L2 Mathématiques

AlgÃ"bre Linéaire
Combinatoire et Probabilités discrÃ"tes
Analyse approfondie
Diagonalisation
Séries et intégrales généralisées
Fonctions de deux variables
Séries de Fourier

Nicolas Rousseau

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Chapitre 1

1.1 Random Examples

Définition 1.1.1: Limit of Sequence in

Let $\{s_n\}$ be a sequence in \mathbb{R} . We say

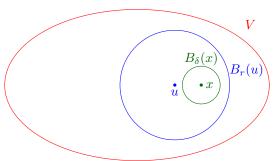
$$\lim_{n \to \infty} s_n = s$$

where $s \in \mathbb{R}$ if \forall real numbers $\epsilon > 0$ \exists natural number N such that for n > N $s - \epsilon < s_n < s + \epsilon$ i.e. $|s - s_n| < s + \epsilon$

Question 1

Is the set x-axis\{Origin} a closed set

Démonstration: By openness of $V, x \in B_r(u) \subset V$



Given $x \in B_r(u) \subset V$, we want $\delta > 0$ such that $x \in B_\delta(x) \subset B_r(u) \subset V$. Let d = d(u,x). Choose δ such that $d + \delta < r$ (e.g. $\delta < \frac{r-d}{2}$)
If $y \in B_\delta(x)$ we will be done by showing that d(u,y) < r but

$$d(u, y) \leqslant d(u, x) + d(x, y) < d + \delta < r$$

Corollaire 1.1.1

By the result of the proof, we can then show...

Lemme 1.1.1

Suppose $\vec{v_1}, \dots, \vec{v_n} \in \mathbb{R}[n]$ is subspace of \mathbb{R}^n .

Proposition 1.1.1

1 + 1 = 2.

1.2 Random

Définition 1.2.1: Normed Linear Space and Norm $\|\cdot\|$

Let V be a vector space over \mathbb{R} (or \mathbb{C}). A norm on V is function $\|\cdot\| V \to \mathbb{R}_{\geq 0}$ satisfying

- (2) $\|\lambda x\| = |\lambda| \|x\| \ \forall \ \lambda \in \mathbb{R} (\text{or } \mathbb{C}), \ x \in V$
- (3) $||x+y|| \le ||x|| + ||y|| \ \forall \ x,y \in V$ (Triangle Inequality/Subadditivity)

And V is called a normed linear space.

• Same definition works with V a vector space over \mathbb{C} (again $\|\cdot\| \to \mathbb{R}_{\geqslant 0}$) where ② becomes $\|\lambda x\| = |\lambda| \|x\|$ $\forall \lambda \in \mathbb{C}, x \in V$, where for $\lambda = a + ib, |\lambda| = \sqrt{a^2 + b^2}$

Exemple 1.2.1 (*p*-Norm)

 $V = \mathbb{R}^m, p \in \mathbb{R}_{\geq 0}$. Define for $x = (x_1, x_2, \cdots, x_m) \in \mathbb{R}^m$

$$||x||_p = (|x_1|^p + |x_2|^p + \dots + |x_m|^p)^{\frac{1}{p}}$$

(In school p = 2)

Special Case $p = 1 : ||x||_1 = |x_1| + |x_2| + \cdots + |x_m|$ is clearly a norm by usual triangle inequality.

Special Case $p \to \infty$ (\mathbb{R}^m with $\|\cdot\|_{\infty}$): $\|x\|_{\infty} = \max\{|x_1|, |x_2|, \cdots, |x_m|\}$

For m=1 these p-norms are nothing but |x|. Now exercise

Question 2

Prove that triangle inequality is true if $p \ge 1$ for p-norms. (What goes wrong for p < 1?)

Solution: For Property (3) for norm-2

When field is \mathbb{R} :

We have to show

$$\sum_{i} (x_i + y_i)^2 \le \left(\sqrt{\sum_{i} x_i^2} + \sqrt{\sum_{i} y_i^2} \right)^2$$

$$\implies \sum_{i} (x_i^2 + 2x_i y_i + y_i^2) \le \sum_{i} x_i^2 + 2\sqrt{\left[\sum_{i} x_i^2\right] \left[\sum_{i} y_i^2\right]} + \sum_{i} y_i^2$$

$$\implies \left[\sum_{i} x_i y_i\right]^2 \le \left[\sum_{i} x_i^2\right] \left[\sum_{i} y_i^2\right]$$

So in other words prove $\langle x, y \rangle^2 \leqslant \langle x, x \rangle \langle y, y \rangle$ where

$$\langle x, y \rangle = \sum_{i} x_i y_i$$

Note:-

$$- ||x||^2 = \langle x, x \rangle$$

$$--\langle x,y\rangle = \langle y,x\rangle$$

— $\langle \cdot, \cdot \rangle$ is R-linear in each slot i.e.

$$\langle rx+x',y\rangle=r\langle x,y\rangle+\langle x',y\rangle$$
 and similarly for second slot

Here in $\langle x, y \rangle$ x is in first slot and y is in second slot.

Now the statement is just the Cauchy-Schwartz Inequality. For proof

$$\langle x, y \rangle^2 \leqslant \langle x, x \rangle \langle y, y \rangle$$

expand everything of $\langle x - \lambda y, x - \lambda y \rangle$ which is going to give a quadratic equation in variable λ

$$\begin{split} \langle x - \lambda y, x - \lambda y \rangle &= \langle x, x - \lambda y \rangle - \lambda \langle y, x - \lambda y \rangle \\ &= \langle x, x \rangle - \lambda \langle x, y \rangle - \lambda \langle y, x \rangle + \lambda^2 \langle y, y \rangle \\ &= \langle x, x \rangle - 2\lambda \langle x, y \rangle + \lambda^2 \langle y, y \rangle \end{split}$$

Now unless $x = \lambda y$ we have $\langle x - \lambda y, x - \lambda y \rangle > 0$ Hence the quadratic equation has no root therefore the discriminant is greater than zero.

When field is \mathbb{C} :

Modify the definition by

$$\langle x, y \rangle = \sum_{i} \overline{x_i} y_i$$

Then we still have $\langle x, x \rangle \geqslant 0$

1.3 Algorithms