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Lagrangian-Antidi↵usive Remap schemes for non-local multi-class traffic flow models

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Abstract

This paper focuses on the numerical approximation of the solutions of a class of non-local systems in one space dimension, arising in traffic modeling. We propose alternative simple schemes by splitting the non-local conservation laws into two di↵erent equations, namely, the Lagrangian and the remap steps. We provide some properties and estimates recovered by approximating the problem with the Lagrangian-Antidi↵usive Remap (L-AR) scheme, and we prove the convergence to weak solutions in the scalar case. Finally, we show some numerical simulations illustrating the efficiency of the L-AR schemes in comparison with classical first and second-order numerical schemes.

1 Introduction

The aim of this paper is to propose an efficient and easy-to-implement numerical scheme to compute approximate solutions of the non-local multi-class traffic flow model introduced in [10]. It consists in the following $M \times M$ system of conservation laws with non-local velocities

$$\partial_t \rho_i(t,x) + \partial_x \rho_i(t,x) v_i((r \star \eta_i)(t,x)) = 0, \qquad i = 1, \dots, M, \tag{1.1}$$

where we set

$$r(t,x) := \sum_{i=1}^{M} \rho_i(t,x), \qquad v_i(\cdot) := v_i^{\max}(\cdot), \tag{1.2}$$

$$(r \star \eta_i)(t,x) := \int_x^{x+\mathcal{H}_i} r(t,y)\,\eta_i(y-x)\,dy, \tag{1.3}$$

$$\rho_i(0,x) = \rho_i^0(x), \qquad i = 1, \dots, M. \tag{1.4}$$

We make the following hypotheses:

(H)
- $\eta_i \in \mathbf{C}^1([0,\mathcal{H}_i];\mathbb{R}^+)$, $\mathcal{H}_i > 0$, s.t. $\eta_i' \le 0$, $\int_0^{\mathcal{H}_i} \eta_i(y)\,dy = J_i$ and $W_0 := \max_{i=1,\dots,M} \eta_i(0)$;
- $0 < v_1^{\max} \le v_2^{\max} \le \dots \le v_M^{\max}$;
- $:\mathbb{R}^+ \to \mathbb{R}^+$ is a smooth non-increasing function s.t. $(0)=1$, $(r)=0$ for $r \ge 1$.

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From the application point of view, ρ_i represents the density of the i-th vehicle class, characterized by its maximal velocity $v_i^{\max}$ and its interaction kernel ω_i, which accounts for the reaction to downstream traffic distribution. In particular, $\mathcal{H}$ is proportional to the look-ahead distance of drivers and J_i represents the interaction strength. Equations (1.1) are coupled through the velocity function, which depends on an integral evaluation of the total traffic density r. For simplicity, in this work we will consider linear decreasing velocities setting $(r)=\max\{1 - r, 0\}$.

Model (1.1) is a multi-class extension of the one-dimensional scalar conservation law with non-local flux proposed in [3, 15]. It can also be seen as the non-local generalization of the "local" multi-population model for traffic flow described in [2]. Indeed, one of the limitations of the standard Lighthill-Whitham-Richards (LWR) traffic flow model [17, 18] is the first in first out rule, conversely in multi-class dynamic faster vehicles can overtake slower ones and slower vehicles slow down the faster ones. In our setting, the non-local dependence ofthe speed functions v describes the reaction of drivers that adapt their velocity with respect to what happens in front of them in terms of downstream traffic, assigning greater importance to closer vehicles, see also [9].

Since solutions to (1.1), (1.4) may be discontinuous, they are intended in the following weak sense.

Definition 1. *A function* $\vec{\rho} = (\rho_1, \dots, \rho_M) \in (\mathbf{L}^1 \cap \mathbf{L}^\infty)([0, T[\times \mathbb{R}; \mathbb{R}^M)$, $T > 0$, *is a weak solution of* (1.1), (1.4) *if*

$$\int_0^T \int_{\mathbb{R}} \rho_i \partial_t \varphi + \rho_i v_i(r * \omega_i) \partial_x \varphi(t, x)\, dx\, dt + \int_{\mathbb{R}} \rho_i^0(x) \varphi(0, x)\, dx = 0$$

for all $\varphi \in \mathbf{C}^1_c([-1, T[\times \mathbb{R}; \mathbb{R})$, $i = 1, \dots, M$.

The existence of solutions, locally in time, was proved in [10] by constructing a sequence of approximate solutions via an up-wind finite volume scheme, and proving its convergence. We observe that, due to the lack of smoothness of the convolutions kernels ω_i on $\mathbb{R}$, the general well-posedness results given in [11] cannot be applied.

In this work, we focus on the numerical aspects. We remark that the computation of numerical solutions for (1.1) is challenging due to the high non-linearity of the system and the dependence of the flux function on convolution terms, which increase the computational cost. For these reasons, the development of efficient numerical strategies is crucial to precisely capture solutions' behaviour. The first-order schemes used in [9, 10] are known to be very diffusive. On the other hand, high-order Discontinuous Galerkin and Finite Volume WENO schemes, which are based on quadratic polynomial reconstruction in each cell to evaluate the non-local terms in order to obtain high-order of accuracy, were constructed and tested for the scalar cases in [8] and extended to multi-class cases in [11].

The aim of this paper is to present a generalization to non-local systems of the Lagrangian-Antidiffusive Remap (L-AR) schemes introduced in [5, 6], in order to compute approximate solutions of model (1.1). This type of schemes are constructed exploiting the concentration-times-velocity form of the fluxes in (1.1). In [5], one step L-AR schemes were applied to multi-class Lighthill-Whitham-Richards (MCLWR) traffic models and they were extended to polydisperse sedimentation models in [6]. L-AR schemes do not rely on spectral (characteristics) information and their implementation is as easy as that one of first- and second-order of accuracy schemes introduced in [7]. These L-AR schemes are supported by a partial analysis

for $M = 1$, with the conclusion that under a suitable CFL condition, these schemes converge to a weak solution. Numerical experiments show that L-AR schemes perform better in terms of resolution, accuracy and efficiency, especially for large values of M.

The paper is organized as follows. Section 2 presents the L-AR schemes. We recover some properties of the schemes in both scalar and multi-class cases. In the scalar case, we obtain uniform $\mathbf{L}^1$, $\mathbf{BV}$ estimates on the approximate solutions computed through the L-AR schemes, which provide an alternative proof of existence of weak solutions. In Section 3, we recall classical first-order schemes used to approximate the solutions of the non-local problem (1.1) and we provide a second-order version of a Godunov type numerical scheme. Finally, in Section 4, we present some numerical simulations, analyzing the $\mathbf{L}^1$ error of the approximate solutions of (1.1) computed with different schemes and considering smooth and discontinuous initial data.

2 Lagrangian-Antidiffusive Remap (L-AR) schemes

2.1 Discretization

First of all, we extend $\rho_i(x) = 0$ for $x > \mathcal{H}_i$. For $j \in \mathbb{Z}$ and $n \in \mathbb{N}$, let $x_{j+1/2} = j\Delta x$ be the cells interfaces, $x_j = (j - 1/2)\Delta x$ the cells centers and $t^n = n\Delta t$ the time mesh. In the paper, we will set $\lambda = \frac{\Delta t}{\Delta x}$. We aim at constructing a finite volume approximate solution $\vec{\rho}^{\,\Delta x} = (\vec{\rho}_1^{\,\Delta x}, \ldots, \vec{\rho}_M^{\,\Delta x})$, with $\vec{\rho}_i^{\,\Delta x}(t, x) = \vec{\rho}_{i,j}^{\,n}$ for $(t, x) \in C_j^n = [t^n, t^{n+1}[\times [x_{j-1/2}, x_{j+1/2}]$ and $i = 1, \ldots, M$.

To this end, we approximate the initial datum $\vec{\rho}_i^{\,0}$ for $i = 1, \ldots, M$ with a piece-wise constant function

$$\vec{\rho}_{i,j}^{\,0} = \frac{1}{\Delta x} \int_{x_{j-1/2}}^{x_{j+1/2}} \vec{\rho}_i^{\,0}(x)\, dx, \qquad j \in \mathbb{Z}. \tag{2.1}$$

Similarly, for the kernel, we set

$$\omega_i^k := \frac{1}{\Delta x} \int_{(k-1)\Delta x}^{k\Delta x} \omega_i(x)\, dx, \qquad k \in \mathbb{N}^*, \tag{2.2}$$

so that $\Delta x \sum_{k=1}^{+\infty} \omega_i^k = \int_0^{\mathcal{H}_i} \omega_i(x)\, dx = J_i$ (the sum is indeed finite since $\omega_i^k = 0$ for $k \geq N_i$ sufficiently large). Moreover, we set $r_{j+k}^n = \sum_{i=1}^{M} \vec{\rho}_{i,j+k}^{\,n}$ for $k \in \mathbb{N}$ and

$$V_{i,j+1/2}^n := v_i^{\max}\left(\Delta x \sum_{k=1}^{+\infty} \omega_i^k r_{j+k}^n\right), \qquad i = 1, \ldots, M, \quad j \in \mathbb{Z}. \tag{2.3}$$

We formally rewrite (1.1) as

$$\partial_t \vec{\rho}_i + \vec{\rho}_i \partial_x(v_i(r * \omega_i)) + v_i(r * \omega_i)\partial_x \vec{\rho}_i = 0, \qquad i = 1, \ldots, M. \tag{2.4}$$

L-AR schemes are obtained splitting (2.4) into two different equations, which are solved successively for each time iteration. To advance the solution from time t to $t + \Delta t$, we first apply a Lagrangian method [16] to solve

$$\partial_t \vec{\rho}_i + \vec{\rho}_i \partial_x(v_i(r * \omega_i)) = 0, \tag{2.5}$$

and we use this solution, evolved over a time interval of length Δt, as initial condition for solving in a second step the transport equation [4]

$$\partial_t \rho_i + v_i(r * \omega_i)\partial_x \rho_i = 0, \tag{2.6}$$

whose solution, again evolved over a time interval of length Δt, provides the sought approximate solution of (1.1) valid for $t + \Delta t$.

2.2 Discretization of the Lagrangian step

We observe that, defining $\tau_i := 1/\rho_i$, one obtains from (2.5) the conservation mass equation in Lagrangian coordinates

$$\rho_i \partial_t \tau_i - \partial_x(v_i(r * \omega_i)) = 0. \tag{2.7}$$

In other words, solving (2.5), or equivalently (2.7), means solving the original equation (1.1) on a moving referential mesh with velocity v_i. Assume now that $\{\rho^n_{i,j}\}_{j\in\mathbb{Z}}$, $i = 1, \ldots, M$ is a approximate solution of (1.1) in the sense of finite volume methods (2.1)-(2.3) at time $t = t^n$, then a numerical solution $\{\rho^{n+1,-}_{i,j}\}_{j\in\mathbb{Z}}$ of the equation (2.7) at time Δt can be naturally computed by

$$\rho^{n+1,-}_{i,j}\left[\Delta x + (V^n_{i,j+1/2} - V^n_{i,j-1/2})\Delta t\right] = \Delta x\, \rho^n_{i,j}, \quad i = 1, \ldots, M, \quad j \in \mathbb{Z}, \tag{2.8}$$

since (2.8) expresses that the initial mass in the cell $[x_{j-1/2}, x_{j+1/2}]$ at time t^n equals the mass in the modified cell $[\bar{x}_{j-1/2}, \bar{x}_{j+1/2}]$ at time Δt, where $\bar{x}_{j+1/2} = x_{j+1/2} + V^n_{i,j+1/2}\Delta t$ are the new interface positions for all j. We have the following properties.

Lemma 1. *Assume that the time step satisfies the following condition:*

$$\Delta t \leq \left(v_M^{\max} \|\psi'\|_1 \|r^n\|_1 W_0 \right)^{-1}. \tag{2.9}$$

If $\{\rho^{n+1,-}_{j}\}_{j\in\mathbb{Z}}$ denotes the numerical solution computed by the scheme (2.8), then the following bounds hold:

(*i*) *If $\rho^n_{i,j} \geq 0$ for all $j \in \mathbb{Z}$, then $\rho^{n+1,-}_{i,j} \geq 0$ for all $j \in \mathbb{Z}$.*

(*ii*) *In the scalar case $M = 1$, the following maximum property holds:*

$$\min\{\rho^n_j, \ldots, \rho^n_{j+N}\} \leq \rho^{n+1,-}_j \leq \max\{\rho^n_j, \ldots, \rho^n_{j+N}\} \quad \forall j \in \mathbb{Z}. \tag{2.10}$$

Proof. (*i*) Suppose that $\rho^n_{i,j} \geq 0$ for all $j \in \mathbb{Z}$ and $i = 1, \ldots, M$. From (2.8) we have

$$\rho^{n+1,-}_{i,j} = \frac{\rho^n_{i,j}}{1 + \frac{\Delta t}{\Delta x}\left(V^n_{i,j+1/2} - V^n_{i,j-1/2}\right)}. \tag{2.11}$$

If $V^n_{i,j+1/2} \geq V^n_{i,j-1/2}$ then it is clear that $\rho^{n+1,-}_{i,j} \geq 0$. Consider now the case $V^n_{i,j+1/2} \leq V^n_{i,j-1/2}$. We can compute

$$V^n_{i,j-1/2} - V^n_{i,j+1/2} = v_i^{\max}\psi'(\xi_{i,j})\, \Delta x \left(\sum_{k=1}^{N} \omega_i^k r^n_{j+k} - \sum_{k=1}^{N} \omega_i^k r^n_{j+k-1} \right)$$

$$\eth v_i^{\max}\ {}^{0}(\leftarrow_{i,j})\ x\ @^{\overset{0}{\ast}\overset{1}{\ast}}_{k=1}\ (!_i^k\ !_i^{k+1})r_{j+k}^n\ !_i^1 r_j^n A$$

$$\eth v_i^{\max}\ xW_0\ {}^{0}{}_1\ kr^n k_1\ ,$$

for some $\leftarrow_{i,j}\ 2\ I\ \overset{\downarrow}{\ }\ x\ P_{k=1}^{+1}\ !_i^k r_{j+k}^n\ {}_1,\ x\ P_{k=1}^{+1}\ !_i^k r_{j+k}^n\ {}^{\mathcal{H}}$, where we have set $I(a, b) = \overset{\rightarrow}{\ }\min\{a, b\}, \max\{a, b\}\overset{\leftarrow}{\ }$. Therefore $\overset{\rightarrow}{\ }^{n+1,}\ 0$ under (2.9).

(*ii*) Assume $M = 1$ and set $V_{j+1/2}^n := V_{1,j+1/2}^n$, $!^k := !_1^k$ and $v^{\max} := v_1^{\max}$. Let us prove the upper bound, the lower one resulting from a symmetric procedure. Define $\overline{\rightarrow}_j = \max\{\rightarrow_j^n, \ldots, \overset{\rightarrow}{\ }_{j+N}^n\}$. Consider first the case $V_{j+1/2}^n\ V_j^n\ {}_{1/2}$. Then it is clear that

$$\overset{\rightarrow}{\ }_j^{n+1,} = \frac{\overset{\rightarrow}{\ }_j^n}{1 + \overset{\downarrow}{\ }V_{j+1/2}^n\ V_j^n\ {}_{1/2}}\ {}^{\mathcal{H}}\ \eth\ \overline{\rightarrow}_j\ ,$$

Consider now the case $V_{j+1/2}^n\ \eth V_j^n\ {}_{1/2}$. We note that

$$\frac{\overset{\rightarrow}{\ }_j^n}{1 + \overset{\downarrow}{\ }V_{j+1/2}^n\ V_j^n\ {}_{1/2}}\ {}^{\mathcal{H}}\ \eth\ \overline{\rightarrow}_j\ ()\ \overline{\ }\ \rightarrow_j\ \overset{\rightarrow}{\ }_j^n + \overline{\ }\ \rightarrow_j\ \overset{\downarrow}{\ }V_{j+1/2}^n\ V_j^n\ {}_{1/2}\ {}^{\mathcal{H}}\ 0.$$

According with (*i*), we have the estimation

$$V_j^n\ {}_{1/2}\ V_{j+1/2}^n = v^{\max}\ {}^{0}(\leftarrow_j)\ x\ @^{\overset{0}{\ast}\overset{1}{\ast}}_{k=1}\ \overset{\downarrow}{\ }!^k\ !^{k+1}\ {}^{\mathcal{H}}\ \overset{\rightarrow}{\ }_{j+k}^n\ !^1\ \overset{\rightarrow}{\ }_j^n A$$

$$\eth v^{\max}\ {}^{0}{}_1\ x\ \overset{\downarrow}{\ }!^1\ \overline{\rightarrow}_j\ !^1\ \overset{\rightarrow}{\ }_j\ {}^{\mathcal{H}}$$

$$\eth v^{\max}\ {}^{0}{}_1\ x\ \overline{\ }\ \rightarrow_j\ \rightarrow_j\ W_0.$$

Finally we obtain that

$$\overline{\rightarrow}_j\ \overset{\rightarrow}{\ }_j^n\ \overline{\rightarrow}_j\ \overset{\downarrow}{\ }V_j^n\ {}_{1/2}\ V_{j+1/2}^n\ {}^{\mathcal{H}}\ (\overline{\rightarrow}_j\ \overset{\rightarrow}{\ }_j^n)\ \overset{\downarrow}{\ }1\ tv^{\max}\ \overline{\rightarrow}_j\ {}^{0}{}_1\ W_0\ {}^{\mathcal{H}}\ 0$$

holds if $t\ \eth\ \overset{\downarrow}{\ }v^{\max}\ {}^{0}{}_1\ k\ \overset{\rightarrow}{\ }^n k_1\ W_0\ {}^{\mathcal{H}}\ {}^1$.

□

Remark 1. *Due to the lack of uniform* $\mathbf{L}^1$ *estimates on approximate solutions [10], the time step should in principle be recomputed at each iteration to comply with (2.9). In practice, since the computed solutions stay uniformly bounded in time, it is possible to choose a fixed time step, as we did in Section 4. Moreover, we remark that, in the particular case* $M = 1$, *the maximum principle (2.10) guarantees that* $k\ \overset{\rightarrow}{\ }^n k_1\ \eth\ \cdots\rightarrow\ {}^{0}{}_1$ *for all* $n\ 2\ N\ \overset{\leftarrow}{\ }$.

2.3 Remap Step: Antidi↵usive scheme

After the Lagrangian step, the new value $\overset{\rightarrow}{\ }_{i,j}^{n+1,}$ represents approximate values of the density of the *i*-th class on a moved mesh with new cells $[\overline{x}_j\ {}_{1/2}, \overline{x}_{j+1/2}]$ for all j. To avoid dealing with moving meshes, a so-called remap step is necessary to define the new approximations

ρ_j^{n+1} on the uniform mesh with cells $[x_{j-1/2}, x_{j+1/2}]$. This averaging step can equivalently be reformulated by using the solution of the transport equation (2.6) with initial data defined by $\rho_{i,j}^{n+1,\prime}$ on each cell $[x_{j-1/2}, x_{j+1/2}]$, i.e. we consider a numerical scheme in the form

$$\rho_{i,j}^{n+1} = \rho_{i,j}^{n+1,\prime} - \bar{V}_{i,j}^{n} \left(\rho_{i,j+1/2}^{n+1,\prime} - \rho_{i,j-1/2}^{n+1,\prime} \right), \qquad i = 1, \ldots, M \quad j \in \mathbb{Z}. \tag{2.12}$$

Here, $\bar{V}_{i,j}^{n}$ is a velocity value, defined in terms of available density, which will be chosen in such a way that the complete scheme (2.5) plus (2.12) is conservative with respect to (1.1). The quantities $\rho_{i,j+1/2}^{n+1,\prime}$ are numerical fluxes associated with the cell interfaces $x_{j+\frac{1}{2}}$ and will be chosen such that the scheme (2.12) has certain stability and consistency properties. In particular, the choice $\rho_{i,j+1/2}^{n+1,\prime} = \rho_{i,j}^{n+1,\prime}$ for all $j \in \mathbb{Z}$ produces a diffusive and stable scheme, while $\rho_{i,j+1/2}^{n+1,\prime} = \rho_{i,j+1}^{n+1,\prime}$ yields an antidiffusive but unstable scheme. For this reason, we proceed as in [4, 12] and we choose $\rho_{i,j+1/2}^{n+1,\prime}$ as close to the anti-diffusive value $\rho_{i,j+1}^{n+1,\prime}$ as possible, subject to the following consistency condition

$$m_{i,j+1/2} := \min\{\rho_{i,j}^{n+1,\prime}, \rho_{i,j+1}^{n+1,\prime}\} \le \rho_{i,j+1/2}^{n+1,\prime} \le M_{i,j+1/2} := \max\{\rho_{i,j}^{n+1,\prime}, \rho_{i,j+1}^{n+1,\prime}\}, \tag{2.13}$$

and maximum principle

$$m_{i,j+1/2} \le \rho_{i,j}^{n+1} \le M_{i,j+1/2}, \tag{2.14}$$

which resume the properties of the scheme defined by (2.12).

Let us now define

$$b_{i,j}^{+} := M_{i,j+1/2} + \frac{\rho_{i,j}^{n+1,\prime} - M_{i,j+1/2}}{\max\{v_{i,j-1/2}^{n}, v_{i,j+1/2}^{n}\}}, \qquad B_{i,j}^{+} := m_{i,j+1/2} + \frac{\rho_{i,j}^{n+1,\prime} - m_{i,j+1/2}}{\max\{v_{i,j-1/2}^{n}, v_{i,j+1/2}^{n}\}}$$

and

$$a_{i,j+1/2} := \max\{b_{i,j}^{+}, m_{i,j+1/2}\}, \qquad A_{i,j+1/2} := \max\{B_{i,j}^{+}, M_{i,j+1/2}\}.$$

In the next lemma, which is a slight modification of Lemma 4.1 in [5], we summarize the existence and properties of the scheme defined by (2.12).

Lemma 2. *Assume that the following CFL condition holds*

$$t \le \frac{x}{v_M^{\max} \|k\|_1}. \tag{2.15}$$

Then $a_{i,j+1/2} \le \rho_{i,j}^{n+1,\prime} \le A_{i,j+1/2}$ *for all* $j \in \mathbb{Z}$, *and for any numerical flux that satisfies*

$$\rho_{i,j+1/2}^{n+1,\prime} \in [a_{i,j+1/2}, A_{i,j+1/2}], \qquad \text{for all } j \in \mathbb{Z}, \tag{2.16}$$

the scheme (2.12) *is* L^1*-stable, that is*

$$\rho_{i,j}^{n+1} \in I(\rho_{i,j}^{n+1,\prime}, \rho_{i,j+1}^{n+1,\prime}), \qquad \text{for all } j \in \mathbb{Z}, \tag{2.17}$$

and TVD, i.e.,

$$\sum_{j\in\mathbb{Z}} \left| \rho_{i,j+1}^{n+1,\prime} - \rho_{i,j}^{n+1,\prime} \right| \le \sum_{j\in\mathbb{Z}} \left| \rho_{i,j+1}^{n} - \rho_{i,j}^{n} \right|. \tag{2.18}$$

In particular, for each $j \in \mathbb{Z}$, *there exist numbers* $\theta_{i,j} \in [0, 1]$ *such that*

$$\rho_{i,j}^{n+1,\prime} = \theta_{i,j} \rho_{i,j-1/2}^{n+1,\prime} + (1 - \theta_{i,j}) \rho_{i,j+1/2}^{n+1,\prime}. \tag{2.19}$$

2.4 Choice of numerical flux

In this subsection, we explain how to compute $\vec{\ }^{n+1,}_{i,j+1/2}$ for the scalar case $M = 1$, the procedure can be applied component-wise in the multi-class case $M > 1$. We here proceed as in [5] and consider the so-called U-Bee method proposed in [12] for linear transport equation, which is defined by

$$\vec{\ }^{n+1,}_{j+1/2} := \vec{\ }^{n+1,}_{j} + \frac{1}{2}\bar{\ }_{j}\,'_{j}\left(\vec{\ }^{n+1,}_{j+1} - \vec{\ }^{n+1,}_{j}\right), \tag{2.20}$$

where $\bar{\ }_{j} = \max\{V^{n}_{j\ 1/2}, V^{n}_{j+1/2}\}$, $'_{j} := '(R_{j}, \bar{\ }_{j})$, with $R_{j} := \dfrac{\vec{\ }^{n+1,}_{j} - \vec{\ }^{n+1,}_{j\ 1}}{\vec{\ }^{n+1,}_{j+1} - \vec{\ }^{n+1,}_{j}}$ and

$$'(R, \bar{\ }) := '^{UB}(R, \bar{\ }) = \max\left(0, \min\left(\frac{2}{1}, \frac{2R}{\ }\right)\right). \tag{2.21}$$

Similarly, the so-called N-Bee method described, in [4], corresponds to a second-order scheme in space and it is more di↵usive that the U-Bee scheme. It is defined as in (2.20) with

$$'(R, \bar{\ }) := '^{NB}(R, \bar{\ }) := \max\left(0, \min\left(1, \frac{2R}{\ }\right), \min\left(R, \frac{2}{1}\right)\right). \tag{2.22}$$

It is proved in [4] that the numerical flux (2.20) for U-Bee and N-Bee methods satisfies the assumptions of Lemma 2.

2.5 Lagrangian-Antidi↵usive Remap scheme

Assume that $\vec{\ }^{n} = \vec{\ }^{n}_{1}, \ldots, \vec{\ }^{n}_{M}$ approximates the solution of (1.1) at time $t = t^{n}$ and we wish to advance this solution to $t^{n+1} = t^{n} + t$. To this end, two steps are performed successively:

1. *Lagrangian step*. Consider that $\vec{\ }^{n}$ are initial data for (2.5). First, we define the intermediate velocities $V^{n}_{i,j+1/2}$ by using (2.3), then we compute the numerical solution $\vec{\ }^{n+1,}_{i,j}$ of equation (2.5) after an evolution over a time interval of length t, using scheme (2.8).
2. *Antidi↵usive remap step*. Solve (2.6) with initial data $\vec{\ }^{n+1,}_{i,j}$ using an antidi↵usive scheme (2.12) for a specific choice of $\bar{V}^{n}_{i,j}$, obtaining a numerical solution $\vec{\ }^{n+1}$ which approximates the solution of (1.1) a time t^{n+1}.

In the next theorem, the choice of $\bar{V}^{n}_{i,j}$ is motivated by the existence of a classical conservative update formula for the whole L-AR scheme (2.5) plus (2.12).

Theorem 1. *Under the stability conditions (2.9) and (2.15), there exists a definition of $\bar{V}^{n}_{i,j}$ 2 $I(v^{n}_{i,j\ 1/2}, v^{n}_{i,j+1/2})$ such that the complete Lagrangian-Antidi↵usive remap scheme can be written in the conservative form*

$$\vec{\ }^{n+1}_{i,j} = \vec{\ }^{n}_{i,j} \left(\vec{\ }^{n+1,}_{i,j+1/2} V^{n}_{i,j+1/2} - \vec{\ }^{n+1,}_{i,j\ 1/2} V^{n}_{i,j\ 1/2}\right), \quad j\ 2\ Z. \tag{2.23}$$

Proof. Let $\{\vec{\ }^{n+1,}_{i,j}\}_{j2Z}$, be a solution of (2.5) obtained by scheme (2.8). Using this solution we solve (2.6) by the scheme (2.12), where the value $\bar{V}^{n}_{i,j}$ still needs to be determined in such a way that the resulting scheme is conservative. Replacing $\vec{\ }^{n+1,}_{i,j}$ in (2.12) we obtain

$$\vec{\ }^{n+1}_{i,j} = \vec{\ }^{n}_{i,j} \left(V^{n}_{i,j+1/2} - V^{n}_{i,j\ 1/2}\right) \vec{\ }^{n+1,}_{i,j} - \bar{V}^{n}_{i,j} \left(\vec{\ }^{n+1,}_{i,j+1/2} - \vec{\ }^{n+1,}_{i,j\ 1/2}\right). \tag{2.24}$$

As $\rightarrow^{n+1,}_{i,j+1/2}$ satisfies the assumptions of Lemma 2, there exist $\leftarrow_{i,j}$ 2 [0, 1] satisfying (2.19). Setting $V^n_{i,j} := \leftarrow_{i,j} V^n_{i,j\ 1/2} + (1 \quad \leftarrow_{i,j})V^n_{i,j+1/2}$ in (2.24), we obtain (2.23). □

Remark 2. *Note that, due to (2.13), the numerical flux* $F^n_{i,j+1/2} := \rightarrow^{n+1,}_{i,j+1/2} V^n_{i,j+1/2}$ *in (2.23) is consistent with the flux* $f_i(\rightarrow) = \rightarrow_i v(r \mid \leftarrow !_i)$.

As a consequence of Lemmas 1 and 2, we have the following property.

Lemma 3.(Positivity) *For any* $T > 0$, *under the stability conditions (2.9) and (2.15), the scheme (2.23) is positivity preserving on* $[0, T] \rightarrow \mathbb{R}$.

Moreover, in the scalar case, we have the following estimates.

Lemma 4 (**L^1 estimate**, case $M = 1$)**.** *Under conditions (2.15) and (2.9), and as a consequence of (2.10) and (2.17), we have*

$$\rightarrow^{n+1}_j \ 2\ I(\rightarrow^n_{j\ 1}, \rightarrow^n_{j+1}) \text{ for all } j\ 2\ \mathbb{Z}.$$

Lemma 5 (**BV estimates**, case $M = 1$)**.** *Assume (2.15) and*

$$t\ \eth\ \frac{1}{v^{\max} k\ {}^0k_1 \rightarrow^0_1 W_0}. \tag{2.25}$$

Let $\rightarrow^x$ *be constructed using (2.23). Then for every* $T > 0$ *the following discrete space BV estimate holds*

$$\mathrm{TV}(\rightarrow^x(T, \cdot))\ \eth\ e^{v^{\max} W_0 k \rightarrow^0 k_1 \downarrow_3 k\ {}^0k_1 +_5 k\ {}^{00}k_1\ k \rightarrow^0 k_1\ J_1 \overset{\text{⌘}}{T}}\ \mathrm{TV}(\rightarrow^0).$$

Proof. Setting $v(\leftarrow) := v^{\max}(\leftarrow)$, from (2.8) we recover

$$\rightarrow^{n+1,}_{j+1} \quad \rightarrow^{n+1,}_j = \rightarrow^n_{j+1} \quad \rightarrow^n_j \quad \rightarrow^{n+1,}_{j+1} \quad \overset{\downarrow}{V}{}^n_{j+3/2} \quad V^n_{j+1/2} \overset{\text{⌘}}{} + \rightarrow^{n+1,}_j \quad \overset{\downarrow}{V}{}^n_{j+1/2} \quad V^n_{j\ 1/2} \overset{\text{⌘}}{}.$$

We have

$$\rightarrow^{n+1,}_{j+1} \quad \overset{\downarrow}{V}{}^n_{j+3/2} \quad V^n_{j+1/2} \overset{\text{⌘}}{} + \rightarrow^{n+1,}_j \quad \overset{\downarrow}{V}{}^n_{j+1/2} \quad V^n_{j\ 1/2} \overset{\text{⌘}}{} \tag{2.26a}$$

$$= \rightarrow^{n+1,}_{j+1} \quad v^0(\leftarrow_{j+1}) \quad x \overset{\overset{\text{*}}{1}}{\underset{k=1}{}} !^k \overset{\downarrow}{\rightarrow}{}^n_{j+k+1} \quad \rightarrow^n_{j+k} \overset{\text{⌘}}{} + \rightarrow^{n+1,}_j \quad v^0(\leftarrow_j) \quad x \overset{\overset{\text{*}}{1}}{\underset{k=1}{}} !^k \overset{\downarrow}{\rightarrow}{}^n_{j+k} \quad \rightarrow^n_{j+k\ 1} \overset{\text{⌘}}{} \tag{2.26b}$$

$$= t^{h} \rightarrow^{n+1,}_{j+1} \quad \rightarrow^{n+1,}_j \quad {}^{i} v^0(\leftarrow_{j+1}) \overset{\overset{\text{*}}{1}}{\underset{k=1}{}} !^k \overset{\downarrow}{\rightarrow}{}^n_{j+k+1} \quad \rightarrow^n_{j+k} \overset{\text{⌘}}{} \tag{2.26c}$$

$$t \rightarrow^{n+1,}_j \quad \vec{v}{}^0(\leftarrow_{j+1}) \quad v^0(\leftarrow_j) \overset{\overset{\text{*}}{1}}{\underset{k=1}{}} !^k \overset{\downarrow}{\rightarrow}{}^n_{j+k+1} \quad \rightarrow^n_{j+k} \overset{\text{⌘}}{} \tag{2.26d}$$

$$t \rightarrow^{n+1,}_j \quad v^0(\leftarrow_j) \overset{2}{4} \overset{\overset{\text{*}}{1}}{\underset{k=1}{}} !^k (\rightarrow^n_{j+k+1} \quad \rightarrow^n_{j+k}) \overset{\overset{\text{*}}{1}}{\underset{k=1}{}} !^k (\rightarrow^n_{j+k} \quad \rightarrow^n_{j+k\ 1}) \overset{3}{5} \tag{2.26e}$$

$$= t^{h} \rightarrow^{n+1,}_{j+1} \quad \rightarrow^{n+1,}_j \quad {}^{i} v^0(\leftarrow_{j+1}) \overset{\overset{\text{*}}{1}}{\underset{k=1}{}} !^k \overset{\downarrow}{\rightarrow}{}^n_{j+k+1} \quad \rightarrow^n_{j+k} \overset{\text{⌘}}{} \tag{2.26f}$$

$$t\,\vec{\jmath}^{\,n+1,}\quad \vec{v}^{0}(\overleftarrow{\jmath}_{j+1})\quad v^{0}(\overleftarrow{\jmath}_{j})\overleftarrow{\sum_{k=1}^{\infty}}\,!^{k}\,\dot{\vec{\jmath}}^{\,n}_{j+k+1}\quad \vec{\jmath}^{\,n}_{j+k}\quad {}^{\Cmd} \tag{2.26g}$$

$$t\,\vec{\jmath}^{\,n+1,}\quad v^{0}(\overleftarrow{\jmath}_{j})\,\Big[\sum_{k=1}^{\infty}\,!^{k}(\vec{\jmath}^{\,n}_{j+k+1}\quad \vec{\jmath}^{\,n}_{j+k})\quad \sum_{k=1}^{\infty}\,!^{k}(\vec{\jmath}^{\,n}_{j+k}\quad \vec{\jmath}^{\,n}_{j+k\quad 1})\Big]. \tag{2.26h}$$

This implies

$$\Big[1+\quad t\,v^{0}(\overleftarrow{\jmath}_{j+1})\sum_{k=1}^{\infty}\,!^{k}\,\dot{\vec{\jmath}}^{\,n}_{j+k+1}\quad \vec{\jmath}^{\,n}_{j+k}\Big]^{\Cmd}\,\dot{\vec{\jmath}}^{\,n+1,}_{j+1}\quad \vec{\jmath}^{\,n+1,}_{j}{}^{\Cmd} \tag{2.27a}$$

$$=\vec{\jmath}^{\,n}_{j+1}\quad \vec{\jmath}^{\,n}_{j} \tag{2.27b}$$

$$t\,\vec{\jmath}^{\,n+1,}\quad \vec{v}^{0}(\overleftarrow{\jmath}_{j+1})\quad v^{0}(\overleftarrow{\jmath}_{j})\overleftarrow{\sum_{k=1}^{\infty}}\,!^{k}\,\dot{\vec{\jmath}}^{\,n}_{j+k+1}\quad \vec{\jmath}^{\,n}_{j+k}{}^{\Cmd} \tag{2.27c}$$

$$t\,\vec{\jmath}^{\,n+1,}\quad v^{0}(\overleftarrow{\jmath}_{j})\,\Big[\sum_{k=1}^{\infty}\,!^{k}(\vec{\jmath}^{\,n}_{j+k+1}\quad \vec{\jmath}^{\,n}_{j+k})\quad \sum_{k=1}^{\infty}\,!^{k}(\vec{\jmath}^{\,n}_{j+k}\quad \vec{\jmath}^{\,n}_{j+k\quad 1})\Big]. \tag{2.27d}$$

$$\tag{2.27e}$$

Observe that

$$1+\quad t\,v^{0}(\overleftarrow{\jmath}_{j+1})\sum_{k=1}^{\infty}\,!^{k}\,\dot{\vec{\jmath}}^{\,n}_{j+k+1}\quad \vec{\jmath}^{\,n}_{j+k}{}^{\Cmd}\quad 1\quad t\quad v^{0}{}_{1}\quad \vec{\jmath}^{\,0}{}_{1}\,W_{0}$$

which is positive if $t\,\delta\,\dot{\vec{\jmath}}\,v^{0}{}_{1}\quad \vec{\jmath}^{\,0}{}_{1}\,W_{0}{}^{\Cmd}{}_{1}$. Moreover, we have that

$$v^{0}(\overleftarrow{\jmath}_{j+1})\quad v^{0}(\overleftarrow{\jmath}_{j})=v^{00}(\dot{\overleftarrow{\jmath}}_{j+1/2})(\overleftarrow{\jmath}_{j+1}\quad \overleftarrow{\jmath}_{j}),$$

with $\dot{\overleftarrow{\jmath}}_{j+1/2}\quad 2\,I\,\overleftarrow{\jmath}_{j},\overleftarrow{\jmath}_{j+1}$. We can compute

$$\begin{aligned}
\overleftarrow{\jmath}_{j+1}\quad \overleftarrow{\jmath}_{j} &= \#\quad x\sum_{k=1}^{\infty}\,!^{k}\,\vec{\jmath}^{\,n}_{j+k+1}\quad +(1\quad \#)\quad x\sum_{k=1}^{\infty}\,!^{k}\,\vec{\jmath}^{\,n}_{j+k}\\
&\quad \mu\quad x\sum_{k=1}^{\infty}\,!^{k}\,\vec{\jmath}^{\,n}_{j+k}\quad (1\quad \mu)\quad x\sum_{k=1}^{\infty}\,!^{k}\,\vec{\jmath}^{\,n}_{j+k\quad 1}\\
&=\#\quad x\sum_{k=2}^{\infty}\,!^{k\quad 1}\,\vec{\jmath}^{\,n}_{j+k}\quad +(1\quad \#)\quad x\sum_{k=1}^{\infty}\,!^{k}\,\vec{\jmath}^{\,n}_{j+k}\\
&\quad \mu\quad x\sum_{k=1}^{\infty}\,!^{k}\,\vec{\jmath}^{\,n}_{j+k}\quad +(1\quad \mu)\quad x\sum_{k=0}^{\infty}\,!^{k+1}\,\vec{\jmath}^{\,n}_{j+k}\\
&= x\sum_{k=2}^{\infty}\Big[\#!^{k\quad 1}+(1\quad \#)!^{k}\quad \mu!^{k}\quad (1\quad \mu)!^{k+1}\Big]\,\vec{\jmath}^{\,n}_{j+k}\\
&\quad +\quad x\Big[(1\quad \#)!^{1}\,\vec{\jmath}^{\,n}_{j+1}\quad \mu!^{1}\,\vec{\jmath}^{\,n}_{j+1}\quad (1\quad \mu)!^{1}\,\vec{\jmath}^{\,n}_{j}\quad (1\quad \mu)!^{2}\,\vec{\jmath}^{\,n}_{j+1}\Big].
\end{aligned}$$

By monotonicity of $!$ we have

$$\#!^{k-1} + (1-\#)!^{k} - \mu!^{k} - (1-\mu)!^{k+1} \geq 0.$$

Taking the absolute values we get

$$\begin{aligned}
\left|\xi_{j+1} - \xi_j\right| &\leq \left\|x\right\|_{0,1} \left\{ \sum_{k=2}^{n} \left[\#!^{k-1} + (1-\#)!^{k} - \mu!^{k} - (1-\mu)!^{k+1} \right] + 3!^{1} \right\} \\
&= \left\|x\right\|_{0,1} \left\{ \#!^{1} + (1-\mu)!^{2} + 3!^{1} \right\} \\
&\leq \left\|x\right\|_{0,1} W_0.
\end{aligned}$$

Taking the absolute values in (2.27) we get

$$\begin{aligned}
&\left(1 - t\,\|v^0\|_{1,\infty} W_0\right) \sum_j \left|u^{n+1,\varepsilon}_{j+1} - u^{n+1,\varepsilon}_{j}\right| \\
&\leq \left(1 + \left(\sum_{k=1}^{n} \left|u^{n+1,\varepsilon}_{j-k} - v^0(\xi_{j+1-k})\right| - \left|v^0(\xi_{j-k})\right| !^{k} - \left|u^{n+1,\varepsilon}_{j-k} - v^0(\xi_j)\right|(!^{k} - !^{k+1})\right)\right) \\
&\quad \left(\left|u^{n+1,\varepsilon}_{j} - v^0(\xi_j)\right| !^{1}\right) + t \sum_j \left|u^{n}_{j+1} - u^{n}_{j}\right| \\
&\leq \left(1 + t\,\|v^0\|_{2,\infty} W_0 \sum_{k=1}^{n} !^{k} + 2\,t\,W_0\,\|v^0\|_{1,\infty} \sum_j \left|u^{n}_{j+1} - u^{n}_{j}\right|\right) \\
&\leq \left(1 + t\,\|v^0\|_{1,\infty} W_0 \left(2\,\|v^0\|_{1,\infty} + 5\,\|v^{00}\|_{1,\infty} \|J_1\|_{\infty,1}\right) \sum_j \left|u^{n}_{j+1} - u^{n}_{j}\right|\right),
\end{aligned}$$

which, together with the TVD property of the remap step [4, 5], implies

$$\begin{aligned}
TV(u_x(T,\cdot)) &\leq \left(\frac{1 + t\,\|v^0\|_{1,\infty} W_0 \left(2\,\|v^0\|_{1,\infty} + 5\,\|v^{00}\|_{1,\infty} \|J_1\|_{\infty,1}\right)}{1 - t\,\|kv^0\|_{k,1} \|v^0\|_{1,\infty} W_0}\right) TV(u^{\varepsilon}_x(0,\cdot)) \\
&\leq e^{\|k\|_{\infty,0} W_0 \left(3\|kv^0\|_{k,1} + 5\|kv^{00}\|_{k,1} \|k\|_{\infty,0} \|J_1\|_{\infty,1}\right) T}\, TV(u^0).
\end{aligned}$$

□

Remark 3. *Following the ideas of Lemmas 2 and 4 in [10], it is possible to find the local time extension to the above estimates for the case $M > 1$.*

Next Theorem follows from Theorem 1 and Lemmas 3, 4 and 5.

Theorem 2 (Convergence to weak solutions, case $M = 1$)**.** *Let us consider the Cauchy problem (1.1)-(1.4) with $M = 1$, $u^0(x) \in$ **BV**(R; [0, 1]), under the assumptions* **(H1)** - **(H3)**. *If (2.15) and (2.9) hold, then the approximate solution u^{ε}_x constructed by the scheme (2.23) converges to a weak solution of (1.1)- (1.4).*

Proof. Under conditions (2.15) and (2.9), the approximate solutions $\rightarrow^{x}$ constructed by the numerical scheme (2.23) are uniformly bounded and uniformly bounded total variation. The result follows by standard application of Helly's Theorem, see [16, Proposition 4.1] or [13, Lemma 5.6]. □

3 Two simple schemes for the non-local multi-class traffic flow model

In Section 4, we consider the following conservative schemes for the multi-class model (1.1) in the form

$$\rightarrow^{n+1}_{i,j} = \rightarrow^{n}_{i,j} \quad \downarrow F^{n}_{i,j+1/2} \quad F^{n}_{i,j} \quad {}_{1/2} \quad \text{⌘}, \qquad i = 1, \ldots, M. \tag{3.1}$$

First, we consider the Godunov-type scheme , which was introduced in [14] in the scalar case and then extended to (1.1) in [10], with numerical flux

$$F^{n}_{i,j+1/2} := \rightarrow^{n}_{i,j} V^{n}_{i,j+1/2} \ . \tag{3.2}$$

We recall that for scheme (3.1)-(3.2) the positivity is guaranteed if

$$\text{ð} \ \frac{1}{v^{\max}_{M} k \, k \ {}_{1}}.$$

We consider also the approximate solutions constructed via the following adapted Lax-Friedrichs flux, that was used [3, 9] in the scalar case and in [10] for system (1.1):

$$F^{n}_{i,j+1/2} := \frac{1}{2} \downarrow \rightarrow^{n}_{i,j} V^{n}_{i,j} \quad {}_{1/2} + \rightarrow^{n}_{i,j+1} V^{n}_{i,j} \quad {}_{3/2} \quad \text{⌘} + \frac{\hookleftarrow}{2} \downarrow \rightarrow^{n}_{i,j} \quad \rightarrow^{n}_{i,j+1} \quad \text{⌘}, \tag{3.3}$$

where $\hookleftarrow \ v^{\max}_{M} k \, k \ {}_{1}$ is the viscosity coefficient and $\hookleftarrow \text{ð} \, 1$ the CFL condition.

3.1 A second-order Godunov scheme

Schemes (3.1)-(3.2) and (3.1)-(3.3) are only first-order accurate. We propose here a second-order accuracy scheme,constructed using MUSCL-type variable extrapolation and Runge-Kutta temporal di↵erencing. To implement it, we approximate $\rightarrow_{i}(x, t^{n})$ by a piecewise linear functions in each cell, i.e. $\hat{\rightarrow}_{i,j}(x, t^{n}) = \rightarrow^{n}_{i,j} + {}^{n}_{i,j}(x \quad x_{j})$, where the slopes ${}^{n}_{i,j}$ are calculated via the generalized minmod limiter, i.e.

$${}^{n}_{i,j} = \frac{1}{x} \text{minmod}(\#(\rightarrow^{n}_{i,j} \quad \rightarrow^{n}_{i,j} \quad {}_{1}), \frac{1}{2}(\rightarrow^{n}_{i,j+1} \quad \rightarrow^{n}_{i,j} \quad {}_{1}), \#(\rightarrow^{n}_{i,j+1} \quad \rightarrow^{n}_{i,j})),$$

where # *2* [1, 2] and

$$\text{minmod}(a, b, c) := \begin{cases} \text{sgn}(a) \min\{|a|, |b|, |c|\} & \text{if sgn}(c) = \text{sgn}(b) = \text{sgn}(a) \\ 0 & \text{otherwise.} \end{cases}$$

This extrapolation enables one to define left and right values at the cell interfaces respectively by

$$\rightarrow^{L}_{i,j+1/2,} := \hat{\rightarrow}_{i,j}(x_{j} + \quad x/2, t^{n}) = \rightarrow^{n}_{i,j} + {}^{n}_{i,j} \quad x/2,$$

$$\rightarrow^{R}_{i,j\ 1/2,} := \hat{\rightarrow}_{i,j}(x_j \quad x/2, t^n) = \rightarrow^{n}_{i,j} \quad {}^{n}_{i,j} \quad x/2.$$

In order to define the corresponding velocity approximations, we set

$$r^{n}_{j+k} = \sum_{i=1}^{M} \rightarrow^{n}_{i,j+k} = r^{n}_{j+k} + \rightarrow^{n}_{j+k}(x \quad x_{j+k}),$$

where $\rightarrow^{n}_{j+k} := \sum_{i=1}^{M} {}^{n}_{i,j+k}$, and

$$\hat{V}^{n}_{i,j+1/2} := v_i((\tau \leftarrow w_i)(t^n, x_{j+1/2})) = v_i^{\max} \left(\sum_{k=1}^{\infty} \mathcal{F}^{k}_{i} r^{n}_{j+k} + \sum_{k=1}^{\infty} \mathcal{F}^{k}_{i,j} \rightarrow^{n}_{j+k} \right) \tag{3.4}$$

for $i = 1, \ldots, M$, $j \in \mathbb{Z}$, where $\mathcal{F}^{k}_{i,j} := \frac{1}{x} \int_{x/2}^{x/2} y!_i(y + (k \quad 1/2) \quad x)\, dy$. The MUSCL version of the i-th flux component thus reads

$$f^{n}_{i,j+1/2} := \rightarrow^{L}_{i,j+1/2} \hat{V}^{n}_{i,j+1/2}.$$

To achieve formal second-order accuracy also in time, we use second-order RungeKutta (RK) time stepping. More precisely, if we write our scheme with first-order Euler time differences and second-order spatial differences formally as

$$\rightarrow^{n+1}_{j} = \rightarrow^{n}_{j} \quad L_j(\rightarrow^{n}) := \rightarrow^{n}_{j} \quad \mathbf{F}^{n}_{j+1/2} \quad \mathbf{F}^{n}_{j\ 1/2}, \tag{3.5}$$

then the RK version takes the following two-step form

$$\begin{cases} \rightarrow^{(1)}_{j} = \rightarrow^{n}_{j} \quad L_j(\rightarrow^{n}) \\ \rightarrow^{n+1}_{j} = \frac{1}{2}(\rightarrow^{n}_{j} + \rightarrow^{(1)}_{j}) \quad \frac{1}{2} L_j(\rightarrow^{(1)}_{j}). \end{cases} \tag{3.6}$$

Lemma 6. *For any $T > 0$, under the CFL condition*

$$t \leq \frac{x}{2 v_M^{\max} k\, k_1}, \tag{3.7}$$

the scheme (3.6) is positivity preserving on $\mathbb{R} \rightarrow [0, T]$.

Proof. Let us assume that $\rightarrow^{n}_{i,j} \quad 0$ for $j \quad 2\mathbb{Z}$ and $i = 1, \ldots, M$. The positivity of the reconstructed values $\rightarrow^{L}_{i,j+1/2}$ and $\rightarrow^{L}_{i,j+1/2}$ is guaranteed by the positivity preserving property of the chosen limiter [19, 20]. It suffices to prove that $\rightarrow^{n+1}_{i,j} \quad 0$ in (3.5). Due to $\rightarrow^{n}_{i,j} = \frac{1}{2}(\rightarrow^{R}_{i,j\ 1/2} + \rightarrow^{L}_{i,j+1/2})$, the i-th term in (3.5) can be written in the form

$$\rightarrow^{n+1}_{i,j} = \frac{1}{2} \rightarrow^{R}_{i,j\ 1/2} + \frac{1}{2} \hat{V}^{n}_{i,j+1/2} \rightarrow^{L}_{i,j+1/2} + \rightarrow^{L}_{i,j\ 1/2} \hat{V}^{n}_{i,j\ 1/2} \quad 0,$$

under the CFL condition (3.7). □

4 Numerical results

In this section, we solve (1.1) numerically in the intervals $x \in [a, b]$ with suitable boundary conditions and $t \in [0, T]$, for values of a, b and T specified later. We propose several test cases to illustrate the behaviour of the L-AR schemes (2.20) and (2.21) in comparison with first-order Lax-Friedrichs (Lax-F) (3.1)-(3.3) and Godunov (3.1)-(3.2) schemes and the second-order Godunov scheme (Godunov2) (3.6). As benchmarks, we consider the numerical examples studied in [14, Section 4.2] for the scalar case and in [10, Sections 4.1-4.2] for the multi-class system case.

From now on, we refer to the numerical scheme (2.23) applied in a component-wise manner with velocities (2.3) and $\rho^{n+1,}_{i,j+1/2}$ computed with (2.20) and (2.21) (respectively (2.22)) as L-UBee (respectively L-NBee) scheme. For each integration, we set t to satisfy the most restrictive CFL condition (3.7). Clearly, the computational bottleneck in these schemes is the discrete convolution term (2.3) corresponding to (1.3). This is a classical problem in scientific computing, which can be effectively evaluated using fast convolution algorithms, mainly based on Fast Fourier Transforms [21].

For each numerical test, we specify the type of boundary conditions that are imposed. Given a uniform partition of $[a, b], \{I_j\}_{j=1}^{N}$ with $x = (b - a)/N$, for absorbing boundary conditions, we define $\rho^{n}_{i,j}$ in the ghost cells for $i = 1, \ldots, M$ as follows:

$$\rho^{n}_{i,0} = \rho^{n}_{i,1}, \quad \text{and} \quad \rho^{n}_{i,N+j} = \rho^{n}_{i,N}, \quad \text{for } j = 1, 2, \ldots, L,$$

and for periodic boundary conditions,

$$\rho^{n}_{i,0} = \rho^{n}_{i,N}, \quad \text{and} \quad \rho^{n}_{i,N+j} = \rho^{n}_{i,j}, \quad \text{for } j = 1, 2, \ldots, L,$$

with $L > \max_{i=1,\ldots,M} \varepsilon_i / x$.

Since we cannot compute the exact solution explicitly, we use the second-order Godunov scheme with a refined mesh to obtain a reference solution. The $\mathbf{L^1}$-error for the cell average is given by

$$L^1(x) = \sum_{i=1}^{M} \left(\frac{1}{N} \sum_{j=1}^{N} |\rho_{i,j} - \rho^{ref}_{i,j}| \right),$$

where $\rho_{i,j}$ and $\rho^{ref}_{i,j}$ are the cell averages of the numerical approximation and the reference solution respectively. The Experimental Order of Accuracy (E.O.A.) is naturally defined by

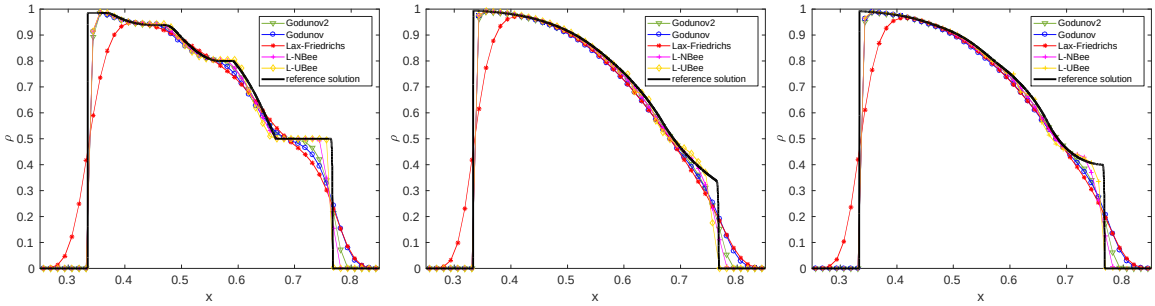
$$(x) = \log_2 \left(L^1(x)/L^1(x/2) \right).$$

4.1 Test 1, scalar case

We consider the problem (1.1) for $M = 1$, with initial datum

$$\rho_0(x) = \begin{cases} 1, & \text{if } 1/3 \le x \le 2/3 \\ \frac{1}{3}, & \text{otherwise,} \end{cases} \tag{4.1}$$

for $x \in [0, 1]$, with absorbing boundary conditions, and different non-increasing kernel functions with $\varepsilon = 0.1$. In Figure 1, we display the numerical approximations obtained with the schemes presented in the previous sections, computed with $1/x = 80$ at $T = 0.1$.

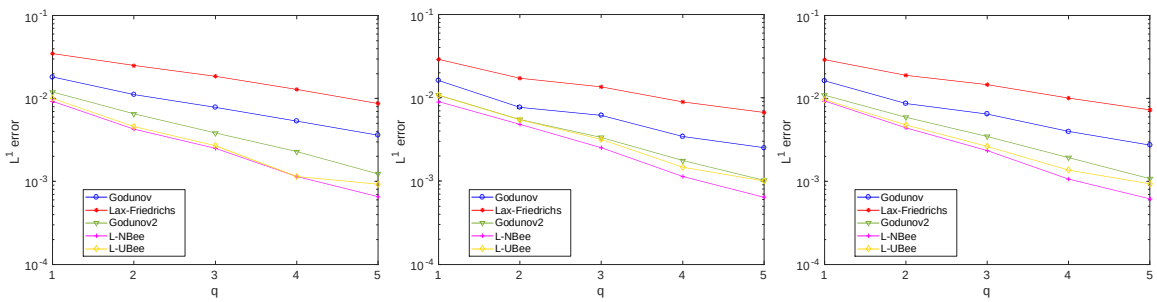


(a) (b) (c)

Figure 1: Test 1: Comparison of the numerical solutions at $T = 0.1$ corresponding to the initial condition (4.1), computed with $1/\quad x = 80$ and di↵erent kernel functions. (a) $!(x) = 1/\mathcal{H}$, (b) $!(x) = 2(\mathcal{H}\quad x)/\mathcal{H}^2$, (c) $!(x) = 3(\mathcal{H}^2\quad x^2)/(2\mathcal{H}^3)$.

Fig. 1a show the result for $!(x) = 1/\mathcal{H}$, Fig. 1b for $!(x) = 2(\mathcal{H}\quad x)/\mathcal{H}^2$ and Fig. 1c for $!(x) = 3(\mathcal{H}^2\quad x^2)/(2\mathcal{H}^3)$. The reference solution is computed with $1/\quad x = 10240$. The numerical solutions obtained with L-UBee and L-NBee approximate adequately shocks and rarefaction waves according to the theoretical results of Theorem 2In particular, concerning the shock waves,L-AR schemes capture the reference solution better than the second-order Godunov scheme, whereas the solutions computed with Lax-Friedrichs and Godunov schemes are more di↵usive. In the presence of rarefaction waves, L-UBee scheme produces "stair-caising" due to the particular choice of the antidi↵usive scheme. We can observe the same "staircaising" phenomena also for the linear advection and other equations [4, 5].

Table 1 shows the approximate $L^1(\quad x)$-errors and the numerical orders of accuracy $(\quad x)$ for the di↵erent schemes.We computed numerical approximations with $1/\quad x = 40 \rightarrow 2^q$ for $q = 1, 2, ..., 5$. Clearly, the error of the L-AR schemes decreases when the mesh is refined and we observe that for each level of refinement, the $\mathbf{L}^1$-error of the L-AR schemes is smaller than the respective errors of Lax-Friedrichs and Godunov schemes.In conclusion, when the solution presents discontinuities, we can compare the performances of the L-AR schemes with those of a second-order scheme.



(a) (b) (c)

Figure 2: Test 1. Initial condition (4.1). Approximate $\mathbf{L}^1$-error for di↵erent numerical schemes with: (a) constant kernel function $!(x) = 1/\mathcal{H}$, (b) decreasing kernelfunction $!(x) = 2(\mathcal{H}\quad x)/\mathcal{H}^2$, (c) concave kernel function $!(x) = 3(\mathcal{H}^2\quad x^2)/(2\mathcal{H}^3)$.

Now, in order to determine the correct order of accuracy of the L-AR schemes, we consider

Scheme		$!(x) = 1/\mathcal{H}$		$!(x) = 2(\mathcal{H} \quad x)/\mathcal{H}^2$		$!(x) = 3(\mathcal{H}^2 \quad x^2)/(2\mathcal{H}^3)$	
	$1/\ x$	$\mathbf{L}^1$ error	$(\ x)$	$\mathbf{L}^1$ error	$(\ x)$	$\mathbf{L}^1$ error	$(\ x)$
Godunov	80	1.81e-02	–	1.62e-02	–	1.64e-02	–
	160	1.12e-02	6.98e-01	7.73e-0.3	1.06	8.72e-03	9.11e-01
	320	7.85e-03	5.10e-01	6.15e-03	3.29e-01	6.53e-03	4.19e-01
	640	5.33e-03	5.58e-01	3.43e-03	8.43e-01	4.01e-03	7.04e-01
	1280	3.62e-03	5.58e-01	2.51e-0.3	4.50e-01	2.76e-03	5.39e-01
Lax-F	80	3.48e-02	–	2.89e-02	–	2.94e-02	–
	160	2.50e-02	4.81e-01	1.72e-02	7.50e-01	1.91e-02	6.23e-01
	320	1.86e-02	4.24e-01	1.35 e-02	3.46e-01	1.48e-02	3.67e-01
	640	1.29e-02	5.28e-01	8.94e-03	5.7e-01	1.02e-02	5.38e-01
	1280	8.72e-03	5.64e-01	6.670e-03	4.23e-01	7.30e-03	4.70e-01
L-NBee	80	9.30e-03	–	8.93e-03	–	9.24e-03	–
	160	4.29e-03	1.11	4.78e-03	9.01e-01	4.50e-03	1.03e-01
	320	2.51e-03	7.47e-01	2.52e-03	9.27e-01	2.37e-03	9.25e-01
	640	1.58e-03	1.11	1.15e-03	1.13	1.08e-03	1.13
	1280	6.57e-04	8.17e-01	6.46e-04	8.31e-01	6.19e-04	8.05e-01
L-UBee	80	1.00e-02	–	8.90e-03	–	9.09e-03	–
	160	4.58e-03	1.13	4.40e-03	1.02	4.82e-03	9.16e-01
	320	2.7e-03	7.62e-01	2.87e-03	6.61e-01	2.62e-03	8.80e-01
	640	1.15e-03	1.23	1.38e-03	1.05	1.37e-03	9.30e-01
	1280	9.48e-04	2.76e-01	9.69e-04	5.13e-01	9.00e-04	6.11e-01
Godunov2	80	1.20e-02	–	1.08e-02	–	1.01e-02	–
	160	6.54e-03	8.70e-01	5.5e-03	9.71e-01	5.96e-03	8.86e-01
	320	3.82e-03	7.73e-01	3.35e-03	7.17e-01	3.51e-03	7.64e-01
	640	2.29e-03	7.42e-01	1.76e-03	9.25e-01	1.94e-03	8.53e-01
	1280	1.23e-03	8.89e-01	1.02e-03	7.87e-01	1.08e-03	8.42e-01

Table 1: Test 1. Approximate $\mathbf{L}^1$-error and E.O.A. for di↵erent numerical schemes and with di↵erent kernel functions and $\mathcal{H} = 0.1$ corresponding to the initial condition (4.1).

a smooth initial datum

$$\rho_0(x) = 0.5 + 0.4 \sin(\pi x) \tag{4.2}$$

for $x \in [\ 1, 1]$, with periodic boundary conditions and compute the numerical approximation at $T = 0.15$ for di↵erent kernel functions with $\mathcal{H} = 0.1$. The reference solution is computed with $1/\ x = 10240$. In Table 2 and Figure 3 we compute the L^1-error and E.O.A. $(\ x)$. We recover the correct order of accuracy for the second-order Godunov scheme. Instead, we obtain just first-order accuracy for L-AR schemes. However, it is worth underlying that the $\mathbf{L}^1$-error of the L-NBee scheme is smaller than the corresponding error for Lax-Friedrichs and Godunov schemes. For the L-UBee scheme, we obtain first order accuracy and the L^1-error for each level of refinement is bigger than the error of the other first order numerical schemes, due to the antidi↵usive property of the UBee scheme.

4.2 Test 2. Cars and trucks mixed traffic

In this test case, we consider a stretch of road populated by cars and trucks as in the example proposed in [10, Section 4.1]. The space domain is given by the interval $[\ 1, 1]$ and we impose absorbing conditions at the boundaries. The dynamics is described by the equation (1.1) with

Scheme		$!(x) = 1/\mathscr{H}$		$!(x) = 2(\mathscr{H}\ x)/\mathscr{H}^2$		$!(x) = 3(\mathscr{H}^2\ x^2)/(2\mathscr{H}^3)$	
	$1/\ x$	$\mathbf{L}^1$ error	$(\ x)$	$\mathbf{L}^1$ error	$(\ x)$	$\mathbf{L}^1$ error	$(\ x)$
Godunov	80	1.28e-03	–	1.33e-03	–	1.33e-03	–
	160	6.44e-04	9.88e-01	6.73e-0.4	9.95e-01	6.68e-04	9.94e-01
	320	3.23e-04	9.94e-01	3.38e-04	9.97e-01	3.34e-04	9.97e-01
	640	1.62e-04	9.97e-01	1.69e-04	9.99e-01	1.67e-04	9.98e-01
	1280	8.11e-05	9.98e-01	8.47e-0.5	9.99e-01	8.38e-05	9.99e-01
Lax-F	80	1.58e-03	–	1.92e-03	–	1.76e-03	–
	160	7.24e-04	1.12	8.14e-04	1.24	7.73e-04	1.18
	320	3.46e-04	1.07	3.70e-04	1.14	3.59e-04	1.10
	640	1.69e-04	1.03	1.77e-04	1.06	1.74e-04	1.05
	1280	8.35e-05	1.02	8.67e-04	1.03	8.55e-05	1.02
L-NBee	80	4.55e-04	–	4.30e-04	–	4.36e-04	–
	160	2.23e-04	1.02	2.24e-04	9.43e-01	2.24e-04	9.65e-01
	320	1.10e-04	1.01	1.14e-04	9.72e-01	1.13e-04	9.83e-01
	640	5.49e-04	1.01	5.76e-05	9.86e-01	5.69e-05	9.92e-01
	1280	2.74e-05	1.00	2.89e-05	9.93e-01	2.85e-05	9.96e-01
L-UBee	80	2.30e-03	–	2.14e-03	–	2.16e-03	–
	160	1.75e-03	3.96e-01	1.23e-03	7.97e-01	1.26e-03	7.72e-01
	320	1.48e-03	2.49e-01	1.18e-03	5.59e-02	1.20e-03	8.05e-02
	640	9.82e-04	5.89e-01	8.39e-04	4.98e-01	8.41e-04	5.09e-01
	1280	5.06e-04	9.56e-01	4.53e-04	8.88e-01	4.63e-04	8.61e-01
Godunov2	80	2.86e-05	–	2.89e-05	–	2.89e-05	–
	160	6.80e-06	2.07	6.74e-06	2.10	6.76e-06	2.09
	320	1.53e-06	2.15	1.53e-06	2.14	1.53e-06	2.14
	640	3.42e-07	2.16	3.42e-07	2.16	3.41e-07	2.16
	1280	7.72e-08	2.15	7.75e-08	2.14	7.73e-08	2.14

Table 2: Test 1. Approximate $\mathbf{L}^1$-error and E.O.A. , with smooth initial condition (4.2) and di↵erent kernels function with $\mathscr{H} = 0.1$.

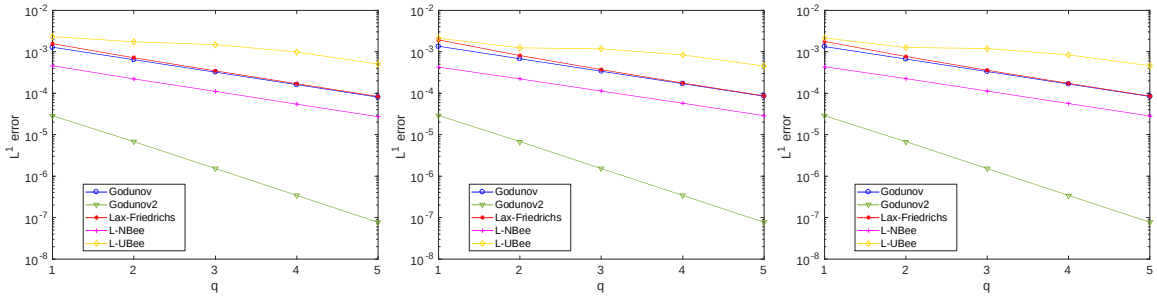
$M = 2$, and the following initial conditions and parameter values

$$\begin{aligned} &\rightarrow_1(0, x) = 0.5\ {}_{[\ 0.6,\ \ 0.1]}(x), \quad !_1(x) = \frac{2}{\mathscr{H}_1}\ {}^{\checkmark}\ 1\ \frac{x}{\mathscr{H}_1}\ {}^{\blacklozenge}, \quad \mathscr{H}_1 = 0.3, \quad v_1^{\max} = 0.8, \\ &\rightarrow_2(0, x) = 0.5\ {}_{[\ 0.9,\ \ 0.6]}(x), \quad !_2(x) = \frac{2}{\mathscr{H}_2}\ {}^{\checkmark}\ 1\ \frac{x}{\mathscr{H}_2}\ {}^{\blacklozenge}, \quad \mathscr{H}_2 = 0.1, \quad v_2^{\max} = 1.3. \end{aligned} \tag{4.3}$$

In this setting, $\rightarrow_1(t, x)$ and $\rightarrow_2(x, t)$ describe the density of trucks and cars respectively. We have a red traffic light located at $x =$ 0.1, which turns green at the initial time $t = 0$. In Figure 4, we display the reference solution of equation (1.1) with initial conditions and parameters (4.3), computed with with $1/\ x = 5120$ at increasing time instants ($T = 0.25$ in Fig. 4a, $T = 0.5$ in Fig. 4b and $T = 1$ in Fig. 4c). We can observe that cars, initially located behind trucks, overtake the slower class while impacting its flow.

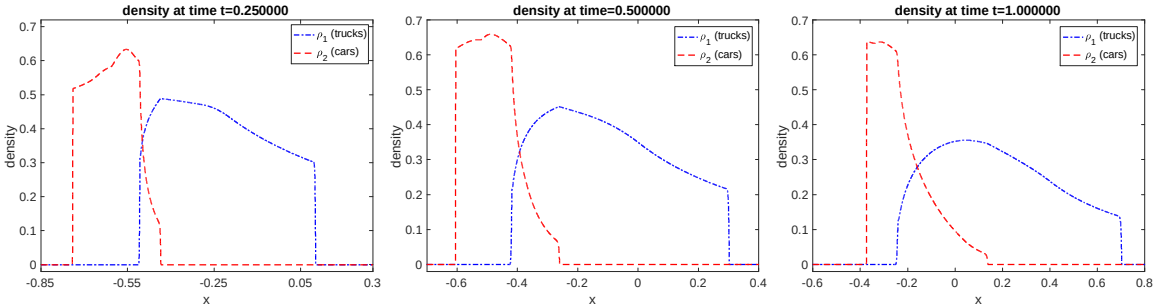
In Figure 5, we display separately the two density components of the approximate solutions computed using all the considered schemes with $1/\ x = 80$, compared to the reference solution of Figure 4. The numerical tests indicate that for $M > 1$, the L-AR solutions are anti-di↵usive for each class and they keep this anti-di↵usive behavior for the whole simulation time. We observe that the L-NBee solution approaches very well the reference solution for each class at di↵erent times. On the other hand, the L-UBee solution shows “stairs” in the presence of rarefaction-waves.

In Table 3 and Figure 6, we compute the approximate $\mathbf{L}^1$-error and the E.O.A at time



(a) (b) (c)

Figure 3: Test 1. Initial condition (4.2). Approximate $\mathbf{L^1}$-error for di↵erent numerical schemes with: (a) constant kernel function $!(x) = 1/\mathscr{H}$, (b) decreasing kernel function $!(x) = 2(\mathscr{H}\ x)/\mathscr{H}^2$, concave kernel function $!(x) = 3(\mathscr{H}^2\ \ x^2)/(2\mathscr{H}^3)$ (c).



(a) (b) (c)

Figure 4: Test 2: Density profiles corresponding to (1.1)-(4.3), computed by second-order Godunov scheme with $1/\ \ x = 5120$, at di↵erent times.

$T = 0.5$. We observe that the performance of L-AR schemes are comparable with those of the second-order Godunov scheme.In particular, we have that the L-NBee $\mathbf{L}^1$-error is the smallest for each level of refinement.

$1/\ \ x$	Godunov		Lax-F		L-NBee		L-UBee		Godunov2	
$1/\ \ x$	$\mathbf{L}^1$-err	$(\ \ x)$	$\mathbf{L}^1$-err	$(\ \ x)$	$\mathbf{L}^1$-err	$(\ \ x)$	$\mathbf{L}^1$-err	$(\ \ x)$	$\mathbf{L}^1$-err	$(\ \ x)$
80	2.7e-02	–	4.8e-02	–	5.2e-03	–	1.6e-02	–	8.5e-03	–
160	1.9e-02	0.53	3.4e-02	0.52	2.9e-03	0.83	5.8e-03	1.5	5.5e-03	0.64
320	1.3e-02	0.57	2.3e-02	0.55	1.2e-03	1.3	2.4e-03	1.2	3.0e-03	0.84
640	8.6e-03	0.58	1.6e-02	0.57	5.1e-04	1.3	1.4e-03	0.74	1.7e-03	0.87
1280	5.7e-03	0.59	1.0e-02	0.58	3.6e-04	0.51	9.4e-04	0.63	8.0e-04	1.0

Table 3: Test 2. Non-local multi-class LWR model. Initial condition (4.3), with decreasing kernel functions, final time $T = 0.5$. The reference solution is computed with $1/\ \ x = 5120$.

4.3 Test 3.Autonomous and human-driven mixed traffic

This test is intended to study how the presence of Connected Autonomous Vehicles (CAVs) impacts road traffic performances, as proposed in [10, Section 4.2]. Let us consider a circular road modeled by the space interval $[\ 1,\ 1]$ with periodic boundary conditions at $x = \pm 1$. All

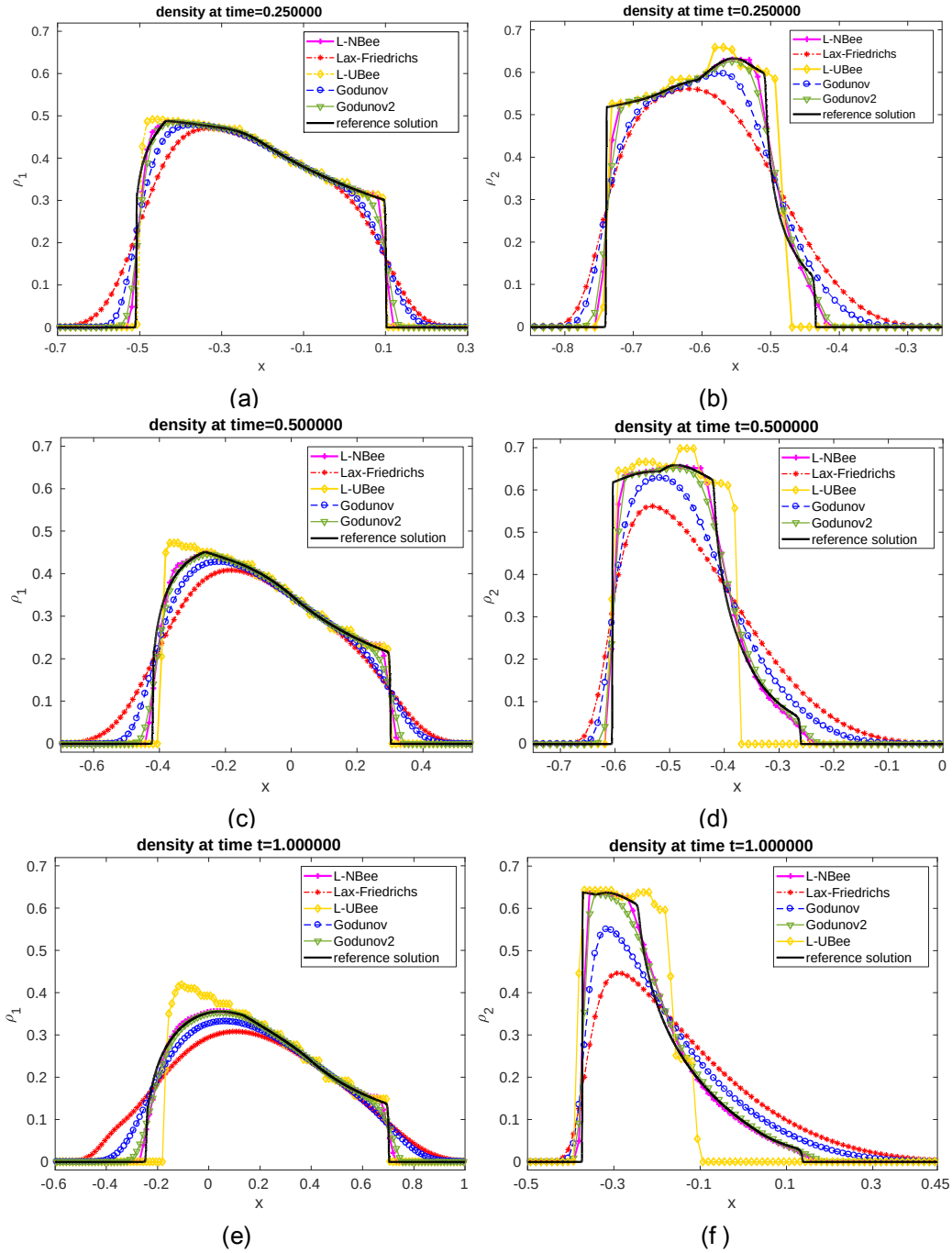


Figure 5: Test 2. (a)-(c)-(e) Profile of ρ_1 (trucks); (b)-(d)-(f) profile of ρ_2 (cars) computed with different numerical schemes at different times and $1/\Delta x = 80$.

the vehicles have the same maximal speed, but the interaction radius of CAVs is much grater than the one of human-driven cars. Moreover, we consider a constant convolution kernel to model the behaviour of CAVs, since the information they get about surrounding traffic is transmitted through wireless connections and it does not deteriorate with distance. We

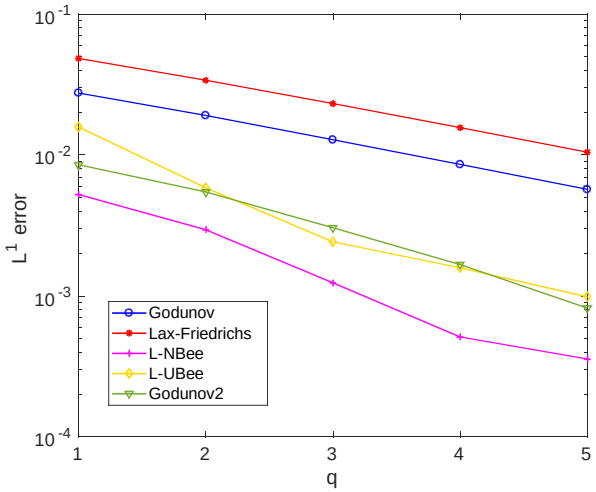


Figure 6: Test 2. Approximate total $\mathbf{L^1}$-error for di↵erent numerical schemes with the decreasing kernel functions $!_1(x) = !\ _2(x) = 2(\mathcal{H}_{1,2}\ \ x)/\mathcal{H}^2_{1,2}$, $\mathcal{H}_1 = 0.3$, $\mathcal{H}_2 = 0.1$

consider the following initial data and parameters

$$\rho_1(0, x) = (0.5 + 0.3\sin(5\pi x)), \qquad !\ _1(x) = \frac{1}{\mathcal{H}_1}, \quad \mathcal{H}_1 = 1.0, \quad v_1^{max} = 1,$$

$$\rho_2(0, x) = (1\ \ \ ) (0.5 + 0.3\sin(5\pi x)), \quad !\ _2(x) = \frac{2}{\mathcal{H}_2}\left(1 - \frac{x}{\mathcal{H}_2}\right), \quad \mathcal{H}_2 = 0.05, \quad v_2^{max} = 1, \tag{4.4}$$

where ρ_1 is the density of autonomous vehicles and ρ_2 the density of human-driven vehicles. The parameter $\ \ 2$ [0, 1] gives the penetration rate of autonomous vehicle. Figure 7a displays the reference solution of (1.1)-(4.4) with $\ \ = 0.9$, computed by the second-order Godunov scheme with $1/\ \ x = 10240$ at time $T = 1.5$. In Table 4 and Figure 7b we compute the approximate $\mathbf{L^1}$-error and the E.O.A at time $T = 1.5$.

In Figure 8, we display separately the two classes and we compare the approximate solutions computed by all the considered schemes with $1/\ \ x = 320$, and the reference solution. Again, the numerical solutions obtained using the L-AR schemes are more anti-di↵usive than those produced by first-order schemes. We observe a good behavior of the L-NBee scheme. On the other hand, the L-UBee scheme approaches the reference solution very well in the presence of shock-waves. On the other hand, the usual "stairs" appear in presence of rarefaction-waves.

We observe that the performances of L-NBee schemes are comparable with those of the second-order Godunov scheme. It is worth pointing out that despite the "staircaising" phenomenon the L-UBee $\mathbf{L^1}$-error is still smaller than the $\mathbf{L^1}$-error of the other first-order schemes.

5 Conclusions

We extended the L-AR schemes proposed in [5, 6] to the non-local multi-class traffic flow model proposed in [10]. We provided some properties of the L-AR scheme and we proved the convergence to weak solutions in the scalar case. The proposed numerical tests indicate that these schemes are competitive with the first and second-order schemes proposed in the

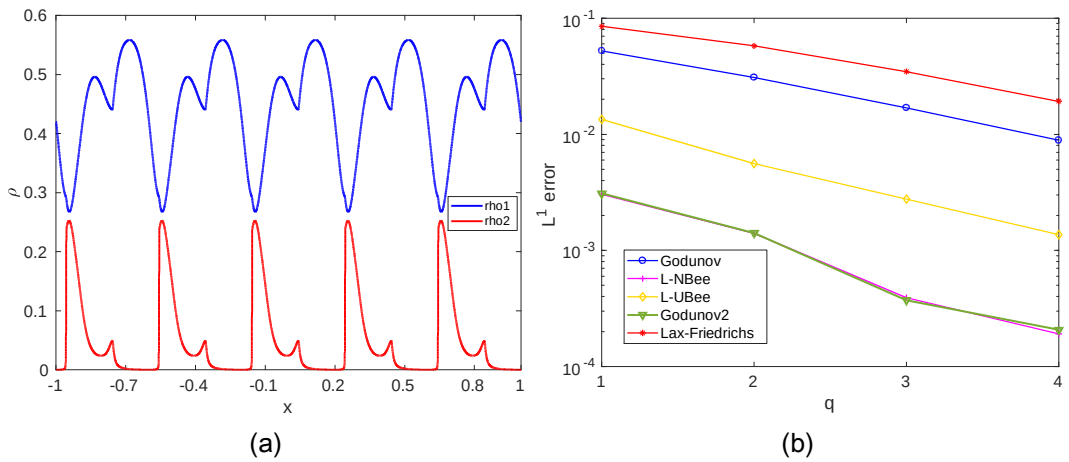


Figure 7: Test 3. (a) Reference solution of Test 3 at time T = 1.5 computed with 1/ x = 10240. (b) Approximate total $\mathbf{L}^1$-error for di↵erent numerical schemes.

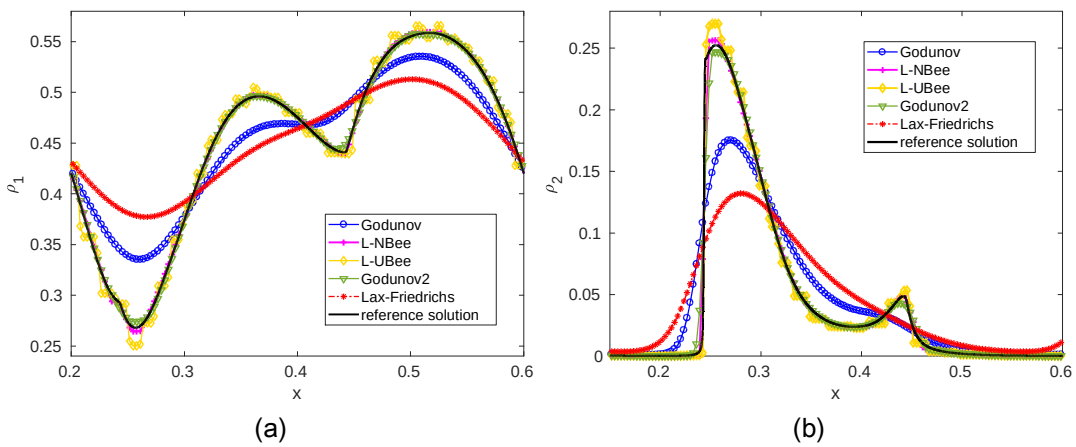


Figure 8: Test 3. (a) Profile of ρ_1; (b) profile of ρ_2, computed with di↵erent numerical schemes at time T = 1.5 and 1/ x = 320.

	Godunov		Lax-F		L-NBee		L-UBee		Godunov2	
$1/\Delta x$	$\mathbf{L^1}$-err	(Δx)	$\mathbf{L^1}$-err	(Δx)	$\mathbf{L^1}$-err	(Δx)	$\mathbf{L^1}$-err	(Δx)	$\mathbf{L^1}$-err	(Δx)
320	5.2e-02	–	8.5e-02	–	3.0e-03	–	1.3e-02	–	3.1e-03	–
640	3.1e-02	0.76	5.8e-02	0.56	1.4e-03	1.1	5.7e-03	1.3	1.4e-03	1.1
1280	1.7e-02	0.87	3.5e-02	0.73	3.9e-04	1.8	2.8e-03	1.0	3.7e-04	1.9
2560	8.9e-03	0.93	1.9e-02	0.85	1.9e-04	1.0	1.4e-03	1.0	2.0e-04	0.84

Table 4: Test 3. Initial condition (4.4), with different kernel functions, final time $T = 1.5$. The solutions are computed with $1/\Delta x = 160 \cdot 2^q$ for $q = 1, \ldots, 4$.

literature, in particular when more than one class are involved. If the initial datum has jump discontinuities, the performance of L-AR schemes are comparable with those of the second-order Godunov scheme. The improvement to higher order of accuracy is an involved issue that can be studied in the near future.

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