

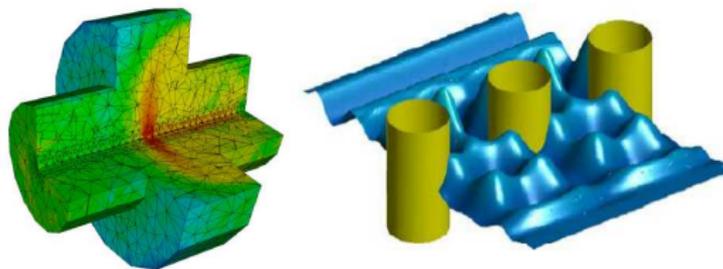
# Lecture - Introduction

Ph.D. Course:  
An Introduction to DG-FEM  
for solving partial differential equations

Allan P. Engsig-Karup  
Scientific Computing Section  
DTU Informatics  
Technical University of Denmark

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# Today!



- ▶ Presentation and practical details
- ▶ Introduction to DG-FEM methods
- ▶ Getting setup for hands-on exercises

## Course content

This course is organized by

- ▶ Ass. Prof. Allan Peter Engsig-Karup  
Building 321, r. 016  
DTU Informatics, Scientific Computing Section, DTU,  
Denmark.
- ▶ Prof. Jan Hesthaven  
Building 321, r. 010  
Division of Applied Mathematics, Brown University, USA.

The course is sponsored by two PhD schools at Technical University of Denmark

- ▶ DTU Informatics Graduate School ITMAN
- ▶ The Danish Center for Applied Mathematics and Mechanics, DCAMM

# Course content

The following topics are covered in the course

- 1 Introduction & DG-FEM in one spatial dimension
- 2 Implementation and numerical aspects (1D)
- 3 Insight through theory
- 4 Nonlinear problems
- 5 Extensions to two spatial dimensions
- 6 Introduction to mesh generation
- 7 Higher-order operators
- 8 Problem with three spatial dimensions and other advanced topics

# Course structure

## Week 1:

Time	Monday	Tuesday	Wednesday	Thursday	Friday
08.30-09.00	Breakfast				
09.00-11.30	1	2	3	4	Project work
12.30-16.00	Hands-on	Hands-on	Hands-on	Hands-on	Project work

## Week 2:

Time	Monday	Tuesday	Wednesday	Thursday	Friday
08.30-09.00	Breakfast				
09.00-11.30	5	5+6	7	8	Project work
12.30-16.00	Hands-on	Hands-on	Hands-on	Hands-on	Project work

- ▶ Lectures: approx. 2.5 h/day, including 15 mins review + 15 mins break.
- ▶ Hand-on exercises: approx. 3.5 h/day.

## Learning objectives

A student who has met the objectives of the course will be able to:

- ▶ Apply the basic ideas underlying discontinuous Galerkin methods.
- ▶ Apply the building blocks of DG-FEM methods for the simulation of phenomena described by partial differential equations.
- ▶ Identify and exploit the properties and structure of the underlying problem.
- ▶ Be able to complete basic analysis to formulate a suitable scheme for a new problem.
- ▶ Implement such methods and extensions in Matlab using the provided Matlab based toolbox.
- ▶ Skillfully perform numerical experiments.
- ▶ Analyse and explain the observed behavior of the methods based on a basic theoretical insight.
- ▶ Apply important principles underlying the use of modern numerical methods in selected applications.

## Coursework and assessment

This 2-week course has approx. 70 scheduled hours

- ▶ Breakfast and coffee/tee (0.5 hours/day)
- ▶ Lectures (approx. 2 hours/day)
- ▶ Discussions (as needed)
- ▶ Hands-on computer exercises (approx. 4 hours/day)
- ▶ Lunch (1 hours/day)

To pass the course and get a diploma the requirements are

- ▶ Completing a written report for assessment of work
- ▶ Satisfactory completion of assignment problems (approx. 40 hours)

The assignment is divided in two parts

- ▶ Each part will be available friday morning of each week
- ▶ Initiate your work on the assignments

# Practical details

- ▶ Background
  - ▶ What is your background?
  - ▶ Why are you here?
- ▶ Access to the databar terminals, software and Internet
- ▶ Access to Matlab codes, <http://www.nudg.org>
- ▶ Access to hands-on exercises/slides/ect.,  
<http://www2.imm.dtu.dk/~apek/DGFEMCourse2009/>
- ▶ Course material:  
*Nodal Discontinuous Galerkin Methods - Algorithms, Analysis, and Applications*  
By J. S. Hesthaven & T. Warburton (2008), Springer.
- ▶ General information

# Course work

The work in the course should be carried out in teams

- ▶ Two persons per team
- ▶ Hands-on exercises and assignment work is made by the team

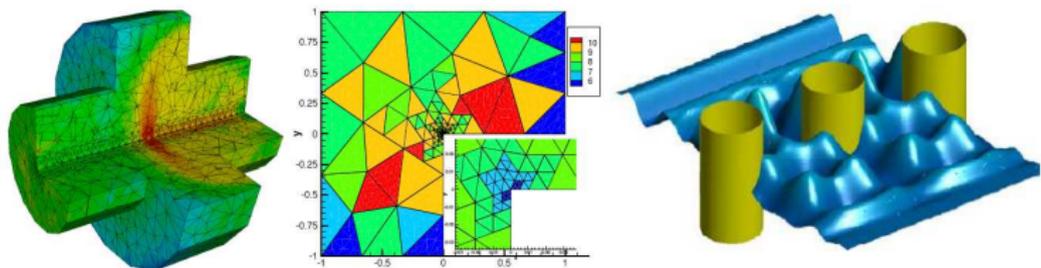
Everyone is encouraged to take the opportunity to

- ▶ Interact!
- ▶ Get to know each other!
- ▶ Discuss the work!
- ▶ Share experiences!

# Introduction

- discussion of numerical schemes and properties

# Our goals



For the application of **numerical methods** we want

- ▶ accuracy at minimal effort
- ▶ flexibility to solve classes of problems with same code
- ▶ easy problem prototyping and code maintenance (avoid adhoc solutions)
- ▶ ensure that numerical results can be thrusted (validation)

# Numerical solution of PDEs

To construct a numerical method for solving PDEs we need to consider

- ▶ How to **represent** the solution  $u(x, t)$  by an approximate solution  $u_h(x, t)$ ?
- ▶ In which sense will the approximate solution  $u_h(x, t)$  **satisfy** the PDE?

The two choices separate and define the properties of different numerical methods...

Bottom line is that we need ways to

- ▶ **Generate** a (coupled) system of equations from the well-posed PDE and incorporate boundary conditions
- ▶ **Solve** the system and equations while minimizing unavoidable errors that are introduced in the process

## Conservation laws

Conservation laws appear in many branches of computational science and engineering and are typically derived from physical conservation principles, e.g. conservation of energy, momentum and mass.

A general nonlinear conservation law (3D) can be stated in **differential form** as

$$\partial_t u + \nabla \cdot \mathbf{F}(u) = S(u)$$

or

$$\partial_t u + \partial_x F(u) + \partial_y G(u) + \partial_z H(u) = S(u)$$

where

$$u = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_m \end{bmatrix}, F(u) = \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_m \end{bmatrix}, G(u) = \begin{bmatrix} g_1 \\ g_2 \\ \vdots \\ g_m \end{bmatrix}, H(u) = \begin{bmatrix} h_1 \\ h_2 \\ \vdots \\ h_m \end{bmatrix}, S(u) = \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_m \end{bmatrix}$$

$u(\mathbf{x}, t)$  is a vector of **conserved variables** and  $F, G, H$  are **flux** vectors.  $S$  is a **source** vector.

# Conservation laws

## Examples of conservation laws

- ▶ Euler equations of compressible gas dynamics (1D)

$$\begin{aligned}\frac{\partial \rho}{\partial t} + \frac{\partial \rho u}{\partial x} &= 0 && \text{(Mass)} \\ \frac{\partial \rho u}{\partial t} + \frac{\partial(\rho u^2 + p)}{\partial x} &= 0 && \text{(Momentum)} \\ \frac{\partial E}{\partial t} + \frac{\partial(E+p)u}{\partial x} &= 0 && \text{(Energy)} \\ p &= (\gamma - 1) \left( E - \frac{1}{2} \rho u^2 \right), \quad c = \sqrt{\frac{\gamma p}{\rho}} && \text{(Ideal gas law)}\end{aligned}$$

- ▶ Nonlinear shallow water equations (1D)

$$\begin{aligned}\frac{\partial h}{\partial t} + \frac{\partial hu}{\partial x} &= 0 && \text{(Mass)} \\ \frac{\partial hu}{\partial t} + \frac{\partial(hu^2 + \frac{1}{2}gh^2)}{\partial x} &= 0 && \text{(Momentum)}\end{aligned}$$

- ▶ and many many more...

# Conservation laws

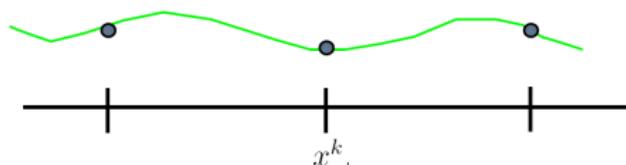
For now, we restrict ourselves to consider the one-dimensional **scalar conservation law**

$$\frac{\partial u}{\partial t} + \frac{\partial f}{\partial x} = g, \quad x \in \Omega$$

where  $f(u)$  is the flux function,  $g(x, t)$  is a source function.

Let's discuss basic ideas, advantages and disadvantages of different classical numerical methods for solving this PDE...

# Finite Difference Method



- ▶ Domain is represented by a set of collocation points
- ▶ Solution is **represented** locally as a polynomial

$$x \in [x^{k-1}, x^{k+1}]: \quad u_h(x, t) = \sum_{i=0}^2 a_i(t)(x - x^k)^i, \quad f_h(x, t) = \sum_{i=0}^2 b_i(t)(x - x^k)^i$$

- ▶ PDE is **satisfied** in a point-wise manner

$$\mathcal{R}_h(x^k) = \frac{du_h(x^k, t)}{dt} + \frac{f_h(x^{k+1}, t) - f_h(x^{k-1}, t)}{h^k + h^{k-1}} - g(x^k, t) = 0$$

- ▶ Local smoothness requirement pose a problem for resolving complex geometries, internal discontinuities and overall grid structure.

# Finite Difference Method

## Main benefits

- ▶ Simple to understand
- ▶ Straightforward implementation on structured meshes
- ▶ High-order accurate approximations feasible
- ▶ Method is local and can be made explicit in time
- ▶ Simple techniques for local adaptivity (upwinding)
- ▶ Extensive body of theoretical and practical work on these methods since 1960's

## Main problems

- ▶ Implementation complexity increases if geometric flexibility is needed
- ▶ Less well-suited for problems with discontinuities
- ▶ Grid smoothness requirements

# Finite Volume Method



- ▶ Domain is represented by non-overlapping cells
- ▶ Solution is **represented** locally as a cell average

$$\bar{u}^k \equiv \frac{1}{h^k} \int_{\Omega^k} u^k dx^k$$

- ▶ PDE is **satisfied** on conservation form

$$h^k \frac{d\bar{u}^k}{dt} + f(x^{k+1/2}, t) - f(x^{k-1/2}, t) = h^k \bar{g}^k$$

- ▶ The flux function needs to be reconstructed on cell interfaces  $x^{k\pm 1/2}$

$$f(x^{k-1/2}, t) = F(\bar{u}^{k-1}, \bar{u}^k), \quad f(x^{k+1/2}, t) = F(\bar{u}^k, \bar{u}^{k+1})$$

# Finite Volume Method

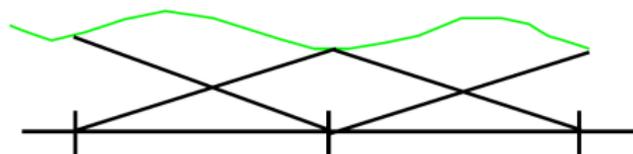
## Main benefits

- ▶ Robust
- ▶ Support resolution of complex geometries
- ▶ Well-suited for hyperbolic conservation laws (local upwinding)
- ▶ Method is local and can be made explicit in time
- ▶ Method is locally conservative (due to telescopic property)
- ▶ Extensive theoretical framework since 1970's

## Main problems

- ▶ Inability to achieve high-order accuracy in a straightforward way on general grids due to requirement for extended stencils (flux reconstruction problem)
- ▶ Grid smoothness requirements

# Finite Element Method



- ▶ Domain is represented by non-overlapping elements
- ▶ Solution is **represented** globally using piecewise polynomials

$$u_h(x) = \sum_{k=1}^K u(x_k, t) N^k(x), \quad N^i(x_j) = \delta_{ij}$$

- ▶ PDE is **satisfied** in a global manner

$$\int_{\Omega_h} \left( \frac{\partial u_h}{\partial t} + \frac{\partial f_h}{\partial x} - g_h \right) N^j(x) dx = 0, \quad j = 1, \dots, K \quad \Rightarrow \quad \mathcal{M} \frac{du_h}{dt} + \mathcal{S} f_h = \mathcal{M} g_h$$

- ▶ The semi-discrete scheme is implicit by construction and reduces overall efficiency

# Finite Element Method

## **Main benefits**

- ▶ Robust
- ▶ High-order accuracy can be combined with complex geometries
- ▶ Well-suited for elliptic problems (global statement)
- ▶ Extensive theoretical framework since 1970's

## **Main problems**

- ▶ Not well-suited for problems with direction (global statement)
- ▶ Implicit in time reduces overall efficiency

# Properties of numerical methods

Numerical methods for solving PDEs can in general be characterized by the properties

- ▶ **Accuracy**

Can we reduce the error? and how fast?

- ▶ **Flexibility**

What is the range of problems that can be solved using the chosen method?

- ▶ **Robustness**

Can we always expect a solution from our numerical model?

- ▶ **Efficiency**

How long does it take to compute our solution?

**Note:** Very often it is difficult to achieve all properties at once!

⇒ **Thus, we need to prioritize!**

- ▶ Choice is often dictated by domain complexity and required levels of accuracy.

# General properties of the numerical methods

Assesment of general properties of some classical numerical methods

	Complex geometries	High-order accuracy and $hp$ -adaptivity	Explicit semi-discrete form	Conservation laws	Elliptic problems
FDM	×	✓	✓	✓	✓
FVM	✓	×	✓	✓	(✓)
FEM	✓	✓	×	(✓)	✓
DG-FEM	✓	✓	✓	✓	(✓)

We want a scheme which have the properties

- ▶ The local high-order elements of FEM.
- ▶ The geometric flexibility of FEM and FVM.
- ▶ The local statement of the FVM.

These are exactly the components of the

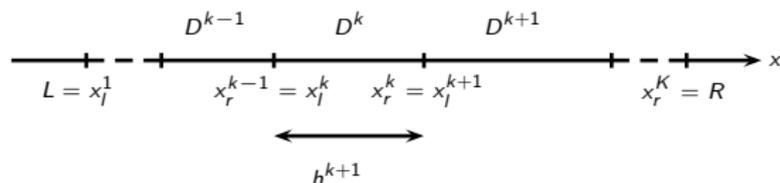
Discontinuous Galerkin Method Finite Element Method

# A first look at DGFEM

## Formulating a DG-FEM scheme

By subdividing the domain  $\Omega \in [L, R]$  similar to FVM/FEM into a union of non-overlapping elements  $D^k$

$$\Omega \cong \Omega_h = \bigcup_{k=1}^K D^k$$



we have the basis for [geometric flexibility](#) (any type of grid).

# Formulating a DG-FEM scheme

We seek to **represent** the global solution using local high-order polynomial approximations similar to FEM

$$u(x, t) \cong u_h(x, t) = \bigoplus_{k=1}^K u_h^k(x, t),$$
$$u_h^k(x, t) = \sum_{j=1}^{N_p} \hat{u}_j^k(t) \psi_j(x) = \sum_{j=1}^{N_p} u_h^k(x_j^k, t) l_j(x)$$

using either a modal or nodal form.

This is the basis for **high-order accurate** approximations.

Note: both low and high-order approximations then an option in the scheme.

## Formulating a DG-FEM scheme

We want to find an **approximation**  $u_h$  to the solution  $u$  of the general scalar conservation law

$$\partial_t u + \partial_x f(u) = g(x, t), \quad x \in \Omega$$

To do this, we form the local residual on the  $k = 1, \dots, K$  elements

$$x \in \mathcal{D}^k : \mathcal{R}_h^k(x, t) = \partial_t u_h^k + \partial_x f_h^k - g_h^k$$

and require this to vanish locally in a Galerkin sense

$$\int_{\mathcal{D}^k} \mathcal{R}_h^k(x, t) l_i^k(x) dx = 0$$

This is the basis for a **nodal DG-FEM scheme**.

However, we are not done yet... all elements are **disconnected** due to the local statement on the residual.

## Formulating a DG-FEM scheme

To **connect** elements, we apply Gauss's Theorem

$$\int_{\mathcal{D}^k} \mathcal{R}_h^k(x, t) l_j^k(x) dx = \int_{\mathcal{D}^k} \left[ \partial_t u_h^k l_j^k + \partial_x u_h^k l_j^k - g_h^k l_j^k \right] dx = 0$$

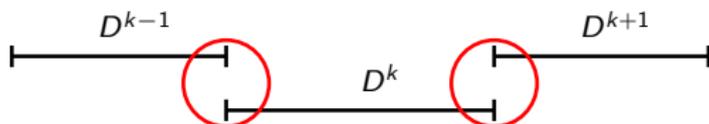
to convert the term with a spatial derivative such that

$$\int_{\mathcal{D}^k} \left[ \partial_t u_h^k l_j^k - u_h^k \partial_x l_j^k - g_h^k l_j^k \right] dx = - \oint_{\partial \mathcal{D}^k} \hat{n} \cdot f_h^k l_j^k dx$$

where the boundary integral in 1D takes the form

$$\oint_{\partial \mathcal{D}^k} \hat{n} \cdot f_h^k l_j^k dx = \left[ f_h^k l_j^k \right]_{x_r^k}^{x_l^k} = f_h^k(x_r^k) \delta_{N_{pj}} - f_h^k(x_l^k) \delta_{1j}$$

The solution is **not unique** at interfaces between adjacent elements.



We have multiple solutions! How can we address this problem?

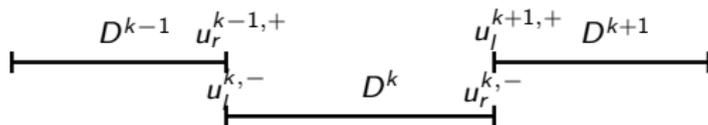
## Formulating a DG-FEM scheme

Similar to FVM, we could introduce a **numerical flux**  $f^*$  which approximate the physical flux, i.e.

$$\hat{n} \cdot f^* \cong \hat{n} \cdot f_h^k$$

to address the lack of solution uniqueness at the interfaces. We require that the numerical flux is somehow defined in terms of **interior (-)** and **exterior (+)** interface states

$$f^* = f^*(u_h^-, u_h^+)$$



Clearly, the choice of the numerical flux must be important!

## Formulating a DG-FEM scheme

So, after having applied Gauss's Theorem we found

$$\int_{\mathcal{D}^k} \left[ \partial_t u_h^k l_j^k - u_h^k \partial_x l_j^k - g_h^k l_j^k \right] dx = - \oint_{\partial \mathcal{D}^k} \hat{n} \cdot f_h^k l_j^k dx$$

With the introduction of a **numerical flux**  $f^*$ , the local scheme in the **weak form** then becomes

$$\int_{\mathcal{D}^k} \left[ \partial_t u_h^k l_j^k - u_h^k \partial_x l_j^k - g_h^k l_j^k \right] dx = - \oint_{\partial \mathcal{D}^k} \hat{n} \cdot f^* l_j^k dx$$

## Formulating a DG-FEM scheme

From the weak form

$$\int_{\mathcal{D}^k} \left[ \partial_t u_h^k l_j^k - u_h^k \partial_x l_j^k - g_h^k l_j^k \right] dx = - \oint_{\partial \mathcal{D}^k} \hat{n} \cdot f^* l_j^k dx$$

we can generate a local linear system by inserting the polynomial approximation  $u_h^k$  arriving at the compact scheme

$$\mathcal{M}^k \frac{du_h^k}{dt} - (\mathcal{S}^k)^T f_h^k - \mathcal{M}^k g_h^k = -f^* \delta_{1j} + f^* \delta_{Npj}$$

where  $\delta_{ij}$  is Kronecker's delta and the **element mass** and **stiffness<sup>1</sup> matrices** have been introduced. These are defined from

$$\mathcal{M}_{ij}^k = \int_{\mathcal{D}^k} l_i^k(x) l_j^k(x) dx, \quad \mathcal{S}_{ij}^k = \int_{\mathcal{D}^k} l_i^k(x) \frac{dl_j^k}{dx} dx$$

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<sup>1</sup>In classical finite element terminology, the discrete operator approximating the first derivative is called a convection/advection matrix.

## Formulating a DG-FEM scheme

It is also possible to derive yet another scheme from the **weak form**

$$\int_{\mathcal{D}^k} \left[ \partial_t u_h^k l_j^k - u_h^k \partial_x l_j^k - g_h^k l_j^k \right] dx = - \oint_{\partial \mathcal{D}^k} \hat{n} \cdot f^* l_j^k dx$$

by applying Gauss's Theorem once more

$$\int_{\mathcal{D}^k} \left[ \partial_t u_h^k l_j^k + \partial_x u_h^k l_j^k - g_h^k l_j^k \right] dx = \oint_{\partial \mathcal{D}^k} \hat{n} \cdot (f_h^k - f^*) l_j^k dx$$

This is the so-called **strong form**.

From this we can generate a local linear system of the form

$$\mathcal{M}^k \frac{du_h^k}{dt} + \mathcal{S}^k f_h^k - \mathcal{M}^k g_h^k = (f_h^k - f^*) \delta_{1j} - (f_h^k - f^*) \delta_{N_p j}$$

We now have two basic DG-FEM schemes. How will they perform?

The first examples...

## Examples: error behavior

Consider the simple advection equation on a periodic domain

$$\partial_t u - 2\pi \partial_x u = 0, \quad x \in [0, 2\pi], \quad u(x, 0) = \sin(lx), \quad l = \frac{2\pi}{\lambda}$$

Exact solution is then  $u(x, t) = \sin(l(x - 2\pi t))$ .

Errors at final time  $T = \pi$ .

$N \setminus K$	2	4	8	16	32	64	Convergence rate
1	-	4.0E-01	9.1E-02	2.3E-02	5.7E-03	1.4E-03	2.0
2	2.0E-01	4.3E-02	6.3E-03	8.0E-04	1.0E-04	1.3E-05	3.0
4	3.3E-03	3.1E-04	9.9E-06	3.2E-07	1.0E-08	3.3E-10	5.0
8	2.1E-07	2.5E-09	4.8E-12	2.2E-13	5.0E-13	6.6E-13	$\cong 9.0$

Error is seen to behave as

$$\|u - u_h\|_{\Omega, h} \leq Ch^{N+1}$$

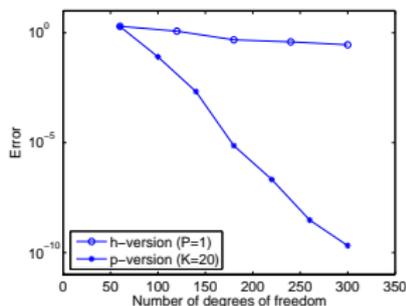
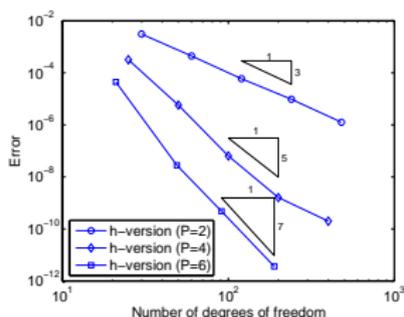
Clearly, paths to convergence are based on adjusting the size of elements ( $h$ -convergence), the polynomial order ( $p$ -convergence) or combinations hereoff.

## Examples: error behavior

Consider the linear shallow water equations in one horizontal dimension on a periodic domain

$$\frac{\partial}{\partial t} \begin{bmatrix} \eta \\ u \end{bmatrix} = \begin{bmatrix} 0 & -h \\ -g & 0 \end{bmatrix} \frac{\partial}{\partial x} \begin{bmatrix} \eta \\ u \end{bmatrix}$$

Tests of  $h$ - and  $p$ -refinement



Again, the error behaves as

$$\|u - u_h\|_{\Omega, h} \leq Ch^{N+1}$$

## Example - High-order makes the difference

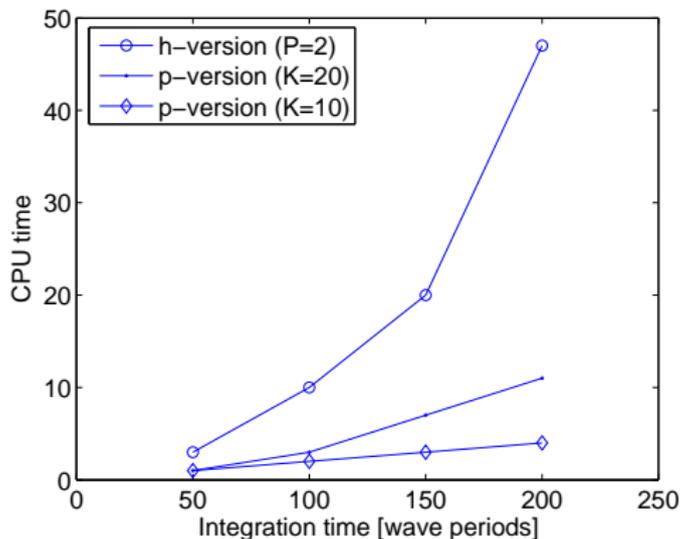


Figure: Optimized CPU-time vs. integration time for a fixed relative error in amplitude of 5%.

- ▶ Conclusion: a significant improvement in performance can be achieved using high-order elements over long times of integration.

# Numerical solution of PDEs

Important reasons for the interest in DG-FEM methods are

- ▶ Need for numerical methods of high accuracy in space and time
- ▶ Support for locally adaptive numerical solutions  
*hp*-adaptivity, meshes can be both non-conforming and unstructured.
- ▶ General and very flexible framework for solving large classes of PDEs
- ▶ Conceptually no difference between 1-D, 2-D or N-D
- ▶ The method is local (to the elements)

Note: For a high-order accurate method demand asymptotic behavior  $\mathcal{O}(h^p)$  of truncation error for  $h \rightarrow 0$  for  $p > 2$ .

## A brief history

- ▶ DG-FEM was first proposed by Reed & Hill in 1973 for a neutron transport equation

$$\sigma u + \nabla \cdot (au) = f$$

- ▶ First analysis by Lesaint & Raviart (1974) showing in general  $\mathcal{O}(h^N)$  and optimal  $\mathcal{O}(h^{N+1})$  for special meshes.
- ▶ Sharp analysis by Johnson (1986) showed  $\mathcal{O}(h^{N+1/2})$  for general meshes
- ▶ However, the schemes did not enjoy much use until further developments...

## A brief history

- ▶ Extension from scalar conservation laws to systems  
1980s-late 1990s, Cockburn/Shu
- ▶ Development of limiters and RKDG for problems with discontinuities  
Late 1980s, Shu/Cockburn
- ▶ Nodes, modes and large codes  
from 1995, Warburton/Karniadakis
- ▶ Maxwell's equations, MHD, water waves, elasticity, etc.  
- last decade has seen an explosion in development and applications
- ▶ Higher order problems
  - ▶ Interior-Penalty (IP), Arnold (1982)
  - ▶ Bassi-Rebay (BR), Bassi & Rebay (1997)
  - ▶ Local Discontinuous Galerkin (LDG), Cockburn & Shu (1998)

## A brief history

The last decade has seen an explosion in activities

- ▶ Hamilton-Jacobi equations
- ▶ Non-coercive problems and spectral accuracy
- ▶ Adaptive solution techniques
- ▶ Improved solvers
- ▶ Advanced time-integration methods
- ▶ Large-scale production codes
- ▶ etc.

## Brief summary

We have established **basic understanding** of DG-FEM

- ▶ How to formulate DG-FEM schemes
- ▶ Local expansions to achieve a high-order accurate basis
- ▶ Geometric flexibility in the spirit of FEM/FVM
- ▶ Explicit scheme and 'problem control' in the spirit of FVM

However, many **questions remains**

- ▶ How do we choose the numerical flux?
- ▶ Is the scheme stable?
- ▶ How does the idea generalize to multi-dimensions?
- ▶ What is the price?
- ▶ etc...