

## Chap. 4-6. Implicit Time Integration

- **AF-ADI Scheme**

- **Approximate factorization along  $\xi$ - and  $\eta$ - directions**

$$\left[ \mathbf{I} + \mu \Delta t \left( D_{\xi}^{-} \bar{\mathbf{A}}^{+} + D_{\xi}^{+} \bar{\mathbf{A}}^{-} \right) \right] \left[ \mathbf{I} + \mu \Delta t \left( D_{\eta}^{-} \bar{\mathbf{B}}^{+} + D_{\eta}^{+} \bar{\mathbf{B}}^{-} \right) \right] \Delta \bar{\mathbf{U}}_{i,j}^n = -\Delta t \bar{\mathbf{R}}_{i,j}^n$$

with the factorization error =  $\mu^2 \Delta t^2 \left( \partial \bar{\mathbf{A}}^n / \partial \xi \right) \left( \partial \bar{\mathbf{B}}^n / \partial \eta \right)$

- **The resulting formulation requires block tri-diagonal matrix inversion along  $\xi$ - and  $\eta$ - directions.**

$$D_{\xi}^{-} \bar{\mathbf{A}}^{+} = \frac{1}{\Delta \xi} \left( \bar{\mathbf{A}}_{i,j}^{+} - \bar{\mathbf{A}}_{i-1,j}^{+} \right), D_{\xi}^{+} \bar{\mathbf{A}}^{-} = \frac{1}{\Delta \xi} \left( \bar{\mathbf{A}}_{i+1,j}^{-} - \bar{\mathbf{A}}_{i,j}^{-} \right); D_{\eta}^{-} \bar{\mathbf{B}}^{+} = \frac{1}{\Delta \eta} \left( \bar{\mathbf{B}}_{i,j}^{+} - \bar{\mathbf{B}}_{i,j-1}^{+} \right), D_{\eta}^{+} \bar{\mathbf{B}}^{-} = \frac{1}{\Delta \eta} \left( \bar{\mathbf{B}}_{i,j+1}^{-} - \bar{\mathbf{B}}_{i,j}^{-} \right)$$

$$\left[ \mathbf{I} + \mu \frac{\Delta t}{\Delta \xi} \left( -\bar{\mathbf{A}}_{i-1,j}^{+} + \left( \bar{\mathbf{A}}_{i,j}^{+} - \bar{\mathbf{A}}_{i,j}^{-} \right) + \bar{\mathbf{A}}_{i+1,j}^{-} \right) \right] \underbrace{\left[ \mathbf{I} + \mu \frac{\Delta t}{\Delta \eta} \left( -\bar{\mathbf{B}}_{i,j-1}^{+} + \left( \bar{\mathbf{B}}_{i,j}^{+} - \bar{\mathbf{B}}_{i,j}^{-} \right) + \bar{\mathbf{B}}_{i,j+1}^{-} \right) \right]}_{\equiv \Delta \bar{\mathbf{U}}_{i,j}^*} \Delta \bar{\mathbf{U}}_{i,j}^n = -\Delta t \bar{\mathbf{R}}_{i,j}^n$$

- $\xi$ -sweep (relaxation along the  $\xi$ -direction)

$$-\left( \mu \frac{\Delta t}{\Delta \xi} \bar{\mathbf{A}}_{i-1,j}^{+} \right) \Delta \bar{\mathbf{U}}_{i-1,j}^* + \left[ \mathbf{I} + \mu \frac{\Delta t}{\Delta \xi} \left( \bar{\mathbf{A}}_{i,j}^{+} - \bar{\mathbf{A}}_{i,j}^{-} \right) \right] \Delta \bar{\mathbf{U}}_{i,j}^* + \left( \mu \frac{\Delta t}{\Delta \xi} \bar{\mathbf{A}}_{i+1,j}^{-} \right) \Delta \bar{\mathbf{U}}_{i+1,j}^* = -\Delta t \bar{\mathbf{R}}_{i,j}^n$$

- $\eta$ -sweep (relaxation along the  $\eta$ -direction)

$$-\left( \mu \frac{\Delta t}{\Delta \eta} \bar{\mathbf{B}}_{i,j-1}^{+} \right) \Delta \bar{\mathbf{U}}_{i,j-1}^n + \left[ \mathbf{I} + \mu \frac{\Delta t}{\Delta \eta} \left( \bar{\mathbf{B}}_{i,j}^{+} - \bar{\mathbf{B}}_{i,j}^{-} \right) \right] \Delta \bar{\mathbf{U}}_{i,j}^n + \left( \mu \frac{\Delta t}{\Delta \eta} \bar{\mathbf{B}}_{i,j+1}^{-} \right) \Delta \bar{\mathbf{U}}_{i,j+1}^n = \Delta \mathbf{U}_{i,j}^*$$

$$\rightarrow \bar{\mathbf{U}}_{i,j}^{n+1} = \bar{\mathbf{U}}_{i,j}^n + \Delta \bar{\mathbf{U}}_{i,j}^n$$

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- For 2-D  $N \times N$  mesh with single unknown at each grid point, operating count =  $O(N^2)$
- $\mu$  for C-N  $\leftrightarrow$  fully implicit :  $\frac{1}{2\Delta\xi} \leftrightarrow \frac{1}{\Delta\xi}, \frac{1}{2\Delta\eta} \leftrightarrow \frac{1}{\Delta\eta}$
- 3 types of errors
  - Linearization error for the implicit part
  - Factorization error to approximate the block penta-diagonal term
  - Discretization error of the residual term
- Factorization error dominant at high frequency range, and parallelization issue
- **LU-SGS scheme**
  - **Instead of dimensional splitting, a variant of Gauss-Seidel relaxation is applied to Eq. (2) in symmetric forward & backward manner.**
  - See the works by Yoon and Jameson(1988), and others
  - **Ex) LU-SGS for 1-D case**

$$\text{From } \frac{\partial \mathbf{U}}{\partial t} + \frac{\partial \mathbf{F}}{\partial x} = \mathbf{0}, \quad \frac{\Delta \mathbf{U}^n}{\Delta t} + (1 - \mu) \frac{\partial \mathbf{F}^n}{\partial x} + \mu \frac{\partial \mathbf{F}^{n+1}}{\partial x} = \mathbf{0}$$

$$\text{With } \mathbf{F}^{n+1} \cong \mathbf{F}^n + \mathbf{A}^n \Delta \mathbf{U}^n, \quad \mathbf{A}^n = \left( \frac{\partial \mathbf{F}}{\partial \mathbf{U}} \right)^n \rightarrow \frac{\Delta \mathbf{U}^n}{\Delta t} + \mu \left( \frac{\partial \mathbf{A}^n}{\partial x} \right) \Delta \mathbf{U}^n = -\frac{\partial \mathbf{F}^n}{\partial x} = -\mathbf{R}(\mathbf{U}^n).$$

$$\text{From } \mathbf{A}^n = \mathbf{A}^+ + \mathbf{A}^-, \quad \left[ \mathbf{I} + \mu \Delta t \left( D^- \mathbf{A}^+ + D^+ \mathbf{A}^- \right) \right] \Delta \mathbf{U}^n = -\Delta t \mathbf{R}^n \quad \text{or}$$

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$$\Delta \mathbf{U}_i^n + \lambda (\mathbf{A}_i^+ \Delta \mathbf{U}_i^n - \mathbf{A}_{i-1}^+ \Delta \mathbf{U}_{i-1}^n) + \lambda (\mathbf{A}_{i+1}^- \Delta \mathbf{U}_{i+1}^n - \mathbf{A}_i^- \Delta \mathbf{U}_i^n) = -\Delta t \mathbf{R}_i^n, \quad \lambda = \mu \Delta t / \Delta x$$

$$\rightarrow -\lambda \mathbf{A}_{i-1}^+ \Delta \mathbf{U}_{i-1}^n + \left[ \mathbf{I} + \lambda (\mathbf{A}_i^+ - \mathbf{A}_i^-) \right] \Delta \mathbf{U}_i^n + \lambda \mathbf{A}_{i+1}^- \Delta \mathbf{U}_{i+1}^n = -\Delta t \mathbf{R}_i^n \quad \text{or} \quad A \Delta \mathbf{U}^n = -\Delta t \mathbf{R}^n$$

Gauss-Seidel relaxations in a symmetric fashion as follows

- (forward sweep)  $-\lambda \mathbf{A}_{i-1}^+ \Delta \mathbf{U}_{i-1}^* + \left[ \mathbf{I} + \lambda (\mathbf{A}_i^+ - \mathbf{A}_i^-) \right] \Delta \mathbf{U}_i^* = -\Delta t \mathbf{R}_i^n$

- (backward sweep)  $\left[ \mathbf{I} + \lambda (\mathbf{A}_i^+ - \mathbf{A}_i^-) \right] \Delta \mathbf{U}_i^n + \lambda \mathbf{A}_{i+1}^- \Delta \mathbf{U}_{i+1}^n - \lambda \mathbf{A}_{i-1}^+ \Delta \mathbf{U}_{i-1}^* = -\Delta t \mathbf{R}_i^n$

By subtracting forward sweep from backward sweep,

$$\left[ \mathbf{I} + \lambda (\mathbf{A}_i^+ - \mathbf{A}_i^-) \right] \Delta \mathbf{U}_i^n + \lambda \mathbf{A}_{i+1}^- \Delta \mathbf{U}_{i+1}^n = \left[ \mathbf{I} + \lambda (\mathbf{A}_i^+ - \mathbf{A}_i^-) \right] \Delta \mathbf{U}_i^* \quad \text{or} \quad U \Delta \mathbf{U}^n = D \Delta \mathbf{U}^*$$

From  $\Delta \mathbf{U}^* = D^{-1} U \Delta \mathbf{U}^n$  and  $L \Delta \mathbf{U}^* = -\Delta t \mathbf{R}^n$ , we have  $LD^{-1}U \Delta \mathbf{U}^n = -\Delta t \mathbf{R}^n$

$\rightarrow$  a variant of LU decomposition to invert  $A \Delta \mathbf{U}^n = -\Delta t \mathbf{R}^n$

with  $L = -\lambda \mathbf{A}_{i-1}^+ + \mathbf{I} + \lambda (\mathbf{A}_i^+ - \mathbf{A}_i^-)$ ,  $D = \mathbf{I} + \lambda (\mathbf{A}_i^+ - \mathbf{A}_i^-)$ ,  $U = \mathbf{I} + \lambda (\mathbf{A}_i^+ - \mathbf{A}_i^-) + \lambda \mathbf{A}_{i+1}^-$ .

- **Symmetric G-S relaxation to Eq. (2)**

- The same procedure for 2-D case with

$$\left[ \mathbf{I} + \mu \Delta t \left( D_\xi^- \bar{\mathbf{A}}^+ + D_\xi^+ \bar{\mathbf{A}}^- + D_\eta^- \bar{\mathbf{B}}^+ + D_\eta^+ \bar{\mathbf{B}}^- \right) \right] \Delta \bar{\mathbf{U}}^n = -\Delta t \bar{\mathbf{R}}_{i,j}^n \quad \text{or} \quad A \Delta \bar{\mathbf{U}}^n = -\Delta t \bar{\mathbf{R}}^n$$

- (forward sweep)  $\left[ \mathbf{I} + \Delta t \left\{ (\bar{\mathbf{A}}_{i,j}^+ - \bar{\mathbf{A}}_{i,j}^-) + (\bar{\mathbf{B}}_{i,j}^+ - \bar{\mathbf{B}}_{i,j}^-) \right\} \right] \Delta \bar{\mathbf{U}}_{i,j}^* - \Delta t \bar{\mathbf{A}}_{i-1,j}^+ \Delta \bar{\mathbf{U}}_{i-1,j}^* - \Delta t \bar{\mathbf{B}}_{i,j-1}^+ \Delta \bar{\mathbf{U}}_{i,j-1}^*$   
 $= -\Delta t \bar{\mathbf{R}}_{i,j}^n$

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- (backward sweep)

$$\left[ \mathbf{I} + \Delta t \left\{ \left( \bar{\mathbf{A}}_{i,j}^+ - \bar{\mathbf{A}}_{i,j}^- \right) + \left( \bar{\mathbf{B}}_{i,j}^+ - \bar{\mathbf{B}}_{i,j}^- \right) \right\} \right] \Delta \bar{\mathbf{U}}_{i,j}^n + \Delta t \bar{\mathbf{A}}_{i+1,j}^- \Delta \bar{\mathbf{U}}_{i+1,j}^n + \Delta t \bar{\mathbf{B}}_{i,j+1}^- \Delta \bar{\mathbf{U}}_{i,j+1}^n - \Delta t \bar{\mathbf{A}}_{i-1,j}^+ \Delta \bar{\mathbf{U}}_{i-1,j}^* - \Delta t \bar{\mathbf{B}}_{i,j-1}^+ \Delta \bar{\mathbf{U}}_{i,j-1}^* = -\Delta t \bar{\mathbf{R}}_{i,j}^n$$

Subtracting forward from backward sweep gives  $LD^{-1}U\Delta\bar{\mathbf{U}}^n = -\Delta t\bar{\mathbf{R}}^n$ .

with  $L = \mathbf{I} + \Delta t \left( D_{\xi}^- \bar{\mathbf{A}}^+ + D_{\eta}^- \bar{\mathbf{B}}^+ - \bar{\mathbf{A}}^- - \bar{\mathbf{B}}^- \right)$ ,  $D = \mathbf{I} + \Delta t \left( \bar{\mathbf{A}}^+ - \bar{\mathbf{A}}^- + \bar{\mathbf{B}}^+ - \bar{\mathbf{B}}^- \right)$ ,

$$U = \mathbf{I} + \Delta t \left( D_{\xi}^+ \bar{\mathbf{A}}^- + D_{\eta}^+ \bar{\mathbf{B}}^- + \bar{\mathbf{A}}^+ + \bar{\mathbf{B}}^+ \right).$$

- Instead of  $\lambda$ -based splitting  $\bar{\mathbf{A}}^{\pm} = \bar{R}_A \Lambda_A^{\pm} \bar{R}_A^{-1}$ ,  $\bar{\mathbf{B}}^{\pm} = \bar{R}_B \Lambda_B^{\pm} \bar{R}_B^{-1}$ , a  $\rho$ -based splitting is introduced to avoid block matrix inversion and to maintain diagonal dominance.

$$\bar{\mathbf{A}}^{\pm} = \frac{1}{2} \left( \bar{\mathbf{A}} \pm \rho(\bar{\mathbf{A}}) \mathbf{I} \right), \quad \bar{\mathbf{B}}^{\pm} = \frac{1}{2} \left( \bar{\mathbf{B}} \pm \rho(\bar{\mathbf{B}}) \mathbf{I} \right) \quad \text{with} \quad \rho(\bar{\mathbf{A}}) = k \left| \lambda(\bar{\mathbf{A}}) \right|_{\max} \quad \text{and} \quad k = 1 + \alpha (> 0)$$

- A variant of LU decomposition to efficiently invert  $A\Delta\mathbf{U}^n = -\Delta t\mathbf{R}^n$
- Complete vectorization along diagonals of  $i+j=const$  (or coloring scheme)
- Easily extendable to complex flows, such as chemical reacting flows
  - $\lambda$ -based splitting can be used for non-calorically perfect gas.
- Parallelization issue and temporal accuracy
- LU-SGS / AF-ADI relaxation combined with MG
  - After transferring solution and residuals onto coarse grid, compute the multi-grid forcing term ( $\mathbf{P}$ ) and update solution  $\rightarrow A\Delta\mathbf{U}^n$  by LU-SGS or AF-ADI =  $-\Delta t(\mathbf{R} + \mathbf{P})$  by multi-grid

## Chap. 4-7. Time-Accurate Schemes

- ***TVD R-K scheme***

- **Higher-order R-K schemes preserving TVD stability**

- See the works by Shu and Osher(1988), and others
- Linearly stable R-K schemes are not sufficient to support monotonic solutions.
- Ex: computed solutions with linearly(LS)/non-linearly(NLS) stable R-K methods

$$u_t + uu_x = 0 \text{ with } u_0(x) = \begin{cases} 1, & \text{if } x \leq 0 \\ -0.5, & \text{if } x > 0 \end{cases}$$

- One-step Euler-forward from the semi-discrete form of  $u_t = L(u)$  by evaluating  $F_{i+1/2}$  with a 2nd-order monotonic flux  $\rightarrow u^{n+1} = u^n + \Delta t L(u^n)$  with  $TV(u^{n+1}) \leq TV(u^n)$
- 2-stage R-K time integrations

$$\text{(LS)} \quad u^{(0)} = u^n, \quad u^{(1)} = u^{(0)} - 20\Delta t L(u^{(0)}), \quad u^{(2)} = u^{(0)} + \frac{41}{40}\Delta t L(u^{(0)}) - \frac{1}{40}\Delta t L(u^{(1)}) \quad \rightarrow \quad u^{n+1} = u^{(2)}$$

$$\text{(NLS)} \quad u^{(0)} = u^n, \quad u^{(1)} = u^{(0)} + \Delta t L(u^{(0)}), \quad u^{(2)} = \frac{1}{2}u^{(0)} + \frac{1}{2}u^{(1)} + \frac{1}{2}\Delta t L(u^{(1)}) \quad \rightarrow \quad u^{n+1} = u^{(2)}$$

$\rightarrow$  both are the same for linear case, but yield different solutions for non-linear case.

- TVD-stable, multi-stage R-K schemes by optimizing standard R-K coefficients
- One-step TVD scheme:  $u^{n+1} = u^n + \Delta t L(u^n) \equiv EL(u^n)$  with  $TV(u^{n+1}) \leq TV(u^n)$
- Multi-stage TVD R-K schemes: from general  $m$ -stage R-K method as

$$u^{(0)} = u^n, \quad u^{(k)} = \sum_{j=0}^{k-1} (\alpha_{jk} u^{(j)} + \beta_{jk} \Delta t L(u^{(j)})) \text{ with } \sum_{j=0}^{k-1} \alpha_{jk} = 1, \quad 1 \leq k \leq m, \quad u^{n+1} = u^{(m)}$$

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If  $\alpha_{jk}$  and  $\beta_{jk}$  are all positive,  $u^{(k)}$  can be regarded as a convex combination of  $EL$  operators as

$$u^{(k)} = \sum_{j=0}^{k-1} \alpha_{jk} \left( u^{(j)} + \Delta t_j L(u^{(j)}) \right) = \sum_{j=0}^{k-1} \alpha_{jk} EL(u^{(j)}) \text{ with } \Delta t_j = \frac{\beta_{jk}}{\alpha_{jk}} \Delta t. \text{ Thus, } TV(u^k) \leq TV(u^{k-1})$$

For  $m$ -stage R-K methods, search optimal positive  $\alpha_{jk}, \beta_{jk}$  yielding a maximal CFL time-step.

- 2-stage/2nd-order TVD R-K schemes

$$u^{(1)} = u^{(0)} + \Delta t L(u^{(0)}), \quad u^{(2)} = \frac{1}{2} u^{(0)} + \frac{1}{2} u^{(1)} + \frac{1}{2} \Delta t L(u^{(1)})$$

- 3-stage/3rd-order TVD R-K schemes

$$u^{(1)} = u^{(0)} + \Delta t L(u^{(0)}), \quad u^{(2)} = \frac{3}{4} u^{(0)} + \frac{1}{4} u^{(1)} + \frac{1}{4} \Delta t L(u^{(1)}), \quad u^{(3)} = \frac{1}{3} u^{(0)} + \frac{2}{3} u^{(2)} + \frac{2}{3} \Delta t L(u^{(2)})$$

- $m$ -stage/ $m$  th-order TVD R-K schemes?

- 5-stage/4th-order SSPRK scheme

- Non-linearly stable R-K schemes with maximal CFL number and minimal complexity

- **Dual Time Stepping with 2<sup>nd</sup>-order Backward Differencing Formula**

- **2<sup>nd</sup>-order A-stable BDF**

$$\frac{d\mathbf{U}}{dt} = -\mathbf{R}(\mathbf{U}) \rightarrow \frac{3}{2\Delta t} \mathbf{U}^{n+1} - \frac{2}{\Delta t} \mathbf{U}^n + \frac{1}{2\Delta t} \mathbf{U}^{n-1} + \mathbf{R}(\mathbf{U}^{n+1}) = \mathbf{0}$$

- (Dahlquist's stability barrier theorem) For  $y_t = f(y, t)$ , general linear multi-step methods

$$\text{of } \sum_{i=0}^k \alpha_i y_{n+i} = \sum_{i=0}^k h \beta_i f_{n+i} \text{ can achieve 'A-stability' with the maxim order of '2'.$$

## Chap. 4-7. Time-Accurate Schemes

- **Instead of linearizing the implicit residual term, solve the full non-linear BDF by introducing a pseudo-time derivative.**

- Define extended residual,  $\mathbf{R}^*(\tilde{\mathbf{U}})$ , in terms of  $\tilde{\mathbf{U}}$

$$\mathbf{R}^*(\tilde{\mathbf{U}}) \equiv \frac{3\tilde{\mathbf{U}} - 4\mathbf{U}^n + \mathbf{U}^{n-1}}{2\Delta t} + \mathbf{R}(\tilde{\mathbf{U}}) = \mathbf{0}$$

$$\xrightarrow[\text{to update solution, } \tilde{\mathbf{U}}, \text{ by sub-iterations in terms of } \tau]{\text{introduce a pseudo-time derivative term}} \frac{d\tilde{\mathbf{U}}}{d\tau} + \mathbf{R}^*(\tilde{\mathbf{U}}) = \mathbf{0} \quad \text{Eq.(3)}$$

- two-time scale: physical-time( $t$ ) for time accuracy and pseudo-time( $\tau$ ) for convergence

- **Solution of the original BDF can be recovered by converging Eq. (3).**
  - Explicit scheme
    - Modified R-K schemes, such as (4,2) and (5,3) schemes, with local time stepping
    - Implicit residual smoothing and multi-grid
  - Implicit schemes, such as LU-SGS or AF-ADI with multi-grid
- **With a fast convergence of inner iterations, an efficient 2<sup>nd</sup>-order time-accurate scheme allowing a larger  $\Delta t$  can be obtained.**