

## Convergence result for the method of steepest descent

Since  $A$  is self-adjoint w.r.t.  $(\cdot, \cdot)$ , the bilinear form

$$(\mathbf{y}, \mathbf{z})_A := (A\mathbf{y}, \mathbf{z})$$

is an inner product. The corresponding norm  $\|\mathbf{y}\|_A := \sqrt{(\mathbf{y}, \mathbf{y})_A}$  is called **energy norm**.

**Lemma 1.58** The method of steepest descent converges  $q$ -linearly with

$$\|\mathbf{x}_{k+1} - \mathbf{x}\|_A \leq q \|\mathbf{x}_k - \mathbf{x}\|_A$$

where  $q = \frac{\kappa-1}{\kappa+1}$  and  $\kappa = \frac{\mu_2}{\mu_1}$  for  $\mu_1 \leq \lambda_{\min}(A)$ ,  $\mu_2 \geq \lambda_{\max}(A)$ .

*Proof:* (1) Auxiliary identity for arbitrary  $y \in \mathbb{R}^n$ :

$$\begin{aligned} \|y - x\|_A^2 &= (Ay, y) - 2(Ax, y) + (Ax, x) = (Ay, y) - 2(\underbrace{Ax}_{=b}, y) - (Ax, x) + 2(\underbrace{Ax}_{=b}, x) \\ &= (Ay, y) - 2(b, y) - [(Ax, x) - 2(b, x)] = 2[J_A(y) - J_A(x)] \end{aligned}$$

(2) After  $k$  steps in the method of steepest descent, we perform one hypothetical step of Richardson's method:

$$\tilde{x}_{k+1} := x_k + \tau r_k = x_k + \tau p_k$$

with the optimal parameter  $\tau$ .

(3) We use Lemma 1.47 to analyze the hypothetical Richardson step.

It is easily seen that  $A$  is self-adjoint w.r.t.  $(\cdot, \cdot)_A$  and that

$$\begin{aligned} \mu_1(y, y) \leq (Ay, y) \leq \mu_2(y, y) \quad \forall y \in \mathbb{R}^n \\ \stackrel{\text{Lemma 1.49}}{\iff} \mu_1 \leq \lambda_{\min}(A), \quad \mu_2 \geq \lambda_{\max}(A) \\ \iff \mu_1(y, y)_A \leq (Ay, y)_A \leq \mu_2(y, y)_A \quad \forall y \in \mathbb{R}^n \end{aligned}$$

Hence, Lemma 1.47 (with  $\|\cdot\| \mapsto \|\cdot\|_A$ ) yields:

$$\|\tilde{x}_{k+1} - x\|_A \leq q \|x_k - x\|_A \quad \text{where} \quad q = \frac{\kappa - 1}{\kappa + 1} \quad \text{and} \quad \kappa = \frac{\mu_2}{\mu_1}.$$

(4) Recall that  $x_{k+1}$  is the next step in the method of steepest descent. Because of the optimal choice of  $\alpha_k$ , we have  $J_A(x_{k+1}) \leq J_A(\tilde{x}_{k+1})$ . Therefore, with the identity of Part (1):

$$\begin{aligned} \|x_{k+1} - x\|_A^2 &= 2[J_A(x_{k+1}) - J_A(x)] \\ &\leq 2[J_A(\tilde{x}_{k+1}) - J_A(x)] = \|\tilde{x}_{k+1} - x\|_A^2 \leq q^2 \|x_k - x\|_A^2 \end{aligned}$$

This concludes the proof. □

### Remark 1.59

1. The method of steepest descent converges as fast as Richardson's method but does not need a-priori information  $\mu_1$ ,  $\mu_2$  or the choice of the damping parameter  $\tau$ . The (optimal) stepsize  $\alpha_k$  is computed automatically using just  $r_k$  and  $A$ .
2. One can show that the method of steepest descent converges in general not faster than Richardson's method (with  $\tau = \tau_{\text{opt}}$ ).
3. The statement of Lemma 1.58 only holds for the energy norm  $\|\cdot\|_A$ !