Well-balanced finite volume schemes of arbitrary order of accuracy for shallow water flows

Sebastian Noelle a, Normann Pankratz a, Gabriella Puppo b, Jostein R. Natvig c

a Institut für Geometrie und Praktische Mathematik, RWTH Aachen, Templergraben 55, 52056 Aachen, Germany
b Dipartimento di Matematica, Politecnico di Torino, Corso Duca degli Abruzzi 24, 10129 Torino, Italy
c SINTEF ICT, Forskningsveien 1, N0314 Oslo, Norway

Received 18 March 2005; received in revised form 18 July 2005; accepted 19 August 2005
Available online 14 October 2005

Abstract

Many geophysical flows are merely perturbations of some fundamental equilibrium state. If a numerical scheme shall capture such flows efficiently, it should be able to preserve the unperturbed equilibrium state at the discrete level. Here, we present a class of schemes of any desired order of accuracy which preserve the lake at rest perfectly. These schemes should have an impact for studying important classes of lake and ocean flows.

© 2005 Elsevier Inc. All rights reserved.

Keywords: Hyperbolic conservation laws; Source terms; Shallow water equations; Well-balanced schemes; Finite volume schemes; WENO reconstruction

1. Introduction

In this introduction, we present some of the key ideas and ingredients of the subsequent sections. We begin with a brief review of the shallow water equations and their equilibrium states, in particular the lake at rest. Then we show an example of a numerical storm produced by a scheme which is not in discrete equilibrium. Next we review the key ingredient of several of the recent well-balanced schemes, and give some related references. We close with a preview of our new high order well-balanced schemes.
1.1. Shallow water equations

Many geophysical flows are modeled by variants of the shallow water equations. In their simplest form these equations read

\[ \begin{align*}
    h_t + (hu)_x &= 0, \\
    (hu)_t + \left( hu^2 + \frac{1}{2} gh^2 \right)_x &= -ghz.
\end{align*} \tag{1} \]

Here \( z(x) \) defines the bottom-topography, \( h(x,t) \) denotes the water height above the bottom, and \( u(x,t) \) is the horizontal component of the water velocity at position \( x \) at time \( t \). The gravity constant is denoted by \( g \). In (1), we have neglected two-dimensional effects, bottom friction, Coriolis forces arising in a rotational frame, wind forces, and, of course, vertical variations of the velocity field. The proper treatment of Coriolis forces is considered, for instance, in [5], see also [6] for a general topography. For an example of more complete shallow water equations which are used in coastal engineering, we refer to Gjevik et al. [9].

1.2. Equilibrium states

In spite of all of these simplifications, Eq. (1) still contain the most fundamental balances of shallow water flows. The convective part on the left-hand-side (LHS) is a hyperbolic system of conservation laws similar to that of compressible fluid flows, and the source term on the right-hand-side (RHS) is due to gravitational acceleration. Let us look at the equilibrium, or stationary, states. They are given by

\[ \begin{align*}
    hu &\equiv \text{const.} \quad \text{and} \quad \frac{1}{2} u^2 + gh \equiv \text{const.,}
\end{align*} \]

where

\[ H := h + z \]

is the water level. In this paper, we are particularly interested in the lake at rest, given by

Fig. 1. Cross section of lake Rursee: bottom topography and quiet water level. 296 cells.
Such a situation is shown in Fig. 1 for a cross-section of lake Rursee near Aachen. Let us pause for a moment and look at this balance once more. From (1) and the assumptions of stationary flow with vanishing velocity we have

\[ 0 = \left( \frac{gh^2}{2} \right)_x + ghz_x, \quad (2) \]

which is called hydrostatic balance. The first term is the hydrostatic pressure, which models the tendency of a column of water to collapse vertically and at the same time expand laterally under the influence of gravity. The second term is the gravitational acceleration down an inclined bottom \( z \). Now use the chain rule of differentiation and divide by \( h \) to obtain

\[ 0 = g(h + z)_x = gH_x. \]

Thus, we see that the effective acceleration can be interpreted as gravitational acceleration down a non-flat water level \( H \).

1.3. Numerical storms

If a numerical scheme does not preserve the fundamental balance (2) at the discrete level, this may result in spurious oscillations, or numerical storms, as seen in Fig. 2. The figure shows a cross-section of lake Rursee near Aachen, and the water should remain at rest as in Fig. 1. Thus, all waves in Fig. 2 are pure numerical artifacts. Some of them are more than a meter high, especially near the edge of the lake. The computation is run with a standard finite volume scheme, a naive treatment of the source term, and 296 spatial grid cells. Clearly, this scheme on the current grid would not be able to resolve waves which are of the order of magnitude of the numerical perturbations. One would therefore have to run such a scheme with a much finer grid, which would make the computation rather costly.

1.4. Well-balanced schemes

The results in Fig. 3, which reproduce the lake at rest perfectly, are obtained with a so-called well-balanced scheme, using the same number of spatial grid cells and time steps. Let us briefly sketch the main ingredient of the discrete balance which makes the scheme successful. The main difficulty for the schemes is to preserve the balance of hydrostatic pressure and gravitational acceleration (hydrostatic balance). Given a cell \([x_L, x_R]\), let

\[ h_L = h(x_L), \quad h_R = h(x_R). \quad (3) \]

Before we proceed, let us note that this notation hides whether these values should represent the left or right limits of piecewise smooth reconstructions at the interface, or the value chosen by an approximate Riemann solver. We postpone this crucial question to Section 2 and continue to outline the main ingredient of well-balancing.

A conservative finite volume discretization of the hydrostatic pressure would then be

\[ \left( \frac{gh^2}{2} \right)_x \approx g \frac{h_R^2 - h_L^2}{\Delta x}. \quad (4) \]

We will now show that this already implies a canonical well-balanced discretization of the source term. Indeed, suppose that the source term is discretized as

\[ ghz_x \approx ghDz, \]

where \( \hat{h} \approx h \) and \( Dz \approx z_x \). Now we suppose that \( u \equiv 0 \) and \( H \equiv \text{const.} \), and we want to enforce the discrete hydrostatic balance

\[ 0 = \frac{g}{\Delta x} \left( \frac{h_R^2 - h_L^2}{2} \right) + ghDz. \quad (5) \]
From (5) we obtain

\[
\hat{h}D_x = - \frac{1}{2} \frac{h_L^2 - h_R^2}{\Delta x} = - \frac{h_L + h_R}{2} \frac{h_R - h_L}{\Delta x} = - \frac{h_L + h_R}{2} \frac{(H_R - z_R) - (H_L - z_L)}{\Delta x} = \frac{h_L + h_R}{2} \frac{z_R - z_L}{\Delta x}.
\]
This discretization of the source term was first proposed by Bermudez and Vazquez [3], and it is also the essential ingredient of the recent well-balanced schemes of Jin [16], Kurganov and Levy [19] and Audusse et al. [1]. Closely related schemes usually try to discretize the derivative of the convective flux and the source term by one and the same finite difference or finite volume operator, see [2,4,29]. Greenberg, LeRoux and coworkers developed schemes based on the solution of the non-homogeneous Riemann-problem, see [13,11,10]. We

Fig. 3. Well-balanced computation of quiet lake Rursee: water level (top) and momentum (bottom) at time $T = 0.2$ (71 time steps). Note that the scale of the momentum axis is $10^{-9}$. 

This discretization of the source term was first proposed by Bermudez and Vazquez [3], and it is also the essential ingredient of the recent well-balanced schemes of Jin [16], Kurganov and Levy [19] and Audusse et al. [1]. Closely related schemes usually try to discretize the derivative of the convective flux and the source term by one and the same finite difference or finite volume operator, see [2,4,29]. Greenberg, LeRoux and coworkers developed schemes based on the solution of the non-homogeneous Riemann-problem, see [13,11,10]. We
would also like to mention the finite volume Roe schemes of Gallouët and coworkers [8] and the Norwegian front tracking approach [14]. This list is by far not exhaustive, and we refer to the papers mentioned above for further references.

We now comment on the ambiguities hidden in (3). The well-balancing in Eqs. (5) and (6) will only work if some continuity property holds at the equilibrium state. Our paper is based on the recent work of Audusse et al. [1], where such a continuity is guaranteed by a hydrostatic reconstruction, plus an additional correction of the source term. Their first and second-order schemes preserve positivity of water height and the lake at rest. The first-order scheme also satisfies a discrete entropy inequality at discontinuities.

In the present paper, we are interested in very high order accurate well-balanced schemes. These more sophisticated schemes are needed if, for instance, one wants to track small waves over long periods of time. Well-balanced finite difference schemes of high order of accuracy were developed by Vukovic and Sopta 2002 [28] and Xing and Shu 2004 [29]. This approach is extended to more general balance laws in [30] for finite differences and [31] for finite volumes. Here, we extend the well-balanced finite volume schemes of Audusse et al. [1] to any desired order of accuracy.

We would like to stress that the approach to achieve high order is rather different in the case of finite difference and finite volume schemes. In the former case, Xing and Shu rewrite the balance law in such a way that the fluxes and source terms can be treated by one and the same difference operator. In the present paper, we observe that the well-balanced quadrature (6) maintains all its desirable properties under numerical extrapolation. Together with standard high order reconstructions and the hydrostatic correction this leads immediately to the desired very high order accurate well-balanced finite volume schemes.

We conjecture that this technique can be applied to many, if not all, of the second-order well-balanced schemes based on (6), once an appropriate continuity condition at the cell interface is enforced. We refer the reader to the recent preprints [30,31] for further examples of interface continuity conditions.

Numerical experiments show the expected convergence rates for a fourth/fifth-order version of our new scheme, and excellent resolution of discontinuities and very small disturbances.

2. High order well-balanced schemes

In this section, we introduce our extrapolation technique. Even though we believe that the approach is rather general, we develop it only for the scheme of Audusse et al. [1], and remark on the more general features as we go along. We first summarize their second-order well-balanced scheme. Then, for any order of accuracy, we introduce our new treatment of the source term. We close the section with a summary of the new algorithm. Details of the WENO reconstruction are given in Appendix A.

2.1. Review of second-order well-balancing via hydrostatic reconstruction

Let \( U := (h, hu)^T \) be the vector of conservative variables. First we formulate a semidiscrete finite volume scheme for the cell averages,

\[
U_i(t) := \frac{1}{\Delta x_i} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} U(x, t) \, dx,
\]

with \( \Delta x_i := x_{i+\frac{1}{2}} - x_{i-\frac{1}{2}} \). Based on these cell averages, one defines a piecewise polynomial reconstruction, which will in general be discontinuous at the interfaces \( x_{i+\frac{1}{2}} \). Oscillations will be suppressed with limiters. Audusse et al. use a linear reconstruction with minmod limiter, which leads to a second-order scheme. Within cell \( i \), the left and right values of each component at position \( x_{i-\frac{1}{2}} + 0 \), respectively, \( x_{i+\frac{1}{2}} - 0 \) are denoted by \( (. )_{i, l} \) and \( (. )_{i, r} \).

Audusse et al. reconstruct \( h \), \( H \), and \( u \). From this, the bottom topography is computed as \( z = H - h \). This leaves the lake at rest unperturbed, but it leads to a discontinuous bottom. To get a stable and well-balanced scheme (compare the discussion in Section 1), the following hydrostatic reconstruction is introduced:
\[ z^*_{i+\frac{1}{2}} := \max(z_{i,r}, z_{i+1,l}), \]
\[ h^*_i := \max(0, h_{i,r} + z_{i,r} - z^*_{i+\frac{1}{2}}), \]
\[ h^*_{i+1,l} := \max(0, h_{i+1,l} + z_{i+1,l} - z^*_{i+\frac{1}{2}}). \]

Eq. (7) recovers a continuous bottom locally at each interface. The new local values \( h^*_i \) and \( h^*_{i+1,l} \) of the height ensure that at steady state, i.e., for \( h_{i,r} + z_{i,r} = h_{i+1,l} + z_{i+1,l}, \) \( h^* \) remains continuous across each cell (compare (3)). On the other hand, one has thereby modified the fluxes at the interface, and we will have to correct this below (see Eqs. (11) and (12)). Together the hydrostatic reconstruction and the correction of the interface fluxes will permit to balance the scheme for any numerical flux consistent with the homogeneous shallow water equations. The values for \( h^* \) are used to construct auxiliary values \( U^*_{i,r} \) and \( U^*_{i+1,l} \) which will enter an approximate Riemann solver (compare (4)):

\[
U^*_{i,r} := \left( \frac{h^*_i}{h^*_r} ; \frac{h^*_i}{h^*_r} U_{i,r} \right),
\]
\[
U^*_{i+1,l} := \left( \frac{h^*_{i+1,l}}{h^*_l} ; \frac{h^*_{i+1,l}}{h^*_l} U_{i+1,l} \right).
\]

Note that at the interface \( x_{i+\frac{1}{2}} \), we have two different reconstructions, namely \( x_{i+\frac{1}{2}}^- \) on the left and \( x_{i+\frac{1}{2}}^+ \) on the right side. As in [1], the semidiscrete finite volume scheme reads

\[
\Delta x_i \frac{d}{dt} U_i(t) + \mathcal{F}^r(U_i, U_{i+1}, z_{i,r}; z_{i+1,l}) - \mathcal{F}^l(U_{i-1}, U_i, z_{i-1,r}; z_{i,l}) = S_i^{(ij)}.
\]

It remains to specify the numerical fluxes and the source term. The fluxes are given by

\[
\mathcal{F}^r(U_i, U_{i+1}, z_{i,r}; z_{i+1,l}) := F(U^*_{i,r}, U^*_{i+1,l}) + \left( \frac{\xi}{2} h^2_{i,r} - \frac{\xi}{2} (h^*_{i,r})^2 \right), \tag{11}
\]
\[
\mathcal{F}^l(U_{i-1}, U_i, z_{i-1,r}; z_{i,l}) := F(U^*_{i-1,r}, U^*_{i,l}) + \left( \frac{\xi}{2} h^2_{i-1,l} - \frac{\xi}{2} (h^*_{i-1,l})^2 \right), \tag{12}
\]

where \( F \) is a conservative numerical flux consistent with the homogeneous shallow water equations and the second term on the RHS is the correction to the interface fluxes due to the modification in the water height introduced by the hydrostatic reconstruction. If the hydrostatic reconstruction leaves the water height untouched, then \( h^*_i = h_{i,r}, \) \( h^*_{i+1,l} = h_{i,l} \) and no correction is required. Because of their robustness, the local Lax-Friedrichs, Harten–Lax-vanLeer or kinetic solvers are used in [1]. In the present paper, we use the local Lax-Friedrichs flux for all our examples.

Let us consider the steady state of the lake at rest. Recall from (7) to (9) that in this case the hydrostatic reconstruction \( U^* \) is continuous across the interfaces and \( u = 0 \). Then any consistent numerical flux \( F(U^*_{i,r}, U^*_{i+1,l}) \) will reduce to the hydrostatic pressure term,

\[
F(U^*_{i,r}, U^*_{i+1,l}) = \left( \frac{0}{2} (h^*_{i,r})^2 \right), \tag{13}
\]
\[
F(U^*_{i-1,r}, U^*_{i,l}) = \left( \frac{0}{2} (h^*_{i-1,l})^2 \right). \tag{14}
\]

Here, \( h^*_i \) and \( h^*_{i+1,l} \) correspond to \( h_R \) and \( h_L \) in Eq. (4). Thus, the second term on the RHS of (11) and (12) cancels the difference of the hydrostatic pressures based on the piecewise polynomial reconstruction \( h_{i,r} \) and \( h_{i,l} \) at the interior of the cell and the hydrostatic reconstruction \( h^*_i \) and \( h^*_{i+1,l} \) at the interfaces \( x_{i+\frac{1}{2}} \) and \( x_{i-\frac{1}{2}}. \)
The index \( j = 1, 2 \) represents the order of the numerical source term \( S_i^{(j)} \). It is given by

\[
S_i^{(1)} := \begin{pmatrix} 0 \\ 0 \end{pmatrix}
\]

for \( j = 1 \). For a first order reconstruction \( h_{ix} = h_i = h_{i,t} \). Substituting this information in (11) and (12), and using (13) and (14), one immediately obtains that at steady state \( U_i \) remains constant, so that the scheme is well balanced. For \( j = 2 \):

\[
S_i^{(2)} := \left( 0 \ g \left[ \frac{h_i + h_{ix}}{2} - \frac{h_i + h_{i,t}}{2} (z_{i,1} - z_{i,r}) \right] \right)
\]

Note that this corresponds to the source term discretization (6), and below we review the argument that shows how this leads to a well-balanced scheme for the lake at rest. Together with a second-order Runge–Kutta time discretization the fully discrete second-order well-balanced scheme of Audusse et al. is now complete. With constant reconstruction and without the Runge–Kutta procedure you get the associated first-order scheme

[1]. Audusse et al. could show for their scheme that it preserves the nonnegativity of the water height \( h_i(t) \), it preserves the steady state of the lake at rest, is consistent with the shallow water system and their first-order scheme does also satisfy an in-cell entropy inequality.

2.2. Second-order well-balancing

To motivate the subsequent development of a well-balanced scheme of very high accuracy, we need to review the well-balanced property of Audusse et al.’s second-order semidiscrete scheme (10). Suppose that \( H = h + z \) is constant at time \( t \), and \( u \equiv 0 \). Since \( H_{ix} = H_{i+1,x} \):

\[
h_{ix}^* = \max(0, H_{ix} - \max(z_{i,x}, z_{i+1,1})) = \max(0, H_{i+1,1} - \max(z_{i,x}, z_{i+1,1})) = h_{i+1,1}^*
\]

and since \( u_{ix} = u_{i+1,1} = 0 \), we also have

\[
U_{ix}^* = U_{i+1,1}^*
\]

Because now the values \( U_{ix}^* \) and \( U_{i+1,1}^* \) are equal, and \( u = 0 \), the numerical fluxes \( F(U_{ix}^*, U_{i+1,1}^*) \) and \( F(U_{i+1,1}^*, U_{i+1,1}^*) \) reduce to the hydrostatic pressure. Substituting this information in (12) and (11), we find

\[
\mathcal{F}^+(U_{i+1,1}, U_{i+1,1}, z_{i,x}, z_{i+1,1}) = \left( \frac{0}{h_{ix}^*} \right) \quad \text{and} \quad \mathcal{F}^+(U_{i-1,1}, U_{i-1,1}, z_{i,x}, z_{i+1,1}) = \left( \frac{0}{h_{ix}^*} \right).
\]

This, together with the definitions (10)–(15) of the semidiscrete scheme implies

\[
\frac{d}{dt} h_i(t) = 0
\]

and

\[
\frac{d}{dt} m_i = -\frac{1}{\Delta x} \left[ g \frac{h_i^2}{2} - g \frac{h_i}{2} - g \frac{h_{i,t} + h_{i,x} (z_{i,1} - z_{i,r})}{2} \right]
\]

\[
= -\frac{1}{\Delta x} \left[ g \frac{h_{ix}^2}{2} - g \frac{h_{i,t} + h_{i,x}}{2} ((H_{i,1} - h_{i,1}) - (H_{i,r} - h_{i,r})) \right] = -\frac{1}{\Delta x} \left[ -g \frac{(h_{i,t} + h_{i,x})}{2} (H_{i,1} - H_{i,r}) \right].
\]

Because of \( H_{i,1} = H_{i,r} = H \),

\[
\frac{d}{dt} m_i(t) = 0,
\]

so

\[
\frac{d}{dt} U_i(t) = 0.
\]

Therefore, the second-order semidiscrete scheme preserves the stationary state of the lake at rest.
2.3. Higher order well-balancing

The project of the present paper is to show how to extend the first and second-order accurate well-balanced schemes to any desired order of accuracy. Most ingredients which we use are well-established in the literature: high order WENO spatial reconstructions [15], high order Runge–Kutta time discretizations [23,25], and appropriate quadrature rules for the initial data. But there is one essential difficulty to be solved: we need to find a quadrature rule for the source term which is both accurate and well-balanced. The remainder of this section is devoted to the solution of this question.

As before, let $U_{i,x}$, $U_{i+1,1}$ be the left and right values of a piecewise polynomial reconstruction at interface $x_{i+\frac{1}{2}}$. Of course, this time we work with polynomials of any desired order of accuracy. Define the hydrostatic reconstruction $h_{i+\frac{1}{2}}$ by (8) and (9) as before, and set

$$U_{i,x}^r := \left( \frac{h_{i,x}}{m_{i,x}} \right), \quad U_{i+1,1}^r := \left( \frac{h_{i+1,1}}{m_{i+1,1}} \right).$$

Note that to achieve orders higher than two, it is convenient to reconstruct in the conservative variable $m$ (which is computed with full accuracy by the finite volume scheme) instead of the primitive variable $u$, which is only derived from the conservative ones. We define the left and right interface fluxes $F^l$ and $F^r$ as before, see (11) and (12). It remains to define a high order, well-balanced numerical quadrature of the source term $S_N$.

$$S := - \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} ghz_j \, dx.$$ 

The main observation of this paper is that this can be done by numerical extrapolation. To do so, we subdivide each cell into $N$ subcells and apply the quadrature (6) to all subcells. This gives the quadrature $S^N$,

$$S^N := g \sum_{j=1}^{N} \frac{h_{j-1} + h_j}{2} (z_{j-1} - z_j) \approx S,$$

where $z_j = z(x_{i-\frac{1}{2}} + j\Delta x/N)$, etc., are local values of the reconstruction at the interfaces of the subcells. In the situation of the lake at rest, where

$$z_{j-1} - z_j = h_j - h_{j-1},$$

the source term reduces to

$$S^N = - \frac{g}{2} \sum_{j=1}^{N} \frac{h_{j-1} + h_j}{2} (h_j - h_{j-1}) = - \frac{g}{2} (h_N^2 - h_0^2) = - \frac{g}{2} (h_{i,x}^2 - h_{i+1,1}^2).$$

By the same arguments as for the second-order case this is well-balanced, but it is still only second-order accurate (see Table 2).

To get higher orders of accuracy we use numerical extrapolation (see e.g. the textbook of Deuflhard and Bornemann [7]). Note that the quadrature (6) is symmetric and second-order accurate. Therefore, from Theorem 4.39 of [7], there exists an asymptotic expansion of the form

$$S^N = S + c_1 \left( \frac{\Delta x}{N} \right)^2 + c_2 \left( \frac{\Delta x}{N} \right)^4 + \cdots$$

(17)

The $S^N$ can be combined for different values of $N$ to compute $S$ with any order of accuracy. For example, to get a source term of order four, simply use

$$4S^2 - S^1 = S + \tilde{c}_2(\Delta x)^4 + \cdots$$

Therefore, we define $S_i^{(4)}$ by

$$S_i^{(4)} := \frac{4}{3} \left( \frac{1}{2} (h_{i,x} + h_{i+1,1})(z_{i,x} - z_{i+1,1}) + \frac{1}{2} (h_{i+1,x} + h_{i+1,1})(z_{i+1,x} - z_{i+1,1}) - \frac{1}{2} (h_{i,x} + h_{i+1,1})(z_{i,x} - z_{i+1,1}) \right).$$

(18)
Thus for the lake at rest:

\[ S^{(4)}_l = -\frac{g}{2} (h_{l,t}^2 - h_{l,i}^2), \]

which leads to a well balanced scheme.

**Remark 1.** Compared with \( S^1 \), the computation of \( S^2 \) uses only one additional reconstruction point per cell, namely the cell center. Thus we can compute \( S \) to fourth-order accuracy using three points per cell, which is analogous to Simpsons rule (which may be obtained by extrapolating the trapezoidal rule). Note that we could not use Simpsons rule directly, because this would not give a well-balanced scheme.

**Remark 2.** Any scheme that is well balanced with the source term \( (6) \) will also be well balanced with the fourth-order source term \( (18) \). Besides our quadrature, one only has to add the correct interface fluxes which couple the reconstruction in the interior of the cell, used for the quadrature, with the hydrostatic reconstruction used by the numerical fluxes.

We summarize our high order well-balanced finite volume schemes in the following theorem:

**Theorem 3.** Consider the fully discrete finite volume scheme given by a \( j \)th order Runge–Kutta time discretization of the semidiscrete scheme \( (10) \), with \( k \)th order spatial reconstruction, hydrostatic reconstruction \( (8), (9) \) and \( (16) \), interface fluxes defined by \( (11), (12) \), and source term \( S^{(l)} \) given by an \( l \)th order extrapolation of \( (17) \). Then

(i) the scheme preserves the stationary state of the lake at rest
(ii) the scheme is consistent of order \( p = \min\{j, k, l\} \) with the shallow water equation \( (1) \).

**Proof.** We have already proved the well-balanced property. The proof of consistency follows closely that of Theorem 3.1 of [1], q.e.d. □

**Definition 4.** In the following, we will denote our well balanced WENO schemes with the triplet \( (j, k, l) \), where \( j, k \) and \( l \) denote, respectively, the accuracy in time of the Runge–Kutta integrator, the accuracy in space of the WENO reconstruction and the accuracy of the quadrature rule \( (18) \).

In the numerical experiments in Section 3, we use a scheme of orders \( (4, 5, 4) \) where the classical 4th order Runge–Kutta scheme is used for time integration, a 5th order WENO reconstruction is used in space (see Appendix A) and the 4th order extrapolation \( (18) \) for the source term. According to Theorem 3, this scheme is formally 4th order accurate. Surprisingly, in experiments with smooth solutions, it clearly gives 5th order convergence, see Table 1 below. Note that we could also have used Shu’s TVD Runge–Kutta time discretizations [23] or the recent SSPRK schemes [12,26].

<table>
<thead>
<tr>
<th>Number of cells</th>
<th>( h ) \text{error}</th>
<th>Order</th>
<th>( hu ) \text{error}</th>
<th>Order</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>( L^1 )</td>
<td></td>
<td>( L^1 )</td>
<td></td>
</tr>
<tr>
<td>Convergence table with fourth-order source term</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>25</td>
<td>1.13E–02</td>
<td></td>
<td>8.22E–02</td>
<td></td>
</tr>
<tr>
<td>50</td>
<td>1.84E–03</td>
<td>2.61</td>
<td>1.71E–02</td>
<td>2.27</td>
</tr>
<tr>
<td>100</td>
<td>2.83E–04</td>
<td>2.70</td>
<td>2.48E–03</td>
<td>2.78</td>
</tr>
<tr>
<td>200</td>
<td>2.07E–05</td>
<td>3.77</td>
<td>1.77E–04</td>
<td>3.81</td>
</tr>
<tr>
<td>400</td>
<td>8.18E–07</td>
<td>4.66</td>
<td>7.02E–06</td>
<td>4.66</td>
</tr>
<tr>
<td>800</td>
<td>2.67E–08</td>
<td>4.94</td>
<td>2.29E–07</td>
<td>4.94</td>
</tr>
</tbody>
</table>

Here the dominating term in the truncation error is the spatial discretization of the convective part, so the scheme converges with fifth order on the grids used.
3. Numerical experiments

3.1. Order of accuracy

To verify the order of accuracy we follow Xing and Shu [29] and choose

\[ z(x) := \sin^2(\pi x), \]
\[ h(x, 0) := 5 + e^{\cos(2\pi x)}, \]
\[ hu(x, 0) := \sin(\cos(2\pi x)). \]

for bottom topography, initial water height and momentum. Here \( x \in [0, 1] \), the boundary conditions are periodic, and the gravitational constant \( g \) is set to 9.812. We compute up to time \( t = 0.1 \) with CFL number 0.4. Since the exact solution for this experiment is not known explicitly, we use the same well-balanced WENO scheme of order (4,5,4) with \( N = 25,600 \) cells to compute a reference solution. We use a fifth-order WENO reconstruction with \( \varepsilon = 10^{-6} \) and optimal weights from (37), together with the weight splitting method [22] to compute the central point values needed in the quadrature (18). Table 1 contains the \( L^1 \) errors and numerical order of accuracy for both components. We achieve full fifth-order convergence in both components. Note that we have used the fifth-order WENO reconstruction in space, but only a fourth-order accurate extrapolation of the source term and the classical fourth-order Runge–Kutta time discretization. Thus not all elements of the algorithm contribute equally to the overall error. However, a standard second-order discretization of the source term does reduce the order of accuracy to two, see Table 2. This shows the relevance of the key new ingredient of our algorithm. We conjecture that on finer grids the quadrature rule for the source and the time discretization error would eventually dominate also for the (4,5,4) scheme, lowering the overall accuracy from fifth to fourth order.

3.2. Perturbation of a lake at rest

The following problem was studied by LeVeque [20]. It shows the behavior of a small perturbation of a lake at rest with variable bottom topography

\[ z(x) = \begin{cases} 0.25(1 + \cos(10\pi(x - 0.5))) & \text{if } 1.2 \leq x \leq 1.4, \\ 0 & \text{else}, \end{cases} \]

where \( x \in [0,2] \). The total initial height is given by

\[ H(x, 0) = \begin{cases} 1 + \Delta H & \text{if } 1.1 \leq x \leq 1.2, \\ 1 & \text{else}. \end{cases} \]

Table 2

<table>
<thead>
<tr>
<th>Number of cells</th>
<th>( h )</th>
<th>( L^1 ) error</th>
<th>Order</th>
<th>( hu )</th>
<th>( L^1 ) error</th>
<th>Order</th>
</tr>
</thead>
<tbody>
<tr>
<td>Convergence table with second-order source term</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>25</td>
<td>1.12E – 02</td>
<td>2.58</td>
<td>2.27</td>
<td>8.29E – 02</td>
<td>2.78</td>
<td>2.77</td>
</tr>
<tr>
<td>50</td>
<td>1.87E – 03</td>
<td>2.71</td>
<td>3.77</td>
<td>1.73E – 02</td>
<td>3.99</td>
<td>4.19</td>
</tr>
<tr>
<td>100</td>
<td>2.86E – 04</td>
<td>3.99</td>
<td>4.19</td>
<td>2.50E – 03</td>
<td>5.05E – 08</td>
<td>2.14</td>
</tr>
<tr>
<td>200</td>
<td>2.18E – 05</td>
<td>2.65</td>
<td>2.77</td>
<td>1.82E – 04</td>
<td>2.11</td>
<td>2.14</td>
</tr>
<tr>
<td>400</td>
<td>1.37E – 06</td>
<td>3.70</td>
<td>3.77</td>
<td>9.99E – 06</td>
<td>3.31E – 07</td>
<td>2.14</td>
</tr>
<tr>
<td>800</td>
<td>2.18E – 07</td>
<td>3.99</td>
<td>4.19</td>
<td>1.46E – 06</td>
<td>2.27</td>
<td>2.77</td>
</tr>
<tr>
<td>1600</td>
<td>5.05E – 08</td>
<td>2.27</td>
<td>2.77</td>
<td>3.31E – 07</td>
<td>2.14</td>
<td>2.14</td>
</tr>
</tbody>
</table>

Now the error of the source term discretization dominates, lowering the overall order of convergence to 2.
LeVeque, who worked with a second-order scheme, used \( \Delta H = 0.1 \). We will use \( \Delta H = 0.001 \) as Xing and Shu [29], Vukovic and Sopta [28] used for their higher order schemes: in this fashion the perturbation becomes smaller and is therefore more challenging to capture. The initial velocity is set to

\[ v(x,0) = 0 \]

and the gravitational constant \( g = 9.81 \). We use a fifth-order WENO reconstruction with optimal weights from (37) and \( \varepsilon = 10^{-12} \) in order to satisfy (39). Indeed, \( \varepsilon = 10^{-6} \) results in oscillations at the shocks. Periodic boundary conditions are used. The CFL number is 0.4 and the final time is \( T = 0.2 \) (Fig. 4).

Figs. 5 and 6 show total height and momentum computed with 200 cells and 157 time steps, and a comparison between first, second and (4,5,4)th order solutions are shown in Fig. 7. At this time, the wave traveling to the right has just passed the hump, and part of it has been reflected. All the schemes are able to produce the physically correct reflected waves (see the interval \([1,1.5]\) around the hump). The new scheme shows remarkably high resolution. Schemes which do not preserve the discrete hydrostatic balance may introduce unphysical waves and high frequency oscillations (see [28, Figs. 8 and 9]).

### 3.3. Dambreak over a rectangular wall

This test case simulates a dambreak over a rectangular wall. It produces a rapidly varying flow over a discontinuous bottom topography. This example was used in [28,29]. The bottom topography is given by

\[ z(x) = \begin{cases} 
8 & \text{if } |x - 1500/2| \leq 1500/8, \\
0 & \text{otherwise},
\end{cases} \]

with \( x \in [0,1500] \). The total initial height is

\[ H(x,0) = \begin{cases} 
20 & \text{if } 0 \leq x \leq 750, \\
15 & \text{otherwise}.
\end{cases} \]

The initial velocity is set to zero \( v(x,0) = 0 \) and the gravitation constant is \( g = 9.81 \). At the left boundary we use reflective boundary conditions and on the right side open boundary conditions. In Fig. 8, we show level lines of the water level, or total height, of the solution up to time \( T = 60 \). In the beginning, one observes the

![Fig. 4. Example 3.2, bottom topography and initial water level.](image-url)
standard rarefaction and shock waves which form the solution of the Riemann problem of the homogeneous shallow water equations. Figs. 9 and 10 show the water level and velocity at $T = 15$. At time $T \approx 17$ the waves cross the two edges of the wall. A part is transmitted, another part reflected, and a remaining part becomes a standing wave. Such standing waves have recently been studied analytically by Klausen and Risebro [17], Towers [27], Klingenberg and Risebro [18], and Seguin and Vovelle [21] who consider the inhomogeneous
one dimensional shallow water equations as a system of three conservation laws for $(h, hu, z)$ with $\partial_t z = 0$. This system has the three wave speeds $u \pm \sqrt{gh}$ and 0. For later times, the wave system keeps interacting. At time $T = 60$, we have six waves in the solution. The main shock and rarefaction waves just hit the boundary of the computational domain. Between them we have, from left to right, a standing wave, a weak rarefaction traveling leftwards, a second standing wave, and a weak compressive wave traveling rightwards. Figs. 11 and 12 show cross sections of total height and velocity. Note that the standing waves (which are contact
discontinuities) are not easy to capture. This is discussed, e.g., in [28]. Here we have almost perfect resolution of all features of this challenging solution.

4. Two dimensional extension

The shallow water equations in 2D are given by
\[ h_t + (hu)_x + (hv)_y = 0, \]
\[ (hu)_t + (hu^2 + \frac{1}{2}gh^2)_x + (huv)_y = -ghz_x, \]
\[ (hv)_t + (huv)_x + \left( hv^2 + \frac{1}{2}gh^2 \right)_y = -ghz_y, \]
where \( h \) is the water height, \( z \) is the bottom topography, \( u \) is the velocity in \( x \)-direction, \( v \) is the velocity in \( y \)-direction, and \( g \) is the gravitational constant. We will now discuss how to extend our scheme to two dimensions.

### 4.1. Overview of the scheme

Rewrite the system (19) in the standard form:

\[
U_t + F_x(U) + G_y(U) = S(U),
\]

where clearly \( U = (h, hu, hv, hv^2 + \frac{1}{2}gh^2, hvw + \frac{1}{2}gh^2)^T, \) \( F = (hu, hv, hu^2 + \frac{1}{2}gh^2, hvw)^T, \) \( G = (hv, hv, hv^2 + \frac{1}{2}gh^2, hvw + \frac{1}{2}gh^2)^T, \) and \( S = (0, -gzh_x, -gzh_y)^T. \)

We define the cell averages over grid cells \( I_{ij} = (x_{i-\frac{1}{2}, j}, x_{i+\frac{1}{2}, j}) \times (y_{j-\frac{1}{2}}, y_{j+\frac{1}{2}}) \) by

\[
U_{ij} = \frac{1}{\Delta x \Delta y} \int_{I_{ij}} U(x, y) \, dx \, dy,
\]

where \( \Delta x = x_{i+\frac{1}{2}} - x_{i-\frac{1}{2}}, \Delta y = y_{j+\frac{1}{2}} - y_{j-\frac{1}{2}} \). Suppose for simplicity that the cells are square: let \( \delta = \Delta x = \Delta y \). Integrating each term in (19) over the cell \( I_{ij} \) and invoking the divergence theorem, we get the following semidiscrete scheme for the evolution of the cell averages \( U_{ij} \):

\[
\frac{\delta^2}{\delta t} \frac{d}{d t} U_{ij}(t) + \int_{\partial I_{ij}} (F, G) \cdot \mathbf{n} \, ds = \int_{I_{ij}} S(x, y) \, dx \, dy,
\]

Rewrite the system as:

\[
\frac{d}{d t} U_{ij}(t) + \frac{F_{i+\frac{1}{2},j}^+ - F_{i-\frac{1}{2},j}^-}{\delta} + \frac{G_{i,j+\frac{1}{2}}^- - G_{i,j-\frac{1}{2}}^+}{\delta} = S_{ij},
\]

where

\[
F_{i-\frac{1}{2},j}^+ = \frac{1}{\delta} \int_{y_{j-\frac{1}{2}}}^{y_{j+\frac{1}{2}}} F(U(x_{i-\frac{1}{2}, y})) \, dy,
\]

\[
G_{i,j-\frac{1}{2}}^+ = \frac{1}{\delta} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} G(U(x, y_{j-\frac{1}{2}})) \, dx,
\]

\[
S_{ij} = \frac{1}{\delta^2} \int_{I_{ij}} S(x, y) \, dx \, dy.
\]

In analogy with the 1D case, we reconstruct the variables \( h, hu, hv, \) and \( H \), while the bottom topography is given by \( z = H - h \). In general, this yields a discontinuous approximation of \( z \). Let \( U_{ij}(x, y) \) denote the reconstruction computed in the cell \( I_{ij} \) with \( U \) denoting any of the reconstructed variables. Again, to preserve the equilibrium states, a hydrostatic reconstruction is needed on the quadrature points on the boundary of the cell, which will be denoted by \( h^*: \)

\[
\begin{align*}
h_{i-\frac{1}{2},j}^+(x_{i+\frac{1}{2}}, \cdot) &= \max \left( 0, H_{i+1,j} - \max \left( z_{ij}(x_{i+\frac{1}{2}}, \cdot), z_{i+1,j}(x_{i+\frac{1}{2}}, \cdot) \right) \right), \\
h_{i,j}^+(x_{i+\frac{1}{2}}, \cdot) &= \max \left( 0, H_{ij} - \max \left( z_{ij}(x_{i+\frac{1}{2}}, \cdot), z_{i+1,j}(x_{i+\frac{1}{2}}, \cdot) \right) \right), \\
h_{i,j+1}(\cdot, y_{j+\frac{1}{2}}) &= \max \left( 0, H_{ij+1} - \max \left( z_{ij}(\cdot, y_{j+\frac{1}{2}}), z_{i,j+1}(\cdot, y_{j+\frac{1}{2}}) \right) \right), \\
h_{i,j}^+(\cdot, y_{j+\frac{1}{2}}) &= \max \left( 0, H_{ij} - \max \left( z_{ij}(\cdot, y_{j+\frac{1}{2}}), z_{i,j+1}(\cdot, y_{j+\frac{1}{2}}) \right) \right).
\end{align*}
\]

To approximate the quantities \( F_{i-\frac{1}{2},j}^+ \) and \( G_{i,j-\frac{1}{2}}^+ \) in (23) and (24), we use a quadrature

\[
F_{i-\frac{1}{2},j}^+ \simeq \sum_k \omega_k F(U(x_{i-\frac{1}{2}}, y_j + \zeta_k \delta)),
\]

\[
G_{i,j-\frac{1}{2}}^+ \simeq \sum_k \omega_k G(U(x_{i-\frac{1}{2}}, y_j + \zeta_k \delta)).
\]
where $\omega_k$ and $\xi_k$ are the weights and nodes of the quadrature formula. For a fourth-order scheme we use the classical two-point Gaussian formula

$$F_{i-\frac{1}{2},j} \simeq \frac{1}{2} \left( F \left( U \left( x_{i-\frac{1}{2}},y_j - \alpha \delta \right) \right) + F \left( U \left( x_{i-\frac{1}{2}},y_j + \alpha \delta \right) \right) \right),$$

(26)

where $\alpha = 1/(2\sqrt{3})$. A similar formula holds for $\bar{G}_{i,j-\frac{1}{2}}$.

We still need to construct a well balanced approximation to each of the flux evaluations required in (26). As in 1D, the numerical flux is composed of two contributions. The first contribution ($F^h$ for $F$ and $G^h$ for $G$) is consistent with the flux of the homogeneous shallow water equations, the second contribution compensates the perturbation introduced by the hydrostatic correction.

The modified state variables that will be applied in the flux computations are

$$U^*_{ij} = \begin{pmatrix} h_{ij}^* \\ (hu)_{ij} \\ (hv)_{ij} \end{pmatrix}.$$

Along the edge $(x_{i-\frac{1}{2}},y)$ for instance the numerical fluxes are

$$\mathcal{F}^i(U_{i-1,j}, U_{i,j}, z_{i-1,j}, z_{i,j})_{i-\frac{1}{2},j=\pm x}$$

$$:= F^h(U^*_{i-1,j} \left( x_{i-\frac{1}{2}},y_j \pm \alpha \delta \right), U^*_{ij} \left( x_{i-\frac{1}{2}},y_j \pm \alpha \delta \right)) + \frac{g}{2} \begin{pmatrix} 0 \\ h_{ij}^2 \left( x_{i-\frac{1}{2}},y_j \pm \alpha \delta \right) - (h^*)_{ij}^2 \left( x_{i-\frac{1}{2}},y_j \pm \alpha \delta \right) \\ 0 \end{pmatrix},$$

(27)

and

$$\mathcal{F}^r(U_{i,j}, U_{i+1,j}, z_{i,j}, z_{i+1,j})_{i+j=\pm x}$$

$$:= F^h(U^*_{i,j} \left( x_{i+\frac{1}{2}},y_j \pm \alpha \delta \right), U^*_{i+1,j} \left( x_{i+\frac{1}{2}},y_j \pm \alpha \delta \right)) + \frac{g}{2} \begin{pmatrix} 0 \\ h_{ij}^2 \left( x_{i+\frac{1}{2}},y_j \pm \alpha \delta \right) - (h^*)_{ij}^2 \left( x_{i+\frac{1}{2}},y_j \pm \alpha \delta \right) \\ 0 \end{pmatrix},$$

(28)

with similar formulas for $\mathcal{F}^l$ and $\mathcal{F}^r$.

Thus the semidiscrete scheme can be written as,

$$\frac{d}{dt} U_{ij}(t) = -\frac{1}{2\delta} \left( \mathcal{F}^r_{i+\frac{1}{2},j+\frac{1}{2},z} + \mathcal{F}^r_{i+\frac{1}{2},j-\frac{1}{2},z} - \mathcal{F}^l_{i-\frac{1}{2},j+\frac{1}{2},z} - \mathcal{F}^l_{i-\frac{1}{2},j-\frac{1}{2},z} + \mathcal{F}^r_{i+z,j+\frac{1}{2},z} + \mathcal{F}^r_{i+z,j-\frac{1}{2},z} - \mathcal{F}^l_{i-z,j+\frac{1}{2},z} - \mathcal{F}^l_{i-z,j-\frac{1}{2},z} \right) + S_{ij}. $$

(29)

The construction of the source term $S_{ij}$ is carried out as follows. First we write the source term component-wise: $S_{ij} = (0, S^x_{ij}, S^y_{ij})^T$. Note that the component of the source term in the $x$-momentum equation contains only the derivative of $z$ along the $x$ direction. Thus, we employ the well-balanced quadrature (18) of the previous section to integrate in the $x$-direction and apply the Gaussian rule in the $y$-direction. For the fourth-order case,

$$S^y_{ij} = \frac{\delta}{2} \left( s^y_i(y_j + \alpha \delta) + s^y_i(y_j - \alpha \delta) \right),$$

(30)

where

$$s^y_i(y) = \frac{4g}{6} \left( \left( h_{ij} \left( x_{i-\frac{1}{2}},y \right) + h_{ij} \left( x_{i+\frac{1}{2}},y \right) \right) \left( z_{ij} \left( x_{i-\frac{1}{2}},y \right) - z_{ij} \left( x_{i-\frac{1}{2}},y \right) \right) + \left( h_{ij} \left( x_{i+\frac{1}{2}},y \right) + h_{ij} \left( x_{i+\frac{1}{2}},y \right) \right) \left( z_{ij} \left( x_{i+\frac{1}{2}},y \right) - z_{ij} \left( x_{i+\frac{1}{2}},y \right) \right) \right)$$

$$- \frac{g}{6} \left( h_{ij} \left( x_{i-\frac{1}{2}},y \right) + h_{ij} \left( x_{i+\frac{1}{2}},y \right) \right) \left( z_{ij} \left( x_{i-\frac{1}{2}},y \right) - z_{ij} \left( x_{i+\frac{1}{2}},y \right) \right).$$
In the same fashion, we compute the source \( S_{ij}^y \) using (18) in the \( y \)-direction and the Gaussian rule in the \( x \)-direction. Again, in the fourth-order case:

\[
S_{ij}^y = \frac{\delta}{2} \left( s_j^y(x_i + \alpha \delta) + s_j^y(x_i - \alpha \delta) \right),
\]

where now,

\[
s_j^y(x) = \frac{4g}{6} \left( (h_{ij}(x, y_{j-\frac{1}{2}}) + h_{ij}(x, y_{j+\frac{1}{2}})) \left( z_j(x, y_{j-\frac{1}{2}}) - z_j(x, y_{j+\frac{1}{2}}) \right) + (h_{ij}(x, y_{j-\frac{1}{2}}) + h_{ij}(x, y_{j+\frac{1}{2}})) \left( z_j(x, y_{j-\frac{1}{2}}) - z_j(x, y_{j+\frac{1}{2}}) \right) \right) - \frac{g}{6} \left( h_{ij}(x, y_{j-\frac{1}{2}}) + h_{ij}(x, y_{j+\frac{1}{2}}) \right) \left( z_j(x, y_{j-\frac{1}{2}}) - z_j(x, y_{j+\frac{1}{2}}) \right)\]

Using the same arguments as in the proof of Theorem 3 one can show:

**Corollary 5.** The 2D scheme is fourth-order accurate and preserves the stationary state of the lake at rest.

### 4.2. 2D reconstruction

In order to evaluate the numerical flux functions \( \mathcal{F} \) and \( \mathcal{G} \) and the source term \( S \), we need to reconstruct point values of \( H, h, hu \) and \( hv \) at 12 integration points, 8 on the boundary \((x_{i\pm\frac{1}{2}}, y_{j\pm\frac{1}{2}}), (x_{i\pm\frac{1}{2}}, y_{j\pm\frac{1}{2}})\) and 4 in the interior \((x_i,y_{j\pm\frac{1}{2}})\) as shown on the right of Fig. 13. Note that the interior points are required only to compute the source term, which is fourth-order accurate. As in the 1D case we apply a WENO procedure to find these data.

In 2D this reconstruction is somewhat more involved. Our approach is to reconstruct each variable dimension by dimension. For each cell, the one-dimensional WENO procedure has to be applied six times to produce point values in all quadrature points.

To fix ideas, we illustrate the algorithm for the reconstruction of the variable \( h \). In this section only, we denote the cell averages as \( \bar{h}_{ij} \), to distinguish the averages computed on a cell (a double integral) from the averages computed along only one segment (a single integral).

We start applying the WENO reconstruction procedure in the \( y \) direction, starting from the cell averages \( \bar{h}_{ij} \). We apply the reconstruction defined by the constants in Appendix A (38) and we find approximations in the points \((x_i, y_{j+\frac{1}{2}})\) and \((x_i, y_{j-\frac{1}{2}})\) to the function:

\[
\bar{h}_{ij}(x_i, y_{j+\frac{1}{2}}) = \frac{1}{\delta} \int_{x_i-\frac{\alpha}{2}}^{x_i+\frac{\alpha}{2}} h(x, y_{j+\frac{1}{2}}) \, dx.
\]

![Fig. 13. (Left) The positions of the reconstructed cross section averages \((x_{i\pm\frac{1}{2}}, \cdot)\) (dashed lines), \((\cdot, y_{j\pm\frac{1}{2}})\) (black lines) for the first reconstruction step. (Right) The Gaussian integration points for the edges \((x_{i\pm\frac{1}{2}}, y_{j\pm\frac{1}{2}})\) and for the interior, \((x_{i\pm\frac{1}{2}}, y_{j\pm\frac{1}{2}})\) and \((x_{i\pm\frac{1}{2}}, y_{j\pm\frac{1}{2}})\). The point values in these locations are computed in the second step of the reconstruction.](image-url)
Once these data are available for all $i$, we apply again the WENO reconstruction along the $x$ axis to get the required point values, i.e., $h_{ij}(x_{i-1/2}, y_{j})$, $h_{ij}(x_{i+1/2}, y_{j})$ and $h_{ij}(x_{i+1/2}, y_{j+1/2})$, starting from $\tilde{h}_{ij}(x_{i}, y_{j})$. With another reconstruction, we find $h_{ij}(x_{i-1/2}, y_{j-1/2})$, $h_{ij}(x_{i+1/2}, y_{j-1/2})$ and $h_{ij}(x_{i+1/2}, y_{j+1/2})$, starting from $\tilde{h}_{ij}(x_{i}, y_{j-1/2})$. This set of operations will be called $y$–$x$ sweep, see Fig. 14.

To get the quadrature points along the dashed lines on the right in Fig. 13, we perform the same operations in the reversed order. This will be called $x$–$y$ sweep, see Fig. 15.

Now, all the quantities appearing in the semidiscrete scheme (22) have been defined. Finally, to get a fully discrete scheme, we need to specify a method to march forward in time. As in the 1D scheme, we apply the classic fourth-order Runge–Kutta method. For other cases, it might be advantageous to use the recent TVD or SSP Runge–Kutta schemes [23,12,26].

### 4.3. Well-balanced test in two dimensions

The two dimensional experiments we present here follow closely the work of Xing and Shu [29]. We check the behavior of the two dimensional scheme in a lake at rest situation on a rectangular domain $[0,1] \times [0,1]$, with a non-flat, fully two-dimensional, bottom topography

$$z(x, y) = 0.8e^{-50((x-0.5)^2+(y-0.5)^2)}.$$ (32)

Fig. 14. In the first step of the $y$–$x$-sweep we compute averages over cross sections ($\cdot, y_{j+\alpha}$) marked with black lines in the figure on the left side. In the second step of the $y$–$x$-sweep we use the cross section averages to compute point values at quadrature points $(x_{i+\frac{1}{2}}, y_{j+\alpha})$, $(x_{i}, y_{j+\alpha})$, $(x_{i+\frac{1}{2}}, y_{j})$, $(x_{i}, y_{j})$ (right figure).

Fig. 15. In the first step of the $x$–$y$-sweep we compute averages over cross sections ($\cdot, x_{i+\alpha}$) marked with dotted lines in the figure on the left side. In the second step of the $x$–$y$-sweep we use the cross section averages to compute point values at quadrature points $(x_{i+\alpha}, y_{j+\frac{1}{2}})$, $(x_{i+\alpha}, y_{j})$, $(x_{i}, y_{j})$, $(x_{i+\alpha}, y_{j-\frac{1}{2}})$ (right figure).
The initial water height is

\[ h(x, y) = 1 - z(x, y), \]  

so that the water surface level \( H \) is constant, with \( H = 1.0 \). The momentum in \( x \) and \( y \) direction is set to zero:

\[ hu(x, y, t = 0) = 0 \quad \text{and} \quad hv(x, y, t = 0) = 0. \]  

The lake is at rest initially, and should remain at rest indefinitely. In this situation, a scheme without well-balancing would produce unphysical waves. For this test we use a uniform \( 100 \times 100 \) grid and compute the solution at time \( t = 0.1 \). We get the following \( L^1 \)-errors for the conservative components:

\[ \|h\|_1 = 1.23 \times 10^{-16}, \]  
\[ \|hu\|_1 = 2.20 \times 10^{-16} \]  
\[ \|hv\|_1 = 2.22 \times 10^{-16}. \]  

The errors are all of the magnitude of the rounding error \( 10^{-16} \) thus the scheme is indeed perfectly well-balanced.

### 4.4. Testing the order of accuracy

To check the numerical order of accuracy we use the same experiment as Xing and Shu [29]. On the unit square \([0,1] \times [0,1]\) we choose the bottom topography:

\[ z(x, y) = \sin(2\pi x) + \cos(2\pi y) \]

the initial water surface level:

\[ h(x, y, t = 0) = 10 + e^{20}(2\pi x) \cos(2\pi y) \]

and the initial momentum in the \( x \) and \( y \) directions, respectively:

\[ hu(x, y, t = 0) = \sin(\cos(2\pi x)) \sin(2\pi y), \]
\[ hv(x, y, t = 0) = \cos(2\pi x) \cos(\sin(2\pi y)). \]

We compute up to time \( T = 0.05 \) with CFL-number 0.8. For the WENO reconstruction we use the optimal weights of 37,38 and set \( \epsilon = 10^{-6} \). The reference solution is computed with the same scheme and \( 1600 \times 1600 \) cells, since the exact solution is unknown.

For this experiment we expect fourth-order of accuracy in all conservative components. The applied standard Runge–Kutta time integration, the integration of the numerical fluxes with Gaussian rule and the cell centered source term are all formally fourth-order accurate, while the applied WENO reconstruction is fifth-order accurate. Table 3 contains the \( L^1 \)-errors and orders of accuracy. We can clearly see that for this two dimensional test case, fourth-order accuracy (in fact almost fifth-order) is indeed achieved in all components.

### 4.5. A small perturbation of a two dimensional steady-state lake

This classical problem is given by LeVeque [20] and is also computed in [29]. For this problem we consider the rectangular domain \([0,2] \times [0,1]\). The bottom topography is displayed in Fig. 16 and it is given by:

\[ z(x, y) = 0.8e^{-(5(x-0.9)^2-50(y-0.5)^2)}. \]

<table>
<thead>
<tr>
<th>Number of points</th>
<th>CFL</th>
<th>( h ) ( L^1 ) error</th>
<th>Order</th>
<th>( hu ) ( L^1 ) error</th>
<th>Order</th>
<th>( hv ) ( L^1 ) error</th>
<th>Order</th>
</tr>
</thead>
<tbody>
<tr>
<td>25</td>
<td>0.5</td>
<td>8.77E – 03</td>
<td></td>
<td>3.42E – 02</td>
<td></td>
<td>6.71E – 02</td>
<td></td>
</tr>
<tr>
<td>50</td>
<td>0.5</td>
<td>1.10E – 03</td>
<td>3.00</td>
<td>2.73E – 03</td>
<td>3.65</td>
<td>9.40E – 03</td>
<td>2.84</td>
</tr>
<tr>
<td>100</td>
<td>0.5</td>
<td>9.84E – 05</td>
<td>3.48</td>
<td>1.56E – 04</td>
<td>4.13</td>
<td>7.85E – 04</td>
<td>3.58</td>
</tr>
<tr>
<td>200</td>
<td>0.5</td>
<td>4.91E – 06</td>
<td>4.32</td>
<td>6.58E – 06</td>
<td>4.57</td>
<td>3.93E – 05</td>
<td>4.32</td>
</tr>
<tr>
<td>400</td>
<td>0.5</td>
<td>1.82E – 07</td>
<td>4.76</td>
<td>2.41E – 07</td>
<td>4.77</td>
<td>1.46E – 06</td>
<td>4.75</td>
</tr>
<tr>
<td>800</td>
<td>0.5</td>
<td>6.06E – 09</td>
<td>4.91</td>
<td>7.94E – 09</td>
<td>4.92</td>
<td>4.90E – 08</td>
<td>4.90</td>
</tr>
</tbody>
</table>
The initial water surface level is given by:

\[ h(x, y, t = 0) = \begin{cases} 
1.01 - z(x, y) & \text{if } 0.05 \leq x \leq 0.15, \\
1 - z(x, y) & \text{otherwise},
\end{cases} \]

so the initial surface level is almost flat, only in the region \(0.05 < x < 0.15\) it is perturbed upward by the displacement 0.01. The initial momentum in the \(x\) and \(y\) directions is:

\[ hu(x, y, t = 0) = 0, \]
\[ hv(x, y, t = 0) = 0. \]

We compute using two different uniform meshes with \(200 \times 100\) cells and \(600 \times 300\) cells.

Fig. 17 shows 30 uniformly spaced contour lines of the surface level \(H\) at times \(t = 0.12, 0.24, 0.36, 0.48\) and final time \(T = 0.6\). The results obtained with the coarse grid appear on the left side, while on the right we find the numerical solution obtained with the fine grid.

In the simulation with the coarse grid we use \(\varepsilon = 10^{-6}\) and for the fine grid \(\varepsilon = 10^{-9}\). In both experiments we choose the CFL-number equal to 0.5. As we can see, we get results comparable to the finite difference approach in [29].

5. Conclusion

In this paper, we have constructed well-balanced finite volume schemes for the shallow water equations, which are of any desired order of accuracy. The new schemes generalize a class of second-order schemes proposed by Audusse et al. [1]. A (4,5,4)th order version of the new scheme gives the expected high resolution both for smooth and non-smooth flows, and perfect balance for the lake at rest in one and two spatial directions. The key technique, a new quadrature formula for the source term, can be applied to a wide variety of first and second-order well balanced schemes, to raise their order of accuracy. Work on stable schemes for flows with dry areas is in progress.

We would like to thank an unknown referee for pointing us to the recent preprints [30,31], where a class of more general balance laws is treated via different techniques. Our high order accurate quadrature technique...
Fig. 17. Contour lines of the surface level $h + z$ for the experiment of Section 4.5 at times $t = 0.12, 0.24, 0.36, 0.48, 0.6$. Left: $200 \times 100$ grid, right: $600 \times 300$ grid. There are 30 uniformly spaced contour lines in each plot. At time $t = 0.12$ the contour lines go from 0.999837 to 1.005974; at time $t = 0.24$ from 0.996091 to 1.014523; at time $t = 0.36$ from 0.988829 to 1.011245; at time $t = 0.48$ from 0.990559 to 1.004614; at time $t = 0.6$ from 0.995244 to 1.005207.
using extrapolation carries over to that class as well, and it will be well-balanced exactly if the continuity condition (4.8) of [31] holds. For the lake at rest, (4.8) is given by the hydrostatic reconstruction, but for each system of balance laws, an analogous technique has to be established (separately for each scheme). We will leave this to future work.

Acknowledgements

We are grateful to two unknown referees for asking inspiring questions about the deeper mechanism of our balancing approach. Moreover, we would like to thank Francois Bouchut for lively and stimulating discussions. Francois gave us a version of his first and second-order well-balanced scheme. We would also like to thank the Institute of Hydraulic Engineering and Water Resources Management of RWTH Aachen for providing the topographical data of lake Rursee. Part of this work was done while the first author was in residence at CMA, “Center of Mathematics for Application” at Oslo University. The first and second authors would like to thank CMA and its members for their generous hospitality.

Appendix A. WENO reconstruction

For completeness, we review the WENO reconstruction [15, 24] for uniform grids in 1D. We aim to give sufficient details of all parameters so that the reader could reconstruct our algorithm and recover the numerical experiments. Necessarily, we present this as concisely as possible. For more details, we refer to the relevant literature.

A main ingredient that we were not able to find elsewhere in the literature are the accuracy constants for the points in the Gaussian quadrature at the edges of the two-dimensional cells.

Given cell averages

\[ u_i = \frac{1}{|C_i|} \int_{C_i} u(x) \, dx \]
on cell \( C_i \) of a smooth function \( u(x) \) and a fixed point \( x^* \in C_j \), the WENO procedure provides a highly accurate piecewise polynomial approximation \( R(x^*) \) of \( u(x^*) \).

\[ R(x^*) = u(x^*) + \mathcal{O}(\Delta x^{2r-1}) \]

1. Define \( r \) small stencils, composed of \( r \) cells, around the cell containing \( x_j \)

\[ S_k := (x_{j+k-r+1}, x_{j+k-r+2}, \ldots, x_{j+k}), \quad k = 0, \ldots, r-1 \]

and one large stencil

\[ \mathcal{S} := \bigcup_{k=0}^{r-1} S_k \]

which contains all the cells from the \( r \) smaller stencils.

2. Given cell averages \( u_j \) compute the interpolation polynomials \( p_k(x) \) of degree \( (r - 1) \) associated with the stencils \( S_k \) for \( k = 0, \ldots, r - 1 \) and the higher order reconstruction polynomial \( Q(x) \), of degree \( (2r-1) \) associated with the large stencil \( \mathcal{S} \). Here, interpolation is understood in the sense of cell averages.

3. Find the linear weights \( C_0, \ldots, C_{r-1} \) such that:

\[ Q(x^*) = \sum_{k=0}^{r-1} C_k^* p_k(x^*). \]
For the fifth-order WENO reconstruction used in this paper, solving the linear system (36) leads to the following weights:

\[
\begin{array}{cccc}
\chi^* & C_0^r & C_1^r & C_2^r \\
x_j & 0.3 & 0.6 & 0.1 \\
x_j-\frac{1}{2} & -0.1125 & 1.225 & -0.1125 \\
x_j+\frac{1}{2} & 0.1 & 0.6 & 0.3 \\
\end{array}
\]  

(37)

For the 2D extension we need also weights for the Gaussian points on the edges. Again we obtain them by solving (36):

\[
\begin{array}{cccc}
\chi^\ast & C_0^r & C_1^r & C_2^r \\
x_j - \frac{\Delta x}{2\sqrt{3}} & \frac{70\sqrt{3}+1}{360\sqrt{3}} & \frac{11}{18} & \frac{70\sqrt{3}-1}{360\sqrt{3}} \\
x_j + \frac{\Delta x}{2\sqrt{3}} & \frac{70\sqrt{3}-1}{360\sqrt{3}} & \frac{11}{18} & \frac{70\sqrt{3}+1}{360\sqrt{3}} \\
\end{array}
\]  

(38)

4. Compute the smoothness indicators

\[ IS_k = \sum_{l=1}^{r-1} \int_{x_j-\frac{1}{2}}^{x_j+\frac{1}{2}} \Delta x^{2l-1} (p_k^{(l)})^2 \, dx, \]

where \( l \) denotes the \( l \)th order derivative of \( p_k \).

5. Compute the nonlinear weights based on the smoothness indicators

\[ \omega_k := \frac{\alpha_k}{\alpha_0 + \cdots + \alpha_{r-1}}, \]

where

\[ \alpha_k := \frac{C_r^k}{(\varepsilon + IS_k)^2}, \quad k = 0, 1, \ldots, r - 1. \]

Here \( \varepsilon \) is a real number which is introduced to prevent the denominator from becoming zero. To preserve accuracy, \( \varepsilon \) should satisfy the constraints

\[ 0 < \varepsilon, \quad \text{everywhere}, \]

\[ \varepsilon \ll IS_k, \quad \text{in regions where the solution is smooth}. \]

In the numerical experiments we use \( \varepsilon = 10^{-6} \), except for the experiment in Section 3.2, where the \( IS_k \) are extremely small and we need to use \( \varepsilon = 10^{-12} \) and for the fine grid computation of the 2D test of Section 4.5. This is due to the fact that the perturbation introduced in the lake at rest in these particular tests is so small that the smoothness indicators become of the same order as \( 10^{-6} \) even when the solution is non smooth.

6. The final WENO reconstruction is given by:

\[ R(x^*) = \sum_{k=0}^{r-1} \omega_k p_k(x^*). \]

It is well known that the negative weights appearing in (37) may lead to oscillations at discontinuities [22]. Note however that this problem occurs only in the reconstruction of the point values at the locations \((x, y/\sqrt{3})\) and \((x\pm x, y)\) which are needed only in the computation of \( s^x_i(y) \) and \( s^y_j(x) \), see (30) and (31). We have tested the following two approaches: first, we simply replaced the weights by the fourth-order accurate choice of \( C_0^3 := 0.25, C_1^3 := 0.5 \) and \( C_2^3 := 0.25. \) Since these data midpoint appear only in the quadrature \( S^4 \)
for the source term, which is only fourth-order accurate anyway, this does not decrease the overall order (4,5,4) of the algorithm. The second cure is the splitting technique of Shi, Hu and Shu [22]. For the problems presented in this paper, this more expensive approach did not lead to superior results.

References