

# FINITE DIFFERENCE METHODS

①

## for ELLIPTIC PDES

A. DONEV, COURANT

We will consider boundary value problems  
BVPs, mostly linear ones:

$$\begin{cases} \mathcal{L} u(\vec{x} \text{ : } \cdot) = f(\vec{x} \text{ : } \cdot) & \text{in } \Omega \\ \mathcal{B} u(\vec{x} \in \partial\Omega \text{ : } \cdot) = g(\vec{x} \in \partial\Omega, \cdot) & \text{on } \partial\Omega \end{cases}$$

where  $\mathcal{L}$  and  $\mathcal{B}$  are linear  
(differential) operators.

Note that domain  $\Omega$  can be unbounded  
But this is usually harder

Another important example in 1D are (2)  
Sturm-Liouville (SL) problems

$$\begin{cases} \mathcal{L}u = - (p(x)u'(x))' + q(x)u(x) \\ p(x) > 0, \quad q(x) > 0 \quad \text{on } [a, b] \end{cases}$$

with either periodic "boundary" conditions (PBCs), or BC's of the form:

$$\underbrace{\alpha u'(a)}_{\text{Neumann}} + \underbrace{\beta u(a)}_{\text{Dirichlet}} = \underbrace{f(t)}_{\text{If non-zero inhomogeneous}}$$

Mixed / Robin

On paper, it does not really make sense to write something like (3)

$$u = \mathcal{L}^{-1} f$$

because  $\mathcal{L}$  is an infinite dimensional operator and inverse is unclear (no Gaussian elimination)

Instead, we use eigenfunctions and eigenvalues of  $\mathcal{L}$

Consider homogeneous BCs first:

$$\left\{ \begin{array}{l} \mathcal{L}u = f \\ Bu = 0 \end{array} \right. \implies \text{eigenproblem} \quad \left\{ \begin{array}{l} \mathcal{L}u_k = \lambda_k u_k \\ Bu = 0 \end{array} \right.$$

$(\lambda_k, u_k)$  are eigensolutions,  $k = 0, 1, 2, 3, \dots$

If  $\mathcal{L}$  is Hermitian or self-adjoint (4)  
(symmetric),  $\boxed{\mathcal{L}^* = \mathcal{L}}$ , as is the  
case for SL BVPs, then the eigenvalues  
are real, and eigenvectors/functions are  
enumerable and orthogonal and a complete  
basis (let's not be too technical here).

If  $\mathcal{L}$  is symmetric positive definite  
(SPD), as it is for SL BVPs, then  
all  $\lambda > 0$  ( $\lambda \geq 0$  if SPD)

BVP has unique solution if  $\lambda \neq 0$

Since eigenspace is a complete basis for  $L_2$ , we can expand both solution & r.h.s. into a generalized Fourier Series: (5)

$$u = \sum_n a_n u_n$$

$$f = \sum_n b_n u_n$$

$$\mathcal{L}u = \sum_n a_n (\mathcal{L}u_n) = \sum_n \lambda_n a_n u_n = f = \sum_n b_n u_n$$

$$\Rightarrow \boxed{a_n = \frac{b_n}{\lambda_n}}$$

If we have Inhomogeneous BCs (6)  
then we only need to find one  
particular solution of

$$\begin{cases} \mathcal{L}\bar{u} = 0 \\ B\bar{u} = g \end{cases}$$

$$\Rightarrow u = \bar{u} + \underbrace{\sum_n a_n u_n}_{\tilde{u}}$$
$$= \bar{u} + \tilde{u}$$

where

$$\begin{cases} \mathcal{L}\tilde{u} = f \\ B\tilde{u} = 0 \end{cases}$$


Superposition principle

Another approach on paper is ⑦  
to use the Green's function for the  
PDE.  $G(\vec{x}, \vec{y})$

$$\begin{cases} \Delta G(\vec{x}) = \delta(\vec{y} \in \Omega) \\ \partial G(\vec{x}) = 0 \end{cases}$$

$\Rightarrow$  quadrature

$$\bar{u}(\vec{x} \in \Omega) = \bar{u} + \int_{\vec{y} \in \Omega} f(\vec{y}) G(\vec{x}, \vec{y}) d\vec{y}$$

The problem: We cannot analytically  
compute eigenfunctions or Green's functions  


Poisson eq. in 1D

is easy though:

(8)

$$\left\{ \begin{array}{l} u''(x) = f(x) \end{array} \right.$$

$$\boxed{L = b - a}$$

example:  $u(a) = u_1, \quad u(b) = u_2$

First inhomogeneous:

$$\bar{u}''(x) = 0 \Rightarrow \bar{u}(x) = \alpha x + \beta \quad (\text{linear})$$

$$\bar{u}(a) = u_1, \quad \bar{u}(b) = u_2 \Rightarrow \bar{u}(x) = \frac{u_2 - u_1}{b - a} (x - a) + u_1$$

Now homogeneous:

$$\left\{ \begin{array}{l} u'' = f(x) \\ u(a) = u(b) = 0 \end{array} \right.$$

$$\lambda_k = \left( \frac{2\pi k}{L} \right)^2$$

Eigenfunctions:

$$\begin{aligned} u_k'' &= \lambda_k u_k \\ u_k(a) &= u_k(b) = 0 \end{aligned}$$

$$\Rightarrow \boxed{u_k = \sin\left(\frac{2\pi k}{L}(x - a)\right)}$$



Or, use Green's function

(9)

$$\begin{cases} u'' = \delta(y) \Rightarrow \\ u'(y^+) - u'(y^-) = 1 \\ u(a) = u(b) = 0 \end{cases}$$

$\Rightarrow -G(x, y)$



jump in slope = 1

$$\Rightarrow G(x, y) = \frac{1}{b-a} \begin{cases} (y-b)(x-a), & a \leq x \leq y \\ (y-a)(x-b), & y \leq x \leq b \end{cases}$$

All of these pen-and-paper concepts will appear in our analysis of methods. (10)

However, numerical methods will not be based on either of these approaches,

here is why: in general.

① Green's function is either hard to compute or is not smooth  
( $\delta$  function not good for numerics!)

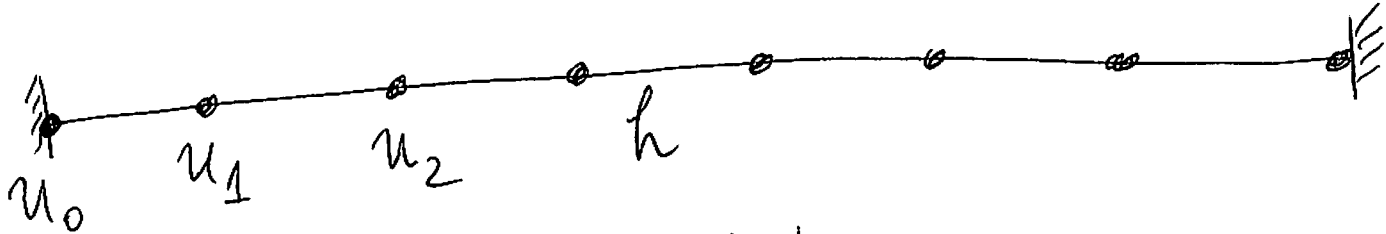
② Computing eigenvalues of a matrix is more expensive than computing  $u = Y^{-1}f$

→ We will solve linear systems on computer

# Finite Difference Operators

(11)

In FD methods, we put down a grid of points with spacing  $h$ , and approximate derivatives using finite differences:



FD pointwise interpretation:

$$u(x = kh) \approx u_k$$

$u'(kh)$ ,  $u''(kh)$  etc. approximated from pointwise values

$$u'(\bar{x}) \approx \left\{ \begin{array}{l} (D_+ u)(\bar{x}) = \frac{u(\bar{x}+h) - u(\bar{x})}{h} + O(h) \quad (12) \\ \uparrow \\ \text{one-sided difference} \quad \text{OR} \end{array} \right.$$

$$(D_- u)(\bar{x}) = \frac{u(\bar{x}) - u(\bar{x}-h)}{h} + O(h) \quad \text{OR}$$

two-sided  
centered  
approximation

$$(D_0 u)(\bar{x}) = \frac{u(\bar{x}+h) - u(\bar{x}-h)}{2h} + O(h^2)$$

OR

$$(D_3 u)(\bar{x}) = \frac{1}{6h} [2u(\bar{x}+h) + 3u(\bar{x}) - 6u(\bar{x}-h) + u(\bar{x}-2h)] + O(h^3)$$

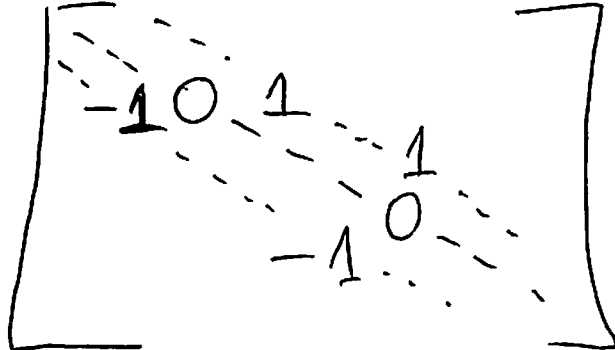
OR

...

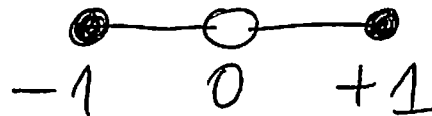
Each formula is characterized by its stencil, the set of neighboring points and their coefficients, e.g.

(13)

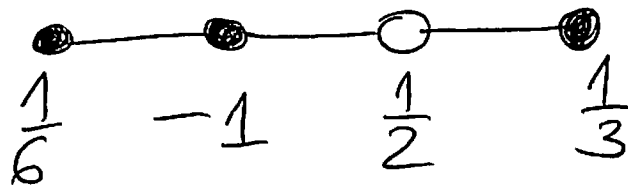
$$(D_0 \vec{u})_j = \frac{u_{j+1} - u_{j-1}}{h}$$

$\Rightarrow D_0 =$    $\leftarrow$  Circulant or Toeplitz matrix far from boundaries or with PBCs

$D_0$  stencil is



$D_3$  stencil is



Truncation error can be computed easily using Taylor series

(14)

E.g.

$$(D_3 u)(\bar{x}) = u'(\bar{x}) + \frac{h^3}{12} u^{(4)}(\bar{x}) + O(h^4)$$

One can most easily construct finite-difference formulas by fitting a polynomial through the points of the stencil and then differentiating that.

If the polynomial is of order  $p$  ( $p+1$  points in stencil) then error =  $O(h^{p-1})$

# Second-order derivatives

In 1D, by far the simplest and most common is the centered second order

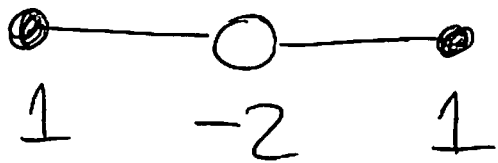
(15)

$$D^2 = D_+ D_-$$

$$D^2 u(\bar{x}) = \frac{u(\bar{x}-h) - 2u(\bar{x}) + u(\bar{x}+h)}{h^2}$$

$$= u''(\bar{x}) + \frac{h^2}{12} u^{(3)}(\bar{x}) + O(h^4)$$

$\frac{1}{h^2}$



No  $O(h^3)$  because of symmetry - only even powers

$$D^2 = \begin{bmatrix} -2 & 1 & & & \\ & 1 & -2 & 1 & \\ & & 1 & -2 & 1 \\ & & & 1 & -2 \\ & & & & 1 & -2 \\ & & & & & 1 & -2 \\ & & & & & & 1 & -2 \\ & & & & & & & 1 & -2 \\ & & & & & & & & 1 & -2 \end{bmatrix}$$

as a matrix

Observe how this is different and better than

(16)

$$\tilde{D}^2 = D_0^2 \equiv \frac{1}{h^2} \begin{array}{ccccc} \textcircled{\bullet} & \bullet & \circ & \bullet & \textcircled{\bullet} \\ \frac{1}{2} & & -1 & & \frac{1}{2} \end{array}$$

In this wide stencil, odd and even points on the grid are uncoupled:

We can change all even values and not change the result for odd points.

$\tilde{D}^2$  has a non-trivial null space of checkerboard solutions!

Accuracy is not everything!



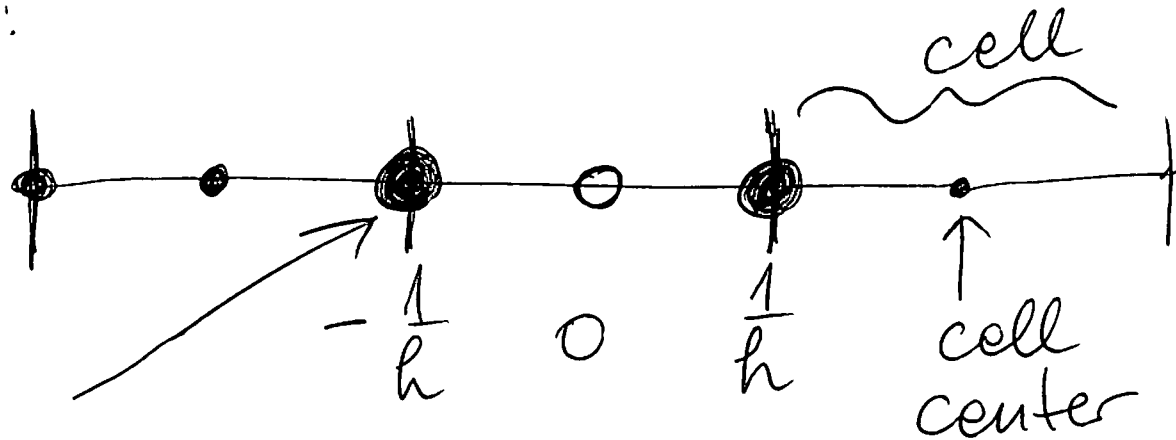
An alternative view of  $D^2$ :

(17)

Define a staggered centered difference

$$u'(\bar{x}) \approx (\hat{D}_0 u)(\bar{x}) = \frac{1}{h} \left[ u\left(x + \frac{h}{2}\right) - u\left(x - \frac{h}{2}\right) \right]$$

stencil:



node or  
face center  
or vertex  
depending on  
context.

Claim (check)

$$\hat{D}_0^T = -\hat{D}_0$$

on an unbounded domain  
(formal)

More generally define:

(18)

$\hat{D}_0$  : cell centers  $\rightarrow$  nodes  
 $u^{(c)}$   $u^{(F)}$

$\hat{D}_0^*$  : nodes  $\rightarrow$  cell centers  
adjoint

$$\left(\hat{D}_0 \vec{u}^{(c)}\right)_{j+\frac{1}{2}} = \frac{u_{j+\frac{1}{2}}^{(c)} - u_j^{(c)}}{h}$$

$$\left(\partial_x\right)^* = -\partial_x \quad (\text{integration by parts})$$

$$\left(\hat{D}_0^* \vec{u}^{(F)}\right)_j = -\frac{u_{j+\frac{1}{2}}^{(F)} - u_{j-\frac{1}{2}}^{(F)}}{h}$$

Claim:

$$D^2 = -\hat{D}_0 \hat{D}_0^*$$

(19)

which is a discrete equivalent of

$$\nabla^2 u = \nabla \cdot (\nabla u)$$

$$\Delta = \text{div} \cdot \text{grad}$$

This can be generalized to higher dimensions

$$L = D G = -D D^*$$

Laplacian      divergence      gradient

(including boundary conditions)

$$(-\hat{D}_0 \hat{D}_0^* u)_j = (D^2 u)_j = \frac{(u_{j+1} - u_j) - (u_j - u_{j-1}))}{h}$$

Immediate corollary:

(20)

①  $D^2$  is negative semidefinite  
(as a matrix)

just as the Laplacian as an operator

②  $D^2$  has only constant fields  
as null-vectors  $\vec{u}_j = \text{const}$

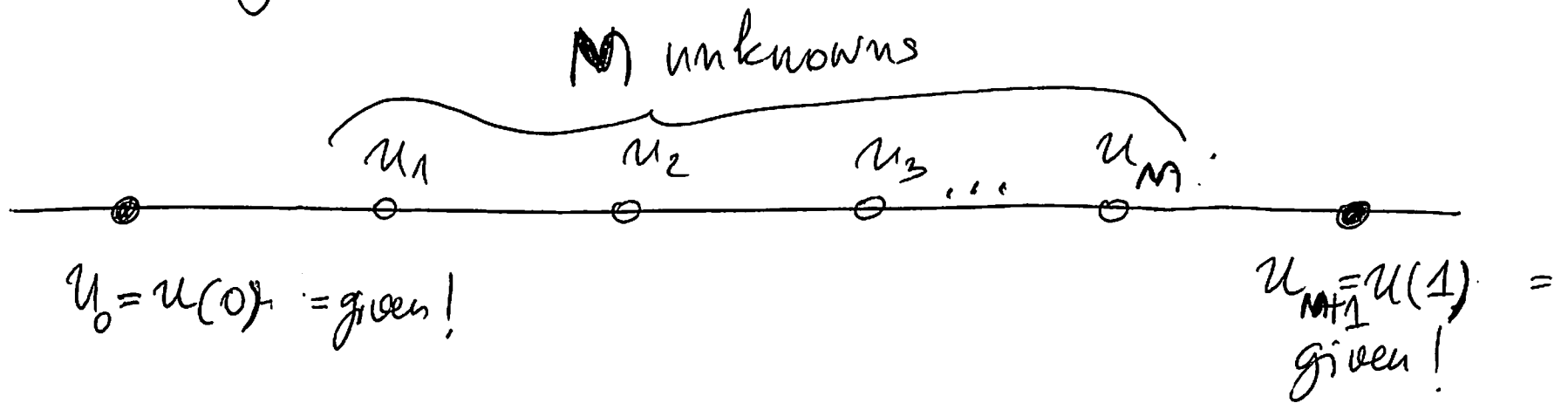
also like the continuous Laplacian

DIY: at home  
Write a finite difference discretization  
of  $\mathcal{L} = \frac{d}{dx} c(x) \frac{d}{dx}$  where  $c(x) > 0$   
that you can prove is negative  
(semi) definite

What about boundary conditions?

(21)

The easiest is a Dirichlet BC with an FD grid aligned with the boundary:



$u_0$  and  $u_{M+1}$  are not real variables, they are known constants (or functions of time for parabolic PDEs)

E.g. 
$$\begin{cases} u''(x) = f(x) & x \in [0, 1] \\ u(x=0) = \alpha \\ u(x=1) = \beta \end{cases} \quad (22)$$

Convert into:

$$\begin{cases} (D^2 u)_j = \frac{1}{h^2} (u_{j-1} - 2u_j + u_{j+1}) = f(x_j) \\ j=1, \dots, m \\ u_0 = \alpha \\ u_{m+1} = \beta \end{cases}$$

This has the form

$$\tilde{A} \underline{\underline{u}} = \underline{\underline{f}}$$

$$\underline{\underline{u}} = \begin{bmatrix} u_0 \\ \vdots \\ u_{m+1} \end{bmatrix}$$

But here  $\tilde{A}$  is not a matrix  
 but rather an affine linear operator (23)  
 because  $u_0$  and  $u_{m+1}$  are constants

$$\tilde{A} \underline{\vec{u}} = \tilde{A} \begin{bmatrix} 0 \\ u_1 \\ \vdots \\ u_m \\ 0 \end{bmatrix} + \tilde{A} \begin{bmatrix} \alpha \\ \vdots \\ \beta \end{bmatrix}$$

(or ghost)  
 boundary  
 interior

$$= \tilde{A} \underline{\vec{u}}_{int} + \tilde{A} \underline{\vec{u}}_{BC}$$

where now  $\underline{\vec{u}}_{int} = (0, u_1, \dots, u_m, 0)$  so

that  $\tilde{A} \underline{\vec{u}}$  contains just the part

from homogeneous BCs

We can instead consider

$$\vec{u} = \begin{bmatrix} u_1 \\ \vdots \\ u_m \end{bmatrix} \text{ as the unknown}$$

(24)

and write

$$A \vec{u} = \vec{f} \text{ (inhomogeneous part)} = \vec{f}$$

$$A = \begin{bmatrix} -2 & 1 & & & & \\ & 1 & \ddots & & & \\ & & \ddots & \ddots & & \\ & & & \ddots & 1 & \\ & & & & 1 & -2 \end{bmatrix} \rightarrow \vec{f} = \begin{bmatrix} f_1 - \alpha/h^2 \\ f_2 \\ \vdots \\ f_{m-1} \\ f_m - \beta/h^2 \end{bmatrix}$$

OR  $\tilde{A} \vec{u} = \vec{f} + \begin{cases} u_0 = \alpha \\ u_{m+1} = \beta \end{cases}$  (extended)



Note that in many "real life" codes one uses  $\sim A$  and not  $A$ , just so that the code can handle arbitrary BCs with little changes, and for efficiency (no if statements in loops). (25)

The values  $u_0$  and  $u_{m+1}$  are then called ghost or virtual cells.

E.g. useful for periodic BCs where ghost cells are images (copies) of interior values on the other side of the domain.

We have converted the BVP  
into a linear system

(26)

$$\boxed{Au = \tilde{f}}$$

the tricky part?

The size of  $A$  (dimension) grows as  
we add more points, Is there a  
clear limit as  $M \rightarrow \infty$ ?

We will answer this next lecture.

For now let's discuss a bit  
other BCs

Periodic BCs :

$$A = \begin{bmatrix} -2 & 1 & & & \textcircled{1} \\ & 1 & \dots & & \\ & & \dots & & \\ \textcircled{1} & & & & \\ & & & & 1 \\ & & & 1 & -2 \end{bmatrix}$$

← Symmetric  
(negative definite!)  
just as Dirichlet

②⑦

A is a circulant matrix.

Observe A is diagonalized by the Fourier transform (why?):

$$A = F \Lambda F^*$$

(i) FFT      ↑      (i) FFT

↑

diagonal  
of eigenvalues.

Exercise:  
Compute  $\Lambda$

(will appear in homework X) later

How about Neumann BCs, e.g.

(28)

$$u'(0) = 6$$

Now  $u_0$  is a real variable:



$$\frac{du}{dx} = 6$$

First idea (section 2.12 in LeVeque)

$$\frac{u_1 - u_0}{h} = 6 \quad \leftarrow \text{use one-sided difference to implement BC (first order)}$$

Instead of  $u_1 = \alpha$

In implementation, we can either add  $u_0$  as a variable and this extra equation, or, eliminate

(29)

$$u_0 = u_1 - h\sigma$$

as a ghost cell value. This really means we are extrapolating from the interior to the ghost cells using the BCs.

What error do we make?

Local truncation error (LTE)

$$\tau_0 = \frac{u(x_1) - u(x_0)}{h} - \sigma = \frac{1}{2} h u''(x_0) + O(h^2)$$

So this is only first order accurate at the boundary. Will this ruin the overall second-order global accuracy? (30)

We will figure that out next time.

### Second idea

Use a ghost point  $u_{-1}$  and a second-order centered difference

$$\frac{1}{2h} (u_1 - u_{-1}) = \sigma \Rightarrow$$

$$\Rightarrow u_{-1} = u_1 - 2h\sigma$$

Extrapolate linearly to the ghost cell!

Let's see if this works.

(31)

$$\frac{1}{h^2} (u_{-1} - 2u_0 + u_1) = \tau_0 = f(x_0)$$

$$\uparrow u_{-1} = u_1 - 2h\tau$$

Becomes

$$\frac{1}{h} (u_1 - u_0) = \tau + \frac{h}{2} f(x_0)$$

Compare to Taylor series:

$$\frac{1}{h} (u(x_1) - u(x_0)) = \tau + \frac{h}{2} u''(x_0)$$

So now it is second-order!

|||  
 $f(x_0)$  by PDE

### Third idea

Use a second-order three-point  
one-sided difference

(32)

$$u'(x_0) \approx \frac{1}{h} \left( \frac{3}{2}u_0 - 2u_1 + \frac{u_2}{2} \right) = \sigma$$

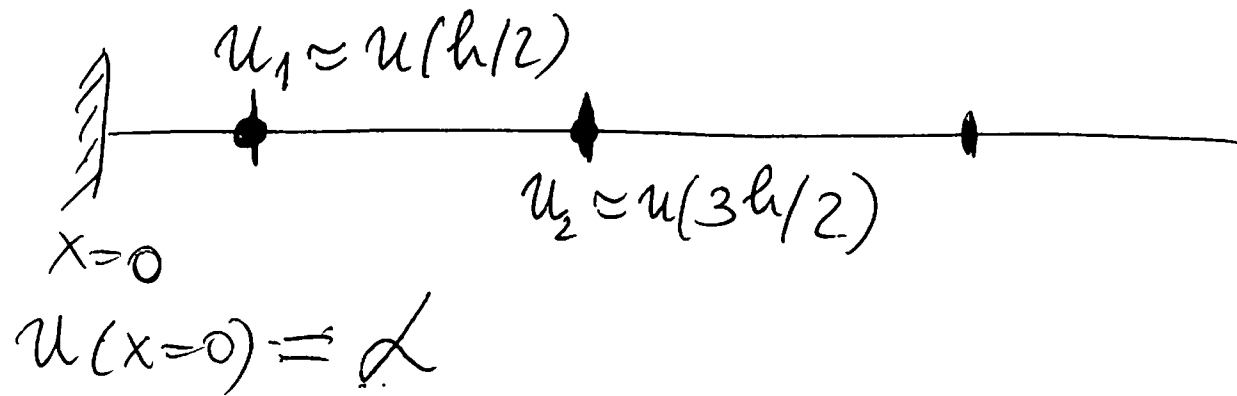
Matrix becomes

$$\frac{1}{h^2} \begin{bmatrix} 3h/2 & -2h & h/2 & & & \\ 1 & -2 & 1 & & & \\ & 1 & -2 & 1 & & \\ & & & \ddots & \ddots & \ddots & \ddots \end{bmatrix} \leftarrow \text{NOT symmetric!}$$

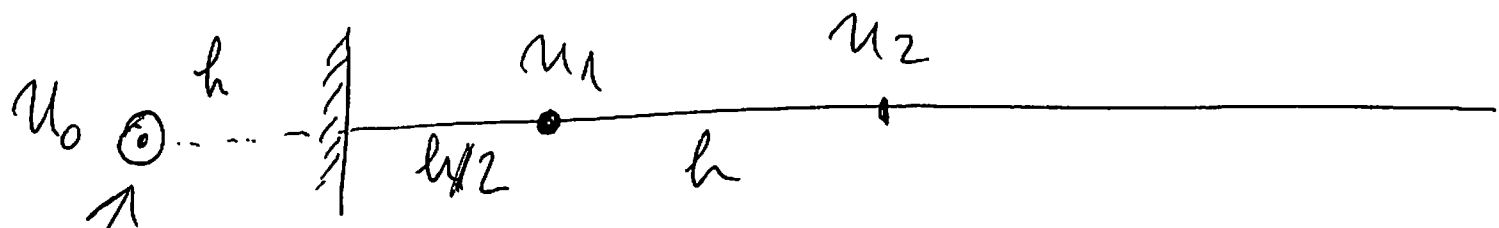
Which method is "best"? Depends...  
Neither matrix symmetric!?!



Note that sometimes even for Dirichlet BCs we may need extrapolation/ghosts or one-sided, because the grid may be staggered by  $h/2$  from the boundary



Examples: Finite volume methods for mixed parabolic-hyperbolic laws, (Navier-) Stokes with staggered pressure/velocity grids, magnetic fields and charges, etc.



$$u(x=0) = \alpha$$

ghost cell  $u_0 \approx u(x = -h/2)$  (makes no sense, but...)

B.C: 
$$\frac{u_1 + u_0}{2} \approx \alpha = u(x=0)$$

$$u_0 = 2\alpha - u_1$$

Now

$$\frac{1}{h^2}(u_0 - 2u_1 + u_2) = f(h/2)$$

$$\frac{1}{h^2}(-3u_1 + u_2) = f_{1/2} + \frac{2\alpha}{h^2}$$

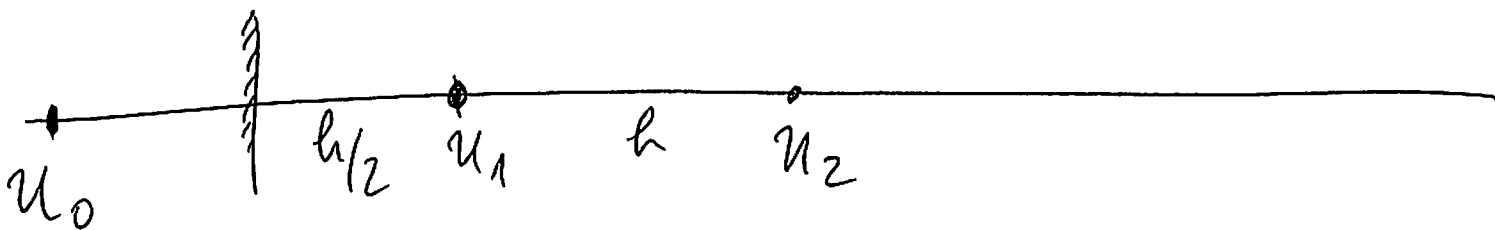
Now the matrix is

(35)

$$\frac{1}{h^2} \begin{bmatrix} -3 & 1 & & \\ 1 & -2 & 1 & \\ & 1 & -2 & \dots \\ & & & \dots & 1 \end{bmatrix}$$

so it is symmetric  
(and in fact negative definite!)

If we had Neumann with staggered



$$\frac{u_1 - u_0}{h} \approx u'(0) = \beta$$

$$\Rightarrow u_0 = u_1 - \beta h$$

$$\frac{1}{h^2} (u_0 - 2u_1 + u_2) = \frac{1}{h^2} (-u_1 + u_2) - \frac{\beta}{h} = f_{1/2}$$

ALSO SYMMETRIC

Of course, we can use higher-order differences, e.g.

(36)

$$\frac{1}{12h^2} [-u_{j-2} + 16u_{j-1} - 30u_j + 16u_{j+1} - u_{j+2}] = u''(jh) + O(h^4) = f(jh)$$

Matrix is  $1/12h^2 *$

$$\begin{bmatrix} -30 & 16 & -1 & & & & & & 1 & 16 \\ 16 & -30 & 16 & -1 & & & & & 0 & 1 \\ -1 & 16 & -30 & 16 & -1 & & & & & \\ 0 & -1 & 16 & -30 & 16 & 1 & & & & \\ & & & & & & & & & \\ & & & & & & & & & \\ & & & & & & & & & \\ & & & & & & & & & \\ & & & & & & & & & \\ & & & & & & & & & \\ & & & & & & & & & \end{bmatrix}$$

Periodic BCS

Is this NEGATIVE SEMIDEFINITE?

TRY to prove it

Will numerical solution satisfy MAXIMUM PRINCIPLE?

LESS SPARSE!

# Addendum to FD for 1D BVPs

(37)

I asked in class for you to think about how you would discretize using finite differences the BVP

$$(k(x) u'(x))' = f(x)$$

which models heat conduction in a non-uniform rod.

The wrong thing to do (go back to lecture on spectral methods) is to use the chain rule

~~$k u'' + k' u' = f(x)$~~  !  
wrong

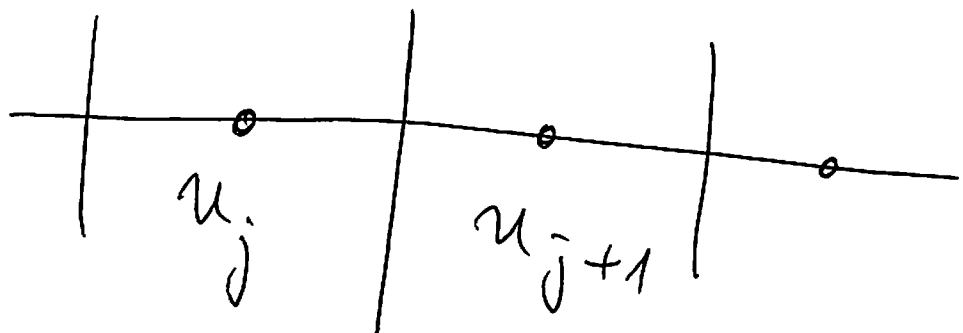
Instead, use physics

(38)

$$(ku')' = (\psi)'$$

where  $\psi = ku' = \text{flux}$

Put flux on a staggered grid  
(related to finite volume methods)



$$\psi_{j+1/2} = K_{j+1/2} \frac{u_{j+1} - u_j}{h}$$

where  $K_{j+1/2} = K(x_{j+1/2})$

Now

$$(\varphi')_j = \frac{\varphi_{j+1/2} - \varphi_{j-1/2}}{h}$$

(39)

$$\Rightarrow (Ku')'_i = \frac{1}{h} \left[ K_{i+1/2} \left( \frac{u_{i+1} - u_i}{h} \right) - K_{i-1/2} \left( \frac{u_i - u_{i-1}}{h} \right) \right]$$

$$(Ku')' = Au \quad \text{where}$$

$A$  is a discretization of the operator

$$\mathcal{L} = \frac{d}{dx} K(x) \frac{d}{dx} \quad \text{where } K(x) > 0$$

$\mathcal{L}$  is a negative semidefinite operator  
with constant fields in its null space  
(depends on BCs)

Claim :

- ①  $A$  is a symmetric and negative definite (for Dirichlet BCs)

40

How to prove it? (in class on board)

- ② The numerical solution satisfies a maximum principle, i.e., the extremal values are achieved on the boundary

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For simple Poisson, observe for  $f(x)=0$

$$\frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} = 0 \quad \text{implies}$$

the DISCRETE mean value property  $u_i = \frac{u_i + u_{i+1}}{2}$



$$u_i = \frac{K_{i+1/2} u_{i+1} + K_{i-1/2} u_{i-1}}{K_{i+1/2} + K_{i-1/2}}$$

(41)

$$\Rightarrow \boxed{u_i = w_i u_{i-1} + (1-w_i) u_{i+1}}$$

where  $0 < w_i < 1$

So indeed  $u_i$  cannot be larger than the larger of  $u_{i-1}$  and  $u_{i+1}$  (same for smaller), and so extremum is on the boundary.

LOOK AT PHYSICS TO INFORM  
GOOD DISCRETIZATIONS