

CHAPTER 5.

Multi-Dimensional Parabolic Problems.

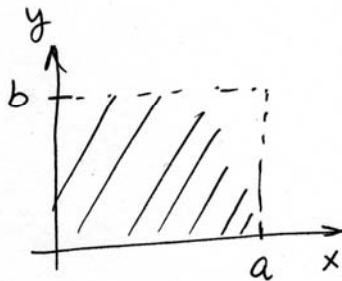
Heat Conduction (3-D)

$$\begin{cases} \rho c u_t = \nabla \cdot (k \nabla u), & (x, y, z) \in \Omega. \\ u(x, y, z, 0) = \phi(x, y, z), & (x, y, z) \in \Omega \cup \partial \Omega. \\ \alpha u + \beta \frac{\partial u}{\partial n} = \mathcal{H}, & (x, y, z) \in \partial \Omega, t > 0. \end{cases}$$

We will restrict our attention to 2-D problems on a rectangular domain. $k \equiv \text{const.}$

$$\begin{cases} u_t = \sigma (u_{xx} + u_{yy}), & (x, y) \in \Omega \\ u(x, y) = \phi(x, y), & (x, y) \in \Omega \cup \partial \Omega \\ u(x, y, t) = \mathcal{H}(x, y, t), & (x, y) \in \partial \Omega, t > 0 \end{cases}$$

$$\Omega = \{ (x, y) : 0 \leq x < a, 0 < y < b \}.$$



Introduce a uniform rectangular grid with space step sizes given by

$$\Delta x = \frac{a}{N_x}, \quad \Delta y = \frac{b}{N_y}.$$

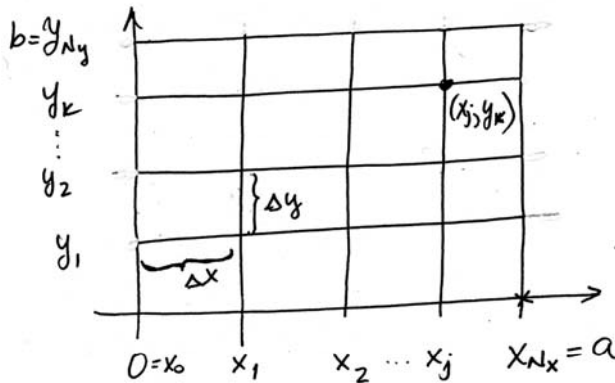
and Δt as a time step size.

Let $U_{j,k}^n \approx u(x_j, y_k, t_n)$ where

$$x_j = j * \Delta x$$

$$y_k = k * \Delta y$$

$$t_n = n * \Delta t.$$



Explicit scheme FT-CS

$$\frac{U_{j,k}^{n+1} - U_{j,k}^n}{\Delta t} = \sigma \left[\frac{U_{j-1,k}^n - 2U_{j,k}^n + U_{j+1,k}^n}{(\Delta x)^2} + \frac{U_{j,k-1}^n - 2U_{j,k}^n + U_{j,k+1}^n}{(\Delta y)^2} \right] \quad (1)$$

Equ. (1) can also be written as

$$U_{j,k}^{n+1} = U_{j,k}^n + r_x [U_{j+1,k}^n - 2U_{j,k}^n + U_{j-1,k}^n] + r_y [U_{j,k+1}^n - 2U_{j,k}^n + U_{j,k-1}^n] \quad (2)$$

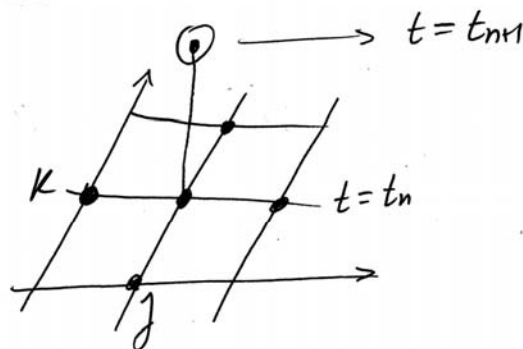
or
$$U_{j,k}^{n+1} = U_{j,k}^n + r_x S_x^2 U_{j,k}^n + r_y S_y^2 U_{j,k}^n$$

Where $r_x = \frac{\sigma \Delta t}{\Delta x^2}$, and $r_y = \frac{\sigma \Delta t}{\Delta y^2}$

To study stability, it's convenient to write equ. (2) as

$$U_{j,k}^{n+1} = r_x U_{j-1,k}^n + r_y U_{j,k-1}^n + (1-2r_x-2r_y) U_{j,k}^n + r_x U_{j+1,k}^n + r_y U_{j,k+1}^n \quad (3)$$

Computational stencil



Stability

If $r_x + r_y \leq 1/2 \Rightarrow 1 - 2(r_x + r_y) \geq 0$

And $1 - 2(r_x + r_y) + 2r_x + 2r_y = 1$

Max. norm principle applies

In particular, if $\Delta x = \Delta y \Rightarrow r_x + r_y \leq 1/2 \Leftrightarrow 2r_x \leq 1/2$

$\Rightarrow \boxed{r_x \leq 1/4}$ More restrictive than in 1-D.

Crank-Nicholson.

(Approxs are performed at $(x_j, y_k, t_{n+1/2})$
and average betw t_n and t_{n+1}
is taken)

$$\frac{U_{j,k}^{n+1} - U_{j,k}^n}{\Delta t} = \sigma \left[\frac{\delta_x^2}{\Delta x^2} U_{j,k}^{n+1/2} + \frac{\delta_y^2}{\Delta y^2} U_{j,k}^{n+1/2} \right]$$

$$= \sigma \left[\frac{\delta_x^2}{\Delta x^2} \left[\frac{U_{j,k}^{n+1} + U_{j,k}^n}{2} \right] + \frac{\delta_y^2}{\Delta y^2} \left[\frac{U_{j,k}^{n+1} + U_{j,k}^n}{2} \right] \right] \quad (4)$$

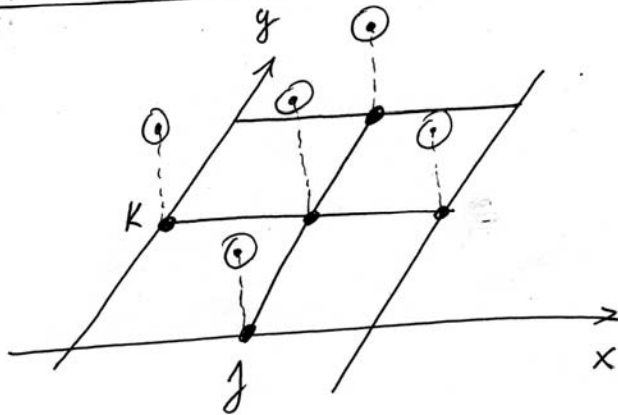
This can also be written as

$$\left(1 - \frac{r_x \delta_x^2 + r_y \delta_y^2}{2} \right) U_{j,k}^{n+1} = \left(1 + \frac{r_x \delta_x^2 + r_y \delta_y^2}{2} \right) U_{j,k}^n$$

$$\delta_x^2 U_{j,k}^n = U_{j-1,k}^n - 2U_{j,k}^n + U_{j+1,k}^n$$

$$\delta_y^2 U_{j,k}^n = U_{j,k-1}^n - 2U_{j,k}^n + U_{j,k+1}^n$$

Computational Stencil



Matrix Representation

Equ. (4) can also be written as

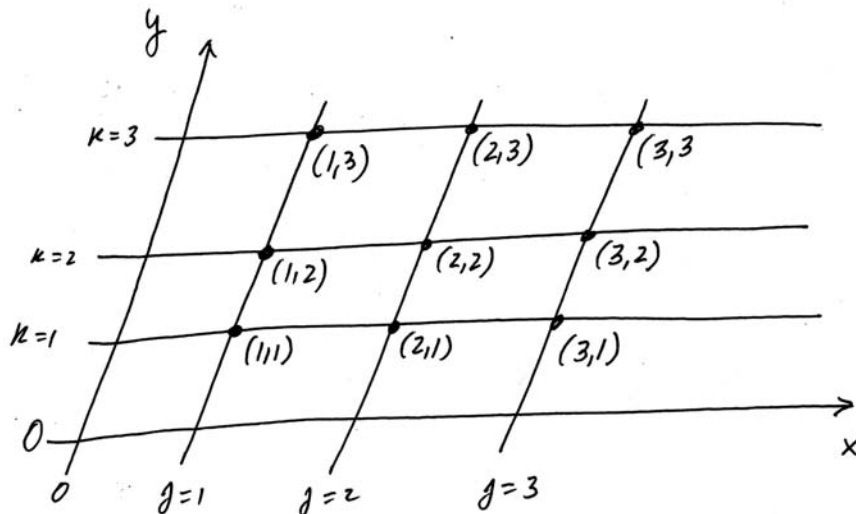
$$U_{j,k}^{n+1} - \frac{r_x}{2} \delta_x^2 U_{j,k}^{n+1} - \frac{r_y}{2} U_{j,k}^{n+1} = \left[1 + \frac{r_x \delta_x^2 + r_y \delta_y^2}{2} U_{j,k}^n \right] \equiv \text{rhs}$$

or expanding $\delta_x^2 U_{j,k}^{n+1}$ in the lhs.

$$U_{j,k}^{n+1} + \frac{1}{2} \left[-r_x U_{j-1,k}^{n+1} + 2r_x U_{j,k}^{n+1} - r_x U_{j+1,k}^{n+1} \right] + \frac{1}{2} \left[-r_y U_{j,k-1}^{n+1} + 2r_y U_{j,k}^{n+1} - r_y U_{j,k+1}^{n+1} \right] = \text{rhs}$$

$$\text{or } U_{j,k}^{n+1} + \frac{1}{2} \left[-r_y U_{j,k-1}^{n+1} - r_x U_{j-1,k}^{n+1} + 2(r_x + r_y) U_{j,k}^{n+1} - r_x U_{j+1,k}^{n+1} - r_y U_{j,k+1}^{n+1} \right] = \text{rhs.}$$

To construct the matrix consider



If BC's are homogeneous (DIRICHLET). 6

$k=1$

$$U_{11}^{n+1} + \frac{1}{2} [2(r_x+r_y)U_{11}^{n+1} - r_x U_{2,1}^{n+1} - r_y U_{1,2}^{n+1}] = \text{rhs.}$$

$$U_{2,1}^{n+1} + \frac{1}{2} [-r_x U_{11}^{n+1} + 2(r_x+r_y)U_{2,1}^{n+1} - r_x U_{3,1}^{n+1} - r_y U_{2,2}^{n+1}] = \text{rhs.}$$

$$\vdots$$

$$U_{N_x-1,1}^{n+1} + \frac{1}{2} [-r_x U_{N_x-2,1}^{n+1} + 2(r_x+r_y)U_{N_x-1,1}^{n+1} - r_y U_{N_x-1,2}^{n+1}] = \text{rhs}$$

$k=2$

$$U_{1,2}^{n+1} + \frac{1}{2} [-r_y U_{11}^{n+1} + 2(r_x+r_y)U_{1,2}^{n+1} - r_x U_{2,2}^{n+1} - r_y U_{1,3}^{n+1}] = \text{rhs}$$

$$U_{2,2}^{n+1} + \frac{1}{2} [-r_y U_{2,1}^{n+1} - r_x U_{1,2}^{n+1} + 2(r_x+r_y)U_{2,2}^{n+1} - r_x U_{3,2}^{n+1} - r_y U_{2,3}^{n+1}] = \text{rhs}$$

$$\vdots$$

$$U_{N_x-1,2}^{n+1} + \frac{1}{2} [-r_y U_{N_x-1,1}^{n+1} - r_x U_{N_x-2,2}^{n+1} + 2(r_x+r_y)U_{N_x-1,2}^{n+1} - r_y U_{N_x-1,3}^{n+1}] = \text{rhs}$$

$k=N_y-1$

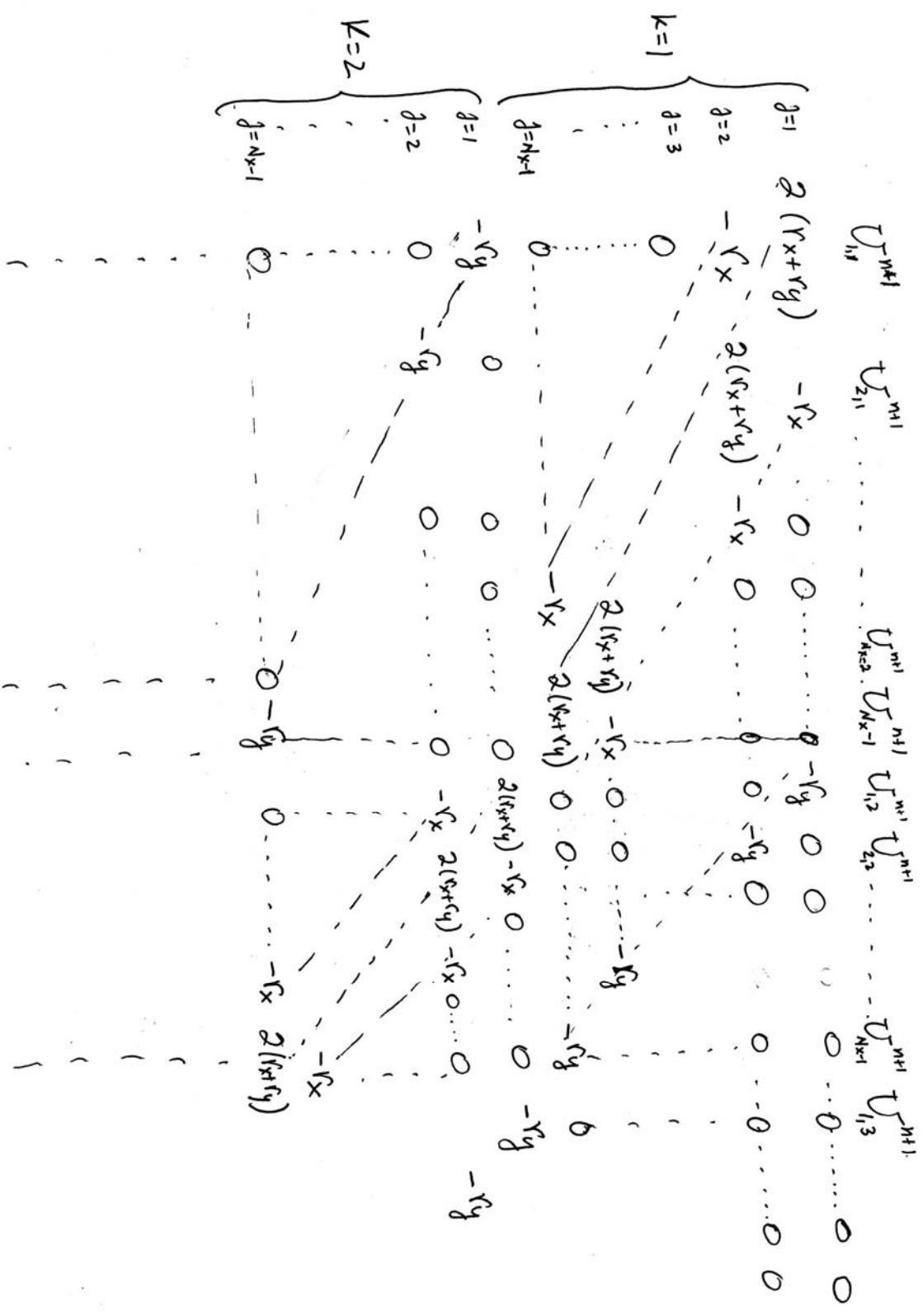
It's convenient to define the unknown vector \vec{U}^n by rows.

$$\vec{U}^n = \left[\begin{array}{c} U_{11}^n \\ U_{21}^n \\ \vdots \\ U_{N_x-1,1}^n \\ U_{1,2}^n \\ U_{2,2}^n \\ \vdots \\ U_{N_x-1,2}^n \\ \vdots \\ \vdots \end{array} \right] \left. \begin{array}{l} \text{1st row} \\ \text{in } x\text{-direction} \\ \\ \text{2nd row} \\ \text{in } x\text{-direction} \end{array} \right\}$$

$$\left[\begin{array}{c} \vdots \\ U_{1,N_y-1}^n \\ U_{2,N_y-1}^n \\ \vdots \\ U_{N_x-1,N_y-1}^n \end{array} \right] \left. \begin{array}{l} \text{last row} \\ \text{in} \\ x\text{-direction.} \end{array} \right\}$$

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MATRIX REPRESENTATION.



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The matrix Representation of the previous page
 Can be written in compact form as

$$\boxed{[I + \frac{1}{2} G] \hat{U}^{n+1} = [I - \frac{1}{2} G] \hat{U}^n} \quad (8.1)$$

where

$$C \equiv \begin{bmatrix} D_x & D_y & 0 & 0 & \dots & 0 \\ D_y & D_x & D_y & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & D_y & D_x & D_y \\ 0 & \dots & 0 & D_y & D_x \end{bmatrix}_{[(N_x-1)(N_y-1)]_x [(N_x-1)(N_y-1)]}$$

$$D_x \equiv \begin{bmatrix} 2(r_x+r_y) & -r_x & 0 & \dots & 0 \\ -r_x & 2(r_x+r_y) & -r_x & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & -r_x & 2(r_x+r_y) & -r_x \\ 0 & \dots & 0 & -r_x & 2(r_x+r_y) \end{bmatrix}_{(N_x-1)_x (N_x-1)}$$

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$$D_y \equiv \begin{pmatrix} -r_y & 0 & \dots & 0 \\ 0 & -r_y & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & -r_y \end{pmatrix}$$

The linear system (8.1) can be solved by an extension of the tridiagonal algorithm to block systems.

It requires $\frac{5}{3} N_y (N_x)^3$ multiplications per time step.

Too expensive!

where

$$a = -\frac{\alpha \Delta t}{2(\Delta y)^2} = -\frac{1}{2} r_y$$

$$b = -\frac{\alpha \Delta t}{2(\Delta x)^2} = -\frac{1}{2} r_x$$

$$c = 1 + r_x + r_y$$

$$d_{i,j}^n = u_{i,j}^n + \frac{\alpha \Delta t}{2} (\hat{\delta}_x^2 + \hat{\delta}_y^2) u_{i,j}^n$$

If we apply Eq. (4-101) to the two-dimensional (6 × 6) computational mesh shown in Fig. 4-15, the following system of 16 linear algebraic equations must be solved at each (n + 1) time level.

$$\begin{bmatrix}
 c & b & 0 & 0 & a & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 b & c & b & 0 & 0 & a & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & b & c & b & 0 & 0 & a & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & b & c & 0 & 0 & 0 & a & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 a & 0 & 0 & c & b & 0 & 0 & 0 & a & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & a & 0 & 0 & b & c & b & 0 & 0 & a & 0 & 0 & 0 & 0 & 0 & 0 \\
 & a & 0 & 0 & b & c & b & 0 & 0 & 0 & a & 0 & 0 & 0 & 0 & 0 \\
 & & a & 0 & 0 & b & c & 0 & 0 & 0 & 0 & a & 0 & 0 & 0 & 0 \\
 & & & a & 0 & 0 & c & b & 0 & 0 & 0 & 0 & a & 0 & 0 & 0 \\
 & & & & a & 0 & 0 & b & c & b & 0 & 0 & 0 & a & 0 & 0 \\
 & & & & & a & 0 & 0 & b & c & b & 0 & 0 & 0 & a & 0 \\
 & & & & & & a & 0 & 0 & c & b & 0 & 0 & 0 & 0 & 0 \\
 & & & & & & & a & 0 & 0 & b & c & b & 0 & 0 & 0 \\
 & & & & & & & & a & 0 & 0 & 0 & b & c & 0 & 0 \\
 0 & & & & & & & & & 0 & a & 0 & 0 & 0 & b & c
 \end{bmatrix}
 \begin{bmatrix}
 u_{2,2}^{n+1} \\
 u_{3,2}^{n+1} \\
 u_{4,2}^{n+1} \\
 u_{5,2}^{n+1} \\
 u_{2,3}^{n+1} \\
 u_{3,3}^{n+1} \\
 u_{4,3}^{n+1} \\
 u_{5,3}^{n+1} \\
 u_{2,4}^{n+1} \\
 u_{3,4}^{n+1} \\
 u_{4,4}^{n+1} \\
 u_{5,4}^{n+1} \\
 u_{2,5}^{n+1} \\
 u_{3,5}^{n+1} \\
 u_{4,5}^{n+1} \\
 u_{5,5}^{n+1}
 \end{bmatrix}
 =
 \begin{bmatrix}
 d_{2,2}''' \\
 d_{3,2}' \\
 d_{4,2}' \\
 d_{5,2}''' \\
 d_{2,3}'' \\
 d_{3,3} \\
 d_{4,3} \\
 d_{5,3}'' \\
 d_{2,4}'' \\
 d_{3,4} \\
 d_{4,4} \\
 d_{5,4}'' \\
 d_{2,5}''' \\
 d_{3,5}' \\
 d_{4,5}' \\
 d_{5,5}'''
 \end{bmatrix}
 \quad (4-102)$$

where $d' = d - au_0$
 $d'' = d - bu_0$
 $d''' = d - (a + b)u_0$

A system of equations, like Eq. (4-102), requires substantially more computer time to

ADI METHOD.

It's a predictor-corrector scheme

For example, Peaceman-Rachford version of it.

PREDICTOR STEP:

$$\boxed{U_{j,k}^{n+1/2} - \frac{r_x}{2} \delta_x^2 U_{j,k}^{n+1/2} = U_{j,k}^n + \frac{r_y}{2} \delta_y^2 U_{j,k}^n} \quad (1)$$

Formula (1) can be obtained this way
Split the continuous wave equation in 2. as

$$\frac{1}{2} U_{tt} = \sigma U_{xx} \quad \text{and} \quad \frac{1}{2} U_{tt} = \sigma U_{yy}$$

Use BT-CS at $(x_j, y_k, t_{n+1/2})$
time step $\Delta t/2$

$$\frac{1}{2} \left[\frac{U_{j,k}^{n+1/2} - U_{j,k}^n}{\Delta t/2} \right] = \sigma \delta_x^2 U_{j,k}^{n+1/2}$$

$$\Rightarrow U_{j,k}^{n+1/2} - r_x \delta_x^2 U_{j,k}^{n+1/2} = U_{j,k}^n \quad (2)$$

Use FT-CS at (x_j, y_k, t_n)
with time step size $\Delta t/2$.

$$\frac{1}{2} \left[\frac{U_{j,k}^{n+1/2} - U_{j,k}^n}{\Delta t/2} \right] = \sigma \delta_y^2 U_{j,k}^n$$

$$\Rightarrow U_{j,k}^{n+1/2} = r_y \delta_y^2 U_{j,k}^n + U_{j,k}^n \quad (3)$$

Adding (2) and (3).

$$2 U_{j,k}^{n+1/2} - r_x \delta_x^2 U_{j,k}^{n+1/2} = r_y \delta_y^2 U_{j,k}^n + 2 U_{j,k}^n$$

$$\Rightarrow \boxed{U_{j,k}^{n+1/2} - \frac{r_x}{2} \delta_x^2 U_{j,k}^{n+1/2} = \frac{r_y}{2} \delta_y^2 U_{j,k}^n + U_{j,k}^n} \quad (1)$$

IMPLICIT in the x-direction.

(2)

Corrector step:

$$\boxed{U_{j,k}^{n+1} - \frac{r_y}{2} \delta_y^2 U_{j,k}^{n+1} = U_{j,k}^{n+1/2} + \frac{r_x}{2} \delta_x^2 U_{j,k}^{n+1/2}} \quad (4)$$

Formula (4) can be obtained as follows:

$$\frac{1}{2} U_{tt} = \sigma U_{xx} \quad \text{and} \quad \frac{1}{2} U_{tt} = \sigma U_{yy}$$

F.T-CS at $(x_j, y_k, t_{n+1/2})$
with time step size $\Delta t/2$

$$\frac{1}{2} \left[\frac{U_{j,k}^{n+1} - U_{j,k}^{n+1/2}}{\Delta t/2} \right] = \sigma \delta_x^2 U_{j,k}^{n+1/2}$$

$$\Rightarrow U_{j,k}^{n+1} = r_x \delta_x^2 U_{j,k}^{n+1/2} + U_{j,k}^{n+1/2} \quad (5)$$

B.T-CS at (x_j, y_k, t_{n+1})
with time step size $\Delta t/2$

$$\frac{1}{2} \left[\frac{U_{j,k}^{n+1} - U_{j,k}^{n+1/2}}{\Delta t/2} \right] = \sigma \delta_y^2 U_{j,k}^{n+1}$$

$$\Rightarrow U_{j,k}^{n+1} - r_y \delta_y^2 U_{j,k}^{n+1} = U_{j,k}^{n+1/2} \quad (6)$$

Adding (5) and (6).

$$2U_{j,k}^{n+1} - r_y \delta_y^2 U_{j,k}^{n+1} = r_x \delta_x^2 U_{j,k}^{n+1/2} + 2U_{j,k}^{n+1/2}$$

$$\Rightarrow \boxed{U_{j,k}^{n+1} - \frac{r_y}{2} \delta_y^2 U_{j,k}^{n+1} = \frac{r_x}{2} \delta_x^2 U_{j,k}^{n+1/2} + U_{j,k}^{n+1/2}} \quad (7)$$

Implicit in the y-direction.

③

The two finite difference equations to be solved are

(I) Implicit in the x-direction:

$$\boxed{U_{j,k}^{n+1/2} - \frac{r_x}{2} \delta_x^2 U_{j,k}^{n+1/2} = \frac{r_y}{2} \delta_y^2 U_{j,k}^n + U_{j,k}^n} \quad \text{Predictor.}$$

(II) Implicit in the y-direction:

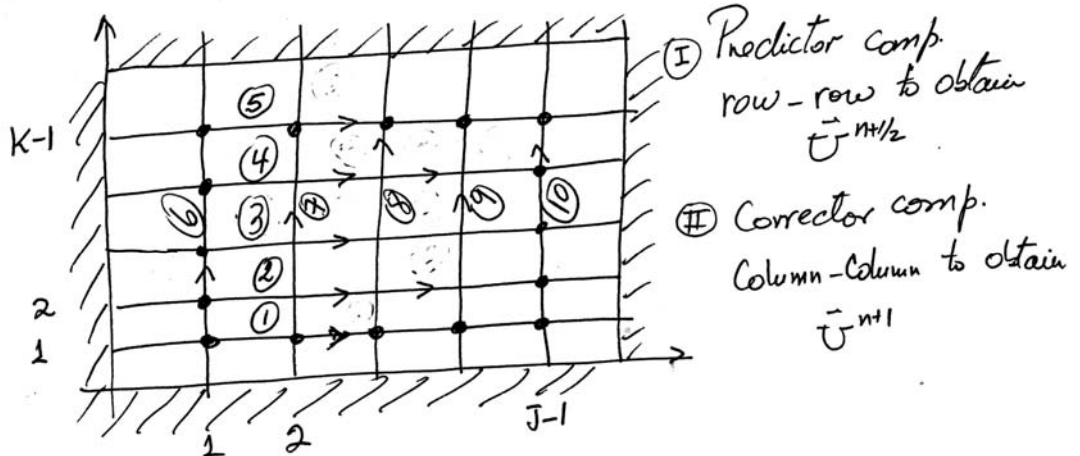
$$\boxed{U_{j,k}^{n+1} - \frac{r_y}{2} \delta_y^2 U_{j,k}^{n+1} = \frac{r_x}{2} \delta_x^2 U_{j,k}^{n+1/2} + U_{j,k}^{n+1/2}} \quad \text{Corrector.}$$

Solution:

(I) is solved using the tridiagonal algorithm row by row.

That way, we can obtain all the approximate values at time level $t_{n+1/2}$. Each row solution is obtained independently from all other rows.

Order in the computation



(4)

① Matrix Representation of Predictor Step.

Row k :
$$[I + C_x] \vec{U}_k^{n+1/2} = \vec{g}_{y,k}^n \quad (8)$$

where

$$\vec{U}_k^{n+1/2} = \begin{bmatrix} U_{1,k}^{n+1/2} \\ U_{2,k}^{n+1/2} \\ \vdots \\ U_{J-1,k}^{n+1/2} \end{bmatrix},$$

$$C_x = \frac{r_x}{2} \begin{bmatrix} 2 & -1 & 0 & \dots & 0 \\ -1 & 2 & -1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & \vdots & -1 & 2 \end{bmatrix}$$

$$\vec{g}_{y,k}^n = \begin{bmatrix} U_{1,k}^n + \frac{r_y}{2} S_y^2 U_{1,k}^n \\ \vdots \\ U_{J-1,k}^n + \frac{r_y}{2} S_y^2 U_{J-1,k}^n \end{bmatrix}$$

We need to solve this system for each row $k=1, 2, \dots, K-1$.
using the tridiagonal algorithm.

Similarly, Once all values at ^{time} level $t^{n+1/2}$ have been obtained, the next corrector step will compute all values at ^{time} level t^{n+1} . This is done column by column using

$$\vec{U}_j^{n+1} = \begin{bmatrix} U_{j,1}^{n+1} \\ U_{j,2}^{n+1} \\ \vdots \\ U_{j,K-1}^{n+1} \end{bmatrix}, \quad C_y = \frac{r_y}{2} \begin{bmatrix} 2 & -1 & 0 & \dots & 0 \\ -1 & 2 & -1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & \vdots & -1 & 2 \end{bmatrix}, \quad \vec{g}_{x,j}^{n+1/2} = \begin{bmatrix} U_{j,1}^{n+1/2} + \frac{r_x}{2} S_x^2 U_{j,1}^n \\ \vdots \\ U_{j,K-1}^{n+1/2} + \frac{r_x}{2} S_x^2 U_{j,K-1}^n \end{bmatrix}$$

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of operations (Complexity of algorithm).

Predictor: Each tridiag. system requires $\approx 5J$ operations

operation: One mult/division + One add/subtr.

\Rightarrow A tridiag. system is solved $K-1$ times, then

$$\# \text{ operations Predictor step} = 5J(K-1) \approx 5JK.$$

Corrector: $J-1$ tridiag. systems require $\approx 5K$ operations of $K-1$ dimension. This is done $J-1$ times

$$\Rightarrow \# \text{ operations Corrector step} \approx 5K(J-1) \approx 5JK$$

TOTAL # operations $\approx 10JK$. much less than $\frac{5}{3}KJ^3$

employed when solving Crank-Nicholson tridiag. block system.

It can be proved (Homework).

a) Local discretization error of ADI is given by

$$\tau_{j,k}^n = (\tau_{j,k}^n)_{C-N} + O(\Delta t^2) = O(\Delta x^2) + O(\Delta y^2) + O(\Delta t^2).$$

b) Using Von Neumann method. that Peaceman-Rachford numerical scheme is unconditionally stable. Stable for all ^{positive} values of r_x and r_y .