

# Numerical Simulation Methods in Geophysics, Part 5: Finite Elements

## 1. MGPy+MGIN

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# Recap Finite Differences

- elliptic (Poisson) or parabolic PDE problems
- replace partial differential operators  $\partial$  by finite differences  $\Delta$
- transfer PDE into a matrix-vector equation  $\mathbf{A}\mathbf{u} = \mathbf{b}$
- finite-difference stencil spatial or temporal
- spatial derivative  $\Rightarrow$  system matrix  $\mathbf{A}$ , temporal  $\Rightarrow$  identity matrix  $\mathbf{I}$
- time-stepping explicit, implicit or mixed (stable & accurate)
- accuracy depends on discretization & parameter contrast

# The Finite Element Method

# History and background

- [1943] Courant: Variational Method
- [1956] Turner, Clough, Martin, Topp: Stiffness
- [1960] Clough: Finite Elements for static elasticity
- [1970-80] extension to structural, thermic and fluid dynamics
- [1990] computational improvements
- now main method for almost all PDE types

Geophysics: Poisson equation in 1970s, revival in 1990s and predominant from 2000s up to now

# Variational formulation of Poisson equation

$$-\nabla \cdot a \nabla u = f$$

Multiplication with test function  $w$  and integration  $\Rightarrow$  weak form

$$-\int_{\Omega} w \nabla \cdot a \nabla u d\Omega = \int_{\Omega} w f d\Omega$$

$$\nabla \cdot (b\mathbf{c}) = b \nabla \cdot \mathbf{c} + \nabla b \cdot \mathbf{c}$$

$$\int_{\Omega} a \nabla w \cdot \nabla u d\Omega - \int_{\Omega} \nabla \cdot (w a \nabla u) d\Omega = \int_{\Omega} w f d\Omega$$

# Variational formulation of Poisson equation

$$\int_{\Omega} a \nabla w \cdot \nabla u d\Omega - \int_{\Omega} \nabla \cdot (w a \nabla u) d\Omega = \int_{\Omega} w f d\Omega$$

use Gauss' law  $\int_{\Omega} \nabla \cdot \mathbf{A} = \int_{\Gamma} \mathbf{A} \cdot \mathbf{n}$

$$\int_{\Omega} a \nabla w \cdot \nabla u d\Omega - \int_{\Gamma} aw \nabla u \cdot \mathbf{n} d\Gamma = \int_{\Omega} fw d\Omega$$

Let  $u$  be constructed by shape functions  $v$ :  $u = \sum_i u_i v_i$

$$\int_{\Omega} a \nabla w \cdot \nabla v_i d\Omega - \int_{\Gamma} aw \nabla v_i \cdot \mathbf{n} d\Gamma = \int_{\Omega} fw d\Omega$$

# Galerkin's method

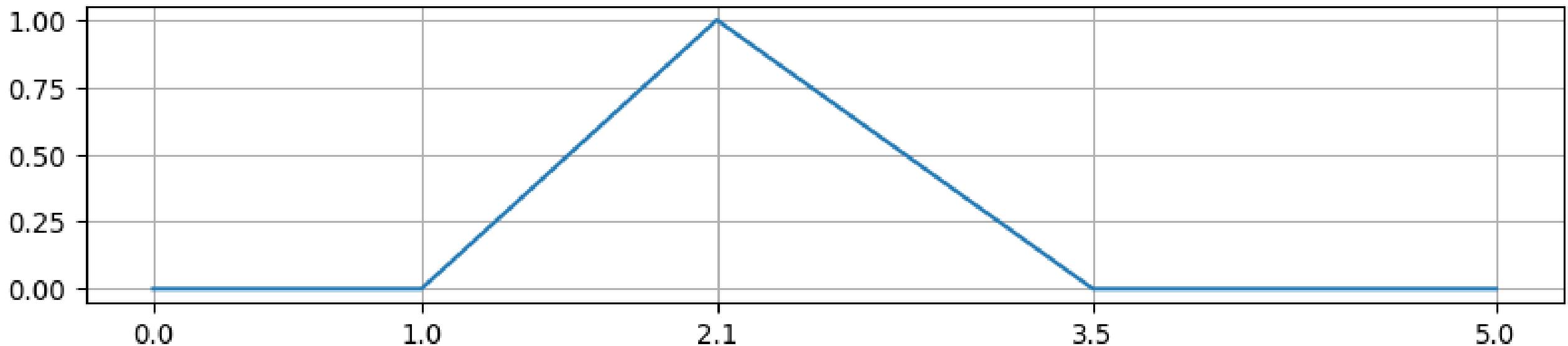
$$\int_{\Omega} a \nabla w \cdot \nabla v_i d\Omega - \int_{\Gamma} aw \nabla v_i \cdot \mathbf{n} d\Gamma = \int_{\Omega} f w d\Omega$$

Test functions the same as shape (trial) functions  $w \in v_i$

$$\int_{\Omega} a \nabla v_j \cdot \nabla v_i d\Omega - \int_{\Gamma} av_j \nabla v_i \cdot \mathbf{n} d\Gamma = \int_{\Omega} f v_j d\Omega$$

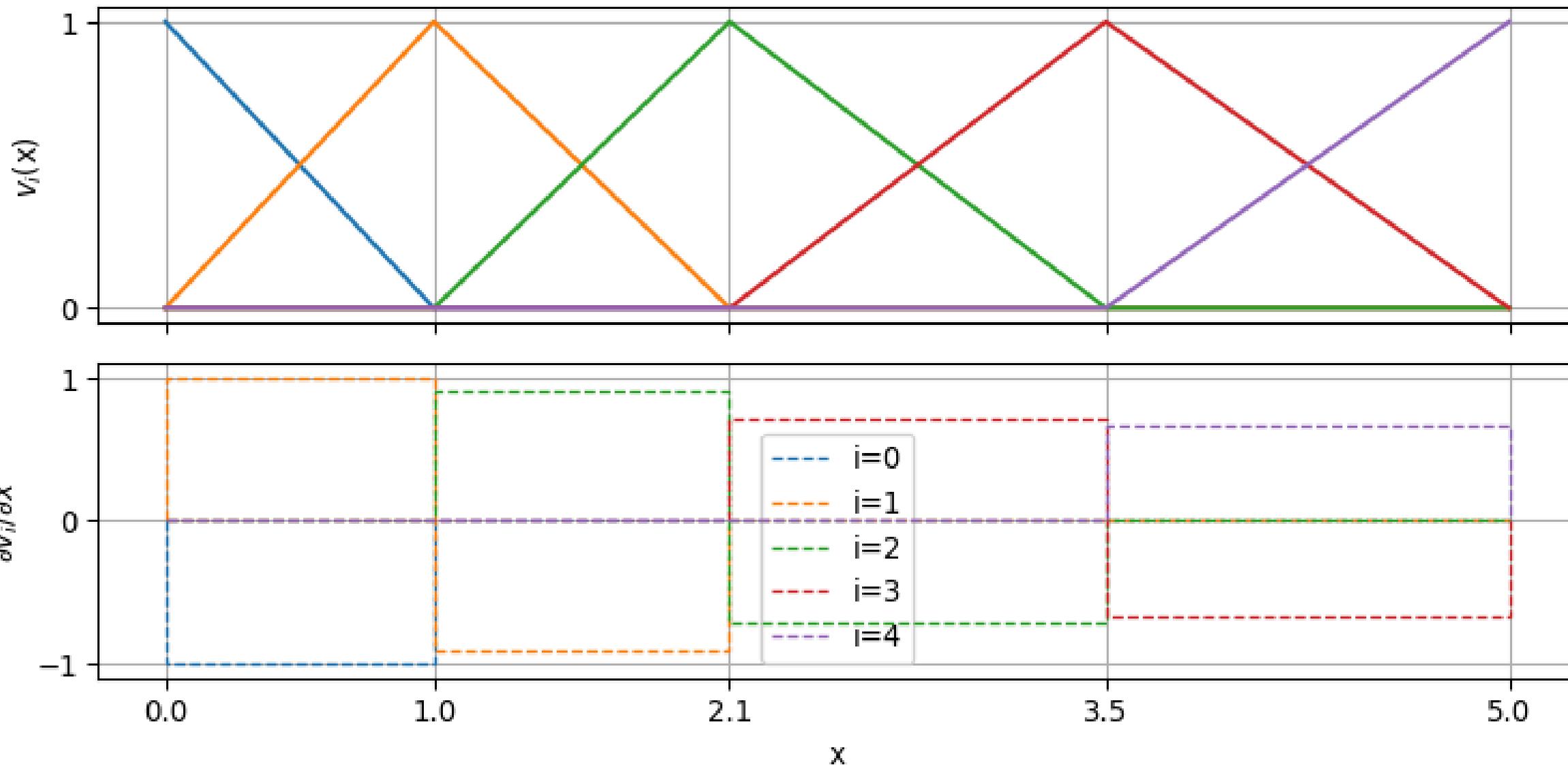
- choose  $v_i$  so that  $\nabla v_i$  is simple and  $\nabla v_i \cdot \nabla v_j$  mostly 0
- divide subsurface in sub-volumes  $\Omega_i$  with constant  $a_i$  (&  $\nabla v_j$ )

# FE for 1D Poisson PDE



Every node carries a hat

# 1st-order nodal shape functions (hats)



# Gradients for hat functions

- every element is surrounded by two nodes “carrying” a hat.
- the gradients  $v'_i$  are piece-wise constant  $\pm 1/\Delta x_i$
- neighboring functions  $v_i$  &  $v_{i+1}$  only meet between  $x_i$  &  $x_{i+1}$

$$\int_{\Omega} a \nabla v_i \cdot \nabla v_{i+1} d\Omega = \int_{x_i}^{x_{i+1}} a_i v'_i v'_{i+1} d\Omega = -\frac{a_i}{\Delta x_i^2} \Delta x_i = -\frac{a_i}{\Delta x_i}$$

$$\int_{x_{i-1}}^{x_{i+1}} a v'_i v'_i d\Omega = \frac{a_{i-1}}{\Delta x_{i-1}^2} \Delta x_{i-1} + \frac{a_i}{\Delta x_i^2} \Delta x_i = \frac{a_{i-1}}{\Delta x_{i-1}} + \frac{a_i}{\Delta x_i}$$

# Integration

Let's write the equation for the first and second nodes in 1D

$$\int_{x_0}^{x_1} u_0 a_0 v'_0 v'_0 + \int_{x_0}^{x_1} u_1 a_1 v'_0 v'_1 = \int_{x_0}^{x_1} v_0 f$$

$$\int_{x_0}^{x_1} u_0 a v'_0 v'_1 + \int_{x_0}^{x_2} u_1 a v'_1 v'_1 + \int_{x_1}^{x_2} a u_2 v'_2 v'_1 = \int_{x_0}^{x_2} v_1 f$$

$$u_{i-1} a_{i-1} \int_{x_{i-1}}^{x_i} v'_i v'_{i-1} + u_i a_{i-1} \int_{x_{i-1}}^{x_i} v'_i v'_i + u_i a_i \int_{x_i}^{x_{i+1}} v'_i v'_i + u_{i+1} a_i \int_{x_i}^{x_{i+1}} v'_i v'$$

# The stiffness matrix

Matrix integrating gradients of base functions for neighbors with  $a$

$$\mathbf{A}_{i,i+1} = -\frac{a_i}{\Delta x_i^2} \cdot \Delta x_i = -\frac{a_i}{\Delta x_i}$$

$$A_{i,i} = \int_{\Omega} a \nabla v_i \cdot \nabla v_i d\Omega = -A_{i,i+1} - A_{i+1,i}$$

$\Rightarrow$  matrix-vector equation  $\mathbf{A}\mathbf{u} = \mathbf{b}$  with bending&shear stiffness in  $\mathbf{A}$

# Boundary conditions

second term

$$-\int_{\Gamma} av_j \nabla v_i \cdot \mathbf{n} d\Gamma$$

reads in 1D as

$$[av_i v'_j]_{x_0}^{x_N} = a_{N-1} u_N v'_N - a_0 u_0 v'_0$$

⇒ Homogeneous Neumann BC ( $v'_0 = 0$ ) are automatically implemented

# Right-hand side vector

The right-hand-side vector  $b = \int v_i f d\Omega$  also scales with  $\Delta x$

$$\text{e.g. } f = \nabla \cdot \mathbf{j}_s \Rightarrow b = \int v_i \nabla \cdot \mathbf{j}_s d\Omega = \int_{\Gamma} v_i \mathbf{j}_S \cdot \mathbf{n}$$

(system identical to FD for  $\Delta x=1$ )

## Difference of FE to FD

Any source function  $f(x)$  can be integrated on the whole space!

# Solution

$\mathbf{u}$  holds the coefficient  $u_i$  creating  $u(x) = \sum u_i v_i(x)$

## Difference of FE to FD

$u$  is described on the whole space and approximates the solution, not the PDE!

Hat functions:  $u_i$  potentials on nodes,  $u$  piece-wise linear

## Generality of FE

Arbitrary base functions  $v_i$  can be used to describe  $u$

# Method of weighted residuals

PDE  $\mathcal{L}(u) = f \Rightarrow$  approximated by  $u_h$

residual  $R = L_h(u) - f$  to be minimized, integrating over modelling domain

$$\int_{\Omega} w R d\Omega = \int_{\Omega} w \mathcal{L}(u_h) d\Omega - \int_{\Omega} w f d\Omega = 0$$

with approximation  $u_h(\mathbf{r}) = \sum_j^M u_j \mathbf{v}_j(\mathbf{r})$

( $\mathbf{v}$  basis / shape functions,  $\mathbf{w}$  test / trial functions)

# Bilinear form for Poisson equation

Solve  $\mathbf{Ax} = \mathbf{b}$  with  $A_{ij} = (\nabla v_i, a \nabla v_j)$  and  $b_i = (\mathbf{v}_i, f)$ , where

$$(\mathbf{a}, \mathbf{b}) = \int_{\Omega} \mathbf{a} \cdot \mathbf{b} \, d\Omega = \sum_{c=i}^M \int_{\Omega_c} \mathbf{a} \cdot \mathbf{b} \, d\Omega_c$$

Solve the integrals either analytically or numerically

# Coordinate transformation

1D: local coordinate  $\xi = \frac{x-x_i}{x_{i+1}-x_i}$  (0..1)

$$u(\xi) = c_1 + c_2\xi$$

$$u_0 = u(0) = c_1, u_1 = u(1) = c_1 + c_2 \Rightarrow c_2 = u_1 - u_0$$

$$\Rightarrow u(\xi) = u_0 + \xi(u_1 - u_0) = u_0(1 - \xi) + u_1\xi = u_0v_0 + u_1v_1$$

# Quadratic elements

$$u(\xi) = c_1 + c_2\xi + c_3\xi^2$$

nodes at  $x_0, x_{1/2}, x_1$

$$u_i = u(0) = c_1, u_1 = c_1 + c_2 + c_3$$

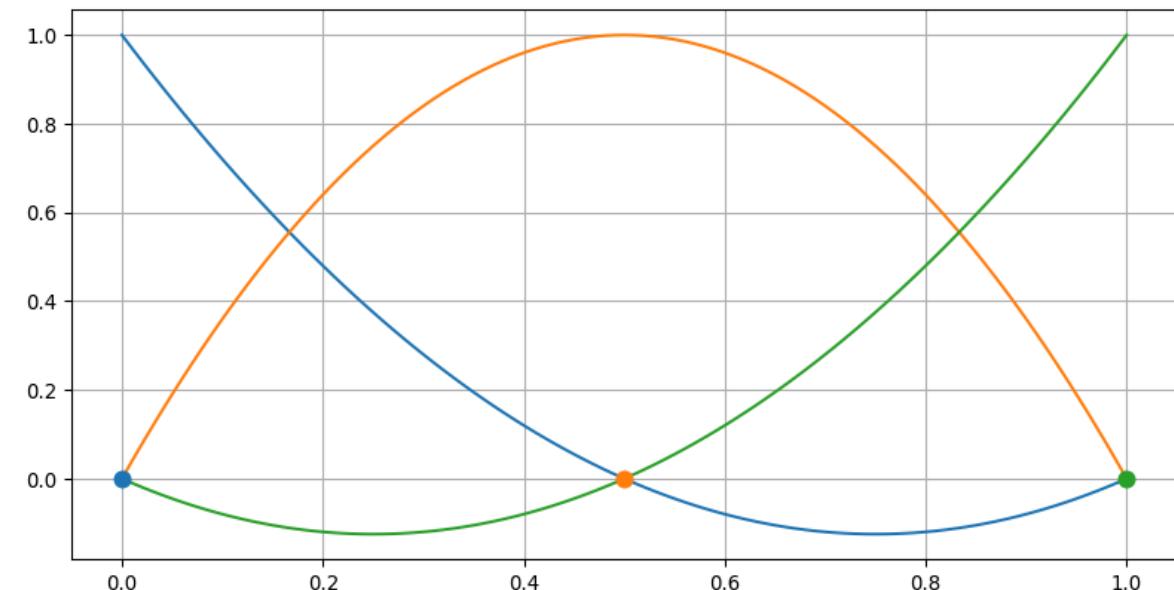
$$u_{1/2} = c_1 + c_2/2 + c_3/4$$

$$u(\xi) = u_0(1 - 3\xi + 2\xi^2) + u_{1/2}(4\xi - 4\xi^2) + u_1(-\xi + 2\xi^2)$$

# Quadratic elements

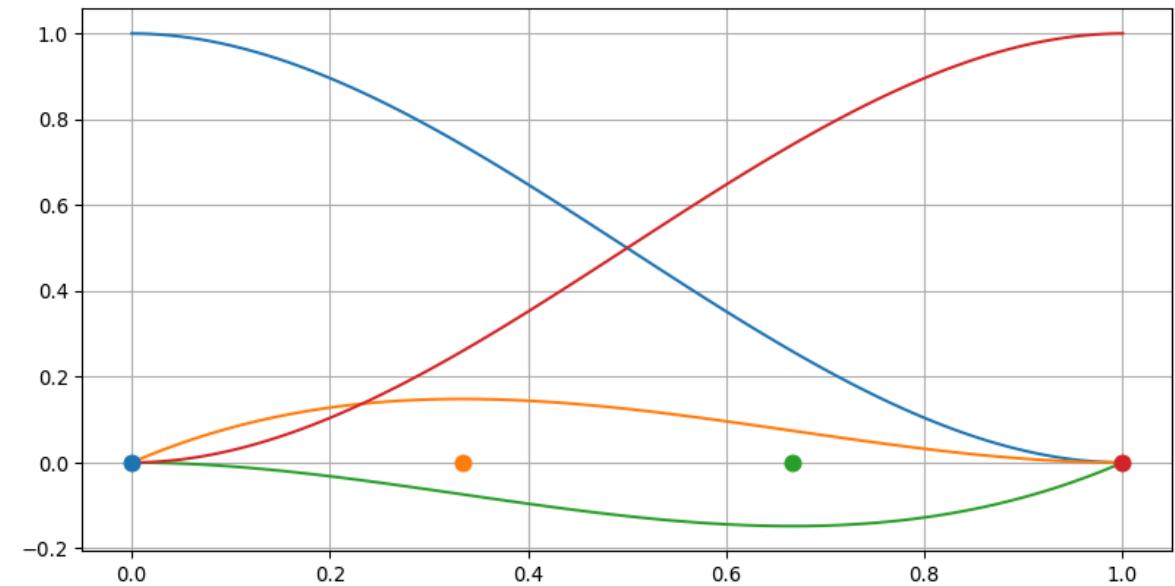
$$u(\xi) = u_0(3\xi + 2\xi^2) + u_{1/2}(4\xi - 4\xi^2) + u_1(-\xi + 2\xi^2)$$

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 x=np.linspace(0, 1, 101)
4
5 # Plot velocity distribution.
6 plt.plot(x, 1-3*x+2*x**2)
7 plt.plot(x, 4*x-4*x**2)
8 plt.plot(x, -x+2*x**2)
9 plt.plot(0, 0, "o", color="C0", ms=8)
10 plt.plot(0.5, 0, "o", color="C1", ms=8)
11 plt.plot(1, 0, "o", color="C2", ms=8)
12 plt.grid()
```



# Cubic elements

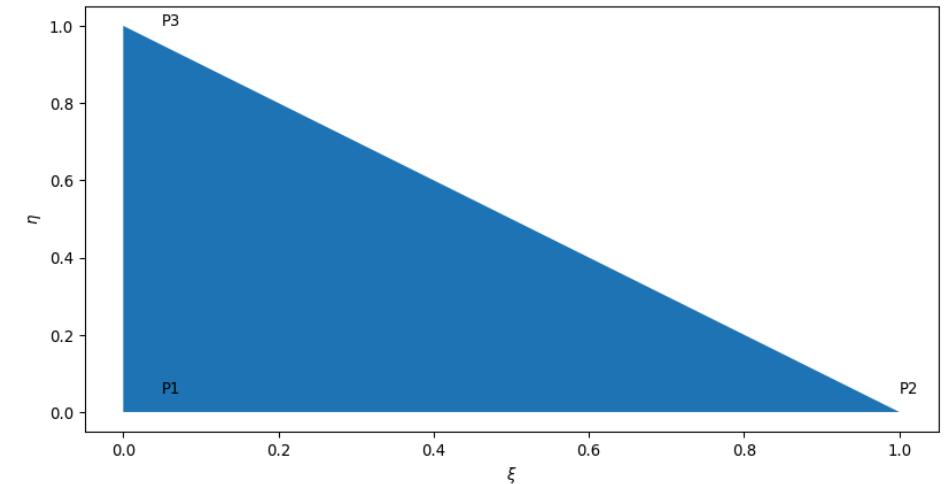
```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 x=np.linspace(0, 1, 101)
4
5 # Plot velocity distribution.
6 plt.plot(x, 1-3*x**2+2*x**3)
7 plt.plot(x, x-2*x**2+x**3)
8 plt.plot(x, -x**2+x**3)
9 plt.plot(x, 3*x**2-2*x**3)
10 for i in range(4):
11     plt.plot(i/(3), 0, "o",
12               color=f"C{i}", ms=8)
13 plt.grid()
```



# Triangles with linear shape functions

$$x = x_1 + (x_2 - x_1)\xi + (x_3 - x_1)\eta$$

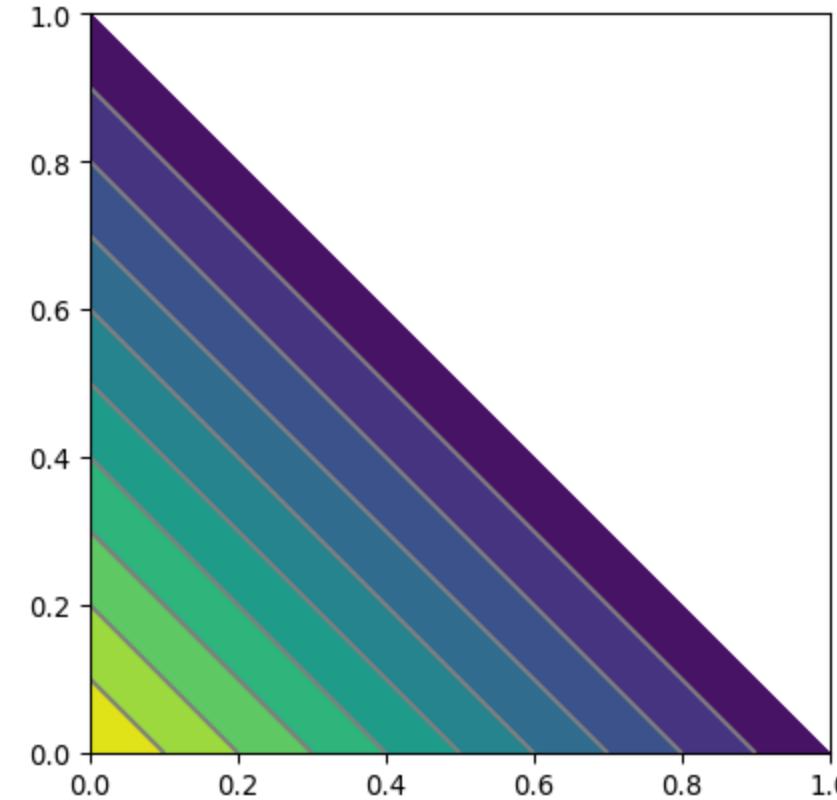
$$y = y_1 + (y_2 - y_1)\xi + (y_3 - y_1)\eta$$



# Triangle

$$u(\xi) = u_1(1 - \xi - \eta) + u_2\xi + u_3\eta$$

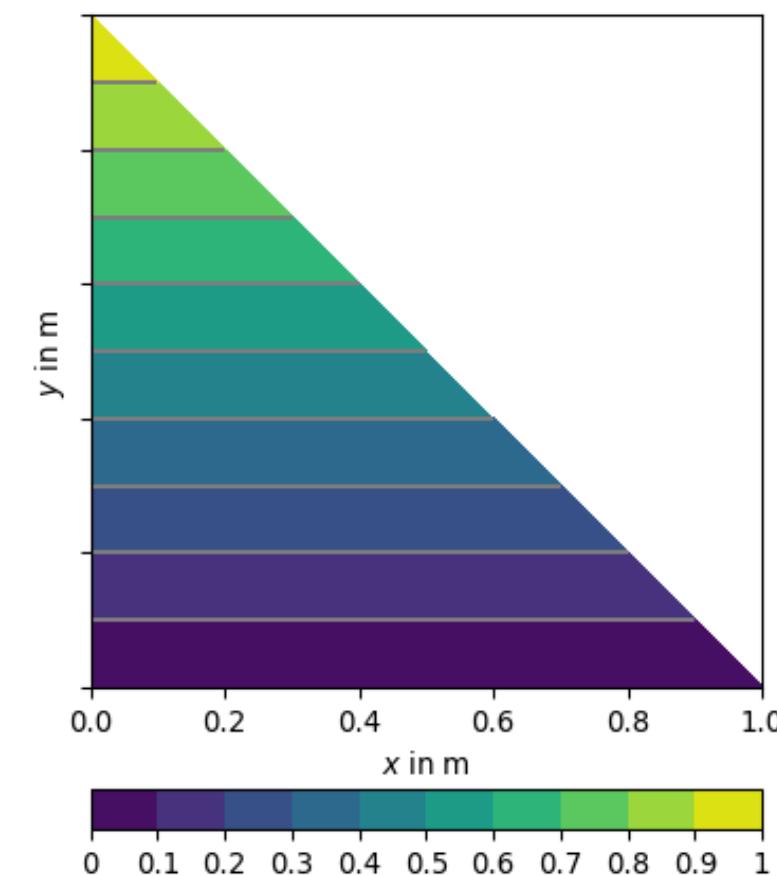
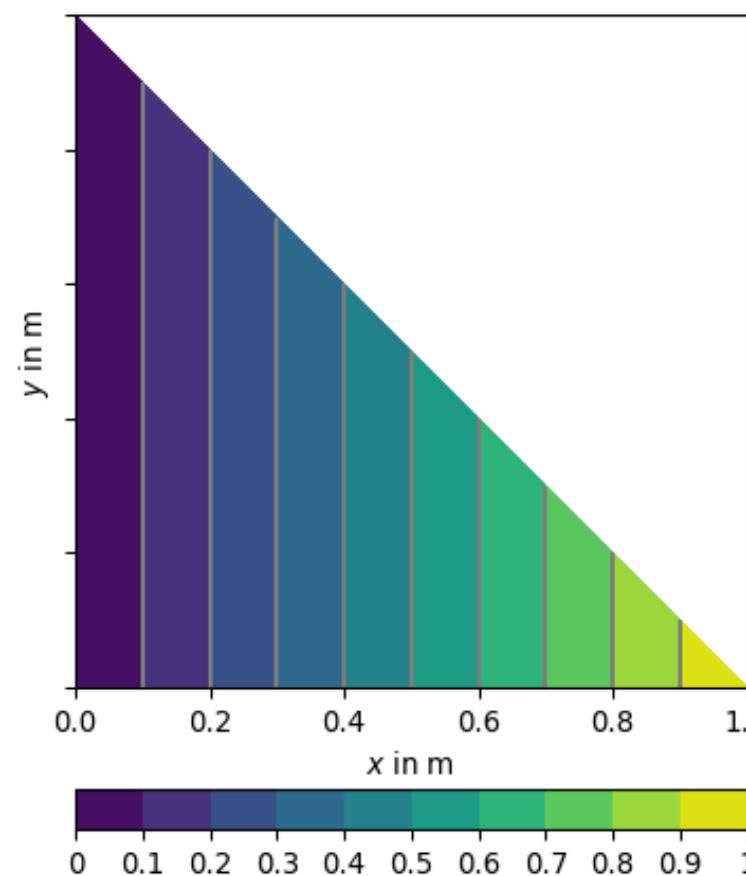
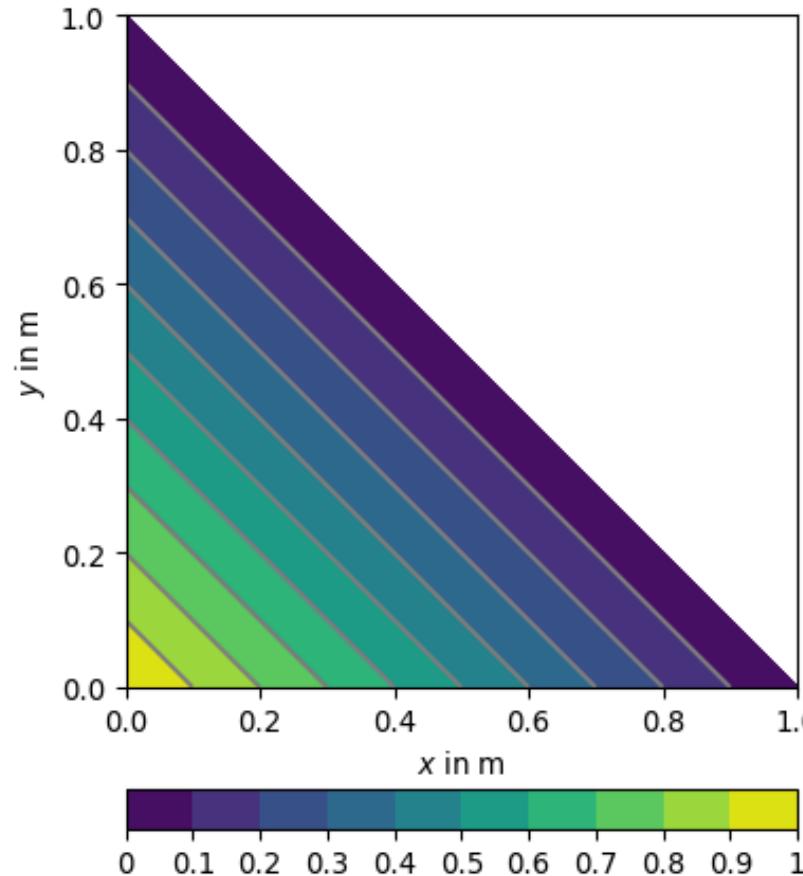
```
1 import pygimli as pg
2 import pygimli.meshTools as mt
3
4 shape = mt.createPolygon(
5     [[0, 0], [1, 0], [0, 1]],
6     isClosed=True)
7 mesh = mt.createMesh(shape, area=0.01)
8 mx = pg.x(mesh)
9 my = pg.y(mesh)
10 # Plot velocity distribution.
11 fig, ax = plt.subplots()
12 pg.show(mesh, 1-mx-my, ax=ax,
13         nLevs=11, label="u");
```



<Figure size 960x480 with 0 Axes>

# Triangle linear shape functions

$$u(\xi) = u_1(1 - \xi - \eta) + u_2\xi + u_3\eta$$



# The general solution

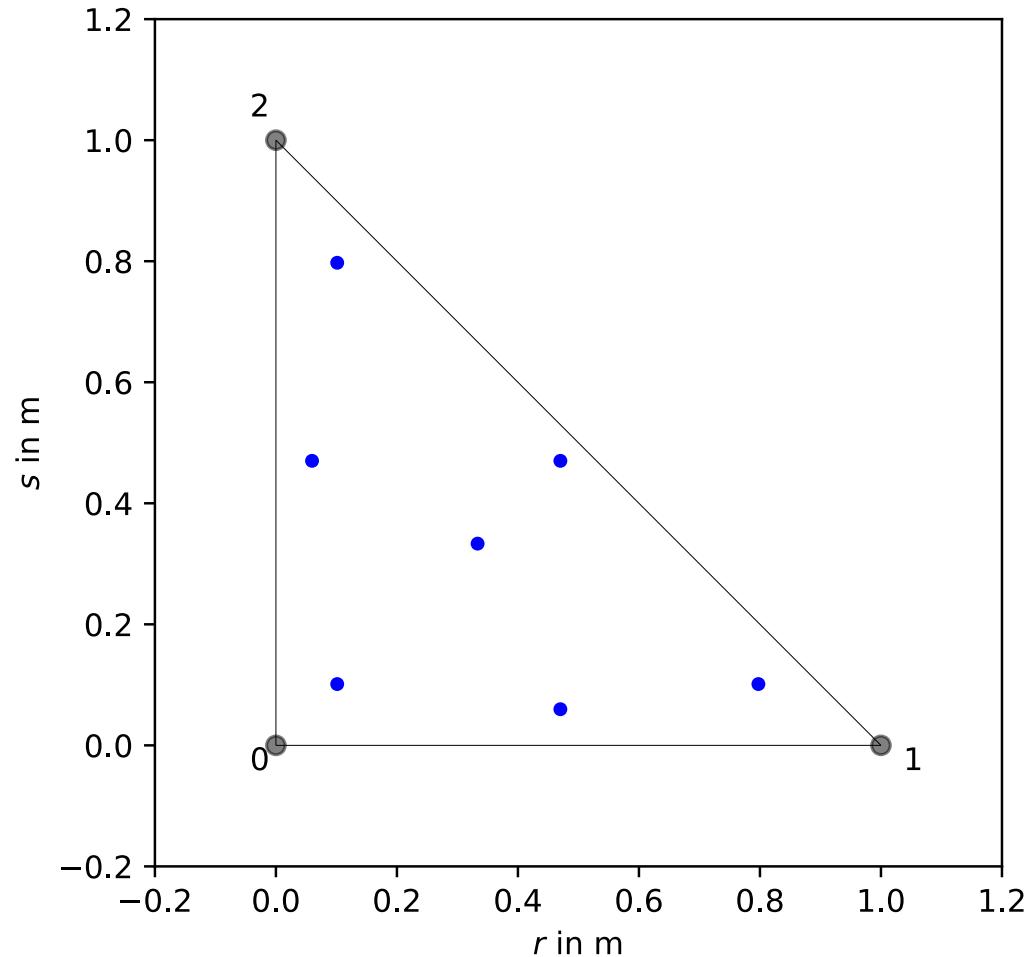
Solving any integral using (Gaussian) quadrature

$$\int g(x)dx \approx \sum_q g(x_q)w_q$$

$$f_i^c = \int_{\Omega_c} v_i f dx \approx \sum_q v(x_q^c) f(x_q^c) w_q^c$$

$$a_{ij}^c = \int_c a_c \nabla v_i \cdot \nabla v_j = \sum a_c \nabla v_i(x_q^c) \cdot \nabla v_j(x_q^c) w_q^c$$

# Gaussian quadrature

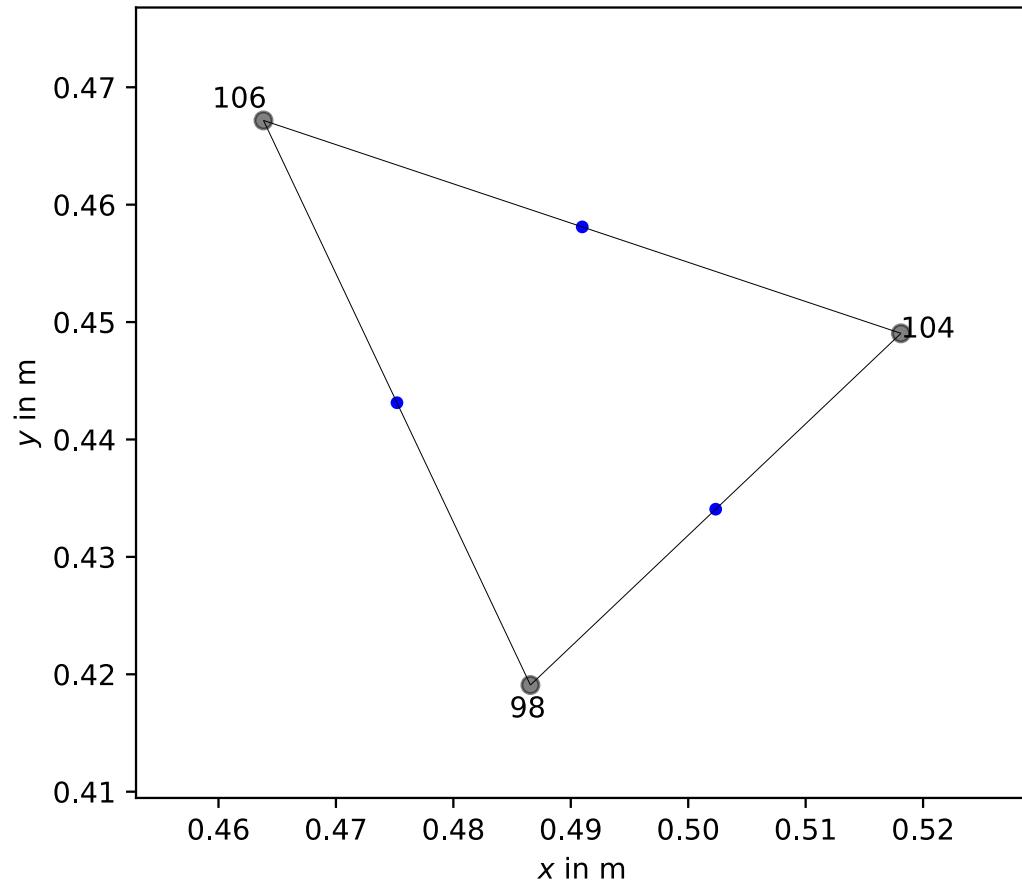


Quadrature points

```
quadratureRules(c.shape(), 5)
```

- optimum quadrature on reference triangle for a given order (5)

# Gaussian quadrature



Quadrature points

`quadratureRules(c, 2)`

- optimum quadrature on arbitrary triangle for order 2

# Verification

## 1. Method of Manufactured Solutions (MMS)

- manufacture a smooth  $u$
- generate  $f$  matching approximation of  $u$

## 2. Method of Exact Solutions (MES)

- find parameters for which an analytic solution exists

## 3. Perform convergence tests for increasingly smaller $h$

- approximation error  $E(h) < Ch^n$  test for some  $h$

# Green's functions

The Green's function  $G$  is the solution for a Dirac source  $\delta$

$$\mathcal{L}G = \delta(\mathbf{r})$$

The solution can then be obtained by convolution

$$u(\mathbf{r}) = G(\mathbf{r} - \mathbf{r}') * f(\mathbf{r}') =$$

# Time-stepping in FE

# Recap time-stepping in FD

**Explicit:**

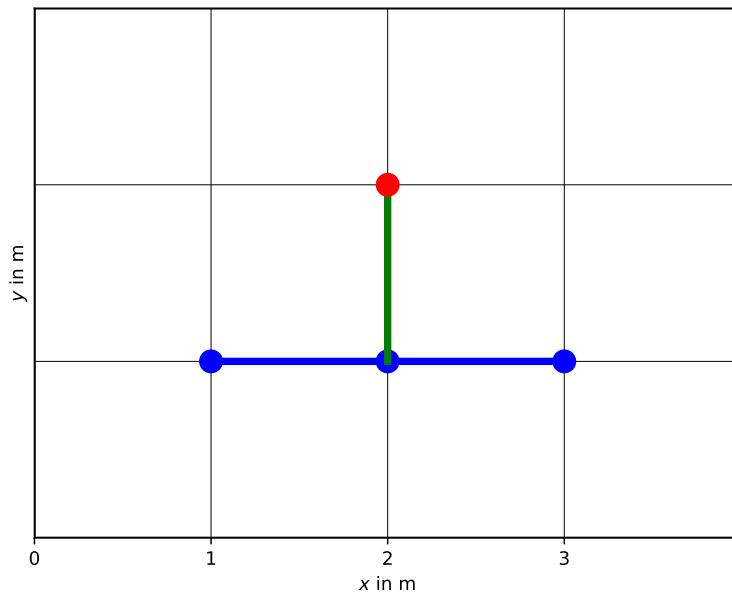
$$\mathbf{u}^{n+1} = (\mathbf{I} - \Delta t \mathbf{A}) \mathbf{u}^n$$

**Implicit:**

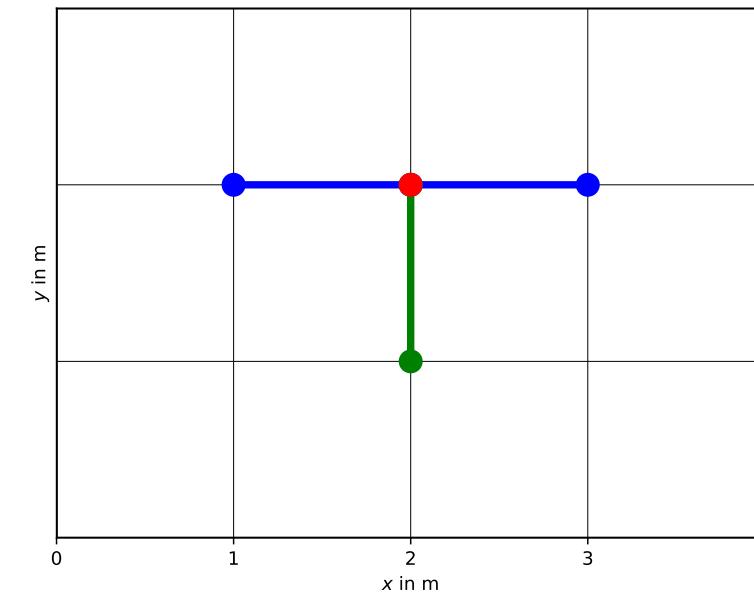
$$(\mathbf{I} + \Delta t \mathbf{A}) \mathbf{u}^{n+1} = \mathbf{u}^n$$

**Mixed:**

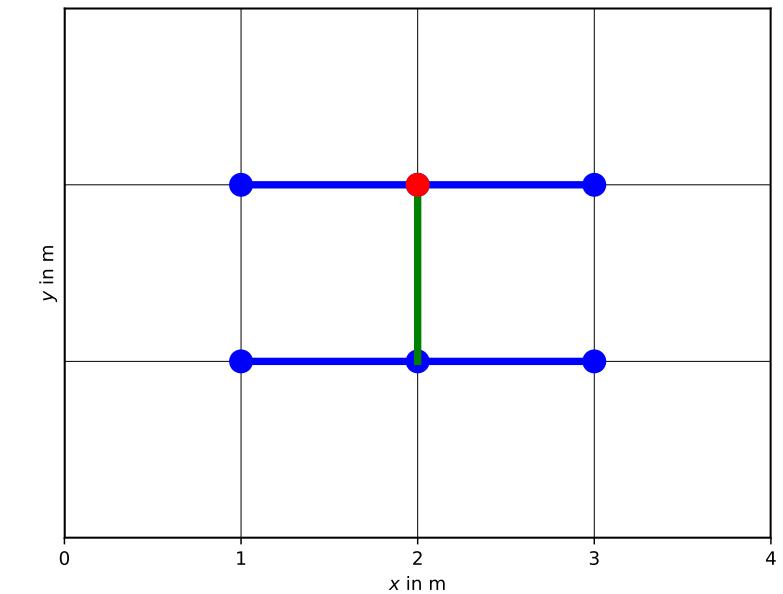
$$(2\mathbf{I} + \Delta t \mathbf{A}) \mathbf{u}^{n+1} = (2\mathbf{I} - \Delta t \mathbf{A}) \mathbf{u}^n$$



Explicit



Implicit



Mixed

# Variational formulation of Diffusion equation

$$\frac{\partial u}{\partial t} - \nabla \cdot a \nabla u = f$$

Finite Difference in Time (NOT in space)

$$\frac{u^{n+1} - u^n}{\Delta t} - \nabla \cdot a \nabla u = f$$

# Variational formulation

$$\frac{u^{n+1} - u^n}{\Delta t} - \nabla \cdot a \nabla u = f$$

Multiplication with test function  $w$  and integration  $\Rightarrow$  weak form

$$1/\Delta t \left( \int_{\Omega} w u^{n+1} d\Omega - \int_{\Omega} w u^n d\Omega \right) - \int_{\Omega} w \nabla \cdot a \nabla u d\Omega = \int_{\Omega} w f d\Omega$$

$$1/\Delta t \left( \int_{\Omega} w u^{n+1} d\Omega - \int_{\Omega} w u^n d\Omega \right) - \int_{\Omega} a \nabla w \cdot \nabla u d\Omega = \int_{\Omega} w f d\Omega$$

# Variational formulation of diffusion equation

$u$  is constructed of shape functions  $\mathbf{v}_i$  that are identical to  $w$

The integral over the Poisson term

$$-\int_{\Omega} a \nabla w \cdot \nabla u d\Omega$$

is represented by  $\mathbf{A}\mathbf{v}$  using the stiffness matrix

$$\mathbf{A}_{i,j} = \int_{\Omega} \sigma \nabla v_i \cdot \nabla v_j d\Omega$$

# Variational formulation of diffusion equation

Weighted integrals over both  $u$  are represented by the mass matrix  $\mathbf{Mv}$

$$\mathbf{M}_{i,j} = \int_{\Omega} v_i \cdot v_j d\Omega$$

explicit method (use  $u^n$ ):  $\mathbf{Mu}^{n+1} = (\mathbf{M} - \Delta t \mathbf{A})\mathbf{u}^n$

implicit method (use  $u^{n+1}$ ):  $(\mathbf{M} + \Delta t \mathbf{A})\mathbf{u}^{n+1} = \mathbf{Mu}^n$

mixed method (mix  $u^n/u^{n+1}$ ):

$(2\mathbf{M} + \Delta t \mathbf{A})\mathbf{u}^{n+1} = (2\mathbf{M} - \Delta t \mathbf{A})\mathbf{u}^n$

# Time-stepping in FE

**Explicit:**

$$\mathbf{M} \mathbf{u}^{n+1} = (\mathbf{M} - \Delta t \mathbf{A}) \mathbf{u}^n$$

**Implicit:**

$$(\mathbf{M} + \Delta t \mathbf{A}) \mathbf{u}^{n+1} = \mathbf{M} \mathbf{u}^n$$

**Mixed:**

$$(2\mathbf{M} + \Delta t \mathbf{A}) \mathbf{u}^{n+1} = (2\mathbf{M} - \Delta t \mathbf{A}) \mathbf{u}^n$$



**Tip**

same as in FD but with FE mass matrix  $\mathbf{M}$  instead of  $\mathbf{I}$

# The mass matrix in 1D

$$\mathbf{M}_{i,j} = \int_{\Omega} v_i \cdot v_j d\Omega$$

$$\mathbf{M}_{i,i+1} = \int_{x_i}^{x_{i+1}} \frac{x_{i+1} - x}{\Delta x_i} \frac{x - x_i}{\Delta x_i} dx = \int_0^1 (1 - \xi) \xi \Delta x_i d\xi$$

$$\Rightarrow \mathbf{M}_{i,i+1} = \Delta x_i \int_0^1 (\xi - \xi^2) = \Delta x_i \left| \frac{1}{2} \xi^2 - \frac{1}{3} \xi^3 \right|_0^1 = \frac{\Delta x_i}{6}$$

# The mass matrix in 1D

$$\mathbf{M}_{i,i} = \Delta x_{i-1} \int_0^1 \xi^2 d\xi + \Delta x_i \int_0^1 (1 - \xi)^2 d\xi$$

$$\mathbf{M}_{i,i} = \Delta x_{i-1} \left| \frac{1}{3} \xi^3 \right|_0^1 - \Delta x_i \left| \frac{1}{3} \xi^3 \right|_1^0$$

$$\Rightarrow \mathbf{M}_{i,i} = \frac{\Delta x_{i-1}}{3} + \frac{\Delta x_i}{3} = 2(\mathbf{M}_{i,i-1} + \mathbf{M}_{i,i+1})$$

$$\Delta x = 1 \quad \Rightarrow \quad [1, 4, 1] \text{ (stiffness was [-1, 2, -1])}$$

# Inner vs. outer nodes

distinguish dofs into inner and outer  $[\mathbf{u}_i, \mathbf{u}_o]^T$

$$\mathbf{A} = \begin{pmatrix} \mathbf{A}_{ii} & \mathbf{A}_{io} \\ \mathbf{A}_{oi} & \mathbf{A}_{oo} \end{pmatrix}$$

$$\mathbf{A} \begin{bmatrix} \mathbf{u}_i \\ \mathbf{u}_o \end{bmatrix} = \begin{bmatrix} \mathbf{f}_i \\ \mathbf{f}_o \end{bmatrix}$$

$$\Rightarrow \mathbf{A}_{ii}\mathbf{u}_i = \mathbf{f}_i - \mathbf{A}_{oi}\mathbf{u}_o$$

# Tasks

1. Write a function computing the FE stiffness matrix for 1D discretization
2. Test it by solving the Poisson equation with  $f = 1$  (analytical solution)
3. Compare with analytical and FD solutions
4. Write a function computing the FE mass matrix for 1D discretization
5. Repeat the time-stepping tasks from FD with FE