Economics 204 Summer/Fall 2016 Lecture 3–Wednesday July 27, 2016

Section 2.1. Metric Spaces and Normed Spaces

Here we seek to generalize notions of distance and length in \mathbf{R}^n to abstract settings.

Definition 1 A *metric space* is a pair (X, d), where X is a set and $d : X \times X \to \mathbf{R}_+$ a function satisfying

- 1. $d(x,y) \ge 0, \ d(x,y) = 0 \Leftrightarrow x = y \ \forall x, y \in X$
- 2. $d(x,y) = d(y,x) \ \forall x, y \in X$
- 3. triangle inequality:

$$\begin{array}{ccc} d(x,z) \leq d(x,y) + d(y,z) & \forall x,y,z \in X \\ & \swarrow & & \searrow \\ & \swarrow & & \searrow \\ & x & \longrightarrow & z \end{array}$$

A function $d: X \times X \to \mathbf{R}_+$ satisfying 1-3 is called a *metric* on X.

A metric gives a notion of distance between elements of X.

Definition 2 Let V be a vector space over **R**. A *norm* on V is a function $\|\cdot\|: V \to \mathbf{R}_+$ satisfying

- 1. $||x|| \ge 0 \ \forall x \in V$
- 2. $||x|| = 0 \Leftrightarrow x = 0 \ \forall x \in V$
- 3. triangle inequality:

$$\begin{aligned} \|x+y\| &\leq \|x\| + \|y\| \ \forall x, y \in V \\ x & \searrow y \\ 0 & \rightarrow & x+y \\ y &\searrow & \swarrow x \\ y & & \swarrow y \end{aligned}$$

4. $\|\alpha x\| = |\alpha| \|x\| \ \forall \alpha \in \mathbf{R}, x \in V$

A normed vector space is a vector space over \mathbf{R} equipped with a norm.

A norm gives a notion of length of a vector in V.

Example: In \mathbb{R}^n , the standard notion of distance between two vectors x and y measures the length of the difference x - y, i.e., $d(x, y) = ||x - y|| = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$.

In an abstract normed vector space, the norm can be used analogously to define a notion of distance.

Theorem 3 Let $(V, \|\cdot\|)$ be a normed vector space. Let $d: V \times V \Rightarrow \mathbf{R}_+$ be defined by $d(v, w) = \|v - w\|$

Then (V, d) is a metric space.

Proof: We must verify that *d* satisfies all the properties of a metric.

1. Let $v, w \in V$. Then by definition, $d(v, w) = ||v - w|| \ge 0$ (why?), and

$$\begin{split} d(v,w) &= 0 &\Leftrightarrow ||v-w|| = 0 \\ &\Leftrightarrow v-w = 0 \\ &\Leftrightarrow (v+(-w))+w = w \\ &\Leftrightarrow v+((-w)+w) = w \\ &\Leftrightarrow v+0 = w \\ &\Leftrightarrow v = w \end{split}$$

2. First, note that for any $x \in V$, $0 \cdot x = (0+0) \cdot x = 0 \cdot x + 0 \cdot x$, so $0 \cdot x = 0$. Then $0 = 0 \cdot x = (1-1) \cdot x = 1 \cdot x + (-1) \cdot x = x + (-1) \cdot x$, so we have $(-1) \cdot x = (-x)$. Then let $v, w \in V$.

$$d(v,w) = \|v - w\|$$

= $|-1|\|v - w\|$
= $\|(-1)(v + (-w))\|$
= $\|(-1)v + (-1)(-w)\|$
= $\|-v + w\|$
= $\|w + (-v)\|$
= $\|w - v\|$
= $\|w - v\|$
= $d(w, v)$

3. Let $u, w, v \in V$.

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

Thus d is a metric on V.

Examples of Normed Vector Spaces

• \mathbf{E}^n : *n*-dimensional Euclidean space.

$$V = \mathbf{R}^{n}, \ \|x\|_{2} = |x| = \sqrt{\sum_{i=1}^{n} (x_{i})^{2}}$$

- $V = \mathbf{R}^n$, $||x||_1 = \sum_{i=1}^n |x_i|$ (the "taxi cab" norm or L^1 norm)
- $V = \mathbf{R}^n$, $||x||_{\infty} = \max\{|x_1|, \dots, |x_n|\}$ (the maximum norm, or sup norm, or L^{∞} norm)
- $C([0,1]), ||f||_{\infty} = \sup\{|f(t)| : t \in [0,1]\}$
- $C([0,1]), ||f||_2 = \sqrt{\int_0^1 (f(t))^2 dt}$
- $C([0,1]), ||f||_1 = \int_0^1 |f(t)| dt$

Theorem 4 (Cauchy-Schwarz Inequality)

If $v, w \in \mathbf{R}^n$, then

$$\begin{pmatrix} \sum_{i=1}^{n} v_i w_i \end{pmatrix}^2 \leq \left(\sum_{i=1}^{n} v_i^2 \right) \left(\sum_{i=1}^{n} w_i^2 \right) |v \cdot w|^2 \leq |v|^2 |w|^2 |v \cdot w| \leq |v||w|$$

Proof: Read the proof in de La Fuente.

The Cauchy-Schwarz Inequality is essential in proving the triangle inequality in \mathbf{E}^n . Deriving the triangle inequality in \mathbf{E}^n from the Cauchy-Schwarz inequality is a good exercise. The Cauchy-Schwarz inequality can also be viewed as a consequence of geometry in \mathbf{R}^2 , in particular the law of cosines. Note that for $v, w \in \mathbf{R}^2$, $v \cdot w = |v||w| \cos \theta$ where θ is the angle between v and w; see Figure 1.¹

Notice that a given vector space may have many different norms. As a trivial example, if $\|\cdot\|$ is a norm on a vector space V, so are $2\|\cdot\|$ and $3\|\cdot\|$ and $k\|\cdot\|$ for any k > 0. Less trivially, \mathbf{R}^n supports many different norms as in the examples above. Different norms on a given vector space yield different geometric properties; for example, see Figure 2 for different norms on \mathbf{R}^2 .

¹From the law of cosines, $(v-w) \cdot (v-w) = v \cdot v + w \cdot w - 2|v||w| \cos \theta$. On the other hand, $(v-w) \cdot (v-w) = v \cdot v - 2v \cdot w + w \cdot w$, so $v \cdot w = |v||w| \cos \theta$.

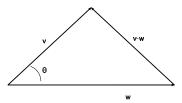


Figure 1: θ is the angle between v and w.

Definition 5 Two norms $\|\cdot\|$ and $\|\cdot\|^*$ on the same vector space V are said to be *Lipschitz-equivalent* (or *equivalent*) if $\exists m, M > 0$ s.t. $\forall x \in V$,

$$m\|x\| \le \|x\|^* \le M\|x\|$$

Equivalently, $\exists m, M > 0$ s.t. $\forall x \in V, x \neq 0$,

$$m \le \frac{\|x\|^*}{\|x\|} \le M$$

If two norms are equivalent, then they define the same notions of convergence and continuity. For topological purposes, equivalent norms are indistinguishable. For example, suppose two norms $\|\cdot\|$ and $\|\cdot\|^*$ on the vector space V are equivalent, and fix $x \in V$. Let $B_{\varepsilon}(x, \|\cdot\|)$ denote the $\|\cdot\|$ -ball of radius ε about x; similarly, let $B_{\varepsilon}(x, \|\cdot\|^*)$ denote the $\|\cdot\|^*$ -ball of radius ε about x. That is,

$$B_{\varepsilon}(x, \|\cdot\|) = \{y \in V : \|x-y\| < \varepsilon\}$$

$$B_{\varepsilon}(x, \|\cdot\|^*) = \{y \in V : \|x-y\|^* < \varepsilon\}$$

Then for any $\varepsilon > 0$,

$$B_{\frac{\varepsilon}{M}}(x, \|\cdot\|) \subseteq B_{\varepsilon}(x, \|\cdot\|^*) \subseteq B_{\frac{\varepsilon}{m}}(x, \|\cdot\|)$$

See Figure 3.

In \mathbf{R}^n (or any finite-dimensional normed vector space), all norms are equivalent. This says roughly that, up to a difference in scaling, for topological purposes there is a unique norm in \mathbf{R}^n .

Theorem 6 All norms on \mathbb{R}^n are equivalent.²

 $^{^2{\}rm The}$ statement of the theorem in de la Fuente (Theorem 10.8, p. 107) is correct, but the proof has a problem.

However, infinite-dimensional spaces support norms that are not equivalent. For example, on C([0, 1]), let f_n be the function

$$f_n(t) = \begin{cases} 1 - nt & \text{if } t \in \left[0, \frac{1}{n}\right] \\ 0 & \text{if } t \in \left(\frac{1}{n}, 1\right] \end{cases}$$

Then

$$\frac{\|f_n\|_1}{\|f_n\|_{\infty}} = \frac{\frac{1}{2n}}{1} = \frac{1}{2n} \to 0$$

Definition 7 In a metric space (X, d), a subset $S \subseteq X$ is *bounded* if $\exists x \in X, \beta \in \mathbf{R}$ such that $\forall s \in S, d(s, x) \leq \beta$.

In a metric space (X, d), define

$$B_{\varepsilon}(x) = \{ y \in X : d(y, x) < \varepsilon \}$$

= open ball with center x and radius ε
$$B_{\varepsilon}[x] = \{ y \in X : d(y, x) \le \varepsilon \}$$

= closed ball with center x and radius ε

We can use the metric d to define a generalization of "radius". In a metric space (X, d), define the *diameter* of a subset $S \subseteq X$ by

$$\operatorname{diam}\left(S\right) = \sup\{d(s,s') : s, s' \in S\}$$

Similarly, we can define the distance from a point to a set, and distance between sets, as follows:

$$d(A, x) = \inf_{a \in A} d(a, x)$$

$$d(A, B) = \inf_{a \in A} d(B, a)$$

$$= \inf\{d(a, b) : a \in A, b \in B\}$$

Note that d(A, x) cannot be a metric (since a metric is a function on $X \times X$, the first and second arguments must be objects of the same type); in addition, d(A, B) does not define a metric on the space of subsets of X (why?).³

Section 2.2. Convergence of Sequences in Metric Spaces

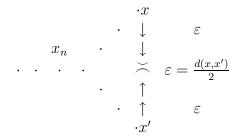
Definition 8 Let (X, d) be a metric space. A sequence $\{x_n\}$ converges to x (written $x_n \to x$ or $\lim_{n\to\infty} x_n = x$) if

$$\forall \varepsilon > 0 \; \exists N(\varepsilon) \in \mathbf{N} \; \text{s.t.} \; n > N(\varepsilon) \Rightarrow d(x_n, x) < \varepsilon$$

³Another, more useful notion of the distance between sets is the Hausdorff distance, given by $d(A, B) = \max \{ \sup_{a \in A} \inf_{b \in B} d(a, b), \sup_{b \in B} \inf_{a \in A} d(a, b) \}.$

Notice that this is exactly the same as the definition of convergence of a sequence of real numbers, except we replace the standard measure of distance $|\cdot|$ in **R** by the general metric d.

Theorem 9 (Uniqueness of Limits) In a metric space (X, d), if $x_n \to x$ and $x_n \to x'$, then x = x'.



Proof: Suppose $\{x_n\}$ is a sequence in $X, x_n \to x, x_n \to x', x \neq x'$. Since $x \neq x', d(x, x') > 0$. Let

$$\varepsilon = \frac{d(x, x')}{2}$$

Then there exist $N(\varepsilon)$ and $N'(\varepsilon)$ such that

$$n > N(\varepsilon) \Rightarrow d(x_n, x) < \varepsilon$$

 $n > N'(\varepsilon) \Rightarrow d(x_n, x') < \varepsilon$

Choose

$$n > \max\{N(\varepsilon), N'(\varepsilon)\}$$

Then

$$d(x, x') \leq d(x, x_n) + d(x_n, x')$$

$$< \varepsilon + \varepsilon$$

$$= 2\varepsilon$$

$$= d(x, x')$$

$$d(x, x') < d(x, x')$$

a contradiction. \blacksquare

Definition 10 An element c is a *cluster point* of a sequence $\{x_n\}$ in a metric space (X, d) if $\forall \varepsilon > 0$, $\{n : x_n \in B_{\varepsilon}(c)\}$ is an infinite set. Equivalently,

$$\forall \varepsilon > 0, N \in \mathbf{N} \ \exists n > N \text{ s.t. } x_n \in B_{\varepsilon}(c)$$

Example:

$$x_n = \begin{cases} 1 - \frac{1}{n} & \text{if } n \text{ even} \\ \frac{1}{n} & \text{if } n \text{ odd} \end{cases}$$

For n large and odd, x_n is close to zero; for n large and even, x_n is close to one. The sequence does not converge; the set of cluster points is $\{0, 1\}$.

If $\{x_n\}$ is a sequence and $n_1 < n_2 < n_3 < \cdots$ then $\{x_{n_k}\}$ is called a subsequence.

Note that a subsequence is formed by taking some of the elements of the parent sequence, in the same order.

Example: $x_n = \frac{1}{n}$, so $\{x_n\} = (1, \frac{1}{2}, \frac{1}{3}, \ldots)$. If $n_k = 2k$, then $\{x_{n_k}\} = (\frac{1}{2}, \frac{1}{4}, \frac{1}{6}, \ldots)$.

Theorem 11 (2.4 in De La Fuente, plus ...) Let (X, d) be a metric space, $c \in X$, and $\{x_n\}$ a sequence in X. Then c is a cluster point of $\{x_n\}$ if and only if there is a subsequence $\{x_{n_k}\}$ such that $\lim_{k\to\infty} x_{n_k} = c$.

Proof: Suppose c is a cluster point of $\{x_n\}$. We inductively construct a subsequence that converges to c. For k = 1, $\{n : x_n \in B_1(c)\}$ is infinite, so nonempty; let

$$n_1 = \min\{n : x_n \in B_1(c)\}$$

Now, suppose we have chosen $n_1 < n_2 < \cdots < n_k$ such that

$$x_{n_j} \in B_{\frac{1}{2}}(c)$$
 for $j = 1, ..., k$

 $\{n: x_n \in B_{\frac{1}{k+1}}(c)\}$ is infinite, so it contains at least one element bigger than n_k , so let

$$n_{k+1} = \min\left\{n : n > n_k, \ x_n \in B_{\frac{1}{k+1}}(c)\right\}$$

Thus, we have chosen $n_1 < n_2 < \cdots < n_k < n_{k+1}$ such that

$$x_{n_j} \in B_{\frac{1}{i}}(c)$$
 for $j = 1, \dots, k, k+1$

Thus, by induction, we obtain a subsequence $\{x_{n_k}\}$ such that

$$x_{n_k} \in B_{\frac{1}{k}}(c)$$

Given any $\varepsilon > 0$, by the Archimedean property, there exists $N(\varepsilon) > 1/\varepsilon$.

$$k > N(\varepsilon) \implies x_{n_k} \in B_{\frac{1}{k}}(c)$$
$$\implies x_{n_k} \in B_{\varepsilon}(c)$$

SO

$$x_{n_k} \to c \text{ as } k \to \infty$$

Conversely, suppose that there is a subsequence $\{x_{n_k}\}$ converging to c. Given any $\varepsilon > 0$, there exists $K \in \mathbb{N}$ such that

$$k > K \Rightarrow d(x_{n_k}, c) < \varepsilon \Rightarrow x_{n_k} \in B_{\varepsilon}(c)$$

Therefore,

$$\{n: x_n \in B_{\varepsilon}(c)\} \supseteq \{n_{K+1}, n_{K+2}, n_{K+3}, \ldots\}$$

Since $n_{K+1} < n_{K+2} < n_{K+3} < \cdots$, this set is infinite, so c is a cluster point of $\{x_n\}$.

Section 2.3. Sequences in \mathbf{R} and \mathbf{R}^m

Definition 12 A sequence of real number $\{x_n\}$ is *increasing* (*decreasing*) if $x_{n+1} \ge x_n$ $(x_{n+1} \le x_n)$ for all n.

Definition 13 If $\{x_n\}$ is a sequence of real numbers, $\{x_n\}$ tends to infinity (written $x_n \to \infty$ or $\lim x_n = \infty$) if

$$\forall K \in \mathbf{R} \exists N(K) \text{ s.t. } n > N(K) \Rightarrow x_n > K$$

Similarly define $x_n \to -\infty$ or $\lim x_n = -\infty$.

Notice we don't say the sequence *converges* to infinity; the term "converge" is limited to the case of finite limits.

Theorem 14 (Theorem 3.1') Let $\{x_n\}$ be an increasing (decreasing) sequence of real numbers. Then $\lim_{n\to\infty} x_n = \sup\{x_n : n \in \mathbf{N}\}$ ($\lim_{n\to\infty} x_n = \inf\{x_n : n \in \mathbf{N}\}$). In particular, the limit exists.

Proof: Read the proof in the book, and figure out how to handle the unbounded case.

Lim Sups and Lim Infs:⁴

Consider a sequence $\{x_n\}$ of real numbers. Let

$$\alpha_n = \sup\{x_k : k \ge n\}$$

=
$$\sup\{x_n, x_{n+1}, x_{n+2}, \ldots\}$$

$$\beta_n = \inf\{x_k : k \ge n\}$$

Either $\alpha_n = +\infty$ for all n, or $\alpha_n \in \mathbf{R}$ and $\alpha_1 \ge \alpha_2 \ge \alpha_3 \ge \cdots$. Either $\beta_n = -\infty$ for all n, or $\beta_n \in \mathbf{R}$ and $\beta_1 \le \beta_2 \le \beta_3 \le \cdots$.

⁴See the handout for this material.

Definition 15

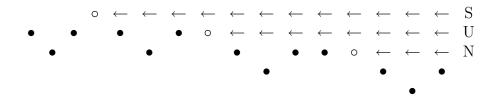
$$\limsup_{n \to \infty} x_n = \begin{cases} +\infty & \text{if } \alpha_n = +\infty \text{ for all } n \\ \lim \alpha_n & \text{otherwise.} \end{cases}$$
$$\lim_{n \to \infty} x_n = \begin{cases} -\infty & \text{if } \beta_n = -\infty \text{ for all } n \\ \lim \beta_n & \text{otherwise.} \end{cases}$$

Theorem 16 Let $\{x_n\}$ be a sequence of real numbers. Then

$$\lim_{n \to \infty} x_n = \gamma \in \mathbf{R} \cup \{-\infty, \infty\}$$

$$\Leftrightarrow \limsup_{n \to \infty} x_n = \liminf_{n \to \infty} x_n = \gamma$$

Theorem 17 (Theorem 3.2, Rising Sun Lemma) Every sequence of real numbers contains an increasing subsequence or a decreasing subsequence or both.



Proof: Let

$$S = \{s \in \mathbf{N} : x_s > x_n \quad \forall n > s\}$$

Either S is infinite, or S is finite.

If S is infinite, let

$$n_1 = \min S$$

$$n_2 = \min (S \setminus \{n_1\})$$

$$n_3 = \min (S \setminus \{n_1, n_2\})$$

$$\vdots$$

$$n_{k+1} = \min (S \setminus \{n_1, n_2, \dots, n_k\})$$

Then $n_1 < n_2 < n_3 < \cdots$.

$$x_{n_1} > x_{n_2} \quad \text{since } n_1 \in S \text{ and } n_2 > n_1$$

$$x_{n_2} > x_{n_3} \quad \text{since } n_2 \in S \text{ and } n_3 > n_2$$

$$\vdots$$

$$x_{n_k} > x_{n_{k+1}} \quad \text{since } n_k \in S \text{ and } n_{k+1} > n_k$$

$$\vdots$$

so $\{x_{n_k}\}$ is a strictly decreasing subsequence of $\{x_n\}$.

If S is finite and nonempty, let $n_1 = (\max S) + 1$; if $S = \emptyset$, let $n_1 = 1$. Then

$$n_{1} \notin S \quad \text{so} \quad \exists n_{2} > n_{1} \text{ s.t. } x_{n_{2}} \ge x_{n_{1}}$$

$$n_{2} \notin S \quad \text{so} \quad \exists n_{3} > n_{2} \text{ s.t. } x_{n_{3}} \ge x_{n_{2}}$$

$$\vdots$$

$$n_{k} \notin S \quad \text{so} \quad \exists n_{k+1} > n_{k} \text{ s.t. } x_{n_{k+1}} \ge x_{n_{k}}$$

$$\vdots$$

so $\{x_{n_k}\}$ is a (weakly) increasing subsequence of $\{x_n\}$.

Theorem 18 (Thm. 3.3, Bolzano-Weierstrass) Every bounded sequence of real numbers contains a convergent subsequence.

Proof: Let $\{x_n\}$ be a bounded sequence of real numbers. By the Rising Sun Lemma, find an increasing or decreasing subsequence $\{x_{n_k}\}$. If $\{x_{n_k}\}$ is increasing, then by Theorem 3.1', $\lim x_{n_k} = \sup\{x_{n_k} : k \in \mathbf{N}\} \le \sup\{x_n : n \in \mathbf{N}\} < \infty$, since the sequence is bounded; since the limit is finite, the subsequence converges. Similarly, if the subsequence is decreasing, it converges.

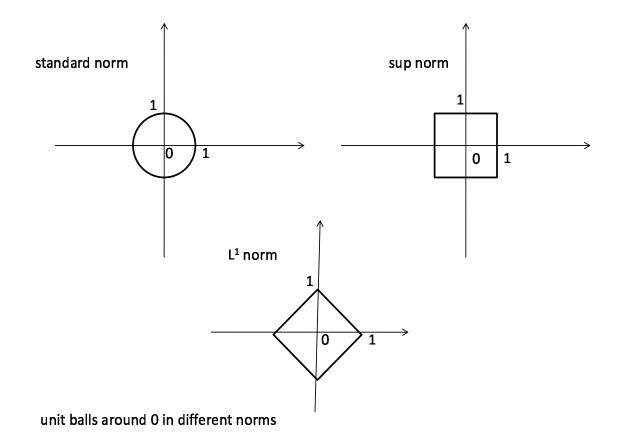
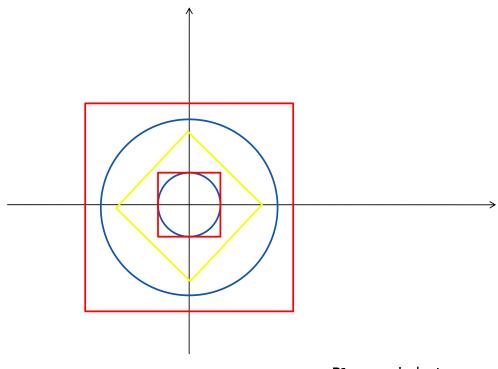


Figure 2: The unit ball around 0 in different norms on \mathbf{R}^2 : standard $\|\cdot\|_2$, $\|\cdot\|_1$ (L^1 or taxi cab norm) and $\|\cdot\|_{\infty}$ (sup norm or L^{∞} norm).



norms on Rⁿ are equivalent

Figure 3: All norms on \mathbf{R}^n are equivalent.