

Compatible Discretizations of Partial Differential Equations

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Outline

- 1 Functions
- 2 Approximation of BVP
- 3 Compatibility
- 4 Mimetic Finite Differences

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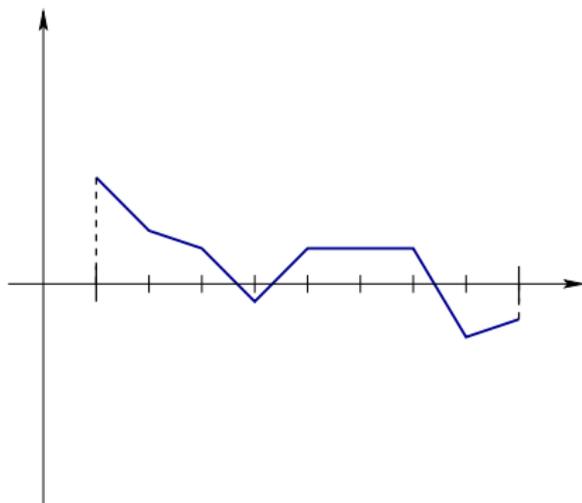
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This corresponds to use either **a basis** or **a dual basis**

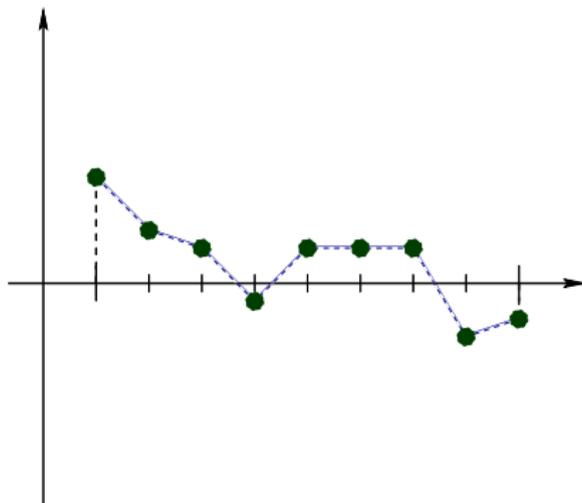
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You can think of a piecewise linear function or of its nodal values



Fourier and Wavelets

A **trigonometric polynomial** could be identified by its **Fourier Coefficients**. In other cases one might use **Wavelet Coefficients**, or **Legendre Coefficients**, and so on.

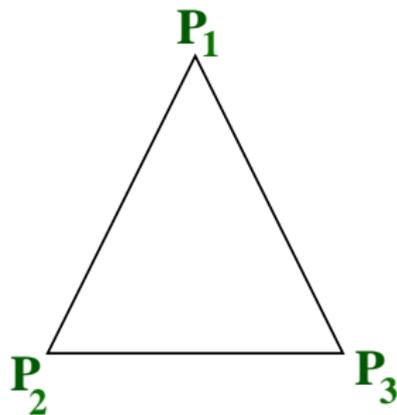
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Valuable instruments are often available in order to pass from one way of identification (e.g. the **point values**) to another (e.g. the **Fourier Coefficients**). For instance the **Fast Fourier Transform (FFT)**, the **Fast Wavelet Transform (FWT)**, etc.

Dual bases

On a given triangle T with vertices P_1, P_2, P_3 , in the space of linear functions, one could identify three **basis functions** $\lambda^1, \lambda^2, \lambda^3$ such that



$$\lambda^i(P_j) = \delta_j^i$$

Other types of values

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or by

- the values of the integral of its normal components on two given edges (non parallel)

$$\int_{e_1} \mathbf{v} \cdot \mathbf{n}^1 de \quad \int_{e_2} \mathbf{v} \cdot \mathbf{n}^2 de$$

The RT Spaces

A vector \mathbf{v} of the form

$$RT \equiv \text{span}\{(1, 0), (0, 1), (x, y)\}$$

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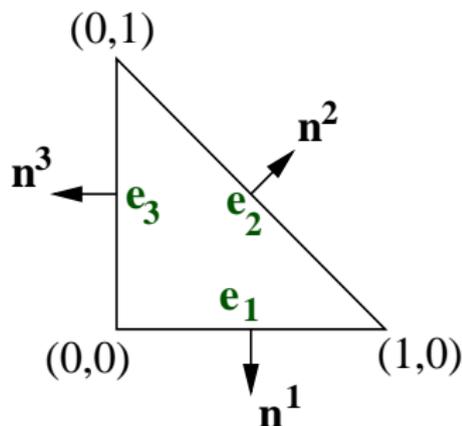
or by

- the values of the integral of its normal components on the edges of a given triangle.

$$\int_{e_1} \mathbf{v} \cdot \mathbf{n}^1 de, \quad \int_{e_2} \mathbf{v} \cdot \mathbf{n}^2 de, \quad \int_{e_3} \mathbf{v} \cdot \mathbf{n}^3 de$$

Dual bases

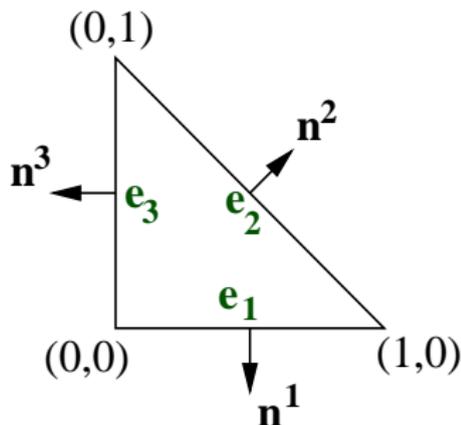
On a given triangle T with edges e_1, e_2, e_3 , in the space of RT vectors, one could identify three **basis functions** τ^1, τ^2, τ^3 such that



$$\int_{e_j} \tau^i \cdot \mathbf{n}^j \, de = \delta_j^i$$

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$$\tau^1 = (x, y - 1), \quad \tau^2 = (x, y), \quad \tau^3 = (x - 1, y)$$

The Mixed Zoo

(Arnold-Falk-Winther, Acta Numerica 2006)

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- scalars given by **element integrals** $\int_{\mathit{element}_j} f$

Two Philosophies - Unknowns

We can identify two different philosophies:

The unknown is a function, to be chosen within a given finite dimensional space (Ex: Finite Elements, Spectral Methods, Wavelets,...)

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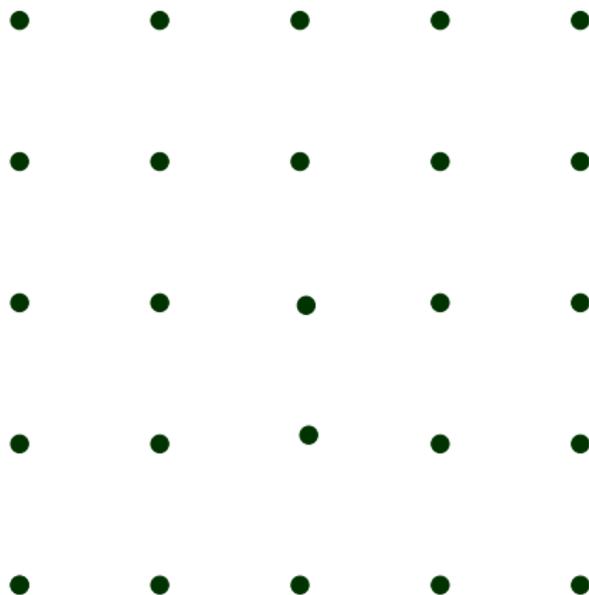
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The unknown is a set of values (point values, fluxes, works, volume integrals) (Ex: Finite Differences, Finite Volumes,.....)

Once a finite dimensional space has been chosen, the difference in the two approaches is just *psychological*. But this is not always the case.

Nodal Functions

In finite difference methods, the *unknowns* are the *nodal values*. We do not assume to have an underlying finite dimensional space



Differential Operators on Nodal Functions

How can we write *differential operators* on *nodal functions*?

For instance, the second derivative at C can be approximated as

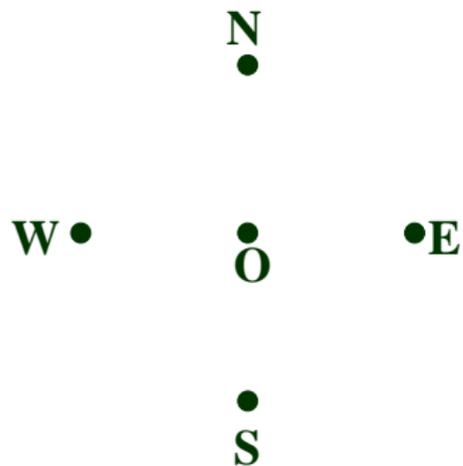
$$\bullet \quad \bullet \quad \bullet \quad \frac{d^2 f}{dx^2}(C) \simeq \frac{f(R) - 2f(C) + f(L)}{|R - C|^2}$$

\mathbf{L} \mathbf{C} \mathbf{R}

and the approximation is *exact* whenever f is a polynomial of degree ≤ 3 .

The Five-point Scheme

Similarly, the famous *five point method* reads

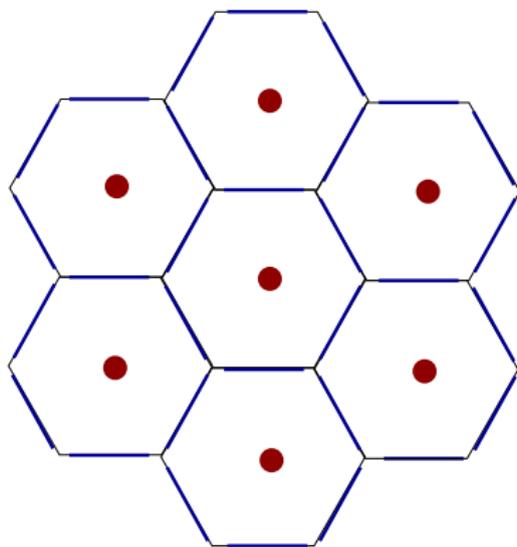


The diagram illustrates the five-point stencil. A central point is labeled 'O'. Four other points are positioned around it: 'N' is directly above 'O', 'S' is directly below 'O', 'W' is to the left of 'O', and 'E' is to the right of 'O'. Each point is represented by a small black dot with its corresponding letter next to it.

$$-h^2 \Delta_h f(O) = 4f(O) - f(N) - f(S) - f(W) - f(E)$$

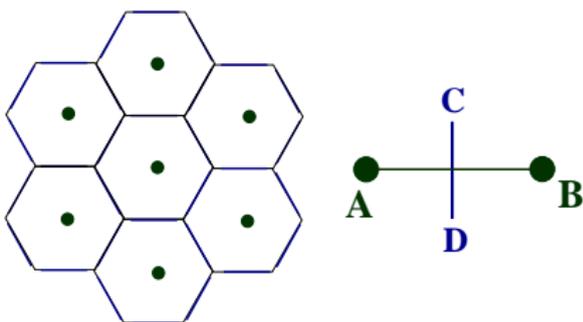
Finite Volumes-1

In other situations one can mix, e.g., **fluxes** and **volume integrals**. For instance, in Finite Volumes, one applies the divergence theorem to get (*exactly*) the integral of the divergence on each cell, given the fluxes on every edge.



Finite Volumes-2

However, the **fluxes** can be expressed as a function of the **nodal values at the centers of the cells** only in an approximate way



$$\begin{aligned} p(B) - p(A) &\simeq \int_A^B \frac{\partial p}{\partial x} \simeq \\ &\simeq \frac{|B - A|}{|C - D|} \int_{[C,D]} \frac{\partial p}{\partial \mathbf{n}} \end{aligned}$$

Approximating the space

Revisiting the *Two Philosophies* we can say that typically function-based methods use integral (weak) formulations. For instance the problem $-\Delta p = f$ in Ω with $p = 0$ on $\partial\Omega$ becomes: find $p_h \in L_h$ such that

$$\int_{\Omega} \nabla p_h \cdot \nabla q_h = \int_{\Omega} f q_h \quad \forall q_h \in L_h$$

where L_h is the space of piecewise linear functions vanishing on $\partial\Omega$.

Approximating the operator

Revisiting the *Two Philosophies* we can say that, on the other hand, value-based methods use a discretization of the differential operator. For instance the problem $-\Delta p = f$ in Ω with $p = 0$ on $\partial\Omega$ will be approximated by: Find a nodal function p_h such that

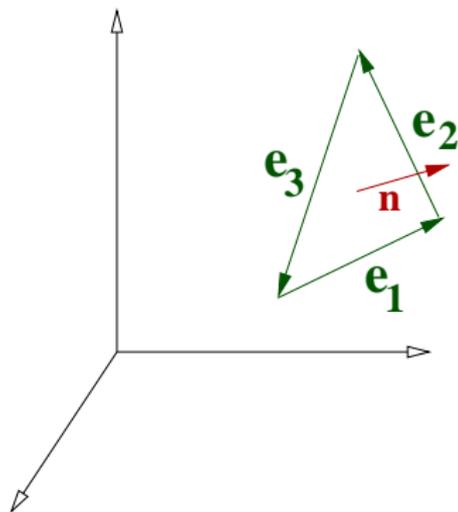
$$-\Delta_h p_h = f_h \quad \text{and} \quad p_h = 0 \quad \text{on} \quad \partial\Omega$$

where f_h is made by the nodal values of f .

Fundamental Laws

$$\text{curl } \mathbf{H} = \mathbf{j}$$

\mathbf{H} = Magnetic field \mathbf{j} = current density

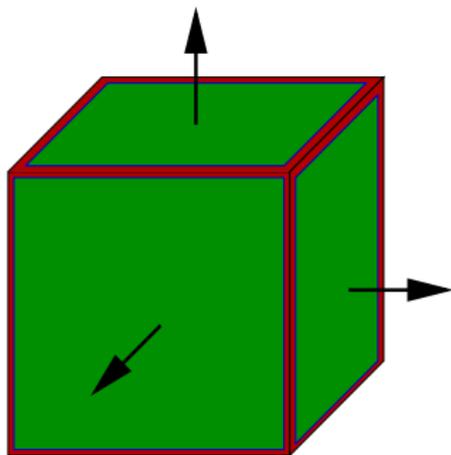


$$\sum_{k=1}^3 \int_{e_k} \mathbf{H} \cdot \mathbf{t}_k = \int_{\text{face}} \mathbf{j} \cdot \mathbf{n}$$

Fundamental Laws

$$\operatorname{div} \mathbf{D} = \rho$$

\mathbf{D} = Electric Displacement ρ = Charge density



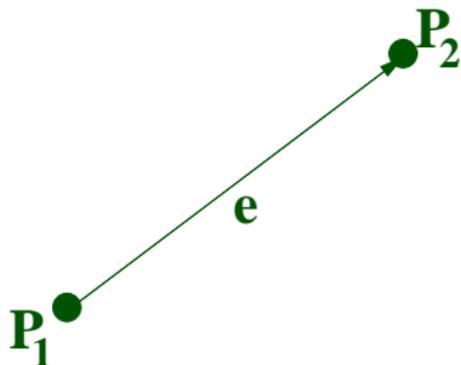
$$\sum_{k=1}^6 \int_{\text{face}_k} \mathbf{D} \cdot \mathbf{n}_k = \int_{\text{element}} \rho$$

Fundamental Laws

$$\mathbf{curl} \mathbf{E} = 0 \quad \mathbf{E} = -\nabla V$$

\mathbf{E} =Electric Field

V =Electric Potential



$$V(P_1) - V(P_2) = \int_{edge} \mathbf{E} \cdot \mathbf{t}$$

Less Fundamental Laws

$$\mathbf{D} = \epsilon \mathbf{E}$$

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Compatible Methods (as Mixed FEM)

try, as much as possible, to reproduce
exactly the **Fundamental Laws**

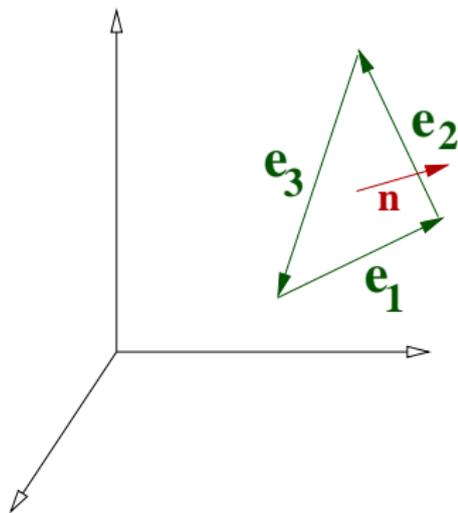
The Mixed Zoo - Compatible Animals

In the literature on **Mixed FEM**, the finite dimensional spaces are made (in 3 dimensions) by **four** types of **basic ingredients**:

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From edges to faces

$$\text{curl } \mathbf{H} = \mathbf{j}$$

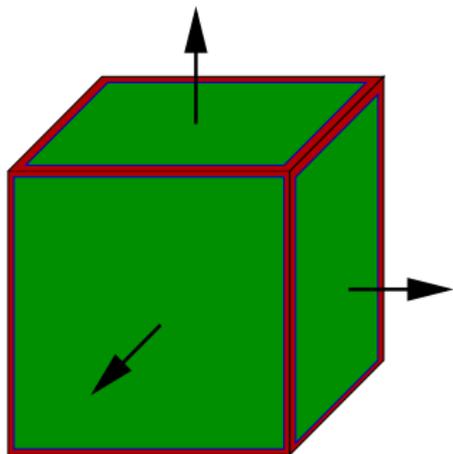


$$\sum_{k=1}^3 \int_{e_k} \mathbf{H} \cdot \mathbf{t}_k = \int_{\text{face}} \mathbf{j} \cdot \mathbf{n}$$

$\int_{\text{edge}} \mathbf{H} \cdot \mathbf{t}_k$ on each edge \Rightarrow $\int_{\text{face}} \mathbf{j}$ on each face

From faces to elements

$$\operatorname{div} \mathbf{D} = \rho$$

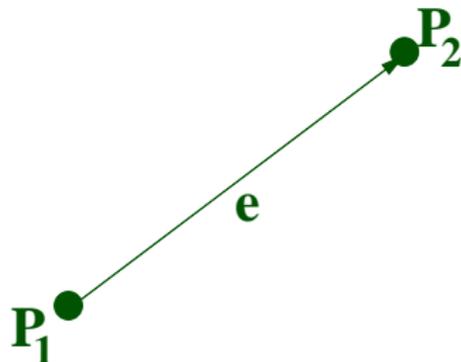


$$\sum_{k=1}^6 \int_{\text{face}_k} \mathbf{D} \cdot \mathbf{n}_k = \int_{\text{element}} \rho$$

$\int_{\text{face}} \mathbf{D} \cdot \mathbf{n}$ on each face \Rightarrow $\int_{\text{ele}} \rho$ on each element

From vertices to edges

$$\text{curl } \mathbf{E} = 0 \quad \mathbf{E} = -\nabla V$$



$$V(P_1) - V(P_2) = \int_{\text{edge}} \mathbf{E} \cdot \mathbf{t}$$

$$V \text{ on each node} \Rightarrow \int_e \mathbf{E} \cdot \mathbf{t} \text{ on each edge}$$

An *Ideally Compatible* Scheme

Ideally, the *most compatible* scheme should be:

$$\text{Minimize } \|\mathbf{B} - \mathbf{H}\|$$

under the constraints

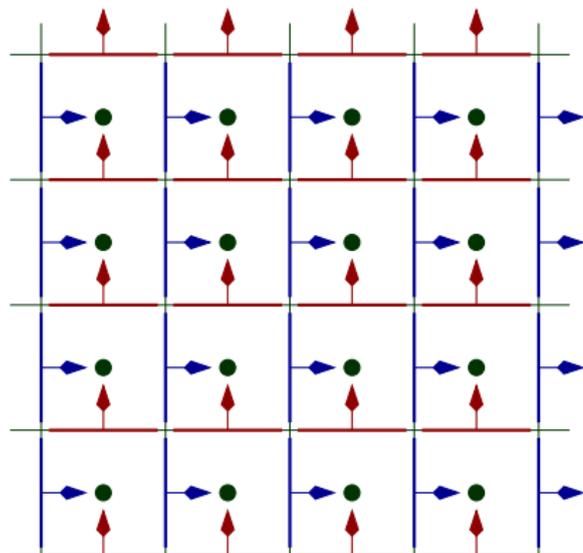
$$\text{curl}\mathbf{H} = \mathbf{j} \quad \text{and} \quad \text{div}\mathbf{B} = 0$$

(J.Rikabi, A. Bossavit, I. Perugia)

But it is very expensive...

Other Compatible Schemes

The **Marker and Cell (MaC)** scheme (Harlow & Welch, 1965) for Darcy flows uses *horizontal velocities* on the vertical edges, *vertical velocities* on horizontal edges, and *pressure* at the center of each cell.

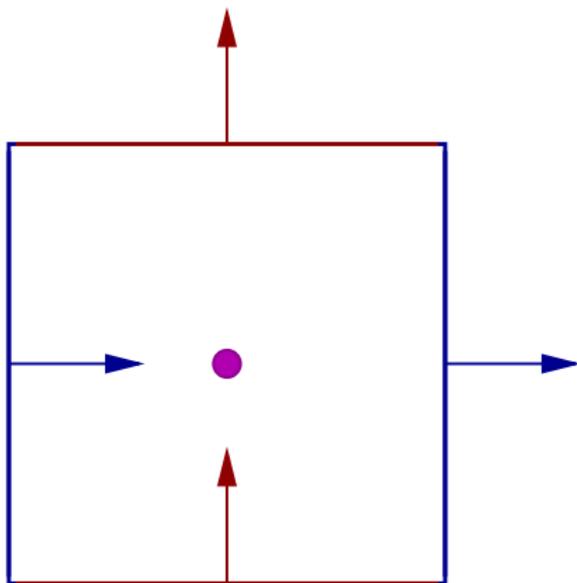


$$-\nabla p = \mathbf{u} \equiv (u, v)$$

$$\operatorname{div} \mathbf{u} = \text{source}$$

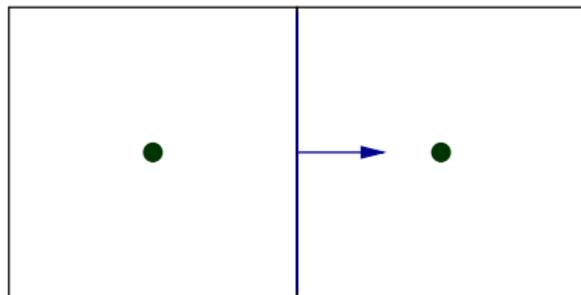
The Conservation equation in MaC

The *conservation equation*, e.g., $\operatorname{div} \mathbf{u} = 0$, is satisfied in a strong sense. This is *the compatible part* of the scheme.



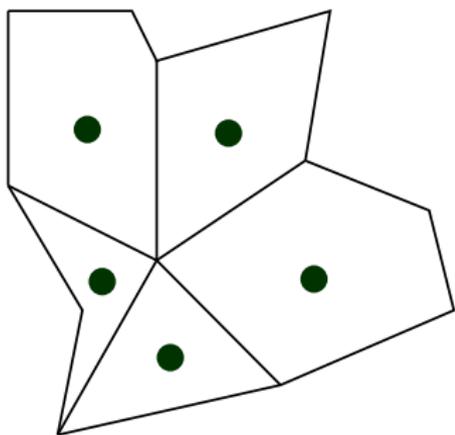
The constitutive equation in MaC

On a regular grid (like ours) the fluxes are easily computed as functions of the pressure



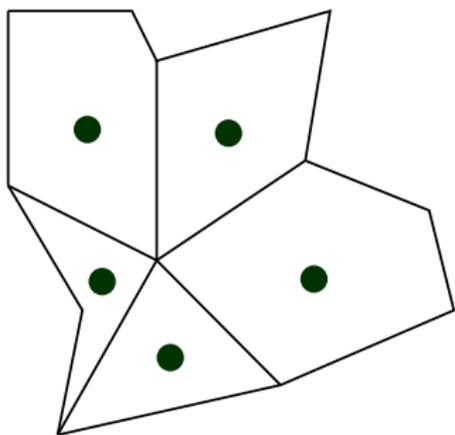
The constitutive equation in MaC

What if the grid is like this?



The constitutive equation in MaC

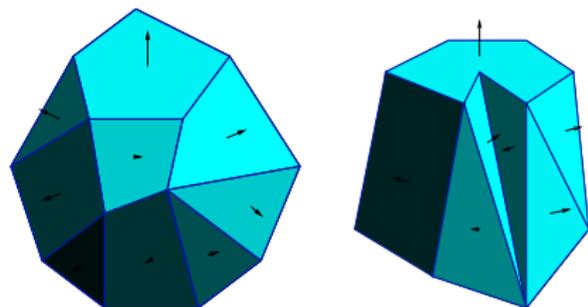
What if the grid is like this?



On a general geometry one needs all sorts of tricks in order to do it. All in a *quite incompatible* way!

General geometries

On the other hand also for Mixed Finite Element Methods, it is **very hard** to find the proper **finite dimensional spaces** in General geometries

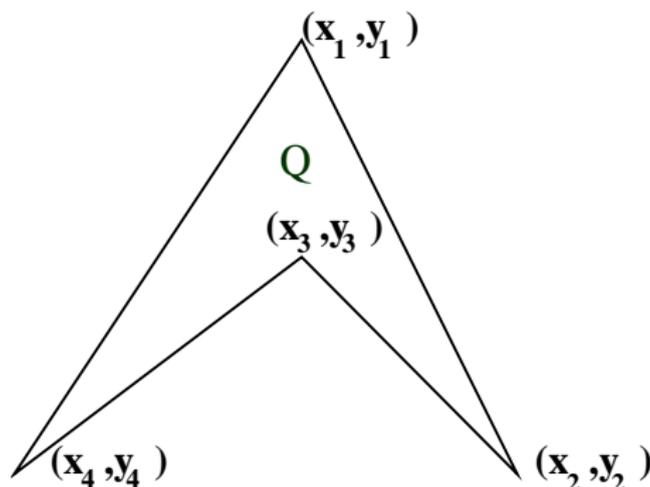


Two examples of polyhedral elements

Mimetic Finite Differences

(Hyman-Scovel 1988)

Let us consider the **most simple** example of a **nodal function** on a quadrilateral.



Towards a local Scalar Product

We want to mimic the *scalar product*

$$\int_Q \nabla u \cdot \nabla v \, dQ$$

but we lack the proper **finite dimensional space** (of dimension four, in a one-to-one correspondence with the **nodal values at the vertices**).

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Hence besides the functions "1", "x", "y" we would need a **fourth function** Ψ .

The nodal values of linear functions

We work on **nodal functions**. Hence, in our **world** (a copy of \mathbb{R}^4), the **linear functions** are represented by *the three-dimensional subspace generated by*:

- the nodal values of the function "1": $(1, 1, 1, 1)$
- the nodal values of the function "x": (x_1, x_2, x_3, x_4)
- the nodal values of the function "y": (y_1, y_2, y_3, y_4)

The local matrix-1

	"1"	"X"	"Y"	
"1"	0	0	0	0
"X"	0	 Q 	0	
"Y"	0	0	 Q 	
	0			

If both u and v are one of the functions 1 , x , or y the scalar product $\int_Q \nabla u \cdot \nabla v \, dQ$ is easy to compute.

The local matrix-2

We make the **further assumption that**: On each edge of ∂Q all the functions of our four-dimensional local space are integrated exactly by the *trapezoidal rule*

$$\int_a^b f \simeq \frac{b-a}{2} [f(a) + f(b)]$$

It **follows** that, for p_1 **linear**, the integral

$$\int_Q \nabla p_1 \cdot \nabla v \, dQ = \int_{\partial Q} \frac{\partial p_1}{\partial \mathbf{n}} v$$

depends **only** on the **nodal values** of v .

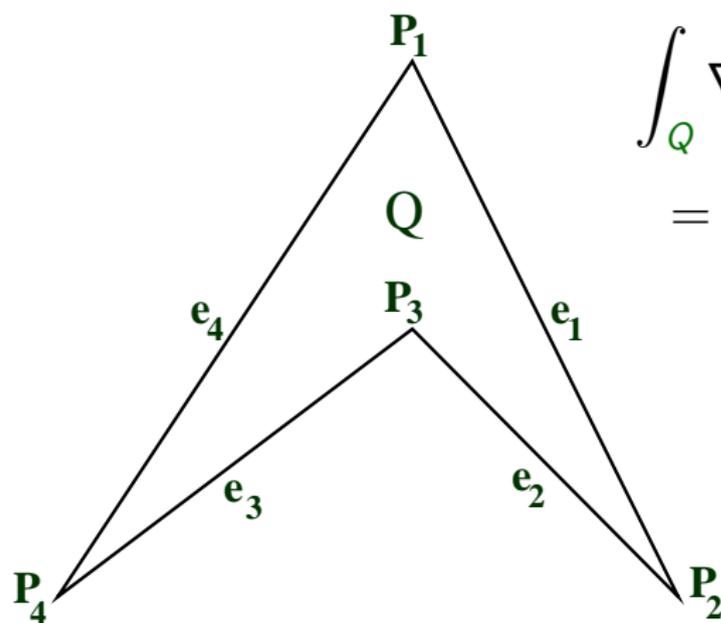
The Local Matrix-3

At this point, for every vector $(\Psi_1, \Psi_2, \Psi_3, \Psi_4)$ of nodal values (of an ideal function Ψ that we don't have), we can compute

$$\int_Q \nabla_x \cdot \nabla \Psi \, dQ \quad \text{and} \quad \int_Q \nabla_y \cdot \nabla \Psi \, dQ$$

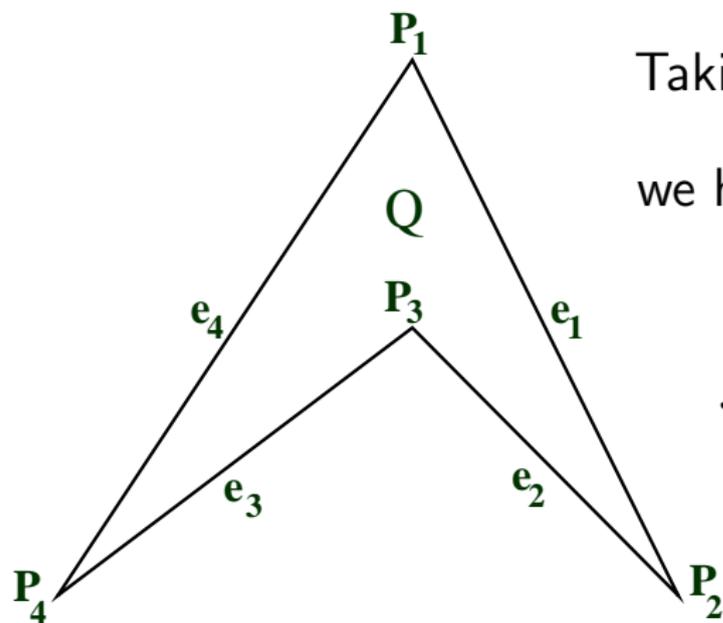
using only the nodal values $(\Psi_1, \Psi_2, \Psi_3, \Psi_4)$.

The Local Matrix-4



$$\begin{aligned}\int_Q \nabla_x \cdot \nabla \Psi \, dQ &= \int_{\partial Q} n_x \Psi = \\ &= |e_1| n_x^1 (\Psi_1 + \Psi_2) / 2 + \dots \\ &\dots + |e_4| n_x^4 (\Psi_4 + \Psi_1) / 2\end{aligned}$$

The Local Matrix-5



Taking $\psi_A := (1, -1, 1, -1)$

we have $\int_Q \nabla x \cdot \nabla \psi_A dQ =$

$$\int_Q \nabla y \cdot \nabla \psi_A dQ = 0$$

The Local Matrix-6

	"1"	"X"	"Y"	" Ψ_A "
"1"	0	0	0	0
"X"	0	$ Q $	0	0
"Y"	0	0	$ Q $	0
" Ψ_A "	0	0	0	

We are left with the evaluation

$$\text{of } \int_Q |\nabla \psi_A|^2 dQ =: \alpha$$

which depends on the values
of ψ_A inside

The Miracle

	"1"	"X"	"Y"	" Ψ_A "
"1"	0	0	0	0
"X"	0	$ Q $	0	0
"Y"	0	0	$ Q $	0
" Ψ_A "	0	0	0	α

Believe it or not, you can take **any positive value** for α , and there will be a function Ψ_A such that your scheme actually comes from a **function-based** approach (using linear functions plus Ψ_A)

In a more general geometry, the role of α is played by a symmetric and positive definite matrix.

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- You **shamelessly cook up** the remaining part of the local matrix.
- You find that you are using a **function based** approach.