

FORMULARY

Basic FEM Analysis

February 10, 2021

1 Linear algebra

1.1 Identity matrix

$$\mathbf{I} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ in 2D or } \mathbf{I} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \text{ in 3D.}$$

1.2 Matrix multiplication

If \mathbf{A} is an $n \times m$ matrix and \mathbf{B} is a $m \times p$ matrix the matrix product \mathbf{AB} is defined to be a $n \times p$ matrix where each i, j entry is given by multiplying the entries A_{ik} (across row i of \mathbf{A}) by entries B_{jk} (down column j of \mathbf{B}), for $k = 1, 2, \dots, m$ and summing the results over k : $(\mathbf{AB})_{ij} = \sum_{k=1}^m A_{ik}B_{kj}$.

1.3 The determinant and inverse of a matrix

If

$$\mathbf{A} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$$

then $\det \mathbf{A} = A_{11}A_{22} - A_{12}A_{21}$ and

$$\mathbf{A}^{-1} = \frac{1}{\det \mathbf{A}} \begin{bmatrix} A_{22} & -A_{12} \\ -A_{21} & A_{11} \end{bmatrix}$$

1.4 The trace of a matrix

The trace of a matrix is the sum of its diagonal components, e.g., $\text{tr} \mathbf{A} = A_{11} + A_{22}$

1.5 Outer product

The outer product $\mathbf{u} \otimes \mathbf{v}$ is equivalent to a matrix multiplication $\mathbf{u}\mathbf{v}^T$, provided that \mathbf{u} is represented as $m \times 1$ column vector and \mathbf{v} as a $n \times 1$ column vector (which makes \mathbf{v}^T a row vector), e.g., $(\mathbf{u} \otimes \mathbf{v})_{ij} = u_i v_j$ or

$$\mathbf{u} \otimes \mathbf{v} = \mathbf{u}\mathbf{v}^T = \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} \begin{bmatrix} v_1 & v_2 & v_3 \end{bmatrix} = \begin{bmatrix} u_1 v_1 & u_1 v_2 & u_1 v_3 \\ u_2 v_1 & u_2 v_2 & u_2 v_3 \\ u_3 v_1 & u_3 v_2 & u_3 v_3 \end{bmatrix}$$

1.6 Matrix contraction (double dot product)

$$\mathbf{A} : \mathbf{B} = \sum_{i,j=1}^n A_{ij} B_{ij}$$

2 Numerical Integration

$$\int_a^b f(x) dx \approx \frac{b-a}{2} (f(a) + f(b))$$

(trapezoidal rule) or

$$\int_a^b f(x) dx \approx (b-a) f((a+b)/2)$$

(midpoint rule), both exact for linear polynomials.

2.1 Gaussian quadrature

$$\int_0^1 f(x) dx \approx \sum_{i=1}^n f(x_i) w_i$$

An n -point Gaussian quadrature yields an exact result for polynomials of degree $2n - 1$ or less. For the range $[0,1]$ the Gauss points and weights are given in Table 1. A function for computing the any order Gauss-points and weights is given in the MATLAB function `gauss.m`¹.

2.2 Gaussian quadrature for triangles

$$\int_K f(\mathbf{x}) dK \approx \text{area}(K) \sum f(\boldsymbol{\xi}_i) w_i$$

where $\text{area}(K)$ is the area of the triangle K , $\boldsymbol{\xi}_i$ are the Gauss points (integration points) and w_i are the Gauss weights. Note that one-point integration is exact for up to linear polynomials, that is linear triangles.

$$\text{area}(K) = \int_K dx dy = \int_{\hat{K}} \det(\mathbf{J}) d\xi d\eta = \sum \det(\mathbf{J}(\boldsymbol{\xi}_i)) w_i d\hat{K}$$

¹Golub-Welsh algorithm taken from L. N. Trefethen, Spectral Methods in Matlab, SIAM, 2000 p.129

Gauss order n	Polynomial degree	w_i	x_i
1	1	1	$1/2$
2	3	$1/2$	$\frac{1}{2} - \frac{\sqrt{3}}{6}$
		$1/2$	$\frac{1}{2} + \frac{\sqrt{3}}{6}$
3	5	$5/18$	$\frac{1}{2} - \frac{\sqrt{3}\sqrt{5}}{10}$
		$4/9$	$\frac{1}{2}$
		$5/18$	$\frac{1}{2} + \frac{\sqrt{3}\sqrt{5}}{10}$

Table 1: Gaussian quadrature for $\xi \in [0, 1]$

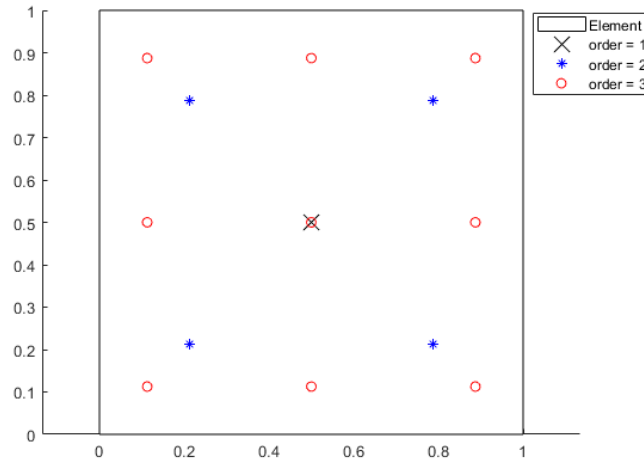


Figure 1: Gauss points for quadrilateral

where $d\hat{K} = 1/2$ is the area of the parametric triangle.

For the quadrature points and weight, see Table 2, for higher order scheme see the function *trigauc.m*.

2.3 Gaussian quadrature for quadrilaterals

$$\int_K f(\mathbf{x}) dK \approx \text{area}(K) \sum f(\mathbf{x}_i) w_i$$

where $\text{area}(K)$ is the area of the quadrilateral K , \mathbf{x}_i are the Gauss points (integration points) and w_i are the Gauss weights. The Gauss-points and weights for each dimension are given by Table 1, see Figure 1 for 2D, use the function **GaussQuadrilateral.m**. Syntax: `[GP,GW] = GaussQuadrilateral(order)`

Number of points n	Polynomial degree	ξ_i	η_i	w_i
1	1	1/3	1/3	1
3	2	1/6	1/6	1/3
		2/3	1/6	1/3
		1/6	2/3	1/3
4	3	1/3	1/3	-9/16
		3/5	1/5	25/48
		1/5	3/5	25/48
		1/5	1/5	25/48
7	4	0	0	1/20
		1/2	0	2/15
		1	0	1/20
		1/2	1/2	2/15
		0	1	1/20
		0	1/2	2/15
		1/3	1/3	9/20

Table 2: Quadrature rules for a triangle in $\xi, \eta \in [0, 1]$

3 Linear map, 1D

Element stretches from x_1 to x_2 , map: $x = \hat{\varphi}_1(\xi)x_1 + \hat{\varphi}_2(\xi)x_2 = (1 - \xi)x_1 + \xi x_2$, $0 \leq \xi \leq 1$. Then $\varphi_i(x) = \varphi_i(\xi(x)) = \hat{\varphi}_i(\xi)$,

$$\frac{dx}{d\xi} = x_2 - x_1 = h, \quad \frac{d\varphi_i}{dx} = \frac{1}{h} \frac{d\hat{\varphi}_i}{d\xi}, \quad \text{and} \quad \int_{x_1}^{x_2} f(x)dx = \int_0^1 f(x(\xi))hd\xi$$

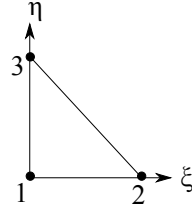
4 Linear map of elements

Isoparametric (linear) map from the reference element \hat{K} in the domain $0 \leq \xi \leq 1$, $0 \leq \eta \leq 1$ to the physical element K , is given by

$$(x, y) = \left(\sum_{i=1}^3 x_i \hat{\varphi}_i(\xi, \eta), \sum_{i=1}^3 y_i \hat{\varphi}_i(\xi, \eta) \right) \quad (1)$$

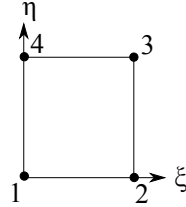
where for the triangle we have the parametric basis functions

$$\hat{\varphi}_1 = 1 - \xi - \eta, \quad \hat{\varphi}_2 = \xi, \quad \hat{\varphi}_3 = \eta,$$



and for the bilinear element we have

$$\hat{\varphi}_1 = (1 - \xi)(1 - \eta), \quad \hat{\varphi}_2 = \xi(1 - \eta), \quad \hat{\varphi}_3 = \eta\xi, \quad \hat{\varphi}_4 = \eta(1 - \xi).$$



Under this map, derivatives transform as

$$\begin{bmatrix} \frac{\partial \varphi_i}{\partial x} \\ \frac{\partial \varphi_i}{\partial y} \end{bmatrix} = \mathbf{J}^{-1} \begin{bmatrix} \frac{\partial \hat{\varphi}_i}{\partial \xi} \\ \frac{\partial \hat{\varphi}_i}{\partial \eta} \end{bmatrix}, \quad \text{where} \quad \mathbf{J} = \begin{bmatrix} \frac{\partial x}{\partial \xi} & \frac{\partial y}{\partial \xi} \\ \frac{\partial x}{\partial \eta} & \frac{\partial y}{\partial \eta} \end{bmatrix},$$

$\frac{\partial x}{\partial \xi} = \frac{\partial \hat{\varphi}}{\partial \xi} \cdot \mathbf{x}_c$, etc., and $\varphi_i(x, y) = \hat{\varphi}_i(\xi, \eta)$ if (1) holds.

Change of variables in integrals:

$$\int_K f(x, y) dx = \int_{\hat{K}} f(x(\xi, \eta), y(\xi, \eta)) \det \mathbf{J} d\xi d\eta.$$

5 FEM, 1D

Consider the following weak form: Find u such that

$$\int_a^b \frac{du}{dx} \frac{dv}{dx} dx + \int_a^b uv dx = \int_a^b fv dx \quad \text{for all } v$$

We want to approximate u and $\frac{du}{dx}$ using Galerkin's method using one or several basis functions φ_i . More precisely we want to find a set of discrete values u_i such that $u \approx \sum_{i=1}^n \varphi_i u_i = \boldsymbol{\varphi} \cdot \mathbf{u}$ and $\frac{du}{dx} \approx \sum_{i=1}^n \frac{d\varphi_i}{dx} u_i = \frac{d\boldsymbol{\varphi}}{dx} \cdot \mathbf{u}$, where $\boldsymbol{\varphi} = [\varphi_1, \varphi_2, \dots, \varphi_n]$ and $\mathbf{u} = [u_1, u_2, \dots, u_n]^T$. We insert the approximation into the weak form and get:

$$\int_a^b \frac{du}{dx} \frac{dv}{dx} dx \approx \int_a^b \sum_{i=1}^n \left(\frac{d\varphi_i}{dx} u_i \right) \frac{d\varphi_j}{dx} dx = \int_a^b \frac{d\varphi_i}{dx} \frac{d\varphi_j}{dx} dx (u_1, \dots, u_n)^T = \int_a^b \left(\frac{d\boldsymbol{\varphi}}{dx} \right)^T \frac{d\boldsymbol{\varphi}}{dx} dx \mathbf{u} = \mathbf{S} \mathbf{u}$$

and

$$\int_a^b uv dx \approx \int_a^b \left(\sum_{i=1}^n \varphi_i u_i \right) \varphi_j dx = \int_a^b \varphi_i \varphi_j dx (u_1, \dots, u_n)^T = \int_a^b \boldsymbol{\varphi}^T \boldsymbol{\varphi} dx \mathbf{u} = \mathbf{M} \mathbf{u}$$

and with $\int_a^b fv dx = \left[\frac{du}{dx} v \right]_a^b = \mathbf{f}$ we arrive at

$$(\mathbf{S} + \mathbf{M}) \mathbf{u} = \mathbf{f}$$

6 FEM, heat conduction 2D

The governing equation for heat conduction: $-\nabla \cdot (k \nabla u) = f$ in Ω , $u = g_D$ on $\partial\Omega_D$, $-k \mathbf{n} \cdot \nabla u = g_F$ on $\partial\Omega_F$.
Weak form: find u , such that

$$\int_{\Omega} k \nabla u \cdot \nabla v d\Omega = \int_{\Omega} fv d\Omega + \int_{\partial\Omega_F} g_F v ds \quad \text{for all } v, v = 0 \text{ on } \partial\Omega_D.$$

Approximation $u \approx U = \sum_{i=1}^n \varphi_i(x, y) u_i = \boldsymbol{\varphi} \cdot \mathbf{u}$, $\nabla U = \mathbf{B} \mathbf{u}$ leading to $\mathbf{S} \mathbf{u} = \mathbf{f}$, where $\boldsymbol{\varphi} = [\varphi_1, \varphi_2, \dots, \varphi_n]$, $\mathbf{u} = [u_1, u_2, \dots, u_n]^T$,

$$\mathbf{B} = \begin{bmatrix} \frac{\partial \varphi_1}{\partial x} & \frac{\partial \varphi_2}{\partial x} & \dots & \frac{\partial \varphi_n}{\partial x} \\ \frac{\partial \varphi_1}{\partial y} & \frac{\partial \varphi_2}{\partial y} & \dots & \frac{\partial \varphi_n}{\partial y} \end{bmatrix}, \quad \mathbf{S}_K = \int_K \mathbf{B}^T k \mathbf{B} dK, \quad \mathbf{f}_K = \int_K \boldsymbol{\varphi}^T f dK + \int_{\partial K_F} \boldsymbol{\varphi}^T g_F ds,$$

where \mathbf{S}_K is the element stiffness matrix for element K and \mathbf{f}_K the element load vector.

7 FEM, elasticity 2D

The governing equation for static linear elasticity:

$$\begin{aligned} -\nabla \cdot \boldsymbol{\sigma} &= \mathbf{f} & \text{in } \Omega & (\text{equilibrium}) \\ \boldsymbol{\sigma} &= \lambda \nabla \cdot \mathbf{u} \mathbf{I} + 2\mu \boldsymbol{\varepsilon}(\mathbf{u}) & \text{in } \Omega & (\text{small strain material behaviour, Hooke's law}) \\ \mathbf{u} &= \mathbf{0} & \text{on } \partial\Omega_D & (\text{prescribed displacements}) \\ \boldsymbol{\sigma} \cdot \mathbf{n} &= \mathbf{t} & \text{on } \partial\Omega_F & (\text{prescribed traction forces}), \end{aligned}$$

where the Lamé parameters are defined as

$$\lambda = \frac{E\nu}{(1+\nu)(1-2\nu)} \quad (\text{Plane strain}) \quad \text{or} \quad \lambda = \frac{E\nu}{1-\nu^2} \quad (\text{Plane stress}),$$

$$\mu = \frac{E}{2(1+\nu)},$$

and where E is the Young's modulus and ν the Poisson's ratio.

Weak form: find \mathbf{u} such that

$$\int_{\Omega} \boldsymbol{\sigma}(\mathbf{v}) : \boldsymbol{\varepsilon}(\mathbf{v}) \, d\Omega = \int_{\Omega} \mathbf{f} \cdot \mathbf{v} \, d\Omega + \int_{\Omega_F} \mathbf{g} \cdot \mathbf{v} \, ds \quad \text{for all } \mathbf{v}, \mathbf{v} = \mathbf{0} \text{ on } \partial\Omega_D.$$

Mandel notation

$$\boldsymbol{\sigma}_M = \begin{bmatrix} \sigma_x \\ \sigma_y \\ \tau_{xy} \end{bmatrix} \quad \boldsymbol{\varepsilon}_M(\mathbf{u}) = \begin{bmatrix} \varepsilon_x \\ \varepsilon_y \\ \sqrt{2}\varepsilon_{xy} \end{bmatrix} = \begin{bmatrix} \frac{\partial}{\partial x} & 0 \\ 0 & \frac{\partial}{\partial y} \\ \frac{1}{\sqrt{2}} \frac{\partial}{\partial y} & \frac{1}{\sqrt{2}} \frac{\partial}{\partial x} \end{bmatrix} \begin{bmatrix} u_x \\ u_y \end{bmatrix} = \begin{bmatrix} \frac{\partial u_x}{\partial x} \\ \frac{\partial u_y}{\partial y} \\ \frac{1}{\sqrt{2}} \frac{\partial u_x}{\partial y} - \frac{1}{\sqrt{2}} \frac{\partial u_y}{\partial x} \end{bmatrix}$$

and we have that $\boldsymbol{\varepsilon} : \boldsymbol{\varepsilon} = \boldsymbol{\varepsilon}_M \cdot \boldsymbol{\varepsilon}_M$ and $\boldsymbol{\sigma} : \boldsymbol{\varepsilon} = \boldsymbol{\varepsilon}_M^T \boldsymbol{\sigma}_M$.

Approximation $\mathbf{u} \approx \mathbf{U} = (\sum_{i=1}^n \varphi_i(x, y) u_x^i, \sum_{i=1}^n \varphi_i(x, y) u_y^i) = \boldsymbol{\Phi} \mathbf{u}$, where

$$\boldsymbol{\Phi} = \begin{bmatrix} \varphi_1 & 0 & \varphi_2 & 0 & \dots & \varphi_n & 0 \\ 0 & \varphi_1 & 0 & \varphi_2 & \dots & 0 & \varphi_n \end{bmatrix}, \quad \mathbf{u} = [u_x^1 \quad u_y^1 \quad u_x^2 \quad u_y^2 \quad \dots]^T.$$

Then

$$\mathbf{S}_K := \int_K (2\mu \mathbf{B}_\varepsilon^T \mathbf{B}_\varepsilon + \lambda \mathbf{B}_{\text{div}}^T \mathbf{B}_{\text{div}}) dK$$

where

$$\mathbf{B}_\varepsilon := \begin{bmatrix} \frac{\partial \varphi_1}{\partial x} & 0 & \frac{\partial \varphi_2}{\partial x} & 0 & \dots \\ 0 & \frac{\partial \varphi_1}{\partial y} & 0 & \frac{\partial \varphi_2}{\partial y} & \dots \\ \frac{1}{\sqrt{2}} \frac{\partial \varphi_1}{\partial y} & \frac{1}{\sqrt{2}} \frac{\partial \varphi_1}{\partial x} & \frac{1}{\sqrt{2}} \frac{\partial \varphi_2}{\partial y} & \frac{1}{\sqrt{2}} \frac{\partial \varphi_2}{\partial x} & \dots \end{bmatrix}$$

and

$$\mathbf{B}_{\text{div}} := \begin{bmatrix} \frac{\partial \varphi_1}{\partial x} & \frac{\partial \varphi_1}{\partial y} & \frac{\partial \varphi_2}{\partial x} & \frac{\partial \varphi_2}{\partial y} & \dots \end{bmatrix}.$$

Similarily the load element vector is defined by

$$\mathbf{f}_K := \int_K \mathbf{\Phi}^T \mathbf{f}(\mathbf{x}) dK + \int_{\partial K_F} \mathbf{\Phi}^T \mathbf{g}_F dK$$

which leads to the linear system $\mathbf{S}\mathbf{u} = \mathbf{f}$.

8 Post processing of linear elasticity

8.1 Computing the strain matrix

$$\boldsymbol{\varepsilon} = \begin{bmatrix} \frac{\partial u_x}{\partial x} & \frac{1}{2} \left(\frac{\partial u_x}{\partial y} + \frac{\partial u_y}{\partial x} \right) \\ \frac{1}{2} \left(\frac{\partial u_x}{\partial y} + \frac{\partial u_y}{\partial x} \right) & \frac{\partial u_y}{\partial y} \end{bmatrix} = \frac{1}{2} (\nabla \mathbf{u}^T + \nabla \mathbf{u})$$

this is approximated, using the basis functions, by

$$\boldsymbol{\varepsilon}(\mathbf{u}) \approx \boldsymbol{\varepsilon}_a(\mathbf{u}) = \begin{bmatrix} \frac{\partial \varphi}{\partial x} \cdot \mathbf{u}_x & \frac{1}{2} \left(\frac{\partial \varphi}{\partial y} \cdot \mathbf{u}_x + \frac{\partial \varphi}{\partial x} \cdot \mathbf{u}_y \right) \\ \frac{1}{2} \left(\frac{\partial \varphi}{\partial y} \cdot \mathbf{u}_x + \frac{\partial \varphi}{\partial x} \cdot \mathbf{u}_y \right) & \frac{\partial \varphi}{\partial y} \cdot \mathbf{u}_y \end{bmatrix}$$

where $\frac{\partial \varphi}{\partial x} = \begin{bmatrix} \frac{\partial \varphi_1}{\partial x} & \frac{\partial \varphi_2}{\partial x} & \dots & \frac{\partial \varphi_n}{\partial x} \end{bmatrix}$, $\varphi = \varphi(x, y)$ and $\mathbf{u}_x = \begin{bmatrix} u_x^1 & u_x^2 & \dots & u_x^n \end{bmatrix}^T$. Using

$$\mathbf{B} = \begin{bmatrix} \frac{\partial \varphi_1}{\partial x} & \frac{\partial \varphi_2}{\partial x} & \dots & \frac{\partial \varphi_n}{\partial x} \\ \frac{\partial \varphi_1}{\partial y} & \frac{\partial \varphi_2}{\partial y} & \dots & \frac{\partial \varphi_n}{\partial y} \end{bmatrix}$$

and $\nabla \mathbf{u} = \mathbf{B}\mathbf{u}$ we arrive at

$$\boldsymbol{\varepsilon}_a(\mathbf{u}) = \frac{1}{2} (\nabla \mathbf{u}^T + \nabla \mathbf{u}) = \frac{1}{2} ((\mathbf{B}\mathbf{u})^T + \mathbf{B}\mathbf{u})$$

8.2 Computing the stress matrix from the strain

$$\boldsymbol{\sigma} = 2\mu \boldsymbol{\varepsilon}(\mathbf{u}) + \lambda \text{tr} \boldsymbol{\varepsilon}(\mathbf{u}) \mathbf{I} \approx 2\mu \boldsymbol{\varepsilon}_a(\mathbf{u}) + \lambda \text{tr} \boldsymbol{\varepsilon}_a(\mathbf{u}) \mathbf{I} \approx \begin{bmatrix} \sigma_x & \tau_{xy} \\ \tau_{xy} & \sigma_y \end{bmatrix}$$

where $\text{tr} \boldsymbol{\varepsilon}(\mathbf{u}) = \varepsilon_x(\mathbf{u}) + \varepsilon_y(\mathbf{u}) = \nabla \cdot \mathbf{u} \approx \frac{\partial \varphi}{\partial x} \cdot \mathbf{u}_x + \frac{\partial \varphi}{\partial y} \cdot \mathbf{u}_y$.

8.3 Computing principle stresses/ strains

Solve $|\boldsymbol{\sigma} - \Sigma \mathbf{I}| = 0$, with $|\cdot| = \det(\cdot)$ which leads to the equation

$$\Sigma^2 - (\sigma_x + \sigma_y)\Sigma + \sigma_x\sigma_y - \tau_{xy}^2 = 0$$

note that $(\sigma_x + \sigma_y) = \text{tr}\boldsymbol{\sigma}$ and $\sigma_x\sigma_y - \tau_{xy}^2 = |\boldsymbol{\sigma}|$, so $\Sigma_{1,2} = \frac{1}{2}\text{tr}\boldsymbol{\sigma} \pm \sqrt{\left(\frac{1}{2}\text{tr}\boldsymbol{\sigma}\right)^2 - |\boldsymbol{\sigma}|}$, where Σ_1 and Σ_2 are the principal stresses. The principal strains can be computed the same way. In MATLAB this is done by the `eig()` function, see the help files for more info.

8.4 Equivalent strain

A scalar quantity called the equivalent strain, or the von Mises strain, is often used to describe the state of strain in solids.

$$\varepsilon_{\text{eq}} = \sqrt{\frac{2}{3}\boldsymbol{\varepsilon}^{\text{dev}} : \boldsymbol{\varepsilon}^{\text{dev}}} = \sqrt{\frac{2}{3}\varepsilon_{ij}^{\text{dev}}\varepsilon_{ij}^{\text{dev}}}; \quad \boldsymbol{\varepsilon}^{\text{dev}} = \boldsymbol{\varepsilon} - \frac{1}{3}\text{tr}(\boldsymbol{\varepsilon})\mathbf{I}$$

8.5 Von Mises stress

State of stress	Boundary Conditions	von Mises Equations
General 3D	No restrictions	$\sigma_v = \sqrt{\frac{1}{2}\left[(\sigma_x - \sigma_y)^2 + (\sigma_y - \sigma_z)^2 + (\sigma_z - \sigma_x)^2 + 6(\sigma_{xy}^2 + \sigma_{yz}^2 + \sigma_{zx}^2)\right]}$
General plane stress (2D)	$\sigma_z = \sigma_{yz} = \sigma_{zx} = 0$	$\sigma_v = \sqrt{\sigma_x^2 - \sigma_x\sigma_y + \sigma_y^2 + 3\sigma_{xy}^2}$

8.6 L_2 projection

If we want to take a field that exists inside the elements and put it in the nodes we basically want to minimize $u - u_h$, where u_h is the element field and u is the nodal field. We chose to minimize the error on average, this means that we want to minimize

$$\epsilon := \sqrt{\int_{\Omega} (u - u_h)^2 d\Omega}$$

We can achieve this by setting up the problem: Find u such that

$$\int_{\Omega} (u - u_h) v d\Omega = 0$$

Rewrite this by multiplying v into the parenthesis

$$\int_{\Omega} uv d\Omega = \int_{\Omega} u_h v d\Omega$$

Now apply Galerkin and approximate $u \approx \sum_i \varphi_i u_i$ and insert into the equation above

$$\int_{\Omega} \varphi_i \varphi_j d\Omega u_i = \int_{\Omega} u_h \varphi_j d\Omega$$

or

$$\mathbf{M}\mathbf{u} = \mathbf{f}$$

we can then get the averaged nodal values by

$$\mathbf{u} = \mathbf{M}^{-1}\mathbf{f}$$