

Compendium MAT260

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1 Preliminaries

1.1 Norms

Definition 1. Let V be a linear space over \mathbb{R} . A function $\|\cdot\| : V \rightarrow \mathbb{R}$ is a norm on V if it satisfies the following:

$$(i) \quad \|\lambda x\| = |\lambda| \|x\| \quad \forall x \in \mathbb{R}$$

$$(ii) \quad x + y \leq \|x\| + \|y\| \quad \forall x, y \in V$$

$$(iii) \quad \|x\| = 0 \Leftrightarrow x = 0$$

Example 1. p - norm: $\|x\|_p = (\sum_{i=1}^n |x_i|^p)^{1/p}$

$$p = 2: \|x\|_2 = (\sum_{i=1}^n |x_i|^2)^{1/2}$$

$$p = \infty: \|x\|_\infty = \max_{i \in \{1, \dots, n\}} |x_i|$$

1.1.1 Matrix norms

A matrix norm has the following properties:

$$(i) \quad \|\lambda A\| = |\lambda| \|A\| \quad \forall \lambda \in \mathbb{R}, \quad \forall A \in \mathbb{R}^{\kappa, \kappa}$$

$$(ii) \quad \|A + B\| \leq \|A\| + \|B\| \quad \forall A, B \in \mathbb{R}^{\kappa, \kappa}$$

$$(iii) \quad \|A\| = 0 \rightarrow A = 0_n$$

$$(iv) \quad \|AB\| \leq \|A\| \|B\| \quad \forall A, B \in \mathbb{R}^{\kappa, \kappa}$$

Example 2. 1. $\|A\|_1 = \sup_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1} = \max_{j=1, \dots, n} \sum_{i=1}^n |a_{ij}|$. Max column sum

$$2. \quad \|A\|_\infty = \sup_{x \neq 0} \frac{\|Ax\|_\infty}{\|x\|_\infty} = \max_{i=1, \dots, n} \sum_{j=1}^n |a_{ij}|. \text{ Max row sum}$$

$$3. \quad \|A\|_2 = \sup_{x \neq 0} \frac{\|Ax\|_2}{\|x\|_2}. \text{ Also: } \sqrt{\rho(A^T A)}, \rho \text{ is the spectrum of } A. \text{ Euclidean norm}$$

$$4. \quad \|A\|_F = (\sum_i \sum_j |a_{ij}|^2)^{1/2}. \text{ Frobenius norm}$$

1.1.2 Function norm

Function norms we have used in this course are among others:

1. $\|f\|_p = (\int_a^b |f(x)|^p \omega(x) dx)^{1/p}$ where $\omega(x)$ is some weight function.
2. $\|f\|_\infty = \max_{a \leq x \leq b} |f(x)|$

1.2 Banach stuff

Definition 2 (Banach space). *A Banach space is a normed space that is complete. I.e every Cauchy sequence is convergent (to an element of the space).*

Theorem 1. *Let $(X, \|\cdot\|)$ be a Banach space and let $U \subseteq X$ be a subset of X and $f : U \rightarrow X$ be a function. If*

- (i) U is closed
- (ii) f is a contraction
- (iii) $f(U) \subseteq U$

then f has a unique fixed point $x^ \in U$, i.e $x^* = f(x^*)$. Moreover, the sequence $x_n = f(x_{n-1})$, with $x_0 \in U$ arbitrary, converges to x^* .*

1.3 Inner product

Definition 3. *Let V be a vector space. $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}(\mathbb{C})$ is called an inner product on V over $\mathbb{R}(\mathbb{C})$ if:*

- (i) $\langle x, x \rangle \geq 0 \forall x \in \mathbb{R}(\mathbb{C})$
- (ii) $\langle x, x \rangle = 0 \Rightarrow x = 0$
- (iii) $\langle \alpha x + \beta y, z \rangle = \alpha \langle x, z \rangle + \beta \langle y, z \rangle \quad \forall \alpha, \beta \in \mathbb{R}(\mathbb{C}), \quad \forall x, y, z \in V$
- (iv) $\langle x, y \rangle = \overline{\langle y, x \rangle} \quad \forall x, y \in V$

Example 3. 1. \mathbb{R} : $\langle x, y \rangle = \sum_{i=1}^n x_i y_i$

2. \mathbb{C} : $\langle x, y \rangle = \sum_{i=1}^n x_i \overline{y_i}$

3. $\langle f, g \rangle = \int_a^b f(x)g(x)dx$. Inner product of functions

1.4 Properties of a matrix

1.4.1 Positive definite

A matrix is positive definite if $\langle Ax, x \rangle \geq 0$ ($\langle Ax, x \rangle = 0 \Rightarrow x = 0$).

A matrix that is symmetric is positive definite if all its eigenvalues are positive.

To check if a symmetric (Hermitian in the complex case) matrix is positive definite, we can do two things:

1. Check if all eigenvalues are greater than 0
2. Check if all main minors are greater than 0 (i.e. $\det(A_{ii}) > 0$)

1.4.2 Diagonally dominant

If A is diagonally dominant, then A is regular, i.e. A is non-singular, i.e. $\det(A) \neq 0$.

2 Condition of a problem

2.1 Condition analysis by solving linear systems

We have the system $A\vec{x} = \vec{b}$ where $A \in \mathbb{R}^{\kappa, \kappa}$, $b \in \mathbb{R}^{\kappa}$ and $\det(A) \neq 0$

2.1.1 Condition number of a matrix A

1. $K_{abs} = \|A^{-1}\|$
2. $K_{rel} = \|A\| \|A^{-1}\|$

2.2 Preconditioning

Having a system $A\vec{x} = \vec{b}$, preconditioning means to multiply our system with a matrix $B \in \mathbb{R}^{\kappa, \kappa}$ such that $K(BA) < K(A)$. We are solving $BA\vec{x} = B\vec{b}$, $\det(B) \neq 0$.

The easiest preconditioning is to take $B = \text{diag}(d_i)$