

# Bedlam

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# 1. Topology

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## Definition 1.1 (Topology):

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

1.  $A \subseteq \Sigma \implies \bigcup_{E \in A} E$ . Infinite or finite unions of sets.
2.  $A, B \in \Sigma \implies A \cap B \in \Sigma$ . 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.  $\forall E \in \Sigma, E \neq \emptyset : H \cap E \neq \emptyset$ . 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.  $d : X \times X \longrightarrow \mathbb{R}_{\geq 0}$  such that (**d is a distance**):
  1.  $\forall x, y \in X : d(x, y) = 0 \implies x = y$ . there are no different elements at zero-distance.
  2.  $\forall x, y \in X : d(x, y) = d(y, x)$ . **symmetry**.
  3.  $\forall x, y, z \in X : d(x, z) \leq d(x, y) + d(y, z)$ . **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** (Metrizible):

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

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**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.  $E \in \Sigma \implies X \setminus E \in \Sigma$ . **close under complement.**
3.  $\{A_n \in \Sigma\}_{n=1}^\infty \implies \bigcup_{i=1}^\infty A_i \in \Sigma$ . **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 :  $G \subseteq \Sigma \implies \sigma(G) \subseteq \Sigma$ . **Every other  $\sigma$ -algebra that contains  $G$  contains also the generated one,  $\sigma(G)$ .**

**Definition 2.3** (Borel  $\sigma$ -algebra):

The **Borel  $\sigma$ -algebra**, denoted  $\mathcal{B}(\mathbb{R}^n)$ , is  $\sigma(\{(a_0, b_0) \times \dots \times (a_n, b_n) \mid a_0 \in \mathbb{R}, b_0 \in \mathbb{R}, a_0 < b_0, \dots, a_n \in \mathbb{R}, b_n \in \mathbb{R}, a_n < b_n\})$ . **The smallest  $\sigma$ -algebra generated by the set of all open intervals on  $\mathbb{R}^n$ .**

Further, we extend the **Borel  $\sigma$ -algebra** to open intervals.  $\mathcal{B}(I = (a_0, b_0) \times \dots \times (a_n, b_n)) = \sigma(\{A \text{ in } \mathcal{B}(\mathbb{R}^n) \mid A \subseteq I\})$  **TODO: Show that this is indeed another sigma algebra**

**Definition 2.4** ( $\sigma$ -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 2.5** (measurable space):

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



**Definition 2.6** (measure):

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

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

**Definition 2.7** (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

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**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.  $p : \Sigma \rightarrow [0, 1]$  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.  $\forall E \in \Sigma_1 : \gamma(E \times \Omega_2) = \mu_1(E)$ .
2.  $\forall E \in \Sigma_2 : \gamma(\Omega_1 \times E) = \mu_2(E)$ .

## 4. Wasserstein Distance

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