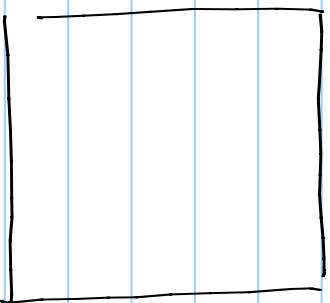


## 6.3 Mixed Method

Model Problem :  $-\Delta u = f$  in  $\Omega$

$$u|_{\partial\Omega} = 0$$

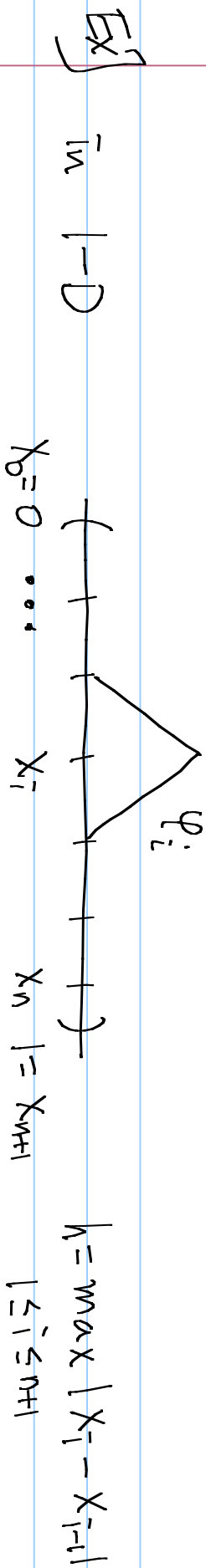


$$\Omega = (0,1)^2 \quad \text{in } 2D$$

$$\Omega = (0,1) \quad \text{in } 1D.$$

Find  $u \in H_0^1(\Omega)$  s.t  $a(u,v) = (f,v) \quad \forall v \in H_0^1(\Omega)$

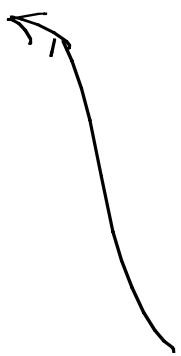
where  $a(u, v) = \int_{\Omega} \nabla u \nabla v \, dx$ .



$$V_h = \text{Span} \{ \varphi_1, \varphi_2, \dots, \varphi_n \} \quad \varphi_i(x_j) = \delta_{ij} = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$

$$u_h \in V_h \quad \text{s.t.} \quad a(u_h, v_h) = (f, v_h) \quad \forall v_h \in V_h$$

$$u_h = \sum_{i=1}^n u_h(x_i) \varphi_i := \underbrace{\sum_{i=1}^n u_i^h}_{u_i^h = u_h(x_i)} \varphi_i$$



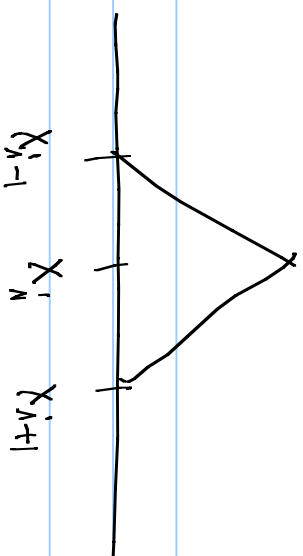
$$\Leftrightarrow a(u_h, \varphi_j) = (f, \varphi_j) \quad \forall j=1:n \quad \text{ith eq}$$

$$\Leftrightarrow \sum_{i=1}^n u_i^T a(\varphi_i, \varphi_j) = (f, \varphi_j) = \int_{\Omega} f \varphi_j dx = f_j^D$$

$$A_h = \left( (A_h)_{ij}^n \right)_{i,j}^n = (A_h)_{ij}^n = a(\varphi_j, \varphi_i) \quad \forall i,j=1:n$$

Ex]  $(1-D)$   $A_h = \begin{pmatrix} 2/h & -1/h & & 0 \\ -1/h & 2/h & -1/h & \\ 0 & -1/h & \ddots & \ddots \\ \ddots & \ddots & -1/h & 2/h \end{pmatrix}$   $f_j^D = \int_0^1 f \varphi_j dx$   
 $x_i = ih$

$$(A_n)_{i,i} = \int_0^1 \varphi_i' \varphi_i' dx$$



$$= \int_{x_{i-1}}^{x_i} \frac{1}{h} dx + \int_{x_i}^{x_{i+1}} \frac{1}{h} dx = \frac{1}{h} \cdot h + \frac{1}{h} \cdot h = \frac{2}{h}$$

$\tilde{S}$ th local eq :

$$A_n U = F$$

$$U =$$

$$\begin{pmatrix} u^1 \\ \vdots \\ u^n \end{pmatrix}$$

$$F = \begin{pmatrix} f^1 \\ \vdots \\ f^n \end{pmatrix}$$

$$(A_n U)_j = f_j^i$$

$$\textcircled{1} \quad \frac{-U^{j+1} + 2U^j - U^{j-1}}{h} = f_j^s \quad j=1:n.$$

For given  $U_k = (U_k^1, U_k^2, \dots, U_k^n)$ ,  $k$ th iterate

Solve for  $U_{k+1}^j$

$$\textcircled{2} \quad \frac{-U_k^{j+1} + 2U_{k+1}^j - U_k^{j-1}}{h} = f_j^s \quad j=1:n$$

→ Jacobi method

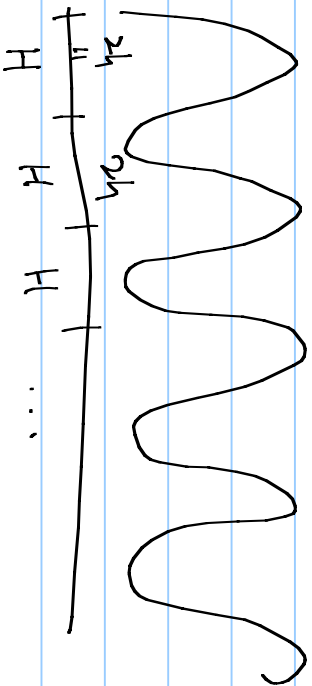
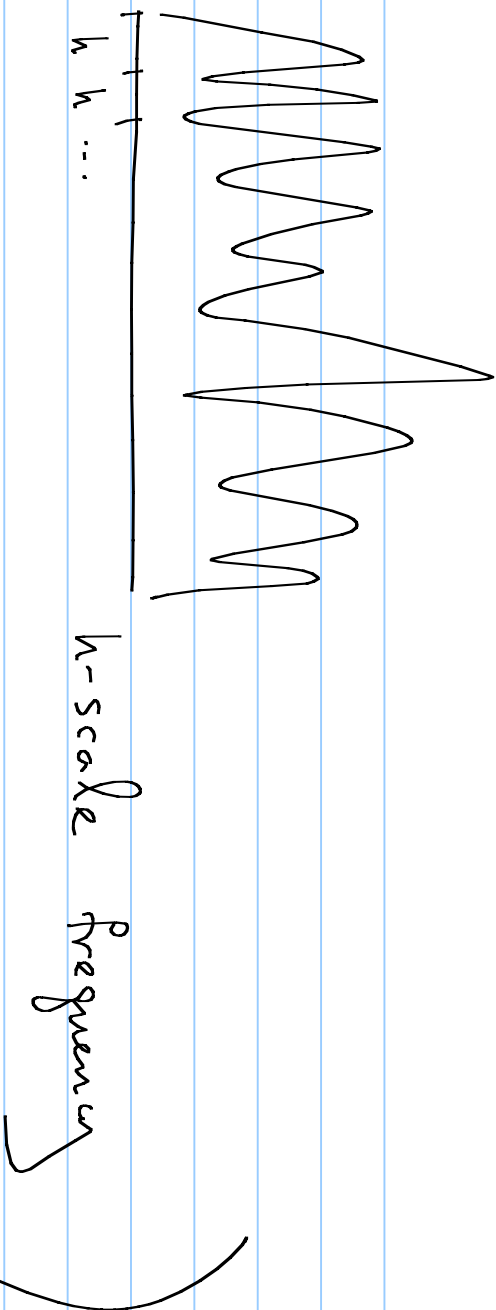
Properties of Jacobi method

$$e_k^j = U_k^j - U_k^{j-1} \quad \forall j=1:n$$

$$\frac{-e_k^{j+1} + 2e_k^j - e_k^{j-1}}{h} = 0.$$

$$e_k^{j+1} = \frac{e_k^j + e_k^{j-1}}{2}.$$

R.K.]



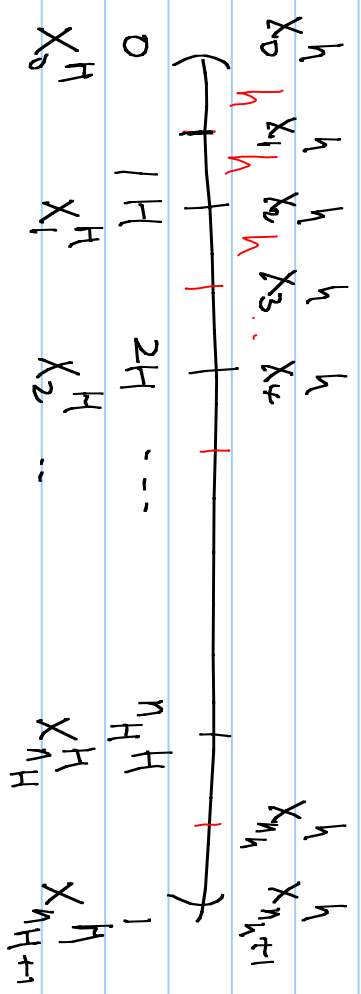
After applying a  
classical iterative method

$$P = \sum_{n=1}^m c_n k^n \quad \mu_n \quad n\text{-scale frequency}$$

can be handled by Jenkins efficiently. However, smooth component of error is difficult to handle.

### Ingredient I]

[Ex]  $1-D, \quad \Omega = (0, 1)$





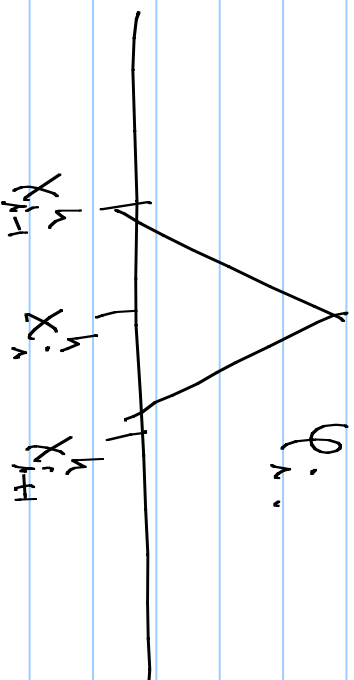
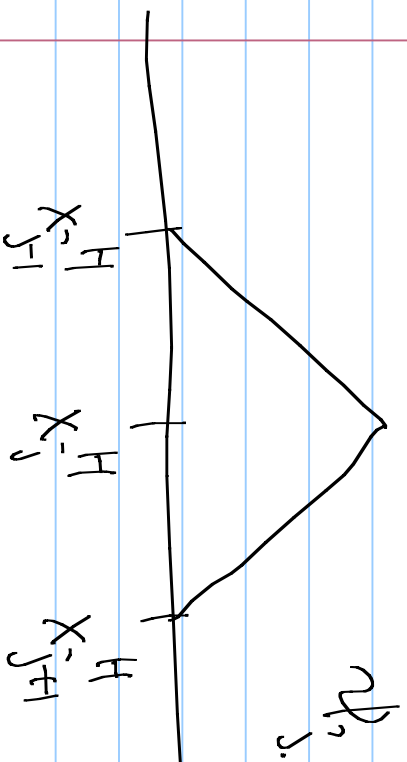
$$\mathcal{O}_H = \sum_{j=1}^{n_{H+1}} \mathcal{O}_j^H \quad \mathcal{O}_j^H = \begin{pmatrix} x_j^H & x_{j-1}^H \end{pmatrix} \quad \bar{\Sigma} = \bigcup_{j=1}^{n_{H+1}} \bar{\Sigma}_j^H.$$

$\mathcal{O}_n$ : obtained by forming 2 subtriangles in each triangle in  $\mathcal{O}_H$

$$= \sum_{j=1}^{n_{n+1}} \mathcal{O}_j^n \quad \mathcal{O}_j^n = (x_j^n, x_{j-1}^n) \quad \bar{\Sigma} = \bigcup_{j=1}^{n_{n+1}} \bar{\Sigma}_j^n.$$

$$V_H = \text{span} \{ \mathcal{P}_1, \mathcal{P}_2, \dots, \mathcal{P}_{n_H} \} \quad \text{where}$$

$$\mathcal{P}_j \text{ is a hat fit s.t. } \mathcal{P}_j(x_i^H) = \delta_{ij}.$$

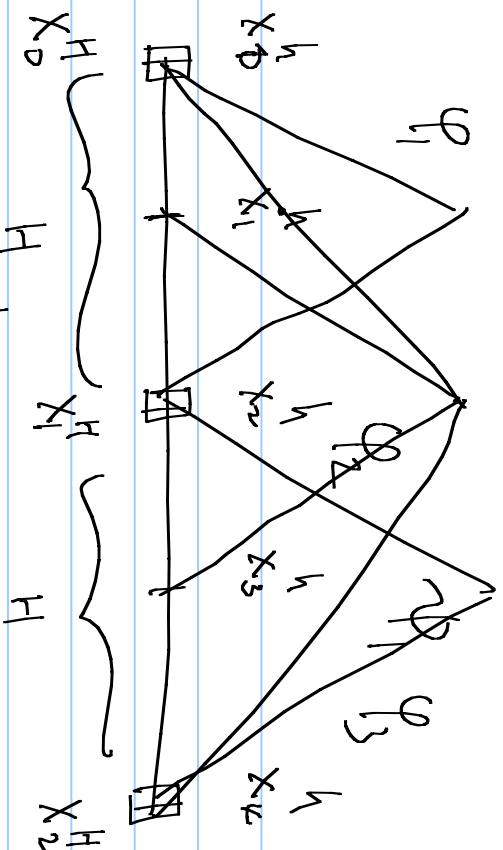


$V_n = \text{Span} \{ \varphi_1, \varphi_2, \dots, \varphi_{m_n} \}$ , where

$$\varphi_i \text{ is a hat fct s.t } \varphi_i(x_j^N) = \delta_{ij}.$$

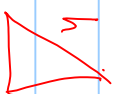
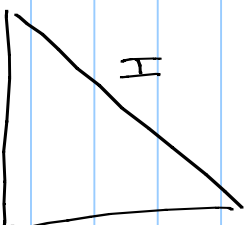
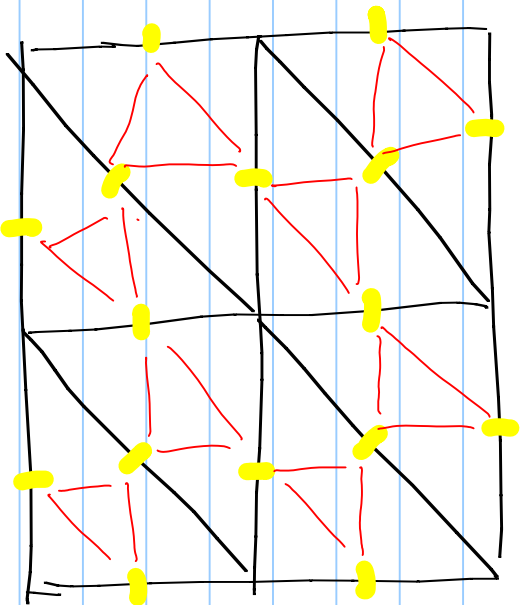
$$V_H \subset V_n$$

why.



Choose  $\varphi_1 = \frac{1}{2} \varphi_1 + \varphi_2 + \frac{1}{2} \varphi_3$ .

$$\Rightarrow \forall w_H \in V_H, \quad w_H \in V_n$$



$$H = 2h$$

Notation:  $X^h = \{ X_n^h \}_{n=1}^{N_h}$

Nodes for  $\mathcal{T}_h$

$$X^H = \{ X_n^H \}_{n=1}^{N_H}$$

Nodes for  $\mathcal{T}_H$ .

Let  $V_n = \text{span}\{\phi_1, \phi_2, \dots, \phi_{n_H}\}$  and  $V_H = \text{span}\{\psi_1, \dots, \psi_{n_H}\}$  where  $\phi_i$ 's usual basis (hat ft) defined on  $\mathcal{F}_n$ . s.t

$\phi_i(x_j^n) = \delta_{ij}$ .  $\forall i, j$ . and  $\psi_i$ 's usual basis (hat ft) defined on  $\mathcal{F}_H$ .

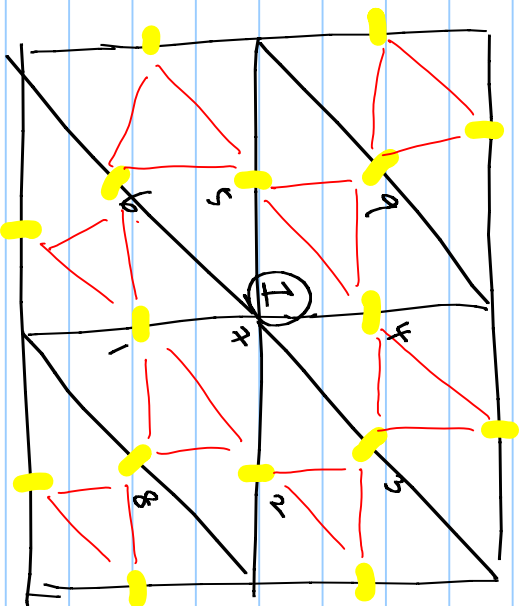
s.t  $\psi_i(x_j^H) = \delta_{ij}$ .  $\forall i, j$ .  $\Rightarrow V_H \subset V_n$ .

Under this settings  $\forall v \in V_n$  and  $w_H \in V_H$  can be expressed as

$$v_n = \sum_{i=1}^{n_H} v_n(x_i^n) \phi_i \quad \text{and} \quad w_H = \sum_{i=1}^{n_H} w_H(x_i^H) \psi_i$$

$$V_H \subset V_h$$

[Ex] 2-D cone      Only 1 - interior node for  $\mathcal{P}_1$ .



$$\mathcal{P}_1 = C_1^1 \varphi_1 + C_1^2 \varphi_2 + \dots + C_1^9 \varphi_9.$$

$$C_1^1 = \frac{1}{2}, \quad C_1^2 = \frac{1}{2}, \quad C_1^3 = \frac{1}{2},$$

$$C_1^4 = \frac{1}{2}, \quad C_1^5 = \frac{1}{2}, \quad C_1^6 = \frac{1}{2},$$

$$C_1^7 = 1, \quad C_1^8 = C_1^9 = 0.$$

$$U_h \in V_h, \quad U_h = \sum_{j=1}^{n_h} u_j^{\tilde{}} \phi_j^{\tilde{}}, \quad \tilde{u} = \begin{pmatrix} u_1^{\tilde{}} \\ \vdots \\ u_{n_h}^{\tilde{}} \end{pmatrix} \in \mathbb{R}^{n_h}$$

Goal: To solve  $A_h U_h = \tilde{f}_h$ .

Ingredient (II). With  $B_h \approx A_h^T$ ,  $(A_h)_j^i = a(\varphi_j, \varphi_i)$ .

$$\tilde{u}_h^{KH} = \tilde{u}_h^k + B_h (\tilde{f}_h - A_h \tilde{u}_h^k).$$

$$\begin{pmatrix} \tilde{p}_h \\ \tilde{f}_h \end{pmatrix}_j = (f, \varphi_j^{\tilde{}}).$$

Recall Classical Iterative Method to solve

$$Ax = b, \Leftrightarrow x = A^{-1}b. \quad A \text{ spd}$$

Step 1] Given  $x_i$  with iterate

compute the residual  $r_i = b - Ax_i$ .

Form the residual eq



$$\textcircled{*} \quad Ae = v_n.$$

Step 2] Solve  $\textcircled{*}$  approximately,

$$\hat{e} = Bv_n, \quad \text{where } B \approx A^{-1}.$$

Step 3] Update  $x_{i+1} = x_i + \hat{e}.$

If  $B = A^{-1}$ , then  $X_{i,t+1} = X_i$ .

why?)  $X_{i,t+1} = \cancel{X_i} + A^{-1}(b - \cancel{A}X_i) = A^{-1}b = X_i$

$$A = D - L - L^T,$$

$$B = D^{-1} \rightarrow \text{Jacobi} \quad B = \omega I.$$

$$B = (D-L)^{-1} \rightarrow \text{Gauss-Seidel} \quad \text{— Richardson}$$

## Two grids method

Let  $u_n^k$  be an  $k$ -th iterate approximating

$$u_n \text{ in } V_{h_n}$$

$$T. \quad \tilde{u}_n^k = \text{Smoothing} (u_n^k, A_{h_n}, f_{h_n}, \# \text{Smoothing})$$

For  $j = \# \text{Smoothing}$  (1 or 2)

$$\tilde{u}_n^k = u_n^k + B_{h_n} (f_{h_n} - A_{h_n} u_n^k)$$

$$\begin{aligned} \tilde{u}_h &\stackrel{\text{End}}{\approx} \tilde{u}_h^k, & \bar{u}_h &= \sum_{j=1}^{m_h} \bar{u}_h^j \phi_j \\ u_h &= u_h, \end{aligned}$$

2, Coarse-grid correction: Compute the solution  $w_H \in V_H$  to the problem

: Find  $w_H \in V_H$  s.t

$$a(\bar{u}_h + w_H, v_H) = (f, v_H) \quad \forall v_H \in V_H.$$

$$\Rightarrow u_h^{k+1} = \bar{u}_h + w_H.$$

and obtain  $u_n^{k+1}$ , a vector representation of  $u_n^{k+1}$ .

Ingredient (II) - (For a convenient implementation)

Define intergrid transfer operator.

$$I_n^H : V_H \rightarrow V_h, \quad I_n^H : V_h \rightarrow V_H.$$

$\uparrow$  prolongation  
 $\uparrow$  restriction.

Note  $\forall j, 1 \leq j \leq N_H$ .

$$\mathcal{P}_j = \sum_{n=1}^{M_H} c_j^n \phi_n.$$

$$c_j =$$

$$\begin{bmatrix} c_j^1 \\ c_j^2 \\ \vdots \\ c_j^{M_H} \end{bmatrix}.$$

$$\mathcal{P} = [c_1, c_2, \dots, c_{N_H}] \in \mathbb{R}^{M_H \times N_H}.$$

For any given  $w_H \in V_H$

$$w_H = \sum_{j=1}^{M_H} w_{Hj}^H \phi_j = \sum_{j=1}^{M_H} \tilde{w}_j^H \phi_j.$$

$$\begin{aligned} &= \sum_{j=1}^{M_H} \tilde{w}_j^H \sum_{k=1}^{M_H} c_{jk}^H \phi_k = \sum_{k=1}^{M_H} \left[ \sum_{j=1}^{M_H} \tilde{w}_j^H c_{jk}^H \right] \phi_k \\ &= \end{aligned}$$

$$I_n^H W_H = \sum_{i=1}^{M_H} [P_{\tilde{W}}]_i \phi_i \quad \tilde{W} = \begin{pmatrix} W_1 \\ \vdots \\ W_{M_H} \end{pmatrix} \in V_H.$$

Note that

$$Q(\bar{u}_H + W_H, \sigma_H) = (f, \sigma_H) \quad \forall \sigma_H \in V_H.$$



$$a(W_H, \psi_j^r) = \underbrace{(f, \psi_j^r)}_{M_H} - \underbrace{a(\bar{w}_H, \psi_j^r)}_{j=1: M_H},$$

Set  $W_H = \sum_{j=1}^{M_H} w_j^i \psi_j^r$ .

$$a(\psi_i, \psi_j^r) = (P^T A_H P)_{ij} \quad (A_H)_{kl} = a(\varphi_k, \varphi_l)$$

$$(f, \psi_j^r) = (f, \sum_{k=1}^{M_H} C_j^k \phi_k^r) = \sum_{k=1}^{M_H} C_j^k (f, \phi_k^r) =$$

$[P^T \tilde{r}]_{j_i}$  , To compute  $a(\bar{u}_n, \psi_j)$  ,

$$a(\phi_i, \psi_j) = a(\phi_i, \sum_{k=1}^{m_n} c_j^k \phi_k)$$

$$\begin{aligned} &= \sum_{k=1}^{m_n} c_j^k a(\phi_i, \phi_k) = (P^T A_n)_{j_i} \Rightarrow \\ &= \sum_{k=1}^{m_n} c_j^k a(\phi_i, \phi_k) = (P^T A_n)_{j_i} \Rightarrow \end{aligned}$$

$$\alpha(\bar{u}_n, \mathcal{P}_1) = (P^T A_n \tilde{u})_j, \quad \text{where}$$

$\tilde{u}$  is the vector representation of  $\bar{u}_n$ .

$$(P^T A_n P) \tilde{w} = [P^T f_n - P^T A_n \tilde{u}]$$