

One Dimensional Burgers' Equation

J.M. Burgers, *Adv. Appl. Mech.* **1**, 171 (1948), introduced the equation

$$\frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} = \nu \frac{\partial^2 u}{\partial x^2} ,$$

as a simple model of shock propagation. This is basically a Navier-Stokes equation in one dimension without a pressure term. The convective term on the left is nonlinear. The diffusive term on the right represents the effects of viscosity.

The development of a shock can be seen by letting the kinematic viscosity $\nu = 0$. This gives the *inviscid* Burgers' equation

$$\frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} = 0 .$$

Compare this with the linear equation

$$\frac{\partial u}{\partial t} + c \frac{\partial u}{\partial x} = 0 ,$$

where c is a constant. The linear equation has the solution

$$u(x, t) = f(x - ct) ,$$

where f is any differentiable function. This solution represents a wave form with shape $f(x)$ moving to the right with constant speed c .

Now, in the inviscid Burgers' equation, the “speed” $c = u$, i.e., the instantaneous speed of the wave form is proportional to its amplitude u . This implies that a peak in the wave travels faster than a trough, which implies that the wave will tend to *break*. This is not allowed mathematically because breaking implies that the solution $u(x, t)$ becomes multiple valued. What actually happens is that a *shock front* develops: this is a moving point at which the solution is discontinuous.

The viscous term in Burgers' equation has two effects. First, it causes the wave amplitude to damp to zero in a diffusive fashion. Secondly, it prevents the development of a mathematical singularity at the shock front: the amplitude is continuous albeit varying very rapidly through the front.

Finite Difference Algorithms and their Stability

Consider the simpler *advection* equation

$$\frac{\partial u}{\partial t} + c \frac{\partial u}{\partial x} = 0 .$$

We discretize the variable $x = x_0 + jh$, $j = 0, 1, 2, \dots$ and the time $t = t_0 + n\tau$, $n = 0, 1, 2, \dots$. The solution $u(x, t)$ is represented by u_j^n .

Forward Time Centered Space (FTCS) algorithm

$$u_j^{n+1} = u_j^n - \frac{c\tau}{2h} (u_{j+1}^n - u_{j-1}^n) .$$

This algorithm happens to be unstable. This can be seen from a *von Neumann stability analysis*, which employs an approximate solution of the form

$$u(x, t) = z^t e^{ikx} ,$$

where k is the wave number of a spatial Fourier component of the solution, and z is an *amplification factor*. Substituting this form into the discretized equation gives

$$z^\tau = 1 - \frac{c\tau}{2h} (e^{ikh} - e^{-ikh}) = 1 - i \frac{c\tau}{h} \sin(kh) .$$

The magnitude of the amplification per time step is

$$|z^\tau| = \sqrt{1 + \left(\frac{c\tau}{h}\right)^2 \sin^2(kh)} ,$$

which is greater than unity. This shows that the algorithm is unconditionally unstable: the solution grows exponentially as a function of time if $\sin(kh) \neq 0$.

The Lax differencing scheme

The mathematician Peter Lax discovered a simple solution to the instability problem with the FTCS scheme:

$$u_j^{n+1} = \frac{1}{2} (u_{j+1}^n + u_{j-1}^n) - \frac{c\tau}{2h} (u_{j+1}^n - u_{j-1}^n) .$$

It is easy to see that

$$z^\tau = \frac{1}{2} (e^{ikh} + e^{-ikh}) - \frac{c\tau}{2h} (e^{ikh} - e^{-ikh}) = \cos(kh) - i \frac{c\tau}{h} \sin(kh) .$$

The amplification per time step is now

$$|z^\tau| = \sqrt{\cos^2(kh) + \left(\frac{c\tau}{h}\right)^2 \sin^2(kh)} ,$$

which is less than unity only if the *Courant-Friedrichs-Lewy* (CFL) stability criterion

$$\left| \frac{c\tau}{h} \right| \leq 1 ,$$

is satisfied.

Program to solve the 1-D Burgers' Equation

```
// Program to solve the 1-D Burgers' Equation
```

```
#include <cmath>
```

```
#include <cstdio>
#include <cstdlib>
#include <cstring>
#include <iostream>

using namespace std;

#include <GL/glut.h>

const double pi = 4 * atan(1.0); // value of pi

double L = 1;                // size of periodic region
int N = 200;                 // number of grid points
double h;                   // lattice spacing
double tau;                 // time step
double CFLRatio = 1;        // Courant-Friedrichs-Lewy ratio tau/h
enum {SINE, STEP};
int initialWaveform = SINE;  // sine function, step, etc.

double nu = 1e-6;           // kinematic viscosity
double *u;                  // the solution
double *uNew;               // for updating
double *F;                  // the flow
double *uPlus, *uMinus;     // for Godunov scheme
int step;                   // integration step number

void allocate() {
    static int oldN = 0;
    if (oldN != N) {
```

```
        if (u != 0)
            delete [] u; delete [] uNew; delete [] F;
            delete [] uPlus; delete [] uMinus;
    }
    oldN = N;
    u = new double [N];
    uNew = new double [N];
    F = new double [N];
    uPlus = new double [N];
    uMinus = new double [N];
}

void initialize() {

    allocate();
    h = L / N;

    double uMax = 0;
    for (int i = 0; i < N; i++) {
        double x = i * h;
        switch (initialWaveform) {
            case SINE:
                u[i] = sin(2 * pi * x) + 0.5 * sin(pi * x);
                break;
            case STEP:
                u[i] = 0;
                if (x > L / 4 && x < 3 * L / 4)
                    u[i] = 1;
                break;
        }
    }
}
```

```
        default:
            u[i] = 1;
            break;
    }
    if (abs(u[i]) > uMax)
        uMax = abs(u[i]);
}

tau = CFLRatio * h / uMax;
step = 0;
}
```

Integration algorithms

```
void (*integrationAlgorithm)();
void redraw();

void takeStep() {
    integrationAlgorithm();
    double *swap = u;
    u = uNew;
    uNew = swap;
    redraw();
    ++step;
}

void FTCS() {
    for (int j = 0; j < N; j++) {
```

```

    int jNext = j < N - 1 ? j + 1 : 0;
    int jPrev = j > 0 ? j - 1 : N - 1;
    uNew[j] = u[j] * (1 - tau / (2 * h) * (u[jNext] - u[jPrev])) +
              nu * tau / h / h * (u[jNext] + u[jPrev] - 2 * u[j]);
  }
}

```

Lax algorithm

```

void Lax() {
  for (int j = 0; j < N; j++) {
    int jNext = j < N - 1 ? j + 1 : 0;
    int jPrev = j > 0 ? j - 1 : N - 1;
    uNew[j] = (u[jNext] + u[jPrev]) / 2
              - u[j] * tau / (2 * h) * (u[jNext] - u[jPrev])
              + nu * tau / h / h * (u[jNext] + u[jPrev] - 2 * u[j]);
  }
}

```

Lax-Wendroff algorithm

The Lax-Wendroff algorithm is constructed in two steps. First, the time and convective derivatives are expressed in terms of a *flow function* F as follows:

$$\frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} = \frac{\partial u}{\partial t} + \frac{\partial F}{\partial x}, \quad F(x, t) = \frac{1}{2} u^2(x, t).$$

This is the form of a conservation equation with F representing the *current* of the quantity u .

Second, a Taylor series expansion in the time step τ of all variables is made and terms up to and including $\mathcal{O}(\tau^2)$ are retained, e.g.,

$$u(x, t + \tau) = u(x, t) + \tau \frac{\partial u}{\partial t} + \frac{\tau^2}{2} \frac{\partial^2 u}{\partial t^2} + \mathcal{O}(\tau^3) .$$

The resulting algorithm can be expressed as a two-step formula:

$$u_{j+\frac{1}{2}}^* = \frac{1}{2} (u_j^n + u_{j+1}^n) - \frac{\tau}{2h} (F_{j+1}^n - F_j^n) + \frac{\nu\tau}{2h^2} \left[\frac{1}{2} (u_{j+1}^n + u_{j-1}^n - 2u_j^n) + \frac{1}{2} (u_{j+2}^n + u_j^n - 2u_{j+1}^n) \right] ,$$

$$u_j^{n+1} = u_j^n - \frac{\tau}{h} \left(F_{j+\frac{1}{2}}^* - F_{j-\frac{1}{2}}^* \right) + \frac{\nu\tau}{h^2} (u_{j+1}^n + u_{j-1}^n - 2u_j^n) .$$

```
void LaxWendroff() {
    for (int j = 0; j < N; j++)
        F[j] = u[j] * u[j] / 2;
    for (int j = 0; j < N; j++) {
        int jMinus1 = j > 0 ? j - 1 : N - 1;
        int jPlus1 = j < N - 1 ? j + 1 : 0;
        int jPlus2 = jPlus1 < N - 1 ? jPlus1 + 1 : 0;
        uNew[j] = (u[j] + u[jPlus1]) / 2 -
            (tau / 2 / h) * (F[jPlus1] - F[j]) +
            (nu * tau / (2 * h * h)) * (
                (u[jPlus1] + u[jMinus1] - 2 * u[j]) / 2 +
                (u[jPlus2] + u[j] - 2 * u[jPlus1]) / 2 );
    }
    for (int j = 0; j < N; j++)
        F[j] = uNew[j] * uNew[j] / 2;
```



```

for (int j = 0; j < N; j++) {
    int jMinus1 = j > 0 ? j - 1 : N - 1;
    int jPlus1 = j < N - 1 ? j + 1 : 0;
    uNew[j] = u[j] - (tau / h) * (F[j] - F[jMinus1]) +
              (nu * tau / (h * h)) * (u[jPlus1] + u[jMinus1] - 2 * u[j]);
}
}

```

Godunov Scheme

This type of scheme was introduced by S.K. Godunov, *Mat. Sb.* **47**, 271 (1959). This is an *upwind* differencing scheme which makes use of the solution to a local *Riemann problem*.

A *Riemann problem* is an initial value problem for a partial differential equation with a *piecewise constant* initial value function, i.e., the solution at $t = 0$ is a step function. A *Riemann solver* is an exact or approximate algorithm for solving a Riemann problem.

The basic formula for updating u is

$$u_j^{n+1} = u_j^n - \frac{\tau}{h} \left[F_{j+\frac{1}{2}} - F_{j-\frac{1}{2}} \right] + \frac{\nu\tau}{h^2} [u_{j+1} + u_{j-1} - 2u_j] ,$$

where $F_{j\pm\frac{1}{2}}$ represents the average flux on the cells to the right and left of the lattice point j respectively. These average flux values are computed from Riemann problems in the cells to the right and left of j using *upwind* initial data

$$u_j^{(+)} = \begin{cases} u_j & \text{if } u_j > 0 \\ 0 & \text{otherwise} \end{cases} \quad u_j^{(-)} = \begin{cases} u_j & \text{if } u_j < 0 \\ 0 & \text{otherwise} \end{cases}$$

The solution to the Riemann problem on the left cell is

$$F_{j-\frac{1}{2}} = \max \left\{ \frac{1}{2} (u_{j-1}^{(+)})^2, \frac{1}{2} (u_j^{(-)})^2 \right\} ,$$

and for the cell on the right

$$F_{j+\frac{1}{2}} = \max \left\{ \frac{1}{2} \left(u_j^{(+)} \right)^2, \frac{1}{2} \left(u_{j+1}^{(-)} \right)^2 \right\} .$$

```
void Godunov() {

    for (int j = 0; j < N; j++) {
        uPlus[j] = u[j] > 0 ? u[j] : 0;
        uMinus[j] = u[j] < 0 ? u[j] : 0;
    }
    for (int j = 0; j < N; j++) {
        int jNext = j < N - 1 ? j + 1 : 0;
        int jPrev = j > 0 ? j - 1 : N - 1;
        double f1 = uPlus[jPrev] * uPlus[jPrev] / 2;
        double f2 = uMinus[j] * uMinus[j] / 2;
        F[jPrev] = f1 > f2 ? f1 : f2;
        f1 = uPlus[j] * uPlus[j] / 2;
        f2 = uMinus[jNext] * uMinus[jNext] / 2;
        F[j] = f1 > f2 ? f1 : f2;
        uNew[j] = u[j] + nu * tau / h / h * (u[jNext] + u[jPrev] - 2 * u[j]);
        uNew[j] -= (tau / h) * (F[j] - F[jPrev]);
    }
}
```

Graphics

```
int mainWindow, solutionWindow, controlWindow;
int margin = 10;
```

```
int controlHeight = 30;

void reshape(int w, int h) {
    glViewport(0, 0, w, h);
    glMatrixMode(GL_PROJECTION);
    glLoadIdentity();
    gluOrtho2D(0, w, 0, h);
    glMatrixMode(GL_MODELVIEW);
    glLoadIdentity();
}

void redraw() {
    glutSetWindow(solutionWindow);
    glutPostRedisplay();
}

void display() {
    glClear(GL_COLOR_BUFFER_BIT);

    glutSwapBuffers();
}

void displaySolution() {
    glClear(GL_COLOR_BUFFER_BIT);
    glColor3ub(255, 255, 255);
    glBegin(GL_LINE_STRIP);
        for (int i = 0; i < N; i++) {
            int iNext = i < N - 1 ? i + 1 : 0;
            glVertex2d(i * h, u[i]);
        }
    glEnd();
}
```

```

        glVertex2d((i + 1) * h, u[iNext]);
    }
    glEnd();
    char str[100];
    sprintf(str, "CFL Ratio = %.4f      nu = %.4g      t = %.4f",
            CFLRatio, nu, step * tau);
    glRasterPos2d(0.02, -0.95);
    for (int j = 0; j < strlen(str); j++)
        glutBitmapCharacter(GLUT_BITMAP_HELVETICA_12, str[j]);
    glutSwapBuffers();
}

void (*method[])() = {FTCS, Lax, LaxWendroff, Godunov};
char methodName[][20] = {"FTCS", "Lax", "Lax Wendroff", "Godunov"};

void displayControl() {
    glClear(GL_COLOR_BUFFER_BIT);
    int w = glutGet(GLUT_WINDOW_WIDTH);
    int h = glutGet(GLUT_WINDOW_HEIGHT);
    for (int i = 0; i < 4; i++) {
        if (method[i] == integrationAlgorithm)
            glColor3ub(255, 0, 0);
        else
            glColor3ub(0, 0, 255);
        glRectd((i + 0.025) * w / 4, 0.1 * h, (i + 0.975) * w / 4, 0.9 * h);
        glColor3ub(255, 255, 255);
        glRasterPos2d((i + 0.2) * w / 4, 0.3 * h);
        for (int j = 0; j < strlen(methodName[i]); j++)
            glutBitmapCharacter(GLUT_BITMAP_HELVETICA_12, methodName[i][j]);
    }
}

```

```
    }  
    glutSwapBuffers();  
}  
  
void reshapeMain(int w, int h) {  
    reshape(w, h);  
  
    glutSetWindow(solutionWindow);  
    glutPositionWindow(margin, margin);  
    glutReshapeWindow(w - 2 * margin, h - 3 * margin - controlHeight);  
  
    glutSetWindow(controlWindow);  
    glutPositionWindow(margin, h - margin - controlHeight);  
    glutReshapeWindow(w - 2 * margin, controlHeight);  
}  
  
void reshapeSolution(int w, int h) {  
    glViewport(0, 0, w, h);  
    glMatrixMode(GL_PROJECTION);  
    glLoadIdentity();  
    gluOrtho2D(0, 1, -1, +1.5);  
    glMatrixMode(GL_MODELVIEW);  
    glLoadIdentity();  
}  
  
void mouseSolution(int button, int state, int x, int y) {  
    static bool running = false;  
  
    switch (button) {
```

```
case GLUT_LEFT_BUTTON:
    if (state == GLUT_DOWN) {
        if (running) {
            glutIdleFunc(NULL);
            running = false;
        } else {
            glutIdleFunc(takeStep);
            running = true;
        }
    }
    break;
default:
    break;
}
}

void mouseControl(int button, int state, int x, int y) {

    if (button == GLUT_LEFT_BUTTON && state == GLUT_DOWN) {
        int w = glutGet(GLUT_WINDOW_WIDTH);
        int algorithm = int(x / double(w) * 4);
        if (algorithm >= 0 && algorithm < 4)
            integrationAlgorithm = method[algorithm];
        glutPostRedisplay();
    }
}

void makeMainWindow() {
    glutInitDisplayMode(GLUT_DOUBLE | GLUT_RGB);
```

```
    glutInitWindowSize(600, 400);
    glutInitWindowPosition(100, 100);
    mainWindow = glutCreateWindow("One-dimensional Burgers' Equation");
    glClearColor(1.0, 1.0, 1.0, 0.0);
    glShadeModel(GL_FLAT);
    glutDisplayFunc(display);
    glutReshapeFunc(reshapeMain);
}

void solutionMenu(int menuItem) {
    switch (menuItem) {
        case 1:
            initialWaveform = SINE;
            break;
        case 2:
            initialWaveform = STEP;
            break;
        default:
            break;
    }
    initialize();
    glutPostRedisplay();
}

void makeSolutionWindow() {
    glutSetWindow(mainWindow);
    int w = glutGet(GLUT_WINDOW_WIDTH);
    int h = glutGet(GLUT_WINDOW_HEIGHT);
    glutInitDisplayMode(GLUT_DOUBLE | GLUT_RGB);
```

```
solutionWindow = glutCreateSubWindow(mainWindow, margin, margin,
                                     w - 2 * margin, h - 3 * margin - controlHeight);
glClearColor(0.0, 0.0, 0.0, 0.0);
glShadeModel(GL_FLAT);
glutDisplayFunc(displaySolution);
glutReshapeFunc(reshapeSolution);
glutMouseFunc(mouseSolution);
integrationAlgorithm = Lax;
glutCreateMenu(solutionMenu);
glutAddMenuEntry("Initial Sine Waveform", 1);
glutAddMenuEntry("Initial Step Waveform", 2);
glutAttachMenu(GLUT_RIGHT_BUTTON);
}

void makeControlWindow() {
    glutSetWindow(mainWindow);
    int w = glutGet(GLUT_WINDOW_WIDTH);
    int h = glutGet(GLUT_WINDOW_HEIGHT);
    glutInitDisplayMode(GLUT_DOUBLE | GLUT_RGB);
    controlWindow = glutCreateSubWindow(mainWindow,
                                       margin, h - margin - controlHeight,
                                       w - 2 * margin, controlHeight);
    glClearColor(0.0, 1.0, 0.0, 0.0);
    glShadeModel(GL_FLAT);
    glutDisplayFunc(displayControl);
    glutReshapeFunc(reshape);
    glutMouseFunc(mouseControl);
}
```



```
int main(int argc, char *argv[]) {  
    glutInit(&argc, argv);  
    if (argc > 1)  
        CFLRatio = atof(argv[1]);  
    if (argc > 2)  
        nu = atof(argv[2]);  
    initialize();  
    makeMainWindow();  
    makeSolutionWindow();  
    makeControlWindow();  
    glutMainLoop();  
}
```