

Finite Element Methods for PDEs

FEM for Poisson's problem in 2D

2024 AJOU-KYUSHU SUMMER SCHOOL ON APPLIED MATHEMATICS

Outline

- 1 Introduction
- 2 Affine mapping
- 3 Triangulation
- 4 Basis functions of V_h^k
- 5 Mass matrix and Stiffness matrix

2D Poisson problem

Consider the two dimensional Poisson problem

$$\begin{aligned} -\Delta u &= f && \text{in } \Omega \\ u &= u_D && \text{on } \partial\Omega \end{aligned}$$

Weak formulation

Find $u(x, y) \in H_0^1(\Omega)$ such that

$$\int_{\Omega} \nabla u \cdot \nabla v \, d\mathbf{x} = \int_{\Omega} f v \, d\mathbf{x}, \quad \forall v \in H_0^1(\Omega).$$

Variational formulation

Find $u_h \in V_h^k$ such that

$$\int_{\Omega} \nabla u_h \cdot \nabla v_h \, d\mathbf{x} = \int_{\Omega} f v_h \, d\mathbf{x}, \quad \forall v_h \in V_h^k$$

where

$$V_h^k = \{v_h \in H_0^1(\Omega) \mid v_h|_T \in P_k(T), \, \forall T \in \mathcal{T}_h\},$$

and $P_k(T)$ is the polynomial function space of degrees $\leq k$.

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Affine mapping

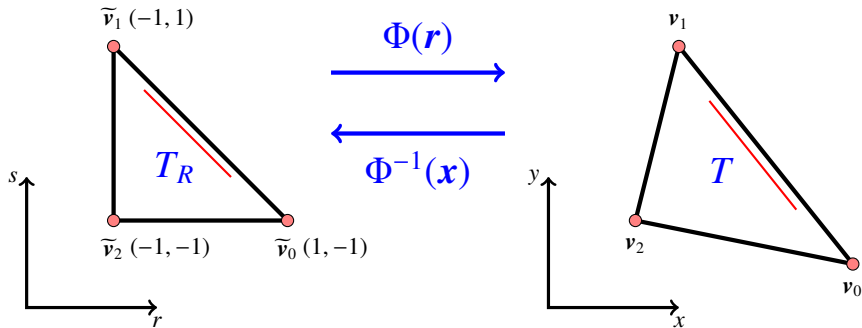


Figure: An affine mapping from the reference triangle T_R to a triangle T .

Barycentric coordinates

Barycentric coordinates $(\tilde{\lambda}_0, \tilde{\lambda}_1, \tilde{\lambda}_2)$ have the properties

$$\begin{cases} 0 \leq \tilde{\lambda}_i(\mathbf{r}) \leq 1, & i = 0, 1, 2 \\ \tilde{\lambda}_0(\mathbf{r}) + \tilde{\lambda}_1(\mathbf{r}) + \tilde{\lambda}_2(\mathbf{r}) = 1. \end{cases}$$

Affine mapping

Let us define a triangle T as

$$T = \text{span}\{\mathbf{v}_0, \mathbf{v}_1, \mathbf{v}_2\}, \quad \mathbf{v}_i = (v_i^{(1)}, v_i^{(2)}),$$

where $\mathbf{v}_0, \mathbf{v}_1, \mathbf{v}_2$ are vertices of T .

Then, we have an affine mapping Φ such that

$$\Phi(\mathbf{r}) = \frac{r+1}{2}\mathbf{v}_0 + \frac{s+1}{2}\mathbf{v}_1 - \frac{r+s}{2}\mathbf{v}_2 = \mathbf{x}.$$

Properties

$$\textcircled{1} \quad \lambda_0 = \frac{r+1}{2}, \quad \lambda_1 = \frac{s+1}{2}, \quad \lambda_2 = -\frac{r+s}{2}$$

$$\textcircled{2} \quad x_r = \frac{v_0^{(1)} - v_2^{(1)}}{2}, \quad y_r = \frac{v_0^{(2)} - v_2^{(2)}}{2}, \quad x_s = \frac{v_1^{(1)} - v_2^{(1)}}{2}, \quad y_s = \frac{v_1^{(2)} - v_2^{(2)}}{2}$$

$$\textcircled{3} \quad r_x = \frac{y_s}{J}, \quad r_y = -\frac{x_s}{J}, \quad s_x = -\frac{y_r}{J}, \quad s_y = \frac{x_r}{J} \quad (J = x_r y_s - x_s y_r)$$

$$\textcircled{4} \quad \lambda_i(\mathbf{x}) = \widetilde{\lambda}_i(\Phi^{-1}(\mathbf{x})), \quad \widetilde{\lambda}_i(\mathbf{r}) = \lambda_i(\Phi(\mathbf{r}))$$

$$\textcircled{5} \quad \frac{d}{dx} \lambda_i(\mathbf{x}) = r_x \frac{d}{dr} \widetilde{\lambda}_i(\mathbf{r}) + s_x \frac{d}{ds} \widetilde{\lambda}_i(\mathbf{r})$$

$$\textcircled{6} \quad \frac{d}{dy} \lambda_i(\mathbf{x}) = r_y \frac{d}{dr} \widetilde{\lambda}_i(\mathbf{r}) + s_y \frac{d}{ds} \widetilde{\lambda}_i(\mathbf{r})$$

Properties

Using the property $I = \frac{\partial \mathbf{x}}{\partial \mathbf{r}} \frac{\partial \mathbf{r}}{\partial \mathbf{x}}$, we have

$$I = \frac{\partial \mathbf{x}}{\partial \mathbf{r}} \frac{\partial \mathbf{r}}{\partial \mathbf{x}} = \begin{bmatrix} x_r & x_s \\ y_r & y_s \end{bmatrix} \begin{bmatrix} r_x & r_y \\ s_x & s_y \end{bmatrix} \Rightarrow \begin{bmatrix} r_x & r_y \\ s_x & s_y \end{bmatrix} = \frac{1}{x_r y_x - s_x y_r} \begin{bmatrix} y_s & -x_s \\ -y_r & x_r \end{bmatrix}.$$

1 Introduction

2 Affine mapping

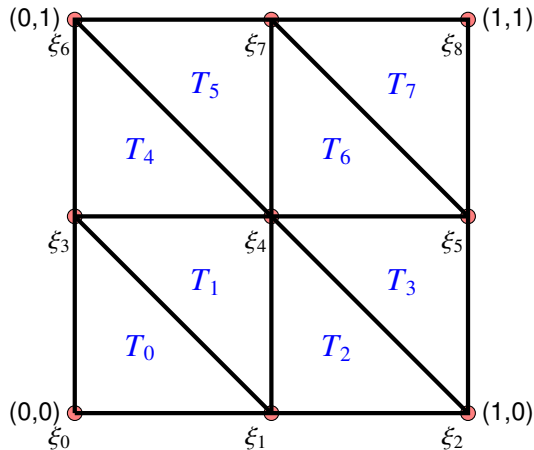
3 **Triangulation**

4 Basis functions of V_h^k

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Triangulation data (P_1)

If $\Omega = [0, 1]^2$, $M = 2$ and $k = 1$, data are stored as follows.



$$\begin{aligned} \text{c4n} &= [[0, 0], [1/2, 0], [1, 0], \\ &\quad [0, 1/2], [1/2, 1/2], [1, 1/2], \\ &\quad [0, 1], [1/2, 1], [1, 1]], \end{aligned}$$

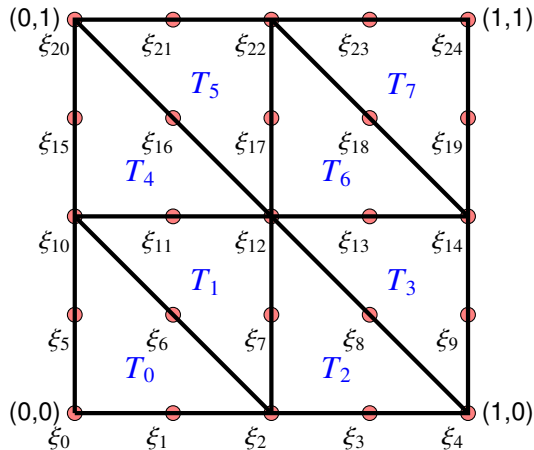
$$\begin{aligned} \text{n4e} &= [[1, 3, 0], [3, 1, 4], [2, 4, 1], [4, 2, 5], \\ &\quad [4, 6, 3], [6, 4, 7], [5, 7, 4], [7, 5, 8]], \end{aligned}$$

$$\begin{aligned} \text{ind4e} &= [[0, 1, 3], [4, 3, 1], [1, 2, 4], [5, 4, 2], \\ &\quad [3, 4, 6], [7, 6, 4], [4, 5, 7], [8, 7, 5]], \end{aligned}$$

$$\text{n4db} = [0, 1, 2, 3, 5, 6, 7, 8]$$

Triangulation data (P_2)

If $\Omega = [0, 1]^2$, $M = 2$ and $k = 2$, data are stored as follows.

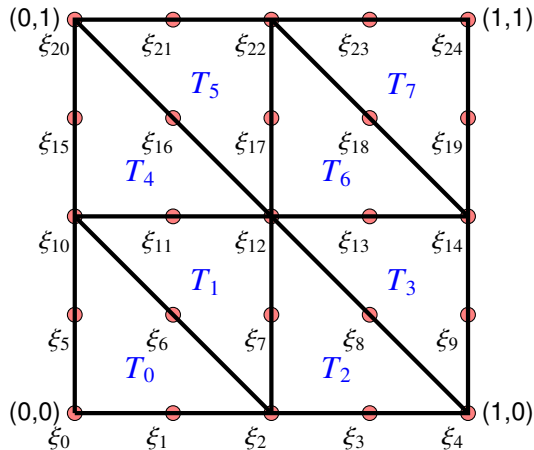


$$\begin{aligned} \text{c4n} = & [[0, 0], [1/4, 0], [1/2, 0], [3/4, 0], [1, 0], \\ & [0, 1/4], [1/4, 1/4], [1/2, 1/4], [3/4, 1/4], \\ & [1, 1/4], [0, 1/2], [1/4, 1/2], [1/2, 1/2], \\ & [3/4, 1/2], [1, 1/2], [0, 3/4], [1/4, 3/4], \\ & [1/2, 3/4], [3/4, 3/4], [1, 3/4], [0, 1], \\ & [1/4, 1], [1/2, 1], [3/4, 1], [1, 1]], \end{aligned}$$

$$\begin{aligned} \text{n4e} = & [[2, 10, 0], [10, 2, 12], [4, 12, 2], [12, 4, 14], \\ & [12, 20, 10], [20, 12, 22], [14, 22, 12], \\ & [22, 14, 24]] \end{aligned}$$

Triangulation data (P_2)

If $\Omega = [0, 1]^2$, $M = 2$ and $k = 2$, data are stored as follows.



ind4e = $[[0, 1, 2, 5, 6, 10], [12, 11, 10, 7, 6, 2],$
 $[2, 3, 4, 7, 8, 12], [14, 13, 12, 9, 8, 4],$
 $[10, 11, 12, 15, 16, 20], [22, 21, 20, 17, 16, 12],$
 $[12, 13, 14, 17, 18, 22], [24, 23, 22, 19, 18, 14]]$

n4db = $[0, 1, 2, 3, 5, 9, 10, 14, 15, 19, 20, 21, 22, 23, 24].$

This python code generates an uniform triangular mesh on the domain $[x_\ell, x_r] \times [y_\ell, y_r]$ in 2D with $2M_x$ elements along x-direction and $2M_y$ elements along y-direction. Also this code returns an index matrix for continuous k -th order polynomial approximations.

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Basis functions

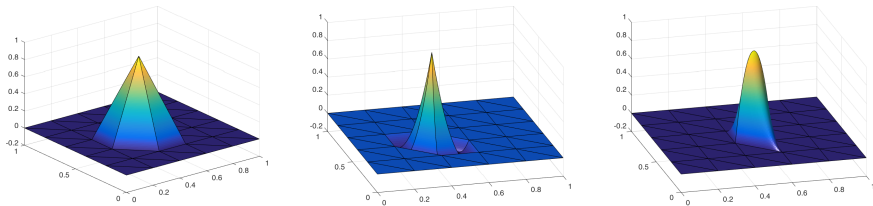
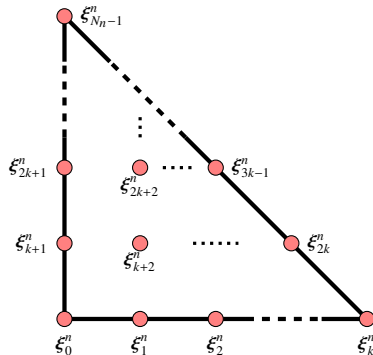


Figure: Global basis functions of V_h^1 (left) and V_h^2 (others) on an interval.

$$\psi_i(\xi_j) = \delta_{ij} = \begin{cases} 1 & \text{if } i = j, \\ 0 & \text{if } i \neq j, \end{cases}$$

$$\sum_{i=0}^{N-1} \psi_i(\mathbf{x}) = 1 \quad \forall \mathbf{x} \in \Omega$$

Local basis functions



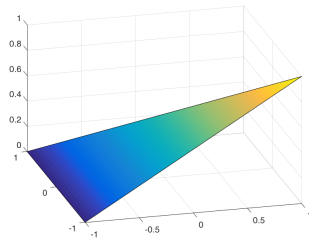
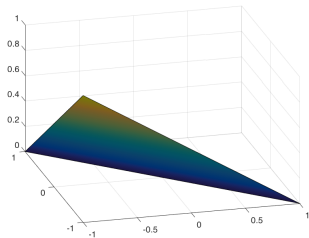
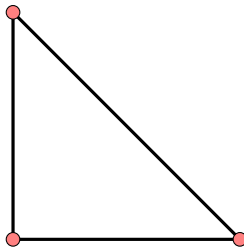
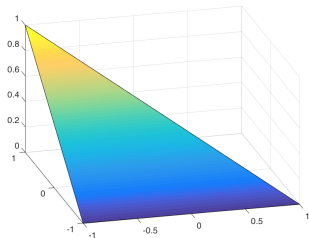
$$N_n = \frac{(k+1)(k+2)}{2}$$

$$\xi^n_j = \xi^n_{(\ell-1)(k+2-\ell/2)+m-1}$$

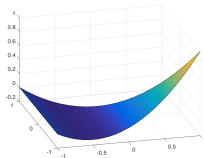
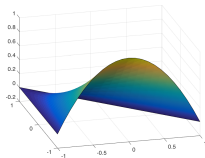
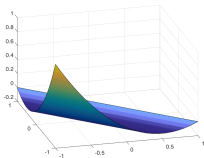
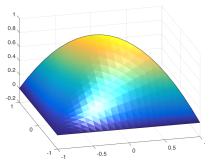
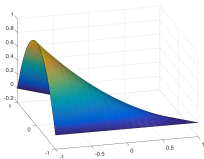
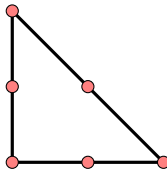
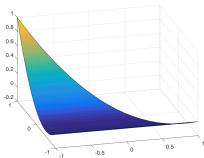
(ℓ : row number, m : column number, $\ell + m \leq k + 2$)

Figure: Node numbering in the n -th triangular element

Local basis functions



Local basis functions



Local basis functions

$$\psi_i^n(\mathbf{x}) = 0 \quad \text{if } \mathbf{x} \in \Omega \setminus T_n$$

$$\sum_{i=0}^{N_n-1} \psi_i^n(\mathbf{x}) = 1 \quad \forall \mathbf{x} \in T_n,$$

Interpolation

Using these basis functions, the interpolate function of a function $f(\mathbf{x})$ can be written as follows

$$\mathcal{I}f(\mathbf{x}) = \sum_{i=0}^{N-1} f_i \psi_i(\mathbf{x})$$

where $f_i = f(\xi_i)$.

Numerical solutions

- Numerical solution

$$u_h = \sum_{i=0}^{N-1} u_i \psi_i$$

- Local solution

$$u_h|_{T_n} = \sum_{i=0}^{N_n-1} u_i^n \psi_i^n$$

- Derivatives

$$\nabla u_h = \sum_{i=0}^{N-1} u_i \nabla \psi_i,$$

$$\nabla u_h|_{T_n} = \sum_{i=0}^{N_n-1} u_i^n \nabla \psi_i^n$$

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Variational formulation

For a test function $\psi_i(\mathbf{x}) \in V_h^k$, the variational formulation can be rewritten as

$$\sum_{j=0}^{N-1} u_j \int_{\Omega} \nabla \psi_i \cdot \nabla \psi_j \, d\mathbf{x} = \int_{\Omega} f \psi_i \, d\mathbf{x}.$$

Finite element system

$$\mathbf{A}\mathbf{u} = \mathbf{b}$$

where

$$(\mathbf{A})_{ij} = \int_{\Omega} \nabla \psi_{i-1} \cdot \nabla \psi_{j-1} \, d\mathbf{x}$$

$$(\mathbf{b})_i = \int_{\Omega} f \psi_{i-1} \, d\mathbf{x}$$

$$(\mathbf{u})_j = u_{j-1}$$

Finite element system

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

where

$$(\mathbf{A})_{ij} = \int_{\Omega} \nabla \psi_{i-1} \cdot \nabla \psi_{j-1} \, d\mathbf{x}$$

$$(\mathbf{u})_j = u_{j-1}$$

$$(\mathbf{M})_{ij} = \int_{\Omega} \psi_{i-1} \psi_{j-1} \, d\mathbf{x}$$

$$(\mathbf{f})_j = f_{j-1}$$

Matrices

$$\mathbf{A} = \sum_{n=1}^M \mathbf{A}_{T_n},$$
$$\mathbf{M} = \sum_{n=1}^M \mathbf{M}_{T_n}$$

where N_n by N_n matrices \mathbf{A}_{T_n} and \mathbf{M}_{T_n} are defined as

$$(\mathbf{A}_{T_n})_{ij} = \int_{T_n} \nabla \psi_{i-1}^n(x) \cdot \nabla \psi_{j-1}^n \, dx$$
$$(\mathbf{M}_{T_n})_{ij} = \int_{T_n} \psi_{i-1}^n \psi_{j-1}^n \, dx$$

where $1 \leq i, j \leq N_n$.

Lemma 2.1

For a given interval $T = \text{conv}\{\mathbf{v}_0, \mathbf{v}_1, \mathbf{v}_2\}$ with barycentric coordinates λ_0 , λ_1 and λ_2 , it holds for $a, b, c \in \mathbb{N}_0$ that

$$\int_T \lambda_0^a \lambda_1^b \lambda_2^c d\mathbf{x} = 2|T| \frac{a! b! c!}{(a + b + c + 2)!} \quad (1)$$

where $|T|$ is the area of T .

Differentiation matrix

$$(\mathbf{D}\mathbf{x})_{ij} = \frac{\partial \psi_{j-1}}{\partial x}(\xi_{i-1}), \quad (\mathbf{D}\mathbf{y})_{ij} = \frac{\partial \psi_{j-1}}{\partial y}(\xi_{i-1})$$

\Rightarrow

$$\frac{\partial}{\partial x} \psi_{i-1}(\mathbf{x}) = \sum_{j=0}^{N-1} \frac{\partial \psi_{i-1}}{\partial x}(\xi_j) \psi_j(\mathbf{x}) = (\mathbf{D}\mathbf{x}^t)_i \boldsymbol{\psi}$$

$$\frac{\partial}{\partial y} \psi_{i-1}(\mathbf{x}) = \sum_{j=0}^{N-1} \frac{\partial \psi_{i-1}}{\partial y}(\xi_j) \psi_j(\mathbf{x}) = (\mathbf{D}\mathbf{y}^t)_i \boldsymbol{\psi}$$

where

$$(\mathbf{D}\mathbf{x}^t)_i = \left[\frac{\partial \psi_{i-1}}{\partial x}(\xi_0) \cdots \frac{\partial \psi_{i-1}}{\partial x}(\xi_{N-1}) \right],$$

$$(\mathbf{D}\mathbf{y}^t)_i = \left[\frac{\partial \psi_{i-1}}{\partial y}(\xi_0) \cdots \frac{\partial \psi_{i-1}}{\partial y}(\xi_{N-1}) \right]$$

$$\boldsymbol{\psi} = [\psi_0(x) \cdots \psi_{N-1}(x)]^t$$

Properties of D

$$\bullet D\mathbf{x} = \sum_{n=0}^{2M^2-1} D\mathbf{x}_{T_n}, \quad (D\mathbf{x}_{T_n})_{ij} = \frac{\partial \psi_{j-1}^n}{\partial x}(\xi_{i-1}^n)$$

$$\bullet D\mathbf{y} = \sum_{n=0}^{2M^2-1} D\mathbf{y}_{T_n}, \quad (D\mathbf{y}_{T_n})_{ij} = \frac{\partial \psi_{j-1}^n}{\partial y}(\xi_{i-1}^n)$$

$$\bullet \nabla \psi_i^n(\mathbf{x}) = \left((D\mathbf{x}_{T_n}^t)_i \psi^n, (D\mathbf{y}_{T_n}^t)_i \psi^n \right)$$

$$\bullet \nabla u_h(\xi_m) = \left((D\mathbf{x})_m \mathbf{u}, (D\mathbf{y})_m \mathbf{u} \right)$$

$$\bullet \nabla u_h(\xi_m^n) = \left((D\mathbf{x}_{T_n})_m \mathbf{u}, (D\mathbf{y}_{T_n})_m \mathbf{u} \right)$$

Mass matrix

$$(\mathbf{M})_{ij} = \int_{R_n} \psi_{i-1}(\mathbf{x}) \psi_{j-1}(\mathbf{x}) d\mathbf{x} = J \int_{R_R} \tilde{\psi}_{i-1}(\mathbf{r}) \tilde{\psi}_{j-1}(\mathbf{r}) d\mathbf{r} = J(\mathbf{M}_R)_{ij}$$

$$\Rightarrow \quad \mathbf{M} = J\mathbf{M}_R$$

Stiffness matrix

$$(\mathbf{S})_{ij} = J \left[(r_x^2 + r_y^2)(\mathbf{S}_R^{rr})_{ij} + (r_x s_x + r_y s_y) \left((\mathbf{S}_R^{rs})_{ij} + (\mathbf{S}_R^{sr})_{ij} \right) + (s_x^2 + s_y^2)(\mathbf{S}_R^{ss})_{ij} \right]$$

where

$$(\mathbf{S}_R^{rr})_{ij} = (\mathbf{D} \mathbf{r}_R^t \mathbf{M}_R \mathbf{D} \mathbf{r}_R)_{ij}$$

$$(\mathbf{S}_R^{rs})_{ij} = (\mathbf{D} \mathbf{r}_R^t \mathbf{M}_R \mathbf{D} \mathbf{s}_R)_{ij}$$

$$(\mathbf{S}_R^{sr})_{ij} = (\mathbf{D} \mathbf{s}_R^t \mathbf{M}_R \mathbf{D} \mathbf{r}_R)_{ij}$$

$$(\mathbf{S}_R^{ss})_{ij} = (\mathbf{D} \mathbf{s}_R^t \mathbf{M}_R \mathbf{D} \mathbf{s}_R)_{ij}$$

Lemma

For a given interval $T = \text{conv}\{\mathbf{v}_0, \mathbf{v}_1, \mathbf{v}_2\}$ with barycentric coordinates λ_0 , λ_1 and λ_2 , it holds for $a, b, c \in \mathbb{N}_0$ that

$$\int_T \lambda_0^a \lambda_1^b \lambda_2^c d\mathbf{x} = 2|T| \frac{a! b! c!}{(a + b + c + 2)!}$$

where $|T|$ is area of T .

P_1 matrices

$$\widetilde{\psi}_0(\mathbf{r}) = \widetilde{\lambda}_2(\mathbf{r}), \quad \nabla \widetilde{\psi}_0(\mathbf{r}) = \left(-\frac{1}{2}, -\frac{1}{2} \right),$$

$$\widetilde{\psi}_1(\mathbf{r}) = \widetilde{\lambda}_0(\mathbf{r}), \quad \nabla \widetilde{\psi}_1(\mathbf{r}) = \left(\frac{1}{2}, 0 \right),$$

$$\widetilde{\psi}_2(\mathbf{r}) = \widetilde{\lambda}_1(\mathbf{r}), \quad \nabla \widetilde{\psi}_2(\mathbf{r}) = \left(0, \frac{1}{2} \right).$$

P_1 matrices

$$\mathbf{M}_R = \frac{1}{6} \begin{pmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{pmatrix}.$$

P_1 matrices

$$\mathbf{D}r_R = \frac{1}{2} \begin{pmatrix} -1 & 1 & 0 \\ -1 & 1 & 0 \\ -1 & 1 & 0 \end{pmatrix}$$

$$\mathbf{D}s_R = \frac{1}{2} \begin{pmatrix} -1 & 0 & 1 \\ -1 & 0 & 1 \\ -1 & 0 & 1 \end{pmatrix}.$$

P_1 matrices

$$\mathbf{S}_R^{rr} = \mathbf{D}\mathbf{r}_R^t \mathbf{M}_R \mathbf{D}\mathbf{r}_R = \frac{1}{2} \begin{pmatrix} 1 & -1 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\mathbf{S}_R^{rs} = \mathbf{D}\mathbf{r}_R^t \mathbf{M}_R \mathbf{D}\mathbf{s}_R = \frac{1}{2} \begin{pmatrix} 1 & 0 & -1 \\ -1 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\mathbf{S}_R^{sr} = \mathbf{D}\mathbf{s}_R^t \mathbf{M}_R \mathbf{D}\mathbf{r}_R = \frac{1}{2} \begin{pmatrix} 1 & -1 & 0 \\ 0 & 0 & 0 \\ -1 & 1 & 0 \end{pmatrix}$$

$$\mathbf{S}_R^{ss} = \mathbf{D}\mathbf{s}_R^t \mathbf{M}_R \mathbf{D}\mathbf{s}_R = \frac{1}{2} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{pmatrix}.$$

P_2 matrices

$$\begin{aligned}\tilde{\psi}_0(\mathbf{r}) &= \tilde{\lambda}_2(\mathbf{r})(2\tilde{\lambda}_2(\mathbf{r}) - 1), & \nabla\tilde{\psi}_0(\mathbf{r}) &= \left(-2\tilde{\lambda}_2(\mathbf{r}) + \frac{1}{2}, -2\tilde{\lambda}_2(\mathbf{r}) + \frac{1}{2}\right), \\ \tilde{\psi}_1(\mathbf{r}) &= 4\tilde{\lambda}_0(\mathbf{r})\tilde{\lambda}_2(\mathbf{r}), & \nabla\tilde{\psi}_1(\mathbf{r}) &= \left(2\tilde{\lambda}_2(\mathbf{r}) - 2\tilde{\lambda}_0(\mathbf{r}), -2\tilde{\lambda}_0(\mathbf{r})\right), \\ \tilde{\psi}_2(\mathbf{r}) &= \tilde{\lambda}_0(\mathbf{r})(2\tilde{\lambda}_0(\mathbf{r}) - 1), & \nabla\tilde{\psi}_2(\mathbf{r}) &= \left(2\tilde{\lambda}_0(\mathbf{r}) - \frac{1}{2}, 0\right), \\ \tilde{\psi}_3(\mathbf{r}) &= 4\tilde{\lambda}_1(\mathbf{r})\tilde{\lambda}_2(\mathbf{r}), & \nabla\tilde{\psi}_3(\mathbf{r}) &= \left(-2\tilde{\lambda}_1(\mathbf{r}), 2\tilde{\lambda}_1(\mathbf{r}) - 2\tilde{\lambda}_2(\mathbf{r})\right), \\ \tilde{\psi}_4(\mathbf{r}) &= 4\tilde{\lambda}_0(\mathbf{r})\tilde{\lambda}_1(\mathbf{r}), & \nabla\tilde{\psi}_4(\mathbf{r}) &= \left(2\tilde{\lambda}_1(\mathbf{r}), 2\tilde{\lambda}_0(\mathbf{r})\right), \\ \tilde{\psi}_5(\mathbf{r}) &= \tilde{\lambda}_1(\mathbf{r})(2\tilde{\lambda}_1(\mathbf{r}) - 1), & \nabla\tilde{\psi}_5(\mathbf{r}) &= \left(0, 2\tilde{\lambda}_1(\mathbf{r}) - \frac{1}{2}\right),\end{aligned}$$

P_2 matrices

$$\mathbf{M}_R = \frac{1}{90} \begin{pmatrix} 6 & 0 & -1 & 0 & -4 & -1 \\ 0 & 32 & 0 & 16 & 16 & -4 \\ -1 & 0 & 6 & -4 & 0 & -1 \\ 0 & 16 & -4 & 32 & 16 & 0 \\ -4 & 16 & 0 & 16 & 32 & 0 \\ -1 & -4 & -1 & 0 & 0 & 6 \end{pmatrix},$$

P_2 matrices

$$Dr_R = \frac{1}{2} \begin{pmatrix} -3 & 4 & -1 & 0 & 0 & 0 \\ -1 & 0 & 1 & 0 & 0 & 0 \\ 1 & -4 & 3 & 0 & 0 & 0 \\ -1 & 2 & -1 & -2 & 2 & 0 \\ 1 & -2 & 1 & -2 & 2 & 0 \\ 1 & 0 & -1 & -4 & 4 & 0 \end{pmatrix}$$

$$Ds_R = \frac{1}{2} \begin{pmatrix} -3 & 0 & 0 & 4 & 0 & -1 \\ -1 & -2 & 0 & 2 & 2 & -1 \\ 1 & -4 & 0 & 0 & 4 & -1 \\ -1 & 0 & 0 & 0 & 0 & 1 \\ 1 & -2 & 0 & -2 & 2 & 1 \\ 1 & 0 & 0 & -4 & 0 & 3 \end{pmatrix}$$

P_2 matrices

$$\mathbf{S}_R^{rr} = \frac{1}{6} \begin{pmatrix} 3 & -4 & 1 & 0 & 0 & 0 \\ -4 & 8 & -4 & 0 & 0 & 0 \\ 1 & -4 & 3 & 0 & 0 & 0 \\ 0 & 0 & 0 & 8 & -8 & 0 \\ 0 & 0 & 0 & -8 & 8 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\mathbf{S}_R^{rs} = \frac{1}{6} \begin{pmatrix} 3 & 0 & 0 & -4 & 0 & 1 \\ -4 & 4 & 0 & 4 & -4 & 0 \\ 1 & -4 & 0 & 0 & 4 & -1 \\ 0 & 4 & 0 & 4 & -4 & -4 \\ 0 & -4 & 0 & -4 & 4 & 4 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

P_2 matrices

$$\mathbf{S}_R^{sr} = \frac{1}{6} \begin{pmatrix} 3 & -4 & 1 & 0 & 0 & 0 \\ 0 & 4 & -4 & 4 & -4 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ -4 & 4 & 0 & 4 & -4 & 0 \\ 0 & -4 & 4 & -4 & 4 & 0 \\ 1 & 0 & -1 & -4 & 4 & 0 \end{pmatrix}$$

$$\mathbf{S}_R^{ss} = \frac{1}{6} \begin{pmatrix} 3 & 0 & 0 & -4 & 0 & 1 \\ 0 & 8 & 0 & 0 & -8 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ -4 & 0 & 0 & 8 & 0 & -4 \\ 0 & -8 & 0 & 0 & 8 & 0 \\ 1 & 0 & 0 & -4 & 0 & 3 \end{pmatrix}$$

This python code generates the mass matrix \mathbf{M}_R , the stiffness matrices \mathbf{S}_R^{rr} , \mathbf{S}_R^{rs} , \mathbf{S}_R^{sr} , \mathbf{S}_R^{ss} and the differentiation matrices $\mathbf{D}\mathbf{r}_R$, $\mathbf{D}\mathbf{s}_R$ for continuous k -th order polynomial approximations on the reference triangle T_R .

Programming

Assignments

1. Add the matrices for the cubic approximations ($k = 3$) in `get_matrices_2d_triangle` and check the convergence rate.

2. Modify `fem_for_poisson_2d_triangle_ex2` to solve the Poisson problem with non-homogeneous Dirichlet boundary condition,

$$\begin{aligned} -\Delta u(\mathbf{x}) &= f(\mathbf{x}) && \text{in } \Omega \\ u(\mathbf{x}) &= u_D(\mathbf{x}) && \text{on } \partial\Omega. \end{aligned}$$

3. Modify `fem_for_poisson_2d_ex3` to solve the Poisson problem with mixed boundary condition,

$$\begin{aligned} -\Delta u(\mathbf{x}) &= f(\mathbf{x}) && \text{in } \Omega \\ u(\mathbf{x}) &= u_D(\mathbf{x}) && \text{on } \Gamma_D \\ \nabla u(\mathbf{x}) \cdot \mathbf{n} &= u_N(\mathbf{x}) && \text{on } \Gamma_N, \end{aligned}$$

where Γ_D denotes the Dirichlet boundary, Γ_N denotes the Neumann boundary, and \mathbf{n} is the outward unit normal vector.

4. Prove Lemma 2.1.