

## Math 55a: Honors Advanced Calculus and Linear Algebra

### Metric topology I: basic definitions and examples

**Definition.** Metric topology is concerned with the properties of and relations among *metric spaces*. In general, “space” is used in mathematics for a set with a specific kind of structure; in Math 55 we’ll also encounter vector spaces, function spaces, inner-product spaces, and more. The structure that makes a set  $X$  a metric space is a *distance*  $d$ , which we think of as telling how far any two points  $p, q \in X$  are from each other. That is,  $d$  is a function from  $X \times X$  to  $\mathbf{R}$ . [NB: This is often indicated by the notation  $d : X \times X \rightarrow \mathbf{R}$ ; to emphasize that  $d$  is a function of two variables one might write  $d = d(\cdot, \cdot)$ . Cf. the old symbol “ $\div$ ” for division!] The *axioms* that formalize the notion of a distance are:

**Nonnegativity:**  $d(p, q) \geq 0$  for all  $p, q \in X$ . Moreover,  $d(p, p) = 0$  for all  $p$ , and  $d(p, q) > 0$  if  $p \neq q$ .

**Symmetry:**  $d(p, q) = d(q, p)$  for all  $p, q \in X$ .

**Triangle inequality:**  $d(p, q) \leq d(p, r) + d(r, q)$  for all  $p, q, r \in X$ .

Any function satisfying these three properties is called a *distance function*, or *metric*, on  $X$ . A set  $X$  together with a metric becomes a *metric space*. Note that strictly speaking a metric space is thus an ordered pair  $(X, d)$  where  $d$  is a distance function on  $X$ . Usually we’ll simply call this space  $X$  when  $d$  is understood.

**Examples.** The prototypical example of a metric space is  $\mathbf{R}$  itself, with the metric  $d(x, y) := |x - y|$ . (Check that this in fact satisfies all the axioms required of a metric.) Two trivial examples are an empty set, and a one-point set  $\{x\}$  with  $d(x, x) = 0$ .

Having introduced a new mathematical structure one often shows how to construct new examples from known ones. For the structure of a metric space, the easiest such construction is to take an arbitrary *subset*  $Y$  of a known metric space  $X$ , using the same distance function — more formally, the *restriction* of  $d$  to  $Y \times Y \subset X \times X$ . It should be clear that this is a distance function on  $Y$ , which thus becomes a metric space in its own right, known as a metric *subspace* of  $X$ . So, for instance, the single metric space  $\mathbf{R}$  gives as a huge supply of further metric spaces: simply take any subset, use  $d(x, y) = |x - y|$  to make it a subspace of  $X$ .

Another construction is the *Cartesian product*  $X \times Y$  of two known metric spaces  $X, Y$ . This consists of all ordered pairs  $(x, y)$  with  $x \in X$  and  $y \in Y$ . There are several choices of metric on  $X \times Y$ , of which the simplest is the *sup metric* defined by

$$d_{X \times Y}((x, y), (x', y')) = \max(d_X(x, x'), d_Y(y, y')). \quad (1)$$

(Note that we use subscripts to distinguish the distance functions on  $X$ ,  $Y$ , and  $X \times Y$ .) So, for instance, taking  $X = Y = \mathbf{R}$  we obtain a new metric space

$\mathbf{R} \times \mathbf{R}$ , otherwise known as  $\mathbf{R}^2$ . [Warning: the sup metric

$$d((x, y), (x', y')) = \max(|x - x'|, |y - y'|)$$

on  $\mathbf{R}^2$  is *not* the Euclidean metric you are familiar with. It is a bit tricky to prove analytically that the Euclidean metric satisfies the triangle inequality; we shall do this when we study inner-product spaces a few weeks hence. For an alternative proof, see Rudin, Thms. 1.35 and 1.37e (pages 15,17).] Of course any subset of  $\mathbf{R}^2$  then becomes a metric space as well. We can also iterate the product construction, obtaining for instance the metric spaces  $\mathbf{R}^2 \times \mathbf{R}$  and  $\mathbf{R} \times \mathbf{R}^2$ .

Shouldn't both of these simply be called  $\mathbf{R}^3$ ? True, both sets can be identified with ordered triples  $(x, y, z)$  of real numbers, arising as  $((x, y), z)$  in  $\mathbf{R}^2 \times \mathbf{R}$  and as  $(x, (y, z))$  in  $\mathbf{R} \times \mathbf{R}^2$ . But we haven't defined a metric on the Cartesian product  $X \times Y \times Z$  of three metric spaces  $X, Y, Z$ , and meanwhile we have two metrics coming from the definition (1): one from  $(X \times Y) \times Z$ , the other from  $X \times (Y \times Z)$ . Fortunately the two metrics coincide: both tell us that the distance between  $(x, y, z)$  and  $(x', y', z')$  should be

$$\max(d_X(x, x'), d_Y(y, y'), d_Z(z, z')).$$

In other words, the function  $i : (X \times Y) \times Z \rightarrow X \times (Y \times Z)$  taking  $((x, y), z)$  to  $(x, (y, z))$  is not only a bijection of sets but an *isomorphism* of metric spaces, a.k.a. an *isometry*. In general an *isometry* is a bijection  $i : X \rightarrow X'$  between metric spaces such that  $d_X(p, q) = d_{X'}(i(p), i(q))$  for all  $p, q \in X$ . This definition captures the notion that  $X, X'$  are "the same" metric space, and  $i$  effects an identification between  $X, X'$ . This justifies our identification of  $(X \times Y) \times Z$  with  $X \times (Y \times Z)$  as *metric spaces*, and calling them both  $X \times Y \times Z$ . In particular, we have a natural isometry between  $\mathbf{R}^2 \times \mathbf{R}$  and  $\mathbf{R} \times \mathbf{R}^2$  and may call them both  $\mathbf{R}^3$ . Likewise we may inductively construct  $\mathbf{R}^n$  ( $n = 2, 3, 4, \dots$ ) as  $\mathbf{R}^m \times \mathbf{R}^{n-m}$  for any integer  $m$  with  $0 < m < n$ ; the choice does not matter, because we always get the same distance function

$$d((x_1, x_2, \dots, x_n), (x'_1, x'_2, \dots, x'_n)) = \max_{1 \leq i \leq n} |x_i - x'_i|.$$

We shall give several further basic constructions of metric spaces and examples of isometries in the first problem set.

**Bounded metric spaces and function spaces.** Perhaps the simplest property a metric space might have is boundedness. A metric space  $X$  is said to be *bounded* if there exists a real number  $B$  such that  $d(p, q) < B$  for all  $p, q \in X$ . Note that this is not quite the definition given by Rudin (2.18i, p.32). However, the two definitions are equivalent by the following easy

**Proposition.** *Let  $E$  be a nonempty subset of a metric space  $X$ . The following are equivalent:*

- i)  $E$ , considered as a subspace of  $X$ , is bounded.
- ii) There exists  $p \in E$  and a real number  $M$  such that  $d(p, q) < M$  for all  $q \in E$ .
- iii) There exists  $p \in X$  and a real number  $M$  such that  $d(p, q) < M$  for all  $q \in E$ .

*Proof:* We show (i) $\Rightarrow$ (ii) $\Rightarrow$ (iii) $\Rightarrow$ (i). (i) $\Rightarrow$ (ii) is clear: let  $M = B$ , and choose for  $p$  an arbitrary point of  $E$ . The implication (ii) $\Rightarrow$ (iii) is even easier, for we may use the same  $M, p$ . Finally (iii) $\Rightarrow$ (i) is a consequence of the triangle inequality, with  $B = 2M$ : for  $q, q' \in E$  we have  $d(q, q') \leq d(p, q) + d(p, q') < M + M = 2M = B$ .  $\square$

Why did we require  $E$  to be nonempty? Note that the empty metric space is bounded by our definition. It is also bounded by Rudin's definition, except when regarded as a subset of the empty metric space — which is surely an oversight. We shall always regard  $\emptyset$  as bounded regardless of where we found it, even if nowhere!

Further examples: a finite metric space is bounded; so is an interval  $[a, b] := \{x \in \mathbf{R} : a \leq x \leq b\}$ , considered as a subspace of  $\mathbf{R}$ . If  $X$  is bounded then so is any subspace; if  $X, Y$  are bounded, so is  $X \times Y$ . The metric space  $\mathbf{R}$  is *not* bounded. If  $X, Y$  are metric spaces, and  $X$  is not bounded, then neither is  $X \times Y$ , unless  $Y$  is empty. (Verify all these!)

Given a bounded metric space  $X$  and any set  $S$  we may construct a new kind of metric space, a *function space*. We shall call it  $X^S$ . As a set, this is simply the space of functions  $f : S \rightarrow X$ . (Do you see why we use the notation  $X^S$  for this?) To make it a metric space we define the distance between two functions  $f, g$  by

$$d_{X^S}(f, g) := \sup_{s \in S} d_X(f(s), g(s)).$$

[NB this doesn't quite work when  $S = \emptyset$ ; what is  $X^S$  then, and what goes wrong with the definition of  $d_{X^\emptyset}$ ? How should we fix it?] This makes sense because  $d_X(f(s), g(s)) < B$  for all  $s$ , so the set  $\{d_X(f(s), g(s)) : s \in S\}$  is bounded and thus has a least upper bound. (See Rudin, Chapter 1 to review this notion if necessary.) Verify that this is in fact a metric space. Actually this is not an entirely new example, since if  $S$  is the finite set  $\{1, 2, \dots, n\}$  the space  $X^S$  is isometric with  $X^n$ . The more general  $X^S$  will be a starting point for many important constructions later in the course.