

① BBSI notes

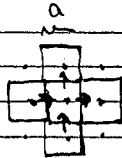
Partial Differential Equations (PDE's)

- Real world applications:
- i) recovering oil from the ground
 - ii) modeling blood flow in the heart
 - iii) modeling flow (current) of ions through protein channels

The Continuity Equation

Fig 1

Illustrate in 2D



arrows indicate flux through "faces"

For flow into (central) point (i,j) :

x -flux = $\frac{\# \text{ particles}}{\text{cm} \cdot \text{sec}}$

$$a^2 \dot{\rho}_{ij} = [j_x(i+\frac{1}{2}, j) - j_x(i-\frac{1}{2}, j)] a$$

$$- [j_y(i, j+\frac{1}{2}) - j_y(i, j-\frac{1}{2})] a$$

thus:

$$\dot{\rho}_{ij} = - \left(\frac{\partial j_x}{\partial x} \right)_{ij} - \left(\frac{\partial j_y}{\partial y} \right)_{ij} \iff \frac{\partial \rho(x,y,t)}{\partial t} = - \frac{\partial j_x(x,y,t)}{\partial x} - \frac{\partial j_y(x,y,t)}{\partial y}$$

2D

In 3D:

$$\frac{\partial \rho(x,y,z,t)}{\partial t} = - \frac{\partial j_x(x,y,z,t)}{\partial x} - \frac{\partial j_y(x,y,z,t)}{\partial y} - \frac{\partial j_z(x,y,z,t)}{\partial z} = - \vec{\nabla} \cdot \vec{j}(x,y,z,t)$$

Application to Diffusion: Fick's Law: $\vec{j}(\vec{r}, t) = -D \vec{\nabla} \rho(\vec{r}, t)$

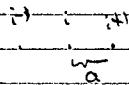
↑
diffusion constant

②

how, in 3D: $\frac{\partial \rho(\vec{x}, t)}{\partial t} = D \nabla^2 \rho(\vec{x}, t)$

Solving PDEs via finite difference (real-space lattice) methods

Illustrate 1st w/ 1D Diffusion $\left[\frac{\partial \rho(x, t)}{\partial t} = D \frac{\partial^2 \rho(x, t)}{\partial x^2} \right]$



$$j\left(i+\frac{1}{2}\right) \approx \frac{-(\rho_{i+1} - \rho_i) D}{a}$$

↑
1D flux = #particles/sec

thus, for interior points:

$$\dot{\rho}_i = \frac{(\rho_{i+1} - \rho_i) D}{a^2} - \frac{(\rho_i - \rho_{i-1}) D}{a^2} = \frac{D}{a^2} [\rho_{i+1} - 2\rho_i + \rho_{i-1}]$$

(3)

What about points ^{at} next to the boundary?

For concreteness, consider absorbing b.c.'s ← "boundary conditions"

(0) 1 2 3 ... 6=N (N+1)
 $\frac{\Delta x}{a}$

$$p_0(t) = 0 = p_{N+1}(t)$$

$$\text{Thus, } \dot{p}_1 = \frac{D}{a^2} [p_2 - 2p_1] \quad ; \quad \dot{p}_N = \frac{D}{a^2} [-2p_N + p_{N+1}]$$

Put all this together:

$$\begin{pmatrix} \dot{p}_1 \\ \dot{p}_2 \\ \dot{p}_3 \\ \dot{p}_4 \\ \dot{p}_5 \\ \dot{p}_6 \end{pmatrix} = \frac{D}{a^2} \begin{bmatrix} -2 & 1 & 0 & 0 & 0 & 0 \\ 1 & -2 & 1 & 0 & 0 & 0 \\ 0 & 1 & -2 & 1 & 0 & 0 \\ 0 & 0 & 1 & -2 & 1 & 0 \\ 0 & 0 & 0 & 1 & -2 & 1 \\ 0 & 0 & 0 & 0 & 1 & -2 \end{bmatrix} \begin{pmatrix} p_1 \\ p_2 \\ p_3 \\ p_4 \\ p_5 \\ p_6 \end{pmatrix}$$

Reduction to coupled ODE's

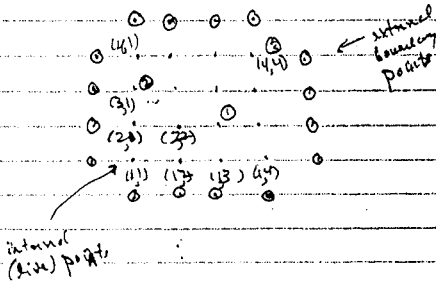
(sparsely)

Note: special techniques for Δ solutions:

$$\vec{p}(t) = \exp\left\{ \frac{D}{a^2} \Delta t \right\} \vec{p}(0)$$

Comment on multi-d finite-diff. eqs:

In 2D:



For concreteness,

take the coordinates on the external boundary points as fixed and, for simplicity, $p=0$ ← "Boundary surfaces are sinks" external boundary points

(4)

Then, for completely internal points like ①

$$\frac{D^2}{D} \rho_{ij} = [\rho_{i+1,j} - 2\rho_{ij} + \rho_{i-1,j}] + [\rho_{i,j+1} - 2\rho_{ij} + \rho_{i,j-1}]$$

For a point on one edge, e.g. ②

$$\frac{D^2}{D} \rho_{31} = [\rho_{3,2} - 2\rho_{3,1}] + [\rho_{4,1} - 2\rho_{3,1} + \rho_{2,1}]$$

For a point on one of the corners, e.g. ③

$$\frac{D^2}{D} \rho_{44} = [-2\rho_{4,4} + \rho_{4,3}] + [-2\rho_{4,4} + \rho_{3,4}]$$

Drift-Diffusion: when there is

a systematic (external) force $\vec{F}(x)$,
there is a drift flux:

$$\vec{j}_{\text{drift}} = \beta D \vec{F}(x) \rho(x)$$

\uparrow
 $(\frac{1}{kT})$

Total flux is then:

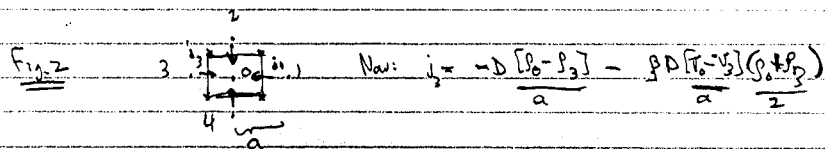
$$\vec{j} = \vec{j}_{\text{diff}} + \vec{j}_{\text{drift}} = -D \vec{\nabla} \rho + \beta D \vec{F} \rho = -D [\vec{\nabla} \rho - \beta \vec{F} \rho]$$

Then:

$$\frac{\partial \rho}{\partial t} = -\vec{\nabla} \cdot \vec{j} = D \vec{\nabla} \cdot [\vec{\nabla} \rho - \beta \vec{F} \rho]$$

Comment on the discretization procedure when $\vec{F}(x) = -\vec{\nabla} V(x)$

potential energy function



⑤ note direction of fluxes in Fig 2
same basic structure as
diff. Eq.
(conv)

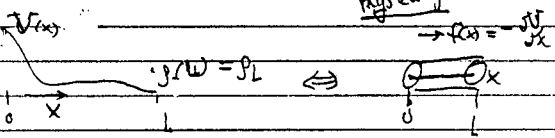
Completing the exercise $\Rightarrow \rho_0 = i_1 + i_2 + i_3 + i_4$

the special case of steady state: $\underbrace{\nabla \cdot i}_{\text{gross}} = -\frac{d\rho}{dt} = 0$

For diff-diffusion: $0 = \nabla \cdot [D \nabla \rho - \rho F \rho] + \rho(\nabla \cdot V)$

Example:

1st steady 1D process: $\rho = \rho_0$



what is?: $\rho(x, t \rightarrow \infty) \equiv \rho(x)$; what is?: $j(x, t \rightarrow \infty) \equiv j$

$0 = \frac{j}{x} (\frac{j}{D} + \rho \frac{d\rho}{dx})$; let $\rho = e^{-\beta \rho x}$
 $-\frac{j}{D} = j_0 \text{ constant!}$

$-j_0 = -\beta x e^{-\beta x} j + e^{-\beta x} j + \beta x e^{-\beta x} j$
 $f' = -\frac{j}{D} e^{-\beta x} \Rightarrow f = -\frac{j}{D} \int_0^x e^{-\beta x'} dx' + f_0$
 $f_0 = j_0 e^{-\beta x_0}$

$f(L) = j_0 e^{-\beta L} = -\frac{j}{D} \int_0^L e^{-\beta x'} dx' + j_0 e^{-\beta L}$

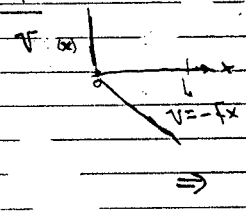
This:

$j = -\frac{(j_0 e^{-\beta L} - j_0 e^{-\beta x_0})}{\int_0^L e^{-\beta x'} dx'}$

Note: In Fig 3,

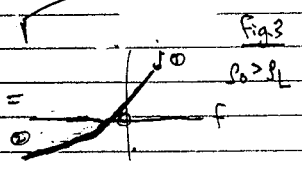
- ① \Leftrightarrow asymptotic slope of: $D \beta F \rho_0$
- ② \Leftrightarrow asymptotic slope of: $-D \beta F \rho_0$

finally, specific to:



Note: $\int_0^L e^{-\beta F x} dx = \frac{1}{\beta F} (1 - e^{-\beta F L})$

$j = \frac{(j_0 - j_0 e^{-\beta F L})}{(1 - e^{-\beta F L})} \Delta \rho F$



ⓐ

In the case of 2, 3... Dim., use "Relaxation" principles. In 2D, for Diff-Diff Eq:

$$0 = [j_1 + j_2 + j_3 + j_4] / D \Rightarrow$$

$$0 = (\rho_0 - \rho_1) + \beta (V_0 - V_1) \rho_0 + (\rho_0 - \rho_2) + \beta (V_0 - V_2) \rho_0 +$$

$$(\rho_0 - \rho_3) + \beta (V_0 - V_3) \rho_0 + (\rho_0 - \rho_4) + \beta (V_0 - V_4) \rho_0$$

$$0 = 4\rho_0 - \sum_{k=1}^4 \rho_k + \beta \rho_0 \sum_{k=1}^4 (V_0 - V_k)$$

$$\rho_0 = \frac{\sum_{k=1}^4 \rho_k [1 + \beta (V_k - V_0)]}{4 + \beta \sum_{k=1}^4 (V_0 - V_k)}$$

Relaxation Method

In practice: $\rho_{new} = (1 - \alpha) \rho_{old} + \alpha \bar{\rho}$

$0 < \alpha < 2$

← cycle around lattice of "live" points; update them one by one.

Final version of the PDE/relaxation thm: Poisson Eq.

$$\nabla^2 \phi = -4\pi \rho$$

← in vacuum (using Gaussian/CGS units)

Consider the discretized 2-D version:

$$\frac{1}{a^2} (\phi_1 - 2\phi_0 + \phi_3) + \frac{1}{a^2} (\phi_2 - 2\phi_0 + \phi_4) = -4\pi \rho_0$$

$\phi(x)$ = electrostatic potential

$$\frac{4\pi \rho_0 a^2 + \sum_{k=1}^4 \phi_k}{4} = \phi_0$$

from Mathcad manual:

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Solve  $Y'' = \begin{pmatrix} Y_1 \\ Y_2 \end{pmatrix}$  for  $x \in 0$ 
      for  $x \in 2 \cdot 0$ 
      where  $Y(1) = 1$  and  $Y(1) = 2$ 
 $O(x, y) = \begin{cases} (x < 0) \cdot Y_0 + (x > 0) \cdot -Y_0 \\ Y_1 \\ Y_2 \end{cases}$ 
 $Y_0 = 1$  ← guess value for  $Y(x)$ 
 $Y_1 = 1$  ← guess value for  $Y'(x)$ 
 $Y_2 = 1$  ← guess value for  $Y''(x)$ 
 $loadl(x1, v1) = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$  ← guess value for  $Y(x)$ 
 $loadl(x2, v2) = \begin{pmatrix} 2 \\ 0 \\ 0 \end{pmatrix}$  ← guess value for  $Y(x)$ 
score(xf, yf) := Y ← tells Mathcad to match the two halves of the solution at x=0
S := bndint(v1, v2, -1, 1, 0, 0, loadl, loadl, score)
S = ( 0.882 -0.878 )
      ← contains ( Y(x) Y'(x) )

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Figure 16-8: Using bndint to match solutions in the middle of the integration interval.

Partial differential equations

A second type of boundary value problem arises when you are solving a partial differential equation. Rather than fixing the value of a solution at two points as was done in the previous section, we now fix the solution at a whole continuum of points representing some boundary.

Two partial differential equations that arise often in the analysis of physical systems are Poisson's equation:

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

and its homogeneous form, Laplace's equation.

Mathcad has two functions for solving these equations over a square boundary. You should use the relax function if you know the value taken by the unknown function $u(x, y)$ on all four sides of a square region.

If $u(x, y)$ is zero on all four sides of the square, you can use multigrd function instead. This function will often solve the problem faster than relax. Note that if the boundary condition is the same on all four sides, you can simply transform the equation to an equivalent one in which the value is zero on all four sides.

The relax function returns a square matrix in which:

- An element's location in the matrix corresponds to its location within the square region, and
 - Its value approximates the value of the solution at that point.
- This function uses the relaxation method to converge to the solution. Poisson's equation on a square domain is represented by:

$$a_{i,j} u_{i+1,k} + b_{i,j} u_{i-1,k} + c_{i,j} u_{i,k+1} + d_{i,j} u_{i,k-1} + e_{i,j} u_{i,k} = f_{i,j,k}$$

The arguments taken by these functions are shown below:

Pro relax(a, b, c, d, e, f, u, rjac)

a, b, c, d, e = Square matrices all of the same size containing coefficients of the above equation.

f = Square matrix containing the source term at each point in the region in which the solution is sought.

u = Square matrix containing boundary values along the edges of the region and initial guesses for the solution inside the region.

rjac = Spectral radius of the Jacobi iteration. This number between 0 and 1 controls the convergence of the relaxation algorithm. Its optimal value depends on the details of your problem.

If the boundary condition is zero on all four sides of the square integration domain, use the multigrd function instead. An example is shown in Figure 16-9. The same problem solved with the relax function instead is shown in Figure 16-10.

Pro multigrd(M, ncycle)

M = (1 + 2ⁿ) row square matrix whose elements correspond to the source term at the corresponding point in the square domain.

ncycle = The number of cycles at each level of the multigrd iteration. A value of 2 will generally give a good approximation of the solution.