

## Relaxation Notes

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General strategy: elliptic eq 'static' form (my term) such as  $\nabla^2 f = 0$  also has a corresponding formulation which minimizes an energy,  $\int (\nabla f)^2 dxy$ . This can be shown by starting with the energy minimization form and adding a variational calculus 'variation', integrate by parts, get the 'static' form.

Easier, see below, is to discretize, then minimize wrt  $f_k$ , you get the static form as a result.

$$\begin{aligned} \min_f \int (\nabla f)^2 dxy &= \min_f \sum (\nabla f)^2 \\ \nabla f = \left(\frac{df}{dx}, \frac{df}{dy}\right)' &\approx \frac{1}{h}(f_{10} - f_{00}), \frac{1}{h}(f_{01} - f_{00}) \text{ using forward first diff} \\ (\nabla f)^2 &= \frac{1}{h^2}((f_{10} - f_{00})^2 + (f_{01} - f_{00})^2) \\ E = \sum (\nabla f)^2 &= \frac{1}{h^2} \sum \dots + (f_{k,j} - f_{k-1,j})^2 + (f_{k,j} - f_{k,j-1})^2 + (f_{k+1,j} - f_{k,j})^2 + (f_{k,j+1} - f_{k,j})^2 \\ \frac{dE}{df_{k,j}} &= \frac{1}{h^2} [(f_{k,j} - f_{k-1,j}) + (f_{k,j} - f_{k,j-1}) - (f_{k+1,j} - f_{k,j}) - (f_{k,j+1} - f_{k,j})] \\ &= \frac{1}{h^2} (4f_{k,j} - f_{k+1,j} - f_{k-1,j} - f_{k,j+1} - f_{k,j-1}) \end{aligned}$$

now iterate, reduce the error by changing f:  $f_{k,j} \leftarrow f_{k,j} - \alpha \frac{dE}{df_{k,j}}$

$\alpha$  needs to be 1/4 or less for stability, see NR

$$\begin{aligned} f_{k,j} &\leftarrow f_{k,j} - 1/4(4f_{k,j} - f_{k+1,j} - f_{k-1,j} - f_{k,j+1} - f_{k,j-1}) \\ f_{k,j} &\leftarrow 1/4(f_{k+1,j} + f_{k-1,j} + f_{k,j+1} + f_{k,j-1}) \end{aligned}$$

Now compare this to the solution starting from the static form:

$$\begin{aligned} \text{forward first difference} & \frac{1}{h}(f_+ - f_0) \\ \text{2nd difference = difference of first diff} & \frac{1}{h}\left(\frac{1}{h}(f_+ - f_0) - \frac{1}{h}(f_0 - f_-)\right) \\ &= \frac{1}{h^2}(f_+ - 2f_0 + f_-) \\ \text{so } \nabla^2 f = 0 &= \frac{d^2 f}{dx^2} + \frac{d^2 f}{dy^2} = \frac{1}{h^2}[f_{+0} - 2f_{00} + f_{-0}] + \frac{1}{h^2}[f_{0+} - 2f_{00} + f_{0-}] \\ &= \frac{1}{h^2}(f_{+0} + f_{-0} + f_{0+} + f_{0-} - 4f_{00}) = 0 \\ \text{since this equals zero, can mult by } h^2 &= f_{+0} + f_{-0} + f_{0+} + f_{0-} - 4f_{00} = 0 \end{aligned}$$

Now write this as a linear system and show its solution with Jacobi iteration. It can be written as  $Af = 0$  with f unrolled into a vector and A a  $\text{len}(f)^2$  matrix. Rows of this matrix look like

$$\dots 1, \dots * \dots, 1, -4, 1, \dots * \dots, 1, \dots$$

where \* is the offset from one row to the previous or next, and -4 is on the diagonal.

Jacobi iteration sets each  $f_k$  to the solution of its row of the matrix equation, independent of all other rows:

$$\begin{aligned} \sum A_{rc} f_c &= b_r \\ \rightarrow A_{rk} f_k &= b_k - \sum_{j \neq k} A_{rj} f_j \end{aligned}$$

$$f_k \leftarrow \frac{b_k}{A_{kk}} - \sum_{j \neq k} A_{kj} / A_{kk} f_j$$

Apply this here:

$$\text{eqn of row } k: \quad -4f_k + \text{surrounding points} = 0$$

$$f_k \leftarrow 1/4 \text{ surrounding points}$$

$$\text{converting back to 2d, gives same result as energy min} \quad f_{k,j} \leftarrow 1/4(f_{k+1,j} + f_{k-1,j} + f_{k,j+1} + f_{k,j-1})$$

The Jacobi iteration converges if the matrix is ‘diagonally dominant’, meaning that the diagonal elements are much larger than the off diagonal elements. Strict row-diagonal dominance requires that the abs of the diagonal be greater than the sum-abs of the off-diagonal elements.

For *interpolation* some  $f_{r,c}$  are specified rather than free. In setting up the linear system, subtract these from both sides of the eq, so the known quantities move to the rhs. CHECK:

$$\begin{aligned} \frac{1}{h^2} (f_{+0} + f_{-0} + f_{0+} + f_{0-} - 4f_{00}) &= 0 && \text{say } f_{+0} \text{ is known/fixed, then} \\ \frac{1}{h^2} (f_{-0} + f_{0+} + f_{0-} - 4f_{00}) &= -\frac{1}{h^2} f_{+0} \end{aligned}$$

TODO: show interp derivative

### can square of gradient be expressed as a matrix?

$$|\nabla u|^2 = \left(\frac{du}{dx}\right)^2 + \left(\frac{du}{dy}\right)^2$$

say  $K_x$  is the matrix that produces  $du/dx$ , then

$$u' K_x' K_x u = (du/dx)^2$$

$$\text{so } u'(K_x' K_x + K_y' K_y)u = (u' K_x' K_x + u' K_y' K_y)u$$

$$= u' K_x' K_x u + u' K_y' K_y u = \left(\frac{du}{dx}\right)^2 + \left(\frac{du}{dy}\right)^2 \text{ try this in maple}$$

### minimizing gradient squared same as laplace

Dan Ruderman derives this, requires several integration by parts. A simpler approach is (in 1d):

$$\text{roughness} \quad R = \int |\nabla u|^2 du \approx \sum (u_{k+1} - u_k)^2$$

$$\text{for a particular } k: \quad \frac{dR}{du_k} = \frac{d}{du_k} [(u_k - u_{k-1})^2 + (u_{k+1} - u_k)^2]$$

$$= 2(u_k - u_{k-1}) - 2(u_{k+1} - u_k) = 0$$

$$u_{k+1} - 2u_k + u_{k-1} = 0 \rightarrow \nabla^2 u = 0$$