In [1]: %autosave 0

Autosave disabled

(Custom CSS files are not reliable for controlling Jupyter font style. To establish the same appearance as the original notebook, depend on the browser to control the font, by setting the desired font faces in the browser settings. For example, Chrome 135 or Firefox 134 can do this. In this notebook series, Bookerly font is for markdown and Monaco is for code.)

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 [ ]: |clear all
        close all
        % animation of the solution to burgers equation
        % using different finite-difference methods
        r=0.50;
        N=2*2*32;
        a=-2;b=2;
        Dx=(b-a)/N;
        Dt=r*Dx;
        method=1; % Lax
        method=2; % Lax_Wendroff
        %---
        % initial condition
        %----
        for i=1:N+1
         x(i)=a+(i-1)*Dx;
         f(i) = \exp(-x(i)*x(i));
         q(i) = 0.5*f(i)^2;
        end
        %-----
        for step=1:30000
        for i=2:N
         if(method==1)%--- Lax
          fnew(i) = 0.5*(1+r*f(i-1))*f(i-1)+0.5*(1-r*f(i+1))*f(i+1);
         elseif(method==2)%--- Lax-Wendroff
           A = f(i-1)+f(i);
            B = f(i-1)+2*f(i)+f(i+1);
           C = f(i)+f(i+1);
            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));
         end
        %---
        end
        f(2:N)=fnew(2:N);
        for i=1:N+1
         q(i) = 0.5*f(i)^2;
        if(step==1)
          Handle1 = plot(x, f, 'o-');
          set(Handle1, 'erasemode', 'xor');
          set(gca, 'fontsize', 15)
          axis([a b 0.0 1.2])
          xlabel('x','fontsize',15)
ylabel('f','fontsize',15)
        else
          set(Handle1, 'XData', x, 'YData', f);
          pause(0.1)
          drawnow
        end
         end
```

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

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

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

$$u(x, 0) = \exp(-c(x - 0.5)2)$$
 $0 \le x \le 1$

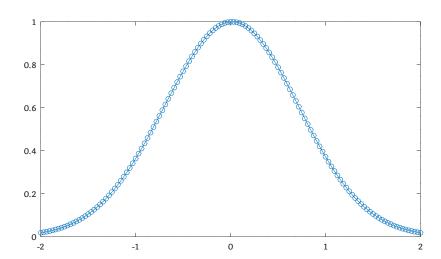
such that the analytic solution is a smooth Gaussian wave

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

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 ut + ux = $0.0 \le x \le 1$, $0.5 \le x \le 1$, $0.5 \le x \le 1$, 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

$$ut + ux = 00 \le x \le 1, 0 < t < 0.5$$

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

$$u(x, 0) = exp(-c(x - 0.5)2), 0 \le x \le 1$$

such that the analytic solution is a smooth Gaussian wave

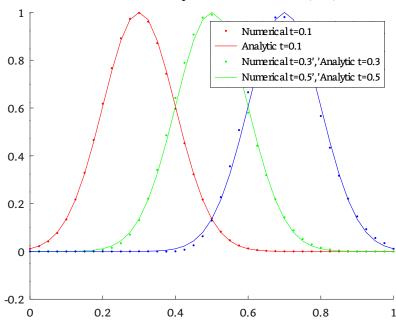
$$u(x, t) = exp(-c(x - t - 0.5)2)$$

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

```
In [ ]: |%-----
        % hyper1d.m:
        % use Lax-Wendroff scheme to solve the hyperbolic equation
        u_t(x,t) + u_x(x,t) = 0, xl < x < xr, 0 < t < tf
        % u(x, 0) = f(x), xl < x < xr
        % u(0, t) = g(t), 0 < t < tf
        %% A special case is choosing f and g properly such that the
        % The analytic solution is:
        % u(x,t) = f(x-t) = e^{-10(x-t-0.2)}
        clear all; % clear all variables in memory
        xl=0; xr=1; % x domain [xl,xr]
        J = 40; % J: number of division for x
        dx = (xr-xl) / J; % dx: mesh size
        tf = 0.5; % final simulation time
        Nt = 50; % Nt: number of time steps
        dt = tf/Nt;
        c = 50; % parameter for the solution
        mu = dt/dx;
        if mu > 1.0 % make sure dt satisy stability condition
          error('mu should < 1.0!')
        end
        % Evaluate the initial conditions
        x = xl : dx : xr; % generate the grid point
        f = \exp(-c^*(x-0.2).^2); % dimension f(1:J+1)
        % store the solution at all grid points for all time steps
        u = zeros(J+1,Nt);
        % Find the approximate solution at each time step
          for n = 1:Nt
          t = n*dt; % current time
          gl = exp(-c*(xl-t-0.2)^2); % BC at left side
          gr = exp(-c*(xr-t-0.2)^2); % BC at right side
          if n==1 % first time step
            for j=2:J % interior nodes
            u(j,n) = f(j) - 0.5*mu*(f(j+1)-f(j-1)) + ...
                          0.5*mu^2*(f(j+1)-2*f(j)+f(j-1));
            u(1,n) = gl; % the left-end point
            u(J+1,n) = gr; % the right-end point
            for j=2:J % interior nodes
              u(j,n) = u(j,n-1) - 0.5*mu*(u(j+1,n-1)-u(j-1,n-1)) + ...
                          0.5*mu^2*(u(j+1,n-1)-2*u(j,n-1)+u(j-1,n-1));
            u(1,n) = gl; % the left-end point
            u(J+1,n) = gr; % the right-end point
          % calculate the analytic solution
          for j=1:J+1
            xj = xl + (j-1)*dx;
            u_ex(j,n)=exp(-c*(xj-t-0.2)^2);
          end
        % plot the analytic and numerical solution at different times
        figure;
        hold on;
        n=10;
        plot(x,u(:,n),'r.',x,u_ex(:,n),'r-'); % r for red
        n=30;
        plot(x,u(:,n),'g.',x,u_ex(:,n),'g-');
        n=50;
        plot(x,u(:,n), 'b.',x,u_ex(:,n), 'b-');
        legend('Numerical t=0.1','Analytic t=0.1',...
               'Numerical t=0.3','Analytic t=0.3',...
              'Numerical t=0.5', 'Analytic t=0.5');
        title('Numerical and Analytic Solutions at t=0.1, 0.3, 0.5');
```

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

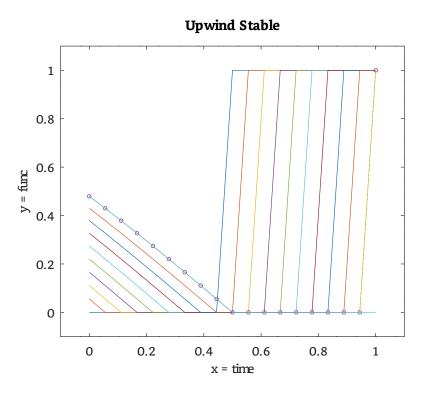
Numerical and Analytic Solutions at t=0.1, 0.3, 0.5



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 Δt = h and the scheme is stable.

```
In [ ]: |clear; close all
         a = 0; b=1; tfinal = 0.5
         m = 20;
         m=18;
         h = (b-a)/m;
         k = h; mu = k/h;
         t = 0;
         n = fix(tfinal/k);
         y1 = zeros(m+1,1); y2=y1; x=y1;
         for i=1:m+1,
           x(i) = a + (i-1)*h;
           y1(i) = uexact(t,x(i));
           y2(i) = 0;
         end
        figure(1); plot(x,y1); hold
        axis([-0.1 1.1 -0.1 1.1]);
        title('Upwind Stable');
        xlabel('x = time', 'fontsize', 10)
        ylabel('y = func', 'fontsize', 10)
         t = 0;
         for j=1:n,
           y1(1)=bc(t); y2(1)=bc(t+k);
        % for i=2:m+1
            y2(i) = y1(i) - mu*(y1(i)-y1(i-1));
           for i=2:m
             y2(i) = y1(i) - mu*(y1(i+1)-y1(i-1))/2;
             y2(i) = 0.5*(y1(i+1)+y1(i-1)) - mu*(y1(i+1)-y1(i-1))/2;
           end
           i = m+1;
           y2(i) = y1(i) - mu*(y1(i)-y1(i-1));
           t = t + k;
           y1 = y2;
           plot(x,y2); pause(0.5)
         end
         plot(x,y2,'o', "markersize", 3)
         u_e = zeros(m+1,1);
         for i=1:m+1
           u_e(i) = uexact(t,x(i));
         max(abs(u_e-y2))
         plot(x,y2,':',x,u_e)
        figure(2); plot(x,u_e-y2)
```



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 [ ]: clear; close all
         a = 0; b=1; tfinal = 0.5; % Input the domain and final time.
         m = 20; h = (b-a)/m; k = h; mu = 1.5*k/h; % Set mesh and time step.
         t = 0; n = fix(tfinal/k); % Find the number of time steps.
         y1 = zeros(m+1,1); y2=y1; x=y1;
         figure(1); hold
         %axis([-0.1 1.1 -0.1 1.1]);
         for i=1:m+1,
           x(i) = a + (i-1)*h;
           y1(i) = uexact(t,x(i)); % Initial data
           y2(i) = 0;
         end
         % Time marching
         for j=1:n,
           y1(1)=bc(t); y2(1)=bc(t+k);
           for i=2:m+1
             y2(i) = y1(i) - mu*(y1(i)-y1(i-1));
           end
           t = t + k;
           y1 = y2;
           plot(x,y2); pause(0.5); % Add the solution plot to the history.
         u_e = zeros(m+1,1);
         for i=1:m+1
           u_e(i) = uexact(t,x(i));
         end
         max(abs(u_e-y2))
         plot(x,y2,'o',"markersize", 3,x,u_e )
         title('Upwind Unstable');
        xlabel('x = time', 'fontsize', 10)
ylabel('y = u_e', 'fontsize', 10)
```

