



AE/ME 339

## Computational Fluid Dynamics (CFD)

K. M. Isaac

9/1/2005

topic4: Implicit method, Stability,  
ADI method

1

Computational Fluid Dynamics (AE/ME 339)

K. M. Isaac

MAEEM Dept., UMR

### Implicit form of difference equation

In the previous explicit method, the solution at time level  $n$ ,  $u_{i,n}$ , depended only on the known values of  $u_{i-1,n-1}$ ,  $u_{i,n-1}$ , and  $u_{i+1,n-1}$ , all which are at time level  $n-1$ .

Note:  $u_{i,n}$  and  $u_i^n$  mean the same.

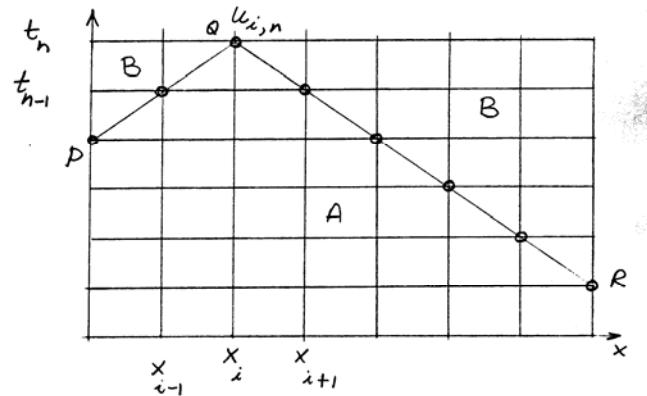
We will be using these interchangeably.

9/1/2005

topic4: Implicit method, Stability,  
ADI method

2

Nature of solution in explicit method can be illustrated graphically as shown below.



9/1/2005

topic4: Implicit method, Stability,  
ADI method

3

In this formulation, solution at  $(i,n)$ ,  $u_{i,n}$  is affected only by the values along and below boundary PQR (region A) in the previous figure. Values in region above PQR (region B) do not influence  $u_{i,n}$ .

Exact solution  $u(x,y)$  at Q depends on the values at all times earlier than  $t_n$ , a property of parabolic PDE.

Is a limitation of the explicit method.

$$\text{Stability criterion: } 0 \leq \frac{\Delta t}{(\Delta x)^2} \leq \frac{1}{2}$$

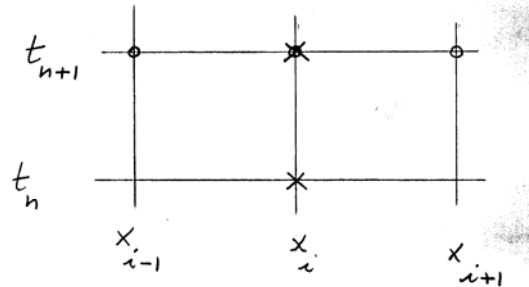
9/1/2005

topic4: Implicit method, Stability,  
ADI method

4

### Fully Implicit Method

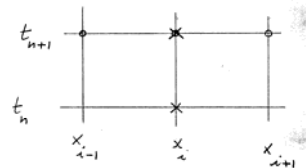
The nodes used in implicit method are illustrated in the figure below



As before, crosses (x) denote grid points used for  $\frac{\partial u}{\partial t}(u_t)$  and circles (o) for  $\frac{\partial^2 u}{\partial x^2}(u_{xx})$

The equation now becomes

$$\frac{u_{i,n+1} - u_{i,n}}{\Delta t} = \frac{u_{i-1,n+1} - 2u_{i,n+1} + u_{i+1,n+1}}{(\Delta x)^2}$$



Defining  $\lambda = \Delta t / \Delta x^2$ , the above equation can be rewritten as

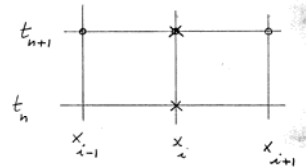
$$-\lambda u_{i-1,n+1} + (1 + 2\lambda)u_{i,n+1} - \lambda u_{i+1,n+1} = u_{i,n}$$

IC and BC are the same as before.

$$u_{0,n+1} = g_0(n+1)$$

$$u_{M,n+1} = g_1(n+1)$$

$$u_{i,0} = f(x_i)$$



Equations similar to the above should be written for each grid point

$$1 \leq i \leq M-1$$

Note: left boundary has  $i=0$  and right boundary has  $i=M$ , thus total of  $(M+1)$  grid points (also known as nodes) are present.

Thus we have  $(M-1)$  linear simultaneous equations with  $(M-1)$  unknowns. Explicit solution is not possible.

9/1/2005

topic4: Implicit method, Stability,  
ADI method

7

### Convergence of Implicit form

Can show using Taylor series

$$\frac{u_{i,n+1} - u_{i,n}}{\Delta t} = \frac{u_{i-1,n+1} - 2u_{i,n+1} + u_{i+1,n+1}}{(\Delta x)^2} + z_{i,n}$$

Where

$$z_{i,n} = \underbrace{\left[ \frac{\Delta t}{2} u_{tt} + \frac{(\Delta x)^2}{12} u_{xxxx} \right]}_{\text{leading terms in truncation error}} + O[(\Delta t)^2] + O[(\Delta x)^4]$$

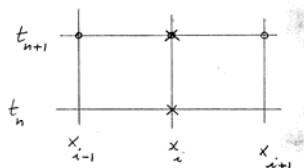
9/1/2005

topic4: Implicit method, Stability,  
ADI method

8

From the leading terms in the above

$$z_{i,n} = O\left[\Delta t + (\Delta x)^2\right]$$



It can be shown that implicit method converges to the exact solution of the PDE as  $\Delta t \rightarrow 0$  and  $\Delta x \rightarrow 0$  for any value of  $\frac{\Delta t}{(\Delta x)^2}$

The difference equation is now written for  $1 \leq i \leq M - 1$  as follows.

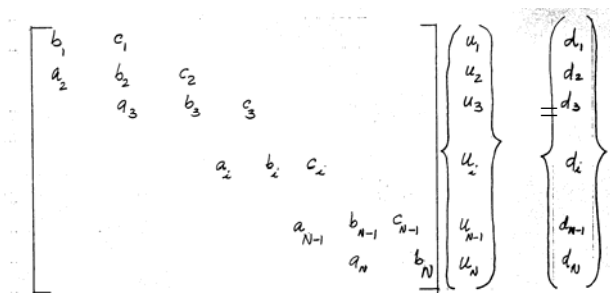
$$(1 + 2\lambda)u_{1,n+1} - \lambda u_{2,n+1} = u_{1,n} + \lambda g_0(t_{n+1}) \quad \text{for } i = 1$$

$$-\lambda u_{i-1,n+1} + (1 + 2\lambda)u_{i,n+1} - \lambda u_{i+1,n+1} = u_{i,n} \quad \text{for } 2 \leq i \leq M - 2$$

Finally, (show M, N in figure)

$$-\lambda u_{M-2,n+1} + (1 + 2\lambda)u_{M-1,n+1} = u_{M-1,n} + \lambda g_1(t_{n+1})$$

for  $i = M$



Note: In the matrix shown above, change in subscript  $N=M-1$  is adopted for simplicity.

The RHS terms  $d_1, d_2, \dots, d_N$  are known quantities. All matrix elements not shown are zero.

The matrix above is called a tridiagonal matrix i.e., only sub-diagonal, diagonal, and super diagonal terms are non-zero.

### Solution Procedure

Solution can be obtained by Gauss elimination procedure

### Recursion Relation

$$u_i = \gamma_i - \frac{c_i}{\beta_i} u_{i+1}$$

Constants  $\beta_i$  and  $\gamma_i$  are to be determined.

Substituting into the  $i^{\text{th}}$  equation of the set for  $u_{i-1}$  gives

$$a_i \left( \gamma_{i-1} - \frac{c_{i-1}}{\beta_{i-1}} u_i \right) + b_i u_i + c_i u_{i+1} = d_i$$

Rewrite as

$$u_i = \frac{d_i - a_i \gamma_{i-1}}{b_i - \frac{a_i c_{i-1}}{\beta_{i-1}}} - \frac{c_i u_{i+1}}{b_i - \frac{a_i c_{i-1}}{\beta_{i-1}}} \quad u_i = \gamma_i - \frac{c_i}{\beta_i} u_{i+1}$$

Comparing the two equations we have recursion relations for  $\beta$  and  $\gamma$

$$\beta_i = b_i - \frac{a_i c_{i-1}}{\beta_{i-1}} \quad u_i = \gamma_i - \frac{c_i}{\beta_i} u_{i+1}$$

$$\gamma_i = \frac{d_i - a_i \gamma_{i-1}}{\beta_i}$$

From the first equation (see matrix in slide 10)

$$u_1 = \frac{d_1}{b_1} - \frac{c_1}{b_1} u_2$$

$$\text{Where } \beta_1 = b_1 \text{ and } \gamma_1 = \frac{d_1}{\beta_1}$$

$$u_i = \gamma_i - \frac{c_i}{\beta_i} u_{i+1}$$

From the last equation (see matrix in slide 10)

$$u_N = \frac{d_N - a_N u_{N-1}}{b_N} = \frac{d_N - a_N \left( \gamma_{N-1} - \frac{c_{N-1}}{\beta_{N-1}} u_N \right)}{b_N}$$

Rearranging yields

$$u_N = \frac{d_N - a_N \gamma_{N-1}}{b_N - \frac{a_N c_{N-1}}{\beta_{N-1}}} = \gamma_N$$

$$u_i = \gamma_i - \frac{c_i}{\beta_i} u_{i+1}$$



Algorithm summary:

Recursion formulas for  $\beta_i$  and  $\gamma_i$

$$\beta_1 = b_1, \gamma_1 = \frac{d_1}{\beta_1}$$

$$\beta_i = b_i - \frac{a_i c_{i-1}}{\beta_{i-1}}, \quad i = 2, 3, \dots, N$$

$$\gamma_i = \frac{d_i - a_i \gamma_{i-1}}{\beta_i}, \quad i = 1, 2, 3, \dots, N$$

Once the coefficients have been calculated, the solution vector  $u$  can be calculated starting with  $u_N$  and going backwards, as follows:

$$u_N = \gamma_N$$

$$u_i = \gamma_i - \frac{c_i u_{i+1}}{\beta_i}, \quad i=N-1, N-2, \dots, 1$$

Note:  $n$  denotes time level and  $N$  denotes the last but one node in the  $i$ -direction.

Recursion formulas for  $\beta_i$  and  $\gamma_i$

$$\beta_1 = b_1, \gamma_1 = \frac{d_1}{\beta_1}$$

$$\beta_i = b_i - \frac{a_i c_{i-1}}{\beta_{i-1}}, \quad i = 2, 3, \dots, N$$

$$\gamma_i = \frac{d_i - a_i \gamma_{i-1}}{\beta_i}, \quad i = 1, 2, 3, \dots, N$$

Gauss elimination can cause large round-off errors.

Implicit schemes usually require more computational steps, but the ratio  $\Delta t/(\Delta x)^2$  has no restrictions, is a definite advantage.

### Stability (7.10)

A finite difference form is convergent if the solution tends to the exact solution as  $(\Delta t, \Delta x) \rightarrow 0$  (in the absence of round off error).

Stability refers to amplification of information present in IC, BC or introduced by errors in the numerical procedure such as round off error.

### Von-Neumann's stability analysis:

Stability implies only boundedness, not the magnitude of deviation from the true solution.

### Key features of stability analysis:

Assume that: i) At any stage,  $t=0$  here, a Fourier expansion can be made of some initial function  $f(x)$ , and a typical term in the expansion can be written as  $e^{j\beta x}$  where  $\beta$  is a positive constant and  $j = \sqrt{-1}$

ii) Separation of time and space dependence can be made.

At time  $t$ , the term becomes  $\psi(t) e^{j\beta x}$

By substituting in the difference equation, the form of  $\psi(t)$  can be determined and stability criterion established.

### Example

Explicit finite difference form

$$\frac{u_{i,n+1} - u_{i,n}}{\Delta t} = \frac{u_{i-1,n} - 2u_{i,n} + u_{i+1,n}}{(\Delta x)^2}$$

Substitute for each  $u$ .

$$\frac{\psi(t+\Delta t) e^{j\beta x} - \psi(t) e^{j\beta x}}{\Delta t} = \frac{\psi(t)}{(\Delta x)^2} \left[ e^{j\beta(x-\Delta x)} - 2e^{j\beta x} + e^{j\beta(x+\Delta x)} \right]$$

Cancel exp ( $j\beta x$ ) throughout

$$\psi(t + \Delta t) = \psi(t) \left[ 1 + \lambda (e^{-j\beta\Delta x} - 2 + e^{j\beta\Delta x}) \right]$$

Trig. identity:  $\cos(\theta) = \frac{e^{i\theta} + e^{-i\theta}}{2}$  can be used to get

$$\psi(t + \Delta t) = \psi(t) \left[ 1 + \lambda (-2 + 2\cos(\beta\Delta x)) \right]$$

Since  $\cos \theta = 1 - 2 \sin^2(\theta/2)$

$$\psi(t + \Delta t) = \psi(t) \left[ 1 - 2\lambda \left( 2 \sin^2 \frac{\beta \Delta x}{2} \right) \right]$$

$$\psi(t + \Delta t) = \psi(t) \left[ 1 - 4\lambda \sin^2 \left( \frac{\beta \Delta x}{2} \right) \right]$$

If we choose  $\psi(0) = 1$ , this has the solution

$$\psi(t) = \left( 1 - 4\lambda \sin^2(\beta \Delta x / 2) \right)^{(t/\Delta t)}$$

And can be proven by substitution (see next slide).

$$\begin{aligned} \psi(t + \Delta t) &= \left( 1 - 4\lambda \sin^2(\beta \Delta x / 2) \right)^{(t+\Delta t)/\Delta t} \\ &= \psi(t) \left[ 1 - 4\lambda \sin^2(\beta \Delta x / 2) \right] \end{aligned}$$

For stability,  $\psi(t)$  must be bounded as  $(\Delta t, \Delta x) \rightarrow 0$

This requires

$$\left| 1 - 4\lambda \sin^2(\beta \Delta x / 2) \right| \leq 1$$

An amplification factor  $\xi$  is usually defined as follows:

$$\xi = 1 - 4\lambda \sin^2(\beta\Delta x / 2)$$

Which shows  $|\xi| \leq 1$  for stability.

In the Fourier expansion we considered only one term corresponding to 1 value of  $\beta$ . When all possible values of  $\beta$  are considered  $\sin(\beta\Delta x/2)$  could become 1.

Therefore, the stability condition becomes

$$\lambda = \frac{\Delta t}{(\Delta x)^2} \leq \frac{1}{2} \quad \begin{array}{l} \lambda = 0.5, \quad |\xi| = 1 \\ \lambda = 1.0, \quad |\xi| = 3 \text{ (unstable)} \end{array}$$

Intuitively it implies that  $u_{i,n}$  affects  $u_{i,n+1}$  in a “non-negative” manner.

A similar analysis for the implicit method would give

$$\xi = \frac{1}{1 + 4\lambda \sin^2(\beta\Delta x / 2)}$$

Since  $\xi \leq 1$  for all  $\lambda$ , the procedure is unconditionally stable.

Consistency means that the procedure may in fact approximate the solution of the PDE under study and not the solution of some other PDE.