

Homework Problem Sheet 6

Introduction. This exercise sheet focuses on the stable FVM discretizations of the *non-linear* conservation laws. The first problem is the implementation of various numerical flux functions for the approximate Riemann solver and application of the methods for the one-dimensional Burgers' equation. The second problem shows how to apply many of the FVM schemes that you have already learned to a more practical setting. In particular, you will solve an example of a *non-linear multi-dimensional system* of hyperbolic conservation laws - the so-called shallow water equations. Such equations are used to model dynamics in fluids, where depth is very small, compared to the width and length of a given physical domain. In particular, flows in oceans, rivers, lakes, as well as atmospheric flows of air masses are often modeled using shallow water equations. In this problem, we will use a very simplified model of a “dambreak”, in one and two space dimensions, modeling the corresponding physical 2-D and 3-D flows, respectively. Upon multiple requests, the master solution for the second problem will be provided in Python (however, MATLAB users should be able to easily understand the code, as many things are similar).

Problem 6.1 Burgers' equation: Rankine-Hugoniot condition and Riemann solvers (Core problem)

Consider the one-dimensional Burger's equation:

$$U_t + \left(\frac{U^2}{2} \right)_x = 0, \quad \forall (x, t) \in D \times \mathbb{R}_+, \quad (6.1.1)$$

with domain $D = (-1, 3)$ and following initial conditions:

$$\textbf{(Shock)} \quad U(x, 0) = \begin{cases} 1, & \text{if } x < 0, \\ 0, & \text{otherwise,} \end{cases} \quad (6.1.2)$$

$$\textbf{(Rarefaction)} \quad U(x, 0) = \begin{cases} -1, & \text{if } x < 0, \\ 1, & \text{otherwise,} \end{cases} \quad (6.1.3)$$

$$\textbf{(Rarefaction and Shock)} \quad U(x, 0) = \begin{cases} 1, & \text{if } 0 < x < 1, \\ 0, & \text{otherwise.} \end{cases} \quad (6.1.4)$$

(6.1a) Using *Rankine-Hugoniot conditions* (for shock waves) and *entropy conditions* (to distinguish between shocks and rarefaction waves), derive the expression for *weak entropy* solutions of (6.1.1) and all three initial conditions (6.1.2) - (6.1.4).

Solution: For a convex flux $f(\cdot)$ and Riemann initial data of the general form

$$U(x, 0) = \begin{cases} U_l, & \text{if } x < 0, \\ U_r, & \text{otherwise,} \end{cases} \quad (6.1.5)$$

with constant left and right states U_l and U_r .

For $U_l > U_r$ (which is the case for (6.1.2)), the solution is given as a shock

$$U(x, t) = \begin{cases} U_l & x < st, \\ U_r & \text{otherwise,} \end{cases} \quad (6.1.6)$$

moving with the wave speed s given by the Rankine-Hugoniot condition:

$$s = \frac{f(U_r) - f(U_l)}{U_r - U_l} = \frac{1}{2}.$$

For $U_l < U_r$ (which is the case for (6.1.3)), the solution is given as a rarefaction wave

$$U(x, t) = \begin{cases} -1, & \text{if } x < -t, \\ x/t, & \text{if } -t \leq x < t, \\ 1 & \text{otherwise.} \end{cases} \quad (6.1.7)$$

For the case (6.1.4), initially we have rarefaction wave joining state 0 and 1, starting around point $x = 0$, and shock wave separating 1 and 0, starting around point $x = 0$. The solution is given by

$$U(x, t) = \begin{cases} 0, & \text{if } x < 0, \\ x/t, & \text{if } 0 \leq x < t, \\ 1, & \text{if } t \leq x < 1 + \frac{t}{2}, \\ 0 & \text{otherwise.} \end{cases} \quad (6.1.8)$$

At $t = 2$ the rarefaction wave meet the shock wave so solution after this time is given by a 0 left state followed by a rarefaction wave that form a shock with 0 on the right state. The shock speed for this shock is given by,

$$\sigma'(t) = \frac{\frac{1}{2} \left(\frac{\sigma(t)}{t} \right)^2}{\frac{\sigma(t)}{t}} = \frac{\sigma(t)}{2t} \quad (t \geq 2),$$

with initial condition $\sigma(2) = 2$. This results in $\sigma(t) = (2t)^{1/2}$. So finally solution for $t \geq 2$ is given by,

$$U(x, t) = \begin{cases} 0, & \text{if } x < 0, \\ x/t, & \text{if } 0 \leq x < (2t)^{1/2}, \\ 0 & \text{otherwise.} \end{cases} \quad (t \geq 2) \quad (6.1.9)$$

.

(6.1b) Implement a finite volume code for the Burgers' equation (6.1.1), using Godunov's, Roe, Lax-Friedrichs and Rusanov's numerical flux functions. Compute the numerical solution for the initial conditions (6.1.2), (6.1.3), and (6.1.4), using *outflow boundary conditions*. Plot the solutions at time $T = 1.0$ for 100 mesh cells. Explain the observed results.

HINT: *Outflow boundary conditions* can be easily implemented by extrapolating the values of the variable to the ghost cell i.e. $U_0^n = U_1^n$ and $U_{N+1}^n = U_N^n$.

Solution:

See codes 6.1 - 6.14 and figures 6.1 - 6.3 for the exact and the approximate FVM solutions.

Listing 6.1: Implementation for solve

```

1 function [ x, u, dx ] = solve( f_prime, flux, u0, l, r, T,
   mesh_size )
2
3 [ x, dx ] = getMesh( r, l, mesh_size );
4
5 u = feval(getCellAverages(u0, dx), x);
6
7 t = 0;
8 while (t < T)
9     % calculating time step from CFL condition
10    dt = 1/2 * dx / max(abs(f_prime(u)));
11    t = min(t + dt, T);
12    v = u;
13    v = apply_outflow_bc(v);
14    for j = 2:length(x)-1
15        u(j) = v(j) - dt/dx * (flux(j, v, dx, dt) - flux(j-1,
           v, dx, dt));
16    end
17 end
18
19 u = u(2:end-1);
20 x = x(2:end-1);
21
22 end

```

Listing 6.2: Implementation for solve_all

```

1 function [ x, u, dx ] = solve_all( f, f_prime, f_type,
   f_extrema, u0, l, r, T, mesh_size )
2
3 % Godunov scheme
4 flux = Godunov( f, f_prime, f_type, f_extrema);
5 [x, u(1,:), dx] = solve(f_prime, flux, u0, l, r, T,
   mesh_size);
6
7 % Roe scheme
8 flux = Roe( f, f_prime, f_type, f_extrema);

```

```

9  [~, u(2,:), ~] = solve(f_prime, flux, u0, l, r, T, mesh_size);
10
11 % Lax-Friedrichs scheme
12 flux = Lax_Friedrichs( f, f_prime, f_type, f_extrema);
13 [~, u(3,:), ~] = solve(f_prime, flux, u0, l, r, T, mesh_size);
14
15 % Rusanov scheme
16 flux = Rusanov( f, f_prime, f_type, f_extrema);
17 [~, u(4,:), ~] = solve(f_prime, flux, u0, l, r, T, mesh_size);
18
19 end

```

Listing 6.3: Implementation for Godunov

```

1  function [ flux ] = Godunov( f, f_prime, f_type, f_extrema )
2
3  flux = @(j, u, dx, dt) flux_eval(j, u, dx, dt, f, f_prime,
4      f_type, f_extrema);
5
6  end
7
8  function [flux] = flux_eval( j, u, ~, ~, f, ~, f_type,
9      f_extrema )
10
11 switch f_type
12     case 'convex',
13         flux = max( f(max(u(j), f_extrema)), f(min(u(j+1),
14             f_extrema)) );
15     case 'concave',
16         flux = min( f(min(u(j), f_extrema)), f(max(u(j+1),
17             f_extrema)) );
18     otherwise
19         flux = 0;
20
21 end
22
23 end

```

Listing 6.4: Implementation for LaxFriedrichs

```

1  function [ flux ] = Lax_Friedrichs( f, f_prime, f_type,
2      f_extrema )
3
4  flux = @(j, u, dx, dt) flux_eval(j, u, dx, dt, f, f_prime,
5      f_type, f_extrema);
6
7  end
8
9  function [flux] = flux_eval( j, u, dx, dt, f, ~, ~, ~ )

```

```

8
9 flux = ( f(u(j)) + f(u(j+1)) )/2 - dx/(2*dt) * ( u(j+1) -
    u(j) );
10
11 end

```

Listing 6.5: Implementation for Roe

```

1 function [ flux ] = Roe( f, f_prime, f_type, f_extrema )
2
3 flux = @(j, u, dx, dt) flux_eval(j, u, dx, dt, f, f_prime,
    f_type, f_extrema);
4
5 end
6
7 function [flux] = flux_eval( j, u, ~, ~, f, f_prime, ~, ~ )
8
9 % calculating Roe average
10 if abs(u(j+1) - u(j)) < eps
11     Roe_avg = f_prime(u(j));
12 else
13     Roe_avg = ( f(u(j+1)) - f(u(j)) ) / ( u(j+1) - u(j) );
14 end
15
16 % calculating flux
17 if Roe_avg >= 0
18     flux = f(u(j));
19 else
20     flux = f(u(j+1));
21 end
22
23 end

```

Listing 6.6: Implementation for Rusanov

```

1 function [ flux ] = Rusanov( f, f_prime, f_type, f_extrema )
2
3 flux = @(j, u, dx, dt) flux_eval(j, u, dx, dt, f, f_prime,
    f_type, f_extrema);
4
5 end
6
7 function [flux] = flux_eval( j, u, ~, ~, f, f_prime, ~, ~ )
8
9 flux = ( f(u(j)) + f(u(j+1)) )/2 - max( abs(f_prime(u(j))),
    abs(f_prime(u(j+1))) )/2 * ( u(j+1) - u(j) );
10
11 end

```

Listing 6.7: Implementation for `apply_outflow_bc`

```

1 function [ out ] = apply_outflow_bc( u )
2
3 out = u;
4
5 for i=1:size(u,1)
6     out(i, 1) = out(i, 2);
7     out(i, end) = out(i, end - 1);
8 end
9
10 end

```

Listing 6.8: Implementation for `shock`

```

1 function [ u0 ] = shock()
2
3 u0 = @(x) x < 0;
4
5 end

```

Listing 6.9: Implementation for `rarefaction`

```

1 function [ u0 ] = rarefaction()
2
3 u0 = @(x) -2*(x < 0) + 1;
4
5 end

```

Listing 6.10: Implementation for `rare_shock`

```

1 function [ u0 ] = rare_shock()
2
3 u0 = @(x) (x > 0) .* (x < 1);
4
5 end

```

Listing 6.11: Implementation for `getMesh`

```

1 function [ x, dx ] = getMesh( r, l, mesh_size )
2
3 dx = abs(r-l)/mesh_size;
4
5 x = linspace(l + dx/2, r - dx/2, mesh_size);    % space
6         discretization
7 x = [x(1)-dx x x(end)+dx];                      % ghost cells
8         for bc
9
10 end

```

Listing 6.12: Implementation for getRarefaction

```

1 function [ u_rare ] = getRarefaction( x, u_l, u_r, f_prime,
   f_prime_inv, T )
2
3 u_rare = x;
4
5 for i=1:length(x)
6
7     if (x(i) < (f_prime(u_l)*T))
8         u_rare(i) = u_l;
9     elseif (x(i) < (f_prime(u_r)*T))
10        u_rare(i) = f_prime_inv(x(i)/T);
11    else
12        u_rare(i) = u_r;
13    end
14
15 end
16
17 end

```

Listing 6.13: Implementation for plotCase

```

1 function [] = plotCase(
   ic_str,N,plot_res,axis_limits,T,l,r,f,f_prime,f_type,f_extrema,u0,u_e)
2
3 figure;
4
5 % solutions
6 [ x, u, dx ] = solve_all( f, f_prime, f_type, f_extrema, u0,
   l, r, T, plot_res );
7
8 subplot(2,2,[1 2]);
9 plot(x, u, '-o', x, feval(getCellAverages(u_e, dx), x));
10 title(strcat('Solution for', ic_str, ' initial conditions'));
11 xlabel('x');
12 ylabel('u(x,1)');
13 if strcmp(ic_str, ' rarefaction')
14     axis([l r -1.5 1.5]);
15 else
16     axis([l r -0.5 1.5]);
17 end
18 legend('Godunov','Roe','Lax-Friedrichs','Rusanov','Exact
   solution');
19
20 % convergence rates
21 [err_L1, err_Li] = geterrors (u_e, f, f_prime, f_type,
   f_extrema, u0, l, r, T, N);

```

```

22
23 subplot(2,2,3);
24 loglog(N, err_L1);
25 title(strcat('L^1 errors for', ic_str, ' initial
    conditions'));
26 legend('Godunov','Roe','Lax-Friedrichs','Rusanov');
27 xlabel('no. of cells');
28 ylabel('error');
29 grid on;
30 axis(axis_limits);
31
32 subplot(2,2,4);
33 loglog(N, err_Li);
34 title(strcat('L^\infty errors for', ic_str, ' initial
    conditions'));
35 legend('Godunov','Roe','Lax-Friedrichs','Rusanov');
36 xlabel('no. of cells');
37 ylabel('error');
38 grid on;
39 axis([1e2 2e3 1e-2 1e1]);
40
41 end

```

Listing 6.14: Implementation for main

```

1 function [] = main()
2
3 %%% === PROBLEM 1 ===
4
5 N = 100*(2.^(0:4));
6 plot_res = 100;
7 T = 1;
8 l = -1;
9 r = 3;
10
11 f = @(x) x.^2 / 2;
12 f_prime = @(x) x;
13 f_prime_inv = @(x) x;
14 f_type = 'convex';
15 f_extrema = 0;
16
17 %%% shock initial conditions
18
19 u0 = shock();
20 s = 1/2;
21 u_e = @(x) x < s*T;
22
23 ic_str = ' shock';

```



```

24
25 axis_limits = [1e2 2e3 1e-3 1e-1];
26 plotCase(ic_str, N, plot_res, axis_limits, T, l, r, f,
    f_prime, f_type, f_extrema, u0, u_e)
27
28 print -depsc '../fig/plot_shock.eps'
29
30 %%% rarefaction initial conditions
31
32 u0 = rarefaction();
33 u_l = -1;
34 u_r = 1;
35 u_e = @(x) getRarefaction(x, u_l, u_r, f_prime, f_prime_inv,
    T);
36
37 ic_str = ' rarefaction';
38
39 axis_limits = [1e2 2e3 1e-2 2e-0];
40 plotCase(ic_str, N, plot_res, axis_limits, T, l, r, f,
    f_prime, f_type, f_extrema, u0, u_e)
41
42 print -depsc '../fig/plot_rarefaction.eps'
43
44 %%% rarefaction and shock initial conditions
45
46 u0 = rare_shock();
47 s = 1/2;
48 u_l = 0;
49 u_r = 1;
50 u_e = @(x) (x>=1).*((x-1) < s*T) + (x<1).*getRarefaction(x,
    u_l, u_r, f_prime, f_prime_inv, T);
51
52 ic_str = ' rarefaction and shock';
53
54 axis_limits = [1e2 2e3 0.2e-2 1e-0];
55 plotCase(ic_str, N, plot_res, axis_limits, T, l, r, f,
    f_prime, f_type, f_extrema, u0, u_e)
56
57 print -depsc '../fig/plot_rarefaction-shock.eps'
58
59 end

```

(6.1c) Use the exact solutions derived in sub-problem (6.1a) to calculate $L^1(D)$ and $L^\infty(D)$ errors. Plot $L^1(D)$ and $L^\infty(D)$ errors vs numbers of cells (for no. of cells 100, 200, 400, 800, 1600), for each case of the initial conditions in (6.1.2) - (6.1.4).

HINT: The runs of the FVM solver for meshes with 800 and 1600 cells can take several minutes.

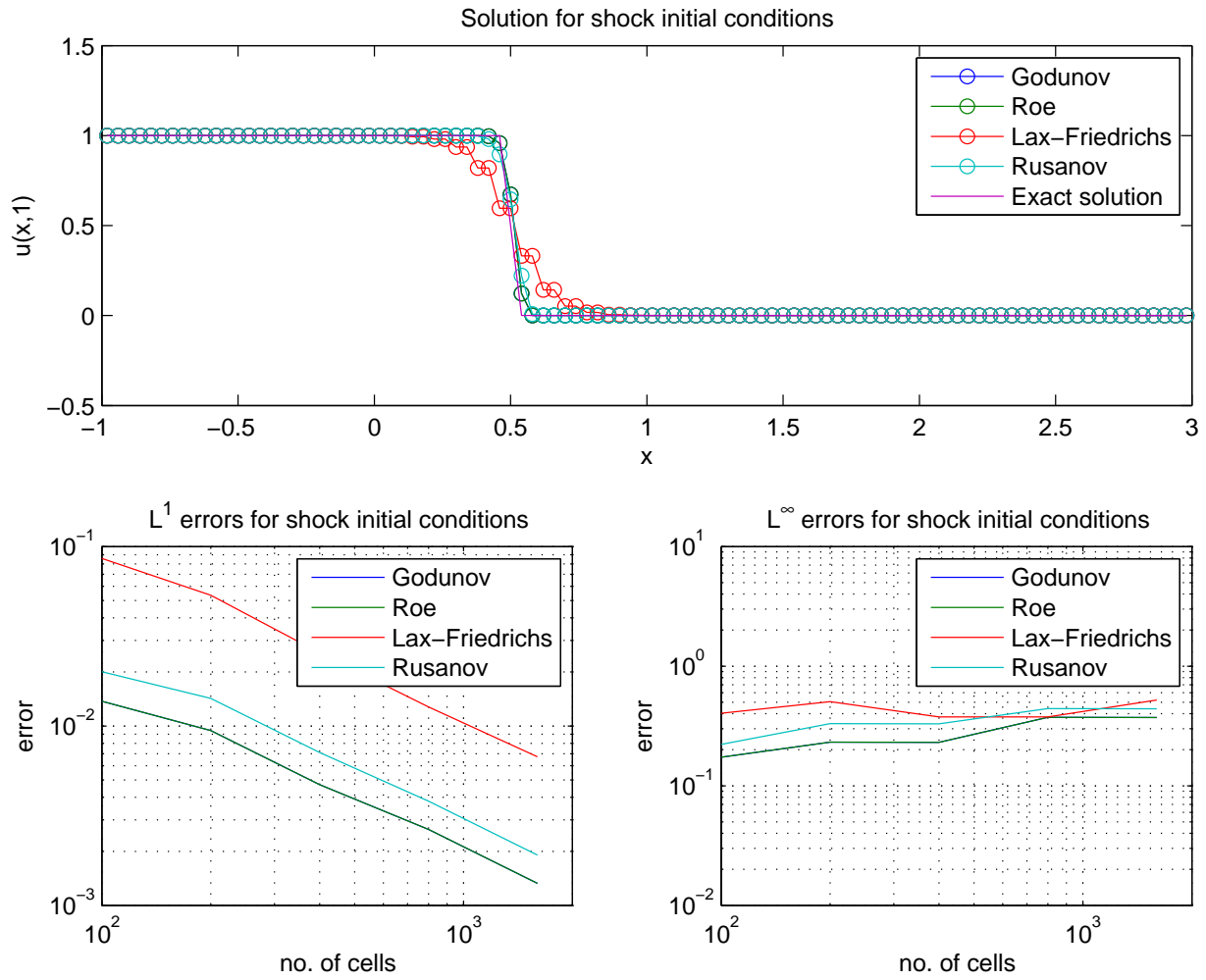


Figure 6.1: Exact solutions and FVM approximations for the shock solution (6.1.2).

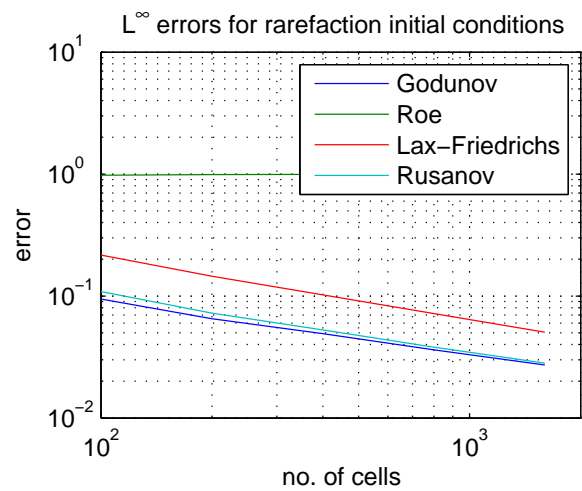
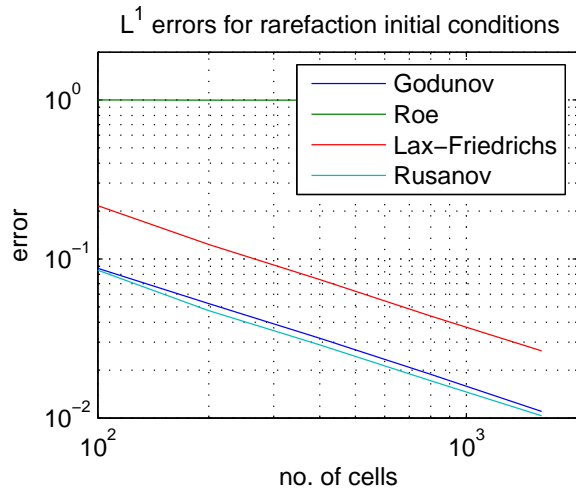
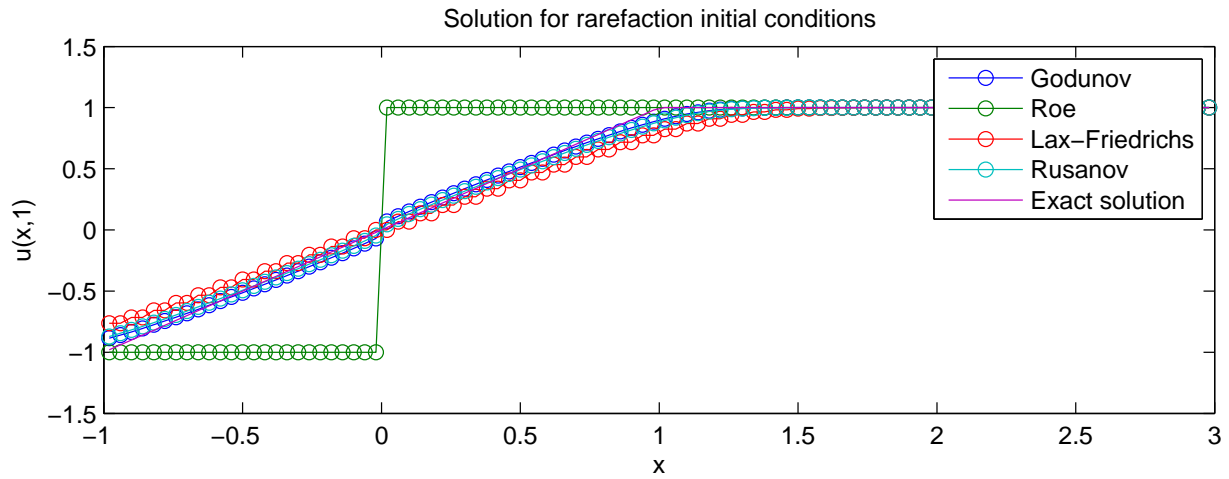


Figure 6.2: Exact solutions and FVM approximations for the rarefaction solution (6.1.3).

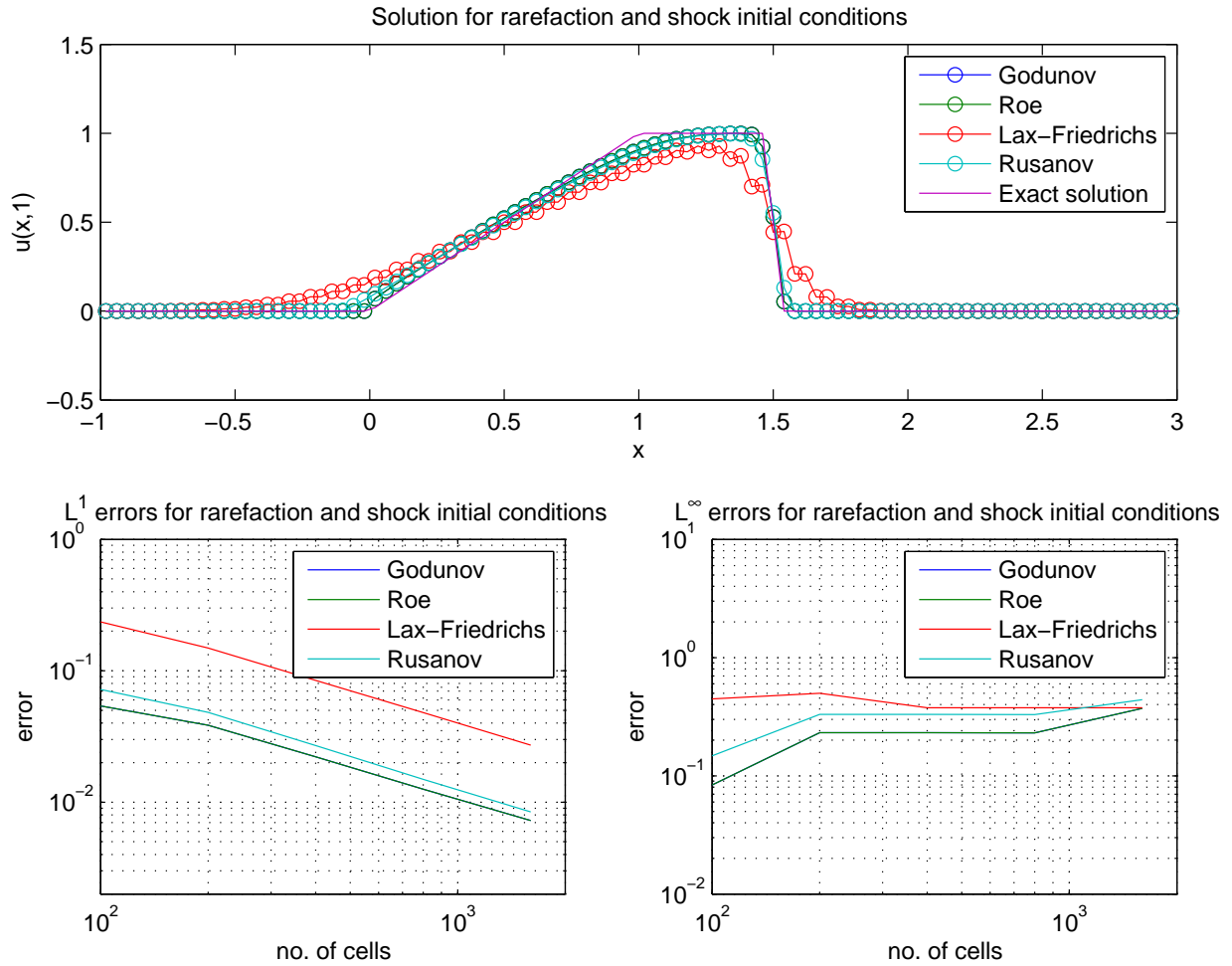


Figure 6.3: Exact solutions and FVM approximations for the rarefaction and shock solution (6.1.4).

Solution:

See the additional codes [6.15](#) - [6.18](#). For the error convergence plots, see [6.1](#) - [6.3](#).

Listing 6.15: Implementation for geterrors

```
1 function [err_L1, err_Li] = geterrors(u_e, f, f_prime,  
   f_type, f_extrema, u0, l, r, T, N )  
2  
3 err_L1 = [];  
4 err_Li = [];  
5  
6 for mesh_size=N  
7     [x, u, dx] = solve_all(f, f_prime, f_type, f_extrema, u0,  
   l, r, T, mesh_size);  
8     u_avg = ones(size(u,1) , 1) * feval(getCellAverages(u_e,  
   dx), x);  
9     err_L1 = [err_L1 get_L1(u, u_avg, dx)]; %#ok<AGROW>  
10    err_Li = [err_Li get_Li(u, u_avg, dx)]; %#ok<AGROW>  
11    display(mesh_size);  
12 end  
13  
14 end
```

Listing 6.16: Implementation for get_L1

```
1 function [ err ] = get_L1( u, u_e, dx )  
2  
3 err = dx*sum(abs(u-u_e), 2);  
4  
5 end
```

Listing 6.17: Implementation for get_Li

```
1 function [ err ] = get_Li( u, u_e, ~ )  
2  
3 err = max(abs(u-u_e), [], 2);  
4  
5 end
```

Listing 6.18: Implementation for getCellAverages

```
1 function [ u_avg ] = getCellAverages( u, dx )  
2  
3 u_avg = @(x) avg_eval(x, u, dx);  
4  
5 end  
6  
7 function u_avg = avg_eval(x, u, dx)  
8
```

```

9 u_avg = zeros(1, length(x));
10
11 for i=1:length(x)
12     u_avg(i) = quad(u, x(i) - dx/2, x(i) + dx/2)/dx;
13 end
14
15 end

```

Problem 6.2 Shallow water equations in two dimensions

Consider the 2-dimensional shallow water equations

$$\begin{cases} h_t + (hu)_x + (hv)_y = 0, \\ (hu)_t + \left(hu^2 + \frac{1}{2}gh^2\right)_x + (huv)_y = -ghb_x, \\ (hv)_t + (huv)_x + \left(hv^2 + \frac{1}{2}gh^2\right)_y = -ghb_y. \end{cases} \quad (6.2.1)$$

Here, for a point $(x, y) \in \mathbb{R}^2$ and time instance t , variable $h(x, y, t)$ denotes the height of the fluid column above the bottom topography $b = b(x, y)$ over which the fluid flows and $(u, v) = (u(x, y, t), v(x, y, t))$ is the vertically averaged (or depth averaged) horizontal fluid velocity field. The constant g denotes the size of the negative vertical acceleration due to gravity and is set to $g = 9.812$ here. For now, we also consider flat bottom topography, i.e. we set $b \equiv \text{const} = 0$, and hence the right hand side of (6.2.1) becomes zero.

Denoting the vectors of conserved variables $(h, hu$ and $hv)$ as $\mathbf{U} = \mathbf{U}(\mathbf{x}, t) : \mathbb{R}^2 \times \mathbb{R}_+ \rightarrow \mathbb{R}^3$, and the directional (in x and y directions) fluxes as $\mathbf{F}, \mathbf{G} : \mathbb{R}^3 \rightarrow \mathbb{R}^3$, i.e.

$$\mathbf{U} = \begin{bmatrix} h \\ hu \\ hv \end{bmatrix}, \quad \mathbf{F} = \begin{bmatrix} hu \\ hu^2 + \frac{1}{2}gh^2 \\ huv \end{bmatrix}, \quad \mathbf{G} = \begin{bmatrix} hv \\ huv \\ hv^2 + \frac{1}{2}gh^2 \end{bmatrix}, \quad (6.2.2)$$

the system (6.2.1) with given initial data \mathbf{U}_0 is rewritten as a *system of conservation laws*,

$$\begin{cases} \mathbf{U}(\mathbf{x}, t)_t + \mathbf{F}(\mathbf{U})_x + \mathbf{G}(\mathbf{U})_y = 0, \\ \mathbf{U}(\mathbf{x}, 0) = \mathbf{U}_0(\mathbf{x}). \end{cases} \quad \mathbf{x} = (x, y) \in \mathbf{D}, \quad t > 0, \quad (6.2.3)$$

where we restrict the computational domain to some bounded Cartesian domain $\mathbf{D} \subset \mathbb{R}^2$.

The maximum directional wave speeds (in directions x and y), corresponding to the maximal eigenvalues of the matrices $\partial \mathbf{F} / \partial \mathbf{U}$ and $\partial \mathbf{G} / \partial \mathbf{U}$, are given by the

$$\lambda_x(\mathbf{U}) = |u| + \sqrt{gh}, \quad \lambda_y(\mathbf{U}) = |v| + \sqrt{gh}. \quad (6.2.4)$$

(6.2a) As a starting point, we first consider the one-dimensional version of the shallow water equations, obtained by setting $v \equiv 0$ in (6.2.1), removing flux \mathbf{G} , and denoting $\mathbf{U} = (h, hu)^\top$,

$$\begin{cases} h_t + (hu)_x = 0, \\ (hu)_t + \left(hu^2 + \frac{1}{2}gh^2\right)_x = 0. \end{cases} \quad (6.2.5)$$

For the finite domain $\mathbf{D} = I_1 = (0, 2)$, implement the Finite Volume solver using the Rusanov numerical flux and the Forward Euler time stepping, i.e.

$$\mathbf{U}_i^{n+1} := \mathbf{U}_i^n - \frac{\Delta t}{\Delta x} (\mathbf{F}_{i+\frac{1}{2}}^n - \mathbf{F}_{i-\frac{1}{2}}^n), \quad (6.2.6)$$

where \mathbf{U}_i^n are the vectors $(h_i^n, (hu)_i^n)^\top$ denoting the cell averages of the solution $\mathbf{U} = (h, hu)^\top$, and $\mathbf{F}_{i+\frac{1}{2}}^n$ are approximated by the Rusanov flux function

$$\mathbf{F}_{i+\frac{1}{2}}^n(\mathbf{U}_i^n, \mathbf{U}_{i+1}^n) \approx \mathbf{F}^{\text{Rus}}(\mathbf{U}_L, \mathbf{U}_R) = \frac{(\mathbf{F}_L + \mathbf{F}_R)}{2} - \frac{\lambda_{\max}}{2}(\mathbf{U}_R - \mathbf{U}_L), \quad (6.2.7)$$

with *local* maximum wave speeds

$$\lambda_{\max} := \max(\lambda_x(\mathbf{U}_L), \lambda_x(\mathbf{U}_R)).$$

The mesh width is $\Delta x = |I_1|/N_x$, where N_x denotes the number of the mesh cells, and the time step size respects the CFL condition (with CFL number C_{CFL} set to 0.9)

$$\Delta t = C_{\text{CFL}} \frac{\Delta x}{\bar{\lambda}_x} \leq \frac{\Delta x}{\bar{\lambda}_x} = \frac{\Delta x}{\max_i \lambda_x(\mathbf{U}_i)}. \quad (6.2.8)$$

Use the “outflow” boundary conditions, which can be easily implemented by extrapolating the values of the variable to the ghost cell i.e. $\mathbf{U}_0^n = \mathbf{U}_1^n$ and $\mathbf{U}_{N+1}^n = \mathbf{U}_N^n$.

Run your code for the “dambreak” initial data given by

$$\mathbf{U}_0(x) = \left(h_0(x), u_0(x) \right)^\top = \begin{cases} (2 - b(x), 0)^\top, & \text{if } x < 1, \\ (1.5 - b(x), 0)^\top, & \text{otherwise,} \end{cases} \quad (6.2.9)$$

for $N_x = 512$ mesh cells up to final time $T = 0.1$ and plot the results (both water column height h and velocity u).

HINT: Consider the evaluations of \mathbf{U} at the cell mid-points x_i for the cell averages \mathbf{U}_i .

HINT: Implement the Rusanov flux (6.2.7) as a separate function; this way you will be able to quickly extend and adapt it for the 2-dimensional FVM solver in the next sub-problem.

HINT: The solution consists of the left-moving rarefaction wave and the right-moving shock wave, as depicted in Figure 6.4.

HINT: It is advisable to implement and use the conversion functions between the *observable* (also called *primitive*) variables (h, u) and the *conserved* variables (h, hu) .

HINT: You might find the MATLAB function `diff` or the Python function `numpy.diff` useful.

HINT: For debugging, it is always a good idea to plot the initial data and the solution after *one* time step, and verify that your numerical flux is correct and provides stable approximations.

Solution:

See codes 6.19 - 6.22, ignoring the code lines dealing with bottom topography, and Figure 6.4 for the approximate FVM solutions.

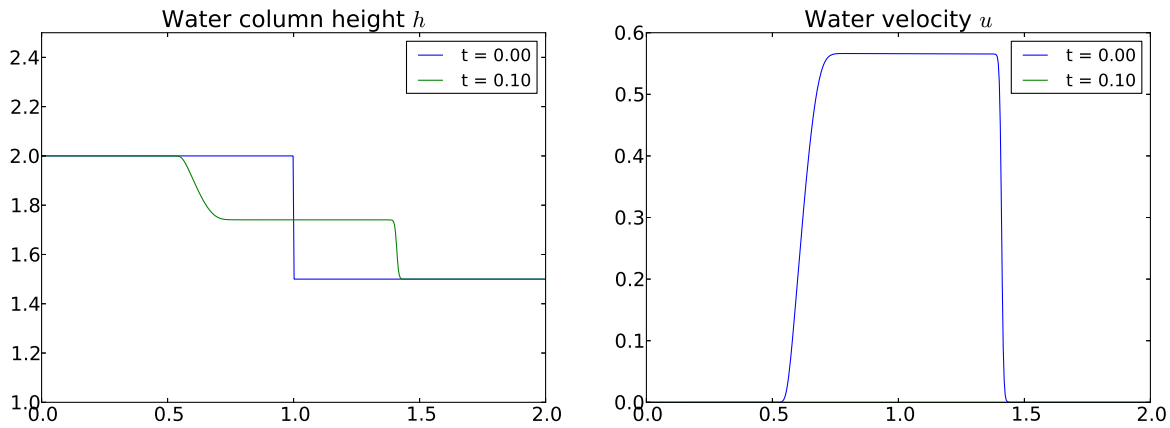


Figure 6.4: FVM approximations of the 1-D shallow water equations (6.2.5) and the dambreak problem (6.2.9) with flat bottom topography $b \equiv 0$.

Listing 6.19: Implementation for shallow_water_1d

```

1  # global imports
2  import numpy
3  import pylab
4
5  # local imports
6  from rusanov import rusanov_flux
7  from compute_flux_1d import compute_flux_1d
8  from variables import ConservedVars, PrimitiveVars
9
10 # plotting configuration
11 import matplotlib
12 matplotlib.rcParams['font.size'] = 18
13 matplotlib.rcParams['legend.fontsize'] = 16
14
15 def shallow_water_1d (D, g, T, h0, u0, F, DF, N, num_flux,
16                       CFL, b):
17
18     # mesh width
19     dx = numpy.fabs ( numpy.diff(D) ) / N
20
21     # cell mid-points
22     xr = numpy.linspace ( D[0] + dx/2, D[1] - dx/2, N)
23
24     # cell averages of the initial data
25     h = h0 (xr)
26     u = u0 (xr)
27
28     # modify 'h', if bottom topography is present
29     if b != None:

```



```

29     h -= b(xr)
30
31     # convert to conserved variables
32     [h, hu] = ConservedVars (h, u)
33
34     # cell averages of the bottom topography (if present)
35     # and the 2nd order accurate approximation of derivative DB
36     of B
37     # where at the boundary (first and last cells), we set DB =
38     0
39     if b != None:
40         B = b(xr)
41         DB = (B[2:] - B[:-2]) / (2 * dx)
42         DB = numpy.insert(DB, 0, [0])
43         DB = numpy.append(DB, [0])
44
45     # Forward Euler time-stepping
46     t = 0
47     while t < T:
48
49         # display current time 't'
50         print t
51
52         # compute the numerical flux
53         [Fh, Fhu, lmax] = compute_flux_1d (h, hu, F, DF, g,
54             num_flux)
55
56         # compute the time step size
57         dt = min (T - t, CFL * dx / lmax)
58
59         # compute source terms, if bottom topography is present
60         if b != None:
61             S = - g * h * DB
62
63         # perform time integration
64         h -= dt/dx * numpy.diff (Fh)
65         hu -= dt/dx * numpy.diff (Fhu)
66
67         # add source terms, if present
68         if b != None:
69             hu += dt * S
70
71         # update the time step size
72         t += dt;
73
74     # return the solution
75     return [xr] + PrimitiveVars (h, hu)

```

```

73
74 if __name__ == '__main__':
75
76     # problem data
77     D = [0, 2]
78     T = 0.1
79     g = 8.912
80     h0 = lambda x : numpy.where (x < 1, 2.0, 1.5)
81     u0 = lambda x : numpy.zeros (x.shape)
82     F = lambda h, u, g : [h*u, h*u*u + 0.5*g*h*h]
83     DF = lambda h, u, g : numpy.fabs(u) + numpy.sqrt(g*h)
84
85     # discretization parameters
86     N = 512
87     CFL = 0.9
88
89     # run the FVM scheme
90     [xr, h, u] = shallow_water_1d (D, g, T, h0, u0, F, DF, N,
91                                     rusanov_flux, CFL, None)
92
93     # plot the results
94     pylab.figure (figsize=(2*8,6))
95     pylab.subplot (121)
96     pylab.plot (xr, h0(xr), label='t = %.2f' % 0)
97     pylab.plot (xr, h, label='t = %.2f' % T)
98     pylab.legend (loc='upper right')
99     pylab.ylim([1, 2.5])
100    pylab.title (r'Water column height h')
101    pylab.subplot (122)
102    pylab.plot (xr, u, label='t = %.2f' % 0)
103    pylab.plot (xr, u0(xr), label='t = %.2f' % T)
104    pylab.legend (loc='upper right')
105    pylab.title (r'Water velocity u')
106    pylab.subplots_adjust(left=0.05,right=0.95)
107    pylab.savefig ('../fig/shallow_water_1d.eps')
108    pylab.show()
109
110    # run the FVM scheme with bottom topography
111    b = lambda x : 0.1*numpy.sin(5*numpy.pi*x) - 0.2*x + 1.4
112    [xr, h, u] = shallow_water_1d (D, g, T, h0, u0, F, DF, N,
113                                    rusanov_flux, CFL, b)
114
115    # plot the results
116    pylab.figure (figsize=(2*8,6))
117    pylab.subplot (121)
118    pylab.plot (xr, h0(xr), label='t = %.2f' % 0)
119    pylab.plot (xr, h + b(xr), label='t = %.2f' % T)

```

```

118 pylab.plot (xr, b(xr), color='k', label='topography')
119 pylab.fill_between(xr, b(xr), 0, color=(0.5, 0.5, 0.5))
120 pylab.legend (loc='upper right')
121 pylab.ylim([1, 2.5])
122 pylab.title (r'Water column height h')
123 pylab.subplot (122)
124 pylab.plot (xr, u, label='t = %.2f' % 0)
125 pylab.plot (xr, u0(xr), label='t = %.2f' % T)
126 pylab.legend (loc='upper right')
127 pylab.title (r'Water velocity u')
128 pylab.subplots_adjust(left=0.05,right=0.95)
129 pylab.savefig ('../fig/shallow_water_1d_topography.eps')
130 pylab.show()

```

Listing 6.20: Implementation for variables

```

1  # convert primitive variables (h, u, v) to conserved
   variables (h, hu, hv)
2  def ConservedVars (h, u, v=None):
3      return [h, h*u, h*v] if v != None else [h, h*u]
4
5  # convert conserved variables (h, hu, hv) to primitive
   variables (h, u, v)
6  def PrimitiveVars (h, hu, hv=None):
7      return [h, hu/h, hv/h] if hv != None else [h, hu/h]

```

Listing 6.21: Implementation for rusanov

```

1  # Rusanov flux
2  def rusanov_flux (FL, FR, lmax, UR, UL):
3      return 0.5 * (FL + FR) - 0.5 * lmax * (UR - UL)

```

Listing 6.22: Implementation for compute_flux_1d

```

1  import numpy
2  from variables import ConservedVars, PrimitiveVars
3
4  # compute all fluxes using the numerical flux 'num_flux'
5  def compute_flux_1d (h, hu, F, DF, g, num_flux):
6
7      # get number of cells and increment by 1
8      N = h.shape[0] + 1
9
10     # allocate arrays for flux
11     Fh = numpy.zeros (N)
12     Fhu = numpy.zeros (N)
13
14     # allocate array for the wave speeds
15     lmax = numpy.zeros (N)

```

```

16
17 # compute primitive variables
18 [h, u] = PrimitiveVars (h, hu)
19
20 # compute fluxes at each cell interface
21 for i in xrange(0, N):
22
23     # set left and right indices
24     # incorporating 'outflow' boundary conditions
25     L = max (0, i - 1)
26     R = min (i, N - 2)
27
28     # compute the maximum wave speed
29     lmax[i] = max ( DF (h[L], u[L], g), DF (h[R], u[R], g) )
30
31     # compute the left and the right fluxes
32     [FhL, FhuL] = F (h[L], u[L], g)
33     [FhR, FhuR] = F (h[R], u[R], g)
34
35     # compute the Rusanov flux
36     Fh[i] = num_flux (FhL, FhR, lmax[i], h [R], h [L])
37     Fhu[i] = num_flux (FhuL, FhuR, lmax[i], hu[R], hu[L])
38
39     # compute the global maximal wave speed
40     lmax = numpy.max(lmax)
41
42     # return the numerical flux and the global maximal wave
43     # speed
44     return [Fh, Fhu, lmax]

```

(6.2b) Extend your code of [subproblem \(6.2a\)](#) to non-constant bottom topography, given by

$$b(x) = 0.1 \sin(5\pi x) - 0.2x + 1.4. \quad (6.2.10)$$

The resulting FVM scheme then also needs to incorporate the source term S_i ,

$$\mathbf{U}_i^{n+1} := \mathbf{U}_i^n - \frac{\Delta t}{\Delta x} (\mathbf{F}_{i+\frac{1}{2}}^n - \mathbf{F}_{i-\frac{1}{2}}^n) + \Delta t \mathbf{S}_i, \quad (6.2.11)$$

which you can approximate by using second order accurate central differences,

$$\mathbf{S}_i = \begin{bmatrix} 0 \\ gh_i \frac{B_{i+1} - B_{i-1}}{2\Delta x} \end{bmatrix}. \quad (6.2.12)$$

In order to deal with the boundary conditions for b , simply set

$$\mathbf{S}_0 = (0, 0)^\top, \quad \mathbf{S}_{N_x} = (0, 0)^\top. \quad (6.2.13)$$

Run your code for the “dambreak” initial data (6.2.9) for $N_x = 512$ mesh cells up to final time $T = 0.1$ and plot the results (plot water surface $h + b$ and bottom topography b in the same plot).

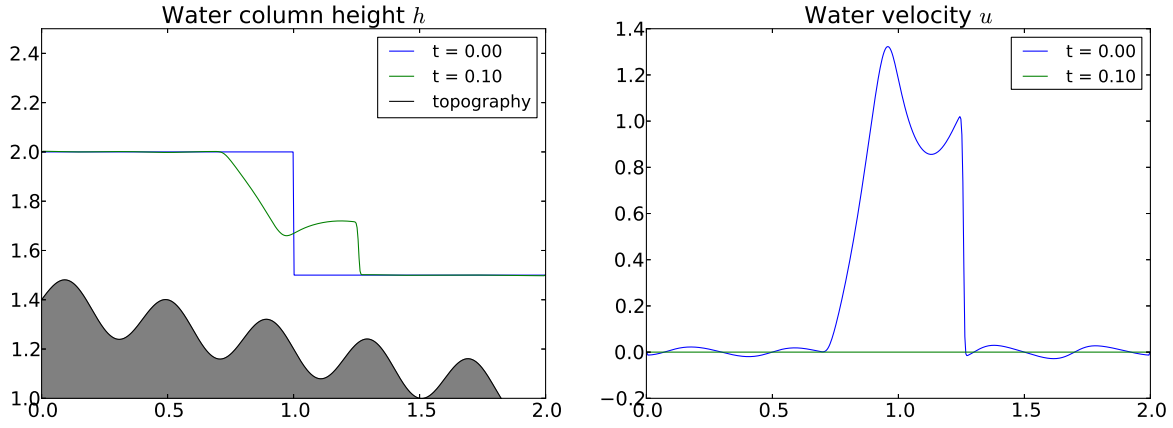


Figure 6.5: FVM approximations of the 1-D shallow water equations (6.2.5) and the dambreak problem (6.2.9) with varying bottom topography (6.2.10).

HINT: Do not forget to *subtract* bottom topography $b(x)$ from $h(x)$ in (6.2.9).

HINT: For plotting, do not forget to *add* bottom topography b to water column h to get the actual water surface.

HINT: The solution, analogously as in subproblem (6.2a), consists of the left-moving rarefaction wave and the right-moving shock wave, as depicted in Figure 6.5.

Solution:

See codes 6.19 - 6.22 and Figure 6.5 for the approximate FVM solutions.

(6.2c) In this sub-problem, we extend the 1-dimensional Finite Volume scheme from the subproblem (6.2a) with flat bottom topography $b \equiv 0$ to the 2-dimensional shallow water equations (6.2.3). Consider the finite domain $\mathbf{D} = I_1 \times I_2 = (0, 2) \times (0, 1)$ and the corresponding uniform axiparallel equidistant mesh with N_x cells in x direction and N_y cells in y direction. Implement the Finite Volume solver using the dimension splitting, Rusanov numerical flux and the Forward Euler time stepping, i.e.

$$\mathbf{U}_{i,j}^{n+1} := \mathbf{U}_{i,j}^n - \frac{\Delta t}{\Delta x} (\mathbf{F}_{i+\frac{1}{2},j}^n - \mathbf{F}_{i-\frac{1}{2},j}^n) - \frac{\Delta t}{\Delta y} (\mathbf{G}_{i,j+\frac{1}{2}}^n - \mathbf{G}_{i,j-\frac{1}{2}}^n), \quad (6.2.14)$$

where $\mathbf{U}_{i,j}^n$ are the vectors $(h_{i,j}^n, (hu)_{i,j}^n, (hv)_{i,j}^n)^\top$ denoting the cell averages of the solution $\mathbf{U} = (h, hu, hv)^\top$, and the *directional* fluxes $\mathbf{F}_{i+\frac{1}{2},j}^n$ and $\mathbf{G}_{i,j+\frac{1}{2}}^n$ are approximated by the Rusanov flux function from (6.2.7) in the *corresponding* direction, i.e.

$$\mathbf{F}_{i+\frac{1}{2},j}^n \approx \mathbf{F}^{\text{Rus}}(\mathbf{F}_{i,j}^n, \mathbf{F}_{i+1,j}^n), \quad \mathbf{G}_{i,j+\frac{1}{2}}^n \approx \mathbf{F}^{\text{Rus}}(\mathbf{F}_{i,j}^n, \mathbf{F}_{i,j+1}^n). \quad (6.2.15)$$

The time step size respects the 2-dimensional version of the CFL condition (6.2.8)

$$\Delta t = C_{\text{CFL}} \left(\frac{\bar{\lambda}_x}{\Delta x} + \frac{\bar{\lambda}_y}{\Delta y} \right)^{-1}, \quad \bar{\lambda}_x = \max_{i,j} \lambda_x(\mathbf{U}_{i,j}), \quad \bar{\lambda}_y = \max_{i,j} \lambda_y(\mathbf{U}_{i,j}). \quad (6.2.16)$$

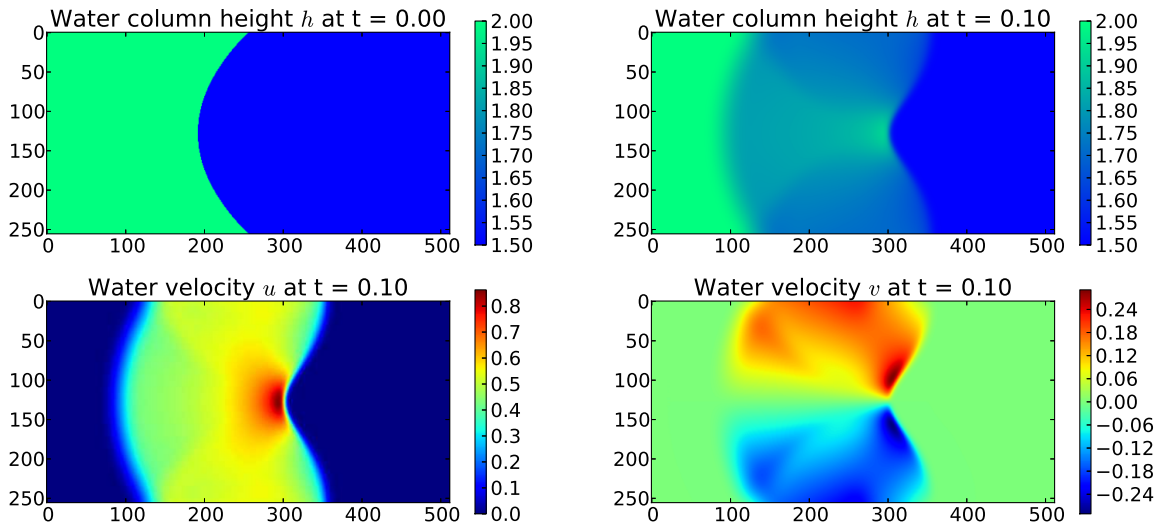


Figure 6.6: FVM approximations of the 2-D shallow water equations (6.2.1) and the dambreak problem (6.2.17).

Run your code for the 2-dimensional “dambreak” initial data given by

$$\mathbf{U}_0(x, y) = \left(h_0(x, y), u_0(x, y), v_0(x, y) \right)^\top = \begin{cases} (2 - b(x, y), 0, 0)^\top, & \text{if } x < (y - 0.5)^2 + 0.75, \\ (1.5 - b(x, y), 0, 0)^\top, & \text{otherwise,} \end{cases} \quad (6.2.17)$$

for $N_x = 64$ and $N_y = 32$ mesh cells up to final time $T = 0.1$ and plot the results (both water column height h and velocities u and v). You might also want (this is optional) to generate a 3D plot with the water surface h being plotted on the z -axis. Use the “outflow” boundary conditions, which can be easily implemented by extrapolating the values of the variable to the ghost cell, i.e.

$$\mathbf{U}_{0,j}^n = \mathbf{U}_{1,j}^n, \quad \mathbf{U}_{N+1,j}^n = \mathbf{U}_{N,j}^n \quad (6.2.18)$$

in the x direction (i.e. for the evaluations of the flux \mathbf{F}), and

$$\mathbf{U}_{i,0}^n = \mathbf{U}_{i,1}^n, \quad \mathbf{U}_{i,N+1}^n = \mathbf{U}_{i,N}^n. \quad (6.2.19)$$

in the y direction (i.e. for the evaluations of the flux \mathbf{G}).

HINT: You might find the MATLAB function `meshgrid` or the Python function `meshgrid` (with option `indexing='ij'`) from `numpy` useful.

HINT: If you are eager to test the limits of your machine and prepared to wait longer (an hour or so), run your 2-D code with $N_x = 512$ and $N_y = 256$. For even larger resolutions (and hence smaller errors), parallel implementations are usually needed.

HINT: For code testing purposes, the reference plots (using $N_x = 512$ and $N_y = 256$) are given in Figure 6.6.

Solution:

See codes 6.23 - 6.24 and Figure 6.6 for the approximate FVM solutions.

Listing 6.23: Implementation for shallow_water_2d

```

1  # global imports
2  import numpy
3  import pylab
4
5  # local imports
6  from rusanov import rusanov_flux
7  from compute_flux_2d import compute_flux_2d
8  from variables import ConservedVars, PrimitiveVars
9
10 # plotting configuration
11 import matplotlib
12 matplotlib.rcParams['font.size'] = 18
13 matplotlib.rcParams['legend.fontsize'] = 16
14
15 # constants for 'x' and 'y' directions
16 X = 0
17 Y = 1
18
19 def shallow_water_2d (D, g, T, h0, u0, v0, F, G, DF, DG, N,
    num_flux, CFL):
20
21     # mesh widths
22     dx = numpy.fabs ( numpy.diff(D[X]) ) / N[X]
23     dy = numpy.fabs ( numpy.diff(D[Y]) ) / N[Y]
24
25     # cell mid-points
26     xr = numpy.linspace (D[X][0] + dx/2, D[X][1] - dx/2, N[X])
27     yr = numpy.linspace (D[Y][0] + dy/2, D[Y][1] - dy/2, N[Y])
28
29     # generate the mesh grid
30     [xg, yg] = numpy.meshgrid(xr, yr, indexing='ij')
31
32     # cell averages of the initial data
33     h = h0 (xg, yg)
34     u = u0 (xg, yg)
35     v = v0 (xg, yg)
36     [h, hu, hv] = ConservedVars (h, u, v)
37
38     # Forward Euler time-stepping
39     t = 0
40     while t < T:
41
42         # display current time 't'
43         print t
44
45         # compute the numerical fluxes

```

```

46     [Fh, Fhu, Fhv, lmax_x] = compute_flux_2d (h, hu, hv, F,
47         DF, g, num_flux, X)
48
49     [Gh, Ghu, Ghv, lmax_y] = compute_flux_2d (h, hu, hv, G,
50         DG, g, num_flux, Y)
51
52     # compute the time step size
53     dt = min ( T - t, CFL * (lmax_x / dx + lmax_y / dy)**(-1)
54         )
55
56     # perform time integration
57     h -= dt/dx * numpy.diff (Fh, axis=X) + dt/dy *
58         numpy.diff (Gh, axis=Y)
59     hu -= dt/dx * numpy.diff (Fhu, axis=X) + dt/dy *
60         numpy.diff (Ghu, axis=Y)
61     hv -= dt/dx * numpy.diff (Fhv, axis=X) + dt/dy *
62         numpy.diff (Ghv, axis=Y)
63
64     # update the time step size
65     t += dt;
66
67     # return the solution
68     return [xg, yg] + PrimitiveVars (h, hu, hv)
69
70 if __name__ == '__main__':
71
72     # problem data
73     D = [[0, 2], [0, 1]]
74     T = 0.1
75     g = 8.912
76     h0 = lambda x, y : numpy.where (x < (y - 0.5)**2 + 0.75,
77         2.0, 1.5)
78     u0 = lambda x, y : numpy.zeros (x.shape)
79     v0 = lambda x, y : numpy.zeros (x.shape)
80     F = lambda h, u, v, g : [h*u, h*u*u + 0.5*g*h*h, h*u*v]
81     G = lambda h, u, v, g : [h*v, h*u*v, h*v*v + 0.5*g*h*h]
82     DF = lambda h, u, v, g : numpy.fabs(u) + numpy.sqrt(g*h)
83     DG = lambda h, u, v, g : numpy.fabs(v) + numpy.sqrt(g*h)
84
85     # discretization parameters
86     #N = [512, 256]
87     N = [64, 32]
88     CFL = 0.9
89
90     # run the FVM scheme
91     [xg, yg, h, u, v] = shallow_water_2d (D, g, T, h0, u0, v0,
92         F, G, DF, DG, N, rusanov_flux, CFL)

```



```

85     # == plot the results
86
87     # create figure of a desired size
88     pylab.figure (figsize=(2*8,2*4))
89
90     pylab.subplot (221)
91     pylab.imshow (h0(xg,yg).T)
92     pylab.colorbar()
93     pylab.title (r'Water column height  $h$  at  $t = %.2f$ ' % 0)
94
95     # choose color scheme
96     pylab.winter()
97
98     pylab.subplot (222)
99     pylab.imshow (h.T)
100    pylab.colorbar()
101    pylab.title (r'Water column height  $h$  at  $t = %.2f$ ' % T)
102
103    pylab.subplot (223)
104    pylab.imshow (u.T)
105    pylab.colorbar()
106    pylab.title (r'Water velocity  $u$  at  $t = %.2f$ ' % T)
107
108    # choose color scheme
109    pylab.jet()
110
111    pylab.subplot (224)
112    pylab.imshow (v.T)
113    pylab.colorbar()
114    pylab.title (r'Water velocity  $v$  at  $t = %.2f$ ' % T)
115
116    # adjust the margins, save figure and display in screen
117    pylab.subplots_adjust(left=0.05,right=0.95)
118    pylab.savefig ('../fig/shallow_water_2d.eps')
119    pylab.show()

```

Listing 6.24: Implementation for compute_flux_2d

```

1  import numpy
2  from variables import ConservedVars, PrimitiveVars
3
4  # constants for 'x' and 'y' directions
5  X = 0
6  Y = 1
7
8  # compute all fluxes using the numerical flux 'num_flux' (in
   direction 'd')
9  def compute_flux_2d (h, hu, hv, F, DF, g, num_flux, d):

```

```

10
11 # get number of cells (in each direction)
12 N = h.shape
13
14 # for fluxes, increment the direction 'd' by 1
15 NF = list(N[:])
16 NF[d] += 1
17
18 # allocate arrays for fluxes
19 Fh = numpy.zeros (NF)
20 Fhu = numpy.zeros (NF)
21 Fhv = numpy.zeros (NF)
22
23 # allocate array for the wave speeds
24 lmax = numpy.zeros (NF)
25
26 # compute primitive variables
27 [h, u, v] = PrimitiveVars (h, hu, hv)
28
29 # compute fluxes at each cell interface (in direction 'd')
30 for i in xrange(0, NF[X]):
31     for j in xrange(0, NF[Y]):
32
33         # set left and right index pairs
34         # incorporating 'outflow' boundary conditions
35         Li = max (0, i - (1 if d == X else 0) )
36         Lj = max (0, j - (1 if d == Y else 0) )
37         Ri = min (i, N[X] - 1)
38         Rj = min (j, N[Y] - 1)
39
40         # compute the maximum wave speed
41         lmax[i][j] = max ( DF (h[Li][Lj], u[Li][Lj], v[Li][Lj],
42                                g), DF (h[Ri][Rj], u[Ri][Rj], v[Ri][Rj], g) )
43
44         # compute the left and the right fluxes
45         [FhL, FhuL, FhvL] = F (h[Li][Lj], u[Li][Lj], v[Li][Lj],
46                                g)
47         [FhR, FhuR, FhvR] = F (h[Ri][Rj], u[Ri][Rj], v[Ri][Rj],
48                                g)
49
50         # compute the Rusanov flux
51         Fh [i][j] = num_flux (FhL, FhR, lmax[i][j], h
52                                [Ri][Rj], h [Li][Lj])
53         Fhu [i][j] = num_flux (FhuL, FhuR, lmax[i][j],
54                                hu[Ri][Rj], hu[Li][Lj])
55         Fhv [i][j] = num_flux (FhvL, FhvR, lmax[i][j],
56                                hv[Ri][Rj], hv[Li][Lj])

```

```
51 |  
52 | # compute the global maximal wave speed  
53 | lmax = numpy.max(lmax)  
54 |  
55 | # return the numerical flux and the global maximal wave  
56 | speed  
   | return [Fh, Fhu, Fhv, lmax]
```

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