

Lecture 19: Discretization in 2D

Logistics: - HW7 problem 1 posted more will be added

Last time: - Numerical solution for unconfined flow

general non-linear diffusion eqn

$$s(u) \frac{\partial u}{\partial t} - \nabla \cdot [f(u) \nabla u] = f_s$$

$$s(u) = \phi_0 u^m \quad f(u) = \frac{k_0}{n+1} u^{n+1}$$

- Backward Euler

$$\underline{\Gamma} = \{ \underline{s}(u^{n+1}) \}_c (u^{n+1} - u^n) - \Delta t \underline{D} [\{ \underline{M} f(u) \}_f \underline{G} u] - \Delta t f_s$$

$$\underline{J}(u, u^n) = \underline{dS}(\underline{u}) \underline{U}(u, u^n) + \underline{S}(u) + \Delta t \underline{D} [\underline{G} \underline{U}(\underline{u}) \underline{M} \underline{dF}(\underline{u}) + \underline{F}(\underline{u}) \underline{G}]$$

Application to highland aquifer

- to match published crustal structures

$$\underline{m} \in [2, 3] \quad \underline{n} \in [5, 10]$$

⇒ highly non-linear



- Looks like drainage rates are extremely

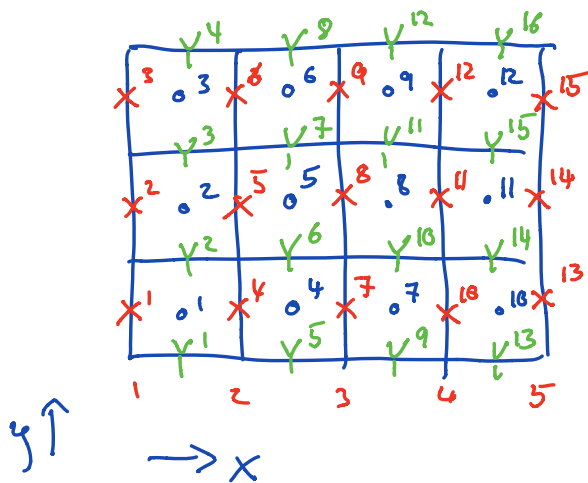
slow $\sim Ga$ time scales!

Today: Introduction to 2D numerics

- look at some Matlab functions
- Discrete operators
 - tensor products to go from 1D → 2D

Discrete operators ($\underline{D}, \underline{G}$) in 2D

Staggered grid in 2D



$$N_x = 4 \quad N_y = 3 \quad N = N_x N_y = 12$$

$$\text{faces in x-dir: } N_{fx} = (N_x + 1) N_y = 15$$

$$\text{faces in y-dir: } N_{fy} = N_x (N_y + 1) = 16$$

$$\text{total faces: } N_f = N_{fx} + N_{fy} = 31$$

Discrete gradient in 2D

$$\nabla h = \begin{pmatrix} \frac{\partial h}{\partial x} \\ \frac{\partial h}{\partial y} \end{pmatrix}$$

$$\frac{\partial h}{\partial x} \sim \underline{dh}_x \quad \text{on x-faces}$$

$$\frac{\partial h}{\partial y} \sim \underline{dh}_y \quad \text{on y-faces}$$

$$\nabla h \sim \underline{dh} = \begin{bmatrix} \underline{dh}_x \\ \underline{dh}_y \end{bmatrix} \quad (\text{choice})$$

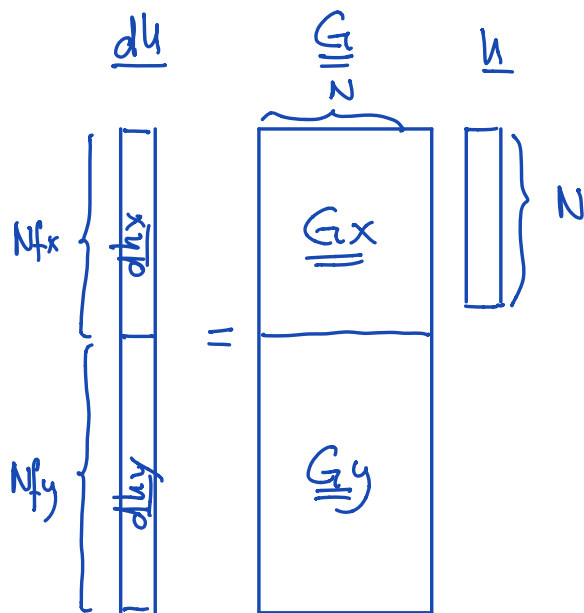
⇒ 2D gradient matrix can be decomposed

$$\underline{dh} = \underline{G} \underline{h}$$

$$\begin{bmatrix} \underline{dh}_x \\ \underline{dh}_y \end{bmatrix} = \begin{bmatrix} \underline{G}_x \\ \underline{G}_y \end{bmatrix} \underline{h}$$

$$\underline{dh}_x = \underline{G}_x \underline{h}$$

$$\underline{dh}_y = \underline{G}_y \underline{h}$$



Matrix dimensions:

\underline{G} is N_f by N

\underline{G}_x is N_{fx} by N

\underline{G}_y is N_{fy} by N

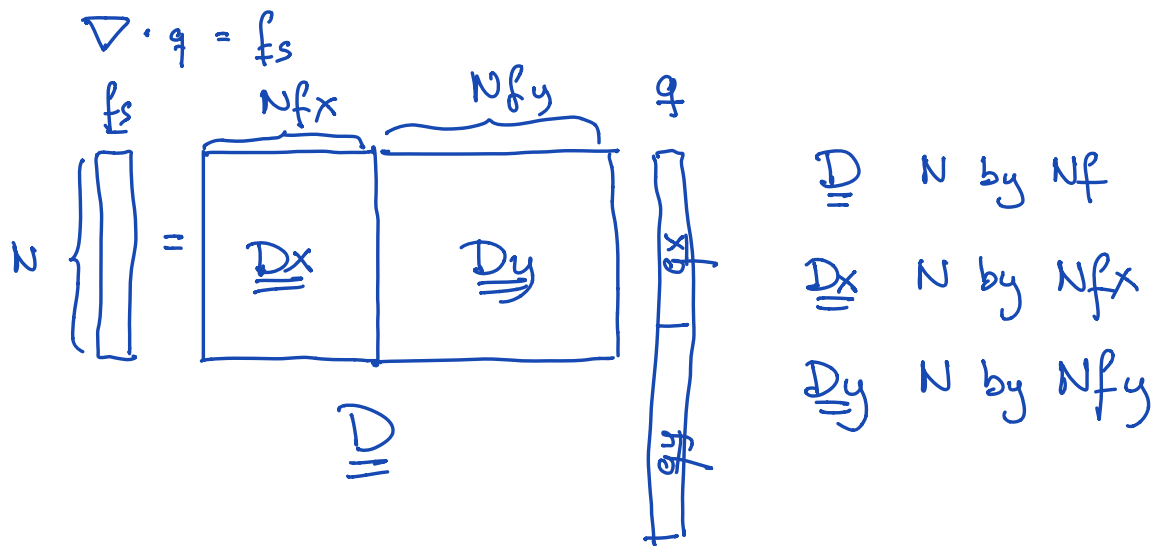
Gradient takes values at cell centers and returns derivatives on cell faces

2D discrete divergence

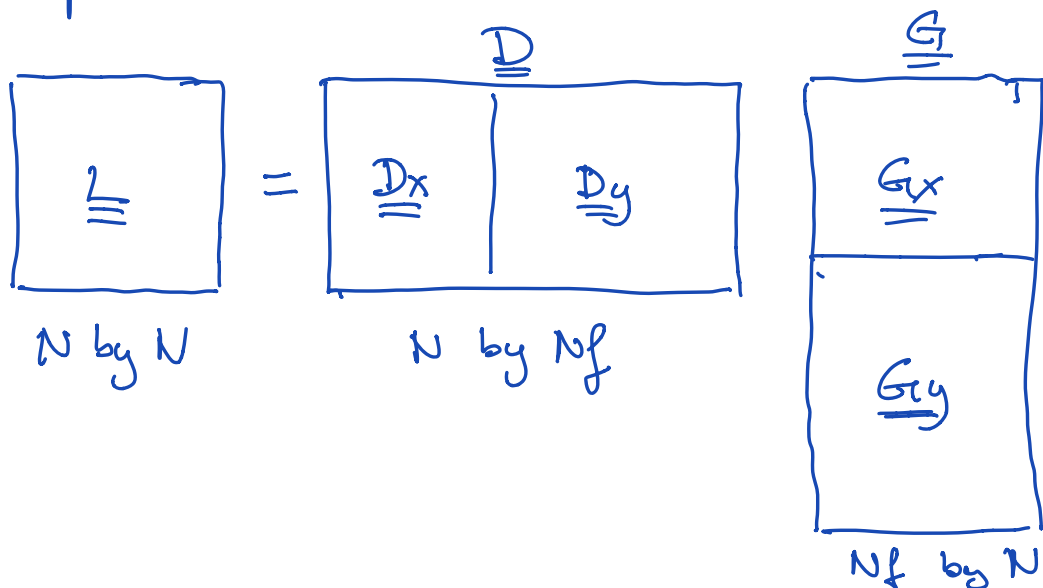
$$\nabla \cdot \mathbf{q} = \frac{\partial q_x}{\partial x} + \frac{\partial q_y}{\partial y} \approx \underline{\underline{D}} \mathbf{q} = \underline{\underline{D_x}} q_x + \underline{\underline{D_y}} q_y$$

$\mathbf{q} = \begin{pmatrix} q_x \\ q_y \end{pmatrix} \leftarrow \text{continuum}$

\uparrow
discrete flux vector $\mathbf{q} = \begin{bmatrix} q_x \\ q_y \end{bmatrix}$

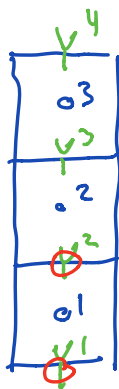


Laplacian: $\nabla^2 = \nabla \cdot \nabla$



Building the 2D discrete divergence

Start with D_y in 1D $\nabla \cdot q = \underline{f_s}$



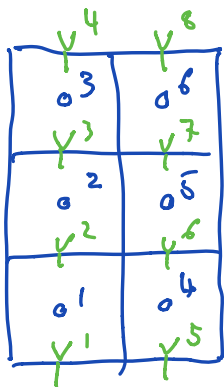
$$\underline{f_s} = \frac{1}{\Delta y} \begin{bmatrix} -1 & 1 \\ & -1 & 1 \\ & & -1 & 1 \end{bmatrix} q = q_y$$

N_y by $(N_y + 1)$

$$\frac{\partial q}{\partial y} =$$

$$\frac{q_2 - q_1}{\Delta y}$$

Suppose we add second column



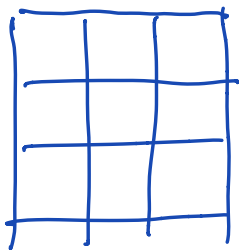
$$\underline{f_s} = \frac{1}{\Delta y} \begin{bmatrix} -1 & 1 & & & & \\ & -1 & 1 & & & \\ & & -1 & 1 & & \\ & & & -1 & 1 & \\ & & & & -1 & 1 \end{bmatrix} q$$

$$\frac{q_6 - q_5}{\Delta y}$$

$$\underline{D_y}^2 = \begin{bmatrix} \underline{D_y}^1 & 0 \\ 0 & \underline{D_y}^1 \end{bmatrix}$$

2 by 2 block matrix with $\underline{D_y}^1$ on diagonal

Suppose we add another row:



$$\underline{D_y}^2 = \begin{bmatrix} \underline{D_y}^1 & & \\ & \underline{D_y}^1 & \\ & & \underline{D_y}^1 \end{bmatrix}$$

3 by 3 block matrix

In general:

$\underline{\underline{D}}_y^2$ is a N_x by N_x block matrix with blocks of size N_y by (N_y+1) . Diagonal entries are $\underline{\underline{D}}_y^1$ and all others are zero.

Tensor product construction of $\underline{\underline{D}}_y^2$

The discrete 2D operators can easily and efficiently be assembled using Kronecker/tensor products.

Definition:

If $\underline{\underline{A}}$ is a $m \times n$ matrix and $\underline{\underline{B}}$ is a $p \times q$ matrix, then the Kronecker product $\underline{\underline{A}} \otimes \underline{\underline{B}}$ is the $mp \times nq$ block matrix:

$$\underline{\underline{A}} \otimes \underline{\underline{B}} = \begin{bmatrix} a_{11} \underline{\underline{B}} & a_{12} \underline{\underline{B}} & \dots & a_{1n} \underline{\underline{B}} \\ a_{21} \underline{\underline{B}} & & & \vdots \\ \vdots & & & \\ a_{m1} \underline{\underline{B}} & - & - & a_{mn} \underline{\underline{B}} \end{bmatrix}$$

Hence, we can construct $\underline{\underline{Dy}}^2$ as

$$\underline{\underline{Dy}}^2 = \underline{\underline{Ix}} \otimes \underline{\underline{Dy}}' = \begin{bmatrix} \underline{\underline{Dy}}' & & & \\ & \underline{\underline{Dy}}' & & \\ & & \underline{\underline{Dy}}' & \\ & & & \dots & \\ & & & & \underline{\underline{Dy}}' \end{bmatrix}$$

$\underline{\underline{Ix}}$ is N_x by N_x identity

In matlab: $\underline{\underline{Dy}} = \text{kron}(\underline{\underline{Ix}}, \underline{\underline{Dy}})$

Next time \rightarrow complete this

