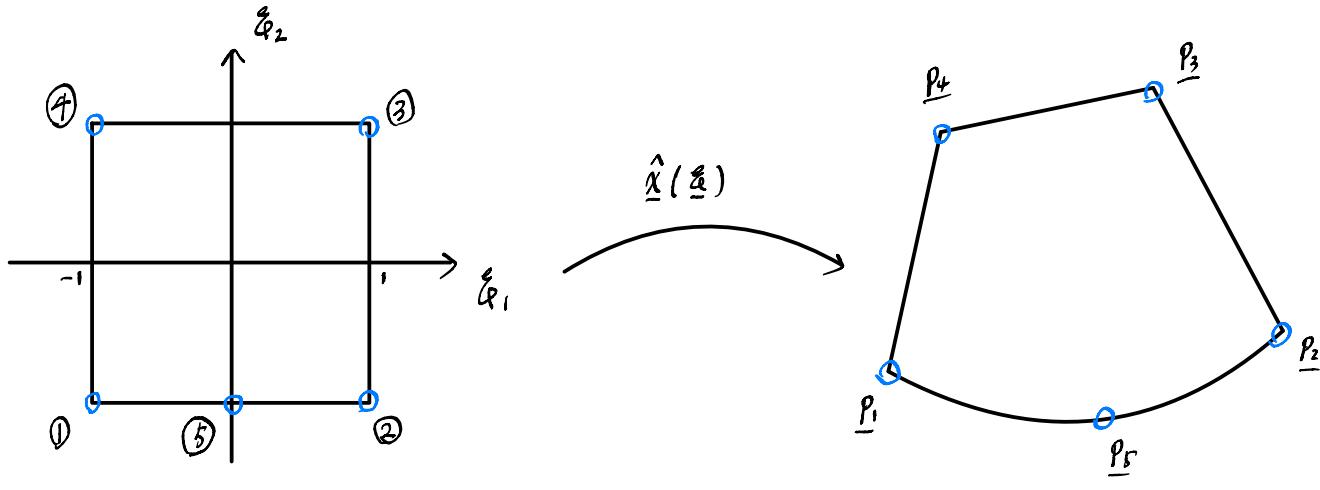


## Lec 04 : 2D linear problem

- Element technology
- Numerical integration
- Constraints

### Element technology

#### Designing a customized element



$$\hat{x}(\xi) = \sum_{i=1}^5 p_i \hat{\phi}_i(\xi)$$

Goal: Design shape functions  $\{\hat{\phi}_i\}_{i=1}^5$  for the 5-node element s.t. Kronecker-δ and partition of unity properties are satisfied.

\* Let  $\hat{\phi}_5(\xi_1, \xi_2) = l_2'(\xi_1) l_1'(\xi_2) = \frac{1}{2}(1+\xi_1)(1-\xi_1)(1-\xi_2)$

Note that  $\hat{\phi}_5(-1, -1) = \hat{\phi}_5(1, -1) = \hat{\phi}_5(1, 1) = \hat{\phi}_5(-1, 1) = 0$ , and  $\hat{\phi}_5(0, -1) = 1$ . ✓

\* Let  $\hat{\phi}_1(\xi_1, \xi_2) = l_1'(\xi_1) l_1'(\xi_2) = \frac{1}{4}(1-\xi_1)(1-\xi_2)$ , notice that  $\hat{\phi}_1(0, -1) = \frac{1}{2} \neq 0$ , not good.

Let  $\hat{\phi}_1(\xi_1, \xi_2) = l_1'(\xi_1) l_1'(\xi_2) - \frac{1}{2} \hat{\phi}_5(\xi_1, \xi_2) = \frac{1}{4}(1-\xi_1)(1-\xi_2) - \frac{1}{4}(1+\xi_1)(1-\xi_2)(1-\xi_2)$ . ✓

\* Let  $\hat{\phi}_2(\xi_1, \xi_2) = l_2'(\xi_1) l_1'(\xi_2) - \frac{1}{2} \hat{\phi}_5(\xi_1, \xi_2)$ . ✓

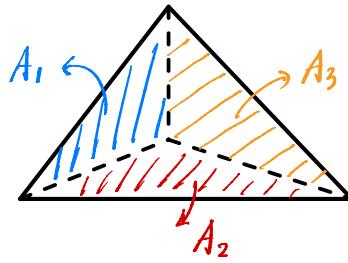
\* Let  $\hat{\phi}_3(\xi_1, \xi_2) = l_2'(\xi_1) l_2'(\xi_2)$  ✓ and  $\hat{\phi}_4(\xi_1, \xi_2) = l_1'(\xi_1) l_2'(\xi_2)$  ✓

Verify that  $\sum_{i=1}^5 \hat{\phi}_i = 1$  and Kronecker-δ holds. More reading: Hughes book Section 3.7

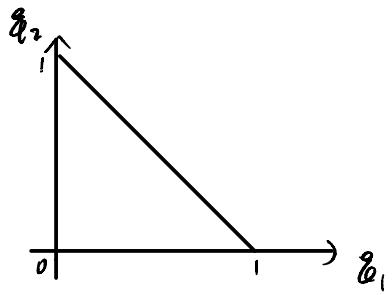
## Shape functions for triangular elements

### Area coordinate

$$r = \frac{A_1}{A}, \quad s = \frac{A_2}{A}, \quad t = \frac{A_3}{A}$$



Consider parametric domain  $\hat{\Omega}$ :



We have a map from  $(\hat{r}_1, \hat{r}_2)$  to  $(r, s, t)$ :

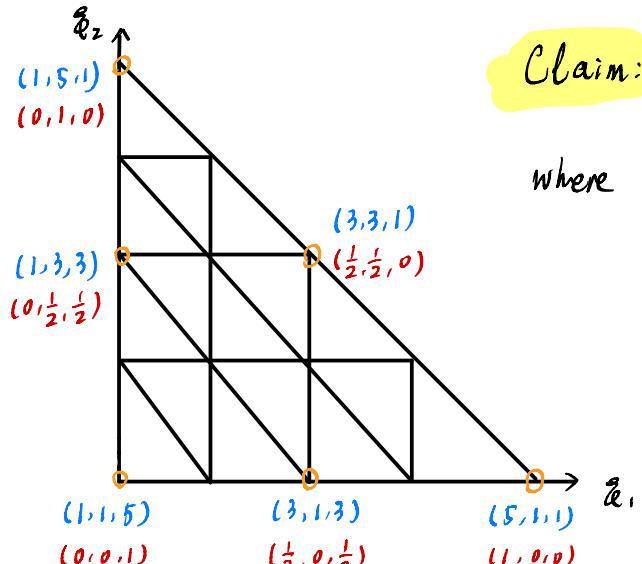
$$r = \hat{r}_1$$

$$s = \hat{r}_2$$

$$t = 1 - \hat{r}_1 - \hat{r}_2$$

$$\hat{\Omega} = \{ \underline{\hat{r}} : \hat{r}_1 \geq 0, \hat{r}_2 \geq 0, \hat{r}_1 + \hat{r}_2 \leq 1 \}$$

Goal: Design shape functions  $\{\hat{\phi}_i(\underline{\hat{r}})\}$  with the help of area coordinate  $(r, s, t)$ .



$(I, J, K)$

$(\bar{r}_I, \bar{s}_J, \bar{t}_K)$

$p$ : order (Here,  $p=4$ )

# Dof's:  $\frac{1}{2}(p+1)(p+2)$

**Claim:**  $\hat{\phi}_i(\underline{\hat{r}}) = T_I(r(\underline{\hat{r}})) T_J(s(\underline{\hat{r}})) T_K(t(\underline{\hat{r}}))$

where

$$T_I(r) = \begin{cases} l_I^{I-1}(r) & \text{if } I \neq 1 \\ 1 & \text{if } I = 1 \end{cases}$$

Similar definitions for  $T_J(s)$  and  $T_K(t)$ .

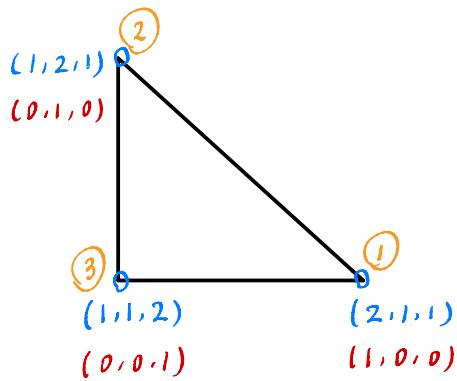
$$l_I^{I-1}(r) = \frac{(r - \bar{r}_1)(r - \bar{r}_2) \cdots (r - \bar{r}_{I-1})}{(\bar{r}_I - \bar{r}_1)(\bar{r}_I - \bar{r}_2) \cdots (\bar{r}_I - \bar{r}_{I-1})}$$

$I, J, K$ : node index

$(I=1, 2, \dots, p+1, J=1, 2, \dots, p+1, K=1, 2, \dots, p+1, I+J+K \leq p+3)$

$$\bar{r}_I = \frac{I-1}{p}, \bar{s}_J = \frac{J-1}{p}, \bar{t}_K = \frac{K-1}{p} \quad (\text{knot coordinate})$$

Ex. Shape functions for 3-node linear triangle:

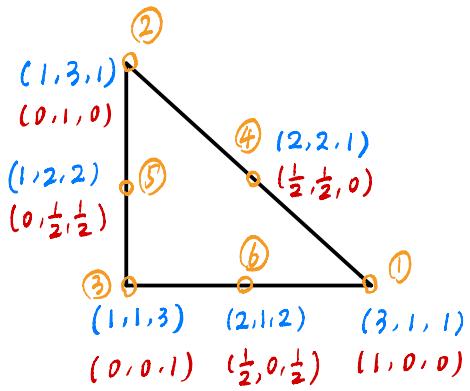


$$\hat{\phi}_1(\underline{x}) = T_1(r) T_1(s) T_1(t) = \ell'_1(r) \cdot 1 \cdot 1 = \frac{r - \bar{r}_1}{\bar{r}_2 - \bar{r}_1} = r$$

$$\hat{\phi}_2(\underline{x}) = T_1(r) T_2(s) T_1(t) = 1 \cdot \ell'_2(s) \cdot 1 = \frac{s - \bar{s}_1}{\bar{s}_2 - \bar{s}_1} = s$$

$$\hat{\phi}_3(\underline{x}) = T_1(r) T_1(s) T_2(t) = 1 \cdot 1 \cdot \ell'_2(t) = \frac{t - \bar{t}_1}{\bar{t}_2 - \bar{t}_1} = t$$

Ex. Shape functions for 6-node triangle:



$$\hat{\phi}_1(\underline{x}) = T_3(r) T_1(s) T_1(t) = \ell'_3(r) \cdot 1 \cdot 1 = \frac{(r - \bar{r}_1)(r - \bar{r}_2)}{(\bar{r}_3 - \bar{r}_1)(\bar{r}_3 - \bar{r}_2)} = \frac{(r - 0)(r - \frac{1}{2})}{(1 - 0)(1 - \frac{1}{2})} = 2r(r - \frac{1}{2}) = r(2r - 1)$$

$$\hat{\phi}_5(\underline{x}) = T_1(r) T_2(s) T_2(t) = 1 \cdot \ell'_2(s) \ell'_2(t) = \frac{(s - \bar{s}_1)}{(\bar{s}_2 - \bar{s}_1)} \cdot \frac{(t - \bar{t}_1)}{(\bar{t}_2 - \bar{t}_1)} = \frac{s - 0}{\frac{1}{2} - 0} \cdot \frac{t - 0}{\frac{1}{2} - 0} = 4st$$

## Numerical integration

Motivation: Let  $f: \Omega^e \subset \mathbb{R}^d \rightarrow \mathbb{R}$  be a given function. We are interested in computing

$$\int_{\Omega^e} f(\underline{x}) d\underline{x}$$

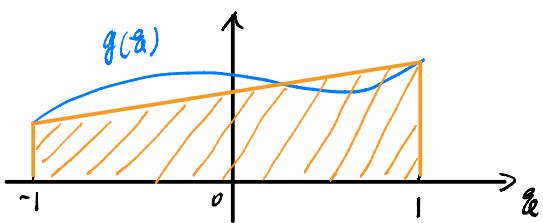
$$\text{Recall } K_{\alpha\beta}^e = \int_{\Omega^e} k \nabla \phi_\alpha(\underline{x}) \cdot \nabla \phi_\beta(\underline{x}) d\underline{x} = \int_{\hat{\Omega}} k \left( \nabla_{\hat{\underline{x}}} \hat{\phi}_\alpha(\underline{x}) \cdot \nabla_{\hat{\underline{x}}}^{-1} \hat{\phi}_\beta(\underline{x}) \right) \cdot \left( \nabla_{\hat{\underline{x}}} \hat{\phi}_\beta(\underline{x}) \cdot \nabla_{\hat{\underline{x}}}^{-1} \hat{\phi}_\beta(\underline{x}) \right) / j(\underline{x}) / d\underline{x}$$

Consider 1D case:  $\int_{-1}^1 g(\underline{x}) d\underline{x}$ , this can be approximated by numerical integration:

$$\int_{-1}^1 g(\underline{x}) d\underline{x} = \sum_{i=1}^{N_{int}} g(\tilde{\underline{x}}_i) w_i + R \approx \sum_{i=1}^{N_{int}} g(\tilde{\underline{x}}_i) w_i \quad \text{Quadrature formula}$$

## Ex. Trapezoidal rule

$$\int_{-1}^1 g(\tilde{x}) d\tilde{x} \approx g(-1) + g(1)$$



$$N_{\text{int}} = 2, \tilde{x}_1 = -1, \tilde{x}_2 = 1, W_1 = 1, W_2 = 1$$

$$R = -\frac{2}{3} \frac{d^2 g}{dx^2}(\bar{x}) \text{ at some } \bar{x}$$

## Ex. Simpson's rule

$$\int_{-1}^1 g(\tilde{x}) d\tilde{x} \approx \frac{1}{3} g(-1) + \frac{4}{3} g(0) + \frac{1}{3} g(1)$$

$$N_{\text{int}} = 3, \tilde{x}_1 = -1, \tilde{x}_2 = 0, \tilde{x}_3 = 1, W_1 = \frac{1}{3}, W_2 = \frac{4}{3}, W_3 = \frac{1}{3}$$

$$R = -\frac{g^{(4)}(\bar{x})}{90} \text{ at some } \bar{x}$$

We want to save computational cost: The fewer the integration points, the less the cost.

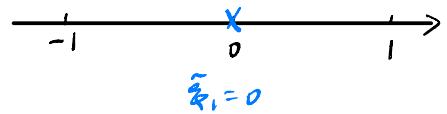
## Gaussian quadrature

$$1. N_{\text{int}} = 1$$

$$\int_{-1}^1 g(\tilde{x}) d\tilde{x} \approx 2g(0)$$

$$W_1 = 2$$

$$\tilde{x}_1 = 0, W_1 = 2, R = \frac{g^{(2)}(\bar{x})}{3}$$

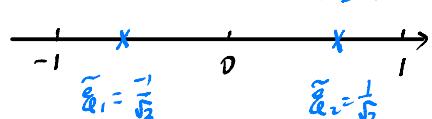


$$2. N_{\text{int}} = 2$$

$$\int_{-1}^1 g(\tilde{x}) d\tilde{x} \approx g\left(\frac{-1}{\sqrt{3}}\right) + g\left(\frac{1}{\sqrt{3}}\right)$$

$$W_1 = 1 \quad W_2 = 1$$

$$\tilde{x}_1 = \frac{-1}{\sqrt{3}}, \tilde{x}_2 = \frac{1}{\sqrt{3}}, W_1 = 1, W_2 = 1, R = \frac{g^{(4)}(\bar{x})}{135}$$

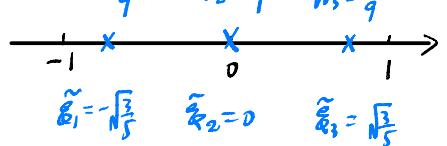


$$3. N_{\text{int}} = 3$$

$$\int_{-1}^1 g(\tilde{x}) d\tilde{x} \approx \frac{5}{9} g\left(-\sqrt{\frac{3}{5}}\right) + \frac{8}{9} g(0) + \frac{5}{9} g\left(\sqrt{\frac{3}{5}}\right)$$

$$W_1 = \frac{5}{9} \quad W_2 = \frac{8}{9} \quad W_3 = \frac{5}{9}$$

$$\tilde{x}_1 = -\sqrt{\frac{3}{5}}, \tilde{x}_2 = 0, \tilde{x}_3 = \sqrt{\frac{3}{5}}, W_1 = \frac{5}{9}, W_2 = \frac{8}{9}, W_3 = \frac{5}{9}, R = \frac{g^{(6)}(\bar{x})}{15750}$$



Proposition Gaussian quadrature scheme with  $n_{\text{int}}$  integration points can exactly integrate polynomial function  $g(\xi)$  of order  $p$ , if  $2n_{\text{int}} \geq p+1$ .

Proof Let  $g(\xi) = a_0 + a_1 \xi + a_2 \xi^2 + \dots + a_p \xi^p$

$$\int_{-1}^1 g(\xi) d\xi = \int_{-1}^1 a_0 d\xi + \int_{-1}^1 a_1 \xi + \dots + \int_{-1}^1 a_p \xi^p = 2a_0 + 0 \cdot a_1 + \dots = \underline{g}^T \underline{a}$$

$$\text{where } \underline{g} = \begin{bmatrix} 2 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \underline{a} = \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_p \end{bmatrix}$$

$$\int_{-1}^1 g(\xi) d\xi \approx \sum_{\ell=1}^{n_{\text{int}}} g(\tilde{\xi}_\ell) w_\ell = \underbrace{\begin{bmatrix} w_1 & w_2 & \dots & w_{n_{\text{int}}} \end{bmatrix}}_{W^T} \begin{bmatrix} g(\tilde{\xi}_1) \\ g(\tilde{\xi}_2) \\ \vdots \\ g(\tilde{\xi}_{n_{\text{int}}}) \end{bmatrix}$$

$$= \underbrace{\begin{bmatrix} w_1 & w_2 & \dots & w_{n_{\text{int}}} \end{bmatrix}}_{W^T} \underbrace{\begin{bmatrix} 1 & \tilde{\xi}_1 & \tilde{\xi}_1^2 & \dots & \tilde{\xi}_1^p \\ 1 & \tilde{\xi}_2 & \tilde{\xi}_2^2 & \dots & \tilde{\xi}_2^p \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \tilde{\xi}_{n_{\text{int}}} & \tilde{\xi}_{n_{\text{int}}}^2 & \dots & \tilde{\xi}_{n_{\text{int}}}^p \end{bmatrix}}_M \underbrace{\begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_p \end{bmatrix}}_{\underline{a}} = W^T M \underline{a}$$

We want  $\underline{g}^T \underline{a} = W^T M \underline{a}$  to hold for arbitrary  $\underline{a}$ .  $\Rightarrow (\underline{g}^T - W^T M)^T \underline{a} = 0 \quad \forall \underline{a}$

$\Rightarrow \underline{g}^T - W^T M = 0 \Rightarrow M^T W = \underline{g}$  ( $p+1$  equations to be satisfied by  $2n_{\text{int}}$  parameters)

Note: The  $2n_{\text{int}}$  parameters include  $\{w_1, w_2, \dots, w_{n_{\text{int}}}\}$  and  $\{\tilde{\xi}_1, \tilde{\xi}_2, \dots, \tilde{\xi}_{n_{\text{int}}}\}$ .

Ex. Show that the Gaussian quadrature scheme for  $n_{\text{int}}=2$  can exactly integrate  $g(\xi) = a_0 + a_1 \xi + a_2 \xi^2 + a_3 \xi^3$ .

Sol. ① Since  $2n_{\text{int}} = 4 \geq 4 = p+1$ , the argument holds.

② Go through the reasoning (see homework).

## Gaussian quadrature in several dimensions

Construct by employing 1D Gaussian rules on each coordinate separately:

$$\int_{-1}^1 \int_{-1}^1 g(\xi, y) d\xi dy \approx \int_{-1}^1 \left( \sum_{\ell^{(1)}=1}^{N_{int}^{(1)}} g(\tilde{\xi}_{\ell^{(1)}}, y) W_{\ell^{(1)}} \right) dy \approx \sum_{\ell^{(1)}=1}^{N_{int}^{(1)}} \sum_{\ell^{(2)}=1}^{N_{int}^{(2)}} g(\tilde{\xi}_{\ell^{(1)}}, \tilde{y}_{\ell^{(2)}}) W_{\ell^{(1)}} W_{\ell^{(2)}}$$

↑
↑

Gaussian rule (1)  
 applied to  $\xi$ 
Gaussian rule (2)  
 applied to  $y$

$$\Rightarrow \int_{-1}^1 \int_{-1}^1 g(\tilde{x}, y) d\tilde{x} dy \approx \sum_{e=1}^{n_{int}} g(\tilde{x}_e, \tilde{y}_e) w_e, \text{ where}$$

$$N_{\text{int}} = N_{\text{int}}^{(1)} N_{\text{int}}^{(2)}, \quad \tilde{\mathcal{E}}_c = \tilde{\mathcal{E}}_c^{(1)}, \quad \tilde{g}_c = \tilde{g}_c^{(2)}, \quad W_c = W_c^{(1)} W_c^{(2)}$$

Ex. If 1D one-point rule is used in each direction:

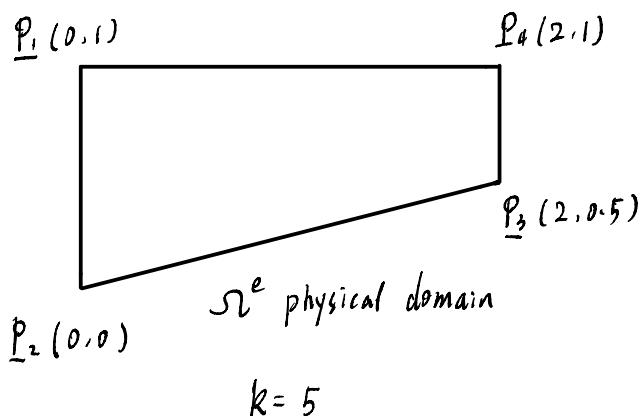
$$\int_1^1 \int_{-1}^1 g(x,y) dx dy \approx 4 g(0,0)$$

Ex. If 1D two-point rule is used in each direction:

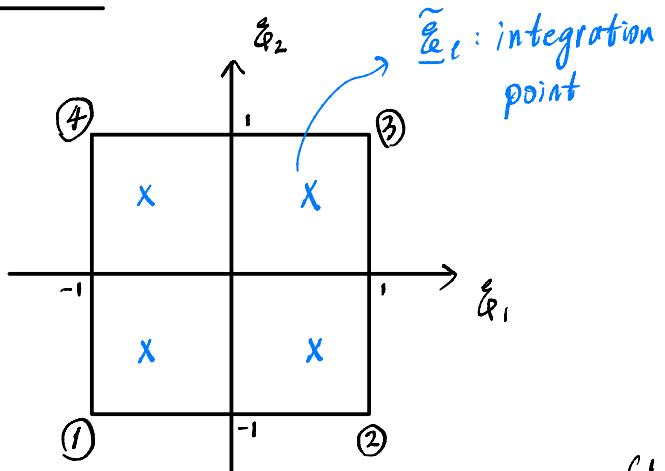
$$\int_{-1}^1 \int_{-1}^1 g(x, y) dx dy \approx g\left(\frac{-1}{\sqrt{3}}, \frac{-1}{\sqrt{3}}\right) + g\left(\frac{-1}{\sqrt{3}}, \frac{1}{\sqrt{3}}\right) + g\left(\frac{1}{\sqrt{3}}, \frac{-1}{\sqrt{3}}\right) + g\left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}\right)$$

Question: Why can we "decouple" the rule in several dims to 1D respectively?

## A practical example



Recall:



$$\hat{\Omega} = [-1, 1] \times [-1, 1]$$

parametric / parent domain

$$\hat{\phi}_1(\underline{\xi}) = (\underline{\xi}_1 - 1)(\underline{\xi}_2 - 1)/4$$

$$\hat{\phi}_2(\underline{\xi}) = -(\underline{\xi}_1 + 1)(\underline{\xi}_2 - 1)/4$$

$$\hat{\phi}_3(\underline{\xi}) = (\underline{\xi}_1 + 1)(\underline{\xi}_2 + 1)/4$$

$$\hat{\phi}_4(\underline{\xi}) = -(\underline{\xi}_1 - 1)(\underline{\xi}_2 + 1)/4$$

$$\int_{-1}^1 \int_{-1}^1 g(\underline{\xi}) d\underline{\xi} \approx \sum_{\ell=1}^{N_{\text{int}}} g(\hat{\underline{\xi}}_\ell) W_\ell$$

$$K_{\alpha\beta}^e = \int_{\hat{\Omega}} k \left( \begin{bmatrix} \frac{\partial \hat{\phi}_\alpha}{\partial \hat{\xi}_1} & \frac{\partial \hat{\phi}_\alpha}{\partial \hat{\xi}_2} \\ \frac{\partial \hat{\phi}_\beta}{\partial \hat{\xi}_1} & \frac{\partial \hat{\phi}_\beta}{\partial \hat{\xi}_2} \end{bmatrix} \underbrace{\begin{bmatrix} \frac{\partial \hat{x}_1}{\partial \hat{\xi}_1} & \frac{\partial \hat{x}_1}{\partial \hat{\xi}_2} \\ \frac{\partial \hat{x}_2}{\partial \hat{\xi}_1} & \frac{\partial \hat{x}_2}{\partial \hat{\xi}_2} \end{bmatrix}^{-1}}_{\left[ \underline{\underline{J}} \hat{\underline{x}}(\underline{\xi}) \right]^{-1}} \right) \cdot \left( \begin{bmatrix} \frac{\partial \hat{\phi}_\beta}{\partial \hat{\xi}_1} & \frac{\partial \hat{\phi}_\beta}{\partial \hat{\xi}_2} \\ \frac{\partial \hat{\phi}_\alpha}{\partial \hat{\xi}_1} & \frac{\partial \hat{\phi}_\alpha}{\partial \hat{\xi}_2} \end{bmatrix} \cdot \begin{bmatrix} \frac{\partial \hat{x}_1}{\partial \hat{\xi}_1} & \frac{\partial \hat{x}_1}{\partial \hat{\xi}_2} \\ \frac{\partial \hat{x}_2}{\partial \hat{\xi}_1} & \frac{\partial \hat{x}_2}{\partial \hat{\xi}_2} \end{bmatrix}^{-1} \right) \cdot |j(\underline{\xi})| d\underline{\xi}_1 d\underline{\xi}_2 / \left| \det \left( \underline{\underline{J}} \hat{\underline{x}}(\underline{\xi}) \right) \right|$$

$$\hat{\underline{x}}(\underline{\xi}) = \sum_{i=1}^4 p_i \hat{\phi}_i(\underline{\xi}), \Rightarrow \left[ \underline{\underline{J}} \hat{\underline{x}}(\underline{\xi}) \right] = \sum_{i=1}^4 \begin{bmatrix} p_{i1} \\ p_{i2} \end{bmatrix} \cdot \begin{bmatrix} \frac{\partial \hat{\phi}_i}{\partial \hat{\xi}_1} & \frac{\partial \hat{\phi}_i}{\partial \hat{\xi}_2} \end{bmatrix}_{2 \times 2}.$$

In matrix format:

$$\left[ \underline{\underline{J}} \hat{\underline{x}}(\underline{\xi}) \right]_{2 \times 2} = \left[ p_1 \ p_2 \ p_3 \ p_4 \right]_{2 \times 4} \begin{bmatrix} \frac{\partial \hat{\phi}_1}{\partial \hat{\xi}_1} & \frac{\partial \hat{\phi}_1}{\partial \hat{\xi}_2} \\ \frac{\partial \hat{\phi}_2}{\partial \hat{\xi}_1} & \frac{\partial \hat{\phi}_2}{\partial \hat{\xi}_2} \\ \frac{\partial \hat{\phi}_3}{\partial \hat{\xi}_1} & \frac{\partial \hat{\phi}_3}{\partial \hat{\xi}_2} \\ \frac{\partial \hat{\phi}_4}{\partial \hat{\xi}_1} & \frac{\partial \hat{\phi}_4}{\partial \hat{\xi}_2} \end{bmatrix}_{4 \times 2} = \begin{bmatrix} 0 & 0 & 2 & 2 \\ 1 & 0 & 0.5 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{4}(\hat{\xi}_2 - 1) & \frac{1}{4}(\hat{\xi}_1 - 1) \\ \frac{-1}{4}(\hat{\xi}_2 - 1) & \frac{-1}{4}(\hat{\xi}_1 + 1) \\ \frac{1}{4}(\hat{\xi}_2 + 1) & \frac{1}{4}(\hat{\xi}_1 + 1) \\ \frac{-1}{4}(\hat{\xi}_2 + 1) & \frac{-1}{4}(\hat{\xi}_1 - 1) \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 \\ \frac{\hat{\xi}_2 - 3}{8} & \frac{\hat{\xi}_1 + 1}{8} \end{bmatrix} \Rightarrow j = \frac{3 - \hat{\xi}_2}{8}, \quad \left[ \underline{\underline{J}} \hat{\underline{x}}(\underline{\xi}) \right]^{-1} = \begin{bmatrix} \frac{\hat{\xi}_1 + 1}{8} & \frac{8}{8 - \hat{\xi}_2} \\ -\frac{\hat{\xi}_2 + 3}{8} & 1 \end{bmatrix}$$

$$\text{Consider } K_{12}^e = \int_{\Omega^e} k \nabla \phi_1(\underline{x}) \nabla \phi_2(\underline{x}) d\underline{x}$$

$$K_{12}^e = \int_{\Omega} 5 \left( \begin{bmatrix} \frac{1}{4}(\xi_2-1) & \frac{1}{4}(\xi_1-1) \\ -\frac{1}{4}(\xi_2+1) & \frac{1}{4}(\xi_1+1) \end{bmatrix} \begin{bmatrix} \frac{\xi_1+1}{8} & \frac{8}{8-3} \\ -\frac{\xi_2+3}{8} & \frac{\xi_2-3}{8} \\ 1 & 0 \end{bmatrix} \right) \cdot \left( \begin{bmatrix} -\frac{1}{4}(\xi_2-1) & -\frac{1}{4}(\xi_1+1) \\ -\frac{1}{4}(\xi_2+1) & -\frac{1}{4}(\xi_1-1) \end{bmatrix} \begin{bmatrix} \frac{\xi_1+1}{8} & \frac{8}{8-3} \\ -\frac{\xi_2+3}{8} & \frac{\xi_2-3}{8} \\ 1 & 0 \end{bmatrix} \right) \cdot \frac{3-\xi_2}{8} d\xi$$

$$K_{12}^e \approx \sum_{i=1}^4 g(\tilde{\xi}_i) W_i = g\left(\frac{-1}{\sqrt{3}}, \frac{-1}{\sqrt{3}}\right) + g\left(\frac{-1}{\sqrt{3}}, \frac{1}{\sqrt{3}}\right) + g\left(\frac{1}{\sqrt{3}}, \frac{-1}{\sqrt{3}}\right) + g\left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}\right)$$

Consider the contribution from quadrature point  $(-\frac{1}{\sqrt{3}}, -\frac{1}{\sqrt{3}})$ :

$$g\left(-\frac{1}{\sqrt{3}}, -\frac{1}{\sqrt{3}}\right) = 5 \left( \left[ \frac{1}{4}(-\frac{1}{\sqrt{3}}-1) \quad \frac{1}{4}(-\frac{1}{\sqrt{3}}-1) \right] \dots \right) = -1.68$$

Homework: Complete this calculation and find  $K_{\alpha\beta}^e$  for  $\alpha=1, 2$  and  $\beta=1, 2$ .

### Constraints

Recall our 1D problem  $[K][v]=[f]$ , where

$$[K] = \frac{AE}{L/3} \begin{bmatrix} 1 & -1 & & \\ -1 & 2 & -1 & \\ & -1 & 2 & -1 \\ & & -1 & 1 \end{bmatrix}, \quad [v] = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{bmatrix}, \quad [f] = \begin{bmatrix} b \frac{L}{6} - \frac{AE}{L/3} g \\ b \frac{L}{3} + \frac{AE}{L/3} g \\ b \frac{L}{3} \\ b \frac{L}{6} + f \end{bmatrix}$$

We want to impose  $v_1=0$  (from Dirichlet B.C.) and solve for  $[v]$ .

For simplicity, assume  $A=1, E=1, L=3, b=2, g=1, f=1$ , so that

$$[K] = \begin{bmatrix} 1 & -1 & & \\ -1 & 2 & -1 & \\ & -1 & 2 & -1 \\ & & -1 & 1 \end{bmatrix}, \quad [f] = \begin{bmatrix} 0 \\ 3 \\ 2 \\ 2 \end{bmatrix}. \quad \text{Previously, we considered reduced system:}$$

$$[K_{\text{red}}] = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 1 \end{bmatrix}, \quad [F_{\text{red}}] = \begin{bmatrix} 3 \\ 2 \\ 2 \end{bmatrix}. \Rightarrow [v_{\text{red}}] = \begin{bmatrix} v_2 \\ v_3 \\ v_4 \end{bmatrix} = \begin{bmatrix} 7 \\ 11 \\ 13 \end{bmatrix}.$$

Observations:

- \*  $[K]$  is symmetric positive semi-definite ( $[x]^T [K] [x] \geq 0 \quad \forall [x] \in \mathbb{R}^n$ , and if  $[x] = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$ ,  $[x]^T [K] [x] = 0$ , physical interpretation: rigid body motion)
- \*  $[K_{\text{red}}]$  is symmetric positive definite (it is invertible).
- \* Do we have other ways to impose the constraint?

### Lagrange multiplier method

In the optimization context:

$$\min_{[v]} \bar{E}([v])$$
$$\text{s.t. } g([v]) = 0$$

where  $\bar{E}([v]) = \frac{1}{2} [v]^T [K] [v] - [v]^T [\bar{F}]$  is the energy function to be minimized,  
 $g([v]) = v_i - 0 = 0$  is the constraint that must be satisfied.

The Lagrange multiplier method tries to find the stationary points of  $\lambda$ :

$$\lambda([v], \lambda) := \bar{E}([v]) + \lambda g$$

that is:

$$\frac{\partial \lambda}{\partial [v]} = 0 \quad \text{and} \quad \frac{\partial \lambda}{\partial \lambda} = 0$$

$$\Rightarrow \begin{cases} [K][v] - [\bar{F}] + \lambda \frac{\partial g}{\partial [v]} = 0 \\ g([v]) = 0 \end{cases} \quad (\text{N+1 equations, N+1 variables})$$

For our problem:

$$\begin{bmatrix} 1 & -1 & 0 & 0 & 1 \\ -1 & 2 & -1 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 \\ 0 & 0 & -1 & 1 & 0 \\ 1 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ n \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \\ 2 \\ 2 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ n \end{bmatrix} = \begin{bmatrix} 0 \\ 7 \\ 11 \\ 13 \\ 7 \end{bmatrix}$$