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# 1. Topology

## **Definition 1.1** (Topology):

Let X bet a set. A **topology over** X is a subset  $\Sigma$  of  $2^X$  such that:

- 1.  $A \subseteq \Sigma \Longrightarrow \bigcup_{E \in A} E$ . Infinite or finite unions of sets.
- 2.  $A, B \in \Sigma \Longrightarrow A \cap B \in \Sigma$ . Finite intersections of sets.
- 3.  $X \in \Sigma$

## **Definition 1.2** (Topological Space):

 $(X, \Sigma)$  is a **topological space** iff.  $\Sigma$  is a topology of X.

## **Definition 1.3** (Everywhere dense):

Let  $(X, \Sigma)$  topological space, and  $H \subseteq X$ . H is said everywhere dense in  $\Sigma$  iff.  $\forall E \in \Sigma, E \neq \emptyset : H \cap E = \emptyset$ . We can find a bit of H in every corner of the topology  $\Sigma$ .

# **Definition 1.4** (Separable):

Let  $(X, \Sigma)$  be a topological space.  $(X, \Sigma)$  is said **separable** iff  $\exists H \subseteq X, H$  is countable: H is everywhere dense  $\in \Sigma$ . There is a sequence of elements  $\{x_n \in X\}_{n=1}^{\infty}$  such that every set in the topology contains at least one element  $x_i$ .

## **Definition 1.5** (Metric Space):

(X,d) is a metric space iff.

- 1.  $X \neq \emptyset$
- 2.  $d: X \times X \longrightarrow \mathbb{R}_{>0}$  such that (d is a distance):
  - 1.  $\forall x, y \in X : d(x, y) = 0 \Longrightarrow x = y$ . there are no different elements at zero-distance.
  - 3.  $\forall x, y \in X : d(x, y) = d(y, x)$ . symmetry.
  - 2.  $\forall x, y, z \in X : d(x, z) \leq d(x, y) + d(y, z)$ . triangular inequality.

## **Definition 1.6** (open $\varepsilon$ -ball):

Let (X, d) be a metric space,  $x \in X$ , and  $\varepsilon \in \mathbb{R}_{>0}$ . We call  $B_{\varepsilon}(x) = \{y \in X \mid d(x, y) < \varepsilon\}$  an open  $\varepsilon$ -ball. A ball of  $\varepsilon$  radius centered at some point.

## **Definition 1.7** (Neighborhood):

Let (X, d) be a metric space,  $S \subseteq X$ ,  $x \in S$ , and  $\varepsilon \in \mathbb{R}_{>0}$  such that the open  $\varepsilon$ -ball  $B_{\varepsilon}(x) \subseteq S$ . Then S is said a **neighborhood of** x. A neighborhood of an element is simply a set that contains an open ball containing the element.

# **Definition 1.8** (Open Set):

Let (X, d) be a metric space and  $U \subseteq X$ . U is an open set iff.  $\forall u \in U : \exists \varepsilon \in \mathbb{R}_{>0} : B_{\varepsilon}(u) \subseteq U$ . An open set is simply a set which is also neighborhood for all its points.

## **Definition 1.9** (Induced Topology):

Let (X,d) be a metric space.  $\Sigma$  is said an **induced topology** iff.  $\Sigma = \{U \subseteq X \mid U \text{ is an open set in } (X,d)\}$ 

## **Definition 1.10** (Metrizable):

Let  $(X, \Sigma)$  be a topological space.  $(X, \Sigma)$  is said **metrizable** iff.  $\exists (X, d)$  metric space :  $\Sigma$  is a topology induced by (X, d).

## **Definition 1.11** (Cauchy Sequence):

Let (X,d) be a metric space,  $[x_n \in X]$  a sequence.  $[x_n]$  is said a **cauchy sequence** iff.  $\forall \varepsilon \in \mathbb{R}_{>0}: \exists N \in \mathbb{N}: \forall m,n \in \mathbb{N}: d(x_n,x_m) \leq \varepsilon$ . There is a point after which all pairs of elements are close to each other.

### **Definition 1.12** (Convergent Sequence):

Let (X,d) be a metric space,  $l \in X$ ,  $[x_n \in X]$  a sequence.  $[x_n]$  is said a **convergent sequence** to the limit l iff.  $\forall \varepsilon \in \mathbb{R}_{>0}: \exists N \in \mathbb{R}_{>0}: \forall n > N: d(x_n,l) < \varepsilon$ . If such a limit exists the sequence is simply said **convergent**.

### **Definition 1.13** (Complete Metric Space):

Let (X, d) be a metric space. (X, d) is said a **complete metric space** iff. every cauchy sequence is convergent.

# **Definition 1.14** (Polish Space):

Let  $(X, \Sigma)$  be a topological space.  $(X, \Sigma)$  is said a **Polish Space** iff.  $(X, \Sigma)$  is separable, metrizable, and a complete metric space for some metric.

# 2. Measure Theory

## **Definition 2.1** ( $\sigma$ -algebra):

Let X be a set.  $\Sigma \subseteq 2^X$  is said a sigma algebra of X iff.:

- 1.  $X \in \Sigma$
- 2.  $E \in \Sigma \Longrightarrow X \setminus E \in \Sigma$ . close under complement.
- 3.  $\{A_n \in \Sigma\}_{n=1}^{\infty} \Longrightarrow \bigcup_{i=1}^{\infty} A_i \in \Sigma$ . close under infinite unions.

# **Definition 2.2** (generate $\sigma$ -algebra):

Let X be a set and  $G \subseteq 2^X$ . The  $\sigma$ -algebra generated by G, denoted  $\sigma(G)$ , is the smallest  $\sigma$ -algebra such that:

- 1.  $G \subseteq \sigma(G)$ .
- 2.  $\forall \Sigma$   $\sigma$ -algebra :  $G \subseteq \Sigma \Longrightarrow \sigma(G) \subseteq \Sigma$ . Every other  $\sigma$ -algebra that contains G contains also the generated one,  $\sigma(G)$ .

## **Definition 2.3** ( $\sigma$ -algebra product):

Let  $\Sigma_1$  and  $\Sigma_2$  be  $\sigma$ -algebras on  $X_1$  and  $X_2$  respectively. The **product**  $\sigma$ -algebra denoted  $\Sigma_1 \otimes \Sigma_2$  is defined as  $\sigma(\{S_1 \times S_2 \mid S_1 \in \Sigma_1, S_2 \in \Sigma_2\})$ 

## **Definition 2.4** (measurable space):

 $(X, \Sigma)$  is said **measurable** iff.  $\Sigma$  is a sigma-algebra of X.

#### **Definition 2.5** (measure):

Given  $(X, \Sigma)$  measurable space.  $\mu : \Sigma \longrightarrow \mathbb{R} \cup \{+\infty, -\infty\}$  is said a **measure** iff.

- 1.  $E \in \Sigma \Longrightarrow \mu(E) \geq 0$ . positive.
- 2.  $\{E_n \in \Sigma\}_{n=1}^{\infty}$  such that  $E_i \cap E_j$  for  $i \neq j \Longrightarrow \mu(\bigcup_{i=1}^{\infty} E_i) = \sum_{i=1}^{\infty} \mu(E_i)$ . The measure of disjoint sets is is the sum of the measures of each set.
- 3.  $\exists E \in \Sigma : \mu(E) \in \mathbb{R}_{\geq 0}$ . For at least an element  $\mu$  is finite.

### **Definition 2.6** (measure space):

 $(X, \Sigma, \mu)$  is said a **measure space** iff.  $(X, \Sigma)$  is a sigma algebra and  $\mu$  is a measure of  $(X, \Sigma)$ .

# 3. Probability Theory

# **Definition 3.1** (Probability Space):

- $(\Omega, \Sigma, p)$  is said a **probability space** iff.
- 1.  $(\Omega, \Sigma, p)$  is a measure space.
- 2.  $p(\Omega) = 1$ .

Intuitively,  $\Omega$  represents the set of all possible outcomes, it is also known as **sample space**.  $\Sigma$  represents the set of all possible events. These are nothing more than set of outcomes. It is also known as **event space**. p is a measure on the event space, it is also known as **probability function**. It maps events to their likelihood.

### Example 3.1 (Fair Die):

Consider the probability space  $(\Omega, \Sigma, p)$ , where:

- 1.  $\Omega = \{1, 2, 3, 4, 5, 6\}$  is the sample space, representing the possible outcomes of rolling a standard six-sided die.
- 2.  $\Sigma = 2^{\Omega}$  is the event space.
- 3.  $p: \Sigma \longrightarrow [0,1]$  is the probability measure function, defined as  $P(E) = \frac{|E|}{6}$  for any event  $E \in \Sigma$ .

For example, consider the event  $A = \{1, 2, 3\}$ , which represents rolling a 1, 2, or 3. This event is an element of  $\Sigma$ . The probability of event A occurring is  $p(A) = \frac{|A|}{6} = \frac{3}{6} = \frac{1}{2}$ .

## **Definition 3.2** (Coupling):

Let  $(\Omega_1, \Sigma_1, \mu_1)$  and  $(\Omega_2, \Sigma_2, \mu_2)$  be probability spaces. A **coupling** is a probability space  $(\Omega_1 \times \Omega_2, \Sigma_1 \otimes \Sigma_2, \gamma)$  such that:

- 1.  $\forall E \in \Sigma_1 : \gamma(E \times \Omega_2) = \mu_1(E)$ .
- $\text{2. }\forall E\in \Sigma_2: \gamma(\Omega_1\times E)=\mu_2(E).$

# 4. Wasserstein Distance