

Def

Measures

M-1

We say that μ is a measure on a space X if

(i) $\mu(\emptyset) = 0$

(ii) $\mu(A) \geq 0$ for $A \subset X$

(iii) If A_1, A_2, \dots are disjoint subsets of X ,

then

$$\mu\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} \mu(A_n).$$

REMARKS. In general $\mu(A)$ is not defined for all subsets A of X . We denote the collection of subsets A for which $\mu(A)$ is defined by \mathcal{Q} . For the above definition to make sense, it is sufficient that \mathcal{Q} have the following three properties:

(i) $\emptyset \in \mathcal{Q}$

(ii) $A \in \mathcal{Q} \Rightarrow A^c \in \mathcal{Q}$

(iii) A_1, A_2, \dots in $\mathcal{Q} \Rightarrow \bigcup_{n=1}^{\infty} A_n \in \mathcal{Q}$.

Such a collection of subsets is called a σ -algebra or a σ -field.

Example 1. Fix any $x_0 \in X$, and put

$$\mu(A) := I_A(x_0) = \begin{cases} 1, & x_0 \in A, \\ 0, & x_0 \notin A. \end{cases}$$

Example 2. If μ is a measure and $C \geq 0$, M - \mathbb{Z} then $\nu(A) := C\mu(A)$ defines another measure.

Example 3. If μ_1, μ_2, \dots are measures, then so are

$$\mu(A) := \sum_{n=1}^{\infty} \mu_n(A).$$

and

$$\nu(A) := \sum_{n=1}^{\infty} c_n \mu_n(A),$$

assuming all $c_n \geq 0$.

Proposition. If $A \subset B$, $\mu(A) \leq \mu(B)$

Proof. Write $B = A \cup (B \setminus A)$. Since this is a disjoint union,

$$\begin{aligned} \mu(B) &= \mu(A) + \mu(B \setminus A) \\ &\geq \mu(A) \end{aligned}$$

□

Def. If $\mu(\mathbb{X}) < \infty$, μ is called a finite measure. If $\mu(\mathbb{X}) = 1$, μ is called a probability measure.

Def. The pair $(\mathbb{X}, \mathcal{A})$ is called a measurable space, and the triple $(\mathbb{X}, \mathcal{A}, \mu)$ is called a measure space. If $\mu(\mathbb{X}) = 1$, $(\mathbb{X}, \mathcal{A}, \mu)$ is called a probability space.

In a probability space, $0 \leq \mu(A) \leq 1$ for all $A \in \mathcal{A}$.

Example 4. Let x_1, x_2, \dots be distinct points $M-\mathbb{Z}$ in \mathbb{I} , and let p_1, p_2, \dots be nonnegative numbers.

Then

$$\mu(A) = \sum_{n=1}^{\infty} p_n \mathbb{I}_A(x_n)$$

defines a measure. Also note that

$$\mu(\{x_k\}) = \sum_{n=1}^{\infty} p_n \mathbb{I}_{\{x_k\}}(x_n) = p_k.$$

If A contains x_i and x_j but no other x_n , then

$$\mu(A) = p_i + p_j.$$

Fact: Lebesgue measure \mathcal{L} on \mathbb{R}^d is characterized as follows:

$$d=1: \mathcal{L}([a, b]) = b - a$$

$$d=2: \mathcal{L}([a, b] \times [c, d]) = (b - a)(d - c)$$

$$d=3: \mathcal{L}([a, b] \times [c, d] \times [e, f]) = (b - a)(d - c)(f - e)$$

...

In other words, the Lebesgue measure of an interval is its length, the Lebesgue measure of a rectangle is its area, and the Lebesgue measure of rectangular parallelepiped is its volume, ...

Def. A counting measure is a measure that takes only ^{nonneg.} integer values (or $+\infty$). M-4

Example 5. Fix any x_1, x_2, \dots in \mathbb{X} and put,

$$\mu(A) := \sum_{k=1}^{\infty} \mathbb{I}_A(x_k)$$

Then $\mu(A) = \#x_k \in A$.

Example 6. If X_1, X_2, \dots are \mathbb{X} -valued random variables, then

$$N(A) := \sum_{k=1}^{\infty} \mathbb{I}_A(X_k)$$

is a random counting measure. Its mean measure is

$$\Lambda(A) := E[N(A)].$$

Example 7. (Poisson Counting Measure). We say N is a Poisson counting measure if whenever A_1, \dots, A_m are disjoint, $N(A_1), \dots, N(A_m)$ are independent (II), and

$$P(N(A)=n) = \frac{\Lambda(A)^n e^{-\Lambda(A)}}{n!}, \quad n=0,1,\dots$$

and Λ is a measure with no point masses.