Computational Convergence Study

As $h \to 0$ we want $||\mathsf{E}^h|| \to 0$

$$\underline{\text{Uu}}: \begin{cases} u'' = f(x) & \text{on } (0,1) \Rightarrow (ku')' = f(x_j) + \varepsilon \\ u(0) = \alpha \\ u(1) = \beta \end{cases}$$

For
$$j = 1, 2, ..., m$$

$$\frac{1}{h^2}(U_{j-1} - 2U_j + U_{j+1} = f(x_j)$$

We know if everything goes okay, that this process of discretization will generate an $O(h^2)$ approximation to the exact result. This is a mathematical result that leads to convergence, but there are errors occur in any program we might write/implement.

- Roundoff errors (you cant do shit about that)
- measurement error
- this will work as $h \rightarrow 0$
- what if the problem is more complicated?

Computational convergence analysis

• chose a decreasing sequence of h that makes

$$\{h_0,\frac{h_0}{2},\frac{h_0}{2^2},\frac{h_0}{2^3}\}$$

• for each $h \le h_0$, we can compute an approximation of U

$$U = \begin{bmatrix} U_1 \\ U_2 \\ \dots \\ U_m \end{bmatrix}$$

• evaluate u(x) of parts

$$\hat{U} = \begin{bmatrix} u(x_j1) \\ u(x_j2) \\ u(x_j3) \\ ... \\ u(x_jm) \end{bmatrix}$$

- Compute $||E_h|| = ||U \hat{U}||$
- this gives us the data we need to determine convergence

$$\begin{split} \|E^h\| &\leq Ch^2 \\ \log \|E^h\| &\leq \log C + p \log h \end{split}$$

We can fit the error to ?????

$$\begin{array}{c|cccc} h & E^h & log\,h & log\,\|E_h\| \\ \hline h_0 & \|E^{h_0}\| & log(h_0) & log\,\|E^{h_0} \\ h_1 & \|E^{h_1}\| & log(h_1) & log\,\|E^{h_1} \\ & ... \\ h_n & \|E^{h_n}\| & log(h_n) & log\,\|E^{h_n} \\ \hline (x_i,y_i), i = 0,1,2,...,n \Rightarrow & \text{fit to a function} \\ & y(x) = \alpha + px \\ & y(x_0) = \alpha + px_0 \\ & y(x_1) = \alpha + px_1 \\ & ... y(x_n) = \alpha + px_n \end{array}$$

Which can be represented as

$$\begin{bmatrix} 1 & x_0 \\ 1 & x_1 \\ \dots & 1 & x_n \end{bmatrix} \begin{bmatrix} a \\ p \end{bmatrix} = \begin{bmatrix} y(x_0) \\ y(x_1) \\ \dots \\ y(x_m) \end{bmatrix} \Rightarrow X \begin{bmatrix} a \\ p \end{bmatrix} = Y$$

We can project this to the column space of X using

$$X^{-1}X\begin{bmatrix} \alpha \\ p \end{bmatrix} = X^{-1}Y$$

Which gives us the result

$$\begin{bmatrix} \sum_{k=0}^n 1 & \sum_{k=0}^n x_n \\ \sum_{k=0}^n h & \sum_{k=0}^n x_h^2 \end{bmatrix} \begin{bmatrix} a \\ p \end{bmatrix} = \begin{bmatrix} \sum_{k=0}^n y_k \\ \sum_{k=0}^n y_k x_k \end{bmatrix}$$

We can write the following code to achieve this

matrix inversion

$$\begin{bmatrix} \alpha_{11} & \alpha_{12} \\ \alpha_{21} & \alpha_{22} \end{bmatrix} \begin{bmatrix} \alpha \\ p \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \Rightarrow \begin{bmatrix} \alpha \\ p \end{bmatrix} = \frac{1}{\alpha_{11}\alpha_{21} - \alpha_{21}\alpha_{12}} \begin{bmatrix} \alpha_{22} & -\alpha_{12} \\ -\alpha_{21} & \alpha_{11} \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

$$\begin{split} \det A &= a_{11} a_{22} - a_{12} a 21 \\ a &= \frac{1}{\det A} (b_1 a_{22} - b_2 a_{12}) \\ p &= \frac{1}{\det A} (-b_1 a_{21} + b_2 a_{11}) \end{split}$$

what if we don't have u?

$$\begin{array}{c|cccc} h & E^h & log\,h & log\,\|E_h\| \\ \hline h_0 & \|E^{h_0}\| & ... & ... \\ h_1 & \|E^{h_1}\| & ... & ... \\ ... & & & \\ h_m = \frac{h^0}{2^m} & \|E^{h_n}\| & log(h_n) & log\,\|E^{h_n} \end{array}$$

We can use the $h_{\mathfrak{m}}$ to approximate $\mathfrak{u}.$ We can then do the following to get the other needed approximations

$$||u - u_k + u_k - \hat{u}|| \le ||u - u_h|| + ||u_h - \hat{u}||$$

February 14, 2024

Errors:

$$\begin{split} \|E^h\|_{\infty} &= \max_{1 \leq i \leq m} |U_i - \hat{U}_i| \approx \|e(x)\| = \max_{\alpha \leq x \leq b} |u(x) - \hat{u}(x)| \\ \|E^h\|_1 &= \sum_{j=1}^m |U_j - \hat{U}_j| \approx \|e(x)\|_1 = \int |u(x) - \hat{u}(x)| dx \\ \|E^h\|_2 &= (\sum_{j=1}^m |U_j - \hat{U}_j|^2)^{1/2} \approx \|e(x)\|_2 = (\int_b^\alpha |u(x) - \hat{u}(x)|^2 dx)^{1/2} \end{split}$$

PDEs

$$\alpha_1(x,y)u_{xx} + \alpha_2(x,y)u_{xy} + \alpha_3(x,y)u_{yy} + ... + \alpha_6(x,y) = f$$

Def: $\rho=\alpha_2^2(x,y)-4\alpha_1(x,y)*\alpha_3(x,y)$

$$\label{eq:continuous_problem} \begin{split} \text{if}: \begin{cases} \rho < 0 & \to \text{ elliptic equation} \\ \rho = 0 & \to \text{ parabolic equation} \\ \rho > 0 & \to \text{ hyperbolic equation} \\ \end{split}$$

Ex: Elliptic equation

$$\begin{split} \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} &= f & \text{poisson equation} \\ \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} &= 0 & \text{laplace equation} \\ &\Rightarrow \alpha_1 = 1, \alpha_2 = 0, \alpha_3 = 1 \\ &\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial x^2} & \text{heat equation} \\ &\Rightarrow \alpha_1 = 1, \alpha_2 = 0 \alpha_3 = 0 \end{split}$$

Notation:

$$\Delta u = \frac{\partial^2 u}{\partial x^2} + \frac{\partial u^2}{\partial y^2} = \Delta \Delta u$$
$$= (\frac{\partial}{\partial x}, \frac{\partial}{\partial y})(\frac{\partial u}{\partial x}, \frac{\partial u}{\partial y})$$
$$= \nabla^2 u$$

Mesh: In 1-d we have

$$x_j = a + j * h = a + jx\Delta x$$

in 2-d we have

. Δy
Δx Δx Δx Δx

probabilities

$$\Delta \mathfrak{u} = f(x,y) \qquad \qquad \text{on } \Omega \in \mathbb{R}^m$$

$$\mathfrak{u}\big|_{\partial\Omega} = g(x,y)$$

February 16, 2024

We still talking about heat equations (Potential equation)

$$\Delta u = f$$
 on Ω
$$u\big|_{\partial\Omega} = g \tag{\eth is the boundary)}$$

Solutions are called <u>harmonic functions</u>. Building a linear system alike to the 1D problem, where are finding the unknowns between two bounds. When we *step it up* to 2D we are still finding the unknowns at specific points, we just add an axis.

Where ind is a conversion from 2d to 1d indices.

We can use the following variables to convert from 1D to 2D indices

$$ind \rightarrow (i,j) \Rightarrow j = \frac{ind}{m} \% 1, i = ind - (j-1) * m$$

spout stencil

$$\Delta u = \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = \rho$$

Which means the equation at (i,j) is

$$\frac{1}{h^2}(u(x_{i+1},y_j)-2u(x_i,y_j)+u(x_{i-1},y_j)+u(x_i,y_{j+1})-2u(x_i,y_j)+u(x_i,y_{j+1}))=f(x_i,y_j)$$

When we build a system of equations it will look like

We then can worry about stability and truncation error to determine convergence.

LTE:
$$\tau_{i,j}(h) = ? = O(h^2)$$

stability: $\lambda_{i,1} = -2\pi + O(h^2)$

$$\therefore contains \ number \ \to K_1(A) = (\frac{8}{h^2} \frac{1}{2\varepsilon}) = O\left(\frac{1}{h^2}\right)$$

Each row represents the approximate equation at a point. We can then take the top 4x4 subset of this matrix to solve our system

$$\Rightarrow \begin{bmatrix} T & I & 0 & 0 \\ I & T & I & 0 \\ 0 & I & T & I \\ 0 & 0 & I & T \end{bmatrix}$$