

**Supplement to
 SUMMATION BY PARTS, PROJECTIONS, AND STABILITY. I**

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Let $P = I - \Sigma^{-1}L(L^T\Sigma^{-1}L)^{-1}L^T$ where $L = (L_{11} \dots L_{1r} \ L_{0\chi} \ L_{21} \dots L_{2r})$ represents homogeneous Neumann conditions (localized to the origin), cf. section on the heat equation. For convenience we have $\chi = 0.5$ and $h_1 = h_2 = 1$. We shall show that $L_{10}^T P = 0$, which will follow if we can prove that $L_{10}^T \Sigma^{-1}L(L^T\Sigma^{-1}L)^{-1}L^T = L_{10}^T$. Straightforward computations show that

$$L^T \Sigma^{-1} L = \begin{pmatrix} D_{11} & D_{12} & D_{13} \\ D_{12}^T & D_{22} & D_{23} \\ D_{13}^T & D_{23}^T & D_{33} \end{pmatrix},$$

where

$$D_{11} = D_{33} = \kappa \begin{pmatrix} \frac{1}{\sigma_1} & & \\ & \ddots & \\ & & \frac{1}{\sigma_r} \end{pmatrix}, \quad D_{12} = D_{23}^T = \tau^{-1} \begin{pmatrix} \frac{d_{01}}{\sigma_1} \\ \vdots \\ \frac{d_{0r}}{\sigma_r} \end{pmatrix},$$

and

$$D_{13} = \tau^2 D_{12} D_{23}, \quad D_{22} = \frac{1}{2} \left(\frac{\kappa}{\sigma_0} + \frac{d_{00}^2}{\sigma_0^2} \right), \quad \kappa = \sum_{k=0}^r \frac{d_{0k}^2}{\sigma_k}, \quad \tau = \frac{2\sigma_0}{d_{00}}.$$

The inverse is given by

$$(L^T \Sigma^{-1} L)^{-1} = \begin{pmatrix} T_{11} & T_{12} & T_{13} \\ T_{12}^T & T_{22} & T_{23} \\ T_{13}^T & T_{23}^T & T_{33} \end{pmatrix}$$

with

$$T_{11} = T_{33} = D_{11}^{-1} + \frac{\mu}{1 - \mu\sigma} D_{11}^{-1} D_{12} D_{12}^T D_{11}^{-1},$$

$$T_{12} = T_{23}^T = \frac{-(\mu - \sigma\tau^4)}{(1 - \mu\sigma)(1 - \sigma\tau^2)} D_{11}^{-1} D_{12},$$

$$T_{13} = T_{13}^T = \frac{\mu - \tau^2}{(1 - \mu\sigma)(1 - \sigma\tau^2)} D_{11}^{-1} D_{12} D_{23} D_{33}^{-1},$$

$$T_{22} = \frac{(\mu - \sigma\tau^4)(1 + \sigma\tau^2)}{(1 - \mu\sigma)(1 - \sigma\tau^2)},$$

and

$$\begin{aligned} \sigma &= D_{12}^T D_{11}^{-1} D_{12}, \\ \mu &= \sigma\tau^4 + \frac{1}{\nu} (1 - \sigma\tau^2)^2, \\ \nu &= D_{22} - \sigma. \end{aligned}$$

Let $\tilde{L} = (L_{10} \dots L_{1r} \ L_{20} \dots L_{2r})$. Then

$$L_{10}^T \Sigma^{-1} L (L^T \Sigma^{-1} L)^{-1} L^T = L_{10}^T \Sigma^{-1} \tilde{L} \begin{pmatrix} R & S \\ S & R \end{pmatrix} \tilde{L}^T,$$

where (using $T_{12}^T = T_{23}$ and $T_{11} = T_{33}$)

$$R = \begin{pmatrix} T_{22}/4 & T_{12}^T/2 \\ T_{12}/2 & T_{13} \end{pmatrix}, \quad S = \begin{pmatrix} T_{22}/4 & T_{12}^T/2 \\ T_{12}/2 & T_{13} \end{pmatrix}.$$

Furthermore,

$$L_{10}^T \Sigma^{-1} \tilde{L} = (\kappa/\sigma_0 \ 0 \ d_{60}^2/\sigma_0^2 \ 2D_{12}^T).$$

Using $D_{22} = (\kappa/\sigma_0 + d_{60}^2/\sigma_0^2)/2$, we have

$$L_{10}^T \Sigma^{-1} \tilde{L} \begin{pmatrix} R & S \\ S & R \end{pmatrix} =$$

$$\begin{pmatrix} \frac{1}{2} D_{22} T_{22} + D_{12}^T T_{12} & D_{22} T_{12}^T + 2D_{12}^T T_{13} & \frac{1}{2} D_{22} T_{22} + D_{12}^T T_{12} & D_{22} T_{12}^T + 2D_{12}^T T_{13} \end{pmatrix}.$$

But

$$\frac{1}{2} D_{22} T_{22} + D_{12}^T T_{12} = \frac{1}{2} (D_{12}^T T_{12} + D_{22} T_{22} + D_{23} T_{33}) = \frac{1}{2},$$

and

$$D_{22} T_{12}^T + 2D_{12}^T T_{13} = 2(D_{12}^T T_{13} + D_{22} T_{12}^T + D_{12}^T T_{11}) - (D_{23} T_{12}^T + 2D_{12}^T T_{11}).$$

Observing that $T_{12}^T = T_{23}$, $D_{12}^T T_{11} = D_{23} T_{33}$, we conclude that the first parenthetical expression vanishes. Thus,

$$L_{10}^T \Sigma^{-1} \tilde{L} \begin{pmatrix} R & S \\ S & R \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & D_{22} T_{12}^T + 2D_{12}^T T_{13} & \frac{1}{2} & - (D_{22} T_{12}^T + 2D_{12}^T T_{13}) \end{pmatrix}.$$

Also,

$$D_{22} T_{12}^T + 2D_{12}^T T_{13} = \rho D_{12}^T D_{11}^{-1}, \quad \rho = \frac{2\sigma(\mu - \tau^2) - (\sigma + \nu)(\mu - \sigma\tau^2)}{(1 - \mu\sigma)(1 - \sigma\tau^2)}.$$

Substituting $1 - \sigma\tau^2 = d_{60}^2/(\kappa\sigma_0)$ and the expression for μ yields

$$\rho = -\frac{\kappa\sigma_0}{d_{60}^2}.$$

and so

$$\rho D_{12}^T D_{11}^{-1} = -\frac{1}{2d_{60}} (d_{01} \dots d_{0r}).$$

Consequently,

$$L_{10}^T \Sigma^{-1} L (L^T \Sigma^{-1} L)^{-1} L^T = \frac{1}{2d_{60}} (d_{00} \ -d_{01} \ \dots \ -d_{0r} \ d_{00} \ d_{01} \ \dots \ d_{0r}) \tilde{L}^T. \quad (3)$$

The j th block column of \tilde{L}^T reads

$$c_j^T = \begin{pmatrix} 0 & \dots & 0 & \sum_{k=0}^r d_{0k} c_k^T & 0 & \dots & 0 & d_{0j} c_0^T & \dots & d_{0r} c_r^T \end{pmatrix},$$

where the sum is the j th block element of c_j , $j = 0, \dots, r$. Hence,

$$\frac{1}{2d_{60}} (d_{00} \ -d_{01} \ \dots \ -d_{0r} \ d_{00} \ d_{01} \ \dots \ d_{0r}) c_j = \delta_{j0} \sum_{k=0}^r d_{0k} c_k^T,$$

i. e.,

$$L_{10}^T \Sigma^{-1} L (L^T \Sigma^{-1} L)^{-1} L^T = L_{10}^T.$$

In a similar fashion one shows that L_{20} also satisfies the above equation.

The simplest example is obtained by discretizing the Neumann conditions using the standard divided difference D_+ in both coordinate directions, i. e.,

$$\begin{aligned} v_{11} - v_{01} &= 0, \\ v_{11} - v_{10} &= 0, \\ (v_{10} - v_{00})/2 + (v_{01} - v_{00})/2 &= 0, \end{aligned} \quad (1)$$

which leads to

$$\Sigma^{-1} L (L^T \Sigma^{-1} L)^{-1} L^T = \frac{1}{18} \begin{pmatrix} 16 & -4 & 0 & \dots & 0 & -4 & -8 & 0 & \dots & 0 \\ -2 & 14 & 0 & \dots & 0 & -4 & -8 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 0 & 0 & 0 & 0 & \dots & 0 \\ -2 & -4 & 0 & \dots & 0 & 14 & -8 & 0 & \dots & 0 \\ -2 & -4 & 0 & \dots & 0 & -4 & 10 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 & 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 & 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 & 0 & 0 & 0 & \dots & 0 \end{pmatrix}. \quad (2)$$

Evidently, any vector $v = Pu$, $P = I - \Sigma^{-1} L (L^T \Sigma^{-1} L)^{-1} L^T$, satisfies (1). Furthermore, by (2), $v_{10} - v_{00} = v_{01} - v_{00} = 0$, which also follows directly from (1).

We conclude this Supplement by proving a number of technical lemmas that were used in the paper. Let D be a difference operator satisfying a summation-by-parts rule with respect to the scalar product $(u, v)_h = hu^T \Sigma v$, where

Proposition 1.1. Let D be as above, and define the norm $\|\cdot\|_h = \sqrt{\langle \cdot, \cdot \rangle}_h$. Then

$$\|v\|_\infty^2 \leq c \|Dv\|_h^2 + (c^{-1} + 1 + \mathcal{O}(h)) \|v\|_h^2,$$

where $c > 0$.

Proof. Choose k, l such that

$$\begin{aligned} |v_k|^2 &= \min_j (|v_j|^2), \\ |v_l|^2 &= \max_j (|v_j|^2) \equiv |v|_\infty^2. \end{aligned}$$

Eq. (3) implies that

$$\|v\|_h^2 \geq h \left(\lambda_1 |v^{(1)}|^2 + \lambda_2 |v^{(2)}|^2 \right) + h \sum_{j=r_1+1}^{r_2} |v_j|^2,$$

where $\lambda_{1,2} > 0$ are the smallest eigenvalues of $\Sigma^{(1,2)}$. Note that $\lambda_{1,2}$ are independent of h . Hence,

$$\|v\|_h^2 \geq (1 - h(r_1(1 - \lambda_1) + r_2(1 - \lambda_2))) |v_k|^2,$$

where we have used $h\nu = L = 1$. If $c \equiv r_1(1 - \lambda_1) + r_2(1 - \lambda_2) \leq 0$, one immediately gets $|v_k|^2 \leq \|v\|_h^2$. Otherwise, we choose h such that $hc < 1$. Hence,

$$|v_k|^2 \leq \frac{1}{1 - hc} \|v\|_h^2 \leq (1 + Kh) \|v\|_h^2, \quad K = \frac{c}{1 - h_0 c}. \quad (4)$$

for $h \leq h_0$, where h_0 is a fixed number such that $h_0 c < 1$.

Next, we define a family of norms, which is obtained by shrinking the interior of (3); $\Sigma^{(1,2)}$ remain constant. Allowing a slight abuse of notation, we write these norms as

$$(u, v)_{h,r,s} = h \sum_{j=r}^s \sigma_{ij} u_i^T v_j,$$

where $r \geq 0$ and $s \leq \nu$. Shrinking the interior of D accordingly, one has

$$(v, Dv)_{h,k,l} = |v|^2 - |v_k|^2 - (Dv, v)_{h,k,l},$$

i. e.,

$$|v|^2 \leq |v_k|^2 + 2|Dv|_{h,k,l} \|v\|_{h,k,l}.$$

Obviously, $\|v\|_{h,k,\nu} \leq \|v\|_{h,0,\nu} \equiv \|v\|_h$, whence

$$|v|_\infty^2 \leq c \|Dv\|_h^2 + (c^{-1} + 1 + \mathcal{O}(h)) \|v\|_h^2,$$

where (4) and the standard algebraic inequality have been used. \square

Lemma 1.1. Let $A = \text{diag}(A_0, \dots, A_\nu)$, $A_j = A(jh) \in \mathbb{R}^d$. Then

$$\|(u, Av)_h - (A^T u, v)_h\| \leq \mathcal{O}(h) \|u\|_h \|v\|_h.$$

Proof. Denote the commutator of Σ and A by $[\Sigma, A]$. Then

$$(u, Av)_h = (A^T u, v)_h + h u^T [\Sigma, A] v,$$

where

$$[\Sigma, A] = \begin{pmatrix} [\Sigma^{(1)}, A^{(1)}] & & \\ & 0 & \\ & & [\Sigma^{(2)}, A^{(2)}] \end{pmatrix}$$

with

$$[\Sigma^{(1)}, A^{(1)}] = \begin{pmatrix} 0 & \sigma_{0l}(A_1 - A_0) & \dots & \sigma_{0r}(A_r - A_0) \\ -\sigma_{0l}(A_1 - A_0) & 0 & \dots & \sigma_{1r}(A_r - A_1) \\ \vdots & \vdots & \ddots & \vdots \\ -\sigma_{0r}(A_r - A_0) & -\sigma_{1r}(A_r - A_1) & \dots & 0 \end{pmatrix}.$$

The other nonzero block has a similar structure. Assuming that $A(x)$ is differentiable, we can apply the mean value theorem:

$$[\Sigma^{(1)}, A^{(1)}] = h \begin{pmatrix} 0 & \sigma_{0l} A'_{l0} & \dots & \sigma_{0r} A'_{r0} \\ -\sigma_{0l} A'_{l0} & 0 & \dots & \sigma_{1r}(r-1) A'_{r1} \\ \vdots & \vdots & \ddots & \vdots \\ -\sigma_{0r} A'_{r0} & -\sigma_{1r}(r-1) A'_{r1} & \dots & 0 \end{pmatrix}.$$

Hence,

$$|(u, Av)_h - (A^T u, v)_h| \leq c |A'|_\infty h^2 (|u^{(1)}| |v^{(1)}| + |u^{(2)}| |v^{(2)}|) \leq \mathcal{O}(h) \|u\|_h \|v\|_h,$$

which proves the lemma. \square

Lemma 1.2. Let A be as in the previous lemma. Then

$$|(u, Av)_h| \leq |A|_\infty (1 + \mathcal{O}(h)) \|v\|_h \|u\|_h.$$

where $|A|_\infty = \sup |A(x)|$.

Proof. The definition of $(\cdot, \cdot)_h$ implies that $(u, Av)_h = h \tilde{u}^T \tilde{A} \tilde{v}$, where $\tilde{u} = \Sigma^{1/2} u$, $\tilde{v} = \Sigma^{1/2} v$, and $\tilde{A} = \Sigma^{1/2} A \Sigma^{-1/2}$. Taylor expansion yields $\tilde{A} = A + R$,

$$R = \begin{pmatrix} R^{(1)} & \\ & 0 \\ & & R^{(2)} \end{pmatrix}$$

with $R^{(l)} = \mathcal{O}(h)$, $l = 1, 2$. Thus,

$$|(u, Av)_h| \leq |A|_\infty \|\tilde{u}\| \|\tilde{v}\| + \mathcal{O}(h) (\|\tilde{u}^{(1)}\| \|\tilde{v}^{(1)}\| + \|\tilde{u}^{(2)}\| \|\tilde{v}^{(2)}\|),$$

where we have used $[\Sigma^{-1/2}D_s\Sigma^{-1/2}, A] = 0$. Since D_s is anti-symmetric and A symmetric, we have $C^T = C$, i. e.,

$$\| [D_s A] \|_k^2 = \max_{\|w\|=1} h w^T C^2 w = \rho(C)^2.$$

Finally, $C = \Sigma^{1/2}[D_s A]\Sigma^{-1/2}$ implies that

$$\| [D_s A] \|_k = \rho([D_s A]),$$

which proves the lemma. \square

i. e., $\| (u, A^r)_k \| \leq (1+A)^{r-1} \mathcal{O}(h) \| \tilde{u} \| \| \tilde{r} \|$, where $\| \cdot \|$ denotes the standard Euclidean norm. Since $\| \tilde{u} \| = \| u \|_k$, $\| \tilde{r} \| = \| r \|_k$, the lemma follows. \square

Corollary 1.1. *If, in addition to the hypotheses of Lemma 1.2, one of the following conditions holds:*

- (i) Σ is diagonal $\text{diag}(\sigma_0 I \dots \sigma_r I)$,
- (ii) The blocks of A satisfy $A_0 = \dots = A_{r_1}$ and $A_{p-r_2} = \dots = A_p$,

then
$$\| (u, A^r)_k \| \leq |A|_\infty \| u \|_k \| r \|_k.$$

Proof. The hypotheses imply that $\tilde{A} = A$, and the corollary follows. \square

Lemma 1.3. *Let A be as in Lemma 1.1 and assume that one of the following conditions holds:*

- (i) Σ is diagonal $\text{diag}(\sigma_0 I \dots \sigma_r I)$,
- (ii) The blocks of A satisfy $A_0 = \dots = A_{r_1}$ and $A_{p-r_2} = \dots = A_p$.

If A is symmetric, then

$$(u, [D_s A]^r)_k \leq \rho([D_s A]) \| u \|_k \| r \|_k.$$

Proof. According to the definition of the operator norm we have

$$\| [D_s A] \|_k^2 = \max_{\|w\|=1} \| [D_s A] r \|_k^2 = \max_{\|w\|=1} h w^T C^T C w,$$

where $w = \Sigma^{1/2}v$, $C = \Sigma^{1/2}[D_s A]\Sigma^{-1/2}$. Because of the assumptions on A (or Σ) we have $C = DA - AD$, where $D = \Sigma^{1/2}D_s\Sigma^{-1/2}$. Summation by parts implies that

$$\Sigma D = D_s + D_s, \quad D_s = \frac{1}{2} \begin{pmatrix} -I & & \\ & 0 & \\ & & I \end{pmatrix}, \quad I \in \mathbb{R}^{d \times d},$$

and D_s is an anti-symmetric matrix. Consequently,

$$C = [\Sigma^{-1/2}D_s\Sigma^{-1/2}, A],$$