ANALYSIS II Introduction Metric and Normed Linear Spaces

<u>Defn</u> A *metric space* is a pair (X,d) where **X** is a set and **d** : $X^2 \rightarrow [0,\infty)$ with the properties that, for each x,y,z in X:

- 1. d(x,y)=0 if and only if x=y,
- 2. d(x,y) = d(y,x),
- 3. $d(x,y) \leq d(x,z) + d(z,y)$.

d is called the *distance function* and d(x,y) denotes the *distance between x and y*.

Note: A given set X may be measured by various distances in order to study the set in different ways.

Examples

- X is any set and d(x,y) := 1 if and only if x is not y.
- The real numbers with absolute value: i.e., $X = \mathbf{R}$ and d(x,y) := |x-y|.
- The complex numbers with <u>modulus</u>: i.e. X = C and $d(z_1, z_2) := |z_1 z_2|$.
- $(X, \|\cdot\|)$ a normed linear space (see <u>below</u>) and $d(x,y) := \|x-y\|$. (Verify!)
- $X = \mathbf{R}^k$ and for $\mathbf{x} = (x_1, x_2, ..., x_k)$ define
 - 1. the standard *Euclidean* distance as $d_2(\mathbf{x}, \mathbf{y}) := (\sum_i |\mathbf{x}_i \mathbf{y}_i|^2)^{1/2}$
 - 2. $\mathbf{d}_{\mathbf{p}}(\mathbf{x},\mathbf{y}) := (\sum_{i} |\mathbf{x}_{i} \mathbf{y}_{i}|^{p})^{1/p}, 1 \le p$
 - 3. $\mathbf{d}_{\infty}(\mathbf{x},\mathbf{y}) := \max_{i} |\mathbf{x}_{i} \mathbf{y}_{i}|$

We will concentrate our studies on the cases $p=1,2,\infty$. We prove that each of the above are metric spaces by showing that they are <u>normed linear spaces</u>, where the obvious candidates are used for norms. The following metrics do not arise as norms [otherwise we must have d(a x, a y) = |a| d(x,y)].

1. $d_p(\mathbf{x}, \mathbf{y}) := (\sum_i |x_i - y_i|^p), \ 0$ $2. <math>\mathbf{X} = \mathbf{R}$ and $d(\mathbf{x}, \mathbf{y}) := |\mathbf{x} - \mathbf{y}|/(1 + |\mathbf{x} - \mathbf{y}|)$ (Homework)

<u>Defn</u> A *normed linear space* is a vector space X and a non-negative valued mapping $||\cdot||$ on X, called the *norm*, which satisfies the properties

- 1. $||\mathbf{x}||=0$ if and only if $\mathbf{x}=0$.
- 2. ||a x|| = |a| ||x||, for all scalars a.
- 3. $||x+y|| \le ||x|| + ||y||$

Here $||\mathbf{x}||$ is thought of as the *length of x* or the distance from \mathbf{x} to $\mathbf{0}$. Notice that for a given vector \mathbf{x} , if \mathbf{y} is defined as $(1/||\mathbf{x}||) \mathbf{x}$, then \mathbf{y} has unit length and is called the *normalized* vector for \mathbf{x} .

Examples

- $X = \mathbf{R}$ and ||x|| := |x|.
- X = C. For z in C, the modulus of z, $|z| := (|\text{Re } z|^2 + |\text{Im } z|^2)^{1/2}$ is a norm for the complex numbers.
- $X = \mathbf{R}^k$ and for $\mathbf{x} = (x_1, x_2, ..., x_k)$ define
 - 1. the standard *Euclidean* distance as $||\mathbf{x}||_2 := (\sum_i |\mathbf{x}_i|^2)^{1/2}$.

Pf: Although this is a special case of the p-norms, it is instructive to demonstrate this separately: First we establish

 $\frac{\text{Lemma}}{\sum_{i} |x_{i} y_{i}|} (\text{Cauchy-Schwarz Inequality})$

Pf: We may assume, without loss of generality, that neither x nor y are the zero vector. First assume that x and y are unit vectors, i.e. ||x|| = ||y|| = 1. Observe by expanding out the square that the inequality $|ab| \le (a^2+b^2)/2$

holds. (This is essentially the famous arithemetic-geometric inequality (using x=a² and y=b²).) It then follows that $\sum_{i} |x_i y_i| \ge 1/2 \sum_{i} (|x_i|^2 + |y_i|^2) = 1$

proving Cauchy's inequality in the special case of unit vectors. For general nonzero vectors, apply this inequality to the normalized vectors for x and y. \Box

Continuing the proof that $\|.\|_2$ is a norm, we observe

$$(\sum_{i} |\mathbf{x}_{i} + \mathbf{y}_{i}|^{2}) \leq ||\mathbf{x}||^{2} + ||\mathbf{y}||^{2} + 2 (\sum_{i} |\mathbf{x}_{i} \mathbf{y}_{i}|) \leq ||\mathbf{x}||^{2} + ||\mathbf{y}||^{2} + 2 ||\mathbf{x}||_{2} ||\mathbf{y}||_{2} = (||\mathbf{x}||_{2} + ||\mathbf{y}||_{2})^{2}$$

and complete the proof by taking square roots. \Box

2. $\|\mathbf{x}\|_{\mathbf{p}} := (\sum_{i} |\mathbf{x}_{i}|^{p})^{1/p}, 1 \le p$

The subadditivity of this norm is known as Minkowski's inequality and relys on

<u>Lemma</u> (Holder's Inequality) Suppose that 1/p+1/q = 1 where $1 \le p$, then $(\sum_{i} |x_{i} y_{i}|) \le ||\mathbf{x}||_{p} ||\mathbf{y}||_{q}$.

Pf: We first show that for nonnegative a and b, that

(*)
$$a^{c}b^{d} \leq c a + d b$$

is true where we have set c = 1/p and d = 1-c = 1/q. Notice that the statement is symmetric in p and q (that is they can

be interchanged in the statement of the Lemma). If a and b are nonnegative real numbers, then we may assume without loss of generality that neither vanishes. We may also assume by the symmetry of p and q that a is larger than b. Observe by the first derivative test that the function $f(u)=1+u-u^c$ is a strictly monotone increasing function for u greater or equal 1. This shows in particular that f(a/b)>0 if a/b>1 since f(1)=0. The inequality (*) follows by multiplying through by b. We apply (*) to $a = x^p$ and $b = y^q$ to obtain for nonnegative numbers x and y

(**)
$$x y \leq x^p / p + y^q / q$$

Hence it follows that if **x** and **y** are normalized vectors $(||\mathbf{x}||_p = ||\mathbf{y}||_q = 1)$, then

 $(\sum_{i} |x_{i} y_{i}|) \leq 1/p (\sum_{i} |x_{i}|^{p}) + 1/q (\sum_{i} |y_{i}|^{q}) = 1.$

Holder's inequality follows by normalizing general nonzero vectors as was done in the Cauchy-Schwarz inequality. 🗖

Pf: To prove the subadditivity of the p-norm, we let $z_i := |x_i + y_i|^{p-1}$. Then

$$\|x+y\|_{p}^{p} \leq \left(\sum_{i} |x_{i}| |z_{i}\right) + \left(\sum_{i} |y_{i}| |z_{i}\right) \leq \left(\|x\|_{p} + \|y\|_{p}\right) \|z\|_{q}$$

where in the second inequality from the left, Holder's inequality is applied to both $(|x_i|, z_i)$ and $(|y_i|, z_i)$. The inequality follows by noticing that

 $||z||_q = ||x+y||_p p^{p/q}$ and that p/q = p-1.

1. $\|\mathbf{x}\|_{\infty} := \max_{i} |\mathbf{x}_{i}|$

Pf: The first two properties of norm follow directly from the properties of absolute value. To establish the subadditve property (3), we observe that for each i between 1 and k,

$$|(x+y)_i| \le |x_i| + |y_i| \le ||x|| + ||y||$$

and then take the maximum over i.

Defn C[a,b] is the set of continuous functions on [a,b]. If f belongs to C[a,b], then $||f||_{\infty} := \max_{x} |f(x)|$ is defined as the norm of f. Sometimes, this is referred to as the *sup norm*.

(Note that the max is always attained in the norm by the extreme value theorem.)

<u>Proposition</u> C[a,b] is a normed linear space.

Pf: By the properties of continuous functions, C[a,b] is a vector space. Since $|f(x)+g(x)| \le ||f|| + ||g||$ for all x in [a,b], taking the maximum over all such x, the subadditivity of the norm is established.

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