ADER and DeC: arbitrarily high order (explicit) methods for PDEs and ODEs





Davide Torlo

*MathLab, Mathematics Area, SISSA International School for Advanced Studies, Trieste, Italy davidetorlo.it

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Outline

- Motivation
- 2 DeC
- 3 ADER
- 4 Similarities
- **5** ADER stability and accuracy
- **6** Simulations
- 7 Efficient DeC (ADER)
- 8 An efficient Deferred Correction

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Motivation: high order accurate explicit method

Methods used to solve a hyperbolic PDE system for $u: \mathbb{R}^+ imes \Omega o \mathbb{R}^D$

$$\partial_t u + \nabla_{\mathbf{x}} \mathcal{F}(u) = 0. \tag{1}$$

Or ODE system for ${\it {f u}}: \mathbb{R}^+
ightarrow \mathbb{R}^{\it {f S}}$

$$\partial_t \mathbf{u} = F(\mathbf{u}). \tag{2}$$

Applications:

- Fluids/transport
- Chemical/biological processes

How?

- Arbitrarily high order accurate
- .

Motivation: high order accurate explicit method

Methods used to solve a hyperbolic PDE system for $u: \mathbb{R}^+ imes \Omega o \mathbb{R}^D$

Or ODE system for u:

Fluids/transportChemical/biologica

Applications:

10⁰ 10⁻² 10-4 10°6 - order 1 order 2 order 3 10⁻⁸ Threshold

Discretization Scale

10⁰

10⁻¹

(1)

(2)

How?

Arbitrarily high orc

.

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Applications:

- Fluids/transport
- Chemical/biological processes

How?

- Arbitrarily high order accurate
- Explicit (if nonstiff problem)

Classical time integration: Runge-Kutta

$$\boldsymbol{u}^{(1)} := \boldsymbol{u}^n, \tag{3}$$

$$\boldsymbol{\sigma}^{(k)} := \boldsymbol{u}^n + \sum_{k=1}^{N} a_{ks} F\left(t^n + c_s \Delta t, \boldsymbol{u}^{(s)}\right), \quad \text{for } k = 2, \dots, K,$$

$$\mathbf{u}^{(k)} := \mathbf{u}^n + \sum_{s=1}^K a_{ks} F\left(t^n + c_s \Delta t, \mathbf{u}^{(s)}\right), \quad \text{for } k = 2, \dots, K,$$

$$\mathbf{u}^{n+1} := \mathbf{u}^n + \sum_{s=1}^K b_s F\left(t^n + c_s \Delta t, \mathbf{u}^{(s)}\right).$$
(5)

Classical time integration: Explicit Runge-Kutta

$$\boldsymbol{u}^{(k)} := \boldsymbol{u}^n + \sum_{s=1}^{k-1} a_{ks} F\left(t^n + c_s \Delta t, \boldsymbol{u}^{(s)}\right), \quad \text{for } k = 2, \ldots, K.$$

- Easy to solve
- High orders involved:
 - o Order conditions: system of many equations
 - Stages $K \ge d$ order of accuracy (e.g. RK44, RK65)

Classical time integration: Implicit Runge-Kutta

$$oldsymbol{u}^{(k)} := oldsymbol{u}^n + \sum_{s=1}^K a_{ks} F\left(t^n + c_s \Delta t, oldsymbol{u}^{(s)}
ight), \quad ext{for } k=2,\ldots,K.$$

- More complicated to solve for nonlinear systems
- High orders easily done:
 - Take a high order quadrature rule on $[t^n, t^{n+1}]$
 - · Compute the coefficients accordingly, see Gauss-Legendre or Gauss-Lobatto polynomials
 - Order up to d = 2K

ADER and DeC

Two iterative explicit arbitrarily high order accurate methods.

- ADER¹ for hyperbolic PDE, after a first analytic more complicated approach.
- Deferred Correction (DeC): introduced for explicit ODE², extended to implicit ODE³ and to hyperbolic PDE⁴.

¹M. Dumbser, D. S. Balsara, E. F. Toro, and C.-D. Munz. A unified framework for the construction of one-step finite volume and discontinuous galerkin schemes on unstructured meshes. Journal of Computational Physics, 227(18):8209–8253, 2008.

²A. Dutt, L. Greengard, and V. Rokhlin. Spectral Deferred Correction Methods for Ordinary Differential Equations. BIT Numerical Mathematics, 40(2):241–266, 2000.

³M. L. Minion. Semi-implicit spectral deferred correction methods for ordinary differential equations. Commun. Math. Sci., 1(3):471–500, 09 2003.

⁴R. Abgrall. High order schemes for hyperbolic problems using globally continuous approximation and avoiding mass matrices. Journal of Scientific Computing, 73(2):461–494, Dec 2017.

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DeC high order time discretization: \mathcal{L}^2

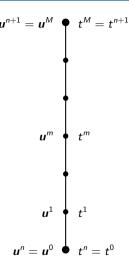
High order in time: we discretize our variable on $[t^n, t^{n+1}]$ in M substeps (u^m) .

$$\partial_t \mathbf{u} = F(\mathbf{u}(t)).$$

Thanks to Picard-Lindelöf theorem, we can rewrite

$$\boldsymbol{u}^m = \boldsymbol{u}^0 + \int_{t^0}^{t^m} F(\boldsymbol{u}(t)) dt.$$

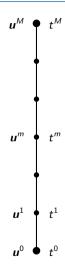
and if we want to reach order r + 1 we need M = r.



More precisely, for each σ we want to solve $\mathcal{L}^2(\boldsymbol{u}^{n,0},\ldots,\boldsymbol{u}^{n,M})=0$, where

$$\mathcal{L}^{2}(\boldsymbol{u}^{0},\ldots,\boldsymbol{u}^{M}) = \begin{pmatrix} \boldsymbol{u}^{M} - \boldsymbol{u}^{0} + \sum_{r=0}^{M} \int_{t^{0}}^{t^{M}} F(\boldsymbol{u}^{r}) \varphi_{r}(s) ds \\ \vdots \\ \boldsymbol{u}^{1} - \boldsymbol{u}^{0} + \sum_{r=0}^{M} \int_{t^{0}}^{t^{1}} F(\boldsymbol{u}^{r}) \varphi_{r}(s) ds \end{pmatrix}$$

- $\mathcal{L}^2 = 0$ is a system of $M \times S$ coupled (non)linear equations
- \mathcal{L}^2 is an implicit method (collocation method: Gauss, LobattoIIIA)
- Not easy to solve directly $\mathcal{L}^2(\underline{\boldsymbol{u}}^*)=0$
- High order ($\geq M+1$), depending on points distribution



DeC high order time discretization: \mathcal{L}^2

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$$\mathcal{L}^2(oldsymbol{u}^0,\ldots,oldsymbol{u}^M) = egin{pmatrix} oldsymbol{u}^M - oldsymbol{u}^0 + \Delta t \sum_{r=0}^M heta_r^M F(oldsymbol{u}^r) \ dots \ oldsymbol{u}^1 - oldsymbol{u}^0 + \Delta t \sum_{r=0}^M heta_r^1 F(oldsymbol{u}^r) \end{pmatrix}$$

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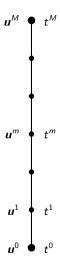


DeC low order time discretization: \mathcal{L}^1

Instead of solving the implicit system directly (difficult), we introduce a first order scheme $\mathcal{L}^1(\boldsymbol{u}^{n,0},\ldots,\boldsymbol{u}^{n,M})$:

$$\mathcal{L}^1(oldsymbol{u}^0,\ldots,oldsymbol{u}^M) = egin{pmatrix} oldsymbol{u}^M - oldsymbol{u}^0 + \Delta t eta^M F(oldsymbol{u}^0) \ dots \ oldsymbol{u}^1 - oldsymbol{u}^0 + \Delta t eta^1 F(oldsymbol{u}^0) \end{pmatrix}$$

- First order approximation
- Explicit Euler
- Easy to solve $\mathcal{L}^1(\underline{\boldsymbol{u}})=0$



How to combine two methods keeping the accuracy of the second and the stability and simplicity of the first one?

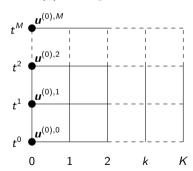
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Theorem (Convergence DeC)

- $\mathcal{L}^2(\underline{\boldsymbol{u}}^*)=0$
- If \mathcal{L}^1 coercive with constant C_1
- If $\mathcal{L}^1 \mathcal{L}^2$ Lipschitz with constant $C_2 \Delta t$

Then
$$\|\underline{\boldsymbol{u}}^{(K)} - \underline{\boldsymbol{u}}^*\| \leq C(\Delta t)^K$$

- $\mathcal{L}^1(\underline{\boldsymbol{u}}) = 0$, first order accuracy, easily invertible.
- $\mathcal{L}^2(\underline{\boldsymbol{u}}) = 0$, high order M+1.



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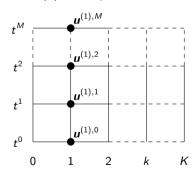
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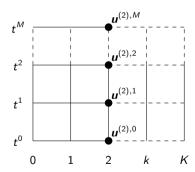
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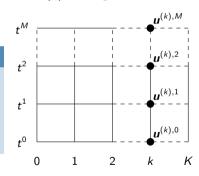
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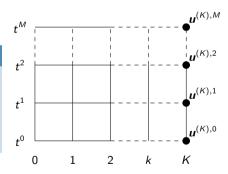
$$\begin{split} & \boldsymbol{u}^{0,(k)} := \boldsymbol{u}(t^n), \quad k = 0, \dots, K, \\ & \boldsymbol{u}^{m,(0)} := \boldsymbol{u}(t^n), \quad m = 1, \dots, M \\ & \mathcal{L}^1(\underline{\boldsymbol{u}}^{(k)}) = \mathcal{L}^1(\underline{\boldsymbol{u}}^{(k-1)}) - \mathcal{L}^2(\underline{\boldsymbol{u}}^{(k-1)}) \text{ with } k = 1, \dots, K. \end{split}$$

Theorem (Convergence DeC)

- $\mathcal{L}^2(u^*) = 0$
- If C^1 coercive with constant C_1
- If $\mathcal{L}^1 \mathcal{L}^2$ Lipschitz with constant $C_2\Delta t$

Then $\|{\bf u}^{(K)} - {\bf u}^*\| < C(\Delta t)^K$

- $\mathcal{L}^1(\mathbf{u}) = 0$, first order accuracy, easily invertible.
- $\mathcal{L}^2(\mathbf{u}) = 0$, high order M+1.



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⁵A. Dutt, L. Greengard, and V. Rokhlin. BIT Numerical Mathematics, 40(2):241–266, 2000.

DeC - Proof

Proof.

Let f^* be the solution of $\mathcal{L}^2(\underline{u}^*)=0$. We know that $\mathcal{L}^1(\underline{u}^*)=\mathcal{L}^1(\underline{u}^*)-\mathcal{L}^2(\underline{u}^*)$, so that

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$$\begin{split} \mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k+1)}) - \mathcal{L}^{1}(\underline{\boldsymbol{u}}^{*}) &= \left(\mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k)}) - \mathcal{L}^{2}(\underline{\boldsymbol{u}}^{(k)})\right) - \left(\mathcal{L}^{1}(\underline{\boldsymbol{u}}^{*}) - \mathcal{L}^{2}(\underline{\boldsymbol{u}}^{*})\right) \\ & \frac{\boldsymbol{C}_{1}||\underline{\boldsymbol{u}}^{(k+1)} - \underline{\boldsymbol{u}}^{*}|| \leq ||\mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k+1)}) - \mathcal{L}^{1}(\underline{\boldsymbol{u}}^{*})|| = \\ &= ||\mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k)}) - \mathcal{L}^{2}(\underline{\boldsymbol{u}}^{(k)}) - (\mathcal{L}^{1}(\underline{\boldsymbol{u}}^{*}) - \mathcal{L}^{2}(\underline{\boldsymbol{u}}^{*}))|| \leq \\ &\leq \frac{\boldsymbol{C}_{2}\Delta||\underline{\boldsymbol{u}}^{(k)} - \underline{\boldsymbol{u}}^{*}||.} \\ & ||\underline{\boldsymbol{u}}^{(k+1)} - \underline{\boldsymbol{u}}^{*}|| \leq \left(\frac{\boldsymbol{C}_{2}}{C_{1}}\Delta\right)||\underline{\boldsymbol{u}}^{(k)} - \underline{\boldsymbol{u}}^{*}|| \leq \left(\frac{\boldsymbol{C}_{2}}{C_{1}}\Delta\right)^{k+1}||\underline{\boldsymbol{u}}^{(0)} - \underline{\boldsymbol{u}}^{*}||. \end{split}$$

After K iteration we have an error at most of $\left(\frac{C_2}{C_1}\Delta\right)^K||\underline{\pmb{u}}^{(0)}-\underline{\pmb{u}}^*||$.

In practice

$$\mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k)}) = \mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k-1)}) - \mathcal{L}^{2}(\underline{\boldsymbol{u}}^{(k-1)}), \qquad k = 1, \dots, K,$$

$$\mathbf{u}^{(k),m} - \mathbf{u}^0 - \beta^m \Delta t F(\mathbf{u}^0) - \mathbf{u}^{(k-1),m} + \mathbf{u}^0 + \beta^m \Delta t F(\mathbf{u}^0)$$
$$+ \mathbf{u}^{(k-1),m} - \mathbf{u}^0 - \Delta t \sum_{r=0}^{M} \theta_r^m F(\mathbf{u}^{(k-1),r}) = 0$$

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DeC and residual distribution

Deferred Correction + Residual distribution

- Residual distribution (FV \Rightarrow FE) \Rightarrow High order in space
- Prediction/correction/iterations ⇒ High order in time
- Subtimesteps ⇒ High order in time

$$U_{\xi}^{m,(k+1)} = U_{\xi}^{m,(k)} - |C_{\rho}|^{-1} \sum_{\mathbf{E}|\xi \in \mathbf{E}} \left(\int_{\mathbf{E}} \Phi_{\xi} \left(U^{m,(k)} - U^{n,0} \right) d\mathbf{x} + \Delta t \sum_{r=0}^{M} \theta_{r}^{m} \mathcal{R}_{\xi}^{\mathbf{E}} (U^{r,(k)}) \right),$$
with
$$\sum_{\xi \in \mathbf{E}} \mathcal{R}_{\xi}^{\mathbf{E}}(u) = \int_{\mathbf{E}} \nabla_{\mathbf{x}} F(u) d\mathbf{x}.$$

- The \mathcal{L}^2 operator contains also the complications of the spatial discretization (e.g. mass matrix)
- \mathcal{L}^1 operator further simplified up to a first order approximation (e.g. **mass lumping**)

 \mathcal{L}^1 with mass lumping

Define \mathcal{L}^1 as

$$\mathcal{L}^1(oldsymbol{u}^0,\ldots,oldsymbol{u}^M) = egin{pmatrix} oldsymbol{u}^M - oldsymbol{u}^0 - \Delta t eta^M F(oldsymbol{u}^0) \ dots \ oldsymbol{u}^1 - oldsymbol{u}^0 - \Delta t eta^1 F(oldsymbol{u}^0) \end{pmatrix}$$

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$$\mathcal{L}^{1}(\boldsymbol{u}^{0},\ldots,\boldsymbol{u}^{M}) = \begin{pmatrix} \boldsymbol{u}^{M} - \boldsymbol{u}^{0} - \Delta t \beta^{M} \left(F(\boldsymbol{u}^{0}) + \partial_{\boldsymbol{u}} F(\boldsymbol{u}^{0}) (\boldsymbol{u}^{M} - \boldsymbol{u}^{0}) \right) \\ \vdots \\ \boldsymbol{u}^{1} - \boldsymbol{u}^{0} - \Delta t \beta^{1} \left(F(\boldsymbol{u}^{0}) + \partial_{\boldsymbol{u}} F(\boldsymbol{u}^{0}) (\boldsymbol{u}^{1} - \boldsymbol{u}^{0}) \right) \end{pmatrix}$$
$$= \begin{pmatrix} \boldsymbol{u}^{M} - \boldsymbol{u}^{0} - \Delta t \beta^{M} \partial_{\boldsymbol{u}} F(\boldsymbol{u}^{0}) \boldsymbol{u}^{M} \\ \vdots \\ \boldsymbol{u}^{1} - \boldsymbol{u}^{0} - \Delta t \beta^{1} \partial_{\boldsymbol{u}} F(\boldsymbol{u}^{0}) \boldsymbol{u}^{1} \end{pmatrix}$$

$$\mathcal{L}^{1,m}(\boldsymbol{u}^0,\ldots,\boldsymbol{u}^M) = \boldsymbol{u}^m - \boldsymbol{u}^0 - \Delta t \beta^m \partial_{\boldsymbol{u}} F(\boldsymbol{u}^0) \boldsymbol{u}^m$$
$$\mathcal{L}^{2,m}(\boldsymbol{u}^0,\ldots,\boldsymbol{u}^M) = \boldsymbol{u}^m - \boldsymbol{u}^0 - \Delta t \sum \theta_r^m F(\boldsymbol{u}^r)$$

$$\boldsymbol{u}^{(k),m} - \boldsymbol{u}^0 - \Delta t \sum_{r=0}^{M} \theta_r^m F(\boldsymbol{u}^{(k-1),r}) = 0$$

DeC as RK

We can write DeC as RK defining $\underline{\theta}_0 = \{\theta_0^m\}_{m=1}^M$, $\underline{\theta}^M = \theta_r^M$ with $r \in 1, \ldots, M$, denoting the vector $\underline{\theta}_r^{M,T} = (\theta_1^M, \ldots, \theta_M^M)$. The Butcher tableau for an arbitrarily high order DeC approach is given by:

Idea: study the RK version!

$$u' = \lambda u \qquad \Re(\lambda) < 0. \tag{7}$$

$$u_{n+1} = R(\lambda \Delta t)u_n, \qquad R(z) = 1 + zb^T (I - zA)^{-1} \mathbf{1}, \qquad z = \lambda \Delta t$$
 (8)

Goal: find $z \in \mathbb{C}$ such that |R(z)| < 1.

Recall: stability function for explicit RK methods is a polynomial, indeed the inverse of (I-zA) can be written in Taylor expansion as

$$(I-zA)^{-1} = \sum_{r=0}^{\infty} z^r A^s = I + zA + z^2 A^2 + \dots,$$
 (9)

and, since A is strictly lower triangular, it is nilpotent. Hence, R(z) is a polynomial in z with degree at most equal to S.

Theorem

If the RK method is of order P, then

$$R(z) = 1 + z + \frac{z^2}{2!} + \dots + \frac{z^P}{P!} + O(z^{P+1}). \tag{10}$$

The first P+1 terms of the stability functions $R(\cdot)$ for explicit DeCs of order P are known.

Theorem

The stability function of any explicit DeC of order P (with P iterations) is

$$R(z) = \sum_{r=0}^{P} \frac{z^r}{r!} = 1 + z + \frac{z^2}{2!} + \dots + \frac{z^P}{P!}$$
 (11)

and does not depend on the distribution of the subtimenodes.

Proof (1/3)

$$A = \begin{pmatrix} 0 & 0 & 0 & \dots & 0 & 0 \\ \star & 0 & 0 & \dots & 0 & 0 \\ \star & \star & 0 & \dots & 0 & 0 \\ \star & 0 & \star & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \star & 0 & 0 & \dots & \star & 0 \end{pmatrix},$$

Block structure of the matrix A

 \star are some non-zero block matrices and the 0 are some zero block matrices.

The number of blocks in each line and row of these matrices is P, the order of the scheme.

Proof (2/3)

By induction, A^k has zeros in the upper triangular part, in the main block diagonal, and in all the k-1 block diagonals below the main diagonal, i.e.,

$$(A^k)_{i,j} = 0$$
 , if $i < j + k$,

where the indexes here refer to the blocks. Indeed, it is true that $A_{i,j} = 0$ if i < j + 1. Now, let us consider the entry $(A^{k+1})_{i,j}$ with i < j + k + 1, i.e., i - k < j + 1. It is defined as

$$(A^{k+1})_{i,j} = \sum_{w} (A^k)_{i,w} A_{w,j}.$$
 (12)

Now, we can prove that all the terms of the sum are 0. Let w < j + 1, then $A_{w,j} = 0$ because of the structure of A; while, if $w \ge j + 1 > i - k$, we have that i < w + k, so $(A^k)_{i,w} = 0$ by induction.

Proof (3/3)

In particular, this means that $A^P = \underline{0}$, because i is always smaller than j + P as P is the number of the block matrices that we have. Hence,

$$(I-zA)^{-1} = \sum_{r=0}^{\infty} z^r A^s = \sum_{r=0}^{P-1} z^r A^s = I + zA + z^2 A^2 + \dots + z^{P-1} A^{P-1}.$$
 (13)

Plugging this result into $R(z) = 1 + zb^T(I - zA)^{-1}\mathbf{1}$, the stability function R(z) is a polynomial of degree P, the order of the scheme. All terms of order lower or equal to P must agree with the expansion of the exponential function, so it must be

$$R(z) = \sum_{r=0}^{P} \frac{z^r}{r!} = 1 + z + \frac{z^2}{2!} + \dots + \frac{z^P}{P!}.$$
 (14)

Note: no assumption on the distribution of the subtimenodes.

CODE

- Choice of iterations (P) and order
- Choice of point distributions t⁰,...,t^M
- Computation of θ
- Loop for timesteps
- Loop for correction
- Loop for subtimesteps

Outline

- Motivation
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ADER

- Cauchy–Kovalevskava theorem
- Modern automatic version.
- Space/time DG
- Prediction/Correction
- Fixed-point iteration process Prediction: iterative procedure

Modern approach is DG in space time for hyperbolic problem

$$\partial_t u(x,t) + \nabla \cdot F(u(x,t)) = 0, x \in \Omega \subset \mathbb{R}^d, t > 0.$$
 (15)

Correction step: communication between cells

$$\int_{V_i} \Phi_r \left(u(t^{n+1}) - u(t^n) \right) dx + \int_{\mathcal{T}^n \times \partial V_i} \Phi_r(x) \mathcal{G}(z^-, z^+) \cdot \mathbf{n} dS dt - \int_{\mathcal{T}^n \times V_i} \nabla_x \Phi_r \cdot F(z) dx dt = 0,$$

ADER: space-time discretization

Defining $\theta_{rs}(x,t) = \Phi_r(x)\phi_s(t)$ basis functions in space and time

$$\int_{T^n \times V_i} \theta_{rs}(x,t) \partial_t \theta_{pq}(x,t) u^{pq} dx dt + \int_{T^n \times V_i} \theta_{rs}(x,t) \nabla \cdot F(\theta_{pq}(x,t) u^{pq}) dx dt = 0.$$
 (16)

ADER: space-time discretization

Defining $\theta_{rs}(x,t) = \Phi_r(x)\phi_s(t)$ basis functions in space and time

$$\int_{T^n \times V_i} \theta_{rs}(x,t) \partial_t \theta_{pq}(x,t) u^{pq} dx dt + \int_{T^n \times V_i} \theta_{rs}(x,t) \nabla \cdot F(\theta_{pq}(x,t) u^{pq}) dx dt = 0.$$
 (16)

This leads to

$$\underline{\underline{\underline{M}}}_{rspq} u^{pq} = \underline{\underline{\underline{r}}}(\underline{\underline{\underline{u}}})_{rs}, \tag{17}$$

solved with fixed point iteration method.

+ Correction step where cells communication is allowed (derived from (16)).

ADER: time integration method

Simplify! Take
$$u(t) = \sum_{m=0}^{M} \phi_m(t) u^m = \underline{\phi}(t)^T \underline{u}$$

$$\begin{split} &\int_{\mathcal{T}^n} \psi(t) \partial_t \boldsymbol{u}(t) dt - \int_{\mathcal{T}^n} \psi(t) F(\boldsymbol{u}(t)) dt = 0, \quad \forall \psi: T^n = [t^n, t^{n+1}] \to \mathbb{R}. \\ \mathcal{L}^2(\underline{\boldsymbol{u}}) &:= \int_{\mathcal{T}^n} \underline{\phi}(t) \partial_t \underline{\phi}(t)^T \underline{\boldsymbol{u}} dt - \int_{\mathcal{T}^n} \underline{\phi}(t) F(\underline{\phi}(t)^T \underline{\boldsymbol{u}}) dt = 0 \\ &\underline{\phi}(t) = (\phi_0(t), \dots, \phi_M(t))^T \end{split}$$

Quadrature. . .

$$\mathcal{L}^{2}(\underline{\boldsymbol{u}}) := \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}} - \underline{r}(\underline{\boldsymbol{u}}) = 0 \Longleftrightarrow \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}} = \underline{r}(\underline{\boldsymbol{u}}). \tag{18}$$

Nonlinear system of $M \times S$ equations

What goes into the mass matrix? Use of the integration by parts

$$\mathcal{L}^{2}(\underline{\boldsymbol{u}}) := \int_{\mathcal{T}^{n}} \underline{\phi}(t) \partial_{t} \underline{\phi}(t)^{T} \underline{\boldsymbol{u}} dt + \int_{\mathcal{T}^{n}} \underline{\phi}(t) F(\underline{\phi}(t)^{T} \underline{\boldsymbol{u}}) dt =$$

$$\underline{\phi}(t^{n+1}) \underline{\phi}(t^{n+1})^{T} \underline{\boldsymbol{u}} - \underline{\phi}(t^{n}) \boldsymbol{u}^{n} - \int_{\mathcal{T}^{n}} \partial_{t} \underline{\phi}(t) \underline{\phi}(t)^{T} \underline{\boldsymbol{u}} - \int_{\mathcal{T}^{n}} \underline{\phi}(t) F(\underline{\phi}(t)^{T} \underline{\boldsymbol{u}}) dt$$

$$\underline{\underline{M}} = \underline{\phi}(t^{n+1}) \underline{\phi}(t^{n+1})^{T} - \int_{\mathcal{T}^{n}} \partial_{t} \underline{\phi}(t) \underline{\phi}(t)^{T}$$

$$\underline{r}(\underline{\boldsymbol{u}}) = \underline{\phi}(t^{n}) \boldsymbol{u}^{n} + \int_{\mathcal{T}^{n}} \underline{\phi}(t) F(\underline{\phi}(t)^{T} \underline{\boldsymbol{u}}) dt$$

$$\underline{\underline{M}} \underline{\boldsymbol{u}} = \underline{r}(\underline{\boldsymbol{u}})$$

ADER: Fixed point iteration

Iterative procedure to solve the problem for each time step

$$\underline{\underline{\boldsymbol{u}}}^{(k)} = \underline{\underline{\underline{M}}}^{-1} \underline{\underline{r}}(\underline{\underline{\boldsymbol{u}}}^{(k-1)}), \quad k = 1, \dots, \text{convergence}$$
 (19)

with $\underline{\boldsymbol{u}}^{(0)} = \boldsymbol{u}(t^n)$. Reconstruction step

$$\boldsymbol{u}(t^{n+1}) = \boldsymbol{u}(t^n) - \int_{T^n} F(\boldsymbol{u}^{(K)}(t)) dt.$$

- Convergence?
- How many steps K?
- Accuracy \mathcal{L}^2 ?

ADER 2nd order

Example with 2 Gauss Legendre points, Lagrange polynomials and 2 iterations Let us consider the timestep interval $[t^n, t^{n+1}]$, rescaled to [0, 1]. Gauss-Legendre points quadrature and interpolation (in the interval [0, 1])

$$egin{aligned} \underline{t}_q &= \left(t_q^0, t_q^1\right) = \left(t^0, t^1\right) = \left(rac{\sqrt{3} - 1}{2\sqrt{3}}, rac{\sqrt{3} + 1}{2\sqrt{3}}
ight), \quad \underline{w} = (1/2, 1/2). \ &\underline{\phi}(t) = \left(\phi_0(t), \phi_1(t)\right) = \left(rac{t - t^1}{t^0 - t^1}, rac{t - t^0}{t^1 - t^0}
ight). \end{aligned}$$

Then, the mass matrix is given by

$$\underline{\underline{\underline{M}}}_{m,l} = \phi_m(1)\phi_l(1) - \phi'_m(t')w_l, \quad m, l = 0, 1,$$

$$\underline{\underline{\underline{M}}} = \begin{pmatrix} 1 & \frac{\sqrt{3}-1}{2} \\ -\frac{\sqrt{3}+1}{2} & 1 \end{pmatrix}.$$

ADER 2nd order

The right hand side is given

$$r(\underline{\boldsymbol{u}})_m = \alpha(0)\phi_m(0) + \Delta t F(\alpha(t^m))w_m, \quad m = 0, 1.$$

$$\underline{r}(\underline{\boldsymbol{u}}) = \alpha(0)\underline{\phi}(0) + \Delta t \begin{pmatrix} F(\alpha(t^1))w_1 \\ F(\alpha(t^2))w_2. \end{pmatrix}.$$

Then, the coefficients $\underline{\boldsymbol{u}}$ are given by

$$\underline{\boldsymbol{u}}^{(k+1)} = \underline{\underline{\underline{M}}}^{-1}\underline{r}(\underline{\boldsymbol{u}}^{(k)}).$$

Finally, use $\underline{u}^{(k+1)}$ to reconstruct the solution at the time step t^{n+1} :

$$oldsymbol{u}^{n+1} = \underline{\phi}(1)^T \underline{oldsymbol{u}}^{(k+1)} = oldsymbol{u}^n + \int_{T^n} \underline{\phi}(t)^T dt \, F(\underline{oldsymbol{u}}^{(k)}).$$

CODE

- ullet Choice: ϕ Lagrangian basis functions
- Different subtimesteps: Gauss-Legendre, Gauss-Lobatto, equispaced
- $\bullet \ \, \mathsf{Precompute} \,\, \underline{M}$
- Precompute the rhs vector part using quadratures after a further approximation

$$\underline{r}(\underline{\boldsymbol{u}}) = \underline{\phi}(t^n)\boldsymbol{u}^n + \int_{T^n} \underline{\phi}(t)F(\underline{\phi}(t)^T\underline{\boldsymbol{u}})dt \approx \underline{\phi}(t^n)\boldsymbol{u}^n + \underbrace{\int_{T^n} \underline{\phi}(t)\underline{\phi}(t)^Tdt}_{Can \text{ be stored}}F(\underline{\boldsymbol{u}})$$

• Precompute the reconstruction coefficients $\phi(1)^T$

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ADER⁶ and DeC⁷: immediate similarities

- High order time-space discretization
- Start from a well known space discretization (FE/DG/FV)
- FE reconstruction in time
- System in time, with M equations
- Iterative method / K corrections

⁶M. Dumbser, D. S. Balsara, E. F. Toro, and C.-D. Munz. A unified framework for the construction of one-step finite volume and discontinuous galerkin schemes on unstructured meshes. Journal of Computational Physics, 227(18):8209–8253, 2008.

 $^{^7}$ R. Abgrall. High order schemes for hyperbolic problems using globally continuous approximation and avoiding mass matrices. Journal of Scientific Computing, 73(2):461–494, Dec 2017.

ADER⁶ and DeC⁷: immediate similarities

- High order time-space discretization
- Start from a well known space discretization (FE/DG/FV)
- FE reconstruction in time
- System in time, with M equations
- Iterative method / K corrections
- Both high order explicit time integration methods (neglecting spatial discretization)

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ADER as DeC

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$$\mathcal{L}^{2}(\underline{\boldsymbol{u}}) := \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}} - r(\underline{\boldsymbol{u}}),$$

$$\mathcal{L}^{1}(\underline{\boldsymbol{u}}) := \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}} - r(\boldsymbol{u}(t^{n})).$$

$$\mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k)}) = \mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k-1)}) - \mathcal{L}^{2}(\underline{\boldsymbol{u}}^{(k-1)}), \qquad k = 1, \dots, K,$$

$$\underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}}^{(k)} - r(\underline{\boldsymbol{u}}^{(k),0}) - \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}}^{(k-1)} + r(\underline{\boldsymbol{u}}^{(k-1),0}) + \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}}^{(k-1)} - r(\underline{\boldsymbol{u}}^{(k-1)}) = 0$$

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$$\mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k)}) = \mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(k-1)}) - \mathcal{L}^{2}(\underline{\boldsymbol{u}}^{(k-1)}), \qquad k = 1, \dots, K,$$

$$\underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}}^{(k)} - r(\underline{\boldsymbol{u}}^{(k)}) - \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}}^{(k-1)} + r(\underline{\boldsymbol{u}}^{(k-1)}) + \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}}^{(k-1)} - r(\underline{\boldsymbol{u}}^{(k-1)}) = 0$$

$$\underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}}^{(k)} - r(\underline{\boldsymbol{u}}^{(k-1)}) = 0.$$

ADER as DeC

$$\mathcal{L}^{2}(\underline{\boldsymbol{u}}) := \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}} - r(\underline{\boldsymbol{u}}),$$

$$\mathcal{L}^{1}(\underline{\boldsymbol{u}}) := \underline{\underline{\mathbf{M}}}\underline{\boldsymbol{u}} - r(\boldsymbol{u}(t^{n})).$$

Apply the DeC Convergence theorem!

- \mathcal{L}^1 is coercive because \underline{M} is always invertible
- ullet $\mathcal{L}^1-\mathcal{L}^2$ is Lipschitz with constant $C\Delta t$ because they are consistent approx of the same problem
- Hence, after K iterations we obtain a Kth order accurate approximation of u^*

$$\mathcal{L}^2(\boldsymbol{u}^0,\ldots,\boldsymbol{u}^M) := egin{cases} \boldsymbol{u}^M - \boldsymbol{u}^0 - \sum_{r=0}^M \int_{t^0}^{t^M} F(\boldsymbol{u}^r) arphi_r(s) \mathrm{d}s \ \ldots \ \boldsymbol{u}^1 - \boldsymbol{u}^0 - \sum_{r=0}^M \int_{t^0}^{t^1} F(\boldsymbol{u}^r) arphi_r(s) \mathrm{d}s \end{cases}$$

DeC as ADER

DeC as ADER

$$\mathcal{L}^2(\boldsymbol{u}^0,\ldots,\boldsymbol{u}^M) := egin{cases} \boldsymbol{u}^M - \boldsymbol{u}^0 - \sum_{r=0}^M \int_{t^0}^{t^M} F(\boldsymbol{u}^r) arphi_r(s) \mathrm{d}s \ \ldots \ \boldsymbol{u}^1 - \boldsymbol{u}^0 - \sum_{r=0}^M \int_{t^0}^{t^1} F(\boldsymbol{u}^r) arphi_r(s) \mathrm{d}s \end{cases}.$$

$$\mathcal{L}^2(oldsymbol{u}^0,\ldots,oldsymbol{u}^M) := egin{dcases} oldsymbol{u}^M - oldsymbol{u}^0 - \sum_{r=0}^M \int_{t^0}^{t^M} F(oldsymbol{u}^r) arphi_r(s) \mathrm{d}s \ \ldots \ oldsymbol{u}^1 - oldsymbol{u}^0 - \sum_{r=0}^M \int_{t^0}^{t^1} F(oldsymbol{u}^r) arphi_r(s) \mathrm{d}s \end{cases}.$$

$$\chi_{[t^0,t^m]}(t^m)\boldsymbol{u}^m - \chi_{[t^0,t^m]}(t_0)\boldsymbol{u}^0 - \int_{t^0}^{t^m} \chi_{[t^0,t^m]}(t) \sum_{r=0}^M F(\boldsymbol{u}^r)\varphi_r(t) dt = 0$$

$$\int_{t^0}^{t^M} \chi_{[t^0,t^m]}(t)\partial_t (\boldsymbol{u}(t)) dt - \int_{t^0}^{t^M} \chi_{[t^0,t^m]}(t) \sum_{r=0}^M F(\boldsymbol{u}^r)\varphi_r(t) dt = 0,$$

$$\int_{T^n} \psi_m(t)\partial_t \boldsymbol{u}(t) dt - \int_{T^n} \psi_m(t)F(\boldsymbol{u}(t)) dt = 0.$$

Runge Kutta vs DeC-ADER

Classical Runge Kutta (RK)

- One step method
- Internal stages

Explicit Runge Kutta

- + Simple to code
- Not easily generalizable to arbitrary order
- Stages > order

Implicit Runge Kutta

- + Arbitrarily high order
- Require nonlinear solvers for nonlinear systems
- May not converge

DeC - ADER

- · One step method
- Internal subtimesteps
- Can be rewritten as explicit RK (for ODE)
- + Explicit
- + Simple to code
- + Iterations = order
- + Arbitrarily high order
 - Large memory storage

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Stability

Since ADER can be written as a DeC, the stability functions are given by the same formula as for DeC and the stability regions are the following.

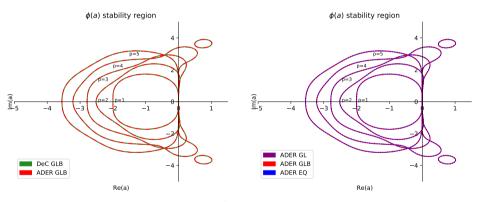


Figure: Stability region

Accuracy of ADER \mathcal{L}^2 operators

The two things that determine the accuracy of the ADER method are the iterations P and the accuracy of \mathcal{L}^2 .

Accuracy of ADER \mathcal{L}^2 for different distributions

- Equispaced: boring, minimum accuracy possible M+1 nodes p=M+1
- Guass–Lobatto: this generates the LobattoIIIC methods, M+1 nodes p=2M
- Gauss–Legendre: this does not generate Gauss methods, M+1 nodes p=2M+1

\mathcal{L}^2 ADER as RK

Here, we see \mathcal{L}^2 as an implicit RK

$$\mathcal{L}^{2,m}(\underline{\boldsymbol{u}}) = \underline{\underline{\mathbb{H}}}_{j}^{m} \boldsymbol{u}^{(j)} - \underline{\phi}^{m}(t^{n}) \boldsymbol{u}^{n} - \underbrace{\int_{T^{n}} \underline{\phi}^{m}(t) \underline{\phi}(t)_{j} dt}_{\Delta t \underline{\underline{\mathbb{H}}}_{j}^{m}} F(\boldsymbol{u}^{(j)}) = 0$$

$$\tilde{\mathcal{L}}^{2,z}(\underline{\boldsymbol{u}}) = \boldsymbol{u}^{(z)} - (\underline{\underline{\mathbb{M}}}^{-1})_{m}^{z} \underline{\phi}^{m}(t^{n}) \boldsymbol{u}^{n} - \Delta t (\underline{\underline{\mathbb{M}}}^{-1})_{m}^{z} \underline{\underline{\mathbb{H}}}_{j}^{m} F(\boldsymbol{u}^{(j)}) = 0$$

$$\boldsymbol{u}^{(z)} = \boldsymbol{u}^{n} + \Delta t a_{z,j} F(\boldsymbol{u}^{(j)})$$

- $a_{mj} = (\underline{\underline{\mathbf{M}}}^{-1})_m^z \underline{\underline{\mathbf{R}}}_j^m$
- ullet Prove that $(\underline{\underline{\mathrm{M}}}^{-1})_m^z \underline{\phi}^m(t^n) = 1$ for every z
- $c^m = \sum_r a_{mr} = t^m$
- $b_r = \frac{1}{\Delta t} \int_{T^m} \phi_r(t) dt = w_r$ quadrature weights

BCD conditions (Butcher 1964)

Define the conditions

$$B(p): \sum_{i=1}^{s} b_i c_i^{z-1} = \frac{1}{z},$$
 $z = 1, \dots, p;$ (20)

$$C(\eta): \sum_{j=1}^{s} a_{ij} c_{j}^{z-1} = \frac{c_{i}^{z}}{z}, \qquad i = 1, \dots, s, z = 1, \dots, \eta;$$
 (21)

$$D(\zeta): \qquad \sum_{i=1}^{s} b_i c_i^{z-1} a_{ij} = \frac{b_j}{z} (1 - c_j^z), \qquad \qquad j = 1, \dots, s, \ z = 1, \dots, \zeta.$$
 (22)

Theorem (Butcher 1964)

If the coefficients b_i , c_i , a_{ij} of a RK scheme satisfy B(p), $C(\eta)$ and $D(\zeta)$ with $p \leq \eta + \zeta + 1$ and $p \leq 2\eta + 2$, then the method is of order p.

$$C(s-1) D(s-1)$$

Lemma

 \mathcal{L}^2 operator of ADER defined by Gauss–Lobatto or Gauss–Legendre points and quadrature (they coincide) with s = M + 1 stages satisfies C(s - 1) and D(s - 1).

Proof (1/4).

Interpolation with ϕ^{j} is exact for polynomials of degree s-1.

The quadrature is exact for polynomials of degree 2s - 3.

Recall that $\underline{A} = \underline{\mathrm{MR}}$, Condition C(s-1) reads

$$\underline{\underline{A}} \underline{\underline{c}^{z-1}} = \frac{1}{z} \underline{\underline{c}^z} \Longleftrightarrow \underline{\underline{R}} \underline{\underline{c}^{z-1}} = \frac{1}{z} \underline{\underline{\underline{M}}} \underline{\underline{c}^z} \Longleftrightarrow \underline{\mathcal{X}} := \underline{\underline{\underline{R}}} \underline{\underline{c}^{z-1}} - \frac{1}{z} \underline{\underline{\underline{M}}} \underline{\underline{c}^z} = \underline{\underline{0}}, \qquad z = 1, \dots, s-1.$$

Recall $b_m=t^m$, $c_m=w_m$, $\underline{\underline{R}}_{i,j}=\delta_{i,j}w_i$ and the definition of $\underline{\underline{M}}$

$$\mathcal{X}_m := w_m(t^m)^{z-1} - rac{1}{z} \left(\phi^m(1) \phi^j(1) (t^j)^z - \int_0^1 rac{d}{d\xi} \phi^m(\xi) \phi^j(\xi) (t^j)^z d\xi
ight).$$

$$C(s-1) D(s-1)$$

Proof (2/4).

Now, the interpolation of t^z with $z \le s-1$ with basis functions ϕ^j is exact. Hence, we can substitute $\phi^j(\xi)(t^j)^z=\xi^z$ for all $z=1,\ldots,s-1$, obtaining

$$\mathcal{X}_m = w_m(t^m)^{z-1} - \frac{1}{z} \left(\phi^m(1) 1^z - \int_0^1 \frac{d}{d\xi} \phi^m(\xi) \xi^z d\xi \right).$$

Using the exactness of the quadrature for polynomials of degree 2s-3, both true for Gauss–Lobatto and Gauss–Legendre, we know that the previous integral is exactly computed as $\frac{d}{d\xi}\phi^m(\xi)$ is of degree at most s-2 and ξ^z is at most s-1. So, we can use integration by parts and obtain

$$\mathcal{X}_{m} = w_{m}(t^{m})^{z-1} - \frac{1}{z} \left(\phi^{m}(0)0^{z} + \int_{0}^{1} \phi^{m}(\xi) \frac{d}{d\xi} \xi^{z} d\xi \right) = w_{m}(t^{m})^{z-1} - \int_{0}^{1} \phi^{m}(\xi) \xi^{z-1} d\xi = 0$$

by the exactness of the quadrature rule and the definition of w_m . Note that the condition is sharp, since the interpolation is not anymore exact for z = s, hence C(s) is not satisfied.

$$C(s-1) D(s-1)$$

Proof (3/4).

To prove D(s-1), we write explicitly the condition in matricial form, for all $z=1,\ldots,s-1$

$$\underline{bc^{z-1}}\underline{\underline{A}} = \frac{1}{z}\underline{b(1-c^z)} \iff \underline{bc^{z-1}}\underline{\underline{\underline{M}}}^{-1}\underline{\underline{\underline{R}}} = \frac{1}{z}\underline{b(1-c^z)} \iff \underline{bc^{z-1}} = \frac{1}{z}\underline{b(1-c^z)}\underline{\underline{\underline{R}}}^{-1}\underline{\underline{\underline{M}}}.$$

Note that $b^m = w_m$ and $\underline{\underline{\mathbb{R}}}_r^m = w_m \delta_r^m$, so $\underline{b(1-c^z)}\underline{\underline{\mathbb{R}}}^{-1} = \underline{(1-c^z)}$. It is left to prove that

$$\mathcal{Y} := \underline{bc^{z-1}} - \frac{1}{z} (1 - c^z) \underline{\mathbf{M}} = \underline{\mathbf{0}}.$$

$$\mathcal{Y}_m = w_m(t^m)^{z-1} - \frac{1}{z} \sum_{i=1}^s \left(1 - (t^i)^z\right) \left(\phi^j(1)\phi^m(1) - \int_0^1 \frac{d}{d\xi} \phi^j(\xi)\phi^m(\xi)d\xi\right).$$

$$C(s-1) D(s-1)$$

Proof (4/4).

Let us observe that, since $z \le s-1$, the polynomial is exactly represented by the Lagrangian interpolation $t^z = \sum_{j=1}^s \phi(t)(t^m)^z$. Hence, using the exactness of the quadrature for polynomials of degree at most 2s-3, we have

$$\mathcal{Y}_{m} = w_{m}(t^{m})^{z-1} - \frac{1}{z}(1 - (1)^{z})\phi^{m}(1) + \frac{1}{z}\int_{0}^{1}\frac{d}{d\xi}(1 - (\xi)^{z})\phi^{m}(\xi)d\xi$$
$$= w_{m}(t^{m})^{z-1} - \frac{1}{z}\int_{0}^{1}z\xi^{z-1}\phi^{m}(\xi)d\xi = w_{m}(t^{m})^{z-1} - w_{m}(t^{m})^{z-1} = 0.$$

Hence, ADER-Legendre and ADER-Lobatto satisfy D(s-1). Note that the condition is sharp, since the interpolation is not anymore exact for z=s, hence D(s) is not satisfied.

ADER Gauss-Legendre \mathcal{L}^2

Remark (ADER-Legendre is no collocation method)

From the proof of previous Lemma, we can observe that ADER-Legendre methods do not satisfy C(s), hence, the methods are not collocation methods and they do not coincide with Gauss-Legendre implicit RK methods.

Theorem

 \mathcal{L}^2 of ADER with Gauss–Legendre is of order 2s-1.

Proof.

ADER-Legendre with s=M+1 stages satisfies B(2s) for the quadrature rule and, hence, it satisfies B(2s-1). For previous Lemma it also satisfies C(s-1) and D(s-1). Hence, Butcher's (1964) Theorem $(p \leq \eta + \zeta + 1 \text{ and } p \leq 2\eta + 2)$ guarantees that the method is of order 2s-1, since it is satisfied with p=2s-1 and $q=\zeta=s-1$.

ADER Gauss-Lobatto \mathcal{L}^2

Theorem

 \mathcal{L}^2 of ADER with Gauss-Lobatto is of order 2s-2.

Proof.

The condition for B(2s-2) is satisfied as (c,b) is the Gauss–Lobatto quadrature with order 2s-2. Previous Lemma guarantees that ADER-Lobatto satisfies B(2s-2), C(s-1) and D(s-1), so Butcher's (1964) Theorem $(p \le \eta + \zeta + 1 \text{ and } p \le 2\eta + 2)$ is satisfied for order p = 2s-2 and $\eta = \zeta = s-1$.

ADER Gauss-Lobatto \mathcal{L}^2

Theorem

 \mathcal{L}^2 of ADER with Gauss-Lobatto is LobattoIIIC.

The Lobatto IIIC method is defined using the condition

$$a_{i1} = b_1$$
 for $i = 1, \dots, s$. (23)

Lemma

 \mathcal{L}^2 of ADER with Gauss-Lobatto satisfies (23).

Theorem (Chipman 1971)

Lobatto IIIC schemes (in particular RK a_{ij}) are uniquely determined by Gauss–Lobatto quadrature rule (c,b), condition (23) and by C(s-1).

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ADER Gauss-Lobatto \mathcal{L}^2

Lemma

 \mathcal{L}^2 of ADER with Gauss-Lobatto satisfies (23).

Proof.

$$egin{aligned} egin{aligned} egin{aligned} eta_{i1} &= \sum_{j} (\underline{\underline{\mathbb{M}}}^{-1})_{ij} \mathbb{R}_{j1} = b_1 = w_1 & \Longrightarrow \ & \sum_{i,j} \underline{\underline{\mathbb{M}}}_{ki} (\underline{\underline{\mathbb{M}}}^{-1})_{ij} \mathbb{R}_{j1} = \sum_{i} \underline{\underline{\mathbb{M}}}_{ki} w_1 & \Longleftrightarrow \ & \delta_{k1} w_1 = \mathbb{R}_{k1} = \sum_{i} \underline{\underline{\mathbb{M}}}_{ki} w_1 \ & \sum_{i} \underline{\underline{\mathbb{M}}}_{ki} w_1 = \phi^m(1) w_1 - \int_0^1 rac{d}{dt} \phi^m(\xi) w_1 dt = w_1 \phi^m(0) = w_1 \delta_{m,1}. \end{aligned}$$

Outline

- 1 Motivation
- 2 DeC
- ADER
- 4 Similarities
- **5** ADER stability and accuracy
- **6** Simulations
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- An efficient Deferred Correction

Applications

Usages

- Hyperbolic PDEs as explicit iterative methods (ADER: Toro, Dumbser, Klingenberg, Boscheri; DeC: Abgrall, Ricchiuto)
- IMEX solvers for hyperbolic with stiff sources (ADER: Dumbser, Boscheri; DeC: Abgrall, Torlo)
- IMEX solvers for hyperbolic with viscosity (treated implicitly) as compressible Navier Stokes (DeC: Minion, Dumbser, Zeifang)

IMEX

$$\partial_t u = F(u) + S(u)$$

 $S(u)$ stiff to be treated implicitly

Advantages

- Arbitrary high order
- Unique framework to have matching between implicit and explicit terms
- Easy to code
- Iterative solver automatically included

Disadvantages

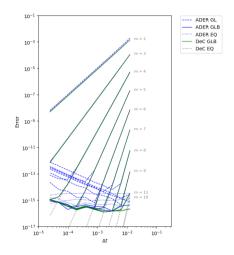
- Explicit solver: many many stages
- Implicit: many stages
- Explicit: not amazing stability property (wrt SSP RK e.g.)

Convergence

$$y'(t) = -|y(t)|y(t),$$

 $y(0) = 1,$
 $t \in [0, 0.1].$ (24)

Convergence curves for ADER and DeC, varying the approximation order and collocation of nodes for the subtimesteps for a scalar nonlinear ODE



Lotka-Volterra

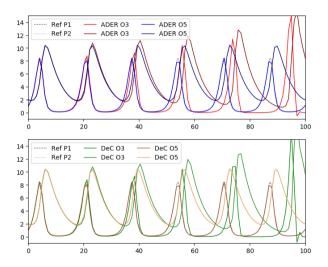
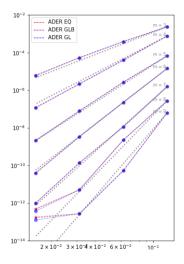


Figure: Numerical solution of the Lotka-Volterra system using ADER (top) and DeC (bottom) with Gauss-Lobatto nodes with timestep $\Delta T=1$.

PDE: Burgers with spectral difference



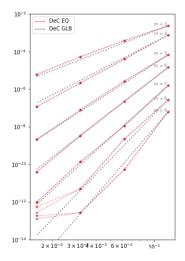


Figure: Convergence error for Burgers equations: Left

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Reduce computational cost for explicit DeC

Literature

- Micalizzi, L., Torlo, D. A new efficient explicit Deferred Correction framework: analysis and applications to hyperbolic PDEs and adaptivity. arxiv.org/abs/2210.02976
- Micalizzi, L., Torlo, D., Boscheri, W. Efficient iterative arbitrary high order methods: an adaptive bridge between low and high order. arxiv.org/abs/2212.07783

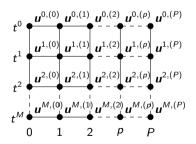
Goal

Reduce computational costs of explicit DeC.

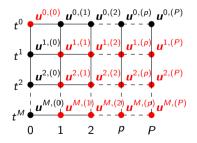
$$\mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(p)}) = \mathcal{L}^{1}(\underline{\boldsymbol{u}}^{(p-1)}) - \mathcal{L}^{2}(\underline{\boldsymbol{u}}^{(p-1)}) \text{ with } p = 1, \dots, P.$$

$$\boldsymbol{u}^{m,(p)} = \boldsymbol{u}^{0} + \sum_{r=1}^{M} \theta_{r}^{m} F(t^{r}, \boldsymbol{u}^{r,(p-1)}), \qquad \forall m = 1, \dots, M, \ p = 1, \dots, P$$

$$\mathcal{L}^1(\underline{m{u}}^{(
ho)}) = \mathcal{L}^1(\underline{m{u}}^{(
ho-1)}) - \mathcal{L}^2(\underline{m{u}}^{(
ho-1)}) \text{ with }
ho = 1, \dots, P.$$
 $m{u}^{m,(
ho)} = m{u}^0 + \sum_{r=0}^M heta_r^m F(t^r, m{u}^{r,(
ho-1)}), \qquad orall m = 1, \dots, M, \
ho = 1, \dots, P$

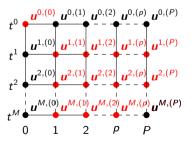


$$\mathcal{L}^1(\underline{m{u}}^{(
ho)}) = \mathcal{L}^1(\underline{m{u}}^{(
ho-1)}) - \mathcal{L}^2(\underline{m{u}}^{(
ho-1)}) \text{ with }
ho = 1, \dots, P.$$
 $m{u}^{m,(
ho)} = m{u}^0 + \sum_{r=0}^M heta_r^m F(t^r, m{u}^{r,(
ho-1)}), \qquad orall m = 1, \dots, M, \
ho = 1, \dots, P$



$$\mathcal{L}^1(\underline{\boldsymbol{u}}^{(p)}) = \mathcal{L}^1(\underline{\boldsymbol{u}}^{(p-1)}) - \mathcal{L}^2(\underline{\boldsymbol{u}}^{(p-1)}) \text{ with } p = 1, \dots, P.$$

$$\boldsymbol{u}^{m,(p)} = \boldsymbol{u}^0 + \sum_{r=0}^M \theta_r^m F(t^r, \boldsymbol{u}^{r,(p-1)}), \qquad \forall m = 1, \dots, M, \ p = 1, \dots, P.$$



<u>c</u>	u ⁰	u ⁽¹⁾	u ⁽²⁾	u ⁽³⁾		$\mathbf{u}^{(M-1)}$	$\mathbf{u}^{(M)}$	Α
0	0							u ⁰
$\beta_{1:}$	$\underline{\beta}_{1:}$	0						u ⁽¹⁾
β_{\perp}^{1}	$\Theta_{1:,0}$	$\Theta_{1:,1:}^{-}$	0					u ⁽²⁾
$\frac{\underline{\beta}_{1:}}{\underline{\beta}_{1:}}$	Θ _{1:,0}	<u>o</u>	$\Theta_{1:,1:}^{=}$	<u>o</u>				u ⁽³⁾
	:	:		٠.	٠			:
	:	:			٠	٠		:
$\underline{\beta}_{1:}$	Θ _{1:,0}	<u>0</u>			<u>0</u>	$\Theta_{1:,1:}$	<u>0</u>	u ^(M)
<u>b</u>	Өм,0	<u>0</u>				<u>0</u>	Өм,1:	$\mathbf{u}^{M,(M+1)}$

Large costs!

Large costs!

• DeC
$$S = M \cdot (P-1) + 1$$

• DeC equi $S = (P-1)^2 + 1$
• DeC GLB $S = \left\lceil \frac{P}{2} \right\rceil (P-1) + 1$

Equispaced P M DeC

2	1	2
3	1 2 3	5
4 5		10
5	4	17
6	4 5 6	26
7	6	37
8	7 8	50
9		65
10	9	82

Gauss-	ماما	++-
Gauss-	LOD_c	LLC

Gauss-Lobatto								
Ρ	M	DeC						
2	1	2						
3	2	5						
4	2 2 3	7						
5	3	13						
6	3	16						
7	4	25						
8	4	29						
9	5	41						
10	5	46						

Large costs!

• DeC
$$S=M\cdot (P-1)+1$$

• DeC equi $S=(P-1)^2+1$
• DeC GLB $S=\left\lceil\frac{P}{2}\right\rceil (P-1)+1$

Equispaced							
P	М	DeC					
2	1	2					
3	2	5					
4	2 3 4 5	10					
5	4	17					
6	5	26					
7	6	37					
8	7	50					
9	8	65					
10	9	82					

Equienced

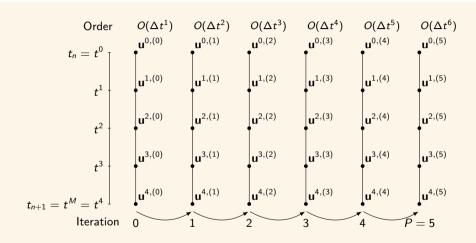
Gauss–Lobatto							
Ρ	M	DeC					
2	1	2					
3	2	5					
4	2	7					
5	3	13					
6	3	16					
7	4	25					
8	4	29					
9	5	41					
10	5	46					

How can we save computational time?

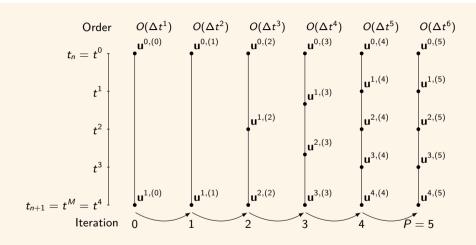
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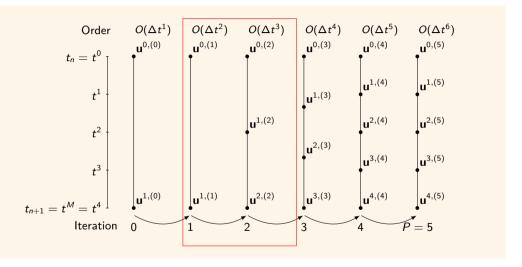
Idea for reduction of stages

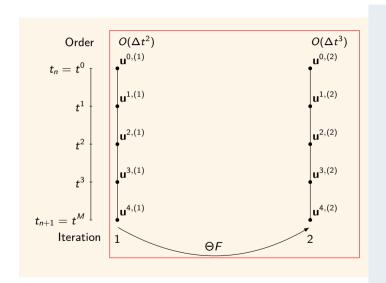


Idea for reduction of stages



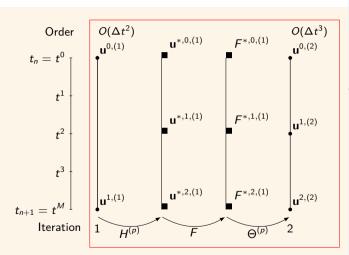
Idea for reduction of stages





DeC

$$\underline{\boldsymbol{\mathit{u}}}^{(\rho)} = \underline{\boldsymbol{\mathit{u}}}^0 + \Delta t \Theta F(\underline{\boldsymbol{\mathit{u}}}^{(\rho-1)})$$



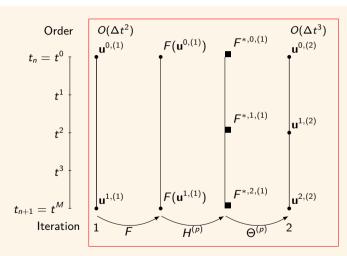
DeC

$$\underline{\boldsymbol{u}}^{(p)} = \underline{\boldsymbol{u}}^0 + \Delta t \Theta F(\underline{\boldsymbol{u}}^{(p-1)})$$

DeCu

$$\underline{{m u}}^{(
ho)} = \underline{{m u}}^0 + \Delta t \Theta^{(
ho)} {m F} ({m H}^{(
ho)} \underline{{m u}}^{(
ho-1)})$$

$$H_{ij}^{(p)} = \phi_j^{(p-1)}(t^{i,(p)})$$



DeC

$$\underline{\boldsymbol{u}}^{(\rho)} = \underline{\boldsymbol{u}}^0 + \Delta t \Theta F(\underline{\boldsymbol{u}}^{(\rho-1)})$$

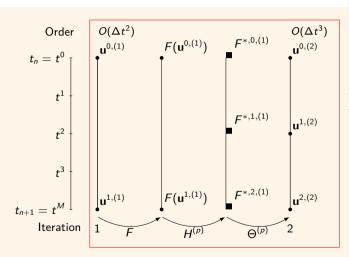
DeCu

$$\underline{\underline{\textit{u}}}^{(\rho)} = \underline{\underline{\textit{u}}}^0 + \Delta t \Theta^{(\rho)} F(H^{(\rho)} \underline{\underline{\textit{u}}}^{(\rho-1)})$$

DeCdu

$$\underline{\boldsymbol{u}}^{(p)} = \underline{\boldsymbol{u}}^0 + \Delta t \Theta^{(p)} H^{(p)} F(\underline{\boldsymbol{u}}^{(p-1)})$$

$$H_{ij}^{(p)} = \phi_j^{(p-1)}(t^{i,(p)})$$



DeC

$$\underline{\underline{u}}^{(\rho)} = \underline{\underline{u}}^0 + \Delta t \Theta F(\underline{\underline{u}}^{(\rho-1)})$$

DeCu

$$\underline{\underline{\boldsymbol{u}}}^{(p)} = \underline{\underline{\boldsymbol{u}}}^0 + \Delta t \Theta^{(p)} F(H^{(p)} \underline{\underline{\boldsymbol{u}}}^{(p-1)})$$
$$\underline{\underline{\boldsymbol{u}}}^{*(p)} = \underline{\underline{\boldsymbol{u}}}^0 + \Delta t H^{(p)} \Theta^{*(p-1)} F(\underline{\underline{\boldsymbol{u}}}^{*(p-1)})$$

DeCdu

$$\underline{\underline{\boldsymbol{u}}}^{(p)} = \underline{\underline{\boldsymbol{u}}}^0 + \Delta t \Theta^{(p)} H^{(p)} F(\underline{\underline{\boldsymbol{u}}}^{(p-1)})$$

$$H_{ij}^{(p)} = \phi_j^{(p-1)}(t^{i,(p)})$$

Efficient DeC into RK framework

$$DeC S = M \cdot (P-1) + 1$$

<u>c</u>	u ⁰	$u^{(1)}$	u ⁽²⁾	u ⁽³⁾		$\mathbf{u}^{(M-1)}$	$\mathbf{u}^{(M)}$	А	dim
0	0							u ⁰	1
β_1	$\beta_{1:}$	<u>0</u>						$u^{(1)}$	М
$\left \frac{\overline{\beta}_{1}}{\beta_{1}} \right $	$\Theta_{1:,0}$	$\Theta_{1:,1:}^-$	<u>o</u>					u ⁽²⁾	М
$\begin{array}{c c} \underline{\beta}_{1:} \\ \underline{\beta}_{1:} \\ \underline{\beta}_{1:} \end{array}$	$\Theta_{1:,0}$	<u>0</u>	$\Theta_{1:,1:}^-$	<u>0</u>				u ⁽³⁾	М
	:	:		٠.	٠.			:	М
	:	:			٠.	٠		:	М
$\beta_{1:}$	$\Theta_{1:,0}$	<u>0</u>			<u>0</u>	$\Theta_{1:,1:}$	<u>o</u>	$\mathbf{u}^{(M)}$	М
<u>b</u>	$\Theta_{M,0}$	<u>0</u>				<u>0</u>	$\Theta_{M,1:}$	$\mathbf{u}^{M,(M+1)}$	

Efficient DeC into RK framework

DeCu
$$S = M \cdot (P-1) + 1 - \frac{(M-1)(M-2)}{2}$$

<u>c</u>	u ⁰	$u^{*(1)}$	u* ⁽²⁾	$u^{*(3)}$		$u^{*(M-2)}$	$\mathbf{u}^{*(M-1)}$	$\mathbf{u}^{(M)}$	А	dim
0	0								u ⁰	1
$\beta_1^{(2)}$	$\beta_1^{(2)}$	<u>0</u>							$\mathbf{u}^{*(1)}$	2
$\beta_1^{(3)}$	$W_{1:,0}^{(2)}$	$W^{\underline{\underline{0}}}_{1:,1:}$	<u>0</u>						u*(2)	3
$ \frac{\beta_{1:}^{(2)}}{\beta_{1:}^{(3)}} \\ \frac{\beta_{1:}^{(4)}}{\beta_{1:}^{(4)}} $	$W_{1:,0}^{(2)}$ $W_{1:,0}^{(3)}$	<u>o</u>	$W_{1:,1:}^{\underline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{\overline{$	<u>0</u>					u *(3)	4
	:	:		٠.	٠.,				:	:
		•		•	•					.
	:	:			٠	·			i :	:
$\frac{\beta_{1:}^{(M)}}{\beta_{1:}^{(M)}}$	$W_{1:,0}^{(M-1)} \ W_{1:,0}^{(M)}$	<u>0</u>			<u>o</u>	$W_{1:,1:}^{(M-1)}$	<u>0</u>	<u>0</u>	$\mathbf{u}^{*(M-1)}$	M
$\beta_{1:}^{(M)}$	$W_{1:,0}^{(M)}$	<u>0</u>	• • • •			<u>0</u>	$W_{1:,1:}^{\underline{\underline{0}}(M)}$	<u>0</u>	u ^(M)	М
<u>b</u>	$W_{M,0}^{(M+1)}$	<u>0</u>	• • •				<u>0</u>	$W_{M,1:}^{(M+1)}$	$\mathbf{u}^{M,(M+1)}$	

$$W^{(p)} := \begin{cases} H^{(p)} \Theta^{(p)} \in \mathbb{R}^{(p+2) \times (p+1)}, & \text{if } p = 2, \dots, M-1, \\ \Theta^{(M)} \in \mathbb{R}^{(M+1) \times (M+1)}, & \text{if } p \geq M. \end{cases}$$

Efficient DeC into RK framework

DeCdu
$$S = M \cdot (P - 1) + 1 - \frac{M(M - 1)}{2}$$

<u>c</u>	u ⁰	$u^{(1)}$	u ⁽²⁾	u ⁽³⁾		$\mathbf{u}^{(M-2)}$	$\mathbf{u}^{(M-1)}$	$\mathbf{u}^{(M)}$	А	dim
0	0								u ⁰	1
$\beta_1^{(1)}$	$\beta_1^{(1)}$	<u>0</u>							$\mathbf{u}^{(1)}$	1
$\beta_{1}^{(2)}$	$Z_{1:,0}^{(2)}$	$Z_{1:,1:}^{\underline{\underline{0}}}$	<u>0</u>						u ⁽²⁾	2
$ \frac{\beta_{1:}^{(2)}}{\beta_{1:}^{(3)}} $ $ \frac{\beta_{1:}^{(3)}}{\beta_{1:}^{(3)}} $	$Z_{1:,0}^{(3)}$ $Z_{1:,0}^{(3)}$	<u>0</u>	$Z_{1:,1:}^{\overline{\overline{(3)}}}$	<u>0</u>					u ⁽³⁾	3
	:	:		٠.	٠.				l :	:
	:	:			٠.	••			:	:
$\beta_{1}^{(M-1)}$	$Z_{1:,0}^{(M-1)}$	<u>0</u>			<u>o</u>	$Z_{1:,1:}^{(M-1)}$	<u>0</u>	<u>o</u>	$\mathbf{u}^{(M-1)}$	M-1
$\frac{\underline{\beta}_{1:}^{(M-1)}}{\underline{\beta}_{1:}^{(M)}}$	$Z_{1:,0}^{(M)}$	<u>0</u>				<u>0</u>	$Z_{1:,1:}^{\underline{\underline{S}}(M)}$	<u>0</u>	u ^(M)	М
<u>b</u>	$Z_{M,0}^{(M+1)}$	<u>0</u>					<u>0</u>	$Z_{M,1:}^{(M+1)}$	$\mathbf{u}^{M,(M+1)}$	

$$Z^{(p)} := \begin{cases} \Theta^{(p)} H^{(p-1)} \in \mathbb{R}^{(p+1) \times p}, & \text{if } p = 1, \dots, M, \\ \Theta^{(M)} \in \mathbb{R}^{(M+1) \times (M+1)}, & \text{if } p > M. \end{cases}$$

Equispaced

Р	М	DeC	DeCu	DeCdu
2	1	2	2	2
2 3	2	5	5	4
4	3	10	9	7
5	4	17	14	11
6	5	26	20	16
7	6	37	27	22
8	7	50	35	29
9	8	65	44	37
10	9	82	54	46
11	10	101	65	56
12	11	122	77	67
13	12	145	90	79

Gauss-Lobatto

Γ	Р	М	DeC	DeCu	DeCdu
	2	1	2	2	2
	3	2	5	5	4
	4	2 2 3	7	7	6
	5	3	13	12	10
	6	3	16	15	13
	7	4	25	22	19
	8	4	29	26	23
	9	5	41	35	31
	10	5	46	40	36
	11	6	61	51	46
	12	6	67	57	52
	13	7	85	70	64

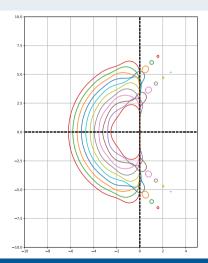
Stability Properties

DeC-DeCu-DeCdu

The stability function of DeC, DeCu, DeCdu of order *P* for any nodes distribution is

$$R(z) = 1 + z + \frac{z^2}{2!} + \cdots + \frac{z^P}{P!}.$$

DeC, DeCu, DeCdu



Efficient DeC

- Code DeCu or DeCdu
- Check order of accuracy
- Write a code to obtain its RK matrix
- Check the stability function with nodepy
- Compare computational costs with original DeC