

In [1]:

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Chapter 31-7: Hyperbolic PDEs Using Finite Difference Method.

In mathematics, a hyperbolic partial differential equation of order n is a partial differential equation (PDE) that, roughly speaking, has a well-posed initial value problem for the first $n - 1$ derivatives. More precisely, the Cauchy problem can be locally solved for arbitrary initial data along any non-characteristic hypersurface. Many of the equations of mechanics are hyperbolic, and so the study of hyperbolic equations is of substantial contemporary interest. The model hyperbolic equation is the wave equation. In one spatial dimension, this is

$$\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2}$$

The equation has the property that, if u and its first time derivative are arbitrarily specified initial data on the line $t = 0$ (with sufficient smoothness properties), then there exists a solution for all time t .

The solutions of hyperbolic equations are "wave-like". If a disturbance is made in the initial data of a hyperbolic differential equation, then not every point of space feels the disturbance at once. Relative to a fixed time coordinate, disturbances have a finite propagation speed. They travel along the characteristics of the equation. This feature qualitatively distinguishes hyperbolic equations from elliptic partial differential equations and parabolic partial differential equations. A perturbation of the initial (or boundary) data of an elliptic or parabolic equation is felt at once by essentially all points in the domain.

A finite difference is a mathematical expression of the form $f(x + b) - f(x + a)$. If a finite difference is divided by $b - a$, one gets a difference quotient. The approximation of derivatives by finite differences plays a central role in finite difference methods for the numerical solution of differential equations, especially boundary value problems.

The difference operator, commonly denoted Δ , is the operator that maps a function f to the function $\Delta[f]$ defined by

$$\Delta[f](x) = f(x + 1) - f(x)$$

A difference equation is a functional equation that involves the finite difference operator in the same way as a differential equation involves derivatives. There are many similarities between difference equations and differential equations, specially in the solving methods. Certain recurrence relations can be written as difference equations by replacing iteration notation with finite differences.

From Wikipedia, the free encyclopedia

The Lax–Wendroff method, named after Peter Lax and Burton Wendroff, is a numerical method for the solution of hyperbolic partial differential equations, based on finite differences. It is second-order accurate in both space and time. This method is an example of explicit time integration where the function that defines the governing equation is evaluated at the current time.

6. Solve Burgers equation and provide a plot of the resulting solution.

In []:

```
1 clear all
2 close all
3
4 %-----
5 % animation of the solution to burgers equation
6 % using different finite-difference methods
7 %-----
8
9 r=0.50;
10
11 N=2*2^32;
12 a=-2;b= 2;
13 Dx=(b-a)/N;
14
15 Dt=r*Dx;
16
17 method=1; % Lax
18 method=2; % Lax_Wendroff
19
20 %---
```

```

21 % initial condition
22 %----
23
24 for i=1:N+1
25     x(i)=a+(i-1)*Dx;
26     f(i) = exp(-x(i)*x(i));
27     q(i) = 0.5*f(i)^2;
28 end
29
30
31 %-----
32 for step=1:30000
33
34     for i=2:N
35         if(method==1)%-- Lax
36             fnew(i) = 0.5*(1+r*f(i-1))*f(i-1)+0.5*(1-r*f(i+1))*f(i+1);
37         elseif(method==2)%-- Lax-Wendroff
38             A = f(i-1)+f(i);
39             B = f(i-1)+2*f(i)+f(i+1);
40             C = f(i)+f(i+1);
41             fnew(i) = f(i) + 0.5*r*(q(i-1)-q(i+1)) + 0.25*r*r*(A*q(i-1)-B*q(i)+C*q(i+1));
42         end
43     %---
44 end
45
46 f(2:N)=fnew(2:N);
47
48 for i=1:N+1
49     q(i) = 0.5*f(i)^2;
50 end
51
52 if(step==1)
53     Handle1 = plot(x,f,'o-');
54     set(Handle1, 'erasemode', 'xor');
55     set(gca,'fontsize',15)
56     axis([a b 0.0 1.2])
57     xlabel('x','fontsize',15)
58     ylabel('f','fontsize',15)
59 else
60     set(Handle1,'XData',x,'YData',f);
61     pause(0.1)
62     drawnow
63 end
64
65 end
66
67

```

7. Use the Lax-Wendroff scheme to solve the hyperbolic equation

$$ut + ux = 0 \quad 0 \leq x \leq 1, \quad 0 < t < 0.5$$

with proper boundary condition at $x = 0$ and initial condition:

$$u(x, 0) = \exp(-c(x - 0.5)^2) \quad 0 \leq x \leq 1$$

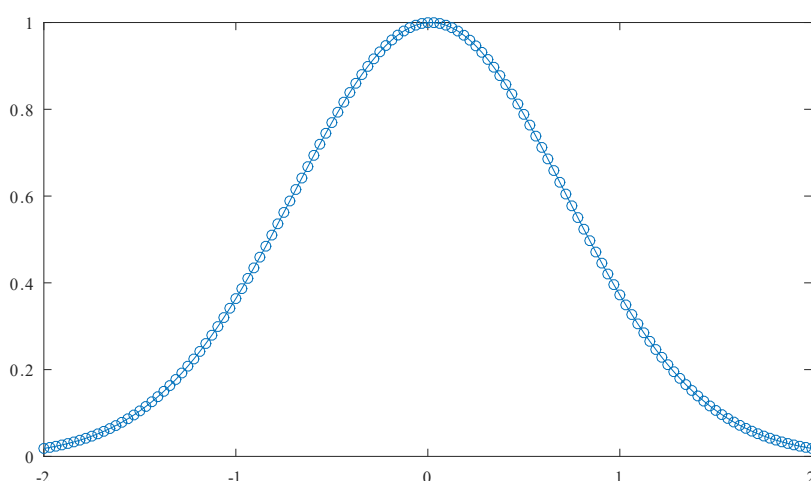
such that the analytic solution is a smooth Gaussian wave

$$u(x, t) = \exp(-c(x - t - 0.5)^2)$$

where $c > 0$ is a constant determining the narrowness of the wave.

(The larger c is, the narrower the wave will be.)

The close correlation shown by the tracks of the two marker symbols testifies that in this case the two methods employed produce essentially identical results.



The example shown above uses the Lax-Wendroff scheme to solve the hyperbolic equation $u_t + ux = 0$ $0 \leq x \leq 1$, $0 < t < 0.5$, with proper boundary condition at $x = 0$ and initial condition: $u(x, 0) = \exp(-c(x - 0.5)^2)$, $0 \leq x \leq 1$, such that the analytic solution is a smooth Gaussian wave $u(x, t) = \exp(-c(x - t - 0.5)^2)$, where $c > 0$ is a constant determining the narrowness of the wave. The larger c is, the narrower the wave will be.

Following is another (Matlab) example of Lax-Wendroff on a hyperbolic equation, using finite differences.

Show an example using the Lax-Wendroff scheme to solve the hyperbolic equation

$$u_t + ux = 0 \quad 0 \leq x \leq 1, 0 < t < 0.5$$

with proper boundary condition at $x = 0$, and initial condition:

$$u(x, 0) = \exp(-c(x - 0.5)^2), \quad 0 \leq x \leq 1$$

such that the analytic solution is a smooth Gaussian wave

$$u(x, t) = \exp(-c(x - t - 0.5)^2)$$

where $c > 0$ is a constant determining the narrowness of the wave. (The larger c is, the narrower the wave will be.)

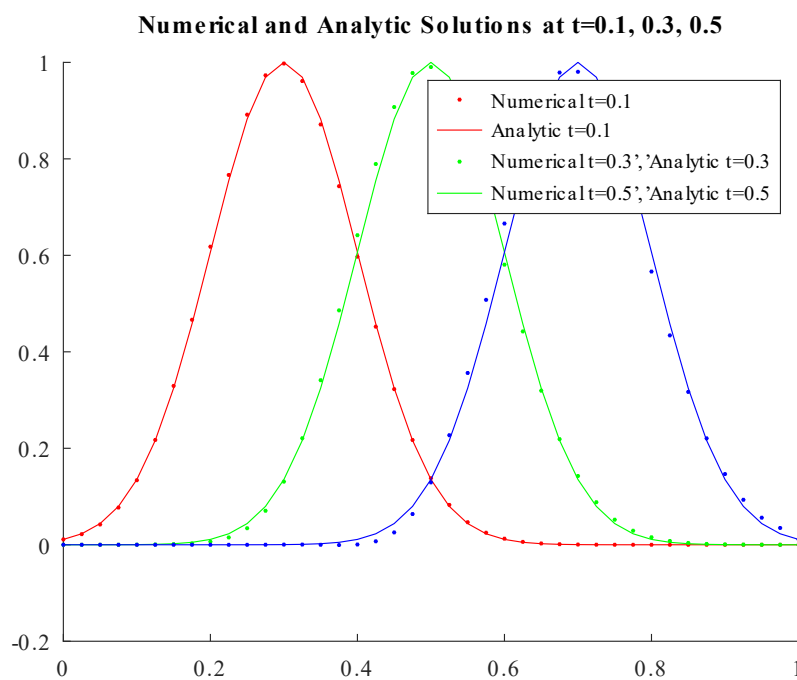
```
In [ ]: 1 %-----
2 % hyper1d.m:
3 % use Lax-Wendroff scheme to solve the hyperbolic equation
4 %  $u_t(x,t) + u_x(x,t) = 0$ ,  $x_l < x < x_r$ ,  $0 < t < t_f$ 
5 %  $u(x, 0) = f(x)$ ,  $x_l < x < x_r$ 
6 %  $u(0, t) = g(t)$ ,  $0 < t < t_f$ 
7 %% A special case is choosing f and g properly such that the
8 % The analytic solution is:
9 %  $u(x,t) = f(x-t) = \exp(-10(x-t-0.2)^2)$ 
10 %-----
11 clear all; % clear all variables in memory
12  $x_l=0$ ;  $x_r=1$ ; % x domain [ $x_l, x_r$ ]
13 J = 40; % J: number of division for x
14 dx = ( $x_r - x_l$ ) / J; % dx: mesh size
15  $t_f = 0.5$ ; % final simulation time
16 Nt = 50; % Nt: number of time steps
17 dt =  $t_f / Nt$ ;
18 c = 50; % parameter for the solution
19 mu = dt/dx;
20 if mu > 1.0 % make sure dt satisfy stability condition
21     error('mu should < 1.0!')
22 end
23 % Evaluate the initial conditions
24 x =  $x_l$  : dx :  $x_r$ ; % generate the grid point
25 f =  $\exp(-c*(x-0.2)^2)$ ; % dimension f(1:J+1)
26 % store the solution at all grid points for all time steps
27 u = zeros(J+1,Nt);
28 % Find the approximate solution at each time step
29 for n = 1:Nt
30     t = n*dt; % current time
31     gl =  $\exp(-c*(x_l-t-0.2)^2)$ ; % BC at left side
32     gr =  $\exp(-c*(x_r-t-0.2)^2)$ ; % BC at right side
33     if n==1 % first time step
34         for j=2:J % interior nodes
35             u(j,n) = f(j) - 0.5*mu*(f(j+1)-f(j-1)) + ...
36                     0.5*mu^2*(f(j+1)-2*f(j)+f(j-1));
37         end
38         u(1,n) = gl; % the left-end point
39         u(J+1,n) = gr; % the right-end point
40     else
41         for j=2:J % interior nodes
42             u(j,n) = u(j,n-1) - 0.5*mu*(u(j+1,n-1)-u(j-1,n-1)) + ...
43                     0.5*mu^2*(u(j+1,n-1)-2*u(j,n-1)+u(j-1,n-1));
44         end
45         u(1,n) = gl; % the left-end point
46         u(J+1,n) = gr; % the right-end point
47     end
48 % calculate the analytic solution
49 for j=1:J+1
50      $x_j = x_l + (j-1)*dx$ ;
51     u_ex(j,n) =  $\exp(-c*(x_j-t-0.2)^2)$ ;
52 end
53 end
54
55 % plot the analytic and numerical solution at different times
56 figure;
```

```

57 hold on;
58 n=10;
59 plot(x,u(:,n),'r.',x,u_ex(:,n),'r-'); % r for red
60 n=30;
61 plot(x,u(:,n),'g.',x,u_ex(:,n),'g-');
62 n=50;
63 plot(x,u(:,n),'b.',x,u_ex(:,n),'b-');
64 legend('Numerical t=0.1','Analytic t=0.1',...
65        'Numerical t=0.3','Analytic t=0.3',...
66        'Numerical t=0.5','Analytic t=0.5');
67 title('Numerical and Analytic Solutions at t=0.1, 0.3, 0.5');
68
69

```

In the following plot, the analytical track does not adhere as closely as before to the numerical track.



6. Develop an upwind scheme (within finite differences) for the equation $u_t + au_x = 0$.

Below is a plot of the initial and consecutive approximation of the upwinding method for an advection equation. The time step is $\Delta t = h$ and the scheme is stable.

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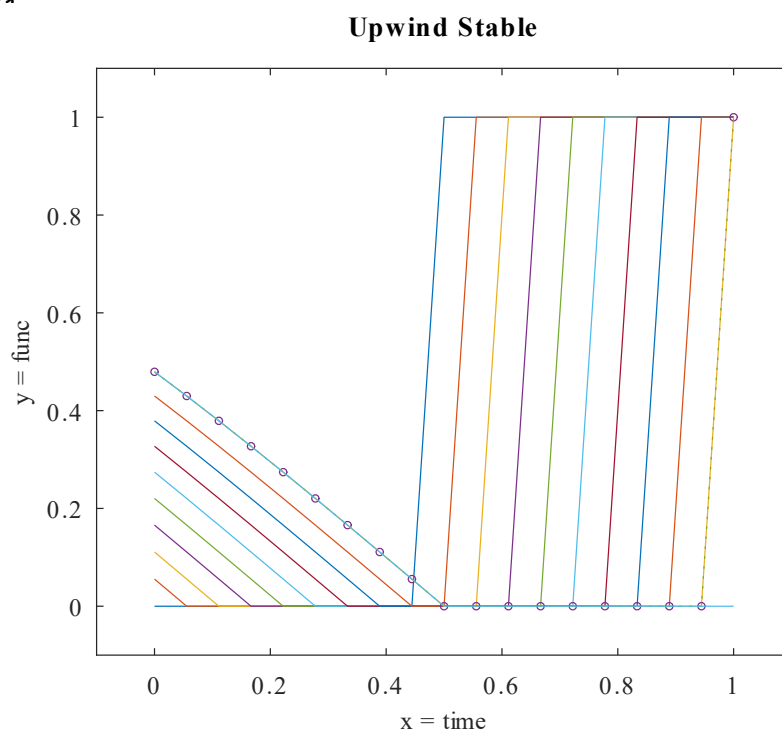
In [ ]: 1 clear; close all
2
3 a = 0; b=1; tfinal = 0.5
4
5 m = 20;
6 m=18;
7
8 h = (b-a)/m;
9 k = h; mu = k/h;
10
11 t = 0;
12 n = fix(tfinal/k);
13 y1 = zeros(m+1,1); y2=y1; x=y1;
14
15 for i=1:m+1,
16     x(i) = a + (i-1)*h;
17     y1(i) = uexact(t,x(i));
18     y2(i) = 0;
19 end
20
21 figure(1); plot(x,y1); hold
22 axis([-0.1 1.1 -0.1 1.1]);
23 title('Upwind Stable');
24 xlabel('x = time','fontsize',10)
25 ylabel('y = func','fontsize',10)
26
27
28 t = 0;
29 for j=1:n,
30
31     y1(1)=bc(t); y2(1)=bc(t+k);
32 % for i=2:m+1
33 %     y2(i) = y1(i) - mu*(y1(i)-y1(i-1));
34     for i=2:m
35 %         y2(i) = y1(i) - mu*(y1(i+1)-y1(i-1))/2;
36         y2(i) = 0.5*(y1(i+1)+y1(i-1)) - mu*(y1(i+1)-y1(i-1))/2;
37     end

```

```

38     i = m+1;
39     y2(i) = y1(i) - mu*(y1(i)-y1(i-1) );
40
41     t = t + k;
42     y1 = y2;
43
44     plot(x,y2); pause(0.5)
45
46 end
47
48 plot(x,y2,'o', "markersize", 3)
49
50 u_e = zeros(m+1,1);
51 for i=1:m+1
52     u_e(i) = uexact(t,x(i));
53 end
54
55 max(abs(u_e-y2))
56
57 plot(x,y2,':',x,u_e)
58
59 figure(2); plot(x,u_e-y2)
60
61

```



Below is a plot of the initial and consecutive approximation of the upwinding method for the same advection equation. The time step is $\Delta t = 1.5h$ and the scheme is unstable, which leads to a blowing-up quantity.

```

In [ ]: 1 clear; close all
2
3     a = 0; b=1; tfinal = 0.5; % Input the domain and final time.
4
5     m = 20; h = (b-a)/m; k = h; mu = 1.5*k/h; % Set mesh and time step.
6
7     t = 0; n = fix(tfinal/k); % Find the number of time steps.
8     y1 = zeros(m+1,1); y2=y1; x=y1;
9
10    figure(1); hold
11    %axis([-0.1 1.1 -0.1 1.1]);
12
13    for i=1:m+1,
14        x(i) = a + (i-1)*h;
15        y1(i) = uexact(t,x(i)); % Initial data
16        y2(i) = 0;
17    end
18
19    % Time marching
20
21    for j=1:n,
22        y1(1)=bc(t); y2(1)=bc(t+k);
23        for i=2:m+1
24            y2(i) = y1(i) - mu*(y1(i)-y1(i-1) );
25        end
26        t = t + k;
27        y1 = y2;
28        plot(x,y2); pause(0.5); % Add the solution plot to the history.
29    end
30
31    u_e = zeros(m+1,1);
32    for i=1:m+1
33        u_e(i) = uexact(t,x(i));
34    end
35
36    max(abs(u_e-y2))

```

```
37
38 plot(x,y2,'o',"markersize", 3,x,u_e )
39 title('Upwind Unstable');
40 xlabel('x = time','fontsize',10)
41 ylabel('y = u_e','fontsize',10)
42
```

