

Burgers' Equation

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Abstract

This paper covers some topics about Burgers equation. Starting from a traffic flow model, Burgers equation emerges. It is then solved by Cole-Hopf transformation before giving asymptotic results of the exact solution. Finally, the Input-to-State Stability(ISS) properties of Burgers equation are analyzed, and numerical experiments concludes this course project.

A simple model of traffic flow

Consider a fleet of cars driving on a highway, and let $\rho(x, t), v(x, t)$ denote the density and the velocity respectively. For any interval $[a, b]$ on the road, the total number of cars in this segment is $\int_a^b \rho(x, t) dx$. Therefore, from time t to $t + \Delta t$, we have

$$\begin{aligned} \int_a^b \rho(x, t + \Delta t) dx - \int_a^b \rho(x, t) dx &= \int_t^{t+\Delta t} \rho(a, t) v(a, t) dt - \int_t^{t+\Delta t} \rho(b, t) v(b, t) dt \\ &= \rho(a, t_1) v(a, t_1) \Delta t - \rho(b, t_2) v(b, t_2) \Delta t. \quad (\text{Mean Value Theorem}) \end{aligned}$$

Therefore,

$$\int_a^b \frac{\rho(x, t + \Delta t) - \rho(x, t)}{\Delta t} dx = \rho(a, t_1) v(a, t_1) - \rho(b, t_2) v(b, t_2).$$

Let $\Delta t \rightarrow 0$, we have

$$\int_a^b \rho_t dx = \rho(a, t) v(a, t) - \rho(b, t) v(b, t) = - \int_a^b (\rho v)_x dx,$$

i.e.

$$\int_a^b [\rho_t + (\rho v)_x] dx = 0.$$

As a, b are arbitrary, we have

$$\rho_t + (\rho v)_x = 0. \quad (\text{continuity equation})$$

Assume that $v = v_{\max}(1 - \frac{\rho}{\rho_{\max}}) - C \frac{\rho_x}{\rho}$, where ρ_{\max} is the maximum density, v_{\max} is the maximum velocity, and C is a positive constant (this term is in order to reflect the fact that drivers would reduce their speed for an increasing density ahead), then

$$\rho_t + v_{\max} \rho_x - 2 \frac{\rho_{\max}}{v_{\max}} \rho \rho_x - C \rho_{xx} = 0.$$

Make

$$\begin{cases} \hat{x} = -x + v_{\max} t \\ \hat{t} = t \end{cases},$$

then

$$\begin{aligned}\rho_t(\widehat{x}, \widehat{t}) &= \rho_{\widehat{x}} \cdot (v_{\max}) + \rho_{\widehat{t}} \\ \rho_x(\widehat{x}, \widehat{t}) &= \rho_{\widehat{x}} \cdot (-1) = -\rho_{\widehat{x}} , \\ \rho_{xx}(\widehat{x}, \widehat{t}) &= \rho_{\widehat{xx}}\end{aligned}$$

hence, $\rho_t + 2\frac{v_{\max}}{\rho_{\max}}\rho\rho_{\widehat{x}} - C\rho_{\widehat{xx}} = 0$

Let

$$\begin{cases} \rho' = \frac{2}{\rho_{\max}}\rho \\ x' = \frac{1}{v_{\max}}\widehat{x} , \\ t' = \widehat{t} \end{cases}$$

then

$$\rho'_{t'} + \rho'\rho'_{x'} = \mu\rho'_{x'x'}, \quad \text{where } \mu = \frac{C}{v_{\max}^2}.$$

When $C = 0$, we get the inviscid Burgers equation $\rho'_{t'} + \rho'\rho'_{x'} = 0$. For more information about Burgers' equation and traffic flow model, refer to [1, 2].

A simplification of Navier-Stokes Equation

Consider the incompressible Navier-Stokes equation, see [3],

$$\rho\left(\frac{\partial \mathbf{v}}{\partial t} + \mathbf{v} \cdot \nabla \mathbf{v}\right) = -\nabla p + v\Delta \mathbf{v} + \mathbf{F}$$

where ρ is the density, \mathbf{v} is the velocity, p is the pressure, v is the fluid viscosity and \mathbf{F} is an external force.

Assume that there are no external forces, and the pressure term is negligible, then the Navier-Stokes equation for 1D problem becomes the viscid Burgers equation, shown as

$$u_t + uu_x = \mu u_{xx}, \tag{1}$$

where $\mu = \frac{v}{\rho}$ is the kinematic viscosity.

When the viscosity μ is zero, then equation (1) becomes the following inviscid Burgers equation, shown as

$$u_t + uu_x = 0. \tag{2}$$

Cole-Hopf transformation and the exact solution

Consider now the viscid Burgers equation (1) with initial condition

$$u(x, 0) = u_0(x). \tag{3}$$

Define the Cole-Hopf transformation(see [4]), as

$$u = -2\mu \frac{w_x}{w}, \tag{4}$$

then

$$u_t = \frac{2\mu(w_t w_x - w w_{xt})}{w^2}, \quad uu_x = \frac{4\mu w_x(w w_{xx} - w_x^2)}{w^3}$$

and

$$\mu u_{xx} = -\frac{2\mu^2(2w_x^3 - 3w w_{xx} w_x + w^2 w_{xxx})}{w^3}.$$

Substituting these expressions into equation (1), we get

$$w_x(w_t - \mu w_{xx}) = w(w_{xt} - \mu w_{xxx}) = w(w_t - \mu w_{xx})_x.$$

Therefore, if $w(x, t)$ solves the heat equation

$$w_t - \mu w_{xx} = 0, \tag{5}$$

then $u(x, t)$ given by transformation (4) solves the viscid Burgers equation (1).

As for the initial condition (3), the new variable $w(x, t)$ must satisfy

$$-2\mu \frac{w_x(x, 0)}{w(x, 0)} = u_0(x)$$

i.e.

$$\frac{dw(x, 0)}{w(x, 0)} = -\frac{1}{2\mu} u_0(x) dx.$$

Integrating from both sides from 0 to x ,

$$\int_0^x \frac{dw(s, 0)}{w(s, 0)} = -\frac{1}{2\mu} \int_0^x u_0(s) ds, \quad (6)$$

hence,

$$w(x, 0) = e^{-\frac{1}{2\mu} \int_0^x u_0(s) ds}. \quad (7)$$

Note that the lower limit of the integral in (6) can be changed from 0 to any other convenient value. As a result of the transformation, we only need to deal with the heat diffusion problem, satisfying equation (5) and (7). With the heat kernel expression

$$E(x, t) = \frac{1}{\sqrt{4\pi\mu t}} e^{-\frac{x^2}{4\mu t}},$$

we have

$$\begin{aligned} w(x, t) &= \int_{-\infty}^{+\infty} w(\xi, 0) E(x - \xi, t) d\xi \\ &= \frac{1}{\sqrt{4\pi\mu t}} \int_{-\infty}^{+\infty} e^{-\frac{1}{2\mu} \int_0^\xi u_0(\eta) d\eta - \frac{(x-\xi)^2}{4\mu t}} d\xi \\ &=: \frac{1}{\sqrt{4\pi\mu t}} \int_{-\infty}^{+\infty} e^{-\frac{1}{2\mu} G} d\xi, \end{aligned}$$

where

$$G(x, t, \xi) = \int_0^\xi u_0(\eta) d\eta + \frac{(x - \xi)^2}{2t}.$$

It then follows that

$$\frac{\partial w}{\partial x} = -\frac{1}{2\mu\sqrt{4\pi\mu t}} \int_{-\infty}^{+\infty} \frac{x - \xi}{t} e^{-\frac{1}{2\mu} G} d\xi.$$

and the exact solution of Burgers initial value problem is obtained, which is

$$u(x, t) = \frac{\int_{-\infty}^{+\infty} \frac{x - \xi}{t} e^{-\frac{1}{2\mu} G(x, t, \xi)} d\xi}{\int_{-\infty}^{+\infty} e^{-\frac{1}{2\mu} G(x, t, \xi)} d\xi}. \quad (8)$$

Asymptotic behaviour of the solution

Suppose $G(x, t, \xi)$ has only one stationary point ξ_0 , with Taylor's formula, we expand $G(x, t, \xi)$ and $\frac{x - \xi}{t}$ at ξ_0 respectively,

$$\begin{aligned} G(x, t, \xi) &= G(x, t, \xi_0) + G'(x, t, \xi_0)(\xi - \xi_0) + \frac{1}{2} G''(x, t, \xi_0)(\xi - \xi_0)^2 + \dots \\ &= G(x, t, \xi_0) + \frac{1}{2} G''(x, t, \xi_0)(\xi - \xi_0)^2 + \dots \\ \frac{x - \xi}{t} &= \frac{x - \xi_0}{t} + \left(-\frac{1}{t}\right)(\xi - \xi_0) + \dots \end{aligned}$$

Thus, asymptotically,

$$\int_{-\infty}^{+\infty} \frac{x-\xi}{t} e^{-\frac{1}{2\mu}G(x,t,\xi)} d\xi \sim \frac{x-\xi_0}{t} e^{-\frac{1}{2\mu}G(x,t,\xi_0)} \int_{-\infty}^{+\infty} e^{-\frac{1}{4\mu}G''(x,t,\xi)(\xi-\xi_0)^2} d\xi$$

The symbol \sim means in a limiting or asymptotic sense. Let $z^2 = \frac{1}{4\mu}|G''(x,t,\xi)|(\xi-\xi_0)^2$, then $dz = \sqrt{\frac{|G''(x,t,\xi)|(\xi-\xi_0)^2}{4\mu}} d\xi$, thus,

$$\begin{aligned} \int_{-\infty}^{+\infty} \frac{x-\xi}{t} e^{-\frac{1}{2\mu}G(x,t,\xi)} d\xi &\sim \frac{x-\xi_0}{t} e^{-\frac{1}{2\mu}G(x,t,\xi_0)} \sqrt{\frac{4\mu\pi}{|G''(x,t,\xi)|}} \\ \int_{-\infty}^{+\infty} G(x,t,\xi) d\xi &\sim e^{-\frac{1}{2\mu}G(x,t,\xi_0)} \sqrt{\frac{4\mu\pi}{|G''(x,t,\xi)|}}. \end{aligned}$$

Therefore, we conclude that

$$u(x,t) \sim \frac{x-\xi_0}{t} \quad \text{as } \mu \rightarrow 0.$$

For this asymptotic estimation, refer to Method of Steepest Descent in [5, 6].

As $G'(x,t,\xi_0) = u_0(\xi_0) - \frac{x-\xi_0}{t} = 0$, then the asymptotic solution may be rewritten as

$$u(x,t) = u_0(\xi_0), \quad x = \xi_0 + u_0(\xi_0)t. \quad (9)$$

Notice that this is the same solution we obtained with method of characteristics for the inviscid Burgers' equation. This solution is only valid before t_b , where t_b denotes the breaking time of a gradient catastrophe. In some cases, (9) gives multivalued solution after a sufficient time, and discontinuities or a shock wave solution must be introduced. (Weak solution, shock wave solution, Rankine Hugoniot (jump) condition, entropy condition). When this stage is reached, The explanation is that there are two stationary points for $G(x,t,\xi)$, denoted by ξ_1 and ξ_2 with $\xi_1 > \xi_2$ respectively. Then

$$u(x,t) \sim \frac{\frac{x-\xi_1}{t}|G''(x,t,\xi_1)|^{-\frac{1}{2}}e^{-\frac{1}{2\mu}G(x,t,\xi_1)} + \frac{x-\xi_2}{t}|G''(x,t,\xi_2)|^{-\frac{1}{2}}e^{-\frac{1}{2\mu}G(x,t,\xi_2)}}{|G''(x,t,\xi_1)|^{-\frac{1}{2}}e^{-\frac{1}{2\mu}G(x,t,\xi_1)} + |G''(x,t,\xi_2)|^{-\frac{1}{2}}e^{-\frac{1}{2\mu}G(x,t,\xi_2)}} \quad (10)$$

When $G(x,t,\xi_1) \neq G(x,t,\xi_2)$, one or the other term of $e^{-\frac{1}{2\mu}G(x,t,\xi)}$ is overwhelmingly large when $\mu \rightarrow 0$. Suppose when $G(x,t,\xi_1) < G(x,t,\xi_2)$, we have

$$u(x,t) \sim \frac{x-\xi_1}{t},$$

then when $G(x,t,\xi_1) > G(x,t,\xi_2)$,

$$u(x,t) \sim \frac{x-\xi_2}{t}.$$

The criteria of $G(x,t,\xi_1) \leq G(x,t,\xi_2)$ will determine the asymptotic solution for given (x,t) . The changeover will occur at those (x,t) for which $G(x,t,\xi_1) = G(x,t,\xi_2)$, i.e.

$$\int_0^{\xi_1} u_0(\eta)d\eta + \frac{(x-\xi_1)^2}{2t} = \int_0^{\xi_2} u_0(\eta)d\eta + \frac{(x-\xi_2)^2}{2t}.$$

As $G'(x,t,\xi_1) = 0$ and $G'(x,t,\xi_2) = 0$, this condition may be rewritten as

$$\int_{\xi_1}^{\xi_2} u_0(\eta)d\eta = \frac{1}{2}(u_0(\xi_1) + u_0(\xi_2))(\xi_1 - \xi_2).$$

This is exactly the shock fitting rule for determining the shock path, which can be derived from the jump condition analytically, refer to [7].

Geometrically, it has a more intuitive explanation based on the equal area principle, refer to Figure 1 and Figure 2.

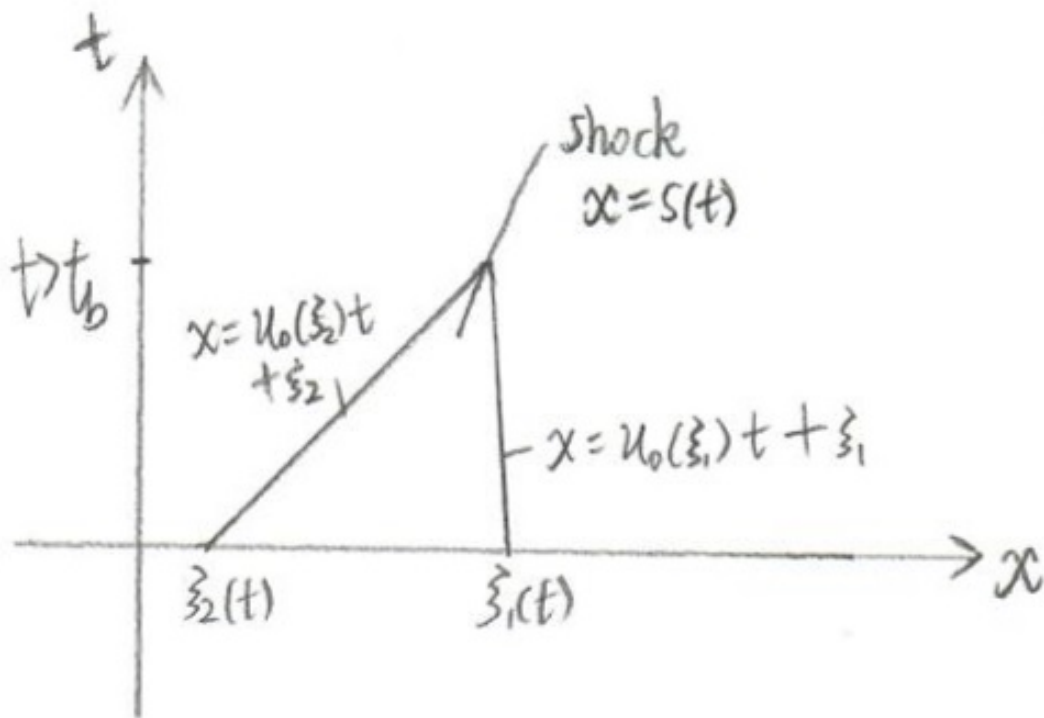


Figure 1: Two characteristics intersect on the shock path

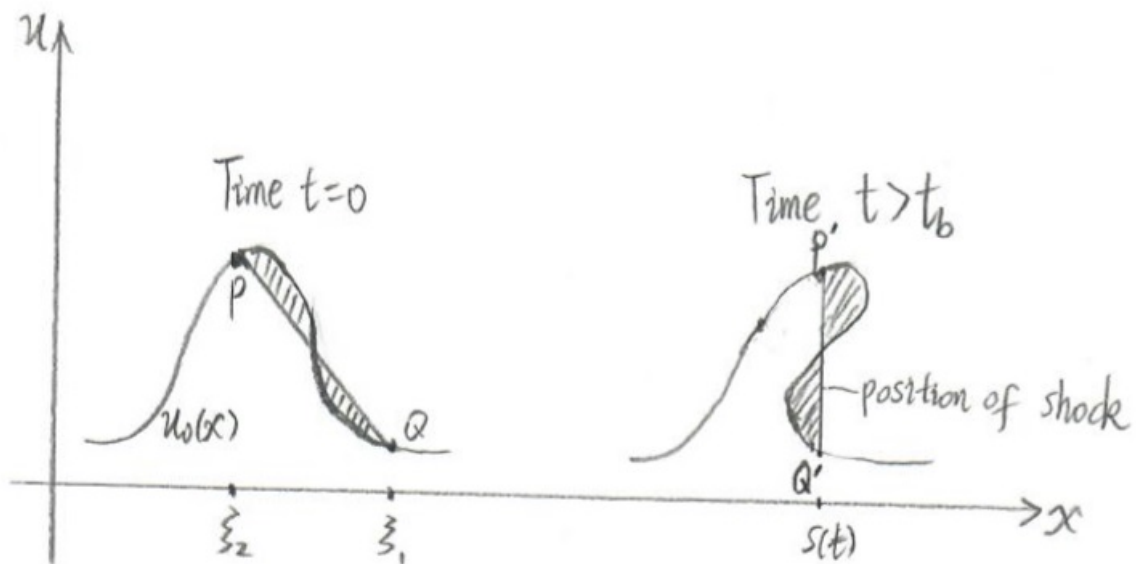


Figure 2: initial wave profile evolving into a multivalued wavelet

Input-to-State Stability(ISS) properties for Burgers Equation

Let us move on to another topic. Consider the following system for Burgers' equation with Dirichlet boundary conditions:

$$\begin{cases} u_t - \mu u_{xx} + \nu u u_x = u_0(x, t) & \text{in } (0, 1) \times \mathbb{R}_+ \\ u(0, t) = 0, u(1, t) = d(t), \\ u(x, 0) = u_0(x), \end{cases} \quad (11)$$

where $\mu > 0, \nu > 0$ are constants, $d(t)$ is the disturbance on the boundary, which can represent actuation or sensing errors, and $f(x, t)$ is the disturbance distributed over the domain. We assume that $f \in \mathcal{H}^{\theta, \frac{\theta}{2}}([0, 1], \overline{R_+})$, and $d \in \mathcal{H}^{1+\frac{\theta}{2}}(\overline{R_+})$ for some $\theta \in (0, 1)$, where \mathcal{H} represents the Holder space ([8, 9]). For the existence and uniqueness of the solution to system (11), the following theorem holds.

Theorem 1. *Assume that $u_0 \in \mathcal{H}^{2+\theta}([0, 1])$ with $u_0(0) = 0, u_0(1) = d(0), \mu u_0''(0) + f(0, 0) = 0$ and $\mu u_0''(1) + f(1, 0) = d'(0)$. For any $T > 0$, there exists a unique classical solution $u \in \mathcal{H}^{2+\theta, 1+\frac{\theta}{2}}([0, 1] \times [0, T]) \subset C^{2,1}([0, 1] \times [0, T])$ of system (11).*

The proof of this theorem follows from Theorem 6.1 in [8](pages 452-453). It is based on the linearization of the considered system and the Leray-Schauder theorem on fixed points.

Before listing the well-posedness result of system (11), we begin with some concepts and definitions, refer to [10, 11].

i **De Giorgi class** Let $\Omega \subset R^n$ be an open bounded set, and γ be a constant. The De Giorgi class $DG^+(\Omega, \gamma)$ consists of functions $u \in W^{1,2}(\Omega)$ which satisfy, for every ball $B_r(y) \subset \Omega$, every $0 < r' < r$, and $k \in R$, the following inequality:

$$\int_{B_{r'}(y)} |\nabla(u - k)_+|^2 dx \leq \frac{\gamma}{(r - r')^2} \int_{B_r(y)} |(u - k)_+|^2 dx,$$

where $(u - k)_+ = \max\{u - k, 0\}$.

The main idea of De Giorgi iteration is to estimate $|A_k|$, the measure of $\{x \in \Omega; u(x) \geq k\}$, and derive $|A_k| = 0$ with some k for functions u in De Giorgi class. The following iteration given in [12] is useful.

Lemma 1. *Suppose that ϕ is a non-negative decreasing function on $[k_0, \infty)$ satisfying*

$$\phi(h) \leq \left(\frac{M}{h - k}\right)^\alpha \phi^\beta(k), \quad \forall h > k \geq k_0,$$

where $M > 0, \alpha > 0, \beta > 1$ are constants. Then the following holds

$$\phi(k_0 + l_0) = 0,$$

with $l_0 = 2^{\frac{\beta}{\beta-1}} M \phi(k_0)^{\frac{\beta-1}{\alpha}}$.

i **Class \mathcal{K} and Class \mathcal{K}_∞** A continuous function $\alpha : [0, a) \rightarrow [0, \infty)$ belongs to class \mathcal{K} if α is strictly increasing and $\alpha(0) = 0$. If, in addition, $a = \infty$ and $\alpha(r) \rightarrow \infty$ as $r \rightarrow \infty$, then α is called a class \mathcal{K}_∞ function.

i **Class \mathcal{KL}** A continuous function $\beta : [0, a) \times [0, \infty) \rightarrow [0, \infty)$ belongs to class \mathcal{KL} if for each fixed s , the mapping $r \mapsto \beta(r, s)$ is a class \mathcal{K} function, and for each fixed r , the mapping $s \mapsto \beta(r, s)$ is decreasing in s and $\beta(r, s) \rightarrow 0$ as $s \rightarrow \infty$.

Here are some examples of class \mathcal{KL} functions:

1. $\beta(r, s) = \frac{r}{1+rs}, r \geq 0, s \geq 0;$
2. $\beta(r, s) = \frac{r}{\sqrt{2r^2s+1}}, r \geq 0, s \geq 0.$

i

ISS and EISS System (11) is said to be Input-to-State Stable (ISS) in $L^q (q \geq 2)$ with respect to (w.r.t.) boundary disturbances $d(t)$ and in-domain disturbances $f(x, t)$, if there exist functions $\beta \in \mathcal{KL}$ and $\gamma_1, \gamma_2 \in \mathcal{K}$ such that the solution to (11) satisfies

$$\begin{aligned} \|u(\cdot, t)\|_{L^q(0,1)} &\leq \beta(\|u_0\|_{L^q(0,1)}, t) + \gamma_1(\max_{[0,t]} |d(s)|) \\ &\quad + \gamma_2(\max_{[0,1] \times [0,t]} |f(x, s)|), \quad \forall t \geq 0. \end{aligned} \quad (12)$$

Moreover, it is said to be exponential input-to-state stable (EISS) if there exist $\beta' \in \mathcal{K}_\infty$ and a constant $\lambda > 0$ such that

$$\beta(\|u_0\|_{L^q(0,1)}, t) < \beta'(\|u_0\|_{L^q(0,1)})e^{-\lambda t}$$

in (12).

In order to use the technique of splitting and the method of De Giorgi iteration in the investigation of the ISS properties for the considered system, while guaranteeing the well-posedness of Theorem 1, we assume the compatibility condition $u_0(0) = u_0''(0) = u_0(1) = u_0''(1) = d(0) = d'(0) = f(0, 0) = f(1, 0) = 0$ always holds, then the ISS property for system (11) is stated in the following theorem.

Theorem 2. System (11) is EISS in L^2 norm w.r.t boundary disturbances $d(t) \in \mathcal{H}^{1+\frac{\theta}{2}}(\overline{R_+})$ and in-domain disturbances $f(x, t) \in \mathcal{H}^{\theta, \frac{\theta}{2}}([0, 1], \overline{R_+})$ satisfying $\sup_{R_+} |d(s)| + \frac{18\sqrt{2}}{\mu} \sup_{[0,1] \times R_+} |f(x, s)| < \frac{\mu}{\nu}$. And we have

$$\|u(\cdot, t)\|^2 \leq 2\|u_0\|^2 + 4 \max_{[0,t]} |d(s)|^2 + \frac{2592}{\mu^2} \sup_{[0,1] \times [0,t]} |f(x, s)|^2.$$

Let w be the unique solution of the following system:

$$\begin{cases} w_t - \mu w_{xx} + \nu w w_x = f(x, t) & \text{in } (0, 1) \times R_+ \\ w(0, t) = 0, w(1, t) = d(t), \\ w(x, 0) = 0. \end{cases} \quad (13)$$

Then $v = u - w$ is the unique solution of the following system:

$$\begin{cases} v_t - \mu v_{xx} + \nu v v_x + \nu (wv)_x = 0 & \text{in } (0, 1) \times R_+ \\ v(0, t) = 0, v(1, t) = 0, \\ v(x, 0) = u_0(x). \end{cases} \quad (14)$$

For system (13), the following estimate holds.

Lemma 2. Suppose that $\mu > 0, \nu > 0$. For every $t > 0$, one has

$$\max_{[0,1] \times [0,t]} |w(x, s)| \leq \max_{[0,t]} |d(s)| + \frac{18\sqrt{2}}{\mu} \max_{[0,1] \times [0,t]} |f(x, s)|.$$

For system (14), we have the following estimate.

Lemma 3. Suppose that $\mu > 0, \nu > 0$, and $\sup_{R_+} |d(t)| + \frac{18\sqrt{2}}{\mu} \sup_{[0,1] \times R_+} |f(x, t)| < \frac{2\mu}{5\nu}$. For every $t > 0$, one has

$$\|v(\cdot, t)\|^2 \leq \|u_0\|^2.$$

Proof of Theorem 2:

Proof. Note that $u = w + v$, we get from Lemma 2 and Lemma 3 that:

$$\begin{aligned} \|u(\cdot, t)\|^2 &\leq 2\|w(\cdot, t)\|^2 + 2\|v(\cdot, t)\|^2 \\ &\leq 2\left(\max_{[0,1] \times [0,t]} |w(x, s)|\right)^2 + 2\|v(\cdot, t)\|^2 \\ &\leq 2\|u_0\|^2 + 2\left(\max_{[0,t]} |d(s)| + \frac{18\sqrt{2}}{\mu} \max_{[0,1] \times [0,t]} |f(x, s)|\right)^2. \end{aligned}$$

Thus, the estimate holds and system (11) is EISS. \square

Proof of Lemma 2:

Proof. For any $t > 0$, let $k_0 = \max\{\max_{[0,t]} d(s), 0\}$, then $(w(0, s) - k)_+ = (w(1, s) - k)_+ = 0$ for $k \geq k_0$.

Let $I_k(s) = \int_0^1 ((w(x, s) - k)_+)^2 dx$, and suppose that $I_k(t_0) = \max_{[0,t]} I_k(s)$. Due to $I_k(0) = 0$ and $I_k(s) \geq 0$, we can assume that $t_0 > 0$ without loss of generality. Define $\eta(x, s) = (w(x, s) - k)_+ \chi_{[t_1, t_2]}(s)$, where $\chi_{[t_1, t_2]}(s)$ is the character function on $[t_1, t_2]$ with $0 \leq t_1 < t_2 \leq t_0$. For ϵ sufficiently small, choose $t_1 = t_0 - \epsilon, t_2 = t_0$, and multiply system (13) by η , we get

$$\begin{aligned} & \frac{1}{2} \int_{t_0-\epsilon}^{t_0} \frac{d}{dt} \int_0^1 ((w - k)_+)^2 dx ds + \mu \int_{t_0-\epsilon}^{t_0} \int_0^1 |((w - k)_+)_x|^2 dx ds \\ & + \nu \int_{t_0-\epsilon}^{t_0} \int_0^1 w w_x (w - k)_+ dx ds \leq \int_{t_0-\epsilon}^{t_0} \int_0^1 |f| (w - k)_+ dx ds. \end{aligned}$$

Note that

$$\frac{1}{2} \int_{t_0-\epsilon}^{t_0} \frac{d}{dt} \int_0^1 ((w - k)_+)^2 dx ds = \frac{1}{2} (I_k(t_0) - I_k(t_0 - \epsilon)) \geq 0,$$

and

$$\lim_{\epsilon \rightarrow 0^+} \int_{t_0-\epsilon}^{t_0} \int_0^1 w w_x (w - k)_+ dx ds = 0,$$

we get

$$\mu \int_0^1 |((w(x, t_0) - k)_+)_x|^2 dx \leq \int_0^1 |f(x, t_0)| (w(x, t_0) - k)_+ dx.$$

Using the fact that

$$\left(\int_a^b |u|^p dx \right)^{\frac{1}{p}} \leq (b-a)^{\frac{1}{p}} \left(\frac{2}{b-a} \|u\|^2 + (b-a) \|u_x\|^2 \right)^{\frac{1}{2}} \quad \forall p \geq 1,$$

when $u \in C^1([a, b]; R)$, refer to [13], and Pioncare's inequality, see [14], we have

$$\begin{aligned} \left(\int_0^1 |((w(x, t_0) - k)_+)_x|^p dx \right)^{\frac{2}{p}} & \leq 9 \int_0^1 |((w(x, t_0) - k)_+)|^2 dx \\ & \leq \frac{9}{\mu} \int_0^1 |f(x, t_0)| (w(x, t_0) - k)_+ dx. \quad (\forall p > 2) \end{aligned}$$

Let $A_k(s) = \{x \in (0, 1); w(x, s) > k\}$, and $\phi_k = \sup_{(0,t)} |A_k(s)|$, then

$$\begin{aligned} \left(\int_{A_k(t_0)} |((w(x, t_0) - k)_+)|^p dx \right)^{\frac{2}{p}} & \leq \frac{9}{\mu} \int_{A_k(t_0)} |f(x, t_0)| (w(x, t_0) - k)_+ dx \\ & \leq \frac{9}{\mu} \left(\int_{A_k(t_0)} |((w(x, t_0) - k)_+)|^p dx \right)^{\frac{1}{p}} \left(\int_{A_k(t_0)} |f(x, t_0)|^q dx \right)^{\frac{1}{q}} \\ & \quad \text{(Holder's Inequality)} \end{aligned}$$

Thus,

$$\begin{aligned} \left(\int_{A_k(t_0)} |((w(x, t_0) - k)_+)|^p dx \right)^{\frac{1}{p}} & \leq \frac{9}{\mu} \left(\int_{A_k(t_0)} |f(x, t_0)|^q dx \right)^{\frac{1}{q}} \\ & \leq \frac{9}{\mu} |A_k(t_0)|^{\frac{1}{q}} \max_{[0,1] \times [0,t]} |f(x, s)| \\ & \leq \frac{9}{\mu} \max_{[0,1] \times [0,t]} |f(x, s)| \phi_k^{\frac{1}{q}} \end{aligned}$$

Moreover, with the definition of $I_k(s)$, we get

$$\begin{aligned} I_k(t_0) & \leq \left(\int_{A_k(t_0)} |((w(x, t_0) - k)_+)|^p dx \right)^{\frac{2}{p}} |A_k(t_0)|^{\frac{p-2}{p}} \quad \text{(Holder's Inequality)} \\ & \leq \left(\frac{9}{\mu} \max_{[0,1] \times [0,t]} |f(x, s)| \right)^2 \phi_k^{3-\frac{4}{p}}. \end{aligned}$$

Recalling that $I_k(t_0) = \max_{[0,t]} I_k(s)$, we have

$$I_k(s) \leq I_k(t_0) \leq \left(\frac{9}{\mu} \max_{[0,1] \times [0,t]} |f(x,s)|\right)^2 \phi_k^{3-\frac{4}{p}}. \quad (15)$$

On the other hand, noticing that $|A_h(s)| < A_k(s)$ when $h > k$, we have

$$I_k(s) \geq \int_{A_h(s)} ((w(x,s) - k)_+)^2 dx \geq (h - k)^2 |A_h(s)| \quad (16)$$

Then we infer from (15) and (16) that

$$(h - k)^2 \phi_h \leq \left(\frac{9}{\mu} \max_{[0,1] \times [0,t]} |f(x,s)|\right)^2 \phi_k^{3-\frac{4}{p}},$$

i.e.

$$\phi_h \leq \left(\frac{9 \max_{[0,1] \times [0,t]} |f(x,s)|}{\mu (h - k)}\right)^2 \phi_k^{3-\frac{4}{p}}.$$

As $p > 2$, we have $3 - \frac{4}{p} > 1$. By De Giorgi iteration in Lemma 1, we obtain

$$\phi_{k_0+l_0} = \sup_{[0,t]} |A_{k_0+l_0}| = 0,$$

where $l_0 = 2^{\frac{3p-4}{2p-4}} \frac{9}{\mu} \max_{[0,1] \times [0,t]} |f(x,s)| \phi_{k_0}^{1-\frac{2}{p}} \leq \frac{9}{\mu} 2^{\frac{3p-4}{2p-4}} \max_{[0,1] \times [0,t]} |f(x,s)|$. Then

$$\begin{aligned} w(x,s) &\leq k_0 + l_0 \\ &\leq \max\{\max_{[0,t]} d(s), 0\} + \frac{9}{\mu} 2^{\frac{3p-4}{2p-4}} \max_{[0,1] \times [0,t]} |f(x,s)| \\ &\leq \max\{\max_{[0,t]} d(s), 0\} + \frac{18\sqrt{2}}{\mu} \max_{[0,1] \times [0,t]} |f(x,s)| \quad \text{as } p \rightarrow \infty. \end{aligned}$$

In order to prove the lower boundedness of $w(x,t)$, set $\bar{w} = -w$, we have

$$\begin{cases} \bar{w}_t - \mu \bar{w}_{xx} + \nu \bar{w} \bar{w}_x = -f(x,t) & \text{in } (0,1) \times R_+ \\ \bar{w}(0,t) = 0, \bar{w}(1,t) = -d(t), \\ \bar{w}(x,0) = 0. \end{cases}$$

Proceeding as above, the De Giorgi iteration gives

$$-w(x,s) = \bar{w}(x,s) \leq \max\{\max_{[0,t]} -d(s), 0\} + \frac{18\sqrt{2}}{\mu} \max_{[0,1] \times [0,t]} |f(x,s)|.$$

Hence,

$$\max_{[0,1] \times [0,t]} |w(x,s)| \leq \max_{[0,t]} |d(s)| + \frac{18\sqrt{2}}{\mu} \max_{[0,1] \times [0,t]} |f(x,s)|.$$

□

Proof of Lemma 3:

Proof. Multiplying System (14) by v and Integrating over $(0,1)$, we get

$$\int_0^1 v_t v dx + \mu \int_0^1 v_x^2 dx + \nu \int_0^1 v^2 v_x dx + \nu \int_0^1 (wv)_x v dx = 0$$

□

Note that $\int_0^1 v^2 v_x dx = 0$ and $\int_0^1 (wv)_x v dx = wv^2|_{x=0}^{x=1} - \int_0^1 wv v_x dx = - \int_0^1 wv v_x dx$, we deduce that

$$\begin{aligned}
& \frac{1}{2} \frac{d}{dt} \|v(\cdot, t)\|^2 + \mu \|v_x(\cdot, t)\|^2 \leq \nu \int_0^1 |wv v_x| dx \\
& \leq \frac{\nu}{2} \max_{[0,1] \times [0,t]} |w(x, s)| (\|v(\cdot, t)\|^2 + \|v_x(\cdot, t)\|^2) \\
& \leq \frac{\nu}{2} (\max_{[0,t]} |d(s)| + \frac{18\sqrt{2}}{\mu} \max_{[0,1] \times [0,t]} |f(x, s)|) (\|v(\cdot, t)\|^2 + \|v_x(\cdot, t)\|^2) \\
& < \frac{\nu}{2} \frac{2\mu}{5\nu} (\|v(\cdot, t)\|^2 + \|v_x(\cdot, t)\|^2) \\
& < \mu \|v_x(\cdot, t)\|^2 \quad (\text{Poincaré's inequality})
\end{aligned}$$

Thus,

$$\frac{d}{dt} \|v(\cdot, t)\|^2 < 0,$$

hence,

$$\|v(\cdot, t)\|^2 \leq \|v(\cdot, 0)\|^2 = \|u_0\|^2.$$

❶

iISS System (11) is said to be ISS w.r.t. boundary disturbances $d(t)$ and integral input-to-state stable (iISS) w.r.t. in-domain disturbances $f(x, t)$ in L^q -norm ($q \geq 2$), if there exist functions $\beta \in \mathcal{KL}$, $\theta \in \mathcal{K}_\infty$ and $\gamma_1, \gamma_2 \in \mathcal{K}$ such that the solution to (11) satisfies

$$\begin{aligned}
\|u(\cdot, t)\|_{L^q(0,1)} & \leq \beta(\|u_0\|_{L^q(0,1)}, t) + \gamma_1(\max_{[0,t]} |d(s)|) \\
& + \theta\left(\int_0^t \gamma_2(\|f(x, s)\|) ds\right), \quad \forall t \geq 0.
\end{aligned} \tag{17}$$

Theorem 3. System (11) is EISS in L^2 norm w.r.t boundary disturbances $d(t) \in \mathcal{H}^{1+\frac{\theta}{2}}(\overline{R_+})$ satisfying $\sup_{t \in \overline{R_+}} |d(t)| < \frac{\mu}{\nu}$, and EiiSS w.r.t in-domain disturbances $f(x, t) \in \mathcal{H}^{\theta, \frac{\theta}{2}}([0, 1], \overline{R_+})$, with the following estimate for any $t > 0$:

$$\|u(\cdot, t)\|^2 \leq 2\|u_0\|^2 + 2 \max_{[0,t]} |d(s)|^2 + \frac{2}{\epsilon} \int_0^t \|f(\cdot, s)\|^2 ds.$$

In order to prove Theorem 3, consider the following two systems:

$$\begin{cases} w_t - \mu w_{xx} + \nu w w_x = 0 & \text{in } (0, 1) \times R_+ \\ w(0, t) = 0, w(1, t) = d(t), \\ w(x, 0) = 0. \end{cases} \tag{18}$$

and

$$\begin{cases} v_t - \mu v_{xx} + \nu v v_x + \nu(wv + vw)_x = f(x, t) & \text{in } (0, 1) \times R_+ \\ v(0, t) = 0, v(1, t) = 0, \\ v(x, 0) = u_0(x). \end{cases} \tag{19}$$

where $v = u - w$.

For system (18), it is a special case of system (13). And for system (19), we have the following estimate:

Lemma 4. Suppose that $\mu > 0, \nu > 0$, and $\sup_{R_+} |d(t)| < \frac{\mu}{\nu}$. For every $t > 0$, one has

$$\|v(\cdot, t)\|^2 \leq \|u_0\|^2 + \frac{1}{\epsilon} \int_0^t \|f(\cdot, s)\|^2 ds, \quad \forall \epsilon \in (0, \mu).$$

Based on the results of Lemma 2 and Lemma 4, the estimate in Theorem 3 holds.

Proof of Lemma 4:

Proof. Multiply system (19) by v and integrating over $(0, 1)$, we get

$$\int_0^1 v v_t dx + \mu \int_0^1 v_x^2 dx + \nu \int_0^1 v^2 v_x dx + \nu \int_0^1 (wv)_x v dx = \int_0^1 f(x, t) dx.$$

Then

$$\begin{aligned} & \frac{1}{2} \frac{d}{dt} \|v(\cdot, t)\|^2 + \mu \|v_x(\cdot, t)\|^2 \\ & \leq \nu \int_0^1 |w v v_x| dx + \int_0^1 f(x, t) v dx \\ & \leq \frac{\nu}{2} \max_{[0, t]} |d(s)| (\|v(\cdot, t)\|^2 + \|v_x(\cdot, t)\|^2) \\ & \quad + \frac{1}{2\epsilon} \|f(\cdot, t)\|^2 + \frac{\epsilon}{2} \|v(\cdot, t)\|^2 \quad (\text{Young's Inequality}) \\ & < \frac{\mu}{5} (\|v(\cdot, t)\|^2 + \|v_x(\cdot, t)\|^2) + \frac{1}{2\epsilon} \|f(\cdot, t)\|^2 + \frac{\epsilon}{2} \|v(\cdot, t)\|^2 \end{aligned}$$

where we choose ϵ sufficiently small.

Thus,

$$\frac{d}{dt} \|v(\cdot, t)\|^2 \leq \frac{1}{\epsilon} \|f(\cdot, t)\|^2.$$

Integrating from 0 to t , we get

$$\begin{aligned} \|v(\cdot, t)\|^2 & \leq \|v(\cdot, 0)\|^2 + \frac{1}{\epsilon} \int_0^t \|f(\cdot, s)\|^2 ds \\ & \leq \|u_0\|^2 + \frac{1}{\epsilon} \int_0^t \|f(\cdot, s)\|^2 ds. \end{aligned}$$

□

Numerical experiments of Burgers Equation

In the section, we use finite differences and the Lax-Wendroff method(see [15]) to obtain the solution of inviscid time-dependent Burgers equation. The source code written in matlab is in the end of this article. Here we only give partial result of this method, shown as follows:

21-Apr-2019 14:29:01

FD1D_BURGERS_LAX:

MATLAB version

Solve the non-viscous time-dependent Burgers equation,
using the Lax-Wendroff method.

Equation to be solved:

$$du/dt + u * du/dx = 0$$

for x in [a, b], for t in [t_init, t_last]

with initial conditions:

$$u(x, 0) = u_{init}$$

and boundary conditions:

```

u(a,t) = u_a(t), u(b,t) = u_b(t)

-1.000000 <= X <= 1.000000
Number of nodes = 41
DX = 0.050000

0.000000 <= t <= 1.000000
Number of time steps = 80
DT = 0.012500

```

X:

-1.000000	-0.950000	-0.900000	-0.850000	-0.800000
-0.750000	-0.700000	-0.650000	-0.600000	-0.550000
-0.500000	-0.450000	-0.400000	-0.350000	-0.300000
-0.250000	-0.200000	-0.150000	-0.100000	-0.050000
0.000000	0.050000	0.100000	0.150000	0.200000
0.250000	0.300000	0.350000	0.400000	0.450000
0.500000	0.550000	0.600000	0.650000	0.700000
0.750000	0.800000	0.850000	0.900000	0.950000
1.000000				

```

STEP = 0
TIME = 0.000000
STABILTY = 0.125000

```

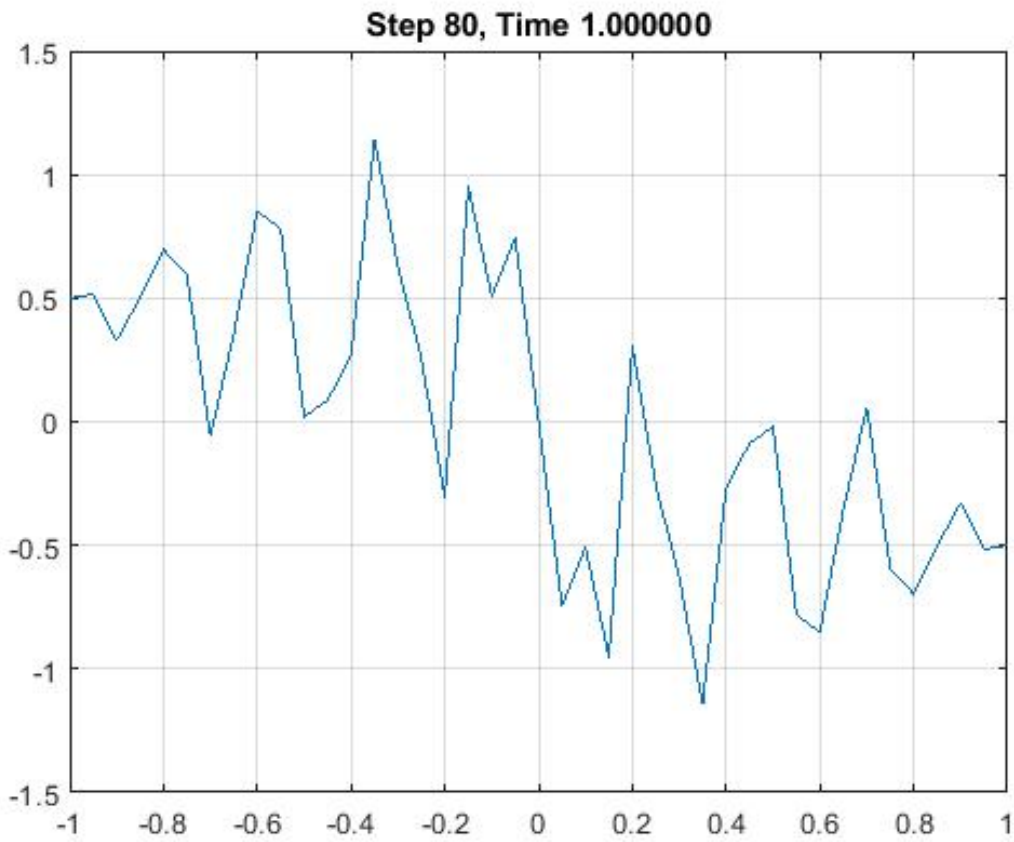


Figure 3:

0.5	0.48368	0.466525	0.448495	0.429553
0.409666	0.3888	0.366932	0.344042	0.32012
0.295167	0.269197	0.242238	0.214334	0.185547
0.155958	0.125666	0.0947863	0.063451	0.0318045
0	-0.0318045	-0.063451	-0.0947863	-0.125666
-0.155958	-0.185547	-0.214334	-0.242238	-0.269197
-0.295167	-0.32012	-0.344042	-0.366932	-0.3888
-0.409666	-0.429553	-0.448495	-0.466525	-0.48368
-0.5				

STEP = 1
TIME = 0.012500
STABILTY = 0.125000

0.5	0.491773	0.47473	0.456785	0.437893
0.418014	0.39711	0.375146	0.352098	0.327947
0.302688	0.276327	0.248889	0.220416	0.190969
0.160634	0.129515	0.0977389	0.065452	0.032815
5.98005e-19	-0.032815	-0.065452	-0.0977389	-0.129515
-0.160634	-0.190969	-0.220416	-0.248889	-0.276327
-0.302688	-0.327947	-0.352098	-0.375146	-0.39711
-0.418014	-0.437893	-0.456785	-0.47473	-0.491773
-0.5				

STEP = 2
TIME = 0.025000
STABILTY = 0.125000

0.5	0.497809	0.483033	0.465196	0.446381
0.426538	0.405622	0.383591	0.36041	0.336053
0.310505	0.283766	0.255853	0.226805	0.196683
0.165575	0.133592	0.100874	0.0675796	0.0338904
1.19612e-18	-0.0338904	-0.0675796	-0.100874	-0.133592
-0.165575	-0.196683	-0.226805	-0.255853	-0.283766
-0.310505	-0.336053	-0.36041	-0.383591	-0.405622
-0.426538	-0.446381	-0.465196	-0.483033	-0.497809
-0.5				

.....
.....
.....

STEP = 79
TIME = 0.987500
STABILTY = 0.284547

0.5	0.486051	0.350438	0.585817	0.674504
0.478441	-0.0459906	0.540337	0.882731	0.578412
-0.0601122	0.161666	0.560493	1.13819	0.292521
0.23613	-0.078433	1.01461	0.355727	0.725162
1.63746e-15	-0.725162	-0.355727	-1.01461	0.078433
-0.23613	-0.292521	-1.13819	-0.560493	-0.161666
0.0601122	-0.578412	-0.882731	-0.540337	0.0459906
-0.478441	-0.674504	-0.585817	-0.350438	-0.486051

-0.5

STEP = 80

TIME = 1.000000

STABILTY = 0.285560

0.5	0.516579	0.32841	0.501757	0.696799
0.594191	-0.0579838	0.354658	0.851096	0.779783
0.0197099	0.0872451	0.265955	1.14224	0.62925
0.256108	-0.306763	0.950214	0.507699	0.744095
1.72333e-15	-0.744095	-0.507699	-0.950214	0.306763
-0.256108	-0.62925	-1.14224	-0.265955	-0.0872451
-0.0197099	-0.779783	-0.851096	-0.354658	0.0579838
-0.594191	-0.696799	-0.501757	-0.32841	-0.516579
-0.5				

FD1D_BURGERS_LAX:

Normal end of execution.

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Source code(Matlab)

```

function fd1d_burgers_lax ( )

%*****80
%
%% FD1D_BURGERS_LAX solves the nonviscous Burgers equation using Lax-Wendroff.
%
% Discussion:
%
% The non-viscous time-dependent Burgers equation is:
%
%  $du/dt + u du/dx = 0$ 
%
% which can be written in conservative form as
%
%  $du/dt + 1/2 d/dx ( u^2 ) = 0$ 
%
% or
%
%  $du/dt + dF/dx = 0$ 
%
% For the Burgers equation, we define
%
%  $F(x,t) = 1/2 u^2,$ 
%  $A(x,t) = dF/dx = u$ 
%
% and then the Lax-Wendroff method approximates the solution
% using the iteration:
%
%  $u(x,t+dt) = u(t) - dt dF/dx + 1/2 dt^2 d/dx A dF/dx$ 
%
% which can be written:
%
%  $u(x,t+dt) = u(x,t) - dt ( F(x+dx,t) - F(x-dx,t) ) / ( 2 * dx )$ 
%  $+ 1/2 dt^2/dx^2 ( A(x+dx/2,t) * ( F(x+dx,t) - F(x,t) )$ 
%  $- A(x-dx/2,t) * ( F(x,t) - F(x-dx,t) ) )$ 
%
% where we approximate:
%
%  $A(x+dx/2,t) = 1/2 ( u(x+dx,t) + u(x,t) )$ 
%  $A(x-dx/2,t) = 1/2 ( u(x,t) + u(x-dx,t) )$ 
%
% There is a stability condition that applies here, which requires that
%
%  $dt * \max ( \text{abs} ( u ) ) / dx \leq 1$ 
%
% Licensing:
%
% This code is distributed under the GNU LGPL license.
%
% Modified:
%
% 21 August 2010
%
% Author:
%
% John Burkardt

```



```

%
% Parameters:
%
%   None
%
timestamp ( );

fprintf ( 1, '\n' );
fprintf ( 1, 'FD1D_BURGERS_LAX:\n' );
fprintf ( 1, '  MATLAB version\n' );
fprintf ( 1, '  Solve the non-viscous time-dependent Burgers equation,\n' );
fprintf ( 1, '  using the Lax-Wendroff method.\n' );
fprintf ( 1, '\n' );
fprintf ( 1, '  Equation to be solved:\n' );
fprintf ( 1, '\n' );
fprintf ( 1, '    du/dt + u * du/dx = 0\n' );
fprintf ( 1, '\n' );
fprintf ( 1, '  for x in [ a, b ], for t in [t_init, t_last]\n' );
fprintf ( 1, '\n' );
fprintf ( 1, '  with initial conditions:\n' );
fprintf ( 1, '\n' );
fprintf ( 1, '    u(x,0) = u_init\n' );
fprintf ( 1, '\n' );
fprintf ( 1, '  and boundary conditions:\n' );
fprintf ( 1, '\n' );
fprintf ( 1, '    u(a,t) = u_a(t), u(b,t) = u_b(t)\n' );
%
% Set and report the problem parameters.
%
n = 41;
a = -1.0;
b = +1.0;
dx = ( b - a ) / ( n - 1 );
step_num = 80;
t_init = 0.0;
t_last = 1.0;
dt = ( t_last - t_init ) / step_num;

fprintf ( 1, '\n' );
fprintf ( 1, '  %f <= X <= %f\n', a, b );
fprintf ( 1, '  Number of nodes = %d\n', n );
fprintf ( 1, '  DX = %f\n', dx );
fprintf ( 1, '\n' );
fprintf ( 1, '  %f <= t <= %f\n', t_init, t_last );
fprintf ( 1, '  Number of time steps = %d\n', step_num );
fprintf ( 1, '  DT = %f\n', dt );

x = r8vec_even ( n, a, b );

fprintf ( 1, '\n' );
fprintf ( 1, '  X:\n' );
fprintf ( 1, '\n' );
for ilo = 1 : 5 : n
    ihi = min ( ilo + 4, n );
    for i = ilo : ihi
        fprintf ( 1, '  %16f', x(i,1) );
    end
end

```

```

    fprintf ( 1, '\n' );
end
%
% Set the initial condition,
% and apply boundary conditions to first and last entries.
%
step = 0;
t = t_init;
un(1:n,1) = u_init ( n, x, t );
un(1,1) = u_a ( x(1,1), t );
un(n,1) = u_b ( x(n,1), t );

stability = ( dt / dx ) * max ( abs ( un(1:n,1) ) );
report ( step, step_num, n, x, t, un, stability );

if ( true )
    plot ( x, un );
    grid ( 'on' );
    title ( sprintf ( 'Step %d, Time %f', step, t ) );
end
%
% March in time.
%
c1 = - ( 0.5 * dt / dx );
c2 = - ( 0.5 * dt^2 / dx^2 );

for step = 1 : step_num

    t = ( ( step_num - step ) * t_init ...
          + ( step ) * t_last ) ...
        / ( step_num );

    uo(1:n,1) = un(1:n,1);

    un(2:n-1,1) = uo(2:n-1,1) ...
        - ( dt / dx ) * ( uo(3:n,1).^2 - uo(1:n-2,1).^2 ) ...
        + 0.5 * ( dt^2 / dx^2 ) * ( 0.5 * ( uo(3:n,1) + uo(2:n-1,1) ) ...
            .* ( uo(3:n,1).^2 - uo(2:n-1,1).^2 ) ...
            - 0.5 * ( uo(2:n-1,1) + uo(1:n-2,1) ) ...
            .* ( uo(2:n-1,1).^2 - uo(1:n-2,1).^2 ) );

    un(1,1) = u_a ( x(1,1), t );
    un(n,1) = u_b ( x(n,1), t );

    stability = ( dt / dx ) * max ( abs ( un(1:n,1) ) );
    report ( step, step_num, n, x, t, un, stability );

    if ( true )
        plot ( x, un );
        grid ( 'on' );
        title ( sprintf ( 'Step %d, Time %f', step, t ) );
    end

end

end
%
% Terminate.
%
```

```

fprintf ( 1, '\n' );
fprintf ( 1, 'FD1D_BURGERS_LAX:\n' );
fprintf ( 1, ' Normal end of execution.\n' );

return
end
function a = r8vec_even ( n, alo, ahi )

%*****80
%
%% R8VEC_EVEN returns N real values, evenly spaced between ALO and AHI.
%
% Licensing:
%
% This code is distributed under the GNU LGPL license.
%
% Modified:
%
% 24 January 2004
%
% Author:
%
% John Burkardt
%
% Parameters:
%
% Input, integer N, the number of values.
%
% Input, real ALO, AHI, the low and high values.
%
% Output, real A(N), N evenly spaced values.
% Normally, A(1) = ALO and A(N) = AHI.
% However, if N = 1, then A(1) = 0.5*(ALO+AHI).
%
if ( n == 1 )

    a(1,1) = 0.5 * ( alo + ahi );

else

    a(1:n,1) = ( (n-1:-1:0) * alo + (0:n-1) * ahi ) / ( n - 1 );

end

return
end
function report ( step, step_num, n, x, t, u, stability )

%*****80
%
%% REPORT prints or plots or saves the data at the current time step.
%
% Licensing:
%
% This code is distributed under the GNU LGPL license.
%
% Modified:

```

```

%
%   18 August 2010
%
% Author:
%
%   John Burkardt
%
% Parameters:
%
%   Input, integer STEP, the index of the current step,
%   between 0 and STEP_NUM.
%
%   Input, integer STEP_NUM, the number of steps to take.
%
%   Input, integer N, the number of nodes.
%
%   Input, real X(N), the coordinates of the nodes.
%
%   Input, real T, the current time.
%
%   Input, real U(N), the initial values U(X,T).
%
%   Input, real STABILITY, the stability factor, which should be
%   no greater than 1.
%
fprintf ( 1, '\n' );
fprintf ( 1, ' STEP = %d\n', step );
fprintf ( 1, ' TIME = %f\n', t );
fprintf ( 1, ' STABILTY = %f\n', stability )
fprintf ( 1, '\n' );
for ilo = 1 : 5 : n
    ihi = min ( ilo + 4, n );
    for i = ilo : ihi
        fprintf ( 1, ' %14g', u(i) );
    end
    fprintf ( 1, '\n' );
end

return
end
function ua = u_a ( x, t )

%*****80
%
%% U_A sets the boundary condition for U at A.
%
% Licensing:
%
%   This code is distributed under the GNU LGPL license.
%
% Modified:
%
%   18 August 2010
%
% Author:
%
%   John Burkardt

```

```

%
% Parameters:
%
%   Input, real X, T, the position and time.
%
%   Output, real UA, the prescribed value of U(X,T).
%
ua = + 0.5;

return
end
function ub = u_b ( x, t )

%*****80
%
%% U_B sets the boundary condition for U at B.
%
% Licensing:
%
%   This code is distributed under the GNU LGPL license.
%
% Modified:
%
%   18 August 2010
%
% Author:
%
%   John Burkardt
%
% Parameters:
%
%   Input, real X, T, the position and time.
%
%   Output, real UB, the prescribed value of U(X,T).
%
ub = - 0.5;

return
end
function u = u_init ( n, x, t )

%*****80
%
%% U_INIT sets the initial condition for U.
%
% Licensing:
%
%   This code is distributed under the GNU LGPL license.
%
% Modified:
%
%   18 August 2010
%
% Author:
%
%   John Burkardt
%

```

```

% Parameters:
%
%   Input, integer N, the number of nodes.
%
%   Input, real X(N), the coordinates of the nodes.
%
%   Input, real T, the current time.
%
%   Output, real U(N), the initial values U(X,T).
%
ua = u_a ( x(1,1), t );
ub = u_b ( x(n,1), t );

q = 2.0 * ( ua - ub ) / pi;
r = ( ua + ub ) / 2.0;
%
% S can be varied. It is the slope of the initial condition at the midpoint.
%
s = 1.0;
u(1:n,1) = ( 2 * x(1:n,1) - x(n,1) - x(1,1) ) ...
           / ( x(n,1) - x(1,1) );

u(1:n,1) = - q * atan ( s * u(1:n,1) ) + r;

return
end

function timestamp ( )

%*****80
%
%% TIMESTAMP prints the current YMDHMS date as a timestamp.
%
% Licensing:
%
%   This code is distributed under the GNU LGPL license.
%
% Modified:
%
%   14 February 2003
%
% Author:
%
%   John Burkardt
%
t = now;
c = datevec ( t );
s = datestr ( c, 0 );
fprintf ( 1, '%s\n', s );

return
end

```