# Statistical Machine Learning Reading Assignment 1 Report

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## 1 A Taste of Real Analysis

Real Analysis is the field of mathematics on which probability theory is founded, it is therefore convinient to first introduce some basic concepts from real analysis. We begin the this chapter with the definiation of metric space and use the notion of metric to define two core concepts: convergence and continuity. In ?? we revie the natural topoly defined on a metric space based on the metric function and then in ?? we provide a more general view of topological spaces.

## 1.1 Metric Space

A metric space is a set M together with a metric  $d: M \times M \to \mathcal{R}$  satisfying following four properties

- d is non-negative
- d(x, y) = 0 iff x = y
- Symmetry: d(x, y) = d(y, x)
- Triangle Inequality:  $d(x, z) \le d(x, y) + d(y, z)$

Strictly speaking, the pair (M,d) is the metric space as different metric functions can be defined on same M, consider for example  $\mathcal{R}^n$ ; a well-known class of metrics defined on  $\mathcal{R}^n$  is **Minkowski Norm**:

$$d_p(x,y) = \left(\sum_{i=1}^n (x_i - y_i)^p\right)^{1/p}$$

which for all values of  $p \ge 1$  is a valid metric function.

#### 1.1.1 Convergence & Continuity

There are different ways for defining convergence, we follow [?] and use sequence/subsequence approach. A sequence  $(p_n)$  is a list of points  $p_1, p_2, \ldots$  in M. Formaly, a sequence is a function  $f: \mathbb{N} \to M$  in which  $f(n) = p_n$ . The sequence  $(p_n)$  converges to the limit p in M (and denote this by  $(p_n) \to p$  if:

$$\forall \epsilon > 0 \quad \exists N \in \mathbb{N} \quad \text{such that}$$
  
 $n \geq N \Rightarrow d(p_n, p) < \epsilon$ 

Having defined convergence, contiuity can be described: For a function  $f: M \to N$  between two metric spaces  $(M, d_M)$  and  $(N, d_N)$ , we say that function is continuous if it preserves sequential convergence, that is if  $(p_n) \to p$  then  $(f(p_n)) \to f(p)$ .

The sequence definitation of continuitity stated above is equivalent with the more familiar definiation using  $(\epsilon, \delta)$  condition:

**Theorem 1**  $f: M \to N$  is continuous if and only if for each  $\epsilon > 0$  and  $p \in M$  there exists  $\delta > 0$  such  $\forall x \in M: d_M(x,p) < \delta \Rightarrow d_N(f(x),f(p)) < \epsilon$ 

#### 1.1.2 Topology of Metric Space

Althogh topology can be defined on non-metric spaces (as we will do so in the next section) there is a *natural* topology induced on metric spaces induced by the distance function. To this end we need to define notion of **openness** and **closeness** based metric and convergences. We say that point  $p \in M$  is a limit of  $S \subset M$  if there exists a sequence in S like  $(p_n)$  that  $(p_n) \to p$ 

**Closeness:** S is a closed set if it contains all it limits. **Openness:** S is an open set if for each  $p \in S$ , r > 0 exists such that

$$d(p,q) < r \Rightarrow q \in S$$
.

that is for each point in S an small ball around it is also in S.

One can simply prove that complement of an open set is closed and vice versa. However (like doors) sets can be neighter open nor closed and unlike doors they can be both at the same time.

**Theorem 2** Now the collection  $\mathcal{T}$  of all open sets of M is the topology of M, i.e. is satisfies the following three properties:

- M, Φ ∈ T
- The intersection of finitely many open sets is an open set
- The union of arbitrarily many open sets is an open set.

### 1.2 Topology: a More General Perspective

The three properties stated in ?? are the definiation of topology, one can hand-craft a collection  $\mathcal{T}$  that satisfies these properties and call it the collection of open sets of M, even if they does not satisfy the definiation of openness based on metric or even M is not a metric space at all.

- 2 Theory of Probability
- 2.1 Measure Theory
- 2.2 Measurable Functions
- 2.3 Absolute Continuity
- 2.4 Mathematical Probability
- 2.5 Random Variable
- 2.6 Conditioning
- 3 Stochastic Processes
- 3.1 Definition
- 3.2 Random Functions