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Numerical Methods for
Differential Equations in Physics

Discretization in Two or Three Dimensions

Classification of second order linear partial differential equations

$$au_{xx} + bu_{xy} + c_{yy} + du_x + eu_y + fu = g$$

Depends on coefficients a, b, c . Compare to quadratic functions:

$$ax^2 + bxy + cy^2 + dx + ey = g$$

Ellipse: $b^2 - 4ac < 0$

Examples: $x^2 + y^2 = g$

Poisson $(\partial_{xx} + \partial_{yy}) u = g$

Laplace $(\partial_{xx} + \partial_{yy}) u = 0$

Helmholtz $(\partial_{xx} + \partial_{yy} - k^2 I) u = 0$

Parabola: $b^2 - 4ac = 0$

Examples: $x^2 + y = g$

Heat $\partial_t u = \partial_{xx} u$

Hyperbola: $b^2 - 4ac > 0$

Examples: $x^2 - y^2 = g$

Wave $\partial_{tt} u = \partial_{xx} u$

Elliptic Partial Differential Equations: Poisson or Laplace equation

$$b^2 - 4ac < 0$$

$$\Delta u = u_{xx} + u_{yy} = g$$

Periodic boundary conditions in x and y : Fourier-Fourier

$$u(x, y) = \sum_{k,m} \hat{u}_{k,m} e^{ikx} e^{imy} \quad g(x, y) = \sum_{k,m} \hat{g}_{k,m} e^{ikx} e^{imy}$$

$$\Delta u = \sum_{k,m} (-k^2 - m^2) \hat{u}_{k,m} e^{ikx} e^{imy} = \sum_{k,m} \hat{g}_{k,m} e^{ikx} e^{imy}$$

$$\hat{u}_{k,m} = \frac{-\hat{g}_{k,m}}{k^2 + m^2}$$

Have used interval $[0, 2\pi)$ for simplicity. More generally,

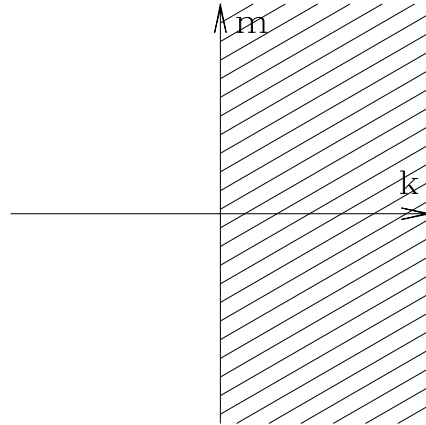
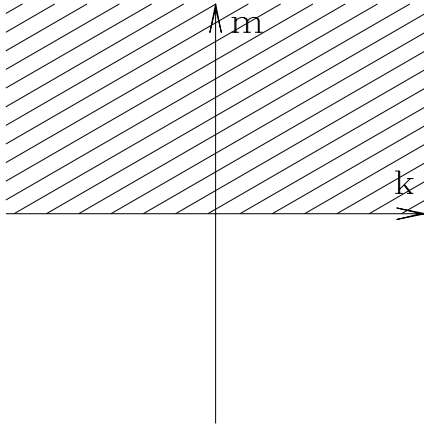
domain is $[0, L_x) \times [0, L_y)$ and basis functions are $e^{ikx2\pi/L_x} e^{imy2\pi/L_y}$

For $u(x)$ real, \hat{u}_k is conjugate symmetric: $\hat{u}_{-k} = \hat{u}_k^*$ so that

$$u(x) \sim \hat{u}_k e^{ikx} + \hat{u}_{-k} e^{-ikx} = \hat{u}_k e^{ikx} + (\hat{u}_k e^{ikx})^* = 2\mathcal{R}e(\hat{u}_k e^{ikx})$$

is real.

For $u(x, y)$ real, $\hat{u}_{-k, -m} = \hat{u}_{k, m}^*$ so need half of (k, m) plane.



Multidimensional Fourier Transform:

Transform in x : N_y independent transforms, each takes time $O(N_x \log N_x)$.

Transform in y : N_x independent transforms, each takes time $O(N_y \log N_y)$.

Total: $N_y N_x \log N_x + N_x N_y \log N_y = N_x N_y \log(N_x N_y)$.

Three dimensions (x, y, z):

Total: $N_x N_y N_z (\log N_x + \log N_y + \log N_z) = N_x N_y N_z \log(N_x N_y N_z)$

Even without FFT (SFT= Slow Fourier Transform), multidimensional Fourier transform would be fast because the different dimensions are decoupled:

$$N_x N_y N_z (N_x + N_y + N_z)$$

Decoupling also applies in other non-Fourier contexts.

Fourier transform in x is action with matrix $F_{k_x, k'_x}^x \delta_{k_y, k'_y} \delta_{k_z, k'_z}$

Fourier transform in y is action with matrix $F_{k_y, k'_y}^y \delta_{k_x, k'_x} \delta_{k_z, k'_z}$

Fourier transform in z is action with matrix $F_{k_z, k'_z}^z \delta_{k_y, k'_y} \delta_{k_x, k'_x}$

**Periodic boundary conditions in x , Dirichlet boundary conditions in y :
Fourier-Finite Differences**

$$u(x, y) = \sum_k \hat{u}_k(y) e^{ikx} \qquad g(x, y) = \sum_k \hat{g}_k(y) e^{ikx}$$

$$\begin{aligned} \Delta u &= \sum_k \left(-k^2 \hat{u}_k(y) + \frac{\hat{u}_k(y + \Delta y) - 2\hat{u}_k(y) + \hat{u}_k(y - \Delta y)}{(\Delta y)^2} \right) e^{ikx} \\ &= \sum_k \frac{\hat{u}_k(y + \Delta y) - (2 + k^2(\Delta y)^2) \hat{u}_k(y) + \hat{u}_k(y - \Delta y)}{(\Delta y)^2} e^{ikx} \end{aligned}$$

$$\frac{1}{(\Delta y)^2} \begin{pmatrix} -(2 + k^2 \Delta y) & 1 & & & & & \\ 1 & -(2 + k^2 \Delta y) & & & & & \\ & 1 & -(2 + k^2 \Delta y) & & & & \\ & & 1 & -(2 + k^2 \Delta y) & & & \\ & & & \ddots & & & \\ & & & & 1 & -(2 + k^2 \Delta y) & \\ & & & & & & \ddots & \\ & & & & & & & 1 & -(2 + k^2 \Delta y) \end{pmatrix} \begin{pmatrix} \hat{u}_k(y_1) \\ \hat{u}_k(y_2) \\ \hat{u}_k(y_3) \\ \vdots \\ \hat{u}_k(y_N) \end{pmatrix}$$

Boundary conditions needed, e.g.

$$\frac{1}{(\Delta y)^2} \begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ 1 & -(2 + k^2 \Delta y) & 1 & & \\ & 1 & -(2 + k^2 \Delta y) & 1 & \\ & & & \ddots & \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix} \begin{pmatrix} \hat{u}_k(y_1) \\ \hat{u}_k(y_2) \\ \hat{u}_k(y_3) \\ \vdots \\ \hat{u}_k(y_N) \end{pmatrix} = \begin{pmatrix} \gamma_- \\ \hat{g}_k(y_2) \\ \hat{g}_k(y_3) \\ \vdots \\ \gamma_+ \end{pmatrix}$$

Elliptic equations need boundary conditions (Dirichlet, Neumann or periodic) along all boundaries of domain

How should a system like this be solved?

LU decomposition \iff Gaussian elimination

$$\begin{pmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{pmatrix} = \begin{pmatrix} 1 & & & \\ \ell_{21} & 1 & & \\ \ell_{31} & \ell_{32} & 1 & \\ \ell_{41} & \ell_{42} & \ell_{43} & 1 \end{pmatrix} \begin{pmatrix} u_{11} & u_{12} & u_{13} & u_{14} \\ & u_{22} & u_{23} & u_{24} \\ & & u_{33} & u_{34} \\ & & & u_{44} \end{pmatrix}$$

$$\begin{pmatrix} 1 & & & \\ \ell_{21} & 1 & & \\ \ell_{31} & \ell_{32} & 1 & \\ \ell_{41} & \ell_{42} & \ell_{43} & 1 \end{pmatrix} \underbrace{\left[\begin{pmatrix} u_{11} & u_{12} & u_{13} & u_{14} \\ & u_{22} & u_{23} & u_{24} \\ & & u_{33} & u_{34} \\ & & & u_{44} \end{pmatrix} \begin{pmatrix} f_1 \\ f_2 \\ f_3 \\ f_4 \end{pmatrix} \right]}_h = \begin{pmatrix} g_1 \\ g_2 \\ g_3 \\ g_4 \end{pmatrix}$$

$$\begin{pmatrix} u_{11} & u_{12} & u_{13} & u_{14} \\ & u_{22} & u_{23} & u_{24} \\ & & u_{33} & u_{34} \\ & & & u_{44} \end{pmatrix} \begin{pmatrix} f_1 \\ f_2 \\ f_3 \\ f_4 \end{pmatrix} = \begin{pmatrix} h_1 \\ h_2 \\ h_3 \\ h_4 \end{pmatrix}$$

Find $f_4 = h_4/u_{44}$, then $f_3 = (h_3 - u_{34}f_4)/u_{33}, \dots$

LU decomposition preserves bandedness

$$\begin{pmatrix} a_{11} & a_{12} & & & \\ a_{21} & a_{22} & a_{23} & & \\ & a_{32} & a_{33} & a_{34} & \\ & & a_{43} & a_{44} & \end{pmatrix} = \begin{pmatrix} 1 & & & & \\ \ell_{21} & 1 & & & \\ & \ell_{32} & 1 & & \\ & & \ell_{43} & 1 & \end{pmatrix} \begin{pmatrix} u_{11} & u_{12} & & & \\ & u_{22} & u_{23} & & \\ & & u_{33} & u_{34} & \\ & & & u_{44} & \end{pmatrix}$$

For $N \times N$ matrix, operation count is
 $O(N^3)$ for LU decomposition and $O(N^2)$ for backsolve.

For $N \times N$ matrix with J diagonal bands, operation count is
 $O(J^2N)$ for LU decomposition and $O(JN)$ for backsolve.

LU decomposition is done once, backsolve done for each right-hand-side.

**Periodic boundary conditions in x , Dirichlet boundary conditions in y :
Fourier-Chebyshev**

$$\begin{aligned}
 u(x, y) &= \sum_{k,n} u_{k,n} e^{ikx} T_n(y) & g(x, y) &= \sum_{k,n} g_{k,n} e^{ikx} T_n(y) \\
 \Delta u &= \sum_{k,n} -k^2 u_{k,n} e^{ikx} T_n(y) + \sum_{k,n} u_{k,n} e^{ikx} T_n''(y) \\
 &= \sum_{k,n} -k^2 u_{k,n} e^{ikx} T_n(y) + \sum_{k,n} u_{k,n} e^{ikx} \sum_m R_{m,n} T_m(y) \\
 &= \sum_{k,n} -k^2 u_{k,n} e^{ikx} T_n(y) + \sum_{k,m} \left(\sum_n R_{m,n} u_{k,n} \right) e^{ikx} T_m(y) \\
 &= \sum_{k,n} -k^2 u_{k,n} e^{ikx} T_n(y) + \sum_{k,m} (Ru_k)_m e^{ikx} T_m(y) \\
 &= \sum_{k,n} -k^2 u_{k,n} e^{ikx} T_n(y) + \sum_{k,n} (Ru_k)_n e^{ikx} T_n(y) \\
 &= \sum_{k,n} (-k^2 u_{k,n} + (Ru_k)_n) e^{ikx} T_n(y)
 \end{aligned}$$

where $(Ru_k)_m \equiv \sum_n R_{m,n} u_{k,n}$

$$\Delta u(x, y) = \sum_{k,n} \left(\sum_{k',n'} \Delta_{k,n,k',n'} u_{k',n'} \right) e^{ikx} T_n(y)$$

where $\Delta_{k,n,k',n'} = -k^2 \delta_{k,k'} \delta_{n,n'} + R_{n,n'} \delta_{k,k'}$

Operators in x commute with operators in y .

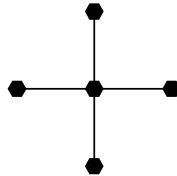
Odd and even Chebyshev polynomials are decoupled.

**Recall: second-derivative Chebyshev matrix R is upper triangular
tridiagonal matrix B is such that BR is tridiagonal.**

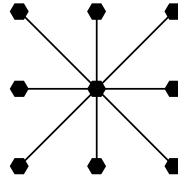
$$B\Delta u(x, y) = \sum_{k,n,n'} (-k^2 B_{n,n'} + (BR)_{n,n'}) u_{k',n'} e^{ikx} T_n(y)$$

Dirichlet BCs in x and y : Finite differences/Finite differences

five-point



nine-point



Five-point stencil: $\frac{d^2u}{dx^2}(x_j, y_k) + \frac{d^2u}{dy^2}(x_j, y_k)$

$$\approx \frac{1}{h^2} (u(x_{j+1}, y_k) - 2u(x_j, y_k) + u(x_{j-1}, y_k))$$
$$+ \frac{1}{h^2} (u(x_j, y_{k+1}) - 2u(x_j, y_k) + u(x_j, y_{k-1}))$$

Error is $\frac{h^2}{24}(\partial_x^4 + \partial_y^4)u + \dots$

This error is not isotropic, unlike the Laplacian itself.

We show that the continuous Laplacian IS isotropic:

Rotate $(x, y) \rightarrow (x', y')$: $x' = \alpha x + \beta y$, $y' = -\beta x + \alpha y$, $\alpha^2 + \beta^2 = 1$

$$\frac{\partial}{\partial x} = \frac{\partial x'}{\partial x} \frac{\partial}{\partial x'} + \frac{\partial y'}{\partial x} \frac{\partial}{\partial y'} = \alpha \frac{\partial}{\partial x'} - \beta \frac{\partial}{\partial y'}$$

$$\begin{aligned} \frac{\partial^2}{\partial x^2} &= \left(\alpha \frac{\partial}{\partial x'} - \beta \frac{\partial}{\partial y'} \right)^2 \\ &= \alpha^2 \frac{\partial^2}{\partial x'^2} - 2\alpha\beta \frac{\partial^2}{\partial x' \partial y'} + \beta^2 \frac{\partial^2}{\partial y'^2} \end{aligned}$$

$$\frac{\partial}{\partial y} = \frac{\partial x'}{\partial y} \frac{\partial}{\partial x'} + \frac{\partial y'}{\partial y} \frac{\partial}{\partial y'} = \beta \frac{\partial}{\partial x'} + \alpha \frac{\partial}{\partial y'}$$

$$\begin{aligned} \frac{\partial^2}{\partial y^2} &= \left(\beta \frac{\partial}{\partial x'} + \alpha \frac{\partial}{\partial y'} \right)^2 \\ &= \beta^2 \frac{\partial^2}{\partial x'^2} + 2\alpha\beta \frac{\partial^2}{\partial x' \partial y'} + \alpha^2 \frac{\partial^2}{\partial y'^2} \end{aligned}$$

$$\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} = \frac{\partial^2}{\partial x'^2} + \frac{\partial^2}{\partial y'^2}$$

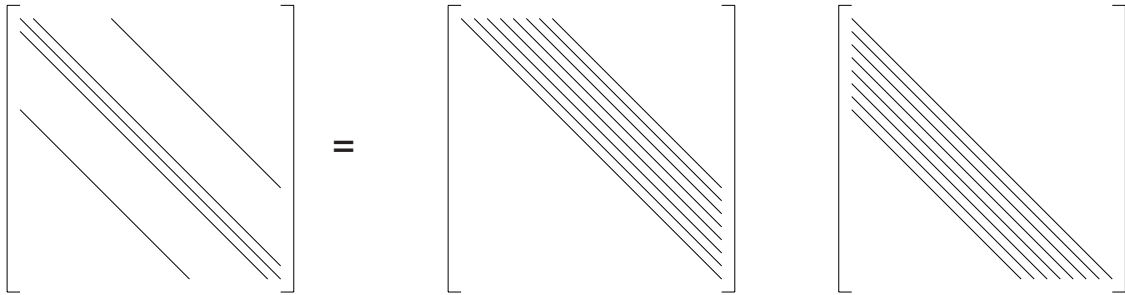
Nine-point stencil is isotropic:

$$\frac{d^2u}{dx^2}(x_j, y_k) + \frac{d^2u}{dy^2}(x_j, y_k) \approx$$

$$\frac{1}{6h^2} [4(u(x_{j+1}, y_k)u(x_{j-1}, y_k) + u(x_j, y_{k+1}) + u(x_j, y_{k-1})) \\ + u(x_{j+1}, y_{k+1}) + u(x_{j-1}, y_{k-1}) + u(x_{j+1}, y_{k-1}) + u(x_{j-1}, y_{k+1}) \\ - 20u(x_j, y_k)]$$

$$\text{Error: } \frac{h^2}{12} (\partial_x^2 + \partial_y^2)^2 = \frac{h^2}{12} \Delta^2$$

Two-dimensional finite-difference Laplacian:



Fill-in until maximal bandwidth: bandedness not preserved

Alternative way of inverting: diagonalize in one or both directions.

**Recall that operators in x commute with operators in y
so they are simultaneously diagonalizable.**

$$\begin{aligned}\Delta(n, m, n', m') &= D_{n,n'}^{xx} \delta_{m,m'} + D_{m,m'}^{yy} \delta_{n,n'} \\ &= V^{xx} \Lambda^{xx} (V^{xx})^{-1} \delta_{m,m'} + V^{yy} \Lambda^{yy} (V^{yy})^{-1} \delta_{n,n'}\end{aligned}$$

Transform to x and y eigenspace $O(N_x^2 N_y + N_y^2 N_x)$

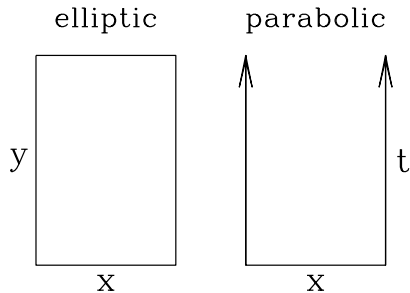
Invert Laplacian in eigenspace: $O(N_x N_y)$

Inverse transform back from x and y eigenspace $O(N_x^2 N_y + N_y^2 N_x)$

Total: $O(N_x N_y (N_x + N_y))$.

What about other kinds of PDEs?

Parabolic, example of heat equation: $\partial_t u = \partial_{xx} u$



Need:

**boundary conditions on boundary in x
(possibly also y, z in higher dimensions)**

initial condition at $t = 0$

Spatial discretization: Finite differences, Fourier, Chebyshev

Temporal discretization: implicit or explicit multistep or Runge-Kutta

Implicit Euler:

$$U(t + \Delta t) = U(t) + \Delta t D^{xx} U(t + \Delta t)$$
$$(I - \Delta t D^{xx}) U(t + \Delta t) = U(t)$$

Helmholtz equation – like Poisson equation

Same questions of discretization and solution of linear systems

Example of Navier-Stokes equations

$$\partial_t \mathbf{U} = -(\mathbf{U} \cdot \nabla) \mathbf{U} - \nabla P + Re^{-1} \Delta \mathbf{U}$$

$$\nabla \cdot \mathbf{U} = 0$$

$$\mathbf{U} = \mathbf{b} \text{ at boundaries}$$

Four equations for four fields: U, V, W, P .

Derive pressure Poisson equation using $\nabla \cdot \mathbf{U} = 0$

$$\underbrace{\nabla \cdot \partial_t \mathbf{U}}_0 = -\nabla \cdot (\mathbf{U} \cdot \nabla) \mathbf{U} - \nabla \cdot \nabla P + Re^{-1} \underbrace{\nabla \cdot \Delta \mathbf{U}}_0$$

$$\boxed{\Delta P = -\nabla \cdot (\mathbf{U} \cdot \nabla) \mathbf{U}}$$

Use implicit Euler for viscous term, explicit Euler for other terms

$$\mathbf{U}(t + \Delta t) = \mathbf{U}(t) - \Delta t \nabla P(t) - \Delta t (\mathbf{U} \cdot \nabla) \mathbf{U}(t) + Re^{-1} \Delta t \Delta \mathbf{U}(t + \Delta t)$$

$$\boxed{(I - Re^{-1} \Delta t \Delta) \mathbf{U}(t + \Delta t) = \mathbf{U}(t) - \Delta t \nabla P(t) - \Delta t (\mathbf{U} \cdot \nabla) \mathbf{U}(t)}$$

Need to solve Poisson and Helmholtz equations at each timestep.

$$\partial_t \mathbf{U} = -(\mathbf{U} \cdot \nabla) \mathbf{U} - \nabla P + Re^{-1} \Delta \mathbf{U}$$

Timestepping the advective term explicitly:

$$U(t + \Delta t) = U(t) - \Delta t (U(t) \cdot \nabla) U(t)$$

$$\text{or} \quad = U(t) - \frac{\Delta t}{2} (3 [(U \cdot \nabla) U](t) - [(U \cdot \nabla) U](t - \Delta t))$$

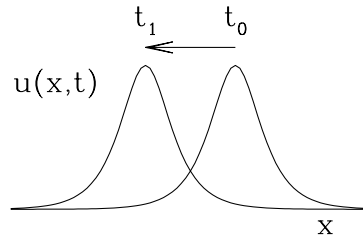
$$(U(t) \cdot \nabla) U(t) \sim C \partial_x U(t)$$

Motivates study of methods for solving model of first-order wave equation:

$$\partial_t u = c \partial_x u$$

Analytic solution: traveling wave

$$u(x, t) = u(x + ct, 0)$$



General definition of hyperbolic equation in one dimension:

$$\partial_t \mathbf{u} = \partial_x \mathbf{f}(\mathbf{u})$$

First order wave equation has $\mathbf{f}(\mathbf{u}) = c\mathbf{u}$.

Can generalize to $\mathbf{x} \in \mathcal{R}^d$, $\mathbf{u} \in \mathcal{R}^s$, $\mathbf{f}^j : \mathcal{R}^s \rightarrow \mathcal{R}^s$

$$\partial_t \mathbf{u}(x_1, \dots, x_d) = \sum_{j=1}^d \partial_{x_j} \mathbf{f}^j(\mathbf{u}(x_1, \dots, x_d)) = \nabla \cdot \mathbf{f}(\mathbf{u})$$

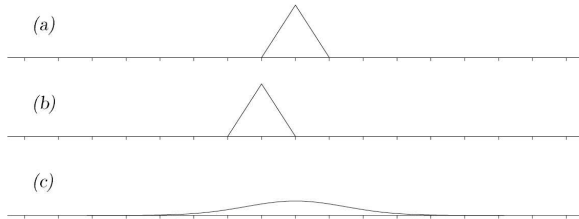
such that partial derivatives $\partial_{x_j} \mathbf{f}_s^j$ have certain properties.

Think of $d = 1, 2, 3$ as dimensionality of spatial domain. Think of s as number of variables, e.g. (u, v, w) but also temperature, concentration.

Conservation law for scalar u :

$$\int_{\Omega} d\Omega \partial_t u(x, t) = \int_{\Omega} d\Omega \nabla \cdot \mathbf{f}(u)$$
$$\frac{d}{dt} \int_{\Omega} d\Omega u(x, t) = \int_{\partial\Omega} \mathbf{f}(u) \cdot \mathbf{n}$$

i.e. $\mathbf{f}(u)$ is the flux of u and change in quantity of u in domain is only due to flux in or out through boundary $\partial\Omega$ (no source or sink terms in bulk).



Initial condition

Wave: $\partial_t u = \partial_x u$

Heat: $\partial_t u = \partial_{xx} u$

Hyperbolic PDE of order n is well-posed initial value problem for first $n - 1$ derivatives.

For first order

$$\partial_t u = c \partial_x u$$

must specify $u(x, t = 0)$.

For second-order

$$\partial_{tt} u = c^2 \partial_{xx} u$$

must specify initial conditions $u(x, t = 0)$ and $\partial_x u(x, t = 0)$.

Finite difference methods for solving hyperbolic problems

$$\partial_t u = \partial_x u$$

Leapfrog:

$$u(x, t + \Delta t) = u(x, t - \Delta t) + \frac{\Delta t}{\Delta x} (u(x + \Delta x, t) - u(x - \Delta x, t))$$

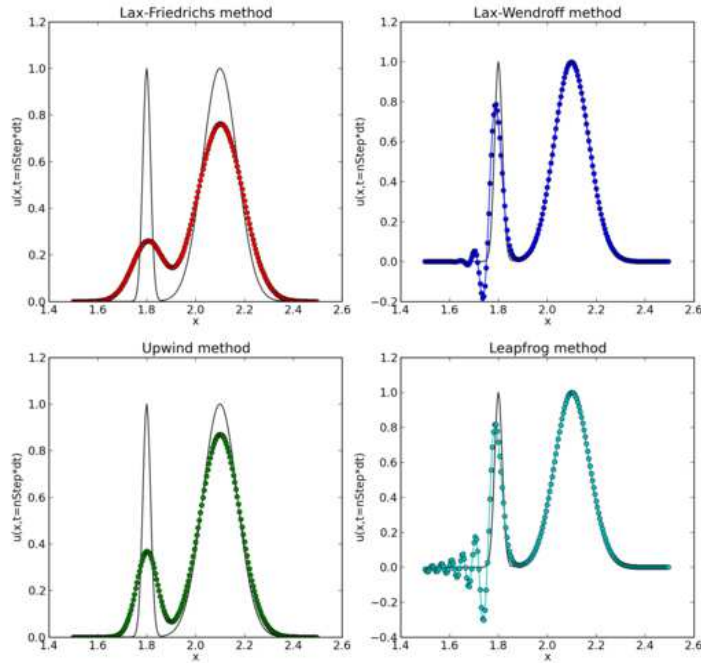
Upwind:

$$u(x, t + \Delta t) = u(x, t) + \frac{\Delta t}{\Delta x} (u(x + \Delta x, t) - u(x, t))$$

Lax-Wendroff:

$$\begin{aligned} u(x, t + \Delta t) = & u(x, t) + \frac{\Delta t}{2\Delta x} (u(x + \Delta x, t) - u(x - \Delta x, t)) \\ & + \frac{\Delta t^2}{2\Delta x^2} (u(x + \Delta x, t) - 2u(x, t) + u(x - \Delta x, t)) \end{aligned}$$

Advection equation: $\partial_t u + a \partial_x u = 0$



Leapfrog suffers from spurious high-wavenumber oscillations.
Lax-Wendroff damps these oscillations.

Accuracy of Leapfrog Method

$$u(x, t + \Delta t) = u(x, t - \Delta t) + \frac{\Delta t}{\Delta x} (u(x + \Delta x, t) - u(x - \Delta x, t))$$

$$u(x, t - \Delta t) = (u - \Delta t u_t + \frac{1}{2}(\Delta t)^2 u_{tt} + \dots) (x, t)$$

$$\frac{\Delta t}{\Delta x} u(x + \Delta x, t) = \frac{\Delta t}{\Delta x} (u + \Delta x u_x + \frac{1}{2}(\Delta x)^2 u_{xx} + \dots) (x, t)$$

$$-\frac{\Delta t}{\Delta x} u(x - \Delta x, t) = -\frac{\Delta t}{\Delta x} (u - \Delta x u_x + \frac{1}{2}(\Delta x)^2 u_{xx} + \dots) (x, t)$$

$$\text{Sum} = (u - \Delta t u_t + 2\Delta t u_x + \frac{1}{2}(\Delta t)^2 u_{tt} + \dots) (x, t)$$

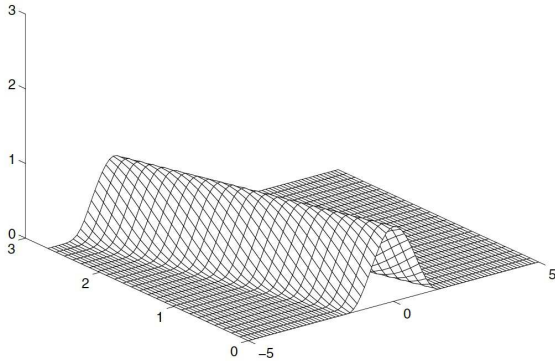
$$u(x, t + \Delta t) = (u + \Delta t u_t + \frac{1}{2}(\Delta t)^2 u_{tt} + \dots) (x, t)$$

$$\text{Error} = 2\Delta t (u_t - u_x) + O(\Delta x, \Delta t)^3 = O(\Delta x, \Delta t)^3$$

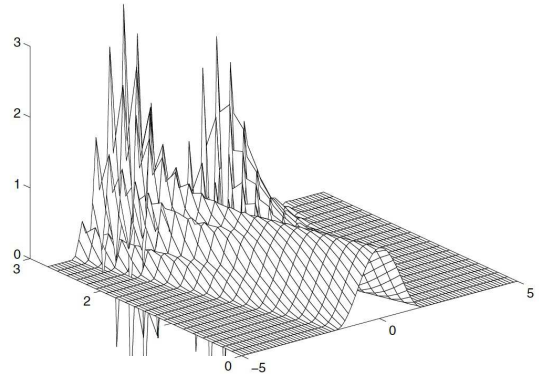
Stability of Leapfrog Method

Leapfrog method used to solve $\partial_t u = \partial_x u$

with initial condition: $u(x, t = 0) = \begin{cases} \cos^2 x & |x| \leq \frac{\pi}{2} \\ 0 & |x| \leq \frac{\pi}{2} \end{cases}$



$$\lambda = \frac{\Delta t}{\Delta x} = 0.9$$



$$\lambda = \frac{\Delta t}{\Delta x} = 1.1$$

Fourier or von Neumann stability analysis: insert $e^{i\xi x}$

$$u(j\Delta x, n\Delta t) = z^n e^{i\xi x_j} \quad \lambda \equiv \Delta t / \Delta x$$

Leapfrog: $z^{n+1} e^{i\xi x_j} = z^{n-1} e^{i\xi x_j} + \lambda (z^n e^{i\xi x_{j+1}} - z^n e^{i\xi x_{j-1}})$

$$z = z^{-1} + \lambda (e^{i\xi\Delta x} - e^{-i\xi\Delta x})$$

$$z - z^{-1} = 2i\lambda \sin(\xi\Delta x)$$

$$u(x, t + \Delta t) = u(x, t - \Delta t) + \frac{\Delta t}{\Delta x} (u(x + \Delta x, t) - u(x - \Delta x, t))$$

Upwind: $z^{n+1} e^{i\xi x_j} = z^n e^{i\xi x_j} + \lambda (z^n e^{i\xi x_{j+1}} - z^n e^{i\xi x_j})$

$$z = 1 + \lambda (e^{i\xi\Delta x} - 1)$$

Lax-Wendroff: $z^{n+1} e^{i\xi x_j} = z^{n-1} e^{i\xi x_j} + \frac{1}{2} z^n \lambda (e^{i\xi x_{j+1}} - e^{i\xi x_{j-1}})$

$$+ \frac{1}{2} \lambda^2 z^n (e^{i\xi x_{j+1}} - 2e^{i\xi x_j} + e^{i\xi x_{j-1}})$$

$$z = z^{-1} + \frac{1}{2} \lambda (e^{i\xi\Delta x} - e^{-i\xi\Delta x}) + \frac{1}{2} \lambda^2 (e^{i\xi\Delta x} - 2 + e^{-i\xi\Delta x})$$

$$z = z^{-1} + \lambda i \sin(\xi\Delta x) + \lambda^2 (\cosh(\xi\Delta x) - 1)$$

$$z - z^{-1} = 2i\lambda \sin(\xi\Delta x)$$

$$0 = z^2 - 1 - 2i\lambda \sin(\xi\Delta x)$$

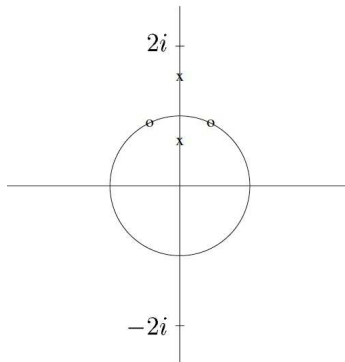
$$z = i\lambda \sin(\xi\Delta x) \pm \sqrt{1 - \lambda^2 \sin^2(\xi\Delta x)}$$

$$\lambda < 1 \implies 1 - \lambda^2 \sin^2(\xi\Delta x) > 0 \implies a = \sqrt{1 - \lambda^2 \sin^2(\xi\Delta x)} \quad \text{real}$$

$$|z|^2 = \lambda^2 \sin^2(\xi\Delta x) + 1 - \lambda^2 \sin^2(\xi\Delta x) = 1$$

$$\lambda > 1 \implies 1 - \lambda^2 \sin^2(\xi\Delta x) < 0 \implies a = \sqrt{1 - \lambda^2 \sin^2(\xi\Delta x)} \quad \text{imag}$$

$$|z|^2 = (i\lambda \sin(\xi\Delta x) + a)^2 > 1 \text{ for values of } \xi\Delta x \text{ surrounding } \pi/2$$

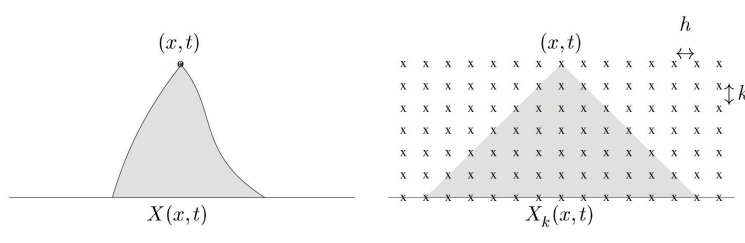


$|z| > 1 \implies$ **growth in time**

o $\lambda = 0.9$

x $\lambda = 1.1$

Domain of Dependence (Courant, Friedrichs, Lewy = CFL)



continuous system

discretized system

$X(x, t)$ is the set of all points x where the initial data $u(x, t = 0)$ may have some effect on the solution $u(x, t)$.

In parabolic systems such as $u_t = u_{xx}$, information travels infinitely fast. Magnitude of influence of faraway data may decay exponentially with distance but influence will still be present. $X(x, t)$ is the entire real line.

In hyperbolic systems such as $u_t = u_x$ or $u_{tt} = u_{xx}$, perturbations travel at finite speed and so $X(x, t)$ is finite for each x and t . Curves which bound $X(x, t)$ are the characteristic curves.

For convergence of a numerical approximation of a PDE, the continuous domain of dependence must be contained in the limiting numerical domain of dependence as $k = \Delta t \rightarrow 0$.