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Positivity-preserving scheme for two-dimensional advection-diffusion equations including mixed derivatives

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Positivity-preserving scheme for two-dimensional advection-diffusion equations including mixed derivatives

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Abstract

In this work, we propose a positivity-preserving scheme for solving two-dimensional advection-diffusion equations including mixed derivative terms, in order to improve the accuracy of lower-order methods. The solution of these equations, in the absence of mixed derivatives, has been studied in detail, while positivity-preserving schemes for mixed derivative terms have received much less attention. A two-dimensional diffusion equation, for which the analytical solution is known, is solved numerically to show the applicability of the scheme. It is further applied to the Fokker-Planck collision operator in two-dimensional cylindrical coordinates under the assumption of local thermal equilibrium. For a thermal equilibration problem, it is shown that the scheme conserves particle number and energy, while the preservation of positivity is ensured and the steady-state solution is the Maxwellian distribution.

Keywords: Advection-diffusion, Fokker-Planck equation, low-order positivity-preserving scheme

1. Introduction

Two-dimensional advection-diffusion equations have widespread applications in physics, engineering and finance, and can generally be written as

\[ u_t = A u_{xx} + B u_{xy} + C u_{yy} + D u_x + E u_y + F u \]  

where \( u = u(x, y, t) \). As these equations are often too difficult to solve analytically, numerical solutions are required. For \( F = 0 \), these equations can be written in a two-dimensional advection-diffusion form,

\[ \frac{\partial u}{\partial t} = \nabla \cdot (\hat{a} \cdot \nabla u) \]  

where \( u = u(x, y, t) \) is advected by the 2D vector \( \hat{a}(x, y, t) \) and diffused by the tensor \( \hat{k}(x, y, t) \). A particular application of two-dimensional advection-diffusion equations is the Fokker-Planck collision operator, which can be written in the form (2) and has a wide range of applications in plasmas in the laboratory (e.g. magnetic and inertial thermonuclear fusion), space (e.g. Earth’s magnetosphere), and astrophysics (e.g. solar coronal mass ejections) [1].

If the initial condition \( u(x, y, 0) \geq 0 \) holds for all \( (x, y) \), then the solution must always be positive, i.e. \( u(x, y, t) \geq 0 \) for all \( (x, y, t) \) when \( F = 0 \). Moreover, a good numerical method has to preserve the monotonicity of the initial condition. The conservation of such properties poses a particular challenge if a change of coordinates, adopted to eliminate the mixed derivative terms \( u_{xy} \) throughout \( (x, y) \) space, is not possible.

Positivity-preserving schemes for two-dimensional advection-diffusion equations, in the absence of mixed derivative terms, have been studied in detail [2, 3], but schemes for problems where mixed derivative terms are present have received much less attention. Recent research has focused on developing improved and refined higher-order methods for solving advection-diffusion equations, as these methods have the potential of providing accurate solutions at reasonable cost. Implementing these methods, however, can be complicated. In contrast, lower-order methods, while being less accurate, are easier to implement and generally more robust and reliable, and are therefore routinely employed in practical calculations [4].

In this paper, we propose a scheme for improving the accuracy of lower-order methods, in particular with respect to the preservation of positivity, when solving two-dimensional advection-diffusion equations in the presence of mixed derivatives, as lower-order methods are routinely used for solving these equations, yet fail to preserve...
In our proposed scheme, in order to preserve positivity, the mixed
derivative term is rewritten as an advective equation, for which many positivity-preserving approximations exist
[2, 3, 4, 5, 6, 7]. We show, using an example, that this scheme exhibits the same order of accuracy as central-finite
difference methods, while ensuring the preservation of positivity. The scheme is then applied to the Fokker-Planck
collision operator in cylindrical coordinates.

One-dimensional solutions to the Fokker-Planck collision operator have been studied for decades, but detailed
numerical discretizations in two-dimensions, particularly in cylindrical coordinates where mixed derivative terms are
present, have only recently been studied. A numerical approximation to the Fokker-Planck collision operator should
ensure the conservation of particle number, momentum and energy, the preservation of positivity and a steady-state
Maxwellian distribution under thermal equilibration. The most successful approaches consist of an extension of
the one-dimensional Chang and Cooper scheme [8] to two-dimensions, but this method does not guarantee the
preservation of positivity if the solution is far from equilibrium [9]. A fully implicit finite element algorithm, using
appropriate flux limiters to ensure the preservation of positivity, the conservation of particle number, momentum,
and energy, has also been developed, but is intensive [1]. A good review of other numerical methods can be found
in [1].

Here, we present an alternative approach to the Fokker-Planck collision operator in two-dimensional cylindrical
coordinates. The solution is based on the proposed positivity-preserving scheme and extends the Chang and
Cooper scheme, based on the assumption of local thermal equilibrium, to ensure an accurate steady-state solution
is obtained. We approximate the collision operators by assuming the distribution collides with a background
Maxwellian distribution, and it is shown that, if this approximation holds, thermal equilibration occurs at the
theoretically predicted rate. The proposed scheme conserves particle number and energy, while the preservation of
positivity is ensured.

The paper is structured as follows: Section 2 introduces our proposed positivity-preserving scheme for solving
two-dimensional diffusion equations, including the treatment of mixed derivative terms. Section 3 discusses the
Fokker-Planck collision operator in cylindrical coordinates, including the assumption of local thermal equilibrium
and thermal equilibration tests, and is followed by a short summary in Section 4.

2. 2D diffusion with mixed derivatives

Positivity-preserving schemes to two-dimensional advection-diffusion equations have been studied in detail [2, 3],
but solving problems with mixed derivative terms have received much less attention. The reason for this is that
typically a change of coordinate system can be performed in order to eliminate the mixed derivative terms, or the
mixed derivative terms are weak compared to the advection-diffusion terms and can therefore be neglected. If this
is not possible, however, a positivity-preserving scheme is required for the unmodified equation.

Consider, as an example, the two-dimensional diffusion equation

\[ u_t = u_{xx} + u_{xy} + u_{yx} + u_{yy} \] (3)

with initial condition

\[ u(x, y, t = 0) = \exp(-x^2 - y^2) \]

and open boundary conditions, such that the discretized domain on which we solve the problem must be large
enough to ensure \( u = 0 \) at the boundaries always. Typically, this equation will be solved by performing a change
of coordinates in order to eliminate the mixed derivative terms \( u_{xy} \) and \( u_{yx} \). In this way, an analytical solution can
be obtained,

\[ U(x, y, t) = \frac{1}{\sqrt{1 + 8t}} \exp \left[ -\left( \frac{1}{2}x^2 + \frac{1}{2}y^2 - xy \right) - \frac{1}{1 + 8t} \left( \frac{1}{2}x^2 + \frac{1}{2}y^2 + xy \right) \right] \] (4)

The aim of this section is to introduce a low-order positivity-preserving scheme for solving problems with mixed
derivative terms, thereby solving equation (3) without a change of coordinate system. The numerical approximation
\( u_{i,j} \) is then compared to the analytical solution through an RMS error given by

\[ E_{rms} = \sqrt{\frac{1}{N^2} \sum_{i,j} (U_{i,j} - u_{i,j})^2} \] (5)

where \( N \) is the number of grid points in both the \( x \)- and \( y \)-directions and the labels \( i, j \) refers to the \( i^{th} \) node in
the \( x \)-direction and the \( j^{th} \) node in the \( y \)-direction.
Although higher-order methods for approximating (3) can be obtained with the use of flux limiters, these are often complicated and less robust than lower-order methods. Lower-order methods, on the other hand, tend to be less accurate, and, as will be shown, do not guarantee the preservation of positivity. Our proposed scheme improves the accuracy of lower-order methods, especially by ensuring the preservation of positivity. For this reason, we only consider two second-order accurate finite-difference methods. The diffusion terms \(u_{xx}\) and \(u_{yy}\) are straightforward to approximate by employing central finite differences, which are second-order accurate and preserve positivity. For instance, the derivative \(u_{xx}\) is discretized as follows,

\[
u_{xx} = \frac{\partial}{\partial x} \frac{\partial u}{\partial x} = \frac{u_{i+1,j} - 2u_{i,j} + u_{i-1,j}}{\Delta x^2} + O(\Delta x^2)
\]  

(6)

and is second-order accurate and preserves positivity.

The mixed derivatives can be discretized with a similar central finite-difference method, where the boundary values are determined as an average, i.e.

\[
u_{i+1/2} = \frac{1}{2} \left( u_{i} + u_{i+1} \right)
\]  

(7)

but, although such a method is second-order accurate, it does not guarantee positivity.

Positivity-preserving schemes exist for linear advection equations, and we therefore rewrite the mixed derivative terms as advection equations in order to employ these schemes and preserve positivity.

2.1. Positivity-preserving scheme for the advection equation

Positivity-preserving schemes for linear advection equations have been studied in detail and various schemes for approximating these equations exist [2, 3, 5, 6, 7]. In this work the approach of Hundsdorfer [3], which results in a second-order positivity-preserving approximation to the linear advection equation, will be used.

To illustrate this positivity-preserving scheme, consider the linear advection equation,

\[
u_t + (au)_x = 0
\]  

(8)

which can be discretized,

\[
(au)_x = \frac{1}{\Delta x} (F_{i+1/2} - F_{i-1/2})
\]

where \(F_{i\pm1/2} = (au)_{i\pm1/2}\) and the values \(a_{i\pm1/2}\) are assumed known. The values of the function \(u\) are therefore required at the node boundaries \(i\pm1/2\). In a central finite-difference scheme, the boundary values can be determined as an average of neighbouring nodes (7), but this scheme does not preserve positivity in two-dimensions.

An alternative scheme is the donor-cell-upwind (DCU) scheme, which is a first-order flux-conserving scheme. Consider the discretized linear advection equation with \(a > 0\),

\[
u_t = \frac{1}{\Delta x} (F_{i+1/2} - F_{i-1/2})
\]

(9)

where the first term on the right-hand side describes a flow across the boundary \(i + 1/2\) out of node \(i\), while the second-term describes a flow across the boundary \(i - 1/2\) into node \(i\). To first-order, this flow can be approximated with

\[
u_t = -\frac{1}{\Delta x} \left( a_{i+1/2}u_i - a_{i-1/2}u_{i-1} \right) + O(\Delta x)
\]

where the first term on the right-hand side describes the decrease in the amount of fluid at node \(i\), and the second term describes the increase in the amount of fluid at node \(i\) due to the inflow from node \(i - 1\). This scheme, known as the DCU scheme, is first-order accurate and preserves positivity, while having the added advantage of keeping the solution monotonically varying in regions where it should be monotonic, even though the accuracy may not be very good [2].

The idea is therefore to find a second-order approximation to the flux function \(F_{i\pm1/2}\), as second-order methods give much better accuracy on smooth oscillations. However, second-order methods often fail near discontinuities, where oscillations may appear due to their dispersive nature. The best approach would therefore be to combine the favourable qualities of both first- and second-order methods, such that the second-order method ensures better accuracy in regions where the solution is smooth, while the first-order method ensures the preservation of monotonicity in regions where the solution is discontinuous [2]. This is done by introducing the flux limiter \(\phi_{i\pm1/2}\), and
writing the flux functions \( F_{i+1/2} \), for \( a > 0 \), as

\[
F_{i+1/2} = a_{i+1/2} \left( \frac{u_i + \frac{1}{2} \phi_{i+1/2}(u_i - u_{i-1})}{u_i - u_{i-1}} \right)
\]

\[
F_{i-1/2} = a_{i-1/2} \left( \frac{u_{i-1} + \frac{1}{2} \phi_{i-1/2}(u_{i-1} - u_{i-2})}{u_{i-1} - u_{i-2}} \right)
\]

(10)

If \( \phi = 0 \), the DCU scheme (9) is recovered, and the scheme is first-order accurate. For \( \phi = 1 \), the scheme is second-order accurate, but does not preserve monotonicity in regions where the solution is discontinuous [2]. The flux limiter helps to ensure that we get the best of both second- and first-order methods. There exists a number of flux limiter functions (see, for example [2]), but in this work we use the flux limiter introduced by Hundsdorfer [3], which gives

\[
\phi_{i+1/2} = \max(0, \min(2r, \min(2, K(r))))
\]

(11)

where

\[
r_{i+1/2} = \frac{u_{i+1} - u_i}{u_i - u_{i-1}}
\]

(12)

and

\[
K(r) = \frac{1 + 2r}{3}
\]

(13)

which provides a second-order accurate approximation to the advection equation (8) that preserves positivity in two dimensions, while also preserving monotonicity in regions where the solution is discontinuous.

### 2.2. Positivity-preserving scheme for mixed derivative terms

Positivity-preserving schemes for linear advection equations have been studied in detail [2, 3, 5, 6, 7], while such approximations for problems with mixed derivative terms have received much less attention. As central finite-difference methods for equations including mixed derivatives do not preserve positivity, we rewrite these terms as advection equations, and then apply a positivity-preserving approximation. In order to rewrite the mixed derivative terms as advection equations, we define the function

\[
v = \frac{1}{u} \frac{\partial u}{\partial y}
\]

(14)

so that we can write

\[
 u_{x'y} = \frac{\partial}{\partial x} \frac{\partial}{\partial y} = \frac{\partial}{\partial x} v \cdot u
\]

(15)

where \( v = v(x, y, u) \). In this way, the mixed derivative terms can be written as non-linear advection equations, which can be discretized using positivity-preserving schemes, such as the second-order scheme of Hundsdorfer [3], introduced in Section 2.1. Higher-order schemes exist, but as these schemes are more complicated and less robust than lower-order methods, and as we aim to compare our numerical approximation to a second-order central finite-difference method, we use a second-order positivity-preserving scheme.

The non-linearity can be resolved through Picard iteration, discussed later. A more straightforward solution is through Picard linearizing, where the function \( v \) is determined from the known values of \( u^n \) in order to find the solution \( u^{n+1} \) at the next time step.

Further, the function \( v \) contains singularities, as the function \( 1/u \to \infty \) as \( u \to 0 \). In order to deal with these singularities, it is assumed that the grid is fine enough and the function smooth enough such that the difference between neighbouring nodes is small, \( u_{j+1} - u_{j-1} \ll u_j \), and the function

\[
v = \frac{1}{u} \frac{\partial u}{\partial y} = \frac{1}{2\Delta y} \frac{u_{j+1} - u_{j-1}}{u_j} \approx 0
\]

(16)

if \( u_j < \epsilon \), where \( \epsilon < 1 \). The value of \( \epsilon \) will depend on the particular problem, as well as the relative values of \( u \) and the coarseness of the grid. In the problems we consider, the solutions approach zero exponentially and \( \epsilon = 10^{-16} \) is sufficiently small.

### 2.3. Time evolution

The time evolution will be discretized implicitly. For example, the one-dimensional diffusion equation \( u_t = u_{xx} \) is discretized as

\[
 u_t = \frac{1}{\Delta t} \left[ u_{i+1}^{n+1} - u_i^n \right] = \frac{1}{\Delta x^2} \left[ u_{i+1,j}^{n+1} - 2u_{i,j}^{n+1} + u_{i-1,j}^{n+1} \right] + \mathcal{O}(\Delta t)
\]

(16)

which is first-order accurate in time, unconditionally stable and independent of the Courant number [10].
2.4. Numerical results

We approximate equation (3) on a uniform grid with \(x, y \in (-10, 10)\) and \(\Delta x = \Delta y = 20/N\), with \(\Delta t = 0.1\) for 20 timesteps. We use only a single Picard iteration, in which the coefficients \(u\) are calculated from the known \(u^n\) to obtain \(u^{n+1}\). Comparisons between the analytical and numerical solutions are shown in figure 1.

As expected, the central finite-difference scheme does not preserve positivity, while both the first-order DCU scheme (9) and second-order Hundsdorfer scheme (11 - 13) preserve positivity. The central finite-difference scheme is second-order accurate, while the DCU scheme is first-order accurate. The Hundsdorfer scheme transitions from first-order to second-order accuracy as the number of grid points \(N\) increases. For small \(N\) the grid is very coarse, such that the flux limiter \(\phi \approx 0\) due to large variations between neighbouring nodes and the Hundsdorfer scheme reduces to the first-order DCU scheme. As the number of grid points increases, the variation between neighbouring nodes decreases, the function becomes smoother, and \(\phi \to 1\), such that the Hundsdorfer scheme becomes second-order accurate.

![Figure 1: (a) The comparison of the analytical solution to numerical approximations shows that the central finite-difference scheme does not preserve positivity, while the other schemes do. (b) The difference between the analytical solution and numerical approximations for \(y = 0\), with \(N = 100\) grid points, shows that the second-order central finite-difference and Hundsdorfer schemes are more accurate than the first-order DCU scheme, a fact further illustrated by (c) the comparison of \(E_{RMS}\) for the various numerical methods as a function of the number of grid points \(N\).](image)

3. Fokker-Planck collision operator in cylindrical coordinates

The Fokker-Planck collision operator describes the local collisional relaxation process of distribution functions in plasmas under the assumption of binary, small-angle collisions [11, 12, 13] and is regarded, along with Vlasov and Maxwell’s equations, as the basis for studying weakly-coupled plasmas in all collisionality regimes. It conserves particle number, momentum and energy, preserves positivity of the distribution function, and satisfies the Boltzmann H-theorem, such that the steady-state solution is given by the Maxwellian distribution function. Despite this, however, it is a stiff advection-diffusion operator in velocity space, and nonlinear when solving the collision operators using the Landau integral [12] or the Rosenbluth potentials [11], which leads to several difficulties in dealing with it numerically [1].

In the presence of a strong magnetic field, the gyrofrequency exceeds all frequencies of interest, such that the dependence of the distribution function on the gyroangle can be neglected. The Fokker-Planck collision operator is normally solved in spherical coordinates \((r, \theta)\) where it contains no mixed derivative terms. In some cases, however, the distribution function must be solved under multiple effects, some of which may be best described in cylindrical coordinates \((v_\|, v_\perp)\), such as radiofrequency current drive [14, 15, 16], and an approximation to the collision operator in cylindrical coordinates is therefore required. The Fokker-Planck collision operator in cylindrical coordinates is therefore considered,

\[
\frac{\partial f}{\partial t} = \frac{1}{v_\| \partial v_\perp} \left[ \frac{D_\perp}{v_\perp} \frac{\partial f}{\partial v_\perp} + D_\parallel \frac{\partial f}{\partial v_\|} - F_\perp f \right] + \frac{\partial}{\partial v_\|} \left[ D_\parallel \frac{\partial f}{\partial v_\|} + D_\perp \frac{\partial f}{\partial v_\perp} - F_\| f \right] \tag{17}
\]

where \(v_\|\) is the velocity parallel and \(v_\perp\) is the velocity perpendicular to the magnetic field and \(f = f(v_\|, v_\perp, t)\) is the electron distribution function. For simplicity, the collision operators \(D\) and \(F\) are taken to be those for an
isotropic, background Maxwellian [15]. This assumption implies that energy will be lost to or gained from that background Maxwellian; otherwise only particle number will be conserved, while the preservation of positivity is ensured by the numerical scheme. In general, the collision operators could be determined from the Landau integral [12] or the Rosenbluth potentials [11], but this is not considered in this work.

The solution to the Fokker-Planck equation (17) is obtained using the positivity-preserving scheme discussed earlier, but using the first-order DCU scheme for solving the advective equations. This simplifies the extension of the Chang and Cooper scheme, which ensures the correct equilibrium solution under the assumption of local thermal equilibrium. The result is a first-order solution that always conserves particle number and preserves positivity, while energy is conserved if the assumptions in calculating the collision operator are satisfied.

### 3.1. Local thermal equilibrium

The equilibrium solution to the collision operator, in the absence of other effects, is a Maxwellian distribution function, which is a strongly (exponentially) varying function of \( v \). Central finite-difference schemes require a very fine grid in order to accurately resolve this strongly varying nature of the Maxwellian distribution, therefore Chang and Cooper [8] introduced a weighted averaging scheme which ensures the preservation of positivity and the conservation of particle number. This scheme has been extended to two-dimensions, but the resulting approximation does not guarantee the preservation of positivity [9].

Here, we employ a similar weighted averaging scheme, but, as we employ a flux-conserving scheme for discretizing the advection-diffusion equation, particle number is conserved by construction. In this case, the assumption of local thermal equilibrium ensures the correct steady-state distribution and greatly improves the accuracy of the time evolution of the distribution function.

In general, the time evolution of the distribution function can be solved if multiple effects are present, such as collisions, electric fields and plasma-wave interactions [14, 15, 16], and this method could easily be extended to include these additional terms. Here, we only consider the effect of collisions, for which the Fokker-Planck equation can be written as the divergence of a flux,

\[
\frac{\partial f}{\partial t} = \nabla \cdot \vec{S}_c
\]

where the flux \( \vec{S}_c \) will approach zero as the distribution approaches steady-state. The approach introduced by Chang and Cooper attempts to modify the numerical scheme in such a way that it gives the best outcome if \( \vec{S}_c = 0 \). If the numerical scheme used for approximating the time evolution of the distribution function does not guarantee the conservation of particle number under the divergence of a flux, this approach will ensure the conservation of particle number. For numerical schemes that ensure the conservation of particle number by construction, however, such as the flux-conserving scheme used in this work, this approach improves the accuracy of the numerical approximation to the distribution function, as will be shown later.

The scheme introduced by Chang and Cooper therefore assumes local thermal equilibrium and \( \vec{S}_c = 0 \), in order to ensure the conservation of particle number. This condition can be used to derive a weighting parameter for the advective terms of the two-dimensional collision operator in cylindrical coordinates, \( F_\parallel \) and \( F_\perp \).

First, consider the \( v_\perp \) part of the collision operator (17),

\[
\frac{\partial f}{\partial t} = \frac{1}{v_\perp} \frac{\partial}{\partial v_\perp} v_\perp \left[ D_\perp \frac{\partial f}{\partial v_\perp} + D_\parallel \frac{\partial f}{\partial v_\parallel} - F_\perp f \right]
\]

for which the flux,

\[
D_\perp \frac{\partial f}{\partial v_\perp} + D_\parallel \frac{\partial f}{\partial v_\parallel} - F_\perp f
\]

will equal zero if the distribution is in local thermal equilibrium. If this is the case, the flux can be rewritten as

\[
\frac{\partial f}{\partial v_\perp} = \frac{1}{D_\perp} \left( F_\perp - D_\parallel \frac{\partial f}{\partial v_\parallel} \right) f
\]

(19)

where

\[
g = \frac{1}{f} \frac{\partial f}{\partial v_\parallel}
\]

(20)

is analogous to equation (14). Equation (19) can be discretized to obtain the solution

\[
f_{i,j+1} \sim f_{i,j} \exp \left[ \frac{F_\perp - D_\parallel \frac{\partial f}{\partial v_\parallel}}{D_\perp} \Delta v_\perp \right] = f_{i,j} \exp \left[ A_g - B_g \right]
\]

(21)
where the labels \((i, j)\) refers to the \(i^{th}\) node in the parallel direction and the \(j^{th}\) node in the perpendicular direction. The functions \(A_g\) and \(B_g\) are defined as

\[
A_g = \frac{F_1}{D_{\perp \perp}} \Delta v_\perp; \quad B_g = \frac{D_{\perp \perp}}{D_{\perp \perp}} \Delta v_\perp
\]  

(22)

The aim is to write the discretized form of the \(F_1\)-term (for \(F_1 < 0\)), using the first-order DCU scheme, as

\[
\frac{\partial f}{\partial t} = \frac{1}{\Delta v} \left[ \left( \frac{v_\perp - \Delta v/2}{v_\perp} \right) F_\perp \left( v_{\parallel}, v_\perp - \frac{\Delta v}{2} \right) \delta_{ij,j} + \left( -\frac{v_\perp + \Delta v/2}{v_\perp} \right) F_\perp \left( v_{\parallel}, v_\perp + \frac{\Delta v}{2} \right) \delta_{i,j+1,1} \right]
\]

by introducing, and solving, for \(\delta_{\perp}\). The discretized form of the flux in the \(v_\perp\)-direction (18), after rewriting the mixed derivative as an advective equation using (20) and the first-order DCU scheme for discretizing the advection equations, gives

\[
\frac{1}{\Delta p} D_{\perp \perp,j+1/2} (f_{i,j+1} - f_{i,j}) + D_{\perp \perp,j+1/2} \cdot g_{j+1/2} \cdot f_{i,j+1} - F_{\perp \perp,j+1/2} \cdot f_{i,j+1} \cdot \delta_{\perp} = 0
\]

\[
\Rightarrow f_{ij} e^{A_g e^{B_g}} - f_{ij} + B_g f_{ij} e^{A_g e^{B_g}} - A_g e^{A_g e^{B_g}} f_{ij} \delta_{\perp} = 0
\]

at the \((v_\perp + \Delta v/2)\)-boundary, for \(D_{\perp \perp} \cdot g > 0\). It equals zero for local thermal equilibrium, and therefore we choose

\[
\delta_{\perp} = \frac{1}{A_g} \left( 1 + B_g - e^{-A_g e^{B_g}} \right)
\]

(23)

with the same result obtained at the \((v_\perp - \Delta v/2)\)-boundary.

For \(D_{\perp \perp} \cdot g < 0\) (and \(F_\perp < 0\), the solution to \(\delta_{\perp}\) is given by

\[
\delta_{\perp} = \frac{1}{A_g} \left( 1 + B_g e^{-A_g e^{B_g}} - e^{-A_g e^{B_g}} \right)
\]

(24)

with \(A_g\) and \(B_g\) given by (22). A similar approach is used for the flux in the \(v_\parallel\)-direction in order to obtain solutions for \(\delta_{\parallel}\), but in this case there are four possibilities,

| \(F_\parallel\) | \(D_{\parallel \parallel} \cdot h\) | \(\delta_{\parallel}\) |
|----------------|----------------|----------------|
| >0 <0 | \(\frac{1}{A_{h}} \left( e^{A_{h}} e^{B_{h}} + B_{h} e^{A_{h}} e^{B_{h}} - 1 \right) \) |
| <0 >0 | \(\frac{1}{A_{h}} \left( 1 + B_{h} - e^{B_{h}} - A_{h} \right) \) |
| >0 <0 | \(\frac{1}{A_{h}} \left( e^{A_{h}} - B_{h} - 1 + B_{h} \right) \) |
| <0 <0 | \(\frac{1}{A_{h}} \left( 1 - e^{B_{h}} - A_{h} + B_{h} e^{B_{h}} - A_{h} \right) \) |

with the functions \(A_h\) and \(B_h\) defined as

\[
A_h = \frac{F_\parallel}{D_{\parallel \parallel}} \Delta v_\parallel; \quad B_h = \frac{D_{\parallel \parallel} \cdot h}{D_{\parallel \parallel}} \Delta v_\parallel
\]

and

\[
h(v_\parallel, v_\perp) = \frac{1}{f} \frac{\partial f}{\partial v_\perp}
\]

which is equivalent to equation (14).

Note that, as \(N \rightarrow \infty, \Delta v \rightarrow 0\), so \(A, B \rightarrow 0\) and \(\delta \rightarrow 1\), such that the effect of \(\delta\)-splitting is negligible. Of course, as \(N \rightarrow \infty\), the differences in \(f\) between two neighbouring nodes becomes negligible, and therefore \(\delta\)-splitting is no longer required. The introduction of this \(\delta\)-splitting technique therefore acts to improve the solution for coarse grids, while \(\delta \approx 1\) for sufficiently fine grids.
3.2. Numerical stability

The δ-splitting method has been introduced to ensure that the steady-state solution of the electron distribution function under the effect of only the Fokker-Planck collision operator is a Maxwellian distribution, due to the distribution being a strongly (exponentially) varying function of \( v \). The introduction of \( \delta \), however, leads to a stability issue. Consider, for example, if \( f \) is a Maxwellian,

\[
f \sim \exp(-v^2)
\]

and therefore

\[
g \sim -v
\]

In the limit \( v \to \infty \), we have \( A \to 0 \), but \( B \to \mp \infty \), such that \( \delta \to \infty \). Numerically, this introduces problems, as large terms lead to instabilities by creating ill-conditioned matrices. The value of \( \delta \) must therefore be limited to some maximum \( \delta_{\text{max}} \). Fortunately there is the competing effect that \( \delta \to 1 \) for increasing \( N \), such that, if \( N \) is large enough, the value of \( \delta_{\text{max}} \) is irrelevant as \( \delta \) is always small enough for a stable solution, while the cutoff value \( \delta_{\text{max}} \) only comes into effect at large \( v \) where there are very few particles.

As an example, consider the case of an initial Maxwellian distribution with \( T_e = 20 \text{eV} \) colliding with a fixed background distribution at \( T_b = 10 \text{eV} \) with density \( n_e = 10^{14} \text{m}^{-3} \). The number of grid points \( (N = 150 \text{ such that } v_t = 10\Delta v) \) is chosen in order to demonstrate the effect of \( \delta_{\text{max}} \), as increasing the number of grid points can eliminate the need for \( \delta_{\text{max}} \). Temperature is defined as an integral over the distribution and is related to the thermal velocity of the electrons.

The temperature evolution reaches equilibrium at \( t = 0.4 \text{s} \). However, as electrons at large \( v \) collide infrequently, the distribution functions only reach equilibrium at \( t = 1 \text{s} \). As there are very few electrons at large \( v \), the temperature does not change markedly after the bulk electrons reach equilibrium at \( t = 0.4 \text{s} \). The temperature evolution up to \( t = 0.4 \text{s} \) and the distribution functions at \( t = 1 \text{s} \) are therefore compared in figure 2 for three cases.

Without δ-splitting (\( \delta = 1 \)), the wrong equilibrium temperature is reached, while for \( \delta_{\text{max}} = 2 \) and for \( \delta_{\text{max}} = 10 \) there are no differences in the temperature evolution, and the correct equilibrium temperature is reached, as expected, since δ-splitting ensures that the solution is a Maxwellian.

Comparing the distribution functions to the background distribution shows the effect of \( \delta_{\text{max}} \). Firstly, for \( \delta = 1 \) the final distribution is different from the background distribution, due to the wrong equilibrium temperature being reached. Comparing the \( \delta_{\text{max}} = 2 \) and \( \delta_{\text{max}} = 10 \) distributions to the background distribution, it is clear that there are only small differences for \( v_{\bot} < 4v_t \), where the majority of electrons are, and therefore the correct temperature is obtained. The differences are in the high \( v \) tail, with \( \delta_{\text{max}} = 2 \) underestimating the relaxation, and \( \delta_{\text{max}} = 10 \) overestimating the relaxation.

Larger values of \( \delta_{\text{max}} \) lead to ill-conditioned matrices as equilibrium is approached, so for numerical stability, the value of \( \delta_{\text{max}} \) must be relatively small (\( \delta \geq 1 \) always). Of course, as \( N \to \infty \) the value of \( \delta \to 1 \), and there will be no need for \( \delta_{\text{max}} \). In practice, however, the grid will hardly ever be large enough to allow this to happen, so it will be necessary to specify a value for \( \delta_{\text{max}} \). Fortunately, this value will only impact regions of large \( v \), where there are very few electrons which does not influence the low-order moments, such as temperature, and a relatively small value for \( \delta_{\text{max}} \) suffices.

3.3. Temperature equilibration

Consider two electron distributions of equal density colliding with each other. The distributions will equilibrate according to

\[
\frac{dT_a}{dt} = \nu(T_b - T_a)
\]  \hspace{1cm} (25)

with \( \nu \) the collision frequency, and

\[
\frac{dT_b}{dt} = -\frac{dT_a}{dt}
\]

The collision frequency is given by [17],

\[
\nu = \frac{8}{3\sqrt{\pi}} \left( \frac{e^2}{4\pi\varepsilon_0} \right)^2 \frac{4\pi n_e \lambda}{m_e^3 \sqrt{\left( v_{t,a}^2 + v_{t,b}^2 \right) 2}}
\]  \hspace{1cm} (26)

where \( v_{t,a}^2 = 2T_a/m_e \) is the thermal velocity of distribution \( a \), and the Coulomb logarithm is taken to be constant \( \lambda = 15 \).
Consider two distributions, with temperatures $T_a = 20 \text{ eV}$ and $T_b = 10 \text{ eV}$ and densities $n_a = n_b = 10^{14} \text{ m}^{-3}$ colliding with each other. The resultant temperature evolution, when using $\delta$-splitting, is shown in figure 3. Of course, the assumption that both distributions collide with a background Maxwellian is not always true; cold electrons undergo more collisions than warm electrons, such that cold electrons gain energy faster than warm electrons lose energy, and the distribution is not always Maxwellian, resulting in the wrong equilibrium temperature, as energy is not conserved.

By constraining both distributions to always be Maxwellian (by replacing each distribution with a Maxwellian of the same temperature at each time step), the assumption of collisions with a background Maxwellian remains true, and the correct equilibrium temperature is reached at the predicted rate, while energy is conserved.

If the collision operators are calculated through the Rosenbluth potentials [11] or the Landau integrals [12], it will no longer be necessary to constrain the distributions to be Maxwellian, as this constraint is only necessary to ensure the assumption of background Maxwellsians is satisfied. The calculation of the collision operators from the distribution functions, however, adds additional numerical evaluations, and therefore is not considered here.

3.4. Time evolution

The treatment of the mixed derivatives introduces a non-linearity which can be solved through a Picard iteration. Alternatively, by Picard linearizing, an accurate solution can be obtained by using a smaller time step [9]. To
illustrate this, consider two initial Maxwellian distributions with $T_a = 20\, \text{eV}$ and $T_b = 10\, \text{eV}$ colliding with each other. The number of grid points is $N = 150$, such that $v = 10\Delta v$, while the density $n_e = 10^{14}\, \text{m}^{-3}$ and the collision time $\tau \approx 20\, \text{ms}$.

The time evolution of the temperature for different values of $\Delta t$ is shown in figure 4. As expected, convergence is achieved for larger time steps when approximating the time evolution of the distribution function through Picard iteration as compared to Picard linearizing. However, obtaining a solution through Picard iteration is computationally more expensive as we have to iterate over the solution $f^{n+1}$.

![Figure 4: The time evolution of the temperature of a distribution function at $T_a = 20\, \text{eV}$ colliding with a distribution at $T_b = 10\, \text{eV}$, both with density $n_e = 10^{14}\, \text{m}^{-3}$, for different choices of $\Delta t$ when solving the distribution through (a) Picard linearizing and (b) Picard iterating. (c) The comparison of the difference between the analytical and numerical temperatures for the two methods for $\Delta t = 20\, \text{ms}$, shows that Picard linearizing leads to more accurate approximations.](image)

### 4. Conclusion

Lower-order numerical methods are less accurate, but generally more robust and reliable, than higher-order methods, which tend to be more complicated. In this paper we propose a scheme that focuses on improving the accuracy of lower-order methods, in particular with respect to the preservation of positivity, for solving two-dimensional advection-diffusion equations of the form

$$\frac{\partial u}{\partial t} = \nabla \cdot (\vec{a}u + \hat{k} \cdot \nabla u)$$

where $u = u(x, y, t)$ is advected by the vector $\vec{a}(x, y, t)$ and diffused by the tensor $\hat{k}(x, y, t)$. Numerical solutions to these equations, in the absence of mixed derivatives, have been studied in detail [2, 3], but solutions of problems where mixed derivative terms are present have received much less attention. The scheme proposed in this paper allows the mixed derivative terms to be written as advection-type equations, after which a lower-order positivity-preserving scheme can be applied. This was discussed by using an example and then applied to the Fokker-Planck collision operator in cylindrical coordinates. Compared to central finite-difference methods, the proposed scheme has the same order of accuracy, with the added advantage of the solution being non-negative.

The Fokker-Planck collision operator has been approximated to get the most accurate solution when the solution is in local thermal equilibrium, which introduces an averaging parameter to ensure an accurate steady-state distribution. The solution is tested with the thermal equilibration of two colliding Maxwellian distributions and compares well with the theoretically predicted rate. The scheme conserves particle number and preserves positivity, but only conserves energy and agrees with the theoretical equilibration rate if the distributions are constrained to be Maxwellian, as the collision operators are calculated under the assumption of a background Maxwellian.

The treatment of the mixed derivatives introduces a non-linearity, which can be solved through Picard iteration. This iteration allows for larger time steps to be taken, but also increases the computational time. For the considered example of studying the thermal equilibration of two electron distribution functions under the Fokker-Planck collision operator, Picard linearizing provides an accurate approximation to the time evolution when the time-step is shorter than the collision time.

In conclusion, the scheme proposed in this paper is suitable when a positive solution under the conservation of particle number is required, when a small time step is used such that the non-linearity in the mixed derivatives can be approximated by Picard linearizing. Although higher-order methods and flux limiting can be used to obtain
more accurate solutions, the proposed scheme uses lower-order methods to ensure a reliable and robust method which is less complicated than higher-order methods. The δ-splitting method can, in general, be applied to any advection-diffusion problem to ensure the correct steady-state solution. We show that, for equilibration under the Fokker-Planck collision operator, the scheme ensures the correct equilibrium distribution while conserving particle number and preserving positivity.

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