

# MULTIGRID LECTURE NOTES <sup>\*†</sup>

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## Abstract

Multigrid (MG) methods in numerical analysis are a group of algorithms for solving differential equations using a hierarchy of discretizations. MG method is based on the two facts: High frequency will be damped by smoother and Low frequency can be approximated well by coarse grid, since the low-frequency errors on a fine mesh can become high-frequency errors on a coarser mesh. MG methods can be used as solvers as well as preconditioners.

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## 0 Preliminaries

Let  $\mathcal{M}_{m \times n} := \mathcal{M}_{m \times n}(\mathbb{C})$  be the set of all matrices with  $m$  rows and  $n$  columns and  $\mathcal{M}_{m \times n}(\mathbb{R})$  be the subset of  $\mathcal{M}_{m \times n}$  composed of matrices with only real entries. Denote by  $\mathcal{M}_n := \mathcal{M}_{n \times n}$  the set of all square matrices of size  $n \times n$ , and by  $\mathcal{M}_n(\mathbb{R})$  the subset of  $\mathcal{M}_n$  composed of matrices with only real entries.

### 0.1 Vector norms

**Definition 0.1. (Vector Norms)** A vector norm is a function  $\|\cdot\| : \mathbb{R}^n \rightarrow \mathbb{R}$  satisfying the following conditions for all  $x, y \in \mathbb{R}^n$  and  $\alpha \in \mathbb{R}$

1. *nonnegative* :  $\|x\| \geq 0$ , ( $\|x\| = 0 \Leftrightarrow x = 0$ ),
2. *homogeneity* :  $\|\alpha x\| = |\alpha| \|x\|$ ,
3. *triangle inequality* :  $\|x + y\| \leq \|x\| + \|y\|$ ,  $\forall x, y \in \mathbb{R}^n$ ,

### 0.2 Matrix norms

**Definition 0.2. (Matrix Norms)** A matrix norm is a function  $\|\cdot\| : \mathbb{R}^{m \times n} \rightarrow \mathbb{R}$  satisfying the following conditions for all  $A, B \in \mathbb{R}^{m \times n}$  and  $\alpha \in \mathbb{R}$

1. *nonnegative* :  $\|A\| \geq 0$ , ( $\|A\| = 0 \Leftrightarrow A = 0$ ),  $\forall A \in \mathcal{M}_n$ ,
2. *homogeneity* :  $\|\lambda A\| = |\lambda| \|A\|$ ,  $\forall \lambda \in \mathbb{C}$  and  $\forall A \in \mathcal{M}_n$
3. *triangle inequality* :  $\|A + B\| \leq \|A\| + \|B\|$ ,  $\forall A, B \in \mathcal{M}_n$ ,
4. *submultiplicativity* :  $\|AB\| \leq \|A\| \|B\|$ ,  $\forall A, B \in \mathcal{M}_n$ .

**Definition 0.3.** For  $A \in \mathbb{R}^{m \times n}$ , some of the most frequently matrix vector norms are

1. *F-norm* :  $\|A\|_F = \sqrt{\sum_{i=1}^m \sum_{j=1}^n |a_{ij}|^2}$ ,
2. *1-norm* :  $\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^m |a_{ij}|$ ,
3.  *$\infty$ -norm* :  $\|A\|_\infty = \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}|$ ,
4. *induced-norm* :  $\|A\|_p = \sup_{x \in \mathbb{R}^n, x \neq 0} \frac{\|Ax\|_p}{\|x\|_p}$ .

**Lemma 0.4.** Let  $\|\cdot\|$  is a matrix norm on  $\mathcal{M}_n$ , then

$$\|A^k\| \leq \|A\|^k$$

*Proof.*

$$\|A^k\| = \sup_{0 \neq x \in \mathbb{C}^n} \frac{\|A^k x\|}{\|x\|} \leq \sup_{0 \neq x \in \mathbb{C}^n} \frac{\|A\| \|A^{k-1} x\|}{\|x\|} \leq \dots \leq \|A\|^k.$$

□

**Theorem 0.5. (Neumann Series)** Suppose that  $A \in \mathbb{R}^{n \times n}$ . If  $\|A\| < 1$ , then  $(I - A)$  is nonsingular and

$$(I - A)^{-1} = \sum_{k=0}^{\infty} A^k \quad (0.1)$$

with

$$\frac{1}{1 + \|A\|} \leq \|(I - A)^{-1}\| \leq \frac{1}{1 - \|A\|}. \quad (0.2)$$

Moreover, if  $A$  is nonnegative, then  $(I - A)^{-1} = \sum_{k=0}^{\infty} A^k$  is also nonnegative.

*Proof.* 1.  $(I - A)$  is nonsingular, i.e.  $(I - A)^{-1}$  exists.

$$\begin{aligned} \|(I - A)x\| &\geq \|Ix\| - \|Ax\| \\ &\geq \|x\| - \|A\| \|x\| \\ &= (1 - \|A\|) \|x\| \\ &= C \|x\|. \end{aligned}$$

So, we get if  $(I - A)x = 0$ , then  $x = 0$ . Therefore,  $\ker(I - A) = \{0\}$ , then  $(I - A)^{-1}$  exists.

2. Let  $S_N = \sum_{k=0}^N A^k$ , we want to show  $(I - A)S_N \rightarrow I$ , as  $N \rightarrow \infty$ .

$$(I - A)S_N = S_N - AS_N = \sum_{k=0}^N A^k - \sum_{k=1}^{N+1} A^k = A^0 - A^{N+1} = I - A^{N+1}.$$

So by Lemma 0.4

$$\|(I - A)S_N - I\| = \| -A^{N+1} \| \leq \|A\|^{N+1}.$$

Since  $\|A\| < 1$ , then  $\|A\|^{N+1} \rightarrow 0$ . Therefore,

$$(I - A) \sum_{k=0}^{\infty} A^k = I.$$

and

$$(I - A)^{-1} = \sum_{k=0}^{\infty} A^k$$

3. bounded norm

Since

$$1 = \|I\| = \|(I - A) * (I - A)^{-1}\|.$$

So,

$$(1 - \|A\|) \|(I - A)^{-1}\| \leq 1 \leq (1 + \|A\|) \|(I - A)^{-1}\|.$$

Therefore,

$$\frac{1}{1 + \|A\|} \leq \|(I - A)^{-1}\| \leq \frac{1}{1 - \|A\|}.$$

□

**Lemma 0.6.** Suppose that  $A \in \mathbb{R}^{n \times n}$ . If  $(I - A)$  is singular, then  $\|A\| \geq 1$ .

*Proof.* Converse-negative proposition of If  $\|A\| < 1$ , then  $(I - A)$  is nonsingular.  $\square$

**Theorem 0.7.** Let  $A$  be a nonnegative matrix. then  $\rho(A) < 1$  if only if  $I - A$  is nonsingular and  $(I - A)^{-1}$  is nonnegative.

*Proof.* 1.  $\Rightarrow$  By theorem (0.5).

2.  $\Leftarrow$  since  $I - A$  is nonsingular and  $(I - A)^{-1}$  is nonnegative, by the Perron- Frobenius theorem, there is a nonnegative eigenvector  $u$  associated with  $\rho(A)$ , which is an eigenvalue, i.e.

$$Au = \rho(A)u$$

or

$$(I - A)^{-1}u = \frac{1}{1 - \rho(A)}u.$$

since  $I - A$  is nonsingular and  $(I - A)^{-1}$  is nonnegative, this show that  $1 - \rho(A) > 0$ , which implies

$$\rho(A) < 1.$$

$\square$

### 0.3 Eigenvalues

#### 0.3.1 Eigenvalues of Tridiagonal Matrices

**Theorem 0.8.** (*Eigenvalues of Tridiagonal Matrices*) If  $A = \text{diag}(b, a, b) \in \mathcal{M}_n$  is an tridiagonal matrix, then the eigenvalues of  $A$  are

$$\lambda_k = a + 2b \cos(\theta_k), \quad k = 1, 2, 3, \dots, N$$

and its corresponding eigenvector are

$$\vec{\xi}_k = \sqrt{2} (\sin(1\theta_k), \sin(2\theta_k), \dots, \sin(N\theta_k))$$

where

$$\theta_k = k\theta = k\pi h = \frac{k\pi}{N+1}.$$

*Proof.* It can be easily verified by the trigonometric identities

$$\sin(2\theta_k) = 2 \sin(\theta_k) \cos(\theta_k),$$

and

$$2 \sin(ki\theta_k) \cos(k\theta_k) = \sin(k(i-1)\theta_k) + \sin(k(i+1)\theta_k).$$

$\square$

### 0.3.2 Eigenvalues of Hermitians Matrices

**Definition 0.9.** (*Hermitian Matrix*) A matrix is *Hermitian* , if

$$A^H = A.$$

**Definition 0.10.** Let  $A$  be *Hermitian* , then the spectral of  $A$ ,  $\sigma(A)$ , is real.

*Proof.* Let  $\lambda \in \sigma(A)$  with corresponding eigenvector  $v$ . Then

$$\begin{aligned} \langle Av, v \rangle &= \langle \lambda v, v \rangle = \lambda \langle v, v \rangle \\ \langle Av, v \rangle &= \langle v, A^H v \rangle = \langle v, \bar{\lambda} v \rangle = \bar{\lambda} \langle v, v \rangle. \end{aligned}$$

Since  $\langle v, v \rangle \neq 0$ , therefore  $\lambda = \bar{\lambda}$ . Hence  $\lambda$  is real. □

**Definition 0.11.** Let  $A$  be *Hermitian* , then the different eigenvector are orthogonal i.e.

$$\langle v_i, v_j \rangle = 0, i \neq j. \quad (0.3)$$

*Proof.* Let  $\lambda_1, \lambda_2$  be the arbitrary two different eigenvalues with corresponding eigenvector  $v_1, v_2$ . Then

$$\begin{aligned} \langle Av_1, v_2 \rangle &= \langle \lambda_1 v_1, v_2 \rangle = \lambda_1 \langle v_1, v_2 \rangle \\ \langle Av_1, v_2 \rangle &= \langle v_1, A^H v_2 \rangle = \langle v_1, \lambda_2 v_2 \rangle = \lambda_2 \langle v_1, v_2 \rangle. \end{aligned}$$

Since  $\lambda_1 \neq \lambda_2$ , therefore  $\langle v_1, v_2 \rangle = 0$ . □

**Theorem 0.12.** (*Spectral Theorem for Hermitian matrices*)  $A$  is *Hermitian* , then  $A$  is *unitary diagonalizable* .

$$A = UDU^{-1} = UDU^H, \quad (0.4)$$

where  $U$  is a unitary matrix ,  $D$  is an diagonal matrix .

### 0.4 Spectral radius

**Definition 0.13.** The spectral radius of a matrix  $A \in \mathcal{M}_n$ , defined as

$$\rho(A) := \max \{ |\lambda|, \lambda \text{ eigenvalue of } A \}.$$

**Theorem 0.14.** If  $\|\cdot\|$  is a matrix norm on  $\mathcal{M}_n$ , then for any  $A \in \mathcal{M}_n$ ,

$$\rho(A) \leq \|A\|.$$

*Proof.* Let  $\lambda$  be an eigenvalue of  $A$ , and let  $x \neq 0$  be a corresponding eigenvector. Then  $Ax = \lambda x$ , we have

$$AX = \lambda X, \text{ where } X := [x | \cdots | x] \in \mathcal{M}_n \setminus \{0\}.$$

By the properties of Matrix norm, then we have

$$|\lambda| \|X\| = \|\lambda X\| = \|AX\| \leq \|A\| \|X\|.$$



Since,  $\|X\| \neq 0$ , then we have

$$|\lambda| \leq \|A\|.$$

Taking the maximum over all eigenvalues  $\lambda$  yields

$$\rho(A) \leq \|A\|.$$

□

**Theorem 0.15.** If  $A = A^*$ , then  $\rho(A) = \|A\|_2$ .

*Proof.* Since  $A$  is self-adjoint, there an orthonormal basis of eigenvector  $x \in \mathbb{C}^n$ , s.t.

$$x = \alpha_1 e_1 + \alpha_2 e_2 + \cdots + \alpha_n e_n.$$

Moreover,  $Ae_i = \lambda_i e_i$ ,  $\|e_i\| = 1$  and  $(e_i, e_j) = 0$  when  $i \neq j$ ,  $(e_j, e_j) = 1$ . So,

$$\|x\|_{\ell^2}^2 = \sum_{i=1}^n |\alpha_i|^2,$$

since,

$$\begin{aligned} (x, x) &= \left( \sum_{i=1}^n \alpha_i e_i, \sum_{j=1}^n \alpha_j e_j \right) \\ &= \sum_{i=1}^n \sum_{j=1}^n \alpha_i \bar{\alpha}_j e_i e_j \\ &= \sum_{i=1}^n |\alpha_i|^2. \end{aligned}$$

Since,  $Ax = A(\alpha_1 e_1 + \alpha_2 e_2 + \cdots + \alpha_n e_n) = \alpha_1 \lambda_1 e_1 + \alpha_2 \lambda_2 e_2 + \cdots + \alpha_n \lambda_n e_n$ , then

$$\|Ax\|_{\ell^2}^2 = \sum_{i=1}^n |\lambda_i \alpha_i|^2 = \sum_{i=1}^n |\lambda_i|^2 |\alpha_i|^2 \leq \max\{|\lambda_i|\}^2 \sum_{i=1}^n |\alpha_i|^2.$$

Therefore,

$$\|Ax\|_{\ell^2} \leq \rho(A) \|x\|_{\ell^2},$$

i.e.

$$\|A\|_2 = \sup_{x \in \mathbb{C}^n} \frac{\|Ax\|_{\ell^2}}{\|x\|_{\ell^2}} \leq \rho(A).$$

Let  $k$  be the index, s.t:  $|\lambda_n| = \rho(A)$  and  $x = e_k$ ,  $Ax = Ae_k = \lambda_n e_k$ , so  $\|Ax\|_{\ell^2} = |\lambda_n| = \rho(A)$  and

$$\|A\|_2 = \sup_{x \in \mathbb{C}^n} \frac{\|Ax\|_{\ell^2}}{\|x\|_{\ell^2}} \geq \frac{\|Ax\|_{\ell^2}}{\|x\|_{\ell^2}} = \rho(A).$$

□

## 0.5 Hermitian and Symmetric

**Definition 0.16.** (*Hermitian and Symmetric*) Let  $\vec{u}, \vec{v} \in \mathbb{C}^n$ . Define

$$(\vec{u}, \vec{v}) := \vec{u}^H \vec{v} = \sum_{i=1}^n \bar{u}_i v_i.$$

1. Let  $A \in \mathbb{C}^{n \times n}$ , we say that  $A$  is *Hermitian* iff

$$A = A^H,$$

where  $A^H \in \mathbb{C}^{n \times n}$  is the matrix satisfying

$$(\vec{u}, A\vec{v}) = (A^H \vec{u}, \vec{v}),$$

for all  $\vec{u}, \vec{v} \in \mathbb{C}^n$ .

2.  $A \in \mathbb{R}^{n \times n}$  is called *Symmetric* iff

$$A = A^T,$$

where  $A^T \in \mathbb{R}^{n \times n}$  is the matrix satisfying

$$(\vec{u}, A\vec{v}) = (A^T \vec{u}, \vec{v}),$$

for all  $\vec{u}, \vec{v} \in \mathbb{R}^n$ .

**Definition 0.17.** (*HPD and SPD*)  $A$  is called *Hermitian Positive Definite (HPD)* iff  $A = A^H$  and

$$(\vec{u}, A\vec{u}) > 0, \quad \forall \vec{u} \in \mathbb{C}_*^n := \mathbb{C}^n \setminus \{\vec{0}\}.$$

And  $A$  is called *Symmetric Positive Definite (SPD)* iff  $A = A^T$  and

$$(\vec{u}, A\vec{u}) > 0, \quad \forall \vec{u} \in \mathbb{R}_*^n := \mathbb{R}^n \setminus \{\vec{0}\}.$$

**Remark 0.18.** Of course, if  $A \in \mathbb{R}^{n \times n}$ , then HPD and SPD mean the same things. We also have the usual simple formulas for the Hermitian and real transposes. For  $A \in \mathbb{C}^{n \times n}$ ,

$$[A^H]_{ij} = \overline{[A]_{ji}},$$

and if  $A \in \mathbb{R}^{n \times n}$

$$A^H = A^T.$$

**Definition 0.19. (Orthogonal and Orthonormal)** Two vectors  $\vec{u}, \vec{v} \in \mathbb{C}^n$  are called *orthogonal* (Orthogonal in canonical inner product) iff

$$(\vec{u}, \vec{v}) = 0.$$

The set  $S = \{\vec{u}_1, \dots, \vec{u}_\ell\} \in \mathbb{C}^n$  is called *orthogonal* iff

$$(\vec{u}_i, \vec{u}_j) = 0, \quad 1 \leq i, j \leq \ell, \quad i \neq j.$$

$S$  is called *orthonormal* iff it is orthogonal and

$$(\vec{u}_i, \vec{u}_i) = 1, \quad 1 \leq i \leq \ell.$$

In this case,

$$(\vec{u}_i, \vec{u}_j) = \delta_{ij}, \quad 1 \leq i, j \leq \ell.$$

**Theorem 0.20. (Properties of Hermitian)** Suppose that  $A \in \mathbb{C}^{n \times n}$  is Hermitian, i.e.  $A = A^H$ ,

1. All of the eigenvalues are real,
2. Eigenvector of  $A$  associated to distinct eigenvalues of  $A$  are orthogonal,
3.  $A$  has a full set of  $n$  eigenvectors that forms an orthonormal basis for  $\mathbb{C}^n$ .

*Proof.* 1. Let  $\lambda \in \sigma(A)$  with corresponding eigenvector  $\vec{v}$ . Then

$$\begin{aligned} \langle A\vec{v}, \vec{v} \rangle &= \langle \lambda\vec{v}, \vec{v} \rangle = \lambda \langle \vec{v}, \vec{v} \rangle \\ \langle A\vec{v}, \vec{v} \rangle &= \langle \vec{v}, A^H \vec{v} \rangle = \langle \vec{v}, \bar{\lambda}\vec{v} \rangle = \bar{\lambda} \langle \vec{v}, \vec{v} \rangle. \end{aligned}$$

Since  $\langle \vec{v}, \vec{v} \rangle \neq 0$ , therefore  $\lambda = \bar{\lambda}$ . Hence  $\lambda$  is real.

2. Let  $\lambda_1, \lambda_2$  be the arbitrary two different eigenvalues with corresponding eigenvector  $\vec{v}_1, \vec{v}_2$ . Then

$$\begin{aligned} \langle A\vec{v}_1, \vec{v}_2 \rangle &= \langle \lambda_1 \vec{v}_1, \vec{v}_2 \rangle = \lambda_1 \langle \vec{v}_1, \vec{v}_2 \rangle \\ \langle A\vec{v}_1, \vec{v}_2 \rangle &= \langle \vec{v}_1, A^H \vec{v}_2 \rangle = \langle \vec{v}_1, A\vec{v}_2 \rangle = \langle \vec{v}_1, \lambda_2 \vec{v}_2 \rangle = \lambda_2 \langle \vec{v}_1, \vec{v}_2 \rangle. \end{aligned}$$

Since  $\lambda_1 \neq \lambda_2$ , therefore  $\langle \vec{v}_1, \vec{v}_2 \rangle = 0$ .

3. Gram-Schmidt Orthonormalization. □

**Theorem 0.21.** Let  $A \in \mathbb{C}^{n \times n}$  be Hermitian,

1. If  $A$  is HPD then the eigenvalues of  $A$  are positive.
2. Conversely, if the eigenvalues are positive, then  $A$  is HPD.

*Proof.* 1. Let  $(\lambda, \vec{v})$  be arbitrary eigen-pair of  $A$ , then we have

$$\vec{v}^H A \vec{v} = \vec{v}^H \lambda \vec{v} = \lambda \vec{v}^H \vec{v} = \lambda(v_1^2 + v_2^2 + \dots + v_n^2) > 0.$$

for all  $\vec{v} \neq 0$ . Hence,  $\lambda$  is positive.

2. Since  $A \in \mathbb{C}_{her}^{n \times n}$ , then the eigenvalue of A are real. Let  $\lambda$  be arbitrary eigenvalue of A, then

$$\begin{aligned} (A\vec{v}, \vec{v}) &= (\lambda\vec{v}, \vec{v}) = \lambda(\vec{v}, \vec{v}), \\ (A\vec{v}, \vec{v}) &= (\vec{v}, A^H \vec{v}) = (\vec{v}, A\vec{v}) = (\vec{v}, \lambda\vec{v}) = \bar{\lambda}(\vec{v}, \vec{v}), \end{aligned}$$

and then  $\lambda = \bar{\lambda}$ , so  $\lambda$  is real. Moreover, we have  $\lambda$  is positive, so

$$\vec{v}^H A \vec{v} = \vec{v}^H \lambda \vec{v} = \lambda \vec{v}^H \vec{v} = \lambda(v_1^2 + v_2^2 + \cdots + v_n^2) > 0.$$

for all  $\vec{v} \neq 0$ . Hence, A is Hermitian Positive Definite. □

**Definition 0.22. (*A-inner product*)** Let  $A \in \mathbb{C}^{n \times n}$  be HPD. Define for all  $\vec{u}, \vec{v} \in \mathbb{C}^n$

$$(\vec{u}, \vec{v}) := (A\vec{u}, \vec{v}).$$

**Theorem 0.23.** 1. Let  $A \in \mathbb{C}^{n \times n}$  be HPD. Then  $(\cdot, \cdot)_A$  defines a bona-fide inner product on  $\mathbb{C}^n$ .  
2. Conversely, suppose  $(\cdot, \cdot)_*$  is an inner product on  $\mathbb{C}^n$ . Then there exists an HPD matrix (w.r.t. the canonical inner product)  $C \in \mathbb{C}^{n \times n}$  such that

$$(\vec{u}, \vec{v})_C = (\vec{u}, \vec{v})_*, \quad \forall \vec{u}, \vec{v} \in \mathbb{C}^n.$$

**Theorem 0.24.** Suppose that A and B are  $n \times n$  HPD matrices with respect to  $(\cdot, \cdot)$ . Then BA is HPD with respect to  $(\cdot, \cdot)_A$  and  $(\cdot, \cdot)_{B^{-1}}$ .

*Proof.* 1. Hermitian-ness:

$$\begin{aligned} (BA\vec{u}, \vec{v})_A &= (BA\vec{u}, A\vec{v}) \\ &= (A\vec{u}, BA\vec{v}) \\ &= (\vec{u}, BA\vec{v})_A, \end{aligned}$$

for all  $\vec{u}, \vec{v} \in \mathbb{C}^n$ .

2. Positive Definite-ness: Suppose  $\vec{u} \in \mathbb{C}_*^n$ ,  $\vec{z} := A\vec{u}$ . Then  $\vec{z} \in \mathbb{C}_*^n$ . Hence

$$\begin{aligned} (BA\vec{u}, \vec{u})_A &= (BA\vec{u}, A\vec{u}) \\ &= (A\vec{z}, \vec{z}) \\ &> 0. \quad (\vec{z} \neq \vec{0}) \end{aligned}$$

The proof for  $(\cdot, \cdot)_{B^{-1}}$  is similar. □

# 1 Classical Iterative Methods

In this section, we shall discuss the classical linear iterative methods on solving the linear operator equation

$$A\vec{u} = \vec{f}, \quad (1.1)$$

which arises from the Finite Difference Method (FDM) or Finite Element Method (FEM). We will assume that  $A$  is invertible, at least, often will assume that  $A$  is HPD.

## 1.1 General Iterative Methods

**Definition 1.1.** (*General Iterative Scheme*) A general linear two layer iterative scheme reads

$$B_k^{-1} \left( \frac{\vec{u}^{k+1} - \vec{u}^k}{\alpha_k} \right) + A\vec{u}^k = \vec{f}.$$

1.  $\alpha_k \in \mathbb{R}, B_k^{-1} \in \mathbb{C}^{n \times n}$ —iterative parameters
2. If  $\alpha_k = \alpha, B_k^{-1} = B^{-1}$ , then the method is stationary.
3. If  $B_k^{-1} = I$ , then the method is explicit.

From now on, we consider the stationary scheme ( $\alpha > 0$ ), i.e

$$B^{-1} \left( \frac{\vec{u}^{k+1} - \vec{u}^k}{\alpha} \right) + A\vec{u}^k = \vec{f}. \quad (1.2)$$

Then we get

$$\vec{u}^{k+1} = \vec{u}^k + \alpha B(\vec{f} - A\vec{u}^k). \quad (1.3)$$

**Definition 1.2.** (*General Linear Iterative Scheme (GLIS)*) Suppose that  $B \in \mathbb{C}^{n \times n}$  consider the iteration: Given  $\vec{u}^0$ , find  $\vec{u}^1, \vec{u}^2, \dots$ , such that

$$\vec{u}^{k+1} = \vec{u}^k + B(\vec{f} - A\vec{u}^k). \quad (1.4)$$

This is known as a *General Linear Iterative Scheme (GLIS)*.

**Remark 1.3.** The application of  $B$  should be simple and cheap (computationally inexpensive). In some sense  $B$  should approximate  $A^{-1}$ .

**Remark 1.4.** In the General Linear Iteration Scheme (GLIS) (1.4), the matrix  $B$  which should approximate  $A^{-1}$  is often called a preconditioner of  $A$ , especially, when  $A$  and  $B$  are both HPD. Suppose  $\vec{u}^0 = \vec{0}$ . Then

$$\vec{u}^1 = B\vec{f}.$$

Thus the action of the preconditioner  $B$  on a vector  $f$  is equivalent to 1 iteration of the GLIS with  $\vec{u}^0 = \vec{0} \in \mathbb{C}^n$ . A GLIS (where  $B$  is HPD) may be used (viewed as) a precondition for a Krylov method. A "good" preconditioned has the property that

$$\kappa(BA) = \frac{\lambda_n(BA)}{\lambda_1(BA)} = \mathcal{O}(1).$$

where

$$0 < \lambda_1(BA) \leq \lambda_2(BA) \leq \dots \leq \lambda_n(BA),$$

are the eigenvalues of the HPD matrix  $BA$ . (This is HPD w.r.t.  $(\cdot, \cdot)_A$  and  $(\cdot, \cdot)_{B^{-1}}$ .)

**Definition 1.5.** (*Error Transfer Operator of the GLIS*) Let  $\vec{u}$  be the exact solution and  $\vec{u}^k$  be the approximate solution at  $k$  step. Then we define *error*, *residual* and *error Transfer Operator* as follows:

$$\text{Error} : \vec{e}^k := \vec{u} - \vec{u}^k. \quad (1.5)$$

$$\text{Residual} : \vec{r}^k := \vec{f} - A\vec{u}^k. \quad (1.6)$$

$$\text{Error transfer operator} : \vec{e}^{k+1} := (I - BA)\vec{e}^k := T\vec{e}^k. \quad (1.7)$$

**Remark 1.6.** Since from the general iterative scheme we can get

$$\begin{aligned} \vec{u} &= \vec{u} + B(\vec{f} - A\vec{u}), \\ \vec{u}^{k+1} &= \vec{u}^k + B(\vec{f} - A\vec{u}^k). \end{aligned}$$

Then, subtracting the above two equations yields

$$\vec{e}^{k+1} = \vec{e}^k - BA\vec{e}^k = (I - BA)\vec{e}^k := T\vec{e}^k.$$

$T = I - BA$  is the error transfer operator.

**Definition 1.7.** (*Energy norm w.r.t A*) The *Energy norm* associated with  $A$  is

$$\|\vec{u}\|_A = (A\vec{u}, \vec{u});$$

**Theorem 1.8.** (*convergence in energy norm*) Suppose  $A$  is HPD and  $B$  is invertible. If  $Q := B^{-1} - \frac{\alpha}{2}A > 0$ , then GLIS (1.3) converges with  $A$  norm, i.e.

$$\|\vec{e}^k\|_A \rightarrow 0.$$

*Proof.* Let  $\vec{u}$  be the exact solution and  $\vec{u}^k$  be the approximate solution at  $k$  step, then

$$\begin{aligned} \vec{u} &= \vec{u} + \alpha B(\vec{f} - A\vec{u}), \\ \vec{u}^{k+1} &= \vec{u}^k + \alpha B(\vec{f} - A\vec{u}^k). \end{aligned}$$

From the definite of Error (1.5), we have

$$\begin{aligned} \vec{e}^{k+1} &= \vec{e}^k + \alpha B(\vec{f} - A\vec{u} - \vec{f} + A\vec{u}^k) \\ &= \vec{e}^k - \alpha BA\vec{e}^k. \end{aligned}$$

Let  $\vec{v}^{k+1} = \vec{e}^{k+1} - \vec{e}^k$ , then

$$\frac{1}{\alpha}B^{-1}\vec{v}^{k+1} + A\vec{e}^k = 0.$$

Taking inner product with  $\vec{v}^{k+1}$  gives

$$\frac{1}{\alpha}(B^{-1}\vec{v}^{k+1}, \vec{v}^{k+1}) + (A\vec{e}^k, \vec{v}^{k+1}) = 0.$$

Since

$$\vec{e}^k = \frac{1}{2}(\vec{e}^{k+1} + \vec{e}^k) - \frac{1}{2}(\vec{e}^{k+1} - \vec{e}^k) = \frac{1}{2}(\vec{e}^{k+1} + \vec{e}^k) - \frac{1}{2}\vec{v}^{k+1},$$

then

$$\begin{aligned} 0 &= \frac{1}{\alpha}(B^{-1}\vec{v}^{k+1}, \vec{v}^{k+1}) + (A\vec{e}^k, \vec{v}^{k+1}) \\ &= \frac{1}{\alpha}(B^{-1}\vec{v}^{k+1}, \vec{v}^{k+1}) + \frac{1}{2}(A(\vec{e}^{k+1} + \vec{e}^k), \vec{v}^{k+1}) - \frac{1}{2}(A\vec{v}^{k+1}, \vec{v}^{k+1}) \\ &= \frac{1}{\alpha}((B^{-1} - \frac{\alpha}{2}A)\vec{v}^{k+1}, \vec{v}^{k+1}) + \frac{1}{2}(A(\vec{e}^{k+1} + \vec{e}^k), \vec{v}^{k+1}) \\ &= \frac{1}{\alpha}((B^{-1} - \frac{\alpha}{2}A)\vec{v}^{k+1}, \vec{v}^{k+1}) + \frac{1}{2}(A(\vec{e}^{k+1} + \vec{e}^k), \vec{e}^{k+1} - \vec{e}^k) \\ &= \frac{1}{\alpha}((B^{-1} - \frac{\alpha}{2}A)\vec{v}^{k+1}, \vec{v}^{k+1}) + \frac{1}{2}(\|\vec{e}^{k+1}\|_A^2 - \|\vec{e}^k\|_A^2) \end{aligned}$$

By assumption,  $Q := B^{-1} - \frac{\alpha}{2}A > 0$ , i.e. there exists a  $m > 0$ , s.t.

$$(Q\vec{u}, \vec{u}) \geq m\|\vec{u}\|_2^2.$$

Therefore,

$$\frac{m}{\alpha}\|\vec{v}^{k+1}\|_2^2 + \frac{1}{2}(\|\vec{e}^{k+1}\|_A^2 - \|\vec{e}^k\|_A^2) \leq 0.$$

i.e.

$$\frac{2m}{\alpha}\|\vec{v}^{k+1}\|_2^2 + \|\vec{e}^{k+1}\|_A^2 \leq \|\vec{e}^k\|_A^2.$$

Hence

$$\|\vec{e}^{k+1}\|_A^2 \leq \|\vec{e}^k\|_A^2.$$

and

$$\|\vec{e}^{k+1}\|_A^2 \rightarrow 0.$$

□

**Definition 1.9. (Convergence)** Let  $C \in \mathbb{C}^{n \times n}$ .  $C$  is called convergent iff for every  $\epsilon > 0$ , there exists a  $K \in \mathbb{N}$  such that if  $k > K$  then for all  $1 \leq i, j \leq n$

$$\left| [C^k]_{i,j} \right| \leq \epsilon.$$

We have the following equivalences:

**Theorem 1.10.** *The following statements are equivalent*

1.  $C \in \mathbb{C}^{n \times n}$  is convergent,
- 2.

$$\lim_{k \rightarrow \infty} \|C^k\| = 0,$$

*for some induced matrix norm (operator norm),*

- 3.

$$\lim_{k \rightarrow \infty} \|C^k\| = 0,$$

*for all induced matrix norms.*

4.  $\rho(C) < 1$ , where

$$\rho(C) = \max_{1 \leq i \leq n} |\lambda_i|,$$

*and  $\lambda_i \in \mathbb{C}$ ,  $1 \leq i \leq n$ , are the eigenvalues of  $C$ .*

- 5.

$$\lim_{k \rightarrow \infty} C^k \vec{x} = \vec{0},$$

*for every  $\vec{x} \in \mathbb{C}^n$ .*

Thus we have the following convergence results.

**Theorem 1.11.** *(Sufficient & necessary condition for convergence) The sufficient & necessary condition for convergence of (1.4) is*

$$\rho(T) < 1,$$

*for any starting vector  $\vec{u}^0 \in \mathbb{C}^n$ . Where  $\rho(T)$  is the spectral radius of  $T = I - BA$ .*

*Proof.* Let  $\vec{u}$  be the exact solution and  $\vec{u}^k$  be the approximate solution at  $k$  step. Then, from Error transfer operator, we have  $\vec{e}^{k+1} = T\vec{e}^k$ . Hence

$$\vec{e}^k = T^k \vec{e}^0.$$

- $\Rightarrow$  Suppose (1.4) converges (it must converge to  $\vec{u} \in \mathbb{C}^n$ , incidentally). Then

$$\lim_{k \rightarrow \infty} \vec{e}^k = \vec{0},$$

for any  $\vec{u}^0 \in \mathbb{C}^n$ . Let  $(\lambda, \vec{w}) \in \mathbb{C} \times \mathbb{C}^n$  be the eigen-pair of  $T$  and set  $\vec{e}^0 = \vec{w}$ . Then

$$\vec{e}^k = \lambda^k \vec{e}^0.$$

This implies that  $|\lambda| < 1$ . Since  $\lambda$  was an arbitrary eigenvalue,

$$\rho(T) < 1$$



- $\Leftarrow$  If  $\rho(T) < 1$ , then T by theorem 1.10,

$$\lim_{k \rightarrow \infty} T^k \vec{x} = \vec{0},$$

for any  $\vec{x} \in \mathbb{C}^n$ . Hence

$$\lim_{k \rightarrow \infty} \vec{e}^k = \lim_{k \rightarrow \infty} T^k \vec{e}^0 = \vec{0},$$

□

**Theorem 1.12. (Sufficient condition for convergence)** The sufficient condition for convergence of (1.4) is

$$\|T\| < 1.$$

Where  $\|\cdot\|$  is any induced matrix norm.

*Proof.* By theorem 0.14, we have

$$\rho(T) \leq \|T\| < 1.$$

Using theorem 1.11 gives the result. □

**Theorem 1.13. (Error estimates)** If (1.3) is convergent, then we have

1.  $\|\vec{u} - \vec{u}^k\| \leq \|T\|^k \|\vec{u} - \vec{u}^0\|$
2.  $\|\vec{u} - \vec{u}^k\| \leq \frac{\|T\|^k}{1 - \|T\|} \|\vec{u}^1 - \vec{u}^0\|$

*Proof.* 1. By Lemma 0.4

$$\begin{aligned} \|\vec{e}^k\| &= \|T^k \vec{e}^0\| \\ &\leq \|T^k\| \|\vec{e}^0\| \\ &\leq \|T\|^k \|\vec{e}^0\|, \end{aligned}$$

i.e.

$$\|\vec{u} - \vec{u}^k\| \leq \|T\|^k \|\vec{u} - \vec{u}^0\|.$$

2. Now, write

$$\begin{aligned} \|\vec{e}^0\| &= \|\vec{u} - \vec{u}^0\| \\ &= \|\vec{u} - \vec{u}^1 + \vec{u}^1 - \vec{u}^0\| \\ &\leq \|\vec{u} - \vec{u}^1\| + \|\vec{u}^1 - \vec{u}^0\| \end{aligned}$$

From part 1, we have

$$\|\vec{u} - \vec{u}^1\| \leq \|T\| \|\vec{u} - \vec{u}^0\| = \|T\| \|\vec{e}^0\|.$$

Therefore,

$$\|\vec{e}^0\| \leq \|T\| \|\vec{e}^0\| + \|\vec{u}^1 - \vec{u}^0\|.$$

Hence

$$\|\vec{e}^0\| \leq \frac{1}{1-\|T\|} \|\vec{u}^1 - \vec{u}^0\|.$$

Substituting into first estimate yields

$$\|\vec{e}^k\| \leq \|T\|^k \|\vec{u} - \vec{u}^0\| = \|T\|^k \|\vec{e}^0\| \leq \frac{\|T\|^k}{1-\|T\|} \|\vec{u}^1 - \vec{u}^0\|.$$

□

Let  $A \in \mathbb{C}^{n \times n}$ , we start with the following matrix decomposition

$$A = D - L - U, \quad (1.8)$$

where  $D$  is the diagonal of  $A$ ,  $-L$  and  $-U$  are the strict lower part and the strict upper part, respectively, as illustrated in (Figure.1).

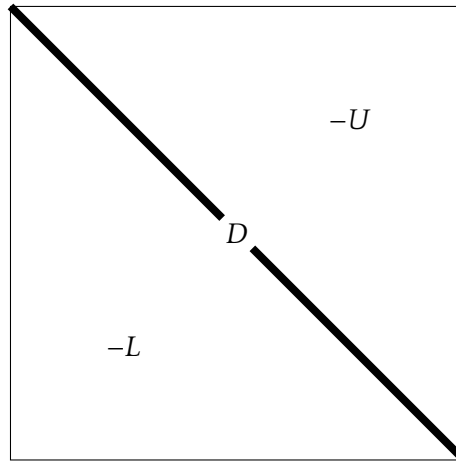


Figure 1: The initial decomposition of matrix  $A$ .

If  $A$  is HPD, then  $D$  is real and  $U = L^H$ .

## 1.2 Jacobi Iterative Method

Assume  $D$  is invertible, then the linear system (1.1) can be rewritten as

$$D\vec{u} = (U + L)\vec{u} + \vec{f}.$$

**Definition 1.14.** (*Jacobi Method*) Define

$$D\vec{u}^{k+1} = (U + L)\vec{u}^k + \vec{f}.$$

for a given starting value  $\vec{u}^0 \in \mathbb{C}^n$ , then Jacobi Method scheme reads

$$\vec{u}^{k+1} = \vec{u}^k + D^{-1}(\vec{f} - A\vec{u}^k). \quad (1.9)$$

**Remark 1.15.** Since the matrix decomposition (1.8) and

$$D\vec{u}^{k+1} = (U + L)\vec{u}^k + \vec{f},$$

we have

$$\begin{aligned}\vec{u}^{k+1} &= D^{-1}(U + L)\vec{u}^k + D^{-1}\vec{f} \\ &= D^{-1}(D - A)\vec{u}^k + D^{-1}\vec{f} \\ &= \vec{u}^k + D^{-1}(\vec{f} - A\vec{u}^k).\end{aligned}$$

Hence Jacobi Method is a GLIS and  $B_J = D^{-1}$ . Moreover, from the definition of (1.14), we have

$$D(\vec{u}^{k+1} - \vec{u}^k) + (L + D + U)\vec{u}^k = L\vec{u}^k + D\vec{u}^{k+1} + U\vec{u}^k = \vec{f}.$$

So, the Jacobi iterative method can be written as

$$\sum_{j<i} a_{ij}u_j^k + a_{ii}u_i^{k+1} + \sum_{j>i} a_{ij}u_j^k = f_i,$$

or

$$u_i^{k+1} = \frac{1}{a_{ii}} \left( f_i - \sum_{j \neq i} a_{ij}u_j^k \right).$$

**Definition 1.16.** (*Preconditioner and Error Transfer Operator for Jacobi Method*) The preconditioner of Jacobi Method is

$$B_J = D^{-1}.$$

And the error transfer operator for Jacobi Method is as follows

$$T_J = I - D^{-1}A.$$

**Remark 1.17.** Since  $B_J = D^{-1}$  and the definition of the error transform operator (1.5), we have

$$T_J = I - D^{-1}A.$$

**Theorem 1.18.** (*convergence of the Jacobi Method*) If  $A$  is *diagonal dominant*, then the Jacobi Method convergences.

*Proof.* We want to show If  $A$  is *diagonal dominant*, then  $\|T_J\| < 1$ , then Jacobi Method convergences. From the definition of  $T$ , we know that  $T$  for Jacobi Method is as follows

$$T_J = I - D^{-1}A.$$

In the matrix form is

$$T = \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix} - \begin{pmatrix} \frac{1}{a_{11}} & & 0 \\ & \ddots & \\ 0 & & \frac{1}{a_{nn}} \end{pmatrix} \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{pmatrix} = [t_{ij}] = \begin{cases} t_{ij} = 0, & i = j, \\ t_{ij} = -\frac{a_{ij}}{a_{ii}}, & i \neq j. \end{cases}$$

So,

$$\|T\|_{\infty} = \max_i \sum_j |t_{ij}| = \max_i \sum_{i \neq j} \left| \frac{a_{ij}}{a_{ii}} \right|.$$

Since A is diagonal dominant, so

$$|a_{ii}| \geq \sum_{j \neq i} |a_{ij}| + \delta.$$

Therefore,

$$1 \geq \sum_{j \neq i} \frac{|a_{ij}|}{|a_{ii}|} + \frac{\delta}{|a_{ii}|}.$$

Hence,  $\|T\|_{\infty} < 1$  □

**Definition 1.19.** (*Weighted Jacobi Method*) Assume  $D$  in (1.8) is invertible, define for  $0 < \omega \leq 1$ ,

$$\begin{aligned} \vec{z} &= \vec{u}^k + D^{-1}(\vec{f} - A\vec{u}^k) \\ \vec{u}^{k+1} &= \omega \vec{z} + (1 - \omega) \vec{u}^k. \end{aligned}$$

for a given starting value  $\vec{u}^0 \in \mathbb{C}^n$ , then *Weighted Jacobi Method* scheme reads

$$\vec{u}^{k+1} = \vec{u}^k + \omega D^{-1}(\vec{f} - A\vec{u}^k). \quad (1.10)$$

**Remark 1.20.** *Weighted Jacobi Method* is a GLIS with

$$B_{WJ} = \omega D^{-1}.$$

### 1.3 Gauss-Seidel Method

Assume  $D - L$  is invertible, define

$$\begin{aligned} \vec{u}^{k+1} &= (D - L)^{-1} U \vec{u}^k + (D - L)^{-1} \vec{f} \\ &= \vec{u}^k + (D - L)^{-1} (\vec{f} - A\vec{u}^k). \end{aligned}$$

**Definition 1.21.** (*Forward Gauss-Seidel Method*) Suppose  $D - L$  is invertible, *Forward Gauss-Seidel Method* scheme reads

$$\vec{u}^{k+1} = \vec{u}^k + B_{GS}(\vec{f} - A\vec{u}^k),$$

where

$$B_{GS} = (D - L)^{-1}.$$

**Definition 1.22.** (*Error Transfer Operator of Forward Gauss-Seidel Method*) The error transfer operator for *Forward Gauss-Seidel Method* is as follows

$$T_{GS} = (D - L)^{-1} U.$$

**Remark 1.23.**

$$\begin{aligned} T_{GS} &= I - (D - L)^{-1}A \\ &= I - (D - L)^{-1}(D - L - U) \\ &= (D - L)^{-1}U. \end{aligned}$$

**Theorem 1.24.** (*convergence of the Forward Gauss-Seidel Method*) If  $A$  is *diagonal dominant*, then the Forward Gauss-Seidel Method converges.

*Proof.* We want to show If  $A$  is *diagonal dominant*, then  $\|T_{GS}\| < 1$ , then the Forward Gauss-Seidel Method converges. From the definition of  $T$ , we know that  $T$  for Forward Gauss-Seidel Method is as follows

$$T_{GS} = (D - L)^{-1}U.$$

Next, we will show  $\|T_{GS}\| < 1$ . Since  $A$  is diagonal dominant, so

$$|a_{ii}| \geq \sum_{j \neq i} |a_{ij}| + \delta = \sum_{j > i} |a_{ij}| + \sum_{j < i} |a_{ij}| + \delta.$$

So,

$$|a_{ii}| - \sum_{j < i} |a_{ij}| \geq \sum_{j > i} |a_{ij}| + \delta,$$

which implies

$$\gamma = \max_i \left\{ \frac{\sum_{j > i} |a_{ij}|}{|a_{ii}| - \sum_{j < i} |a_{ij}|} \right\} \leq 1.$$

Now, we will show  $\|T_{GS}\| < \gamma$ . Let  $x \in \mathbb{C}^n$  and  $y = Tx$ , i.e.

$$y = T_{GS}x = (D - L)^{-1}Ux.$$

Let  $i_0$  be the index such that  $\|y\|_\infty = |y_{i_0}|$ , then we have

$$|((D - L)y)_{i_0}| = |(Ux)_{i_0}| = \left| \sum_{j > i_0} a_{i_0j}x_j \right| \leq \sum_{j > i_0} |a_{i_0j}| |x_j| \leq \sum_{j > i_0} |a_{i_0j}| \|x\|_\infty.$$

Moreover

$$|((D - L)y)_{i_0}| = \left| \sum_{j < i_0} a_{i_0j}y_j + a_{i_0i_0}y_{i_0} \right| \geq |a_{i_0i_0}y_{i_0}| - \left| \sum_{j < i_0} a_{i_0j}y_j \right| = |a_{i_0i_0}| \|y\|_\infty - \left| \sum_{j < i_0} a_{i_0j}y_j \right| \geq |a_{i_0i_0}| \|y\|_\infty - \sum_{j < i_0} |a_{i_0j}| \|y\|_\infty.$$

Therefore, we have

$$|a_{i_0i_0}| \|y\|_\infty - \sum_{j < i_0} |a_{i_0j}| \|y\|_\infty \leq \sum_{j > i_0} |a_{i_0j}| \|x\|_\infty,$$

which implies

$$\|y\|_\infty \leq \frac{\sum_{j > i_0} |a_{i_0j}|}{|a_{i_0i_0}| - \sum_{j < i_0} |a_{i_0j}|} \|x\|_\infty.$$

So,

$$\|T_{GS}x\|_{\infty} \leq \gamma \|x\|_{\infty},$$

which implies

$$\|T_{GS}\|_{\infty} \leq \gamma < 1.$$

□

**Definition 1.25. (Backward Gauss-Seidel Method)** Suppose  $D - U$  is invertible, *Backward Gauss-Seidel Method* scheme reads

$$\vec{u}^{k+1} = \vec{u}^k + B_{BGS}(\vec{f} - A\vec{u}^k),$$

where

$$B_{BGS} = (D - U)^{-1}.$$

**Definition 1.26. (Error Transfer Operator of Backward Gauss-Seidel Method)** The error transfer operator for Backward Gauss-Seidel Method is as follows

$$T_{BGS} = (D - U)^{-1}L.$$

**Remark 1.27.**

$$\begin{aligned} T_{BGS} &= I - (D - U)^{-1}A \\ &= I - (D - U)^{-1}(D - L - U) \\ &= (D - U)^{-1}L. \end{aligned}$$

**Theorem 1.28. (convergence of the Backward Gauss-Seidel Method)** If  $A$  is *diagonal dominant*, then the Backward Gauss-Seidel Method converges.

*Proof.* We want to show If  $A$  is *diagonal dominant*, then  $\|T_{BGS}\| < 1$ , then the Forward Gauss-Seidel Method converges. From the definition of  $T$ , we know that  $T$  for Forward Gauss-Seidel Method is as follows

$$T_{BGS} = (D - U)^{-1}L.$$

Next, we will show  $\|T_{BGS}\| < 1$ . Since  $A$  is diagonal dominant, so

$$|a_{ii}| \geq \sum_{j \neq i} |a_{ij}| + \delta = \sum_{j > i} |a_{ij}| + \sum_{j < i} |a_{ij}| + \delta.$$

So,

$$|a_{ii}| - \sum_{j > i} |a_{ij}| \geq \sum_{j < i} |a_{ij}| + \delta,$$

which implies

$$\gamma = \max_i \left\{ \frac{\sum_{j < i} |a_{ij}|}{|a_{ii}| - \sum_{j > i} |a_{ij}|} \right\} \leq 1.$$

Now, we will show  $\|T_{BGS}\| < \gamma$ . Let  $x \in \mathbb{C}^n$  and  $y = Tx$ , i.e.

$$y = T_{BGS}x = (D - U)^{-1}Lx.$$

Let  $i_0$  be the index such that  $\|y\|_\infty = |y_{i_0}|$ , then we have

$$|((D - U)y)_{i_0}| = |(Lx)_{i_0}| = \left| \sum_{j < i_0} a_{i_0 j} x_j \right| \leq \sum_{j < i_0} |a_{i_0 j}| |x_j| \leq \sum_{j < i_0} |a_{i_0 j}| \|x\|_\infty.$$

Moreover

$$|((D - U)y)_{i_0}| = \left| \sum_{j > i_0} a_{i_0 j} y_j + a_{i_0 i_0} y_{i_0} \right| \geq |a_{i_0 i_0} y_{i_0}| - \left| \sum_{j > i_0} a_{i_0 j} y_j \right| = |a_{i_0 i_0}| \|y\|_\infty - \left| \sum_{j > i_0} a_{i_0 j} y_j \right| \geq |a_{i_0 i_0}| \|y\|_\infty - \sum_{j > i_0} |a_{i_0 j}| \|y\|_\infty.$$

Therefore, we have

$$|a_{i_0 i_0}| \|y\|_\infty - \sum_{j > i_0} |a_{i_0 j}| \|y\|_\infty \leq \sum_{j < i_0} |a_{i_0 j}| \|x\|_\infty,$$

which implies

$$\|y\|_\infty \leq \frac{\sum_{j < i_0} |a_{i_0 j}|}{|a_{i_0 i_0}| - \sum_{j > i_0} |a_{i_0 j}|} \|x\|_\infty.$$

So,

$$\|T_{BGS}x\|_\infty \leq \gamma \|x\|_\infty,$$

which implies

$$\|T_{BGS}\|_\infty \leq \gamma < 1.$$

□

## 1.4 Richardson Method

Suppose  $\omega \in \mathbb{C}_* := \mathbb{C} \setminus \{0\}$ , consider the splitting

$$A = \omega I + A - \omega I,$$

suppose

$$A\vec{u} = \vec{f}.$$

Then

$$A\vec{u} = (\omega I - A)\vec{u} + \vec{f}.$$

**Definition 1.29.** (*Richardson Method*) Define Richardson Method scheme as

$$A\vec{u}^{k+1} = (\omega I - A)\vec{u}^k + \vec{f}.$$

So that

$$\vec{u}^{k+1} = \vec{u}^k + \omega^{-1}(\vec{f} - A\vec{u}^k). \quad (1.11)$$

**Definition 1.30.** (*Preconditioner and Error Transfer Operator for Gauss-Seidel Method*) The preconditioner  $B_{RC}$  and error transfer operator  $T_{RC}$  of Gauss-Seidel Method are as follows

$$\begin{aligned} B_{RC} &= \omega^{-1}I, \\ T_{RC} &= I - \omega A. \end{aligned}$$

**Theorem 1.31.** (*convergence of the Richardson Method*) Let  $A = A^H > 0$  (HPD). If  $0 < \omega < \frac{2}{\lambda_{\max}}$ , then the Richardson Method converges. Moreover, the best acceleration parameter is given by

$$\omega_{opt} = \frac{2}{\lambda_{\min} + \lambda_{\max}},$$

in which, similarly,  $\lambda_{\min}$  is the smallest eigenvalue of  $A^T A$ .

*Proof.* 1. From the above lemma, we know that the error transform operator is as follows

$$T_{RC} = I - \omega(B)^{-1}A = I - \omega A.$$

Let  $\lambda \in \sigma(A)$ , then  $\nu := 1 - \omega\lambda \in \sigma(T)$ . From the sufficient and necessary condition for convergence, we know if  $\sigma(T) < 1$ , then Richardson Method converges, i.e.

$$|1 - \omega\lambda| < 1,$$

which implies

$$-1 < 1 - \omega\lambda_{\max} \leq 1 - \omega\lambda_{\min} < 1.$$

So, we get  $-1 < 1 - \omega\lambda_{\max}$ , i.e.

$$\omega < \frac{2}{\lambda_{\max}}.$$

2. The minimum is attachment at  $|1 - \omega\lambda_{\max}| = |1 - \omega\lambda_{\min}|$  (Figure.2), i.e.

$$\omega\lambda_{\max} - 1 = 1 - \omega\lambda_{\min}.$$

Therefore, we get

$$\omega_{opt} = \frac{2}{\lambda_{\min} + \lambda_{\max}}.$$

□

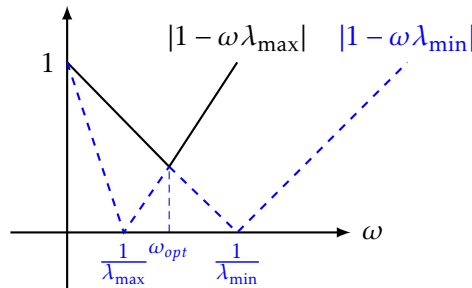


Figure 2: The curve of  $\rho(T_{RC})$  as a function of  $\omega$



## 1.5 Symmetrization

### 1.5.1 symmetrized multiplication

**Definition 1.32.** (*symmetrized multiplication GLIS*) Suppose  $A, B \in \mathbb{C}^{n \times n}$ ,  $\vec{f} \in \mathbb{C}^n$ , the *Symmetrized Multiplication GLIS (SMGLIS)* is defined as follows: given  $\vec{u}^0$ , find  $\vec{u}^1, \vec{u}^2, \dots$  via

$$\vec{u}^{k+1/2} = \vec{u}^k + B(\vec{f} - A\vec{u}^k), \quad (1.12)$$

$$\vec{u}^{k+1} = \vec{u}^{k+1/2} + B^H(\vec{f} - A\vec{u}^{k+1/2}). \quad (1.13)$$

**Remark 1.33.** If the blue part in (1.13) is  $\vec{u}^k$ , then the symmetrization is called symmetrized additive method.

**Lemma 1.34.** (*symmetrized multiplication GLIS*) The SMGLIS can be written as

$$\vec{u}^{k+1} = \vec{u}^k + B_{SM}(\vec{f} - A\vec{u}^k), \quad (1.14)$$

where

$$B_{SM} = B + B^H - B^H A B. \quad (1.15)$$

If  $B$  is invertible, then

$$B_{SM} = B^H(B^{-H} + B^{-1} - A)B.$$

If  $A$  is Hermitian, then  $B_{SM}$  is as well.

*Proof.* 1. Plugging (1.12) into (1.13) yields

$$\begin{aligned} \vec{u}^{k+1} &= \vec{u}^k + B(\vec{f} - A\vec{u}^k) + B^H(\vec{f} - A[\vec{u}^k + B(\vec{f} - A\vec{u}^k)]) \\ &= \vec{u}^k + B(\vec{f} - A\vec{u}^k) + B^H(\vec{f} - A\vec{u}^k - AB(\vec{f} - A\vec{u}^k)) \\ &= \vec{u}^k + B(\vec{f} - A\vec{u}^k) + B^H(\vec{f} - A\vec{u}^k) - B^H AB(\vec{f} - A\vec{u}^k) \\ &= \vec{u}^k + (B + B^H - B^H AB)(\vec{f} - A\vec{u}^k). \end{aligned}$$

2. If  $B$  is invertible, then

$$\begin{aligned} B_{SM} &= B + B^H - B^H AB \\ &= (I + B^H B^{-1} - B^H A)B \\ &= (B^H B^{-H} + B^H B^{-1} - B^H A)B \\ &= B^H(B^{-H} + B^{-1} - A)B. \end{aligned}$$

3. If  $A$  is Hermitian, then  $B_{SM}$  is Hermitian, since

$$B_{SM} = B + B^H - B^H AB,$$

then

$$\begin{aligned} B_{SM}^H &= B^H + B - B^H A^H B \\ &= B^H + B - B^H AB \\ &= B_{SM}. \end{aligned}$$

□

**Remark 1.35.** We could also consider a symmetrized additive method?

$$\vec{u}^{k+1/2} = \vec{u}^k + B(\vec{f} - A\vec{u}^k), \quad (1.16)$$

$$\vec{u}^{k+1} = \vec{u}^{k+1/2} + B^H(\vec{f} - A\vec{u}^k). \quad (1.17)$$

Then

$$B_{SA} = B + B^H.$$

But this is not as useful to us.

### 1.5.2 Symmetric Gauss-Seidel Method

**Definition 1.36.** (*Symmetric Gauss-Seidel Method*) Assume  $A = A^H$ , then

$$U^H = L, \quad D \in \mathbb{R}^{n \times n},$$

in the splitting (1.8). And

$$B_{GS} = (D - L)^{-1}.$$

So that

$$\begin{aligned} B_{GS}^H &= (D - L)^{-H} \\ &= (D - U)^{-1} \\ &= B_{BGS}. \end{aligned}$$

We have

$$\begin{aligned} B_{SM} : &= B_{SGS} \\ &= B_{GS}^H (B_{GS}^{-H} + B_{GS}^{-1} - A) B_{GS} \\ &= (D - U)^{-1} (D - U + D - L - A) (D - L)^{-1}, \end{aligned}$$

with cancellation, we get

$$B_{SGS} = (D - U)^{-1} D (D - L)^{-1}. \quad (1.18)$$

Of course,

$$B_{SGS}^H = B_{SGS}.$$

as desired.

### 1.5.3 Regular Splittings

In this subsection, we consider the following matrix splitting

$$A = M - N \in \mathbb{C}^{n \times n}.$$

where A is associate with the linear system (1.1).

**Theorem 1.37.** Consider the iterative scheme

$$M\vec{u}^{k+1} = N\vec{u}^k + \vec{f}, \quad (1.19)$$

where

$$A = M - N \in \mathbb{C}^{n \times n}.$$

If both  $A$  and  $M + M^H - A$  are HPD, then (1.19) converges.

There are two ways to prove this theorem, one way is based on computing the spectral radius, the other way is energy method.

1. *Proof.* We can rewrite (1.19) as a GLIS, i.e.

$$\begin{aligned} \vec{u}^{k+1} &= M^{-1}N\vec{u}^k + M^{-1}\vec{f} \\ &= \vec{u}^k - M^{-1}M\vec{u}^k + M^{-1}N\vec{u}^k + M^{-1}\vec{f} \\ &= \vec{u}^k + M^{-1}(-M\vec{u}^k + N\vec{u}^k + \vec{f}) \\ &= \vec{u}^k + M^{-1}(\vec{f} - A\vec{u}^k). \end{aligned}$$

Observe that  $M$  is nonsingular, otherwise  $M + M^H - A$  is not HPD. Then by the definition of Error Transfer operator (1.7),

$$T = I - BA = I - M^{-1}A. \quad (1.20)$$

If we can prove that  $\rho(T) < 1$ , then this method converges. Let  $(\lambda, \vec{\omega})$  be an eigenpair of  $T$  in (1.20). Then we have

$$T\vec{\omega} = \lambda\vec{\omega} \Rightarrow (I - M^{-1}A)\vec{\omega} = \lambda\vec{\omega} \Rightarrow (M - A)x = \lambda Mx \Rightarrow (1 - \lambda)M\vec{\omega} = A\vec{\omega}.$$

- (a)  $\lambda \neq 1$ . Otherwise  $\lambda = 1$ , then  $A\vec{\omega} = 0$ , contradicting  $A$  is HPD.
- (b)  $\lambda \leq 1$ . Since,  $(1 - \lambda)M\vec{\omega} = A\vec{\omega}$ , multiplying  $\vec{\omega}^H$  yields

$$(1 - \lambda)\vec{\omega}^H M\vec{\omega} = \vec{\omega}^H A\vec{\omega}.$$

Since  $\lambda \neq 1$ , we have

$$\vec{\omega}^H M\vec{\omega} = \frac{1}{1 - \lambda} \vec{\omega}^H A\vec{\omega}.$$

taking conjugate transpose of which gives

$$\vec{\omega}^H M^H \vec{\omega} = \frac{1}{1 - \lambda} \vec{\omega}^H A^H \vec{\omega} = \frac{1}{1 - \lambda} \vec{\omega}^H A\vec{\omega}.$$

Adding the above two equations and subtracting  $\vec{\omega}^H A\vec{\omega}$ , we have

$$\begin{aligned} \vec{\omega}^H (M + M^H - A)\vec{\omega} &= \left( \frac{1}{1 - \lambda} + \frac{1}{1 - \lambda} - 1 \right) \vec{\omega}^H A\vec{\omega} \\ &= \left( \frac{\lambda}{1 - \lambda} + \frac{1}{1 - \lambda} \right) \vec{\omega}^H A\vec{\omega} \\ &= \frac{1 - \lambda^2}{|1 - \lambda|^2} \vec{\omega}^H A\vec{\omega}. \end{aligned}$$

Since  $M + M^H - A$  and  $A$  are SPD, then  $\vec{\omega}^H(M + M^H - A)\vec{\omega} > 0$  and  $\vec{\omega}^H A \vec{\omega} > 0$ . It must be that

$$1 - \lambda^2 > 0.$$

Hence

$$|\lambda| < 1.$$

□

2. *Proof.* Let  $\vec{u}$  be the exact solution and  $\vec{u}^k$  be the approximate solution at  $k$  step, then

$$\begin{aligned}\vec{u} &= \vec{u} + M^{-1}(\vec{f} - A\vec{u}), \\ \vec{u}^{k+1} &= \vec{u}^k + M^{-1}(\vec{f} - A\vec{u}^k).\end{aligned}$$

From the definite of Error (1.5), we have

$$\begin{aligned}\vec{e}^{k+1} &= \vec{e}^k + M^{-1}(\vec{f} - A\vec{u} - \vec{f} + A\vec{u}^k) \\ &= \vec{e}^k - M^{-1}A\vec{e}^k.\end{aligned}$$

Let  $\vec{v}^{k+1} = \vec{e}^{k+1} - \vec{e}^k$ , then

$$M\vec{v}^{k+1} + A\vec{e}^k = 0.$$

Taking the conjugate transport of the above equation, then we get

$$M^H\vec{v}^{k+1} + A^H\vec{e}^k = M^H\vec{v}^{k+1} + A\vec{e}^k = 0.$$

Adding the last two equations yields

$$\frac{M + M^H}{2}\vec{v}^{k+1} + A\vec{e}^k = 0.$$

Let  $B_s = \frac{M + M^H}{2}$  and taking the inner product of both sides with  $\vec{v}^{k+1}$  gives

$$(B_s\vec{v}^{k+1}, \vec{v}^{k+1}) + (A\vec{e}^k, \vec{v}^{k+1}) = 0.$$

Since

$$\vec{e}^k = \frac{1}{2}(\vec{e}^{k+1} + \vec{e}^k) - \frac{1}{2}(\vec{e}^{k+1} - \vec{e}^k) = \frac{1}{2}(\vec{e}^{k+1} + \vec{e}^k) - \frac{1}{2}\vec{v}^{k+1},$$

then

$$\begin{aligned}0 &= (B_s\vec{v}^{k+1}, \vec{v}^{k+1}) + (A\vec{e}^k, \vec{v}^{k+1}) \\ &= (B_s\vec{v}^{k+1}, \vec{v}^{k+1}) + \frac{1}{2}(A(\vec{e}^{k+1} + \vec{e}^k), \vec{v}^{k+1}) - \frac{1}{2}(A\vec{v}^{k+1}, \vec{v}^{k+1}) \\ &= ((B_s - \frac{1}{2}A)\vec{v}^{k+1}, \vec{v}^{k+1}) + \frac{1}{2}(A(\vec{e}^{k+1} + \vec{e}^k), \vec{v}^{k+1}) \\ &= ((B_s - \frac{1}{2}A)\vec{v}^{k+1}, \vec{v}^{k+1}) + \frac{1}{2}(A(\vec{e}^{k+1} + \vec{e}^k), \vec{e}^{k+1} - \vec{e}^k) \\ &= ((B_s - \frac{1}{2}A)\vec{v}^{k+1}, \vec{v}^{k+1}) + \frac{1}{2}(\|\vec{e}^{k+1}\|_A^2 - \|\vec{e}^k\|_A^2)\end{aligned}$$

By assumption,  $Q := B_s - \frac{1}{2}A = \frac{M + M^T - A}{2} > 0$ , i.e. there exists  $m > 0$ , s.t.

$$(Q\vec{u}, \vec{u}) \geq m\|\vec{u}\|_2^2.$$

Therefore,

$$m \|\vec{v}^{k+1}\|_2^2 + \frac{1}{2} (\|\vec{e}^{k+1}\|_A^2 - \|\vec{e}^k\|_A^2) \leq 0.$$

i.e.

$$2m \|\vec{v}^{k+1}\|_2^2 + \|\vec{e}^{k+1}\|_A^2 \leq \|\vec{e}^k\|_A^2.$$

Hence

$$\|\vec{e}^{k+1}\|_A^2 \leq \|\vec{e}^k\|_A^2.$$

and

$$\|\vec{e}^{k+1}\|_A^2 \rightarrow 0.$$

□

**Theorem 1.38.** Suppose that  $A$  is HPD, then the Forward, Backward and Symmetric Gauss-Seidel Method converge.

*Proof.* In the language in Theorem.1.37.

1. For Forward Gauss-Seidel Method,

$$M = B_{GS} = D - L.$$

So, we have

$$M + M^H - A = D - L + (D - L)^H - A = D - L + D^H - U - A = D.$$

Since  $A$  is HPD, so  $D$  is HPD. Now we can apply Theorem.1.37.

2. For Backward Gauss-Seidel Method,

$$M = B_{BGS} = D - U.$$

So, we have

$$M + M^H - A = D - U + (D - U)^H - A = D - U + D^H - L - A = D.$$

Since  $A$  is HPD, so  $D$  is HPD. Now we can apply Theorem.1.37.

3. For Symmetric Gauss-Seidel Method, we have

$$M = B_{SGS} = (D - U)^{-1} D (D - L)^{-1}.$$

So, we have

$$M + M^H - A = B_{SGS} + B_{SGS}^H - A.$$

Since  $A$  and  $B_{SGS}$  are HPD, so  $M + M^H - A$  is HPD. Now we can apply Theorem.1.37.

□

**Definition 1.39.** (*Regular Splittings*)[3] Let  $A, M, N$  be three given matrices satisfying

$$A = M - N \in \mathbb{C}^{n \times n}. \quad (1.21)$$

The pair of matrices  $M, N$  is a regular splitting of  $A$ , if  $M$  is nonsingular and  $M^{-1}$  and  $N$  are nonnegative.

**Theorem 1.40.** (*The spectral radius estimation of Regular Splittings*[3]) Let  $M, N$  be a regular splitting of  $A$ . Then

$$\rho(M^{-1}N) < 1$$

if only if  $A$  is nonsingular and  $A^{-1}$  is nonnegative.

*Proof.* 1. Define  $G = M^{-1}N$ , since  $\rho(G) < 1$ , then  $I - G$  is nonsingular. And then  $A = M(I - G)$ , so  $A$  is nonsingular. So, by Theorem.0.7 satisfied, since  $G = M^{-1}N$  is nonsingular and  $\rho(G) < 1$ , then we have  $(I - G)^{-1}$  is nonnegative as is  $A^{-1} = (I - G)^{-1}M^{-1}$ .

2.  $\Leftarrow$ : since  $A, M$  are nonsingular and  $A^{-1}$  is nonnegative, then  $A = M(I - G)$  is nonsingular. Moreover

$$\begin{aligned} A^{-1}N &= \left(M(I - M^{-1}N)\right)^{-1}N \\ &= (I - M^{-1}N)^{-1}M^{-1}N \\ &= (I - G)^{-1}G. \end{aligned}$$

Clearly,  $G = M^{-1}N$  is nonnegative by the assumptions, and as a result of the Perron-Frobenius theorem, there is a nonnegative eigenvector  $x$  associated with  $\rho(G)$  which is an eigenvalue, such that

$$Gx = \rho(G)x.$$

Therefore

$$A^{-1}Nx = \frac{\rho(G)}{1 - \rho(G)}x.$$

Since  $x$  and  $A^{-1}N$  are nonnegative, this shows that

$$\frac{\rho(G)}{1 - \rho(G)} \geq 0.$$

and this can be true only when  $0 \leq \rho(G) \leq 1$ . Since  $I - G$  is nonsingular, then  $\rho(G) \neq 1$ , which implies that  $\rho(G) < 1$ . □

## 2 The Two-Grid Algorithm

**Definition 2.1.** (*Two-Grid Algorithm Components*) Suppose  $n_0, n_1 \in \mathbb{Z}$  and  $n_1 \geq n_0 > 0$ . Suppose  $A_1 \in \mathbb{R}^{n_1 \times n_1}$  is SPD and  $\vec{f} \in \mathbb{R}^{n_1}$ ,  $A_1$  is called the *fine grid operator (matrix)*. Define some

$$R_0 \in \mathbb{R}^{n_0 \times n_1}, \quad R_0 \vec{v}_1 \in \mathbb{R}^{n_0}, \quad \forall \vec{v}_1 \in \mathbb{R}^{n_1}.$$

$R_0$  is called a *restriction operator (matrix)*. We assume that  $R_0$  has full rank  $\text{rank}(R_0) = n_0$  and

$$A_0 = \underbrace{R_0}_{n_0 \times n_1} \underbrace{A_1}_{n_1 \times n_1} \underbrace{R_0^T}_{n_1 \times n_0} \in \mathbb{R}^{n_0 \times n_0}.$$

$A_0$  is called the *coarse grid operator (matrix)*. Define

$$P_0 = R_0^T \in \mathbb{R}^{n_1 \times n_0}.$$

Then, as Figure 3 described

$$A_0 = R_0 A_1 P_0.$$

$P_0$  is called the *Prolongation Operator (matrix)*.

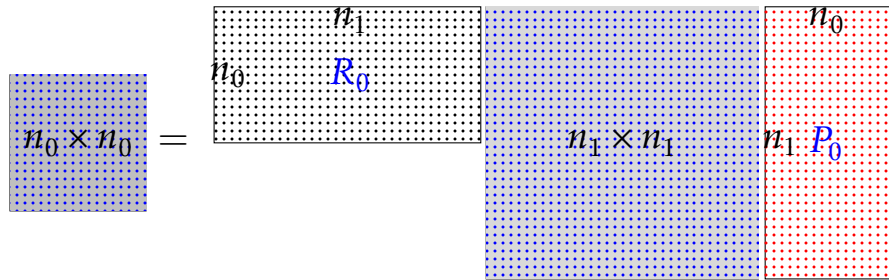


Figure 3: The action of Prolongation and Restriction Operator.

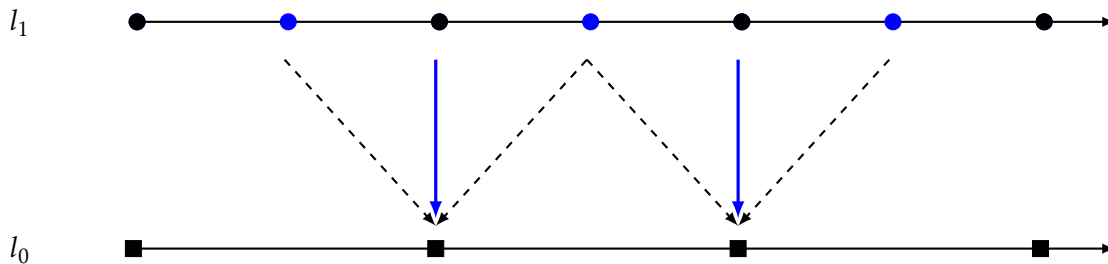


Figure 4: Restriction from fine grid to coarse grid on 1D uniform mesh.

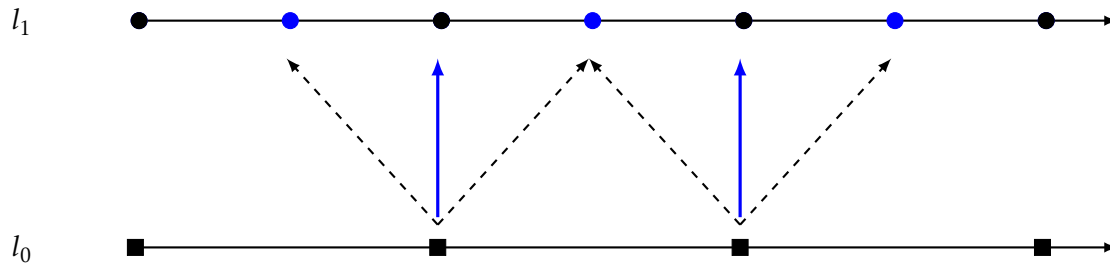


Figure 5: Prolongation from coarse grid to fine grid on 1D uniform mesh.

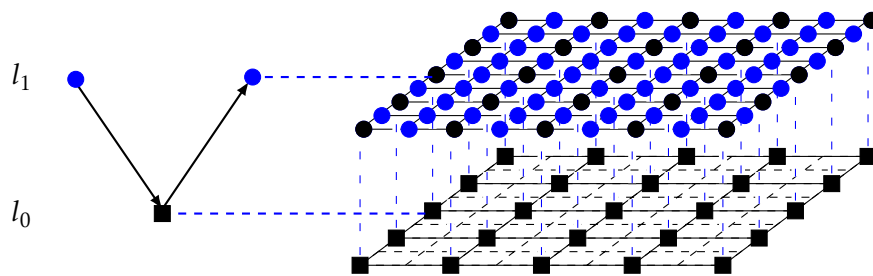


Figure 6: Mesh on the fine and coarse grid of the two-grid method in 2D.

**Proposition 2.2.**  $A_0 \in \mathbb{R}^{n_0 \times n_0}$  is SPD.

*Proof.* 1. Symmetry-ness: Since  $A_0 = R_0 A_1 R_0^T$  and  $A_1$  is SPD, then

$$A_0^T = R_0 A_1^T R_0^T = R_0 A_1 R_0^T = A_0.$$

2. Positive Definite-ness: Let  $\vec{\omega} \in \mathbb{R}^{n_0}$  an arbitrary vector and  $\vec{\omega} \neq 0$ . Since  $R_0$  has full rank, then  $R_0^T \vec{\omega} \neq 0$ . Moreover, since  $A_1$  is SPD,

$$\vec{\omega}^T A_0 \vec{\omega} = \vec{\omega}^T R_0 A_1 R_0^T \vec{\omega} = (R_0^T \vec{\omega})^T A_1 (R_0^T \vec{\omega}) > 0.$$

□

We wish to solve the following: find  $\vec{u}^1 \in \mathbb{R}^{n_1}$  such that

$$A_1 \vec{u}_1 = \vec{f}_1.$$



**Algorithm 2.3.** Given  $\vec{u}_1^k \in \mathbb{R}^n$ , compute

$$\vec{u}_1^{k+1} := TG(\vec{f}_1, \vec{u}_1^k, m_1, m_2)$$

1. *Pre-smoothing* (Figure.7-Figure.11):

- $\vec{u}_1^{(1,0)} := \vec{u}_1^k$
- $\vec{u}_1^{(1,\sigma+1)} := \vec{u}_1^{(1,\sigma)} + S_1(\vec{f}_1 - A_1 \vec{u}_1^{(1,\sigma)}), \quad 1 \leq \sigma \leq m_1 - 1$
- $\vec{u}_1^{(1)} := \vec{u}_1^{(1,m_1)}$

2. *Coarse-Grid correction* (Figure.7-Figure.11):

- $\vec{r}_1^{(1)} := \vec{f}_1 - A_1 \vec{u}_1^{(1)}$
- $\vec{r}_0^{(1)} := R_0 \vec{r}_1^{(1)} = R_0(\vec{f}_1 - A_1 \vec{u}_1^{(1)})$
- $\vec{q}_0^{(1)} := A_0^{-1} \vec{r}_0^{(1)}$
- $\vec{u}_1^{(2)} := \vec{u}_1^{(1)} + R_0^T \vec{q}_0^{(1)}$

3. *Post-smoothing* (Figure.7-Figure.11):

- $\vec{u}_1^{(3,0)} := \vec{u}_1^{(2)}$
- $\vec{u}_1^{(3,\sigma+1)} := \vec{u}_1^{(3,\sigma)} + S_1^T(\vec{f}_1 - A_1 \vec{u}_1^{(3,\sigma)}), \quad 0 \leq \sigma \leq m_2 - 1$
- $\vec{u}_1^{(3)} := \vec{u}_1^{(3,m_2)}$
- $\vec{u}_1^{k+1} := \vec{u}_1^{(3)}$

**Remark 2.4.** Let us examine the *Coarse-Grid Correction* in a very special case suppose  $n_0 = n_1$  and

$$R_0 = I_1, \quad (n_1 \times n_1 \text{ Identity}).$$

Then

$$A_0 = R_0 A_1 P_0 = R_0 A_1 R_0^T.$$

*Coarse-Grid Correction:*

- $\vec{r}_1^{(1)} := \vec{f}_1 - A_1 \vec{u}_1^{(1)}$
- $\vec{r}_0^{(1)} := R_0 \vec{r}_1^{(1)} = \vec{r}_1^{(1)}$
- $\vec{q}_0^{(1)} := A_0^{-1} \vec{r}_0^{(1)} = A_0^{-1} \vec{r}_1^{(1)} = \vec{e}_1^{(1)}$
- $\begin{aligned} \vec{u}_1^{(2)} &:= \vec{u}_1^{(1)} + R_0^T \vec{q}_0^{(1)} \\ &= \vec{u}_1^{(1)} + R_0^T \vec{e}_1^{(1)} \\ &= \vec{u}_1^{(1)} + \vec{e}_1^{(1)} \\ &= \vec{u}_1^{(1)} + \vec{u}_1 - \vec{u}_1^{(1)} = \vec{u}_1 \end{aligned}$

Where

$$A_1 \vec{u}_1 = \vec{f}_1.$$

Of course, the Algorithm.2.3 should terminate at this stage, because we have the exact solution.

Now, in general

$$\vec{q}_0^{(1)} = \vec{e}_1^{(1)}.$$

But

$$R_0^T \vec{q}_0^{(1)} \approx \vec{e}_1^{(1)}.$$

can be a quite good approximation in some sense. We will revisit this issue later.

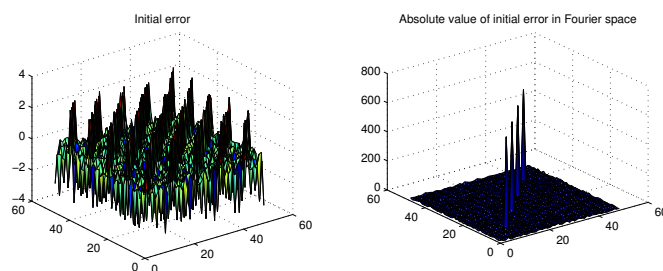


Figure 7: The initial error and the absolute value of initial error in Fourier space.

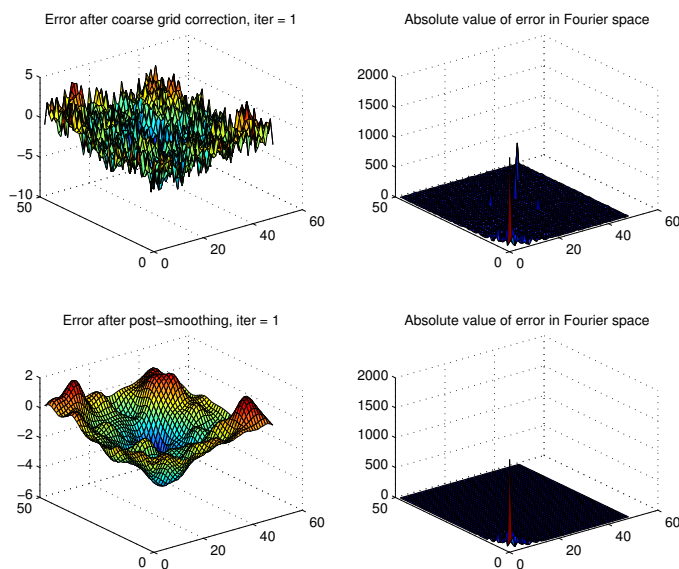


Figure 8: The error after 1<sub>st</sub> coarse grid correction and post-smoothing with the absolute value of error in Fourier space.

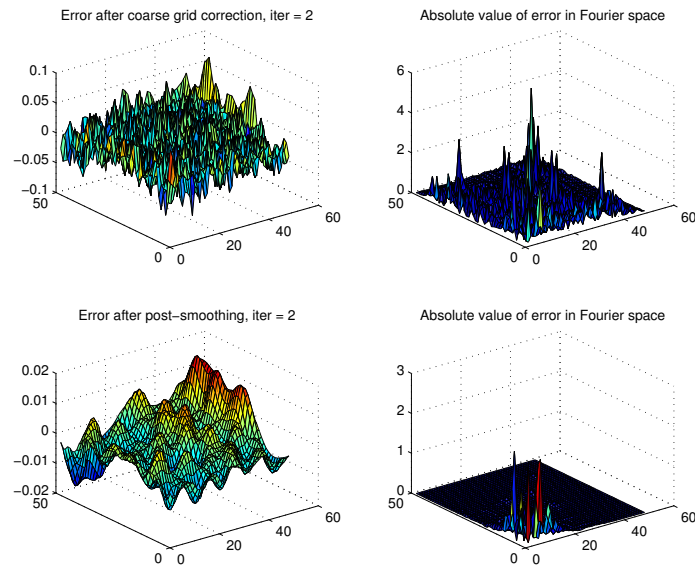


Figure 9: The error after  $2_{nd}$  coarse grid correction and post-smoothing with the absolute value of error in Fourier space.

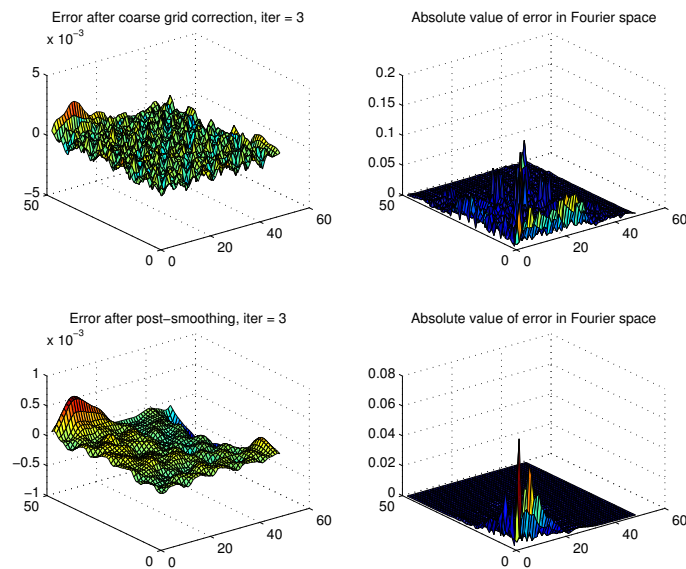


Figure 10: The error after  $3_{rd}$  coarse grid correction and post-smoothing with the absolute value of error in Fourier space.

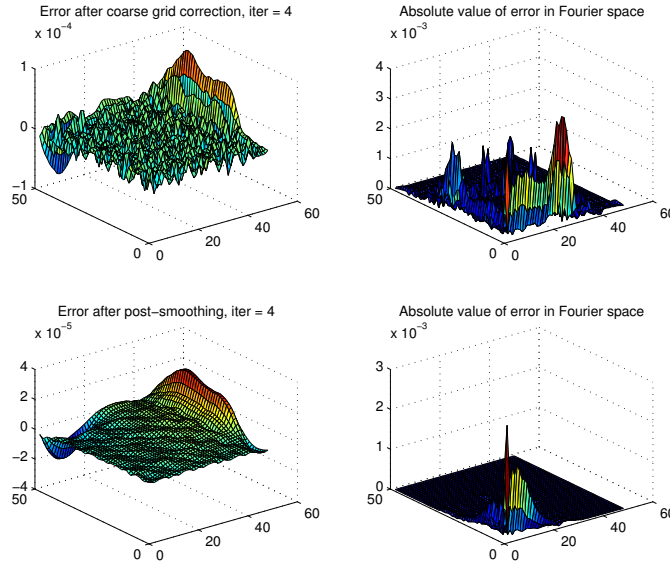


Figure 11: The error after 4<sub>th</sub> coarse grid correction and post-smoothing with the absolute value of error in Fourier space.

From now, let us show that the two-grid method is a GLIS and also calculate the error propagation matrix, we need the following definitions first.

**Definition 2.5.** (*Coarse-Grid Ritz Projection matrix and Galerkin Condition*) Referring to the definitions in Definition. (2.1), we define the matrix

$$\tilde{\Pi}_1 = R_0^T \Pi_0 \in \mathbb{R}^{n_1 \times n_1},$$

where

$$\Pi_0 = \underbrace{A_0^{-1}}_{n_0 \times n_0} \underbrace{R_0}_{n_0 \times n_1} \underbrace{A_1}_{n_1 \times n_1} \in \mathbb{R}^{n_0 \times n_1}.$$

$\tilde{\Pi}_1$  is called the *Coarse-Grid Ritz Projection matrix*. We say that the coarse-grid matrix,  $A_0$ , satisfies the *Galerkin Condition* iff

$$A_0 = R_0 A_1 R_0^T.$$

**Proposition 2.6.**  $\tilde{\Pi}_1$  as defined in Definition.2.6 is a "bona fides" projection matrix, i.e.

$$(\tilde{\Pi}_1)^2 = \tilde{\Pi}_1,$$

provided  $A_0$  satisfies the Galerkin Condition.

*Proof.*

$$(\tilde{\Pi}_1)^2 = R_0^T \Pi_0 R_0^T \Pi_0$$

$$\begin{aligned}
&= R_0^T A_0^{-1} \underbrace{R_0 A_1 R_0^T}_{A_0} A_0^{-1} R_0 A_1 \\
&= R_0^T A_0^{-1} R_0 A_1 \\
&= R_0^T \Pi_0 = \tilde{\Pi}_1.
\end{aligned}$$

□

**Corollary 2.7.** *If the Galerkin Condition holds  $I - \tilde{\Pi}_1$  is a projection matrix, i.e.*

$$(I_1 - \tilde{\Pi}_1)^2 = I_1 - \tilde{\Pi}_1,$$

**Definition 2.8.** (*Smoothing error matrix*) Define the *smoothing error matrix* as

$$K_1 = I_1 - S_1 A_1.$$

**Definition 2.9.** (*Adjoint*) Let  $M \in \mathbb{R}^{n_1 \times n_1}$  and  $\vec{u}_1, \vec{v}_1 \in \mathbb{R}^{n_1}$ . Define the *adjoint* with respect to the inner product

$$(\vec{u}_1, \vec{v}_1)_1 := \vec{u}_1^T \vec{v}_1,$$

via

$$(M \vec{u}_1, \vec{v}_1)_1 := (\vec{u}_1, M^T \vec{v}_1)_1.$$

Define the *adjoint* with respect to the inner product

$$(\vec{u}_1, \vec{v}_1)_{A_1} := (\vec{u}_1, A_1 \vec{v}_1)_1,$$

via

$$(M \vec{u}_1, \vec{v}_1)_{A_1} := (\vec{u}_1, M^* \vec{v}_1)_{A_1}.$$

It follows by a simple calculation that

$$K_1^* = I_1 - S_1^T A_1.$$

In other words, for any  $\vec{u}_1, \vec{v}_1 \in \mathbb{R}^{n_1}$ ,

$$\begin{aligned}
((I_1 - S_1 A_1) \vec{u}_1, \vec{v}_1)_{A_1} &= ((I_1 - S_1 A_1) \vec{u}_1, A_1 \vec{v}_1)_1 \\
&= (\vec{u}_1, (I_1 - S_1 A_1)^T A_1 \vec{v}_1)_1 \\
&= (\vec{u}_1, (I_1 - A_1^T S_1^T) A_1 \vec{v}_1)_1 \\
&= (\vec{u}_1, (I_1 - A_1 S_1^T) A_1 \vec{v}_1)_1 \\
&= (\vec{u}_1, (A_1 - A_1 S_1^T A_1) \vec{v}_1)_1 \\
&= (\vec{u}_1, A_1 (I_1 - S_1^T A_1) \vec{v}_1)_1 \\
&= (\vec{u}_1, (I_1 - S_1^T A_1) \vec{v}_1)_{A_1}.
\end{aligned}$$

**Theorem 2.10.** For the two-grid Algorithm.2.3, we have

$$\vec{e}_1^{k+1} = E_1 \vec{e}_1^k,$$

where

$$E_1 = (K_1^*)^{m_2} (I_1 - \tilde{\Pi}_1) (K_1)^{m_2},$$

and

$$\vec{e}_1^k := \vec{u}_1^E - \vec{u}_1^k$$

where  $\vec{u}_1^E$  is the exact solution. Furthermore, the two-grid method is a GLIS, i.e., there is some  $B_1 \in \mathbb{R}^{n_1 \times n_1}$  such that

$$E_1 = I_1 - B_1 A_1.$$

*Proof.* Set

$$\vec{e}_1^{(i)} := \vec{u}_1^E - \vec{u}_1^{(i)}, \quad i = 1, 2, 3.$$

Then

$$\vec{e}_1^{(1)} := (I_1 - S_1 A_1)^{m_1} \vec{e}_1^k = K_1^{m_1} \vec{e}_1^k.$$

after the pre-smoothing step. Next we have the coarse-grid correction step.

$$\begin{aligned} \vec{e}_1^{(2)} &= \vec{u}_1^E - \vec{u}_1^{(2)} \\ &= \vec{u}_1^E - \vec{u}_1^{(1)} - R_0^T \vec{q}_0^{(1)} \\ &= \vec{e}_1^{(1)} - R_0^T A_0^{-1} \vec{r}_0^{(1)} \\ &= \vec{e}_1^{(1)} - R_0^T A_0^{-1} R_0 \vec{r}_1^{(1)} \\ &= \vec{e}_1^{(1)} - R_0^T A_0^{-1} R_0 A_1 \vec{e}_1^{(1)} \\ &= (I_1 - \tilde{\Pi}_1) \vec{e}_1^{(1)}. \end{aligned}$$

So

$$\vec{e}_1^{(2)} = (I_1 - \tilde{\Pi}_1) K_1^{m_1} \vec{e}_1^k.$$

Finally, from the post-smoothing step

$$\vec{e}_1^{k+1} = \vec{e}_1^{(3)} = (I_1 - S_1^T A_1)^{m_2} \vec{e}_1^{(2)}.$$

So

$$\vec{e}_1^{k+1} = (K_1^*)^{m_2} (I_1 - \tilde{\Pi}_1) K_1^{m_1} \vec{e}_1^k,$$

as desired.

We leave the second part as an exercise. □

**Remark 2.11.** In the case that  $m_1 = m_2 = 1$ , we have the following simple form for  $B_1$ :

$$B_1 = \underbrace{S_1^T (S_1^{-T} + S_1^{-1} + A_1)}_{\text{symmetric smoothing}} \underbrace{+ K_1^* R_0^T A_0^{-1} R_0 K_1}_{\text{error correction}}.$$

In general, we can show that  $B_1$  is symmetric with respect to

$$(\vec{u}_1, \vec{v}_1)_1 = \vec{u}_1^T \vec{v}_1,$$

iff  $m_1 = m_2 = m$ .

**Theorem 2.12.** The error propagation matrix for the two-grid method,  $E_1$ , is symmetric with respect to  $(\cdot, \cdot)_{A_1}$  iff  $m_1 = m_2 = m$ . Furthermore, if  $m_1 = m_2 = m$  and if the Galerkin condition holds for  $A_0$ , then  $E_1$  is *Symmetric Positive Semi-Definite (SPSD)* with respect to  $(\cdot, \cdot)_{A_1}$ .

*Proof.* Let  $\vec{u}_1, \vec{v}_1 \in \mathbb{R}^{n_1}$  be arbitrary. Then we have

$$\begin{aligned} (E_1 \vec{u}_1, \vec{v}_1)_{A_1} &= ((K_1^*)^{m_2} (I_1 - \tilde{\Pi}_1) K_1^{m_1} \vec{u}_1, \vec{v}_1)_{A_1} \\ &= ((I_1 - \tilde{\Pi}_1) K_1^{m_1} \vec{u}_1, K_1^{m_2} \vec{v}_1)_{A_1} \\ &= (K_1^{m_1} \vec{u}_1, (I_1 - \tilde{\Pi}_1)^* K_1^{m_2} \vec{v}_1)_{A_1} \\ &= (\vec{u}_1, (K_1^*)^{m_1} (I_1 - \tilde{\Pi}_1)^* K_1^{m_2} \vec{v}_1)_{A_1}. \end{aligned}$$

Observe that

$$(I_1 - \tilde{\Pi}_1)^* = (I_1 - \tilde{\Pi}_1).$$

Indeed,

$$((I_1 - \tilde{\Pi}_1) \vec{u}_1, \vec{v}_1)_{A_1} = (\vec{u}_1, \vec{v}_1)_{A_1} - (\tilde{\Pi}_1 \vec{u}_1, \vec{v}_1)_{A_1}$$

and

$$\begin{aligned} (\tilde{\Pi}_1 \vec{u}_1, \vec{v}_1)_{A_1} &= (R_0^T A_0^{-1} R_0 A_1 \vec{u}_1, A_1 \vec{v}_1)_1 \\ &= (A_1 \vec{u}_1, (R_0^T A_0^{-1} R_0)^T A_1 \vec{v}_1)_1 \\ &= (A_1 \vec{u}_1, R_0^T A_0^{-1} R_0 A_1 \vec{v}_1)_1 \\ &= (A_1 \vec{u}_1, \tilde{\Pi}_1 \vec{v}_1)_1 \\ &= (\vec{u}_1, \tilde{\Pi}_1 \vec{v}_1)_{A_1}. \end{aligned}$$

Therefore

$$\tilde{\Pi}_1^* = \tilde{\Pi}_1,$$

and

$$(I_1 - \tilde{\Pi}_1)^* = (I_1 - \tilde{\Pi}_1).$$

The symmetry of  $E_1$  follows iff  $m_1 = m_2$ .

PSD-ness: let  $\vec{u}_1 \in \mathbb{R}^{n_1}$  be arbitrary and suppose  $m_1 = m_2 = m$ . Since the Galerkin Condition holds, from Corollary 2.7, we have

$$(I_1 - \tilde{\Pi}_1)^2 = (I_1 - \tilde{\Pi}_1).$$

Now

$$\begin{aligned} (E_1 \vec{u}_1, \vec{u}_1)_{A_1} &= ((K_1^*)^m (I_1 - \tilde{\Pi}_1) K_1^m \vec{u}_1, \vec{u}_1)_{A_1} \\ &= ((I_1 - \tilde{\Pi}_1) K_1^m \vec{u}_1, K_1^m \vec{u}_1)_{A_1} \\ &= ((I_1 - \tilde{\Pi}_1)^2 K_1^m \vec{u}_1, K_1^m \vec{u}_1)_{A_1} \\ &= ((I_1 - \tilde{\Pi}_1) K_1^m \vec{u}_1, (I_1 - \tilde{\Pi}_1)^* K_1^m \vec{u}_1)_{A_1} \\ &= ((I_1 - \tilde{\Pi}_1) K_1^m \vec{u}_1, (I_1 - \tilde{\Pi}_1) K_1^m \vec{u}_1)_{A_1} \\ &= \|(I_1 - \tilde{\Pi}_1) K_1^m \vec{u}_1\|_{A_1}^2 \geq 0. \end{aligned}$$

□

**Remark 2.13.** We remark that the Galerkin Condition is not required for the symmetry of  $E_1$ . All that was used to establish symmetry was the definition

$$R_0^T A_0^{-1} R_0 A_1.$$

On the other hand, we used the Galerkin condition to establish that  $\tilde{\Pi}_1$  and  $(I_1 - \tilde{\Pi}_1)$  are projections and these fact played a role in showing that  $E_1$  is SPSD.

It may be that the Galerkin Condition will fail in some multigrid application even where the definition of  $\tilde{\Pi}_1$  above remains true.

### 3 Rigorous (Quantitative) Fourier Analysis of the Two-Grid Method

We will consider the Model problem:

$$\begin{cases} -\Delta u &= f, & \text{in } \Omega = (0, 1), \\ u &= 0, & \text{on } \partial\Omega = \{0, 1\}. \end{cases}$$

**FINITE DIFFERENCE METHOD:**

Set

$$h := \frac{1}{n_1 + 1}, \quad n_1 \in \mathbb{Z}^+, \quad x_{1,i} = i \cdot h, \quad i = 0, 1, 2, \dots, n_1 + 1.$$

Find  $u_{1,1}, u_{1,2}, \dots, u_{1,n_1}$ , such that

$$\begin{cases} \frac{-u_{1,i-1} + 2u_{1,i} - u_{1,i+1}}{h^2} &= f(x_{1,i}) := f_{1,i}, \\ u_{1,0} = u_{1,n_1+1} &= 0. \end{cases} \quad (3.1)$$

Now set

$$\vec{u}_1^{FD} = \begin{bmatrix} u_{1,1} \\ u_{1,2} \\ \vdots \\ u_{1,n_1-1} \\ u_{1,n_1} \end{bmatrix}, \quad \vec{f}_1^{FD} = \begin{bmatrix} hf_{1,1} \\ hf_{1,2} \\ \vdots \\ hf_{1,n_1-1} \\ hf_{1,n_1} \end{bmatrix} \in \mathbb{R}^{n_1},$$



and

$$A_1 := A_h = \begin{bmatrix} \frac{2}{h} & -\frac{1}{h} & & & \\ -\frac{1}{h} & \frac{2}{h} & -\frac{1}{h} & & \\ & \ddots & \ddots & \ddots & \\ & & -\frac{1}{h} & \frac{2}{h} & -\frac{1}{h} \\ & & & -\frac{1}{h} & \frac{2}{h} \end{bmatrix} \in \mathbb{R}^{n_1 \times n_1}.$$

Then in matrix form, the Finite Difference Approximation is as follows: Find  $\vec{u}_1^{FD} \in \mathbb{R}^{n_1}$ , such that

$$A_1 \vec{u}_1^{FD} = \vec{f}_1^{FD}. \quad (3.2)$$

#### FINITE ELEMENT METHOD:

We will use **piecewise linear** finite element weak formula: find  $u \in H_0^1(0,1)$ , such that

$$\left( \frac{du}{dx}, \frac{dv}{dx} \right)_{L^2(0,1)} = (f, v)_{L^2(0,1)} \quad \forall v \in H_0^1(0,1).$$

Define

$$V_h := \{v \in C_0^0([0,1]) \mid v|_{T_{1,i}} \in \mathbb{P}_1(T_{1,i}), \quad 1 \leq i \leq n_1 + 1\}.$$

The  $n_1$  hat functions form a basis function  $V_h$ .

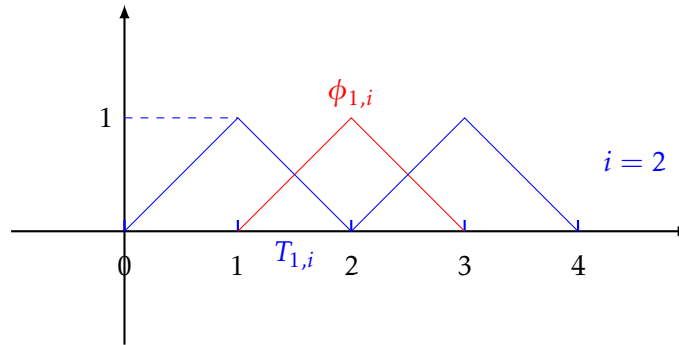


Figure 12: Piecewise linear basis function in 1D.

Then finite element approximation of the model problem is as follows: find  $u_h V_h$ , such that

$$\left( \frac{du_h}{dx}, \frac{d\phi_{1,i}}{dx} \right)_{L^2(0,1)} = (f, \phi_{1,i})_{L^2(0,1)},$$

for all  $i = 1, 2, \dots, n_1$ .

Now set

$$\vec{f}_1^{FE} = \begin{bmatrix} (f, \phi_{1,1})_{L^2(0,1)} \\ \vdots \\ (f, \phi_{1,n_1})_{L^2(0,1)} \end{bmatrix} \in \mathbb{R}^{n_1}.$$

We expand  $u_h$  in the basis of hat functions.

$$u_h = \sum_{j=1}^{n_1} u_{1,j} \phi_{1,j}(x) \in V_h.$$

Then, we set

$$\vec{u}_1^{FE} = \begin{bmatrix} u_{1,1} \\ \vdots \\ u_{1,n_1} \end{bmatrix} \in \mathbb{R}^{n_1}.$$

Define the  $n_1 \times n_1$  stiffness matrix  $A$ , via,

$$a_{1,i,j} := [A_1]_{ij} = \left( \frac{d\phi_{1,i}}{dx}, \frac{d\phi_{1,j}}{dx} \right)_{L^2(0,1)} = \left( \frac{d\phi_{1,i}}{dx}, \frac{d\phi_{1,j}}{dx} \right)_{L^2(0,1)}.$$

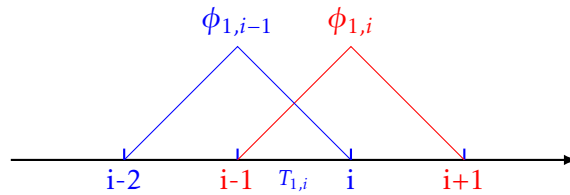


Figure 13: Piecewise linear basis function  $\phi_i$  in 1D.

So,

$$\begin{aligned} a_{1,i,i-1} &= \left( \frac{d\phi_{1,i-1}}{dx}, \frac{d\phi_{1,i}}{dx} \right)_{L^2(0,1)} \\ &= \left( \frac{d\phi_{1,i-1}}{dx}, \frac{d\phi_{1,j}}{dx} \right)_{L^2(T_{1,i})} \\ &= \left( -\frac{1}{h}, \frac{1}{h} \right)_{L^2(T_{1,i})} \\ &= -\frac{1}{h}. \end{aligned}$$

Likewise,

$$a_{1,i,i} = \left( \frac{d\phi_{1,i}}{dx}, \frac{d\phi_{1,i}}{dx} \right)_{L^2(0,1)} = \frac{2}{h}.$$

Let, we have, as before,

$$A_1 := A_h = \begin{bmatrix} \frac{2}{h} & -\frac{1}{h} & & & \\ -\frac{1}{h} & \frac{2}{h} & -\frac{1}{h} & & \\ & \ddots & \ddots & \ddots & \\ & & -\frac{1}{h} & \frac{2}{h} & -\frac{1}{h} \\ & & & -\frac{1}{h} & \frac{2}{h} \end{bmatrix} \in \mathbb{R}^{n_1 \times n_1}.$$

Therefore Finite Element Approximation matrix, is as follows: Find  $\vec{u}_1^{FE} \in \mathbb{R}^{n_1}$ , such that

$$A_1 \vec{u}_1^{FE} = \vec{f}_1^{FE}. \quad (3.3)$$

**Theorem 3.1.**  $A_1 \in \mathbb{R}^{n_1 \times n_1}$  is SPD. Its eigenvalue are

$$\lambda_1^{(k)} = \frac{4}{h} \sin^2\left(\frac{k\pi h}{2}\right) = \frac{2}{h} (1 - \cos k\pi h). \quad (3.4)$$

for  $k = 1, 2, \dots, n_1$ , and the corresponding eigenvectors are

$$\left[ \vec{v}_1^{(k)} \right]_i = v_{1,i}^{(k)} = \sin k\pi x_{1,i}. \quad (3.5)$$

*Proof.* Let  $1 \leq i \leq n_1$ ,

$$\begin{aligned} \left[ A_1 \vec{v}_1^{(k)} \right]_i &= -\frac{1}{h} \sin(k\pi x_{1,i-1}) + \frac{1}{h} \sin(k\pi x_{1,i}) - \frac{1}{h} \sin(k\pi x_{1,i+1}) \\ &= \frac{2}{h} [1 - \cos(k\pi h)] \sin(k\pi x_{1,i}) \end{aligned}$$

Thus

$$\left[ A_1 \vec{v}_1^{(k)} \right]_i = \lambda_1^{(k)} v_{1,i}^{(k)},$$

for  $1 \leq i \leq n_1$ . □

**Theorem 3.2.** The spectral condition number of  $A_1$ , i.e.

$$\kappa_2(A_1) := \|A_1\|_2 \|A_1^{-1}\|_2,$$

satisfies the estimates

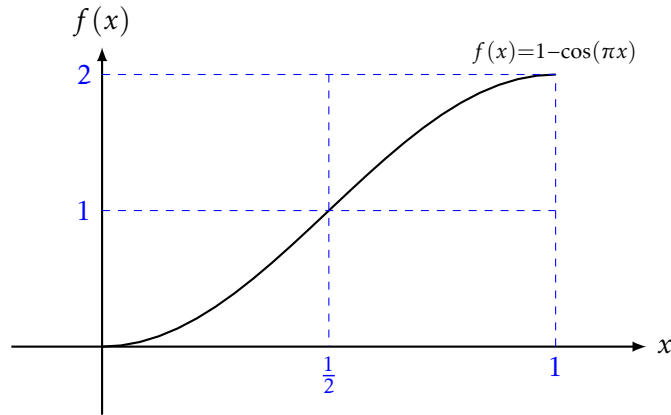
$$C_1 h^{-2} \leq \kappa_2(A_1) \leq C_2 h^{-2},$$

for some constants  $0 < C_1 \leq C_2$ .

*Proof.* Since  $A_1$  is SPD, it follows that

$$\kappa_2(A_1) = \frac{\lambda_1^{(n_1)}}{\lambda_1^{(1)}} = \frac{1 - \cos(n_1 \pi h)}{1 - \cos(\pi h)}.$$

Consider the function  $f(x) = 1 - \cos(\pi x)$  (Figure.15).

Figure 14: The value of  $1 - \cos(\pi x)$ .

Now, observe that, if  $n_1 \geq 1$ , then

$$0 < h \leq \frac{1}{2}, \quad \left(0 < \pi h \leq \frac{\pi}{2}\right).$$

Since, we have  $1 \leq 1 - \cos(n_1 \pi h) \leq 2$ , it follows that

$$\frac{1}{1 - \cos(\pi h)} \leq \kappa_2(A_1) \leq \frac{2}{1 - \cos(\pi h)}.$$

Now, setting  $x = \pi h$ , we have, by Taylor Theorem (expanding at  $x = 0$ ),

$$\cos(x) = 1 - \frac{1}{2}x^2 + \frac{1}{4!}x^4 \cos(\xi),$$

for some

$$0 < \xi < x = \pi h \leq \frac{\pi}{2}.$$

Hence

$$0 < \xi < \frac{\pi}{2},$$

and

$$0 < \cos(\xi) < 1.$$

So,

$$0 < \frac{1}{4!}x^4 \cos(\xi) = \cos(x) - 1 + \frac{1}{2}x^2$$

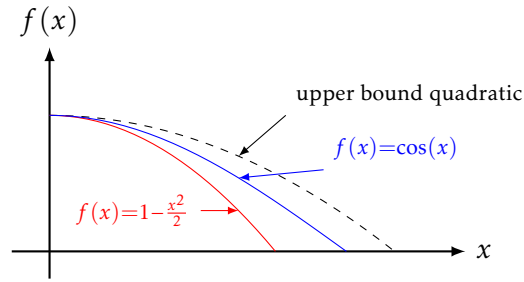
or

$$1 - \cos(\pi h) < \frac{\pi^2}{2}h^2, \quad 0 < \pi h \leq \frac{\pi}{2}.$$

We conclude the lower bound

$$\frac{2}{\pi^2 h^2} \leq \frac{1}{1 - \cos(\pi h)}, \quad 0 < h \leq \frac{1}{2}.$$

The last bound may be viewed graphically as

Figure 15: The upper and lower bound of  $1 - \cos(\pi x)$ .

To go further, again by Taylor's Theorem, we have, for  $0 < x \leq \frac{\pi}{2}$ ,

$$\cos(x) = 1 - \frac{x^2}{2} + \frac{x^4}{24} - \frac{x^6}{6!} \cos(\xi),$$

for some  $0 < \xi < \frac{\pi}{2}$ . So

$$\cos(x) - 1 + \frac{x^2}{2} - \frac{x^4}{24} = -\frac{x^6}{6!} \cos(\xi) < 0,$$

and

$$\frac{x^2}{2} - \frac{x^4}{24} < 1 - \cos(x).$$

But

$$\frac{x^2}{2} - \frac{x^4}{24} \geq \frac{x^2}{4}, \quad 0 < x \leq \frac{\pi}{2}.$$

Thus

$$\frac{x^2}{4} < 1 - \cos(x), \quad 0 < x \leq \frac{\pi}{2}.$$

and

$$\frac{x^2}{2} - \frac{x^4}{24} < 1 - \cos(x), \quad 0 < x \leq \frac{\pi}{2},$$

and

$$\frac{1}{1 - \cos(\pi h)} < \frac{4}{\pi^2 h^2}, \quad 0 < \pi h \leq \frac{\pi}{2}.$$

Thus

$$\frac{2}{\pi^2 h^2} \leq \kappa_2(A_1) \leq \frac{8}{\pi^2 h^2}.$$

□

### 3.1 The Damped Jacobi Smoother

To approximate the solution of

$$A_1 \vec{u}_1 = \vec{f}_1^\square \in \mathbb{R}^{n_1},$$

we apply the damped Jacobi method. This requires a splitting (1.8) of  $A_1$ , i.e.:

$$A_1 = D - U - L.$$

where

$$D = \begin{bmatrix} \frac{2}{h} & & & \\ & \frac{2}{h} & & \\ & & \ddots & \\ & & & \frac{2}{h} \end{bmatrix} \in \mathbb{R}^{n_1 \times n_1}, \quad U = \begin{bmatrix} 0 & -\frac{1}{h} & & \\ & 0 & -\frac{1}{h} & \\ & & \ddots & \ddots \\ & & & 0 & -\frac{1}{h} \\ & & & & 0 \end{bmatrix} \in \mathbb{R}^{n_1 \times n_1}.$$

and  $L = U^T$ . The damped Jacobi method reads

$$\begin{cases} \vec{z}_1 &= D^{-1}(U + U^T)\vec{u}_1^{(\sigma)} + D^{-1}\vec{f}_1^{\square}, \\ \vec{u}_1^{(\sigma+1)} &= \omega \vec{z}_1 + (1 - \omega)\vec{u}_1^{(\sigma)}, \end{cases} \quad (3.6)$$

where  $0 < \omega \leq 1$ .

**Theorem 3.3.** *The eigenvectors of  $K_1$  are the same as those for  $A_1$ , namely,*

$$\left[ \vec{v}_1^{(k)} \right]_i = \vec{v}_{1,i}^{(k)} = \sin(k\pi x_{1,i})$$

for  $k = 1, \dots, n_1$ . The eigenvalues of  $K_1$  are

$$\mu_1^{(k)}(\omega) = \omega \cos(k\pi h) + 1 - \omega = 1 - 2\omega \sin^2\left(\frac{k\pi h}{2}\right), \quad 1 \leq k \leq n_1. \quad (3.7)$$

*Proof.*

$$\begin{aligned} K_1 \vec{v}_1^{(k)} &= \vec{v}_1^{(k)} - \omega D^{-1} A_1 \vec{v}_1^{(k)} \\ &= \vec{v}_1^{(k)} - \omega \frac{h}{2} \lambda_1^{(k)} \vec{v}_1^{(k)} \\ &= \left( 1 - \omega \frac{h}{2} (1 - \cos(k\pi h)) \right) \vec{v}_1^{(k)} \\ &= (1 - \omega(1 - \cos(k\pi h))) \vec{v}_1^{(k)} \\ &= \left( 1 - 2\omega \sin^2\left(\frac{k\pi h}{2}\right) \right) \vec{v}_1^{(k)} \end{aligned}$$

So,

$$\mu_1^{(k)}(\omega) = 1 - \omega + \omega \cos(k\pi h) = 1 - 2\omega \sin^2\left(\frac{k\pi h}{2}\right).$$

□

**Remark 3.4.** *In the multi grid setting, we want the Jacobi method (the smoother) to have a "smooth" effect on the error. In other words, we want to damper high-frequency modes of the error faster than those of low-frequency. Recall, for the damped Jacobi method*

$$\vec{e}_1^{(\sigma+1)} = K_1 \vec{e}_1^{(\sigma)},$$

where

$$K_1 = I_1 - \frac{\omega h}{2} A_1.$$

Now, expand  $\vec{e}_1^{(\sigma)}$  in the basis of eigenvectors  $\{\vec{v}_1^{(k)}\}_{k=1}^{n_1}$ : There exist unique number

$$\epsilon_k^{(\sigma)} \in \mathbb{R}, \quad k = 1, 2, \dots, n_1.$$

such that

$$\vec{e}_1^{(\sigma)} = \sum_{k=1}^{n_1} \epsilon_k^{(\sigma)} \vec{v}_1^{(k)}.$$

Then

$$\vec{e}_1^{(\sigma+1)} = \sum_{k=1}^{n_1} \mu_1^{(k)}(\omega) \epsilon_k^{(\sigma)} \vec{v}_1^{(k)}. \quad (3.8)$$

### 3.2 A Basic Mesh Assumption

We assume that  $n_1 + 1 \geq 2$  is even. In this case, we can define

$$n_0 := \frac{n_1 + 1}{2} - 1.$$

For example, suppose  $n_1 = 3$ . Then

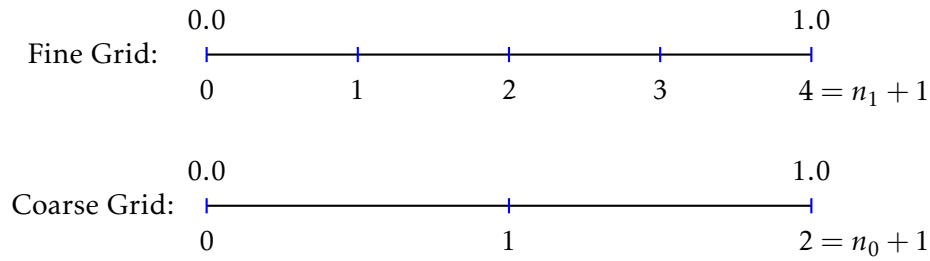


Figure 16: Fine and coarse grid for  $n_1 = 3$ .

**Definition 3.5. (High- and low- Frequency)** Suppose  $n_1 + 1$  is even, we say mode  $k, 1 \leq k \leq n_1$ , is of high frequency iff

$$n_0 + 1 = \frac{n_1 + 1}{2} \leq k \leq n_1.$$

Otherwise, it is of low frequency.

**Theorem 3.6.** The quantity

$$S(\omega) = \max_{\frac{n_1+1}{2} \leq k \leq n_1} |\mu_1^{(k)}(\omega)|,$$

is minimized by

$$\omega = \omega_0 = \frac{2}{3},$$

in which case

$$|\mu_1^{(k)}(\omega_0)| \leq \frac{1}{3} \quad (3.9)$$

for all  $\frac{n_1+1}{2} \leq k \leq n_1$ . More generally, if  $0 < \omega \leq 1$ , then

$$|\mu_1^{(k)}(\omega)| < 1. \quad (3.10)$$

*Proof.* Proof by plots of  $\mu_1^{(k)}(\omega) = 1 - 2\omega \sin^2\left(\frac{k\pi h}{2}\right)$  (Figure. 17):

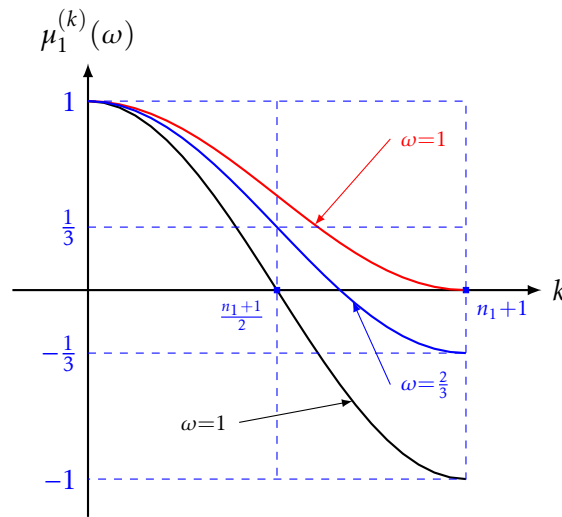


Figure 17: The eigenvalue of  $\mu_1^{(k)}(\omega) = 1 - 2\omega \sin^2\left(\frac{k\pi h}{2}\right)$ .

□

**Remark 3.7.** Recall, with  $\omega = \omega_0$ , we have

$$\vec{e}_1^{(\sigma+1)} = \sum_{k=1}^{n_1} \mu_1^{(k)}(\omega_0) \epsilon_k^{(\sigma)} \vec{v}_1^{(k)}.$$

Thus high-frequency modes will be damped faster than those of low-frequency. In fact the modes  $\frac{n_1+1}{2} \leq k \leq n_1$  will be reduced by at least  $\frac{1}{3}$  after a single smoothing iteration.

**1<sup>st</sup> MULTIGRID PRINCIPLE:** Many classical iterative methods have error smoothing property (High-frequency modes are damped more rapidly than those of low-frequency), but converge very slowly, especially as



$h \rightarrow 0$ .

2<sup>nd</sup> MULTIGRID PRINCIPLE: Low-frequency information(modes) is well approximated on a coarse grid.

**Theorem 3.8.** Let  $K_1 = K(\omega) = I_1 - \omega D^{-1}A_1$  be the error propagation matrix for damped Jacobi method for the model problem (3.2). Then

$$\rho(K_1) = \mu_1^{(1)}(\omega) = 1 - \mathcal{O}(h^2),$$

for all  $0 < \omega \leq 1$ , i.e., there exist  $0 < C_1 \leq C_2$  such that

$$0 \leq C_1 h^2 \leq 1 - \rho(K_1) \leq C_2 h^2.$$

*Proof.* Since  $\mu_1^{(k)}(\omega) = 1 - 2\omega \sin^2\left(\frac{k\pi h}{2}\right)$ , then

$$\rho(K_1) = \mu_1^{(1)}(\omega) = 1 - 2\omega \sin^2\left(\frac{\pi h}{2}\right).$$

By Taylor expansion, we have

$$\sin^2(x) = x^2 - \frac{1}{3}x^4 + \mathcal{O}(x^6).$$

Then

$$\rho(K_1) = \mu_1^{(1)}(\omega) = 1 - 2\omega \left( \frac{\pi^2}{4}h^2 - \frac{1}{3} \frac{\pi^4}{2^4}h^4 + \mathcal{O}(h^6) \right) = 1 - \mathcal{O}(h^2).$$

□

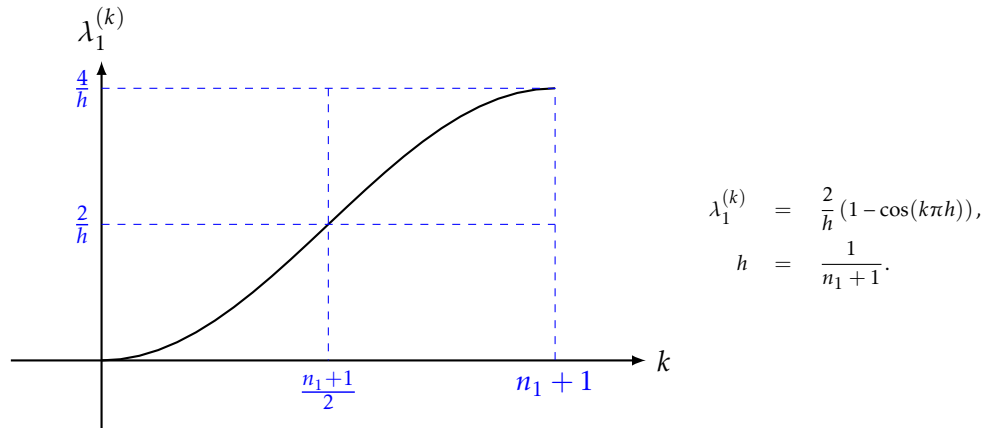
**Remark 3.9.** We observed that a smoothed error is well approximated on a coarse grid. We can show that if the error is smooth, then the residual is also smooth.

Recall (3.8) with  $\omega = \omega_0$ ,

$$\vec{e}_1^{(\sigma+1)} = \sum_{k=1}^{n_1} \mu_1^{(k)}(\omega_0) \epsilon_k^{(\sigma)} \vec{v}_1^{(k)}.$$

Since  $\vec{r}_1^{(\sigma+1)} = A_1 \vec{e}_1^{(\sigma+1)}$ ,

$$\vec{r}_1^{(\sigma+1)} = A_1 \vec{e}_1^{(\sigma+1)} = \sum_{k=1}^{n_1} \mu_1^{(k)}(\omega_0) \epsilon_k^{(\sigma)} A_1 \vec{v}_1^{(k)} = \sum_{k=1}^{n_1} \mu_1^{(k)}(\omega_0) \epsilon_k^{(\sigma)} \lambda_1^{(k)} \vec{v}_1^{(k)}. \quad (3.11)$$



Thus  $\vec{r}_1^{(\sigma+1)}$  will be well approximated on the coarse grid iff  $\vec{e}_1^{(\sigma+1)}$  has this property.

### 3.3 Prolongation and Restriction Operators

We first define the prolongation matrix  $P_0$ , then we set  $R_0 = P_0^T$ .

We approach this from the FEM point of view. Suppose  $u_H \in V_H$  is piecewise linear.

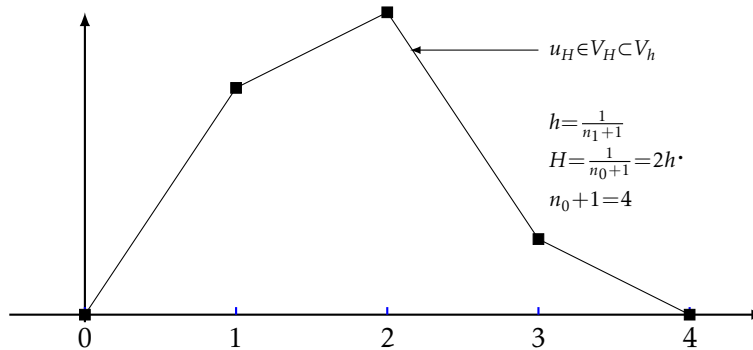
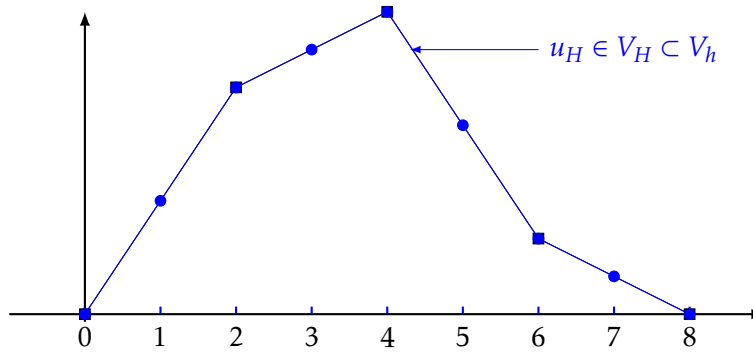


Figure 18: Coarse grid  $n_0 = 3$  in 1D.

Recall, based on our basic mesh assumption

$$n_0 = \frac{n_1 + 1}{2} - 1 \Leftrightarrow n_1 = 2(n_0 + 1) - 1.$$

In this case  $n_1 = 7$ .

Figure 19: Fine grid  $n_1 = 7$  in 1D.

Find a matrix that maps the 3 coarse grid DOF's into the 7 fine grid DOF's, i.e. find a matrix  $P_0 \in \mathbb{R}^{n_1 \times n_0}$  such that

$$P_0 \vec{u}_0 = \vec{u}_1 \in \mathbb{R}^{n_1},$$

where  $\vec{u}_0 \in \mathbb{R}^{n_0}$  is coordinate representation of  $u_H = u_0 \in V_H = V_0$  and  $\vec{u}_1 \in \mathbb{R}^{n_1}$  is the representation of  $u_h = u_1 \in V_h = V_1$  such that

$$u_h = u_H. \quad (\text{same piece-wise linear function})$$

For this example ( $n_0 = 3, n_1 = 7$ )

$$P_0 = \begin{bmatrix} \frac{1}{2} & & \\ \frac{1}{2} & \frac{1}{2} & \\ \frac{1}{2} & 1 & \frac{1}{2} \\ & \frac{1}{2} & 1 \\ & & \frac{1}{2} \end{bmatrix} \in \mathbb{R}^{7 \times 3}.$$

Therefore,

$$R_0 = \begin{bmatrix} \frac{1}{2} & 1 & \frac{1}{2} & & & & \\ & \frac{1}{2} & 1 & \frac{1}{2} & & & \\ & & \frac{1}{2} & 1 & \frac{1}{2} & & \\ & & & \frac{1}{2} & 1 & \frac{1}{2} & \\ & & & & \frac{1}{2} & 1 & \frac{1}{2} \\ & & & & & \frac{1}{2} & 1 \\ & & & & & & \frac{1}{2} \end{bmatrix} \in \mathbb{R}^{3 \times 7}.$$

**Definition 3.10.** Suppose that the positive integers  $n_0$  and  $n_1$  satisfy

$$n_1 = 2(n_0 + 1) - 1,$$

let  $\vec{u}_0 \in \mathbb{R}^{n_0}$ . The action of  $P_0 \in \mathbb{R}^{n_1 \times n_0}$  on  $\vec{u}_0$  is defined as follows

$$\begin{aligned} [P_0 \vec{u}_0]_1 &= \frac{1}{2} u_{0,1}, \\ [P_0 \vec{u}_0]_{n_1} &= \frac{1}{2} u_{0,n_0}, \\ [P_0 \vec{u}_0]_{2i} &= u_{0,i}, \quad 1 \leq i \leq n_0 \\ [P_0 \vec{u}_0]_{2i+1} &= \frac{1}{2} (u_{0,i} + u_{0,i+1}), \quad 1 \leq i \leq n_0 - 1. \end{aligned}$$

We define

$$R_0 = P_0^T \in \mathbb{R}^{n_0 \times n_1}. \quad (3.12)$$

**Theorem 3.11.** Let  $P_0, R_0$  be defined as above, with  $A_1$  defined as in (3.3). Suppose  $A_0 \in \mathbb{R}^{n_0 \times n_0}$  satisfies the Galerkin condition, i.e.

$$A_0 := R_0 A_1 P_0.$$

Then

$$A_0 = \begin{bmatrix} \frac{2}{H} & -\frac{1}{H} & & & \\ -\frac{1}{H} & \frac{2}{H} & -\frac{1}{H} & & \\ & \ddots & \ddots & \ddots & \\ & & -\frac{1}{H} & \frac{2}{H} & -\frac{1}{H} \\ & & & -\frac{1}{H} & \frac{2}{H} \end{bmatrix} \in \mathbb{R}^{n_0 \times n_0},$$

where

$$H = 2h = \frac{1}{n_0 + 1},$$

$A_0$  is clearly SPD and has the eigen-pairs

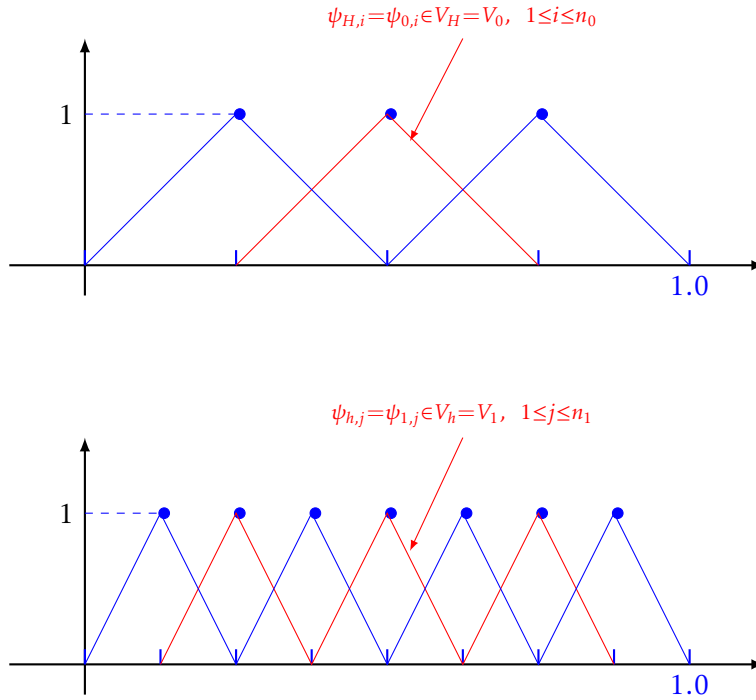
$$\begin{aligned} \left[ \vec{v}_0^{(k)} \right]_i &= \sin(k\pi x_{0,i}), \quad 1 \leq i \leq n_0, \\ \lambda_0^{(k)} &= \frac{2}{H} (1 - \cos(k\pi H)), \end{aligned} \quad (3.13)$$

for  $k = 1, \dots, n_0$ , where

$$x_{0,i} = iH, \quad 0 \leq i \leq n_0 + 1.$$

*Proof.* Here we give a proof that works in much more general settings. It is based on the FEM picture using piecewise linear elements.

Consider the Lagrange nodal basis for the coarse space



Since  $V_0 \subset V_1$ , each  $\psi_{0,i} \in V_1$ . Therefore, since  $B_1 = \{\psi_{1,j}\}_{j=1}^{n_1}$  is a basis for  $V_1$ ,  $\exists!$  numbers

$$p_{0,k,i} \in \mathbb{R}, \quad 1 \leq k \leq n_1,$$

such that

$$\psi_{0,i} = \sum_{k=1}^{n_1} p_{0,k,i} \psi_{1,k}. \quad (3.14)$$

Observe that, in present case where  $P_0$  is as defined in Definition.3.10,

$$[P_0]_{k,i} = p_{0,k,i}.$$

In fact, we could have used (3.14) in Definition.3.10. For instance, it should be clear that

$$\psi_{0,1} = \frac{1}{2} \psi_{1,1} + \psi_{1,2} + \frac{1}{2} \psi_{1,3}.$$

Now, as usual in the FEM setting, we define stiffness matrices via

$$[A_1]_{i,j} := [A_h]_{i,j} = a(\psi_{h,j}, \psi_{h,i}) = a(\psi_{1,j}, \psi_{1,i}).$$

On the coarse grid, we proceed similarly

$$\begin{aligned} [A_0]_{i,j} := [A_H]_{i,j} &= a(\psi_{H,j}, \psi_{H,i}) \\ &= a(\psi_{0,j}, \psi_{0,i}) \\ &\stackrel{(3.14)}{=} a\left(\sum_{k=1}^{n_1} p_{0,k,i} \psi_{1,k}, \sum_{\ell=1}^{n_1} p_{0,\ell,j} \psi_{1,\ell}\right) \end{aligned}$$

$$\begin{aligned}
&= \sum_{k=1}^{n_1} \sum_{\ell=1}^{n_1} p_{0,k,i} a(\psi_{1,k}, \psi_{1,\ell}) p_{0,\ell,j} \\
&= P_0^T A_1 P_0 \\
&= R_0 A_1 P_0.
\end{aligned} \tag{3.15}$$

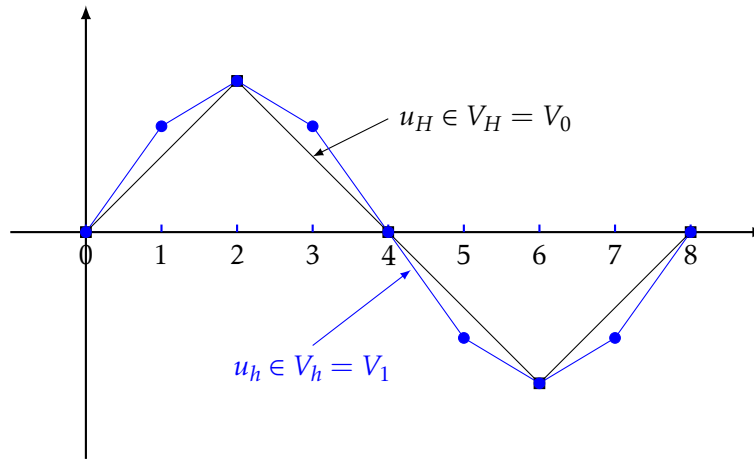
But this is the same as the matrix  $A_0$  which is defined to satisfy the Galerkin condition, i.e.

$$A_0 := R_0 A_1 P_0 \stackrel{(3.15)}{=} A_H := \begin{bmatrix} \frac{2}{H} & -\frac{1}{H} & & & \\ -\frac{1}{H} & \frac{2}{H} & -\frac{1}{H} & & \\ & \ddots & \ddots & \ddots & \\ & & -\frac{1}{H} & \frac{2}{H} & -\frac{1}{H} \\ & & & -\frac{1}{H} & \frac{2}{H} \end{bmatrix} \in \mathbb{R}^{n_0 \times n_0}.$$

□

### 3.4 Restriction Operator

Consider the fine grid function



Suppose  $\vec{u}_1 \in \mathbb{R}^{n_1}$  is the coordinate vector of  $u_h \in V_h = V_1$  w.r.t the lagrange nodal basis  $B_1$ . What is the action of  $R_0$  on  $\vec{u}_1$ ?

$$R_0 \vec{u}_1 = \begin{bmatrix} \frac{1}{2} & 1 & \frac{1}{2} & & & & \\ & \frac{1}{2} & 1 & \frac{1}{2} & & & \\ & & \frac{1}{2} & 1 & \frac{1}{2} & & \\ & & & \frac{1}{2} & 1 & \frac{1}{2} & \\ & & & & \frac{1}{2} & 1 & \frac{1}{2} \\ & & & & & \frac{1}{2} & 1 \\ & & & & & & \frac{1}{2} \end{bmatrix} \begin{bmatrix} u_{1,1} \\ u_{1,2} \\ u_{1,3} \\ u_{1,4} \\ u_{1,5} \\ u_{1,6} \\ u_{1,7} \end{bmatrix} = \begin{bmatrix} \frac{1}{2}u_{1,1} + u_{1,2} + \frac{1}{2}u_{1,3} \\ \frac{1}{2}u_{1,3} + u_{1,4} + \frac{1}{2}u_{1,5} \\ \frac{1}{2}u_{1,5} + u_{1,6} + \frac{1}{2}u_{1,7} \end{bmatrix}$$

**Definition 3.12.** The matrix  $\frac{1}{2}R_0$ , where  $R_0$  is defined as above, is the matrix representing the full weighting restriction operator.

**Remark 3.13.**  $\frac{1}{2}R_0$  gives a more geometrically meaningful result, its a proper weighted average.

### 3.5 Quantitative Error Analysis

Recall the result of Theorem.2.10 concerning the two grid algorithm:

$$\vec{e}_1^{k+1} = E_1 \vec{e}_1^k,$$

where

$$E_1 = (K_1^*)^{m_2} (I_1 - \tilde{\Pi}_1) (K_1)^{m_2}.$$

Recall that, for the present case,

$$K_1 = I_1 - \omega_0 \frac{h}{2} A_1.$$

$\tilde{\Pi}_1$  is the coarse grid Ritz projection matrix

$$\tilde{\Pi}_1 = R_0^T A_0^{-1} R_0 A_1.$$

Since our  $A_0$  satisfies the Galerkin condition

$$\tilde{\Pi}_1^2 = \tilde{\Pi}_1.$$

### 3.6 Some Technical Results

**Proposition 3.14.** *With the basic mesh assumption in place, et cetera,*

$$R_0 \vec{v}_1^{(k)} = \begin{cases} 2 \cos^2\left(\frac{k\pi h}{2}\right) \vec{v}_0^{(k)}, & 1 \leq k \leq n_0, \\ -2 \sin^2\left(\frac{(n_1+1-k)k\pi h}{2}\right) \vec{v}_0^{(n_1+1-k)}, & n_0 + 1 \leq k \leq n_1. \end{cases} \quad (3.16)$$

*Proof.* Consider for  $1 \leq i \leq n_0$ ,  $1 \leq k \leq n_0$ ,

$$\begin{aligned} \left[ R_0 \vec{v}_1^{(k)} \right]_i &= \frac{1}{2} \left\{ \sin(k\pi x_{1,2i-1}) + 2 \sin(k\pi x_{1,2i}) + \sin(k\pi x_{1,2i+1}) \right\} \\ &= \frac{1}{2} \left\{ \sin(k\pi x_{1,2i} - k\pi h) + 2 \sin(k\pi x_{1,2i}) + \sin(k\pi x_{1,2i} + k\pi h) \right\} \\ &= \frac{1}{2} \left\{ \sin(k\pi x_{1,2i}) \cos(k\pi h) - \cos(k\pi x_{1,2i}) \sin(k\pi h) \right. \\ &\quad + 2 \sin(k\pi x_{1,2i}) \\ &\quad + \left. \sin(k\pi x_{1,2i}) \cos(k\pi h) + \cos(k\pi x_{1,2i}) \sin(k\pi h) \right\} \\ &= \frac{1}{2} \left\{ 2 \sin(k\pi x_{1,2i}) \cos(k\pi h) + 2 \sin(k\pi x_{1,2i}) \right\} \\ &= \sin(k\pi x_{1,2i}) (\cos(k\pi h) + 1) \\ &= 2 \cos^2\left(\frac{k\pi h}{2}\right) \sin(k\pi x_{1,2i}) \\ &= 2 \cos^2\left(\frac{k\pi h}{2}\right) \left[ \vec{v}_0^{(k)} \right]_i, \quad 1 \leq k \leq n_0. \end{aligned}$$

This complete the first part. For the second part, assume  $n_0 + 1 \leq k \leq n_1$ . Now

$$\left[ \vec{v}_0^{(n_1+1-k)} \right]_i = \sin((n_1 + 1 - k)\pi i H)$$

$$\begin{aligned}
&= \sin\left((n_1 + 1 - k)\pi \frac{2i}{n_1 + 1}\right) \\
&= \sin(2\pi i - k\pi x_{0,i}) \\
&= \sin(-k\pi x_{0,i}) \\
&= -\sin(k\pi x_{0,i}) \\
&= -\left[\vec{v}_0^{(k)}\right]_i.
\end{aligned}$$

Similarly,

$$\cos((n_1 + 1 - k)\pi h) = -\cos(k\pi h).$$

The first calculation is still valid for  $n_0 + 1 \leq k \leq n_1$ .

$$\begin{aligned}
\left[R_0 \vec{v}_1^{(k)}\right]_i &= \frac{1}{2} \left\{ \sin(k\pi x_{1,2i-1}) + 2\sin(k\pi x_{1,2i}) + \sin(k\pi x_{1,2i+1}) \right\} \\
&= \frac{1}{2} \left\{ \sin(k\pi x_{1,2i} - k\pi h) + 2\sin(k\pi x_{1,2i}) + \sin(k\pi x_{1,2i} + k\pi h) \right\} \\
&= \frac{1}{2} \left\{ \sin(k\pi x_{1,2i}) \cos(k\pi h) - \cos(k\pi x_{1,2i}) \sin(k\pi h) \right. \\
&\quad + 2\sin(k\pi x_{1,2i}) \\
&\quad + \left. \sin(k\pi x_{1,2i}) \cos(k\pi h) + \cos(k\pi x_{1,2i}) \sin(k\pi h) \right\} \\
&= \frac{1}{2} \left\{ 2\sin(k\pi x_{1,2i}) \cos(k\pi h) + 2\sin(k\pi x_{1,2i}) \right\} \\
&= \sin(k\pi x_{1,2i}) (\cos(k\pi h) + 1) \\
&= (\cos(k\pi h) + 1) \left[\vec{v}_0^{(k)}\right]_i \\
&= -(1 - \cos((n_1 + 1 - k)\pi h)) \left[\vec{v}_0^{(n_1+1-k)}\right]_i \\
&= -2\sin^2\left(\frac{(n_1 + 1 - k)\pi h}{2}\right) \left[\vec{v}_0^{(n_1+1-k)}\right]_i.
\end{aligned}$$

□

**Proposition 3.15.** *With the usual assumption, for  $1 \leq k \leq n_0$ ,*

$$P_0 \vec{v}_0^{(k)} = \cos^2\left(\frac{k\pi h}{2}\right) \vec{v}_1^{(k)} - \sin^2\left(\frac{k\pi h}{2}\right) \vec{v}_1^{(n_1+1-k)}. \quad (3.17)$$

*Proof.* Observe, for  $1 \leq k \leq n_0$ , a simple calculation shows

$$\left[\vec{v}_1^{(n_1+1-k)}\right]_i = (-1)^{i+1} \left[\vec{v}_1^{(k)}\right]_i.$$

Now, set

$$\begin{aligned}
C_k &:= \cos^2\left(\frac{k\pi h}{2}\right) = \frac{1}{2}(1 + \cos(k\pi h)), \\
S_k &:= \sin^2\left(\frac{k\pi h}{2}\right) = \frac{1}{2}(1 - \cos(k\pi h)).
\end{aligned}$$



Then, for  $1 \leq i \leq n_1$ , it follows that

$$\begin{aligned}
 & C_k \left[ \vec{v}_1^{(k)} \right]_i - S_k \left[ \vec{v}_1^{(n_1+1-k)} \right]_i \\
 &= \begin{cases} \cos(k\pi h) \sin(k\pi x_{1,i}), & i \text{ odd}, \\ \sin(k\pi x_{1,i}), & i \text{ even}, \end{cases} \\
 &= \begin{cases} \cos(k\pi h) \sin(k\pi x_{1,2j-1}), & 1 \leq j \leq n_0 + 1, \\ \sin(k\pi x_{1,2j}), & 1 \leq j \leq n_0, \end{cases} \\
 &= \begin{cases} \frac{1}{2} \sin(k\pi h + k\pi x_{1,2j-1}) + \frac{1}{2} \sin(k\pi x_{1,2j-1} - k\pi h), & 1 \leq j \leq n_0 + 1, \\ \sin(k\pi x_{1,2j}), & 1 \leq j \leq n_0, \end{cases} \\
 &= \begin{cases} \frac{1}{2} \sin(k\pi x_{1,2j}) + \frac{1}{2} \sin(k\pi x_{1,2j-2}), & 1 \leq j \leq n_0 + 1, \\ \sin(k\pi x_{1,2j}), & 1 \leq j \leq n_0, \end{cases} \\
 &= \begin{cases} \frac{1}{2} \sin(k\pi x_{0,j}) + \frac{1}{2} \sin(k\pi x_{0,j-1}), & 1 \leq j \leq n_0 + 1, \\ \sin(k\pi x_{0,j}), & 1 \leq j \leq n_0, \end{cases} \\
 &= \begin{cases} \left[ P_0 \vec{v}_0^{(k)} \right]_{2j-1}, & 1 \leq j \leq n_0 + 1, \\ \left[ P_0 \vec{v}_0^{(k)} \right]_{2j}, & 1 \leq j \leq n_0, \end{cases}
 \end{aligned}$$

□

**Proposition 3.16.** *With the usual assumption,*

$$(I_1 - \tilde{\Pi}_1) \vec{v}_1^{(k)} = S_k \vec{v}_1^{(k)} + S_k \vec{v}_1^{(n_1+1-k)}, \quad (3.18)$$

for  $1 \leq k \leq n_0$ . Similarly,

$$(I_1 - \tilde{\Pi}_1) \vec{v}_1^{(n_1+1-k)} = C_k \vec{v}_1^{(k)} + C_k \vec{v}_1^{(n_1+1-k)}, \quad (3.19)$$

for  $1 \leq k \leq n_0 + 1$ , where, as before

$$\begin{aligned}
 C_k &: = \cos^2\left(\frac{k\pi h}{2}\right) = \frac{1}{2}(1 + \cos(k\pi h)), \\
 S_k &: = \sin^2\left(\frac{k\pi h}{2}\right) = \frac{1}{2}(1 - \cos(k\pi h)).
 \end{aligned}$$

*Proof.* Recall that

$$\tilde{\Pi}_1 = P_0 A_0^{-1} R_0 A_1 \in \mathbb{R}^{n_1 \times n_1} \quad \text{Ritz Projection,}$$

where

$$A_0 = R_0 A_1 P_0 \in \mathbb{R}^{n_0 \times n_0} \quad \text{Galerkin Condition.}$$

Then, for  $1 \leq k \leq n_0$ ,

$$\begin{aligned}
 \tilde{\Pi}_1 \vec{v}_1^{(k)} &= P_0 A_0^{-1} R_0 A_1 \vec{v}_1^{(k)} \\
 &= \lambda_1^{(k)} P_0 A_0^{-1} R_0 \vec{v}_1^{(k)}
 \end{aligned}$$

$$\begin{aligned}
& \stackrel{(3.16)}{=} 2C_k \lambda_1^{(k)} P_0 A_0^{-1} \vec{v}_0^{(k)} \\
& = \frac{2C_k \lambda_1^{(k)}}{\lambda_0^{(k)}} P_0 \vec{v}_0^{(k)} \\
& \stackrel{(3.17)}{=} \frac{2C_k \lambda_1^{(k)}}{\lambda_0^{(k)}} \left\{ C_k \vec{v}_1^{(k)} - S_k \vec{v}_1^{(n_1+1-k)} \right\}.
\end{aligned}$$

Observe,

$$\begin{aligned}
\frac{\lambda_1^{(k)}}{\lambda_0^{(k)}} &= \frac{\frac{2}{h} (1 - \cos(k\pi h))}{\frac{2}{H} (1 - \cos(k\pi H))} \\
&= \frac{H \sin^2\left(\frac{k\pi h}{2}\right)}{h \sin^2(k\pi h)} \\
&= \frac{2S_k}{\sin^2(k\pi h)}.
\end{aligned}$$

Thus

$$\tilde{\Pi}_1 \vec{v}_1^{(k)} = \frac{4C_k S_k}{\sin^2(k\pi h)} \left\{ C_k \vec{v}_1^{(k)} - S_k \vec{v}_1^{(n_1+1-k)} \right\}.$$

But

$$C_k S_k = \frac{1}{4} \sin^2(k\pi h),$$

as the reader can easily check. We have

$$\tilde{\Pi}_1 \vec{v}_1^{(k)} = C_k \vec{v}_1^{(k)} - S_k \vec{v}_1^{(n_1+1-k)}, \quad 1 \leq k \leq n_0. \quad (3.20)$$

Therefore, for  $1 \leq k \leq n_0$ ,

$$\begin{aligned}
(I_1 - \tilde{\Pi}_1) \vec{v}_1^{(k)} &= (1 - C_k) \vec{v}_1^{(k)} - S_k \vec{v}_1^{(n_1+1-k)} \\
&= S_k \left( \vec{v}_1^{(k)} - \vec{v}_1^{(n_1+1-k)} \right),
\end{aligned}$$

which is (3.18). Equation (3.19) is established in an analogous way.  $\square$

**Theorem 3.17.** (*Convergence of the one-sided two-grid method in  $\|\cdot\|_2$* ) Suppose  $m_2 = 0$  and  $\omega_0 = \frac{2}{3}$ , with  $m_1 \geq 1$ . Then

$$\|\vec{e}_1^{\ell+1}\|_2^2 \leq \left( \frac{1}{2} + \frac{1}{3m_1} \right) \|\vec{e}_1^\ell\|_2^2.$$

*Proof.* We begin with the basis expansion

$$\vec{e}_1^\ell = \sum_{k=1}^{n_1} \epsilon_k \vec{v}_1^{(k)}.$$

The error propagation matrix,  $E_1$ , for the two-grid scheme, in this case, satisfies

$$\vec{e}_1^{\ell+1} = E_1 \vec{e}_1^\ell,$$

where

$$E_1 = (I_1 - \tilde{\Pi}_1)K_1^{m_1}.$$

Hence,

$$\vec{e}_1^{\ell+1} = \sum_{k=1}^{n_1} \epsilon_k (I_1 - \tilde{\Pi}_1)K_1^{m_1} \vec{v}_1^{(k)}.$$

By Theorem.3.3,

$$\left| \mu_1^{(k)}(\omega_0) \right| \leq \frac{1}{3}, \quad n_0 + 1 \leq k \leq n_1,$$

and

$$\left| \mu_1^{(k)}(\omega) \right| < 1, \quad 1 \leq k \leq n_1.$$

By proposition.3.19,

$$(I_1 - \tilde{\Pi}_1)K_1^{m_1} \vec{v}_1^{(k)} = \alpha_k \left\{ \vec{v}_1^{(k)} + \vec{v}_1^{(n_1+1-k)} \right\}, \quad (3.21)$$

for  $1 \leq k \leq n_0 + 1$ , and , for  $1 \leq k \leq n_0 + 1$ ,

$$(I_1 - \tilde{\Pi}_1)K_1^{m_1} \vec{v}_1^{(n_1+1-k)} = \beta_k \left\{ \vec{v}_1^{(k)} + \vec{v}_1^{(n_1+1-k)} \right\}, \quad (3.22)$$

where

$$\alpha_k := \left( \mu_1^{(k)}(\omega_0) \right)^{m_1} S_k, \quad 1 \leq k \leq n_0 + 1,$$

and

$$\beta_k := \left( \mu_1^{(n_1+1-k)}(\omega_0) \right)^{m_1} C_k, \quad 1 \leq k \leq n_0 + 1.$$

Observe that we have extended the upper limit of the index k in (3.21) by 1, up to  $k = n_0 + 1$ . The result is still valid in this case. Further, observe that

$$\alpha_{n_0+1} = \beta_{n_0+1},$$

as the reader can easily check.

Now, notice that the eigenvalue satisfy

$$\begin{aligned} \left( \vec{v}_1^{(i)}, \vec{v}_1^{(j)} \right)_1 &= \sum_{m=1}^{n_1} \sin(i\pi x_{1,m}) \sin(j\pi x_{1,m}) \\ &= \delta_{ij} \frac{n_1 + 1}{2}. \end{aligned}$$

Now, we can estimate  $\alpha_k, \beta_k$  as follows:

$$|\alpha_k| \leq 1^{m_1} S_k \leq 1 \cdot \frac{1}{2}, \quad 1 \leq k \leq n_0 + 1,$$

and

$$|\beta_k| \leq \left( \frac{1}{2} \right)^{m_1} C_k \leq \left( \frac{1}{3} \right) \cdot 1, \quad 1 \leq k \leq n_0 + 1.$$

We can represent the error as

$$\begin{aligned} \vec{e}_1^{\ell+1} &= E_1 \vec{e}_1^{\ell} \\ &= \sum_{k=1}^{n_0+1} \delta_k \left( \epsilon_k \alpha_k + \epsilon_{n_1+1-k} \beta_k \right) \left( \vec{v}_1^{(k)} + \vec{v}_1^{(n_1+1-k)} \right), \end{aligned}$$

where

$$\delta_k = \begin{cases} \frac{1}{2}, & k = n_0 + 1 \\ 1, & \text{otherwise.} \end{cases}$$

Therefore,

$$\begin{aligned} \|\vec{e}_1^{\ell+1}\|_2^2 &= (\vec{e}_1^{\ell+1}, \vec{e}_1^{\ell+1})_2 \\ &= (n_1 + 1) \sum_{k=1}^{n_0+1} \delta_k (\epsilon_k^2 \alpha_k^2 + \epsilon_{n_1+1-k}^2 \beta_k^2 + 2\epsilon_k \epsilon_{n_1+1-k} \alpha_k \beta_k) \\ &\stackrel{a.g.m.i}{\leq} (n_1 + 1) \sum_{k=1}^{n_0+1} \delta_k (\epsilon_k^2 \alpha_k^2 + \epsilon_{n_1+1-k}^2 \beta_k^2 + (\epsilon_k^2 + \epsilon_{n_1+1-k}^2) |\alpha_k| |\beta_k|) \\ &\leq (n_1 + 1) \sum_{k=1}^{n_0+1} \delta_k \left\{ \epsilon_k^2 \left(\frac{1}{2}\right)^2 + \epsilon_{n_1+1-k}^2 \left(\frac{1}{3}\right)^{2m_1} + (\epsilon_k^2 + \epsilon_{n_1+1-k}^2) \left(\frac{1}{2}\right) \left(\frac{1}{2}\right)^{m_1} \right\}. \end{aligned}$$

Since,

$$\left(\frac{1}{3}\right)^{2m_1} \leq \left(\frac{1}{2}\right)^2,$$

we have

$$\begin{aligned} \|\vec{e}_1^{\ell+1}\|_2^2 &\leq \frac{n_1 + 1}{2} \left\{ \frac{1}{2} + \left(\frac{1}{3}\right)^{2m_1} \right\} \sum_{k=1}^{n_0+1} \delta_k (\epsilon_k^2 + \epsilon_{n_1+1-k}^2) \\ &= \left\{ \frac{1}{2} + \left(\frac{1}{3}\right)^{2m_1} \right\} \|\vec{e}_1^\ell\|_2^2. \end{aligned}$$

Finally,

$$\|\vec{e}_1^{\ell+1}\|_2 \leq \left\{ \frac{1}{2} + \left(\frac{1}{3}\right)^{2m_1} \right\}^{1/2} \|\vec{e}_1^\ell\|_2 := \gamma(m_1) \|\vec{e}_1^\ell\|_2.$$

□

$m_1$	1	2	3	...	$\infty$
$\gamma(m_1)$	0.91287	0.78176	0.73283	...	0.70710

**Remark 3.18.** The key observation is that  $\gamma(m_1)$  is  $h$ -independent.

### 3.7 Qualitative Two-Grid Method Convergence

Recall, for any  $\vec{v}_1 \in \mathbb{R}^{n_1}$ ,

$$\begin{aligned} \|\vec{v}_1\|_{A_1} &:= \sqrt{(\vec{v}_1, \vec{v}_1)_{A_1}} \\ &= \sqrt{(\vec{v}_1, A_1 \vec{v}_1)_1} \end{aligned}$$

$$\begin{aligned}
&= \sqrt{(A_1^{1/2} \vec{v}_1, A_1^{1/2} \vec{v}_1)_1} \\
&= \|A_1^{1/2} \vec{v}_1\|_1.
\end{aligned}$$

In a similar way, we can define

$$\begin{aligned}
\|\vec{v}_1\|_{A_1^2} &:= \sqrt{(\vec{v}_1, A_1^2 \vec{v}_1)_1} \\
&= \sqrt{(A_1 \vec{v}_1, A_1 \vec{v}_1)_1} \\
&= \|A_1 \vec{v}_1\|_1.
\end{aligned}$$

**Lemma 3.19.** *With the usual assumptions in this section, there exists a positive constant  $C_1 > 0$  such that*

$$\|(I_1 - \tilde{\Pi}_1) \vec{v}_1\|_1 \leq C_1 \sqrt{\Lambda_1^{-1}} \|(I_1 - \tilde{\Pi}_1) \vec{v}_1\|_{A_1}, \quad (3.23)$$

for all  $\vec{v}_1 \in \mathbb{R}^{n_1}$ , where

$$\Lambda_1 = \frac{4}{h_1}.$$

Here  $h_1 = h, h_0 = H = 2h$ .

*Proof.* Exercise. □

**Remark 3.20.** *We will prove this later in the general FEM framework. Question: What is the smallest (optimal) value of  $C_1$  in the present setting?*

**Theorem 3.21.** (*Approximation property*) *For any  $\vec{v}_1 \in \mathbb{R}^{n_1}$ ,*

$$\|(I_1 - \tilde{\Pi}_1) \vec{v}_1\|_{A_1} \leq C_1 \sqrt{\Lambda_1^{-1}} \|\vec{v}_1\|_{A_1^2}, \quad (3.24)$$

where

$$\Lambda_1 = \frac{4}{h_1}.$$

*Proof.* Let  $\vec{v}_1 \in \mathbb{R}^{n_1}$  be arbitrary. Then

$$\begin{aligned}
\|(I_1 - \tilde{\Pi}_1) \vec{v}_1\|_{A_1}^2 &= ((I_1 - \tilde{\Pi}_1) \vec{v}_1, A_1 (I_1 - \tilde{\Pi}_1) \vec{v}_1)_1 \\
&= ((I_1 - \tilde{\Pi}_1) \vec{v}_1, A_1 \vec{v}_1)_1 - ((I_1 - \tilde{\Pi}_1) \vec{v}_1, A_1 \tilde{\Pi}_1 \vec{v}_1)_1.
\end{aligned}$$

The second term on the RHS is zero, as we now show.

$$\begin{aligned}
((I_1 - \tilde{\Pi}_1) \vec{v}_1, \tilde{\Pi}_1 \vec{v}_1)_{A_1} &= ((I_1 - \tilde{\Pi}_1) \vec{v}_1, A_1 \tilde{\Pi}_1 \vec{v}_1)_1 \\
&= ((I_1 - \tilde{\Pi}_1) \vec{v}_1, A_1 (P_0 A_0^{-1} R_0 A_1) \vec{v}_1)_1 \\
&= ((I_1 - \tilde{\Pi}_1) \vec{v}_1, (A_1 P_0 A_0^{-1} R_0) A_1 \vec{v}_1)_1 \\
&= ((A_1 P_0 A_0^{-1} R_0)^T (I_1 - \tilde{\Pi}_1) \vec{v}_1, A_1 \vec{v}_1)_1 \\
&= (\tilde{\Pi}_1 (I_1 - \tilde{\Pi}_1) \vec{v}_1, \vec{v}_1)_{A_1}
\end{aligned}$$

$$= 0.$$

Since,  $\tilde{\Pi}_1(I_1 - \tilde{\Pi}_1) = \tilde{\Pi}_1 - \tilde{\Pi}_1^2 = \tilde{\Pi}_1 - \tilde{\Pi}_1 = 0$ . Therefore,

$$\begin{aligned} \|(I_1 - \tilde{\Pi}_1)\vec{v}_1\|_{A_1}^2 &= ((I_1 - \tilde{\Pi}_1)\vec{v}_1, A_1 \vec{v}_1)_{A_1} \\ &\stackrel{C.S.}{\leq} \|(I_1 - \tilde{\Pi}_1)\vec{v}_1\|_{A_1} \|A_1 \vec{v}_1\|_{A_1} \\ &= \|(I_1 - \tilde{\Pi}_1)\vec{v}_1\|_{A_1} \|\vec{v}_1\|_{A_1^2} \\ &\stackrel{(3.23)}{\leq} C_1 \sqrt{\Lambda_1^{-1}} \|(I_1 - \tilde{\Pi}_1)\vec{v}_1\|_{A_1} \|\vec{v}_1\|_{A_1^2}. \end{aligned}$$

The result follows. □

**Corollary 3.22.** Suppose that the usual assumptions from this section are in place, in particular, the Galerkin condition is satisfied in the definition of  $A_0$ . Then

$$((I_1 - \tilde{\Pi}_1)\vec{v}_1, (I_1 - \tilde{\Pi}_1)\vec{v}_1)_{A_1} = ((I_1 - \tilde{\Pi}_1)\vec{v}_1, \vec{v}_1)_{A_1},$$

or, equivalently,

$$((I_1 - \tilde{\Pi}_1)\vec{v}_1, \tilde{\Pi}_1 \vec{v}_1)_{A_1} = 0.$$

**Remark 3.23.** Estimate (3.24) is called an approximation property (or condition) with constant  $C_1$ . Of course this estimate was directly required by estimate (3.23).

### 3.8 Smoothing Revisited

In this discussion, we want to change our smooth a bit. Let's use Richardson's "smoothing" method.

$$K_1 = I_1 - \frac{1}{\Lambda_1} A_1, \quad \Lambda_1 = \frac{4}{h_1}.$$

Recall

$$\lambda_1^{(k)} = \frac{2}{h_1} (1 - \cos(k\pi h)).$$

So,

$$0 < \lambda_1^{(1)} < \lambda_1^{(2)} < \dots < \lambda_1^{(n_1)} < \frac{4}{h_1} = \Lambda_1.$$

$\Lambda_1$  is almost the spectral radius of  $A_1$ , the bound is asymptotically sharp.

For damped Jacobi, recall that

$$\begin{aligned} K_1 = K_1(\omega) &= I_1 - \omega D^{-1} A_1 \\ &= I_1 - \omega \frac{h_1}{2} A_1. \end{aligned}$$

If we take  $\omega = \frac{1}{2}$  in damped Jacobi, we get our Richardson's smoother.

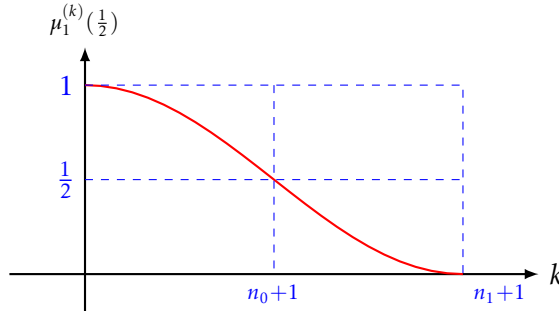
The eigenvalues of

$$K_1 = I_1 - \Lambda_1^{-1} A_1 = I_1 - \frac{h_1}{4} A_1,$$

are simply.

$$\mu_1^{(k)}\left(\frac{1}{2}\right) = \omega \cos k\pi h_1 + 1 - \omega$$

$$= 1 - \frac{\lambda_1^{(k)}}{\Lambda_1}.$$



So

$$\left| \mu_1^{(k)}\left(\frac{1}{2}\right) \right| = \mu_1^{(k)}\left(\frac{1}{2}\right) \leq \frac{1}{2}, \quad n_0 + 1 \leq k \leq n_1.$$

**Lemma 3.24.** (*Smoothing property for Richardson's smoother*) There is some constant  $C_2 > 0$ , such that

$$\|K_1^{m_1} \vec{v}_1\|_{A_1^2} \leq C_2 \sqrt{\Lambda_1} m_1^{-\frac{1}{2}} \|\vec{v}_1\|_{A_1}. \quad (3.25)$$

for all  $\vec{v}_1 \in \mathbb{R}^{n_1}$ . ( $\Lambda_1 = \frac{4}{h_1}$ ).

*Proof.*

$$\begin{aligned} \|K_1^{m_1} \vec{v}_1\|_{A_1^2} &= \|A_1 K_1^{m_1} \vec{v}_1\|_1^2 \\ &= (A_1 K_1^{m_1} \vec{v}_1, A_1 K_1^{m_1} \vec{v}_1)_1. \end{aligned}$$

Let us write

$$\vec{v}_1 = \sum_{k=1}^{n_1} v_k \vec{v}_1^{(k)}.$$

Then

$$\begin{aligned} \|K_1^{m_1} \vec{v}_1\|_{A_1^2} &= \frac{n_1 + 1}{2} \sum_{k=1}^{n_1} \left( \lambda_1^{(k)} v_k \right)^2 \left( \mu_1^{(k)}\left(\frac{1}{2}\right) \right)^{2m} \\ &= \Lambda_1 \left( \frac{n_1 + 1}{2} \right) \sum_{k=1}^{n_1} \left( \frac{\lambda_1^{(k)}}{\Lambda_1} \right) \left( 1 - \frac{\lambda_1^{(k)}}{\Lambda_1} \right)^{2m} \lambda_1^{(k)} v_k^2 \\ &= \Lambda_1 \max_{1 \leq k \leq n_1} \left\{ \left( \frac{\lambda_1^{(k)}}{\Lambda_1} \right) \left( 1 - \frac{\lambda_1^{(k)}}{\Lambda_1} \right)^{2m} \right\} \left( \frac{n_1 + 1}{2} \right) \sum_{k=1}^{n_1} \lambda_1^{(k)} v_k^2 \\ &= \Lambda_1 G(m) \|\vec{v}_1\|_{A_1}^2, \end{aligned}$$

where

$$G(m) := \max_{1 \leq k \leq n_1} \left\{ \left( \frac{\lambda_1^{(k)}}{\Lambda_1} \right) \left( 1 - \frac{\lambda_1^{(k)}}{\Lambda_1} \right)^{2m} \right\}.$$

Observe that, upon resealing

$$G(m) \leq \sup_{0 \leq x \leq 1} x(1-x)^{2m}.$$

Set

$$\begin{aligned} f(x) &= x(1-x)^{2m} = x(x-1)^{2m}. \\ 0 &= f'(x_0) = (x_0-1)^{2m} + x_0(2m)(x_0-1)^{2m-1} \\ \Leftrightarrow x_0 - 1 + x_0(2m) &= 0 \\ \Leftrightarrow (1+2m)x_0 &= 1 \\ \Leftrightarrow x_0 &= \frac{1}{2m+1} \\ \Leftrightarrow f(x_0) &= \frac{1}{2m+1} \left( \frac{2m}{2m+1} \right)^{2m}. \end{aligned}$$

Thus

$$G(m) \leq \frac{1}{2m+1} \left( \frac{2m}{2m+1} \right)^{2m} \leq \frac{1}{2m+1} \leq \frac{1}{2} m^{-1}.$$

Therefore

$$\|K_1^{m_1} \vec{v}_1\|_{A_1^2} \leq \sqrt{\frac{1}{2}} \sqrt{\Lambda_1} m_1^{-\frac{1}{2}} \|\vec{v}_1\|_{A_1}. \quad (3.26)$$

and  $C_2 = \sqrt{\frac{1}{2}}$ . □

**Theorem 3.25.** (*Convergence of the one-sided Two-grid Method with Richardson Smoothing*) Suppose  $m_2 = 00$ , and with  $\omega = \frac{1}{2}$  (Richardson). Then the two grid method obtained in this section converges provided  $m_1$  is sufficiently large, and we have the qualitative error estimate

$$\|\vec{e}_1^{\ell+1}\|_{A_1} \leq C_1 C_2 m_1^{-1/2} \|\vec{e}_1^{\ell}\|_{A_1}.$$

Where  $C_1, C_2 > 0$  are as given in Theorem.3.21 and Lemma.3.24, respectively.

*Proof.* Recall, the error propagation matrix in this case is

$$E_1 = (I_1 - \tilde{\Pi}_1) K_1^{m_1}.$$

So

$$\begin{aligned} \|\vec{e}_1^{\ell+1}\|_{A_1} &= \|(I_1 - \tilde{\Pi}_1) K_1^{m_1} \vec{e}_1^{\ell}\|_{A_1} \\ &\stackrel{(3.24)}{\leq} C_1 \sqrt{\Lambda_1^{-1}} \|K_1^{m_1} \vec{e}_1^{\ell}\|_{A_1^2} \\ &\stackrel{(3.25)}{\leq} C_1 \sqrt{\Lambda_1^{-1}} C_2 \sqrt{\Lambda_1} m_1^{-\frac{1}{2}} \|\vec{e}_1^{\ell}\|_{A_1} \\ &= C_1 C_2 m_1^{-\frac{1}{2}} \|\vec{e}_1^{\ell}\|_{A_1}. \end{aligned}$$

□

**Remark 3.26.** The method converges at a uniform,  $h$ -independent rate provide  $m_1 \geq 1$  is large enough so that

$$0 < C_1 C_2 m_1^{-\frac{1}{2}} < 1.$$

We have shown that  $C_2 = \sqrt{\frac{1}{2}}$  works, but since we do not have a quantitative estimate of  $C_1$  (yet), the convergence is just "qualitative".



## 4 Multigrid

The idea behind multigrid is to replace the exact solution in the coarse-grid connection by a recursive application of a "two-grid" method.

Suppose

$$1 \leq n_0 < n_1 < \cdots < n_\ell < \cdots < n_L \in \mathbb{Z}.$$

Suppose

$$R_{\ell-1} \in \mathbb{R}^{n_{\ell-1} \times n_\ell}, \quad 1 \leq \ell \leq L,$$

is full rank, i.e.

$$\text{rank}(R_{\ell-1}) = n_{\ell-1}.$$

Set

$$P_{\ell-1} = R_{\ell-1}^T \in \mathbb{R}^{n_\ell \times n_{\ell-1}}, \quad 1 \leq \ell \leq L.$$

Assume that we have a family of SPD matrices

$$A_\ell \in \mathbb{R}^{n_\ell \times n_\ell}, \quad 0 \leq \ell \leq L.$$

Symmetry here is understood with respect to the canonical product  $(\cdot, \cdot)_\ell : \mathbb{R}^{n_\ell} \times \mathbb{R}^{n_\ell} \rightarrow \mathbb{R}$ , which is defined as

$$(\vec{u}_\ell, \vec{v}_\ell)_\ell = \vec{u}_1^T \vec{v}_\ell = \sum_{j=1}^{n_\ell} u_{\ell,j} v_{\ell,j},$$

for all  $\vec{u}_\ell, \vec{v}_\ell \in \mathbb{R}^{n_\ell}$ . We define  $(\cdot, \cdot)_{A_\ell}$  via

$$(\vec{u}_\ell, \vec{v}_\ell)_{A_\ell} = (\vec{u}_\ell, A_\ell \vec{v}_\ell)_\ell.$$

Our goal is to find an efficient iterative solver for the equation

$$A_L \vec{u}_L^E = \vec{f}_L,$$

given  $\vec{f}_L \in \mathbb{R}^{n_L}$ . As usual, we define

$$\vec{e}_L^\square = \vec{u}_L^E - \vec{u}_L^\square,$$

where  $\vec{u}_L^\square \in \mathbb{R}^{n_L}$  is an approximate solution. Similarly,

$$\vec{r}_L^\square := \vec{f}_L - A_L \vec{u}_L^\square = A_L \vec{e}_L^\square.$$

**Definition 4.1.** We say that the Galerkin Condition holds iff

$$A_{\ell-1} = R_{\ell-1} A_\ell P_{\ell-1},$$

for all  $1 \leq \ell \leq L$ . Otherwise, the Galerkin Condition fails.

**Definition 4.2.** Suppose  $0 \leq \ell \leq L$  and  $\vec{g}_\ell, \vec{u}_\ell^{(0)} \in \mathbb{R}^{n_\ell}$  are given. The recursive multi grid operator MG is defined as follows:

$$\vec{u}_\ell^{(3)} := \text{MG}(\vec{g}_\ell, \ell, \vec{u}_\ell^{(0)})$$

- If  $(\ell = 0)$  then

$$\vec{u}_\ell^{(3)} := A_0^{-1} \vec{g}_0$$

- else if  $(1 \leq \ell \leq L)$  then

– Pre-Smoothing:

$$* \vec{u}_\ell^{(1,0)} := \vec{u}_\ell^{(0)}$$

$$* \vec{u}_\ell^{(1,\sigma+1)} = \vec{u}_\ell^{(1,\sigma)} + S_\ell(\vec{g}_\ell - A_\ell \vec{u}_\ell^{(1,\sigma)}), \quad 0 \leq \sigma \leq m_1 - 1$$

$$* \vec{u}_\ell^{(1)} := \vec{u}_\ell^{(1,m_1)}$$

– Coarse-Grid Correction:

$$* \vec{r}_\ell^{(1)} := \vec{g}_\ell - A_\ell \vec{u}_\ell^{(1)}$$

$$* \vec{r}_{\ell-1}^{(1)} := R_{\ell-1} \vec{r}_\ell^{(1)}$$

$$* \vec{q}_{\ell-1}^{(1,0)} := \vec{0}$$

$$* \vec{q}_{\ell-1}^{(1,\sigma+1)} := \text{MG}(\vec{r}_{\ell-1}^{(1)}, \ell-1, \vec{q}_{\ell-1}^{(1,\sigma)}), \quad 0 \leq \sigma \leq p-1$$

$$* \vec{q}_{\ell-1}^{(1)} := \vec{q}_{\ell-1}^{(1,p)}$$

$$* \vec{u}_\ell^{(2)} := \vec{u}_\ell^{(1)} + P_{\ell-1} \vec{q}_{\ell-1}^{(1)}$$

– Post-Smoothing:

$$* \vec{u}_\ell^{(3,0)} := \vec{u}_\ell^{(2)}$$

$$* \vec{u}_\ell^{(3,\sigma+1)} := \vec{u}_\ell^{(3,\sigma)} + S_1^T(\vec{g}_\ell - A_\ell \vec{u}_\ell^{(3,\sigma)}), \quad 0 \leq \sigma \leq m_2 - 1$$

- $\vec{u}_\ell^{(3)} = \vec{u}_\ell^{(3,m_2)}$

**Algorithm 4.3.** Let  $m_1, m_2, p$  be nonnegative integers. Let  $\vec{u}_L^k \in \mathbb{R}^{n_L}$  be given. Then

$$\vec{u}_L^{k+1} = \text{MG}(\vec{f}_L, L, \vec{u}_L^k)$$

defines the genetic multigrid iteration method for solving

$$A_L \vec{u}_L^E = \vec{f}_L.$$

**Definition 4.4.** (*one-sided multigrid method*) The multigrid (multilevel) algorithm is called one-sided iff  $m_2 = 0$  and  $m_1 \geq 1$ . The algorithm is called a W-cycle iff  $p = 2$ , and a V-cycle iff  $p = 1$ . The algorithm is symmetric iff  $m_1 = m_2 = m$ .

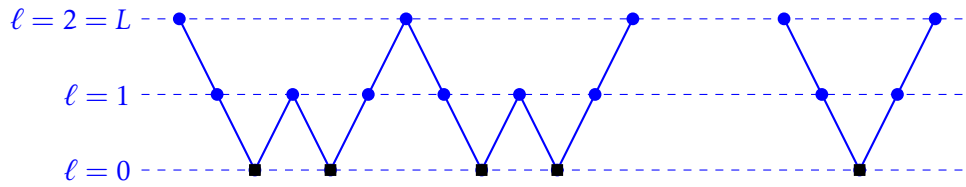


Figure 20: One full W- and V- Cycle for three levels. Left One full W-Cycle, Right: One full V- Cycle .

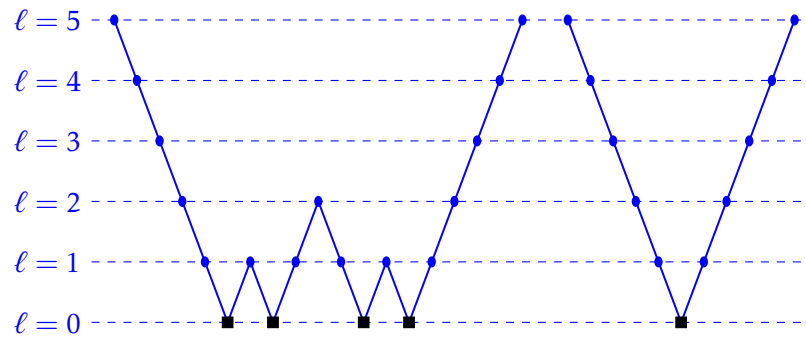


Figure 21: One full W- and V- Cycle for six levels. Left One full W-Cycle, Right: One full V- Cycle .

**Remark 4.5.** In the case  $L = 0$ , we just do a direct solve. When  $L = 1$ , we recover the two-grid (two-level) method.

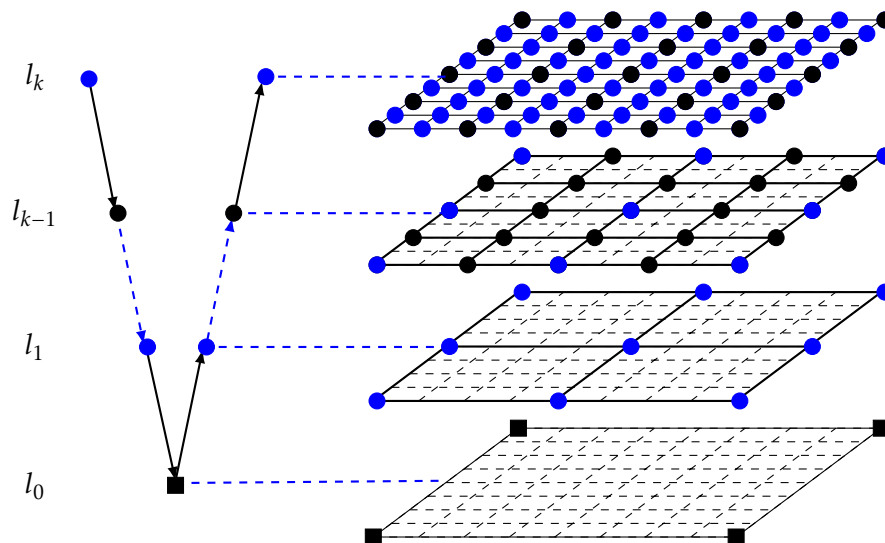


Figure 22: Mesh on the fine and coarse grid of the multigrid method in 2D.

We now want to find the error propagation matrix for the general multilevel algorithm.

**Definition 4.6.** Let  $\ell \geq 0$ . Define for  $\ell = 0$

$$E_0 := 0 \in \mathbb{R}^{n_0 \times n_0},$$

and, for  $\ell \geq 1$ ,

$$E_\ell := (K_\ell^*)^{m_2} (I_\ell - P_{\ell-1} (I_{\ell-1} - E_{\ell-1}^p) \Pi_{\ell-1}) K_\ell^{m_1} \in \mathbb{R}^{n_\ell \times n_\ell},$$

where

$$\left. \begin{aligned} K_\ell &:= I_\ell - S_\ell A_\ell \\ K_\ell^* &= I_\ell - S_\ell^T A_\ell \end{aligned} \right\} \in \mathbb{R}^{n_\ell \times n_\ell},$$

and

$$\Pi_{\ell-1} := A_{\ell-1}^{-1} R_{\ell-1} \in \mathbb{R}^{n_{\ell-1} \times n_\ell}.$$

For further use, let us also define

$$\tilde{\Pi}_\ell := P_{\ell-1} \Pi_{\ell-1} \in \mathbb{R}^{n_\ell \times n_\ell}.$$

**Remark 4.7.** Observe that, for any  $\vec{u}_\ell, \vec{v}_\ell \in \mathbb{R}^{n_\ell}$ ,

$$(\vec{u}_\ell, K_\ell \vec{v}_\ell)_{A_\ell} = (K_\ell^* \vec{u}_\ell, \vec{v}_\ell)_{A_\ell}.$$

More generally, we use "\*" to denote the adjoint w.r.t  $(\cdot, \cdot)_{A_\ell}$ .

**Theorem 4.8.** Suppose that  $\vec{u}_\ell^E, \vec{g}_\ell \in \mathbb{R}^{n_\ell}$  satisfy

$$A_\ell \vec{u}_\ell^E = \vec{g}_\ell.$$

Then, given  $\vec{u}_\ell^{(0)} \in \mathbb{R}^{n_\ell}$ ,

$$\vec{u}_\ell^E - \text{MG}(\vec{g}_\ell, \ell, \vec{u}_\ell^{(0)}) = E_\ell (\vec{u}_\ell^E - \vec{u}_\ell^{(0)}),$$

where  $E_\ell$  is the recursively -defined matrix from Definition.4.6. In particular

$$\vec{e}_L^{k+1} = E_L \vec{e}_L^k.$$

*Proof.* The proof is by induction. Cases  $\ell = 0$  and  $\ell = 1$  (two-grid method) are clear.

Induction Hypothesis: Assume that the result is true for level  $\ell - 1$ . Suppose that  $\vec{q}_{\ell-1} \in \mathbb{R}^{n_{\ell-1}}$  solves

$$A_{\ell-1} \vec{q}_{\ell-1} = \vec{r}_{\ell-1}^{(1)}.$$

Then

$$\vec{q}_{\ell-1} - \text{MG}(\vec{r}_{\ell-1}^{(1)}, \ell - 1, \vec{0}) = E_{\ell-1} (\vec{q}_{\ell-1} - \vec{0}) = E_{\ell-1} \vec{q}_{\ell-1}.$$

Written in the language of Algorithm.4.3,

$$\vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1,1)} = E_{\ell-1} \vec{q}_{\ell-1}.$$

So

$$\vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1,p)} = E_{\ell-1}^p \vec{q}_{\ell-1},$$

or

$$\vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)} = E_{\ell-1}^p \vec{q}_{\ell-1}.$$

Therefore

$$\vec{q}_{\ell-1}^{(1)} = (I_{\ell-1} - E_{\ell-1}^p) \vec{q}_{\ell-1}.$$

Now,

$$\begin{aligned} \vec{q}_{\ell-1} &:= A_{\ell-1}^{-1} \vec{r}_{\ell-1}^{(1)} \\ &= A_{\ell-1}^{-1} R_{\ell-1} \left( \vec{g}_{\ell} - A_{\ell} \vec{u}_{\ell}^{(1)} \right) \\ &= A_{\ell-1}^{-1} R_{\ell-1} A_{\ell} \left( \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(1)} \right) \\ &= \Pi_{\ell-1} \left( \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(1)} \right). \end{aligned}$$

Hence

$$\vec{q}_{\ell-1}^{(1)} = (I_{\ell-1} - E_{\ell-1}^p) \Pi_{\ell-1} \left( \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(1)} \right).$$

Putting it all together,

$$\begin{aligned} \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(3)} &= (K_{\ell}^*)^{m_2} \left( \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(2)} \right), \\ \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(2)} &= \vec{u}_{\ell}^E - \left( \vec{u}_{\ell}^{(1)} + P_{\ell-1} \vec{q}_{\ell-1}^{(1)} \right) \\ &= \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(1)} - P_{\ell-1} (I_{\ell-1} - E_{\ell-1}^p) \Pi_{\ell-1} \left( \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(1)} \right) \\ &= (I_{\ell} - P_{\ell-1} (I_{\ell-1} - E_{\ell-1}^p) \Pi_{\ell-1}) \left( \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(1)} \right), \\ \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(1)} &= K_{\ell}^{m_1} \left( \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(0)} \right), \end{aligned}$$

we have

$$\vec{u}_{\ell}^E - \vec{u}_{\ell}^{(3)} = E_{\ell-1} \left( \vec{u}_{\ell}^E - \vec{u}_{\ell}^{(0)} \right).$$

□

**Definition 4.9.** (*Assumption (A0) and Assumption (A1)*) We say that the stiffness matrices satisfy the Galerkin Condition, Assumption (A0), iff

$$A_{\ell-1} = R_{\ell-1} A_{\ell} P_{\ell-1}, \quad 1 \leq \ell \leq L. \quad (4.1)$$

We say that the *Weakened Galerkin Condition, Assumption (A1)*, holds iff

$$(\vec{v}_{\ell}, \vec{v}_{\ell})_{A_{\ell}} \geq (\Pi_{\ell-1} \vec{v}_{\ell}, \Pi_{\ell-1} \vec{v}_{\ell})_{A_{\ell-1}}, \quad (4.2)$$

for all  $\vec{v}_{\ell} \in \mathbb{R}^{n_{\ell}}$  and all  $1 \leq \ell \leq L$ .

We will show that A0  $\Rightarrow$  A1. To do this, we first need the following:

**Lemma 4.10.** *Assumption (A0), the (strong) Galerkin condition, implies that  $\tilde{\Gamma}_{\ell} = \tilde{\Gamma}_{\ell}^2$ . But  $\tilde{\Gamma}_{\ell}^* = \tilde{\Gamma}_{\ell}$  holds even without (A0).*

*Proof.* If the Galerkin Condition holds then

$$A_{\ell-1} = R_{\ell-1} A_{\ell} P_{\ell-1}, \quad 1 \leq \ell \leq L.$$

Recall that

$$\tilde{\Pi}_\ell = P_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1} A_\ell, \quad 1 \leq \ell \leq L.$$

Consequently,

$$\begin{aligned} \tilde{\Pi}_\ell^2 &= P_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1} A_\ell P_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1} A_\ell \\ &= P_{\ell-1} A_{\ell-1}^{-1} (R_{\ell-1} A_\ell P_{\ell-1}) A_{\ell-1}^{-1} R_{\ell-1} A_\ell \\ &= P_{\ell-1} A_{\ell-1}^{-1} A_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1} A_\ell \\ &= P_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1} A_\ell \\ &= \tilde{\Pi}_\ell. \end{aligned}$$

Next, let  $\vec{u}_\ell, \vec{v}_\ell \in \mathbb{R}^{n_\ell}$  be arbitrary. Then just using the definition of  $\Pi_{\ell-1}, \tilde{\Pi}_\ell$ , which hold independent of (A0),

$$\begin{aligned} (\vec{u}_\ell, \tilde{\Pi}_\ell \vec{v}_\ell)_{A_\ell} &= (\vec{u}_\ell, A_\ell \tilde{\Pi}_\ell \vec{v}_\ell)_\ell \\ &= (\vec{u}_\ell, A_\ell P_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1} A_\ell \vec{v}_\ell)_\ell \\ &= (\vec{u}_\ell, (A_\ell P_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1}) A_\ell \vec{v}_\ell)_\ell \\ &= \left( (A_\ell P_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1})^T \vec{u}_\ell, A_\ell \vec{v}_\ell \right)_\ell \\ &= (\tilde{\Pi}_\ell \vec{u}_\ell, A_\ell \vec{v}_\ell)_\ell \\ &= (\tilde{\Pi}_\ell \vec{u}_\ell, \vec{v}_\ell)_{A_\ell}. \end{aligned}$$

So,

$$\tilde{\Pi}_\ell^* = \tilde{\Pi}_\ell.$$

□

**Corollary 4.11.**  $(I_\ell - \tilde{\Pi}_\ell)^2 = I_\ell - \tilde{\Pi}_\ell$  and  $(I_\ell - \tilde{\Pi}_\ell)^* = I_\ell - \tilde{\Pi}_\ell$ .

**Lemma 4.12.** (*Assumption (A0) and Assumption (A1)*) *Assumption (A0) implies Assumption (A1).*

*Proof.* First, for any  $\vec{v}_\ell \in \mathbb{R}^{n_\ell}$ , consider

$$\begin{aligned} (\Pi_{\ell-1} \vec{v}_\ell, \Pi_{\ell-1} \vec{v}_\ell)_{A_{\ell-1}} &= (\Pi_{\ell-1} \vec{v}_\ell, A_{\ell-1} \Pi_{\ell-1} \vec{v}_\ell)_{\ell-1} \\ &= (\Pi_{\ell-1} \vec{v}_\ell, A_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1} A_\ell \vec{v}_\ell)_{\ell-1} \\ &= (\Pi_{\ell-1} \vec{v}_\ell, R_{\ell-1} A_\ell \vec{v}_\ell)_{\ell-1} \\ &= (R_{\ell-1}^T \Pi_{\ell-1} \vec{v}_\ell, A_\ell \vec{v}_\ell)_\ell \\ &= (P_{\ell-1} \Pi_{\ell-1} \vec{v}_\ell, A_\ell \vec{v}_\ell)_\ell \\ &= (P_{\ell-1} \Pi_{\ell-1} \vec{v}_\ell, \vec{v}_\ell)_{A_\ell} \\ &= (\tilde{\Pi}_\ell \vec{v}_\ell, \vec{v}_\ell)_{A_\ell}. \end{aligned}$$

Using the last calculation

$$\begin{aligned} &(\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - (\Pi_{\ell-1} \vec{v}_\ell, \Pi_{\ell-1} \vec{v}_\ell)_{A_{\ell-1}} \\ &= (\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - (\tilde{\Pi}_\ell \vec{v}_\ell, \vec{v}_\ell)_{A_\ell} \end{aligned}$$

$$\begin{aligned}
&= \left( (I_\ell - \tilde{\Pi}_\ell) \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell} \\
&\stackrel{\text{Cor. (4.11)}}{=} \left( (I_\ell - \tilde{\Pi}_\ell)^2 \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell} \\
&\stackrel{\text{Cor. (4.11)}}{=} \left( (I_\ell - \tilde{\Pi}_\ell) \vec{v}_\ell, (I_\ell - \tilde{\Pi}_\ell) \vec{v}_\ell \right)_{A_\ell} \\
&= \left\| (I_\ell - \tilde{\Pi}_\ell) \vec{v}_\ell \right\|_{A_\ell}^2 \geq 0.
\end{aligned}$$

Thus (A0) implies (A1).  $\square$

**Corollary 4.13.** (*Assumption (A2)*) *Assumption (A1) is equivalent to the following statement (Assumption (A2)): for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$  and  $1 \leq \ell \leq L$*

$$\left( (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell, \vec{u}_\ell \right)_{A_\ell} \geq 0. \quad (4.3)$$

*Proof.* We showed in the proof of lemma 4.12, using only the definition of  $\Pi_{\ell-1}, \tilde{\Pi}_\ell$ , that

$$\begin{aligned}
&(\vec{u}_\ell, \vec{u}_\ell)_{A_\ell} - (\Pi_{\ell-1} \vec{u}_\ell, \Pi_{\ell-1} \vec{u}_\ell)_{A_{\ell-1}} \\
&= (\vec{u}_\ell, \vec{v}_\ell)_{A_\ell} - (\tilde{\Pi}_\ell \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} \\
&= \left( (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell, \vec{u}_\ell \right)_{A_\ell} \\
&= \left( \vec{u}_\ell, (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell \right)_{A_\ell}.
\end{aligned}$$

So (A1) holds iff statement (4.3) holds.  $\square$

**Definition 4.14.** We say that Assumption (A2) holds for a multilevel method iff statement (4.3) holds.

**Theorem 4.15.** Suppose that the weakened Galerkin Condition, Assumption (A1), holds, or equivalently that Assumption (A2) holds. Then if  $m_1 = m_2 = m$ , for all  $0 \leq \ell \leq L$

$$E_\ell = E_\ell^*,$$

and

$$(E_\ell \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} \geq 0, \quad \forall \vec{u}_\ell \in \mathbb{R}^{n_\ell}.$$

Assumption (A1) is required only for the second statement.

*Proof.* By induction: The case  $\ell = 0$  is trivial. For  $\ell = 1$  (the two-grid algorithm) the proof follows as in the proof of Theorem (2.12).

Induction Hypothesis: Assume that

$$(E_{\ell-1} \vec{u}_{\ell-1}, \vec{v}_{\ell-1})_{A_{\ell-1}} = (\vec{u}_{\ell-1}, E_{\ell-1} \vec{v}_{\ell-1})_{A_{\ell-1}},$$

and

$$(E_{\ell-1} \vec{u}_{\ell-1}, \vec{v}_{\ell-1})_{A_{\ell-1}} \geq 0$$

for all  $\vec{u}_{\ell-1}, \vec{v}_{\ell-1} \in \mathbb{R}^{n_{\ell-1}}$ . We will make use of the definition of  $\Pi_{\ell-1}, \tilde{\Pi}_\ell$ , a number of times:

$$\Pi_{\ell-1} = A_{\ell-1}^{-1} R_{\ell-1} A_\ell.$$

So

$$A_{\ell-1}\Pi_{\ell-1} = R_{\ell-1}A_{\ell}.$$

And

$$\tilde{\Pi}_{\ell} = P_{\ell-1}\Pi_{\ell-1} = P_{\ell-1}A_{\ell-1}^{-1}R_{\ell-1}A_{\ell}.$$

We have

$$\begin{aligned}
& (E_{\ell}\vec{u}_{\ell}, \vec{v}_{\ell})_{A_{\ell}} \\
&= \left( (K_{\ell}^*)^{m_2} (I_{\ell} - P_{\ell-1} (I_{\ell-1} - E_{\ell-1}^P) \Pi_{\ell-1}) K_{\ell}^{m_1} \vec{u}_{\ell}, \vec{v}_{\ell} \right)_{A_{\ell}} \\
&= \left( (K_{\ell}^*)^m (I_{\ell} - P_{\ell-1} (I_{\ell-1} - E_{\ell-1}^P) \Pi_{\ell-1}) K_{\ell}^m \vec{u}_{\ell}, \vec{v}_{\ell} \right)_{A_{\ell}} \\
&= \left( (I_{\ell} - P_{\ell-1} (I_{\ell-1} - E_{\ell-1}^P) \Pi_{\ell-1}) K_{\ell}^m \vec{u}_{\ell}, K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} \\
&= \left( (I_{\ell} - P_{\ell-1} \Pi_{\ell-1} + P_{\ell-1} E_{\ell-1}^P \Pi_{\ell-1}) K_{\ell}^m \vec{u}_{\ell}, K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} \\
&= \left( (I_{\ell} - P_{\ell-1} \Pi_{\ell-1}) K_{\ell}^m \vec{u}_{\ell}, K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( P_{\ell-1} E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} \\
&= \left( (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{u}_{\ell}, K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( P_{\ell-1} E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( P_{\ell-1} E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( P_{\ell-1} E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, A_{\ell} K_{\ell}^m \vec{v}_{\ell} \right)_{\ell} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, P_{\ell-1}^T A_{\ell} K_{\ell}^m \vec{v}_{\ell} \right)_{\ell-1} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, R_{\ell-1} A_{\ell} K_{\ell}^m \vec{v}_{\ell} \right)_{\ell-1} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, A_{\ell-1} \Pi_{\ell-1} K_{\ell}^m \vec{v}_{\ell} \right)_{\ell-1} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, \Pi_{\ell-1} K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell-1}} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell-1}} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( A_{\ell-1} \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{v}_{\ell} \right)_{\ell-1} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( R_{\ell-1} A_{\ell} K_{\ell}^m \vec{u}_{\ell}, E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{v}_{\ell} \right)_{\ell-1} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( A_{\ell} K_{\ell}^m \vec{u}_{\ell}, R_{\ell-1}^T E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{v}_{\ell} \right)_{\ell} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( A_{\ell} K_{\ell}^m \vec{u}_{\ell}, P_{\ell-1} E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{v}_{\ell} \right)_{\ell} \\
&= \left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} + \left( K_{\ell}^m \vec{u}_{\ell}, P_{\ell-1} E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{v}_{\ell} \right)_{A_{\ell}} \\
&= (\vec{u}_{\ell}, E_{\ell} \vec{v}_{\ell})_{A_{\ell}}.
\end{aligned}$$

Symmetry is proven. Notice we have not yet used Assumption (A1), only the definition of  $\Pi_{\ell-1}, \tilde{\Pi}_{\ell}$ , which are assumed to always hold.

Now, we setting  $\vec{v}_{\ell} = \vec{u}_{\ell}$  in the last calculation, we have

$$\begin{aligned}
(\vec{u}_{\ell}, E_{\ell} \vec{u}_{\ell})_{A_{\ell}} &= \underbrace{\left( K_{\ell}^m \vec{u}_{\ell}, (I_{\ell} - \tilde{\Pi}_{\ell}) K_{\ell}^m \vec{u}_{\ell} \right)_{A_{\ell}}}_{\geq 0, \text{ by (A2)}} + \underbrace{\left( E_{\ell-1}^P \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell}, \Pi_{\ell-1} K_{\ell}^m \vec{u}_{\ell} \right)_{A_{\ell-1}}}_{\geq 0, \text{ by Induction Hypothesis}} \geq 0.
\end{aligned}$$

for any  $\vec{u}_{\ell} \in \mathbb{R}^{n_{\ell}}$ . The result is proven.  $\square$



**Remark 4.16.** In the proof above we used the fact that any positive integer power of a self-adjoint positive semi-definite matrix is also positive semi-definite. This is easy to prove and is left as an exercise.

**Definition 4.17.** (Assumption (A3): Strong approximation property and Assumption (A4): weakened approximation property) We say that the multilevel algorithm satisfies the Strong approximation property, or Assumption (A3), iff for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$  and  $1 \leq \ell \leq L$

$$\|(I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell\|_\ell^2 \leq C_3^2 \rho_\ell^{-1} \|(I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell\|_{A_\ell}^2, \quad (4.4)$$

for some  $C_3 > 0$  that is independent of  $\ell$ , where  $\rho_\ell = \rho(A_\ell)$ . The multilevel algorithm satisfies the Weakened approximation property, or Assumption (A4), iff for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$  and  $1 \leq \ell \leq L$

$$\left( (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell, \vec{u}_\ell \right)_{A_\ell} \leq C_4^2 \rho_\ell^{-1} \|A_\ell \vec{u}_\ell\|_\ell^2, \quad (4.5)$$

for some  $C_4 > 0$  that is independent of  $\ell$ .

**Theorem 4.18.** If the Galerkin condition, Assumption (A0), holds, then (A3) implies (A4).

*Proof.* Since (A0) holds

$$(I_\ell - \tilde{\Pi}_\ell)^2 = I_\ell - \tilde{\Pi}_\ell.$$

Therefore

$$\begin{aligned} \|(I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell\|_{A_\ell}^2 &= \left( (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell, (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell \right)_{A_\ell} \\ &= \left( (I_\ell - \tilde{\Pi}_\ell)^2 \vec{u}_\ell, \vec{u}_\ell \right)_{A_\ell} \\ &= \left( (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell, \vec{u}_\ell \right)_{A_\ell} \\ &= \left( (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell, A_\ell \vec{u}_\ell \right)_\ell \\ &\stackrel{c.s.}{\leq} \|(I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell\|_\ell \|A_\ell \vec{u}_\ell\|_\ell \\ &\stackrel{(A3)}{\leq} C_3 \rho_\ell^{-1/2} \|(I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell\|_{A_\ell} \|A_\ell \vec{u}_\ell\|_\ell. \end{aligned}$$

Thus

$$\|(I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell\|_{A_\ell} \leq C_3 \rho_\ell^{-1/2} \|A_\ell \vec{u}_\ell\|_\ell. \quad (4.6)$$

Squaring, we have

$$\left( (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell, \vec{u}_\ell \right)_{A_\ell} \leq C_3^2 \rho_\ell^{-1} \|A_\ell \vec{u}_\ell\|_\ell^2,$$

which is the desired result with  $C_4 = C_3$ . □

## 4.1 Richardson's Smoother

**Definition 4.19.** Recall that Richardson's method is defined via

$$\vec{u}_\ell^{(\sigma+1)} = \vec{u}_\ell^{(\sigma)} + \omega^{-1} \left( \vec{g}_\ell - A_\ell \vec{u}_\ell^{(\sigma)} \right),$$

where  $\omega$  is a parameter, In this case

$$K_\ell = I_\ell - \omega^{-1} A_\ell = K_\ell^*.$$

Choosing

$$C_s \rho_\ell \geq \omega := \Lambda_\ell \geq \rho_\ell = \rho(A_\ell), \quad 1 \leq \ell \leq L, \exists C_s \geq 1,$$

we obtain Richardson's smoother. In this case

$$K_\ell = I_\ell - \Lambda_\ell^{-1} A_\ell = K_\ell^*.$$

**Definition 4.20.** (Assumption (A5): First Smoothing Property) We say that the multilevel scheme satisfies the First Smoothing Property, Assumption (A5), iff

$$\|K_\ell^m \vec{u}_\ell\|_{A_\ell^2} \leq C_5 \rho_\ell^{1/2} m^{-1/2} \|\vec{u}_\ell\|_{A_\ell}, \quad (4.7)$$

for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$  and  $1 \leq \ell \leq L$ , for some  $C_5 > 0$  that is independent of  $\ell$ .

**Theorem 4.21.** Richardson's smoother satisfies that first smoothing property, Assumption (A5).

*Proof.* The proof is similar to the proof of Lemma 3.24. □

## 4.2 Convergence of the Two-level Method Revisited

**Theorem 4.22.** Suppose that  $L = 1$  (Two-level)  $m_1 = m \geq 1$  and  $m_2 = 0$  (once-sided). Suppose that Assumption (A0) (Galerkin Condition), Assumption (A3) (Strong Approximation property) and (A5) (1<sup>st</sup> smoothing property) all hold. Then

$$\left\| \vec{u}_1^E - \text{TG}_1 \left( \vec{f}_1, \vec{u}_1^{(0)} \right) \right\|_{A_1} \leq C_3 C_5 m^{-1/2} \left\| \vec{u}_1^E - \vec{u}_1^{(0)} \right\|_{A_1}$$

where

$$A_1 \vec{u}_1^E = \vec{f}_1.$$

Written another way,

$$\left\| \vec{e}_1^{k+1} \right\|_{A_1} \leq C_3 C_5 m^{-1/2} \left\| \vec{e}_1^k \right\|_{A_1}$$

*Proof.* Recall that, in the present case,

$$E_1 = (I_1 - \tilde{\Pi}_1) K_1^m,$$

and

$$\vec{e}_1^{k+1} = E_1 \vec{e}_1^k,$$

or, equivalently

$$\vec{u}_1^E - \text{TG}_1 \left( \vec{f}_1, \vec{u}_1^{(0)} \right) = E_1 \left( \vec{u}_1^E - \vec{u}_1^{(0)} \right).$$

We showed in the proof of Theorem 4.18 that Assumption (A0) and (A3) imply that

$$\left\| (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell \right\|_{A_\ell} \leq C_3 \rho_\ell^{-1/2} \|A_\ell \vec{u}_\ell\|_\ell, \quad (4.8)$$

for any  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$ . Applying (4.8) (with  $\ell = 1$ ), and using Assumption (A5), we have

$$\begin{aligned} \|\vec{e}_1^{k+1}\|_{A_1} &= \left\| (I_1 - \tilde{\Pi}_1) K_1^m \vec{e}_1^k \right\|_{A_1} \\ &\stackrel{(4.8)}{\leq} C_3 \rho_1^{-1/2} \|A_1 K_1^m \vec{e}_1^k\|_1 \\ &\stackrel{(A5)}{\leq} C_3 \rho_1^{-1/2} C_5 \rho_1^{1/2} m^{-1/2} \|\vec{e}_1^k\|_{A_1} \\ &= C_3 C_5 m^{-1/2} \|\vec{e}_1^k\|_{A_1}. \end{aligned}$$

□

Before we go to the next result, we need a technical lemma:

**Lemma 4.23.** *For Richardson's smoother we have*

$$\|K_\ell \vec{v}_\ell\|_{A_\ell} \leq \|\vec{v}_\ell\|_{A_\ell} \quad (4.9)$$

$$(K_\ell \vec{v}_\ell, \vec{v}_\ell)_\ell \leq (\vec{v}_\ell, \vec{v}_\ell)_\ell. \quad (4.10)$$

for all  $\vec{v}_\ell \in \mathbb{R}^{n_\ell}, \ell \geq 0$

*Proof.* 1. Let  $\vec{v}_\ell \in \mathbb{R}^{n_\ell}$  be arbitrary. Suppose  $\{\vec{\omega}_\ell^{(1)}\}, \{\vec{\omega}_\ell^{(2)}, \dots, \{\vec{\omega}_\ell^{(n_\ell)}\}\}$  is an orthonormal basis of eigenvalue of  $A_\ell$  w.r.t  $\|\cdot\|_\ell$ . Then there exist unique  $\alpha_1, \alpha_2, \dots, \alpha_{n_\ell} \in \mathbb{R}$  such that

$$\vec{v}_\ell = \sum_{k=1}^{n_\ell} \alpha_k \vec{\omega}_\ell^{(k)}.$$

Recall

$$K_\ell = I_\ell - \Lambda_\ell^{-1} A_\ell,$$

with

$$\rho(A_\ell) =: \rho_\ell \leq \Lambda_\ell \leq C_s \rho_\ell,$$

where  $C_s \geq 1$  is independent of  $\ell$ . Then

$$K_\ell \vec{\omega}_\ell^{(k)} = \left(1 - \frac{\lambda_\ell^{(k)}}{\Lambda_\ell}\right) \vec{\omega}_\ell^{(k)} =: \mu_\ell^{(k)} \vec{\omega}_\ell^{(k)}.$$

Thus

$$\begin{aligned} \|K_\ell \vec{v}_\ell\|_{A_\ell}^2 &= (K_\ell \vec{v}_\ell, A_\ell K_\ell \vec{v}_\ell)_\ell \\ &= \sum_{i=1}^{n_\ell} \left(\mu_\ell^{(i)}\right)^2 \lambda_\ell^{(i)} \alpha_i^2. \end{aligned}$$

Since

$$0 \leq \left(\mu_\ell^{(i)}\right)^2 = \left(1 - \frac{\lambda_\ell^{(k)}}{\Lambda_\ell}\right)^2 \leq 1,$$

we have

$$\|K_\ell \vec{v}_\ell\|_{A_\ell}^2 \leq \|\vec{v}_\ell\|_{A_\ell}^2.$$

Hence

$$\|K_\ell \vec{v}_\ell\|_{A_\ell} \leq \|\vec{v}_\ell\|_{A_\ell}.$$

2.

$$\begin{aligned} (K_\ell \vec{v}_\ell, \vec{v}_\ell)_\ell &= \left( K_\ell \sum_{k=1}^{n_\ell} \alpha_k \vec{\omega}_\ell^{(k)}, \sum_{k=1}^{n_\ell} \alpha_k \vec{\omega}_\ell^{(k)} \right)_\ell \\ &= \left( \sum_{k=1}^{n_\ell} \alpha_k K_\ell \vec{\omega}_\ell^{(k)}, \sum_{k=1}^{n_\ell} \alpha_k \vec{\omega}_\ell^{(k)} \right)_\ell \\ &= \left( \sum_{k=1}^{n_\ell} \alpha_k \mu_\ell^{(k)} \vec{\omega}_\ell^{(k)}, \sum_{k=1}^{n_\ell} \alpha_k \vec{\omega}_\ell^{(k)} \right)_\ell \\ &= \sum_{k=1}^{n_\ell} \alpha_k^2 \mu_\ell^{(k)} \\ &\leq \sum_{k=1}^{n_\ell} \alpha_k^2 \\ &= (\vec{v}_\ell, \vec{v}_\ell)_\ell. \end{aligned}$$

□

**Theorem 4.24.** (*Convergence of the one-sided W-cycle with Richardson Smoothing*) Suppose that  $p \geq 2$  (Two-level)  $m_1 = m \geq 1$  and  $m_2 = 0$  (once-sided). Suppose, further that Assumption (A0) (Galerkin Condition), Assumption (A3) (Strong Approximation property) hold and the smoothing is done by Richardson's method. Then for any  $0 < \gamma < 1$ ,  $m$  can be chosen large enough so that

$$\left\| \vec{u}_\ell^E - \text{MG}(\vec{g}_\ell, \ell, \vec{u}_\ell^{(0)}) \right\|_{A_\ell} \leq \gamma \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell}$$

for any  $\ell \geq 0$ , where

$$A_\ell \vec{u}_\ell^E = \vec{g}_\ell.$$

*Proof.* First, observe that Richardson's smoother satisfies Assumption (A5) (Theorem.4.21).

The proof is by induction. The cases  $\ell = 0$  and  $\ell = 1$  (Theorem.4.22) are clearly true. Now, define  $\vec{q}_{\ell-1} \in \mathbb{R}^{n_{\ell-1}}$  such that

$$A_{\ell-1} \vec{q}_{\ell-1} = \vec{r}_{\ell-1}^{(1)}.$$

Then

$$\begin{aligned} \vec{u}_\ell^E - \text{MG}(\vec{g}_\ell, \ell, \vec{u}_\ell^{(0)}) &= \vec{u}_\ell^E - \vec{u}_\ell^{(2)} \\ &= \vec{u}_\ell^E - \left\{ \vec{u}_\ell^{(1)} + P_{\ell-1} \vec{q}_{\ell-1}^{(1)} \right\} \\ &= \vec{u}_\ell^E - \left\{ \vec{u}_\ell^{(1)} + P_{\ell-1} \vec{q}_{\ell-1} - P_{\ell-1} \vec{q}_{\ell-1} + P_{\ell-1} \vec{q}_{\ell-1}^{(1)} \right\} \\ &= \vec{u}_\ell^E - \left( \vec{u}_\ell^{(1)} + P_{\ell-1} \vec{q}_{\ell-1} \right) + P_{\ell-1} \left( \vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)} \right) \end{aligned}$$

$$= \vec{u}_\ell^E - \text{TG}_\ell(\vec{g}_\ell, \vec{u}_\ell^{(0)}) + P_{\ell-1}(\vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)}).$$

Suppose that  $m$  satisfies

$$m \geq \left( \frac{C_3 C_5}{\gamma - \gamma^p} \right)^2.$$

Then

$$\begin{aligned} \left\| \vec{u}_\ell^E - \text{MG}(\vec{g}_\ell, \ell, \vec{u}_\ell^{(0)}) \right\|_{A_\ell} &\leq \left\| \vec{u}_\ell^E - \text{TG}_\ell(\vec{g}_\ell, \vec{u}_\ell^{(0)}) \right\|_{A_\ell} + \left\| P_{\ell-1}(\vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)}) \right\|_{A_\ell} \\ &\stackrel{\text{Thm. 4.22}}{\leq} C_3 C_5 m^{-1/2} \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell} + \left\| P_{\ell-1}(\vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)}) \right\|_{A_\ell}. \end{aligned} \quad (4.11)$$

Now, observe that, for any  $\vec{\omega}_{\ell-1} \in \mathbb{R}^{n_{\ell-1}}$ ,

$$\begin{aligned} \left\| P_{\ell-1} \vec{\omega}_{\ell-1} \right\|_{A_\ell}^2 &= (P_{\ell-1} \vec{\omega}_{\ell-1}, P_{\ell-1} \vec{\omega}_{\ell-1})_{A_\ell} \\ &= (P_{\ell-1} \vec{\omega}_{\ell-1}, A_\ell P_{\ell-1} \vec{\omega}_{\ell-1})_\ell \\ &= (\vec{\omega}_{\ell-1}, P_{\ell-1}^T A_\ell P_{\ell-1} \vec{\omega}_{\ell-1})_{\ell-1} \\ &= (\vec{\omega}_{\ell-1}, R_{\ell-1} A_\ell P_{\ell-1} \vec{\omega}_{\ell-1})_{\ell-1} \\ &\stackrel{(A0)}{=} (\vec{\omega}_{\ell-1}, A_{\ell-1} \vec{\omega}_{\ell-1})_{\ell-1} \\ &= (\vec{\omega}_{\ell-1}, \vec{\omega}_{\ell-1})_{A_{\ell-1}} \\ &= \left\| \vec{\omega}_{\ell-1} \right\|_{A_{\ell-1}}^2. \end{aligned}$$

In the proof of Theorem 4.8, we showed that

$$\begin{aligned} \vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)} &= E_{\ell-1}^p \vec{q}_{\ell-1} \\ &= E_{\ell-1}^p \Pi_{\ell-1} \left( \vec{u}_\ell^E - \vec{u}_\ell^{(1)} \right) \\ &= E_{\ell-1}^p \Pi_{\ell-1} K_\ell^m \left( \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right). \end{aligned}$$

Induction Hypothesis: Assume

$$\left\| E_{\ell-1} \vec{\omega}_{\ell-1} \right\|_{A_{\ell-1}} \leq \gamma \left\| \vec{\omega}_{\ell-1} \right\|_{A_{\ell-1}}$$

is true for any  $\vec{\omega}_{\ell-1} \in \mathbb{R}^{n_{\ell-1}}$ . Therefore

$$\begin{aligned} \left\| P_{\ell-1}(\vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)}) \right\|_{A_\ell} &= \left\| \vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)} \right\|_{A_{\ell-1}} \\ &= \left\| E_{\ell-1}^p \Pi_{\ell-1} K_\ell^m \left( \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right) \right\|_{A_{\ell-1}} \\ &\stackrel{\text{ind.hyp.}}{\leq} \gamma^p \left\| \Pi_{\ell-1} K_\ell^m \left( \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right) \right\|_{A_{\ell-1}}. \end{aligned}$$

Since we are assuming the Galerkin Condition, it follows that

$$\left\| \Pi_{\ell-1} \vec{\omega}_\ell \right\|_{A_{\ell-1}} = \left\| \tilde{\Pi}_\ell \vec{\omega}_\ell \right\|_{A_\ell}.$$

Furthermore,

$$\left\| \tilde{\Pi}_\ell \vec{\omega}_\ell \right\|_{A_\ell}^2 = (\tilde{\Pi}_\ell \vec{\omega}_\ell, \tilde{\Pi}_\ell \vec{\omega}_\ell)_{A_\ell}$$

$$\begin{aligned}
&= (\tilde{\Pi}_\ell^2 \vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \\
&\stackrel{(A0)}{=} (\tilde{\Pi}_\ell \vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \\
&\leq \|\tilde{\Pi}_\ell \vec{\omega}_\ell\|_{A_\ell} \|\vec{\omega}_\ell\|_{A_\ell}.
\end{aligned}$$

So

$$\|\tilde{\Pi}_\ell \vec{\omega}_\ell\|_{A_\ell} \leq \|\vec{\omega}_\ell\|_{A_\ell}. \quad \text{stability} \quad (4.12)$$

Therefore,

$$\begin{aligned}
\left\| P_{\ell-1} \left( \vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)} \right) \right\|_{A_\ell} &\leq \gamma^p \left\| \Pi_{\ell-1} K_\ell^m \left( \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right) \right\|_{A_{\ell-1}} \\
&= \gamma^p \left\| \tilde{\Pi}_\ell K_\ell^m \left( \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right) \right\|_{A_\ell} \\
&\stackrel{(4.12)}{\leq} \gamma^p \left\| K_\ell^m \left( \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right) \right\|_{A_\ell} \\
&\stackrel{\text{Lem. 4.23}}{\leq} \gamma^p \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell}
\end{aligned}$$

Combining this with estimate (4.11), we have

$$\begin{aligned}
\left\| \vec{u}_\ell^E - \text{MG} \left( \vec{g}_\ell, \ell, \vec{u}_\ell^{(0)} \right) \right\|_{A_\ell} &\leq C_3 C_5 m^{-1/2} \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell} + \left\| P_{\ell-1} \left( \vec{q}_{\ell-1} - \vec{q}_{\ell-1}^{(1)} \right) \right\|_{A_\ell} \\
&\leq C_3 C_5 m^{-1/2} \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell} + \gamma^p \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell} \\
&\leq (C_3 C_5 m^{-1/2} + \gamma^p) \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell} \\
&\leq \left( C_3 C_5 \left( \frac{C_3 C_5}{\gamma - \gamma^p} \right)^2 \right)^{-1/2} + \gamma^p \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell} \\
&= \gamma \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_{\ell-1}}.
\end{aligned}$$

□

### 4.3 Convergence of the V-cycle Algorithm

**Definition 4.25.** (*Assumption (A6): Second Smoothing Property*) We say that the multilevel scheme satisfies the *Second Smoothing Property*, Assumption (A6), iff there is some  $C_6 > 0$  such that

$$\|\vec{v}_\ell\|_\ell^2 \leq \rho_\ell C_6^2 (\bar{K}_\ell \vec{v}_\ell, \vec{v}_\ell)_\ell, \quad (4.13)$$

for all  $\vec{v}_\ell \in \mathbb{R}^{n_\ell}$  and  $\ell \geq 1$ , where

$$\bar{K}_\ell := (I_\ell - K_\ell^* K_\ell) A_\ell^{-1}.$$

**Lemma 4.26.** Let  $J_\ell \in \mathbb{R}^{n_\ell \times n_\ell}$  and  $J_\ell = J_\ell^*$ . Then

$$(J_\ell \vec{v}_\ell, J_\ell \vec{v}_\ell)_{A_\ell} - (J_\ell^2 \vec{v}_\ell, J_\ell^2 \vec{v}_\ell)_{A_\ell} \leq (\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - (J_\ell \vec{v}_\ell, J_\ell \vec{v}_\ell)_{A_\ell}, \quad (4.14)$$

for any  $\vec{v}_\ell \in \mathbb{R}^{n_\ell}$

*Proof.* Since  $A_\ell$  is SPD.

$$\begin{aligned} 0 &\leq \left\| (I_\ell - J_\ell^2) \vec{v}_\ell \right\|_{A_\ell}^2 \\ &= \left( (I_\ell - J_\ell^2) \vec{v}_\ell, (I_\ell - J_\ell^2) \vec{v}_\ell \right)_{A_\ell} \\ &= \left( I_\ell \vec{v}_\ell - J_\ell^2 \vec{v}_\ell, I_\ell \vec{v}_\ell - J_\ell^2 \vec{v}_\ell \right)_{A_\ell} \\ &= (\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - (J_\ell^2 \vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - (\vec{v}_\ell, J_\ell^2 \vec{v}_\ell)_{A_\ell} + (J_\ell^2 \vec{v}_\ell, J_\ell^2 \vec{v}_\ell)_{A_\ell} \\ &= (\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - (J_\ell \vec{v}_\ell, J_\ell \vec{v}_\ell)_{A_\ell} - (J_\ell \vec{v}_\ell, J_\ell \vec{v}_\ell)_{A_\ell} + (J_\ell^2 \vec{v}_\ell, J_\ell^2 \vec{v}_\ell)_{A_\ell} \\ &= (\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - 2(J_\ell \vec{v}_\ell, J_\ell \vec{v}_\ell)_{A_\ell} + (J_\ell^2 \vec{v}_\ell, J_\ell^2 \vec{v}_\ell)_{A_\ell}. \end{aligned}$$

So

$$(J_\ell \vec{v}_\ell, J_\ell \vec{v}_\ell)_{A_\ell} - (J_\ell^2 \vec{v}_\ell, J_\ell^2 \vec{v}_\ell)_{A_\ell} \leq (\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - (J_\ell \vec{v}_\ell, J_\ell \vec{v}_\ell)_{A_\ell}.$$

□

**Lemma 4.27.** For any  $\vec{v}_\ell \in \mathbb{R}^{n_\ell}$

$$(\Pi_{\ell-1} \vec{v}_\ell, \Pi_{\ell-1} \vec{v}_\ell)_{A_{\ell-1}} = (\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - \left( (I_\ell - \tilde{\Pi}_\ell) \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell}. \quad (4.15)$$

*Proof.* Recall that we always have (Note: this is not from Galerkin Condition, but the definition of  $\Pi_{\ell-1}$ .)

$$R_{\ell-1} A_\ell = A_{\ell-1} \Pi_{\ell-1},$$

and

$$\tilde{\Pi}_\ell = P_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1} A_\ell = P_{\ell-1} \Pi_{\ell-1}.$$

Then

$$\begin{aligned} (\Pi_{\ell-1} \vec{v}_\ell, \Pi_{\ell-1} \vec{v}_\ell)_{A_{\ell-1}} &= (\Pi_{\ell-1} \vec{v}_\ell, A_{\ell-1} \Pi_{\ell-1} \vec{v}_\ell)_{\ell-1} \\ &= (\Pi_{\ell-1} \vec{v}_\ell, R_{\ell-1} A_\ell \vec{v}_\ell)_{\ell-1} \\ &= (R_{\ell-1}^T \Pi_{\ell-1} \vec{v}_\ell, A_\ell \vec{v}_\ell)_\ell \\ &= (P_{\ell-1} \Pi_{\ell-1} \vec{v}_\ell, A_\ell \vec{v}_\ell)_\ell \\ &= (\tilde{\Pi}_\ell \vec{v}_\ell, A_\ell \vec{v}_\ell)_\ell \\ &= (\tilde{\Pi}_\ell \vec{v}_\ell, \vec{v}_\ell)_{A_\ell} \\ &= (\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} - \left( (I_\ell - \tilde{\Pi}_\ell) \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell}. \end{aligned}$$

□

**Theorem 4.28.** Suppose that Assumption Weakened Galerkin Condition (A1), Weakened approximation property (A4) and Second smoothing property (A6) hold. Suppose that  $p = 1$ ,  $m_1 = m_2 = m = 1$  and  $S_\ell = S_\ell^T$ . Then

$$0 \leq (E_\ell \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} \leq \frac{C_4^2 C_6^2}{C_4^2 C_6^2 + 1} (\vec{u}_\ell, \vec{u}_\ell)_{A_\ell}$$

for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$ .

*Proof.* Recall, since  $P = 1$ ,  $m_1 = m_2 = m = 1$  and  $S_\ell = S_\ell^T$ ,

$$E_\ell = K_\ell (I_\ell - \tilde{\Pi}_\ell) K_\ell + K_\ell P_{\ell-1} E_{\ell-1} \Pi_{\ell-1} K_\ell.$$

In particular, notice that

$$K_\ell^* = I_\ell - S_\ell^T = I_\ell - S_\ell A_\ell = K_\ell.$$

Now, set

$$T_1 := \left( (I_\ell - \tilde{\Pi}_\ell) \vec{\omega}_\ell, \vec{\omega}_\ell \right)_{A_\ell},$$

and

$$T_2 := (P_{\ell-1} E_{\ell-1} \Pi_{\ell-1} \vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell},$$

where

$$\vec{\omega}_\ell = K_\ell \vec{u}_\ell.$$

Then

$$(E_\ell \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} = T_1 + T_2.$$

Let us first consider  $T_1$ :

$$\begin{aligned} T_1 &= \left( (I_\ell - \tilde{\Pi}_\ell) \vec{\omega}_\ell, \vec{\omega}_\ell \right)_{A_\ell} \\ &\stackrel{(A4)}{\leq} C_4^2 \rho_\ell^{-1} \|A_\ell \vec{\omega}_\ell\|_\ell^2 \\ &= C_4^2 \rho_\ell^{-1} \|A_\ell K_\ell \vec{u}_\ell\|_\ell^2 \\ &\stackrel{(A6)}{\leq} C_4^2 \rho_\ell^{-1} C_6^2 \rho_\ell \left( \bar{K}_\ell A_\ell K_\ell \vec{u}_\ell, A_\ell K_\ell \vec{u}_\ell \right)_\ell \\ &= C_4^2 C_6^2 \left( (I_\ell - K_\ell^* K_\ell) A_\ell^{-1} A_\ell K_\ell \vec{u}_\ell, A_\ell K_\ell \vec{u}_\ell \right)_\ell \\ &= C_4^2 C_6^2 \left( (I_\ell - K_\ell^* K_\ell) K_\ell \vec{u}_\ell, A_\ell K_\ell \vec{u}_\ell \right)_\ell \\ &= C_4^2 C_6^2 \left\{ (K_\ell \vec{u}_\ell, A_\ell K_\ell \vec{u}_\ell)_\ell - (K_\ell^* K_\ell K_\ell \vec{u}_\ell, A_\ell K_\ell \vec{u}_\ell)_\ell \right\} \\ &= C_4^2 C_6^2 \left\{ (K_\ell \vec{u}_\ell, K_\ell \vec{u}_\ell)_{A_\ell} - (K_\ell^* K_\ell K_\ell \vec{u}_\ell, K_\ell \vec{u}_\ell)_{A_\ell} \right\} \\ &= C_4^2 C_6^2 \left\{ (K_\ell \vec{u}_\ell, K_\ell \vec{u}_\ell)_{A_\ell} - (K_\ell^2 \vec{u}_\ell, K_\ell^2 \vec{u}_\ell)_{A_\ell} \right\} \\ &\stackrel{(4.14)}{\leq} C_4^2 C_6^2 \left\{ (\vec{u}_\ell, \vec{u}_\ell)_{A_\ell} - (K_\ell \vec{u}_\ell, K_\ell \vec{u}_\ell)_{A_\ell} \right\}. \end{aligned} \tag{4.16}$$

Now, we turn to the bound for  $T_2$ :

$$\begin{aligned} T_2 &= (E_{\ell-1} \Pi_{\ell-1} \vec{\omega}_\ell, R_{\ell-1} A_\ell \vec{\omega}_\ell)_{\ell-1} \\ &= (E_{\ell-1} \Pi_{\ell-1} \vec{\omega}_\ell, A_{\ell-1} \Pi_{\ell-1} \vec{\omega}_\ell)_{\ell-1} \\ &= (E_{\ell-1} \Pi_{\ell-1} \vec{\omega}_\ell, \Pi_{\ell-1} \vec{\omega}_\ell)_{A_{\ell-1}}. \end{aligned}$$



Set

$$\gamma = \frac{C_4^2 C_6^2}{C_4^2 C_6^2 + 1}.$$

The proof proceeds by induction:

$$\begin{aligned} T_2 &= (E_{\ell-1} \Pi_{\ell-1} \vec{\omega}_\ell, \Pi_{\ell-1} \vec{\omega}_\ell)_{A_{\ell-1}} \\ &\stackrel{\text{ind.hyp.}}{\leq} \gamma (\Pi_{\ell-1} \vec{\omega}_\ell, \Pi_{\ell-1} \vec{\omega}_\ell)_{A_{\ell-1}} \\ &\stackrel{4.15}{=} \gamma \left\{ (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} - ((I_\ell - \tilde{\Pi}_\ell) \vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \right\} \\ &= \gamma (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} - \gamma T_1. \end{aligned} \tag{4.17}$$

To finish up,

$$\begin{aligned} (E_\ell \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} &= T_1 + T_2 \\ &= (1 - \gamma) T_1 + \gamma T_1 + T_2 \\ &\stackrel{(4.17)}{\leq} (1 - \gamma) T_1 + \gamma T_1 + \gamma (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} - \gamma T_1 \\ &= (1 - \gamma) T_1 + \gamma (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \\ &\stackrel{(4.16)}{\leq} (1 - \gamma) C_4^2 C_6^2 \left\{ (\vec{u}_\ell, \vec{u}_\ell)_{A_\ell} - (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \right\} + \gamma (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \\ &= \left( 1 - \frac{C_4^2 C_6^2}{C_4^2 C_6^2 + 1} \right) C_4^2 C_6^2 \left\{ (\vec{u}_\ell, \vec{u}_\ell)_{A_\ell} - (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \right\} + \frac{C_4^2 C_6^2}{C_4^2 C_6^2 + 1} (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \\ &= \frac{C_4^2 C_6^2}{C_4^2 C_6^2 + 1} \left\{ (\vec{u}_\ell, \vec{u}_\ell)_{A_\ell} - (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \right\} + \frac{C_4^2 C_6^2}{C_4^2 C_6^2 + 1} (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \\ &= \gamma \left\{ (\vec{u}_\ell, \vec{u}_\ell)_{A_\ell} - (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \right\} + \gamma (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \\ &= \gamma (\vec{u}_\ell, \vec{u}_\ell)_{A_\ell}. \end{aligned}$$

□

**Lemma 4.29.** *Richardson's smoother satisfies Assumption (A6) with  $S_\ell = \Lambda_\ell^{-1} I_\ell = S_\ell^T$ .*

*Proof.* Recall

$$\rho_\ell \leq \Lambda_\ell \leq C_R \rho_\ell, \quad \exists C_R \geq 1$$

Then

$$\begin{aligned} \bar{K}_\ell &= (I_\ell - K_\ell^* K_\ell) A_\ell^{-1} \\ &= \left\{ I_\ell - (I_\ell - \Lambda_\ell^{-1} A_\ell) (I_\ell - \Lambda_\ell^{-1} A_\ell) \right\} A_\ell^{-1} \\ &= \left( I_\ell - \left\{ I_\ell - 2\Lambda_\ell^{-1} A_\ell + \Lambda_\ell^{-2} A_\ell^2 \right\} \right) A_\ell^{-1} \\ &= 2\Lambda_\ell^{-1} I_\ell - \Lambda_\ell^{-2} A_\ell. \end{aligned}$$

Define

$$J_\ell := \rho_\ell C_R \bar{K}_\ell - I_\ell.$$

If we can show that  $J_\ell$  is SPSPD w.r.t.  $(\cdot, \cdot)_\ell$  then we get (A6) with  $C_6^2 = C_R$ .

$J_\ell$  is clearly symmetric w.r.t.  $(\cdot, \cdot)_\ell$ . Now let  $\{\vec{v}_\ell^{(1)}, \vec{v}_\ell^{(2)}, \dots, \vec{v}_\ell^{(n_\ell)}\}$  be the orthonormal eigenvector of  $A_\ell$  (w.r.t.  $(\cdot, \cdot)_\ell$ ). Then

$$\begin{aligned} J_\ell \vec{v}_\ell^{(k)} &= \rho_\ell C_R \bar{K}_\ell \vec{v}_\ell^{(k)} - \vec{v}_\ell^{(k)} \\ &= \rho_\ell C_R (2\Lambda_\ell^{-1} I_\ell - \Lambda_\ell^{-2} A_\ell) \vec{v}_\ell^{(k)} - \vec{v}_\ell^{(k)} \\ &= 2\rho_\ell C_R \Lambda_\ell^{-1} \vec{v}_\ell^{(k)} - \rho C_R \Lambda_\ell^{-2} \lambda_\ell^{(k)} \vec{v}_\ell^{(k)} - \vec{v}_\ell^{(k)} \\ &= \left( 2\rho_\ell C_R \Lambda_\ell^{-1} - \rho C_R \Lambda_\ell^{-2} \lambda_\ell^{(k)} - 1 \right) \vec{v}_\ell^{(k)}. \end{aligned}$$

Set

$$\eta_\ell^{(k)} = 2\rho_\ell C_R \Lambda_\ell^{-1} - \rho C_R \Lambda_\ell^{-2} \lambda_\ell^{(k)} - 1.$$

We want to show that  $\eta_\ell^{(k)} \geq 0$  for all  $1 \leq k \leq n_\ell$ .

$$\begin{aligned} \eta_\ell^{(k)} &= 2C_R \frac{\rho_\ell}{\Lambda_\ell} - C_R \frac{\rho_\ell \lambda_\ell^{(k)}}{\Lambda_\ell^2} - 1 \\ &\geq 2C_R \frac{\rho_\ell}{\Lambda_\ell} - C_R \frac{\rho_\ell}{\Lambda_\ell} - 1 \quad (-\lambda_\ell^{(k)} \geq -\Lambda_\ell) \\ &= C_R \frac{\rho_\ell}{\Lambda_\ell} - 1 \\ &\geq 1 - 1 = 0. \quad (C_R \rho_\ell \geq \Lambda_\ell) \end{aligned}$$

Thus the eigenvalues of  $J_\ell$ ,  $\eta_\ell^{(k)}$ , are all non-negative and  $J_\ell$  is SPSD. This implies

$$0 \leq (J_\ell \vec{v}_\ell, \vec{v}_\ell)_\ell = \rho_\ell C_R (\bar{K}_\ell \vec{v}_\ell, \vec{v}_\ell)_\ell - (\vec{v}_\ell, \vec{v}_\ell)_\ell,$$

and (A6) follows with  $C_6^2 = C_R$ . □

#### 4.4 Convergence of the V-Cycle Algorithm

**Lemma 4.30.** Suppose that smoothing is done with Richardson's Method, i.e.

$$S_\ell = \Lambda_\ell^{-1},$$

and

$$\rho_\ell \leq \Lambda_\ell \leq C_R \rho_\ell, \quad \exists C_R \geq 1.$$

$$\left( (I_\ell - K_\ell) K_\ell^{2m} \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell} \leq \frac{1}{2m} \left( (I_\ell - K_\ell^{2m}) \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell} \quad (4.18)$$

for any  $m \geq 1$  and  $\ell \geq 1$ .

*Proof.* Suppose  $i, j \in \mathbb{Z}$  with  $0 \leq j \leq i$ . Then

$$\begin{aligned} \left( (I_\ell - K_\ell) K_\ell^i \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell} &= \left( A_\ell (I_\ell - K_\ell) K_\ell^i \vec{v}_\ell, \vec{v}_\ell \right)_\ell \\ &= \Lambda_\ell^{-1} \left( A_\ell^2 K_\ell^i \vec{v}_\ell, \vec{v}_\ell \right)_\ell \\ &= \Lambda_\ell^{-1} \left( K_\ell^i \vec{v}_\ell, A_\ell \vec{v}_\ell \right)_\ell \end{aligned}$$

$$\begin{aligned}
& \stackrel{4.10}{\leq} \Lambda_\ell^{-1} \left( K_\ell^j \vec{v}_\ell, A_\ell \vec{v}_\ell \right)_\ell \\
& = \left( (I_\ell - K_\ell) K_\ell^j \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell}.
\end{aligned}$$

Therefore,

$$\begin{aligned}
& 2m \left( (I_\ell - K_\ell) K_\ell^{2m} \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell} \\
& = \underbrace{\left( (I_\ell - K_\ell) K_\ell^{2m} \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell} + \cdots + \left( (I_\ell - K_\ell) K_\ell^{2m} \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell}}_{2m} \\
& \stackrel{(\text{above})}{\leq} \underbrace{\left( (I_\ell - K_\ell) K_\ell^0 \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell}}_{(j=0)} + \underbrace{\left( (I_\ell - K_\ell) K_\ell^1 \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell}}_{(j=1)} + \cdots + \underbrace{\left( (I_\ell - K_\ell) K_\ell^{2m-1} \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell}}_{(j=2m-1)} \\
& = \left( (I_\ell - K_\ell^{2m}) \vec{v}_\ell, \vec{v}_\ell \right)_{A_\ell}.
\end{aligned}$$

The last equality follows since the sum telescopes.  $\square$

**Theorem 4.31.** Suppose that Assumptions (A1) and (A4) hold. Suppose  $p = 1, m_1 = m_2 = m \geq 1$  and smoothing is done with Richardson's method. Then

$$0 \leq (E_\ell \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} \leq \frac{M}{M+m} (\vec{u}_\ell, \vec{u}_\ell)_{A_\ell},$$

for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$ , where

$$M := \frac{C_4^2 C_R}{2}.$$

*Proof.* The proof is similar to that of Theorem 4.28. We begin with an expression for the error propagation matrix :

$$E_\ell = K_\ell^m (I_\ell - \tilde{\Pi}_\ell) K_\ell^m + K_\ell P_{\ell-1} E_{\ell-1} P_{\ell-1} \Pi_{\ell-1} K_\ell^m,$$

where

$$K_\ell = I_\ell - \Lambda_\ell^{-1} A_\ell = K_\ell^*,$$

and

$$\rho_\ell \leq \Lambda_\ell \leq C_R \rho_\ell, \quad \exists C_R \geq 1.$$

As before, set

$$T_1 := \left( (I_\ell - \tilde{\Pi}_\ell) \vec{\omega}_\ell, \vec{\omega}_\ell \right)_{A_\ell},$$

and

$$T_2 := \left( P_{\ell-1} E_{\ell-1} P_{\ell-1} \vec{\omega}_\ell, \vec{\omega}_\ell \right)_{A_\ell},$$

where

$$\vec{\omega}_\ell = K_\ell^m \vec{u}_\ell.$$

Then

$$(E_\ell \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} = T_1 + T_2.$$

We first estimate  $T_1$ :

$$T_1 = \left( (I_\ell - \tilde{\Pi}_\ell) \vec{\omega}_\ell, \vec{\omega}_\ell \right)_{A_\ell}$$

$$\begin{aligned}
& \stackrel{A4}{\leq} C_4^2 \rho_\ell^{-1} \|A_\ell \vec{\omega}_\ell\|_\ell^2 \\
& = C_4^2 \rho_\ell^{-1} \|A_\ell K_\ell^m \vec{u}_\ell\|_\ell \\
& = C_4^2 \rho_\ell^{-1} (A_\ell K_\ell^m \vec{u}_\ell, A_\ell K_\ell^m \vec{u}_\ell)_\ell \\
& = C_4^2 \rho_\ell^{-1} (A_\ell^2 K_\ell^m \vec{u}_\ell, K_\ell^m \vec{u}_\ell)_\ell \\
& = C_4^2 \rho_\ell^{-1} (A_\ell K_\ell^m \vec{u}_\ell, K_\ell^m \vec{u}_\ell)_{A_\ell} \\
& = C_4^2 \rho_\ell^{-1} \Lambda_\ell ((I_\ell - K_\ell) K_\ell^m \vec{u}_\ell, K_\ell^m \vec{u}_\ell)_{A_\ell} \\
& = C_4^2 \rho_\ell^{-1} \Lambda_\ell ((I_\ell - K_\ell) K_\ell^{2m} \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} \\
& \stackrel{(4.18)}{\leq} \frac{C_4^2 \rho_\ell^{-1} \Lambda_\ell}{2m} ((I_\ell - K_\ell^{2m}) \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} \\
& \leq \frac{C_4^2 C_R}{2m} ((I_\ell - K_\ell^{2m}) \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} \\
& \leq \frac{M}{m} \{(\vec{u}_\ell, \vec{u}_\ell)_{A_\ell} - (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell}\}.
\end{aligned} \tag{4.19}$$

Set

$$\gamma := \frac{M}{M+m} \quad (< 1).$$

Exactly as before, the induction step yields

$$T_2 \leq \gamma (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} - \gamma T_1. \tag{4.20}$$

Therefore,

$$\begin{aligned}
(E_\ell \vec{u}_\ell, \vec{u}_\ell)_{A_\ell} & = T_1 + T_2 \\
& = (1 - \gamma) T_1 + \gamma T_1 + T_2 \\
& \stackrel{4.20}{\leq} (1 - \gamma) T_1 + \gamma T_1 + \gamma (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} - \gamma T_1 \\
& \stackrel{4.19}{\leq} (1 - \gamma) \frac{M}{m} \{(\vec{u}_\ell, \vec{u}_\ell)_{A_\ell} - (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell}\} + \gamma (\vec{\omega}_\ell, \vec{\omega}_\ell)_{A_\ell} \\
& = \gamma (\vec{u}_\ell, \vec{u}_\ell)_{A_\ell}.
\end{aligned}$$

□

**Theorem 4.32.** Suppose that hypotheses of either Theorem 4.28 or 4.31 hold, as appropriate. Suppose that  $\vec{u}_\ell^E, \vec{g}_\ell \in \mathbb{R}^{n_\ell}$  satisfy

$$A_\ell \vec{u}_\ell^E = \vec{g}_\ell.$$

Then, given any  $\vec{u}_\ell^{(0)} \in \mathbb{R}^{n_\ell}$ ,

$$\begin{aligned} \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(3)} \right\|_{A_\ell} &= \left\| \vec{u}_\ell^E - \text{MG}(\vec{g}_\ell, \ell, \vec{u}_\ell^{(0)}) \right\|_{A_\ell} \\ &\leq \frac{M}{M+m} \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell}, \end{aligned}$$

where

$$M = C_4^2 C_6^2, M = 1, \quad (4.28)$$

or

$$M = \frac{C_4^2 C_R}{2}, M \geq 1, \quad (4.31).$$

*Proof.* We need only show that

$$\left\| E_\ell \vec{v}_\ell \right\|_{A_\ell} \leq \frac{M}{M+m} \left\| \vec{v}_\ell \right\|_{A_\ell},$$

is true for any  $\vec{v}_\ell \in \mathbb{R}^{n_\ell}$ . Since  $E_\ell$  is SPSPD w.r.t.  $(\cdot, \cdot)_{A_\ell}$ , for  $\ell \geq 1$ , there is an orthonormal basis of eigenvectors of  $E_\ell$ ,  $\{\vec{\omega}_\ell^{(1)}, \dots, \vec{\omega}_\ell^{(n_\ell)}\}$ , such that

$$E_\ell \vec{\omega}_\ell^{(j)},$$

$$(\vec{\omega}_\ell^{(i)}, \vec{\omega}_\ell^{(j)})_{A_\ell} = \delta_{ij},$$

and

$$0 \leq \epsilon_\ell^{(1)} \leq \epsilon_\ell^{(2)} \leq \dots \leq \epsilon_\ell^{(n_\ell)}.$$

Suppose

$$\vec{v}_\ell = \sum_{k=1}^{n_\ell} C_k \vec{\omega}_\ell^{(k)}.$$

Then

$$(E_\ell \vec{v}_\ell, \vec{v}_\ell)_{A_\ell} = \sum_{k=1}^{n_\ell} C_k^2 \epsilon_\ell^{(k)},$$

and

$$(\vec{v}_\ell, \vec{v}_\ell)_{A_\ell} = \sum_{k=1}^{n_\ell} C_k^2.$$

Theorem 4.31 implies that

$$\sum_{k=1}^{n_\ell} C_k^2 \epsilon_\ell^{(k)} \leq \frac{M}{M+m} \sum_{k=1}^{n_\ell} C_k^2,$$

for any  $C_1, \dots, C_{n_\ell} \in \mathbb{R}$ . This implies that

$$0 \leq \epsilon_\ell^{(k)} \leq \frac{M}{M+m}, \quad 1 \leq k \leq n_\ell.$$

Therefore

$$\begin{aligned}
 (E_\ell \vec{v}_\ell)_{A_\ell}^2 &= (E_\ell \vec{v}_\ell, E_\ell \vec{v}_\ell)_{A_\ell} \\
 &= \sum_{k=1}^{n_\ell} C_k^2 \left( \epsilon_\ell^{(k)} \right)^2 \\
 &= \left( \frac{M}{M+m} \right)^2 \sum_{k=1}^{n_\ell} C_k^2 \\
 &= \left( \frac{M}{M+m} \right)^2 \|\vec{v}_\ell\|_{A_\ell}^2.
 \end{aligned}$$

□

## 5 Conforming Finite Element Method

Consider the following model problem: Let  $\Omega \subset \mathbb{R}^d$ ,  $d = 1$  or  $d = 2$ , be an open polygonal domain. (Often, we will also assume that  $\Omega$  is convex.) Given  $F \in L^2(\Omega)$ , find  $u \in H_0^1(\Omega)$  such that

$$a(u, v) = (f, v)_{L^2(\Omega)} \quad (5.1)$$

where

$$a(u, v) := (\nabla u, \nabla v)_{L^2(\Omega)}.$$

Let  $\mathcal{T}_0$  be a triangulation of  $\Omega$ . We define  $\mathcal{T}_1$  to be the triangulation of  $\Omega$  that results from bisecting ( $d=1$ ) or quadrisecting ( $d=2$ ) the triangulation  $\mathcal{T}_0$ . For  $d = 2$ , we connect the edge midpoints.



Figure 23: Bisecting the triangulation  $\mathcal{T}_0$  in 1D.

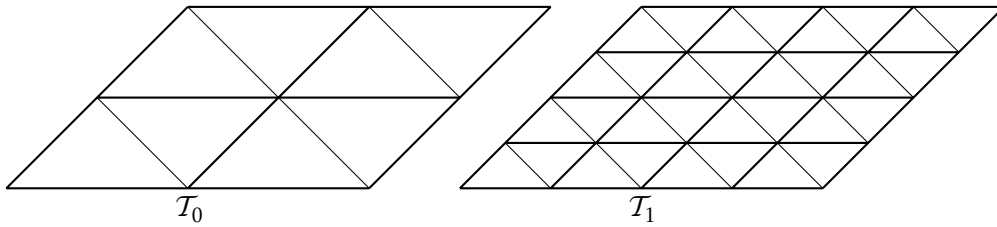


Figure 24: Quadrisecting the triangulation  $\mathcal{T}_0$  in 2D.

Observe that the daughter triangles are similar to the mother. Continuing, we can recursively define a family of the triangulations, indexed by  $\ell$ ,

$$\mathcal{T}_0, \mathcal{T}_1, \mathcal{T}_2, \dots, \mathcal{T}_\ell, \dots, \mathcal{T}_L.$$

**Definition 5.1.** For  $0 \leq \ell \leq L$ , define

$$h_\ell := \max_{K \in \mathcal{T}_\ell} \text{diam}(K).$$

Subordinate to  $\mathcal{T}_\ell$ , define

$$V_\ell := \left\{ v_\ell \in C^0(\overline{\Omega}) \mid v_\ell|_K \in \mathbb{P}_1(K), \forall K \in \mathcal{T}_\ell, v_\ell|_{\partial\Omega} \equiv 0 \right\}.$$

Define

$$n_\ell := \dim V_\ell$$

**Lemma 5.2.** For  $0 \leq \ell \leq L$  and  $v_\ell, \mathcal{T}_\ell$  defined as above, we have

$$V_1 \subset V_2 \subset \cdots \subset V_\ell \subset V_L \subset H_0^1(\Omega),$$

and

$$0 \leq n_0 < n_1 < n_2 < \cdots < n_\ell < \cdots < n_L < \infty.$$

Furthermore,  $n_\ell$  is precisely the number of interior vertices of the triangulation  $\mathcal{T}_\ell$ . Finally,

$$h_{\ell-1} = 2h_\ell, \quad 1 \leq \ell \leq L.$$

*Proof.* Exercise. □

**Remark 5.3.** These ideas can be extended to  $d = 3$  as well, but we neglect this case for simplicity.

**Example.** The Recursively Mesh in 1D and 2D.

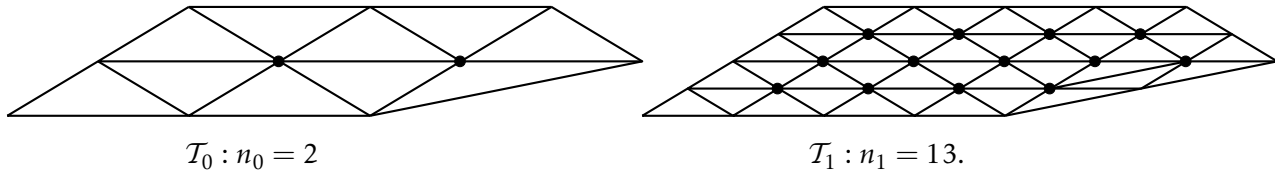


Figure 25: Recursively Mesh in 2D.

**Remark 5.4.** We will assume that  $n_0 > 0$ .

**Definition 5.5.** Let  $0 \leq \ell \leq L$  and  $V_\ell, \mathcal{T}_\ell$  be defined as above. By  $B_\ell$  we denote the Lagrange Nodal basis for  $V_\ell$ , i.e.

$$B_\ell := \{\psi_{\ell,i}\}_{i=1}^{n_\ell},$$

where  $\psi_{\ell,i} \in V_\ell$  is the unique function with the property that

$$\psi_{\ell,i}(\vec{N}_{\ell,j}) = \delta_{ij},$$

and  $\{\vec{N}_{\ell,j}\}_{j=1}^{n_\ell}$  is the set of interior vertices of the triangulation  $\mathcal{T}_\ell$ .

**Example.** Example of  $N_{\ell,j}$  and  $\psi_{\ell,i}$  in 2D.

**Lemma 5.6.**  $B_\ell$  is a bona fide basis for  $V_\ell$ ,  $0 \leq \ell \leq L$ . And for every  $1 \leq \ell \leq L$ , there exist unique numbers

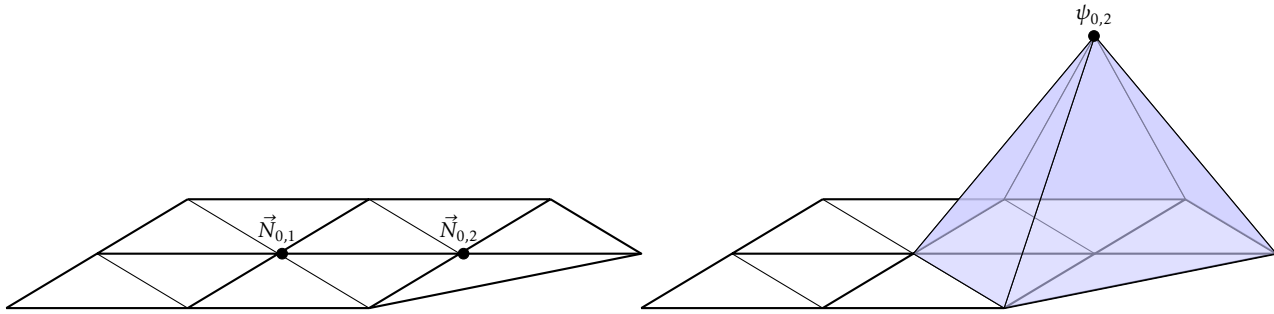
$$p_{\ell-1,j,i} \in \mathbb{R}, \quad 1 \leq j \leq n_\ell, \quad 1 \leq i \leq n_{\ell-1}$$

with the property that

$$\psi_{\ell-1,i} = \sum_{j=1}^{n_\ell} p_{\ell-1,j,i} \psi_{\ell,j}, \tag{5.2}$$

for each  $1 \leq i \leq n_{\ell-1}$ .



Figure 26: Example of  $N_{\ell,j}$  and  $\psi_{\ell,i}$  in 2D.

*Proof.* Since  $V_{\ell-1}$  is a linear subspace of  $V_\ell$  and  $B_\ell$  is a basis for the latter, for every  $\psi_{\ell-1,i} \in B_{\ell-1} \subset V_{\ell-1} \subset V_\ell$ , there exists unique coefficients

$$p_{\ell-1,j,i} \in \mathbb{R}, \quad 1 \leq j \leq n_\ell,$$

such that

$$\psi_{\ell-1,i} = \sum_{j=1}^{n_\ell} p_{\ell-1,j,i} \psi_{\ell,j}.$$

Recall that basis representation are unique. □

**Definition 5.7.** For  $0 \leq \ell \leq L$ , define the prolongation (or injection) matrix

$$P_{\ell-1} \in \mathbb{R}^{n_\ell \times n_{\ell-1}},$$

via

$$[P_{\ell-1}]_{i,j} := p_{\ell-1,i,j}$$

**Lemma 5.8.** Let  $1 \leq \ell \leq L$  and  $v_{\ell-1} \in V_{\ell-1}$  be arbitrary. Suppose  $\vec{v}_{\ell-1}$  is the coordinate vector of  $v_{\ell-1}$  in the Lagrange Nodal basis  $B_{\ell-1}$ . Suppose  $\vec{v}_\ell$  is the coordinate vector of  $v_{\ell-1}$  in  $B_\ell$ . (Recall  $v_{\ell-1} \in v_\ell$ .) Then

$$v_\ell = P_{\ell-1} v_{\ell-1}.$$

*Proof.* We write

$$[\vec{v}_{\ell-1}]_i = v_{\ell-1,i}, \quad 1 \leq i \leq n_{\ell-1},$$

and

$$[\vec{v}_\ell]_i = v_{\ell,i}, \quad 1 \leq i \leq n_\ell.$$

Thus

$$\begin{aligned} v_{\ell-1} &= \sum_{i=1}^{n_{\ell-1}} v_{\ell-1,i} \psi_{\ell-1,i} \\ &\stackrel{5.2}{=} \sum_{i=1}^{n_{\ell-1}} v_{\ell-1,i} \sum_{j=1}^{n_\ell} p_{\ell-1,j,i} \psi_{\ell,j} \\ &= \sum_{j=1}^{n_\ell} \left\{ \sum_{i=1}^{n_{\ell-1}} p_{\ell-1,j,i} v_{\ell-1,i} \right\} \psi_{\ell,j} \end{aligned}$$

$$= \sum_{j=1}^{n_\ell} [P_{\ell-1} \vec{v}_{\ell-1}]_j \psi_{\ell,j}.$$

But

$$v_{\ell-1} = \sum_{j=1}^{n_\ell} [\vec{v}_\ell]_j \psi_{\ell,j}.$$

Since basis representations are unique,

$$\vec{v}_\ell = P_{\ell-1} \vec{v}_{\ell-1}.$$

□

**Definition 5.9.** For  $1 \leq \ell \leq L$ , define the Restriction matrix as

$$R_{\ell-1} := P_{\ell-1}^T \in \mathbb{R}^{n_{\ell-1} \times n_\ell}.$$

**Lemma 5.10.** For  $1 \leq \ell \leq L$  and suppose  $R_{\ell-1}$  and  $P_{\ell-1}$  are defined as above. Then

$$\text{rank}(P_{\ell-1}) = \text{rank}(R_{\ell-1}) = n_{\ell-1}.$$

*Proof.* Suppose

$$P_{\ell-1} \vec{v}_{\ell-1} = \vec{0} \in \mathbb{R}^{n_\ell}.$$

This represents a linear combination of the  $n_{\ell-1}$  columns of  $P_{\ell-1}$ . Using the notation from the last lemma and its proof, we have

$$\vec{v}_\ell = \vec{0} \in \mathbb{R}^{n_\ell},$$

where  $\vec{v}_{\ell-1}$  and  $\vec{v}_\ell$  are coordinate vectors of some function  $\vec{v}_{\ell-1} \in V_{\ell-1}$  in the basis  $B_{\ell-1}$  and  $B_\ell$ , respectively.

The only way that  $\vec{v}_\ell = \vec{0}$  is if  $v_{\ell-1} \equiv 0$  in  $V_{\ell-1}$ . But then  $\vec{v}_{\ell-1} = \vec{0}$ . Thus, the columns of  $P_{\ell-1}$  are linearly independent and

$$\text{rank}(P_{\ell-1}) = n_{\ell-1}.$$

□

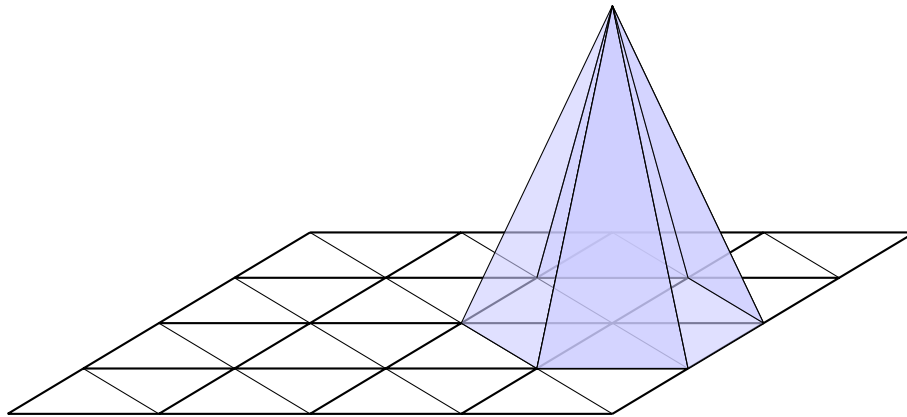


Figure 27: Mesh on the fine and coarse grid of the multigrid method in 2D.

## 5.1 The Stiffness Matrices

**Definition 5.11.** For  $\ell \geq 0$  and  $V_\ell, \mathcal{T}_\ell$  defined as before, we define the Stiffness Matrices  $A_\ell \in \mathbb{R}^{n_\ell \times n_\ell}$  via

$$[A_\ell]_{ij} := a(\psi_{\ell,j}, \psi_{\ell,i})$$

for all  $1 \leq i, j \leq n_\ell$ .

**Lemma 5.12.** The Stiffness Matrices  $A_\ell \in \mathbb{R}^{n_\ell \times n_\ell}$  are all SPD, and, moreover, for  $1 \leq \ell \leq L$ ,

$$A_{\ell-1} = R_{\ell-1} A_\ell P_{\ell-1}. \quad (5.3)$$

for all  $1 \leq i, j \leq n_\ell$ .

*Proof.* 1. Symmetric-ness:

$$\begin{aligned} [A_\ell]_{ij} &= a(\psi_{\ell,j}, \psi_{\ell,i}) \\ &= a(\psi_{\ell,i}, \psi_{\ell,j}) \\ &= [A_\ell]_{ji}. \end{aligned}$$

2. PD-ness: Let  $\vec{v}_\ell \in \mathbb{R}^{n_\ell}$  be arbitrary and suppose  $v_\ell \in V_\ell$  is the unique function with coordinations  $\vec{v}_\ell$ . Then, using the Poincare inequality, there is a  $C > 0$  such that

$$\begin{aligned} 0 \leq C \|v_\ell\|_{L^2(\Omega)}^2 &\leq a(v_\ell, v_\ell) \\ &= a\left(\sum_{j=1}^{n_\ell} v_{\ell,j} \psi_{\ell,j}, \sum_{i=1}^{n_\ell} v_{\ell,i} \psi_{\ell,i}\right) \\ &= \sum_{i,j=1}^{n_\ell} v_{\ell,i} a(\psi_{\ell,j}, \psi_{\ell,i}) v_{\ell,j} \\ &= \vec{v}_\ell^T A_\ell \vec{v}_\ell. \end{aligned}$$

But  $v_\ell \equiv 0$  iff  $\vec{v}_\ell = \vec{0}$ . Hence  $A_\ell$  is SPD.

3. Galerkin Condition: By definition

$$\begin{aligned} [A_{\ell-1}]_{ij} &= a(\psi_{\ell-1,j}, \psi_{\ell-1,i}) \\ &\stackrel{5.2}{=} a\left(\sum_{J=1}^{n_\ell} p_{\ell-1,J,j} \psi_{\ell,J}, \sum_{I=1}^{n_\ell} p_{\ell-1,I,i} \psi_{\ell,I}\right) \\ &= \sum_{I,J=1}^{n_\ell} p_{\ell-1,I,i} a(\psi_{\ell,J}, \psi_{\ell,I}) p_{\ell-1,J,j} \\ &= \sum_{I,J=1}^{n_\ell} [R_{\ell-1}]_{iI} [A_\ell]_{IJ} [P_{\ell-1}]_{J,j}. \end{aligned}$$

Thus

$$A_{\ell-1} = R_{\ell-1} A_\ell P_{\ell-1}.$$

□

**Remark 5.13.** Observe that this last result guarantees that  $P_{\ell-1}$  and  $R_{\ell-1}$  are of full rank. Otherwise  $A_{\ell-1}$  could not be positive definite.

Next we estimate the size of the condition number of  $A_\ell$ . To do this we need a couple of results whose proof can be found in Finite Element books.

**Lemma 5.14.** Suppose  $d = 1, 2, \text{ or } 3$ . let  $v_\ell \in V_\ell$  be arbitrary and  $\vec{v}_\ell \in \mathbb{R}^{n_\ell}$  be its unique coordinate vector in the basis  $B_\ell$ . Then there are constants  $C_2 \geq C_1 > 0$ , independent of  $\ell \geq 0$  and  $v_\ell$ , such that

$$C_1 h_\ell^d \|\vec{v}_\ell\|_2^2 \leq \|v_\ell\|_{L^2(\Omega)}^2 \leq C_2 h_\ell^d \|\vec{v}_\ell\|_2^2. \quad (5.4)$$

**Lemma 5.15.** There is a constant  $C > 0$  independent of  $\ell \geq 0$  such that

$$a(v_\ell, v_\ell) \leq C_3 h_\ell^{-2} \|v_\ell\|_{L^2}^2. \quad (5.5)$$

And, as a consequence of (5.4),

$$a(v_\ell, v_\ell) \leq C_2 C_3 h_\ell^{d-2} \|\vec{v}_\ell\|_2^2. \quad (5.6)$$

**Remark 5.16.** These results require some conditions on the underlying family of conforming meshes. such as global quasi-uniformity and shape regularity, which hold thanks to our construction of the family  $\mathcal{T}_\ell$ ,  $\ell \geq 0$ .

Estimate (5.4) is a norm equivalence, (5.5) is called inverse estimate.

**Lemma 5.17.** (Asymptotic Sharpness of (5.5)) Let  $V_\ell, \mathcal{T}_\ell$  be defined as usual. There exist a constant  $C_4 > 0$ , independent of  $\ell$ , and functions  $v'_\ell \in V_\ell$ , such that, for every  $0 \leq \ell \leq L$

$$C_4 h_\ell^{-2} \|v'_\ell\|_{L^2}^2 \leq a(v'_\ell, v'_\ell). \quad (5.7)$$

Moreover, the Poincare inequality is asymptotic sharp in a similar sense: There exists a constant  $C_5 > 0$ , independent of  $\ell$ , and there exist functions  $v''_\ell \in V_\ell$ , such that, for every  $0 \leq \ell \leq L$

$$a(v''_\ell, v''_\ell) \leq C_5 \|\vec{v}_\ell\|_{L^2}^2. \quad (5.8)$$

**Remark 5.18.** Proofs of these results can be found in Brass's Finite Element book [1].

**Theorem 5.19.** Let  $d = 1, 2, \text{ or } 3$ . There exist constants  $C_7 \geq C_6 > 0$ , independent of  $\ell \geq 0$ , such that

$$C_6 h_\ell^{-2} \leq \kappa_2(A_\ell) = \frac{\lambda_\ell^{(n_\ell)}}{\lambda_\ell^{(1)}} \leq C_7 h_\ell^{-2}. \quad (5.9)$$

In particular, there are constant  $C_7^{(i)}, C_6^{(i)} > 0$  for  $i = 1, n_\ell$ , such that

$$\begin{aligned} C_6^{(n_\ell)} h_\ell^{d-2} &\leq \lambda_\ell^{(n_\ell)} \leq C_7^{(n_\ell)} h_\ell^{d-2}. \\ C_7^{(1)} h_\ell^d &\leq \lambda_\ell^{(1)} \leq C_6^{(1)} h_\ell^d. \end{aligned}$$

*Proof.* First we recall some basis facts for the Rayleigh Quotation.

$$\lambda_\ell^{(1)} = R(\vec{v}_\ell^{(1)}) = \min_{\vec{v}_\ell} R(\vec{v}_\ell) > 0,$$

and

$$\lambda_\ell^{(n_\ell)} = R(\vec{v}_\ell^{(n_\ell)}) = \max_{\vec{v}_\ell} R(\vec{v}_\ell),$$

where

$$A_\ell \vec{v}_\ell^{(k)} = \lambda_\ell^{(k)} \vec{v}_\ell^{(k)}, \quad 1 \leq k \leq n_\ell.$$

UPPER BOUND IN (5.9): As usual let  $v_\ell \in V_\ell, v_\ell \in \mathbb{R}^{n_\ell}$ . Then, for arbitrary  $v_\ell \in V_\ell$ ,

$$\begin{aligned} R(\vec{v}_\ell) : &= \frac{\vec{v}_\ell^T A_\ell \vec{v}_\ell}{\vec{v}_\ell^T \vec{v}_\ell} = \frac{a(v_\ell, v_\ell)}{\|\vec{v}_\ell\|_2^2} \\ &\stackrel{5.6}{\leq} \frac{C_2 C_3 h_\ell^{d-2} \|\vec{v}_\ell\|_2^2}{\|\vec{v}_\ell\|_2^2} \\ &=: C_7^{(n_\ell)} h_\ell^{d-2}. \end{aligned}$$

This implies

$$\lambda_\ell^{(n_\ell)} \leq C_7^{(n_\ell)} h_\ell^{d-2}.$$

Similarly,

$$\begin{aligned} R(\vec{v}_\ell) : &= \frac{a(v_\ell, v_\ell)}{\|\vec{v}_\ell\|_2^2} \\ &\stackrel{\text{Poincare}}{\geq} \frac{C_p \|\vec{v}_\ell\|_{L^2(\Omega)}^2}{\|\vec{v}_\ell\|_2^2} \\ &\stackrel{5.4}{\geq} \frac{C_p C_1 h_\ell^d \|\vec{v}_\ell\|_2^2}{\|\vec{v}_\ell\|_2^2} \\ &=: C_7^{(1)} h_\ell^d. \end{aligned}$$

Therefore,

$$\lambda_\ell^{(1)} \geq C_7^{(1)} h_\ell^d.$$

It follows that

$$\begin{aligned} \lambda_\ell^{(n_\ell)} &= R(\vec{v}_\ell^{(n_\ell)}) \geq R(\vec{v}_\ell') \\ &= \frac{a(v_\ell', v_\ell')}{\|\vec{v}_\ell'\|_2^2} \\ &\stackrel{5.7}{\geq} \frac{C_4 h_\ell^{-2} \|v_\ell'\|_{L^2}^2}{\|\vec{v}_\ell'\|_2^2} \\ &\stackrel{5.4}{\geq} \frac{C_1 C_4 h_\ell^{d-2} \|v_\ell'\|_2^2}{\|\vec{v}_\ell'\|_2^2} \\ &=: C_6^{(n_\ell)} h_\ell^{d-2}. \end{aligned}$$

Likewise

$$\lambda_\ell^{(1)} = R(\vec{v}_\ell^{(1)}) \geq R(\vec{v}_\ell'')$$

$$\begin{aligned}
&= \frac{a(v_\ell'', v_\ell'')}{\|\vec{v}_\ell''\|_2^2} \\
&\stackrel{5.8}{\leq} \frac{C_5 \|v_\ell''\|_{L^2}^2}{\|\vec{v}_\ell''\|_2^2} \\
&\stackrel{5.4}{\leq} \frac{C_2 C_5 h_\ell^d \|v_\ell''\|_2^2}{\|\vec{v}_\ell''\|_2^2} \\
&=: C_6^{(1)} h_\ell^d.
\end{aligned}$$

It then follows that

$$\kappa_2(A_\ell) \geq \frac{C_6^{(n_\ell)} h_\ell^{d-2}}{C_6^{(1)} h_\ell^d} = C_6 h_\ell^{-2}.$$

□

## 5.2 Strong Approximation Property

Now, we want to show that the strong approximate property, Assumption (A3), holds for the correct situation: There is some  $C_{A3}$  such that

$$\|(I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell\|_\ell^2 \leq C_{A3}^2 \rho_\ell^{-1} \|(I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell\|_{A_\ell}^2. \quad (5.10)$$

We need a bit of PDE and FE theory first.

Let  $f \in H^{-1}(\Omega) = (H_0^1(\Omega))'$  be given. Find  $u \in H_0^1(\Omega)$  such that

$$a(u, v) = (f, v) =: \langle f, v \rangle, \quad \forall v \in H_0^1(\Omega).$$

It is well-known that a unique solution  $u \in H_0^1(\Omega)$  exists.

A conforming FE approximation of the problem may be written as follows: Find  $\vec{u}_\ell \in V_\ell$  such that

$$a(\vec{u}_\ell, \vec{v}_\ell) = f(v_\ell), \quad \forall \vec{v}_\ell \in V_\ell, \quad (5.11)$$

where  $V_\ell$  is the family of nested conforming FE subspaces of  $H_0^1(\Omega)$  that we constructed earlier.

Observe that every  $f \in L^2(\Omega)$  gives [figure out what's at here](#) to an  $L_f \in H^{-1}$  in a nature way:

$$L_f(v) := (f, v)_{L^2(\Omega)}, \quad \forall v \in H_0^1(\Omega).$$

In this case we write

$$\langle f, v \rangle = (f, v)_{L^2(\Omega)},$$

and

$$L_f = f.$$

**Definition 5.20.** We say that the model problem satisfies the standard regularity condition iff when  $f \in L^2(\Omega) \cap H^{-1}(\Omega)$ , then  $u \in H_0^1(\Omega) \cap H^2(\Omega)$  and

$$|u|_{H^2(\Omega)} \leq C \|f\|_{L^2(\Omega)}, \quad (5.12)$$

for some universal (regularity) constant  $C > 0$ .

**Theorem 5.21.** If  $\Omega$  is convex and polyhedral, then the standard regularity condition holds.

**Theorem 5.22.** Let  $\Omega \subset \mathbb{R}^d$ ,  $d = 1, 2$ , or  $3$ , be an open polyhedral domain and suppose  $\mathcal{T}_h$  is a shape regular family of triangulation of  $\Omega$  parameterized by

$$h := \max_{K \in \mathcal{T}_h} \text{diam}(K).$$

Then, if  $\Omega$  is convex (so that the standard regularity condition holds) and  $f \in H^{-1}(\Omega)$ , there is a constant  $C > 0$ , such that

$$\|u - u_h\|_{L^2(\Omega)} \leq ch |u - u_h|_{H^1(\Omega)}, \quad (5.13)$$

where  $u_h \in V_h = \{v \in C^0(\Omega) \mid v|_K \in \mathbb{P}_1(K), K \in \mathcal{T}_h, v|_{\partial\Omega=0}\}$  is the unique solution to

$$a(u_h, v_h) = f(v_h), \quad v_h \in V_h.$$

If, in addition, it is known that  $f \in L^2(\Omega)$ , so that  $u \in H_0^1(\Omega) \cap H^2(\Omega)$ , then

$$|u - u_h|_{H^1(\Omega)} \leq ch |u|_{H^2(\Omega)}, \quad (5.14)$$

for some  $C > 0$ . And, all together,

$$\|u - u_h\|_{L^2(\Omega)} \leq ch^2 |u|_{H^2(\Omega)}. \quad (5.15)$$

**Definition 5.23.** Let  $\mathcal{T}_h$  and  $V_h$  be as in the last theorem. Let  $u \in H_0^1(\Omega)$  be arbitrary. Define the Ritz projection as follows:  $\mathcal{R}_h u \in V_h$  is the unique solution to

$$a(\mathcal{R}_h u, v_h) = a(u, v_h), \quad \forall v_h \in V_h.$$

in the case where  $V_h = V_\ell$ ,  $\mathcal{T}_h = \mathcal{T}_\ell$ , we write  $\mathcal{R}_h =: \mathcal{R}_\ell$ :

$$a(\mathcal{R}_\ell u, v_\ell) = a(u, v_\ell), \quad \forall v_\ell \in V_h.$$

**Remark 5.24.**

$$a(u, v) = (-\Delta u, v)_{L^2}, \quad \forall v \in H_0^1(\Omega).$$

In any case, it should be clear that  $\mathcal{R}_h u \in V_h$  is just the FE approximation of  $u$ .

**Lemma 5.25.** Let  $\mathcal{T}_\ell$  and  $V_\ell$  as usual, and suppose  $\vec{u}_\ell \in V_\ell$  is given. Then,  $\Omega$  is convex, as follows:  $\mathcal{R}_h u \in V_h$  is the unique solution to

$$\|\vec{u}_\ell - \mathcal{R}_{\ell-1} \vec{u}_\ell\|_{L^2} \leq Ch_\ell |\vec{u}_\ell - \mathcal{R}_{\ell-1} \vec{u}_\ell|_{H^1}, \quad (5.16)$$

for some constant  $C > 0$  that is independent of  $\ell \geq 0$ .

*Proof.* Observe that  $\vec{u}_\ell \in V_\ell \subset H_0^1(\Omega)$ . But  $\vec{u}_\ell \notin H^2(\Omega)$ .  $\vec{u}_\ell$  plays the role of the exact PDE solution in Theorem.(5.22), but it is not  $H^2$ -regular. But this does not matter. We may still apply (5.13), since  $\Omega$  is convex, to conclude

$$\|\vec{u}_\ell - \mathcal{R}_{\ell-1} \vec{u}_\ell\|_{L^2} \leq Ch_{\ell-1} |\vec{u}_\ell - \mathcal{R}_{\ell-1} \vec{u}_\ell|_{H^1}$$

for some  $C > 0$  that is independent of  $\ell$ . Now, note that

$$h_{\ell-1} = 2h_\ell,$$

and the result follows.  $\square$

**Theorem 5.26.** Let  $\mathcal{T}_\ell$  and  $V_\ell$  as usual, and suppose that  $\Omega$  is convex polyhedral. Then the strong approximate property is satisfied. In particular, there is some  $C_{A3} > 0$ , independent of  $\ell$ , such that

$$\|\vec{u}_\ell - \tilde{\Pi}_\ell \vec{u}_\ell\|_\ell^2 \leq C_{A3}^2 \rho_\ell^{-1} \|\vec{u}_\ell - \tilde{\Pi}_\ell \vec{u}_\ell\|_{A_\ell}^2 \quad (5.17)$$

for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$ .

*Proof.* Let  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$  be arbitrary. Suppose  $\vec{u}_\ell \in V_\ell$  is the unique element whose coordinate vector is  $\vec{u}_\ell$  with basis  $\mathcal{B}_\ell$ .

$$\vec{u}_\ell \in V_\ell \xleftrightarrow{\mathcal{B}_\ell} \vec{u}_\ell \mathbb{R}^{n_\ell}.$$

Referring to (5.16)

$$|\vec{u}_\ell - \mathcal{R}_{\ell-1} \vec{u}_\ell|_{H^1}^2 = a(\vec{u}_\ell - \mathcal{R}_{\ell-1} \vec{u}_\ell, \vec{u}_\ell - \mathcal{R}_{\ell-1} \vec{u}_\ell),$$

let  $\vec{w}_\ell \in \mathbb{R}^{n_\ell}$  be the unique coordinate vector of

$$\begin{array}{c} \vec{u}_\ell - \mathcal{R}_{\ell-1} \vec{u}_\ell \in V_\ell, \\ \in V_\ell \quad \in V_{\ell-1} \subset V_\ell \end{array}$$

w.r.t. the Lagrange nodal basis  $\mathcal{B}_\ell$ . We want to show that

$$\vec{w}_\ell = \vec{u}_\ell - \tilde{\Pi}_\ell \vec{u}_\ell = \vec{u}_\ell - P_{\ell-1} A_{\ell-1}^{-1} R_{\ell-1} A_\ell \vec{u}_\ell.$$

We begin with the definition of  $\mathcal{R}_{\ell-1}$ :

$$a(\mathcal{R}_{\ell-1} \vec{u}_\ell, v_{\ell-1}) = a(\vec{u}_\ell, v_{\ell-1}), \quad \forall v_{\ell-1} \in V_{\ell-1}.$$

Set  $\vec{u}'_{\ell-1} := \mathcal{R}_{\ell-1} \vec{u}_\ell \in V_{\ell-1}$

$$\begin{array}{c} \vec{u}'_{\ell-1} \in \mathbb{R}^{n_{\ell-1}} \xleftrightarrow{\mathcal{B}_{\ell-1}} \vec{u}'_{\ell-1} \in V_{\ell-1}. \\ \vec{v}_{\ell-1} \in \mathbb{R}^{n_{\ell-1}} \leftrightarrow \vec{v}_{\ell-1} \in V_{\ell-1}. \end{array}$$

Then

$$\begin{aligned} (\vec{u}'_{\ell-1}, \vec{v}_{\ell-1})_{A_{\ell-1}} &= (\vec{u}_\ell, P_{\ell-1} \vec{v}_{\ell-1})_{A_\ell} \\ &= (A_\ell \vec{u}_\ell, P_{\ell-1} \vec{v}_{\ell-1})_\ell \\ &= (R_{\ell-1} A_\ell \vec{u}_\ell, \vec{v}_{\ell-1})_{\ell-1}. \end{aligned}$$

But

$$(\vec{u}'_{\ell-1}, \vec{v}_{\ell-1})_{A_{\ell-1}} = (A_{\ell-1} \vec{u}'_{\ell-1}, \vec{v}_{\ell-1})_{\ell-1}.$$

So, it follows that

$$A_{\ell-1} \vec{u}'_{\ell-1} = R_{\ell-1} A_\ell \vec{u}_\ell.$$

Thus

$$\vec{u}'_{\ell-1} = A_{\ell-1}^{-1} R_{\ell-1} A_\ell \vec{u}_\ell = \tilde{\Pi}_{\ell-1} \vec{u}_\ell.$$

Therefore

$$\begin{aligned} \vec{w}_\ell &= \vec{u}_\ell - P_{\ell-1} \vec{u}'_{\ell-1} \\ &= \vec{u}_\ell - \tilde{\Pi}_\ell \vec{u}_\ell. \end{aligned}$$



It follows that

$$\begin{aligned}
 \|\vec{u}_\ell - R_{\ell-1}\vec{u}_\ell\|_{H^1}^2 &= (\vec{w}_\ell, A_\ell \vec{w}_\ell)_\ell \\
 &= (\vec{w}_\ell, \vec{w}_\ell)_{A_\ell} \\
 &= \|\vec{w}_\ell\|_{A_\ell}^2 \\
 &= \|\vec{u}_\ell - \tilde{\Pi}_\ell \vec{u}_\ell\|_{A_\ell}^2.
 \end{aligned}$$

Finally, using the norm equivalence in (5.4)

$$\begin{aligned}
 C_1 h_\ell^d \|\vec{u}_\ell - \tilde{\Pi}_\ell \vec{u}_\ell\|_\ell^2 &\stackrel{5.4}{\leq} \|\vec{u}_\ell - R_{\ell-1}\vec{u}_\ell\|_{L^2}^2 \\
 &\stackrel{5.4}{\leq} \|\vec{u}_\ell - R_{\ell-1}\vec{u}_\ell\|_{H^1}^2 \\
 &= C h_\ell^2 \|\vec{u}_\ell - \tilde{\Pi}_\ell \vec{u}_\ell\|_{A_\ell}^2.
 \end{aligned}$$

The proof of Theorem.(5.19) showed that

$$C_6^{n_\ell} h_\ell^{d-2} \leq \rho_\ell \leq C_7^{n_\ell} h_\ell^{d-2},$$

combining this with the last estimate gives (5.17).  $\square$

**Corollary 5.27.** Let  $\mathcal{T}_\ell$  and  $V_\ell$  as usual with  $A_\ell$  the standard stiffness matrix for the model problem. Then, the weak approximation property, Assumption (A4) holds:  $\exists C_{A4}$  such that

$$\left( (I_\ell - \tilde{\Pi}_\ell) \vec{u}_\ell, \vec{u}_\ell \right)_{A_\ell} \leq C_{A4}^2 \rho_\ell^{-1} \|A_\ell \vec{u}_\ell\|_\ell^2 \quad (5.18)$$

for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$ .

*Proof.* This follows immediately from Theorem.(4.18). Since the Galerkin condition (A0) and the strong approximate property hold.  $\square$

**Remark 5.28.** Therefore, using Richardson's smoother, the W and V-Cycle algorithms defined in Chapter.(4) converge. See Theorem(4.24), (4.28), and (4.32).

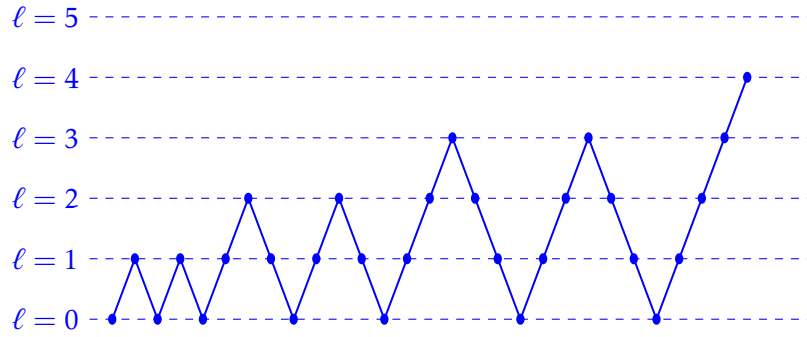
### 5.3 Full Multigrid for Finite Element Methods

**Definition 5.29.** Let MG be as in Definition.(4.2). Define

$$\hat{\vec{u}}_0 := A_0^{-1} \vec{f}_0.$$

For  $\ell = 1$  to  $\ell = L$

$$\begin{aligned}
 \vec{u}_\ell^{(0)} &= P_{\ell-1} \hat{\vec{u}}_{\ell-1} \\
 \vec{u}_\ell^{(\sigma+1)} &= \text{MG}\left(\vec{f}_\ell, \ell, \vec{u}_\ell^{(\sigma)}\right), \quad 0 \leq \sigma \leq r-1 \\
 \hat{\vec{u}}_\ell &= \vec{u}_\ell^{(r)}
 \end{aligned}$$

Figure 28: One full cycle with  $r = 2$ .

Hence

$$\vec{f}_{ell} := \begin{bmatrix} (f, \phi_{\ell,1})_{L^2} \\ (f, \phi_{\ell,2})_{L^2} \\ \vdots \\ (f, \phi_{\ell,n_\ell})_{L^2} \end{bmatrix} \in \mathbb{R}^{n_\ell} \quad (0 \leq \ell \leq L).$$

**Theorem 5.30.** Suppose that, in general, for all  $\vec{u}_\ell^{(0)}$

$$\left\| \vec{u}_\ell^E - \text{MG}(\vec{g}_\ell, \ell, \vec{u}_\ell^{(0)}) \right\|_{A_\ell} \leq \gamma \left\| \vec{u}_\ell^E - \vec{u}_\ell^{(0)} \right\|_{A_\ell}, \quad (5.19)$$

where  $0 < \gamma < 1$  is independent of  $\ell$  and

$$\vec{u}_\ell^E := A_\ell^{-1} \vec{g}_\ell.$$

Suppose that  $r$  in the Full Multigrid algorithm in Definition (5.29) is sufficiently large. Then there exists a constant  $C > 0$  such that

$$\|u_\ell^* - \hat{u}_\ell\|_{H^1} \leq Ch_\ell \|u\|_{H^2}, \quad (5.20)$$

where

$$\vec{v}_{\ell-1} \in \mathbb{R}^{n_{\ell-1}} \xleftrightarrow{B_\ell} \vec{v}_{\ell-1} \in V_{\ell-1}. \quad (\text{Full Multigrid})$$

$$a(u_\ell^*, v_\ell) = (f, v_\ell), \quad \forall v_\ell \in V_\ell,$$

and

$$a(u, v) = (f, v), \quad \forall v_\ell \in H_0^1(\Omega).$$

*Proof.* Define

$$\hat{e}_\ell = u_\ell^* - \hat{u}_\ell \in V_\ell.$$

This is the algebraic error not the FE error. Clearly  $\hat{e}_0 \equiv 0$ . Now

$$|\hat{e}_\ell|_{H^1}^2 = a(\hat{e}_\ell, \hat{e}_\ell) = \|\hat{\tilde{e}}_\ell\|_{A_\ell}^2,$$

where

$$\hat{\tilde{e}}_\ell = \vec{u}_\ell^* - \vec{\hat{u}}_\ell \in \mathbb{R}^{n_\ell} \xleftrightarrow{B_\ell} \hat{e}_\ell = u_\ell^* - \hat{u}_\ell \in V_\ell.$$

Then

$$\begin{aligned}
 |\hat{e}_\ell|_{H^1} &= \|\vec{u}_\ell^* - \hat{u}_\ell\|_{A_\ell} \\
 &\stackrel{(5.19)}{\leq} \gamma^r \|\vec{u}_\ell^* - P_{\ell-1} \hat{u}_{\ell-1}\|_{A_\ell} \\
 &= \gamma^r |u_\ell^* - \hat{u}_{\ell-1}|_{H^1} \\
 &\leq \gamma^r \{ |u_\ell^* - u|_{H^1} + |u - u_{\ell-1}^*|_{H^1} + |u_{\ell-1}^* - \hat{u}_{\ell-1}|_{H^1} \} \\
 &\leq \gamma^r \{ ch_\ell |u|_{H^2} + C \cdot 2h_\ell |u|_{H^2} + |\hat{e}_{\ell-1}|_{H^1} \} \\
 &= \tilde{C} \gamma^r h_\ell |u|_{H^2} + \gamma^r |\hat{e}_{\ell-1}|_{H^1}.
 \end{aligned}$$

By induction

$$\begin{aligned}
 |\hat{e}_\ell|_{H^1} &\leq \{ \tilde{C} h_\ell \gamma^r + \tilde{C} h_{\ell-1} \gamma^{2r} + \dots + \tilde{C} h_0 \gamma^{(\ell+1)r} \} |u|_{H^2} \\
 &\leq h_\ell |u|_{H^2} \frac{\tilde{C} \gamma^r}{1 - 2\gamma^r},
 \end{aligned}$$

provided

$$2\gamma^r < 1.$$

Setting

$$C := \frac{\tilde{C} \gamma^r}{1 - 2\gamma^r},$$

the theorem is proven.  $\square$

**Remark 5.31.** The operation count for the full multigrid algorithm is  $O(n_L)$ . In this sense, multigrid is optimal.

## 5.4 Multigrid and Subspace Corrections

**Definition 5.32.** Suppose  $0 \leq j < \ell$ . Define

$$P_j^\ell := P_{\ell-1} P_{\ell-2} \dots P_j \in \mathbb{R}^{n_\ell \times n_j}$$

In particular,

$$P_{\ell-1}^\ell = P_\ell \in \mathbb{R}^{n_\ell \times n_{\ell-1}}.$$

**Lemma 5.33.** Suppose  $v_j \in V_j$  for some  $0 \leq j < \ell$ , where  $V_0 \subset V_1 \subset \dots \subset V_\ell \subset \dots$  are the usual nested FE spaces. Let  $\vec{v}_j \in \mathbb{R}^{n_j}$  be the coordinate vector of  $v_j$  in basis  $\mathcal{B}_j$ . Then, the unique coordinate vector of  $v_j$  in the basis  $\mathcal{B}_\ell$  is

$$P_j^\ell \vec{v}_j \in \mathbb{R}^{n_\ell}.$$

*Proof.* Simple exercise.  $\square$

**Definition 5.34.** Define  $R_j^\ell \in \mathbb{R}^{n_j \times n_\ell}$ , for  $0 \leq j < \ell$  via

$$R_j^\ell = (P_j^\ell)^T.$$

$R_j^\ell$  is called the multilevel restriction matrix and  $P_j^\ell$  is called the multilevel prolongation matrix.

**Lemma 5.35.** *With the usual construction for the conforming FE method, we have, for any  $0 \leq j < \ell$ ,*

$$A_j = R_j^\ell A_\ell P_j^\ell \in \mathbb{R}^{n_j \times n_j}.$$

*Proof.* This follows because the Galerkin condition holds:

$$\begin{aligned} A_j &= R_j A_{j+1} P_j \\ &= R_j R_{j+1} A_{j+2} P_{j+1} P_j \\ &\vdots \\ &= R_j \cdots R_{\ell-1} A_\ell P_{\ell-1} \cdots P_j \\ &= R_j^\ell A_\ell P_j^\ell. \end{aligned}$$

□

**Definition 5.36.** *For any  $0 \leq j < \ell$  define the multilevel Ritz projection matrix via*

$$\Pi_j^\ell := A_j^{-1} R_j^\ell A_\ell \in \mathbb{R}^{n_j \times n_\ell}.$$

*$R_j^\ell$  is called the multilevel restriction matrix and  $P_j^\ell$  is called the multilevel prolongation matrix.*

**Lemma 5.37.** *We have, for  $0 \leq j < \ell$ ,*

$$\Pi_j^\ell := \Pi_j \times \Pi_{j+1} \times \cdots \times \Pi_{\ell-1}.$$

*Proof.* The matrix product on the RHS is

$$\Pi_j \cdots \Pi_{\ell-1} = A_j^{-1} R_j A_{j+1} \times A_{j+1}^{-1} R_{j+1} A_{j+2} \times \cdots \times A_{\ell-1}^{-1} R_{\ell-1} A_\ell.$$

□

**Definition 5.38.** *Define, for any  $0 \leq j < \ell$*

$$\tilde{\Pi}_{\ell,j} := P_j^\ell \Pi_j^\ell \in \mathbb{R}^{n_\ell \times n_\ell}.$$

*Observe that, for  $j = \ell - 1$ , we have*

$$\tilde{\Pi}_{\ell,j} = \tilde{\Pi}_\ell \in \mathbb{R}^{n_\ell \times n_\ell}.$$

**Theorem 5.39.** Let  $a(\cdot, \cdot)$ ,  $V_\ell$ , et cetera, be defined as usual for the conforming FE method described in Chapter.5. Let  $0 \leq j < \ell$  and  $\vec{u}_\ell \in V_\ell$  be arbitrary. Set

$$u'_j = R_j \vec{u}_\ell \xleftrightarrow{\mathcal{B}_j} \vec{u}''_j \in \mathbb{R}^{n_j}.$$

Then, if  $\vec{u}_\ell$  is the coordinate vector of  $\vec{u}_\ell \in V_\ell$  in the basis  $\mathcal{B}_\ell$ , it follows that the unique representation of  $R_j u_\ell \in V_j$  in the basis  $\mathcal{B}_j$  is precisely

$$\vec{u}''_j = \Pi_j^\ell \vec{u}_\ell \in \mathbb{R}^{n_j}.$$

And the unique representation of  $R_j \vec{u}_\ell \in V_\ell$  in the basis  $\mathcal{B}_\ell$  is precisely

$$\tilde{\Pi}_{\ell,j} \vec{u}_\ell \in \mathbb{R}^{n_\ell}.$$

*Proof.* Let  $\vec{u}_\ell \in V_\ell$  be given.  $R_j \vec{u}_\ell$  is defined as the unique solution to

$$a(\mathcal{R}_j \vec{u}_\ell, v_j) = a(\vec{u}_\ell, v_j), \quad \forall v_j \in V_j.$$

Then

$$a(\mathcal{R}_j \vec{u}_\ell, v_j) = \vec{u}''_j, \vec{v}_{j, A_j}.$$

On the other hand

$$a(\vec{u}_\ell, v_j) = (\vec{u}_\ell, P_j^\ell \vec{v}_j)_{A_\ell},$$

where

$$\vec{v}_j \in \mathbb{R}^{n_j} \xleftrightarrow{\mathcal{B}_\ell} v_j \in V_j,$$

as usual. Going further, we have

$$\begin{aligned} a(\vec{u}_\ell, v_j) &= (A_\ell \vec{u}_\ell, P_j^\ell \vec{v}_j)_\ell \\ &= (R_j^\ell A_\ell \vec{u}_\ell, \vec{v}_j)_j, \end{aligned}$$

and

$$a(\mathcal{R}_j \vec{u}_\ell, v_j) = A_j u'_j, v_{j, j}.$$

Therefore,

$$A_j u'_j = R_j^\ell A_\ell \vec{u}_\ell,$$

or

$$u'_j = A_j^{-1} R_j^\ell A_\ell \vec{u}_\ell = \Pi_j^\ell \vec{u}_\ell.$$

The second part follows from lemma.(5.33). □

**Definition 5.40.** Define, for any  $0 \leq j < \ell$

$$T_j^\ell := \Pi_j^\ell - K_j^m \Pi_j^\ell \in \mathbb{R}^{n_j \times n_\ell},$$

where  $m$  is a given integer exponent. Define

$$\tilde{T}_{\ell,j} = P_j^\ell T_j^\ell \in \mathbb{R}^{n_\ell \times n_\ell},$$

$\tilde{T}_{\ell,j}$  is called a subspace "projection" matrix.

## 5.5 Properties of the "Projections"

**Lemma 5.41.** *Let  $0 \leq j < \ell$ . Then*

$$\Pi_j^\ell P_j^\ell = I_j, \quad (5.21)$$

and

$$\tilde{\Pi}_{\ell,j}^2 = \tilde{\Pi}_{\ell,j}.$$

*Proof.* The Galerkin condition holds in the sense that

$$A_j = R_j^\ell A_\ell P_j^\ell. \quad (5.22)$$

By definition

$$\Pi_j^\ell = A_j^{-1} R_j^\ell A_\ell,$$

so that

$$\begin{aligned} \Pi_j^\ell P_j^\ell &= A_j^{-1} R_j^\ell A_\ell P_j^\ell \\ &= A_j^{-1} A_j = I_j. \end{aligned}$$

Now,

$$\tilde{\Pi}_{\ell,j}^2 = P_j^\ell \Pi_j^\ell P_j^\ell \Pi_j^\ell = P_j^\ell \Pi_j^\ell = \tilde{\Pi}_{\ell,j}.$$

□

**Definition 5.42.** *Let  $0 \leq j < \ell$ . Define*

$$T_j^{\prime\ell} := \Pi_j^\ell - (K_j^*)^m \Pi_j^\ell, \quad \tilde{T}_{\ell,j}' := P_j^\ell T_j^{\prime\ell},$$

where

$$K_j^* = I_j - S_j^T A_j.$$

**Lemma 5.43.** *Let  $0 \leq j < \ell$ . Then*

$$\tilde{T}_{\ell,j}^* = \tilde{T}_{\ell,j}. \quad (5.23)$$

And

$$\tilde{T}_{\ell,j}^* = \tilde{T}_{\ell,j}'.$$

*Proof.* Recall

$$(\tilde{\Pi}_{\ell,j} \vec{u}_\ell, \vec{v}_\ell)_{A_\ell} = (\vec{u}_\ell, \tilde{\Pi}_{\ell,j}^* \vec{v}_\ell)_{A_\ell},$$

for all  $\vec{u}_\ell, \vec{v}_\ell \in \mathbb{R}^{n_\ell}$ . Then

$$\begin{aligned} (\tilde{\Pi}_{\ell,j} \vec{u}_\ell, \vec{v}_\ell)_{A_\ell} &= (P_j^\ell \Pi_j^\ell \vec{u}_\ell, A_\ell \vec{v}_\ell)_\ell \\ &= (\Pi_j^\ell \vec{u}_\ell, R_j^\ell A_\ell \vec{v}_\ell)_j \\ &= (A_j^{-1} R_j^\ell A_\ell \vec{u}_\ell, R_j^\ell A_\ell \vec{v}_\ell)_j \\ &= (R_j^\ell A_\ell \vec{u}_\ell, A_j^{-1} R_j^\ell A_\ell \vec{v}_\ell)_j \end{aligned}$$

$$\begin{aligned}
&= (A_\ell \vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_\ell \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j}^* \vec{v}_\ell)_{A_\ell}.
\end{aligned}$$

Now

$$\tilde{T}_{\ell,j} = \tilde{\Pi}_{\ell,j} - P_j^\ell K_j^m \Pi_j^\ell.$$

Therefore,

$$\begin{aligned}
(\tilde{T}_{\ell,j} \vec{u}_\ell, \vec{v}_\ell)_{A_\ell} &= (\tilde{\Pi}_{\ell,j} \vec{u}_\ell, \vec{v}_\ell)_{A_\ell} - (P_j^\ell K_j^m \Pi_j^\ell \vec{u}_\ell, \vec{v}_\ell)_{A_\ell} \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (P_j^\ell K_j^m \Pi_j^\ell \vec{u}_\ell, A_\ell \vec{v}_\ell)_\ell \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (K_j^m \Pi_j^\ell \vec{u}_\ell, R_j^\ell A_\ell \vec{v}_\ell)_j \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (K_j^m \Pi_j^\ell \vec{u}_\ell, A_j A_j^{-1} R_j^\ell A_\ell \vec{v}_\ell)_j \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (K_j^m \Pi_j^\ell \vec{u}_\ell, A_j^{-1} R_j^\ell A_\ell \vec{v}_\ell)_{A_j} \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (\Pi_j^\ell \vec{u}_\ell, (K_j^m)^* \Pi_j^\ell \vec{v}_\ell)_{A_j} \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (\Pi_j^\ell \vec{u}_\ell, (K_j^*)^m \Pi_j^\ell \vec{v}_\ell)_{A_j} \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (\Pi_j^\ell \vec{u}_\ell, (K_j^*)^m \Pi_j^\ell \vec{v}_\ell)_{A_j} \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (A_j^{-1} R_j^\ell A_\ell \vec{u}_\ell, (K_j^*)^m \Pi_j^\ell \vec{v}_\ell)_{A_j} \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (R_j^\ell A_\ell \vec{u}_\ell, (K_j^*)^m \Pi_j^\ell \vec{v}_\ell)_j \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (A_\ell \vec{u}_\ell, P_j^\ell (K_j^*)^m \Pi_j^\ell \vec{v}_\ell)_\ell \\
&= (\vec{u}_\ell, \tilde{\Pi}_{\ell,j} \vec{v}_\ell)_{A_\ell} - (\vec{u}_\ell, P_j^\ell (K_j^*)^m \Pi_j^\ell \vec{v}_\ell)_{A_\ell} \\
&= (\vec{u}_\ell, \tilde{T}_{\ell,j}' \vec{v}_\ell)_{A_\ell}.
\end{aligned}$$

So,

$$\tilde{\Pi}_{\ell,j}^* = \tilde{\Pi}_{\ell,j},$$

and

$$\tilde{T}_{\ell,j}^* = \tilde{T}_{\ell,j}'.$$

□

**Remark 5.44.** We note that, in general

$$\tilde{T}_{\ell,j}^2 \neq \tilde{T}_{\ell,j}.$$

$\tilde{T}_{\ell,j}$  is not a projection matrix, in this sense. We call  $\tilde{T}_{\ell,j}$  (and  $\tilde{T}_{\ell,j}'$ ) projection-like operators.

**Theorem 5.45.** Let  $0 \leq j < \ell$ . Then

$$(I_\ell - \tilde{\Pi}_{\ell,j})(I_\ell - \tilde{T}_{\ell,j}) = I_\ell - \tilde{\Pi}_{\ell,j}, \quad (5.24)$$

and

$$(I_\ell - \tilde{T}_{\ell,j}^*)(I_\ell - \tilde{\Pi}_{\ell,j}) = I_\ell - \tilde{\Pi}_{\ell,j}. \quad (5.25)$$

*Proof.* The LHS of (5.24) is

$$M_\ell := I_\ell - \tilde{\Pi}_{\ell,j} - \tilde{T}_{\ell,j} + \tilde{\Pi}_{\ell,j} \tilde{T}_{\ell,j}.$$

By definition,

$$\begin{aligned} \tilde{\Pi}_{\ell,j} \tilde{T}_{\ell,j} &= P_j^\ell \Pi_j^\ell P_j^\ell T_j^\ell \\ &\stackrel{5.21}{=} P_j^\ell I_j T_j^\ell = \tilde{T}_{\ell,j}. \end{aligned}$$

So

$$M_\ell := I_\ell - \tilde{\Pi}_{\ell,j} - \tilde{T}_{\ell,j} + \tilde{T}_{\ell,j} = I_\ell - \tilde{\Pi}_{\ell,j}.$$

For the second identity, we denote the LHS of (5.25) as

$$M'_\ell = I_\ell - \tilde{T}_{\ell,j}^* - \tilde{\Pi}_{\ell,j} + \tilde{T}_{\ell,j}^* \tilde{\Pi}_{\ell,j}.$$

Then

$$\begin{aligned} \tilde{T}_{\ell,j}^* \tilde{\Pi}_{\ell,j} &= P_j^\ell T_j^{\ell'} P_j^\ell \Pi_j^\ell \\ &= P_j^\ell (\Pi_j^\ell - (K_j^*)^m \Pi_j^\ell) P_j^\ell \Pi_j^\ell \\ &\stackrel{5.21}{=} P_j^\ell (I_j^\ell - (K_j^*)^m) \Pi_j^\ell = \tilde{T}_{\ell,j}^*. \end{aligned}$$

So

$$M'_\ell = I_\ell - \tilde{\Pi}_{\ell,j}^*,$$

as desired. □

**Lemma 5.46.** Let  $0 \leq j < \ell$ . Then

$$I_\ell - \tilde{\Pi}_{\ell,j} = (I_\ell - \tilde{T}_{\ell,j}^*)(I_\ell - \tilde{\Pi}_{\ell,j})(I_\ell - \tilde{T}_{\ell,j}). \quad (5.26)$$

*Proof.* Since the Galerkin condition holds,  $\tilde{\Pi}_{\ell,j}$  is a bona fide projection matrix (Lemma (5.41))

$$\tilde{\Pi}_{\ell,j}^2 = \tilde{\Pi}_{\ell,j},$$

and

$$(I_\ell - \tilde{\Pi}_{\ell,j})^2 = I_\ell - \tilde{\Pi}_{\ell,j},$$

is a direct consequence. By the last result

$$\begin{aligned} I_\ell - \tilde{\Pi}_{\ell,j} &= (I_\ell - \tilde{\Pi}_{\ell,j})(I_\ell - \tilde{\Pi}_{\ell,j}) \\ &\stackrel{(5.24)(5.25)}{=} (I_\ell - \tilde{T}_{\ell,j}^*)(I_\ell - \tilde{\Pi}_{\ell,j})(I_\ell - \tilde{\Pi}_{\ell,j})(I_\ell - \tilde{T}_{\ell,j}) \\ &= (I_\ell - \tilde{T}_{\ell,j}^*)(I_\ell - \tilde{\Pi}_{\ell,j})(I_\ell - \tilde{T}_{\ell,j}). \end{aligned}$$

□

**Lemma 5.47.** Let  $0 \leq i < j < \ell$ . Then

$$P_j^\ell (I_j - \tilde{T}_{j,i}) = (I_\ell - \tilde{T}_{\ell,i}) P_j^\ell. \quad (5.27)$$



*Proof.*

$$\begin{aligned}
 P_j^\ell (I_j - \tilde{T}_{j,i}) &\stackrel{5.21}{=} P_j^\ell (I_j - \tilde{T}_{j,i}) \Pi_j^\ell P_j^\ell \\
 &= \left\{ P_j^\ell \Pi_j^\ell - P_j^\ell P_i^j T_i^j \Pi_j^\ell \right\} P_j^\ell \\
 &= \left\{ \tilde{\Pi}_{\ell,j} - P_i^\ell (\Pi_i^j - K_i^m \Pi_i^j) \Pi_j^\ell \right\} P_j^\ell \\
 &= \left\{ \tilde{\Pi}_{\ell,j} - P_i^\ell \Pi_i^j + P_i^\ell K_i^m \Pi_i^j \right\} P_j^\ell \\
 &= \left\{ \tilde{\Pi}_{\ell,j} - \tilde{T}_{\ell,i} \right\} P_j^\ell \\
 &= P_j^\ell \Pi_j^\ell P_j^\ell - \tilde{T}_{\ell,i} P_j^\ell \\
 &\stackrel{5.21}{=} (I_\ell - \tilde{T}_{\ell,i}) P_j^\ell.
 \end{aligned}$$

□

**Corollary 5.48.** Let  $0 \leq i < j < \ell$ . Then

$$P_j^\ell (I_j - \tilde{\Pi}_{j,i}) = (I_\ell - \tilde{\Pi}_{\ell,i}) P_j^\ell. \quad (5.28)$$

**Theorem 5.49.** Let  $V_\ell, T_\ell$  and  $a(\cdot, \cdot)$  be defined as usual. Consider the symmetric V-cycle algorithm:  $m = m_1 = m_2$  and  $p = 1$ . The error propagation matrix can be expressed as

$$\begin{aligned}
 E_\ell &= (K_\ell^*)^m (I_\ell - \tilde{T}_{\ell,\ell-1}^*) \times \cdots \times (I_\ell - \tilde{T}_{\ell,1}^*) (I_\ell - \tilde{\Pi}_{\ell,0}^*) \\
 &\quad \times (I_\ell - \tilde{T}_{\ell,1}) \times \cdots \times (I_\ell - \tilde{T}_{\ell,\ell-1}) (K_\ell)^m,
 \end{aligned} \quad (5.29)$$

for all  $\ell \geq 1$ .

*Proof.* Define the quantity

$$M_{\ell,j} := I_\ell - \tilde{\Pi}_{\ell,j} + P_j^\ell E_j \Pi_j^\ell,$$

for any  $0 \leq j < \ell$ . Observe that when  $j = 0$

$$M_{\ell,0} = I_\ell - \tilde{\Pi}_{\ell,0},$$

since  $E_0 = 0$ . Now by Theorem.(4.8),

$$M_{\ell,j} = I_\ell - \tilde{\Pi}_{\ell,j} + P_j^\ell (K_j^*)^m \left( I_j - \tilde{\Pi}_{j,j-1} + P_{j-1}^j E_{j-1} \Pi_{j-1}^j \right) K_j^m \Pi_{j-1}^j. \quad (5.30)$$

In other words,

$$M_{\ell,j} = I_\ell - \tilde{\Pi}_{\ell,j} + P_j^\ell M_{j,j-1} K_j^m \Pi_{j-1}^j.$$

Now, observe that

$$\begin{aligned}
 P_j^\ell (K_j^*)^m &\stackrel{5.21}{=} P_j^\ell (K_j^*)^m \Pi_j^\ell P_j^\ell \\
 &= (P_j^\ell \Pi_j^\ell - P_j^\ell \Pi_j^\ell + P_j^\ell (K_j^*)^m \Pi_j^\ell) P_j^\ell \\
 &= (P_j^\ell \Pi_j^\ell - \tilde{T}_{\ell,j}^*) P_j^\ell \\
 &= P_j^\ell \Pi_j^\ell P_j^\ell - \tilde{T}_{\ell,j}^* P_j^\ell
 \end{aligned} \quad (5.31)$$

$$\stackrel{5.21}{=} (I_\ell - \tilde{T}_{\ell,j}^*) P_j^\ell.$$

Similarly,

$$\begin{aligned} K_j^m \Pi_j^\ell &= \Pi_j^\ell - \Pi_j^\ell + K_j^m \Pi_j^\ell \\ &= \Pi_j^\ell - T_j^\ell \\ &\stackrel{5.21}{=} \Pi_j^\ell P_j^\ell (\Pi_j^\ell - T_{\ell,j}) \\ &= \Pi_j^\ell (P_j^\ell \Pi_j^\ell - \tilde{T}_{\ell,j}) \\ &= \Pi_j^\ell P_j^\ell \Pi_j^\ell - \Pi_j^\ell \tilde{T}_{\ell,j} \\ &\stackrel{5.21}{=} \Pi_j^\ell (I_\ell - \tilde{T}_{\ell,j}). \end{aligned} \tag{5.32}$$

Putting (5.30)- (5.32) together, we have

$$\begin{aligned} M_{\ell,j} &= I_\ell - \tilde{\Pi}_{\ell,j} + (I_\ell - \tilde{T}_{\ell,j}^*) P_j^\ell \left\{ I_j - \tilde{\Pi}_{j,j-1} + P_{j-1}^j E_{j-1} \Pi_{j-1}^j \right\} \Pi_j^\ell (I_\ell - \tilde{T}_{\ell,j}) \\ &\stackrel{5.26}{=} (I_\ell - \tilde{T}_{\ell,j}^*) \left\{ I_\ell - \tilde{\Pi}_{\ell,j} + P_j^\ell (I_j - \tilde{\Pi}_{j,j-1} + P_{j-1}^j E_{j-1} \Pi_{j-1}^j) \Pi_j^\ell \right\} (I_\ell - \tilde{T}_{\ell,j}) \\ &= (I_\ell - \tilde{T}_{\ell,j}^*) \left\{ I_\ell - \tilde{\Pi}_{\ell,j} + \tilde{\Pi}_{\ell,j} - \tilde{\Pi}_{\ell,j-1} + P_{j-1}^\ell E_{j-1} \Pi_{j-1}^\ell \right\} (I_\ell - \tilde{T}_{\ell,j}). \end{aligned}$$

Or

$$M_{\ell,j} = (I_\ell - \tilde{T}_{\ell,j}^*) M_{\ell,j-1} (I_\ell - \tilde{T}_{\ell,j}).$$

Therefore

$$\begin{aligned} M_{\ell,\ell-1} &= (I_\ell - \tilde{T}_{\ell,\ell-1}^*) M_{\ell,\ell-2} (I_\ell - \tilde{T}_{\ell,\ell-1}) \\ &= (I_\ell - \tilde{T}_{\ell,\ell-1}^*) (I_\ell - \tilde{T}_{\ell,\ell-2}^*) M_{\ell,\ell-3} (I_\ell - \tilde{T}_{\ell,\ell-2}) (I_\ell - \tilde{T}_{\ell,\ell-1}) \\ &= (I_\ell - \tilde{T}_{\ell,\ell-1}^*) \times \cdots \times (I_\ell - \tilde{T}_{\ell,1}^*) (I_\ell - \tilde{\Pi}_{\ell,0}) \times (I_\ell - \tilde{T}_{\ell,1}) (I_\ell - \tilde{T}_{\ell,2}) \times \cdots \times (I_\ell - \tilde{T}_{\ell,\ell-1}) \end{aligned}$$

But recall that

$$E_\ell = (K_\ell^*)^m M_{\ell,\ell-1} K_\ell^m,$$

from Theorem.(4.8). The theorem is proven.  $\square$

**Corollary 5.50.** For the one-sided V-cycle with only pre-smoothing ( $p = 1, m := m_1 > 0$  and  $m_2 = 0$ ), we have

$$E_\ell^{pre} = (I_\ell - \tilde{\Pi}_{\ell,0}) (I_\ell - \tilde{T}_{\ell,1}) (I_\ell - \tilde{T}_{\ell,2}) \times \cdots \times (I_\ell - \tilde{T}_{\ell,\ell-1}) K_\ell^m.$$

And for the algorithm with only post-smoothing ( $p = 1, m := m_2 > 0$  and  $m_1 = 0$ ), we have

$$E_\ell^{post} = (K_\ell^*)^m (I_\ell - \tilde{T}_{\ell,\ell-1}^*) \times \cdots \times (I_\ell - \tilde{T}_{\ell,1}^*) (I_\ell - \tilde{\Pi}_{\ell,0}).$$

Therefore

$$E_\ell = E_\ell^{post} \times E_\ell^{pre}.$$

Moreover

$$(E_\ell^{post})^* = E_\ell^{pre}.$$

Clearly

$$E_\ell^* = E_\ell = (E_\ell^{pre})^* E_\ell^{pre}$$

is SPSPD.

**Theorem 5.51.** *Both of the one-sided method converge for any  $m > 0$  if Richardson's method is used for smoothing.*

*Proof.* We have shown, in Theorem.(4.31) that there is some  $C_0$  such that

$$\|E_\ell \vec{u}_\ell\|_{A_\ell} \leq \frac{C_0}{m + C_0} \|\vec{u}_\ell\|_{A_\ell},$$

for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$ . We wish to prove that

$$\|E_\ell^{\text{pre}} \vec{u}_\ell\|_{A_\ell} \leq \gamma,$$

for some  $0 \leq \gamma < 1$ . Observe that

$$\begin{aligned} \|E_\ell^{\text{pre}} \vec{u}_\ell\|_{A_\ell}^2 &= (E_\ell^{\text{pre}} \vec{u}_\ell, E_\ell^{\text{pre}} \vec{u}_\ell)_{A_\ell} \\ &= (\vec{u}_\ell, (E_\ell^{\text{pre}})^* E_\ell^{\text{pre}} \vec{u}_\ell)_{A_\ell} \\ &= (\vec{u}_\ell, E_\ell \vec{u}_\ell)_{A_\ell} \\ &\stackrel{\text{C.S.}}{\leq} \|\vec{u}_\ell\|_{A_\ell} \|E_\ell \vec{u}_\ell\|_{A_\ell} \\ &\leq \|\vec{u}_\ell\|_{A_\ell} \frac{C_0}{m + C_0} \|\vec{u}_\ell\|_{A_\ell} \\ &\leq \frac{C_0}{m + C_0} \|\vec{u}_\ell\|_{A_\ell}^2. \end{aligned}$$

Thus

$$\|E_\ell^{\text{pre}} \vec{u}_\ell\|_{A_\ell} \leq \sqrt{\frac{C_0}{m + C_0}} \|\vec{u}_\ell\|_{A_\ell}.$$

□

**Theorem 5.52.** *For the W-cycle algorithm with only pre-smoothing :  $m := m_1 > 0, m_2 = 0, p = 2$ . The error propagation matrix is*

$$E_\ell^{w, \text{pre}} = F_\ell E_\ell^{\text{pre}},$$

where  $E_\ell^{\text{pre}}$  is defined above, and  $F_\ell \in \mathbb{R}^{n_\ell \times n_\ell}$  is a matrix with

$$\|F_\ell\|_{A_\ell} \leq 1.$$

consequently the one-sided W-cycle method with pre-smoothing converges for any  $m > 0$ .

*Proof.* Exercise.

□

**Theorem 5.53.** For the 2-sided W-cycle algorithm with  $p = 2, m := m_1 = m_2$ , the error propagation matrix is

$$E_\ell^W = (E_\ell^{pre})^* D_\ell E_\ell^{pre}$$

where

$$E_\ell^{pre} = (I_\ell - \tilde{\Pi}_{\ell,0})(I_\ell - \tilde{T}_{\ell,1})(I_\ell - \tilde{T}_{\ell,2}) \times \cdots \times (I_\ell - \tilde{T}_{\ell,\ell-1}) K_\ell^m.$$

and

$$\|D_\ell\|_{A_\ell} \leq 1, \quad \forall \ell \geq 1.$$

The algorithm converges if the symmetric V-cycle algorithm converges with the uniform contraction  $0 < \gamma < 1$ , i.e.

$$\|E_\ell \vec{u}_\ell\|_{A_\ell} \leq \gamma \|\vec{u}_\ell\|_{A_\ell},$$

for all  $\vec{u}_\ell \in \mathbb{R}^{n_\ell}$ , with  $E_\ell = (E_\ell^{pre})^* E_\ell^{pre}$ . Here  $\gamma$  may (and usually does) depend upon  $m$ .

*Proof.* Exercise. □

## 6 Subspace Decompositions

**Definition 6.1.**

$$0 < m_0 < m_1 < \cdots < m_\ell < \cdots < m_L \leq n_L.$$

Let the matrices

$$Q_j^L \in \mathbb{R}^{n_L \times m_j}$$

be defined for  $0 \leq j \leq L$ . We say that Assumption(S1) holds, or, equivalently, that the  $Q_j^L$  prolongation matrices gives us a subspace decomposition of  $\mathbb{R}^{n_L}$ , iff for every  $\vec{u}_L \in \mathbb{R}^{n_L}$ , there exist vector

$$\vec{w}_j \in \mathbb{R}^{m_j}, \quad 0 \leq j \leq L,$$

such that

$$\vec{u}_L = \sum_{j=0}^L Q_j^L \vec{w}_j \quad (6.1)$$

Herein

$$(\vec{u}_L, \vec{u}_L) := \sum_{j=1}^{n_L} (\vec{u}_L)_j (\vec{v}_L)_j \quad \forall \vec{u}_L, \vec{v}_L \in \mathbb{R}^{n_L}.$$

And

$$(\vec{u}_\ell, \vec{u}_\ell)_\ell := \sum_{j=1}^{m_\ell} (\vec{u}_\ell)_j (\vec{v}_\ell)_j \quad \forall \vec{u}_\ell, \vec{v}_\ell \in \mathbb{R}^{m_\ell}.$$

**Definition 6.2.** The matrix  $C \in \mathbb{R}^{n_L \times n_L}$  is called an additive subspace preconditioner iff there are SPD matrices  $C_\ell \in \mathbb{R}^{m_\ell \times m_\ell}$ , for each  $0 \leq \ell \leq L$ , such that

$$C = \sum_{\ell=0}^L Q_\ell^L C_\ell^{-1} Z_\ell^L, \quad (6.2)$$

where

$$Z_\ell^L = (Q_\ell^L)^T \in \mathbb{R}^{m_\ell \times n_L},$$

i.e.

$$(\vec{u}_\ell, Q_\ell^L \vec{v}_L)_\ell = ((Q_\ell^L)^T \vec{u}_\ell, \vec{v}_L) = (Z_\ell^L \vec{u}_\ell, \vec{v}_L)$$

for all  $\vec{u}_\ell \in \mathbb{R}^{m_\ell}$  and  $\vec{v}_L \in \mathbb{R}^{n_L}$ .

**Lemma 6.3.** Suppose that assumption (S1) holds. Then  $C$  is SPD with respect to  $(\cdot, \cdot)$ , and, consequently if  $A$  is also SPD w.r.t  $(\cdot, \cdot)$ , then  $CA$  is SPD w.r.t.  $(\cdot, \cdot)_{C^{-1}}$  and  $(\cdot, \cdot)_A$ .

*Proof.*  $C$  is clearly symmetric, since each  $C_\ell^{-1}$  is symmetric. Now, let  $\vec{u}_L \in \mathbb{R}^{n_L}$  be arbitrary. Then

$$\begin{aligned} (\vec{u}_L, C \vec{u}_L) &\stackrel{(6.2)}{=} \left( \vec{u}_L, \sum_{\ell=0}^L Q_\ell^L C_\ell^{-1} Z_\ell^L \vec{u}_L \right) \\ &= \sum_{\ell=0}^L (\vec{u}_L, Q_\ell^L C_\ell^{-1} Z_\ell^L \vec{u}_L) \end{aligned}$$

$$= \sum_{\ell=0}^L (Z_\ell^L \vec{u}_L, C_\ell^{-1} Z_\ell^L \vec{u}_L)_\ell \geq 0,$$

since  $C_\ell^{-1}$  is SPD w.r.t.  $(\cdot, \cdot)_\ell$ ,  $0 \leq \ell \leq L$ . Suppose

$$\sum_{\ell=0}^L (Z_\ell^L \vec{u}_L, C_\ell^{-1} Z_\ell^L \vec{u}_L)_\ell = 0.$$

Since, again,  $C_\ell^{-1}$  is SPD, it must be that

$$Z_\ell^L \vec{u}_L = \vec{0}, \quad 0 \leq \ell \leq L. \quad (6.3)$$

In this case, since (S1) holds, we have

$$\begin{aligned} \|\vec{u}_L\|^2 &= (\vec{u}_L, \vec{u}_L) \\ &\stackrel{\text{S1}}{=} \left( \vec{u}_L, \sum_{\ell=1}^L Q_\ell^L \vec{w}_\ell \right) \\ &= \sum_{\ell=1}^L (\vec{u}_L, Q_\ell^L \vec{w}_\ell) \\ &= \sum_{\ell=1}^L (Z_\ell^L \vec{u}_L, \vec{w}_\ell)_\ell \\ &\stackrel{(6.3)}{=} \sum_{\ell=1}^L (\vec{0}, \vec{w}_\ell)_\ell = 0. \end{aligned}$$

Thus

$$\|\vec{u}_L\| = 0.$$

Hence  $\vec{u}_L = \vec{0}$ , which shows that C is SPD w.r.t.  $(\cdot, \cdot)$ . The results concerning CA are easy to show.  $\square$

**Theorem 6.4.** Suppose that (S1) holds. Then, for any  $\vec{u}_L \in \mathbb{R}^{n_L}$ ,

$$\begin{aligned} (\vec{u}_L, \vec{u}_L)_{C^{-1}} &= (\vec{u}_L, C^{-1} \vec{u}_L) \\ &= \min_{\vec{u} = \sum_{\ell=0}^L Q_\ell^L \vec{w}_\ell} = \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell. \end{aligned} \quad (6.4)$$

*Proof.* Since each  $C_\ell \in \mathbb{R}^{m_\ell \times m_\ell}$  is SPD w.r.t.  $(\cdot, \cdot)_\ell$ ,  $(\cdot, \cdot)_{C_\ell^{-1}}$  is a bona fide inner product. Therefore

$$\begin{aligned} (\vec{u}_\ell, C_\ell^{-1} \vec{v}_\ell) &=: (\vec{u}_\ell, \vec{v}_\ell)_{C_\ell^{-1}} \\ &\stackrel{\text{C.S.}}{\leq} \|\vec{u}_\ell\|_{C_\ell^{-1}} \|\vec{v}_\ell\|_{C_\ell^{-1}} \\ &= \sqrt{(\vec{u}_\ell, \vec{u}_\ell)_{C_\ell^{-1}}} \sqrt{(\vec{v}_\ell, \vec{v}_\ell)_{C_\ell^{-1}}}. \end{aligned}$$

Let  $\vec{u}_L \in \mathbb{R}^{n_L}$  be arbitrary. Then

$$\vec{u}_L \stackrel{\text{(S1)}}{=} \sum_{\ell=0}^L Q_\ell^L \vec{w}_\ell, \quad \exists \vec{w}_\ell \in \mathbb{R}^{m_\ell}.$$

We have

$$\begin{aligned}
(\vec{u}_L, \vec{u}_L)_{C^{-1}} &= \sum_{\ell=0}^L (\vec{u}_L, Q_\ell^L \vec{w}_\ell)_{C^{-1}} \\
&= \sum_{\ell=0}^L (Z_\ell^L C^{-1} \vec{u}_L, \vec{w}_\ell)_\ell \\
&= \sum_{\ell=0}^L (Z_\ell^L C^{-1} \vec{u}_L, C_\ell \vec{w}_\ell)_{C_\ell^{-1}} \\
&\stackrel{C.S.}{\leq} \sum_{\ell=0}^L \|Z_\ell^L C^{-1} \vec{u}_L\|_{C_\ell^{-1}} \|C_\ell \vec{w}_\ell\|_{C_\ell^{-1}} \\
&\stackrel{C.S.}{\leq} \left( \sum_{\ell=0}^L \|Z_\ell^L C^{-1} \vec{u}_L\|_{C_\ell^{-1}}^2 \right)^{1/2} \left( \sum_{\ell=0}^L \|C_\ell \vec{w}_\ell\|_{C_\ell^{-1}}^2 \right)^{1/2} \\
&\stackrel{C.S.}{\leq} \left( \sum_{\ell=0}^L (Z_\ell^L C^{-1} \vec{u}_L, Z_\ell^L C^{-1} \vec{u}_L)_{C_\ell^{-1}} \right)^{1/2} \left( \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, C_\ell \vec{w}_\ell)_{C_\ell^{-1}} \right)^{1/2} \\
&\leq \left( \sum_{\ell=0}^L (Z_\ell^L C^{-1} \vec{u}_L, C_\ell^{-1} Z_\ell^L C^{-1} \vec{u}_L)_\ell \right)^{1/2} \left( \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell \right)^{1/2} \\
&\stackrel{(6.2)}{=} \left( (C^{-1} \vec{u}_L, C C^{-1} \vec{u}_L) \right)^{1/2} \left( \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell \right)^{1/2} \\
&= \|\vec{u}_L\|_{C^{-1}} \left( \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell \right)^{1/2}.
\end{aligned}$$

So,

$$(\vec{u}_L, \vec{u}_L)_{C^{-1}} \leq \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell.$$

For the particular choice

$$\vec{w}_\ell = C_\ell^{-1} Z_\ell^L C^{-1} \vec{u}_L \in \mathbb{R}^{m_\ell}, \quad 0 \leq \ell \leq L,$$

we have

$$\vec{u}_L = \sum_{\ell=0}^L Q_\ell^L \vec{w}_\ell,$$

and

$$\begin{aligned}
\sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell &= \sum_{\ell=0}^L (C_\ell C_\ell^{-1} Z_\ell^L C^{-1} \vec{u}_L, C_\ell^{-1} Z_\ell^L C^{-1} \vec{u}_L)_\ell \\
&= \sum_{\ell=0}^L (C^{-1} \vec{u}_L, Q_\ell^L C_\ell^{-1} Z_\ell^L C^{-1} \vec{u}_L) \\
&\stackrel{(6.2)}{=} (C^{-1} \vec{u}_L, C C^{-1} \vec{u}_L) \\
&= (\vec{u}_L, \vec{u}_L)_{C^{-1}}.
\end{aligned}$$

□

**Theorem 6.5.** *The eigenvalues of  $CA$  are positive, provided  $A$  is SPD w.r.t.  $(\cdot, \cdot)$  and assumption (S1) holds. Moreover*

$$\lambda_{\max}(CA) = \max_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{(A\vec{u}_L, \vec{u}_L)}{\min_{\vec{u}_L = \sum_{\ell=0}^L Q_\ell^L \vec{w}_\ell} \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell}, \quad (6.5)$$

$$\lambda_{\min}(CA) = \min_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{(A\vec{u}_L, \vec{u}_L)}{\min_{\vec{u}_L = \sum_{\ell=0}^L Q_\ell^L \vec{w}_\ell} \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell}. \quad (6.6)$$

*Proof.* Recall that  $CA$  is SPD w.r.t.  $(\cdot, \cdot)_{C^{-1}}$ . Thus the eigenvalues are positive real and the corresponding eigenvectors may be chosen so that they form an orthonormal basis for  $\mathbb{R}^{n_L}$  w.r.t.  $(\cdot, \cdot)_{C^{-1}}$ . Moreover the Rayleigh quotient formula holds

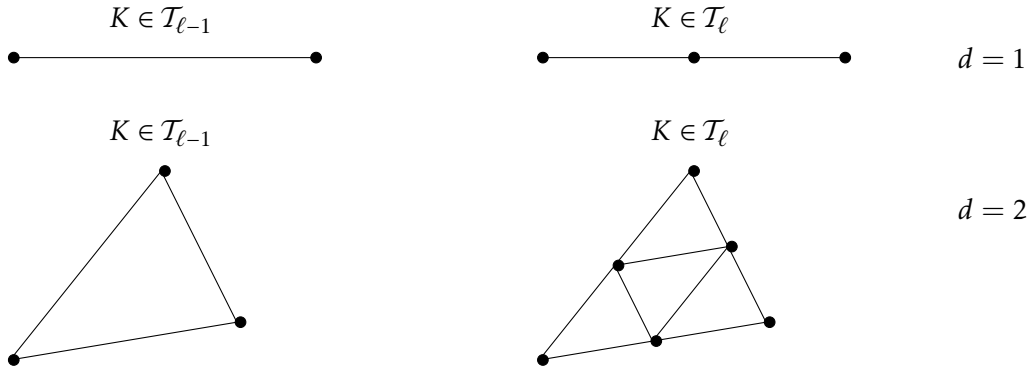
$$\begin{aligned} \lambda_{\max}(CA) &= \max_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{(CA\vec{u}_L, \vec{u}_L)_{C^{-1}}}{(\vec{u}_L, \vec{u}_L)_{C^{-1}}} \\ &\stackrel{(6.4)}{=} \max_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{(A\vec{u}_L, \vec{u}_L)}{\min_{\vec{u}_L = \sum_{\ell=0}^L Q_\ell^L \vec{w}_\ell} \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell}. \end{aligned}$$

The formula for  $\lambda_{\min}(CA)$  is proven in a similar way. □



## 6.1 Hierarchical Basis

**Definition 6.6.** Suppose  $d = 1$  or  $d = 2$ . Suppose  $\Omega$  is an open interval ( $d = 1$ ) or an open convex polygonal domain ( $d = 2$ ). Suppose  $\mathcal{T}_0$  is an initial conforming triangulation ( $d = 2$ ) or partition ( $d = 1$ ) of  $\Omega$  into triangles ( $d = 1$ ) or subintervals ( $d = 1$ ). Let  $\mathcal{T}_\ell$  be the family of triangulations obtained by subdividing each triangular of  $\mathcal{T}_{\ell-1}$  into 4 similar triangles by joining the edge midpoints ( $d = 2$ ) or subdividing each interval of  $\mathcal{T}_{\ell-1}$  into 2 equal subintervals.



Set

$$V_\ell := \left\{ v \in C^0(\overline{\Omega}) \mid v|_K \in \mathbb{P}_1(K), \forall K \in \mathcal{T}_\ell, v|_{\partial\Omega} \equiv 0 \right\} \quad (6.7)$$

for all  $0 \leq \ell \leq L$ . Define

$$n_\ell := \dim(V_\ell).$$

Set  $W_0 := V_0$ , and, for  $1 \leq \ell \leq L$ , define

$$W_\ell := \left\{ v \in V_\ell \mid v(\vec{N}_{\ell-1,j}) = 0, \forall 1 \leq j \leq n_{\ell-1} \right\}. \quad (6.8)$$

Recall that  $\{\vec{N}_{\ell,j}\}_{j=1}^{n_\ell} \subset \Omega$  is set of interior vertices of  $\mathcal{T}_\ell$ . Set

$$m_\ell := \dim(W_\ell).$$

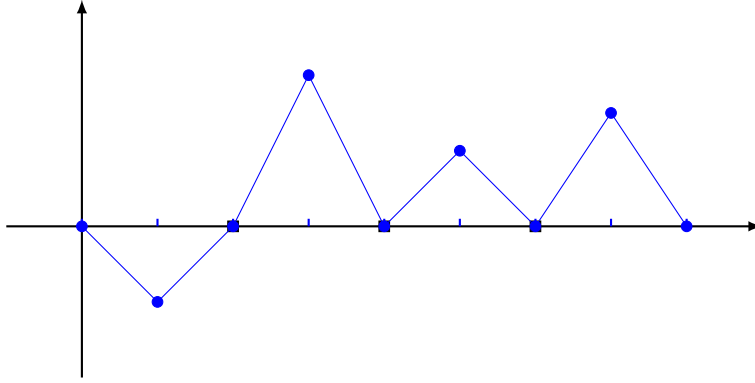
By  $B_\ell^V$ ,  $0 \leq \ell \leq L$ , we denote the family of Lagrange nodal bases of  $V_\ell$

$$B_\ell^V := \left\{ \psi_{\ell,j} \right\}_{j=1}^{n_\ell},$$

with the property that

$$\psi_{\ell,j}(\vec{N}_{\ell,i}) = \delta_{ij}, \quad 1 \leq i, j \leq n_\ell.$$

**Example.**  $d=1$ ,  $\Omega = (0,1)$ ,  $\ell = 2$ ,



**Lemma 6.7.** For the spaces  $W_\ell \subseteq V_\ell$  as in the last definition, we have

$$V_\ell = V_{\ell-1} \oplus W_\ell \quad 1 \leq \ell \leq L, \quad (6.9)$$

and

$$V_L = V_0 \oplus V_1 \oplus \cdots \oplus W_L. \quad (6.10)$$

*Proof.* Let  $\mathcal{I}_\ell : C^0(\overline{\Omega}) \rightarrow V_\ell$  be the standard Lagrange linear nodal interpolation operator. It has the property that

$$\mathcal{I}_\ell(v)(\vec{N}_{\ell,i}) = v(\vec{N}_{\ell,i}), \quad 1 \leq i \leq n_\ell. \quad (6.11)$$

Let  $v_\ell \in V_\ell$  be a given arbitrary function. Write

$$v_\ell = \mathcal{I}_{\ell-1}(v_\ell) + \{v_\ell - \mathcal{I}_{\ell-1}(v_\ell)\}.$$

Clearly

$$\mathcal{I}_{\ell-1}(v_\ell) \in V_{\ell-1},$$

and

$$v_\ell - \mathcal{I}_{\ell-1}(v_\ell) \in W_\ell.$$

Indeed

$$v_\ell(\vec{N}_{\ell-1,j}) - \mathcal{I}_{\ell-1}v_\ell(\vec{N}_{\ell-1,j}) = 0.$$

for every  $1 \leq j \leq n_{\ell-1}$ . And, this decomposition must be unique. Suppose not. Then

$$v_\ell = v_{\ell-1}^{(i)} + w_\ell^{(i)}, \quad i = 1, 2.$$

So,

$$\begin{aligned} 0 &= \left( v_{\ell-1}^{(1)} - v_{\ell-1}^{(2)} \right) + \left( w_\ell^{(1)} - w_\ell^{(2)} \right) \\ &=: v_{\ell-1} - w_\ell. \end{aligned}$$

Therefore

$$V_{\ell-1} \ni v_\ell = w_\ell \in W_\ell.$$

Clearly, both functions must be identically zero. This proves (6.9). Identity (6.10) follows from (6.9).  $\square$

**Definition 6.8.** For  $1 \leq \ell \leq L$ , define  $B_\ell^W := \{\phi_{\ell,i}\}_{i=1}^{m_\ell} \subset W_\ell$  such that

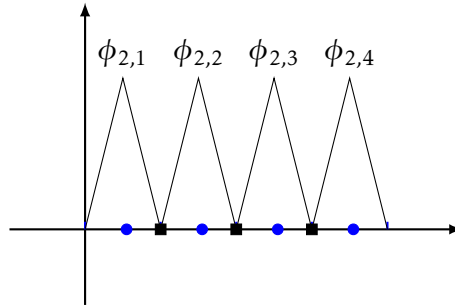
$$W_\ell \ni \phi_{\ell,i}(\vec{N}_{\ell,j}^W) = \delta_{ij}, \quad 1 \leq i, j \leq m_\ell,$$

where

$$\{\vec{N}_{\ell,j}^W\}_{j=1}^{m_\ell} := \{\vec{N}_{\ell,j}\}_{j=1}^{n_\ell} \setminus \{\vec{N}_{\ell,j}\}_{j=1}^{n_{\ell-1}}.$$

$$B_0^W = B_0^V.$$

**Example.**  $d=1$ ,  $\Omega = (0,1)$ ,  $\ell = 2$ ,  $n_1 = 3$ ,  $m_2 = 4$ ,  $n_2 = 7$



**Lemma 6.9.** Definitions (6.6) and (6.8) are consistent in the definition of  $m_\ell$ , and moreover,  $B_\ell^W$  is a basis for  $W_\ell$ , for each  $1 \leq \ell \leq L$ .

*Proof.* Exercise. □

**Lemma 6.10.** Suppose  $W_\ell \subseteq V_\ell$ ,  $0 \leq \ell \leq L$ , are as defined in Definition (6.6). Then

$$H_\ell := \cup_{j=0}^\ell B_j^W$$

is a basis for  $V_\ell$ , for any  $1 \leq \ell \leq L$ .

*Proof.* The result follows if we can show that

$$\text{span}(H_\ell) = V_\ell,$$

and  $H_\ell$  is linearly independent.

Suppose  $v_\ell \in V_\ell$  is arbitrary. Then, there exist unique  $w_j \in W_j$ ,  $0 \leq j \leq \ell$ , such that

$$v_\ell = w_0 + w_1 + \cdots + w_\ell.$$

In fact, we can write down this decomposition explicitly:

$$v_\ell = \mathcal{I}_0 v_\ell + (\mathcal{I}_1 v_\ell - \mathcal{I}_0 v_\ell) + (\mathcal{I}_2 v_\ell - \mathcal{I}_1 v_\ell) + \cdots + (v_\ell - \mathcal{I}_{\ell-1} v_\ell)$$

Setting

$$w_0 := \mathcal{I}_0 v_\ell,$$

$$w_j := \mathcal{I}_j v_\ell - \mathcal{I}_{j-1} v_\ell, \quad 1 \leq j \leq \ell$$

and noticing that

$$\mathcal{I}_\ell v_\ell = v_\ell,$$

gives the result. Now, since  $w_j \in W_j$ ,  $0 \leq j \leq \ell$ , there are unique coefficients  $C_{j1}, \dots, C_{jm_j} \in \mathbb{R}$  such that

$$w_j = \sum_{k=1}^{m_j} C_{jk} \phi_{jk}.$$

Hence

$$v_\ell = \sum_{j=0}^{\ell} w_j = \sum_{j=0}^{\ell} \sum_{k=1}^{m_j} C_{jk} \phi_{jk}$$

and

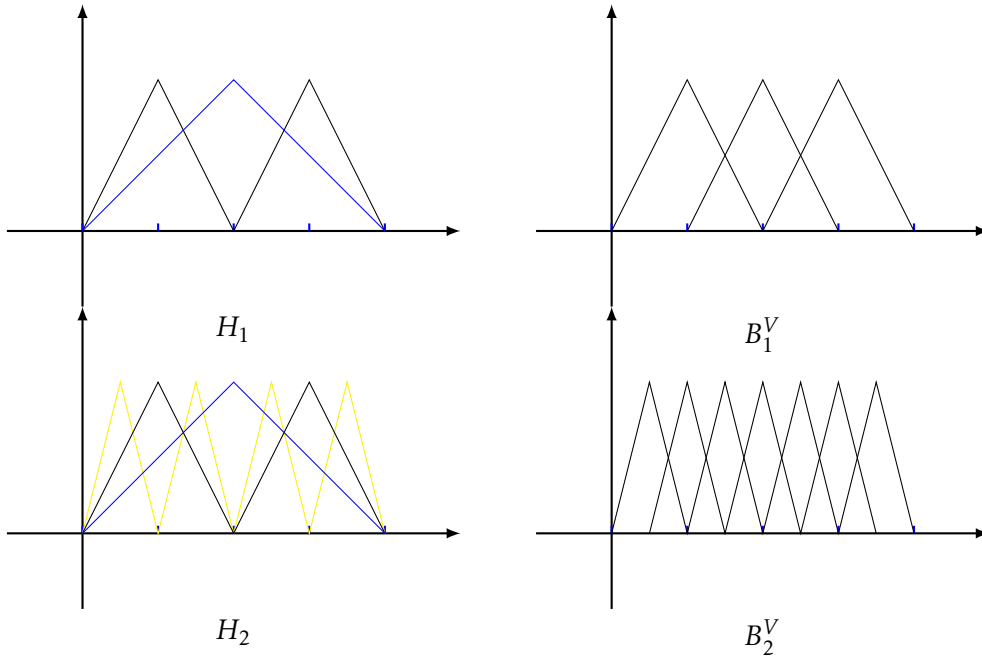
$$V_\ell \subseteq \text{span}(H_\ell).$$

On the other hand, it should be clear that

$$\text{span}(H_\ell) \subseteq V_\ell.$$

Since  $\#(H_\ell) = \#B_\ell^V = n_\ell$ , it follows that  $H_\ell$  is linearly independent, since  $B_\ell^V$  is a basis. □

**Example.**  $d=1$



Now, we need to connect the spaces  $W_j$  to  $V_\ell$  where  $0 \leq j \leq \ell$ .

**Remark 6.11.** Clearly

$$W_j \subset V_j \subset V_\ell, \quad 0 \leq j \leq \ell.$$

Moreover precisely, for each  $0 \leq j \leq \ell$ , and each  $1 \leq i \leq m_j$

$$\phi_{j,i} \in V_\ell.$$

Therefore, there are unique numbers

$$q_{j,k,i}^\ell \in \mathbb{R}, \quad 1 \leq k \leq n_\ell, \quad 1 \leq i \leq m_j,$$

such that

$$\phi_{j,i} = \sum_{k=1}^{n_\ell} q_{j,k,i}^\ell \psi_{\ell,k}.$$

**Definition 6.12.** Define the matrix  $Q_j^\ell \in \mathbb{R}^{n_\ell \times m_j}$  via

$$[Q_j^\ell]_{i,k} := q_{j,i,k}^\ell.$$

$Q_j^\ell$  is called an auxiliary prolongation matrix.

**Lemma 6.13.** Suppose that  $\vec{w}_j \in \mathbb{R}^{m_j}$  is the coordinate vector of the function  $w_j \in W_j$  w.r.t. the basis  $B_j^W$ . The coordinate vector of  $w_j \in V_\ell$  in the basis  $B_\ell^V$  is simply

$$Q_j^\ell \vec{w}_j \in \mathbb{R}^{n_\ell}.$$

*Proof.* Exercise. □

**Remark 6.14.** Note that the family of spaces  $W_j$  are hierarchical, but are not nested

$$W_0 \not\subset W_1 \not\subset W_2 \cdots.$$

It makes no sense to stack the prolongation matrices

$$Q_j^\ell \neq Q_k^\ell Q_j^k,$$

for  $j < k < \ell$ . In fact, the product is not defined.

**Definition 6.15.** Define the operator  $B_j : W_j \rightarrow W_j'$  via

$$B_j[w_1](w_2) = \sum_{i=1}^{m_j} w_1(\vec{N}_{j,i}^W) w_2(\vec{N}_{j,i}^W)$$

Define the matrix  $B_j \in \mathbb{R}^{m_j \times m_j}$  via

$$\begin{aligned} [B_j]_{i,k} &:= B_j[\phi_{j,i}](\phi_{j,k}) \\ &= \sum_{r=1}^{m_j} \phi_{j,i}(\vec{N}_{j,r}^W) \phi_{j,k}(\vec{N}_{j,r}^W) \\ &= \sum_{r=1}^{m_j} \delta_{ir} \delta_{rk} = \delta_{ik}. \end{aligned}$$

**Definition 6.16.** Let  $A_L \in \mathbb{R}^{n_L \times n_L}$  be the SPD matrix defined via

$$[A_L]_{ij} = a(\phi_{L,j}, \phi_{L,i}), \quad 1 \leq i, j \leq n_L,$$

where

$$a(u, v) = (\nabla u, \nabla v)_{L^2}, \quad \forall u, v \in H_0^1(\Omega).$$

The hierarchical basis preconditioner for  $A_L$  is defined as

$$C_H = \sum_{\ell=0}^L Q_\ell^L C_\ell^{-1} Z_\ell^L \stackrel{??}{=} \sum_{\ell=0}^L Q_\ell^L Z_\ell^L, \quad (6.12)$$

where  $Q_L \in \mathbb{R}^{n_L \times m_\ell}$  is the auxiliary prolongation matrix from the Definition.6.12 and

$$Z_\ell^L = (Q_\ell^L)^T.$$

**Lemma 6.17.** Assumption (S1) holds for the hierarchical basis decomposition. In particular, for any  $u_L \in V_L$ , there exist unique  $w_\ell \in W_\ell, 0 \leq \ell \leq L$ , such that

$$u_L = \sum_{\ell=0}^L w_\ell.$$

equivalently, for any  $\vec{u}_L \in \mathbb{R}^{n_L}$ , there are unique vectors  $w_\ell \in \mathbb{R}^{m_\ell}$ , such that

$$\vec{u}_L = \sum_{\ell=0}^L Q_\ell^L w_\ell.$$

Consequently,  $C_H$  defined in (6.12) is SPD.

*Proof.* This follows from lemma.6.7 and lemma.6.3. □

**Remark 6.18.** Our goal is now to show that

$$\lambda_{\min}(C_H A_L) \geq C_1 \left(1 + |\ln h_L|^2\right)^{-1},$$

and

$$\lambda_{\max}(C_H A_L) \leq C_2,$$

where  $C_1, C_2 > 0$  are independent of  $h$ , using Theorem.6.5. In this case

$$\frac{\lambda_{\max}}{\lambda_{\min}} =: \kappa(C_H A_L) \leq \frac{C_2}{C_1} \left(1 + |\ln h_L|^2\right),$$

suppose that

$$h_L = (1/2)^L,$$

then

$$|\ln h_L|^2 = L^2 |\ln(1/2)|^2.$$

This analysis will only work for  $d = 2$ .

Now, we need some technical results, for more details see [2].

**Lemma 6.19.** For any  $0 \leq \ell \leq L$ , with  $\mathcal{I}_{-1} \equiv 0$ ,

$$\|\mathcal{I}_\ell u_L - \mathcal{I}_{\ell-1} u_L\|_{L^2(\Omega)} \leq Ch_\ell (1 + \sqrt{L-\ell}) |u_L|_{H^1(\Omega)}. \quad (6.13)$$

for all  $u_L \in V_L$ , where  $\Omega \subset \mathbb{R}^2$  (i.e.  $d=2$ ).

*Proof.* Define the piecewise constant function  $\bar{u}_L^\ell$  such that

$$\bar{u}_L^\ell|_K := \frac{1}{|K|} \int_K u_L(\vec{x}) d\vec{x}.$$

Then

$$\begin{aligned} \|\mathcal{I}_\ell u_L - \mathcal{I}_{\ell-1} u_L\|_{L^2(\Omega)}^2 &= \|\mathcal{I}_\ell u_L - \mathcal{I}_{\ell-1}(\mathcal{I}_\ell u_L)\|_{L^2(\Omega)}^2 \\ &\leq ch_\ell^2 \sum_{K \in \mathcal{T}_h} |\mathcal{I}_\ell u_L|_{H^1(K)}^2 \quad (\text{interpolation error}) \\ &= ch_\ell^2 \sum_{K \in \mathcal{T}_h} |\mathcal{I}_\ell u_L - \bar{u}_L^\ell|_{H^1(K)}^2 \quad (\bar{u}_L^\ell = \text{constant}) \\ &\leq ch_\ell^2 \sum_{K \in \mathcal{T}_h} \|\mathcal{I}_\ell u_L - \bar{u}_L^\ell\|_{L^\infty(K)}^2 \quad (\text{inverse inequality}) \\ &\leq ch_\ell^2 \sum_{K \in \mathcal{T}_h} \|u_L - \bar{u}_L^\ell\|_{L^\infty(K)}^2 \quad (u_L \in C^0(\bar{\Omega})). \end{aligned}$$

Next, we use the estimate (for  $d=2$ ):

$$\|u_L - \bar{u}_L^\ell\|_{L^\infty(K)}^2 \leq C(1 + \ln(h_\ell)/h_L) |u_L|_{H^1(K)}^2.$$

This estimate requires that

$$\int_K (u_L - \bar{u}_L^\ell) d\vec{x} = 0.$$

See chapter. 7 of [2] for details. Now notice that

$$h_\ell = h_0 2^{-\ell} \quad 1 \leq \ell \leq L.$$

So,

$$\ln(h_\ell/h_L) = \ln(2^{L-\ell}) = (L-\ell) \ln 2.$$

The result follows. □

**Lemma 6.20.** There is some constant  $C_1 > 0$  such that

$$\lambda_{\min}(C_H A_L) \geq C_1 (1 + |\ln h_L|^2)^{-1}. \quad (6.14)$$

*Proof.* By definition for any  $w_{\ell,1}, w_{\ell,2} \in W_\ell$

$$C_\ell[w_{\ell,1}](w_{\ell,2}) = \sum_{i=1}^{m_\ell} w_{\ell,1}(\vec{N}_{\ell,i}^W) w_{\ell,2}(\vec{N}_{\ell,i}^W).$$

Let

$$\vec{w}_{\ell,\alpha} \in \mathbb{R}^{m_\ell} \xleftrightarrow{B_\ell^W} w_{\ell,\alpha}, \quad \alpha = 1, 2.$$

Then

$$\begin{aligned} (C_\ell \vec{w}_{\ell,1}, \vec{w}_{\ell,2})_\ell &= \sum_{i=1}^{m_\ell} [\vec{w}_{\ell,1}]_i [\vec{w}_{\ell,2}]_i \\ &= \sum_{i=1}^{m_\ell} w_{\ell,1}(\vec{N}_{\ell,i}^W) w_{\ell,2}(\vec{N}_{\ell,i}^W) \\ &= C_\ell[w_{\ell,1}](w_{\ell,2}) \\ &= C_\ell[w_{\ell,2}](w_{\ell,1}) \\ &=: \langle w_{\ell,1}, w_{\ell,2} \rangle_{C_\ell}. \end{aligned}$$

This is like a mass-lumping inner product. All that is missing is a factor of  $h_\ell^2$ . As in Lemma 5.14, there are constant  $\tilde{C}_1, \tilde{C}_2 > 0$  such that, for all  $0 \leq \ell \leq L$ ,

$$\tilde{C}_1 h_\ell^2 \langle w_{\ell,\alpha}, w_{\ell,\alpha} \rangle_{C_\ell} \leq \|w_{\ell,\alpha}\|_{L^2(\Omega)}^2 \leq \tilde{C}_2 h_\ell^2 \langle w_{\ell,\alpha}, w_{\ell,\alpha} \rangle_{C_\ell} \quad (6.15)$$

Therefore, for any  $w_\ell \in W_\ell \xleftrightarrow{B_\ell^W} \vec{w}_\ell \in \mathbb{R}^{m_\ell}$ ,

$$\begin{aligned} (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell &= h_\ell^{-2} h_\ell^2 \langle w_\ell, w_\ell \rangle_{C_\ell} \\ &\stackrel{6.15}{\leq} \tilde{C}_1 h_\ell^{-2} \|w_\ell\|_{L^2(\Omega)}^2 \\ &\stackrel{\mathcal{I}_{-1}=0}{\leq} \tilde{C}_1 h_\ell^{-2} \|w_\ell - \mathcal{I}_{\ell-1} w_\ell\|_{L^2(\Omega)}^2 \\ &\leq \tilde{C}_2 |w_\ell|_{H^1(\Omega)}^2 \quad (\text{interp. error}) \\ &\leq \tilde{C}_3 h_\ell^{-2} \|w_\ell\|_{L^2(\Omega)}^2 \quad (\text{inverse ineq.}) \\ &\stackrel{6.15}{\leq} \tilde{C}_4 (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell. \end{aligned} \quad (6.16)$$

Therefore, there are constants  $\tilde{C}_5, \tilde{C}_6 > 0$  such that we have the equivalence

$$\tilde{C}_5 \sum_{\ell=0}^L |w_\ell|_{H^1(\Omega)}^2 \leq \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell \leq \tilde{C}_6 \sum_{\ell=0}^L |w_\ell|_{H^1(\Omega)}^2, \quad (6.17)$$

for any  $w_\ell \in W_\ell$ , in general. Now, let  $u_L \in V_L$  be given and

$$u_L = \sum_{\ell=0}^L w_\ell, \exists! w_\ell \in W_\ell.$$

Recall that

$$w_\ell = \mathcal{I}_\ell u_L - \mathcal{I}_{\ell-1} u_L, \quad 1 \leq \ell \leq L,$$

and

$$w_0 = \mathcal{I}_0 u_L.$$

Then, from (6.16)

$$\sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell \leq \tilde{C}_1 \sum_{\ell=0}^L h_\ell^{-2} \|w_\ell\|_{L^2(\Omega)}^2$$



$$\begin{aligned}
& \stackrel{(6.13)}{\leq} C \sum_{\ell=0}^L (1 + \sqrt{L-\ell})^2 |u_L|_{H^1(\Omega)}^2 \\
& \leq C \sum_{\ell=0}^L (1 + L - \ell) |u_L|_{H^1(\Omega)}^2 \\
& \leq C(1 + L + L^2) |u_L|_{H^1(\Omega)}^2 \\
& \stackrel{L \geq 1}{\leq} CL^2 |u_L|_{H^1(\Omega)}^2.
\end{aligned}$$

But

$$\begin{aligned}
|u_L|_{H^1(\Omega)}^2 &= (\nabla u_L, \nabla u_L) \\
&= a(u_L, u_L) \\
&= (A \vec{u}_L, \vec{u}_L).
\end{aligned}$$

And

$$\begin{aligned}
|\ln h_L|^2 &= |\ln(h_0 2^{-L})|^2 \\
&= |\ln(h_0) - L \ln(2)|^2 \\
&= \ln^2(h_0) - 2 \ln(h_0) L \ln(2) + L^2 \ln^2(2).
\end{aligned}$$

So

$$L^2 \leq C \left(1 + |\ln(h_L)|^2\right), \quad \exists C > 0.$$

Thus,

$$\sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell) \leq C \left(1 + |\ln(h_L)|^2\right) (A \vec{u}_L, \vec{u}_L),$$

and it follows from Theorem 6.5 that

$$\lambda_{\min}(C_H A_L) \geq C_1 \left(1 + |\ln h_L|^2\right)^{-1}.$$

□

**Lemma 6.21.** Let  $a_j, b_j \geq 0, -\infty < j < \infty$ , with

$$s_1 := \sum_{j=-\infty}^{\infty} a_j \leq \infty,$$

and

$$s_2 := \sum_{j=-\infty}^{\infty} b_j \leq \infty.$$

Then

$$\sum_{j=-\infty}^{\infty} \left( \sum_{k=-\infty}^{\infty} a_{j-k} b_k \right) \leq s_1^2 s_2. \quad (6.18)$$

*Proof.* Exercise. □

**Lemma 6.22.** For any  $v_\ell \in V_\ell$  and  $v_k \in V_k$ ,  $0 \leq \ell \leq k \leq L$ , and  $d=2$ ,  $\exists C > 0$  such that

$$\int_{\Omega} \nabla v_\ell \nabla v_k d\vec{x} \leq C 2^{(\ell-k)/2} |v_\ell|_{H^1(\Omega)} \left( h_k^{-1} \|v_k\|_{L^2(\Omega)} \right). \quad (6.19)$$

*Proof.* For any  $K \in \mathcal{T}_h$ , since  $\Delta v_\ell|_K \equiv 0$ ,

$$\begin{aligned} \int_K \nabla v_\ell \nabla v_k d\vec{x} &= \int_{\partial K} \frac{\partial v_\ell}{\partial n} v_k ds \\ &\leq C h_\ell^{-1} |v_\ell|_{H^1(K)} \int_{\partial K} v_k ds \\ &\leq \left( C h_\ell^{-1} |v_\ell|_{H^1(K)} \right) \left( h_k \sum_{\vec{N}_k \in \partial K} v_k(\vec{N}_k) \right) \\ &\stackrel{\text{C.S.}}{\leq} \left( C h_\ell^{-1} |v_\ell|_{H^1(K)} \right) \left( h_k \left( \frac{h_\ell}{h_k} \right)^{1/2} \left( \sum_{\vec{N}_k \in \partial K} |v_k(\vec{N}_k)|^2 \right)^{1/2} \right) \\ &\stackrel{(5.4)}{\leq} C \left( \frac{h_\ell}{h_k} \right)^{1/2} |v_\ell|_{H^1(K)} h_k^{-1} \|v_k\|_{L^2(K)}. \end{aligned}$$

Thus

$$\begin{aligned} \int_{\Omega} \nabla v_\ell \nabla v_k d\vec{x} &= \sum_{K \in \mathcal{T}_h} \int_K \nabla v_\ell \nabla v_k d\vec{x} \\ &\leq C 2^{(\ell-k)/2} \sum_{K \in \mathcal{T}_h} |v_\ell|_{H^1(K)} h_k^{-1} \|v_k\|_{L^2(K)} \\ &\stackrel{\text{C.S.}}{\leq} C 2^{(\ell-k)/2} |v_\ell|_{H^1(\Omega)} h_k^{-1} \|v_k\|_{L^2(\Omega)}. \end{aligned}$$

□

**Lemma 6.23. (Strengthened Cauchy-Schwarz Inequality)** For any  $w_\ell \in W_\ell$  and  $w_k \in W_k$ ,  $0 \leq \ell \leq k \leq L$ ,

$$\int_{\Omega} \nabla w_\ell \nabla w_k d\vec{x} \leq C 2^{(\ell-k)/2} |w_\ell|_{H^1(\Omega)} |w_k|_{H^1(\Omega)}. \quad (6.20)$$

*Proof.* Observe that

$$w_k = w_k - \mathcal{I}_{\ell-1}(w_k).$$

We use the interpolation error estimate

$$\|w_k - \mathcal{I}_{\ell-1}(w_k)\|_{L^2(\Omega)} \leq C h_k |w_k|_{H^1(\Omega)},$$

to conclude that

$$\|w_k\|_{L^2(\Omega)} \leq C h_k |w_k|_{H^1(\Omega)}.$$

Now, we use the last result. Since  $w_\ell \in V_\ell$  and  $w_k \in V_k$ ,

$$\int_{\Omega} \nabla w_\ell \nabla w_k d\vec{x} \leq C^{(\ell-k)/2} |w_\ell|_{H^1(\Omega)} h_k^{-1} \|w_k\|_{L^2(\Omega)}$$

$$\leq C^{(\ell-k)/2} |w_\ell|_{H^1(\Omega)} |w_k|_{H^1(\Omega)}$$

□

**Lemma 6.24.** *There is a constant  $C_2 > 0$  such that*

$$\lambda_{\max}(C_H A_L) \leq C_2,$$

*independent of  $L$ .*

*Proof.* Let  $v_L \in V_L$  be arbitrary.

$$v_L \in V_L \xleftrightarrow{\mathcal{B}_L} \vec{v}_L \in \mathbb{R}^{n_L}.$$

There exist unique  $w_\ell \in W_\ell \xleftrightarrow{\mathcal{B}_\ell^W} \vec{w}_\ell \in \mathbb{R}^{m_\ell}$  such that

$$v_L = \sum_{\ell=0}^L w_\ell \leftrightarrow \vec{v}_L = \sum_{\ell=0}^L Q_\ell^L \vec{w}_\ell.$$

Then

$$\begin{aligned}
 (\vec{v}_L, \vec{v}_L)_{A_L} &= (\vec{v}_L, A_L \vec{v}_L) \\
 &= a(\vec{v}_L, \vec{v}_L) \\
 &= a\left(\sum_{\ell=0}^L w_\ell, \sum_{k=0}^L w_k\right) \\
 &= \int_{\Omega} \left( \nabla \sum_{\ell=0}^L w_\ell \right) \left( \nabla \sum_{k=0}^L w_k \right) d\vec{x} \\
 &= \sum_{\ell,k=0}^L \int_{\Omega} \nabla w_\ell \nabla w_k d\vec{x} \\
 &\stackrel{(6.20)}{\leq} C \sum_{\ell,k=0}^L 2^{-|\ell-k|/2} |w_\ell|_{H^1(\Omega)} |w_k|_{H^1(\Omega)} \\
 &\leq C \sum_{\ell=0}^L \left( \sum_{k=0}^L 2^{-|\ell-k|/2} |w_k|_{H^1(\Omega)} \right) |w_\ell|_{H^1(\Omega)} \\
 &\stackrel{C.S.}{\leq} C \left\{ \sum_{\ell=0}^L \left( \sum_{k=0}^L 2^{-|\ell-k|/2} |w_k|_{H^1(\Omega)} \right)^2 \right\}^{1/2} \left\{ \sum_{\ell=0}^L |w_\ell|_{H^1(\Omega)}^2 \right\}^{1/2} \\
 &\stackrel{(6.18)}{\leq} C \left\{ \sum_{\ell=0}^L |w_\ell|_{H^1(\Omega)}^2 \right\}^{1/2} \left\{ \sum_{\ell=0}^L |w_\ell|_{H^1(\Omega)}^2 \right\}^{1/2} \\
 &= c \sum_{\ell=0}^L |w_\ell|_{H^1(\Omega)}^2 \\
 &\stackrel{(6.17)}{\leq} C_2 \sum_{\ell=0}^L (\vec{w}_\ell, \vec{w}_\ell)_{C_\ell}.
 \end{aligned} \tag{6.21}$$

Recall that, since decomposition are unique

$$\begin{aligned}\lambda_{\max}(C_H A_L) &\stackrel{6.5}{=} \max_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{(\vec{u}_L, \vec{u}_L)_{A_L}}{\sum_{\ell=0}^L (\vec{w}_\ell, \vec{w}_\ell)_{C_\ell}} \\ &\stackrel{6.21}{=} \max_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{C_2 \sum_{\ell=0}^L (\vec{w}_\ell, \vec{w}_\ell)_{C_\ell}}{\sum_{\ell=0}^L (\vec{w}_\ell, \vec{w}_\ell)_{C_\ell}} \\ &\leq C_2.\end{aligned}$$

□

**Theorem 6.25.** *There is a constant  $C > 0$  independent of  $L$ , such that*

$$\kappa(C_H A_L) = \frac{\lambda_{\max}(C_H A_L)}{\lambda_{\min}(C_H A_L)} \leq C(1 + |\ln h_L|^2).$$

*independent of  $L$ .*

*Proof.* Follows from Lemma 6.20 and 6.23

□

## 6.2 The BPX Preconditioner

For this method we choose

$$W_\ell := V_\ell, \quad 0 \leq \ell \leq L.$$

Thus

$$W_L = V_L$$

and

$$m_\ell = n_\ell, \quad 0 \leq \ell \leq L.$$

**Definition 6.26.** *Define the operator  $C_\ell : V_\ell \rightarrow V'_\ell$  via*

$$C_\ell[v_{\ell,1}](v_{\ell,2}) = \sum_{i=1}^{n_\ell} w_{\ell,1}(\vec{N}_{\ell,i}^W) w_{\ell,2}(\vec{N}_{\ell,i}^W)$$

*The matrix  $C_\ell \in \mathbb{R}^{m_\ell \times m_\ell}$  is defined as*

$$[C_\ell]_{jk} = C_\ell[\phi_{\ell,j}](\phi_{\ell,k}) = \delta_{jk}, \quad 1 \leq j, k \leq n_\ell,$$

*where  $\mathcal{B}_\ell = \{\phi_{\ell,j}\}_{j=1}^{n_\ell}$  is the Lagrange nodal basis for the piecewise linear FE space  $V_\ell, 0 \leq \ell \leq L$ . The BPX preconditioner is*

$$C_{BPX} := \sum_{\ell=0}^L P_\ell^L C_\ell^{-1} \mathcal{R}_\ell^L = \sum_{\ell=0}^L P_\ell^L \mathcal{R}_\ell^L, \quad (6.22)$$

*where  $P_\ell^L \in \mathbb{R}^{n_L \times n_\ell}$  is the standard prolongation matrix from Chap. 5 and  $\mathcal{R}_\ell^L = (P_\ell^L)^T$ .*

**Lemma 6.27.** *Assumption (S1) holds for the BPX framework, i.e., for every  $u_L \in V_L$ , there exists  $v_\ell \in V_\ell, 0 \leq \ell \leq L$ , such that*

$$u_L = \sum_{\ell=0}^L v_\ell,$$

*or, equivalently*

$$\vec{u}_L = \sum_{\ell=0}^L P_\ell^l \vec{v}_\ell,$$

*with*

$$V_\ell \ni v_\ell \xleftrightarrow{B_\ell} \vec{v}_\ell \in \mathbb{R}^{n_\ell},$$

*and*

$$V_L \ni u_L \xleftrightarrow{B_\ell} \vec{u}_L \in \mathbb{R}^{n_L}.$$

*Proof.* This is trivial because of the nestedness of the the spaces

$$V_0 \subset V_1 \subset V_2 \subset \cdots \subset V_{L-1} \subset V_L.$$

□

**Remark 6.28.** *Note that the decomposition is no longer unique.*

**Lemma 6.29.** *For any  $v_j \in V_j, v_\ell \in V_\ell$ ,*

$$\int_{\Omega} \nabla v_j \nabla v_\ell d\vec{x} \leq C 2^{-|j-\ell|/2} \left( h_j^{-1} \|v_j\|_{L^2(\Omega)} \right) \left( h_\ell^{-1} \|v_\ell\|_{L^2(\Omega)} \right), \quad (6.23)$$

*for some  $C > 0$ .*

*Proof.* This follows from (6.19) and the inverse inequality

$$\|v_j\|_{H^1(\Omega)} \leq c h_j^{-1} \|v_j\|_{L^2(\Omega)}.$$

□

**Lemma 6.30.** *For some  $C_2 > 0$  that is independent of  $L$ ,*

$$\lambda_{\max}(B_{BPX} A_L) \leq C_2.$$

*for some  $C > 0$ .*

*Proof.* Let  $u_L \in V_L$  be arbitrary. There exists  $v_\ell \in V_\ell, 0 \leq \ell \leq L$ , such that

$$u_L = \sum_{\ell=0}^L v_\ell,$$

or

$$\vec{u}_L = \sum_{\ell=0}^L P_\ell^l \vec{v}_\ell.$$

The decomposition is not unique, however. Then

$$\begin{aligned}
 (\vec{u}_L, \vec{u}_L)_{A_L} &= (\vec{u}_L, A_L \vec{u}_L) \\
 &= a(\vec{u}_L, \vec{u}_L) \\
 &= a\left(\sum_{j=0}^L v_j, \sum_{\ell=0}^L v_\ell\right) \\
 &= \sum_{\ell, j=0}^L a(v_j, v_\ell) \\
 &\stackrel{(6.23)}{\leq} C \sum_{\ell, j=0}^L 2^{-|j-\ell|/2} h_j^{-1} \|v_\ell\|_{L^2(\Omega)} h_\ell \|v_k\|_{L^2(\Omega)} \\
 &\stackrel{(6.18)}{\leq} C \sum_{j=0}^L h_j^{-2} \|v_j\|_{L^2(\Omega)}^2 \\
 &\stackrel{(5.4)}{\leq} C_2 \sum_{j=0}^L (\vec{v}_j, \vec{v}_j)_{C_j} \\
 &= C_2 \sum_{j=0}^L (C_j \vec{v}_j, \vec{v}_j)_j
 \end{aligned}$$

Now,

$$\begin{aligned}
 \lambda_{\max}(C_{BPX} A_L) &\stackrel{6.5}{=} \max_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{(\vec{u}_L, \vec{u}_L)_{A_L}}{\min_{\substack{\vec{u}_L = \sum_{\ell=0}^L p_\ell^L \vec{v}_\ell \\ \vec{v}_\ell \in \mathbb{R}_*^{n_\ell}}} \sum_{\ell=0}^L (\vec{u}_\ell'', \vec{u}_\ell'')_{C_\ell}} \\
 &\leq \max_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{C_2 \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell}{\min_{\vec{v}_\ell''} \sum_{\ell=0}^L (C_\ell \vec{w}_\ell, \vec{w}_\ell)_\ell} \\
 &\leq C_2.
 \end{aligned}$$

Recall that the minimum was achievable, so we could take  $\vec{v}_\ell = \vec{v}_\ell''$ . see the proof of Theorem.6.4.  $\square$

**Lemma 6.31.** *There is a constant  $C_1 > 0$  that is independent of  $L$ , such that*

$$\lambda_{\min}(B_{BPX} A_L) \geq C_1.$$

*for some  $C > 0$ .*

*Proof.* Let  $u_L \in V_L$  be arbitrary. Set

$$v_\ell =: \mathcal{R}_\ell u_L - R_{\ell-1} u_L, \quad 0 \leq \ell \leq L,$$

where  $\mathcal{R}_\ell : H_0^1(\Omega) \rightarrow V_\ell$  is the Ritz projection for  $0 \leq \ell \leq L$  and  $R_{-1} \equiv 0$ . Since

$$\mathcal{R}_\ell u_L = u_L,$$

it follows that

$$u_L = \sum_{\ell=0}^L v_\ell \xleftrightarrow{\mathcal{B}_\ell} \vec{u}_\ell = \sum_{\ell=0}^L P_\ell^L v_\ell.$$

Moreover,

$$a(v_j, v_\ell) = 0, \quad 0 \leq j \neq \ell \leq L. \quad (6.24)$$

To see this, recall that, in general,

$$a(R_j u_L, v'_j) = a(u_L, v'_j), \quad \forall v'_j \in V_j.$$

Suppose  $j < \ell$ , for definiteness. Then

$$a(R_j u_L, v'_\ell) = a(u_L, v'_\ell), \quad \forall v'_\ell \in V_\ell.$$

In particular, since

$$v_j := R_j u_L - R_{j-1} u_L \in V_j \subset V_\ell,$$

and

$$a(\mathcal{R}_\ell u_L, v_j) = a(u_L, v_j),$$

likewise

$$a(R_{\ell-1} u_L, v_j) = a(u_L, v_j),$$

Subtracting, we have

$$a(\mathcal{R}_\ell u_L - R_{\ell-1} u_L, v_j) = 0$$

To make further progress, let us assume that  $\Omega$  is convex. Then the standard regularity condition holds. And, for  $1 \leq \ell \leq L$

$$\begin{aligned} h_\ell^{-2} \|v_\ell\|_{L^2(\Omega)}^2 &= h_\ell^{-2} \|\mathcal{R}_\ell u_L - R_{\ell-1} u_L\|_{L^2(\Omega)}^2 \\ &= h_\ell^{-2} \|\mathcal{R}_\ell u_L - R_{\ell-1} \mathcal{R}_\ell u_L\|_{L^2(\Omega)}^2 \\ &\stackrel{5.13}{\leq} C h_\ell^{-2} h_\ell^2 |\mathcal{R}_\ell u_L - R_{\ell-1} \mathcal{R}_\ell u_L|_{H^1(\Omega)}^2 \\ &= C |\mathcal{R}_\ell u_L - R_{\ell-1} \mathcal{R}_\ell u_L|_{H^1(\Omega)}^2 \\ &= C |v_\ell|_{H^1(\Omega)}^2. \end{aligned} \quad (6.25)$$

To see that  $R_{\ell-1} = R_{\ell-1} \mathcal{R}_\ell$ , let  $u \in H_0^1(\Omega)$  be arbitrary. Then

$$a(R_{\ell-1}(\mathcal{R}_\ell u), v'_{\ell-1}) = a(\mathcal{R}_\ell u, v'_{\ell-1}), \quad \forall v'_{\ell-1} \in V_{\ell-1}.$$

But,

$$a(\mathcal{R}_\ell u, v'_{\ell-1}) = a(u, v'_{\ell-1}), \quad \forall v'_{\ell-1} \in V_{\ell-1}.$$

Since

$$a(\mathcal{R}_\ell u, v'_\ell) = a(u, v'_\ell), \quad \forall v'_\ell \in V_\ell,$$

and

$$V_{\ell-1} \subset V_\ell.$$

But

$$a(R_{\ell-1} u, v'_{\ell-1}) = a(u, v'_{\ell-1}), \quad \forall v'_{\ell-1} \in V_{\ell-1}.$$

Hence

$$a(R_{\ell-1}(\mathcal{R}_\ell u), v'_{\ell-1}) = a(R_{\ell-1} u, v'_{\ell-1}), \quad \forall v'_{\ell-1} \in V_{\ell-1}.$$

And we conclude that  $R_{\ell-1} = R_{\ell-1} \mathcal{R}_\ell$  since

$$R_{\ell-1}(\mathcal{R}_\ell u), R_{\ell-1} u \in V_{\ell-1}.$$

Estimate (6.24) holds trivially for  $\ell = 0$ . Finally,

$$\begin{aligned} \sum_{\ell=0}^L (C_\ell \vec{v}_\ell, \vec{v}_\ell)_\ell &\stackrel{(5.4)}{\leq} C \sum_{\ell=0}^L h_\ell^{-2} \|v_\ell\|_{L^2(\Omega)}^2 \\ &\stackrel{(6.25)}{\leq} C_1^{-1} \sum_{\ell=0}^L |v_\ell|_{H^1(\Omega)}^2 \\ &\stackrel{(6.24)}{=} C_1^{-1} |u_L|_{H^1(\Omega)}^2. \end{aligned} \tag{6.26}$$

Finally,

$$\begin{aligned} \lambda_{\min}(C_{BPX} A_L) &= \min_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{(\vec{u}_L, \vec{u}_L)_{A_L}}{\min_{\vec{u}_L = \sum_{\ell=0}^L P_\ell^L \vec{v}_\ell} \sum_{\ell=0}^L (\vec{u}_\ell^\vee, \vec{u}_\ell^\vee)_{C_\ell}} \\ &\geq \min_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{(A_L \vec{u}_L, \vec{u}_L)_L}{\min_{\vec{v}_\ell} \sum_{\ell=0}^L (C_\ell \vec{v}_\ell, \vec{v}_\ell)} \\ &\geq \min_{\vec{u}_L \in \mathbb{R}_*^{n_L}} \frac{(A_L \vec{u}_L, \vec{u}_L)_L}{C_1^{-1} |u_L|_{H^1(\Omega)}} \\ &= C_1. \end{aligned}$$

□

**Theorem 6.32.**

$$\kappa(B_{BPX} A_L) = \frac{\lambda_{\max}(B_{BPX} A_L)}{\lambda_{\min}(B_{BPX} A_L)} \leq \frac{C_2}{C_1}.$$

*Proof.* Follows from Lemma 6.30 and 6.31. □



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