Module 3: Metric Spaces and Continuity

Loosely speaking, a *space* is a set with some additional structure. For example, a vector space is a set together with a linear structure (i.e. addition and scalar multiplication), and a probability space is a set together with a collection of *events* (i.e. measurable subsets) and a *probability measure* providing the likelihood that a given event occurs. In this chapter we study *metric spaces*, which are comprised of sets together with some associated notion of distance, i.e. a *metric*.

Let *X* be a non-empty set and let $d: X \times X \to \mathbb{R}_+$ have the following properties:

- 1. $d(x,y) = 0 \Leftrightarrow x = y$
- 2. (symmetry) d(x,y) = d(y,x)
- 3. (triangle inequality) $d(x,z) \le d(x,y) + d(y,z)$

The pair (X, d) is a metric space, with d a metric on X.

Intuitively, the metric d measures the distance between points in X. The presence of a metric structure on a given set of interest allows the practitioner to appeal to function continuity and sequential approximation. As we will discuss in more detail in the next chapter, continuous functions preserve nearness: intuitively, if f is a continuous function between metric spaces then x near y implies f(x) near f(y).

Sequential approximation exploits ability to define convergence on a metric space, as we will do below. The idea is as follows: perhaps you would like to say some property holds for a given $x \in X$; and, while you cannot show it holds for x directly, perhaps you can show the property holds for each element in a sequence $\{x_n\}$. If you can also show that $x_n \to x$ and that this property respects limits, you're golden.²

Exercise 1 Show that \mathbb{R}^2_+ is a metric space – be sure to define the metric.

¹More rigorously, it's the local, dual version of this statement that defines continuity: f is continuous at x provided that you can make f(y) as close as you like to f(x) by choosing y sufficiently close to x.

²Continuity can be defined on sets with a more general structure known as a *topology*. In economics, almost all topological spaces are, themselves, metric spaces (a metric space is a special case of a topological space) or, at least, they are naturally metrizable, which means there is a definable metric which reproduces the topology. Sequential approximation can also be generalized, but here it is more subtle: without a metric, the sequences must be replaced with *nets* (or, equivalently, filters), which may be viewed as uncountable analogous to sequences.

1 Normed linear spaces

Most metric spaces encountered in economics are (subsets of) vector spaces coupled with *norms* that measure the *lengths* of their vectors, i.e. *normed linear (or vector) spaces*. If V is a vector space then a norm on V is a map $\|\cdot\|: V \to \mathbb{R}_+$ satisfying the following properties:

- 1. $||v|| = 0 \Leftrightarrow v = 0$
- 2. $\|\alpha v\| = |\alpha| \|v\|$ where, here, $\alpha \in \mathbb{C}$ or \mathbb{R} and $v \in V$
- 3. (triangle inequality) $||v+w|| \le ||v|| + ||w||$

The relationship between a norm and a distance measure is given by the following exercise:

Exercise 2 *Show that if* $(V, \|\cdot\|)$ *is a normed vector space and* $d: V \times V \to \mathbb{R}_+$ *is defined by*

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

then (V,d) is a metric space.

The following are examples of normed vector spaces commonly found in economic analysis. Some of them may not be familiar, and/or include terms not yet defined in these notes, but we will get to the details.

- 1. $(\mathbb{R}^n, \|\cdot\|_2)$ where $\|v\|_2 = (\sum_{i=1}^n v_i^2)^{1/2}$
- 2. $(\mathbb{R}^n, \|\cdot\|_1)$ where $\|v\|_1 = \sum_{i=1}^n |v_i|$
- 3. $(\mathbb{R}^n, \|\cdot\|_{\infty})$ where $\|v\|_{\infty} = \max_{i=1,...,n} |v_i|$
- 4. $(\ell^2(\mathbb{N}), \|\cdot\|_2)$ where

$$\ell^2(\mathbb{N}) = \left\{ f : \mathbb{N} \to \mathbb{R} \text{ such that } \sum_n f(n)^2 < \infty \right\}$$

$$\|f\|_2 = \left(\sum_n f(n)^2 \right)^{1/2}$$

5. $(\ell^1(\mathbb{N}), \|\cdot\|_1)$ where

$$\ell^1(\mathbb{N}) = \left\{ f : \mathbb{N} \to \mathbb{R} \text{ such that } \sum_n |f(n)| < \infty \right\}$$

$$\|f\|_1 = \sum_n |f(n)|$$

6. $(\ell^{\infty}(\mathbb{N}), \|\cdot\|_{\infty})$ where

$$\ell^{\infty}(\mathbb{N}) = \left\{ f : \mathbb{N} \to \mathbb{R} \text{ such that } \sup_{n} |f(n)| < \infty \right\}$$

$$\|f\|_{\infty} = \sup_{n} |f(n)|$$

7. $(C(X), \|\cdot\|_{\infty})$ where *X* is a compact metric space,

$$C(X) = \text{set of continuous functions from } X \text{ to } \mathbb{R}$$

 $||f||_{\infty} = \sup_{x \in X} |f(x)|$

8. $(L^2(\Omega), \|\cdot\|_2)$ where (Ω, μ) is a probability space,

$$L^2(\Omega) = \text{set of random variables on } \Omega \text{ with finite second moments, i.e. } Ex^2 < \infty$$

 $||x||_2 = (Ex^2)^{1/2}$

Hopefully you recognize a broad pattern here.

Exercise 3 Determine the relationships among $\ell^1(\mathbb{N}), \ell^2(\mathbb{N})$ and $\ell^{\infty}(\mathbb{N})$, i.e. which, if any, are included in which?

2 Completeness

A sequence in a metric space (X,d) is just what you think: an infinite list of elements of X, often, but not always (see exercises below) denoted $\{x_n\}$. We say that the sequence converges to a point x, written $x_n \to x$, provided that $d(x_n,x) \to 0$ in \mathbb{R} . Formally, $x_n \to x$ provided that for any $\varepsilon > 0$ there is some N > 0 so that $n \ge N$ implies $d(x_n,x) < \varepsilon$.

Exercise 4 Show that if $\{x_n\}$ is a sequence in a metric space (X,d) and if $x_n \to x$ and $x_n \to y$ then x = y.

Exercise 5 Let $x^n \in \mathbb{R}^{\mathbb{N}}$ be given by $x_m^n = 1/m$ if $m \le n$ and 0 otherwise.³ Let $x \in \mathbb{R}^{\mathbb{N}}$ be given by $x_m = 1/m$.

- 1. Show that $x^n \in \ell^1, \ell^2$ and ℓ^{∞} .
- 2. In which, if any, of these spaces does $x^n \to x$?

³Remember, $\mathbb{R}^{\mathbb{N}}$ is the set of real sequences, with no other restrictions. It is a vector space, but not a natural normed vector space.

Exercise 6 Let $x^n \in \mathbb{R}^{\mathbb{N}}$ be given by $x_m^n = \delta_{mn}$. Does x^n converge to zero in any of ℓ^1, ℓ^2 or ℓ^{∞} ?

Let $\{x_n\}$ be a sequence in a metric space (X,d). The sequence is *Cauchy* provided that for any $\varepsilon > 0$ there is some N > 0 so that m, n > N imply $d(x_n, x_m) < \varepsilon$. Intuitively, the terms in a Cauchy sequence get, and stay close together. A metric space is *complete* provided that Cauchy sequences converge, i.e. if $\{x_n\}$ is a Cauchy sequence then there is a point $x \in X$ such that $x_n \to x$.

Exercise 7 *Show this point is unique.*

The intermediate value theorem (IVT), which applies to continuous functions from \mathbb{R} to \mathbb{R} , is the fundamental tool for establishing the existence of a solution to an equation. Indeed, to show there is a solution to an equation of the form f(x) = 0 for some continuous function f, it is enough, by the IVT, to find some a < b so that f(a) < 0 < f(b) (or vice-versa). The essential step in the proof of this theorem is to form the set \mathscr{A} of all points $z \in [a,b]$ such that $f(z) \geq 0$. Since \mathbb{R} is complete, so too is [a,b], and therefore the infimum of this set \mathscr{A} exists. It is a nice exercise, then, to show that $f(\inf(\mathscr{A})) = 0$.

The concept of completeness generalized to metric spaces allows us to extend the intermediate value theorem in a variety of natural ways, which will provide for solutions to equations of interest. Somewhat more specifically, the intermediate value theorem can be viewed as a fixed point theorem. To see this, recall that if $f:[a,b] \to [a,b]$ then a *fixed point* of f is a real number $x \in [a,b]$ such that f(x) = x. By the IVT, if f is continuous then it has a fixed point. Indeed, suppose f(a) > a and f(b) < b (if either of these conditions doesn't hold then f has a fixed point!) and let g(x) = f(x) - x. Then g is continuous, g(a) > 0 and g(b) < 0. Extensions of the IVT to more general metric spaces come in the form of fixed point theorems.

One more, somewhat conceptually harder notion is needed. Let $\{x_n\}$ be a sequence of real numbers. Two definitions:

- 1. $\limsup x_n = \lim_{n \to \infty} \sup \{x_k : k \ge n\}$
- 2. $\liminf_{n\to\infty} x_n = \lim_{n\to\infty} \inf\{x_k : k \ge n\}$

To make sense of this definition, let

$$\hat{x}^n = \sup\{x_k : k \ge n\} \text{ and } \hat{x}_n = \inf\{x_k : k \ge n\}.$$

Notice that $\{\hat{x}^n\}$ is a *decreasing sequence* and $\{\hat{x}_n\}$ is an *increasing sequence*. Now suppose that the original sequence $\{x_n\}$ is bounded, i.e. there is some interval $[-M,M] \subset \mathbb{R}$ such that $x_n \in [-M,M]$ for all n. Then $\{\hat{x}^n\}$ is a decreasing sequence that is bounded below and $\{\hat{x}_n\}$ is an *increasing sequence* that is bounded above. By the completeness of the real

numbers, it follows that both of these sequences converge, i.e. both $\limsup x_n$ and $\liminf x_n$ exist.

Intuitively, liminfs and lim sups concern subsequences. A subsequence is any infinite list of numbers that comes from a sequence, with the ordering preserved. Thus $\{x_2, x_4, x_6, \ldots\}$ is a subsequence of $\{x_n\}$. Of course a given $\{x_n\}$ cannot be expected to converge to anything. For example, $y_n = n$ diverges to ∞ and $x_n = (-1)^n$ bounces between ± 1 forever, never settling down. On the other hand, if the sequence is bounded, as in the latter case, then it will have convergent subsequences, and liminf and lim sup identify the smallest and largest limits of these convergent subsequences. In particular, if $\{x_n\}$ is a bounded sequence of real numbers then the following hold:

- 1. $\limsup x_n$ and $\liminf x_n$ exist and $\liminf x_n \leq \limsup x_n$
- 2. There is a subsequence of $\{x_n\}$ converging to $\limsup x_n$
- 3. There is a subsequence of $\{x_n\}$ converging to $\liminf x_n$
- 4. If $x \in \mathbb{R}$ is the limit of a subsequence of $\{x_n\}$ then $\liminf x_n \le x \le \limsup x_n$.

Exercise 8 Show that if $\{x_n\}$ is a sequence in (X,d) which converges to $x \in X$ then every subsequence of $\{x_n\}$ also converges to x.

The following exercise shows the power of this concept:

Exercise 9 Let $\{x_n\}$ be a sequence of real numbers.

- 1. Show that $\{x_n\}$ converges to some $x \in \mathbb{R}$ if and only if $\limsup x_n = \liminf x_n$.
- 2. Using our earlier definition of completeness on \mathbb{R} (and the previous part of this exercise) show that Cauchy sequences converge.

3 Openness, closedness, and compactness

First a little notation. If $x \in X$ then the ball of radius ε around x is defined as

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

A subset A of a metric space (X,d) is *open* provided that for each $x \in A$ there is some ball around x that is completely contained in A. A subset A of a metric space (X,d) is *closed* provided that its complement in X is open. Think of open and closed sets as generalizing the concepts of open and closed intervals.

Exercise 10 Some exercises.

- 1. Show that $\bar{B}(x, \varepsilon) = \{ y \in X : d(x, y) \le \varepsilon \}$ is closed.
- 2. Show that $(1,2) \cup (3,\infty)$ is open in \mathbb{R} .
- 3. Show that the intersection of two open sets is open and of two closed sets is closed.
- 4. Are there any subsets of \mathbb{R} that are both open and closed?

Our final metric space notion is compactness. A subset A of a metric space (X,d) is *compact* provided that any sequence in A has a subsequence that converges to some point in A. Intuitively, a compact set is not too big (otherwise it could have sequences with points that are always far from each other) and contains it boundary (otherwise a sequence could converge to some point on the boundary, but this would mean that every subsequences converges to a point on the boundary).

Compact sets constrain the behavior of continuous functions defined on them. Most importantly for us, and as will be examined in detail in later chapters, if f is a real-valued continuous function on a compact set $A \subset X$ then f realizes it maximum and minimum at some points in A, i.e. there exists a_{\max} and a_{\min} in A so that

$$a_{\max} = \arg \max_{x \in A} f(x)$$
 and $a_{\min} = \arg \min_{x \in A} f(x)$.

Thus if we restrict attention to compact choice sets, we will know that our agents' optimization problems have solutions.

It turns out that compact sets in \mathbb{R}^n are nicely characterized. A set is bounded if it can be contained in some open ball. We have the following:

Theorem 1 (Heine-Borel) *The set* $A \subset \mathbb{R}^n$ *is compact if and only if it is closed and bounded.*

Exercise 11 Observe that the "closed unit ball" D in \mathbb{R}^n , as given by

$$D(\mathbb{R}^n) = \{x \in \mathbb{R}^n : ||x||_2 \le 1\}$$

is compact. Show that $D\left(\ell^2(\mathbb{N})\right)$ is not compact.