



# Time parallelization a new (?) and simple (!) method

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# Parallel computing

## Space direction

- **Domain decomposition methods (DDM):**  
efficient approach to parallelization of the **space direction**.
  - divide the whole domain in subdomains,
  - solve iteratively and independantly on each subdomain a smaller problem.
  - Corrections steps to propagate information from one subdomain to the others.
- Parallelization is not limited to space direction  
**Time direction is also a candidate for parallelization**



# Parallel computing

## Time Parallelization

- **Multiple shooting Methods:** Chartier, Philippe (1993), ...
- **Domain Decomposition type:** Saha, Tradel, Tremaine (1996), Gander (1996), Gander, Halpern, Nataf (1999), ...
- **Parareal method:** Lions, Maday, Turinici (2001), ...
- **Direct solvers:** Axelson, Verwer (1985), Worley (1991), Sleen, Sloan, Thomée (1999), Maday, Ronquist (2008), Güttel (2012), ...

M. Gander

*50 Years of Time Parallel Time Integration,  
Multiple Shooting and Time Domain Decomposition  
Springer Verlag, 2015.*



# Outline

Parallel computing

Discretization of Evolution Equations

Parallel computation of Linear Systems

Applications

- Laplace's equation

- Poisson's Equation

Limitations of the method

Conclusion

joint work with F. Hecht

initiated with A. Loumi and P. Parnaudeau



# Evolution Equations

## Implicit discretisation

Consider the linear ODE

$$u'(t) = Au(t)$$

with initial condition  $u(0) = u_0$ .

**FD Discretisation:** implicit scheme (backward Euler)

$$(I - \delta t_n A) \mathbf{u}^{n+1} = \mathbf{u}^n.$$

For  $m$  time steps

$$A_{n+m} \cdots A_{n+1} A_n \mathbf{u}^{n+m} = \mathbf{u}^n.$$

with

$$A_k = I - \delta t_k A$$



# Evolution Equations

## Numerical discretisation

Let us recall what a **sequential procedure** would be.

- Computing  $X = A_1 \cdots A_m$  before solving the linear system  
 $\implies$  **very expensive**
- A **sensible sequential** computation of the solution:

- Compute successively  $\mathbf{u}^{n+j}$  (for  $j = 1, \dots, m$ )  
solution of

$$A_j \mathbf{u}^{n+j} = \mathbf{u}^{n+j-1}$$



### Cost

| **Total cost** =  $m \times$  the cost of solving one linear system.



# Evolution Equations

## Numerical discretisation

$$u'(t) = Au(t),$$

with initial condition  $u(0) = u_0$ .

Implicit scheme

$$X\mathbf{u}^{n+m} = \mathbf{u}^n.$$

with

$$X = (I - \delta t_{n+m}A) \cdots (I - \delta t_{n+1}A)(I - \delta t_n A)$$

Key idea :

$$\mathbf{u}^{n+m} = \mathbf{X}^{-1}\mathbf{u}^n \quad !!!$$



# Evolution Equations

## Numerical discretisation

Take  $m = 2$ .

$$X = (I - h_2 A)(I - h_1 A)$$

For  $h_2 \neq h_1$ , we have

$$X^{-1} = \alpha_1 (I - h_1 A)^{-1} + \alpha_2 (I - h_2 A)^{-1}$$

with

$$\alpha_1 = \frac{h_1}{h_1 - h_2}, \quad \alpha_2 = \frac{h_2}{h_2 - h_1}.$$



# Evolution Equations

## Numerical discretisation

Take  $m = 2$ . The solution of

$$X\mathbf{u}^{n+2} = \mathbf{u}^n$$

is broken down into

$$x = \alpha_1 x_1 + \alpha_2 x_2$$

Each vector  $x_j$  is the unique solution of the linear system

$$A_j x_j = y$$

There are two **independent linear systems** to solve.



### Cost

Computing  $x$  this way is as expensive as solving a single linear system (neglecting the  $\alpha_j x_j$  summation).



# Parallel computation of Linear Systems

$(A_i)_{i=1}^m$  are  $n \times n$  nonsingular real matrices

$$X = A_1 \cdots A_m.$$

**Problem** : how to compute **quickly** the solution  $x \in \mathbb{R}^n$  of the linear system

$$Xx = y$$

where  $y \in \mathbb{R}^n$  is any given vector?



## Parallel computation of Linear Systems

Consider  $m$  **distinct real numbers**  $h_i$  and

$$X = \prod_{i=1}^m (I + h_i A)$$

### Proposition

The inverse matrix of  $X$  (with nonsingular matrices  $A_i$ ) is

$$X^{-1} = \sum_{i=1}^m \alpha_i A_i^{-1}$$

with

$$\alpha_i = \prod_{k \neq i} (1 - h_k/h_i)^{-1}.$$

(partial fraction decomposition of rational functions!!)

## Parallel computation of Linear Systems

Accordingly, the solution  $Xx = y$  is broken down into

$$x = \sum_{i=1}^m \alpha_i x_i$$

Each vector  $x_i$  is the unique solution of the linear system

$$A_i x_i = y$$

There are  $m$  **independent linear systems** to solve.



### Cost

Computing  $x$  this way is as expensive as solving a single linear system (neglecting the  $\alpha_i x_i$  summation).

In practice, the **interconnections between processors are far from being neglectable.**



## Linear Systems arising from FEM

$$M \frac{u^{n+1} - u^n}{h_1} = Bu^{n+1}$$

with

- $M$ : mass matrix
- $B$ : stiffness matrix

For  $m = 2$

$$(I - h_1 A)(I - h_2 A)u^{n+2} = u^n$$

with

$$A = M^{-1}B$$

Exactly the same framework!

## FEM : Laplace equation (2D)

Consider the homogeneous equation

$$u'(t) = Au(t)$$

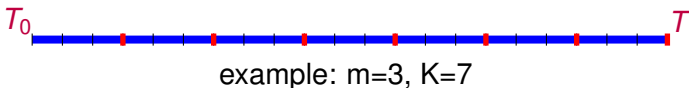
with initial condition  $u(0) = u_0$ .

**Implicit scheme using  $m$  distinct time steps  $h_i$**

**Notations:**

- initial and final times:  $T_0, T$
- $t_n = T_0 + n$  time steps :  $h_1 + h_2 + \dots + h_m + h_1 + h_2 + \dots$

**Remark:** solution computed only at times  $t_{km}, k = 1, \dots, K$



## FEM : Laplace's equation (2D)

Approximation from  $t_{km}$  to  $t_{(k+1)m}$



$$A_1 u_{km+1} = u_{km}$$

$$A_2 u_{km+2} = u_{km+1}$$

$$\vdots \quad \quad \quad \vdots$$

$$A_m u_{km+m} = u_{km+m-1}$$

- $X u_{(k+1)m} = u_{km}$

with  $X = A_1 \cdots A_m$



## FEM : Laplace's equation (2D)

For  $k = 1, \dots, K$

- Compute  $\mathbf{u}_{(k+1)m}$  solution of

$$\mathbf{X}u_{(k+1)m} = u_{km} + \mathbf{g}_{km}$$

$$u_{(k+1)m} = \sum_{i=1}^m \alpha_i x_i \quad (\text{Reduce step})$$

$x_i$  **computed in parallel**



example:  $m=3, K=7$

FreeFem

Domain: unit square

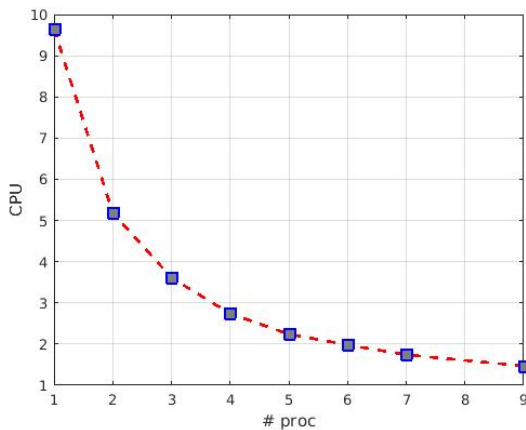
nb vertices =3721 , nb triangles = 7200 , nb boundary edges 240

```
real T = .1; int Ndt=3000;
```



# Applications

FEM : Laplace's equation (2D)





## Efficient parallel solvers

Let us introduce some definitions related to the computation time.

- $f(1, T, N) = f(1, mK\delta t, N)$  is the **sequential time** to compute the solution at the final time  $T = mK\delta t$ , for a problem of size  $N$ :

$$f(1, T, N) = mK f(1, \delta t, N).$$

- $f(m, T, N)$  is the **parallel time** to solve a problem of size  $N$  using  $m$  processors

$$f(m, T, N) = Kf(1, \delta t, N) + KC(m, N).$$

with  $C(m, N)$  the cost of the communications between  $m$  processors sharing a data of size  $N$ .

## Efficient parallel solvers

Two quantities are of interest in parallel computing:

- **Speedup**

$$S(m, T, N) = \frac{f(1, T, N)}{f(m, T, N)} = m \frac{f(1, \delta t, N)}{f(1, \delta t, N) + C(m, N)}.$$

- **Efficiency**

$$E(m, T, N) = \frac{1}{m} S(m, T, N) = \frac{f(1, \delta t, N)}{f(1, \delta t, N) + C(m, N)}.$$

In a perfect world, the communications have no cost  $\implies$  ideal speedup and ideal efficiency

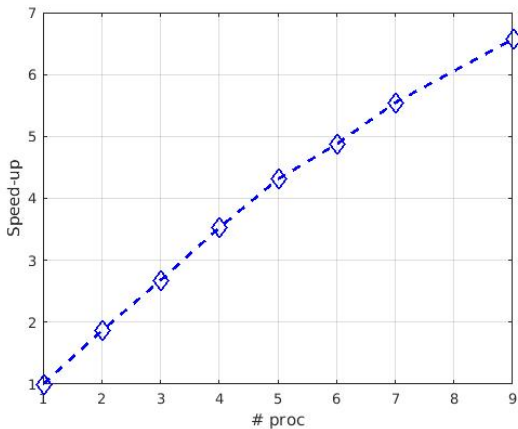
$$S^*(m, T, N) = m, \quad E^*(m, T, N) = 1.$$

**Unfortunately our world is not perfect** and the communication time  $C(m, N)$  is not at all neglectible.

# Applications

FEM : Laplace's equation (2D)

## Speedup

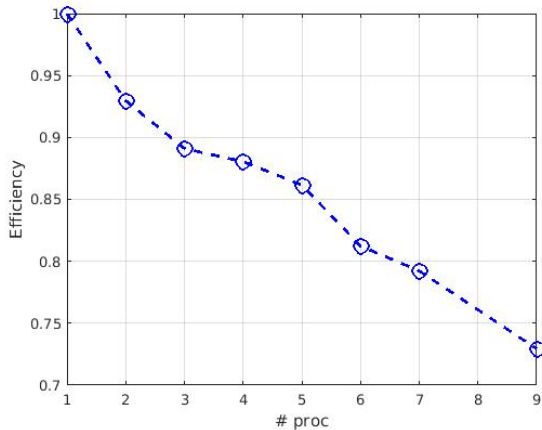




# Applications

FEM : Laplace's equation (2D)

Efficiency



## FEM : Poisson equation (2D)

Consider the non-homogeneous equation

$$u'(t) = Au(t) + f$$

with initial condition  $u(0) = u_0$ .

Implicit scheme using  $m$  distinct time steps  $h_i$

**Notations:**

- initial and final times:  $T_0, T$
- $t_n = T_0 + n$  time steps :  $h_1 + h_2 + \dots + h_m + h_1 + h_2 + \dots$

**Remark:** solution computed only at times  $t_{km}, k = 1, \dots, K$





## FEM : Poisson's equation (2D)

Approximation from  $t_{km}$  to  $t_{(k+1)m}$

$$A_1 u_{km+1} = u_{km} + h_1 f_{km+1}$$

$$A_2 u_{km+2} = u_{km+1} + h_2 f_{km+2}$$

$$\vdots \quad \quad \quad \vdots$$

$$A_m u_{km+m} = u_{km+m-1} + h_m f_{km+m}$$

$$X u_{(k+1)m} = u_{km} + g_{km}$$

with  $X = A_1 \cdots A_m$  and

$$g_{km} = h_1 f_{km+1} + A_1 \{ h_2 f_{km+2} + A_2 \{ h_3 f_{km+3} + \cdots + A_{m-1} \{ h_m f_{km+m} \} \} \}$$



## FEM : Poisson's equation (2D)

- $h = \frac{1}{m} \sum_i h_i, T = Km^2 h$
- Outer loop  $k = 1, \dots$ 
  - input:  $u_{km^2}$ 
    - For  $j = 1, \dots, m$   
compute  $u_{km^2+jm}$  by solving **sequentially**

$$Xu_{km^2+jm} = u_{km^2+(j-1)m} + \underbrace{g_{km^2+(j-1)m}}_{!!}$$

- output:  $u_{(k+1)m^2}$



## FEM : Poisson's equation (2D)

- For  $k = 1, \dots$ 
  - Compute **in parallel**  $g_{km^2+(j-1)m}$  for  $j = 1, \dots, m$
  - For  $j = 1, \dots, m$ 
    - compute **sequentially**  $u_{km^2+jm}$  by solving

$$Xu_{km^2+jm} = u_{km^2+(j-1)m} + g_{km^2+(j-1)m}$$

each  $u_{km^2+jm}$  being computed **in parallel**



See the computer sessions of the workshop



## Limitation(s) of the method

**Accuracy.** In practical situations, the solution  $x_j$  is known only through an approximation  $\tilde{x}_j$  with an error  $\varepsilon_j$  :

$$\|\tilde{x}_j - x_j\| \leq \varepsilon_j.$$

¿From which we deduce an upper bound on the error:

$$\|\tilde{x} - x\| \leq \left( \max_{1 \leq j \leq m} \varepsilon_j \right) \sum_{i=1}^m |\alpha_i|$$

this bound may be very large as  $m$  grows.

## Conclusion

- New method to solve linear time-dependent systems using a fractional decomposition of the matrix resulting from the discretisation of the time evolution operator.
- This method has been applied to solve the bidimensional heat equation. We obtain good results.
- However, for precision and stability reasons, the use of our method should be limited to moderate number of processors.
- We are currently investigating several developpements of the method.
  - **combine low-order methods (computed in parallele) to get a high-order one**
  - **combine with space Domain Decomposition.**
  - ...
- In summary, the method is efficient, but should be limited to a small number of processors



## References

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