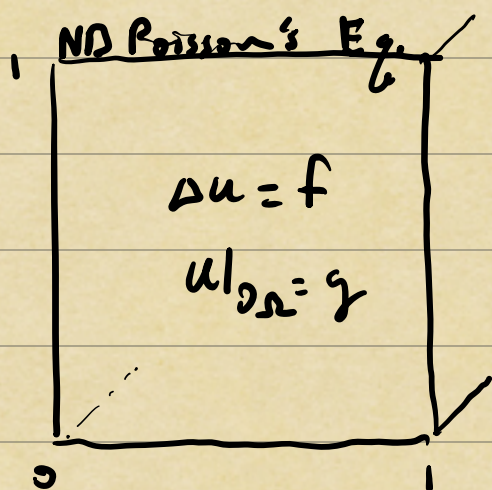


Beyond Poisson's Eq. (pt 1)



Fast FD solvers exploit:
 \Rightarrow FFT-based 1D solvers
 \Rightarrow Separable domain
(Kronecker structure)

What about more complicated PDEs?

- \Rightarrow Variable coefficients
- \Rightarrow Nonseparable Geometry
- \Rightarrow Time-dependence (parabolic, hyperbolic)
- \Rightarrow Nonlinearity

Key Idea 1: Many "more complicated PDE" solvers reduce problem to Boundary Value Problems, or at best to Linear Systems w/ similar structure.

Key Idea 2: Many BVPs can be solved efficiently using preconditioned iterative methods.

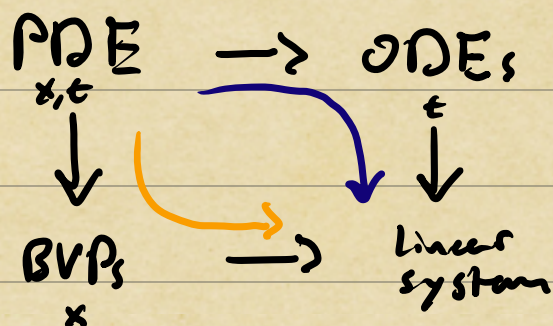
Nonlinear Initial Value Problem

E.g. | KdV Eq. $\overset{\text{linear, time-dep.}}{\partial_t u} + \overset{\text{nonlinear term}}{\partial_x^3 u} = -u \partial_x u$

General first-order systems in time:

$$\begin{array}{c} \text{state} \\ \downarrow \quad \downarrow \\ M \partial_t X + L X = F(X) \\ \uparrow \text{linear diff. ops.} \quad \uparrow \text{nonlinear diff. op.} \end{array}$$

Two useful perspectives:



1) "Propagator"

$$M \left(\frac{X^{n+1} - X^n}{h} \right) + L X^{n+1} = F(X^n)$$

2) "Method of Lines"

\Rightarrow Solve BVPs

discretized ops \downarrow discretized state \uparrow

$$\tilde{M} \partial_t X + L X = F(X)$$

$$\left(\frac{1}{h} M + L \right) X^{n+1} = \frac{1}{h} M X^n + F(X^n)$$

\Rightarrow Integrate system of coupled ODEs

* Need fast matrices/solvers for $A = h^{-1}M + L$.

Iterative Methods for BVPs

1) Stationary Iterative Methods

Goal: Solve $Ax=b$ where $A=B+C$

Initial guess $x^{(0)}$

for $n=0,1,2,\dots$

$$Bx^{(n+1)} = Cx^{(n)} + b$$

end

(1 mat-vec with C
1 linear solve w/ B)

E.g.

$\mathcal{L} = -\Delta + V(x)$ on periodic $[0,1)$.

$$B = \begin{bmatrix} 2 & -1 & & -1 \\ -1 & 2 & -1 & \\ & \ddots & \ddots & \ddots \\ -1 & & -1 & 2 \end{bmatrix}$$

Fast Solve (FFT)

$$C = \begin{bmatrix} & & & \\ & & & \\ & & & \\ & & & \end{bmatrix}$$

Fast multiply

variable
coeff
 $C_{ii} = V(x_i)$

Convergence: $e^{(n)} = x^{(n)} - x$ (error)

from iteration $\Rightarrow Be^{(n+1)} = Ce^{(n)}$
can write

So, the error evolves as

$$e^{(n+1)} = \underbrace{B^{-1}C}_D e^{(n)} = D^{n+1} e^{(0)}$$

The asymptotic growth or decay of $\|D^n e^{(0)}\|$ is governed by spectral radius of D , i.e.,

$$\rho(D) = \max_{1 \leq i \leq n} |\lambda_i| \quad \leftarrow \text{eigenvalues of } D.$$

If D has orthogonal eigenvectors, then

$$\|e^{(n)}\| = \|D^n e^{(0)}\| \leq \rho(D)^n$$

and the iteration converges geometrically if

$$\rho(D) < 1$$

However, if the eigenvectors of D are far from orthonormal, the convergence rate is only asymptotic - there may be large transient growth of $\|e^{(n)}\|$ before convergence.

SPD A

Any A

2) Krylov methods (PCG, PGMRES, ...)

Iterative Approximations constructed from

$$\text{span}\{b, Ab, \dots, A^{n-1}b\} = \text{Krylov Subspace}$$

↑
initial
guess

Want (a) Fast $x \rightarrow Ax$ and (b) Rapid convergence

E.g. PCG Converges geometrically

$$\|e^{(n)}\|_A \leq \left(\frac{\sqrt{\kappa}-1}{\sqrt{\kappa}+1}\right)^n \|e^{(0)}\|_A \leq 2e^{-2n/\kappa} \|e^{(0)}\|$$

↑ error after n iterations

↑ "A-norm"
 $\sqrt{x^T A x}$
weighted
Euclidean
norm

Rate of geometric convergence slows when

$$\kappa = \frac{\lambda_{\max}}{\lambda_{\min}} = \text{ratio of } \frac{\text{largest}}{\text{smallest}} \text{ eigs}$$

is very large $\Rightarrow A$ is ill-conditioned.

Both Stationary : Krylov methods can be

preconditioned:

$$\underbrace{P_L A P_R}_{\tilde{A}} \underbrace{P_R^{-1} x}_{\tilde{x}} = \underbrace{P_L b}_{\tilde{b}}$$

Apply iterative method to preconditioned system - idea is to improve convergence rate by clever selection of P_L and P_R . (E.g. try to reduce $\rho(A)$ or $\kappa(A)$.)

E.g. | Schrödinger Eqn. (time-independent)

$$\Delta u + v(x)u = f$$

high-order
finite-diff

$$\underbrace{(B + C)}_A x = b \quad \Rightarrow \text{discretize}$$

$P_L = B^{-1}$ can be applied efficiently

and $P_L A = I + B^{-1}C$ has a much

smaller condition number K than A does. Each iteration involves

(a) $x \rightarrow (B+C)x$ (fast mat-vec)

(b) $x \rightarrow B^{-1}x$ (fast solve)

And only need $O(\log(\epsilon))$ ^{iterations} k for an $\epsilon > 0$ accurate approximate solution.