Lecture 8

Chebyshev collocation method for differential equations

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Two point boundary value problem Time <u>independent</u> – linear/non-linear

• Time independent boundary value problem in a general form

$$\varepsilon u_{xx}(x) + p(x)u_x(x) + q(x)u_x(x) = f(x,u), \quad -1 \le x \le 1$$

 $\varepsilon > 0$ is a fixed parameter and $p(x)$, $q(x)$ and $f(x)$ are given functions

• Boundary conditions on general (mixed) form:

$$\alpha_{-}u(-1) + \beta_{-}u_{x}(-1) = g_{-}$$

 $\alpha_{+}u(1) + \beta_{+}u_{x}(1) = g_{+}$

- Dirichlet boundary conditions: $u(-1) = g_ u(1) = g_+$
- Neumann boundary conditions: $u_x(-1) = g_ u_x(1) = g_+$ The coefficients α_{\pm} β_{\pm} g_{\pm} are known

Time dependent problems - linear/non-linear

Heat equation

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u_t - \varepsilon u_{xx} = 0 with boundary conditions given
on u(-1,t) = g_- and u(1,t) = g_+ and initial condition u(x,0) = f(x)
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• Linear wave equation

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u_t + u_x = 0 boundary conditions given
on u(-1,t) = g_- and u_x(1,t) = g_+ and initial condition u(x,0) = f(x)
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• Burgers equation

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u_t + uu_x - \varepsilon u_{xx} = 0 with boundary conditions given
on u(-1,t) = g_- and u(1,t) = g_+ and initial condition u(x,0) = f(x)
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Review on the Chebyshev transform

• A function u(x) can be expanded in Chebyshev series

$$u(x) = \sum_{k=0}^{\infty} \hat{u}_k T_k(x) \qquad \hat{u}_k = \frac{2}{\pi c_k} \int_{-1}^{1} u(x) T_k(x) \omega(x) dx$$

$$\omega(x) = (1 - x^2)^{-1/2}$$
 and $c_k = 2$ for $k = 0$ and $c_k = 1$ for $k \ge 1$

- The Chebyshev polynomials are given by $T_k(x) = \cos(k \cos^{-1}(x))$
- In the discrete Chebyshev-Gauss-Lobatto case:

$$x_j = \cos \frac{\pi j}{N}, j = 0, 1, 2..., N$$
 and $\omega_j = \frac{\pi}{2\tilde{c}_k}$ where

$$\tilde{c}_k = 2 \text{ for } j = 0, N \text{ and } \tilde{c}_k = 1 \text{ for } j = 1, 2, ..., N-1$$

• The discrete Chebyshev coefficients are given by

$$\tilde{u}_k = \frac{2}{\tilde{c}_k N} \sum_{j=0}^N \frac{1}{\tilde{c}_j} u(x_j) \cos\left(\frac{\pi k j}{N}\right) = \frac{2}{\tilde{c}_k N} \sum_{j=0}^N \frac{1}{\tilde{c}_j} u(x_j) \Re\left(e^{i\frac{k j \pi}{N}}\right), \quad k = 0, 1, \dots N$$

Review on the Chebyshev transform, cont.

• A function u(x) can be expanded in discrete Chebyshev series

$$u(x_j) = \sum_{k=0}^{N} \tilde{u}_k T_k(x_j), \quad j = 0, 1, ..., N$$

- Chebyshev interpolant $I_N u = \sum_{k=0}^N \tilde{u}_k T_k(x)$ (1)
- Or in terms of Chebyshev lagrangian polynomials $I_N u = \sum_{j=0}^N u(x_j) \Upsilon_j(x)$ (2)

$$\Upsilon_{j}(x) = \frac{(-1)^{j+1} (1 - x^{2}) T_{N}'(x)}{\tilde{c}_{j} N^{2} (x - x_{j})}$$

• Note that $I_N u(x_i) = u(x_i)$ and that $(1) \Leftrightarrow (2)$

Review on the Chebyshev transform - derivatives

• Derivative in transform (polynomial) space (discrete)

$$\left(I_N u(x_j)\right)' := \left(D_N u\right)_j = \sum_{k=0}^N \tilde{u}_k^{(1)} T_k(x_j)$$

$$\left(I_N u(x_j)\right)'' = \left(D_N^2 u\right)_j = \sum_{k=0}^N \tilde{u}_k^{(2)} T_k(x_j)$$

• The coefficients can be found by a recursive relation

$$\mathbf{c}_{k}\tilde{u}_{k}^{(1)} = \underbrace{\tilde{u}_{k+2}^{(1)}}_{k+2} + 2(k+1)\tilde{u}_{k+1}, \quad k = N-1, N-2, \dots 0$$

$$\mathbf{c}_{k}\tilde{u}_{k}^{(2)} = \underbrace{\tilde{u}_{k+2}^{(2)}}_{k+2} + 2(k+1)\tilde{u}_{k+1}^{(1)} \quad k = N-1, N-2, \dots 0$$

$$\tilde{u}_{N+1}^{(1)} = \tilde{u}_{N}^{(1)} = 0 \quad \text{and} \quad \tilde{u}_{N+1}^{(2)} = \tilde{u}_{N}^{(2)} = 0$$

• Cost of $O(N \log N)$ if FFT is used

Review on the Chebyshev transform - derivatives

• Derivative in "physical" space
$$\left(D_N u \right)_j = \sum_{l=0}^N u(x_l) \Upsilon'_l(x_j) = \sum_{l=0}^N D_{jl} u(x_l)$$

- D_{il} are the entries in the Chebyshev derivative matrix, **D**
- The matrix entries for the second order derivative matrix, \mathbf{D}^2 , are given by $(\mathbf{DD})_{il}$ where \mathbf{DD} is a matrix-multiplication
- The Chebyshev derivative matrix at quadrature points is given by

$$D_{jl} = \begin{cases} \frac{\tilde{c}_{j}(-1)^{j+l}}{\tilde{c}_{l}(x_{j} - x_{l})} & l \neq j \\ -\frac{x_{l}}{2(1 - x_{l}^{2})} & 1 \leq l = j \leq N - 1 \end{cases} \qquad D_{jl} = \begin{cases} \frac{2N^{2} + 1}{6} & l = j = 0 \\ -\frac{2N^{2} + 1}{6} & l = j = N \end{cases}$$

• The matrix approach costs $O(N^2)$

Example of a Dirichlet problem

• We wish to solve, by a Chebyshev collocation method,

$$u_{xx} + xu_x - u = \underbrace{(24 + 5x^2)e^{5x} + (2 + 2x^2)\cos(x^2) - (4x^2 + 1)\sin(x^2)}_{f(x)},$$
on $-1 \le x \le 1$ (1)

• With boundary conditions

$$u(-1) = e^{-5} + \sin(1) = g_{-}$$
 and $u(1) = e^{5} + \sin(1) = g_{+}$

- Let $\mathbf{x} = (x_0, x_1, ..., x_N)^T$ where $x_j = \cos\left(\frac{j\pi}{N}\right)$, j = 0, 1, 2, ..., N
 - are the Chebyshev-Gauss-Lobatto points
- Let $\mathbf{f} = (f(x_0), f(x_1), \dots, f(x_N))^T$
- Let $\mathbf{u} = (g_+, u(x_1), u(x_2), \dots, u(x_{N-1}), g_-)^T$ and $\mathbf{u}_M = (u(x_1), u(x_2), \dots, u(x_{N-1}))^T$ be the vector of unknows to be determined
- We will work with the Chebyshev derivative matrices \mathbf{D} and \mathbf{D}^2
- The entries in **D** are given by D_{ij} , $0 \le i, j \le N$ and $\mathbf{D}^2 = \mathbf{DD}$

Example of a Dirichlet problem, cont.

• The approximation to (1) is given by

$$\mathbf{D}_{M}^{2}\mathbf{u}_{M}-\mathbf{x}_{M}\otimes\mathbf{D}_{M}\mathbf{u}_{M}-\mathbf{u}_{M}=\mathbf{F}$$

•
$$\mathbf{x}_{M} = \mathbf{x}(1:N-1), \quad \mathbf{D}_{M} = \mathbf{D}(1:N-1,1:N-1), \quad \mathbf{D}_{M}^{2} = \mathbf{D}^{2}(1:N-1,1:N-1)$$

$$\mathbf{F} = \mathbf{f}(1:N-1) - \left[\mathbf{D}^{2}(1:N-1,0) + \mathbf{x}_{M} \otimes \mathbf{D}(1:N-1,0)\right]g_{+} - \left[\mathbf{D}^{2}(1:N-1,N) + \mathbf{x}_{M} \otimes \mathbf{D}(1:N-1,N)\right]g_{-}$$

- The approximation can also be written on the form $\mathbf{A}\mathbf{u}_M = \mathbf{F}$ where \mathbf{A} is a matrix of size $(N-1)\times(N-1)$
- The matrix **A** is given by $\mathbf{A} = \mathbf{D}^2 + \Lambda \mathbf{D} \mathbf{I}$ where Λ is a diagonal matrix with the values of \mathbf{x}_M on the diagonal and **I** is the identity matrix
- The approximate solution is given as the solution to the linear system of equations $\mathbf{u}_{M} = \mathbf{A}^{-1}\mathbf{F}$

Solution

10⁰

10⁻²

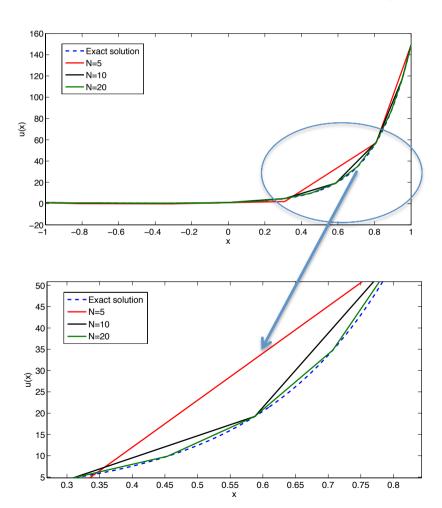
10

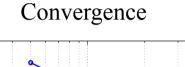
10⁻⁶

10⁻⁸

10-10

- The exact solution is $u(x) = e^{5x} + \sin(x^2)$
- The error is defined by $err = \max_{1 \le j \le N-1} |u_j u(x_j)|$





The error obtained by a second order finite difference approximation with N = 512 is approximately the same as with N = 10 in the spectral method

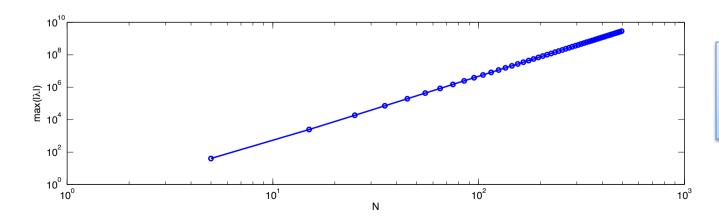
10¹ N 10²

Code example - MATLAB

```
N = 10
                           Chebyshev derivation matrix, D
[D, x] = cheb(N);
                           Chebyshev derivation matrix, D^2
D2=D*D;
                           For the inner points only
DM = D(2:N, 2:N);
D2M=D2(2:N,2:N);
                          Note that the numbering of the
L=diag(x(2:N),0); elements in a vector/matrix is
I=diag(ones(N-1,1),0);
                      from 1 to N+1 (not 0 to N)
                         A as in Au=F
A=D2M+L*DM-I;
f=(24+5*x).*exp(5*x)+(2+2*x.^2).*cos(x.^2);
f=f-(4*x.^2+1).*sin(x.^2);
qminus=exp(-5)+sin(1);
gplus=exp(5)+sin(1);
F=f(2:N)-(D2(2:N,1)+x(2:N).*D(2:N,1))*qplus
F=F-(D2(2:N,end)+x(2:N).*D(2:N,end))*qminus;
sol=A\F
                           u = A^{-1}F
u=[qplus;sol;qminus]; Add the known boundary conditions
```

Remarks

- The collocation method will lead to a full and ill-conditioned linear system
- Gaussian elimination to solve **Au=F** is only feasible for problems with a small number of unknowns (one-dimensional problems)
- For multi-dimensional problems an iterative method together with an appropriate pre-conditioner should be used
- We will get back to this when we speak about multidimensional problems



Eigenvalues of **A** are growing as N⁴