

# Noter til IMO

December 9, 2021

## Kapitel 3

### 0.1 Definitioner, teorem, bemærkninger og propositioner

**Proposition 3.27** The product  $AB$  of two invertible matrices  $A$  and  $B$  is invertible  $(AB)^{-1} = B^{-1}A^{-1}$

**Proposition 3.33** Let  $A$  be an  $m \times r$  matrix and  $B$  an  $r \times n$  matrix. Then

$$(AB)^T = B^T A^T$$

**Definition 3.7.1 Positive definite matrices** En positiv definit matrice er givet ved

$$v^T A v > 0$$

**3.7.2 Positive semi-definite matrices** En positiv semi definit matrice er givet ved

$$v^t A v \leq 0$$

**Proposition 3.41** Let  $A$  be a symmetric  $n \times n$  matrix and  $B$  an invertible  $n \times n$  matrix. Then  $A$  is positive (semi) definite if and only if

$$B^T A B$$

is positive (semi) definite.

### 0.2 Opgaver

**Exercise 3.42** (Overvej kraftigt at bruge 8.26) Beviser at

$$D = \begin{pmatrix} d & 0 \\ 0 & e \end{pmatrix}$$

er positivt definit (ikke semi definit) hvis og kun hvis  $e > 0 \wedge d > 0$

Beviser også at

$$A = \begin{pmatrix} a & c \\ c & b \end{pmatrix}$$

Er positivt definit (ikke semi definit) hvis og kun hvis  $a > 0 \wedge ab - c^2 > 0$

## Kapitel 4

### 0.3 Definitioner, teorem, bemærkninger og propositioner

**Remark 4.6**

$$\max_{x \in C} f(x) = \min_{x \in C} -f(x)$$

**Definition 4.32** Definitionen på en konveks (ikke strengt) er

$$f((1-t)x + ty) \leq (1-t)f(x) + tf(y)$$

for hvert tal  $t$  hvor  $0 \leq t \leq 1$

**Lemma 4.26** Let  $f : \mathbb{R}^d \rightarrow \mathbb{R}$  be a convex function. Then the subset

$$C = \{x \in \mathbb{R}^d \mid f(x) \leq a\}$$

is a convex subset of  $\mathbb{R}^d$ , where  $a \in \mathbb{R}$

### 4.5 Fourier-Motzkin elimination

### 0.4 Opgaver

## Kapitel 5

### 0.5 Definitioner, teorem, bemærkninger og propositioner

**5.2.3 Bounded subsets** A subset  $S \subseteq \mathbb{R}^d$  is called bounded if there exists  $M \in \mathbb{R}$ , such that

$$u \cdot u \leq M$$

for every  $u \in S$

**Theorem 5.19** If  $x_0 \in \mathbb{R}^n$  is a solution to the system

$$(A^T A)x = A^T b$$

of  $n$  linear equations with  $n$  unknowns, then  $x_0$  is an optimal solution to

$$\min_{x \in \mathbb{R}^n} |b - Ax|^2$$

**Cauchy-Schwarz inequality** For two vectors  $u, v \in \mathbb{R}^d$ ,

$$|u \cdot v| \leq |u||v|$$

og

$$-1 \leq \frac{u \cdot v}{|u||v|} \leq 1$$

og

$$\cos(\theta) = \frac{u \cdot v}{|u||v|}$$

**Teorem 5.30** For two vectors  $u, v \in \mathbb{R}^d$ ,

$$|u + v| \leq |u| + |v|$$

og

$$|u - v| = |(u - w) + (w - v)| \leq |u - w| + |w - v|$$

## 0.6 Opgaver