

# High order time stepping and local time stepping for first order wave problems

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September 13, 2011

- Case of second-order hyperbolic problems treated by :  
[Jean-Charles Gilbert and Patrick Joly](#) *Higher order time stepping for second order hyperbolic problems and optimal CFL conditions*,  
[Julien Diaz and Marcus Grote](#), *Energy Conserving Explicit Local Time-Stepping for Second-Order Wave Equations*
- For first-order hyperbolic problems, second-order time scheme :  
[Serge Piperno](#) *Symplectic local time-stepping in non-dissipative DGTD methods applied to wave propagation problems*

# Model problem

First-order hyperbolic problem :

$$\frac{\partial \mathbf{U}}{\partial t} + \sum_{i=1}^d \mathbf{A}_i(\mathbf{x}) \frac{\partial \mathbf{U}}{\partial x_i} = f(\mathbf{x}, t)$$

with  $\mathbf{A}_i(\mathbf{x})$  symmetric matrices.

Use of Local Discontinuous formulation with centered fluxes :

$$\begin{aligned} \int_K \frac{\partial \mathbf{U}}{\partial t} \varphi \, dx - \int_K \sum_{i=1}^d \mathbf{A}_i(\mathbf{x}) \mathbf{U} \frac{\partial \varphi}{\partial x_i} \, dx + \int_{\partial K} \left( \sum_{i=1}^d \mathbf{A}_i(\mathbf{x}) n_i \right) \{ \mathbf{U} \} \varphi \, dx \\ = \int_K f(\mathbf{x}, t) \varphi \, dx \end{aligned}$$

$\mathbf{U}, \varphi \in V_h = \{ u \in (L^2(\Omega))^s \text{ such that } u \circ F_e \in \mathbb{P}_{r_e} \text{ or } \mathbb{Q}_{r_e} \}$

Order of approximation  $r_e$  is different for each element  $e$  of the mesh

First-order hyperbolic problem :

$$\frac{\partial U}{\partial t} + \sum_{i=1}^d A_i(x) \frac{\partial U}{\partial x_i} = f(x, t)$$

with  $A_i(x)$  symmetric matrices.

Associated evolution problem :

$$\frac{dU}{dt} + K_h U = F_h$$

With conservative boundary conditions,  $K_h$  is skew-symmetric.

# Modified equation approach

Second-order leap-frog scheme :

$$\frac{U^{n+1} - U^{n-1}}{2\Delta t} + K_h U^n = \mathcal{F}_h^n,$$

Stability condition of this scheme :

$$\Delta t \|K_h\|_2 \leq 1$$

Small elements in the mesh  $\Rightarrow$  restrictive CFL

In absence of source, the exact solution is given by

$$\frac{U^{n+1} - U^{n-1}}{2} = i \sin(i\Delta t K_h) U^n$$

Taylor expansion of the sinus provide the following scheme of order  $2m + 2$  :

$$\frac{U^{n+1} - U^{n-1}}{2} + \left[ \Delta t K_h + \sum_{q=1}^m \frac{(\Delta t K_h)^{2q+1}}{(2q+1)!} \right] U^n = 0.$$

Cost of this scheme :  $2m + 1$  matrix-vector products with  $K_h$

# Stability condition of modified equation

Let us denote the polynomial :

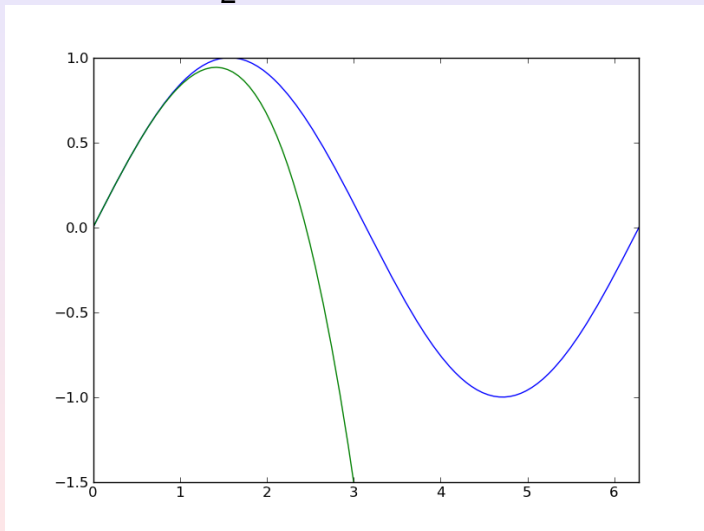
$$\tau_m(x) = x + \sum_{q=1}^m (-1)^q \frac{x^{2q+1}}{(2q+1)!}$$

Stability is obtained if

$$|\tau_m(x)| \leq 1 \Leftrightarrow x \in [0, \alpha_m]$$

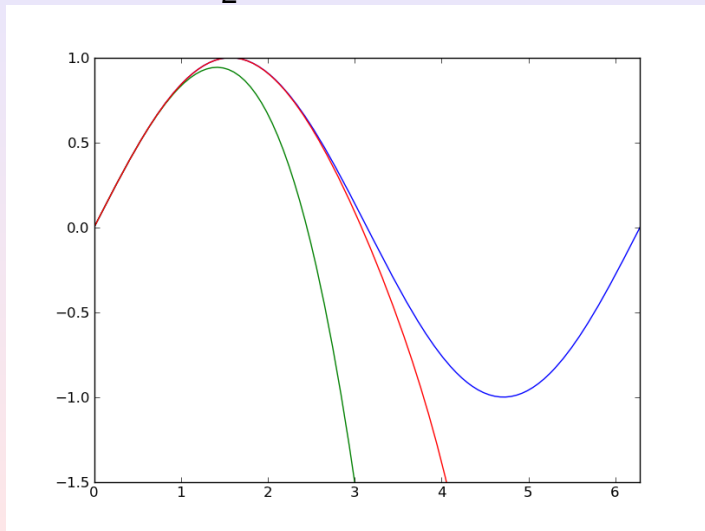
# Stability condition of modified equation

For  $m$  even,  $\alpha_m \leq \frac{3\pi}{2}$



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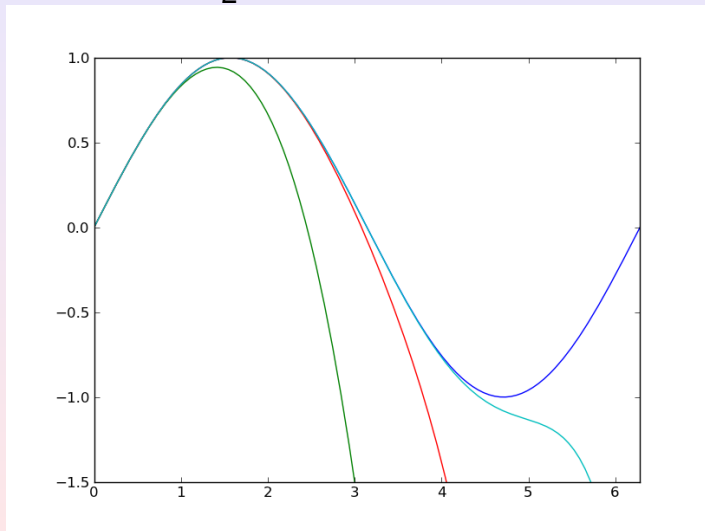
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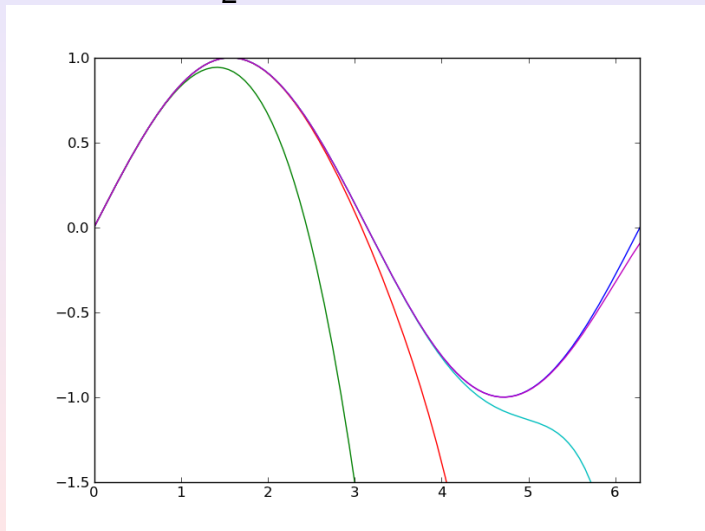
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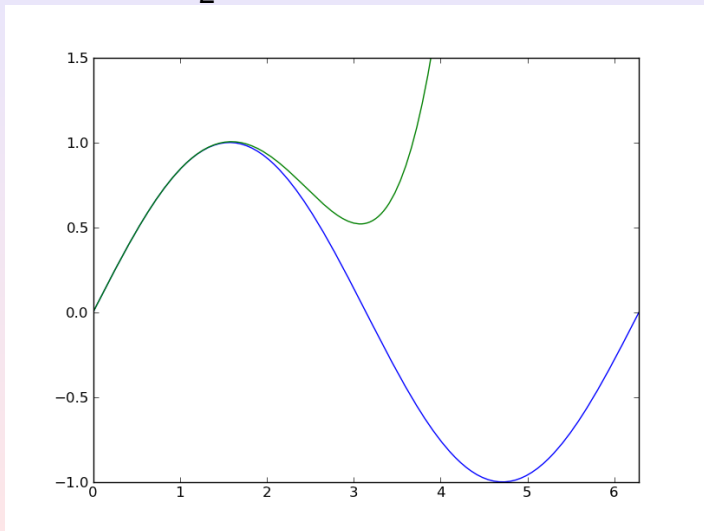
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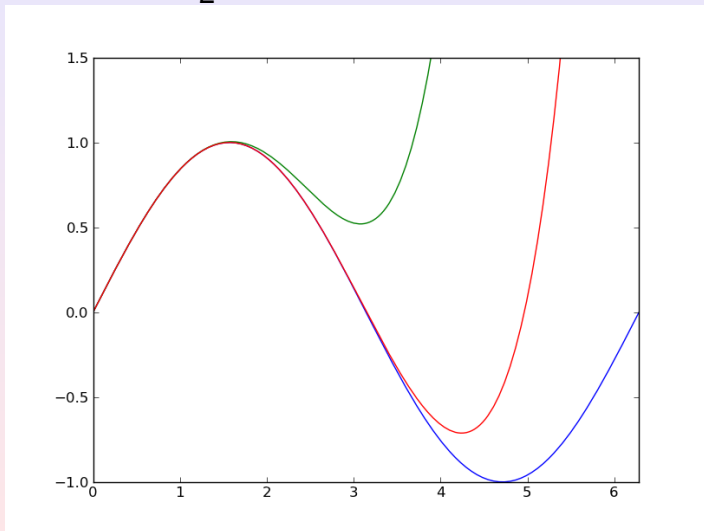
# Stability condition of modified equation

For  $m$  odd,  $\alpha_m \leq \frac{\pi}{2}$



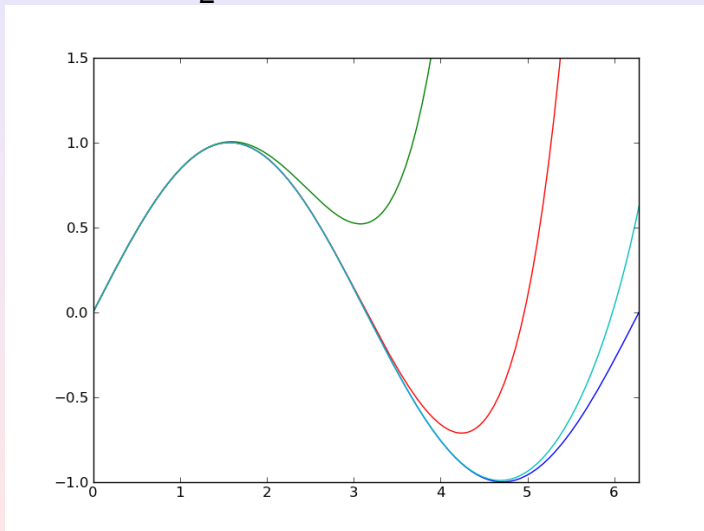
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For  $m$  odd,  $\alpha_m \leq \frac{\pi}{2}$



# Improvement of modified equation

Higher-order terms are added to increase CFL number

$$\frac{U^{n+1} - U^{n-1}}{2} + \left[ \sum_{q=0}^m \frac{(\Delta t K_h)^{2q+1}}{(2q+1)!} \right] U^n + \left[ \sum_{q=m+1}^r \alpha_q (\Delta t K_h)^{2q+1} \right] U^n = 0$$

This scheme is written under the form

$$U^{n+1} - U^{n-1} + 2i \mathcal{T}_{2r+1}(i\Delta t K_h) U^n = 0$$

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Optimal polynomial for  $m = 0$  (second-order), and nearly optimal for  $m = 1$  (fourth-order)

$$\mathcal{T}_{2r+1}^m(x) = \frac{1}{\xi_r} \mathcal{T}_{2r+1}^{Cheb} \left( \frac{(-1)^r \xi_r^m x}{(2r+1)} \right)$$

where  $\mathcal{T}_{2r+1}^{Cheb}$  are Chebyshev polynomials of the first kind and

$$\xi_r^0 = 1, \quad \xi_r^1 = \frac{2r+1}{2\sqrt{r(r+1)}},$$

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Stability condition :

$$\Delta t \|K_h\|_2 \leq \frac{2r+1}{\xi_r^m}$$

with

$$\xi_r^1 = 1 + O\left(\frac{1}{r^2}\right)$$



# Optimal polynomials

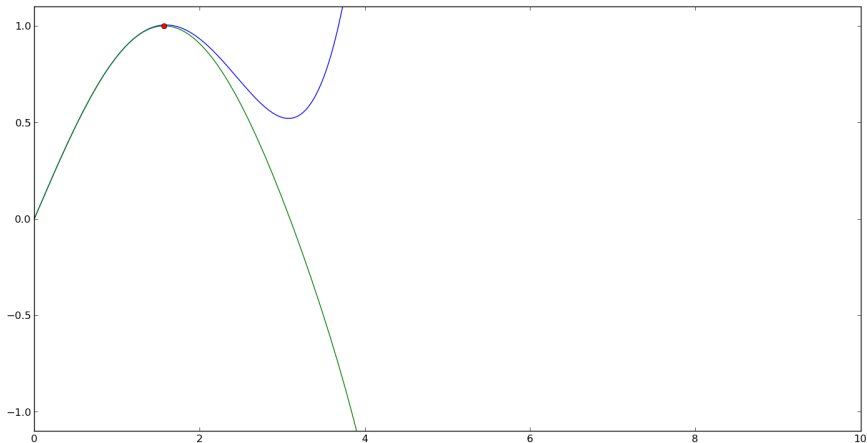
Optimal polynomial is sought, by searching tangent points  $\tau_1, \tau_2, \dots, \tau_k$  such that

$$\begin{cases} \mathcal{T}_{opt}(\tau_i) = -1 \text{ or } 1 \\ \mathcal{T}'_{opt}(\tau_i) = 0 \end{cases}$$

The associated non-linear system with unknowns  $\tau_1, \dots, \tau_k$  is solved numerically with Newton's method.

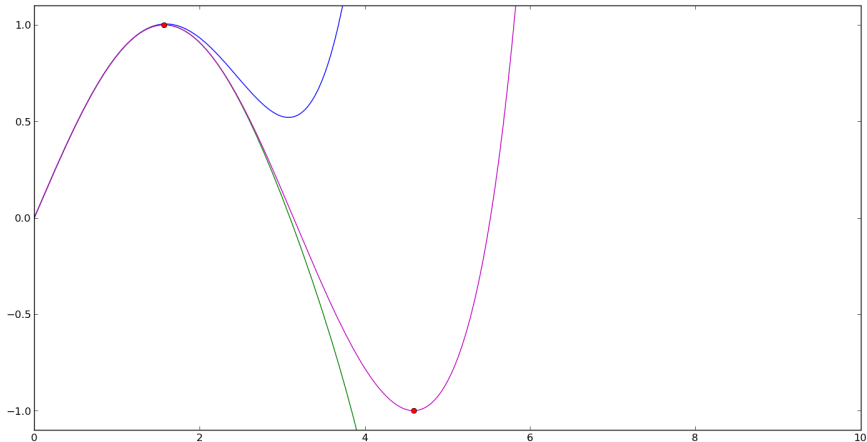
# Optimal polynomials

Tangents points in red :



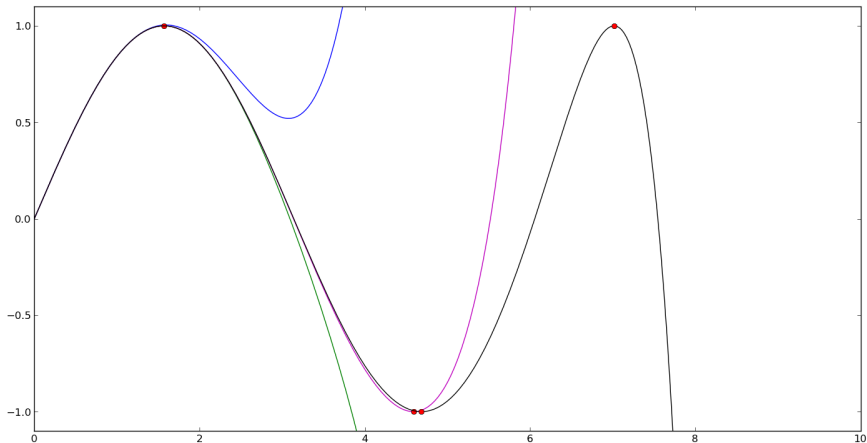
# Optimal polynomials

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# Optimal polynomials

CFL  $\alpha_m$  obtained with this procedure :

$m / r$	0	1	2	3	4	5	6
0	1	3	5	7	9	11	13
1	-	2.85	4.91	6.94	8.95	10.96	12.97
2	-	-	1.49	3.84	5.80	7.71	9.61
3	-	-	-	3.79	5.77	7.69	9.59

# Optimal polynomials

Efficiency compared to leap-frog scheme :

$m / r$	0	1	2	3	4	5	6
0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
1	-	0.95	0.982	0.991	0.994	0.996	0.998
2	-	-	0.298	0.549	0.644	0.701	0.739
3	-	-	-	0.541	0.641	0.699	0.738

# Chebyshev recurrence

- Use of Horner algorithm leads to numerical instabilities for large values of  $r$
- Use of Chebyshev recurrence leads to stable algorithms :

$$Q_0 = U^n$$

$$Q_1 = \frac{\xi_r}{2r+1} \Delta t K_h U^n$$

$$Q_n = \frac{2\xi_r}{2r+1} \Delta t K_h Q_{n-1} + Q_{n-2}$$

...

$$U^{n+1} = U^{n-1} - \frac{2}{\xi_r} Q_{2r+1}$$

# Two-level time stepping

Computational domain split into a “fine region” and a “coarse region”

$P_h$  : projector onto the fine region

$$\frac{U^{n+1} - U^{n-1}}{2} + \left[ \sum_{q=0}^m \frac{(\Delta t K_h)^{2q+1}}{(2q+1)!} \right] U^n \\ + \left[ \sum_{q=m+1}^r \alpha_q (\Delta t K_h P_h)^{2q} \right] \Delta t K_h U^n = 0,$$

Presence of  $P_h \Rightarrow$  terms of second sum are computed only on the “fine region”

Skew-symmetry of the matrices  $K_h P_h K_h \cdots K_h P_h K_h \Rightarrow$  stability of this scheme (CFL not controlled)

$\alpha_q$  are the coefficients defined previously so that CFL is increased.



# Stable two-level algorithm

For  $m = 0$ , it is equivalent to the following scheme (obtained by reproducing the strategy of Diaz and Grote) :

$$\left\{ \begin{array}{l} w_h = K_h(I - P_h)U^n \\ Q_0 = U^n \\ Q_1 = -\frac{\Delta t}{2r+1}(w_h + K_h P_h Q_0) \\ \text{For } k = 1, 2r \\ \quad Q_{k+1} = Q_{k-1} - \frac{2\Delta t}{2r+1}(K_h P_h Q_k + w \delta_k \text{ even}) \\ \text{End For} \\ U^{n+1} = U^{n-1} + 2Q_{2r+1} \end{array} \right.$$

Stable algorithm even for large values of  $r$

Domain split into hierarchical subdomains

$$\Omega = \bigcup \Omega_i = \bigcup K_e$$

with

$$\Omega \supset \Omega_1 \supset \Omega_2 \cdots \supset \Omega_r$$

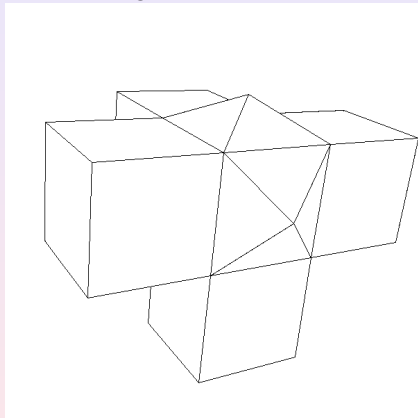
For each element, a nominal time step is computed

$$\Delta t_e = \frac{(2r + 1)c}{\xi_r^m \|\mathcal{P}_e K_h \mathcal{P}_e\|_2}$$

where  $c$  is a safety coefficient depending on the element.

# Multilevel algorithm

by considering only direct neighbors of each element



Global time step  $\Delta t$  is chosen by the user, then a level  $i$  is affected to each element with respect to the rule :

$$\text{if } \Delta t_e \leq \frac{\xi_i^m \Delta t}{2i+1}, \quad \text{then } K_e \in \Omega_i.$$

# Multilevel algorithm

We consider the following time scheme

$$\begin{aligned} \frac{U^{n+1} - U^{n-1}}{2\Delta t} &+ K_h U^n + \Delta t^2 K_h P_1 K_h P_1 K_h U^n \\ &+ \Delta t^4 K_h P_1 K_h P_2 K_h P_2 K_h P_1 K_h U^n \\ &+ \Delta t^6 K_h P_1 K_h P_2 K_h P_3 K_h P_3 K_h P_2 K_h P_1 K_h U^n + \dots = 0 \end{aligned}$$

where  $P_k$  are diagonal matrices :

$$P_k = \begin{pmatrix} \beta_k^0 & \dots & & & \\ \dots & \beta_k^1 & \dots & & \\ & \dots & \dots & \dots & \\ & & \dots & \dots & \dots \\ & & & \dots & \beta_k^r \end{pmatrix}$$

with

$$\beta_k^m = 0, \quad \forall m < k$$

# Multilevel algorithm

If we write the expansion of optimal polynomial  $\tau_{opt}^k(X)$  as :

$$\tau_{opt}^k(X) = X + \gamma_1^k X^3 + \gamma_2^k X^5 + \dots + \gamma_k^k X^{2k+1}$$

Coefficients  $\beta_k^m$  are chosen to coincide with these polynomials for each level

For  $k = 1, r$

For  $m = 1, k-1$

$$\beta_k^m = 0$$

End For

For  $m = k, r$

$$\beta_k^m = \sqrt{\gamma_k^m}$$

For  $n = 1, k-1$

$$\beta_k^m = \beta_k^m / \beta_n^m$$

End For

End For

End For

# Multilevel algorithm

Use of Horner algorithm :

$$Q_0 = \Delta t K_h U^n$$

$$Q_1 = \Delta t K_h P_1 Q_0$$

$$Q_2 = \Delta t K_h P_2 Q_1$$

...

$$Q_r = \Delta t K_h P_r Q_{r-1}$$

$$Q_{r-1} = Q_{r-1} + \Delta t K_h P_r Q_r$$

$$Q_{r-2} = Q_{r-2} + \Delta t K_h P_{r-1} Q_{r-1}$$

...

$$Q_0 = Q_0 + \Delta t K_h P_1 Q_1$$

$$U^{n+1} = U^{n-1} - 2Q_0$$

unstable due to round-off errors when  $r \geq 14$ .



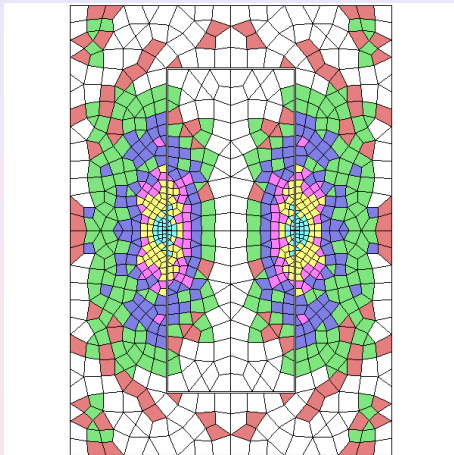
We consider wave equation

$$A_j = \begin{pmatrix} 0 & e_j^* \\ e_j & 0 \end{pmatrix}$$

and Neumann boundary conditions so that  $K_h$  is skew-symmetric

## 2-D numerical results

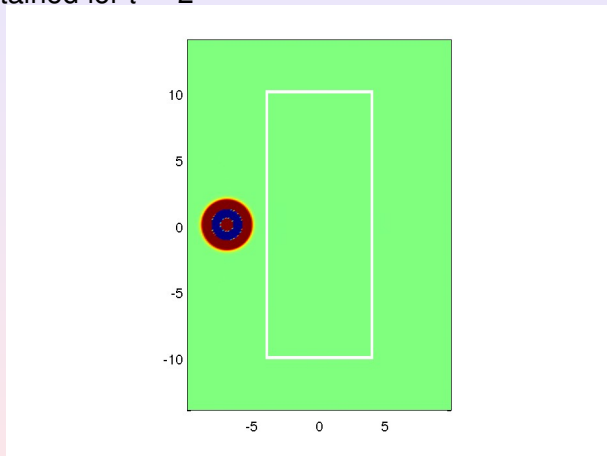
Box pierced with two small holes



each color corresponds to a different order of approximation in space

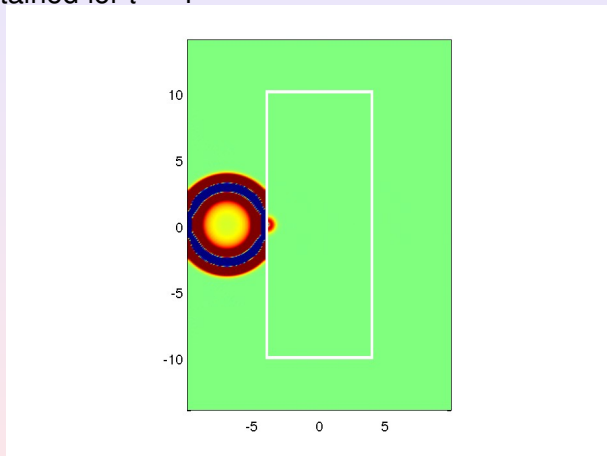
# 2-D numerical results

Solution obtained for  $t = 2$



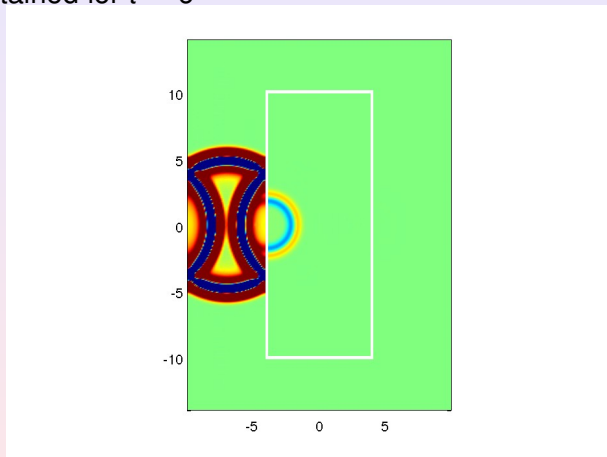
# 2-D numerical results

Solution obtained for  $t = 4$



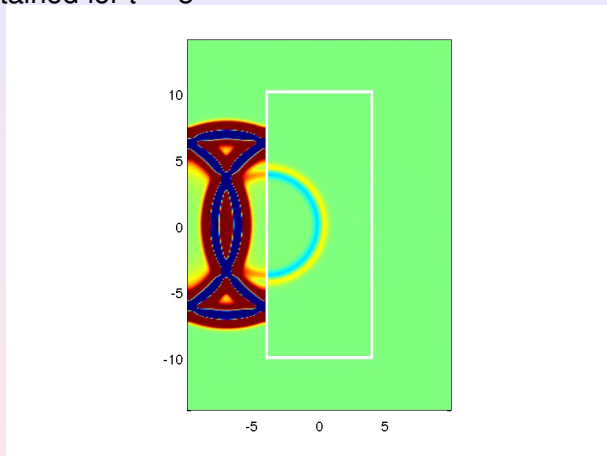
# 2-D numerical results

Solution obtained for  $t = 6$



# 2-D numerical results

Solution obtained for  $t = 8$



## 2-D numerical results

$$\Delta t_{max} = 0.01036, \quad \Delta t_{min} = 0.000737$$

$$\text{Ratio } \frac{\Delta t_{max}}{\Delta t_{min}} = 14.1$$

Computational time with optimized fourth order ( $\Delta t = 0.005$ ): **767s**

## 2-D numerical results

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Computational time with optimized fourth order ( $\Delta t = 0.005$ ): **767s**

Fourth-order local time stepping with the following repartition :

Level	1	2	3	4	5	6	7	8
Number of elements	1024	0	0	0	0	0	16	4

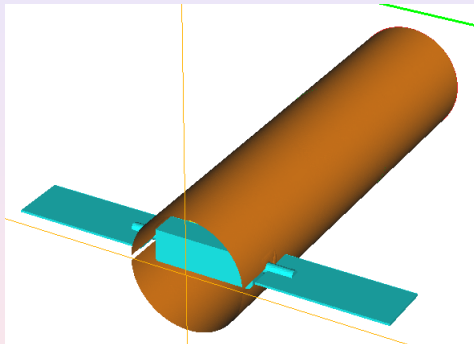
$L^2$  error for  $t = 10$  : 7.78e-6

Computational time ( $\Delta t = 0.01$ ): **177s**



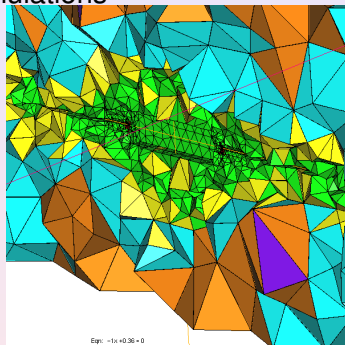
# 3-D numerical results

## Scattering by a satellite



# 3-D numerical results

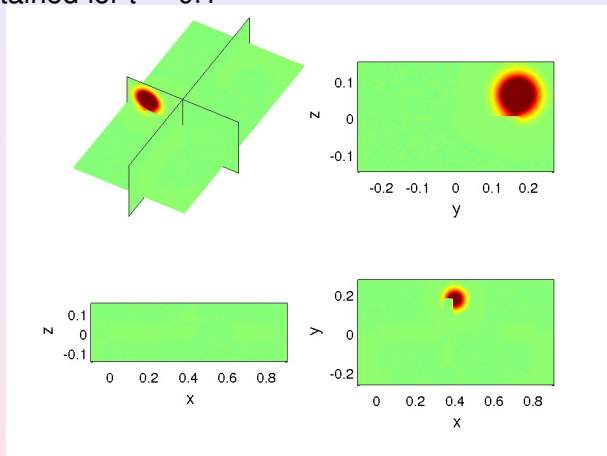
Mesh used for the simulations



each color corresponds to a different order of approximation in space

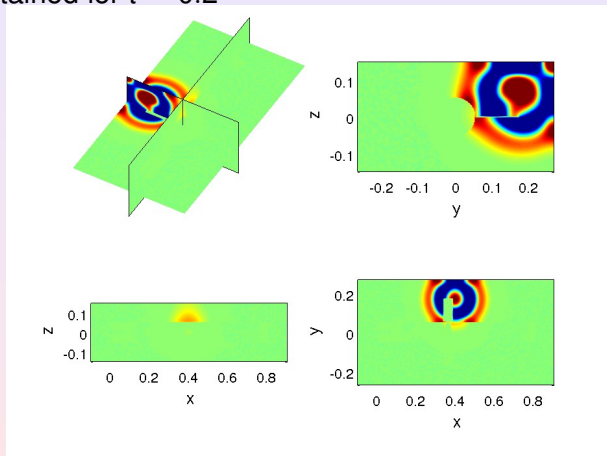
# 3-D numerical results

Solution obtained for  $t = 0.1$



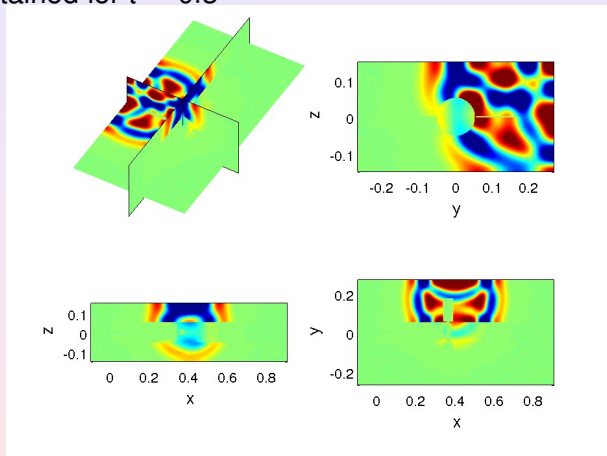
# 3-D numerical results

Solution obtained for  $t = 0.2$



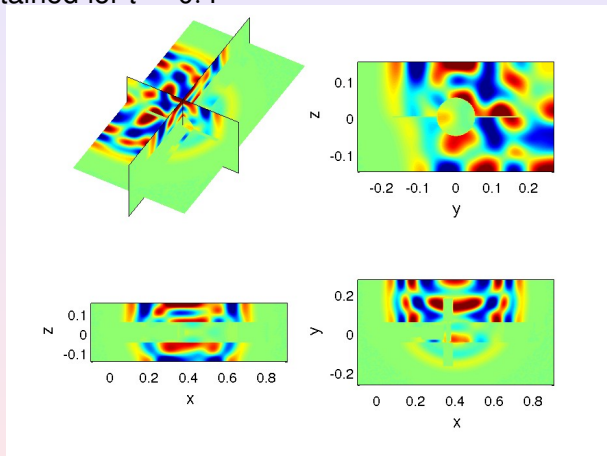
# 3-D numerical results

Solution obtained for  $t = 0.3$



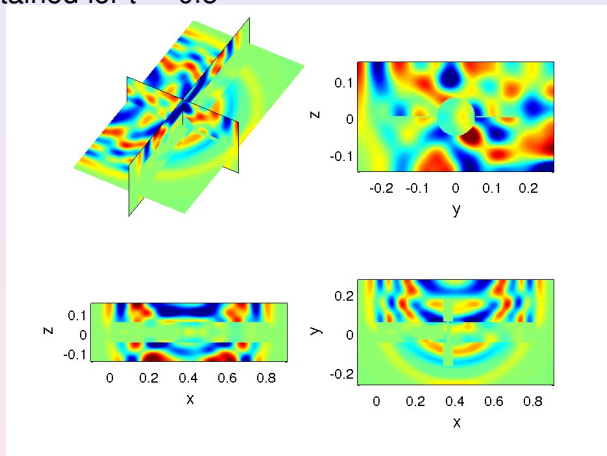
# 3-D numerical results

Solution obtained for  $t = 0.4$



# 3-D numerical results

Solution obtained for  $t = 0.5$



## 3-D numerical results

$$\Delta t_{max} = 1.177e - 3, \quad \Delta t_{min} = 1.442e - 5$$

$$\text{Ratio } \frac{\Delta t_{max}}{\Delta t_{min}} = 81.6$$

Computational time with standard leap frog ( $\Delta t = 1e - 5$ ): **63.4h**



## 3-D numerical results

$$\Delta t_{max} = 1.177e - 3, \quad \Delta t_{min} = 1.442e - 5$$

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Computational time with standard leap frog ( $\Delta t = 1e - 5$ ): **63.4h**

Second-order local time stepping with the following repartition :

Level	1	2	3	4	5	6	7	8	9	10
Number of elements	64468	7629	867	35	3	0	0	3	2	1

$L^2$  error for  $t = 0.5$  : 2.31e-3

Computational time ( $\Delta t = 2.5e - 4$ ) : **9.48h**

Another interesting set of schemes is obtained by considering Taylor expansion of the exponential. In absence of source :

$$U^{n+1} = \sum_{q=0}^m \frac{(\Delta t K_h)^q}{q!} U^n$$

Such schemes coincide with Runge-Kutta schemes for  $m \leq 4$

Advantage :  $K_h$  can be any matrix (not only skew-symmetric matrices)

# Runge-Kutta schemes

Cost of this scheme :  $m$  matrix-vector products

Stability condition for a skew-symmetric matrix :

$$\Delta t \|K_h\|_2 = \alpha_m$$

with :

- $m = 4k + 1$  or  $m = 4k + 2$ ,  $\alpha_m = 0 \Rightarrow$  scheme always unstable
- $m = 4k + 3$ ,  $\alpha_m < \pi$  and tends to  $\frac{\pi}{2}$
- $m = 4k$ ,  $\alpha_m < \frac{3\pi}{2}$  and tends to  $\pi$

# Optimal Runge-Kutta schemes

For second-order schemes, the following polynomials provide a CFL of  $m$  along the imaginary axis :

$$R_m(z) = \frac{1}{2}V_{m-1}(z) + V_m(z) + \frac{1}{2}V_{m+1}(z)$$

with  $V_m(z) = i^m T_m\left(\frac{z}{im}\right)$

associated scheme :

$$U^{n+1} = R_m(\Delta t K_h)U^n$$

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For fourth-order schemes, Kinmark-Gray polynomials :

$$K_m(z) = \frac{1}{\sqrt{\beta^2 + 1}} \left( i^{m+1} \beta T_{m-1}\left(\frac{iz}{\beta}\right) + \right.$$

$$\left. \frac{i^m}{2} \left[ (m-2) T_m\left(\frac{iz}{\beta}\right) - m T_{m-2}\left(\frac{iz}{\beta}\right) \right] \right)$$

$$CFL = \beta = \sqrt{(m-1)^2 - 1}$$

# Local time-stepping with Runge-Kutta approach

Multilevel algorithm :

$$U^{n+1} = U^n + \left( \Delta t K_h + \frac{\Delta t^2}{2} K_h^2 + \frac{\Delta t^3}{6} K_h^3 + \frac{\Delta t^4}{24} K_h^4 \right) U^n \\ + \Delta t^5 K_h P_1 K_h^4 U^n + \Delta t^6 K_h P_2 K_h P_1 K_h^4 U^n$$

coefficients of diagonal matrices  $P_1$ ,  $P_2$ , etc, deduced from coefficients of Kinmark polynomials

# Local time-stepping with Runge-Kutta approach

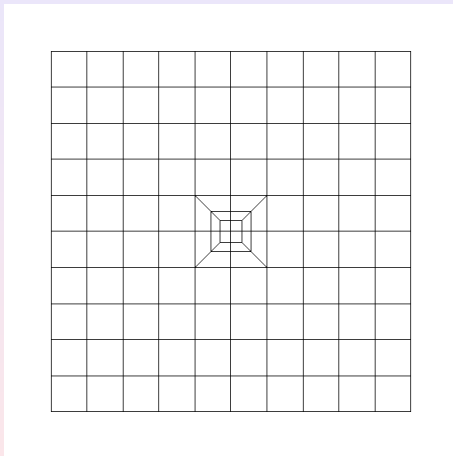
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# Local time-stepping with Runge-Kutta approach

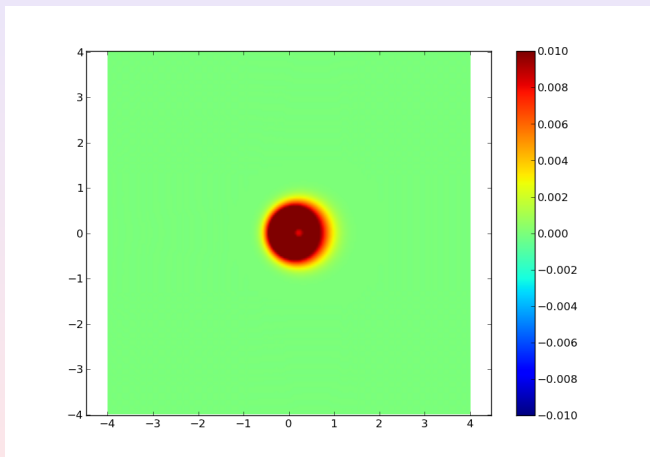
Numerical experiments with a mesh refined at the origin





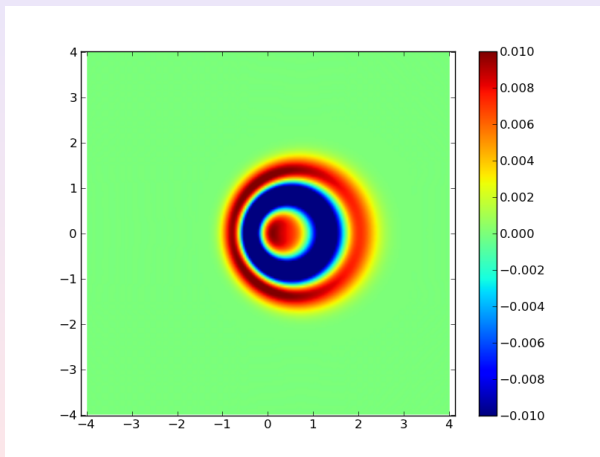
# Local time-stepping with Runge-Kutta approach

Aeroacoustics with an uniform flow and absorbing boundary condition



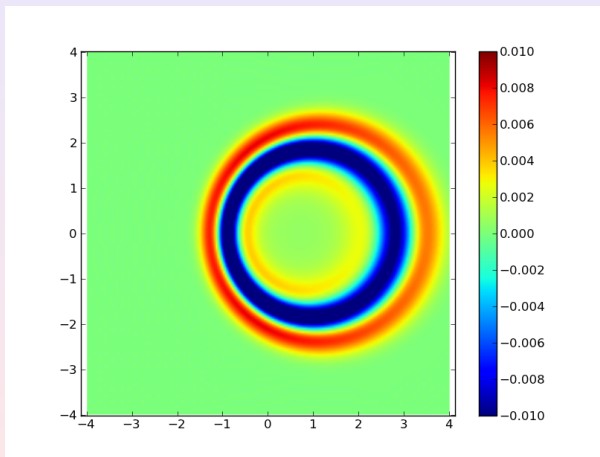
# Local time-stepping with Runge-Kutta approach

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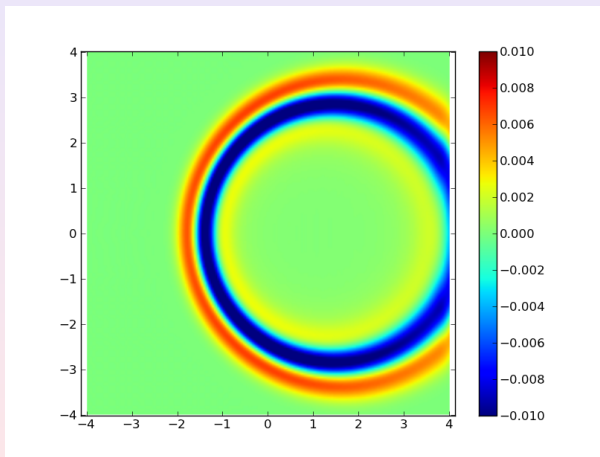
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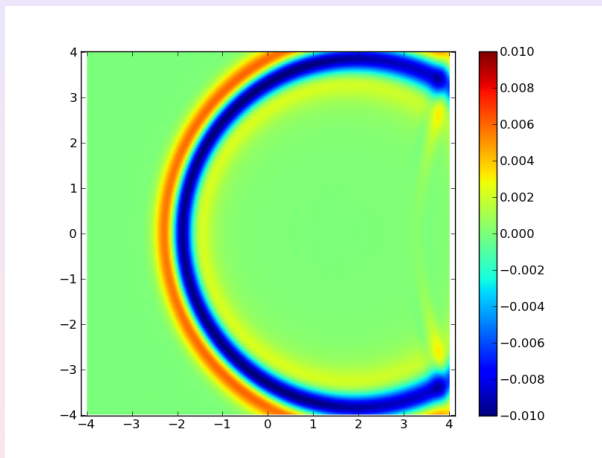
# Local time-stepping with Runge-Kutta approach

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# Local time-stepping with Runge-Kutta approach

Aeroacoustics with an uniform flow and absorbing boundary condition



$L^2$  error :  $1.8e-3$

- Stable multi-level algorithm for large values of  $r$
- Better knowledge of the global CFL from the local time steps  $\Delta t_e$
- Handle dissipative terms (due to upwind fluxes) locally with uncentered approximations