

Implicit time integration

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- Explicit schemes: CFL restricted time step: Courant number $C < 1$ in

$$\Delta t \leq C \min_i \left(\frac{\Delta x_i}{c_i^{\max}} \right)$$

\Rightarrow MHD: $c_i^{\max} = |v_i| + c_i^{\text{fast}}$ fast magnetosonic speed c_i^{fast}

- can be inefficient when

\Rightarrow only interested in steady state $\partial t = 0$ solution

\Rightarrow no fast waves induced (or physically unimportant): still set CFL

\Rightarrow solution: switch to implicit time integration, allowing $C > 1$!

- diffusive type source term: further limit on Δt from

$$\Delta t \leq C_{\text{diff}} \min_i \left[\frac{(\Delta x_i)^2}{\max(\eta, \nu, \kappa)} \right]$$

\Rightarrow for increasing resolution: more restrictive than CFL

\Rightarrow need to treat source term(s) implicitly

Implicit schemes

- after semi-discretization (Finite Volume spatial discretization) solve

$$\frac{d\mathbf{U}}{dt} = \mathbf{R}(\mathbf{U})$$

⇒ two-parameter (α, β) base scheme

$$\begin{aligned} \mathbf{U}^{n+1} = \mathbf{U}^n + \Delta t_n \mathbf{R}(\mathbf{U}^n) + \alpha \Delta t_n \left[\frac{\mathbf{U}^n - \mathbf{U}^{n-1}}{\Delta t_{n-1}} - \mathbf{R}(\mathbf{U}^n) \right] \\ + \beta \Delta t_n [\mathbf{R}_{\text{impl}}(\mathbf{U}^{n+1}) - \mathbf{R}_{\text{impl}}(\mathbf{U}^n)] \end{aligned}$$

⇒ uses two time levels n and $n + 1$ for $\alpha = 0$

⇒ three-level scheme for $\alpha \neq 0$

⇒ all implicitly treated terms in \mathbf{R}_{impl}



- Steady state simulations: fully implicit Backward Euler

⇒ two-level scheme $\alpha = 0, \beta = 1$

⇒ fully implicit: $\mathbf{R}_{\text{impl}} = \mathbf{R}$

$$\mathbf{U}^{n+1} = \mathbf{U}^n + \Delta t_n \mathbf{R}(\mathbf{U}^{n+1})$$

- nonlinear residual $\mathbf{R}(\mathbf{U}^{n+1})$ at unknown time level $n + 1$

⇒ linearize $\mathbf{R}(\mathbf{U}^{n+1}) = \mathbf{R}(\mathbf{U}^n) + \frac{\partial \mathbf{R}}{\partial \mathbf{U}}|_n (\mathbf{U}^{n+1} - \mathbf{U}^n)$

- linearized fully implicit Backward Euler scheme: solve

$$\left[\frac{I}{\Delta t} - \frac{\partial \mathbf{R}}{\partial \mathbf{U}} \right] [\mathbf{U}^{n+1} - \mathbf{U}^n] = \mathbf{R}(\mathbf{U}^n)$$

⇒ large linear system

⇒ Newton's method: $x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$ for $f(x) = 0$ for $\Delta t \rightarrow \infty$

⇒ Jacobian $\partial \mathbf{R} / \partial \mathbf{U}$: numerically evaluated



- two essential ingredients: large linear systems and Jacobian evaluation
- SS: employ nearest neighbour 1st order discretization in $\partial \mathbf{R} / \partial \mathbf{U}$
 - \Rightarrow only interest in high order solution satisfying $\mathbf{R}(\mathbf{U}) = 0$
 - \Rightarrow 1D: block tri-diagonal system; 2D block penta; 3D: block hepta
- 1D block tri-diagonal systems: employ direct solver
- > 1 D problem: storage and accuracy problem for direct linear solvers
 - \Rightarrow switch to iterative linear system solvers, e.g. GMRES

Iterative methods: GMRES (Saad & Schultz 1986)

- Generalized Minimal RESidual for system $A\mathbf{x} = \mathbf{b}$
- guess \mathbf{x}_o and residual $\mathbf{r}_o = \mathbf{b} - A\mathbf{x}_o$
- define Krylov subspace $K^k(A, \mathbf{x}_o) = \{\mathbf{r}_o, A\mathbf{r}_o, \dots, A^{k-1}\mathbf{r}_o\}$
- successive approximations \mathbf{x}_k
 - \Rightarrow minimize $\|\mathbf{b} - A\mathbf{x}_k\|_2$ over $K^k(A, \mathbf{x}_o)$



- iterative linear system solvers for $A\mathbf{x} = \mathbf{b}$
 - ⇒ kernels: matrix-vector products and dot products
 - ⇒ condition number of A = ratio largest/smallest eigenvalue
 - ⇒ fast convergence if condition number close to unity
- for strongly non-symmetric matrix (advection dominated problems)
 - ⇒ needed: pre-conditioning!
 - ⇒ apply iterative solver to 'pre-conditioned' linear system
 - ⇒ pre-multiply system with approximate to inverse of A
 - ⇒ can dramatically improve convergence behavior
 - ⇒ VAC: Modified Block Incomplete LU-preconditioner
- Storage issue (as k increases): Restarted GMRES
 - ⇒ GMRESR (van der Vorst & Vuik 1994)
- Best experience with: Bi-Conjugate Gradient Stabilized
 - ⇒ Bi-CGSTAB (van der Vorst 1992)



Jacobian evaluation

- numerical evaluation: matrix-free (Newton-Krylov) (can be high order!)

⇒ only need directional derivative

⇒ use explicit methods to evaluate $\mathbf{R}(\mathbf{U})$ (CFL limited Δt)

$$\frac{\partial \mathbf{R}}{\partial \mathbf{U}}[\Delta \mathbf{U}] = \frac{\mathbf{R}(\mathbf{U} + \epsilon \Delta \mathbf{U}) - \mathbf{R}(\mathbf{U})}{\epsilon}$$

⇒ action of Jacobian in direction $\Delta \mathbf{U}$ with numerical differentiation

⇒ choice of ϵ crucial

◇ small enough to be physical/accurate

◇ avoid rounding errors (large w.r.t. machine precision)

- PROBLEM: this effectively perturbs the matrix

⇒ errors at each matrix-vector product (kernel for iterative scheme)

⇒ destroys orthogonality properties exploited in iterative schemes!!!



- numerical evaluation: Build the matrix

$$\frac{\partial \mathbf{R}_i^{(u)}}{\partial \mathbf{U}_j^{(w)}} = \frac{\mathbf{R}_i^{(u)}(\mathbf{U} + \epsilon_j^{(w)}) - \mathbf{R}_i^{(u)}(\mathbf{U})}{\epsilon}$$

⇒ still use explicit branch to evaluate $\mathbf{R}(\mathbf{U})$

⇒ low order $\mathbf{R} \rightarrow$ known sparsity pattern (nearest neighbour)

⇒ STRUCTURED GRID: Grid masks (allows data parallelism)

| | | |
|---|---|---|
| 1 | 2 | 3 |
|---|---|---|

1D

| | | | | |
|---|---|---|---|---|
| 1 | 2 | 3 | 4 | 5 |
| 3 | 4 | 5 | 1 | 2 |
| 5 | 1 | 2 | 3 | 4 |
| 2 | 3 | 4 | 5 | 1 |
| 4 | 5 | 1 | 2 | 3 |

2D

$x \rightarrow$

| | | | | | | |
|---|---|---|---|---|---|---|
| 1 | 2 | 3 | 4 | 5 | 6 | 7 |
|---|---|---|---|---|---|---|

$y \rightarrow$

| | | | | | | |
|---|---|---|---|---|---|---|
| 1 | 3 | 5 | 7 | 2 | 4 | 6 |
|---|---|---|---|---|---|---|

$z \rightarrow$

| | | | | | | |
|---|---|---|---|---|---|---|
| 1 | 4 | 7 | 3 | 6 | 2 | 5 |
|---|---|---|---|---|---|---|

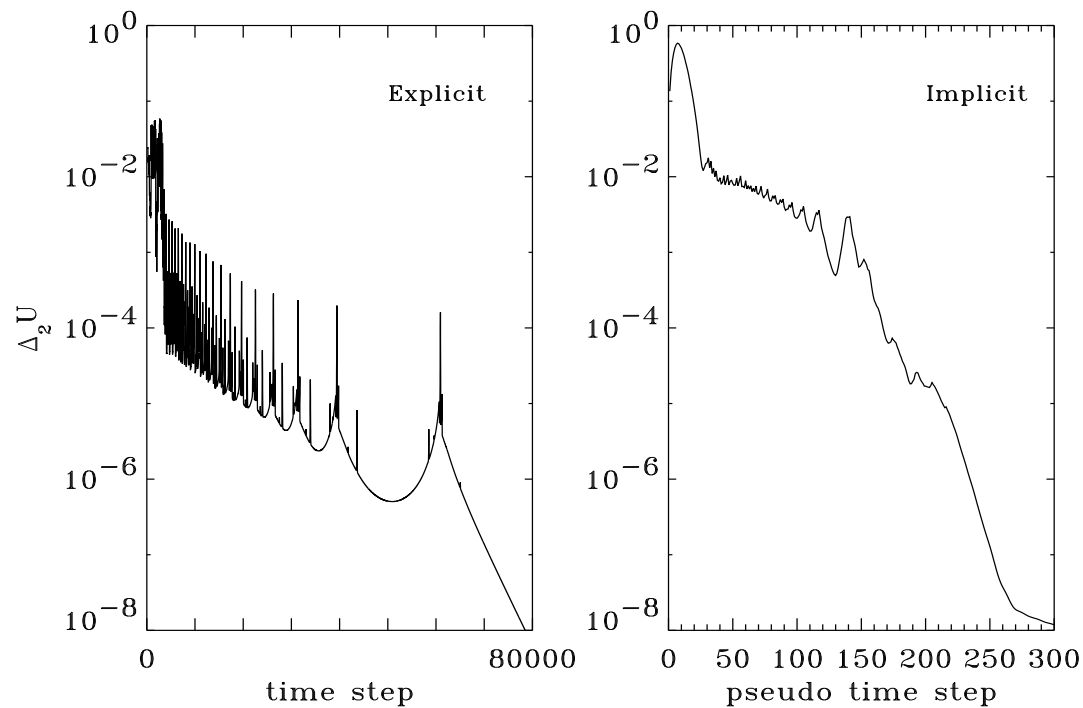
3D

⇒ divide grid into 3, 5^2 , 7^3 blocks, tiled using method stencil



Accretion onto black hole

- 1D HD problem: Steady State transonic accretion onto black hole
- Convergence behaviour Superior to explicit scheme.
- speedup by factor of 40: compare residual behavior explicit/implicit



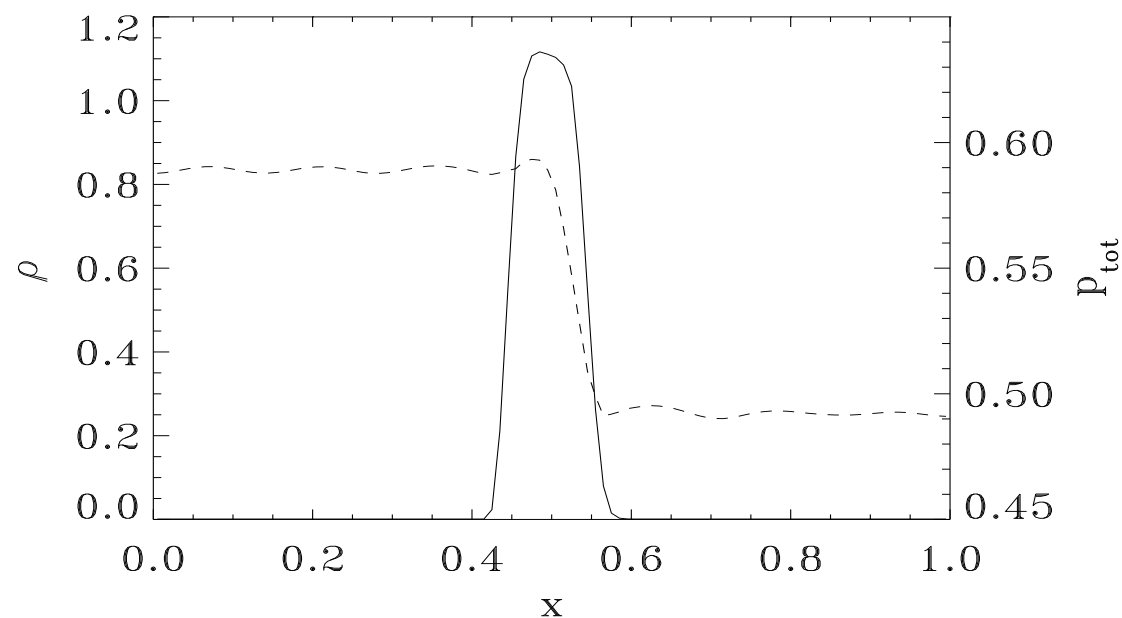
Oscillating plasma sheet

- 1D MHD problem: fully implicit ($\mathbf{R}_{\text{impl}} = \mathbf{R}$) Time Accurate calculation
 - \Rightarrow Models: Oscillating plasma sheet in magnetized Vacuum
 - \Rightarrow Challenge for explicit scheme due to 'infinite' sound speed!
- use three-level BDF2 (Backward Differentiation Formula) scheme

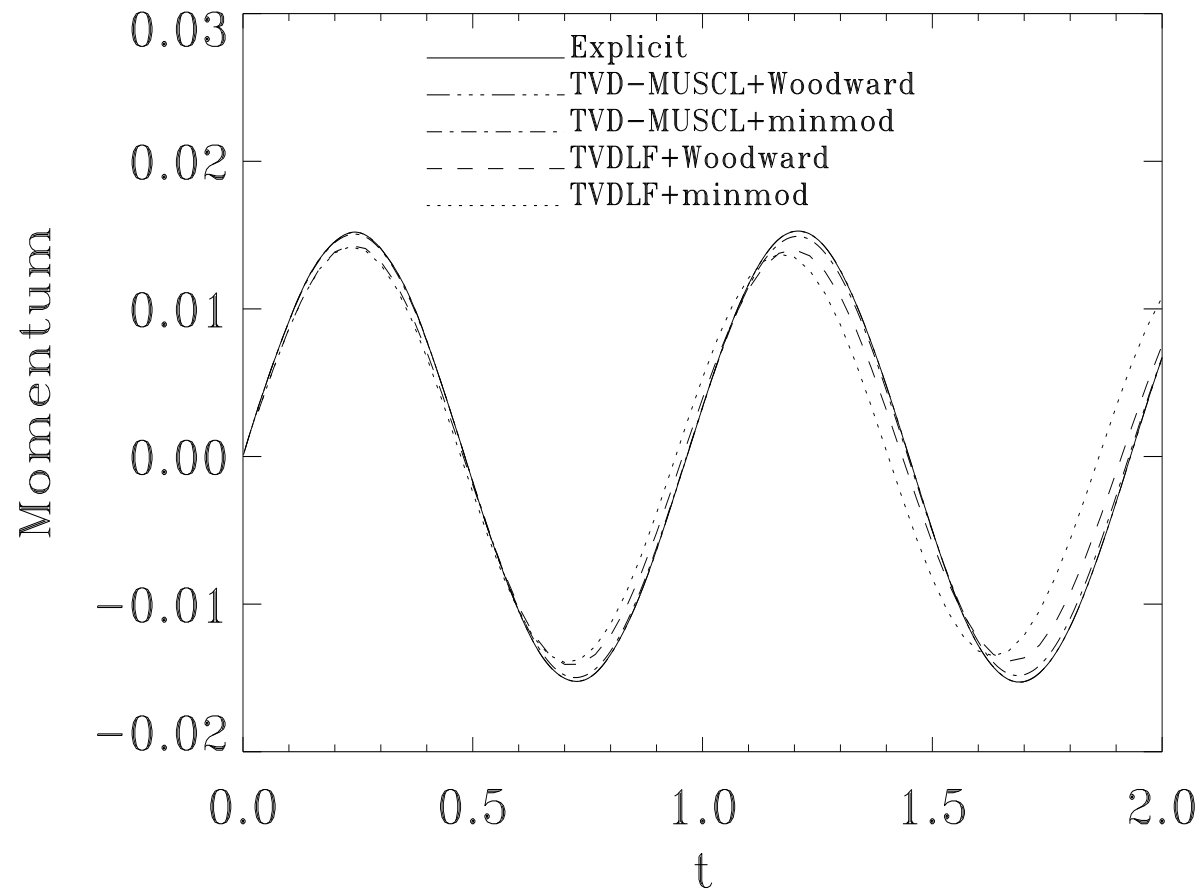
$$\alpha = \frac{\Delta t_n}{2\Delta t_n + \Delta t_{n-1}} = 1 - \beta$$

\Rightarrow first time step: Backward Euler (1st order: $\alpha = 0, \beta = 1$)

\Rightarrow first step at second order: trapezoidal method $\alpha = 0, \beta = 1/2$



- Trade-off efficiency-diffusion: more diffusive solution: faster
 - ⇒ speedup can be up to 30!
 - ⇒ total x -momentum evolution (known amplitude and frequency)



References

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