

Exercise 19. We consider  $f: \mathbb{R}^m \times \mathbb{R}^m \rightarrow \mathbb{R}$  given for a matrix  $B \in \mathbb{R}^{m \times m}$  via

$$f(u, A) = \frac{1}{2} |A|^2 + (Bu) \cdot A + \frac{1}{2} |u|^2$$

$B$  is not symmetric!

$f$  is quadratic, i.e.,  $f(u, A) = \frac{1}{2} \begin{pmatrix} u \\ A \end{pmatrix} \cdot M \begin{pmatrix} u \\ A \end{pmatrix}$

where  $M \in \mathbb{R}^{2m \times 2m}$ ,  $M = \begin{pmatrix} I & B^T \\ B & I \end{pmatrix}$ .

a) We show that  $f$  convex  $\Leftrightarrow \|B\| \leq 1$

Let  $\lambda \in ]0, 1[$  and fix  $(u_0, A_0), (u_1, A_1)$ , then for  $u_\lambda = (1-\lambda)u_0 + \lambda u_1$ ,  $A_\lambda = (1-\lambda)A_0 + \lambda A_1$ , we have (since  $f$  is quadratic)

$$\begin{aligned} f(u_\lambda, A_\lambda) &= (1-\lambda)^2 f(u_0, A_0) + \lambda^2 f(u_1, A_1) \\ &\quad + \frac{(1-\lambda)\lambda}{2} \left( \begin{pmatrix} u_0 \\ A_0 \end{pmatrix} M \begin{pmatrix} u_1 \\ A_1 \end{pmatrix} + \begin{pmatrix} u_1 \\ A_1 \end{pmatrix} M \begin{pmatrix} u_0 \\ A_0 \end{pmatrix} \right) \end{aligned}$$

$$= (1-\lambda) f(u_0, A_0) + \lambda f(u_1, A_1)$$

$$+ \underbrace{(1-2\lambda + \lambda^2 - 1 + \lambda)}_{= \lambda(\lambda-1)} f(u_0, A_0)$$

$$+ \underbrace{(\lambda^2 - \lambda)}_{= \lambda(\lambda-1)} f(u_1, A_1)$$

$$+ \frac{(1-\lambda)\lambda}{2} \left( \begin{pmatrix} u_0 \\ A_0 \end{pmatrix} M \begin{pmatrix} u_1 \\ A_1 \end{pmatrix} + \begin{pmatrix} u_1 \\ A_1 \end{pmatrix} M \begin{pmatrix} u_0 \\ A_0 \end{pmatrix} \right)$$

$$= (1-\lambda) f(u_0, A_0) + \lambda f(u_1, A_1)$$

$$+ \frac{(1-\lambda)\lambda}{2} \left( \begin{pmatrix} u_0 \\ A_0 \end{pmatrix} M \begin{pmatrix} u_1 - u_0 \\ A_1 - A_0 \end{pmatrix} + \begin{pmatrix} u_1 \\ A_1 \end{pmatrix} M \begin{pmatrix} u_0 - u_1 \\ A_0 - A_1 \end{pmatrix} \right)$$

$$= (1-\lambda) f(u_0, A_0) + \lambda f(u_1, A_1)$$

$$- \frac{(1-\lambda)\lambda}{2} \left( \begin{pmatrix} u_1 - u_0 \\ A_1 - A_0 \end{pmatrix} M \begin{pmatrix} u_1 - u_0 \\ A_1 - A_0 \end{pmatrix} \right)$$

Thus, we have convexity if and only

if  $(*) \begin{pmatrix} u \\ A \end{pmatrix}^T M \begin{pmatrix} u \\ A \end{pmatrix} \geq 0 \quad \forall (u, A) \in \mathbb{R}^{2n}$

(essentially, we derived the Taylor expansion off and found that the Hessian has to be positive semidefinite)

$(*)$  holds iff  $\forall (u, A)$

$$|u|^2 + |A|^2 + 2(Bu) \cdot A \geq 0$$

$$\Leftrightarrow |Bu| \leq |u| \quad \forall u \in \mathbb{R}^n.$$

why?

b) Let us assume that  $I$  is convex  
if  $\|B - B^T\| \leq 2$

Note  $B = \underbrace{\frac{1}{2}(B + B^T)}_{\text{symmetric part}} + \underbrace{\frac{1}{2}(B - B^T)}_{\text{skew-symmetric part}}$

In particular, if  $B = B^T$  then  $I$  is

convex, even if  $\|B\| > 1$  and hence  $f$  is not convex!

The crucial point is that the symmetric part of  $B$  doesn't contribute to the functional:

Indeed, let  $C = \frac{1}{2}(B + B^T) = \text{sym}(B)$ ,

then

$$\int_0^l C u \cdot u' dx = \int_0^l \frac{1}{2} \frac{d}{dx} (C u \cdot u) dx$$

used symmetry  
here!

$$= (C u \cdot u) \Big|_{x=0}^l = 0$$

boundary  
conditions.

This means that, we have

$$I(u) = \int_0^l \tilde{f}(u, u') dx \text{ with}$$

$$\hat{f}(u, A) = \frac{1}{2}|u|^2 + \frac{1}{2}|A|^2 + \left( \underset{\substack{\uparrow \\ \text{skew-symmetric} \\ \text{part}}}{B_{\text{skew}}}} u \right) \cdot A$$

if  $B_{\text{skew}} = \frac{1}{2}(B - B^T)$  satisfies condition from a), i.e.,  $\|B - B^T\| \leq 2$ , then  $\hat{f}$  is convex and therefore also  $I$ .

