

High-Fidelity Simulations for Turbulent Flows

Luca Sciacovelli

DynFluid Laboratory
Arts et Métiers Institute of Technology
<http://savoir.ensam.eu/moodle>

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Part VII

Numerical simulation of unsteady flows

1 Introduction

2 Stability

3 One-step Methods

4 Multistep Methods

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Motivation

- ▶ Preliminary design of fluid systems mostly based on nominal conditions
- ▶ CFD of steady flow is now mature, but only **averages** are computed
 - These are not always enough (instabilities, growth rate, vortex shedding..)
 - They are even not always meaningful (steady solution may be different from time-averaged one)
 - Some flows are dominated by vortical structures (e.g., separated flows)

Characteristics of unsteady flows

- ▶ Time is the fourth coordinate direction, which must be discretized
- ▶ Differences w.r.t. spatial discretization: **directionality**
 - A force at a given x may influence the flow anywhere else
 - A force at a given t will affect the flow only in the future
- ▶ Methods very similar to the ones applied to IVP for ODEs
- ▶ The basic problem is to find the solution w a short time Δt after the initial point
 - The solution at $t^1 = t^0 + \Delta t$ is used as new I.C.; the solution is then advanced to $t^2 = t^1 + \Delta t$, ..

Classification of unsteady flows

Unsteady flows classified as **slow** or **rapid** based on the ratio of a characteristic speed of flow unsteadiness to wave speed

Example 1: Inviscid Burgers equation

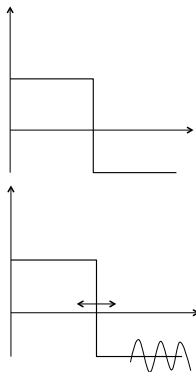
$$\frac{\partial w}{\partial t} + \frac{\partial(w^2/2)}{\partial x} = 0 \quad x \in [-1, 1], \quad \begin{cases} w(-1, t) = 1 \\ w(1, t) = -1 \end{cases}$$

Solution: steady shock located at $x = 0$

Example 2: Inviscid Burgers equation + periodic perturbation

$$\frac{\partial w}{\partial t} + \frac{\partial(w^2/2)}{\partial x} = 0 \quad x \in [-1, 1], \quad \begin{cases} w(-1, t) = 1 \\ w(1, t) = -1 + A \sin(\omega t) \end{cases} \quad \text{with } A \ll 1$$

Solution: shock oscillating around $x = 0$



► Characteristic speed of the unsteady perturbation:

$$u_p = \frac{\Delta x}{t_p} = \frac{\omega \Delta x}{2\pi}$$

► Wave speed (considering $\text{CFL} = \mathcal{O}(1)$):

$$u_0 = u, \quad \Delta t = \text{CFL} \frac{\Delta x}{|u|}$$

► Speed and time ratios:

$$\frac{u_p}{u_0} = \frac{\omega \Delta x}{2\pi u}, \quad \frac{t_p}{\Delta t} = \frac{u}{\text{CFL} \Delta x} \frac{2\pi}{\omega} = \frac{1}{\text{CFL}} \frac{1}{u_p/u_0}$$

$$\text{If } \frac{u_p}{u_0} \ll 1 \implies \frac{t_p}{\Delta t} \gg 1$$

• **Lots of iterations** with an explicit method

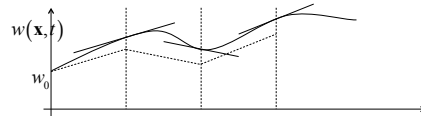
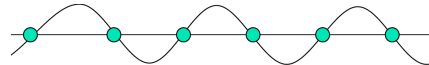
Remember: basics of Time-Marching Methods

Unconditionally stable (implicit) schemes allow large time steps

- ▶ Too large time steps induce **aliasing**
 - **Nyquist barrier**: a signal of frequency f has to be sampled at a frequency $f_s > 2f$
- ▶ Large time steps induce **large truncation errors**
 - **Accuracy barrier**: depends on the numerical scheme
Example for backward Euler scheme (1st-order):
 - Geometric interpretation: solution is linearly extrapolated
 - Quick error accumulation
 - 1st-order accuracy typically not sufficient
 - Achieving high-accurate **and** unconditionally stable schemes is a hard task

How to control the error? 2 possibilities:

1. **Decrease Δt** \Rightarrow very costly if an implicit scheme is used
2. **Increase accuracy**: for a given level of accuracy, increasing the order of the time integration scheme is more efficient than decreasing Δt



Typical methods:

1. One-step methods

- Forward Euler
- Backward Euler
- Trapezoidal (Crank-Nicolson)
- Taylor-series Methods
- Runge-Kutta Methods

2. Multi-step methods

- Leapfrog (midpoint)
- Linear Multistep Methods
 - Adams Methods
 - BDF Methods

3. ..

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Categorizing a Linear ODE: absolute stability

A linear semi-discrete problem yields the ODE

$$\frac{dw}{dt} = R(w)$$

with $R(w) = Aw$ and A a matrix.

Definitions:

- ▶ **Linear system:** $F(w, t) = A(t)w + g(t)$
where $A(t) \in \mathbb{R}^{s \times s}$ and $g(t) \in \mathbb{R}^s$
 - *Constant coefficient* if A is a constant matrix
 - *Homogeneous* if $g(t) \equiv 0$

- ▶ **System of ODEs:** $\frac{dw}{dt} = F(w(t), t)$
for $t > t^0$ and $w(t^0) = w^0$

- ▶ **Solution** of the IVP: Assuming periodic BCs, solutions under the form of Fourier modes:

$$\frac{d\hat{w}}{dt} = \hat{A}\hat{w} \implies \hat{w}(t) = \hat{w}^0 e^{A(t-t^0)} = \hat{w}^0 e^{\lambda_i t}$$

- λ_i eigs of \hat{A} (possibly complex)

Consider the scalar model problem (=homogen. modal eq.)

$$\frac{dw}{dt} = \lambda w(t)$$

Exact (w^0) and **numerical** (\tilde{w}^0) solutions:

$$w(t) = w^0 e^{\lambda t}$$

$$\tilde{w}(t) = \tilde{w}^0 e^{\lambda t}$$

▶ Difference:

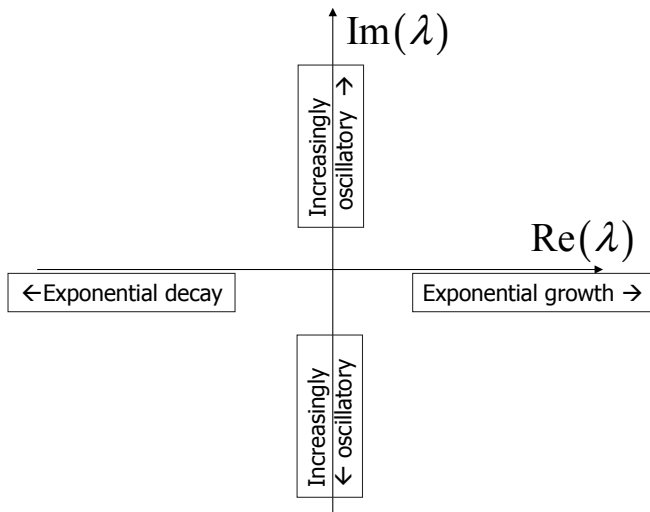
$$|w - \tilde{w}| = |(w^0 - \tilde{w}^0)e^{\lambda t}| = |(w^0 - \tilde{w}^0)|e^{\operatorname{Re}(\lambda)t}$$

- $\operatorname{Re}(\lambda) \leq 0$: difference bounded, **stable**
- $\operatorname{Re}(\lambda) < 0$: difference decays, **asymptotically stable**
- $\operatorname{Re}(\lambda) > 0$: difference unbounded, **unstable**
- For a system, each λ_i is considered

- ▶ **Stability** completely determined by the eigs of the space discretization matrix A :

- In **linear cases**: eigenvalues of the coefficient matrix
- In **nonlinear cases**: eigenvalues of the Jacobian matrix (after linearization)

Stability of the spatial discretization



$$\frac{d\hat{w}}{dt} = \lambda \hat{w} \quad \Rightarrow \quad \hat{w} = \hat{w}^0 e^{\lambda t}$$

Real part: **dissipation** (damping)
Imaginary part: **dispersion**

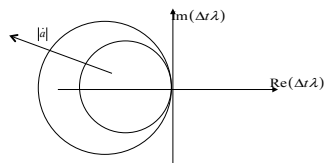
Examples: FOU and CS

- FOU scheme for scalar linear advection:

$$\frac{dw}{dt} = -a \frac{w_{j+1} - w_{j-1}}{2\Delta x} + \frac{1}{2}|a| \frac{w_{j+1} - 2w_j + w_{j-1}}{\Delta x}$$

$$\frac{d\hat{w}}{dt} = -\frac{1}{\Delta t} [i\dot{a} \sin \beta + |\dot{a}|(1 - \cos \beta)] \hat{w}$$

$$\lambda \Delta t = -[i\dot{a} \sin \beta + |\dot{a}|(1 - \cos \beta)], \quad \text{Re}(\lambda \Delta t) = -|\dot{a}|(1 - \cos \beta) \in [-2|\dot{a}|, 0]$$

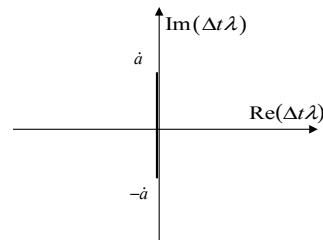


- Locus of FOU in the complex plane: circle of radius $|\dot{a}|$

- Second-order centred scheme for scalar linear advection:

$$\frac{dw_j}{dt} = -a \frac{w_{j+1} - w_{j-1}}{\Delta x} \implies \frac{d\hat{w}}{dt} = -\frac{1}{\Delta t} (i\dot{a} \sin \beta) \hat{w}$$

$$\implies \lambda \Delta t = -i\dot{a} \sin \beta, \quad \text{Re}(\lambda \Delta t) = 0$$



- Zero dissipation!
- Locus of CS in the complex plane: segment of imaginary axis
- The ODE is **marginally stable**

Stability of the time discretization

Applying one-step methods to test problem $w' = \lambda w$, one typically obtains an expression of the form:

$$w^{n+1} = P(\lambda \Delta t) w^n$$

- ▶ Iterations diverge for $|P(\lambda \Delta t)| > 1$!
- ▶ Only the product $\lambda \Delta t$ matters
- ▶ For the solution not to grow without bound, the eigenvalues z of $P(\lambda \Delta t)$ should have $|z| \leq 1 \quad \forall \lambda$

Condition for the stability of **fully discrete scheme**:

The spectrum of the **spatial operator** $A(w)$ must lie in the stability region of the **time integration method**

- ▶ When more than one value is present, consistency requires that one of the eigs should represent an approximation to the physical behaviour
 - This solution of the eigenvalue eq., called **principal solution**, is recognized by the fact that

$$\lim_{\lambda \Delta t \rightarrow 0} z(\lambda) = 1$$

Definition of stability regions:

1. **Absolute stability region** is the set

$$\{z \in \mathbb{C} \mid |P(z)| \leq 1\}$$

2. **Relative stab. region** (or **Order star**) is the set

$$\{z \in \mathbb{C} \mid |P(z)| \leq |e^z|\}$$

It compares the growth of the iteration to the growth of the exact solution $\hat{w}^0 e^{\lambda t}$

A **time discretization** is:

- ▶ **Linearly stable** if it admits an absolute stab. region
- ▶ **A-stable** if its absolute stab. region contains the entire left half-plane ($|P(z)| < 1$ for $\text{Re}(z) < 0$)
- ▶ **$A(\alpha)$ -stable**: $|P(z)| < 1$ for $\text{Re}(z) < -\tan \alpha |\text{Im}(z)|$
 - i.e., the wedge $\pi - \alpha \leq \arg(z) \leq \pi + \alpha$ is entirely contained in the stability region
 - Weaker than A-stable
 - A-stable $\equiv A(\pi/2)$ -stable method

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One-step methods (I)

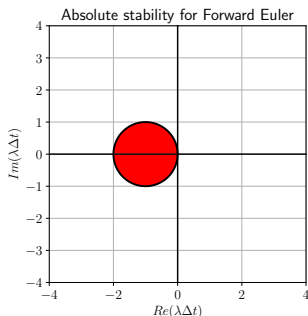
Forward Euler

$$\frac{w^{n+1} - w^n}{\Delta t} = R(w^n) = \lambda w^n$$

$$w^{n+1} = w^n + \lambda \Delta t w^n = (1 + \lambda \Delta t) w^n$$

$$w^{n+1} = P(z) w^n = (1 + z) w^n$$

Absolutely stable $\iff |1 + z| \leq 1$



- ▶ Solve for $z = x + iy$
- ▶ Absolute stability interval for Forward Euler: $[-2, 0]$
- ▶ Backward Euler: A-stable!
- ▶ Both are first-order..

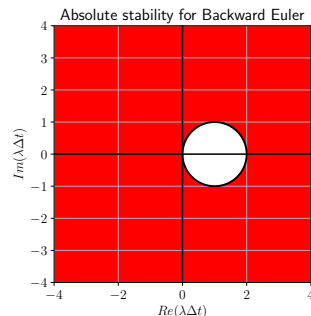
Backward Euler

$$\frac{w^{n+1} - w^n}{\Delta t} = R(w^{n+1}) = \lambda w^{n+1}$$

$$(1 - \lambda \Delta t) w^{n+1} = w^n$$

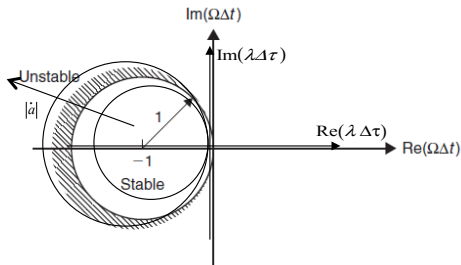
$$w^{n+1} = P(z) w^n = (1 - z)^{-1} w^n$$

Absolutely stable $\iff |1 - z|^{-1} \leq 1$

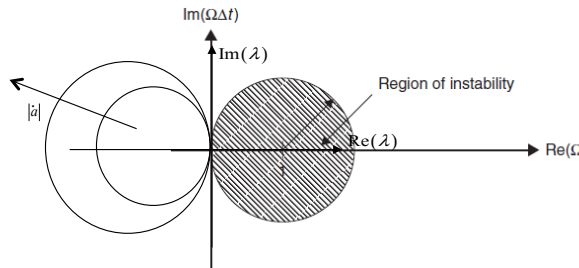


Example: FOU in space + Euler schemes

FOU + Forward Euler: stable for $|\dot{a}| \leq 1$



FOU + Backward Euler: unconditionally stable



- **centred 2nd order + FE:** unconditionally unstable
- **centred 2nd order + BE:** unconditionally stable

An A-stable time integration method leads to a **unconditionally stable** space discretization if combined with a dissipative (stable) operator

One-step methods (II)

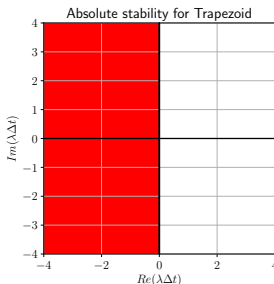
Trapezoidal (CN) Method

$$\frac{w^{n+1} - w^n}{\Delta t} = \frac{1}{2} [R(w^n) + R(w^{n+1})] = \frac{1}{2} [\lambda w^n + \lambda w^{n+1}]$$

$$\left[1 - \frac{1}{2}\lambda\Delta t\right] w^{n+1} = \left[1 + \frac{1}{2}\lambda\Delta t\right] w^n$$

$$w^{n+1} = P(z) = \frac{2+z}{2-z} w^n$$

$$\text{Absolutely stable} \iff \left| \frac{2+z}{2-z} \right| \leq 1$$



$$\text{Re}(z) > 0: \left| \frac{2+z}{2-z} \right| > 1$$

$$\text{Re}(z) \leq 0: \left| \frac{2+z}{2-z} \right| \leq 1$$

- It is the only one-step/single-stage/A-stable 2^{nd} order scheme!

Taylor-series methods

TS method of order p can be derived by keeping the first $p+1$ terms of the TS expansion and dropping the higher ones:

$$w^{n+1} \approx w^n + \Delta t w'^n + \frac{(\Delta t)^2}{2} w''^n + \frac{(\Delta t)^3}{6} w'''^n + \dots + \frac{(\Delta t)^p}{p!} w^{p,n}$$

with

$$w'^n = R(w^n)$$

$$w''^n = \frac{\partial R}{\partial t} + \frac{\partial R}{\partial w} \frac{\partial w}{\partial t} = \frac{\partial R}{\partial t} + \frac{\partial R}{\partial w} R$$

$$w'''^n = \frac{\partial^2 R}{\partial t^2} + 2 \frac{\partial^2 R}{\partial t \partial w} R + 2 \frac{\partial R}{\partial w} \frac{\partial R}{\partial t} + \frac{\partial^2 R}{\partial w^2} R^2 + 2 \left[\frac{\partial R}{\partial w} \right]^2 R$$

- Dropping terms in $(\Delta t)^2$: Forward Euler
- Becomes messy and cumbersome for higher-derivatives!

One-step methods (III)

Runge–Kutta methods

- **Multistage** method: intermediate values of the solution and its derivatives are generated and used within a single timestep

$$\begin{cases} w^{n+1} = w^n + \Delta t \sum_{i=1}^s b_i R(w^{(i)}) \\ w^{(i)} = w^n + \Delta t \sum_{j=1}^s a_{ij} R(w^{(j)}) \end{cases}$$

- Both explicit and implicit methods exist
 - For **explicit**: $a_{ij} = 0$ for $j \geq i$
 - Only **implicit** methods may give A-stability
- Increasing the **number of stages** one can:
 - Increase the **order** of the method
An r -stage ERK can have order at most r (albeit for $r \geq 5$ the order is strictly lower than r)
 - Increase the **stability** of the method
- Explicit schemes are the most used, but lots of combinations possibles

Examples

2nd-order 2-stage RK

$$\begin{cases} w^* = w^n + \frac{1}{2} \Delta t R(w^n) \\ w^{n+1} = w^n + \Delta t R(w^*) \end{cases}$$

It is a 1-step method, can be combined in

$$w^{n+1} = w^n + \Delta t R \left(w^n + \frac{\Delta t}{2} R(w^n) \right)$$

3rd-order 3-stage RK

$$\begin{aligned} w^{(1)} &= w^n + \Delta t R(w^n) \\ w^{(2)} &= \frac{3}{4} w^n + \frac{1}{4} w^{(1)} + \frac{1}{4} \Delta t R(w^{(1)}) \\ w^{n+1} &= \frac{1}{3} w^n + \frac{2}{3} w^{(2)} + \frac{2}{3} \Delta t R(w^{(2)}) \end{aligned}$$

One-step methods (IV)

Examples

2nd-order 4-stage RK (Jameson)

$$\begin{cases} w^{(0)} = w^n \\ w^{(i+1)} = w^{(0)} - \Delta t \alpha_i R(w^{(i)}) \\ w^{n+1} = w^{(4)} \end{cases}$$

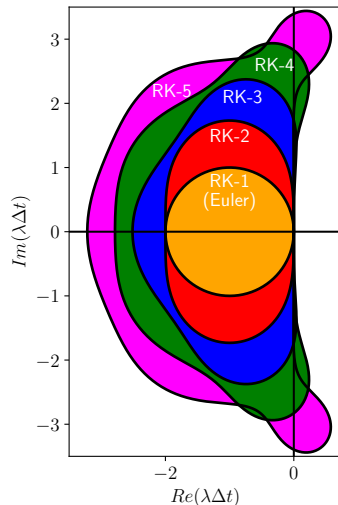
with $i = 1, 2, 3, 4$ and $\alpha = \frac{1}{5-i}$

- 4-stage RK + centred scheme: stable for $CFL \leq 2\sqrt{2}$

For order r , $P(z)$ given by first $r+1$ terms of Taylor expansion of e^z :

$$P(z) = 1 + z + \frac{z^2}{2!} + \frac{z^3}{3!} + \dots + \frac{z^r}{r!} + \dots$$

- if $s = r$, $P(z)$ does not depend on a_{ij} , b_i
 - all ERK with $s=r$ have the same absolute stability region
- $|P(z)| \rightarrow \infty$ as $|z| \rightarrow \infty$
 - Bounded stability region, not good stiff solver
(For stiff problems, Δt restricted by stability, not accuracy)



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Linear Multistep Methods

LMM: values of $R(w)$ computed in previous times are **reused** to obtain higher-order accuracy

$$\sum_{j=0}^r \alpha_j w^{n+j} = \Delta t \sum_{j=0}^r \beta_j R(w^{n+j})$$

- ▶ w^{n+r} computed from previous values w^{n+r-1} , w^{n+r-2} ..
- ▶ Method **explicit** for $\beta_r = 0$, **implicit** for $\beta_r \neq 0$

Examples: Adams methods $w^{n+r} = w^{n+r-1} + \Delta t \sum_{j=0}^r \beta_j R(w^{n+j})$, thus $\alpha_r = 1$, $\alpha_{r-1} = -1$ and $\alpha_j = 0$ for $j < r-1$

Explicit Adams-Bashforth methods (order r)

1-step: $w^{n+1} = w^n + \Delta t R(w^n)$ (forward Euler)

2-step: $w^{n+2} = w^{n+1} + \frac{\Delta t}{2} [-R(w^n) + 3R(w^{n+1})]$

3-step: $w^{n+3} = w^{n+2} + \frac{\Delta t}{12} [5R(w^n) - 16R(w^{n+1}) + 23R(w^{n+2})]$

4-step: $w^{n+4} = w^{n+3} + \frac{\Delta t}{24} [-9R(w^n) + 37R(w^{n+1}) - 59R(w^{n+2}) + 55R(w^{n+3})]$

Implicit Adams-Moulton methods (order $r+1$)

1-step: $w^{n+1} = w^n + \frac{\Delta t}{2} [R(w^n) + R(w^{n+1})]$ (trapezoidal method)

2-step: $w^{n+2} = w^{n+1} + \frac{\Delta t}{12} [-R(w^n) + 8R(w^{n+1}) + 5R(w^{n+2})]$

3-step: $w^{n+3} = w^{n+2} + \frac{\Delta t}{24} [R(w^n) - 5R(w^{n+1}) + 19R(w^{n+2}) + 9R(w^{n+3})]$

4-step: $w^{n+4} = w^{n+3} + \frac{\Delta t}{720} [-19R(w^n) + 106R(w^{n+1}) - 264R(w^{n+2}) + 646R(w^{n+3}) + 251R(w^{n+4})]$

Stability for LMM

Applying LMM to $w' = \lambda w$, one has

$$\sum_{j=0}^r \alpha_j w^{n+j} = \Delta t \sum_{j=0}^r \beta_j \lambda w^{n+j}$$

$$\sum_{j=0}^r (\alpha_j - z \beta_j) w^{n+j} = 0$$

General form of the **solution**:

$$w^n = c_1 \xi_1^n + c_2 \xi_2^n + \dots + c_r \xi_r^n$$

with ξ_j roots of the **stability polynomial**:

$$\pi(\xi; z) = \sum_{j=0}^r (\alpha_j - z \beta_j) \xi^j = \rho(\xi) - z \sigma(\xi)$$

with the characteristics polynomial

$$\rho(\xi) = \sum_{j=0}^r \alpha_j \xi^j \quad \text{and} \quad \sigma(\xi) = \sum_{j=0}^r \beta_j \xi^j$$

- Polynomial in ξ , with coefficients depending on z
- The region of absolute stability is the set of z for which $\pi(\xi; z)$ satisfies the root condition

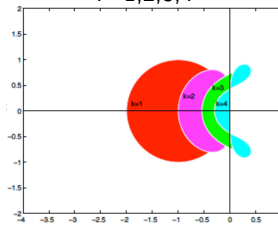
► $\pi(\xi; z)$ satisfies the **root condition** \iff

- $|\xi_j| \leq 1$ for $j = 1, 2, \dots, r$
- $|\xi_j| < 1$ if ξ_j is a repeated root

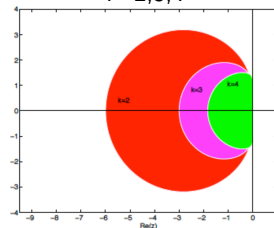
► For $\Delta t \rightarrow 0$: $\sum_{j=0}^r \alpha_j w^{n+j} = 0$

- The formula reduces to the linear recursion with characteristic polynomial $\rho(\xi)$

Adams-Bashforth for
 $r=1,2,3,4$



Adams-Moulton for
 $r=2,3,4$



Stability polynomials for different methods

Forward Euler: $\pi(\xi; z) = \xi - (1 + z)$

Backward Euler: $\pi(\xi; z) = (1 - z)\xi - 1$

Trapezoidal: $\pi(\xi; z) = \left[1 - \frac{z}{2}\right] \xi - \left[1 + \frac{z}{2}\right]$

Midpoint: $\pi(\xi; z) = \xi^2 - 2z\xi - 1$

$$\xi_1 = 1 + z$$

$$\xi_1 = (1 - z)^{-1}$$

$$\xi_1 = \frac{2 + z}{2 - z}$$

$$\xi_{1,2} = z \pm \sqrt{z^2 + 1}$$

For midpoint, root condition never satisfied!

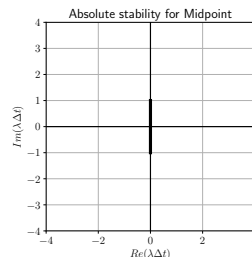
► if $z = \pm i$: $\xi_1 = \xi_2$ repeated root of modulus 1 (Stability region: interval $]-i, i[$)

► A 1-step method is **L-stable** iff it is A-stable and $\lim_{z \rightarrow \infty} |P(z)| = 0$

- Difference in the Riemann sphere, on the right half-plane
- BE is L-stable, Trapezoidal is not! Important in case of rapid transients, that we are not interested in

r-step AB and AM: $\pi(\xi; z)$ of degree $r \implies r$ roots, more complex to find stability region!

- **Dahlquist's second barrier theorem:** any A-stable LMM is at most second-order accurate
- For many stiff problems, eigenvalues are far out in the left half-plane but near the real axis
 - no reason to require A-stability, but $A(\alpha)$ stability is sufficient
 - \implies BDF Methods!



BDF Methods

Backward Differentiation Formula methods: Methods having $\sigma(\xi) = \beta_r \xi^r$:

$$\alpha_0 w^n + \alpha_1 w^{n+1} + \dots + \alpha_r w^{n+r} = \Delta t \beta_r R(w^{n+r}) \quad \text{with} \quad \beta_0 = \beta_1 = \dots \beta_{r-1} = 0$$

$$\text{1-step :} \quad w^{n+1} - w^n = \Delta t R(w^{n+1})$$

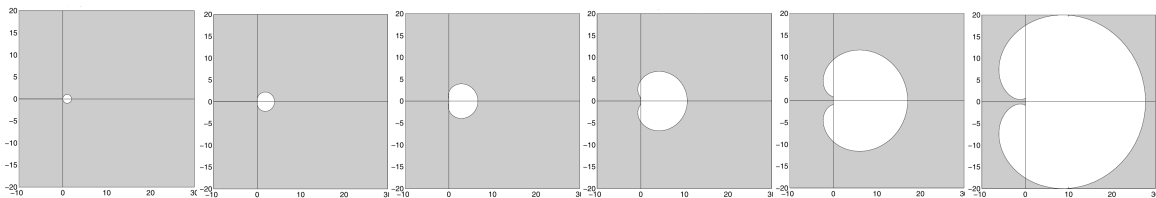
$$\text{2-step :} \quad 3w^{n+2} - 4w^{n+1} + w^n = 2 \Delta t R(w^{n+2})$$

$$\text{3-step :} \quad 11w^{n+3} - 18w^{n+2} + 9w^{n+1} - 2w^n = 6 \Delta t R(w^{n+3})$$

$$\text{4-step :} \quad 25w^{n+4} - 48w^{n+3} + 36w^{n+2} - 16w^{n+1} + 3w^n = 12 \Delta t R(w^{n+4})$$

$$\text{5-step :} \quad 137w^{n+5} - 300w^{n+4} + 300w^{n+3} - 200w^{n+2} + 75w^{n+1} - 12w^n = 60 \Delta t R(w^{n+5})$$

- ▶ r -step BDF method is r -th order accurate
- ▶ They are stable only for $r \leq 6$ (higher orders cannot be used!)
- ▶ A-stable for $r = 1, 2$, $A(\alpha)$ stable for $r > 2$: $\alpha = 90^\circ, 90^\circ, 88^\circ, 73^\circ, 51^\circ, 18^\circ$ for $r = 1, 2, 3, 4, 5, 6$
- ▶ BDF methods sacrifice A-stability for stiff decay (*i.e.*, L-stability)



1 Introduction

2 Stability

3 One-step Methods

4 Multistep Methods

5 Conclusions

Time integration schemes for unsteady flows

One-step methods:

- ✓ Self-starting methods
- ✓ Δt can be changed at any point (more care is needed for LMM)

Practical choice of step size: Δt must be small enough

1. for the LTE to be acceptably small: $\Delta t \leq \Delta t_{\text{acc}}$, where Δt_{acc} depends on several things:
 - What method is being used
 - How smooth the solution is
 - What accuracy is required
2. for the method to be absolutely stable on this particular problem: $\Delta t \leq (\Delta t)_{\text{stab}}$, depending on λ_i 's

Explicit or implicit schemes?

- For problems dominated by **small time scales** or **propagation of discontinuities**, **explicit TVD**
- For problems where dominant time scales are **much larger than the acoustic scale**, **implicit schemes**
- A way for removing stability constraint on the time step is using implicit schemes
 - Generally lead to the solution, at each physical time step, of nonlinear systems of the form:

$$R^*(w) = 0, \quad \text{with} \quad R^*(w) = T(w) + R(w)$$

with $T(w)$ the time discretization operator

- Direct methods cannot be applied without **linearization**, **Jacobian approximation**, **factorization**, ...
⇒ **significant errors!** Alternative: use **iterative methods**, such as **Dual Time Stepping** and **Newton**

Implicit schemes for unsteady flows

Dual Time Stepping

Add to the unsteady residual R^* a derivative with respect to a fictitious or **pseudo time** τ :

$$w_\tau + R^*(w) = 0$$

- ▶ Unsteady problem transformed in **steady** w.r.t. τ
 - Solve the false transient by applying any time-stepping technique
- ✓ Any error allowed in τ , **efficiency** only matters!
- ✓ Can use linearization, factorization, local time stepping, multigrid.. to speed-up convergence in pseudo time
- ✗ Important numerical errors or even instabilities if insufficient convergence
- ✗ Even if large (physical) Δt are allowed, efficiency depends on the number of sub-iterations required

Newton Methods

Solve directly $R^*(w) = 0$ with

$$\frac{\partial R^*}{\partial w}(w^{n+1,m+1} - w^{n+1,m}) = -R^*(w^{n+1,m})$$

- ▶ An approximate (or frozen) Jacobian can be used
- ▶ Factorization and multigrid also allow to speed-up convergence
- ✓ Typically much faster than dual time stepping
- ✗ Similar problems if insufficient convergence

Time advancement

It is chosen following stability and accuracy considerations (unsteady flows)

$$\text{viscous criterion: } \Delta t_v = \sigma \frac{\Delta x^2}{\nu}$$

$$\text{convective criterion: } \Delta t_c = \text{CFL} \frac{\Delta x}{u}$$

- **Incompressible regime:** viscous criterion is often very restrictive (except without solid boundaries)

We can use:

- an **implicit** integration for viscous terms (second-order Crank-Nicolson scheme for ex.)
- an **explicit** integration for convective terms (3rd- or 4th-order Runge Kutta algorithms or linear multistep methods such as 2nd- or 3rd-order Adams-Bashforth).

- **Compressible regime:** the convective CFL criterion becomes limiting since it implies sound speed

$\Delta t_c = \text{CFL} \frac{\Delta x}{u+c}$. An explicit method is generally preferred when strong unsteadiness is present.

- Not always the case, depends on the physics (e.g., chemistry, low-frequency phenomena, ..)