The Computation of Nonclassical Shock Waves with a Heterogeneous Multiscale Method

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THE COMPUTATION OF NONCLASSICAL SHOCK WAVES WITH A HETEROGENEOUS MULTISCALE METHOD

Frederike Kissling1 and Christian Rohde1

Abstract. We consider weak solutions of hyperbolic conservation laws as singular limits of solutions for associated complex regularized problems. We are interested in situations such that undercompressive (Non-Laxian) shock waves occur in the limit. In this setting one can view the conservation law as a macroscale formulation while the regularization can be understood as the microscale model.

With this point of view it appears natural to solve the macroscale model by a heterogeneous multiscale approach in the sense of E&Engquist[7]. We introduce a new mass-conserving numerical method based on this concept and test it on scalar model problems. This includes applications from phase transition theory as well as from two-phase flow in porous media.

1. Introduction. We consider weak solutions of the initial-value problem
\[
U_t + f(U)_x = 0 \quad \text{in} \quad \Omega_T := \mathbb{R} \times (0, T),
\]
\[
U(. , 0) = U_0 \quad \text{in} \quad \mathbb{R} \tag{2}
\]
for the scalar unknown
\[ U : \mathbb{R} \times (0, T) \to U \subseteq \mathbb{R} \]. Hereby
\[ U_0 : \mathbb{R} \to U \] stands for the initial datum and
\[ f \in C^3(U) \] is a nonlinear flux function: we assume that
\[ f'' \] vanishes only in finitely many points. The problem will be called the macroscale problem in what follows.

It is well-known that solutions of (1), (2) can have discontinuous shock wave solutions independently of the regularity of \( U_0 \). These solutions are not uniquely determined. For fluxes with \( f'' \neq 0 \) uniqueness can be enforced if only shock waves are allowed that satisfy the Lax condition. For the case of a discontinuous wave
\[
U(x, t) = U^- \text{ for } x - st < 0 \quad \text{and} \quad U(x, t) = U^+ \text{ for } x - st > 0,
\]
with states \( U^-, U^+ \in U, U^+ \neq U^- \), and shock speed
\[
s = s(U^-, U^+) = (f(U^+) - f(U^-))/(U^+ - U^-),
\]
the Lax condition reads as
\[
f'(U^+) < s < f'(U^-). \tag{5}
\]
Equivalently uniqueness is ensured if the entropy inequality
\[
\eta(U)_t + q(U)_x \leq 0 \tag{6}
\]
holds in the weak sense for at least one entropy pair \((\eta, q) \in (C^2(U))^2\), i.e. functions with \( \eta'' > 0 \) and \( q \) satisfying \( q' = \eta f' \).

If the flux \( f \) has at least one inflection point one can construct discontinuous waves \( U \) that violate exactly one of the conditions in (5) (with \( > \) instead of \( < \)) but still satisfy (6) for one entropy pair. Such waves connect states such that \( f'' \) changes sign and are called undercompressive or nonclassical. There are various applications where they have a crucial physical importance as phase boundaries, detonation waves, infiltration fronts, or thin-film precursors. Unique solvability of the initial value problem (1), (2) can again be guaranteed if only nonclassical waves are allowed that satisfy an additional algebraic condition of type
\[
\varphi(U^-, U^+) = 0, \tag{7}
\]
where the given function \( \varphi : U^2 \to \mathbb{R} \) is called kinetic function. A wellposedness theory based on kinetic functions has been developed by LeFloch ([16] and references therein). Numerical

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schemes for nonclassical solutions of (1), (2) controlled by kinetic functions have just recently been developed (e.g. in [5, 3]). In many cases an explicit kinetic relation that determines the physically relevant solution is either not known or cannot be written in closed algebraic form as in (7). Alternatively one can obtain these solutions as the singular limit

$$U^\lambda := \lim_{\varepsilon \to 0} u^{\varepsilon, \lambda}$$  \hspace{1cm} (8)

of (unique) solutions to a family of associated regularized problem (referred to as the microscale model)

$$u^{\varepsilon, \lambda} + f(u^{\varepsilon, \lambda})_x = R^\varepsilon[\lambda; u^{\varepsilon, \lambda}] \quad (\varepsilon > 0).$$  \hspace{1cm} (9)

Here $R^\varepsilon[\lambda; \cdot]$ stands for a higher-order integro-differential operator which satisfies $R^0[\lambda; \cdot] \equiv 0$ (formally). It is assumed to depend on some parameter $\lambda$ from a subset of $\mathbb{R}$. For many applications operators $R^\varepsilon[\lambda; \cdot]$ have been suggested and it has been proven that the limit $U^\lambda$ in (8) exists (at least for some subsequences) and contains nonclassical waves. We will discuss various examples in Sect. 2.

While the regularization approach is natural from the modeling viewpoint direct numerics for (9) with fixed $0 < \varepsilon << 1$ is computationally expensive: shock wave layers (scaling with $\varepsilon$) have to be resolved and in general time steps become extremely small. Therefore we propose here a new **Heterogeneous Multiscale Method (HMM)** in the sense of E&Engquist ([7] or [19] for a related application for combustion fronts). This approach gives us the solution of the macro model (1) via the limit (8). The micro model (9) is only solved explicitly in a small region around a nonclassical wave in (1) while in the remaining domain any standard solver can be used on a macroscale grid. Using the numerical fluxes developed in [3] we construct a solver that is conservative on the macroscale. Let us emphasize that we are **not** interested in the complete solution of the microscale model rather in the correct behaviour of the solution on the macroscale.

Let us give a short outline on the rest of the paper. After the short discussion of specific regularized models in Sect. 2 in Sect. 3 we present the concept of the new HMM-algorithm. In Sect. 4 we test the HMM-algorithm on Riemann problems for the examples from Sect. 2. In Sect. 5 we validate the HMM-algorithm and test it for a more general initial condition in Sect. 6.


In this section we recall three microscale models which can be written in the form (9). Two of them provide toy problems in the theory of phase transition dynamics, one describes two-phase flow in porous media with rate-dependent capillary pressure.

#### 2.1. Local Diffusive-Dispersive Regularization.

Compressible flow with liquid-vapour transitions can be governed by the Navier-Stokes-Korteweg equations [2], which contain second-order operators to model viscosity effects as well as third-order terms for the effects of capillary forces close to phase boundaries. A standard toy problem is given with the concave-convex flux $f(U) = U^3$ in (1), state space $\mathcal{U} = \mathbb{R}$ and the diffusive-dispersive operator

$$R^\varepsilon[\lambda; u] := \varepsilon u_{xx} - \lambda \varepsilon^2 u_{xxx}, \quad \lambda \in (0, \infty)$$  \hspace{1cm} (10)

in the microscale ansatz (9). For this model it has been proven in [9] that it allows for traveling wave solutions connecting states $U^-$ and $U^+$ such that (3) is undercompressive. Furthermore it is known that a subsequence of solutions $u^{\varepsilon, \lambda}$ converges to a weak solution $U^\lambda$ of (1) for $\varepsilon \to 0$. It is important to note that the limit depends on the parameter $\lambda$. This is also visible for the following special weak solution for Riemann initial data that can be realized as the limit in (8) (cf. [13])

$$U^\lambda(x, t) = \begin{cases} 
U^t & \text{for } x \leq s_1 t, \\
-u^m + \frac{1}{3} \sqrt{2 \lambda} & \text{for } s_1 t < x \leq s_2 t, \\
U^r & \text{for } s_2 t < x.
\end{cases} \quad (U^t > 0 > U^r > u^m)$$  \hspace{1cm} (11)

The shock speeds are given by $s_1 = s(U^t, u^m)$, $s_2 = s(u^m, U^r)$ with the notation from (4). Note that $U^r$ has to be sufficiently close to $u^m$, for details see [13]. We will use (11) as reference solution...
in Sect. 5. An approximation $u^{ε,λ}$ of the solution $U^λ$ with the slow undercompressive shock wave is displayed in Fig. 3 (left viewgraph).

2.2. Nonlocal Diffusive-Dispersive Regularization. In the theory of phase transitions in compressible media local differential operators as in (10) can be viewed as an approximation to more general nonlocal operators [18]. We consider as an instance the macroscale model (1) as in Sect. 2.1 but for the microscale model (9) a nonlocal regularization given through

$$R^ε[λ; u] := εu_{xx} + λ\left[\int_{-∞}^{∞} φ_ε(x − y)(u(y) − u(x)) dy\right], \quad λ ∈ (0, ∞).$$  (12)

Here we have $φ_ε(x) := \frac{1}{ε^2}φ\left(\frac{x}{ε}\right)$ where $φ$ is a nonnegative function which we will always choose as

$$φ(x) := \exp\left(\frac{1}{x^2 - 1}\right) / \int_{-1}^{1} \exp\left(\frac{1}{y^2 - 1}\right) dy \quad \text{for} \quad x ∈ (-1, 1),$$  (13)

and equal to 0 elsewhere. For (12) analogous results as for the local choice are known [17, 15]. We refer to Fig. 5, left viewgraph, for a typical solution with slow undercompressive wave and fast rarefaction wave. However, it is not known whether the weak solution $U^λ$ generated by (8) with operator (12) can also be characterized by a kinetic relation as in (7).

2.3. Two-Phase Flow Process with Rate-Dependent Capillary Pressure. In this part we consider two-phase flow in porous media. In the macroscale model (1) we choose the convex-concave Buckley-Leverett flux

$$f(U) = \frac{U^2}{U^2 + M(1 − U)^2} \quad \text{for} \quad U ∈ U := [0, 1],$$  (14)

where $U$ stands for the unknown water saturation and $M > 1$ describes the oil/water viscosity ratio. For the regularization we use

$$R^ε[λ; u] := εu_{xx} + λε^2u_{xxt} \quad (λ ∈ (0, ∞)).$$  (15)

This type of rate-dependent regularization has been suggested in [12]. It extends the static capillary pressure approach to model dynamic effects in the pressure difference between the two phases. For the regularized problem (9) with (15) it is again known that traveling wave solutions exist as profiles for undercompressive shock waves in the limit $ε → 0$ [6]. A typical solution for Riemann initial data with slow Laxian wave and fast undercompressive wave is illustrated in Fig. 7 (left viewgraph). An explicit kinetic relation as in (7) is not known up to our knowledge.

3. Computation of Nonclassical Waves with a Heterogeneous Multiscale Method. Before we present our numerical approach let us note that a solution which contains nonclassical waves cannot be approximated by a standard scheme as e.g. a monotone Finite Volume method. These are constructed to converge to the Kruzkov solution excluding such waves. Therefore special methods as in [5, 3] have been constructed. However these methods rely on an explicitly given kinetic relation while our approach utilizes the regularized model.

3.1. The Heterogeneous Multiscale Algorithm. We determine the macroscale approximation for (1) and the microscale approximation for (9) on a coarse and a fine space-time grid, respectively.

We denote symbols of the macroscale model by capital letters and the ones of the microscale model with lower case letters.

We use constant grid parameters $ΔX > 0$ and $Δx > 0$ for the space discretization both on the macroscale and on the microscale. Furthermore we set $X_j = jΔX, j ∈ Z$. The approximation of the macroscale solution at discrete times $0 =: T^0 < T^1 < T^2 < . . .$ is a function constant on each cell $C_j = [X_{j−1/2}, X_{j+1/2})$. For the initialization of the Finite Volume type scheme on the macroscale we define the sequence $(U^0_j)_{j ∈ Z}$ of averaged values by

$$U^0_j = \frac{1}{ΔX} \int_{C_j} U_0(x) dx.$$
Furthermore we collect all cell boundaries \(X_{j+1/2}\), say \(\Gamma \in \mathbb{N}\), such that \(f''(U_j^0)\) and \(f''(U_{j+1}^0)\) have different sign in the set \((X_{j+1/2})_{\gamma=1,...,r}\) of discrete front positions.

Algorithm 1 below determines for all times \(T^n\) the macroscale approximation \(U^n := (U^n_j)_{j \in \mathbb{Z}}\) and the approximate front positions \((X^n_{j})_{j=1,...,\Gamma}\) for nonclassical shock waves. The space-time domain for which the microscale problem has to be solved is defined by \(D^n_{\gamma} := (x^n_{\gamma,i}, x^n_{\gamma,r}) \times (T^n, T^n + t^n_{\gamma})\) with \(x^n_{\gamma,i} < X^n_{\gamma} < x^n_{\gamma,r}\) and \(t^n_{\gamma} > 0\). We now give the main HMM-algorithm in a rather general manner. The algorithm will be commented and detailed below. For the sake of simplicity we restrict ourselves to monotone increasing flux functions with at most one inflection point \(U\) with \(f''(U) > 0\).

**Algorithm 1.** Input: \(n = 0, \varepsilon, \lambda, (U^0_j)_{j \in \mathbb{Z}}, (X^0_j)_{\gamma=1,...,\Gamma}, (X_j)_{j \in \mathbb{Z}}\)

DO

1. **The reconstruction.** We reconstruct according to the location of the discrete front \(X^n_{\gamma}\), \(\gamma = 1, \ldots, \Gamma\),

   \[
   u^n_{\gamma}(x, T^n) := \begin{cases} 
   \frac{U^n_{\gamma,s} := u^n_{\gamma}^-}{X^n_{\gamma} \leq x, x \in D^n_{\gamma}}, \\
   \frac{U^n_{\gamma,s+2} := u^n_{\gamma}^+}{x < X^n_{\gamma}, x \in D^n_{\gamma}},
   \end{cases}
   \]

   with \(X^n_{\gamma} \in [X_\gamma, X_{\gamma+1}]\).

2. **Solving the microscale model.** For \(\gamma = 1, \ldots, \Gamma\) we solve (9) on \(D^n_{\gamma}\) with the initial data \(u^n_{\gamma}(. , T^n)\) and denote the solution by \(u^n_{\gamma} : D^n_{\gamma} \rightarrow \mathcal{U}\).

3. **Approximate middle state/data analysis.** For \(\gamma = 1, \ldots, \Gamma\) we determine a middle state \(u^n_{\gamma,m}\) between the two (approximate) shock waves in the solution \(u^n_{\gamma}\) of the microscale model.

4. **Computing the approximate front velocities.** For \(\gamma = 1, \ldots, \Gamma\) we compute the approximate velocity \(v^n_{\gamma}\) of the fronts by

   \[
   v^n_{\gamma} = (f(u^n_{\gamma}^-) - f(u^n_{\gamma}))/ (u^n_{\gamma}^- - u^n_{\gamma}^m).
   \]

5. **Computing the macroscale time step \(\Delta T^n\).** We compute the macroscale time step \(\Delta T^n\) such that

   \[
   \sup_{j \in \mathbb{Z}} \frac{\Delta T^n |f'(U^n_j)|}{\Delta X} < 1 \text{ and } \max_{\gamma=1,...,\Gamma} \frac{\Delta T^n |f'(U^n_\gamma)|}{\Delta X} < 1,
   \]

   and set \(T^{n+1} = T^n + \Delta T^n\).

6. **Moving the fronts.** With the front velocities from step 4 we obtain the new location of the fronts by

   \[
   X^{n+1}_{\gamma} = X^n_{\gamma} + v^n_{\gamma} \Delta T^n, \quad \gamma = 1, \ldots, \Gamma.
   \]

7. **Updating the macroscale variable \(U^n\).** We get the new macroscale variables \((U^0_{j+1})_{j \in \mathbb{Z}}\) by evolving (1) with a Finite Volume scheme

   \[
   U^n_{j+1} = U^n_j - \frac{\Delta T^n}{\Delta X} \left( f^n_{j+\frac{1}{2}} - f^n_{j-\frac{1}{2}} \right), \quad j \in \mathbb{Z},
   \]

   where \(f^n_{j+\frac{1}{2}}\) is a special numerical flux function (see (17) below).

8. \(n \rightarrow n + 1\)

UNTIL \(T^n \geq T\).

It is necessary to make more precise some of the steps in Algorithm 1 but let us make several notes in advance.

**Remark 1.** (i) The scaling parameter \(\varepsilon > 0\) in Algorithm 1 has the role of a discretization parameter: the smaller we choose it the more accurate is the prediction of the front velocities and thus the error of the macroscale approximation. Ideally this error is in the same range as the error induced by the grid parameter \(\Delta X\). We show the effect of the choice of \(\varepsilon\) in Sect. 5.1.
Figure 1. Macroscale computational domain $\Omega_T$ and microscale computational domain $D^n_\gamma$ for a propagating front $\gamma$ intersecting the macroscale time levels $T^n$ in $X^n$.

Figure 2. Approximate middle state $u^n_\gamma$ versus microscale time steps for a whole macroscale time step $\Delta T^n$.

(ii) In the reconstruction (step 1) we use for the right state $u^+_j$ the value $U^n_{s+2}$ (instead of the apparently more natural value $U^n_{s+1}$). This enables us to use a numerical flux (cf. (17) below) in step 7 which guarantees mass conservation (for a detailed explanation see discussion at the end of the section).

(iii) We have not said how to solve numerically the micro model in step 2. This depends in fact on the micro model under study. In Sect. 4 we present some adequate numerical schemes. Let us note that there are completely different possibilities as the numerical use of the Dafermos regularization [1].

(iv) The computational success of Algorithm 1 hinges on the size of the microscale domains $D^n_\gamma$. In particular the microscale time $t^n_\gamma$ has to be small to limit the expensive microscale computations. We illustrate this in Fig. 1. On the other hand we have to determine the middle state $u^n_\gamma$ from the computation. This is done by some data analysis which gives more stable results with an increasing number of microscale time steps. This is visualized in Fig. 2, where the middle state $u^n_\gamma$ is computed over a whole macroscale time step $\Delta T^n$.

(v) In Algorithm 1 we have supposed implicitly that the number of fronts remains constant, equal to $\Gamma$. If two nonclassical shocks join to one Algorithm 1 has to be modified in a straightforward way.

In Step 3 of Algorithm 1 we determine the middle state $u^n_\gamma$ through some data analysis. What is meant is that the results of the microscale computation at time $T^n + t^n_\gamma$ are scanned: the value with the smallest absolute discrete slope (and not close to $u^n_\pm_\gamma$) is chosen as $u^n_\gamma$. This step requires definitively some a-priori knowledge of the exact microscale solution.
It remains to define the numerical fluxes in step 7. Here we follow the construction in [3], precisely

\[
\Delta T^n f_{j+1/2} = \begin{cases} 
\min \left( \Delta T_{j+1/2}, \Delta T^n \right) f(U^n_{j+1}) + \max \left( \Delta T^n - \Delta T_{j+1/2}, 0 \right) f(U^n_{j}), & d^n_j \in [0, 1], \\
\Delta T^n f(U^n_j), & \text{else,}
\end{cases}
\]

with

\[
\Delta T_{j+1/2} = \frac{1 - d^n_j}{s(U^n_j, U^n_{j+1})} \Delta X, \quad U^n_{j+1} = U^n_{j-1} - \epsilon \gamma, \quad U^n_{j,r} = u^n_{j,r}.
\]

We used \(d^n_j = (U^n_{j,r} - U^n_j)/(U^n_{j,r} - U^n_{j+1})\) and \(s(U^n_j, U^n_{j,r})\) is given in (4).

This flux function has been introduced in [3] to compute solutions of (1) with nonclassical waves that satisfy a kinetic relation like in (7). However in the original flux the kinetic relation is just used to compute the middle states like our value \(u^n_m\). Due to this observation we can also make use of the approach and can benefit from its advantages: the Finite Volume like scheme in step 7 conserves mass, i.e. (cf. Property 2 in [3]):

\[
\sum_{j \in \mathbb{Z}} U^n_j = \sum_{j \in \mathbb{Z}} U^0_j \quad \forall n \in \mathbb{N}_0.
\]

Another important feature of the scheme is that by construction the nonclassical wave is only smeared out on three cells with exactly one cell average in this discrete layer (see also Remark 1(ii) on the choice of the reconstruction).


In this part we use Algorithm 1 from Sect. 3 for the model problems introduced in Sect. 2. For every example we give first the discretization of the microscale model, and then present some numerical examples.

For the sake of simplicity we use Finite Difference methods for the microscale problem. Let us stress that the heterogeneous approach can be improved considerably if more efficient schemes are used for the computation of the smooth solution of the microscale problem, as e.g. high-order Discontinuous Galerkin methods (cf. [11] for the application of this method on the model problems for phase transition).

4.1. Local Diffusive-Dispersive Regularization. For the discretization of (9) with (10) we use a fourth-order discretization for the flux function suggested by Hayes and LeFloch [10]. The complete scheme has then the form

\[
u^n_{i+1} = u^n_i - \frac{\Delta t}{\Delta x} (g^n_{i+1/2} - g^n_{i-1/2}) + \varepsilon \frac{\Delta t}{\Delta x^2} (u^n_{i+1} - 2u^n_i + u^n_{i-1}) \]

\[+ \lambda \varepsilon^2 \frac{\Delta t}{2 \Delta x^3} (u^n_{i+2} - 2u^n_{i+1} + 2u^n_{i-1} - u^n_{i-2}) \]

with the fourth-order flux

\[g_{i+1/2} = \frac{1}{12} \left( -(u^n_{i+2})^3 + 7(u^n_{i+1})^3 + 7(u^n_i)^3 - (u^n_{i-1})^3 \right).\]

In Fig. 3 we consider results of Algorithm 1 with the initial datum

\[U_0(x) = U^t = 4, \quad \text{for } x < 0 \quad \text{and} \quad U_0(x) = U^r = -2, \quad \text{for } x > 0, \quad (19)\]

and discretization data

\[\varepsilon = 10^{-5}, \quad \lambda = 4, \quad \Delta x = 4 \cdot 10^{-6}, \quad \Delta X = 0.01. \quad (20)\]

The exact limit solution is given by (11). The second and third viewgraph show for two different times the propagation of a nonclassical and a fast classical shock wave. As (constant) time step on the macroscale we get \(\Delta T = 2.0 \cdot 10^{-4}\). In the first viewgraph we present a result of a separate computation for the microscale model with the same data. One observes that due to the upscaling process in Algorithm 1 details of the microscale solution as the oscillations around the Laxian
wave are suppressed. Note that this calculation leads to the microscale time step $\Delta t = 3.2 \cdot 10^{-9}$.

In Fig. 4 we show analogous results with the same data as in (19), (20) but for the initial condition $U_0(x) = 4,0$ for $x < 0$ and $U_0(x) = -5,0$ for $x > 0$.

The solution consists now of a nonclassical shock wave and a rarefaction wave.

4.2. Nonlocal Diffusive-Dispersive Regularization. For the microscale model (9), (12) with $f(U) = U^3$ we use upwinding for the fluxes and discretize the integral with a one-point rule in the cell boundaries. This results in the scheme

$$u_{i}^{n+1} = u_{i}^{n} - \frac{\Delta t}{\Delta x} \left( f(u_{i}^{n}) - f(u_{i-1}^{n}) \right) + \varepsilon \frac{\Delta t}{(\Delta x)^2} \left( u_{i+1}^{n} - 2u_{i}^{n} + u_{i-1}^{n} \right)$$

$$+ \lambda \Delta t \sum_{k=-m}^{m} u_{i+k}^{n} G_{j}^{n} - \lambda \frac{\Delta t}{\Delta x} \left( u_{i}^{n} - u_{i-1}^{n} \right),$$

with the weights $G_{j}^{n} = \phi_{\varepsilon} \left( x_{j+\frac{1}{2}} \right) - \phi_{\varepsilon} \left( x_{j-\frac{1}{2}} \right)$ (cf. (13)) [14]. One can check that this choice implies that the scheme is in particular conservative.

In the same format as in Sect. 4.1 we display numerical results in Fig. 5 for initial datum

$$U_0(x) = 4,0 \text{ for } x < 0 \quad \text{and} \quad U_0(x) = -2,0 \text{ for } x > 0,$$

and discretization data

$$\varepsilon = 10^{-4}, \lambda = 15, \Delta x = 5,7 \cdot 10^{-6}, \Delta X = 0.01.$$  

(22)

For Fig. 6 the same parameters are used but for initial datum

$$U_0(x) = 4,0 \text{ for } x < 0 \quad \text{and} \quad U_0(x) = -5,0 \text{ for } x > 0.$$

(23)

For the nonlocal case no exact limit solutions like (11) are known but we observe qualitatively the same effects.
4.3. Two-Phase Flow Process with Rate-Dependent Capillary Pressure. To discretize the microscale model (9) with regularization (15) we rewrite it in the form

\[ w = u_t, \quad \left( I - \varepsilon^2 \lambda \frac{d^2}{dx^2} \right) w = \varepsilon u_{xx} + f(u)_x. \]  

(24)

A straightforward Finite Difference discretization of the second equation in (24) leads at each microscale time level to a linear system for the \( w_i^m \)'s. For a uniform grid the system matrix is a tridiagonal matrix for which the inverse can be determined with the so-called HMGIT algorithm from [8]. We use this algorithm and update the \( u_i^m \)'s by an explicit Euler step for the first equation in (24). Note that the flux from (14) is convex-concave, we have \( f''(\bar{U}) < 0 \) in the unique inflection point \( \bar{U} \in (0, 1) \). Therefore we have to exchange \( u^-_i \) with \( u^+_i \) in Step 4 of Algorithm 1.

In Figs. 7 and 8 we consider in the format as in Sect. 4.1 results of Algorithm 1 with the initial data

\[ U_0(x) = 0.8 \text{ for } x < 0 \quad \text{and} \quad U_0(x) = 0 \text{ for } x > 0 \text{ (Fig. 7)}, \]
\[ U_0(x) = 1.0 \text{ for } x < 0 \quad \text{and} \quad U_0(x) = 0 \text{ for } x > 0 \text{ (Fig. 8)}. \]

(25)

and discretization data

\[ \varepsilon = 10^{-5}, \lambda = 5, M = 2, \Delta x = 1, 5 \cdot 10^{-5}, \Delta X = 0, 02. \]

We now see that the nonclassical wave acts as the (fast) precursor front for the invading water phase. It is interesting to see that in the microscale computation in the mostleft viewgraph of Fig. 7 the data do not oscillate und do not take values outside \( U = [0, 1] \). Compare the result with the corresponding plots in Figs. 3, 5.

5. Performance of the Heterogeneous Multiscale Method. In this section we examine Algorithm 1 on aspects of convergence, mass conservation, and computational complexity.

We use the regularized problem (10) with the cubic flux \( f(U) = U^3 \) together with the initial datum from (19) as the test problem to validate our approach. In this case we can use (11) as the exact solution of the macroscale problem in the sense of the limit in (8).
5.1. Convergence and Conservation of Mass. We compute the solution for (1) with the regularization (10) with different macroscale grid sizes $\Delta X \in \{0.5, \ldots, 0.01\}$. The solutions are depicted in Fig. 9 with appropriate zooms into the shock waves. One can see well that for smaller macroscale grid size $\Delta X$ the shock waves are better resolved and we can observe convergence of the scheme.

In Fig. 10 (left) the error between the numerical solution for different macroscale grid sizes and the limit solution (11) is plotted in the $L^1$-norm. We observe that the $L^1$-error decreases followed by a saturation effect. This means that it doesn’t make sense anymore to refine the macroscale grid from $\Delta X = 0.1$ because the error in the $L^1$-norm is dominated by the error which arises by the calculation of the value $u_m^\gamma$ at the microscale, see Remark 1(i), and we should instead scale down the parameter $\varepsilon$. The slight increase of the error in Fig. 10 (left) is caused by the value lying in the layer of the nonclassical shock on the macroscale. Furthermore we have checked the conservation of mass of the scheme for the update on the macroscale of step 7 in Algorithm 1. The result confirms what we know from the construction of the numerical flux in (17), see Fig. 10 (right).

5.2. Computational Complexity for the Heterogeneous Multiscale Approach. As long as an explicit kinetic relation as in (7) is not known the limit solution $U^h$ from (8) can only be approximated by solving the microscale model (9) on the complete macroscale computational domain with some scheme (direct approach). Algorithm 1 using the same scheme on the microscale should be at least more efficient than the direct approach. In Fig. 11 we present a comparison of cpu-times which underlines the efficiency of Algorithm 1. For the grid sizes of the computation we use the same values as in Sect. 4.1. Note that the cpu-times for the microscale model on the macroscale domain are estimated values. They have been extrapolated from computations on a smaller space-time domain. One could argue that the use of an adaptive solver for the microscale computation could lead to a worse performance of Algorithm 1. But note that we can also use this solver within the

![Figure 7. Invading water front with overshoot for regularization (15).](image1)

![Figure 8. Invading water front with trailing wave for regularization (15).](image2)
Figure 9. Macroscale approximation with $\varepsilon = 10^{-5}$, $\lambda = 4$, zooms into the shock waves for different grid sizes.

Figure 10. Error in $L^1$-norm (left) and mass conservation (right).

Heterogeneous Multiscale Method. Moreover it has to be stressed that the time length of the microscale domain $D_n^\gamma$ is much smaller than the macroscale time step. This can not be achieved for the direct approach.

<table>
<thead>
<tr>
<th>$T$</th>
<th>macroscopic model</th>
<th>microscopic model</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$N_T$ $\text{cpu-time}$</td>
<td>$n_t$ $\text{cpu-time}$</td>
</tr>
<tr>
<td>$1.98 \cdot 10^{-4}$</td>
<td>1 $7,52$ sec.</td>
<td>$6.19 \cdot 10^4$ $9.2 \cdot 10^5$ sec. $\approx 11$ days</td>
</tr>
<tr>
<td>0, 1</td>
<td>505 $20,41$ sec.</td>
<td>$3.13 \cdot 10^5$ $4.6 \cdot 10^7$ sec. $\approx 15$ years</td>
</tr>
<tr>
<td>0, 5</td>
<td>2526 $22,76$ sec.</td>
<td>$1.56 \cdot 10^8$ $2.3 \cdot 10^9$ sec. $\approx 73$ years</td>
</tr>
</tbody>
</table>

| $N_X = 1650$, $n_x = 400$ | $n_x = 4\,125\,000$ |

Figure 11. Cpu-time for Algorithm 1 versus cpu-time for direct approach. $N_T$ ($n_T$) stands for the number of macroscale (microscale) time steps, $N_x$ ($n_x$) for the number of macroscale (microscale) cells.

6. **Heterogeneous Multiscale Method for Non-Riemann Initial Data.** Up to now we only have considered Riemann problems. This was helpful since we knew the exact solution in some cases and could validate Algorithm 1. Finally we consider a problem for (10) with more general initial data, namely

$$U_0(x) = 4 \sin(2\pi x) \text{ for } x \in (0, 1) \text{ and } U_0 \equiv 0 \text{ elsewhere}. \quad (26)$$

All other parameters are as in Sect. 4.1.

We can observe from Fig. 12 that the approximation firstly rears up and forms a front from the continuous initial function. After this the solution develops a nonclassical shock wave, forms a nonclassical shock-rarefaction pattern, a classical shock and again a rarefaction and classical shock wave.
Figure 12. Solution of the Cauchy problem for (26) at different times.

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