frac{d}{dt} \sum_{j=1}^N h_j \rho_j(t) = \sum_{j=1}^N \frac{d}{dt} h_j \rho_j(t) = 0,$$

therefore (2.6) holds true for any $t > 0$.

(2) Let $\vec{F}(\vec{\rho})$ be the vector field defined by the right hand side of (2.3), then

$$\frac{d}{dt} \vec{\rho} = \vec{F}(\vec{\rho}). \quad (2.8)$$

Note that the hyperplane $\Sigma = \{\vec{\rho} : \sum_{j=1}^N h_j \rho_j = \sum_{j=1}^N h_j \rho_j(0)\}$ is an invariant region of (2.8). We define a closed set Σ_1 on this hyperplane by

$$\Sigma_1 = \left\{ \vec{\rho} : \rho_j \geq 0, j = 1, 2, \dots, N, \text{ and } \sum_{j=1}^N h_j \rho_j = \sum_{j=1}^N h_j \rho_j(0) \right\}.$$

It suffices to show that Σ_1 is invariant under system (2.8). This is the case if the vector field $\vec{F}(\vec{\rho})$ strictly points to interior of Σ_1 on its boundary $\partial\Sigma_1$: i.e.,

$$\vec{F}(\vec{\rho}) \cdot \vec{v} < 0,$$

where $\vec{v}$ is outward normal vector on any part of $\partial\Sigma_1$.

A direct calculation using (2.3) gives

$$\begin{aligned}\vec{F}(\vec{\rho}) \cdot \vec{v} &= \sum_{j=1}^{N-1} \frac{v_j}{h_j} C_{j+1/2} - \sum_{j=2}^N \frac{v_j}{h_j} C_{j-1/2} \\ &= - \sum_{j=1}^{N-1} \left(\frac{v_{j+1}}{h_{j+1}} - \frac{v_j}{h_j} \right) C_{j+1/2}.\end{aligned}\tag{2.9}$$

For each $\vec{\mu} \in \partial\Sigma_1$, we define the set $S = \{j : 1 \leq j \leq N \text{ and } \mu_j = 0\}$, then the outward normal vector at $\vec{\mu}$ has the form

$$\vec{v} = (v_1, v_2, \dots, v_N)^T \quad \text{with } v_i = \begin{cases} -\alpha_i, & i \in S, \\ 0, & i \notin S, \end{cases}$$

and $\alpha_i > 0$ if $i \in S$.

Note that if $j, j+1 \in S$, then $\rho_j = \rho_{j+1} = 0$ implies $C_{j+1/2} = 0$; if $j, j+1 \notin S$, then $v_{j+1} = v_j = 0$. Therefore nonzero terms in (2.9) are those with $j \in S, j+1 \notin S$ or $j \notin S, j+1 \in S$. Hence

$$\vec{F}(\vec{\rho}) \cdot \vec{v} = - \sum_{j \in S, j+1 \notin S} \frac{\alpha_j}{h_j} \frac{M_{j+\frac{1}{2}}}{h_{j+1/2}} \frac{\rho_{j+1}}{M_{j+1}} - \sum_{j \notin S, j+1 \in S} \frac{\alpha_{j+1}}{h_{j+1}} \frac{M_{j+\frac{1}{2}}}{h_{j+1/2}} \frac{\rho_j}{M_j} < 0.$$

Therefore Σ_1 is an invariant region of (2.3), this completes the proof of (2).

(3) From the fact that $W(x) = W(-x)$, it follows

$$\frac{d}{dt} \sum_{j=1}^N \frac{h_j}{2} g_j \rho_j = \sum_{j=1}^N h_j g_j \frac{d\rho_j}{dt}.\tag{2.10}$$

Differentiating the discrete free energy (2.5) with respect to time and using (2.10) we obtain

$$\begin{aligned}\frac{dE_h(t)}{dt} &= \sum_{j=1}^N (\log(\rho_j) + 1 + V_j + g_j) h_j \frac{d\rho_j}{dt} \\ &= \sum_{j=1}^N (\log(\frac{\rho_j}{M_j}) + 1) (C_{j+1/2} - C_{j-1/2}) \\ &= - \sum_{j=1}^{N-1} C_{j+1/2} \left(\log(\frac{\rho_{j+1}}{M_{j+1}}) - \log(\frac{\rho_j}{M_j}) \right) \\ &= -I_h \leq 0.\end{aligned}$$

Note that

$$\begin{aligned}I_h &= \sum_{j=1}^{N-1} C_{j+1/2} \left(\log(\frac{\rho_{j+1}}{M_{j+1}}) - \log(\frac{\rho_j}{M_j}) \right) \\ &= \sum_{j=1}^{N-1} \frac{1}{h_{j+1/2}} M_{j+1/2} \left(\frac{\rho_{j+1}}{M_{j+1}} - \frac{\rho_j}{M_j} \right) \left(\log(\frac{\rho_{j+1}}{M_{j+1}}) - \log(\frac{\rho_j}{M_j}) \right)\end{aligned}$$

and $(x - y)(\log x - \log y) \geq 0$ for $x, y \in \mathbb{R}^+$, so we have $I_h \geq 0$. □

3. FULLY DISCRETE SCHEME

For time discretization of (2.3), we use an implicit-explicit time discretization in order to construct an easy to implement yet stable numerical scheme without time step restriction.

3.1. Scheme formulation and algorithm. Let τ be time step and ρ_j^n be the numerical solution at $t_n = n\tau$ to approximate $\rho_j(t_n)$. From given ρ_j^n , $j = 1, 2, \dots, N$, we update to get ρ_j^{n+1} by

$$\frac{\rho_j^{n+1} - \rho_j^n}{\tau} = \frac{C_{j+1/2}^{n,*} - C_{j-1/2}^{n,*}}{h_j}, \quad j = 1, 2, \dots, N. \quad (3.1)$$

with

$$C_{j+1/2}^{n,*} = \frac{M_{j+1/2}^n}{h_{j+1/2}} \left(\frac{\rho_{j+1}^{n+1}}{M_{j+1}^n} - \frac{\rho_j^{n+1}}{M_j^n} \right) \quad \text{for } j = 1, 2, \dots, N-1,$$

$$C_{1/2}^{n,*} = C_{N+1/2}^{n,*} = 0,$$

where $M_{j+1/2}^n = Q_1(x_{j+1/2}, \rho^n)$ and $M_j^n = Q_1(x_j, \rho^n)$. The initial data is chosen by

$$\rho_j^0 = \frac{1}{h_j} \int_{I_j} \rho_0(x) dx, \quad j = 1, 2, \dots, N. \quad (3.2)$$

3.2. Scheme properties. Define a fully discrete version E_h^n of the free energy (1.5) as

$$E_h^n = \sum_{j=1}^N h_j \left(\rho_j^n \log(\rho_j^n) + V_j \rho_j^n + \frac{1}{2} g_j^n \rho_j^n \right), \quad (3.3)$$

where $g_j^n = \sum_{i=1}^N h_i W(x_i - x_j) \rho_i^n$.

The following theorem states that the three desired properties are preserved by the scheme (3.1) without strict time step restriction.

Theorem 3.1. The fully discrete scheme (3.1) has the following properties:

(1) Conservation of mass:

$$\sum_{j=1}^N h_j \rho_j^n = \int_{\Omega} \rho_0(x) dx \quad \text{for } n \geq 1. \quad (3.4)$$

(2) Positivity preserving: if $\rho_j^n \geq 0$ for all $j = 1, \dots, N$, then

$$\rho_j^{n+1} \geq 0, \quad j = 1, \dots, N.$$

(3) Entropy dissipation: there exists $\tau^* > 0$ such that if $\tau \in (0, \tau^*)$, then

$$E_h^{n+1} - E_h^n \leq -\frac{\tau}{2} I_h^n,$$

where

$$I_h^n = \sum_{j=1}^{N-1} C_{j+1/2}^{n,*} \left(\log\left(\frac{\rho_{j+1}^{n+1}}{M_{j+1}^n}\right) - \log\left(\frac{\rho_j^{n+1}}{M_j^n}\right) \right) \geq 0.$$

Proof. Set $G_j^{n,*} = \rho_j^{n+1}/M_j^n$ and $\lambda_{j+1/2} = \tau/h_{j+1/2}$, so the fully discrete scheme (3.1) can be rewritten into the following linear system:

$$\begin{aligned} h_1 \rho_1^n &= (h_1 M_1^n + \lambda_{1+1/2} M_{1+1/2}^n) G_1^{n,*} - \lambda_{1+1/2} M_{1+1/2}^n G_2^{n,*}, \\ h_j \rho_j^n &= -\lambda_{j-1/2} M_{j-1/2}^n G_{j-1}^{n,*} + (h_j M_j^n + \lambda_{j-1/2} M_{j-1/2}^n + \lambda_{j+1/2} M_{j+1/2}^n) G_j^{n,*} \\ &\quad - \lambda_{j+1/2} M_{j+1/2}^n G_{j+1}^{n,*} \quad j = 2, 3, \dots, N-1, \\ h_N \rho_N^n &= -\lambda_{N-1/2} M_{N-1/2}^n G_{N-1}^{n,*} + (h_N M_N^n + \lambda_{N-1/2} M_{N-1/2}^n) G_N^{n,*}. \end{aligned} \quad (3.5)$$

Note that the coefficient matrix of linear system (3.5) is strictly diagonally dominant, therefore (3.5) has a unique solution for whatever τ a priori chosen so dose (3.1) because $\rho_j^{n+1} = G_j^{n,*} M_j^n$.

(1) (5.1) follows from adding all equations in system (3.5) and using (3.2).

(2) Since $\rho_j^{n+1} = M_j^n G_j^{n,*}$ and $M_j^n > 0$, it suffices to prove that

$$G_i^{n,*} = \min_{1 \leq j \leq N} \{G_j^{n,*}\} \geq 0.$$

Assume $1 < i < N$, from i -th equation of (3.5) we have

$$\begin{aligned} h_i \rho_i^n &= -\lambda_{i-1/2} M_{i-1/2}^n G_{i-1}^{n,*} + (h_i M_i^n + \lambda_{i-1/2} M_{i-1/2}^n + \lambda_{i+1/2} M_{i+1/2}^n) G_i^{n,*} - \lambda_{i+1/2} M_{i+1/2}^n G_{i+1}^{n,*} \\ &\leq -\lambda_{i-1/2} M_{i-1/2}^n G_i^{n,*} + (h_i M_i^n + \lambda_{i-1/2} M_{i-1/2}^n + \lambda_{i+1/2} M_{i+1/2}^n) G_i^{n,*} - \lambda_{i+1/2} M_{i+1/2}^n G_i^{n,*} \\ &= h_i M_i^n G_i^{n,*}. \end{aligned}$$

Thus $G_i^{n,*} \geq \frac{\rho_i^n}{M_i^n} \geq 0$. A similar argument applies if $i = 1$ or $i = N$.

(3) A direct calculation using (3.3) gives

$$\begin{aligned} E_h^{n+1} - E_h^n &= \sum_{j=1}^N h_j (\rho_j^{n+1} \log(\rho_j^{n+1}) - \rho_j^n \log(\rho_j^n) + V_j \rho_j^{n+1} - V_j \rho_j^n + \frac{1}{2} g_j^{n+1} \rho_j^{n+1} - \frac{1}{2} g_j^n \rho_j^n) \\ &= \sum_{j=1}^N h_j ((\rho_j^{n+1} - \rho_j^n) \log(\rho_j^{n+1}) + (\rho_j^{n+1} - \rho_j^n) V_j + (\rho_j^{n+1} - \rho_j^n) g_j^n \\ &\quad + \frac{1}{2} g_j^n \rho_j^n - g_j^n \rho_j^{n+1} + \frac{1}{2} g_j^{n+1} \rho_j^{n+1} + \rho_j^n \log(\frac{\rho_j^{n+1}}{\rho_j^n})) \\ &\leq \sum_{j=1}^N h_j ((\rho_j^{n+1} - \rho_j^n) \log(G_j^{n,*}) + \frac{1}{2} g_j^n \rho_j^n - g_j^n \rho_j^{n+1} + \frac{1}{2} g_j^{n+1} \rho_j^{n+1}), \end{aligned}$$

here we have used $\rho_j^n \log(\frac{\rho_j^{n+1}}{\rho_j^n}) \leq \rho_j^n (\frac{\rho_j^{n+1}}{\rho_j^n} - 1)$ and mass conservation $\sum_{j=1}^N h_j (\rho_j^{n+1} - \rho_j^n) = 0$. We proceed with

$$\begin{aligned} \tau \sum_{j=1}^N (\frac{h_j \rho_j^{n+1} - h_j \rho_j^n}{\tau}) \log(G_j^{n,*}) &= \tau \sum_{j=1}^N (\log(G_j^{n,*}) (h_{j+1/2}^{-1} M_{j+1/2}^n (G_{j+1}^{n,*} - G_j^{n,*}) \\ &\quad - h_{j-1/2}^{-1} M_{j-1/2}^n (G_j^{n,*} - G_{j-1}^{n,*}))) \\ &= -\tau \sum_{j=1}^{N-1} h_{j+1/2}^{-1} M_{j+1/2}^n (G_{j+1}^{n,*} - G_j^{n,*}) (\log G_{j+1}^{n,*} - \log G_j^{n,*}) \\ &= -\tau I_h^n \leq 0. \end{aligned} \quad (3.6)$$

Here the sign of I_h^n is implied by the monotonicity of the logarithmic function.

It remains to find a sufficient condition on time step τ so that

$$\sum_{j=1}^N h_j \left(\frac{1}{2} g_j^n \rho_j^n - g_j^n \rho_j^{n+1} + \frac{1}{2} g_j^{n+1} \rho_j^{n+1} \right) \leq -\frac{\tau}{2} \sum_{j=1}^N \left(\frac{h_j \rho_j^{n+1} - h_j \rho_j^n}{\tau} \right) \log(G_j^{n,*}). \quad (3.7)$$

From $\sum_{j=1}^N h_j g_j^n \rho_j^{n+1} = \sum_{j=1}^N h_j g_j^{n+1} \rho_j^n$ it follows that

$$\begin{aligned} \sum_{j=1}^N h_j \left(\frac{1}{2} g_j^n \rho_j^n - g_j^n \rho_j^{n+1} + \frac{1}{2} g_j^{n+1} \rho_j^{n+1} \right) &= \frac{1}{2} \sum_{j=1}^N h_j (g_j^{n+1} - g_j^n) (\rho_j^{n+1} - \rho_j^n) \\ &= \frac{1}{2} \sum_{j=1}^N h_j \sum_{i=1}^N h_i W(x_i - x_j) (\rho_i^{n+1} - \rho_i^n) (\rho_j^{n+1} - \rho_j^n) \\ &\leq \frac{\|W\|_\infty}{2} \sum_{j=1}^N h_j \sum_{i=1}^N h_i |\rho_i^{n+1} - \rho_i^n| |\rho_j^{n+1} - \rho_j^n| \\ &\leq \frac{\|W\|_\infty (b-a) \tau^2}{2} \sum_{j=1}^N h_j \left(\frac{\rho_j^{n+1} - \rho_j^n}{\tau} \right)^2, \end{aligned}$$

where we have used the Cauchy-Schwarz inequality and $b-a = \sum_{j=1}^N h_j$. Let $\vec{\xi}, \vec{\eta} \in \mathbb{R}^N$ be vectors defined as $\vec{\xi}_j = \frac{\sqrt{h_j}(\rho_j^{n+1} - \rho_j^n)}{\tau}$, $\vec{\eta}_j = \sqrt{h_j} \log G_j^{n,*}$, then (3.7) is satisfied if

$$\frac{\|W\|_\infty (b-a) \tau^2}{2} |\vec{\xi}|^2 + \frac{\tau}{2} \vec{\xi} \cdot \vec{\eta} \leq 0.$$

We claim that

$$\vec{\xi} \cdot \vec{\eta} = 0 \quad \text{if and only if} \quad \vec{\xi} = 0. \quad (3.8)$$

Therefore

$$0 < c_0 \leq \frac{-\vec{\xi} \cdot \vec{\eta}}{|\vec{\xi}|^2} \leq \frac{|\eta|}{|\xi|} \quad \text{for } \xi \neq 0,$$

where c_0 may depend on numerical solutions at t_n and t_{n+1} . We thus obtain (3.7) by taking

$$\tau \leq \tau^* = \frac{c_0}{\|W\|_\infty (b-a)}.$$

Finally, we verify claim (3.8). If $\vec{\xi} \cdot \vec{\eta} = 0$, then from (3.6) we have

$$0 = \vec{\xi} \cdot \vec{\eta} = -\tau \sum_{j=1}^{N-1} h_{j+1/2}^{-1} M_{j+1/2}^n (\log G_{j+1}^{n,*} - \log G_j^{n,*}) (G_{j+1}^{n,*} - G_j^{n,*}) \leq 0,$$

therefore we must have $G_j^{n,*} = \text{constan}$ for all $j \in \{1, 2, \dots, N\}$. This when inserted into scheme (3.1) leads to

$$\rho_j^{n+1} = \rho_j^n \quad \text{for all } j \in \{1, 2, \dots, N\},$$

thus $\vec{\xi} = 0$. □

Remark 3.1. One could take the Euler forward time discretization to obtain an explicit scheme: From ρ_j^n , $j = 1, 2, \dots, N$, update to get ρ_j^{n+1} by

$$\frac{\rho_j^{n+1} - \rho_j^n}{\tau} = \frac{C_{j+1/2}^n - C_{j-1/2}^n}{h_j}, \quad j = 1, 2, \dots, N.$$

where

$$C_{j+1/2}^n = \frac{M_{j+1/2}^n}{h_{j+1/2}} \left(\frac{\rho_{j+1}^n}{M_{j+1}^n} - \frac{\rho_j^n}{M_j^n} \right) \quad \text{for } j = 1, 2, \dots, N-1,$$

$$C_{1/2}^n = C_{N+1/2}^n = 0,$$

with $M_{j+1/2}^n = Q_1(x_{j+1/2}, \rho^n)$ and $M_j^n = Q_1(x_j, \rho^n)$. One can show that the positivity preserving property is still met yet under a CFL condition like $\tau \leq \gamma h^2$.

4. NUMERICAL METHOD: TWO DIMENSIONAL CASE

In this section, we extend our method to multi-dimensional problems. For simplicity, we only present schemes for the two dimensional initial value problem,

$$\begin{cases} \partial_t \rho = \nabla \cdot (\nabla \rho + \rho \nabla (V(x, y) + W * \rho)), & (x, y) \in \Omega \subset \mathbb{R}^2, \quad t > 0, \\ \rho(x, y, 0) = \rho_0(x, y), & (x, y) \in \Omega, \end{cases} \quad (4.1)$$

on a rectangular domain $\Omega = [a, b] \times [c, d]$ subject to zero flux boundary conditions.

For given positive integers N_x, N_y , we partition Ω by a Cartesian mesh with computational cells

$$I_{i,j} = [x_{i-1/2}, x_{i+1/2}] \times [y_{j-1/2}, y_{j+1/2}],$$

where $i \in \{1, 2, \dots, N_x\}, j \in \{1, 2, \dots, N_y\}$. The mesh size is $|I_{i,j}| = h_i^x h_j^y$ with the cell center at $(x_i, y_j) = (x_{i-1/2} + \frac{1}{2}h_i^x, y_{j-1/2} + \frac{1}{2}h_j^y)$, we set $h_{i+1/2}^x = (h_i^x + h_{i+1}^x)/2$, $h_{j+1/2}^y = (h_j^y + h_{j+1}^y)/2$.

4.1. Semi-discrete scheme. Let $\rho(t) = \{\rho_{i,j}\}$ be the numerical solution, then dimension by dimension spatial discretization of

$$\partial_t \rho = \nabla \cdot \left(M \nabla \left(\frac{\rho}{M} \right) \right), \quad \text{with } M = e^{-V(x,y) - W * \rho},$$

yields the following semi-discrete scheme

$$\frac{d}{dt} \rho_{i,j} = \frac{C_{i+1/2,j} - C_{i-1/2,j}}{h_i^x} + \frac{C_{i,j+1/2} - C_{i,j-1/2}}{h_j^y}, \quad (4.2)$$

where

$$C_{i+1/2,j} = \frac{M_{i+1/2,j}}{h_{i+1/2}^x} \left(\frac{\rho_{i+1,j}}{M_{i+1,j}} - \frac{\rho_{i,j}}{M_{i,j}} \right), \quad i = 1, \dots, N_x - 1, j = 1, \dots, N_y,$$

$$C_{i,j+1/2} = \frac{M_{i,j+1/2}}{h_{j+1/2}^y} \left(\frac{\rho_{i,j+1}}{M_{i,j+1}} - \frac{\rho_{i,j}}{M_{i,j}} \right), \quad i = 1, \dots, N_x, j = 1, \dots, N_y - 1,$$

$$C_{1/2,j} = C_{N_x+1/2,j} = C_{i,1/2} = C_{i,N_y+1/2} = 0, \quad i = 1, \dots, N_x, j = 1, \dots, N_y,$$

with $M_{i+1/2,j} = Q_2(x_{i+1/2}, y_j, \rho)$, $M_{i,j+1/2} = Q_2(x_i, y_{j+1/2}, \rho)$, and $M_{i,j} = Q_2(x_i, y_j, \rho)$. Where

$$Q_2(x, y, v) = e^{-V(x,y) - \sum_{k=1}^{N_x} \sum_{l=1}^{N_y} h_k^x h_l^y W(x_k - x, y_l - y) v_{k,l}}, \quad \text{for } x, y \in \mathbb{R}, v \in \mathbb{R}^{N_x \times N_y}. \quad (4.3)$$

Let

$$E_h(t) = \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y \left(\rho_{i,j} \log(\rho_{i,j}) + V_{i,j} \rho_{i,j} + \frac{1}{2} g_{i,j} \rho_{i,j} \right),$$

be an approximation of the entropy functional (1.5), with

$$g_{i,j} = \sum_{k=1}^{N_x} \sum_{l=1}^{N_y} h_k^x h_l^y W(x_k - x_i, y_l - y_j) \rho_{k,l}.$$

The following theorem states that the semi-discrete scheme (4.2) is conservative, positive, and energy dissipating.

Theorem 4.1. The semi-discrete scheme (4.2) satisfies the following properties:

(1) Conservation of mass: for any $t > 0$,

$$\sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y \rho_{i,j}(t) = \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y \rho_{i,j}(0).$$

(2) Positivity preserving: if $\rho_{i,j}(0) \geq 0$ for all $i \in \{1, \dots, N_x\}$, $j \in \{1, \dots, N_y\}$, then $\rho_{i,j}(t) \geq 0$ for any $t > 0$.

(3) Entropy dissipation: $\frac{dE_h(t)}{dt} \leq -I_h$, where

$$\begin{aligned} I_h &= \sum_{j=1}^{N_y} \sum_{i=1}^{N_x-1} h_j^y C_{i+1/2,j} \left(\log\left(\frac{\rho_{i+1,j}}{M_{i+1,j}}\right) - \log\left(\frac{\rho_{i,j}}{M_{i,j}}\right) \right) \\ &\quad + \sum_{i=1}^{N_x} \sum_{j=1}^{N_y-1} h_i^x C_{i,j+1/2} \left(\log\left(\frac{\rho_{i,j+1}}{M_{i,j+1}}\right) - \log\left(\frac{\rho_{i,j}}{M_{i,j}}\right) \right) \geq 0. \end{aligned}$$

Proof. The proof is similar to that of Theorem 2.1, details are therefore omitted. $\square$

4.2. Fully discrete scheme. Let $\rho_{i,j}^n$ approximate $\rho_{i,j}(t_n)$, then (4.2) gives the following fully discrete scheme,

$$\frac{\rho_{i,j}^{n+1} - \rho_{i,j}^n}{\tau} = \frac{C_{i+1/2,j}^{n,*} - C_{i-1/2,j}^{n,*}}{h_i^x} + \frac{C_{i,j+1/2}^{n,*} - C_{i,j-1/2}^{n,*}}{h_j^y}, \quad (4.4)$$

where

$$\begin{aligned} C_{i+1/2,j}^{n,*} &= \frac{M_{i+1/2,j}^n}{h_{i+1/2}^x} \left(\frac{\rho_{i+1,j}^{n+1}}{M_{i+1,j}^n} - \frac{\rho_{i,j}^{n+1}}{M_{i,j}^n} \right), \quad i = 1, \dots, N_x - 1, j = 1, \dots, N_y, \\ C_{i,j+1/2}^{n,*} &= \frac{M_{i,j+1/2}^n}{h_{j+1/2}^y} \left(\frac{\rho_{i,j+1}^{n+1}}{M_{i,j+1}^n} - \frac{\rho_{i,j}^{n+1}}{M_{i,j}^n} \right), \quad i = 1, \dots, N_x, j = 1, \dots, N_y - 1, \\ C_{1/2,j}^{n,*} &= C_{N_x+1/2,j}^{n,*} = C_{i,1/2}^{n,*} = C_{i,N_y+1/2}^{n,*} = 0, \quad i = 1, \dots, N_x, j = 1, \dots, N_y, \end{aligned}$$

with $M_{i+1/2,j}^n = Q_2(x_{i+1/2}, y_j, \rho^n)$, $M_{i,j+1/2}^n = Q_2(x_i, y_{j+1/2}, \rho^n)$, and $M_{i,j}^n = Q_2(x_i, y_j, \rho^n)$.

The initial data is chosen as

$$\rho_{i,j}^0 = \frac{1}{|I_{i,j}|} \int_{I_{i,j}} \rho_0(x, y) dx dy. \quad (4.5)$$

In 2D case, a discrete version of entropy (4.5) may be defined as

$$E_h^n = \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y \left(\rho_{i,j}^n \log(\rho_{i,j}^n) + V_{i,j} \rho_{i,j}^n + \frac{1}{2} g_{i,j}^n \rho_{i,j}^n \right), \quad (4.6)$$

where

$$g_{i,j}^n = \sum_{k=1}^{N_x} \sum_{l=1}^{N_y} h_k^x h_l^y W(x_k - x_i, y_l - y_j) \rho_{k,l}^n.$$

Theorem 4.2. The fully discrete scheme (4.4) has the following properties:

(1) Conservation of mass:

$$\sum_{l=1}^{N_x} \sum_{j=1}^{N_y} h_l^x h_j^y \rho_{l,j}^n = \int_{\Omega} \rho_0(x, y) dx dy, \quad \text{for all } n \geq 1. \quad (4.7)$$

(2) Positivity preserving: if $\rho_{i,j}^n \geq 0$ for all $i \in \{1, \dots, N_x\}$ and $j \in \{1, \dots, N_y\}$, then

$$\rho_{i,j}^{n+1} \geq 0.$$

(3) Entropy dissipation: there exists $\tau^* > 0$ such that if $\tau \in (0, \tau^*)$, then

$$E_h^{n+1} - E_h^n \leq -\frac{\tau}{2} I_h^n, \quad (4.8)$$

where

$$\begin{aligned} I_h^n &= \sum_{j=1}^{N_y} \sum_{i=1}^{N_x-1} h_j^y C_{i+1/2,j}^{n,*} \left(\log \frac{\rho_{i+1,j}^{n+1}}{M_{i+1,j}^n} - \log \frac{\rho_{i,j}^{n+1}}{M_{i,j}^n} \right) \\ &\quad + \sum_{i=1}^{N_x} \sum_{j=1}^{N_y-1} h_i^x C_{i,j+1/2}^{n,*} \left(\log \frac{\rho_{i,j+1}^{n+1}}{M_{i,j+1}^n} - \log \frac{\rho_{i,j}^{n+1}}{M_{i,j}^n} \right) \geq 0. \end{aligned}$$

Proof. For simplicity of analysis we rewrite the scheme (4.4) as

$$\begin{aligned} h_i^x h_j^y \rho_{i,j}^n &= (h_i^x h_j^y M_{i,j}^n + \tau \tilde{M}_{i+1/2,j}^n + \tau \tilde{M}_{i-1/2,j}^n + \tau \tilde{M}_{i,j+1/2}^n + \tau \tilde{M}_{i,j-1/2}^n) G_{i,j}^{n,*} \\ &\quad - \tau \tilde{M}_{i+1/2,j}^n G_{i+1,j}^{n,*} - \tau \tilde{M}_{i-1/2,j}^n G_{i-1,j}^{n,*} - \tau \tilde{M}_{i,j+1/2}^n G_{i,j+1}^{n,*} - \tau \tilde{M}_{i,j-1/2}^n G_{i,j-1}^{n,*}, \end{aligned} \quad (4.9)$$

with the following notations

$$\tilde{M}_{i+1/2,j}^n = \frac{h_j^y}{h_{i+1/2}^x} M_{i+1/2,j}^n, \quad \tilde{M}_{i,j+1/2}^n = \frac{h_i^x}{h_{j+1/2}^y} M_{i,j+1/2}^n, \quad G_{i,j}^{n,*} = \frac{\rho_{i,j}^{n+1}}{M_{i,j}^n}.$$

Note that the coefficient matrix of the linear system (4.9) (when consider $G_{i,j}^{n,*}$ as unknowns) is strictly diagonally dominant, therefore (4.9) always has a unique solution.

(1) Adding all equations in (4.4) and using (4.5) lead to (4.7).

(2) Since $\rho_{i,j}^{n+1} = M_{i,j}^n G_{i,j}^{n,*}$ and $M_{i,j}^n > 0$, it suffices to prove that $G_{k,l}^{n,*} = \min_{\{i,j\}} G_{i,j}^{n,*} \geq 0$, the corresponding equation is

$$\begin{aligned} h_k^x h_l^y \rho_{k,l}^n &= (h_k^x h_l^y M_{k,l}^n + \tau \tilde{M}_{k+1/2,l}^n + \tau \tilde{M}_{k-1/2,l}^n + \tau \tilde{M}_{k,l+1/2}^n + \tau \tilde{M}_{k,l-1/2}^n) G_{k,l}^{n,*} \\ &\quad - \tau \tilde{M}_{k+1/2,l}^n G_{k+1,l}^{n+1} - \tau \tilde{M}_{k-1/2,l}^n G_{k-1,l}^{n+1} - \tau \tilde{M}_{k,l+1/2}^n G_{k,l+1}^{n+1} - \tau \tilde{M}_{k,l-1/2}^n G_{k,l-1}^{n+1} \\ &\leq h_k^x h_l^y M_{k,l}^n G_{k,l}^{n,*}, \end{aligned}$$

therefore $G_{k,l}^{n,*} \geq 0$.

(3) A direct calculation using (4.6) gives

$$\begin{aligned}
E_h^{n+1} - E_h^n &= \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y (\rho_{i,j}^{n+1} \log(\rho_{i,j}^{n+1}) - \rho_{i,j}^n \log(\rho_{i,j}^{n+1}) + \rho_{i,j}^n \log(\frac{\rho_{i,j}^{n+1}}{\rho_{i,j}^n})) \\
&\quad + V_j \rho_{i,j}^{n+1} + \frac{1}{2} g_{i,j}^{n+1} \rho_{i,j}^{n+1} - V_{i,j} \rho_{i,j}^n - \frac{1}{2} g_{i,j}^n \rho_{i,j}^n \\
&\leq \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y (\log(G_{i,j}^{n,*}) (\rho_{i,j}^{n+1} - \rho_{i,j}^n) + \frac{1}{2} g_{i,j}^n \rho_{i,j}^n - g_j^n \rho_{i,j}^{n+1} + \frac{1}{2} g_{i,j}^{n+1} \rho_{i,j}^{n+1}),
\end{aligned} \tag{4.10}$$

where we have used $\log(x) \leq x - 1$ and mass conservation property. By the symmetrical property of $W(x, y)$ we have

$$\sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y g_{i,j}^n \rho_{i,j}^{n+1} = \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y g_{i,j}^{n+1} \rho_{i,j}^n,$$

so that

$$\begin{aligned}
&\sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y (\frac{1}{2} g_{i,j}^n \rho_{i,j}^n - g_{i,j}^n \rho_{i,j}^{n+1} + \frac{1}{2} g_{i,j}^{n+1} \rho_{i,j}^{n+1}) \\
&= \frac{1}{2} \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y (g_{i,j}^{n+1} - g_{i,j}^n) (\rho_{i,j}^{n+1} - \rho_{i,j}^n) \\
&= \frac{1}{2} \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y (\sum_{k=1}^{N_x} \sum_{l=1}^{N_y} h_k^x h_l^y W(x_i - x_k, y_j - y_l) (\rho_{k,l}^{n+1} - \rho_{k,l}^n)) (\rho_{i,j}^{n+1} - \rho_{i,j}^n) \\
&\leq \frac{\|W\|_\infty}{2} \left(\sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y |\rho_{i,j}^{n+1} - \rho_{i,j}^n| \right)^2 \\
&\leq \frac{\|W\|_\infty |\Omega|}{2} \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y (\rho_{i,j}^{n+1} - \rho_{i,j}^n)^2,
\end{aligned}$$

where $|\Omega| = \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y$. Substitution of the above inequality into (4.10) yields

$$\begin{aligned}
E_h^{n+1} - E_h^n &\leq \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y \log(G_{i,j}^{n,*}) (\rho_{i,j}^{n+1} - \rho_{i,j}^n) + \frac{\|W\|_\infty |\Omega|}{2} \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y (\rho_{i,j}^{n+1} - \rho_{i,j}^n)^2 \\
&:= F_1^n + F_2^n.
\end{aligned}$$

We proceed using summation by parts and boundary conditions so that

$$\begin{aligned}
F_1^n &= \tau \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} \log(G_{i,j}^{n,*}) (\tilde{M}_{i+1/2,j}^n (G_{i+1,j}^{n,*} - G_{i,j}^{n,*}) - \tilde{M}_{i-1/2,j}^n (G_{i,j}^{n,*} - G_{i-1,j}^{n,*})) \\
&\quad + \tau \sum_{i=1}^{N_x} \sum_{j=1}^{N_y} \log(G_{i,j}^{n,*}) (\tilde{M}_{i,j+1/2}^n (G_{i,j+1}^{n,*} - G_{i,j}^{n,*}) - \tilde{M}_{i,j-1/2}^n (G_{i,j}^{n,*} - G_{i,j-1}^{n,*}))
\end{aligned}$$

$$\begin{aligned}
&= -\tau \sum_{i=1}^{N_x-1} \sum_{j=1}^{N_y} \tilde{M}_{i+1/2,j}^n (\log(G_{i+1,j}^{n,*}) - \log(G_{i,j}^{n,*})) (G_{i+1,j}^{n,*} - G_{i,j}^{n,*}) \\
&\quad - \tau \sum_{i=1}^{N_x} \sum_{j=1}^{N_y-1} \tilde{M}_{i,j+1/2}^n (\log(G_{i,j+1}^{n,*}) - \log(G_{i,j}^{n,*})) (G_{i,j+1}^{n,*} - G_{i,j}^{n,*}) \\
&= -\tau I_h^n.
\end{aligned}$$

It remains to figure out a condition on τ so that $F_2^n + \frac{1}{2}F_1^n \leq 0$. Let $\vec{\xi}, \vec{\eta} \in \mathbb{R}^{N_x N_y}$ be vectors defined as:

$$\begin{aligned}
\vec{\xi} &= \left(\frac{\sqrt{h_1^x h_1^y} (\rho_{1,1}^{n+1} - \rho_{1,1}^n)}{\tau}, \dots, \frac{\sqrt{h_{N_x}^x h_1^y} (\rho_{N_x,1}^{n+1} - \rho_{N_x,1}^n)}{\tau}, \dots, \frac{\sqrt{h_{N_x}^x h_{N_y}^y} (\rho_{N_x,N_y}^{n+1} - \rho_{N_x,N_y}^n)}{\tau} \right)^T \\
\vec{\eta} &= (\sqrt{h_1^x h_1^y} \log(G_{1,1}^{n,*}), \dots, \sqrt{h_{N_x}^x h_1^y} \log(G_{N_x,1}^{n,*}), \dots, \sqrt{h_{N_x}^x h_{N_y}^y} \log(G_{N_x,N_y}^{n,*}))^T,
\end{aligned}$$

then $F_2^n + \frac{1}{2}F_1^n \leq 0$ if

$$\tau^2 \|W\|_\infty |\Omega| \vec{\xi}^2 + \tau \vec{\xi} \cdot \vec{\eta} \leq 0.$$

In similar manner as in 1D case, we can show that $\vec{\xi} \cdot \vec{\eta} = 0$ if and only if $\vec{\xi} = 0$. Therefore

$$0 < c_0 \leq \frac{-\vec{\xi} \cdot \vec{\eta}}{|\vec{\xi}|^2} \leq \frac{|\eta|}{|\xi|} \quad \text{for } \xi \neq 0,$$

where c_0 may depend on numerical solutions at t_n and t_{n+1} . We thus obtain the desired result (4.8) by taking $\tau \leq \tau^* = \frac{c_0}{\|W\|_\infty |\Omega|}$. $\square$

Remark 4.1. The schemes presented so far apply well to the general class of nonlinear nonlocal equations (1.6), based on the reformulation

$$\partial_t \rho = \nabla \cdot (M \nabla \frac{\rho}{M}),$$

where $M = \rho e^{-H'(\rho) - V(\mathbf{x}) - W * \rho}$ for ρ away from zero. The numerical solution may be oscillatory at low density, for which one could use either upwind numerical fluxes or non-oscillatory limiters as a remedy [8]. Note that for the aggregation equation (in the absence of diffusion), particle methods have been developed in [11, 36]; Particle methods naturally conserve mass and positivity, yet a large number of particles is often required to resolve finer properties of solutions.

5. SECOND ORDER IN-TIME DISCRETIZATION

The numerical schemes presented so far are only first order in time. In this section we extend these schemes with a second order in time discretization.

5.1. Second order scheme for 1D problem. We replace (3.1) by a two step scheme

$$\frac{\rho_j^* - \rho_j^n}{\tau/2} = \frac{C_{j+1/2}^* - C_{j-1/2}^*}{h_j}, \quad j = 1, 2, \dots, N, \quad (5.1a)$$

$$\rho_j^{n+1} = 2\rho_j^* - \rho_j^n, \quad j = 1, 2, \dots, N, \quad (5.1b)$$

where

$$C_{j+1/2}^* = \frac{M_{j+1/2}^*}{h_{j+1/2}} \left(\frac{\rho_{j+1}^*}{M_{j+1}^*} - \frac{\rho_j^*}{M_j^*} \right), \quad \text{for } j = 1, 2, \dots, N-1,$$

$$C_{1/2}^* = 0, \quad C_{N+1/2}^* = 0,$$

with $M_{j+1/2}^* = Q_1(x_{j+1/2}, \frac{3}{2}\rho^n - \frac{1}{2}\rho^{n-1})$ and $M_j^* = Q_1(x_j, \frac{3}{2}\rho^n - \frac{1}{2}\rho^{n-1})$. The scheme (5.1) has following properties.

Theorem 5.1. Let ρ_j^{n+1} be obtained from (5.1), then

(1) Conservation of mass:

$$\sum_{j=1}^N h_j \rho_j^n = \int_{\Omega} \rho_0(x) dx, \quad \text{for } n \geq 1.$$

(2) Positivity preserving: if $\rho_j^n \geq 0$ for all $j = 1, \dots, N$, then

$$\rho_j^{n+1} \geq 0, \quad j = 1, \dots, N,$$

provided τ is sufficiently small.

Proof. (1) From the scheme construction, the conservation property remains hold.

(2) Setting

$$G_j^n = \frac{\rho_j^n}{M_j^*}, \quad g_{j+1/2}^* = \frac{M_{j+1/2}^*}{h_{j+1/2}},$$

and a careful regrouping leads to the following linear system

$$\begin{aligned} \left(M_1^* + \frac{\tau}{2h_1} g_{3/2}^* \right) G_1^{n+1} - \frac{\tau}{2h_1} g_{3/2}^* G_2^{n+1} &= b_1, \\ \left(M_j^* + \frac{\tau}{2h_j} (g_{j+1/2}^* + g_{j-1/2}^*) \right) G_j^{n+1} - \frac{\tau}{2h_j} g_{j+1/2}^* G_{j+1}^{n+1} - \frac{\tau}{2h_j} g_{j-1/2}^* G_{j-1}^{n+1} &= b_j, \\ \left(M_N^* + \frac{\tau}{2h_N} g_{N-1/2}^* \right) G_N^{n+1} - \frac{\tau}{2h_N} g_{N-1/2}^* G_{N-1}^{n+1} &= b_N, \end{aligned} \quad (5.2)$$

where $j = 1, \dots, N-1$, with the right hand side vector given by

$$\begin{aligned} b_1 &= \left(M_1^* - \frac{\tau}{2h_1} g_{3/2}^* \right) G_1^n + \frac{\tau}{2h_1} g_{3/2}^* G_2^n, \\ b_j &= \left(M_j^* - \frac{\tau}{2h_j} (g_{j+1/2}^* + g_{j-1/2}^*) \right) G_j^n + \frac{\tau}{2h_j} g_{j+1/2}^* G_{j+1}^n + \frac{\tau}{2h_j} g_{j-1/2}^* G_{j-1}^n, \quad j = 1, \dots, N-1, \\ b_N &= \left(M_N^* - \frac{\tau}{2h_N} g_{N-1/2}^* \right) G_N^n + \frac{\tau}{2h_N} g_{N-1/2}^* G_{N-1}^n. \end{aligned}$$

The linear system (5.2) admits a unique solution $\{G_j^{n+1}\}$ since its coefficient matrix is strictly diagonally dominant. Following the proof of (2) in Theorem 3.1, we see that $G_j^{n+1} \geq 0$ is ensured if each $b_j \geq 0$, which is the case provided

$$\tau \leq \min \left\{ \frac{2h_1 M_1^*}{g_{3/2}^*}, \min_{1 < j < N} \frac{2h_j M_j^*}{g_{j+1/2}^* + g_{j-1/2}^*}, \frac{2h_N M_N^*}{g_{N-1/2}^*} \right\}.$$

The stated result thus follows. $\square$

For large time step τ , non-negativity of ρ_j^{n+1} obtained by the second order scheme (5.1) may not be guaranteed, we introduce a local limiter to resolve the solution positivity.

5.2. Local limiter and algorithm. We begin to design a local limiter to restore positivity of $\{c_j\}_{j=1}^N$ if $\sum_{j=1}^N c_j > 0$, but $c_k < 0$ for some k . The idea is to find a neighboring index set S_k such that the local average

$$\bar{c}_k = \frac{1}{|S_k|} \sum_{j \in S_k} c_j > 0,$$

where $|S_k|$ denotes the minimum number of indexes for which $c_j \neq 0$ and $\bar{c}_k > 0$, then use this as a reference to define the following scaling limiter,

$$\tilde{c}_j = \theta c_j + (1 - \theta) \bar{c}_k, \quad j \in S_k, \quad (5.3)$$

where

$$\theta = \min \left\{ 1, \frac{\bar{c}_k}{\bar{c}_k - c_{\min}} \right\}, \quad c_{\min} = \min_{j \in S_k} c_j.$$

Lemma 5.1. This limiter has the following properties:

- (1) $\tilde{c}_j \geq 0$ for all $j \in S_k$,
- (2) $\sum_{j \in S_k} \tilde{c}_j = \sum_{j \in S_k} c_j$, and
- (3) $|\tilde{c}_j - c_j| \leq |S_k|(-\min_{j \in S_k} c_j)$.

Proof. (1) This follows from the definition of θ and (5.3).

(2) By (5.3) and the definition of $\bar{c}_k$, it follows that

$$\sum_{j \in S_k} \tilde{c}_j = \theta |S_k| \bar{c}_k + (1 - \theta) \bar{c}_k |S_k| = \sum_{j \in S_k} c_j.$$

(3) From (5.3) it follows that for all $j \in S_k$,

$$\begin{aligned} |\tilde{c}_j - c_j| &= (1 - \theta) |\bar{c}_k - c_j| = -c_{\min} \frac{|\bar{c}_k - c_j|}{(\bar{c}_k - c_{\min})} \\ &\leq (-c_{\min}) \max \left\{ 1, \frac{c_{\max} - \bar{c}_k}{\bar{c}_k - c_{\min}} \right\}, \end{aligned}$$

where $c_{\max} := \max_{j \in S_k} c_j$ and $c_{\min} := \min_{j \in S_k} c_j$. Note that $\sum_{j \in S_k} (\bar{c}_k - c_j) = 0$ implies

$$\sum_{j \in S_k^+} (c_j - \bar{c}_k) = \sum_{j \in S_k^-} (\bar{c}_k - c_j),$$

in which each term involved on both sides is nonnegative. Hence, $c_{\max} - \bar{c}_k \leq |S_k|(\bar{c}_k - c_{\min})$. Obviously, $|S_k| \geq 1$. Hence the claimed bound follows. $\square$

Remark 5.1. In general, $|S_k|$ may not be bounded. For instance, we let

$$c_j = \frac{1}{2^j} \text{ for } j = 1, \dots, N-1, \text{ and } c_N = -\frac{1}{2},$$

then $\sum_{j=1}^N c_j = \frac{1}{2} - \frac{1}{2^{N-1}} > 0$, but $\sum_{j=2}^N c_j = -\frac{1}{2^{N-1}} < 0$. This implies that $|S_N| = N$ since $S_N = \{1, \dots, N\}$.

The above limiter when applied to $\{\rho_j\}$ with $c_j = h_j \rho_j$ gives

$$\tilde{\rho}_j = \theta \rho_j + (1 - \theta) \frac{\bar{c}_k}{h_j}, \quad (5.4)$$

where

$$\theta = \min \left\{ 1, \frac{\bar{c}_k}{\bar{c}_k - c_{\min}} \right\}, \quad c_{\min} = \min_{j \in S_k} h_j \rho_j, \quad \bar{c}_k = \frac{1}{|S_k|} \sum_{j \in S_k} h_j \rho_j.$$

Such limiter still respects the local mass conservation. In addition, for any sequence g_j with $g_j \geq 0$, we have

$$|\tilde{\rho}_j - g_j| \leq (1 + |S_k| \alpha) \max_{j \in S_k} |\rho_j - g_j|, \quad j \in S_k,$$

where α is the upper bound of mesh ratio h_i/h_j . Let ρ_j be the approximation of $\rho(x) \geq 0$, we let $g_j = \rho(x_j)$ or the average of ρ on I_j , so we can assert that the accuracy is not destroyed by the limiter as long as $|S_k| \alpha$ is uniformly bounded. In practice, it is indeed the case as verified by our numerical tests when using shape-regular meshes.

Indeed, the boundedness of $|S_k|$ can be proved rigorously for shape-regular meshes.

Theorem 5.2. Let $\rho(x) \geq 0$, be in $C^2(\Omega)$, and $\{\rho_j\}$ be an approximation of $\rho(x)$ such that $|\rho_j - \rho(x_j)| \leq Ch^2$, where $h = \min_{1 \leq j \leq N} h_j$ and $h_j \leq \alpha h$ for some $\alpha > 0$. If $\rho_k < 0$ (or only finite number of neighboring values are negative), then there exists $K^* > 0$ finite such that

$$|S_k| \leq K^*.$$

where K^* may depend on the local meshes associated with S_k .

Proof. Under the assumption $\rho_k < 0$, ρ must touch zero near x_k . We discuss the case where $\rho(x^*) = 0$ and $\rho'(x^*) = 0$ with $\rho(x) > 0$ for $x > x^*$ locally with $x^* \in I_k$. The case where $\rho(x) > 0$ for $x < x^*$ can be handled as well. Without loss of generality, we consider $k = 1$ with $x^* \in I_1$, and $\int_{I_1} \rho(x) dx > 0$. It suffices to find K such that

$$\sum_{j=1}^K h_j \rho_j > 0. \quad (5.5)$$

Using the error bound we have

$$\rho_j \geq \rho(x_j) - Ch^2.$$

Also from $\rho \in C^2$ we can deduce that

$$\rho(x_j) \geq \bar{\rho}_j - \lambda h_j^2,$$

with $\lambda = \frac{1}{24} \max_{x \in \Omega} |\rho''|$ and the cell average $\bar{\rho}_j = \frac{1}{h_j} \int_{I_j} \rho(x) dx$. Combining these we see that the left hand side of (5.5) is bounded from below by

$$\begin{aligned} \sum_{j=1}^K h_j \rho_j &\geq \sum_{j=1}^K h_j (\bar{\rho}_j - Ch^2 - \lambda h_j^2) \\ &\geq \int_{x_{1/2}}^{x_{K+1/2}} \rho(x) dx - (\lambda + C) \sum_{j=1}^K h_j^3 \\ &\geq \int_{x_{1/2}}^{x_{K+1/2}} \rho(x) dx - (\lambda + C) h^2 \alpha^2 \sum_{j=1}^K h_j \\ &= \left[\int_0^1 \rho(\theta\eta + x_{1/2}) d\theta - (\lambda + C) h^2 \alpha^2 \right] \eta, \end{aligned}$$

where $\eta := \sum_{j=1}^K h_j$, and we have used $h_j \leq h\alpha$. Using the fact $Kh \leq \eta$, the term in the bracket is bounded below by

$$\int_0^1 \rho(\theta\eta + x_{1/2}) d\theta - (\lambda + C) \eta^2 \alpha^2 / K^2,$$

which is positive if

$$K > \frac{\alpha \sqrt{\lambda + C} \eta}{\sqrt{\int_0^1 \rho(\theta\eta + x_{1/2}) d\theta}}.$$

This can be ensured if we take

$$K = \lfloor A \rfloor + 1,$$

where for $\Omega = [a, b]$,

$$A = \max_{z \in [h_1, b-a]} \frac{\alpha \sqrt{\lambda + C} z}{\sqrt{\int_0^1 \rho(\theta z + x_{1/2}) d\theta}}$$

which is bounded and depends on h_1 . For general cases a different bound can be identified and it may depend on local meshes. $\square$

Note that our numerical solutions feature the following property: if $\rho_j^n = 0$, then $\rho_j^{n+1} = 2\rho_j^* - \rho_j^n \geq 0$ due to the fact that $\rho_j^* \geq 0$ for all $j = 1, \dots, N$. This means that if $\rho_0(x) = 0$ on an interval, then ρ_j^1 cannot be negative in most of nearby cells. Thus negative values appear only where the exact solution turns from zero to a positive value, and the number of these values are finitely many. Our result in Theorem 5.2 is thus applicable.

Algorithm. We have the following algorithm:

- (1) Initialization: From initial data $\rho_0(x)$, obtain $\rho_j^0 = \frac{1}{h_j} \int_{I_j} \rho_0(x) dx$, $j = 1, \dots, N$, by using a second order quadrature.
- (2) Update to get $\{\rho_j^1\}$ by the first order scheme (3.1).
- (3) Marching from $\{\rho_j^n\}$ to $\{\rho_j^{n+1}\}$ for $n = 1, 2, \dots$, based on (5.1).
- (4) Reconstruction: if necessary, locally replace ρ_j^{n+1} by $\tilde{\rho}_j^{n+1}$ using the limiter defined in (5.4).

The following algorithm can be called to find an admissible set S_k used in (5.4).

- (i) Start with $S_k = \{k\}$, $m = 1$.
- (ii) If $k - m \geq 1$ and $c_{k-m} \neq 0$, then set $S_k = S_k \cup \{k - m\}$.
If $\bar{c}_k > 0$, then stop, else go to (iii).
- (iii) If $k + m \leq N$ and $c_{k+m} \neq 0$, then set $S_k = S_k \cup \{k + m\}$.
If $\bar{c}_k > 0$, then stop, else set $m = m + 1$ and go to (ii).

5.3. Second order scheme for 2D problem. A similar two step time-discretization technique can be applied to higher dimensional problems. In the 2D case, that with scheme (4.2) gives the following fully discrete scheme,

$$\frac{\rho_{i,j}^* - \rho_{i,j}^n}{\tau/2} = \frac{C_{i+1/2,j}^* - C_{i-1/2,j}^*}{h_i^x} + \frac{C_{i,j+1/2}^* - C_{i,j-1/2}^*}{h_j^y}, \quad (5.6a)$$

$$\rho_{i,j}^{n+1} = 2\rho_{i,j}^* - \rho_{i,j}^n, \quad (5.6b)$$

where

$$C_{i+1/2,j}^* = \frac{M_{i+1/2,j}^*}{h_{i+1/2}^x} \left(\frac{\rho_{i+1,j}^*}{M_{i+1,j}^*} - \frac{\rho_{i,j}^*}{M_{i,j}^*} \right), \quad i = 1, \dots, N_x - 1, j = 1, \dots, N_y,$$

$$C_{i,j+1/2}^* = \frac{M_{i,j+1/2}^*}{h_{j+1/2}^y} \left(\frac{\rho_{i,j+1}^*}{M_{i,j+1}^*} - \frac{\rho_{i,j}^*}{M_{i,j}^*} \right), \quad i = 1, \dots, N_x, j = 1, \dots, N_y - 1,$$

$$C_{1/2,j}^* = C_{N_x+1/2,j}^* = C_{i,1/2}^* = C_{i,N_y+1/2}^* = 0, \quad i = 1, \dots, N_x, j = 1, \dots, N_y,$$

with $M_{i+1/2,j}^* = Q_2(x_{i+1/2}, y_j, \frac{3}{2}\rho^n - \frac{1}{2}\rho^{n-1})$, $M_{i,j+1/2}^* = Q_2(x_i, y_{j+1/2}, \frac{3}{2}\rho^n - \frac{1}{2}\rho^{n-1})$, and $M_{i,j}^* = Q_2(x_i, y_j, \frac{3}{2}\rho^n - \frac{1}{2}\rho^{n-1})$. In an entirely similar fashion (details are therefore omitted), we can prove the following.

Theorem 5.3. The fully discrete scheme (5.6) has the following properties:

- (1) Conservation of mass:

$$\sum_{i=1}^{N_x} \sum_{j=1}^{N_y} h_i^x h_j^y \rho_{i,j}^n = \int_{\Omega} \rho_0(x, y) dx dy, \quad \text{for } n \geq 1.$$

- (2) Positivity preserving: if $\rho_{i,j}^n \geq 0$ for all $i \in \{1, \dots, N_x\}$ and $j \in \{1, \dots, N_y\}$, then

$$\rho_{i,j}^{n+1} \geq 0,$$

provided τ is sufficiently small.

5.4. Local limiter and algorithm. If the time step τ is not small, positivity of $\rho_{i,j}^n$ is not guaranteed for $n \geq 2$. We use the following limiter to resolve this issue:

$$\tilde{\rho}_{i,j} = \theta \rho_{i,j} + (1 - \theta) \frac{\bar{c}_{k,l}}{h_i^x h_j^y}, \quad (5.7)$$

with

$$\theta = \min \left\{ 1, \frac{\bar{c}_{k,l}}{\bar{c}_{k,l} - c_{min}} \right\}, \quad c_{min} = \min_{(i,j) \in S_{k,l}} h_i^x h_j^y \rho_{i,j}, \quad \bar{c}_{k,l} = \frac{1}{|S_{k,l}|} \sum_{(i,j) \in S_{k,l}} h_i^x h_j^y \rho_{i,j},$$

where $S_{k,l}$ denotes the minimum number of indexes for which $\rho_{i,j} \neq 0$ and $\bar{c}_{k,l} > 0$.

The limiter (5.7) can be shown to be nonnegative and satisfy the local mass conservation. In addition, for any $g_{i,j} \geq 0$ we have

$$|\tilde{\rho}_{i,j} - g_{i,j}| \leq (1 + |S_{k,l}|\alpha) \max_{(i,j) \in S_{k,l}} |\rho_{i,j} - g_{i,j}|, \quad (i,j) \in S_{k,l},$$

where α is the upper bound of 2D mesh ratios. Hence the second order accuracy remains for shape-regular meshes since $|S_{k,l}|$ can be shown bounded as in the one-dimensional case.

Algorithm Our algorithm for 2D problem is given as follows:

- (1) Initialization: From initial data $\rho_0(x, y)$, obtain $\rho_{i,j}^0 = \frac{1}{I_{i,j}} \int_{I_{i,j}} \rho_0(x, y) dx dy$, $i = 1, \dots, N_x$, $j = 1, \dots, N_y$, by using a second order quadrature.
- (2) Update to get $\{\rho_{i,j}^1\}$ by the first order scheme (4.4).
- (3) March from $\{\rho_{i,j}^n\}$ to $\{\rho_{i,j}^{n+1}\}$ based on the scheme (5.6).
- (4) Reconstruction: if necessary, locally replace $\rho_{i,j}^{n+1}$ by $\tilde{\rho}_{i,j}^{n+1}$ using the limiter defined in (5.7).

The following algorithm can be called to find an admissible set $S_{k,l}$ used in (5.7).

- (i) Start with $S_{k,l} = \{(k, l)\}$, $m = 1$.
- (ii) For $d_y = \max\{1, l - m\} : \min\{l + m, N_y\}$ and $d_x = \max\{1, k - m\} : \min\{k + m, N_x\}$,
If $(d_x, d_y) \notin S$ and $c_{k-m} \neq 0$, then set $S_{k,l} = S_{k,l} \cup \{(d_x, d_y)\}$.
If $\bar{c}_{k,l} > 0$, then stop, else go to (iii).
- (iii) Set $m = m + 1$ and go to (ii).

6. NUMERICAL EXAMPLES

In this section, we implement the fully discrete schemes (3.1) and (4.4) and second order extensions (5.1) and (5.6). Errors in 1-D case are measured in the following discrete norms:

$$e_{l^1} = h \sum_{i=1}^N |\rho_i^n - \bar{\rho}_i^n|,$$

$$e_{l^\infty} = \max_{1 \leq i \leq N} |\rho_i^n - \bar{\rho}_i^n|.$$

Here $\bar{\rho}_i^n$ is cell average of the exact solution on I_i at time $t = n\tau$.

6.1. One-dimensional tests.

Example 6.1. (Accuracy test) In this example we test the accuracy of scheme (3.1) and scheme (5.1) Consider the initial value problem with source term

$$\begin{cases} \partial_t \rho = \partial_x(\partial_x \rho + \rho \partial_x(V(x) + W * \rho)) + F(x, t), & t > 0, x \in [-\pi, \pi], \\ \rho(x, 0) = 2 + \cos(x), & x \in [-\pi, \pi], \end{cases} \quad (6.1)$$

subject to zero flux boundary conditions. Here we take $V(x) = \cos(x)$, $W(x) = \cos(x)$, and

$$F(x, t) = \pi e^{-2t}(2 \cos^2(x) + 2 \cos(x) - 1) + e^{-t}(2 \cos^2(x) + 2 \cos(x) - 3).$$

One can check that the exact solution to (6.1) is

$$\rho(x, t) = e^{-t}(2 + \cos(x)).$$

We compute to $t = 1$, first use time step $\tau = 0.1h$ and $\tau = h^2$ to check accuracy of scheme (3.1), then use $\tau = h$ to check accuracy of scheme (5.1), results are reported in Table 1 and Table 2 respectively. We see that the scheme (3.1) is first order accurate in time and second order accurate in space, while the scheme (5.1) is second order accurate both in time and space.

Note that the exact solution is $\rho(x, t) = e^{-t}(2 + \cos(x))$, which is far above 0 for $t \in [0, 1]$. Hence the positivity-preserving limiter is not activated in this test.

TABLE 1. Accuracy of scheme (3.1) with $\tau = 0.1h$ and $\tau = h^2$.

N	errors and orders with $\tau = 0.1h$				errors and orders with $\tau = h^2$			
	l^1 error	order	l^∞ error	order	l^1 error	order	l^∞ error	order
40	0.70474E-01	-	0.26268E-01	-	0.10451E-00	-	0.46075E-01	-
80	0.32212E-01	1.1295	0.15021E-01	0.8063	0.25847E-01	2.0156	0.11397E-01	2.0153
160	0.15796E-01	1.0280	0.79593E-02	0.9163	0.64441E-02	2.0039	0.28433E-02	2.0030
320	0.78955E-02	1.0005	0.40881E-02	0.9612	0.16098E-02	2.0011	0.71027E-03	2.0011

TABLE 2. Accuracy of scheme (5.1) with $\tau = h$.

N	l^1 error	order	l^∞ error	order
40	0.14049E-00	-	0.43022E-01	-
80	0.35941E-01	1.9668	0.10729E-01	2.0036
160	0.90784E-02	1.9851	0.26805E-02	2.0009
320	0.22814E-02	1.9925	0.67108E-03	1.9980

Example 6.2. In this example, we study dynamics of linear Fokker-Plank equations by considering the following problem

$$\partial_t \rho = \partial_x (\partial_x \rho + x \rho), \quad t > 0, \quad x \in [-5, 5], \quad (6.2)$$

with initial condition

$$\rho(x, 0) = \begin{cases} \frac{1}{7} \int_{\Omega} e^{-\frac{x^2}{2}} dx, & x \in [-3.5, 3.5], \\ 0, & \text{otherwise,} \end{cases} \quad (6.3)$$

and zero flux boundary conditions $(\partial_x \rho + x \rho)|_{x=\pm 5} = 0$.

This is (2.1) with $V(x) = \frac{x^2}{2}$ and $W(x) = 0$. The steady state to (6.2) is $\rho_{eq}(x) = e^{-\frac{x^2}{2}}$. We use the time step $\tau = 0.1$ to compute solutions up to $t = 4$, with $N = 200$. In Fig.1(a) are snap shots of ρ at $t = 0, 0.2, 0.5, 1, 4$, and the steady state. Fig.1(b) shows the mass conservation and energy decay. We observe from this figure that the solution of problem (6.2) becomes indistinguishable from the steady state after $t = 2$. Compared in Fig.2 are numerical solutions obtained by the second order scheme (5.1) with and without the local limiter. We see that the limiter produces positive solutions and reduces solution oscillations.

FIGURE 1. First order scheme for Example 6.2.

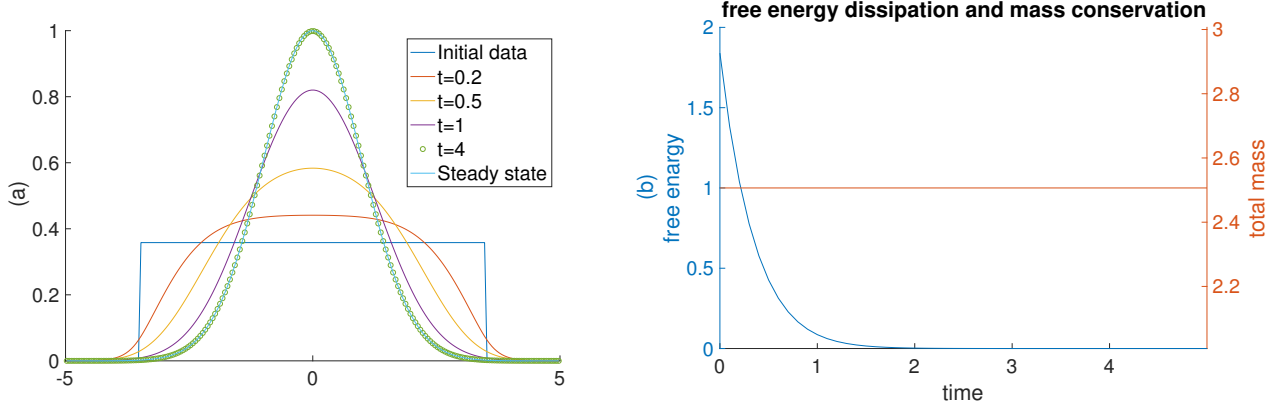
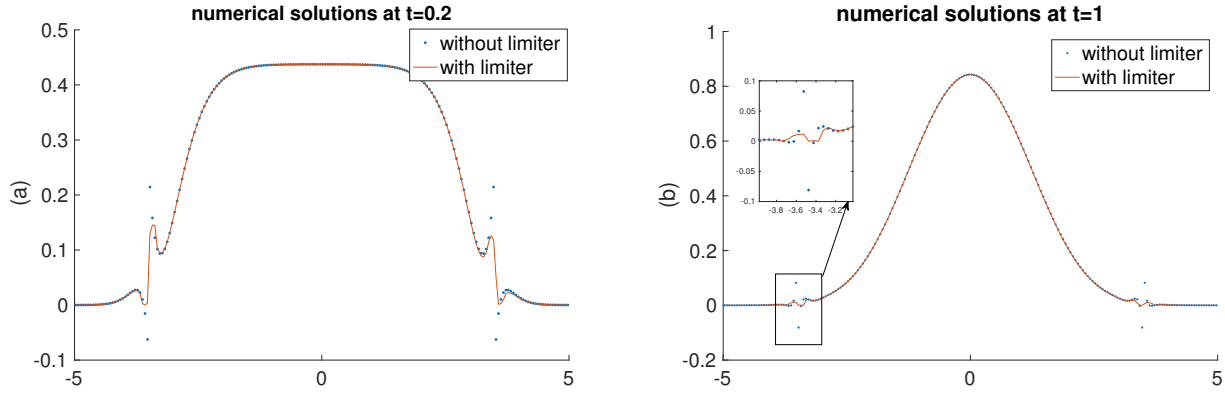


FIGURE 2. Second order scheme (with and without limiter) for Example 6.2.



Example 6.3. (Doi-Onsager equation with the Maier-Saupe potential) In this example, we consider the Doi-Onsager equation with Maier-Saupe potential

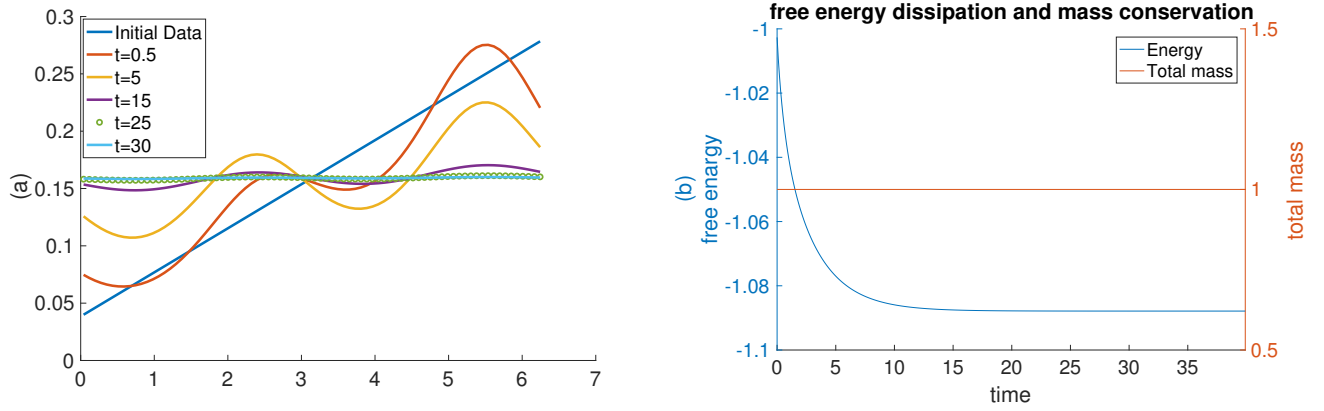
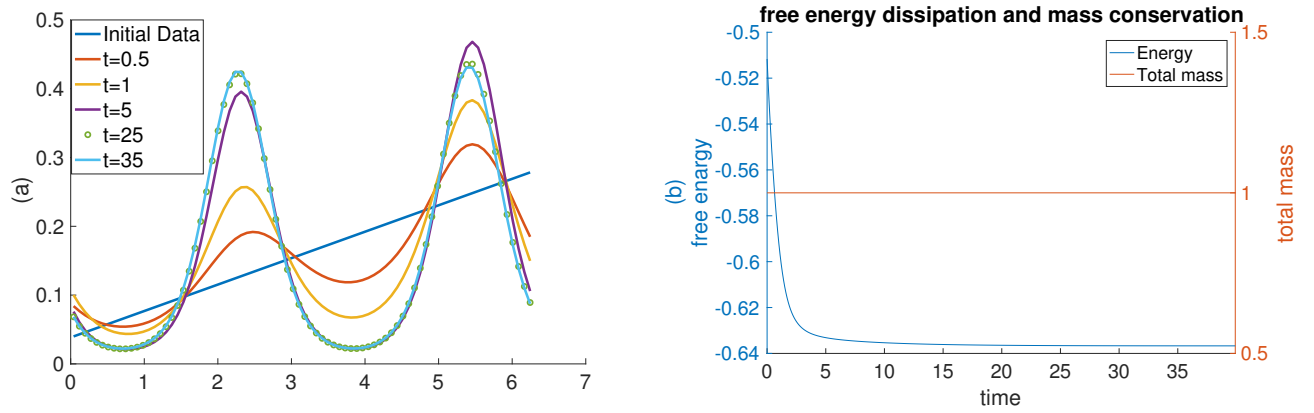
$$\begin{cases} \partial_t \rho = \partial_x (\partial_x \rho + \alpha \rho \partial_x (W * \rho)), & W(x) = \sin^2(x) \quad t > 0, \quad x \in [0, 2\pi] \\ \rho(x, 0) = \frac{x+1}{2\pi(\pi+1)}, \end{cases} \quad (6.4)$$

subject to zero flux boundary conditions. Here α is the intensity parameter. Stationary solutions of (6.4) have been an interesting subject of study, since when α increases, phase transition from isotropic state to nematic state will appear. A detailed characterization of solutions can be found in [30]: for $0 < \alpha \leq \alpha^* = 4$, the only stationary solution is the isotropic state $\rho_{eq}(x) = \frac{1}{2\pi}$. When $\alpha > \alpha^*$ besides the constant solution $\rho_{eq}(x) = \frac{1}{2\pi}$, there are other solutions given by

$$\rho_{eq}(x) = \frac{e^{-\eta^* \cos 2(x-x_0)}}{\int_0^{2\pi} e^{-\eta^* \cos(2x)} dx},$$

where x_0 is arbitrary, $\eta^* > \frac{\alpha}{2} \sqrt{1 - 4/\alpha}$ is uniquely determined by

$$\frac{\int_0^{2\pi} \cos(2x) e^{-\eta^* \cos(2x)} dx}{\int_0^{2\pi} e^{-\eta^* \cos(2x)} dx} + \frac{2\eta^*}{\alpha} = 0.$$

FIGURE 3. Solution evolution and energy dissipation for Example (6.3) with $\alpha = 3$.FIGURE 4. Solution evolution and energy dissipation for Example (6.3) with $\alpha = 5$.

We use scheme (3.1) and choose the time step $\tau = 0.1$ to compute up to $T = 30$ with $N = 80$. In Fig. 3(a) are snap shots of solutions to (6.4) for $\alpha = 3 < \alpha^*$ at $t = 0, 0.5, 5, 15, 25, 30$. Fig.3 (b) shows mass conservation and energy decay, from which we can observe that the problem (6.4) is already at steady state $\rho_{eq}(x) = \frac{1}{2\pi}$ after $t = 20$. In Fig. 4(a) are snap shots of solutions to (6.4) for $\alpha = 5 > \alpha^*$ at $t = 0, 0.5, 1, 5, 25, 35$. Fig.4 (b) shows mass conservation and energy decay, which tells that problem (6.4) is at already steady state after $t = 30$. Our method gives satisfying results for the problem, consistent with the numerical results obtained in [12] by an explicit scheme with Euler forward time discretization.

6.2. Two-dimensional tests.

Example 6.4. (Accuracy test) We consider the initial value problem with source term,

$$\begin{cases} \partial_t \rho = \nabla \cdot (\nabla \rho + \rho \nabla V(x, y)) + F(x, y, t), & t > 0, (x, y) \in [-\frac{\pi}{2}, \frac{\pi}{2}] \times [-\frac{\pi}{2}, \frac{\pi}{2}], \\ \rho(x, y, 0) = 2 + \sin(x) \sin(y), & (x, y) \in [-\frac{\pi}{2}, \frac{\pi}{2}] \times [-\frac{\pi}{2}, \frac{\pi}{2}], \end{cases} \quad (6.5)$$

subject to zero flux boundary conditions, here $V(x, y) = \sin(x) \sin(y)$, and

$$F(x, y, t) = e^{-t}(2 \sin^2(x) \sin^2(y) + 5 \sin(x) \sin(y) - \cos^2(x) \sin^2(y) - \sin^2(x) \cos^2(y) - 2).$$

This problem has the exact solution

$$\rho(x, t) = e^{-t}(2 + \sin(x) \sin(y)).$$

We choose $\tau = 0.1h^2$ in scheme (4.4) and $\tau = 0.1h$ in scheme (5.6). Errors and orders at $t = 1$ are listed in Table 3, in this test uniform meshes with $h = h^x = h^y = \pi/N$ have been used.

TABLE 3. Accuracy of scheme (4.4) and (5.6).

	scheme (4.4) with $\tau = 0.1h^2$				scheme (5.6) with $\tau = 0.1h$			
$N \times N$	l^1 error	order	l^∞ error	order	l^1 error	order	l^∞ error	order
10×10	0.927816E-1	-	0.175767E-1	-	0.31090E-01	-	0.84728E-02	-
20×20	0.232384E-1	1.997	0.446660E-2	1.976	0.77577E-02	2.003	0.22012E-02	1.945
40×40	0.581196E-2	1.999	0.112137E-2	1.994	0.19368E-02	2.002	0.55550E-03	1.986
80×80	0.145297E-2	2.000	0.280607E-3	1.999	0.48558E-03	1.996	0.13975E-03	1.991

Finally we mention that there is a class of equations in which the interaction is modeled through a potential governed by the Poisson equation. The celebrated model is the Patlak-Keller-Segel system of the chemotaxis [18, 19]. The original model is a coupled parabolic system, and the one related to our model equation (1.1) is the parabolic-elliptic version of the form (see e.g., [33])

$$\begin{cases} \partial_t \rho = \Delta \rho - \nabla \cdot (\chi \rho \nabla c), & t > 0, \mathbf{x} \in \mathbb{R}^2, \\ -\Delta c = \rho, \\ \rho(\mathbf{x}, 0) = \rho_0(\mathbf{x}), & \mathbf{x} \in \mathbb{R}^2. \end{cases} \quad (6.6)$$

Here, $\rho(\mathbf{x}, t)$ is the cell density, $c(\mathbf{x}, t)$ is the chemical attractant concentration, the parameter $\chi > 0$ is the sensitivity of bacteria to the chemical attractant. It has been shown in [2] that the solution behavior of problem (6.6) is quite different when crossing a critical mass. If the initial mass $M = \int_{\mathbb{R}^2} \rho_0(x, y) dx dy$ is smaller than a critical value $M_c = 8\pi/\chi$, then the solution exists globally. When $M > M_c$, the solution will blow up in finite time, which is referred to as chemotactic collapse.

Example 6.5. (Patlak–Keller–Segal system). In this example, we test the method's capacity in capturing solution concentrations for the Patlak–Keller–Segal system (6.6). Using the Green function for the Poisson equation, this system can be reformulated as (1.1) with $V = 0$ and

$$W(x, y) = \frac{\chi}{2\pi} \log(\sqrt{x^2 + y^2}). \quad (6.7)$$

In our simulation, we restrict to a bounded domain Ω subject to zero flux boundary conditions, using formulation (4.1) with $V(x, y) = 0$ and W defined in (6.7). We fix $\chi = 1$ and consider both the sub-critical case with

$$\rho_0(x, y) = \begin{cases} 2(\pi - 0.2), & (x, y) \in [-1, 1] \times [-1, 1], \\ 0, & (x, y) \in \Omega \setminus [-1, 1] \times [-1, 1], \end{cases}$$

on $\Omega = [-5, 5] \times [-5, 5]$, and super-critical case with

$$\rho_0(x, y) = \begin{cases} 2(\pi + 0.2), & (x, y) \in [-1, 1] \times [-1, 1], \\ 0, & (x, y) \in \Omega \setminus [-1, 1] \times [-1, 1], \end{cases}$$

on $\Omega = [-1.5, 1.5] \times [-1.5, 1.5]$, for which we know that the solution blows-up at finite time.

We take time step $\tau = 0.01$, and set $N_x = N_y = 51$ so that a single cell is located at the center of the computational domain, where one can view a clear picture of the blow-up phenomena in super-critical case. In Fig.5 are snap shots of numerical solutions in the sub-critical case at $t = 0, 2, 8, 12, 16$, from which we observe that the numerical solution dissipates in time, the last picture in Fig.5 shows mass conservation and energy dissipation. In Fig.6 are snap shots of numerical solutions in super-critical case at $t = 0, 0.5, 1, 1.5, 2$, we observe that numerical solutions tend to concentrate at the origin.

Let us remark that in [42] the same concentration phenomena was observed, using a DG method for this problem with periodic boundary conditions. Different boundary conditions do not affect the concentration profile since the solution is compactly supported in our setting. In the super-critical case, the peak in our result is slightly lower than that captured in [42], this is expected because the solution is concentrated at a single point, and cell averaging near the origin can decrease the height of the peak.

7. CONCLUDING REMARKS

In this paper, we have developed positive and free energy satisfying schemes for diffusion equations with interaction potentials; since such equations are governed by a free energy dissipation law and are featured with non-negative solutions. Based on the non-logarithmic Landau reformulation of the model, we constructed a simple, easy-to-implement fully discrete numerical scheme (first order in time) which proved to satisfy all three desired properties of the continuous model: mass conservation, free energy dissipation and non-negativity, without a strict time step restriction. For a fully second order (in both time and space) scheme, we used a local scaling limiter to restore solution positivity when necessary. Moreover, we rigorously proved that the limiter does not destroy the second order accuracy. Numerical examples have demonstrated the superior performance of these schemes, in particular, the three solution properties numerically confirmed are consistent with our theoretical findings.

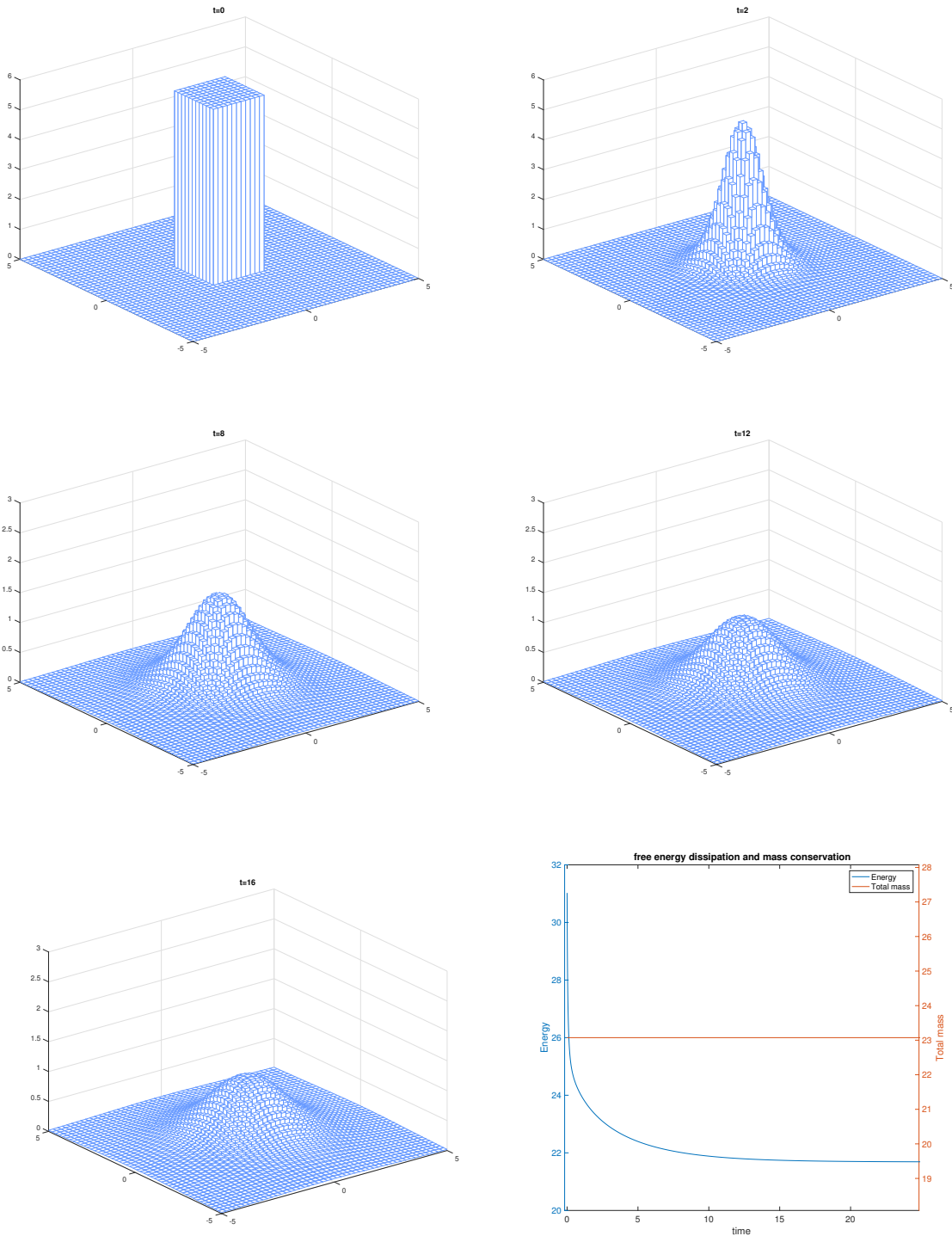
ACKNOWLEDGMENTS

This research was supported by the National Science Foundation under Grant DMS1312636.

REFERENCES

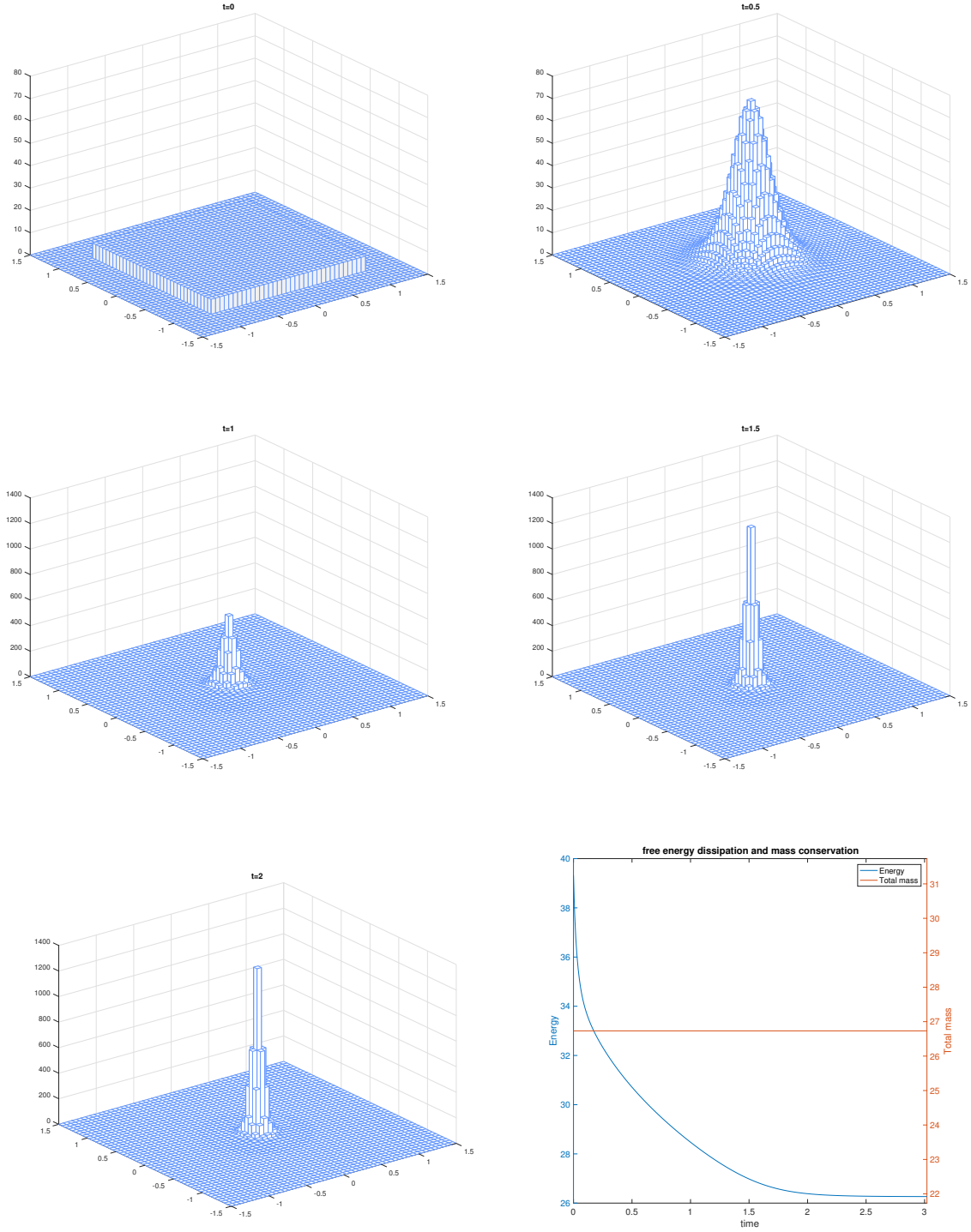
- [1] L. Ambrosio, N. Gigli and G. Savaré. Gradient flows in metric spaces and in the space of probability measures. *Lect. Math., ETH Zürich. Birkhäuser Verlag, Basel*, 2005.
- [2] A. Blanchet, J. Dolbeault and B. Perthame. Two-dimensional Keller-Segel model: optimal critical mass and qualitative properties of the solutions. *Electron.J. Differ. Equ.*, 44:1–32, 2006.

FIGURE 5. Solution evolution for Example 6.5 (sub-critical).



- [3] Y. Brenier. Polar factorization and monotone rearrangement of vector-valued functions. *Comm. Pure Appl. Math.*, 44:375–417, 1991.
- [4] C. Buet and S. Dellacheris. On the Chang and Cooper scheme applied to a linear Fokker–Planck equation. *Commun. Math. Sci.*, 8:1079–1090, 2010.

FIGURE 6. Solution evolution for Example 6.5 (super-critical).



- [5] J.-D. Benamou, G. Carlier, and M. Laborde. An augmented Lagrangian approach to Wasserstein gradient flows and applications. *ESAIM: Proceedings and surveys*, 54:1–17, 2016.
- [6] J.A. Carrillo, A. Jüngel, P.A. Markowich, G. Toscani and A. Unterreiter. Entropy dissipation methods for degenerate parabolic problems and generalized Sobolev inequalities, *Monatshefte Math.*, 133(1): 1–82, 2001.

- [7] J.A. Carrillo, R.J. McCann and C. Villani, Kinetic equilibration rates for granular media and related equations: entropy dissipation and mass transportation estimates, *Rev. Mat. Iberoam.*, 19 (3): 971–1018, 2003.
- [8] J.A. Carrillo, A. Chertock and Y. Huang. A finite-volume method for nonlinear nonlocal equations with a gradient flow structure. *Commun. Computat. Phys.*, 17(1): 233–258, 2015.
- [9] J. S. Chang and G. Cooper. A practical difference scheme for Fokker-Planck equations. *J. Comput. Phys.*, 6(1):1–16, 1970.
- [10] J. A Carrillo, K. Craig, L. Wang, and C.-Z. Wei. Primal dual methods for Wasserstein gradient flows. arXiv preprint arXiv:1901.08081, 2019.
- [11] K. Craig and A. Bertozzi. A blob method for the aggregation equation. *Math. Comp.*, 85:1681–1717, 2016.
- [12] T.J. Chenhall. On the Doi-Onsager model of rigid rod-like polymers, *Masters thesis Iowa State University*, advisor: Hailiang Liu, 2016.
- [13] P. Constantin, I. Kevrekidis and E. S. Titi. Asymptotic states of a Smoluchowski equation. *Arch. Rat. Mech. Anal.*, 174: 365–384, 2004.
- [14] M. Doi and S. F. Edwards. The Theory of Polymer Dynamics. *Oxford University Press*, 1986.
- [15] M. Cuturi and G. Peyre. Semidual regularized optimal transport. *SIAM Review*, 60(4): 941–965, 2018.
- [16] W. Gangbo and R. J. McCann. Optimal maps in Monge’s mass transport problems. *C. R. Acad. Sci. Paris*, 321: 1653–1658, 1995.
- [17] D. Grünbaum and A. Okubo, Modeling social animal aggregations, In: *S. A. Levin (Ed.), Frontiers of Theoretical Biology. Vol. 100 of Lecture Notes in Biomathematics. Springer-Verlag*, 1994.
- [18] D. Horstmann. From 1970 until now: The Keller-Segel model in chemotaxis and its consequences I. *Jahresber. DMV*, 105: 103–165, 2003.
- [19] D. Horstmann. From 1970 until now: The Keller-Segel model in chemotaxis and its consequences II. *Jahresber. DMV*, 106: 51–69, 2004.
- [20] R. Jordan, D. Kinderlehrer, and F. Otto. The variational formulation of the Fokker-Planck equation. *SIAM J. Math. Anal.* 29(1):1–17, 1998.
- [21] K. Kawasaki. Diffusion and the formation of spatial distributions. *Math. Sci.*, 16 (183): 47–52, 1978.
- [22] E. F. Keller and L. A. Segel. Initiation of slime mold aggregation viewed as an instability. *J. Theor. Biol.*, 26: 399–415, 1970.
- [23] H. Liu and W. Maimaitiyiming. Unconditional positivity-preserving and energy stable schemes for a reduced Poisson-Nernst-Planck system. *Commun. Comput. Phys.*, (in press) 2019.
- [24] H. Liu and W. Maimaitiyiming. A second order positive scheme for the reduced Poisson-Nernst-Planck system. *J. Comp. Appl. Math.*, (submitted) 2019.
- [25] H. Liu and H. Yu. An entropy satisfying conservative method for the Fokker-Planck equation of the finitely extensible nonlinear elastic dumbbell model. *SIAM J. Numer. Anal.*, 50(3): 1207–1239, 2012.
- [26] H. Liu and H. Yu. Maximum-principle-satisfying third order discontinuous Galerkin schemes for Fokker-Planck equations. *SIAM J. Sci. Comput.*, 36(5): 2296–2325, 2014.
- [27] H. Liu and H. Yu. The entropy-satisfying discontinuous Galerkin method for Fokker-Planck equations. *J. Sci. Comput.*, 62: 803–830, 2015.
- [28] H. Liu and Z. Wang. A free energy satisfying finite difference method for Poisson-Nernst-Planck equations. *J. Comput. Phys.*, 268: 363–376, 2014.
- [29] H. Liu and Z. Wang. A free energy satisfying discontinuous Galerkin method for one-dimensional Poisson-Nernst-Planck systems. *J. Comput. Phys.*, 328: 413–437, 2017.
- [30] H. Liu, H. Zhang and P. Zhang, Axial symmetry and classification of stationary solutions of Doi-onsager equation on the sphere with Maier-Saupe potential. *Comm. Math. Sci.*, 3(2): 201–218, 2005.
- [31] W.-C. Li, J.-F. Lu, and L. Wang. Fisher information regularization schemes for Wasserstein gradient flows. *arXiv preprint*: 1907.02152v2, 2019.
- [32] R. J. McCann, A convexity principle for interacting gases. *Adv. Math.*, 128: 153–179, 1997.

- [33] V. Nanjundiah. Chemotaxis, signal relaying and aggregation morphology. *Journal of Theoretical Biology.*, 42: 63–105, 1973.
- [34] L. Onsager. The effects of shape on the interaction of colloidal particles. *Ann. N. Y. Acad. Sci.*, 51: 627–659, 1949.
- [35] F. Otto. The geometry of dissipative evolution equations: the porous medium equation. *Comm. Partial Differential Equations*, 26(1-2):101–174, 2001.
- [36] M. Campos-Pinto, J. A. Carrillo, F. Charles, and Y.-P. Choi. Convergence of a linearly transformed particle method for aggregation equations. *Numer. Math.*, 139:743–793, 2018.
- [37] L. Pareschi and M. Zanella. Structure preserving schemes for nonlinear Fokker–Planck equations and applications. *J. Sci. Comput.*, 74(3):1575–1600, 2018.
- [38] C.S. Patlak. Random walk with persistence and external bias. *The Bulletin of Mathematical Biophysics*, 15(3):311–338, 1953
- [39] B. Perthame. Transport Equations in Biology. *Frontiers in Mathematics*, Birkhauser Verlag, Basel, 2007.
- [40] G. Peyré. Entropic approximation of Wasserstein gradient flows. *SIAM Journal on Imaging Sciences*, 8(4):2323–2351, 2015.
- [41] H. Risken. The Fokker-Planck Equation: Methods of Solution and Applications. *Second edition, Springer Series in Synergetics*, vol. 18, Springer-Verlag, Berlin, 1989.
- [42] Z. Sun, J.A. Carrillo and C.-W. Shu. A discontinuous Galerkin method for nonlinear parabolic equations and gradient flow problems with interaction potentials. *J. Comput. Phys.*, 352: 76–104, 2018.
- [43] C. M. Topaz, A. L. Bertozzi and M. A. Lewis. A nonlocal continuum model for biological aggregation. *Bull. Math. Bio.*, 68: 1601-1623, 2006.
- [44] C. Villani. Topics in Optimal Transportation, *American Mathematical Society*, 2003.

IOWA STATE UNIVERSITY, MATHEMATICS DEPARTMENT, AMES, IA 50011

E-mail address: hliu@iastate.edu

E-mail address: wumaierm@iastate.edu