# MAT 228B Notes

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### 1 Implicit-Time methods in Multi-D

 $u_t = b\nabla^2 u$ . Crank-Nicolson is

$$\left(I - \frac{\Delta tb}{2}L\right)u^{n+1} = \left(I + \frac{\Delta tb}{2}L\right)u^n$$

Backward Euler is

$$(I - \Delta tbL)u^{n+1} = u^n$$

BDF-2 is

$$\left(I - \frac{2}{3}\Delta tL\right)u^{n+1} = \text{stuff}$$

so the inversion is pretty similar for BDF-1,2,3,etc. We have to solve

$$(I - \beta \Delta t L)u^{n+1} = r$$

every timestep. Direct-solve (Gaussian elimination) is expensive. We can use iterative methods like SOR, Multigrid, PCG. Or we can use some specialized direct methods like Block-Cyclic Reduction or FFT method (both of these require structure and constant coefficients). Asymptotically, MG and FFT are the best. MG has more overhead, but is more general.

#### How well do the iterative methods work? 1.1

The condition number of L is  $\mathcal{O}(1)\Delta x^2$ . This is pretty bad. the bigger the condition number, the slower the convergence (PCG). Let  $A = I - \beta \Delta t L$ . We can really just look at  $\beta \Delta t L$ . Then  $\kappa(A) = \mathcal{O}(\frac{\Delta t}{\Delta r^2})$  (remember  $\kappa(A)$  is notation for the condition number of A). If  $\Delta t = \mathcal{O}(\Delta x)$ , then  $\kappa(A) = \mathcal{O}(\frac{1}{\Delta x})$ .

So, all of these methods will work one order of magnitude faster than they do for the Poisson equation. The actual convergence rate depends on the size of  $\frac{\beta \Delta t}{\Delta x^2}$ . Two extreme cases:

- $\frac{\beta \Delta t}{\Delta r^2} \to 0$
- $\frac{\beta \Delta t}{\Delta r^2} \to \infty$

For Backward Euler  $(\beta = b)$ ,

$$(I - \Delta tbL)u^{n+1} = u^n + \Delta t f^{n+1}.$$

If f contains boundary conditions, then we expect  $f = \mathcal{O}(\frac{1}{\Lambda r^2})$ .

- If  $\frac{\beta \Delta t}{\Delta x^2} \to 0$ , then  $A \to I$  and the iterative methods converge very fast.
- If  $\frac{\beta \Delta t}{\Delta t^2} \to \infty$ , then  $A \to -\beta \Delta t L$ . In that limit, the equations look like

$$-\Delta t b L u^{n+1} = \Delta t f^{n+1}$$

This is a discrete Poisson equation. So the worst case scenario is that the iterative methods work better (converge faster) than they do for the Poisson equation.

A Multigrid solver on a  $64^2$  periodic domain has convergence factor for the Poisson equation is  $\rho \approx 0.16$ . Then the number of iterations per digit accuracy is  $-\frac{1}{\log_{10} \rho} \approx 1.26$ .

## Results for Various $\beta$ s

 $(I - \Delta t \beta L)$ :

and this is with  $\Delta t = \Delta x$ . Then  $\frac{\beta \Delta t}{\Delta x^2} = \frac{\beta}{64}$  if the grid is  $64^2$ . We actually might do better than this since time-dependent problems give us an initial guess for each step. We'll take  $(u^{n+1})^0 = u^n$ , i.e. the initial guess on the (n+1)th timestep is the nth timestep.

## 2 Exploiting the Time-Dependencies of the Heat Equation

There is another way to solve the equation, which was not available for the Poisson equation because it was timeindependent.

- ADI scheme (Alternating Direction, Implicit)
- LOD scheme (Locally One-Dimensional), which is good for structured grids only.. not often used in practice.

Exploiting the Laplacian in 2D,  $\nabla^2 u = u_{xx} + u_{yy}$ , i.e.  $L = L_x + L_y$ . Intuitively, we diffuse in each dimension sequentially.

Crank-Nicolson is

$$\left(I - \frac{b\Delta t}{2}L_x - \frac{d\Delta t}{2}L_y\right)u^{n+1} = \left(I + \frac{b\Delta t}{2}L_x + \frac{d\Delta t}{2}L_y\right)u^n$$

So, sequentially, the LOD scheme is

$$\left(I - \frac{b\Delta t}{2}L_x\right)u^* = \left(I + \frac{b\Delta t}{2}L_x\right)u^n, \quad \text{followed by} \quad \left(I - \frac{b\Delta t}{2}L_y\right)u^{n+1} = \left(I + \frac{b\Delta t}{2}L_y\right)u^*$$

It turns out this is pretty reasonable. The LOD scheme looks like a fractional stepping method. The "half" step is in the x direction. The "full" step is that, followed by the step in the y direction. In fractional stepping methods, we ignore part of the ODE in each fractional step.. not like RK.