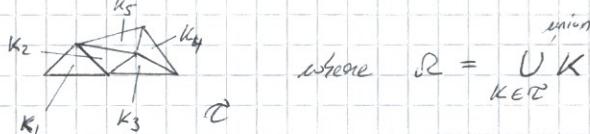


Test

- In 1D we divided our domain Ω into a finite number of points / intervals.
 In 2D our domain Ω represents some area with also confined by the boundary $\partial\Omega$.
 In order to solve our derived variational problems we assume some

Triangulation \mathcal{T}

so our domain Ω is somehow divided into triangles, say

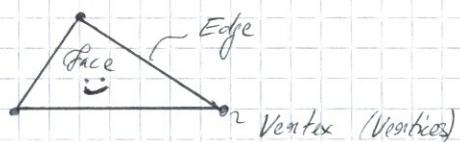


$$\text{where } \Omega = \bigcup_{K \in \mathcal{T}} K$$

This triangulation can take time: Tom de Boer airplane it took a month.

Let's repeat the

Notational language:



elements triangles do not overlap, if two triangles share part of an edge they share the whole edge!

In the case of linear integration i.e. linear base fn's a.k.a. hat-fn's from 1D

let $b = \min_K \text{diam}(K)$ (minimum largest edge of all triangles)

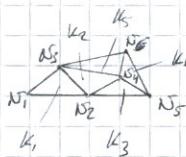
we define a finite dim. subspace $V^h \subset H_0^1(\Omega)$ as

$$V^h = \left\{ v \in H_0^1(\Omega) \mid v \text{ cont., } v|_K \text{ linear} \right\}$$

i.e. base fn



So, how can we represent our triangulation \mathcal{T} , i.e. store it on a computer?



"Naive" approach: store the vertices in terms of coord.

$$\{ (N_1, N_2, N_3) \}$$

$$(N_2, N_4, N_3)$$

$$(N_2, N_5, N_4)$$

...

'where each vertex needs \geq double numbers to store coords'

{

i.e. $\{ (10.2, 0.3), (10.1, \dots), \dots \}$ BUT comparing floating pt. numbers is generally speaking a bad idea

Note:
 counter-
 clockwise
 order
 of vertices

take into acc.
 precision value

Instead we make an array of all vertices

$$Z = (N_1, N_2, \dots, N_n)$$

and store the indices from all triangles in another list

$$\begin{aligned} T = & ((1, 2, 3), \\ & (2, 4, 3), \\ & (2, 5, 4), \\ & \dots \\ &) \end{aligned}$$

The idea is, that then we can access all the information uniformly
i.e. we loop over all triangles and simply 'read out' the vertices
we need i.o. to compute our solution contributions.

Then $Z(q, i)$ is the q^{th} component of the vertex N_i

(Array $Z : 2 \times N$)

and $T(x, m)$ is the x^{th} vertex of the triangle K_m

(Array $T : 3 \times 9$)

\Rightarrow ~~16~~

$Z(q, T(x, m)) = q^{\text{th}}$ component of the x^{th} vertex of K_m

uncertainty

Quadratic Integration

Instead of a linear (1^{st} order) approach to
the Finite El. Method we could also develop a 2^{nd} order
finite el. method.

Of course we can do so in 1 and 2 dimensions.

Say we consider the problem in Series 5 Ex. 2

x contains (x_1, x_2)

2D vector

$$\begin{cases} -\Delta u = f(x) & \text{in } \Omega \subset \mathbb{R}^2 \\ u(x) = 0 & \text{on } \partial\Omega \end{cases}$$

where $f \in L^2(\Omega)$

Note: Norms:

$$\|v\|_{H_0^1(\Omega)} = \left(\int |\nabla v|^2 dx \right)^{1/2}; \|v\|_{L^2(\Omega)} = \left(\int |v|^2 dx \right)^{1/2}$$

hence

$$(u, v)_{H_0^1(\Omega)} = \int_{\Omega} \langle \nabla u, \nabla v \rangle dx;$$

$$(f, v)_{L^2(\Omega)} = \int_{\Omega} f(x) \cdot v(x) dx;$$

Then our Variational Formulation reads:

Find $u \in H_0^1$ st. $\forall v \in H_0^1$ it holds that

$$(u, v)_{H_0^1(\Omega)} = (f, v)_{L^2(\Omega)}$$

Since we now solve this by means of quadratic finite elements on our triangular mesh, we consider the finite-dim. subspace of $H_0^1(\Omega)$, namely

$$V_h = \{ \omega: \Omega \rightarrow \mathbb{R} \mid \omega \text{ is cont., } \frac{\partial \omega}{\partial \nu} = 0 \text{ p. } \text{edge } K \text{ is a } 2^{\text{nd}} \text{ order polynomial } \forall K \in \mathcal{E} \},$$

from BC condition

In general,

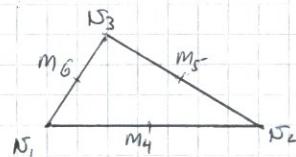
$$\text{we assume } \omega(x, y) = a_{0,0} + a_{1,0}x + a_{0,1}y + a_{1,1}xy + a_{2,0}x^2 + a_{0,2}y^2$$

where $a_{ij} \in \mathbb{R}$.

So, now we have basis functions of two variables and 6 coeff. defined on the whole element K (triangle).

For simplicity we decompose these base fns into two "base fns"

- Contribution of the vertices of K
- Contribution of the mid-points of each edge of K

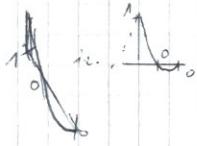


and hence (in the notation which is used in Series 5)

$$x_j \Rightarrow N_j^{\text{vector}}$$

$$\delta_i(x_j) := \begin{cases} 1 & i=j \\ 0 & \text{otherwise} \end{cases} + \{ \delta_i \}_{i=0}^{N_v-1} \quad (N_v \# \text{of vertices})$$

$$\psi_i(m_j) := \begin{cases} 1 & i=j \\ 0 & \text{otherwise} \end{cases} + \{ \psi_i \}_{i=0}^{N_E-1} \quad (N_E \# \text{of edges})$$



δ_i, ψ_i
defined globally

Usually, we are then to determine the coefficients (6 per base fn)
and of course (since we are lazy and rather ~~lazy~~ ~~fast~~)
we do that on the reference element,
where we establish six suitable ops

- 3 from the vertices
- 3 from the midpoints

Similarly, in 1D we add contributions at all $x_{i+\frac{1}{2}}$
and an acc. base fn. To define these fns you are given (or you define) restrictions

Now, we put these contributions back together to get our shape fns

$$\begin{aligned} N &\text{ the \# of base fns} \\ &N_v + N_E \end{aligned}$$

\Rightarrow since our vari. problem holds
for all $v \in V$ it particularly holds
for each $v = \varphi_i$

and since we know that

$$\{ \varphi_i \}_{i=1}^N \text{ form a basis for } V$$

we can write our approx. solution

$$u_N(x) = \sum_{i=1}^N \mu_i \varphi_i^N(x)$$

where

$$\mu = \{ \mu_i \}_{i=0}^{N-1}$$

comprises all coefficients "weighing" the shape fns' contribution.

Our var. formulation reads:

$$(\nabla u_N, \nabla \varphi_i) = (\varphi_i, \varphi_i) \quad \forall i \in \{0, \dots, N-1\}$$

and inserting our approx. solution yields:

$$\begin{aligned} &(\nabla \left(\sum_j \mu_j \varphi_j \right), \nabla \varphi_i) = (\varphi_i, \varphi_i) \quad \text{, where } \mu_i \text{ are const. w.r.t. } x \\ &= \left(\sum_j \mu_j (\nabla \varphi_j, \nabla \varphi_i) \right) \\ &= \sum_j \mu_j (\nabla \varphi_j, \nabla \varphi_i) \end{aligned}$$

Assembly

So, now we get this eqn for each $i=1, \dots, N$

elements

$$N = N_V + N_E$$

$$\sum_{j=1}^N \mu_j \underbrace{(\nabla \phi_j, \nabla \phi_i)}_{A_{ij}} = \underbrace{(\mathbf{f}_i, \phi_i)}_{F_i}$$

hence we get the linear system of eqns

$$A\mu = F, \text{ where } \mu \text{ is the vector of unknowns } \{\mu_i\}_{i=0}^{N-1}$$

From this we see that we can get lucky i.e. in the 1D case, where we only have contributions from directly neighbouring base fns or unlucky:



In order to further discuss the importance of these constraints I continue by repeating the assembly process in more detail before solving an 'example' exercise where we make use of these concepts in a more 'exam-like' landscape.

Going back to our system of equations, where we assumed a triangulation s.t. triangles do not overlap (except for vertices and edges) and if, when we're edges, we had the matrix element

$$(u, v) = u \cdot v$$

dot prod.

u, v_1, v_2, \dots

remember x is a vector

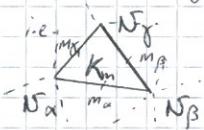
$$A_{ij} = \int (\nabla \phi_i, \nabla \phi_j) dx \quad \text{and} \quad \Omega = \bigcup_{m=1}^M K_m$$

$$= \sum_{m=1}^M \int_{K_m} (\nabla \phi_i, \nabla \phi_j) dx$$

Integration

and it makes sense to loop over all triangles, when computing the entries.

For a specific triangle



The integral is only non-zero if i, j are to be a vertex or midpoint of K . \rightarrow in the linear case i.e. 1st order polynomial basis fns only vertices

$$\int_{K_m} (\nabla \varphi_i, \nabla \varphi_j) \neq 0 \quad \text{only iff } i, j \in K_m \quad \text{if } i, j \in K_m$$

which immediately solidifies this new perspective:

- instead of looping over the basis functions, we loop over triangles K_1, \dots, K_M
- for each triangle we compute the integral for all pairs where $i, j \in K_m$

Of course, to make things easier (remember, we are lazy)
we project our triangles and base functions onto a reference element and
compute our integrals there.

Diego introduced this 'mapping' very nicely last week, so I will
allow myself to skip that step.

Now, for a triangle K (a general triangle in our mesh)
let us define the

$$\begin{aligned} \text{local basis functions } \varphi_\alpha^k & \text{ for } \alpha = \begin{cases} 1, 2, 3 & \text{linear int.} \\ 1, 2, 3, 4, 5, 6 & \text{quadratic int.} \end{cases} \\ \text{so flat } \varphi_\alpha^k(N_i) &= \begin{cases} 1 & T(x, k) = i \\ 0 & \text{else} \end{cases} \end{aligned}$$

where i is a global variable (all nodes).

~~at three vertices~~

You remember, remember, in the linear case (i.e. linear basis functions)
we get 3×3 contributions to the matrix A , so 9 for each triangle.
→ 3 combinations of inner products
Similarly we get $6 \times 6 \sim 36$ in the quadratic case.
→ 6 combinations of inner products

$$\int_K (\nabla \varphi_\alpha^k, \nabla \varphi_\beta^k) dx, \quad \alpha, \beta \in [1, \dots, 6]$$

And we can define the

local stiffness matrix A^k

$$A_{\alpha, \beta}^k = \int_K (\nabla \varphi_\alpha^k, \nabla \varphi_\beta^k) dx, \quad \alpha, \beta = 1, \dots, 6$$

and since the dot prod. is indifferent to the order

A^k is symmetric

→ so no need to compute all entries
(we can skip half the computations)

Let's say some vertex appears twice, hence we get two triplets in an ejon matrix with the same indices,

$$\text{i.e. } \text{Triplet } (i, j, 7.67)$$

$$\text{Triplet } (i, j, e^{x^2})$$

$$\Rightarrow A_{ij} = 7.67 + e^{x^2}.$$