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# ASYMPTOTIC-PRESERVING SCHEME FOR THE RESOLUTION OF EVOLUTION EQUATIONS WITH STIFF TRANSPORT TERMS

BAPTISTE FEDELE <sup>\*</sup>, CLAUDIA NEGULESCU <sup>\*</sup>, AND STEFAN POSSANNER <sup>†</sup>

**Abstract.** We develop an asymptotic-preserving scheme to solve evolution problems containing stiff transport terms. This scheme is based on a micro-macro decomposition of the unknown, coupled with a stabilization procedure. The numerical method is applied to the Vlasov equation in the gyrokinetic regime and to the Vlasov-Poisson 1D1V equation, models occurring in plasma physics. The asymptotic-preserving properties of our procedure permit to study the long-time behavior of these models. Indeed, classical numerical approaches have to cope with large numerical errors in such time asymptotics, whereas our AP-procedure permits to limit this drawback.

**Key words.** Plasma physics, kinetic equations, Vlasov-Poisson system, asymptotic analysis, asymptotic-preserving schemes, BGK equilibria.

**AMS subject classifications.** 35Q83, 65M06, 65F05, 35B40.

**1. Introduction.** The main objective of this work is to introduce and subsequently investigate an efficient numerical scheme for the resolution of evolution equations containing stiff transport terms, namely

$$\partial_t f^\epsilon + \mathcal{L}f^\epsilon + \frac{\mathbf{b}}{\epsilon} \cdot \nabla f^\epsilon = 0, \quad t \in \mathbb{R}^+, \quad \mathbf{x} \in \Omega \subset \mathbb{R}^d, \quad (1.1)$$

where  $\mathbf{b} : \mathbb{R}^+ \times \Omega \rightarrow \mathbb{R}^d$  is a known (passive, linear transport model) or self-consistently computed (active, nonlinear transport model) vector-field satisfying  $\nabla \cdot \mathbf{b} = 0$ , and  $\mathcal{L}$  is a given operator (for ex. transport or diffusion operator). The small parameter  $\epsilon \ll 1$  represents the stiffness of the problem and signifies that we have to cope with a very strong vector-field  $\mathbf{b}$ . It brings up the main difficulties in the numerical resolution of (1.1), due to the introduction of multiple scales in the problem. Indeed, the dynamics along the  $\mathbf{b}$ -field is very rapid, as compared to its perpendicular evolution. In the formal limit  $\epsilon \rightarrow 0$ , the problem reduces to the constraint

$$\mathbf{b} \cdot \nabla f^0 = 0, \quad (1.2)$$

which signifies that the unknown  $f^0$  is constant along the field-lines of  $\mathbf{b}$ . However, in general, problem (1.2) does not permit to determine these different constants, for example when  $\mathbf{b}$  has closed field lines in the considered domain. Thus the reduced problem (1.2) is ill-posed, information has been lost while setting formally  $\epsilon = 0$  in (1.1). This feature is typical for singularly-perturbed problems or multi-scale problems (see [7, 36]).

The study of multi-scale problems is challenging from a mathematical as well a numerical point of view. Standard explicit numerical schemes require very small time steps, dependent on the  $\epsilon$ -parameter, in order to accurately account for the microscopic information (living at the  $\epsilon$ -scale). This procedure, even if accurate, has however the big disadvantage of being numerically very costly in simulation time and memory. Fully implicit schemes or IMEX-schemes are also not of use for  $\epsilon \ll 1$ , due to the ill-conditioned reduced model (1.2). Alternative methodologies are thus required, taking into account for the various scales present in the problem. Asymptotic analysis will be one of the mathematical tools used in this paper, permitting to recover the microscopic information lost in the reduced model (1.2). The numerical scheme presented here is based on such developments.

Evolution equations of the type (1.1) arise often in applications coming from fluid dynamics (see [31]) and plasma physics (see [9, 24, 26]). To mention only some examples, in thermonuclear fusion plasmas, the evolution of ions is described via the non-dimensional Vlasov ( $\eta = 0$ ) or Fokker-Planck ( $\eta > 0$ ) equation

$$\partial_t f_i^\epsilon + \mathbf{v} \cdot \nabla_{\mathbf{x}} f_i^\epsilon + \left( \mathbf{E} + \frac{1}{\epsilon} \mathbf{v} \times \mathbf{B} \right) \cdot \nabla_{\mathbf{v}} f_i^\epsilon = \eta \nabla_{\mathbf{v}} \cdot [\mathbf{v} f_i^\epsilon + \nabla_{\mathbf{v}} f_i^\epsilon], \quad (1.3)$$

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where  $f_i^\epsilon(t, \mathbf{x}, \mathbf{v})$  represents the ion distribution, dependent on time, space and velocity. This equation is coupled via the electromagnetic fields  $(\mathbf{E}(t, \mathbf{x}), \mathbf{B}(t, \mathbf{x}))$  to an equation describing the electron evolution. The coupling is done by means of Maxwell's equations or Poisson's equation in the electrostatic case. The magnetic field is strong in tokamak experiments in the aim to confine the plasma and to render the fusion possible. This feature is translated in (1.3) in the magnitude of the scaling parameter  $\epsilon \ll 1$ .

The further example concerns the long-time asymptotic study of the electron 1D1V Vlasov-Poisson system

$$\begin{cases} \partial_t f_e + v \partial_x f_e - E(t, x) \partial_v f_e = 0, & \forall t \in \mathbb{R}^+, \forall (x, v) \in \Omega \subset \mathbb{R}^2 \\ -\partial_{xx} \varphi = 1 - n_e, & n_e(t, x) = \int_{\mathbb{R}} f_e(t, x, v) dv, \quad E = -\partial_x \varphi. \end{cases} \quad (1.4)$$

Introducing the field  $\mathbf{u} := (v, -E(t, x))^t$  and the stream function  $\Psi := \frac{1}{2}|v|^2 - \varphi(t, x)$ , one has  $\mathbf{u} = {}^\perp \nabla \Psi$ , where  ${}^\perp \nabla := (\partial_v, -\partial_x)$ . Considering additionally long-time scales, the Vlasov-Poisson system (1.4) transforms into the nonlinear, coupled system

$$\begin{cases} \partial_t f_e^\epsilon + \frac{\mathbf{u}^\epsilon}{\epsilon} \cdot \nabla_{x,v} f_e^\epsilon = 0, & \forall (t, x, v) \in \mathbb{R}^+ \times \Omega \\ -\Delta_{x,v} \Psi^\epsilon = n_e^\epsilon - 2, & n_e^\epsilon(t, x) = \int_{\mathbb{R}} f_e^\epsilon(t, x, v) dv, \quad \mathbf{u}^\epsilon = {}^\perp \nabla \Psi^\epsilon. \end{cases} \quad (1.5)$$

Finally a last example comes from fluid mechanics : consider the incompressible Euler equations in the long-time scaling, describing a bi-dimensional, inviscid flow with velocity  $\mathbf{u} := (u_1, u_2, 0)$  and pressure  $p$

$$\begin{cases} \epsilon \partial_t \mathbf{u}^\epsilon + (\mathbf{u}^\epsilon \cdot \nabla) \mathbf{u}^\epsilon + \nabla p^\epsilon = 0, & \forall (t, \mathbf{x}) \in \mathbb{R}^+ \times \Omega, \\ \nabla \cdot \mathbf{u}^\epsilon = 0. \end{cases} \quad (1.6)$$

Introducing the vorticity  $\omega^\epsilon := \nabla \times \mathbf{u}^\epsilon$ , the Euler system leads to the following nonlinear, coupled system

$$\begin{cases} \partial_t \omega^\epsilon + \frac{\mathbf{u}^\epsilon}{\epsilon} \cdot \nabla \omega^\epsilon = 0, \\ -\Delta \Psi^\epsilon = \omega^\epsilon, \quad \mathbf{u}^\epsilon = {}^\perp \nabla \Psi^\epsilon, \end{cases} \quad (1.7)$$

constituted of a transport equation for the vorticity, which is self-consistently coupled with a Poisson equation for the determination of the stream-function  $\Psi^\epsilon$ , result of the divergence-free constraint of  $\mathbf{u}^\epsilon$ . When adding in (1.7) a small viscosity term  $\nu \Delta \omega^\epsilon$ ,  $\nu$  being the reciprocal of the Reynolds number, one obtains an equation related to the incompressible Navier-Stokes equations.

The goal of this work is now to develop an efficient, uniformly accurate and stable (w.r.t.  $\epsilon$ ) numerical scheme for the resolution of the following linear, stiff transport problem

$$(V)^\epsilon \begin{cases} \partial_t f^\epsilon + \frac{\mathbf{b}}{\epsilon} \cdot \nabla f^\epsilon = 0, & \forall t \in (0, T), \forall \mathbf{x} = (x, y) \in \Omega \subset \mathbb{R}^2, \\ f^\epsilon(0, \mathbf{x}) = f_{in}(\mathbf{x}) & \forall \mathbf{x} \in \Omega, \end{cases} \quad (1.8)$$

with given, smooth and time-independent vector-field  $\mathbf{b} : \Omega \rightarrow \mathbb{R}^2$ , satisfying  $\nabla \cdot \mathbf{b} = 0$ . This simplified transport equation contains the numerical difficulties arising, due to stiffness, in equation (1.1). Given an efficient numerical algorithm for the resolution of (1.8), the above examples can be managed as follows: the nonlinear coupling can be treated iteratively, as shall be shown in Section 6 for the Vlasov-Poisson test case, and the discretization of the general non-stiff term  $\mathcal{L}f^\epsilon$  of (1.1) can be done via standard schemes suited for this particular operator. The scheme we propose in this paper shall be verified in two test cases: firstly in a simplified version of the gyrokinetic scaling (1.3), containing only the stiff magnetic term  $\frac{1}{\epsilon}(\mathbf{v} \times \mathbf{B}) \cdot \nabla_{\mathbf{v}} f_i^\epsilon$ , and secondly in the long-time asymptotics of the Vlasov-Poisson system (1.5). The third example (1.7) will be the aim of a forthcoming paper [20].

Due to the divergence constraint of  $\mathbf{b}$ , there exists a stream-function  $\Psi$  such that  $\mathbf{b} = {}^\perp \nabla \Psi$ . Using the Poisson-bracket notation for two functions  $\chi, \theta$ , namely

$$\{\chi, \theta\} := \partial_x \chi \partial_y \theta - \partial_y \chi \partial_x \theta = \nabla \chi \cdot {}^\perp \nabla \theta,$$

the transport equation (1.8) can be simply rewritten as

$$(V)^\epsilon \begin{cases} \partial_t f^\epsilon + \frac{1}{\epsilon} \{f^\epsilon, \Psi\} = 0, & t \in (0, T), \mathbf{x} \in \Omega \subset \mathbb{R}^2, \\ f^\epsilon(0, \cdot) = f_{in}, \end{cases} \quad (1.9)$$

and shall be completed with adequate boundary conditions, depending on the shape of the domain  $\Omega$  and on the vector-field  $\mathbf{b}$ . In order to recover the examples presented above, we shall investigate two different cases, resumed in the following Hypothesis.

**Hypothesis A** : *The domain  $\Omega$  will be either the whole  $\mathbb{R}^2$  (simplified case (1.3)) or an infinite strip  $(L_1, L_2) \times \mathbb{R}$  (case (1.5)) of the  $(x, y)$ -plane. In the second case, we shall assume periodic boundary conditions in  $x$  and the field  $\mathbf{b} : \Omega \mapsto \mathbb{R}^2$  is supposed to be also periodic in  $x$ . Furthermore  $\mathbf{b}$  is supposed to be sufficiently smooth, time-independent and divergence-free  $\nabla \cdot \mathbf{b} = 0$ .*

With regard to the numerical solution of (1.8), we aim to:

- design a simple and robust numerical scheme, working on a Cartesian grid;
- design a scheme which enjoys the Asymptotic-Preserving properties (AP-scheme), in the sense that it has to be uniformly stable and accurate w.r.t.  $\epsilon$ ;
- give a detailed explanation why the proposed AP-method behaves better than standard methods (explicit, implicit, IMEX);
- design a scheme which is simply “generalizable” to more dimensions and various advection fields.

Let us underline at this point one important fact. We are interested in designing a scheme working on a Cartesian grid. One can imagine that for stiff problems of the type (1.8) (or more generally (1.1)), it could be better to adapt the coordinate system, choosing field-aligned variables, and transforming thus the problem into an evolution problem with a strong anisotropy aligned with one coordinate axis, which is much simpler to solve (via IMEX schemes for ex., see [19]). However, our aim is to avoid a coordinate transformation and to design a simple scheme based on a Cartesian grid. The advantage is that the numerical treatment becomes simpler, the disadvantage will be mentioned in Section 2, namely the introduction of a second, auxiliary unknown. The scheme we are going to propose is hence an alternative to the existing schemes for such evolution problems with stiff transport terms, and marries at the same time simplicity and Asymptotic-Preserving property.

Several AP-schemes were designed in the last years for various types of problems, including anisotropic elliptic [14, 15] or parabolic [33] equations, Vlasov equations in the hydrodynamic regime [21], drift-diffusion regime [11, 29] or the high-field limit [12, 28], Euler system in the low-Mach regime [16, 17]. Briefly, an AP-scheme is a numerical scheme specially designed for singularly-perturbed problems  $P^\epsilon$ , containing some small parameter  $\epsilon \ll 1$ , and which enjoy the following properties (see commutative diagram 1.1):

- for fixed  $\epsilon > 0$ , the AP-scheme, denoted in this diagram  $P^{\epsilon, h}$ , is a consistent discretization of the continuous problem  $P^\epsilon$ , where  $h$  denotes the grid parameters;
- the stability condition is independent of  $\epsilon$ ;
- for fixed discretization parameters  $h$ , the AP-scheme  $P^{\epsilon, h}$  provides in the limit  $\epsilon \rightarrow 0$  a consistent discretization of the limit problem  $P^0$ .

One can put these schemes in the category of multi-scale numerical methods. At the end, let us also remark here that standard schemes for the resolution of (1.1) exist in literature, based on Galerkin methods [27], IMEX-techniques [3] or spectral methods [22].

The outline of this paper is the following. In Section 2, the asymptotic-preserving reformulation of the singularly-perturbed advection equation (1.8) is detailed. The reformulation is based on a

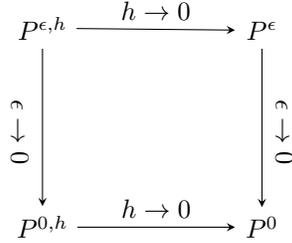


Fig. 1.1: Properties of AP-schemes.

micro-macro decomposition and a stabilization procedure. Section 3 deals with some mathematical aspects of the AP-reformulation, in order to show its well-posedness. Section 4 presents the numerical discretization of our asymptotic-preserving procedure. Section 5 focuses on a mathematical test case and its numerical resolution by our AP-scheme. In particular, we study the stabilization of the numerical scheme. Finally, Section 6 is dedicated to the numerical resolution of the Vlasov-Poisson 1D1V system. We focus notably in this part on the long-time behavior of the two-stream instability, leading to BGK-like equilibria. The last section concludes the paper with some remarks and perspectives.

**2. Asymptotic-Preserving reformulation.** We shall present in this section an AP-reformulation of the singularly-perturbed advection problem (1.8) completed with adequate boundary conditions, explicated in *Hypothesis A*, scheme which shall behave better (regularly) in the limit  $\epsilon \rightarrow 0$ . For this, the well-posed limit-model has firstly to be identified by investigating the asymptotic behaviour of the solutions  $f^\epsilon$ , as  $\epsilon \ll 1$ . We underline here that  $\mathbf{b}$  is time-independent in the following, if not explicitly mentioned, as in Section 6.

**2.1. Identification of the limit model.** As mentioned in the introduction, letting formally  $\epsilon \rightarrow 0$  in (1.8), leads to an ill-posed problem, which does not permit to compute in a unique manner the limit solution  $f^0(t, x, y)$ . The only information we get is that  $f^0$  is constant along the field-lines of  $\mathbf{b}$ .

In order to establish the limit model  $(V)^0$  corresponding to (1.8), let us suppose that  $f^\epsilon$  admits the following Hilbert expansion

$$f^\epsilon = f^0 + \epsilon f^1 + \epsilon^2 f^2 + \dots \quad (2.1)$$

Injecting this Ansatz in (1.8) leads to the infinite hierarchy of equations

$$\mathbf{b} \cdot \nabla f^0 = 0, \quad (2.2)$$

$$\partial_t f^0 + \mathbf{b} \cdot \nabla f^1 = 0, \quad (2.3)$$

$$\partial_t f^1 + \mathbf{b} \cdot \nabla f^2 = 0, \quad (2.4)$$

⋮

Equation (2.2) reveals that  $f^0$  belongs to the kernel of the dominant operator  $\mathcal{T} := \mathbf{b} \cdot \nabla$ . However, this information is not enough to determine completely  $f^0$ . It is necessary to use the next equation (2.3), to get the missing information. To eliminate  $f^1$  from this equation, one projects (2.3) on the kernel of  $\mathcal{T}$ . This projection is nothing else than the average of a quantity  $q$  along the field lines of  $\mathbf{b}$  and will be denoted by  $\langle q \rangle$ . Briefly, if  $Z(s; \mathbf{x})$  is the characteristic flow associated to the field  $\mathbf{b}$ , *i.e.*

$$\begin{cases} \frac{d}{ds} Z(s; \mathbf{x}) = \mathbf{b}(Z(s; \mathbf{x})), \\ Z(0; \mathbf{x}) = \mathbf{x}, \end{cases}$$

the average of a function  $q \in L^2(\Omega)$  over the field lines of  $\mathbf{b}$  is defined as

$$\langle q \rangle(\mathbf{x}) := \lim_{S \rightarrow \infty} \frac{1}{S} \int_0^S q(Z(s; \mathbf{x})) ds \quad \forall \mathbf{x} \in \Omega.$$

One can show (after some hypothesis on the regularity of  $\mathbf{b}$ , see [4]) that  $\langle \cdot \rangle$  is a well-defined application, furthermore that  $\langle q \rangle$  is constant along the field lines of  $\mathbf{b}$  and that one has  $\langle \mathbf{b} \cdot \nabla q \rangle = 0$ . The above mentioned procedure permits then to obtain a well-posed limit model for  $f^0$ . We already know that  $f^0$  belongs to the kernel of  $\mathcal{T}$ , meaning  $f^0 = \langle f^0 \rangle$ , such that the limit model  $(V)^0$  writes

$$(V)^0 \begin{cases} \partial_t f^0 = 0, & \mathbf{b} \cdot \nabla f^0 = 0, & \forall (t, \mathbf{x}) \in (0, T) \times \Omega, \\ f^0(0, \mathbf{x}) = \langle f_{in}(\mathbf{x}) \rangle \in \ker \mathcal{T}, & \mathbf{x} \in \Omega. \end{cases} \quad (2.5)$$

The following theorem proves rigorously the convergence of the solution  $f^\epsilon$  of (1.8) towards the solution  $f^0$  of the limit model (2.5), as  $\epsilon \rightarrow 0$ .

**THEOREM 2.1.** [4] *Consider a subset  $\Omega$  of  $\mathbb{R}^2$  satisfying Hypothesis A. Assume that  $\mathbf{b} \in W_{loc}^{1,\infty}(\mathbb{R}^2)$  (where in the case  $\Omega$  is a strip, we extend  $\mathbf{b}$  periodically to the whole  $\mathbb{R}^2$ ) satisfying  $\nabla \cdot \mathbf{b} = 0$  as well as the growth condition*

$$\exists C > 0 \text{ s.t. } |\mathbf{b}(\mathbf{x})| \leq C(1 + |\mathbf{x}|), \quad \forall \mathbf{x} \in \Omega.$$

*Suppose furthermore that  $f_{in} \in L^2(\Omega)$ . Then (1.8) resp. (2.5) admit unique weak solutions  $f^\epsilon, f^0 \in L^\infty(0, T; L^2(\Omega))$  and one has  $f^\epsilon \xrightarrow{\epsilon \rightarrow 0} f^0$ , weakly- $\star$  in  $L^\infty(0, T; L^2(\Omega))$ .*

*If the initial conditions are well prepared in the sense that  $f_{in}^\epsilon$  is smooth enough and satisfies  $f_{in}^\epsilon \xrightarrow{\epsilon \rightarrow 0} f_{in}^0 \in \ker \mathcal{T}$  in  $L^2(\Omega)$ , then one has even  $f^\epsilon \xrightarrow{\epsilon \rightarrow 0} f^0$  in  $L^\infty(0, T; L^2(\Omega))$ .*

**2.2. Micro-Macro reformulation.** The design of a multiscale numerical procedure for the resolution of problem (1.8) is now inspired by the asymptotic study performed in Section 2.1. To recover the missing microscopic information in the reduced model (1.2), we shall decompose  $f^\epsilon$  into a macroscopic and a microscopic part, as follows

$$f^\epsilon = p^\epsilon + \epsilon q^\epsilon, \quad \text{with } \mathbf{b} \cdot \nabla f^\epsilon = \epsilon \mathbf{b} \cdot \nabla q^\epsilon. \quad (2.6)$$

This signifies that  $p^\epsilon$  belongs to the kernel of the dominant operator  $\mathcal{T} = \mathbf{b} \cdot \nabla$  and is considered as the macroscopic part. This decomposition is not unique as one has still to fix the values of  $p^\epsilon$  (or equivalently  $q^\epsilon$ ) on the field-lines, fact which shall be done in the next subsections.

Plugging for the moment (2.6) into (1.8) leads to the following augmented system for the two unknowns  $(f^\epsilon, q^\epsilon)$

$$\begin{cases} \partial_t f^\epsilon + \mathbf{b} \cdot \nabla q^\epsilon = 0, & \forall (t, \mathbf{x}) \in (0, T) \times \Omega, \\ \mathbf{b} \cdot \nabla f^\epsilon = \epsilon \mathbf{b} \cdot \nabla q^\epsilon, & \forall (t, \mathbf{x}) \in (0, T) \times \Omega, \end{cases} \quad (2.7)$$

associated with the initial condition  $f^\epsilon(0, \cdot) = f_{in}$  and adequate boundary conditions (*Hypothesis A*). Now several possibilities are conceivable to fix the values of  $q^\epsilon$  on the field-lines, rendering thus the decomposition (2.6) unique. Let us observe here that the values of  $q^\epsilon$  on these lines are of no importance for the computation of our physical unknown  $f^\epsilon$ , as only  $\mathbf{b} \cdot \nabla q^\epsilon$  is occurring in the system (2.7). Thus any arbitrary choice could do the work.

**2.3. Zero mean value.** From a purely mathematical point of view, one first idea is to fix the average of  $q^\epsilon$  along the field lines of  $\mathbf{b}$ , by enforcing zero mean, *i.e.*

$$\langle q^\epsilon \rangle = 0. \quad (2.8)$$

Imposing (2.8) can be done by slightly changing the system, adding an additional “subtle” term  $\sigma \langle q^\epsilon \rangle$ , with  $\sigma \in \mathbb{R}$  an arbitrary constant, namely

$$\begin{cases} \partial_t f^\epsilon + \mathbf{b} \cdot \nabla q^\epsilon = 0, \\ \mathbf{b} \cdot \nabla f^\epsilon = \epsilon \mathbf{b} \cdot \nabla q^\epsilon - \sigma \langle q^\epsilon \rangle. \end{cases} \quad (2.9)$$

Indeed, one can remark immediately that taking the average of the second equation over the field-lines yields automatically the constraint  $\langle q^\epsilon \rangle = 0$ , since  $\langle \mathbf{b} \cdot \nabla \theta \rangle = 0$ , for all  $\theta$ . The new introduced term is hence a tricky zero, rendering  $q^\epsilon$  unique by fixing its average values along the  $\mathbf{b}$ -lines to zero. One can

show now that (2.9) is completely equivalent to (1.8), for each  $\epsilon > 0$ . Indeed, two ingredients help to prove this equivalence between both formulations. On one hand, for given  $f^\epsilon$  and  $\epsilon > 0$ , the equation

$$\begin{cases} \mathbf{b} \cdot \nabla q^\epsilon = \frac{1}{\epsilon} \mathbf{b} \cdot \nabla f^\epsilon, \\ \langle q^\epsilon \rangle = 0, \end{cases}$$

has a unique solution  $q^\epsilon$ . On the other hand, the second equation in (2.9) yields immediately, as mentioned above, the constraint  $\langle q^\epsilon \rangle = 0$ .

This idea is very nice from a mathematical point of view, however, if one is thinking at the numerical implementation, one has to average over the field lines of  $\mathbf{b}$ , in order to discretize the new term  $\sigma \langle q^\epsilon \rangle$  in the second equation of (2.9). This procedure is rather hard (we are working on Cartesian grids with not-aligned fields  $\mathbf{b}$ ) and can introduce moreover  $\epsilon$ -dependent error terms in the results. Thus we shall leave this idea behind, and search for a more practical one.

**2.4. Regularization.** In order to render  $q^\epsilon$  unique in (2.7), one can imagine to use a regularization technique. Regularization is a very broad field in mathematics, and is devoted to the design and analysis of methods for obtaining stable solutions of ill-posed problems. In particular, the usual regularization technique consists in replacing the ill-posed problem by a nearby (slightly-perturbed) well-posed problem, whose resolution poses no difficulties (uniqueness, stability of the solution). The original solution is recovered only in the limit of vanishing regularization/perturbation parameter. The choice of the perturbation term as well as the strength of the perturbation parameter is essential and constitutes the key point of the method. There is a rich literature on regularization techniques, we refer the interested reader to the references [2, 6, 8, 18].

Coming now to our problem, one can regularize (2.7) either by adding in the second equation a term of the type “ $\alpha \partial_t q^\epsilon$ ” or of the form “ $\sigma q^\epsilon$ ”. These two regularizations permit to get a unique  $q^\epsilon$ , however the corresponding solutions behave very differently in the desired asymptotic limit  $\epsilon \rightarrow 0$ . To see this difference, let us simplify by putting formally  $\epsilon = 0$  in (2.7) and take a look at both regularizations, namely

$$(W) \begin{cases} \partial_t f^0 + \mathbf{b} \cdot \nabla q^0 = 0, \\ \alpha \partial_t q^0 + \mathbf{b} \cdot \nabla f^0 = 0, \end{cases} \quad (P) \begin{cases} \partial_t f^0 + \mathbf{b} \cdot \nabla q^0 = 0, \\ \mathbf{b} \cdot \nabla f^0 + \sigma q^0 = 0, \end{cases} \quad (2.10)$$

where the constants  $\alpha > 0$  resp.  $\sigma > 0$  have to be sufficiently small in order not to perturb too much the original problem. Now, one can eliminate in both systems the auxiliary unknown  $q^0$  and get an equation involving only  $f^0$ , which reads

$$(W) \partial_{tt} f^0 - \frac{1}{\alpha} \nabla \cdot [(\mathbf{b} \otimes \mathbf{b}) \nabla f^{0,\sigma}] = 0; \quad (P) \partial_t f^0 - \frac{1}{\sigma} \nabla \cdot [(\mathbf{b} \otimes \mathbf{b}) \nabla f^{0,\sigma}] = 0.$$

As one can observe, regularizing (2.7) by adding a term of the form “ $\alpha \partial_t q^\epsilon$ ” leads in the limit  $\epsilon \rightarrow 0$  to a wave-equation, whereas the regularization by adding “ $\sigma q^\epsilon$ ” leads to a parabolic equation.

Which one is better suited for our anisotropic transport problem can be understood by remembering the asymptotic behaviour of the unique solution  $f^\epsilon$  of the original advection equation (1.8) as  $\epsilon$  becomes smaller and smaller. As shown in Section 2.1, one gets in the (weak) limit  $\epsilon \rightarrow 0$  a function  $f^0$  which is constant along the field lines of  $\mathbf{b}$ . This gives us a hint that the regularization (P) is better suited for our problem, as the corresponding limit problem is a diffusion problem, with strong diffusivity along the field lines. Hence one is expecting to get a better approximation of  $f^0$  via (P) than via a wave-equation of the form (W), which describes rather a very rapid wave-motion.

**2.5. The stabilized AP-reformulation.** To summarize, our Asymptotic-Preserving scheme for an efficient resolution of the anisotropic transport equation (1.8) is based on the resolution of the following well-posed reformulated system

$$(MM)_\epsilon^\sigma \begin{cases} \partial_t f^{\epsilon,\sigma} + \mathbf{b} \cdot \nabla q^{\epsilon,\sigma} = 0, & \forall (t, \mathbf{x}) \in (0, T) \times \Omega, \\ \mathbf{b} \cdot \nabla f^{\epsilon,\sigma} = \epsilon \mathbf{b} \cdot \nabla q^{\epsilon,\sigma} - \sigma q^{\epsilon,\sigma}, & \forall (t, \mathbf{x}) \in (0, T) \times \Omega, \end{cases} \quad (2.11)$$

with  $\sigma > 0$  a small parameter to be fixed later on. This system is completed by an initial condition  $f^{\epsilon,\sigma}(0, \cdot) = f_{in}(\cdot)$  and adequate boundary conditions (*Hypothesis A*). Remark that no initial condition is needed for the auxiliary variable  $q^\epsilon$ . Let us underline here the difference between (2.9) and (2.11). Both procedures are fixing the value of the auxiliary variable  $q$  on the field lines of  $\mathbf{b}$  by imposing zero mean  $\langle q \rangle = 0$ . To see this in (2.11), it is enough to take the average  $\langle \cdot \rangle$  of the second equation. However, while (2.9) is completely equivalent to the starting model (1.8), the system (2.11) introduces an error, as the supplementary term we introduced,  $\sigma q^{\epsilon,\sigma}$ , is no more zero but contains also the non-zero fluctuation part of  $q^{\epsilon,\sigma}$ . The big advantage of (2.11) with respect to (2.9) is that a discretization of the average operator  $\langle \cdot \rangle$  is not necessary anymore.

The  $\epsilon$ -regularity of the system (2.11) allows now to pass directly to the  $\epsilon \rightarrow 0$  limit in (2.11) to get the corresponding limit model, *i.e.*

$$(MM)_0^\sigma \begin{cases} \partial_t f^{0,\sigma} + \mathbf{b} \cdot \nabla q^{0,\sigma} = 0, \\ \mathbf{b} \cdot \nabla f^{0,\sigma} + \sigma q^{0,\sigma} = 0. \end{cases} \quad (2.12)$$

Eliminating  $q^{0,\sigma}$  from this system, yields the degenerate diffusion equation with fixed  $\sigma > 0$ )

$$\partial_t f^{0,\sigma} - \frac{1}{\sigma} \nabla \cdot [(\mathbf{b} \otimes \mathbf{b}) \nabla f^{0,\sigma}] = 0, \quad (2.13)$$

which shows clearly what the regularization term is doing in the limit  $\epsilon \rightarrow 0$ . Let us also remark here that (2.13) is a well-posed problem (for  $\sigma > 0$ ), due to the standard parabolic theory associated with the evolution triple  $Q \hookrightarrow V \equiv V^* \hookrightarrow Q^*$ , where the spaces are defined in the next section. Remark that the positive, symmetric bilinear form  $m(u, v) := \frac{1}{\sigma} (\mathbf{b} \cdot \nabla u, \mathbf{b} \cdot \nabla v)_V$  satisfies Garding's inequality (see [37] for more details). For future numerical discretizations, it will be more convenient to rewrite these systems by using the Poisson bracket. Introducing the stream function  $\Psi$  such that  $\mathbf{b} = (\partial_y \Psi, -\partial_x \Psi) =^\perp \nabla \Psi$ , the previous Micro-Macro system (2.11) reads

$$(MM)_\epsilon^\sigma \begin{cases} \partial_t f^{\epsilon,\sigma} + \{q^{\epsilon,\sigma}, \Psi\} = 0, \\ \{f^{\epsilon,\sigma}, \Psi\} = \epsilon \{q^{\epsilon,\sigma}, \Psi\} - \sigma q^{\epsilon,\sigma}. \end{cases} \quad (2.14)$$

Before introducing the numerical discretization of the AP-reformulation (2.14) in Section 4, let us however do in the next section some mathematical investigations.

**3. Some mathematical observations.** The rigorous mathematical study of the existence and uniqueness of a solution to the AP-reformulation (2.11) along with the rigorous study of its limit towards (2.12) is a delicate question and shall be treated in a supplementary, more mathematical work. To give however some ideas about the well-posedness of this model, and underline the difficulties of the mathematical study, we shall concentrate in this paper only on the existence and uniqueness questions corresponding to the implicit time semi-discretization of (2.11), namely

$$(MM)_\epsilon^{\sigma,n} \begin{cases} f^{\epsilon,\sigma,n+1} + \Delta t \mathbf{b} \cdot \nabla q^{\epsilon,\sigma,n+1} = f^{\epsilon,\sigma,n}, \\ \mathbf{b} \cdot \nabla f^{\epsilon,\sigma,n+1} = \epsilon \mathbf{b} \cdot \nabla q^{\epsilon,\sigma,n+1} - \sigma q^{\epsilon,\sigma,n+1}. \end{cases} \quad (3.1)$$

Here, we discretized the time interval  $[0, T]$  with fixed  $T > 0$ , as follows

$$t^n := n \Delta t, \quad \Delta t := T/N_t, \quad n \in \llbracket 0, N_t \rrbracket, \quad N_t \in \mathbb{N},$$

and denoted by  $f^{\epsilon,\sigma,n}$  an approximation of  $f^{\epsilon,\sigma}(t^n, \cdot)$ . This system is associated with boundary conditions for  $(f^{\epsilon,\sigma,n+1}, q^{\epsilon,\sigma,n+1})$  following *Hypothesis A*.

Let us now specify the mathematical framework of problem (3.1). For this, choose firstly the Hilbert spaces  $V = L^2(\Omega)$  and  $Q = \{v \in L^2(\Omega) \mid \mathbf{b} \cdot \nabla v \in L^2(\Omega), v \text{ satisfying the boundary conditions precised in } \textit{Hypothesis A}\}$ , associated with the standard  $L^2$  scalar-product for  $V$  and  $(u, v)_Q := (u, v)_{L^2} + (\mathbf{b} \cdot$

$\nabla u, \mathbf{b} \cdot \nabla v)_{L^2}$  for  $Q$ . We introduce now the following bi-linear forms  $\mathcal{A}, \mathcal{B}, \mathcal{C}_{\epsilon, \sigma}$ :

$$\begin{aligned} \mathcal{A}(u, v) &:= \int_{\Omega} uv \, d\mathbf{x}, & \mathcal{A} : V \times V &\rightarrow \mathbb{R}, \\ \mathcal{B}(v, r) &:= \int_{\Omega} v(\mathbf{b} \cdot \nabla r) \, d\mathbf{x}, & \mathcal{B} : V \times Q &\rightarrow \mathbb{R}, \\ \mathcal{C}_{\epsilon, \sigma}(r, s) &:= -\epsilon \mathcal{B}(s, r) + \sigma(r, s)_{L^2}, & \mathcal{C}_{\epsilon, \sigma} : Q \times Q &\rightarrow \mathbb{R}, \end{aligned} \quad (3.2)$$

and their associated linear operators  $A, B, C_{\epsilon, \sigma}$ :

$$\begin{aligned} A : V &\longrightarrow V^*, & B : V &\longrightarrow Q^*, & C_{\epsilon, \sigma} : Q &\longrightarrow Q^*, \\ \langle Au, v \rangle_{V^*, V} &:= \mathcal{A}(u, v), & \langle Bv, r \rangle_{Q^*, Q} &:= \mathcal{B}(v, r), & \langle C_{\epsilon, \sigma} r, s \rangle_{Q^*, Q} &:= \mathcal{C}_{\epsilon, \sigma}(r, s). \end{aligned}$$

REMARK 3.1. *The bi-linear form  $\mathcal{B}$  defines also the adjoint linear operator  $B^* : Q \longrightarrow V^*$  via  $\mathcal{B}(v, r) = \langle Bv, r \rangle_{Q^*, Q} = \langle v, B^*r \rangle_{V, V^*}$  for all  $(v, r) \in V \times Q$ . Observe also that  $B^*r = \mathbf{b} \cdot \nabla r$  for all  $r \in Q$ , whereas in the distributional sense  $Bv = -\mathbf{b} \cdot \nabla v$  for all  $v \in V$ .*

With these definitions, the variational formulation of the previous problem (3.1) writes now : for fixed  $\epsilon \geq 0, \sigma > 0, \Delta t > 0$  and  $f^{\epsilon, \sigma, n} \in V^*$ , find  $(f^{\epsilon, \sigma, n+1}, q^{\epsilon, \sigma, n+1}) \in V \times Q$ , such that :

$$\begin{cases} \mathcal{A}(f^{\epsilon, \sigma, n+1}, \theta) + \Delta t \mathcal{B}(\theta, q^{\epsilon, \sigma, n+1}) = \langle f^{\epsilon, \sigma, n}, \theta \rangle_{V^* \times V}, & \forall \theta \in V, \\ \mathcal{B}(f^{\epsilon, \sigma, n+1}, \chi) - \mathcal{C}_{\epsilon, \sigma}(q^{\epsilon, \sigma, n+1}, \chi) = 0, & \forall \chi \in Q. \end{cases} \quad (3.3)$$

To prove the existence and uniqueness of a weak solution to (3.3), we shall need some properties of these operators.

LEMMA 3.2. *Let Hypothesis A be satisfied. Then,  $\mathcal{A}(\cdot, \cdot), \mathcal{B}(\cdot, \cdot)$  resp.  $\mathcal{C}_{\epsilon, \sigma}(\cdot, \cdot)$  are continuous bi-linear forms on  $V \times V, V \times Q$  resp.  $Q \times Q$ . Furthermore,  $\mathcal{A}(\cdot, \cdot)$  is coercive on  $V \times V$  and  $\mathcal{C}_{\epsilon, \sigma}(\cdot, \cdot)$  is positive semi-definite on  $Q \times Q$ .*

REMARK 3.3. *Let us remark here that  $\mathcal{C}_{\epsilon, \sigma}(\cdot, \cdot)$  is not coercive on  $Q \times Q$ . However, as we will see later, this hypothesis is not crucial for both existence and uniqueness of a solution of the variational formulation (3.3).*

Without any other hypothesis on  $\mathcal{B}(\cdot, \cdot)$  (as for example an inf-sup condition) we are now able to present the following existence/uniqueness result of a solution to (3.3), and this due to the presence of the regularization term  $\sigma q$ .

THEOREM 3.1. *Let Hypothesis A be satisfied and let  $\mathcal{A}(\cdot, \cdot), \mathcal{B}(\cdot, \cdot)$  and  $\mathcal{C}_{\epsilon, \sigma}(\cdot, \cdot)$  be the continuous bi-linear forms defined in (3.2). Then, for every given  $f^{\epsilon, \sigma, n} \in V^*$  the problem (3.3) has for each fixed  $\epsilon \geq 0, \sigma > 0, \Delta t > 0$  and  $n \in \llbracket 0, N_t \rrbracket$ , a unique weak solution  $(f^{\epsilon, \sigma, n+1}, q^{\epsilon, \sigma, n+1})$  in  $V \times Q$ . This solution satisfies the following estimates, for all  $\epsilon \geq 0, \sigma > 0, \Delta t > 0, n \in \llbracket 0, N_t \rrbracket$ :*

$$\|f^{\epsilon, \sigma, n+1}\|_V \leq \|f^{\epsilon, \sigma, n}\|_{V^*}, \quad (3.4)$$

$$\|q^{\epsilon, \sigma, n+1}\|_V^2 \leq \frac{1}{\sigma \Delta t} \|f^{\epsilon, \sigma, n}\|_{V^*}^2, \quad \|\mathbf{b} \cdot \nabla q^{\epsilon, \sigma, n+1}\|_V \leq \frac{2}{\Delta t} \|f^{\epsilon, \sigma, n}\|_{V^*}. \quad (3.5)$$

*Proof.* Due to the lack of coercivity of  $\mathcal{C}_{\epsilon, \sigma}$ , we shall start by considering the regularized problem : for  $\alpha > 0$ , find  $(f_{\alpha}^{\epsilon, \sigma, n+1} \in V, q_{\alpha}^{\epsilon, \sigma, n+1}) \in V \times Q$ , solution of :

$$\begin{cases} \mathcal{A}(f_{\alpha}^{\epsilon, \sigma, n+1}, \theta) + \Delta t \mathcal{B}(\theta, q_{\alpha}^{\epsilon, \sigma, n+1}) = \langle f^{\epsilon, \sigma, n}, \theta \rangle_{V^* \times V}, & \forall \theta \in V, \\ \mathcal{B}(f_{\alpha}^{\epsilon, \sigma, n+1}, \chi) - \alpha(q_{\alpha}^{\epsilon, \sigma, n+1}, \chi)_Q - \mathcal{C}_{\epsilon, \sigma}(q_{\alpha}^{\epsilon, \sigma, n+1}, \chi) = 0, & \forall \chi \in Q. \end{cases} \quad (3.6)$$

Multiplying the second equation by  $\Delta t$  and subtracting both equations, one can show via Lax-Milgram theorem that (3.6) admits, for each fixed  $\alpha > 0, \epsilon \geq 0, \sigma > 0, \Delta t > 0$  and  $n \in \llbracket 0, N_t \rrbracket$ , a unique weak solution. Indeed, thanks to the term  $\alpha(q_{\alpha}^{\epsilon, \sigma, n+1}, \chi)_Q$  in the second equation of (3.6), the regularized operator  $\mathcal{C}_{\epsilon, \sigma}(\cdot, \cdot) + \alpha(\cdot, \cdot)_Q$  is now coercive on  $Q \times Q$ .

Our aim is now to bound  $f_\alpha^{\epsilon,\sigma,n+1}$  and  $q_\alpha^{\epsilon,\sigma,n+1}$  uniformly in  $\alpha$ . Then, passing to the limit  $\alpha \rightarrow 0$  in (3.6) would permit to conclude about both existence and uniqueness of a weak solution of (3.3).

First, let us choose  $\theta = f_\alpha^{\epsilon,\sigma,n+1}$  and  $\chi = q_\alpha^{\epsilon,\sigma,n+1}$  in (3.6), multiply the second equation with  $\Delta t$  and subtract both equations to get

$$\mathcal{A}(f_\alpha^{\epsilon,\sigma,n+1}, f_\alpha^{\epsilon,\sigma,n+1}) + \alpha \Delta t \|q_\alpha^{\epsilon,\sigma,n+1}\|_Q^2 + \Delta t \mathcal{C}_{\epsilon,\sigma}(q_\alpha^{\epsilon,\sigma,n+1}, q_\alpha^{\epsilon,\sigma,n+1}) = \langle f^{\epsilon,\sigma,n}, f_\alpha^{\epsilon,\sigma,n+1} \rangle_{V^* \times V}.$$

By using the coercivity of  $\mathcal{A}(\cdot, \cdot)$  and the fact that  $\mathcal{C}_{\epsilon,\sigma}(q, q) = \sigma \|q\|_{L^2}^2$  (due to the boundary conditions), we get

$$\|f_\alpha^{\epsilon,\sigma,n+1}\|_V^2 + \alpha \Delta t \|q_\alpha^{\epsilon,\sigma,n+1}\|_Q^2 + \Delta t \sigma \|q_\alpha^{\epsilon,\sigma,n+1}\|_V^2 \leq \|f^{\epsilon,\sigma,n}\|_{V^*} \|f_\alpha^{\epsilon,\sigma,n+1}\|_V,$$

leading to

$$\|f_\alpha^{\epsilon,\sigma,n+1}\|_V \leq \|f^{\epsilon,\sigma,n}\|_{V^*}, \quad \|q_\alpha^{\epsilon,\sigma,n+1}\|_V^2 \leq \frac{1}{\Delta t \sigma} \|f^{\epsilon,\sigma,n}\|_{V^*}^2. \quad (3.7)$$

Now, thanks to the first equation of (3.6), *i.e.*

$$\Delta t \mathcal{B}(\theta, q_\alpha^{\epsilon,\sigma,n+1}) = -\mathcal{A}(f_\alpha^{\epsilon,\sigma,n+1}, \theta) + \langle f^{\epsilon,\sigma,n}, \theta \rangle_{V^* \times V}, \quad \forall \theta \in V,$$

we have

$$\|\mathbf{b} \cdot \nabla q_\alpha^{\epsilon,\sigma,n+1}\|_{L^2(\Omega)} = \sup_{\theta \in L^2(\Omega)} \frac{(\mathbf{b} \cdot \nabla q_\alpha^{\epsilon,\sigma,n+1}, \theta)_{L^2(\Omega)}}{\|\theta\|_{L^2(\Omega)}} \leq \frac{2}{\Delta t} \|f^{\epsilon,\sigma,n}\|_{V^*}. \quad (3.8)$$

The estimates (3.7) as well as (3.8) are independent of  $\alpha > 0$  and thus one can extract weakly convergent subsequences and pass to the limit  $\alpha \rightarrow 0$  in the variational formulation (3.6) to conclude the proof.  $\square$

REMARK 3.4. *One can also observe from the  $\epsilon$ -independent estimates (3.4)-(3.5) that up to extract- ing a subsequence of  $\{f^{\epsilon,\sigma,n+1}, q^{\epsilon,\sigma,n+1}\}_{\epsilon>0}$  we have the  $\epsilon$ -convergences*

$$f^{\epsilon,\sigma,n+1} \xrightarrow{\epsilon \rightarrow 0} f^{0,\sigma,n+1} \quad \text{in } V, \quad q^{\epsilon,\sigma,n+1} \xrightarrow{\epsilon \rightarrow 0} q^{0,\sigma,n+1} \quad \text{in } Q,$$

*which underlines the fact that the AP-reformulation (2.11) is regular and tends towards the limit-model (2.12) as  $\epsilon$  goes to zero.*

*The  $\sigma$ -convergences are more delicate. One has only*

$$f^{\epsilon,\sigma,n+1} \xrightarrow{\sigma \rightarrow 0} f^{\epsilon,0,n+1} \quad \text{in } V,$$

*however there is no convergence for  $q^{\epsilon,\sigma,n+1}$ . This again was to be expected as in the limit  $\sigma \rightarrow 0$  one loses the uniqueness of  $q$ , the term  $\sigma q$  being a regularization term.*

**4. Numerical discretization.** Let us come now to the numerical part of our work and introduce here a numerical discretization of our reformulated system:

$$(MM)_\epsilon^\sigma \begin{cases} \partial_t f^{\epsilon,\sigma} + \{q^{\epsilon,\sigma}, \Psi^{\epsilon,\sigma}\} = 0, \\ \{f^{\epsilon,\sigma}, \Psi^{\epsilon,\sigma}\} = \epsilon \{q^{\epsilon,\sigma}, \Psi^{\epsilon,\sigma}\} - \sigma q^{\epsilon,\sigma}, \end{cases} \quad (4.1)$$

where the stream function  $\Psi^{\epsilon,\sigma}$  is supposed to be given in this section, linked to the given time-independent vector field  $\mathbf{b}$  through  $\mathbf{b} := \nabla^\perp \Psi^{\epsilon,\sigma}$ .

**4.1. Discretization parameters.** In what follows, one assumes a bounded simulation domain  $\Omega_S := (-L_x, L_x) \times (-L_y, L_y)$ . Concerning the time interval  $[0, T]$ ,  $T > 0$ , we employ the discretization:

$$t^n := n \Delta t, \quad \Delta t := T/N_t, \quad n \in \llbracket 0, N_t \rrbracket, \quad N_t \in \mathbb{N}.$$

Similarly, let us supply the domain  $\Omega_S$  with a uniform spatial discretization:

$$x_i := (i-1)\Delta x - L_x, \quad y_j := (j-1)\Delta y - L_y, \quad \Delta x := 2L_x/N_x, \quad \Delta y := 2L_y/N_y,$$

with  $i \in \llbracket 1, N_x + 1 \rrbracket$ , and  $j \in \llbracket 1, N_y + 1 \rrbracket$ . For any function  $f : [0, T] \times \Omega_S \rightarrow \mathbb{R}$ ,  $f_{i,j}^n$  refers to the numerical approximation of  $f(t^n, x_i, y_j)$ , and  $f_h^n$  shall simply denote the discrete grid-function  $(f_{i,j}^n)_{i,j}$ .

The domain  $\Omega_S$  is a truncation of the real physical domain  $\Omega = \mathbb{R}^2$  or  $\Omega = (L_1, L_2) \times \mathbb{R}$ . To be close to the physical reality, we took in our simulations for the variable belonging to  $\mathbb{R}$  a sufficiently large bounded domain and supposed that the distribution function  $f^{\epsilon, \sigma}$  is vanishing on that border, whereas periodic boundary conditions are imposed for the bounded variable. To be more precise, we imposed for the truncation of the physical domain  $\Omega = \mathbb{R}^2$  that

$$f_{i,j}^n = 0 \text{ for } i = 1; j = 1; i = N_x + 1; j = N_y + 1,$$

whereas for the truncation of the physical domain  $\Omega = (L_1, L_2) \times \mathbb{R}$  we imposed

$$f_{i,j}^n = 0 \quad \forall i \text{ and } j = 1; j = N_y + 1 \quad \text{as well as} \quad f_{1,j}^n = f_{N_x+1,j}^n \quad \forall j.$$

**4.2. Space semi-discretization.** For the Poisson brackets appearing in the Micro-Macro reformulation (4.1), let us adopt the second order Arakawa discretization [1]. For two functions  $u, v : \Omega_S \rightarrow \mathbb{R}$ , the discrete version of the Poisson bracket  $\{u, v\}$  calculated at the point  $(x_i, y_j)$  is expressed by:

$$\begin{aligned} [u_h, v_h]_{i,j} := & \frac{1}{12\Delta x \Delta y} \left( u_{i+1,j} \mathcal{A}_{i,j} + u_{i-1,j} \mathcal{B}_{i,j} + u_{i,j+1} \mathcal{C}_{i,j} + u_{i,j-1} \mathcal{D}_{i,j} \right. \\ & \left. + u_{i+1,j+1} \mathcal{E}_{i,j} + u_{i-1,j-1} \mathcal{F}_{i,j} + u_{i-1,j+1} \mathcal{G}_{i,j} + u_{i+1,j-1} \mathcal{H}_{i,j} \right). \end{aligned}$$

where the coefficients write

$$\begin{aligned} \mathcal{A}_{i,j} &:= v_{i,j+1} - v_{i,j-1} + v_{i+1,j+1} - v_{i+1,j-1}, & \mathcal{E}_{i,j} &:= v_{i,j+1} - v_{i+1,j}, \\ \mathcal{B}_{i,j} &:= v_{i,j-1} - v_{i,j+1} - v_{i-1,j+1} + v_{i-1,j-1}, & \mathcal{F}_{i,j} &:= v_{i,j-1} - v_{i-1,j}, \\ \mathcal{C}_{i,j} &:= v_{i-1,j} - v_{i+1,j} - v_{i+1,j+1} + v_{i-1,j+1}, & \mathcal{G}_{i,j} &:= v_{i-1,j} - v_{i,j+1}, \\ \mathcal{D}_{i,j} &:= v_{i+1,j} - v_{i-1,j} + v_{i+1,j-1} - v_{i-1,j-1}, & \mathcal{H}_{i,j} &:= v_{i+1,j} - v_{i,j-1}. \end{aligned}$$

Thus, the semi-discretization in space of the Micro-Macro problem (4.1) reads:

$$(MM)_{\epsilon,h}^{\sigma} \begin{cases} \partial_t f_{i,j}^{\epsilon,\sigma} + [q_h^{\epsilon,\sigma}, \Psi_h^{\epsilon,\sigma}]_{i,j} = 0, \\ [f_h^{\epsilon,\sigma}, \Psi_h^{\epsilon,\sigma}]_{i,j} - \epsilon [q_h^{\epsilon,\sigma}, \Psi_h^{\epsilon,\sigma}]_{i,j} + \sigma q_{i,j}^{\epsilon,\sigma} = 0. \end{cases} \quad (4.2)$$

**4.3. Time discretization.** We shall use a DIRK (diagonally-implicit Runge-Kutta) approach in order to achieve second-order accuracy in time for the problem (4.1). The general form of a RK-method is recalled here for the following equation

$$\partial_t u = Lu,$$

where  $L$  refers to some differential operator. An r-stage Runge-Kutta approach is determined by its Butcher table

$$\begin{array}{c|ccc} c_1 & a_{11} & \dots & a_{1r} \\ \vdots & \vdots & & \vdots \\ c_r & a_{r1} & \dots & a_{rr} \\ \hline & b_1 & \dots & b_r \end{array}$$

For a given  $u^n$ , the subsequent  $u^{n+1}$  is defined by the formula

$$u^{n+1} = u^n + \Delta t \sum_{j=1}^r b_j u_j,$$

where each  $u_i$  is defined by

$$u_i = u^n + \Delta t \sum_{j=1}^r a_{ij} L u_j.$$

In the case where  $b_j = a_{rj}$  for  $j = 1, \dots, r$ , then  $u^{n+1}$  is equal to the last stage of the method, namely  $u_r$ . For the Micro-Macro problem (4.1), we consider the following 2-stage Butcher table

$$\begin{array}{c|cc} \lambda & \lambda & 0 \\ 1 & 1-\lambda & \lambda \\ \hline & 1-\lambda & \lambda \end{array}$$

For  $\lambda := 1 - 1/\sqrt{2}$ , the method is  $L$ -stable. For all  $n \in \llbracket 0, N_t \rrbracket$ , the full discretization of the Micro-Macro problem (4.1) writes now

$$(MM)_{\epsilon, h}^{\sigma, n} \left\{ \begin{array}{l} \text{Stage 1 :} \\ f_{1,i,j}^{\epsilon, \sigma} + \lambda \Delta t [q_{1,h}^{\epsilon, \sigma}, \Psi_h^{\epsilon, \sigma}]_{i,j} = f_{i,j}^{\epsilon, \sigma, n}, \\ [f_{1,h}^{\epsilon, \sigma}, \Psi_h^{\epsilon, \sigma}]_{i,j} - \epsilon [q_{1,h}^{\epsilon, \sigma}, \Psi_h^{\epsilon, \sigma}]_{i,j} + \sigma q_{1,i,j}^{\epsilon, \sigma} = 0. \\ \text{Stage 2 :} \\ f_{2,i,j}^{\epsilon, \sigma} + \lambda \Delta t [q_{2,h}^{\epsilon, \sigma}, \Psi_h^{\epsilon, \sigma}]_{i,j} = f_{i,j}^{\epsilon, \sigma, n} + \frac{1-\lambda}{\lambda} (f_{1,i,j}^{\epsilon, \sigma} - f_{i,j}^{\epsilon, \sigma, n}), \\ [f_{2,h}^{\epsilon, \sigma}, \Psi_h^{\epsilon, \sigma}]_{i,j} - \epsilon [q_{2,h}^{\epsilon, \sigma}, \Psi_h^{\epsilon, \sigma}]_{i,j} + \sigma q_{2,i,j}^{\epsilon, \sigma} = 0. \\ \text{Final Stage :} \\ (f_{i,j}^{\epsilon, \sigma, n+1}, q_{i,j}^{\epsilon, \sigma, n+1}) = (f_{2,i,j}^{\epsilon, \sigma}, q_{2,i,j}^{\epsilon, \sigma}). \end{array} \right. \quad (4.3)$$

REMARK 4.1. *In the following, we shall simply call our Micro-Macro scheme (4.3), obtained with Arakawa space discretization and DIRK time discretization, the (DAMM)-scheme.*

**5. Verification of the AP-scheme in a mathematical test case.** In this section we investigate the numerical properties of our asymptotic-preserving (DAMM)-scheme (4.3) for the resolution of (4.1) with given field  $\mathbf{b}$ . The section is devoted to the investigation of a linear case where the stream function  $\Psi$  does not depend on  $f^\epsilon$  and is static. In mind, we have as an application the Vlasov/Fokker-Planck equation with strong given magnetic field (1.3). Convergence results regarding the discretization parameters and numerical study of the asymptotic limit  $\epsilon \rightarrow 0$  are presented. Moreover, the influence of the stabilization parameter  $\sigma$  on the numerical results is discussed.

**5.1. Analytical solution for both  $\epsilon$ -regimes.** Let us choose in this section the stationary stream function  $\Psi(x, y) = \frac{1}{2}(x^2 + y^2)$ , corresponding to  $\mathbf{b} = (y, -x)^T$ . In this case, we can compute explicitly the characteristics corresponding to the transport equation (1.8). Indeed, the characteristic curve  $\mathcal{C}_\epsilon^{t,x,y}(s) := (X(s), Y(s))$  passing at instant  $t$  through  $(x, y)$ , solves the ODE

$$\left\{ \begin{array}{l} \dot{X}(s) = \frac{Y(s)}{\epsilon}, \\ \dot{Y}(s) = -\frac{X(s)}{\epsilon}, \end{array} \right. \quad (X(t), Y(t)) = (x, y),$$

and has the explicit form

$$\mathcal{C}_\epsilon^{t,x,y}(s) := \begin{pmatrix} X \\ Y \end{pmatrix} (s; t, x, y) = \mathcal{R}\left(\frac{s-t}{\epsilon}\right) \begin{pmatrix} x \\ y \end{pmatrix}.$$

with the rotation matrix given by

$$\mathcal{R}(\theta) := \begin{pmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{pmatrix}.$$

These characteristic curves are nothing else than spirals in the  $(t, x, y)$ -phase-space. All characteristics are  $2\pi\epsilon$ -periodic in  $s$ . The solution  $f^\epsilon$  of the advection equation (1.8) is now constant along these curves, such that

$$f^\epsilon(t, x, y) = f_{in}(X(0; t, x, y), Y(0; t, x, y)), \quad \forall (t, x, y) \in [0, T] \times \Omega,$$

which leads to the following analytic expression of the unique solution to (1.8)

$$f^\epsilon(t, x, y) = f_{in} \left( \cos \left( \frac{t}{\epsilon} \right) x - \sin \left( \frac{t}{\epsilon} \right) y, \sin \left( \frac{t}{\epsilon} \right) x + \cos \left( \frac{t}{\epsilon} \right) y \right). \quad (5.1)$$

Finally, the  $\epsilon \rightarrow 0$  limit solution  $f^0$  solves the problem (2.5), leading to

$$f^0 = \langle f_{in} \rangle = \frac{1}{2\pi} \int_0^{2\pi} f_{in}(R \cos(s), R \sin(s)) ds, \quad (5.2)$$

where  $R := \sqrt{x^2 + y^2}$ . Let us choose now as initial data a Gaussian peak not centered in the origin, i.e.

$$f_{in}(x, y) = \exp \left( - \frac{(x - 0.5)^2 + (y - 0.5)^2}{2 \eta^2} \right), \quad \eta = 0.05, \quad L_x = L_y = 1,$$

and investigate how the numerical scheme is rendering its movement.

In the following, we shall denote by  $\Pi_h(f^\epsilon)$  resp.  $f_h^{\epsilon, \sigma}$  the exact solution (5.1) calculated on the grid mesh resp. the numerical solution of (4.1) obtained with our (DAMM)-scheme, and  $\Pi_h(f^0)$  refers to the exact limit solution (5.2) calculated on the grid mesh. Let  $\mathcal{Q} := (0, T) \times \Omega_S$ . We introduce also the numerical errors

$$\mathcal{L}_{\epsilon, \mathcal{X}}^p := \|\Pi_h(f^\epsilon) - f_h^{\epsilon, \sigma}\|_{L_h^p(\mathcal{X})}, \quad \mathcal{L}_{0, \mathcal{X}}^p := \|\Pi_h(f^0) - f_h^{\epsilon, \sigma}\|_{L_h^p(\mathcal{X})}, \quad (5.3)$$

where  $p \in \{1, 2, \dots, \infty\}$ ,  $\mathcal{X}$  stands for  $\mathcal{Q}$  or  $\Omega_S$ , and  $L_h^p$  denotes the discrete  $L^p$ -norm.

Figure 5.1 shows now the numerical solution obtained with the (DAMM)-scheme and the corresponding exact solution, in the non-limit ( $\epsilon = 1$ ) regime. The solutions related to the limit regime ( $\epsilon=0$ ) are plotted in Figure 5.2. One observes that the numerical scheme we propose in this paper approximates well the exact solutions in both extreme regimes. In the next paragraphs we will try to prove more rigorously this visible correspondence and justify the choice of the stabilization parameter  $\sigma$ .

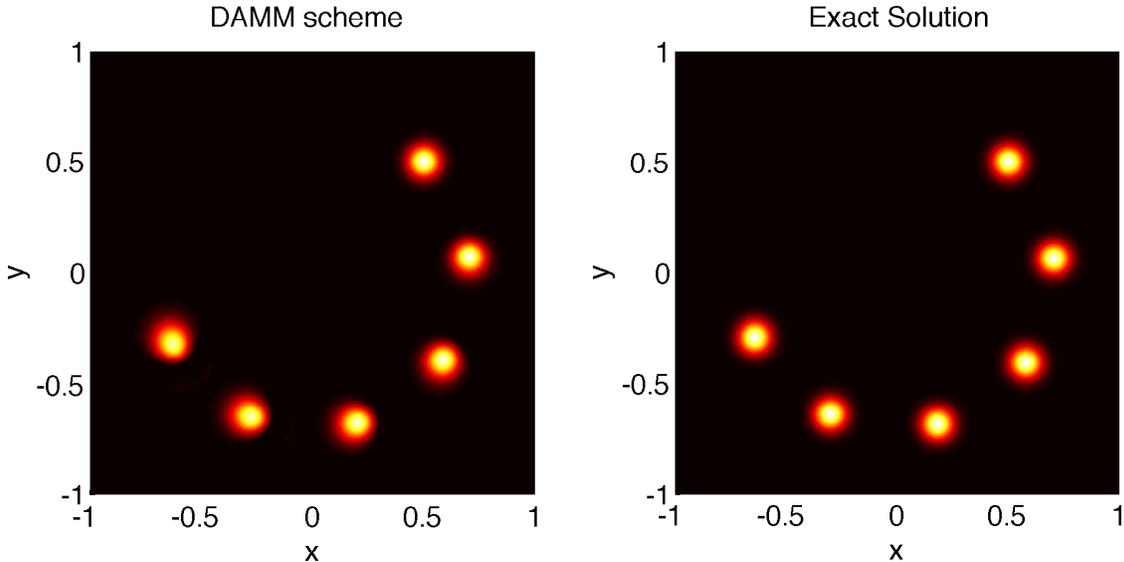


Fig. 5.1: (Non-limit case  $\epsilon = 1$ ). Superposition of the distribution function  $f_h^{\epsilon, \sigma}$  at several time steps  $t^n$  (left-panel) compared to the exact solution  $f^\epsilon$  (right-panel), with  $N_t = 2000$ ,  $T = 3.5$ ,  $N_x = N_y = 200$ ,  $\sigma = \Delta x^2$ .

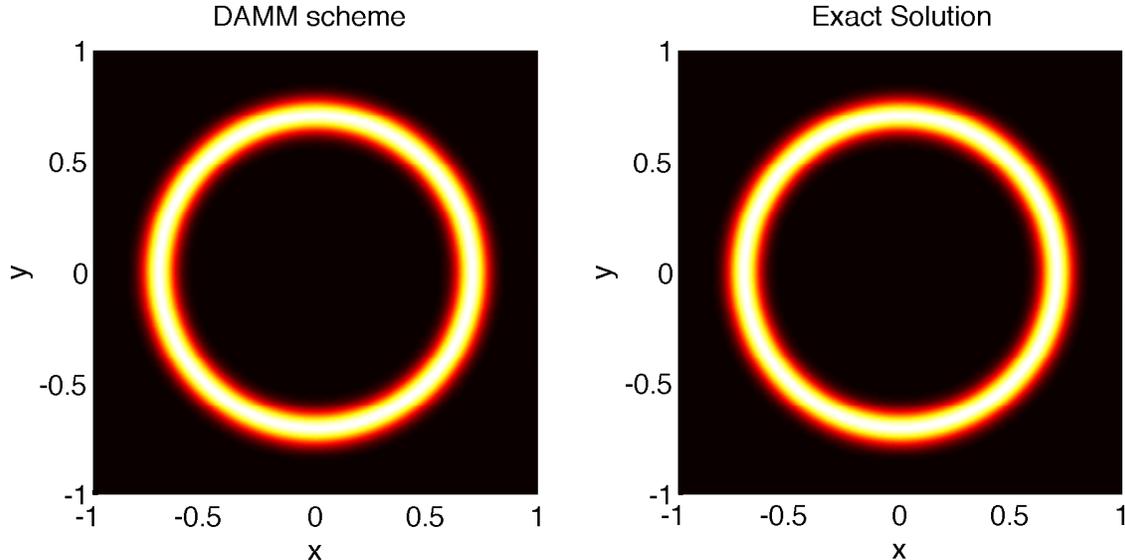


Fig. 5.2: (Limit case  $\epsilon = 0$ ). Plot of  $f_h^{0,\sigma}$  at final time  $T = 1$  with  $N_t = 200$  and  $\sigma = \Delta x$  (left-panel), compared to the limit exact solution  $f^0$  (right-panel). Mesh size :  $N_x = N_y = 200$ .

**5.2. AP-property.** Let us start by discussing the AP-property of our scheme. As a reference scheme, we decided to take a fully implicit DIRK-scheme with Arakawa space discretization, solving (1.8). In Figure 5.3 (a) we plot the condition number of the system matrix of the (DAMM)-scheme and of the implicit reference scheme as a function of  $\epsilon$ . One observes that the condition number of the (DAMM)-scheme is bounded uniformly in  $\epsilon$ , whereas the implicit scheme is ill-conditioned in the limit  $\epsilon \rightarrow 0$ . This underlines one important advantage of our (DAMM)-scheme when compared with standard schemes, namely the fact that the (DAMM)-scheme does not degenerate in the limit  $\epsilon \rightarrow 0$ . From the right panel (b), however, it is evident that the condition number of the (DAMM)-scheme depends on the stabilization parameter  $\sigma$ . This reflects the fact that in the limit  $\sigma \rightarrow 0$ , the solution  $(f^{\epsilon,0,n}, q^{\epsilon,0,n})$  of (3.1) is not unique, and therefore the problem becomes ill-posed. As mentioned later on,  $\sigma$  has to be chosen not too small, such to have a reasonable condition number, and not too large, in order not to modify too much the problem. Concerning the numerical analysis, we do not present any result in this work, since we have proved in a previous publication (see [19]) that the Micro-Macro procedure coupled with a stabilization method permits to avoid an amplification of the truncation error with decreasing  $\epsilon$ .

**5.3. Convergence property.** Next we study the convergence properties of the (DAMM)- scheme. Figure 5.4 displays the convergence rates in  $\Delta t$  and in  $\Delta x = \Delta y$ , obtained by comparison with the exact solutions (5.1) for  $\epsilon = 1$  and  $\sigma = \Delta x^2$ . In panel (a), one observes the expected second-order convergence rates with respect to time. In panel (b), the second-order convergence in space due to the Arakawa discretization of the Poisson brackets is evident. We observe from Table 5.1 that the convergence rate in space is even better for  $\epsilon = 0$ . Indeed, in this simple test-case, in the limit  $\epsilon \rightarrow 0$ , the distribution function is radial, namely  $f^{0,\sigma} = g^\sigma(x^2 + y^2)$ . Following [1], we observe that the leading terms of the truncation error, corresponding to the discretization of the Poisson bracket, vanish for that case.

**5.4. Influence of the stabilization.** Our next objective is to investigate the influence of the stabilization parameter  $\sigma$  on the quality of the numerical solution in the different  $\epsilon$ -regimes. Let us start the discussion with the limit regime  $\epsilon = 0$ . In Figure 5.5 (a), we plot the  $\mathcal{L}_{0,\Omega_S}^1$  error (w.r.t. the exact limit sol.) over time  $t$ , for two different choices of  $\sigma$ , namely  $\sigma = \Delta x$  and  $\sigma = \Delta x^2$ , and for several choices of  $\Delta x$ . One observes two phases: at first the numerical solution relaxes towards the (weak) analytic limit solution, until a plateau is reached and the error remains constant over time. This relaxation is faster for smaller values of  $\sigma$ , fact which can be explained by taking a look at the degenerate diffusion equation (2.13) we are effectively solving in the limit  $\epsilon \rightarrow 0$ . Smaller  $\sigma$  means

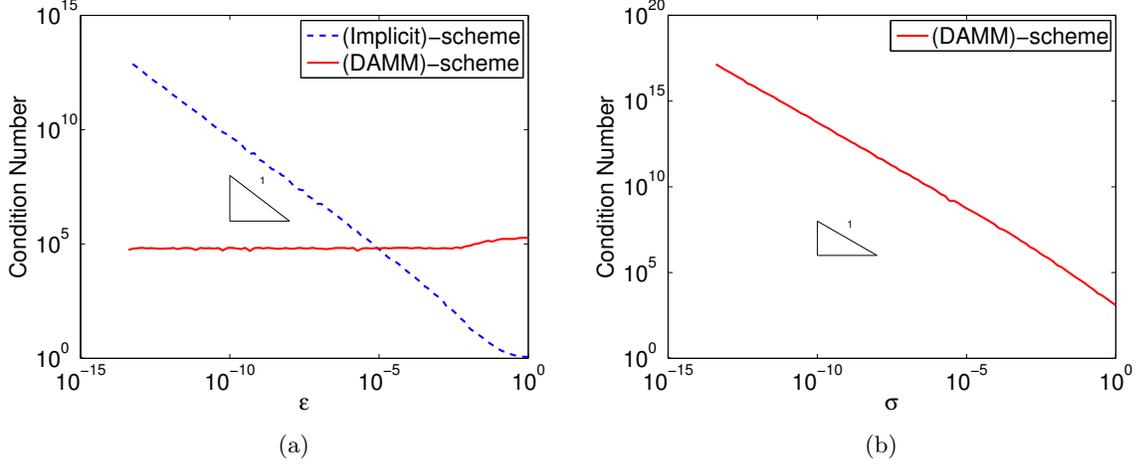


Fig. 5.3: Condition numbers as a function of  $\epsilon$  with  $\sigma = \Delta x^2$  for both (DAMM) and fully implicit schemes (a), and as a function of  $\sigma$  with  $\epsilon = 1$  for the (DAMM)-scheme (b). Here  $T = 1$ ,  $N_t = 200$ ,  $N_x = N_y = 50$ .

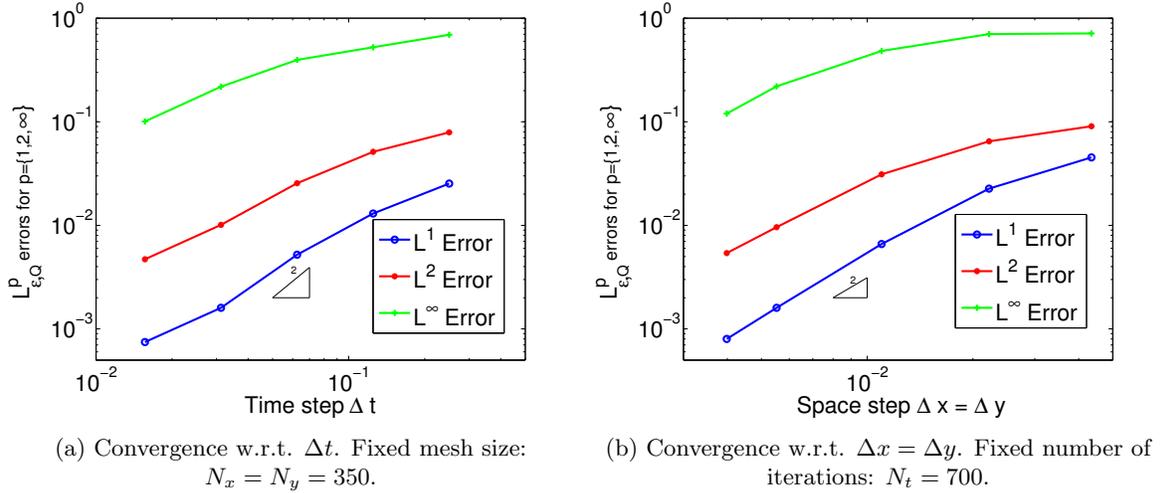


Fig. 5.4: Convergence studies for the (DAMM)-scheme: errors  $\mathcal{L}_{\epsilon, \mathcal{Q}}^1$ ,  $\mathcal{L}_{\epsilon, \mathcal{Q}}^2$ , and  $\mathcal{L}_{\epsilon, \mathcal{Q}}^\infty$  as a function of  $\Delta t$  (a) and  $\Delta x = \Delta y$  (b), respectively. Parameters were  $T = 1$ ,  $\epsilon = 1$  and  $\sigma = \Delta x^2$ .

stronger diffusion along the field lines of  $\mathbf{b}$ , which means that the number of iterations  $n_{eq}$  to reach the equilibrium plateau decreases with  $\Delta x$ , see Table 5.1. Observe also that the error in the equilibrium phase is the same for each  $\sigma = \Delta x^r$  with  $r \geq 1$ , only the relaxation rate strongly depends on the choice of  $r$ .

Let us mention briefly the computational time (CPU time) one needs for reaching the equilibrium plateau for the different cases studied in Figure 5.5 (a). The problem is that the condition number of the system matrix is inversely proportional to  $\sigma$ , as already demonstrated in Figure 5.3 (b). This bad conditioning would lead necessarily to an increase in CPU time for very small  $\sigma$  which has to be evaluated. For example, regarding the case  $\Delta x = 2/80$  (third curve in Figure 5.5 (a) and third line in Table 5.1), one obtains  $t_{CPU}(\sigma = \Delta x) = 53$  s in contrast to  $t_{CPU}(\sigma = \Delta x^2) = 104$  s. Thus, even if Figure 5.5 (a) suggests that a higher  $r$  would be more suitable to attain quickly the equilibrium plateau, this previous study about the CPU time advises us to be more careful and choose  $\sigma$  not too small.

In the regime  $\epsilon = 1$ , displayed in Figure 5.5 (b), the  $\mathcal{L}_{\epsilon, \Omega_S}^1$  error increases linearly with time for all choices of  $\Delta x$ . There is a very weak  $\sigma$  dependence in this regime as shown by the quasi superposition of the curves. The linear increase of the error is due to the dispersive character of the Arakawa discretization which leads to errors in the phase velocities. The not-dependence on  $\sigma$  is due to the fact that the term  $\epsilon \mathbf{b} \cdot \nabla q^\epsilon$  in the second equation of (2.11) is larger for  $\epsilon = 1$  than the term  $\sigma q^\epsilon$ .

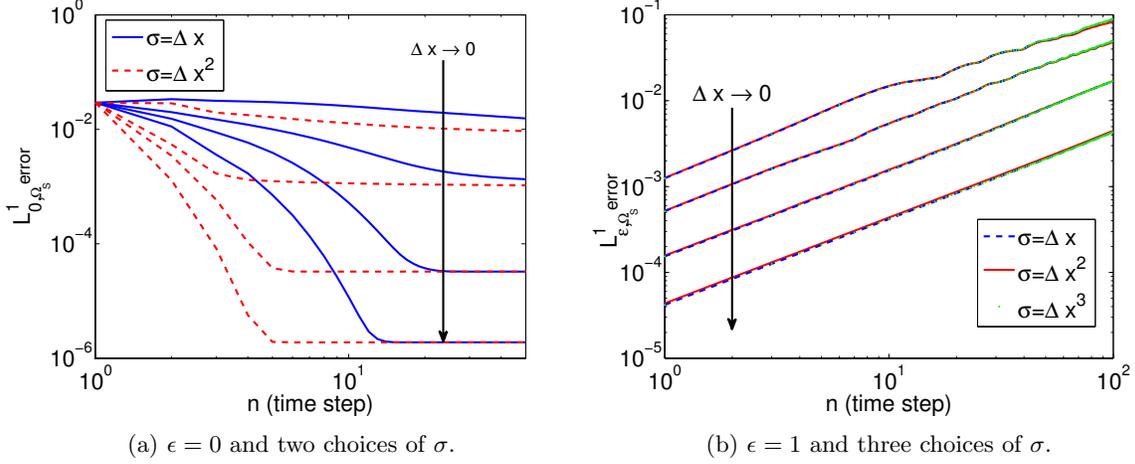


Fig. 5.5:  $\mathcal{L}_{0, \Omega_S}^1$  (panel (a)) and  $\mathcal{L}_{\epsilon, \Omega_S}^1$  (panel (b)) errors over time for the (DAMM)-scheme for two different  $\epsilon$ -regimes and several  $\Delta x$ .

Table 5.1: Corresponding to Figure 5.5 (a) with  $\epsilon = 0$ .  $\mathcal{L}_{0, \Omega_S}^1$ -error at final time  $T = 2$  and  $\Delta t = 0.01$ ; number of iterations  $n_{eq}$  and CPU-time  $t_{CPU}$  for reaching the equilibrium plateau; condition nbr. CN of the linear system.

		$\sigma = \Delta x$				$\sigma = \Delta x^2$			
		$\mathcal{L}_{0, \Omega_S}^1$ at time $T$	$n_{eq}$	$t_{CPU}$	CN	$\mathcal{L}_{0, \Omega_S}^1$ at time $T$	$n_{eq}$	$t_{CPU}$	CN
$\Delta x$	2/20	0.0106	> 200	3.1s	8.4e3	0.0082	175	2.8s	1.3e5
	2/40	0.0011	169	12s	2.6e4	0.0010	66	4.9s	8.5e5
	2/80	3.2737e-5	123	53s	9.1e4	3.2735e-5	10	104s	3.9e6
	2/160	1.8992e-6	19	133s	2.9e5	1.8992e-6	5	1302s	3.5e7

Finally, we shall discuss an intermediate  $\epsilon$ -regime given by  $\epsilon = 0.1$ . We simulate  $n_{final} = 500$  time steps with  $\Delta t = 0.01$ , thus a final time of  $T = 5$  for  $\epsilon = 0.1$ , and choose  $\sigma = \Delta x$  with  $\Delta x \in \{2/40, 2/80, 2/160\}$ . In Figure 5.6, one observes the two previously discussed regimes for the  $\mathcal{L}^1$  errors. The dashed lines refer to the error  $\mathcal{L}_{\epsilon, \Omega_S}^1$  with respect to the exact solution (5.1) whereas the solid lines show the error  $\mathcal{L}_{0, \Omega_S}^1$  w.r.t the limit solution (5.2). In the first phase of the simulation (short times), we recover Figure 5.5 (B), in particular the error  $\mathcal{L}_{0, \Omega_S}^1$  is large and  $\mathcal{L}_{\epsilon, \Omega_S}^1$  increases linearly. At a certain point, these two errors cross, and afterwards (at large times) the solution relaxes towards the weak limit solution (equilibrium plateau), and we recover Figure 5.5 (A), in particular, the error  $\mathcal{L}_{\epsilon, \Omega_S}^1$  is maximal at the end of the simulation, whereas  $\mathcal{L}_{0, \Omega_S}^1$  decreases towards the plateau. To understand Figure 5.6, one has to keep in mind that it corresponds to an intermediate  $\epsilon$ -value ( $\epsilon = 0.1$ ), and that the final simulation time is  $T = 5$ . The physical time is thus  $T_{phys} = 50$ , whereas in Figure 5.5 (B) the physical time was  $T_{phys} = 1$ , so Figure 5.6 combines intermediate  $\epsilon$ -values with long-times asymptotics, which

explains why one recovers there the two regimes of Figure 5.5, in particular, the short-time regime (Figure 5.5 (B)) and the long-time regime (Figure 5.5 (A)). In between, the AP-scheme proceeds to the transition between "non-stiff" and "stiff" regimes.

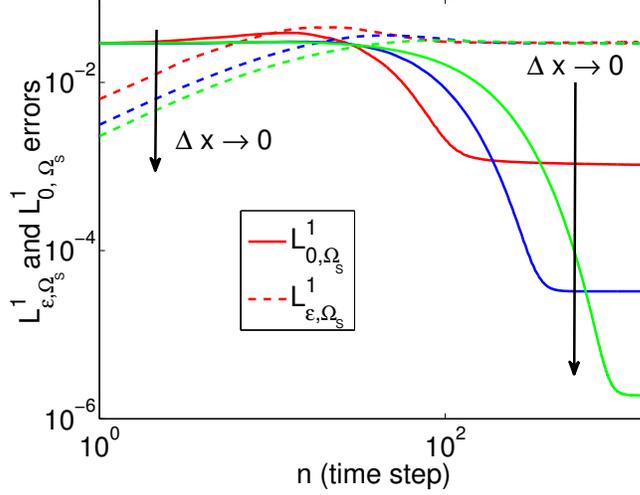


Fig. 5.6:  $\mathcal{L}^1_{\epsilon, \Omega_S}$  and  $\mathcal{L}^1_{0, \Omega_S}$  errors over time for the (DAMM)-scheme for several  $\Delta x$  with  $\epsilon = 0.1$ ,  $\sigma = \Delta x$ . Dashed lines refer to the error  $\mathcal{L}^1_{\epsilon, \Omega_S}$  whereas the solid lines show the error  $\mathcal{L}^1_{0, \Omega_S}$ .

**5.5. Choice of the stabilization parameter  $\sigma$ .** After having given some qualitative observations about the influence of the discretization parameter  $\sigma$  in different  $\epsilon$ -regimes, let us present some ideas to optimize the choice of  $\sigma$ . In Figure 5.7 (a), we plot the  $\mathcal{L}^1_{\epsilon, \Omega_S}$  error at the final time  $T$  for the non-limit regime  $\epsilon = 1$  as a function of  $\sigma$  for several values of  $\Delta x$ . In order to minimize the error  $\mathcal{L}^1_{\epsilon, \Omega_S}$ , the curves suggest to choose a  $\sigma$ -value depending on  $\Delta x$ . To investigate more precisely this dependence, we propose to take as "optimal"  $\sigma$ , for each fixed  $\Delta x$ , the value  $\sigma_h^\epsilon$  computed via

$$\sigma_h^\epsilon := \max \left\{ \sigma \in [\sigma_{min}, 1] \left| \frac{\|\Pi_h(f^\epsilon) - f_h^{\epsilon, \sigma}\|_{L_h^1(\Omega_S)} - \|\Pi_h(f^\epsilon) - f_h^{\epsilon, \sigma_{min}}\|_{L_h^1(\Omega_S)}}{\|\Pi_h(f^\epsilon) - f_h^{\epsilon, \sigma_{min}}\|_{L_h^1(\Omega_S)}} < \eta \right. \right\}, \quad (5.4)$$

(where  $\eta$  is an arbitrary precision) and evaluate how  $\sigma_h^\epsilon$  varies with  $\Delta x$ . In Figure 5.7 (b), we display  $\ln(\sigma_h^\epsilon)$  as a function of  $\ln(\Delta x)$  for  $\eta = 0.01$  and  $\sigma_{min} = 7e - 6$ . The data approach a polynomial line of slope  $p = 1.917$ , suggesting a relation between  $\sigma_h^\epsilon$  and  $\Delta x$  of the form  $\sigma_h^\epsilon = C \Delta x^p$  (with  $C > 0$ ), very close to the relation chosen in the last sections. To end this paragraph, let us briefly analyze the influence of  $\eta$ , the precision criterion, appearing in (5.4). For that, Table 5.2 presents for several values of  $\eta$ , the slope of the polynomial fitting of order 1 of the data  $\ln(\sigma_h^\epsilon) = f(\ln(\Delta x))$ , as well as its determination coefficient  $r^2$ , defined as the square of the correlation coefficient. One notes that the slope is around 2 in each case, validating the choice we have made before, for the stabilization parameter  $\sigma$  in the large  $\epsilon$ -regime, namely  $\sigma_h^\epsilon = (\Delta x)^2$ .

We perform a similar analysis in the limit regime  $\epsilon = 0$ . For that, we display in Figure 5.8 (a) the  $\mathcal{L}^1_{0, \Omega_S}$  error at the final time  $T$  as a function of  $\sigma$ , for several values of  $\Delta x$ . As in the non-limit  $\epsilon$ -regime, the curves suggest us to choose a  $\sigma$  depending on  $\Delta x$ . Let us define the application  $\mathcal{Z} : \sigma \mapsto \|\Pi_h(f^0) - f_h^{0, \sigma}\|_{L_h^1(\Omega_S)}$ . To study the dependence between  $\sigma$  and  $\Delta x$ , we choose for each  $\Delta x$  a  $\sigma_h^0$  defined by:

$$\sigma_h^0 := \arg \max_{\sigma \in [\sigma_{min}, 1]} \left| \frac{d\mathcal{Z}}{d\sigma} \right| (\sigma), \quad (5.5)$$

and we plot in Figure 5.8 (b) the evolution of  $\ln(\sigma_h^0)$  as a function of  $\ln(\Delta x)$ . The corresponding data

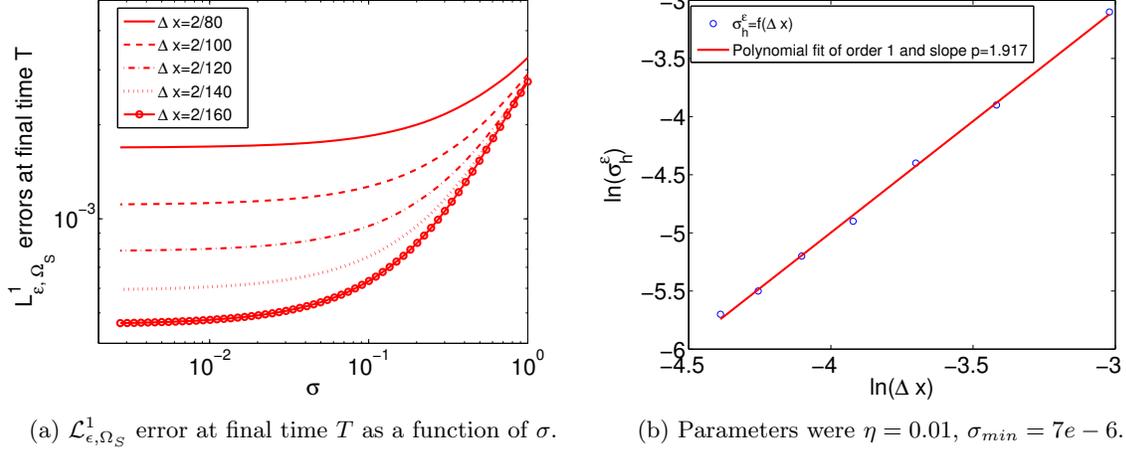


Fig. 5.7: (Non-limit case  $\epsilon = 1$ ). Study of the influence of  $\sigma$  in the  $\mathcal{L}_{\epsilon, \Omega_S}^1$  error (a) and of the relation between  $\sigma_h^\epsilon$  and  $\Delta x$  (b). Time discretization :  $T = 0.1$  and  $\Delta t = 0.01$ .

Table 5.2: Polynomial fitting of order 1 for the data  $\ln(\sigma_h^\epsilon) = f(\ln(\Delta x))$  for several values of the precision  $\eta$ . In each case, the slope  $p$  of the line and the determination coefficient  $r^2$  are written.

$\eta$	$p$	$r^2$
$1e - 1$	1.974	0.9975
$5e - 2$	1.974	0.9975
$1e - 2$	1.9175	0.9988
$5e - 3$	1.8675	0.9988
$1e - 3$	1.8597	0.9989

follow a linear relation, with a slope of  $p = 0.857$ , meaning that  $\sigma$  can be chosen as  $\sigma = \Delta x$  in the limit regime  $\epsilon = 0$  with the aim to reduce the  $\mathcal{L}_{0, \Omega_S}^1$  error and avoid a bad condition number.

To conclude this first numerical part, one can say that this simple test case permits to make a deep analysis of the (DAMM)-scheme. In particular, the AP behavior of the scheme was confirmed, the orders of convergence in both space and time were confirmed, and the influence of  $\sigma$  as well as its delicate choice have been intensively investigated. This study was enabled by the existence of analytic solutions of the problem, rigorously compared to solutions obtained by the (DAMM)-scheme for several  $\epsilon$ -regimes. Thanks to this verification, the (DAMM)-scheme can be used to resolve more complicated models where no analytic solutions are at hand. This is the topic of the next part.

**6. Numerical simulations for the Vlasov-Poisson test case.** The aim of this section is dual: firstly to solve numerically the Vlasov-Poisson system (1.5) using the (DAMM)-scheme and to simulate some particular physical phenomena (such as the Landau damping or the two-stream instability) ; and secondly to study the long-time asymptotics  $\epsilon \rightarrow 0$  of the two-stream instability. Note that the literature on the Vlasov-Poisson system is very rich, some theoretical as well as numerical results can be found in the non-exhaustive list [5, 13, 23, 25].

**6.1. The Vlasov-Poisson system and its numerical discretization.** In this chapter, we set  $\mathcal{Q} := (0, T) \times (0, L_x) \times (-L_v, L_v)$ . Using the Poisson bracket, the Vlasov-Poisson 1D1V system verified by  $f^\epsilon := f^\epsilon(t, x, v)$  reads

$$(VP)^\epsilon \begin{cases} \partial_t f^\epsilon + \frac{1}{\epsilon} \{f^\epsilon, \Psi^\epsilon\} = 0, \\ -\partial_{xx} \varphi^\epsilon(t, x) = 1 - n^\epsilon(t, x), \end{cases} \quad (6.1)$$

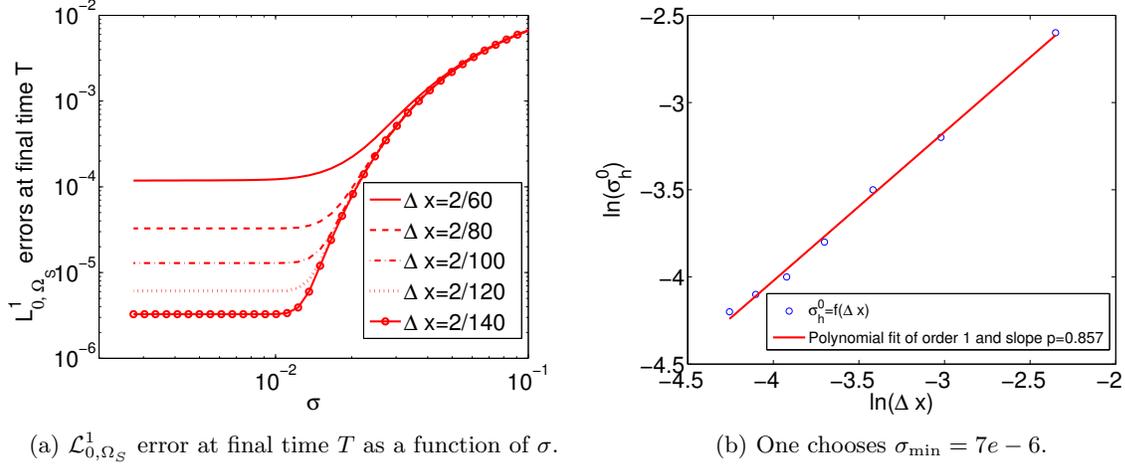


Fig. 5.8: (Limit case  $\epsilon = 0$ ). Study of the influence of  $\sigma$  in the  $\mathcal{L}_{0,\Omega_S}^1$  error (a) and of the relation between  $\sigma_h^0$  and  $\Delta x$  (b). Time discretization :  $T = 0.1$  and  $\Delta t = 0.01$ .

where  $\Psi^\epsilon(t, x, v) := v^2/2 - \varphi^\epsilon(t, x)$  is the stream-function and  $n^\epsilon(t, x) := \int_{\mathbb{R}} f^\epsilon(t, x, v) dv$  denotes the electron density. Due to the fact that this problem is non-linear (unlike the previous one), its study is a more delicate task, and one should think of a linearization procedure.

Following the same reformulation as before, we can construct an Asymptotic-Preserving scheme for the Vlasov-Poisson system by introducing an auxiliary variable  $q^\epsilon(t, x, v)$ . The AP-reformulation of (6.1) is then discretized with the help of the (DAMM)-scheme as before. The determination of the electric field  $E^\epsilon(t, x)$  is guaranteed by the resolution of the discrete Poisson equation. The fully discretized (first order in time) reformulated Vlasov-Poisson system is summarized here for clarity. For each time step  $n$ , one is looking for  $(f_h^{\epsilon,\sigma,n+1}, q_h^{\epsilon,\sigma,n+1})$ , by iterating in  $l \in \mathbb{N}$  like

$$(RVP)_h^{\epsilon,\sigma,n,l} \begin{cases} -\frac{\varphi_{i+1}^{\epsilon,\sigma,n+1,l} - 2\varphi_i^{\epsilon,\sigma,n+1,l} + \varphi_{i-1}^{\epsilon,\sigma,n+1,l}}{\Delta x^2} = 1 - \Delta v \sum_{j=1}^{N_v} f_{i,j}^{\epsilon,\sigma,n+1,l}, \\ \Psi_{i,j}^{\epsilon,\sigma,n+1,l} = \frac{1}{2} v_j^2 - \varphi_i^{\epsilon,\sigma,n+1,l}, \\ f_{i,j}^{\epsilon,\sigma,n+1,l+1} + \Delta t [q_h^{\epsilon,\sigma,n+1,l+1}, \Psi_h^{\epsilon,\sigma,n+1,l}]_{i,j} = f_{i,j}^{\epsilon,\sigma,n}, \\ [f_h^{\epsilon,\sigma,n+1,l+1}, \Psi_h^{\epsilon,\sigma,n+1,l}]_{i,j} = \epsilon [q_h^{\epsilon,\sigma,n+1,l+1}, \Psi_h^{\epsilon,\sigma,n+1,l}]_{i,j} - \sigma q_{i,j}^{\epsilon,\sigma,n+1,l+1}, \end{cases} \quad (6.2)$$

and starting from

$$f_{i,j}^{\epsilon,\sigma,n+1,0} := f_{i,j}^{\epsilon,\sigma,n}.$$

This iterative procedure has been performed to cope with the non-linearity of the Vlasov-Poisson system. In the following simulations, the stopping criterion for these iterations (at  $l = l_f$ ) is

$$\frac{\|f_{i,j}^{\epsilon,\sigma,n+1,l+1} - f_{i,j}^{\epsilon,\sigma,n+1,l}\|_{L_h^1}}{\|f_{i,j}^{\epsilon,\sigma,n+1,l}\|_{L_h^1}} + \frac{\|\varphi_i^{\epsilon,\sigma,n+1,l+1} - \varphi_i^{\epsilon,\sigma,n+1,l}\|_{L_h^1}}{\|\varphi_i^{\epsilon,\sigma,n+1,l}\|_{L_h^1}} < 10^{-2},$$

and we finish by posing

$$f_{i,j}^{\epsilon,\sigma,n+1} := f_{i,j}^{\epsilon,\sigma,n+1,l_f+1}.$$

REMARK 6.1. As briefly mentioned at the beginning of Section 2, in the Vlasov-Poisson test-case, the vector  $\mathbf{b} = \perp \nabla \Psi^\epsilon$  is time-dependent since  $\Psi^\epsilon$  depends on  $f^\epsilon$  via the Poisson equation. Nevertheless, the fixed point iteration procedure that we have employed in (6.2) permits to linearize the problem and during an iteration  $l \rightarrow l+1$  the stream-function  $\Psi^\epsilon$  is considered fixed, given by the previous iteration.

REMARK 6.2. Note that we wrote the previous system (6.2) without the DIRK time discretization in order to simplify its writing, however the following simulations have been implemented with the (DAMM)-scheme, including a DIRK time discretization.

**6.2. Numerical simulations for weak Landau damping and  $\epsilon = 1$ .** In order to validate our numerical procedure, we are interested in the Landau damping, for which analytic results are at hand. The Landau damping represents the exponential decrease of the electric field energy as a function of time (see [30, 35] for more details). For these simulations, the following initial data (see for example [13, 21]) is considered:

$$f_{in}(x, v) = \frac{1}{\sqrt{2\pi}} (1 + \gamma \cos(kx)) e^{-v^2/2}, \quad (6.3)$$

where  $\gamma$  refers to the amplitude and  $k$  to the mode of the perturbation of the homogeneous equilibrium  $\mathcal{M}(v) = (2\pi)^{-1/2} e^{-v^2/2}$ . In the following simulations, we take  $L_v = 10$ ,  $L_x = 2\pi/k$ ,  $\epsilon = 1$  and  $\sigma = (\Delta x/L_x)^2$ . In this section, we investigate the weak Landau damping, choosing a low amplitude of perturbation  $\gamma$ . According to [34], the weak Landau damping manifests for times  $t < 1/\sqrt{\gamma}$ . Beyond this time, the non-linear effects begin to be significant. Thus, we resolve the Vlasov-Poisson system (6.2) with the above initial condition (6.3) for  $\gamma = 0.001$  and  $k = 0.5$ . To simplify the notation, we shall denote in the following simply by  $f^\epsilon$  our numerical solution obtained by the (DAMM)-scheme.

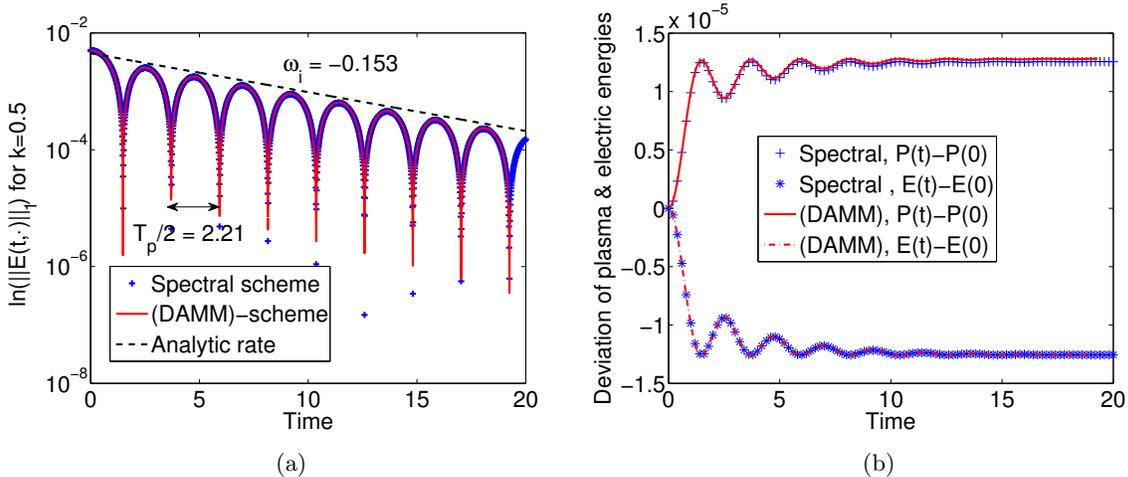


Fig. 6.1: (Weak Landau damping for  $\epsilon = 1$ ).  $L^1$ -norm of the electric field (in log-scale) versus time (a) and deviations of plasma and electric energies versus time (b) for both (DAMM) and spectral schemes.  $N_x = N_y = 256$ ,  $\Delta t = 0.01$ ,  $T = 20$ ,  $\sigma = (\Delta x/L_x)^2$ .

In Figure 6.1 (a) we represent the evolution in time of the  $L^1$ -norm of the electric field  $\|E^\epsilon(t, \cdot)\|_1$  (in *log*-scale) obtained from the (DAMM)-scheme. So as to validate efficiently our (DAMM)-scheme, we plot in the same Figure 6.1 (a) the corresponding evolution with a reference spectral scheme which resolves the system (6.1). The curves obtained from the two numerical schemes coincide perfectly. Moreover, we pay attention to the damping rate  $\omega_i$  and the frequency of oscillations  $\omega_p$ , which depend on the perturbation mode  $k$ . Under certain approximations, several formulae of these latter can be found (see for example [32]). One sees that both schemes approach the analytic values (for  $k = 0.5$ ), namely  $\omega_i = -0.153$  and  $\omega_p = 2\pi/T_p = 1.415$ . In Figure 6.1 (b), we plot the deviation (from their initial value) of both electric and plasma energies. These latter are defined as  $\mathcal{E}^\epsilon(t) := 1/2 \int_{-L_x}^{L_x} |E^\epsilon(t, x)|^2 dx$  and

$\mathcal{P}^\epsilon(t) := 1/2 \int_{\mathbb{R}} \int_{-L_x}^{L_x} v^2 f^\epsilon(t, x, v) dx dv$ . The curves indicate clearly that the total energy  $\mathcal{T}^\epsilon := \mathcal{E}^\epsilon + \mathcal{P}^\epsilon$  is conserved in compliance with the theory, for both (DAMM) and spectral schemes. Thus, the weak Landau damping is well simulated by the (DAMM)-scheme.

In Figure 6.2 we displayed the distribution function in phase-space at time  $t = 0$  (panel (a)),  $t = 20$  (panel (c)) and  $t = 40$  (panel (e)). Note that we have plotted the perturbed part of the distribution function  $f^\epsilon$ , meaning  $f^\epsilon - \mathcal{M}(v)$ . In the panels (b), (d), and (f), we represented the cross-sections at  $x = L_x/2$  of the previous plots, at the same times. These figures show us the continuous filamentation of  $f^\epsilon$  over time.

**6.3. Numerical simulations for strong Landau damping and  $\epsilon = 1$ .** We shall perform now the numerical simulations for the non-linear Landau damping by taking a stronger perturbation as in the previous study. Nevertheless, we stay in the non-limit regime  $\epsilon = 1$ . In Figure 6.3, we plot the distribution function  $f^\epsilon(t, x, v)$  at different times, with  $\gamma = k = 0.3$  and the initial data (6.3). Three levels can be pointed out. Up to  $t = 10$  (panel (a)), the linear effects dominate and the behavior of the electric energy is very close to the linear case. Then, starting from  $t = 20$  (panel (b)), the damping is stopped due to particle trapping, for finally leading to saturation at around  $t = 40$  (panel (d)). The phase-space trapping holes are clearly visible. In Figure 6.4, we plot the space average of the distribution function at several times. The formation of several plateaus is clearly visible at time  $t = 10$  (panel (b)), indicating the trapping of particles in these areas. Over time, this trapping persists, although the numerical diffusion tends to damp these states. Indeed, due to the numerical dissipation, the filamentation is progressively eliminated when the filamentation scale becomes smaller than the velocity grid  $\Delta v$ .

In Figure 6.5 (a), we plot the evolution of the electric energy (in log-scale) as a function of time. Contrary to the weak Landau damping, the growth or decay rates of the oscillations are not known. Nevertheless, we can compare the (DAMM)-scheme to the reference spectral scheme. We observe a good correspondence between these two schemes. In order to carry on the investigations of the strong Landau damping, we look at the evolution of some particular quantities. The Vlasov-Poisson system is well-known to conserve the total particle number (mass), the momentum, the total energy, the  $L^p$ -norms and the entropy. These quantities are given respectively by  $(\Omega_x := (0, L_x))$

$$M^\epsilon(t) := \int_{\mathbb{R} \times \Omega_x} f^\epsilon(t, x, v) dx dv, \quad (6.4)$$

$$\mathcal{M}_o^\epsilon(t) := \int_{\mathbb{R} \times \Omega_x} v f^\epsilon(t, x, v) dx dv, \quad (6.5)$$

$$\mathcal{T}^\epsilon(t) := \mathcal{E}^\epsilon + \mathcal{P}^\epsilon, \quad (6.6)$$

$$C_p^\epsilon(t) := \left( \int_{\mathbb{R} \times \Omega_x} |f^\epsilon(t, x, v)|^p dx dv \right)^{1/p}, \quad (6.7)$$

$$S^\epsilon(t) := \int_{\mathbb{R} \times \Omega_x} -f^\epsilon(t, x, v) \ln(f^\epsilon(t, x, v)) dx dv. \quad (6.8)$$

Due to the presence of the stabilization parameter  $\sigma$ , the conserved quantities introduced previously are no more constant over time when computed via the (DAMM)-scheme. We investigated this in the panels of Figure 6.5. In particular in the panel (b) we see that the total energy is not conserved by the (DAMM)-scheme with 3% of deviation from its initial value. Analogous observations can be done for the mass (panel (c)), the entropy (panel (d)) and the  $L^2$ -norm (panel (f)). Nevertheless, the (DAMM)-scheme conserves the momentum (which is null for the initial condition (6.3)), unlike the spectral scheme. Despite the non-conservation of these quantities, their deviations from their initial value stay weak.

Having carefully considered the Landau damping through several numerical simulations performed by the (DAMM)-scheme, we are interested now in the study of the two-stream instability. Since the Landau damping does not attain an equilibrium (due to the continuous filamentation), it is not suitable for investigating the limit regime  $\epsilon \rightarrow 0$ . As we will see, things are different in the case of the two-stream instability, which permit to investigate the limit  $\epsilon \rightarrow 0$ .

#### 6.4. Numerical simulations of the two-stream instability, study of the limit case $\epsilon \rightarrow 0$ .

**6.4.1. Two-stream instability in the non-limit case  $\epsilon = 1$ .** The two stream instability can be thought of as the inverse of the Landau damping. It occurs when the velocity of the particles is slightly greater than the wave velocity  $v_\phi$ . The instability causes a transfer of energy, from the particles to electric field, unlike the Landau damping phenomenon where the exchange of energy occurs from the electric field to the particles. Thus, to simulate this instability, one imposes the following initial data:

$$f_{in}^1(x, v) = \frac{1}{\sqrt{2\pi}} (1 + \gamma \cos(kx)) \frac{e^{-(v-3)^2/2} + e^{-(v+3)^2/2}}{2}. \quad (6.9)$$

As a first step, we keep  $\epsilon = 1$ . We choose  $\gamma = 0.001$ ,  $k = 0.2$ ,  $L_v = 10$ ,  $L_x = 2\pi/k$  and  $\sigma = (\Delta x/L_x)^2$ . In Figure 6.6, we plot the distribution function  $f^\epsilon(t, x, v)$ , solution of (6.1) at different times, with the previous initial condition (6.9). The panels (a) and (c) refer to the (DAMM)-scheme whereas the panels (b) and (d) correspond to the reference spectral scheme. In both cases, the instability grows until the non-linear effects become significant. Over time, the non-linear effects cause a trapping phenomenon. To push ahead with the investigations, we plot in Figure 6.7 the evolution over time of  $\ln(\|E^\epsilon(t, \cdot)\|_1)$  (a). The analytic value of the growth rate for the electric field, *i.e.*  $\omega_i(k = 0.2) = 0.2548$ , is very close to the numerical value observed and the curves obtained via the two numerical schemes (DAMM and spectral) coincide. As we made for the non-linear Landau damping, we examine in Figure 6.7 the conservation of several quantities over time, for both (DAMM) and spectral schemes. As for the non-linear Landau damping, only the momentum (panel (e)) is well conserved by the (DAMM)-scheme, the total energy (panel (b)), the mass (panel (c)), the entropy (panel (d)) and the  $L^2$ -norm (panel (f)) indicate weak deviations from their initial value. Nevertheless, we will see in the next section the main advantages of the (DAMM)-scheme, when compared to standard schemes.

**6.4.2. Two-stream instability in the limit regime  $\epsilon \rightarrow 0$ .** In order to conclude this section, we show that the AP-property of the (DAMM)-scheme can be useful when an equilibrium is reached. Recall that in the case of the Vlasov-Poisson system (6.1), passing to  $\epsilon \rightarrow 0$  is equivalent to passing to  $t \rightarrow \infty$ . Thus, the (DAMM)-scheme seems suitable to study the long-time behavior of the non-linear two-stream instability. In this part, we modify the initial condition, taking

$$f_{in}^2(x, v) = \frac{1}{\sqrt{2\pi}} v^2 e^{-v^2/2} (1 + \gamma \cos(kx)). \quad (6.10)$$

Although no rigorous proofs exist, the two-stream instability leads (in a certain weak sense) to a BGK (Bernstein-Greene-Kruskal) equilibrium after the growth phase. In Figure 6.8, the qualitative behavior of such equilibrium is visible. We have plotted the initial condition (6.10) in the panel (a), then we resolve the Vlasov-Poisson system with the (DAMM)-scheme for  $\epsilon = 0$ ,  $\sigma = (\Delta x/L_x)^2$ ,  $\Delta t = 0.01$ ,  $N_x = N_y = 256$ ,  $L_x = 2\pi/k$ ,  $L_v = 5$ , and the initial condition (6.10), with  $k = 0.5$  and  $\gamma = 0.05$ . From the first time iteration  $n = 3$  (panel (b)), the equilibrium seems to be attained and the filamentations are smoothed out. Note the formation of the separatrix which connects the saddle points at  $v = 0$  and  $x = 0 = 4\pi$ . Due to the topological conservation of the Vlasov-Poisson equation (see [27]), the distribution function keeps over time the nature and the number of its extrema. Panels (c) and (d) represent the distribution function  $f^\epsilon$  at the time iteration  $n = 15$  and  $n = 50$ , respectively. Note that the separatrix, clearly visible in the panel (b), is progressively smoothed out due to the numerical dissipation of the scheme. Besides, the value of the central extremum in  $(2\pi, 0)$  remains essentially constant in time. Thus, the (DAMM)-scheme conserves the nature and the position of this latter, meaning that the particle trapping is well-reproduced by our scheme. In Figure 6.9, we have plotted the contours of the distribution function  $f^\epsilon$  at the same times. We see clearly in the center the particle trapping on the panels (b), (c) and (d).

In order to confirm this BGK saturation, we shall check if the contours of the distribution function  $f^0$  are aligned with the contours of the stream-function  $\Psi^0 = v^2/2 - \varphi^0$ , as one expects that, in the limit  $\epsilon \rightarrow 0$ ,  $f^0$  depends only on  $\Psi^0$ . Thanks to the AP-property of our (DAMM)-scheme, we can obtain this equilibrium with a very low numerical cost, without too much numerical pollution. Few iterations are effectively needed to reach this equilibrium. To put into evidence the dependence  $f^0(\Psi^0)$ , we use

the fitting proposed by Heath and al [27], namely

$$f_{fit}^0 = a (\Psi^0 + \varphi_M) (\Psi^0 + \Psi^*) e^{-\beta \Psi^0}, \quad (6.11)$$

where  $a$  and  $\beta$  are fitting parameters to be found numerically,  $\varphi_M$  is the maximum of  $\varphi^0$  and  $\Psi^*$  is defined by

$$\Psi^* = \frac{\varphi_M - \beta \Psi_M \varphi_M + 2\Psi_M - \beta \Psi_M^2}{\beta \varphi_M + \beta \Psi_M - 1},$$

where  $\Psi_M$  is the value at which  $f^0$  attains its maximum. We choose  $a = 0.2948$  and  $\beta = 1.20$ . From the numerical simulation, we extract  $\varphi_M = 0.60$ ,  $\Psi_M = 0.93$  and thus  $\Psi^* = 0.90$ . In Figure 6.10, we plot the evolution of  $f^0(\Psi^0)$  as compared to the fitting distribution  $f_{fit}^0$ . Panel (a) represents  $f^0(\Psi^0)$  at time  $t = 0$ , clearly, there is no alignment between  $f^0$  and  $\Psi^0$ , as expected. The panels (c) and (e) which zoom the panel (a) in two regimes confirm this affirmation, we see clearly the non-functional structure of the plot (multi-valued function). However, in panel (b), we plot the same evolution but after fifty time iterations. One notes a very good correspondence between the numerical curve and the fitting one. The panels (d) and (f) show a good alignment of the points, showing that a BGK equilibrium is attained. Nevertheless, we observe an anormal inflexion of the curve near to the point  $f^0(\Psi^0 = 0)$ . A similar phenomenon was observed in [10], with a different numerical scheme. In the panel (f), we examine  $f^0(\Psi^0)$  near to its minimum. We pay attention here that there is no splitting phenomenon, confirming that the saturation is totally achieved.

To summarize, the (DAMM)-scheme permits, by passing to the limit  $\epsilon \rightarrow 0$ , to obtain a BGK equilibrium with a low number of iterations, permitting to control the accumulation of the errors. This is an essential advantage, as compared to standard schemes.

**7. Concluding remarks and perspectives.** The long-time behavior of the Vlasov-Poisson system is a challenging problem, requiring some investigations. Numerically, it is arduous to obtain a solution avoiding numerical pollution in such time asymptotics. We have developed an asymptotic-preserving scheme, based on a micro-macro decomposition coupled with a stabilization procedure in order to limit this difficulty. The analysis of the two-stream instability has shown the remarkable properties of the (DAMM)-scheme, permitting to attain a BGK-like equilibrium in few time-step iterations with low numerical costs and small errors. Nevertheless, the (DAMM)-scheme could be improved, notably through its stabilization part. The circle test case helped us a lot to better understand the choice of the stabilization parameter. But the Vlasov-Poisson test case shows that this parameter alters the conservation properties of the system. One may imagine for a future work to replace the stabilization parameter by a more general operator which could improve the conservation properties of the Vlasov-Poisson equation.

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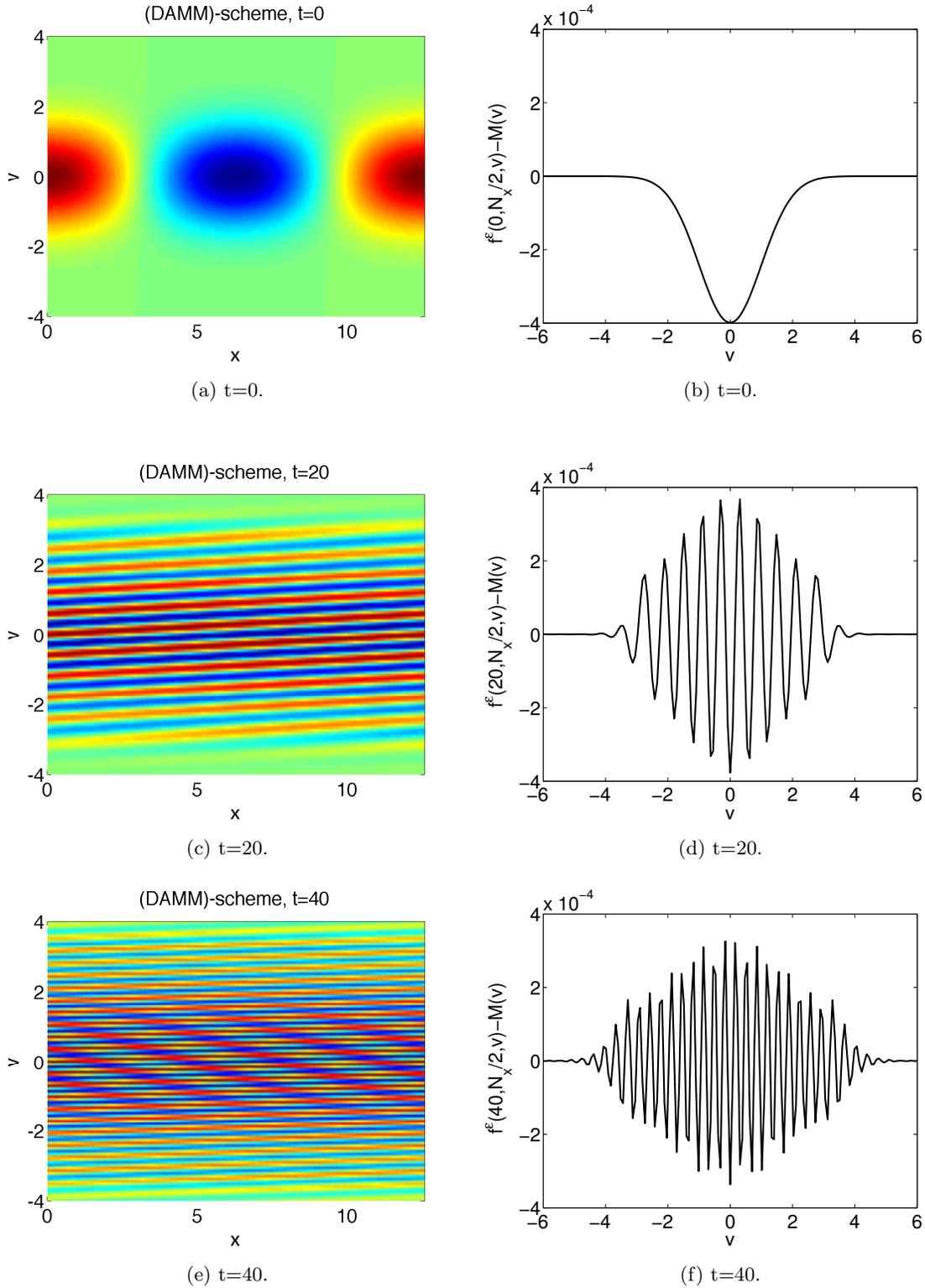


Fig. 6.2: (Weak Landau damping for  $\epsilon = 1$ ) Zoom of the distribution function  $f^\epsilon(t, x, v) - \mathcal{M}(v)$  at different times with  $k = 0.5$  and  $\gamma = 0.001$ , obtained with the (DAMM)-scheme (a), (c) and (e) and cross-sections at  $x = L_x/2$  of this latter (b), (d) and (f). Mesh size :  $N_x = N_y = 256$ . Other parameters were :  $\Delta t = 0.01$ ,  $T = 40$ ,  $\epsilon = 1$  and  $\sigma = (\Delta x/L_x)^2$ .

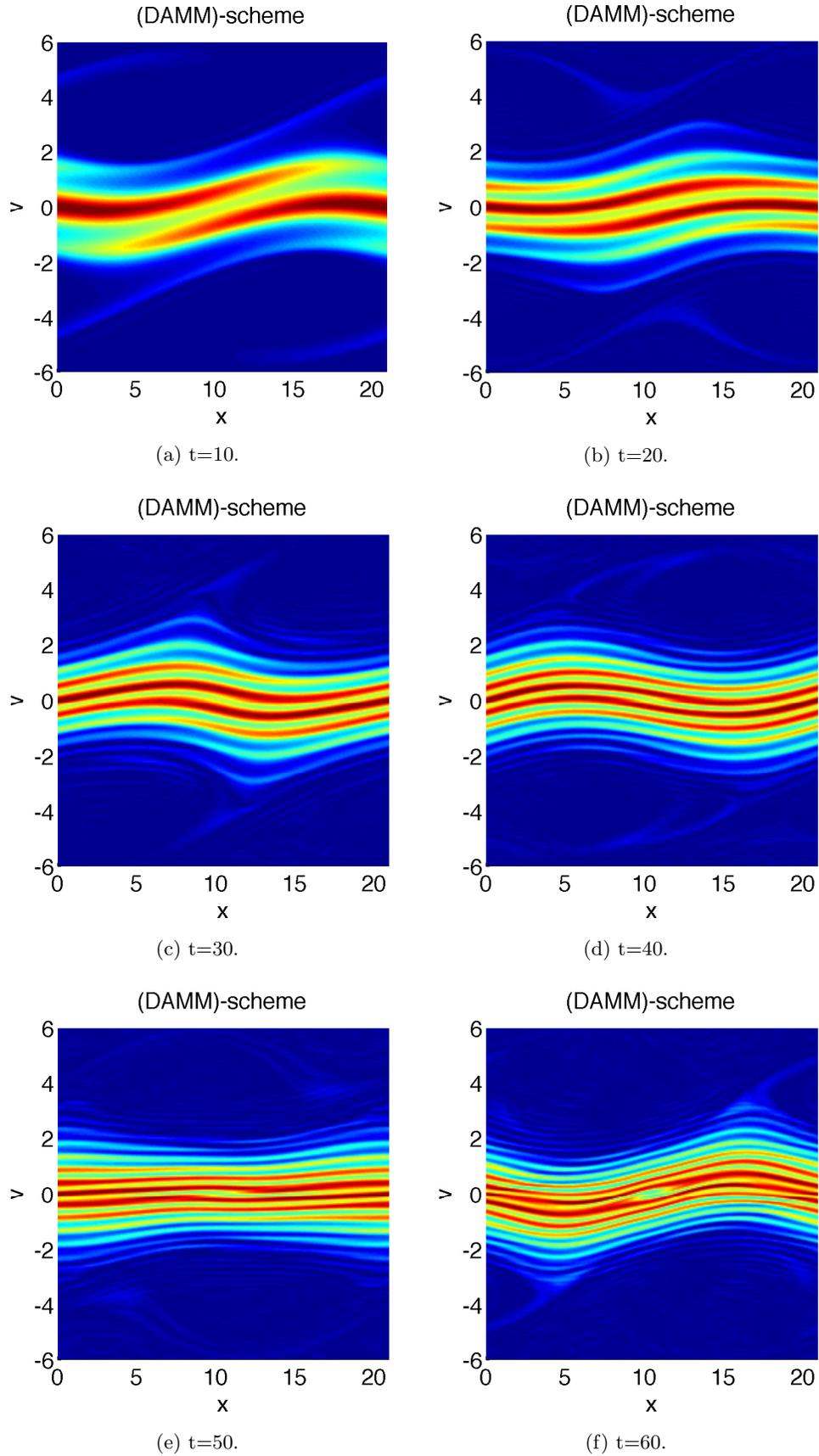
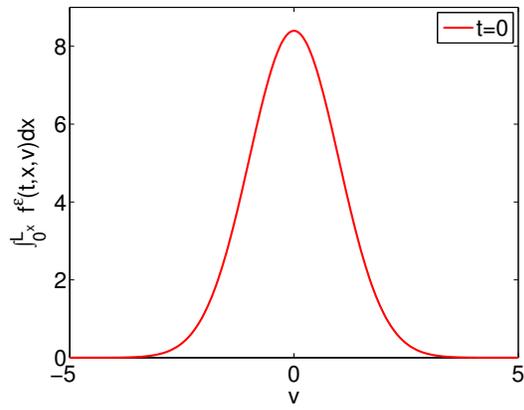
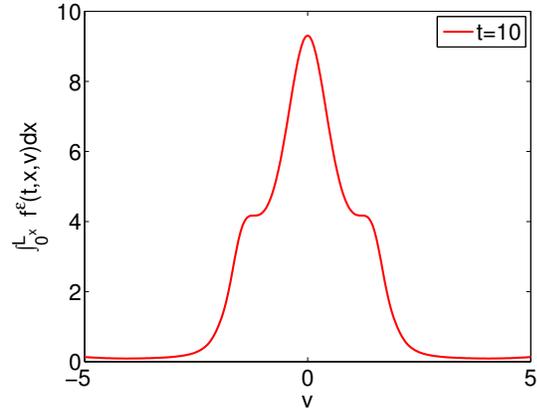


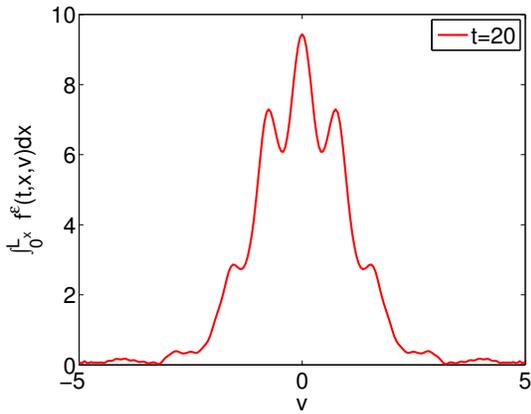
Fig. 6.3: (Strong Landau damping for  $\epsilon = 1$ ) Zoom of the distribution function  $f^\epsilon(t, x, v)$  at different times with  $k = \gamma = 0.3$ , obtained with the (DAMM)-scheme. Mesh size :  $N_x = N_y = 256$ . Parameters were :  $\Delta t = 0.01$ ,  $T = 60$ ,  $\epsilon = 1$  and  $\sigma = (\Delta x/L_x)^2$ .



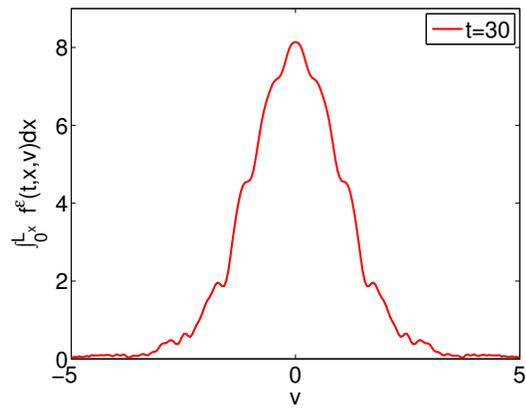
(a)  $t=0$ .



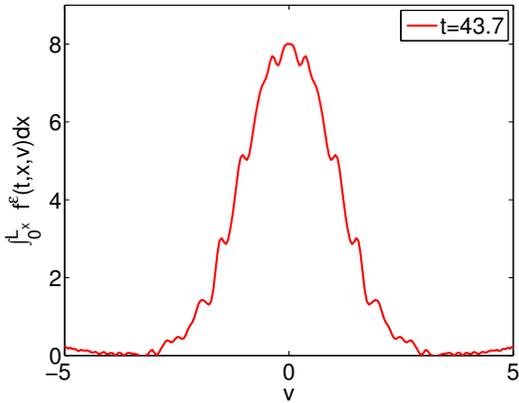
(b)  $t=10$ .



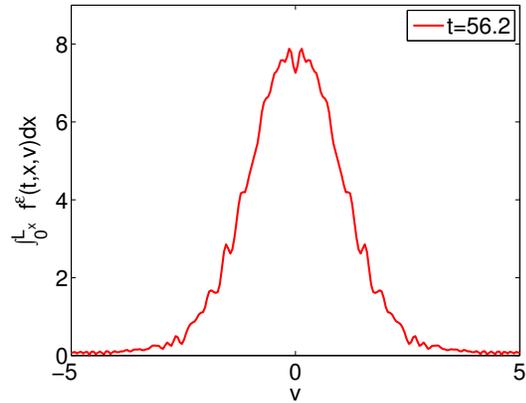
(c)  $t=20$ .



(d)  $t=30$ .

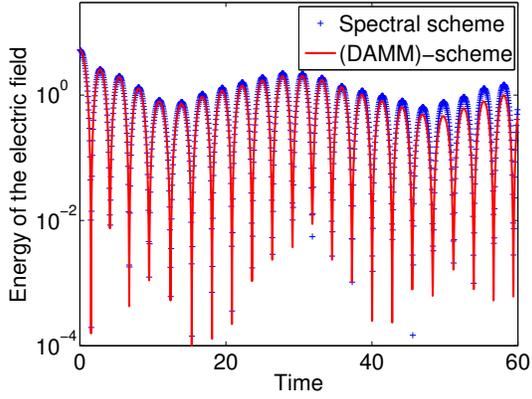


(e)  $t=43.7$ .

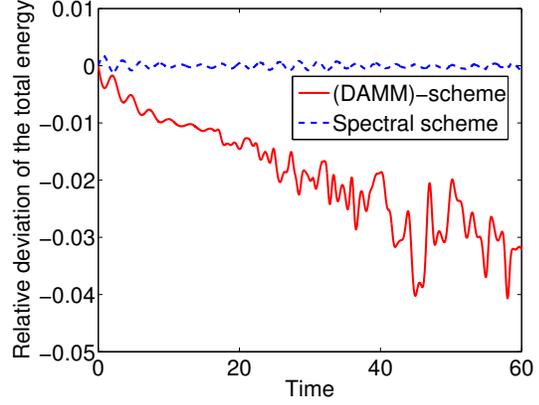


(f)  $t=56.2$ .

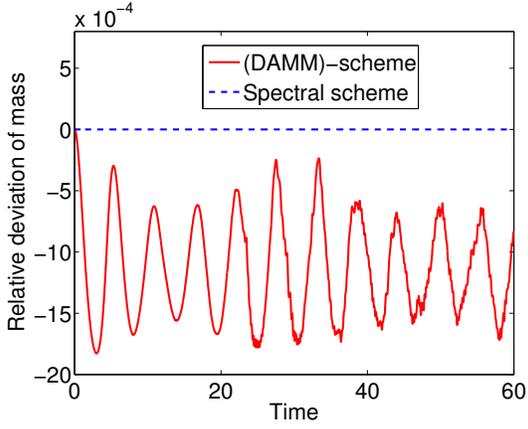
Fig. 6.4: (Strong Landau damping for  $\epsilon = 1$ ) Spatial average of the distribution function  $f^\epsilon(t, x, v)$  at different times with  $k = \gamma = 0.3$ , obtained with the (DAMM)-scheme. Mesh size :  $N_x = N_y = 256$ . Parameters were :  $\Delta t = 0.01$ ,  $T = 60$ ,  $\epsilon = 1$  and  $\sigma = (\Delta x/L_x)^2$ .



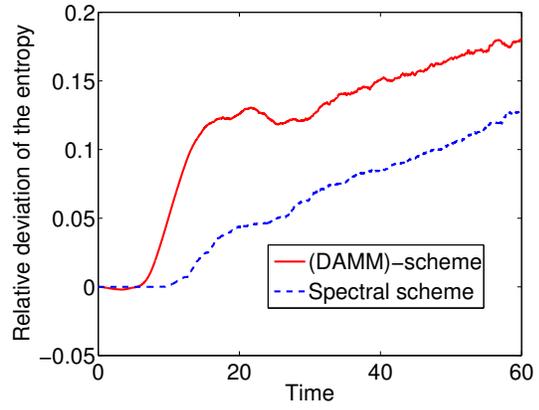
(a) Strong Landau damping.



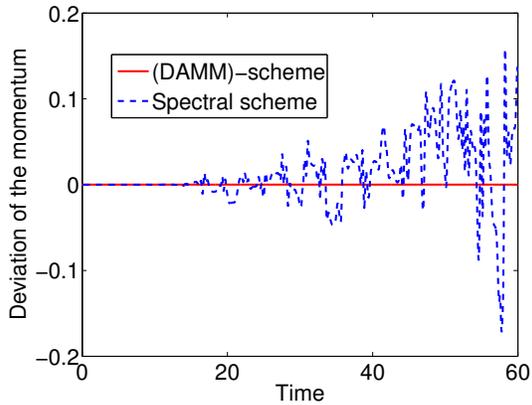
(b) Total energy.



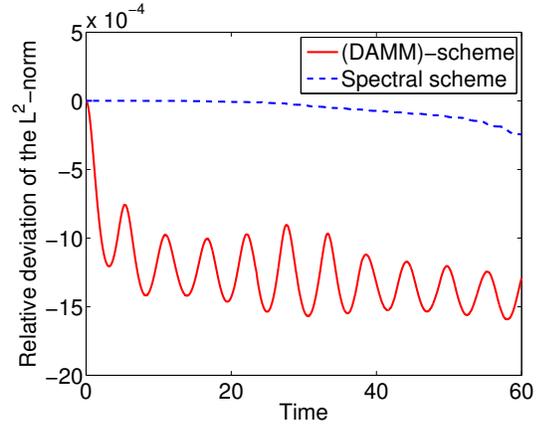
(c) Mass.



(d) Entropy.



(e) Momentum.



(f)  $L^2$ -norm.

Fig. 6.5: (Strong Landau damping for  $\epsilon = 1$ ) Energy of the electric field versus time (a) and deviation over time for both (DAMM) and spectral schemes of several quantities (b), (c), (d), (e) and (f). Mesh size:  $N_x = N_y = 256$ .  $T = 60$ ,  $\Delta t = 0.01$ ,  $\epsilon = 1$ , and  $\sigma = (\Delta x/L_x)^2$ .

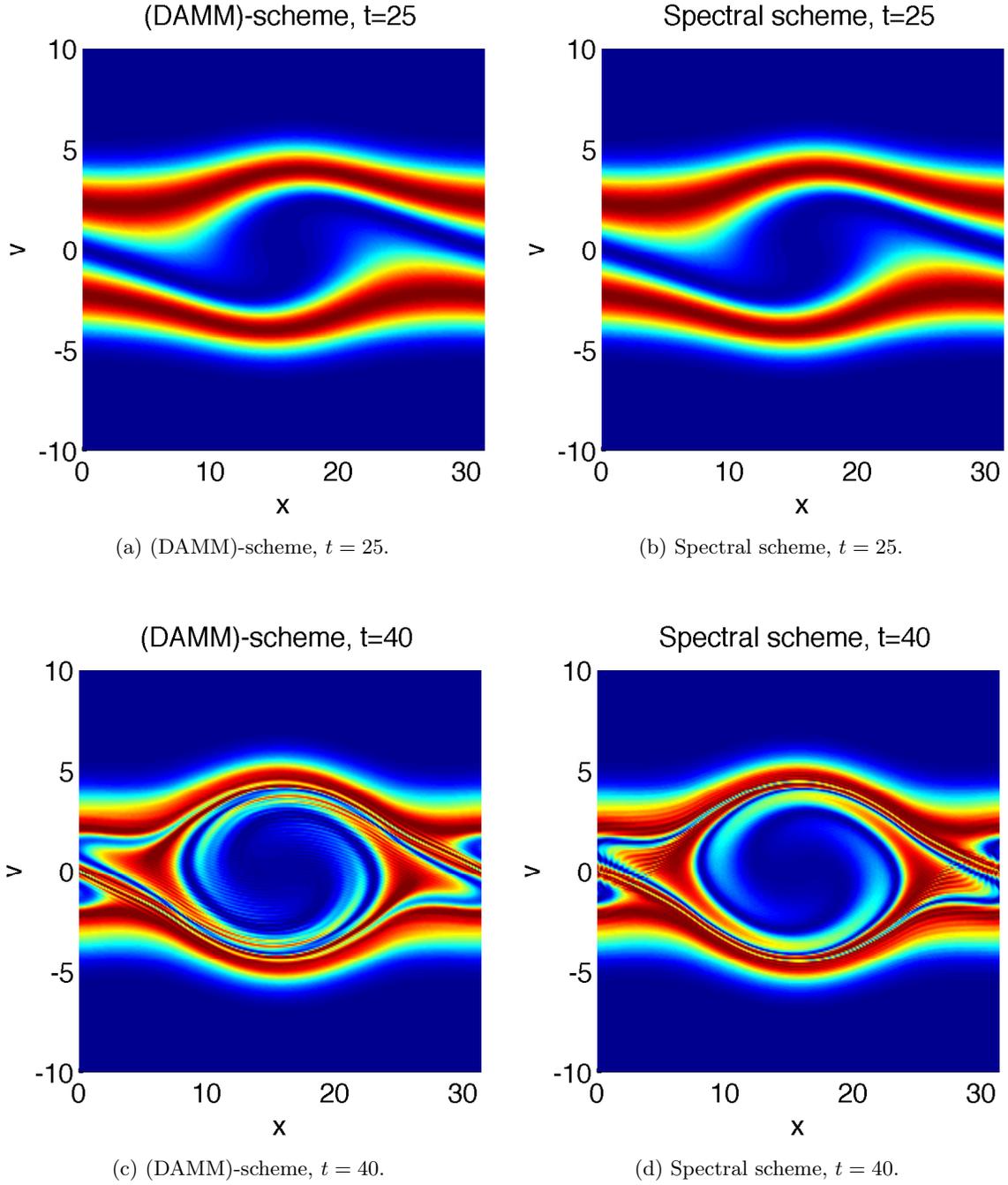
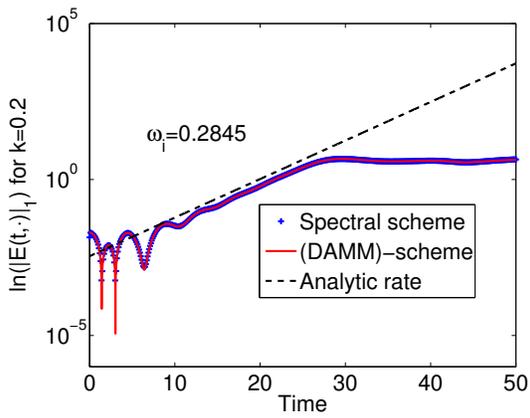
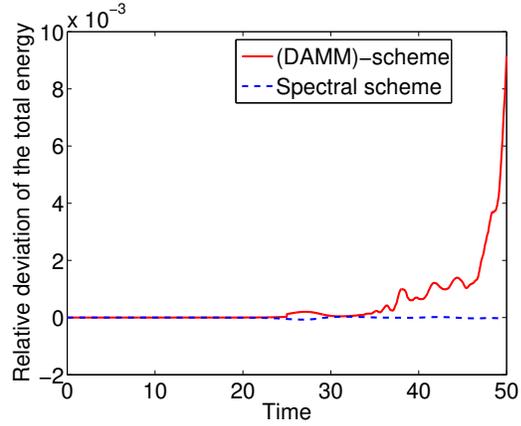


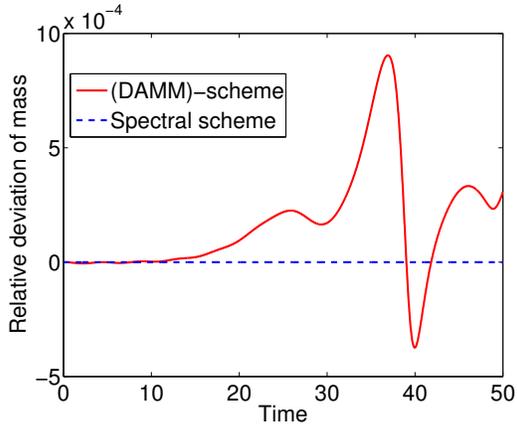
Fig. 6.6: (Two-stream instability for  $\epsilon = 1$  and  $f_{in}^1$ ) Distribution function  $f^\epsilon(t, x, v)$  at different times with  $k = 0.2$  and  $\gamma = 0.001$  for the (DAMM)-scheme ((a) and (c)) and the spectral scheme ((b) and (d)).  $T = 50$ ,  $N_x = 256$ ,  $N_y = 256$ ,  $\Delta t = 0.1$ ,  $\epsilon = 1$  and  $\sigma = (\Delta x/L_x)^2$ .



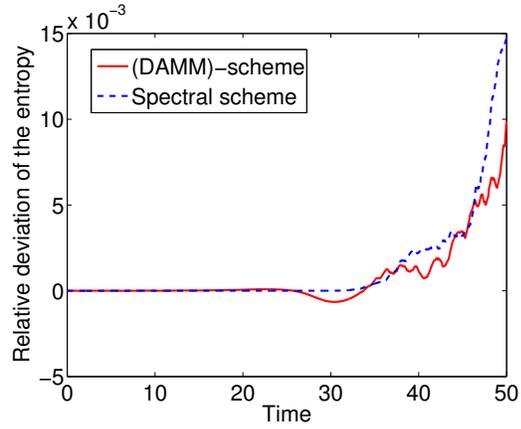
(a) Electric Field (in log value).



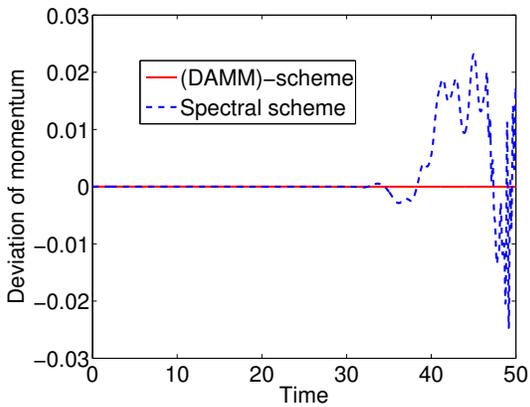
(b) Total energy.



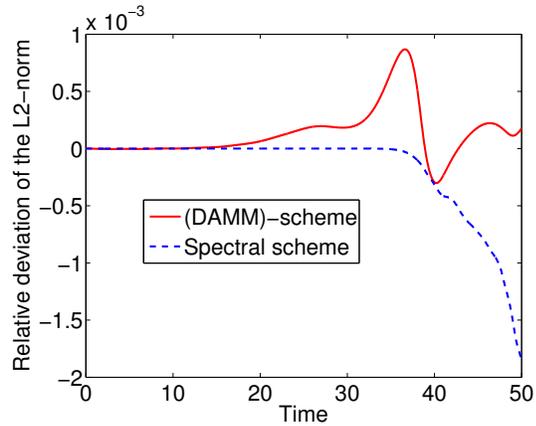
(c) Mass.



(d) Entropy.



(e) Momentum.



(f)  $L^2$ -norm.

Fig. 6.7: (Two-stream instability for  $\epsilon = 1$  and  $f_{in}^1$ ) Electric field versus time (a) and deviation over time for both (DAMM) and spectral schemes of several quantities (b), (c), (d), (e) and (f). Mesh size:  $N_x = N_y = 256$ .  $T = 50$ ,  $\Delta t = 0.01$ ,  $\epsilon = 1$ , and  $\sigma = (\Delta x/L_x)^2$ .

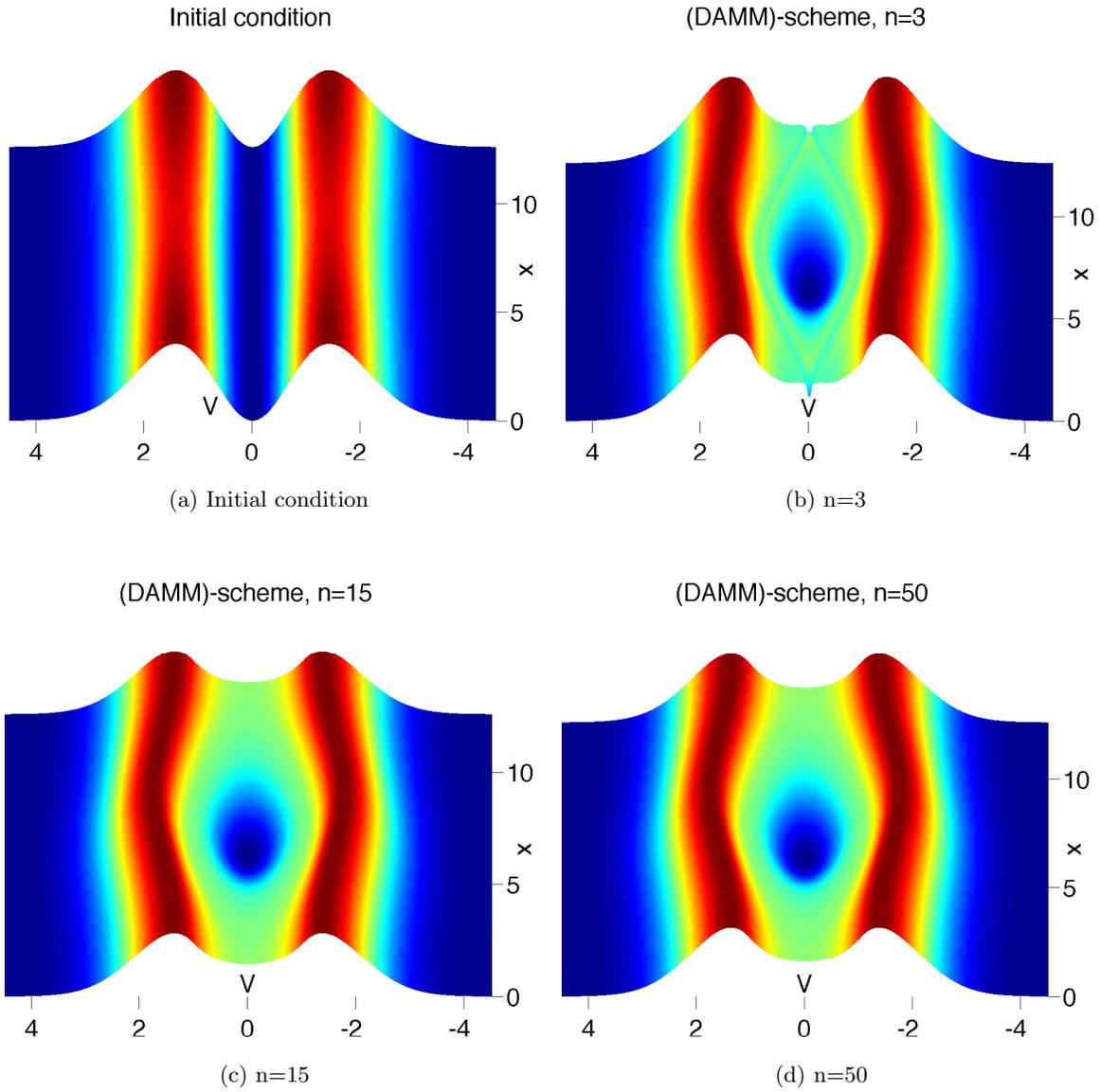


Fig. 6.8: (Two-stream instability for  $\epsilon = 0$  and  $f_{in}^2$ ) Distribution function  $f^0(t, x, v)$  at different times for the two stream instability with  $k = 0.5$  and  $\gamma = 0.05$  via the (DAMM)-scheme. Parameters were  $N_x = N_y = 256$ ,  $\Delta t = 0.1$ , and  $\sigma = (\Delta x/L_x)^2$ .

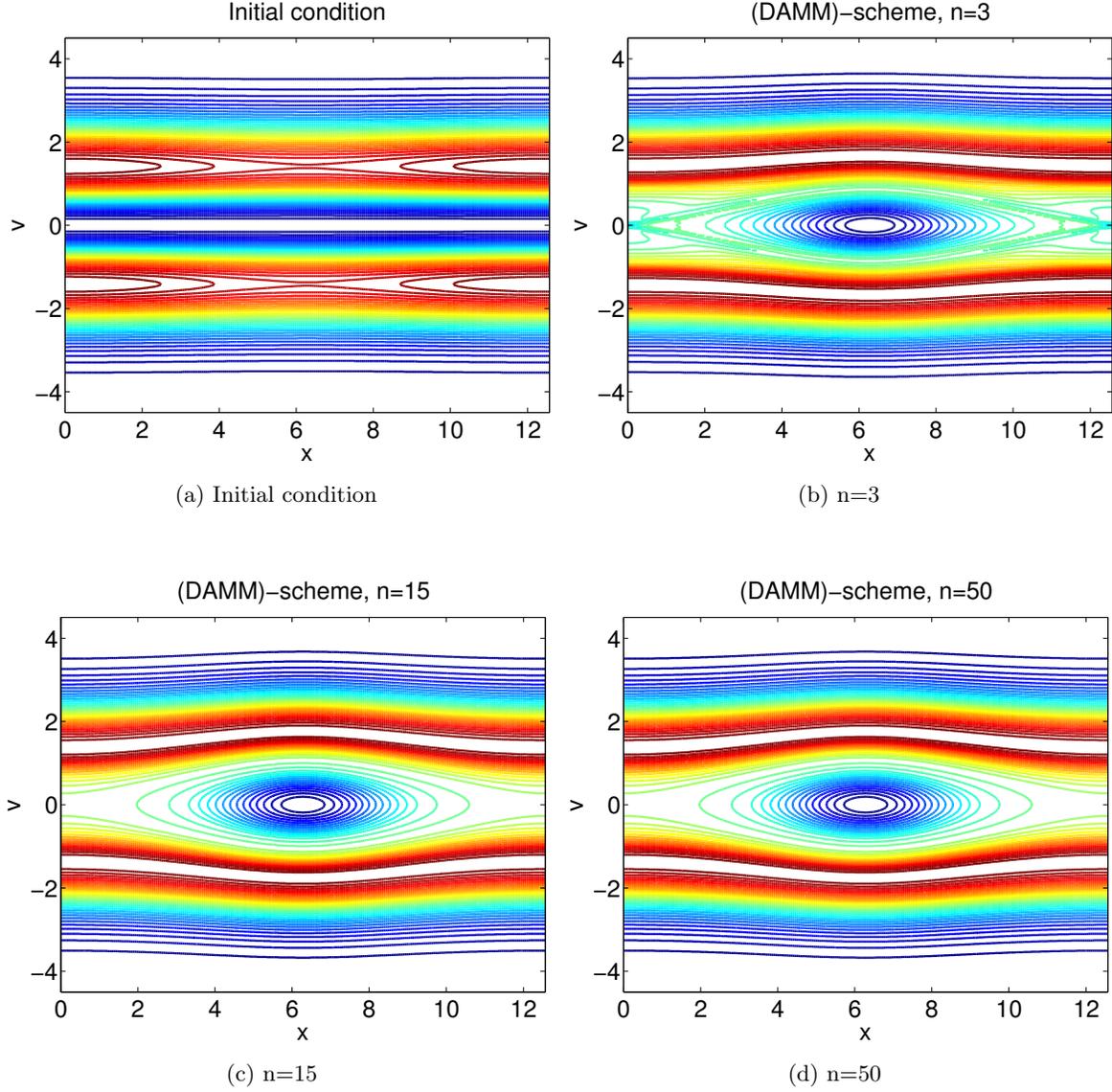


Fig. 6.9: (Two-stream instability for  $\epsilon = 0$  and  $f_{in}^2$ ) Contour plots of the distribution function  $f^0(t, x, v)$  at different times for the two stream instability with  $k = 0.5$  and  $\gamma = 0.05$  via the (DAMM)-scheme. Parameters were  $N_x = N_y = 256$ ,  $\Delta t = 0.1$ , and  $\sigma = (\Delta x/L_x)^2$ .

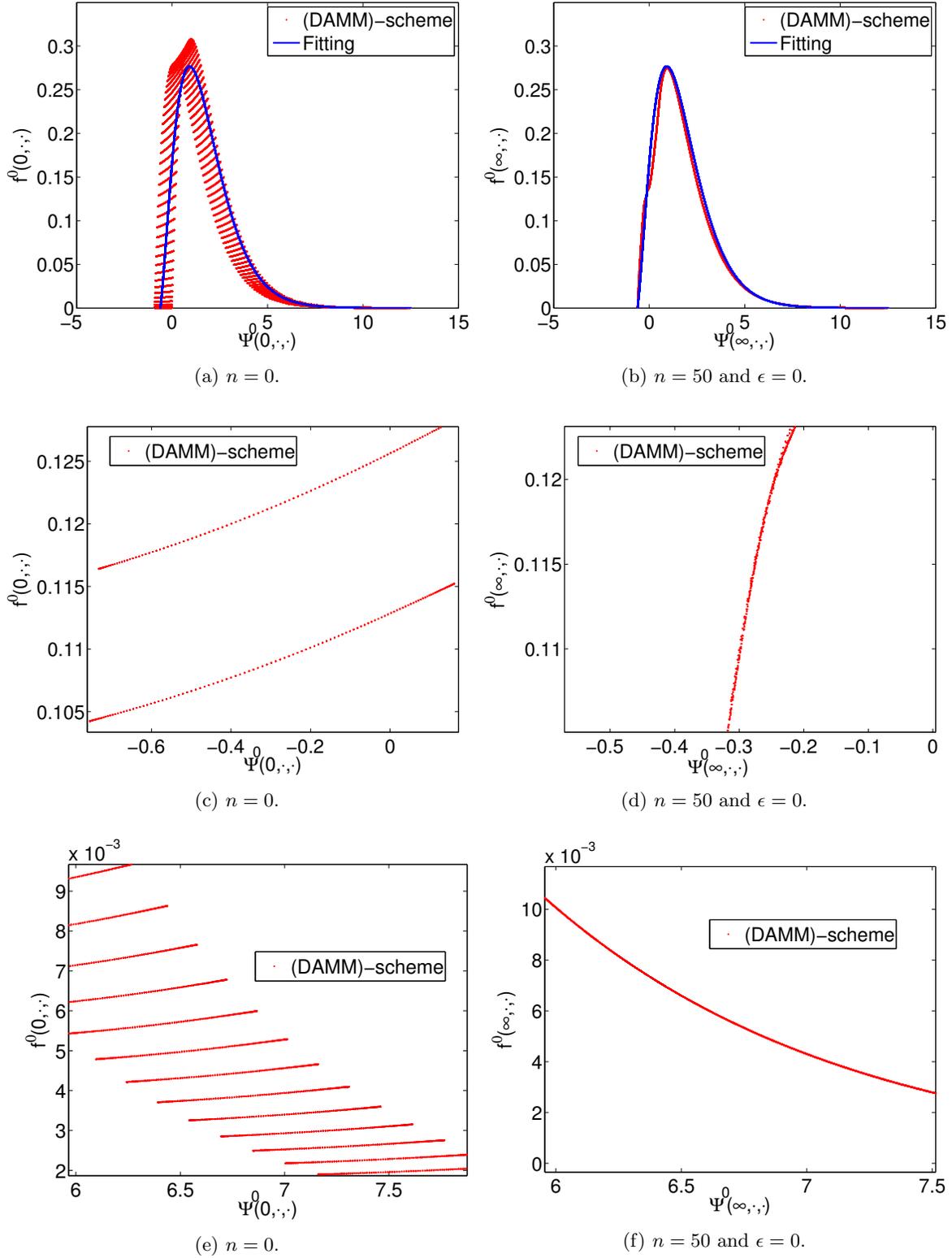


Fig. 6.10: (Two-stream instability for  $\epsilon = 0$  and  $f_{in}^2$ ) Plot of the distribution function at times  $t = 0$  (panels (a), (c), (e)) and  $t = 50 \Delta t$  (panels (b), (d), (f)), with  $\Delta t = 0.01$  as a function of  $\Psi^0$ . Mesh size:  $N_x = N_y = 256$ . Stabilization parameter :  $\sigma = (\Delta x/L_x)^2$ .