

Higher-dimensional problems

Numerics in two dimensions

In order to plot a function of the form $u(x, y)$ we need to specify grid points

$$x_l \quad l = 1, \dots, n_x$$

$$y_k \quad k = 1, \dots, n_y$$

where we evaluate u .

Notational clash

a two dimensional array is a matrix

$(a_{kl}) = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \dots \\ a_{21} & a_{22} & \dots & \\ \vdots & & & \\ a_{n1} & \dots & \dots & \end{pmatrix}$

row ↑ column ↑

first entry increases down
second entry increases →

whereas for a graph

↑ second entry increases up
→ first entry increases

To adjust for this & use the convention

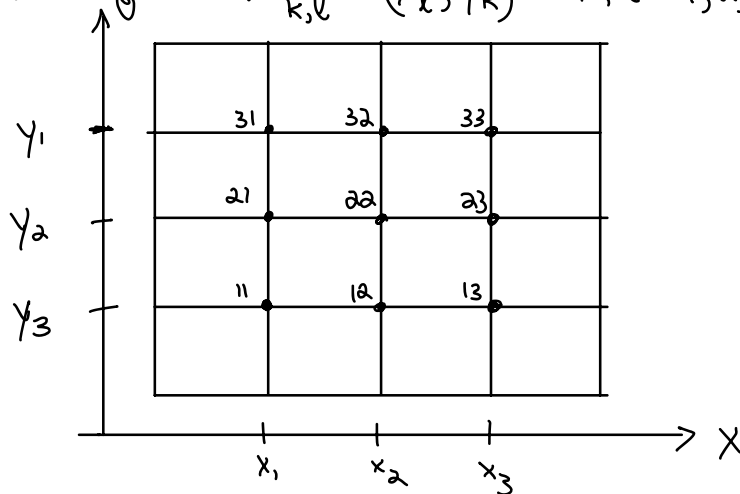
$$(u_{kl}) = u(x_l, y_k)$$

code to graph

$$u(x, y) = \sin \pi x \cos 2\pi y$$

```
x = linspace(-1, 1, Nx);  
y = linspace(2, 3, Ny);  
[xx, yy] = meshgrid(x, y);  
uu = sin(pi * xx) .* cos(2 * pi * yy);  
mesh(xx, yy, uu)
```

Labels for $u_{k,l} = u(x_l, y_k)$ $k, l = 1, 2, 3$ in xx & yy



When working with discretized values it is more convenient to convert matrix u into a vector using the `reshape` command

$$u = \text{reshape}(uu, N^2, 1)$$

Eg. $uu = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ $\text{reshape}(uu, 4, 1) = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$

more generally

$$u = (u_{11}, \dots, u_{N-1,1}, u_{12}, \dots, u_{N-1,2}, \dots, u_{1,N-1}, \dots, u_{N-1,N-1})$$

$$f = (f(x_1, y_1), \dots, f(x_1, y_{N-1}), f(x_2, y_1), \dots, f(x_2, y_{N-1}), \dots, f(x_{N-1}, y_{N-1}))$$

Kronecker Product

$$R = (r_{ij}) \in \mathbb{R}^{a \times b}, S \in \mathbb{R}^{k \times j}$$

$$\text{kron}(R, S) = \begin{bmatrix} r_{11}S & r_{12}S & \dots & r_{1b}S \\ r_{21}S & r_{22}S & \dots & r_{2b}S \\ \vdots & \vdots & \ddots & \vdots \\ r_{a1}S & r_{a2}S & \dots & r_{ab}S \end{bmatrix}$$

Eg. $\text{kron}(A, I)$ vs $\text{kron}(I, B)$

Differentiation in higher dimensions

assume that for $u(x)$ in 1D, $u(x_k) \approx u_k$

there is a 1D differentiation matrix $D \in \mathbb{R}^{N \times N}$

for u on a grid:

$$\begin{pmatrix} u_1' \\ \vdots \\ u_N' \end{pmatrix} \approx D \begin{pmatrix} u_1 \\ \vdots \\ u_N \end{pmatrix}$$

How can we differentiate in the 2D case?

$u_x \approx \text{kron}(D, I) u$ where $u = \text{reshape}(u)$
does the same operation to all of the fixed y points

$$u_y \approx \text{kron}(I, D) u$$

fixed x points are treated the same.

Eg.

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

$$\Delta u \approx \text{kron}(D^2, I) + \text{kron}(I, D^2)$$

Completeness of two-dimensional families

Theorem

Let $\Omega = (0,a) \times (0,b)$ be a rectangular domain. Assume

$\{\chi_m\}$ is a complete orthogonal set on $L^2(0,a)$

$\{\psi_n\}$ is a complete orthogonal set on $L^2(0,b)$.

Then $\varphi_{m,n}(x,y) = \chi_m(x) \psi_n(y)$

is a complete orthogonal set in $L^2((0,a) \times (0,b))$

Proof: suitable application of Fubini's Theorem.

What this means is that you could make complete orthogonal sets out of $\{\cos(k\pi x) \cos(j\pi x)\}$, $\{e^{2\pi i(jx+ky)}\}$ or any mix!

Multidimensional Fourier transforms

Let $\{\varphi_{mn}(x,y)\} = \{\chi_m(x) \psi_n(y)\}$

Let $a_x \leq x_l \leq b_x$ are the grid points associated with interpolation via $\{\chi_m\}$ using transform $d\chi_t$

Let $a_y \leq y_k \leq b_y$ are the grid points associated with interpolation by $\{\psi_n\}$ using transform $d\psi_t$

Then given $u_{kl} = u(x_l, y_k)$, we can find the interpolation coefficients a_{kl} for

$$\varphi(x,y) = \sum_{k=1}^{N_y} \sum_{l=1}^{N_x} a_{kl} \chi_m(x) \psi_n(y)$$

such that this interpolates u :

$$\varphi(x_l, y_k) = u(x_l, y_k)$$

Use $d\chi_t$ on one dim'n of u_{kl}
 $d\psi_t$ on the other dim'n of the result

such as

fft2
dct2
dst2

Example : Stationary Ohta Kawasaki Equation

$$-\Delta(\Delta u + \lambda(u - u^3)) - \lambda \circ u = 0 \quad \text{where } x \in \Omega$$
$$\Omega = (0,1)^2 \text{ or } (0,1)^3$$

with boundary conditions

$$\frac{\partial u}{\partial n}(x) = 0 \quad \text{outward normal on } \partial\Omega$$

and $\frac{\partial \Delta u}{\partial n} = 0$ (4th order equation \Rightarrow extra condn)

$$\varphi_{mn}(x_1, x_2) = \cos(m\pi x_1) \cos(n\pi x_2)$$

$$\Delta \varphi_{mn} = -(m^2 + n^2)\pi^2 \varphi_{mn}(x_1, x_2)$$

The strategy is the same as in 1D:

write

$$u(x_1, x_2) = \sum_{n=0}^{N_y-1} \sum_{m=0}^{N_x-1} a_{mn} \varphi_{mn}(x)$$

Use the 2-dimensional transform:

Let a_{mn} be a starting guess

Let $u = \sqrt{N_x} \sqrt{N_y} \text{idct}(\text{idct}(a_{mn})')'$ (with padding)

Find $u - u \cdot 13$ and unpadded to get w

Transform back and divide by $\sqrt{N_x} \sqrt{N_y}$

Multiply the terms of a_{mn} by
 $-\pi^2(m^2+n^2)$ for each Δ .

Thus we get

$$c2 = \pi(m^2+n^2) \left(-\pi^2(m^2+n^2) + \lambda w_{mn} \right) - \lambda \sigma a_{mn}$$

reshape $c2$ to be a long column.

This is where we apply Newton's method.

What about the Jacobian?

We could compute a derivative, but

it is also possible to sub in an

approximate derivative $e_k = (0 \dots 0 \overset{k\text{th}}{1} 0 \dots 0)^t$

h is small (10^{-6} is about right for double precision)

$$k\text{-column } A(:, k) = (F(a+h \cdot e_k) - F(a-h \cdot e_k)) / 2h$$